

A Dissertation

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The Impact of Time-Based Accounting
on Manufacturing Performance

by

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An Abstract
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The Impact of Time-Based Accounting
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Advanced manufacturing strategies are often undermined by the legacy control mechanisms of another era. In response to today's hypercompetitive global business environment, many manufacturing firms have adopted new business strategies such as time-based manufacturing (TBM). While some firms credit their success in the marketplace to TBM and other advanced manufacturing strategies, still more have realized little or no improvement in performance. Post-implementation failures often find their root not in the manufacturing strategy itself, but in the lack of congruent control mechanisms. One of such important control mechanism is the firm's managerial accounting system (MAS).

Despite evident appeal to both managerial accounting and manufacturing management, few managerial accounting innovations have developed at this interface. There is increasing recognition amongst managerial accounting scholars of the need for innovative studies to fill in the knowledge gaps and provide normative solutions for

managerial accounting practice that has increasingly lost its relevance to modern manufacturing.

This study demonstrates that the design of the MAS can significantly affect key manufacturing performance measures. Market, operational, and financial performance measures are utilized in this study in terms of demand fulfillment rate (DFR), cycle-time (CT), and net operating income (NOI) respectively. Three MAS alternatives – traditional costing systems (TCS), activity-based costing systems (ABC), and the newly proposed time-based accounting (TBA) methods – are all examined. This study employs a computer simulation methodology, which presents an opportunity for the direct comparison and quantitative measurement of this impact. The choice of methodology allows for the controlled introduction of supply and demand stochasticity along with differing levels of product mix complexity, modeled in environments with differing levels of manufacturing overhead burden.

The results provide significant evidence of the importance of MAS design vis-à-vis manufacturing strategy. No single MAS outperformed all others along all performance measures, indicating the need to align the MAS to the manufacturing strategy. While MAS design alone may not ensure the success of an advanced manufacturing strategy, this study clearly demonstrates that it should be considered an integral part of the initiative. The introduction of TBA demonstrates how MAS design can better support a TBM strategy and regain lost relevance.

For Mariana

All thanks be to God

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Chapter One

Introduction

Today's advanced manufacturing strategies are often undermined by legacy control systems designed in another era. In response to today's hypercompetitive global business environment, many manufacturing firms have adopted new business strategies such as time-based manufacturing (TBM) and invested heavily in advanced manufacturing technologies in order to maintain their market position over competitors. While some firms credit their success in the marketplace to TBM and other advanced manufacturing strategies, still more have realized little or no improvement in performance. In some cases, implementation has been reported to have had a negative effect on the performance, and, in extreme pathological cases, some firms have even blamed such strategies for complete demise of the company. Quite often, the failure can be traced back not to the manufacturing strategy and/or technologies, but to the lack of appropriate control systems post-implementation (Womack & Jones 1996).

One of the most crucial control systems is the firm's internal managerial accounting system (MAS), which plays a decisively important role in supporting manufacturing strategy. Yet the blatant insignificance conferred upon it by a great number of academic researchers and practitioners remains paradoxical. Empirical research suggests that MAS design rarely reflects differences in manufacturing strategy, operating environment, or competitive pressures (Hughes & Paulson-Gjerde 2003).

Rather than suggest ways to better incorporate strategic measures within the existing managerial accounting framework, many consultants and researchers have advocated an increased use of non-financial measures, i.e. operational measures, in lieu of traditional cost measures (Chenhall 1997, Sim & Killough 1998, Hoque 2003). The so called 'balanced scorecard' has thus become a popular topic for researchers seeking to restore relevance for cost accounting. Still others believe that in order to influence managers to do the right things, traditional manufacturing cost accounting systems may need to be abandoned altogether (Womack & Jones 2003). Regardless of cost accounting's potential flaws, the bottom line – pun intended – on solutions that would marginalize or even eliminate financial measures is that few manufacturing managers can simply ignore product cost. This presents a real conundrum to manufacturing managers as strategic non-financial performance measures have increasingly come into direct conflict with legacy cost measures that are artifacts of another era of manufacturing.

What is really needed is not a complete abandonment of manufacturing cost accounting systems, but an adaptation that directly links strategic success measures, such as time-based performance measures, to product cost. Alternatives to financial measures, such as the balanced scorecard, remain an enigma for researchers as they are often firm-specific and are seldom identified in practice. The fatal flaw for is that they fail to build on the strengths of manufacturing cost accounting. In the words of Peter Drucker (1990), the world's preeminent authority on management theory:

Cost accounting's strength has always been that it confines itself to the measurable and thus gives objective answers. But if intangibles are brought into its equations, cost accounting will only raise more questions. How to proceed is thus hotly debated, and with good reason. Still, everyone agrees that these business impacts have to be integrated into the measurement of factory performance, that is, into manufacturing accounting.

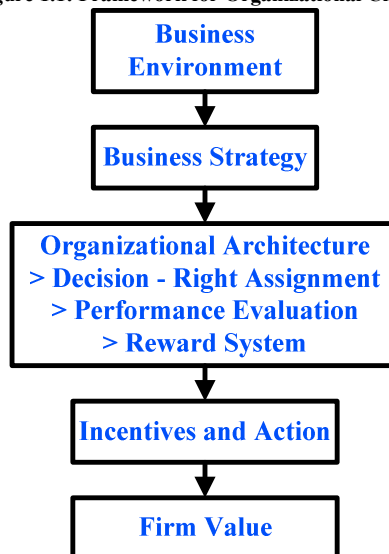
Managerial Accounting Function

Although not usually ranked high in organizational priorities or in academic research models, a firm's MAS provides the ultimate scorecard for management. It is critical that the MAS exhibit the following characteristics (Zimmerman 2003):

1. Provide the information necessary to identify the most profitable products and the pricing and marketing strategies to achieve desired volume levels.
2. Provide information to detect production inefficiencies to ensure that the proposed products and volumes are produced at minimum cost.
3. When combined with the performance evaluation and reward systems, create incentives for managers to maximize firm value.
4. Support the financial accounting and tax accounting reporting functions.
5. Contribute more to firm value than it costs.

The objectives of the managerial accounting function are to assist managers and to influence their behavior in a way that results in goal congruent actions (Anthony 1989). Figure 1.1 below presents a framework for organizational change and managerial accounting's role in driving the action that leads to firm value.

Figure 1.1: Framework for Organizational Change

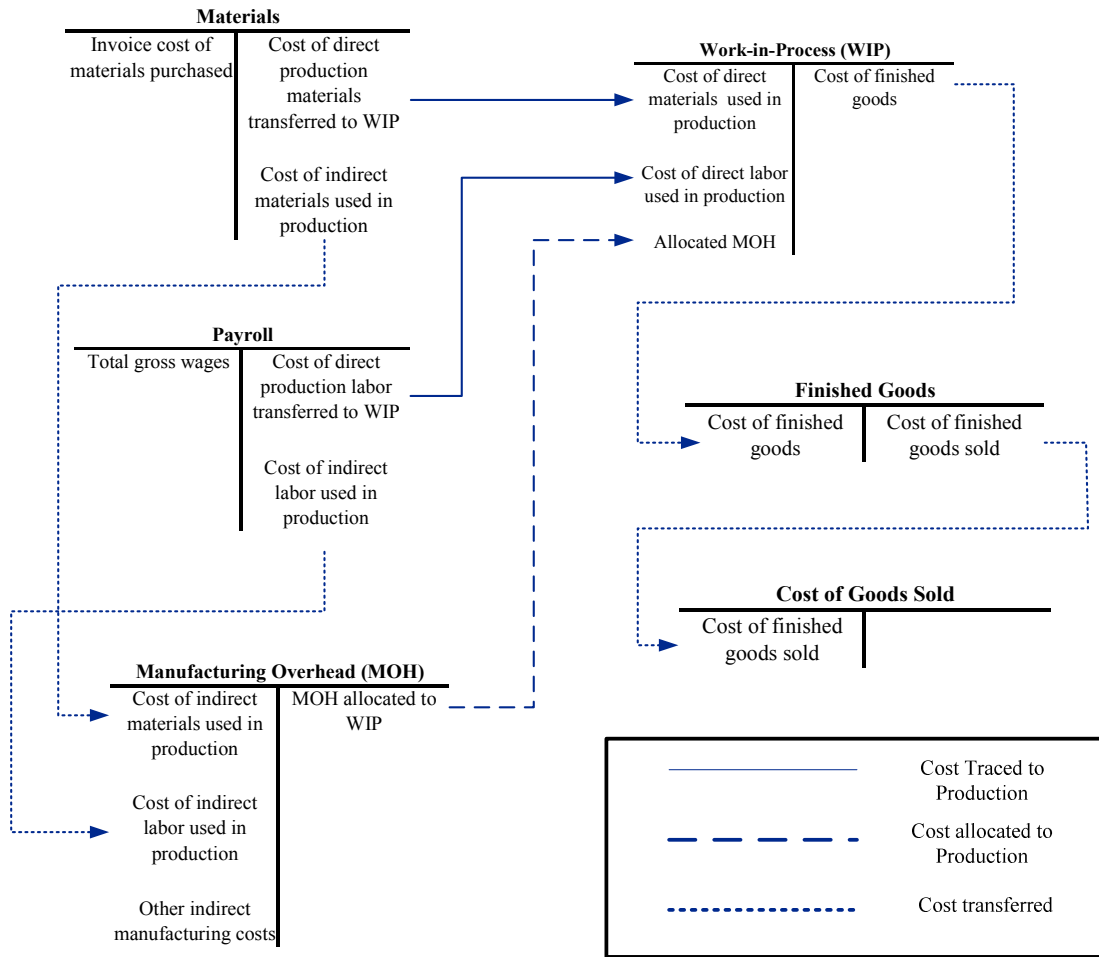


Changes in business environment should lead to the evolution of new strategies and ultimately to changes in the firm's organizational architecture, including changes in the

MAS to better align employee incentives to the objectives of the organization (Zimmerman 2003). The MAS is critical in setting profit goals, establishing departmental targets in the form of budget plans, evaluating the effectiveness of resource usage against those plans, investigating successes and failures in terms of specific manufacturing processes and support tasks, and taking action on adjustments and improvements necessary to keep the entire manufacturing enterprise moving towards the established strategic objectives. This raises the question in today's increasingly time-based competitive environment; to what extent do managerial accounting practices incorporate the strategic importance of time into product cost? The answer for many manufacturing firms is that MAS congruence is an afterthought to strategy, if given any consideration at all (Hughes & Paulson-Gjerde 2003).

Perhaps one of the reasons MAS alternatives are rarely considered in conjunction with implementation of TBM or any other advanced manufacturing strategy, may be a general lack of understanding, particularly for non-accountants, as to the fundamental differences between these alternatives. Essentially, the differences between most MAS alternatives come down largely to how manufacturing overheads are allocated or 'absorbed' into product cost. It is often relatively easy to trace costs such as direct materials, direct labor, and certain conversion costs – such as utilities – to particular products. However, in most modern manufacturing facilities, the proliferation of advanced manufacturing strategies and technologies make it difficult, if not impossible, to trace manufacturing overhead burden directly to any particular product or product class. Figure 1.2 below illustrates the typical flow of material, labor, and overhead costs through the work-in-process and finished goods accounts ending in cost of goods sold.

Figure 1.2: Cost Flow Diagram



For all intents and purposes, the MAS may be thought of as nothing more than a methodology for capturing and allocating costs to products, with the defining difference between alternatives being how manufacturing overhead are allocated. The basic difference then becomes the basis for allocating manufacturing overheads. While this may seem simple enough in concept, in reality it becomes a major issue as direct costs, particularly manual labor, have become increasingly insignificant in relation to manufacturing overheads driven by increasing capital investments in advanced manufacturing strategies and technologies.

With the widespread adoption of advanced manufacturing strategies and technologies, many managers have questioned the ability of traditional cost systems

(TCS) to achieve its objectives of strategic planning, budgeting and cost control, product pricing, profit determination, and especially employee motivation. In fact, TCS has often been charged as one of the major obstacles to the success of any advanced manufacturing strategy. Goldratt, whose ‘theory of constraints’ manufacturing has developed a worldwide cult following, labeled TCS “enemy number one to operational productivity” (Goldratt & Cox 1992). Despite evidence heaped against it from academics and management consultants, most firms in practice continue to use TCS (Garg et al. 2003). Ittner and Larcker (2003) contend that managerial accounting researchers themselves hold much of the responsibility for the failure of managerial accounting practice to effectively adapt.

Kaplan’s (1983) call for greater emphasis on manufacturing performance measurement spurred considerable enthusiasm for research on this topic in the managerial accounting community. Notwithstanding this initial enthusiasm and the potential contribution both to scholarly research and practice, the pursuit of new ‘hot’ topics quickly and substantially reduced research at this interface of accounting and operations management... We are left with an underdeveloped body of [manufacturing accounting] research that fails to build on prior studies to increase our understanding of the topic, leaving many important research topics unexplored, and lacks the critical mass of related studies needed to reconcile conflicting results.

TCS has lost its relevance to the dynamic and fast advancing nature of the modern manufacturing enterprise (Johnson & Kaplan 1987). Nevertheless, case studies suggest that cost remains the primary, and in the majority of cases the only, decision criterion in manufacturing (Veen-Dirks 2006). Continued debate on balancing cost and non-financial measures is of dubious practical value to operations management. The overarching question for research is what effect, if any, different MAS alternatives, with differing methodologies for allocating manufacturing overhead burdens, have on manufacturing

system performance. A particular challenge for management accountant researchers today is developing MAS alternatives for practice that directly link manufacturing strategic measures to product cost.

The Productivity Paradox

Beginning in the 1980s, American industry invested heavily in advanced manufacturing strategies and technologies in order to regain competitive advantage lost to Japanese manufacturers. The majority of manufacturers found implementation of these advanced manufacturing systems problematic and realized that fundamental changes in production must be coupled with corresponding changes in control systems. In particular, managerial accounting research began to consider the ‘perverse’ management incentives created by TCS. Skinner (1986) termed this type of perverse behavior the ‘productivity paradox’, where manufacturing managers strive to better their measure of productivity rather than making true productivity improvements.

Fry and others (1993) contend that continued reliance on TCS may result in dysfunctional behavior vis-à-vis time-based strategic objectives including:

- Batching of orders to minimize setups at the expense of higher priority orders
- Production overruns and excess inventory to create favorable variances
- Cherry picking easy orders at the expense of higher priority orders
- Delaying preventive maintenance to increase machine utilization
- Understaffing in departments to increase worker productivity
- Delay operator training and cross training to prevent under-utilization
- Building a buffer of backorders to ensure that an adequate level of work exists

In their well-known book, *Lean Thinking*, Womack and Jones (1996) combined the core practices of several advanced manufacturing strategies under a broad conceptual

umbrella, which they termed ‘lean manufacturing’. They examined the application of the key practices of lean manufacturing across the value chains of many global manufacturing companies. The authors found commonalities between those companies that were successful not only in implementing a lean initiative but also in sustaining the initiative. The following is a list of the critical success factors they observed at firms that successfully implemented and sustained lean manufacturing programs:

1. *Institutionalizing change* – organizations that accept change as continuous process are more successful in their pursuit of lean manufacturing. Constant change becomes part of the organizational culture, and employees not only accept it as necessary to doing business, but embrace it as a competitive weapon.
2. *Find a leverage point* – organizations on the brink of failure are more likely to accept the radical changes (item 3) in business practices and processes needed to save the company. It gives the employees a sense of urgency knowing that failure of the ‘lean’ program is not an option.
3. *Radical change* – organizations that begin the ‘lean’ process with a complete reengineering of their business systems and processes are more likely to succeed in instituting lasting change throughout the organization. If the firm is on a course of bankruptcy only a 180 degree turn will save it from destruction.
4. *Continuous improvement* – once the firm has set a course for improvement, change must focus on continuous and incremental improvement of business processes.
5. *Install business support systems* – once continuous improvement and change has been institutionalized, support systems must be adjusted to support and encourage a continuation of the program.
 - a. *Accounting systems* – those firms that change their accounting systems to better reflect the goals of the ‘lean’ organization are more likely to succeed. Leaving the legacy accounting system, with its focus on past goals, can create perverse incentives for managers.
 - b. *Compensation systems* – incentives should be given based on performance according to the goals of the company.
6. *Educate* – as part of the system of continuous improvement and employee empowerment, education becomes critical to continued success.
7. *Create transparency* – business systems need to be designed to create transparency throughout the organization, and information should be shared freely among the various organizations and functional departments.

Womack and Jones stressed the importance of installing appropriate business systems (item 5) to the long-term, continued success of lean manufacturing initiatives, yet this is often forgotten or ignored after the hype of installation on the plant floor. Legacy accounting systems, many of which were developed and widely adopted before World War II, when average cost of production was primary focus of manufacturing, remain in common use and may create perverse incentives for managers that undermine the continued performance of the lean system. It is for this reason that managerial accounting plays a critical roll in lean initiatives, and, according to the authors, the fact that it is often ignored may explain a great number of failures cited by companies that have tried to implement some form of advanced manufacturing strategy.

Sustained success of any advanced manufacturing strategy repeatedly comes down to having goal-congruent support systems, such as the MAS, in place post-implementation. Unfortunately, in many cases these ‘backroom’ processes are forgotten or simply ignored after the hype of installation on the plant floor. A recent study of three flexible manufacturers similarly found general incongruence of managerial accounting practices with operational strategy. At one particular facility, only one of the sixteen initiatives being reviewed by management was not considered a cost-reduction, despite a supposed emphasis on manufacturing flexibility measures (Veen-Dirks 2006). Again, this suggests an inherent weakness to the ‘balanced’ scorecard in that cost measures nearly always trump operational measures when they come into conflict.

Managerial accounting plays a critical roll in any advanced manufacturing strategic initiative, and the fact that it is often ignored may explain the great number of failures. The poor performance of many new manufacturing initiatives is due, in part, to

continued reliance on an MAS alternative that fails to provide appropriate goals, performance measures, rewards systems (Kaplan 1983, 1993). Performance measures directly related to business goals are required to ensure that manufacturing processes are in control and continuously improved (Kaplan 1983, Drucker 1990).

Research Objectives, Questions, and Framework

The relationship between time-based manufacturing practices and manufacturing performance is well established in the operations management literature and therefore will not be the focus of this study. However, the strategic importance of time is acknowledged *a priori* as the contextual milieu for this research. *The primary objectives of this study are (1) to examine the impact of different managerial accounting system (MAS) alternatives on manufacturing performance measures and (2) to propose an MAS alternative that better links time-based strategic measures to product cost.* The resultant research questions and a testable research framework are developed along with clearly defined experimental factors, performance measures, and a simulation research methodology is developed in subsequent chapters.

The four managerial accounting systems considered in this study are traditional costing systems (TCS), activity-based costing (ABC), throughput accounting (TA), and the proposed time-based accounting system (TBA). The study is specifically concerned with the impact of various MAS alternatives on production mix decisions (PMD), with the effect of these decisions on manufacturing performance measures (MPM), analyzed in terms of average demand fulfillment rate, throughput-time, and net operating income.

These manufacturing performance measures were chosen because they represent both financial and non-financial and operational and market measures of performance.

In addition, this study examines the impact of MAS choice under varying levels of manufacturing overheads (MOH) and product mix complexity (MIX). Product mix complexity is a multidimensional variable measured through the depth of the bill of material for various products and the variability of demand for those products at a single supply point. The width of a bill of material has been defined within the operations management literature as the maximum number of dependent relationships in a product structure (Veral & Laforge 1985, Sum et al. 1993). Product mix complexity is widely acknowledged in operations management research as one of the primary factors in determining a product cost under different MAS alternatives and has been a primary variable in studies of lot sizing (Blackburn and Millen 1980; 1982a & b, Veral and LaForge 1985, Benton and Srivastava 1985, LaForge 1985), system nervousness strategy (Blackburn et al. 1986), and in capacity control policy (Gutzmann and Wysk 1986).

Product mix complexity defined as the demand variability for different products produced within the same facility has long been acknowledged in the managerial accounting literature as one of the primary drivers of manufacturing cost and may significantly contribute to cost distortions under traditional managerial accounting scenarios (Cooper 1988a , Brimson 1991). Many researchers have posited that TCS may under allocate MOH burden to the low volume, complex products and may over allocate manufacturing overhead burden to the high volume, simple products when both types of products are manufactured in the same facility, because manufacturing overheads are

most often allocated on direct labor hours, machine hours, or some other measure of production volume (Johnson 1991b, Chalos 1992, O'Guin 1991).

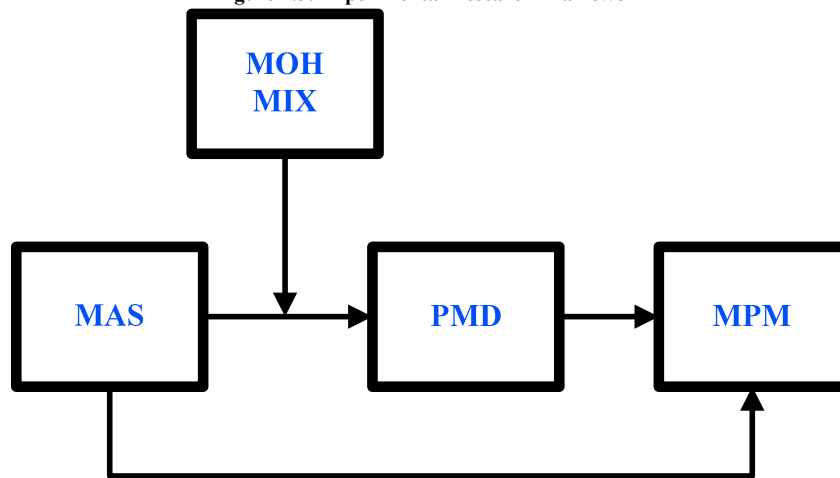
Theory in both managerial accounting and operations management has predicted that producing a heterogeneous product mix within a single facility may increase total cost and reduce operating performance (Skinner 1974, Hayes & Wheelwright 1984, Hill 1985, Johnson & Kaplan 1987, Banker et al. 1988). Empirical studies would later confirm the impact of product variety on manufacturing costs (Anderson 1995, Fisher & Ittner 1999, Randall & Ulrich 2001, Pil & Holweg 2004, Peacock 2005). Yet few studies have integrated these factors from managerial accounting and operations management within a single research model.

The experimental framework below in figure 1.3 shows the hypothesized relationship between the managerial accounting system alternative and the production mix decision. This relationship is mediated by the level of manufacturing overhead and product mix complexity. The production mix decision and the managerial accounting system alternative then together affect the manufacturing performance measures. The following research questions will be answered through the experimental research framework presented below in figure 1.2:

1. Does the use of different managerial accounting systems in production decisions result in differences in average demand fulfillment rate, cycle-time, and/or net operating income?
2. What effect does manufacturing overhead level have on average demand fulfillment rate, cycle-time, and/or net operating income?
3. For a given level of manufacturing overhead level does the use of different managerial accounting systems in production decisions result in differences in average demand fulfillment rate, cycle-time, and/or net operating?
4. What effect does product mix complexity have on average demand fulfillment rate, cycle-time, and/or net operating income?

5. For a given level of product mix complexity does the use of different managerial accounting systems in production decisions result in differences in average demand fulfillment rate, cycle-times, and/or net operating income?
6. For a given level of product mix complexity does manufacturing overhead level affect result in differences in average demand fulfillment rate, cycle-time, and/or net operating income?
7. For a given level of product mix complexity and a given level of manufacturing overhead level does the use of different managerial accounting systems in production decisions result in differences in average demand fulfillment rate, cycle-time, and/or net operating income?

Figure 1.3: Experimental Research Framework



Research Methodology

The impact of various managerial accounting systems on manufacturing performance vis-à-vis a time-based manufacturing strategy is of primary importance to this study and will be tested via computer simulation, an emergent and increasingly accepted methodology in industrial engineering and operations management research. This methodology was chosen for several reasons, not the least of which the fact that few examples of firms utilizing innovative MAS alternatives are to be found in practice.

Traditional or [positivist] social science research methods, such as empirical analysis of large data sets, analytic models of accepted and understood phenomena, and cross-sectional field research, can be effective for studying the universe as it now exists, for understanding “what-is”. But these normal science methods are less helpful for managerial accounting research where major social structural changes are occurring in organizations and in the roles performed by managerial accounting systems within these organizations (Kaplan 1993)

There has been an increasing recognition amongst some of the prominent scholars in the managerial accounting field of the need for innovative studies borrowing novel methodologies from other disciplines to fill in the knowledge gaps in the literature (Kaplan 1993, Ittner & Larcker 2001, 2003, Zimmerman 2001). As has been suggested, at the heart of the problem rests the wider accounting academic community that has largely fortified itself in its accepted research methodologies and avoided the interface with other research traditions altogether.¹ A full decade after his seminal article that questioned the relevance of managerial accounting practice and research to the modern manufacturing, Kaplan surveyed the managerial accounting field finding many opportunities remain undeveloped in relation to operations management (1993):

Managerial accounting scholars, unlike their counterparts in operations management (OM), could not easily find widespread adopters of innovative practices. Our OM colleagues could look to leading Japanese manufactures in the automobile, machine tool, and electronics industries for observable examples of total quality management, just-in-time production, computer-integrated manufacturing, and design for manufacturability... But no ‘Japan’ existed for learning about or studying innovative managerial accounting practices. Therefore, standard cross-sectional field research studies would largely capture traditional managerial accounting systems operating in environments radically different from the ones for which the systems were designed.

¹ It is interesting to note that many of the preeminent scholars in the field of managerial accounting, such as Robert S. Kaplan, H. Thomas Johnson, Robin Cooper, and Eliyahu M. Goldratt, have not been trained as accountants at all. Rather their educational backgrounds include such diverse fields as electrical engineering, operations research, industrial engineering, chemistry, and physics.

Finding few examples of manufacturing organizations exhibiting innovative managerial accounting practice, accounting researchers have largely remained on the sidelines of the cutting edge, leaving the little literature existing on the subject to operations management. Zimmerman (2001: 422) argues that the current research focuses almost exclusively on describing practice rather than testing theories. In his view, managerial accounting field research has become completely descriptive with accounting researchers “wondering the hallways of corporations and manufacturing plants searching for facts unguided by tentative hypotheses.”

Kaplan (1993) predicts that new research in high-payoff managerial accounting topics would likely have to be done in the field with innovative organizations requiring a very different set of research methods than have been used in the past, because of the loss of credibility of managerial accounting researchers amongst practitioners. “Traditional empirical analyses in managerial accounting research is not useless, it’s just difficult to do without credibility.”

Referring back to the research objectives of this study, it is desired not only to better understand the impact of existing managerial accounting practice on manufacturing performance, but also to develop an innovative alternative to traditional cost accounting systems that links time-based manufacturing strategy to product cost. To this end, it becomes necessary to go beyond the positivist tradition of managerial accounting research and its established empirical methodologies and instead take a normative approach employing quasi-experimental methodologies as an active participant. In a substantial departure from the passive, observational role for researchers that has become sacrosanct in the social sciences, and in managerial accounting research in particular

since the 1950s, scholars must become active participants in the change process (Argyris et al. 1985, Argyris 1993, Kaplan 1993).

Contributions of this Research

1) *This study integrates theory and methodologies from industrial engineering, operations management, and managerial accounting within a single research model and directly compares the performance of different MAS alternatives in a controlled environment.* Managerial accounting has yet to produce an integrated set of theories and empirical results as the field is in a somewhat early stage in its evolution (Zimmerman 2001, Ittner & Larcker 2001). As evident in the review of the managerial accounting and POM literature, managerial accounting systems, product complexity, scope of product mix, and manufacturing overhead levels all play a major role in the manufacturing performance of a firm in a time-based competitive environment, yet little research to date has considered all these variable in a single study. Managerial accounting alternatives are seldom considered as a factor in operations management literature, despite the fact that management accountants have received a great deal of criticism from operations managers. In response, managerial accounting researchers have proposed novel alternatives to traditional costing systems. Each alternative has been examined against traditional accounting systems and claims that it will perform better; however, seldom has the manufacturing environment been considered. In the research that has included the manufacturing environment as a factor, it has been difficult to determine whether improvement is due to the new managerial accounting systems or to the new manufacturing system, e.g. Drugdale and Jones (1996) compared traditional costing

systems in an MRP manufacturing environment to throughput accounting in a theory-of-constraints environment.

Managerial accounting has a long tradition of utilizing all types of research methodologies including archival, field studies, case studies, survey, experimental, and even simulation to a limited extent; however, there remains a strong need for the development of interdisciplinary work employing methods such as simulation. Controlled experimentation, such as simulation studies, offers a means for filling in many of the research gaps and for supporting existing theories drawn from theoretical inference, field observations, case studies, and surveys of different companies. Unfortunately, direct comparisons of MAS alternatives are often made across entirely different industries with unique manufacturing settings. Often times it is even difficult to truly determine the extent to which alternatives to TCS have been implemented.

2) This study considers the interaction effects of the various MAS alternatives with factors from operations management. The existing research in both managerial accounting and in production operations management often ignores the interrelationships among important factors. For example, most managerial accounting researchers are in agreement that activity-based costing provides more accurate product cost information than TCS, and most managerial accounting research makes the assumption that more accurate product costs will improve the quality of product mix decision thereby improving firm performance (Brimson 1991). However, this assumption is made without examining important factors such as product complexity, scope of product mix, and manufacturing overhead levels, which in the real world application are as important as managerial accounting alternatives on product costing and product mix decisions.

Cooper and Kaplan (1992) remind us that the measure of an improved managerial accounting system is increased profits, not more accurate product costs.

3) *This study goes a long way towards bridging this gap that has long existed between managerial accounting and operations management and between academic research and practice.* A number of surveys have indicated that potentially up to 90% of all manufacturing companies are considering changes to their internal managerial accounting systems, yet fewer than 20% have actually done so. One possible reason is the lack of knowledge about the various managerial accounting alternatives and their potential impact within various manufacturing environmental settings (Walley et al. 1994). Through evaluating these MAS alternatives with environmental factors that were previously studied separately within a controlled simulation environment, results should create common ground for communication between management accountants and operations managers.

4) *This study goes beyond the more conventional deterministic managerial accounting research with the incorporation of demand and supply stochasticity within the simulation model.* The existing managerial accounting literature often draws conclusions about the performance of managerial accounting alternatives assuming a deterministic manufacturing environment setting such as fixed processing times, no demand uncertainty, and no forecasting error (Goldratt 1986, Bakker and Hellberg 1991, Low 1992). Even simplistic models of pull production systems become extremely difficult to analyze mathematically as the assumptions needed for a closed-loop solution often require deterministic assumptions (Leitch 2001). The production operations management

literature has recognized that most manufacturing operations face considerable uncertainty and should be considered stochastic due to:

1. uncertainty in the timing customer orders,
2. variability in the processing time, rework, and scrap rate,
3. inaccuracy of demand forecasting, and,
4. uncertainty of equipment failure.

For example, Wemmerlov (1979) investigated the effects of forecast errors on inventory and found that it leads to an increased number of stock-outs, declining service levels, increased inventories, and increased ordering activities. He concluded that the manufacturing environments where demand uncertainty is present are fundamentally different from those where there is no uncertainty. He recommended that future experimentation should consider demand uncertainty since it is a more realistic representation of real life settings.

The production operations management literature has accepted stochasticity in most of its research design and has explored how different manufacturing systems cope with and react to stochasticity. However, despite the fact that stochasticity involved in manufacturing operations may cause product cost and performance measurement fluctuations, it has not been generally included in managerial accounting studies. The fact that the robustness of managerial accounting alternatives to uncertainty is largely emergent remains presents a major opportunity for contribution to the literature.

Simulation modeling allows for a direct comparison of alternative managerial accounting schemes under different levels of product complexity and scope and different levels of manufacturing overhead. Simulation modeling makes these complex comparisons in a controlled quasi-experimental setting possible and allows for the

collection and measurement of quantitative data. This is something that empirical methodologies, such as cross sectional surveys and case studies, have not been able to do. Moreover, simulation allows for examination of these factors over an extended, albeit compressed, time period; not just as a snapshot in time. The primary purpose of this study is to examine the four managerial accounting allocation schemes listed above under the exact same environmental settings so that a direct comparison can be made.

Incorporation of supply and demand complexity into the model through stochasticity helps to create a dynamic environment in which to test the robustness of various managerial accounting allocation schemes and provides useful information about its behavior. This will help to bridge the gap between managerial accounting theory often created in academic isolation and real-world practitioners who want to know how these accounting schemes will hold up in practice.

5) This study goes beyond conventional positivist managerial accounting research by taking an active role in developing an innovative alternative to traditional accounting practice, i.e. the proposed time-based accounting system. As was pointed out by Zimmerman (2001), managerial accounting researchers may have lost some credibility in the field because of a failure to positively impact managerial accounting practice. This study addresses the deficiency of simply “describing management consulting practice” as suggested by Zimmerman (2001) and Ittner & Larcker (2001), and may provide helpful and quantifiable information to help struggling manufacturing firms and industries regain their competitive advantage through a better understanding of productivity and profitability. The results of this study will provide useful insights for both the fields of managerial accounting and production operations management research, making

significant contributions both to theory and practice. The introduction of throughput-time accounting demonstrates that, when it comes to managerial accounting schemes in an ever more diverse competitive field of manufacturing, the one-size-fits-all mentality will no longer suffice. Managerial accounting systems are an important consideration with regards to the design and implementation of any manufacturing strategy, if it is to be leveraged to the maximum competitive advantage.

Managerial accounting systems alone may not lead directly to the failure of advanced manufacturing strategies nor will assure its success. However, managerial accounting systems should be viewed as an integral part of the manufacturer's response today's competitive environment (Kaplan 1991). The results of this study will help managers to identify appropriate managerial accounting alternatives, specifically in conjunction with a time-based manufacturing strategy, to evaluate the actual profitability of products and to provide the right motivation for production given product complexities, scope of product variety, and manufacturing overhead levels.

The remaining chapters describe the specifics of this study. Chapter 2 begins with a historiography of the development, evolution, and the more recent stagnation of managerial accounting practice. It later reviews relevant literature on the evolution of time-based competition and managerial accounting alternatives and their potential impact on operations management. In addition, this chapter reviews the literary support from both managerial accounting and operations management on the impact of increasing manufacturing overhead levels and product mix complexity on production mix decisions and ultimately on manufacturing performance. Chapter 3 further develops and operationalizes the experimental factors within the research framework and presents the

statistical hypotheses and methodologies for testing. Chapter 4 addresses the assumptions that were made in the simulation model as well as the issues relating to model verification and validation. Chapter 5 reviews and statistically analyzes the results of the simulation study and the performance of the MAS alternatives. Chapter 6 provides conclusions, applications to practice, limitations of this study, and suggestions for further research at this fertile interface of managerial accounting and operations management.

Chapter Two

Literature Review

The Evolution of Managerial Accounting

The practice and theory of accounting has been in existence for over 6,000 years, even if only in rudimentary form until the late 15th century. A Franciscan Friar, Luca Bartolomeo de Pacioli is considered the father of modern double-entry accounting, known at the time as the Venetian method. The only date during Pacioli's life that is known with absolute certainty is 1494, when nearly fifty years of age, he published the first known work on accounting theory. His famous book was titled *Summa de Arithmetica, Geometria, Proportioni et Proportionalita* – The Collected Knowledge of Arithmetic, Geometry, Proportion and Proportionality. Pacioli wrote the *Summa* in an attempt to redress the poor state of mathematics education at the time, but it is one rather small and obscure section of the book that would make Pacioli famous. This section, a treatise on accounting and reconciliations, was titled *Particularis de Computis et Scripturis* – The Particulars of Reckonings and Writings (Bishop 1995).

His system, a compilation of accounting practices employed by Venetian merchants at the time, included most of the accounting cycle as we know it today. For example, he described the use of journals and ledgers, and he even warned that an accountant should not go to sleep at night until the debits equaled the credits! His ledger included assets – including receivables and inventories, liabilities, capital, income, and

expense accounts. He demonstrated year-end closing entries and proposed that a trial balance be used to prove a balanced ledger. Also, his treatise discussed a wide range of related topics from ethics to cost accounting (Bishop 1995).

This new system of financial accounting was state-of-the-art, and literally revolutionized business and economics of the day. The *Summa* made Pacioli an immediate celebrity throughout the Western World and insured him a place in history as the 'Father of Double-Entry Accounting'. The *Summa* was the most widely read mathematical work of its day in all of Italy, and became one of the first books published on the Gutenberg press.

The *Summa* remained the only published work on accounting and mercantile mathematics for almost a century, providing a solid financial foundation for a period of rapid globalization and economic expansion. The Venetian merchant marine would trade goods as far as the seas would take them, making Venice one of the richest principalities in the world. With this new system of accounting, merchants and venture capitalists could compute profitability on their various ventures upon their completion with little disagreement as to the disposition of cargo, profits, and the ship itself at the end of the voyage. From these early mercantilist beginnings, the generation of standard firm-level accounting information became the practice for most all venture enterprises to this day.

It would be nearly four centuries before cost/managerial accounting would begin to develop as a separate field of study. There exist three major paradigms of contemporary historiography in regards to the origins and evolution of managerial accounting as practice and theory; the economic rationalist, Marxian, and Foucauldian schools (Fleischman 2000). However, most all industrial historians recognize Garcke

and Fells (1887) as the first published accounting text to integrate cost accounts into the double-entry system (Littleton 1933, MacDonald 2002). It became widely accepted that cost/managerial accounting developed as a practice in the latter part of the 19th century to provide information to managers in large capital-intensive firms (Littleton 1933, Johnson 1981). The escalating investments in fixed assets created complex control problems that relatively simplistic 18th century accounting systems could not solve (Garner 1954).

The well-known industrial historian, Alfred Chandler (1966), widened the traditional approach of accounting historiography, with its exclusive reliance on ‘hard, objective’ evidence such as archival artifacts, by considering the development of managerial accounting within its broader historical context. American businesses prior to 1840 were well served by the double-entry mercantile bookkeeping procedures that were introduced a half century earlier. By 1850, however, the technology of production was changing as companies moved from the craft age and adopted factory production systems with increased throughput and fewer workers, resulting in significant productivity gains. Chandler (1977) notes that by the 1880s the focus of managerial accounting systems was prime costs, without any known records of depreciation or overhead costs in most process industries. Johnson (1981) found historical artifacts suggesting that cost accounting may have developed even earlier in the 19th century in certain industries. In his review of records from Lyman Mills and the Boston Manufacturing Company dating back to 1856, he found that management had already developed regular production cost reports to augment the general factory ledger.

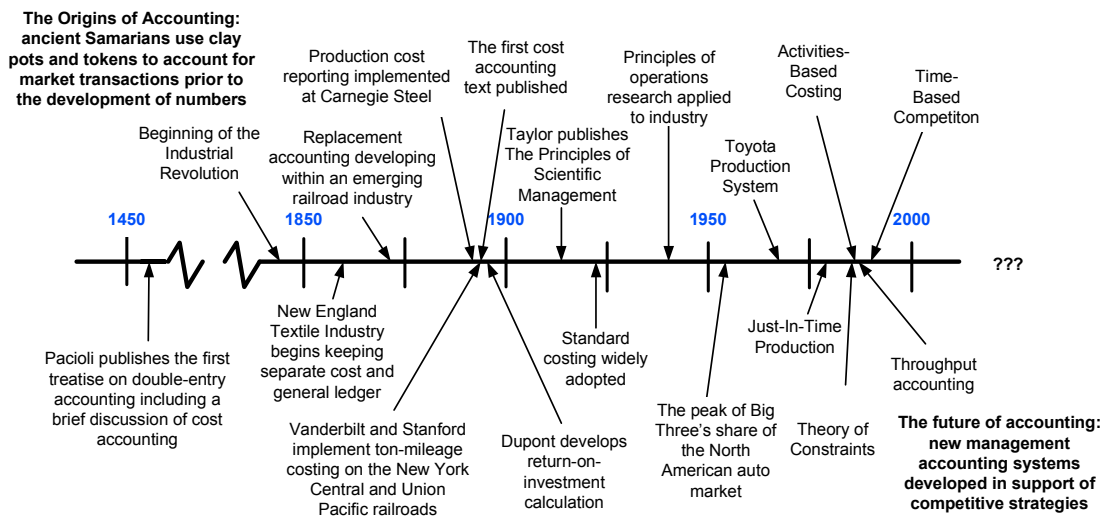
Applying Chandler’s broad historical context approach, Johnson (1981) contends that cost accounting likely developed in conjunction with the development of textile mills

around the very turn of the 19th century. Despite the lack of hard evidence to substantiate this theory, it seems quite probable given the dramatic paradigm shift of the textile industry at that time. Prior to the development of multi-purpose and mechanized mills, the textile industry operated under a ‘domestic’ or ‘putting-out’ business model in which merchant-entrepreneurs provided the raw materials to home-based artisans who received a market-determined piece rate for their production. Accounting records to this time were only utilized to track exchanges and inventories.

Kaplan (1984) attributes the development of modern managerial accounting techniques to the growth of large transportation, production, and distribution enterprises during the period between 1850 and 1925. These enterprises encountered new information needs as they implemented new methods of industrial organization and production. Initially the railroads would provide the greatest source of innovation in managerial accounting as the early tycoons, such as Commodore Cornelius Vanderbilt of the New York Central Railroad and Amasa Leland Stanford of the Central Pacific Railroad, consolidated their power and would seek to gain efficiency in large scale, capital intensive, and geographically dispersed operations (MacDonald 2002).

Sometime between 1870 and 1900, the advent of scientific management ushered in modern manufacturing accounting complete with standard costs, allocation of factory overheads to products, and unit product costing. By 1910 all modern manufacturing accounting was in place with the exception of depreciation accounting, as replacement cost was still being used for book value of plant and equipment in most industries (Chandler 1977). Figure 2.1 below presents a rough time-line of major developments in managerial accounting and industry.

Figure 2.1: A Time-Line of Managerial Accounting and Industrial Developments



Relevance Lost

Traditional managerial accounting practice did not lose relevancy overnight, but progressively over the past century. Many industrial historians claim that managerial accounting practice did not develop as a result of industrialization; rather it was the development of managerial accounting practice that enabled the rapid expansion of industrial production. The relevance of managerial accounting practice was not lost on the industrialists near the turn of the 20th Century, who would leverage the innovative managerial accounting practices of the day to extend their control over vast empires. Many of those methods, such as DuPont's 'return-on-investment', or Carnegie's 'job-costing', and Vanderbilt's 'ton-mileage' calculation, are still widely practiced used control mechanism in industry.

Kaplan (1984) concluded that virtually all of the accounting practices employed by firms today had been fully developed by 1925, and, despite considerable changes in

the nature of organizations and the dimensions of competition during the past 60 years, there has been little innovation in the design and implementation of cost accounting. After a stream of several sole-authored papers, Johnson (1972, 1981, 1983) and Kaplan (1983, 1984) collaborated to provide a more complete picture of the development and stagnation of managerial accounting practice, as summarized in their controversial 1987 book, *Relevance Lost: The Rise and Fall of Managerial accounting*. This book brought into focus a stream of literature begun on the ever widening dissonance between managerial accounting practice and actual operational practices on the production floor. This historiography was yet another call for managerial accounting research to focus its efforts on the increasingly important interface with operations. The authors even suggested that the accounting process has become so detached from the operations of the business that often the accountant can work despite being completely ignorant of the nature of core business processes. The *Relevance Lost* historiography of accounting remains contentious to this day and keeps open the call for an interdisciplinary approach to accounting theory (MacDonald 2002).

The world economy would change for ever on Black Thursday, October 24, 1929, when the world financial markets collapsed. In response to the stock market crash of 1929 and the Great Depression that followed, the accounting establishment in the United States shifted from concern for cost management and operational efficiency to financial accounting and external reporting. Per Zimmerman (2003), supporting the financial accounting and tax accounting reporting functions is one of the primary functions of a managerial accounting system; however, far too often the other functions become subservient to the public reporting needs. Many operations managers have been forced to

base decisions on external financial reporting requirements rather than internal operational performance criteria such as throughput time, inventory turnover, number of defects, or customer service (Fry et al. 1998).

Accountants became overly concerned with the valuation of inventory and not the management of cost (Fry et al. 1993). This concern led to the development of standard absorption costing, as is prevalent today in most industry. Managers began comparing actual results to budgeted figures to assess manufacturing performance, leading to the calculation and development of budget variances. Management by exception became the standard practice, which made sense during this period of mass production with long production runs of standardized products. However, according to Womack and others (1990) the heyday of mass production in America was 1955, when 95 percent of the market of America's largest industry – the automobile industry – was controlled by the Big Three: General Motors, Ford, and Chrysler. The authors point out that 1955 was the year that mass production business model began a steady downhill slide that continues today in the face of advanced manufacturing strategies coupled with flexible automation.

Despite radical changes in modern production strategies and technologies, legacy accounting systems, i.e. TCS, remain firmly in place. Under absorption costing, determining product cost involves adding the costs of direct materials, direct labor, and some percentage of factory overheads. The addition of allocating factory overheads came to be standard practice as United States Generally Accepted Accounting Principles (US GAAP) required that these overheads be allocated to products in inventory valuation for public financial and tax accounting reporting.

Whatever the MAS alternative, US GAAP requires:

In keeping with the principle that accounting is primarily based on cost, there is a presumption that inventories should be stated at cost... It should also be recognized that the exclusion of all overheads from inventory does not constitute an accepted accounting procedure. The exercise of judgment in an individual situation involves a consideration of the adequacy of the procedures of the cost accounting system in use, the soundness of the principles thereof, and their consistent application. (Accounting Research Bulletin No. 43)

Because US GAAP requires some form of absorption costing for inventory valuation – both for public financial and tax reporting – and most companies see maintaining two separate systems as cost prohibitive, variable costing is very seldom used in practice. Most firms began to use the same method of full-absorption costing for both internal management decision making and external reporting.

The most common method of absorption utilizes direct some measure of production volume as a basis for MOH allocation. A study by Price Waterhouse (1989) found that roughly 80% of their American manufacturing clients surveyed continue the use of standard absorption costing with direct labor content as the primary allocation method of overhead costs. Drury and Tayles (1997) conducted a survey with the purpose of investigating the claims of Johnson and Kaplan (1987) that managerial accounting has become subservient to financial accounting. They found 73% of the non-automated and 68% of the automated manufacturers used labor hours as a base for allocations, even though other systems, with claims of better accuracy exist. Only 9% of the surveyed companies had plans or were in the process of implementing some form of ABC, with only 4% having an operational ABC system. The authors concluded that simplistic methods, designed primarily for meeting financial accounting requirements, are being widely used for decision making. A similar survey by Fry and others (1998) of 110

manufacturing companies supported this figure with 75% using TCS with direct labor hours as an allocation basis. An even more recent survey of 131 manufacturing firms shows that little has changed with 88% of firms still relying on TCS at least to some degree (Hughes & Paulson-Gjerde 2003).

Supporting the assessment of the chasm between managerial accounting and operational decision making are numerous articles finding that nothing is wrong with the design of cost systems. The authors of these studies argue that the system is simply being misused in that it was never designed for internal reporting, but rather for external reporting documents to be used in valuing inventory and calculating profits and losses (Baker 1989, Edwards 1985, 1984). These authors suggest that operations managers should not use variance reports against standards as the primary tool for control, but rather something more indicative of the actual manufacturing task, e.g. throughput-time for a time-based manufacturer.

The Dawn of Time-Based Competition

“The way I sell investment in innovation is as a time reduction, not a cost reduction” (Schafrik 2005). This proclamation by the General Manager of GE Aircraft Engines at the International Workshop on Accelerated Radical Innovation underscores both the increasing importance of time and the decreasing relevancy of traditional cost measures in modern manufacturing. Indeed, many managers competing in time-based industries realize that advance manufacturing strategies are often undermined by traditional cost systems developed for another era.

Time is fast becoming the next competitive battleground for American industry (Blackburn 1991). Those manufacturing firms that survive face a major shift in paradigm from mass production systems driven solely by cost efficiency to advanced manufacturing systems driven by quick response to customer demands for a much greater variety of high-quality products (Doll & Vonderembse 1991). A distinctive characteristic of globalized markets has been an increase in customer requirements for ever faster delivery of high quality products without corresponding increases in price (Drucker 1990). This heightened competition has forced most companies to adopt some form of advanced manufacturing strategy in order to remain a viable competitor in the long-term.

The evolution of time-based competition follows a continually evolving global manufacturing environment, where the order winners quickly become order qualifiers. Hayes and others (2005:19) describe this dynamic and evolving basis for competition in the new millennium. According to the authors, “American manufacturers have struggled to keep up with the global competition since the age of mass production, as the basis of competition has shifted from cost, to quality, to variety, and now to speed. It has come to the point where, as one CEO so succinctly stated, ‘time accounting is more important [for us] than cost accounting’”.

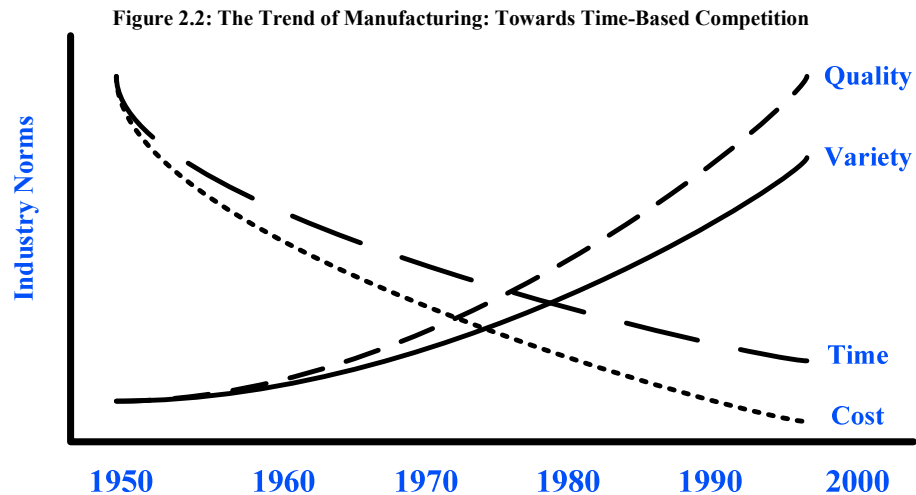
The first major shift in the basis of competition from mass production and its cost efficiencies came with the advent of Total Quality Management (TQM). TQM evolved first as a philosophy in manufacturing that stresses improved quality through continuous process improvement, employee empowerment, and data-based decision making. Closely related to TQM, and evolving shortly thereafter, is the concept of Just-In-Time (JIT) production, which focuses on lot size reductions, minimizing inventory levels,

increasing flexibility, and elimination of all waste in manufacturing and distribution. The relationship between TQM and JIT is very strong, and, although their infrastructural practices have been proven beneficial in isolation, there is an evident synergistic effect in implementing both programs together. TQM delivers the quality levels necessary to support JIT measures for reducing lot sizes and inventory levels, and reduced inventory levels reciprocally expose opportunities for process improvements (Flynn et al. 1995).

The TQM/JIT movement in manufacturing has been described as the genesis of TBM (Blackburn 1991). The first time-based competitors were innovators that applied the TQM/JIT tools and concepts beyond the factory walls and across the entire value chain (Stalk 1988). TBM promises increased customer satisfaction through faster service times and higher quality levels. Time-based competitors found that speed and quality not only allow for premium pricing of products in the marketplace, but it also often leads to reduced internal costs over the long-run and greater profitability.

Figure 2.2 below illustrates long-term trends in manufacturing, adapted from Blackburn (1991: 295-96). Graphing the 1950s, 60s, 70s, 80s, 90s, 2000 and beyond on the x-axis and plotting lines roughly indicating how industry norms have changed from decade to decade during that period presents a revealing picture of the evolution towards time-based competition almost universally across all industries. According to the author:

In most industries – industries as diverse as eyeglasses, package delivery, wholesale distributions, consumer electronics, motorcycles, and automobiles – the patterns are identical. The graph tends to show quickening market response time, improving standards of quality, and increasing product variety... These trends suggest strongly that time-based competition, like quality, is not a fad... Unless a fundamental change occurs in consumer behavior, competition will become increasingly time-based.



Hout & Stalk (1993) present case studies reaffirming Blackburn's hypothesis that across almost every industry time has become the basis for competitive advantage. According to the authors, "time-based competition is a reality, not just a concept. It is rapidly becoming the baseline, not the exception."

Time-based competition focuses on reducing response time to customer demand by squeezing time from every facet of the value-delivery system from research and development, to product development, to manufacturing, to marketing and delivery. TBM is a natural evolution of the TQM/JIT philosophy, albeit with a slightly broader scope of application. Suffice to say that time-based manufacturing is an externally focused production system that emphasizes quick response to changing customer needs and to this end seeks to reduce the end-to-end (throughput) time in manufacturing (Stalk and Hout 1990, Blackburn 1991). According to Blackburn (1991), a clear distinction must be made between TBM and JIT strategies. Time reduction, not inventory reduction, is the real underlying driving force behind the TQM/JIT revolution and naturally leads a firm to become a time-based competitor:

Inventory reduction is a small part of the story; a benefit of JIT, but not the *raison d'être*... Cycle-time compression translates in to faster asset turnover, increased output and flexibility, and satisfied customers. Thus, the diminished inventory often associated with JIT is more of a side benefit than a driving force.

Similarities between JIT and TBM sometimes make it difficult to distinguish the two, and it may be debatable as to whether inventory and cycle-time reductions represent two sides of the same coin. Figure 2.3 below summarizes some of the key conceptual differences between these two strategies described in the literature:

Figure 2.3: A Comparison of the Key Distinctions Between JIT and TBM

	JIT	TBM
	Internal (Operations)	External (Customer)
Focus	Inventory, batch size reduction, and elimination of waste	Time compression and product mix
Scope	Narrowly applied to manufacturing	Broadly applied to value chain
Key Metric	Inventory cost	Throughput time

There are many case studies where American manufacturing firms have redesigned their processes to compress time and improve performance (Stalk 1988, Stalk & Hout 1990, Blackburn 1991). These firms achieved higher productivity, increased market share, charged premium prices, reduced risk, and improved customer service. Substantial and sustainable competitive advantage is the prize when firms achieve speed in all facets of the value-deliver system, and all aspects of the business from accounting to engineering to manufacturing to marketing should be aligned towards this strategic goal.

Despite significant conceptual contributions and numerous case study examples, it would be nearly a decade before empirical studies would identify the key practices of

time-based manufacturers. Koufteros and others (1998) identified the following specific manufacturing practices that focus on time compression:

- (1) *Shop-floor employee involvement in problem solving*: first level employees participate in activities to define and solve problems.
- (2) *Reengineering set up*: efforts are taken to reduce setup time.
- (3) *Cellular manufacturing*: units are produced in a product oriented layout
- (4) *Quality improvement efforts*: methods are developed and used to reduce defects and enhance quality.
- (5) *Preventive maintenance*: equipment is routinely maintained on a proactive basis.
- (6) *Dependable suppliers*: suppliers facilitate customer needs for service quality.
- (7) *Pull production*: production is driven by demand from the next work station and ultimately from the customer.

The results of this study found a statistically significant relationship between the composite measure of these time-based manufacturing practices with reduced throughput-time. Throughput time is the ultimate measure of performance for the time-based competitor, and that is determined by how well an organization is able to move materials to, through, and off the shop floor into finished goods inventory and how well that organization is able to minimize finished goods (Fry et al. 1993). Time-based manufacturing requires an emphasis on time measures, such as throughput-time, inventory turnover, on-time shipments, new product introduction time, and quoted delivery times. Managerial accounting systems that fail to emphasize throughput-time may act as a fetter to sustained TBM practices.

Stalk (1989) points out that although time is a more critical competitive yardstick than traditional financial measures, management seldom monitors its consumption explicitly – almost never with the same precision accorded to sales and cost. Older, cost-based strategies require managers to do whatever is necessary to drive down costs: move production to, or source from, a low-wage country; build new facilities or consolidate old

plants to gain economies of scale; or focus operations down to the most economic subset of activities, all at the expense of responsiveness.

Blackburn (1991) postulates that time-based competition will eventually come to all industries in due time, even those industries that seem to compete on an entirely different basis. It appears that certain industries are immune to or simply resisting time-based competition, such as plastic injection mold producers. The typical mold builder is a small, privately-held machine shop with fewer than 20 employees. Many of these small shops appear to compete for small jobs based mainly on price and the reliability of delivery lead-time, albeit rather long. However; further study in the industry revealed that it would not remain immune to time-based competition for long. Zahorik and others (1989) presented evidence that end users of the molds are increasingly driven by time and willing to pay a premium. According to one manufacturing manager from an IBM plant that is a major consumer of such plastic injection molds, "Time is everything. It's well worth it to pay double to cut production time in half".

Stalk and Blackburn bring to light an interesting paradox for TBM that has largely been ignored by research, as many managers see throughput-time and product cost as an unavoidable tradeoff. This is especially true with regards to traditional managerial accounting systems that often come into conflict with time-based performance measures. While the Koufteros (1998) study provides a firm foundation for research in time-based manufacturing, it focused solely on the plant floor practices of a time-based manufacturer. While it does conclude with suggestions for future research in time-based practices in the areas of product development and marketing, it completely ignores the

question as to whether a particular MAS alternative would also be associated with reduced cycle-times.

The Impact Increasing Manufacturing Overhead Burdens

Adoption of advanced manufacturing strategies, such as TBM, have often been made in conjunction with increased investments in flexible automation, bringing fundamental and lasting change to the modern factory floor. For those successful manufacturing firms, implementation of these advanced manufacturing systems have brought positive changes including lower inventories, reduced production lead times, shortened product lifecycles, and an emphasis on quality. Probably most obvious of these changes has been the movement of the human production interface from the plant floor to the back office. These trends in modern manufacturing result in corresponding changes in manufacturing cost characteristics, including dramatically reduced direct labor costs with correspondingly greater levels of manufacturing overhead (MOH) burden.

The higher levels of capital investment inherent in flexible technologies drive the need for more accurate tracing of costs to individual work centers and products along with more detailed analyses in both capital authorization and recovery decisions. The failure to recognize the impact of these manufacturing trends on managerial accounting information has resulted in inaccurate product costing/pricing, inadequate operations/cost control, poor inventory management, unnecessary capital investments, and poor strategic decision making. An overemphasis on traditional cost-based performance measurement systems, encourage resource utilizations rather than other time-based competitive practices (Fry et al. 1993). The older cost-based strategies require total management

concentration on minimization of costs in an attempt to maximize profits, often in exclusion of other more meaningful performance criteria (Kaplan 1984).

While many managerial accounting researchers have logically inferred the impact of increasing MOH burdens, given outmoded overhead allocation practices left from the mass production era, few have ventured to quantify the level of manufacturing overheads at which TCS begins to grossly misallocate costs. Vokurka and Lummus (2001) performed a simple scenario analysis to determine at what level of manufacturing overhead burden level does the implementation of an ABC system make a significant difference compared to TCS. In this study they compare four fictitious companies producing the same five products all with differing levels of MOH burden, ranging from 6.2% to 40% of total product cost. Their findings were that in general the higher the overhead rate, the greater the difference between traditional and activity-based costing approaches. Given the rather high cost of implementing an ABC system, they conclude that any company with an overhead burden of less than 15% should probably not consider the effort.

A Taxonomy of Managerial Accounting Systems

A firm's MAS is called upon both in strategic and tactical planning, as well as providing support to financial and tax reporting systems. A major concern of management, therefore, is to select the appropriate MAS alternative that would provide accurate product cost information, derive the most profitable production mix decisions, minimize throughput, and maximize capacity utilization in a given manufacturing context.

Most MAS alternatives fall under two broad categories, based on method of inventory valuation; they are absorption and variable costing. The distinction between absorption and variable costing is based on the treatment of overheads. Under absorption costing, fixed overhead is assigned to units of inventory and shows up in the income statement as part of the cost of goods sold when the units are sold. When the units are produced, but not yet sold, allocated MOH stays in finished goods inventory. Under variable costing, no MOH is allocated to inventory. Fixed overhead is treated as a period expense which enters the income statement on a line-item every period regardless of the number of units sold.

Absorption costing makes a primary classification of costs according to manufacturing and non-manufacturing functions, emphasizing the gross margin – sales less cost of goods sold (CGS) – available to cover all fixed and variable selling, general, and administrative expenses (SG&A). Variable costing makes a primary classification of costs into variable and fixed categories, emphasizing the contribution margin (sales less variable costs) available to cover all fixed costs. Figure 2.4 demonstrates how the formats for profit reporting under absorption and variable costing differ:

Figure 2.4: A Comparison of Profit Reporting under Absorption and Variable Costing

<u>Absorption Costing</u>	<u>Variable Costing</u>
Revenues	Revenues
Less: Cost of Goods Sold	Less: Variable Manufacturing

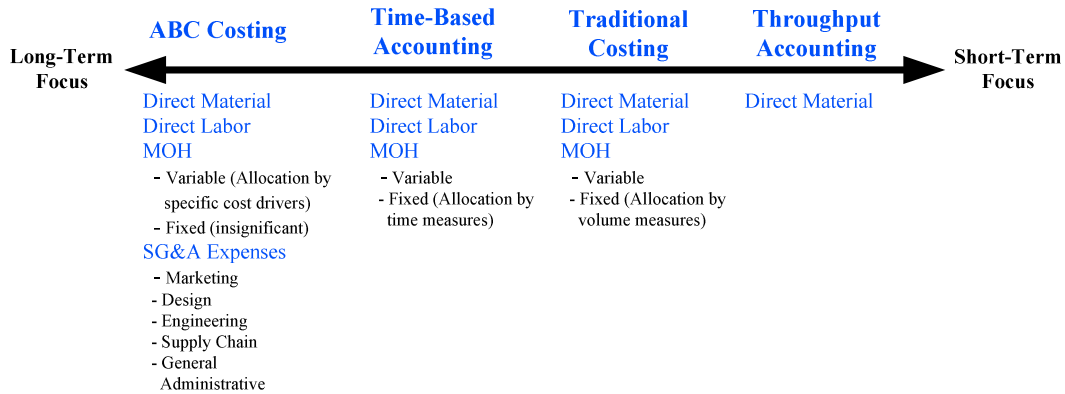
Gross Margin	Less: Variable S&A

Less: Variable S&A	Contribution Margin
Less: Fixed S&A	
-----	Less: Fixed Manufacturing
Profit	Less: Fixed S&A
=====	-----
	Profit
	=====

The difference between the two income measurement approaches is essentially the difference in the timing of the charge to expense for fixed factory overhead costs. In the absorption costing method, fixed factory overhead is first charged to finished goods inventory; thus, it is not charged to expense until the period in which the inventory is sold and included in cost of goods sold. In contrast, under the variable costing method, fixed factory overhead is expensed immediately, and only variable manufacturing costs are included in finished goods inventory. Therefore, if inventories increase during a period, i.e. production exceeds sales, the variable costing method will generally report less operating income than will the absorption costing method. When inventories decrease, the opposite effect occurs with variable costing reporting great operating income than absorption costing.

The four specific managerial accounting systems considered in this research include traditional cost systems (TCS), activity-based costing (ABC), throughput accounting (TA), and the proposed time-based accounting (TBA). A comparison of product costing under these four alternative managerial accounting schemes is presented in figure 2.5. As illustrated, the four accounting alternatives inherently have a different focus in terms of decision making, with the major difference being whether and how overhead costs are allocated to products. Managerial accounting researchers tend to criticize variable costing methods such as throughput accounting for being short-term in their focus, since no attempt is made to allocate manufacturing overheads to product cost.

Figure 2.5: Product Costing Under Different Managerial accounting Schemes



Traditional Costing Systems

Using a TCS, management assigns the rates at which products will ‘absorb’ direct material, direct labor, and MOH. Unlike direct materials and direct labor, MOH is not broken down by actual activity drivers, and is instead assigned through a simple calculation of dividing the total MOH burden by some known allocation basis.

Typically, the allocation basis is some quantifiable measure of volume, such as direct labor hours or direct machine hours.

Though widely used today by roughly three quarters of manufacturing firms, a TCS is most appropriate in specific manufacturing environments (Price Waterhouse 1989, Fry et al. 1998). In particular, standardized processes, similar and limited product lines, high direct cost, and a mature stage in the product cycle all lend themselves to the use of a TCS (Kaplan 1983, Baker 1989, Fry et al. 1998). At the zenith of the industrial age, when labor and materials represented 70-80 % of total production costs, the use of direct labor to allocate overhead seemed logical. Regardless of the apparent logic, if MOH burden was a negligible portion of total production costs, misapplication would not be a concern (Ruhl & Bailey 1994).

This stands in stark contrast to the manufacturing paradigm of today, where overheads, driven by large investments in advanced manufacturing technologies, may represent the single largest percentage of manufacturing costs. Labor, conversely, often represents the smallest share of total costs in most modern manufacturing plants at 10 % or less (Seed 1984, Turk 1990). The allocation of MOH by means of direct labor content or labor dollars has led managers to focus strict attention on direct labor variances, not to control direct labor costs but rather to control the allocation of overhead costs to each product (Fry et al. 1993). Fundamental changes in production also mean that the occurrence of overheads is no longer proportional to production volume. Traditional cost accounting practices tend to influence overproduction, as managers feel the need to maximize standard labor hours in order to spread out MOH (Womack & Jones 2003).

Activities-Based Costing

The increased magnitude of MOH burden and the corresponding decrease of direct labor component in advanced manufacturing facilities have led practitioners and researchers alike to search for a more relevant alternative to TCS to reflect the fundamental changes in manufacturing. The time was ripe for any alternative to established accounting techniques, and activity-based costing (ABC) quickly became a popular focus for consultants and researchers in the early 1990s. ABC was introduced as an alternative to traditional accounting techniques by Cooper and Kaplan (1988) and quickly became a popular focus for both practitioners and researchers alike. ABC proposes that all activities in the organization exist to support the production and distribution of goods and services and that activities consume resources and that products consume activities.

Conceptually, ABC is not radically different from the TCS it is meant to replace. It simply accumulates MOH and SG&A into 'activity pools' and then allocates them to products by specific individual cost drivers as opposed to a single generic and often irrelevant drivers such as direct labor hours or machine hours. ABC attempts to better model the relationships between the final product and all the resources used at all stages in its manufacture, thereby, tying the activity costs to the appropriate activity driver. Many managerial accounting scholars consider ABC superior to TCS because it provides a more accurate and consistent method for calculating manufacturing costs (Andrea et al. 1999). The result of this detailed system is a more accurate cost calculation, which may better highlight the constraints driving costs on the manufacturing floor (Kee & Schmidt 2000).

Although it is conceptually easy to understand and is intuitively logical, actual implementation often proves difficult. Many ABC systems introduced in recent years have been regarded as failures (Cooper et al. 1992, Argyris & Kaplan 1994). In most cases the effort required in identifying and modeling all the individual cost-driving activities within a manufacturing process and back office support functions from beginning to end does not justify the cost. The ability to fully and accurately allocate all overheads, or at least a significant portion, remains a major problem for implementing ABC. According to Johnson (1992), advocates of ABC tell companies that are unable to fully allocate all manufacturing overheads, in effect, to cost products differently for financial reporting purposes than for management purposes, such as planning and decision-support. For financial reporting, the ABC advocates recommend that companies continue allocating overheads using the volume sensitive drivers they have used since the

early years of the past century. The need for maintaining the TCS is that for inventory valuation purposes, US GAAP requires that all overhead costs be fully absorbed into either cost of goods sold or into inventory. The ability to fully allocate all MOH remains the Achilles' heel for ABC advocates.

Kaplan and Anderson (2004) acknowledge that part of the failure of ABC comes from the inability to fully and accurately trace all overhead activities to all products, often leaving large amounts of overhead unallocated. On the other side, they point out an even more sinister demon in the ABC system, that when industrial engineers or management accountants survey individuals in overhead support functions to determine how they spend their time it almost always adds up to 100%. Few people, if any, report any significant amount of their time is spent idly or unproductively. Managers do not typically look favorably on work time that cannot be directly allocated to any particular cost pool, but – as we all know – most operations in the real world run considerably lower than their capacity.

Other factors hypothesized as contributing to the failure of ABC may include the failure to truly leverage all the benefits due to lack of sufficient information system infrastructure and the general resistance or simple lack of appreciation by decision makers (Malmi 1997). Hiromoto (1988) points out another flaw of ABC; that is the failure to leverage the motivational aspects of the MAS. While Japanese companies are well aware of the principle that an allocation system should capture as precisely as possible the reality of shop floor costs, they are more concerned with the motivational potential of the accounting system used in terms of long-term manufacturing strategies than with precise data on costs, variance, and profits.

Johnson (1992) agreed with Hiromoto's assessment of ABC and broke with Kaplan to write a new book entitled *Relevance Regained: From Top-Down Control to Bottom-Up Empowerment*. In this book he was equally critical of ABC as he had been of TCS, describing it as "relevance lost déjà vu". According to Johnson, "the pathway to global competitive excellence is not reached by doing better what we should not be done at all" (1992:149). The point the author makes is that ABC, while certainly potentially more accurate in product costing than TCS, does little to motivate employees towards strategic goals. Johnson follows with two case studies of manufacturing companies, one in the automotive market and one in the electronics market, where ABC information prompted managers to reduce costs and improve short-term profits by altering product mix or process mix, not by altering the way work is performed and the customer is served.

Throughput Accounting

Throughput as the most important measure of manufacturing performance was originally proposed by Eliyahu Goldratt through the concept of "theory of constraints" (TOC) in his 1992 book *The Goal*. TOC has had a significant impact on production scheduling theory and has many similarities with JIT production – with the emphasis of stock reduction and on 'pulling' production only as needed, but TOC advocates stock buffers to protect bottleneck facilities (Drugdale & Jones 1996). TOC also shares many similarities with TBM in that both emphasize the reduction of throughput-time. Goldratt recognized the inherent perversity of TCS that rewarded manufacturing managers for overproduction, while the shop floor and warehouse back up with WIP and aging finished goods

inventory. He argued on many occasions that cost accounting is ‘enemy number one of productivity’, through sub-optimal behavior driven by local efficiency and utilization measures along with the standard accounting practice of ‘recovering’ overhead into stocks as goods are produced – not sold. According to Goldratt & Cox (1992: 91-92):

The goal of a plant is to make money, and the measurements we are seeking should measure progress towards that goal... The result is that our cost accounting measurement have caused either a loss in throughput or an increase in inventory or operating expense – not consistent with the goal of the organization. Measurements should provide incentives for the plant to run more smoothly, but these cost accounting measurements seem to have the opposite effect.

Throughput accounting advocates that only direct materials are included in product cost and all other costs are fixed and not identifiable with products (Goldratt & Cox 1992). The earliest reference to ‘throughput accounting’ can be traced to a series of four articles written by Galloway and Waldron beginning in 1988. Waldron worked for Goldratt’s consulting firm, and contributed to the development of TOC, but would later diverge from Goldratt in attempting to reconcile its principles with more traditional accounting analyses. Throughput accounting (TA) began to appear in some textbooks and has been incorporated into the UK’s Chartered Institute of Management Accountants (CIMA) syllabi. Figure 2.6 below compares TCS with TA:

Figure 2.6: A Comparison of Traditional Cost and Throughput Accounting

Traditional Cost Accounting	Throughput Accounting
There are direct (variable) and indirect (fixed) costs	Distinguishing between direct and indirect costs is impossible and unproductive
Summing component costs to derive product cost and subtracting the result from the sales price used to determine product profitability Inventory is an asset and working on material increases its value	It is the rate at which the factory earns money that determines profitability, the the contribution of each product Inventory is not an asset! It is the result of unsynchronized manufacturing and stands in the way of profits
Reducing component cost directly increases profits	Profit is a function of material cost, total factory cost, and throughput

Throughput, inventory, and operating expense are the three terms used in TOC as well as TA. Because some textbooks may not define these terms exactly as they have been defined by Goldratt & Cox (1992), it is necessary to provide some definitions. “*Throughput (T)* is the rate at which the system generates money through sales”. As interpreted by throughput accounting, it is equivalent to contribution margin, defined as the selling price minus the total variable cost. That is to say that under throughput accounting, product cost is equivalent to variable costs only. From an economic standpoint, the category of variable cost is dependent on the chosen time horizon. As described by Goldratt (2002) in one of his “late night discussions”:

We recognize that at the beginning of the century vast majority of costs were totally variable. You see, I don’t think that differentiating between variable and fixed costs is very useful. At the end, the difference is based on an arbitrary decision. Even the building is not a fixed cost, it can be sold. Look at industry in the past ten years. Overall, what we call variable costs have stayed the same, while during the same time, fixed costs have doubled. So variable is fixed, and fixed is changing.

Under short-term conditions, only material costs, utilities, etc. are variable. Under longer-term horizons all costs are potentially variable (Bakke & Hellberg 1991). Under TA, all costs except material costs are considered fixed and therefore excluded from product costs. This means that only the material costs should be included in the product costs and that labor costs and factory overheads are considered to be part of operating expenses (Miller & Vollmann 1985)

“*Inventory (I)* represents all the money the system invests in purchasing things the system could or intends to sell”. This is the total amount of investment in the system, including things as buildings, equipment, vehicles, and raw material inventory. This does not, however, include added value for labor or factory overheads in the inventory. This view of inventory not as an asset, as is traditional for accounting, rather as a ‘money pit’

is not entirely new. Drucker (1990) had also challenged manufacturing accountants to develop new cost concepts redefining costs and benefits in the manufacturing context.

In the traditional cost accounting system, [inventory] is treated as an asset. In the new manufacturing accounting, however, inventory of finished goods is a ‘sunk cost’ (an economist’s, not an accountant’s term). Stuff that sits in inventory does not earn anything. In fact, it ties down expensive money and absorbs time. As a result, its time costs are high. The new accounting measures these time costs against the benefits of finished-goods inventory (quicker customer service for instance).

“*Operating expense* (OE) is all the money the system spends in turning inventory into throughput”. This is all the money constantly poured into the system to keep it operating, i.e. expenses for labor, supplies, maintenance, depreciation, advertising, etc.

TA is a new tool for managerial accounting that was initially developed to support the TOC. The basic philosophy is that profit is a function of manufacturing response to time and that the long-term objective of management must be to maximize profit. TA has been offered as an alternative to absorption accounting systems such as TCS and ABC systems, for which TOC advocates have little use. The TA measurements provide a method of operationalizing the key concepts of TOC – throughput, inventory, and operating expense – into a functional accounting system (Ruhl 1997). Throughput accounting states:

$$\textit{Net Profit} = \textit{Throughput} - \textit{Operating expenses}$$

When:

$$\textit{Throughput} = \textit{Sales} - \textit{Total variable costs}$$

Then:

$$\textit{Return on investment} = \frac{\textit{Net profit}}{\textit{Inventory}}$$

Proponents of TA laud its simplicity while opponents criticize its complete disregard for fully capturing detailed product costs. Critics point out that TA is short-term focused and does not provide adequate product costing information for long-term decision making. Short-term variable cost may not reflect the true product cost because the decision to make a product creates a long-term commitment to manufacturing, marketing, and post sales support, thereby resulting in an increase in fixed cost (Cooper & Kaplan 1988). TA is not very helpful to strategic decision making, which typically requires the investment of significant capital over a long period of time and is concerned with more than just throughput (Ruhl 1997).

Goldratt and Cox counter these arguments by pointing out that the long-term is nothing more sophisticated than a series of short-terms. According to them, the only relevant element for product cost decisions is variable costs, because all fixed costs have already been incurred, or have been committed to be incurred, and cannot be changed in the short-run. This is the same argument that economists have long made for marginal analysis, i.e. all decisions should be made on the margin.

Despite criticisms from established scholars, the simple lack of novel alternatives to TCS drove some to operationalize TA. Galloway and Waldron (1988a) in their first article on TA claimed that a new language for manufacturing is needed to replace traditional concepts such as direct/indirect costs, economic batch sizes, and the idea of adding value to stock. They later develop a series of articles which make operational the concepts of TOC through development of management tools.

In the second article (1988b) they develop the 'TA ratio', which they suggest can be used to rank individual products.

If:

$$\text{Return per factory hour} = \frac{\text{Sales price} - \text{Material cost}}{\text{Time on key resource}}$$

And:

$$\text{Cost per factory hour} = \frac{\text{Total factory cost}}{\text{Total time on key resource}}$$

Then:

$$\text{TA Ratio} = \frac{\text{Return per factory hour}}{\text{Cost per factory hour}}$$

The third article (1989a) introduces the ‘primary ratio’, where instead of defining profit as throughput less operational expense, they prefer a ratio:

$$\text{Primary ratio} = \frac{\text{Throughput}}{\text{Total factory cost}}$$

Recognizing a need for local measures to drive ‘throughput thinking’ on the plant floor, they develop a set of time-based measures. This is opposed to measures based on departmental sales, which would imply some form of transfer pricing between departments.

$$\text{Departmental throughput} = \text{Standard minutes of throughput} \times \text{Budgeted departmental cost per minute}$$

Waldron (1994) continued to show his commitment to developing departmental measures within the context of TOC.

$$\textit{Departmental ratio} = \frac{\textit{Throughput}}{\textit{Total facility cost}}$$

In their fourth article (1989b), Galloway and Waldron note that in complex manufacturing environments, there are often several manufacturing facilities, and within each facility, the ‘focal’ point needs to be identified.

$$\textit{Cost per focal point minute} = \frac{\textit{Total facility cost}}{\textit{Focal point capacity (minutes)}}$$

In conclusion, product costs can be computed on TA principles:

$$\textit{Product cost} = \frac{\textit{Time required on focal point}}{\textit{focal point}} \times \frac{\textit{Cost per focal point minute}}{\textit{point minute}} + \textit{Material cost}$$

Throughput-based product costs would therefore penalize products that make heavy use of focal points while products not routed through focal points would be cost as materials only, creating a new set of perverse incentives. Waldron (1994) would later agree that TA principles are not conducive to product costing, and he later would propose that ABC is more appropriate for product costing (Drugdale & Jones 1996). He would go on to suggest that companies need both ABC for product costing and TA to tell how many products can be made and how fast. According to Waldron:

Some sectors in the accounting world would want to set TA against ABC... that’s a whole lot of junk, because you need the added information and they’re both adding something... [ABC] doesn’t tell you anything about how the business can make money... It doesn’t tell you how many [products] I can make, or how fast... [but] TA will never tell you the right price to go to the market with for a product. So you need both... TA is not a product costing system.

Given the already low success rate of ABC systems due to the cost and complexity of implementation, adding yet another MAS alternative on top does not present a practical solution. Moreover, careful analysis of the entirety of Waldron and Galloway's ratios reveals that they are in fact analogous to familiar concepts of maximizing contribution per unit of limiting factor. If materials were the only variable cost, contribution and throughput would be identical and there would be no difference between traditional contribution analysis and 'throughput' analysis, therefore there is nothing new in throughput accounting (Willett 1989). Maximizing throughput per bottleneck minute, in the guise of 'contribution per unit of limiting factor', has been standard textbook material for many years, and although he may not succeed in destroying the 'cost world thinking', Goldratt may have succeeded in reinvigorating debate on historical issues such as relevant cost, contribution, and variable cost analysis (Drugdale and Jones 1996).

Furthermore, the debate as to whether to use absorption or variable costing appears to have long been settled by government regulations and the costs of maintaining two separate accounting systems. Because of the US GAAP requirement that some form of absorption costing be used for inventory valuation and tax reporting purposes, TA does not present a viable alternative for most manufacturing firms. Although several studies have included TA or other forms of direct costing, and it is an interesting alternative from an academic standpoint, this study will compare only the three forms of absorption costing that are most likely to be used in practice. To be included in this study are the previously mentioned TCS and ABC system as well as the following proposed TBA

system, which does incorporate the same importance of throughput-time in manufacturing that the TA system does.

Time-Based Accounting

Drucker (1990) predicted that managerial accounting would play an even bigger role than in the past in the American factory of the future, and that time – not labor – would be the critical driver of costs by 1999.

Labor costs are clearly the wrong unit of measure in manufacturing. But – and this is a new insight – so are all the other elements of production. The new measurement unit has to be time. The costs for a given period of time must be assumed to be fixed; there are no ‘variable’ costs. Even material costs are more fixed than variable, since defective output uses as much material as good output does. The only thing that is both variable and controllable is how much time a given process takes. And ‘benefit’ is whatever reduces that time. In one fell swoop, this insight eliminates cost accounting’s traditional limitations.

Unfortunately, managerial accounting practice has failed to advance vis-à-vis Drucker’s prediction. One often cited reason for the lack of innovation in managerial accounting practice is that no ‘Japan’ exists for learning about or studying innovative managerial accounting practices as is the case in operations management (Kaplan 1993). While there may be no example of a single wide-spread managerial accounting development in Japanese industry for cross-sectional study, this is more a reflection of the nature of Japanese managerial accounting practice, which tends to be firm-specific. There are many known case studies of Japanese companies that utilize the allocation of factory overheads to motivate employees towards long-term strategic goals (Hiromoto 1988). For example, a Hitachi factory producing refrigeration and air-conditioning equipment employs an overhead allocation technique based on the number of parts in product

models in order to influence product design decisions towards reducing the number of parts per product.

In large part, Japanese companies have rejected ABC as providing little if any marginal value for management decision making and control.

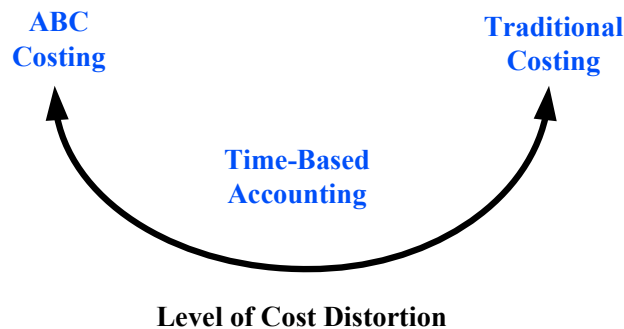
While Japanese management is well aware of the guiding principle that an allocation system should capture as precisely as possible the reality of shop floor costs, they are more concerned with the motivational potential of the MAS in terms of long-term strategy than with absolute precision of cost, variance, and profit data. The allocation basis used commonly in Japan is primarily determined by the activity or resource usage that management wants to minimize. In the case of a time-based competitor, the resource usage to be minimized is time itself – throughput time in manufacturing terms.

One possibility for the time-based manufacturer is to simply change the allocation method to reflect TBM strategy, i.e. time-based accounting (TBA). Rather than volume measures such as direct labor, throughput-time could be used as a basis for allocating overheads. As a time-based manufacturer competes on time, using product throughput time as an allocation basis would encourage managers to constantly reduce time in order to reduce product cost. Since this modification would not violate any aspect of US GAAP or compromise public reporting, TBA could be readily implemented. A simple numerical example of TBA allocations is provided in Appendix B.

TBA accounting would be much easier to implement than ABC, as it does not require radical change in the entire accounting system. However, the real advantage in using product throughput-time to allocate overheads for a time-based manufacturer is that

any investment initiatives are ultimately justified in terms of expected time reductions via product cost. Under TCS justifications on improvements are often made via reduction in labor content that is really *de minimus* in terms of overall product cost (Fry et al. 1993). TBA maintains the strategic decision making strength inherent to absorption costing systems, in that it fully allocates manufacturing overheads to products thereby supporting long-term investment strategy. TBA has the tactical decision making strengths of TA, in that it motivates turning inventory quickly into throughput, i.e. into cash. Furthermore, it presents none of the organizational complexity and cost found in ABC implementations; possibly its greatest strength. Although many accounting experts agree that ABC can generate more accurate cost information, with low manufacturing overhead or a single product line, the cost to implement and adopt ABC may exceed the benefits (Vokurka & Lummus 2001). TBA supports ‘the Goal’ as articulated by Goldratt and Cox (1992) “to make money in the present as well as in the future”. Most importantly, in linking throughput-time directly to product cost, TBA mitigates cost distortions and serves to eliminate potential dysfunctional behavior in terms of TBM strategy (figure 2.7).

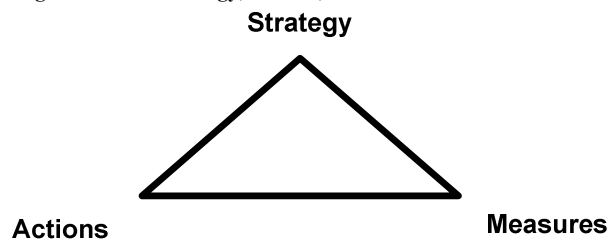
Figure 2.7: Cost Distortion and MAS Design in TBM Strategic Context



Interface of Managerial Accounting and Operations Management

The potential of any advanced manufacturing systems is greatly enhanced by the use of appropriate manufacturing performance measures reflected in the MAS. Hiromoto (1988) points out that a central principle guiding managerial accounting in Japan is that accounting policies should be subservient to corporate strategy, not independent of it. Nanni and others (1992) support the belief that performance measurement, i.e. MAS, is of equal importance to management strategy and actions (figure 2.8). The relationship between strategies and actions is fairly obvious. An organization's actions should be taken to support strategies, but the role played by measures in supporting strategies and actions should be equally obvious.

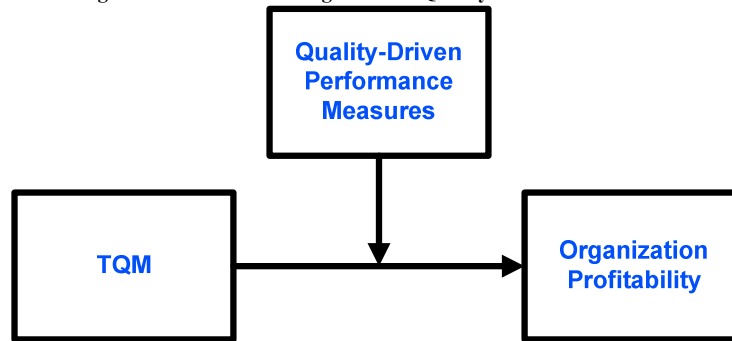
Figure 2.8: The Strategy, Actions, and Measurement Connection



This belief that MAS congruence with manufacturing strategy would lead to higher levels of firm performance has been touted by academic researchers and consultants ever since Kaplan's (1983) seminal article. While very little has been written on MAS congruence in terms of a time-based manufacturing strategy, there have been a handful of studies linking performance of other advanced manufacturing systems with appropriate accounting measures. These empirical studies suggest the need to develop an innovative MAS alternative for a TBM strategy.

Chenhall's (1997) study of 39 manufacturing organizational units exemplifies the reliance on manufacturing measures and its impact on the potential increased organizational profitability of TQM. Given the nature of advanced manufacturing philosophies, such as TQM, those firms whose MAS focuses on key measures of customer satisfaction, on-time delivery, and quality production measures – such as throughput-times, defect rates, and inventory levels – show greater organizational profitability. Figure 2.9 below shows moderating effect of quality-driven performance measures on the relationship between TQM and organizational profitability.

Figure 2.9: The Moderating Effect of Quality-Driven Performance Measures



Firms utilizing an MAS alternative geared towards corporate goals show highest performance. Sim and Killough (1998) showed a disordinal interaction between advanced manufacturing systems, such as TQM/JIT, and the appropriateness of manufacturing performance measures, i.e. the firm's MAS. As seen in figure 2.10 below, those companies demonstrating high levels of TQM/JIT practices and utilizing appropriate performance measures related to customer quality along with performance contingent rewards showed the highest performance. One interesting result was that firms utilizing a low degree of TQM/JIT also performed well as long as their accounting measures also reflected a low degree of TQM/JIT congruent measures. This would indicate that mass production industries may perform best using traditional cost systems.

Figure 2.10: Disordinal Interaction of the Goal-Congruent Measures and Performance

Degree of Use of Goal-Congruent Accounting Measures	High	Low Performance	High Performance
	Low	High Performance	Low Performance
		Low	High
		Degree of Use of Advance Manufacturing Systems	

This empirical evidence indicates that a firm's MAS should be designed contingent on characteristics of the production system, or that the characteristics of the production and managerial accounting systems should be simultaneously designed. It further suggests the possibility that an important reason some firms have not experienced performance gains from implementing advanced manufacturing systems is a reliance on an inappropriate MAS alternative. The following matrices illustrates the match between goal-congruent accounting measures and the degree of use of advanced manufacturing systems as adapted from the findings of Sim & Killough 1998 and clearly reinforces the need for congruence between the manufacturing practices and the performance measures utilized in the form of managerial accounting systems.

Using a survey of top manufacturing executives at 253 US firms, Fullerton and McWatters (2002) found a relationship existed between the level of JIT manufacturing practices implemented and the performance measures and incentive systems inherent to the managerial accounting system. Specifically, the use of non-traditional measures such as bottom-up measures, product quality, and vendor quality, as well as incentive systems of employee empowerment and compensation rewards for quality production are positively related to the degree of JIT manufacturing practices implemented. The factors

which emerged as critical for performance measures and incentive systems and are associated with a high degree of JIT manufacturing practices are:

- The frequency with which quality issues are measured and reported to management strata.
- The importance of quality and teamwork in determining compensation.
- The use of bottom-up data gathering techniques such as Pareto analysis, histograms, and cause-and-effect diagrams to evaluate operations
- The importance of adherence to budget items in determining compensation
- The use of benchmarking to evaluate operations.
- The use of performance measure related to waste and inefficiency in evaluating the manufacturing system.
- The extent to which employees understand the firm's strategic plan.
- The use of performance measures related to timeliness and vendor performance in evaluating the manufacturing system.
- The use of non-financial measures to determine compensation.
- The extent to which line managers and non-management personnel are empowered to make decisions.

The authors suggest that in order to build and strengthen long-term competitive advantage, firms must provide low-cost, high-quality products under time-based demands. The results of this study demonstrate that successful implementation of JIT practices also requires a complementary decision making and control system.

It is clear that the selection of an appropriate accounting system is critical to the performance of the manufacturing system. The question is which accounting system to use, given several options. The answer is to choose the accounting system that best depicts the manufacturing reality. In the cases where manufacturing overhead was high in relation to direct labor and material costs and product variety is high, ABC outperformed the other two systems. In cases where a 'push' manufacturing strategy is used for long production runs of standardized products, in conjunction with centralized resource management systems such as manufacturing resource planning (MRP), MRP II,

or enterprise resource and planning systems (ERP), TCS outperformed the other two accounting systems (Lea & Min 2003).

Other variations of MAS alternatives have been proposed and have shown positive results when applied in the appropriate manufacturing environment. One of these is the so called 'balanced score card', which focuses on four key performance measures: customer, financial, learning, and internal processes (Hoque 2002). Another variation proposed is one of integrated performance measures, which is based on a continuous process of developing performance measures. Based on case studies observed by Nanni and others (1992) propose an 'Integrated Performance Measurement System', which balances cost and other performance knowledge and employs it operationally at every step in the strategic management cycle. The utility of the MAS in supporting advanced manufacturing systems diminishes as the variety of cost data needed increases, the emphasis on future cost increases, and the importance of non-financial measures increases. In a dynamic manufacturing environment, in which the process is continuously changing, dynamic performance measures give the best results.

Although rarely seen in practice, the pathologic case of abolishing the MAS altogether has also been proposed by some researchers and practitioners, citing the fact that in many cases factory managers ignore internal accounting reports all together. It seems that production managers have come to realize that the performance measures being generated by the MAS more often than not reflect little more than a distorted view of shop floor reality. Neely (1999) found much the same reaction in his case study research of UK manufacturing, as he watched a production manager of a small manufacturing business through a freshly-delivered 200 page performance report straight

into the waste bin. The response of the manager to Neely's puzzled look; "we measure everything that walks and moves, but nothing that matters." In their book, *The Race*, Goldratt and Fox (1986) share their experiences consulting to many manufacturing companies and finding that managers often used a combination of cost and intuition in deciding what actions to take, where cost is trumped by intuition.

Currently we are using both cost and intuition to determine what actions to take. The mere fact that we override the cost recommendations so frequently already tells us that cost procedures are not adequate. Intuition often helps to improve the cost recommendation, but unfortunately intuition is not a basis for good communication. Even though this combination of cost and intuition were not sufficient in the past, this new competitive race has now made them totally obsolete.

The Impact of Product Mix Complexity on Manufacturing

Product mix complexity has been a focus of study in both the managerial accounting and the production operations management literature for some time now. Product complexity is commonly defined along two dimensions, breadth complexity and depth complexity (Bentorn and Srivastava 1985). Breadth complexity is represented by the number of components in the parent product (Bentorn and Srivastava 1985). Depth complexity is defined as the maximum number of dependent relationships in a product structure (Veral and Laforge 1985, Sum et al. 1993).

Product mix complexity has commonly been found as a primary variable in various POM studies such as in lot sizing studies (Blackburn and Millen 1980; 1982a & b, Veral and LaForge 1985, Benton and Srivastava 1985, LaForge 1985), in system nervousness strategy studies (Blackburn et al. 1986), and in capacity control policy studies (Gutzmann and Wysk 1986). In all studies it has been shown as a primary driver

of cost as the number of levels in the product BOM increases. In addition, some researchers have shown that JIT strategies work better with a flat BOM structure, while MRP performs better with a deep structure (Vollmann et al. 1992).

Product mix complexity has also been reported as an important factor in product costs amongst various managerial accounting alternatives. Many researchers have suggested that product complexity increases cost distortions under traditional managerial accounting scenarios (Cooper 1988a , Brimson 1991). Their research has shown that the manufacturing of complex products with many part numbers and complex functions places far greater demands on overhead support activities such as production planning, engineering, purchasing, quality control, and logistics than do more simple products. When the quantity of volume-related resources that a product consumes does not vary in direct proportion to the quantity of volume-unrelated resources consumed, a volume-based cost system such as traditional absorption-based costing systems will report distorted product costs (Cooper 1988a). Many researchers in managerial accounting have posited that traditional absorption-based costing systems may under allocate manufacturing overhead costs to the low volume, complex products and may over allocate manufacturing overhead costs to the high volume, simple products when both types of products are manufactured in the same facility, because these manufacturing overheads are most often allocated on direct labor hours, machine hours, or some other volume basis (Johnson 1991b, Chalos 1992, O'Guin 1991).

Increasing complexity is the bane of a factory manager's life. With increasing complexity comes and increased number of parts, greater material handlings and inventories, more diverse process flows, higher supervision requirements, an increase of errors and defects, and smaller batches produced in shorter runs. (Abegglen & Stalk 1985, pg. 81)

Henry Ford was possibly the first industrialist to realize the full cost of variety on operations when he famously ruled that “the American can have a Ford in any color as long as it is black”. Closely related to product complexity is the impact of demand variability production operations. Theories in economics, operations management, and managerial accounting have long predicted that producing a heterogeneous product mix increases costs and reduces operating performance (Skinner 1974, Hayes & Wheelwright 1984, Johnson & Kaplan 1987, Banker et al. 1988, Anderson 1995) as a result of transactions caused by complex material flows, capacity balancing, quality control, and change (Miller & Vollmann 1985).

The impact of product mix complexity has become of increasing importance to researchers in both managerial accounting and production operations management as producers have struggled to cope with the ever increasing demands of consumers for greater product variety in the post-industrial age (Doll & Vonderembse 1991). The scope of product mix complexity is defined as the breadth of product offerings produced at a single facility; the wider the breadth of product offerings, the greater the complexity of the factory operations.

As factory complexity increases, greater overhead resources must be committed to ensure smooth operations. For example, the scheduling, material handling, and expediting efforts required to support a line on which only one product is assembled are significantly lower than the efforts required to support a line on which ten or twenty products are assembled... The more variety of products fabricated or assembled on a line, the less focused, the more difficult to manage, and the greater the required overhead support will be. By inference, the more products a factory manufactures, the less focused, the greater the overhead requirements and costs and the higher its total production costs will be. (Abegglen & Stalk 1985: 81-83)

Theory in both managerial accounting and operations management has predicted that producing a heterogeneous product mix increases costs and reduces operating performance (Skinner 1974, Hayes & Wheelwright 1984, Hill 1985, Johnson & Kaplan 1987, Banker et al. 1988). Later empirical studies confirm the impact of product variety on manufacturing costs (Anderson 1995, Fisher & Ittner 1999, Randall & Ulrich 2001, Pil & Holweg 2004, Peacock 2005). Numerous operations management and accounting studies indicate that product variety's greatest impact may be on indirect and overhead costs rather than on direct labor productivity, yet little attention has been given to the impact of product variety in mixed-model assembly operations (Fisher & Ittner 1999).

Under TCS, the addition of greater product offerings can place an unfair burden on older, less complex products, which often require far less overhead support activities than the newer, more complex products.

Complexity can also be substantially increased when a new product line is introduced, but the older product line is continued for parts or to keep a valued customer. In the early stages of product line expansion, the savings from the product redesigns and from the incremental volume often offset the costs and aggravation of the factory's increased complexity. However, as the expansion continues, the factory becomes increasingly less efficient, more costly, and less profitable. Because most management [accounting] and [financial] accounting systems are used as score cards rather than to demonstrate cause and effect, they are unlikely to show why the performance of the factory is deteriorating. (Abegglen & Stalk 1985, pg. 81)

Simulation Methodologies in Managerial Accounting Research

There has been an increasing recognition amongst some of the top scholars in the managerial accounting field of the need for innovative methodologies to fill in the knowledge gaps (Kaplan 1993, Zimmerman 2001, Ittner & Larcker 2001, 2002). Ittner

and Larcker (2001) highlight that managerial accounting research has developed a somewhat faddish nature and has yet to produce an integrated, substantive body of knowledge. Zimmerman (2001: 422) is even more critical of the state of managerial accounting research and its over reliance of empirical methods. He argues that the state of affairs described by Ittner & Larcker (2001) is caused by a focus of describing practice rather than testing theories. In his view, managerial accounting research has become completely descriptive, and he believes that nothing more substantial can be learned from continued field research. He describes accounting field researchers as “wandering the hallways of corporations and manufacturing plants searching for facts unguided by tentative hypotheses”.

Since this dialogue was brought back to the forefront of managerial accounting research, there has been a renewed interest in looking outside the more established empirical methods, towards other fields. Since managerial accounting interfaces with operations management, it makes sense that it should borrow some of the more established methodologies such as mathematical modeling and linear programming and emerging methodologies such as simulation modeling.

Figure 2.11 below has a listing of recent studies from operations management, industrial engineering, and managerial accounting which have employed some form of simulation methodology within the past ten years. As evident from this listing, with only one journal article greater than five years old, and the rest within the past five years, this methodology is gaining acceptance within the research community.

Figure 2.11: Select Managerial Accounting Studies Using Simulation Methodologies (2001-2006)

No.	Author	Date	Title	Publication	Methods	Findings
1.	Vokurka & Lummus	2001	At What Overhead Level Does Activity-Based Costing Pay Off?	Production & Inventory Management Journal	Scenario analysis with four fictional companies with identical production but different OH levels comparing TCS and ABC	Companies should examine their overhead levels to determine whether ABC is valuable to them. Companies with high overhead levels and companies with a wide product mix and identifiable activities may benefit. Companies with lower overhead and static product mix may find it cost prohibitive to implement.
2.	Boyd & Cox	2002	Optimal Decision Making Using Cost Accounting Information	International Journal of Production Research	Simulation in a resource constrained environment (TOC) compares TCS, ABC, direct costing, and TA.	TA produced the same results linear programming. All others produced suboptimal results. For cost accounting system to provide information for optimal decisions it must be aware of production constraints and not use allocated cost.
3.	Ozbayrak et al.	2004	Activity-Based Cost Estimation in a Push/Pull Advanced Manufacturing System	International Journal of Production Economics	Uses mathematical modeling and simulation to compare TCS and ABC in an automated manufacturing setting.	Pull strategy gives consistently better results under both TCS and ABC, but the optimum is using ABC.
4.	Lea & Fredendall	2002	The Impact of Management Accounting, Product Structure, Product Mix Algorithm, and Planning Horizon on Manufacturing Performance	International Journal of Production Economics	Large-scale simulation examines TCS, ABC, and TA in two shops, one with a flat and one with a deep product structure, in a highly automated industry (high overhead).	No single shop setting is best of all performance measures. The manager must determine which performance measures are most important when choosing an MAS, product mix algorithm, or product structure.
5.	Lea & Min	2003	Selection of Management Accounting Systems in Just-In-Time and Theory of Constraints-Based Manufacturing	International Journal of Production Research	Large-scale simulation examines TCS, ABC, and TA in highly automated JIT and TOC environments with different planning horizons.	The management accounting system that best depicts the manufacturing process provides better product cost information and results in better system performance.
6.	O'Brian & Sivaramakrishnan	1996	Coordinating Order Processing and Production Scheduling in Order Initiated Production Environments	Journal of Management Accounting Research	Simulation model of order initiated environment compares cost (TCS) and cycle time in as order cutoff criteria for coordinating order processing.	A simple cycle-time cutoff decision rule outperforms a cost decision rule.
7.	Leitch	2001	Effect of Stochasticity, Capacity and Lead Time Cost Drivers on WIP and Throughput in a Pull Production Environment	Management Accounting Research	Simulation examines the tradeoffs of cost drivers and its effects on strategic cost management.	The effects of these cost drivers are much different than traditional push manufacturing systems and management may need to model these and other cost drivers to assess their impact on performance.
8.	Burrows et al.	2001	Real-Time Cost Management of Aircraft Operations	Management Accounting Research	Uses flight simulation to model cost impact of pilot inflight decisions on entire flight network.	Provides a unique and detailed description of cost-drivers and cost-behavior information for a non-manufacturing environment.
9.	Balakrishnan & Sivaramakrishnan	2002	A Critical Overview of the Use of Full-Cost Data for Planning and Pricing	Journal of Management Accounting Research	Uses mathematical modeling and simulation in conjunction to determine the relationship between product pricing and product, capacity, and allocation decisions.	Flexible capacity mitigates the potential perverse incentives of full-costing on product, capacity, and capacity allocation decisions.
10.	Banker & Hansen	2002	The Adequacy of Full-Cost-Based Pricing Heuristics	Journal of Management Accounting Research	Uses mathematical modeling and simulation in conjunction to compare three pricing heuristics in a service operation.	Full costing provides optimal performance when demand is greater than capacity in a service operation where there is 'soft' capacity in the form of backorderd service.
11.	Leitch et al.	2005	Opportunity Costing Decision Heuristics for Product Acceptance Decisions	Journal of Management Accounting Research	Uses simulation to evaluate full-cost heuristics under different levels of stochastic demand, lead-time, cost structure, and workstation capacity.	Full costing works well when lead times are long and shop capacity is balanced. Deviations reduce its performance over other heuristics.
12.	Meade	2004	Modeling the Strategic Impact of Management Accounting Methods on the Implementation of Lean Manufacturing	Doctoral Dissertation: Western Michigan University	Evaluates lean implementation ex post facto through the lens of differing MAS alternatives.	Focus on short-term impact (6 month simulation) limited the generalizability of findings.
13.	Whittenberg	2004	Decision Usefulness of Management Accounting Information Systems in Constraints-Based Manufacturing Operations	Doctoral Dissertation: Nova Southeastern University	Human experiment using masters of accountancy students to determine impact of MAS design on decision making.	Results inconclusive as time allowed for only two iterations and difficult to determine if improvements were due to changes in MAS design or simply learning to better play the game.

Vokurka and Lummus (2001) use a simple scenario analysis with a spreadsheet of four fictional companies with identical production and differing levels of factory overhead (6%, 18%, 31%, and 40%) to compare TCS and ABC. The results suggest that companies need to examine their overhead levels in order to determine whether implementing ABC will be valuable. Companies with high levels of manufacturing overhead, wide product mixes, and clearly identifiable activity drivers may benefit, while companies with lower levels of manufacturing overhead and a static product mix may find it cost prohibitive to implement.

Boyd and Cox (2002) use simulation software in a resource constrained environment (TOC) comparing TCS, ABC, direct costing, and TA. They found that TA

produced the same results as a linear programming model, while all others produced suboptimal results. In the TOC environment, for a cost accounting system to provide information for optimal decision making it must be based on production constraints and not use allocated costs.

Özbayrak and others (2004) use a combination of analytical and simulation modeling to compare TCS and ABC in a highly automated manufacturing setting. They found that a pull strategy gives consistently better operational results, regardless of accounting scheme, but is optimized in conjunction with ABC.

Lea and Fredendall (2002) use a large-scale simulation model to examine TCS, ABC, and TA in a highly automated manufacturing environment, i.e. manufacturing overhead. They compare these accounting schemes under different product structures finding that no one performs best under all shop settings. The manager must determine which performance measures are most important when choosing an MAS. Lea and Min (2003) repeated this simulation comparing the same managerial accounting schemes in both a JIT and TOC environment with different planning horizons. They found that the MAS that best depicts the manufacturing process provides better product cost information and results in better system performance.

This research follows a similar methodology as that used in Lea and Fredendall (2002) including two of the same input variables, i.e. MAS alternatives and product mix complexity. The current study develops an additional MAS alternative and addresses another limitation of the Lea and Fredendall (2002) study – namely that it only considered an environment with a high level of manufacturing overhead. One additional limitation, from an accounting standpoint, is that TCS is not operationalized as is most

common in practice, with direct labor hours as an allocation basis. Rather, they use direct machine hours, which would be consistent with a highly automated environment. They note that this may be the reason why their anomalous results were inconsistent with other researchers (Cooper & Kaplan 1992, Johnson 1991 & 1992, Kaplan 1989, O'Guin 1991).

O'Brian and Sivaramakrishnan (1996) used a simulated order initiated environment to compare financial and operational cutoff criteria under a TCS for coordinating order processing. They found that a simple throughput-time cutoff decision rule outperforms a cost decision rule. Although this study is not concerned with different account schemes, the use of throughput-time for the decision rule further supports the use of throughput-time as an allocation basis within the managerial accounting scheme.

Leitch (2001) used a simple simulation model of a pull-production environment to examine the tradeoffs between various cost drivers (stochasticity, capacity, and lead-time constraints) and its effects on strategic cost management. He notes that even relatively simple simulated pull-production environments can offer much complexity and difficulty in analysis. He found that the effects of these cost drivers are much different than in traditional push manufacturing systems, and that management needs to consider the impact of these and other cost drivers on performance. This study supports the use of simulation in accounting studies, so that stochasticity may be introduced to the experimental environment.

Burrows and others (2001) accessed a flight-simulator and utilized its 'real-time' operational data to model the impact of pilot in-flight decisions on the entire network. This research provided a unique and detailed description of cost-drivers and cost-behavior in a non-manufacturing environment.

Balakrishnan and Sivaramakrishnan (2002) used analytical and simulation modeling to determine the relationship between product pricing, production capacity and allocation decisions in a fictional organization. They found that flexible capacity mitigates the potential perverse incentives of full-costing on production capacity and allocation decisions.

Banker and Hansen (2002) used analytical and simulation modeling in order to compare three pricing heuristics in a fictional service operation. They found that full costing provided optimal performance when demand is greater than capacity in a service operation where there is soft capacity in the form of backordered service.

Leitch and others (2005) used simulation modeling to evaluate full-cost heuristics under different levels of stochastic demand, lead-time, cost structure, and work station capacity. They found that full costing works well when lead times are long and shop capacity is balance. Deviations from this ideal scenario reduce its performance over other heuristics.

A search of doctoral dissertations/theses involving managerial accounting (including the terms managerial accounting, cost accounting, costing, and accounting) and simulations was performed in ProQuest since 2000 returning two such dissertations. Meade (2004) uses simulation to model the short-run (six months) impact of the implementation of lean manufacturing on financial performance measures through different managerial accounting methods. The focus of this research was not on how the MAS alternatives affect manufacturing operations, but rather how the view of manufacturing operations is changed through the lenses of different accounting schemes during the implementation phase of lean manufacturing. Although this study provided a

useful look at the impact of lean manufacturing through various accounting treatments, its short-term view did little to show the system-dynamic relationship of managerial accounting on the success of lean initiatives. The author provided some useful suggestions for future research that are incorporated into the study at hand including expansion of the time horizon, incorporation of customer service measures, use of income measures, and the addition of non-normal distributions in supply and demand.

Whittenberg (2004) compared the decision usefulness of various managerial accounting schemes in a constraints-based manufacturing environment. This study utilized a scenario experiment where human participants made production decisions based on information provided, both operational and financial. Although this was a novel approach to introducing 'real' human decision making into the simulation game model, it had a number of limitations due to its nature. First, because of time constraints and the attention span of participants, the simulation could only be run in two iterations, one without the accounting information and one with the accounting information. Even then, the two or three hours needed to play the game precluded most of the operations and accounting professionals originally targeted for the study.

In the end, a group of 159 participants, including 15 industry professionals, was assembled mainly from accounting undergraduate and graduate students. Since time only allowed for two iterations of the game, it was difficult to determine whether improvements in system performance are due to improved decision making or simply to the learning curve of the game itself. In the end, the author notes that the variability introduced by allowing human subjects, from many backgrounds and experience levels, the freedom to make decisions tends to limit the conclusions that can be drawn from the

results. The author suggests that future research would add controls to limit the decision making alternatives; however, this would obviate the purpose of having real human subjects. This conclusion supports the use of integer linear programming in conjunction with the simulation model, which would allow for a realistic decision making process without the inherent variability.

Chapter Three

Research Design

This chapter describes the development of an experimental research design that utilizes a simulation modeling methodology to examine the impact of different MAS alternatives on production decision making and ultimately on manufacturing performance in the context of a TBM strategy. The simulation model is developed and executed within Rockwell Automation Corporation's ARENA software, one of the most powerful commercially available simulation tools on the market today. ARENA is a highly flexible and endlessly reconfigurable tool that allows the research analyst to create an animated simulation model representing virtually any system, build in complex underlying costing and decision logic sub models, and statistically analyze the results of system input modifications.

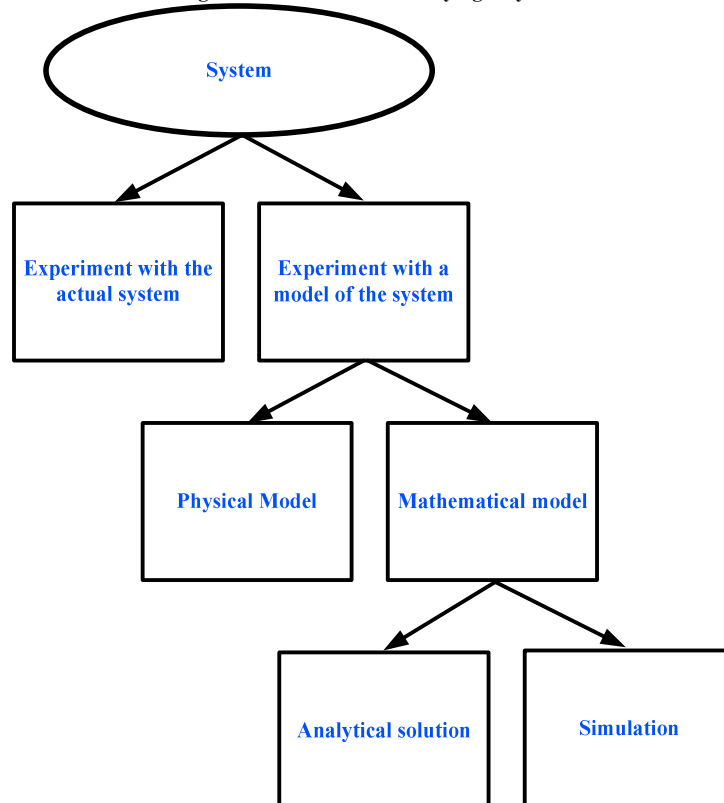
Increasingly researchers are giving less credence to the pejorative old saw often describing simulation as a 'method of last resort'. In fact, researchers are increasingly looking to simulation methods as a method of first resort, in most cases due to the sheer complexity of systems of interest and the models necessary to represent them in a credible and valid way (Law & Kelton 2000). To observe real manufacturing systems is often very expensive and sometimes cumbersome; a simulation model is an easier way to build up models representing real-life scenarios (Ali et al. 2005).

Despite the fact that simulation is an emergent field of research, it has already become one of the most widely used techniques in industrial engineering and

operations research. There have been several meta-analytic studies that support this claim. Lane and other (1993) reported that from 1973 through 1988 that simulation was ranked consistently as one of the three most important operations research techniques. The other two were math – a catch all category as simulation is itself a form a modeling – and linear programming. Gupta (1997) analyzed 1,294 papers from 1970 through 1992 in the journal *Interfaces* – one of the leading journals highlighting real world applications of operations research theory – finding simulation only second to ‘math’ programming out of the 13 research methodologies utilized.

Since the impact of the introduced TBA method of accounting is of primary interest, the scope of this study is limited to experimental methodologies. Law and Kelton (2000) describe the various options to the researcher in terms of experimental methodologies. Figure 3.1 below presents experimental options for studying a system.

Figure 3.1: Methods for Studying a System



Unlike case studies that examine actual companies, simulation studies use hyperphysical companies in purposefully designed experimental environments. The scale, processing times, decision criteria, and other settings are usually carefully designed to examine extreme pathological settings as well as normal settings. Therefore, results from the simulation model demonstrate patterns and trends over a decision space, but managers need to interpret the results carefully to ensure they correspond to their own manufacturing environment (Lea & Fredendall 2002). However, the use of longitudinal data from a single [simulated] plant avoids differences in production functions and other correlated omitted variables that may bias cross-sectional studies (Fisher & Ittner 1999). All other variables can be controlled through the use of simulation, and a sufficiently large sample of outcomes can be obtained to achieve significant comparisons between production systems with differing degrees of variation and levels of stress due to capacity and service constraints (Leitch 2001).

Simulation modeling is becoming a widely used tool in particularly in service operations management studies, where the behavior of human participants in the service system both as servers and customers can be erratic. According to Metters and others (2006), simulation offers several advantages over some more traditional OM techniques such as analytical (mathematical) modeling. One of the biggest advantages is that it allows for the compression of years of experience into just a few seconds or minutes of computer processing time. While it may take multiple simulation runs and scenarios to evaluate a system, with no guarantee of an optimal solution, the results of simulation modeling are far more general than mathematical models. Although both simulation and

mathematical approaches require assumptions, a simulation model can deal with complexity better and a mathematical model affords greater precision (Leitch 2001).

The remainder of this chapter is organized as follows: the first section describes the experimental design, statistical hypotheses, and decision logic foundation of this study. The second section describes the design and parameters of the simulation model and experiment. The third section defines the experimental factors and performance measures used to evaluate the various managerial accounting systems.

Experimental Design

The experimental research design used to address the research questions posed in Chapter 1 includes three experimental factors; the various levels of management accounting system alternatives(MAS), three levels of product mix complexity (MIX), and three levels of manufacturing overhead (MOH). According to Kleijnen (1995) most simulation analysts apply an inferior design of experiments, changing one input at a time as opposed to factorial (2^{K-P}) designs, which controls estimated effects of input changes and shows the importance of interaction effects. For each performance measure the experimental design is a 3 X 3 full factorial with 60 replications, thus resulting in a total of 1620 (3x3x3x6) observations.

The experimental design is then:

$$\begin{aligned} Y_{aom} = & \mu + MAS_a + MOH_o + MIX_m && \text{(Main Effect)} \\ & + MAS_a * MOH_o + MAS_a * MIX_m + MOH_o * MIX_m && \text{(Two-Way Interaction)} \\ & + MAS_a * MOH_o * MIX_m && \text{(Three-Way Interaction)} \\ & + e_{aom} \end{aligned}$$

Where: Y_{aom} = Performance Measurements
 μ = Mean Effect
 MAS_a = Management Accounting System Effect, $a = 1, 2, 3$
 $MAS_1 = TCS$
 $MAS_2 = ABC$
 $MAS_3 = TBA$
 MOH_o = Manufacturing Overhead Level Effect, $o = 1, 2, 3$
 $MOH_1 = Low$
 $MOH_2 = Medium$
 $MOH_3 = High$
 MIX_m = Product Mix Complexity Effect, $m = 1, 2, 3$
 $MIX_1 = Narrow$
 $MIX_2 = Medium$
 $MIX_3 = Wide$
 e_{aom} = Random Effect

Statistical Hypotheses

Research Questions

RQ1: Does the use of different managerial accounting systems in production decisions result in differences in average demand fulfillment rate, cycle-time, and/or net operating income?



Null Hypotheses

H₀₁: The use of different managerial accounting systems has no effect on average demand fulfillment rate, cycle-time, and/or net operating income..

$$\text{MAS}_a = 0$$

RQ2: What effect does manufacturing overhead level have on average demand fulfillment rate, cycle-time, and/or net operating income?



H₀₂: Manufacturing overhead level has no effect on average demand fulfillment rate, cycle-time, and/or net operating income.

$$\text{MOH}_o = 0$$

RQ3: For a given level of manufacturing overhead level does the use of different managerial accounting systems in production decisions result in differences in average demand fulfillment rate, cycle-time, and/or net operating income?



H₀₃: The use of different managerial accounting systems has no effect on average demand fulfillment rate, cycle-time, and/or net operating income for a given level of manufacturing overhead.

$$\text{MAS}_a * \text{MOH}_o = 0$$

RQ4: What effect does product mix complexity have on average demand fulfillment rate, cycle-time, and/or net operating income?



H₀₄: Product mix complexity has no effect on average demand fulfillment rate, cycle-time, and/or net operating income.

$$\text{MIX}_m = 0$$

RQ5: For a given level of product mix complexity does the use of different managerial accounting systems in production decisions result in differences in average demand fulfillment rate, cycle-time, and/or net operating income?



H₀₅: The use of different managerial accounting systems has no effect on average demand fulfillment rate, cycle-time, and/or net operating income for a given level of product mix complexity.

$$MAS_a * MIX_m = 0$$

RQ6: For a given level of product mix complexity does manufacturing overhead level affect result in differences in average demand fulfillment rate, cycle-time, and/or net operating income?



H₀₆: Manufacturing overhead level has no effect on average demand fulfillment rate, cycle-time, and/or net operating income for a given product mix complexity.

$$MOH_o * MIX_m = 0$$

RQ7: For a given level of product mix complexity and a given level of manufacturing overhead level does the use of different managerial accounting systems in production decisions result in differences in average demand fulfillment rate, cycle-time, and/or net operating income?



H₀₇: The use of different managerial accounting systems has no effect on average demand fulfillment rate, cycle-time, and/or net operating income for a given level of product mix complexity and a given manufacturing overhead level.

$$MAS_a * MOH_o * MIX_m = 0$$

Experimental Factors

The three **managerial accounting system (MAS)** alternatives investigated in the simulation model are a traditional costing system (TCS), activity-based costing (ABC), and time-based accounting (TBA). Figure 3.2 below summarizes the product and period cost classifications under these various MAS alternatives:

Figure 3.2: Product and Period Cost Classifications by MAS

Cost Classification	Amount			ABC	TBA	TCS
Manufacturing Direct:						
Direct materials				P (actual cost)	P (actual cost)	P (actual cost)
Direct labor	Varied with production volume and collected from simulation			P (actual cost)	P (actual cost)	P (actual cost)
Depreciation (machinery)				P (actual cost)	P (actual cost)	P (actual cost)
Manufacturing Overhead Costs:	LOW	MED	HIGH			
Depreciation (other)	12,000	18,000	24,000	P [1]	throughput time	MOH (by labor hours)
Data entry	2,200	4,400	8,800	P [1]	throughput time	MOH (by labor hours)
Cost analysis	4,800	7,200	9,600	P [1]	throughput time	MOH (by labor hours)
Production engineering	3,200	4,800	6,400	P [1]	throughput time	MOH (by labor hours)
Scheduling	2,200	3,300	4,400	P [1]	Period cost throughput time	Period cost MOH (by labor hours)
Quality control	1,340	1,444	1,653	P [2]	throughput time	MOH (by labor hours)
Production supervision	4,200	6,300	8,400	P (time/product)	throughput time	MOH (by labor hours)
Utilities	4,500	6,750	9,000	P (actual usage)	throughput time	MOH (by labor hours)
Miscellaneous	1,000	1,000	1,000	P (sales volume)	throughput time	MOH (by labor hours)
Selling, General, & Administrative						
R&D - Basic			1,200	P (time/product)	Period cost	Period cost
Customer support			7,400	P [3]	Period cost	Period cost
General administration			3,000	P (sales volume)	Period cost	Period cost
Advertising costs - general			1,200	P (sales volume)	Period cost	Period cost
Advertising costs - product	A		1	P (actual cost)	Period cost	Period cost
	B		2	P (actual cost)	Period cost	Period cost
	C		4	P (actual cost)	Period cost	Period cost
Commision	A		5	P (actual cost)	Period cost	Period cost
	B		10	P (actual cost)	Period cost	Period cost
	C		20	P (actual cost)	Period cost	Period cost

The first column identifies the common manufacturing and selling, general, and administrative (SG&A) activities. The second column shows the cost associated with each activity followed by the classification and basis for absorption under each of the

MAS alternatives. These classifications are consistent with prior research (Brimson 1991, O'Guin 1991, Low 1992, Lea & Fredendall 2002).

As discussed in the literature review in Chapter 2, manufacturing activities such as raw material purchases, direct labor, and direct manufacturing costs, i.e. machine depreciation, are all likely to vary with production volume and are therefore considered variable. These costs will be collected over the course of the simulation runs. Most of the other costs in manufacturing are generally considered fixed or semi-variable in nature, and these make up the classifications of MOH and SG&A. As seen in figure 12 above, the use of different MAS alternatives entails a different treatment of these costs, either as product or period costs. These activities listed are consistent with the activities studied in the literature (Low 1992, Brimson 1991, O'Guin 1991, Lea & Fredendall 2002).

As the main focus of this research is to make a direct comparison of the various MAS alternatives, budgeting and cost control is not of interest in this study. Therefore, in order to simplify the accounting complexity at the end of each period and avoid tracking variances and adjustments, an actual costing methodology is utilized in this research. The accounting period for this study will be set to one month, i.e. 4 weeks, as this is common practice in industry. All performance measures will be collected at the end of each period.

The direct materials, labor, and manufacturing, as well as the ordering costs, are treated as product costs under each MAS. The cost of raw materials is designed to reflect the greater percentage of standardized and non-standardized parts used in high-volume,

simple products and low-volume, complex products respectively. Figure 3.3 below outlines the costs for the individual parts used in this study.

Figure 3.3: Material Purchase Prices

Associated with		
Item No.	Product(s)	Cost/Unit
110	B & C	\$ 20
220	A & B	\$ 10
230	C	\$ 25
240	C	\$ 25
250	C	\$ 25
260	C	\$ 25

Brimson (1991) indicates that manufacturing overheads are frequently accumulated by department and then allocated to products using a predetermined allocation rate. In this study, TCS and TBA will allocate these costs based on the percentage of total direct labor hours and total cycle-time to each product class respectively. ABC, on the other hand, treats all costs as product costs, attempting to fully allocate all overhead costs by their individual actual activity drivers. Since it is rarely possible to accurately trace 100% of MOH to specific activity drivers and then allocate it to specific products by those drivers (Vokurka & Lummus 2001, Kaplan & Anderson 2004), miscellaneous MOH will be allocated by a single generic basis, the measure of an individual product class' percentage of total production volume.

Under TCS and TBA all SG&A are treated as period costs, i.e. expensed within the period the actually occur. Under ABC, these costs are also treated as product costs and are allocated to the product classes by their individual activity drivers. Again, because it is nearly impossible, in most cases, to accurately trace all overhead costs to

individual activity drivers, general administrative costs are allocated by a single generic basis, the measure of an individual product class' percentage of total production volume.

Under ABC, for the sake of simplicity and generalizability, it is necessary to make certain assumptions in the cost model. It is assumed in this experiment that all the activities shown in figure 12 are primary activities that already reflect all the costs of any secondary activities. Following a similar methodology as Lea & Fredendall (2002) and Lea & Min (2003), the following formula is used to absorb the manufacturing overhead costs to an individual product:

$$R_{k,i} = C_k \frac{Q_i * N_i}{\sum_{i=1}^m Q_i * N_i} \quad [P1]$$

- Where: $R_{k,i}$ - amount of cost k allocated to product i, $i = 1, 2, 3$;
 C_k - $k = 1$ (finished goods storage), 2 (depreciation), 3 (data entry),
 4 (cost analysis), 5 (production engineering), 6 (scheduling)
 Q_i - production quantity of product i
 N_i - total number of parts in product i
 m - number of products, $m = 3$

Quality control costs per period include prevention, appraisal, and internal failure costs. For the purposes of this study, it is assumed that all products passing inspection are good products and therefore the external cost of failure is irrelevant. The activity driver associated with quality control under ABC is the actual number of occurrences weighted by a product's predetermined quality multiplier. The multiplier is selected to reflect the average time needed to inspect and either repair or reject the final product. The following formula is used to absorb the cost of quality control to an individual product:

$$X_i = \text{Quality Cost} * \frac{A_i * AQ_i}{\sum_{i=1}^m (A_i * AQ_i)} \quad [\text{P2}]$$

Where:

X_i - quality control cost allocated to product i

AQ_i - number of defects for product class i

A_i - quality multiplier for product i

m - number of products, m = 3

Customer support is the general cost of providing service to customers after the sale, and this cost varies by product classification. Similar to the quality control costs, the activity driver associated with service under ABC is the actual number of occurrences weighted by a product's predetermined service multiplier. The multiplier is selected to reflect the average time needed to service a customer by product class. The following formula is used to absorb the service costs to an individual product:

$$X_i = \text{Service Cost} * \frac{B_i * AQ_i}{\sum_{i=1}^m (B_i * AQ_i)} \quad [\text{P3}]$$

Where:

Y_i - service cost allocated to product i

BQ_i - actual service calls for product class i

B_i - service multiplier for product i

m - number of products, m = 3

Ever rising **manufacturing overhead levels (MOH)** have long been considered the primary amplifier of cost distortion and have led to the development of ABC. If manufacturing overhead were a negligible portion of total product cost, misapplication of

manufacturing overhead would not be a concern (Ruhl & Baily 1994). However, this does not reflect reality, as MOH has grown to become the single largest product cost component in most modern manufacturing plants. In general, the higher the overhead rate for a company, the greater the amplification of the difference in costs between TCS and ABC approaches. Likewise, the lower the overhead level, the smaller the cost difference between TCS and ABC. Vokurka & Lummus (2001) show an example where the same product manufactured in two different plants with differing levels of overhead burden is costed significantly different under TCS and ABC as the overhead burden increases from 6% to 40% of cost. The difference in product cost between the two MAS alternatives was only 4.1% in the former and 26.1% in the latter.

ABC has been purported to be an overall better manufacturing overhead allocation technique, but many management accounting researchers believe that this is largely contingent upon the percentage of MOH in total production costs and the percentage of that which can be allocated by specific activities or cost drivers. Per Vokurka & Lummus, companies above a certain ratio of overhead to total cost may benefit from adopting ABC; however, it may not be wise for a company with a low overhead burden or a single product line to adopt ABC because of the cost of implementation. The higher the percentage of MOH not allocated by assigned activities, the more the allocation will reflect what could have been calculated through an existing TCS. As a general guideline, the authors suggest that at least 70-75% of total overheads should be assigned by specific activities if ABC is to provide real value in decision making over TCS.

The Vokurka & Lummus study suggests that MOH begins to drive significantly different product costing under TCS and ABC at levels greater than 15%. Using a similar methodology, total MOH burden will be set at three levels – low (roughly 10 %), medium (roughly 20%), and high (roughly 40%). As in the aforementioned study, any statistically significant trend towards greater cost distortions will be assumed to continue as MOH percentage increases.

Product mix complexity (MIX) is defined as the breadth of different products – with varying levels of width in the bill of materials (BOM) for each product – produced at one factory, i.e. narrow, medium, or wide. Product mix complexity has been generally acknowledged as one of the primary drivers of manufacturing cost and a primary cause of cost distortions (Kaplan 1983, Cooper 1988a). Product mix complexity will be examined via three different products under differing levels of demand variability. The three products examined are (A) a high volume product with a narrow BOM structure, (B) a mixed volume product with a mixed BOM structure, and (C) a low volume product with a wide BOM structure.

It is not uncommon in today's competitive environment for a firm's product line to contain a mix of both high and low volume products produced within a single facility. In most cases, the high volume products tend to have a steady demand, utilize cheaper standardized parts, require little marginal overhead support, and traditionally have had lower profit margins reported under TCS. Alternatively, low volume products tend to have higher demand variation, utilize more expensive and specialized parts, require a great deal of overhead support, and traditionally have had higher profit margins under TCS (Bakke and Hellberg 1991). The three products examined in this study are designed

to reflect these characteristics in both their demand variability and BOM structures.

Figure 3.4 below summarizes the characteristics of the products used in this study:

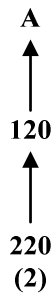
Figure 3.4: Product Characteristics

	A	B	C
Volume	High	Medium	Low
Demand variation	Low	Medium	High
Overhead usage	Low	Medium	High
Main type of parts used	Standard	Mixed	Non-Standard
Profit margin under TCS	Low	Medium	High

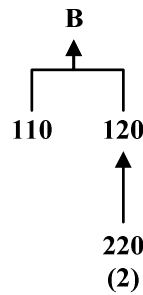
Based on the literature review in Chapter 2, the BOM levels in this study are set two levels. This treatment is consistent with the Lea and Fredendall (2002) study. The mix of differing levels of BOM structure is designed to examine the frequent claims in the management accounting literature that high-volume, simple products often subsidize low-volume, complex products under TCS. As discussed previously in the literature review in Chapter 2, traditionally, high-volume products typically are allocated a greater share of manufacturing overhead than what they marginally drive. This is due mainly to the use of volume measures such as direct labor hours or machine hours. Conversely, low-volume products often bear a share of manufacturing overhead burden considerably less than what they marginally contribute (Johnson 1991 & 1992, Chalos 1992). Figure 3.5 below shows the BOM structures of the product produced in this study:

Figure 3.5: BOM Structure by Product

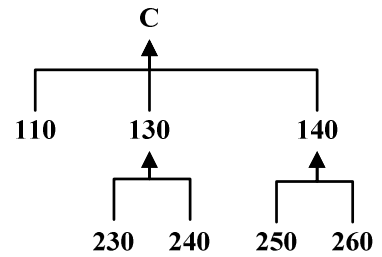
Narrow BOM Structure



Mixed BOM Structure



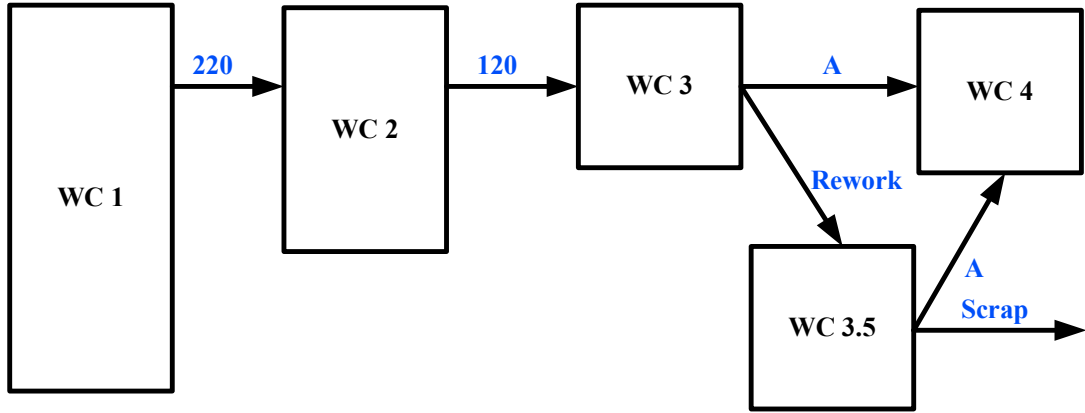
Wide BOM Structure



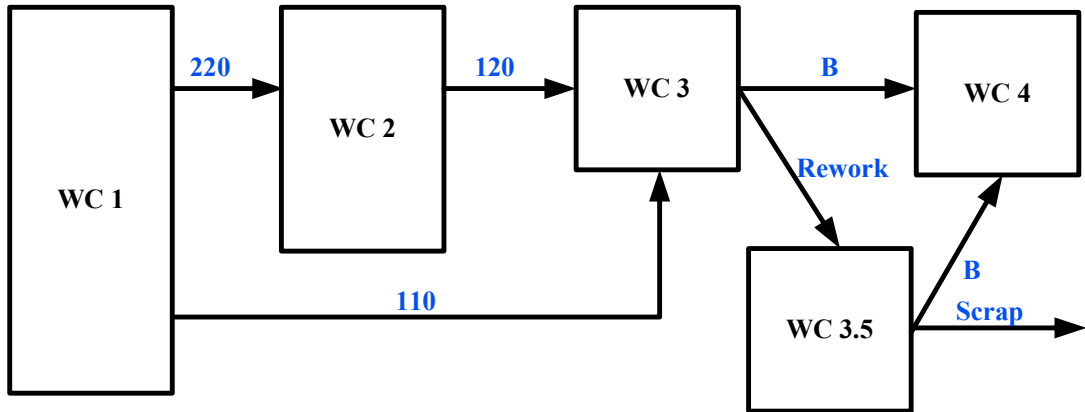
For the purposes of this study, we assume that the market is nearly perfectly competitive, i.e. there are many producers and sellers and total industry capacity is roughly equal to total demand. In this scenario, prices are market driven and any individual firm cannot affect the overall market price. In economic terms, the firm is a ‘price taker’. The selling price in this study is set at \$100 for product A, \$160 for product B, and \$360 for product C. However, at a micro level the individual firm operates in a constrained environment where it has insufficient production capacity to meet its individual market demand. Demand not satisfied by the individual firm is immediately lost to other firms in the market. In this environment variations in product cost caused by different MAS alternatives may result in different product mix decisions that in turn lead to different manufacturing performances (Bakke and Hellberg 1991).

The plant routings for the three products are can be seen in figure 3.6 below:

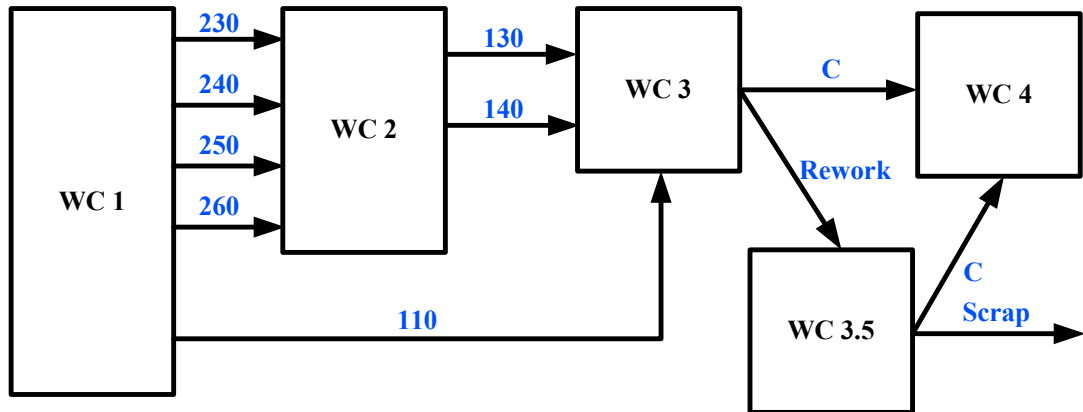
Figure 3.6: Product Routings
Product A



Product B



Product C



Performance Measures

The impact of MAS alternatives on manufacturing performance is evaluated using both internal and external criteria. The internal system performance measurement is the average throughput time and capacity utilization rate. The primary external system performance measure is net operating income.

Demand Fulfillment Rate is the primary customer service measure used in this study. It is defined as the percentage of demand for the three individual products that is accepted for production. It has been used in similar simulation studies (Lea & Min 2003, Lea & Fredendall 2002). This represents one of two non-financial measures for manufacturing performance and the primary market measure used in this study.

Cycle-Time is the primary measure of success for a time-based competitor (Koufteros 1998). In the case of a time-based competitor, the resource usage to be minimized is time itself – throughput time in manufacturing terms. Throughput time is the ultimate measure of performance for the time-based competitor, and that is determined by how well an organization is able to move materials to, through, and off the shop floor into finished goods inventory and how well that organization is able to minimize finished goods (Fry et al. 1993). Time-based manufacturing is an externally focused production system that emphasizes quick response to changing customer needs. Its primary purpose is to reduce end-to-end time in manufacturing (Stalk & Hout 1990, Blackburn 1991, Koufteros 1998).

Net operating income (NOI) is utilized often in the accounting literature to measure performance (Bakke & Hellberg 1991, Low 1992, Dugdale & Jones 1996, Lea & Fredendall 2002, Lea & Min 2003). Regardless of industry type or manufacturing

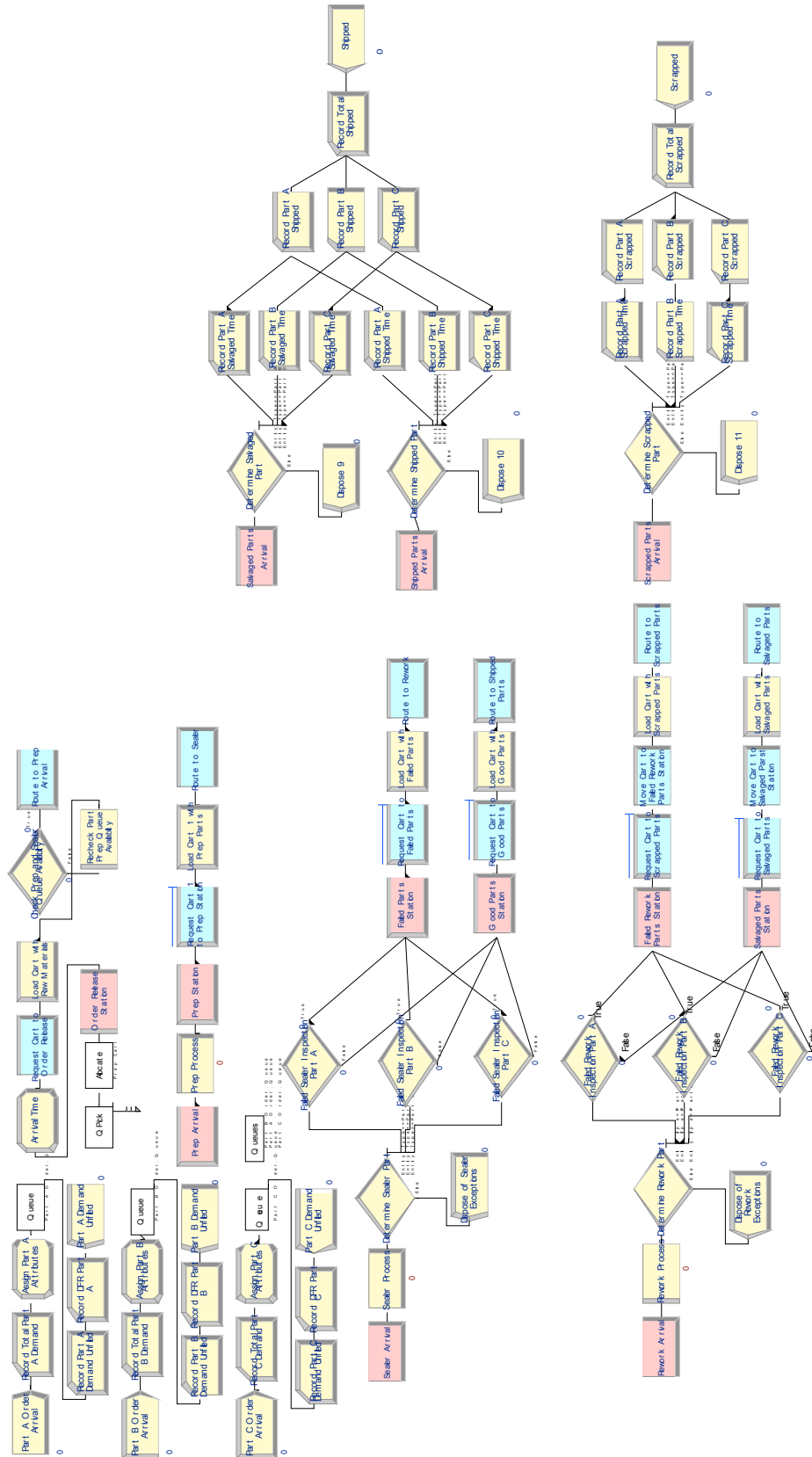
environment, NOI is the bottom line for most organizations and certainly for publicly-held companies. For the purposes of this study, regardless of MAS alternative, NOI is defined as revenues or sales from operations for the accounting period less all operating expenses, including CGS and SG&A. As this study will simulate a make-to-order shop, i.e. with no stocking on finished goods inventory, there is no concern for short-term timing differences between the various MAS alternatives under examination.

Chapter Four

Simulation Design

This study will utilize a modified version of ARENA's existing "Electronic Assembly and Test System with Part Transfers" as a baseline model. The final model used for data generation is shown in figure 4.1 below. This system represents the final operations of the production of different sealed electronic units. The arriving parts are cast metal cases that have already been machined to accept the electronic parts. In the prep area, the cases are machined to ensure a good seal, after which they are deburred, cleaned, and transferred to the sealer. At the sealer operation the electronic components are inserted, the case is assembled and sealed, and the sealed unit is then sent to bet quality control tested. At quality control testing the finished part either passes directly to finished goods to be shipped or is rejected and rerouted to the rework station. After rework, the part is again tested to ensure quality and is either passed, routed to finished goods inventory, and shipped or rejected for a second time and scraped.

Figure 4.1: Simulation Model



The decision logic sub model in ARENA will utilize the following maximization formulation, which includes all constraints for the resources and market demand, in order to determine optimal product mix for the master production schedule:

$$\text{Maximize } Z = \sum_{j=1}^n c_j^{l,k} x_j$$

$$Z = \sum_{j=1}^n a_{ij} x_j \leq b_i \quad i = 1, 2, 3, \dots, m \quad (\text{Resource/ Capacity Constraint})$$

$$x_j \leq d_j \quad \text{For every } j, j = 1, 2, 3, \dots, n \quad (\text{Market Demand Constraint})$$

$$x_j \geq 0$$

Where:

x_j - is the number of product j produced

b_i - is the maximum amount of resource i available

d_j - is the market demand for product j

a_{ij} - is the amount of resource i required to produce product j

$c_j^{l,k}$ - is the contribution margin of product j , with complexity k , under MAS 1

With $m + n$ constraints for this model

Model Assumptions

Based on other simulation studies discussed in Chapter 2, and specifically on the

Krawjewski and others (1987) study, the following assumptions are necessary:

1. No preemption of jobs once work has begun
2. No alternative routings
3. Zero setup times
4. Jobs are not split in the shop. All jobs are moved to the next work center or buffer area when the current work center operation is complete.
5. No backorders. Demand that cannot be filled is lost to the perfectly competitive market.
6. The first work center is never starved for work because raw material supply is not constrained.

Technical Details of Simulation Model

Most simulation models can be classified as either terminating or steady state. This is primarily a methodological issue of intent or goal of the study rather than an issue of internal model logic or construction. Kelton and others (2002) define these two classifications of simulation models according to its fitness for purpose. A terminating simulation is one in which the model dictates specific starting and stopping conditions as a natural reflection of how the target system actually operates.

For the purposes of this study, on the other hand, steady-state determination is used for analyzing the results. A steady-state simulation is one in which the quantities to be estimated are defined in the long-run, i.e. over a theoretically infinite time frame. People often assume that a long-run, steady-state simulation is the thing to do. This largely depends on whether the starting and stopping conditions are part of the essence of the research model, in which case a terminating analysis is probably more appropriate.

In principle, the initial conditions of the simulation are not of interest to this research, but even a steady-state simulation has to stop at some point. Each simulation run is in fact truncated, i.e. the simulation run does end. N.B., there is no 'natural' event that occurs within the simulation model to signal the end of one run and the beginning of another. Rather each simulation run has been programmed to end after 3 years. Buffer size between stations is unbounded, i.e. there is no blocking or balking within the system that may cause bias in certain performance measures.

Steady-state determination addresses one weakness, from an analytical perspective. This is the fact that each simulation run begins in the 'empty' or 'idle' state and this may cause statistical bias especially during the transient priming process. From

an analytical perspective, it is preferable to initiate the collection of statistical data only after the transition from the priming process to a steady state has been accomplished. According to Pritsker (1986), unless the startup behavior is a focus of the research, the analysis of data under a steady state determination leads to an improved mean estimate for performance measures.

According to Kelton and others (2002), there are several ways to achieve a steady-state determination. One cumbersome method would be to place entities in the system at time 0 and start the model. Another, maybe unrealistic, method would be to run the model for so long that any bias at the beginning is overwhelmed by the amount of later data. However, a more commonly used procedure for steady state determination of a stochastic process is observing the behavior of the system and only begin statistical accumulation of data after a certain warm-up period. Work center queue lengths are often used as an indicator of steady-state operation, because queue lengths will continue to increase or decrease before reaching a steady state. Several pilot runs may be executed to establish the steady state for each experimental combination. Fortunately, ARENA has a built-in option for steady-state analysis using a specified warm-up period. Every replication of the model starts as it did before – independent and identical, all statistical accumulators are cleared, and the performance metrics only reflect what happened after the warm-up period ended.

Replications and Variance Reduction

One of the key advantages of simulation modeling is the ease with which experimental conditions may be repeated or reproduced. Replication of the simulation model run is

used to capture the variance of dependent variable means. The number of replications can be estimated via a formula provided by Pritsker (1986) based on a 90% confidence interval for the variance of sample means.

$$I = [(t_{\alpha/2, I-1} S_x) / g]$$

Where:

- I - number of independent replications
- $t_{\alpha/2, I-1}$ - t value with I-1 degrees of freedom
- S_x - sample standard deviation of the dependent variable
- g - half-width confidence interval for the sample mean

The g can be specified in relative terms of σ_x , that is, let $g = v\sigma_x$ for any $v > 0$. In this case, I can be computed with knowledge of σ_x (Pritsker 1986: 754). A 90% confidence interval is desired such that μ_x is within $(\bar{X}_I - 0.8\sigma_x, \bar{X}_I + 0.8\sigma_x)$. Thus, the above equation requires that at least 6 replications be performed, when in fact this study performs 60 replications.

Verification and Validation

Simulation remains a relatively new field for researchers, and unfortunately standards and consensus amongst researchers on the question of verification and validation (V&V) of simulation models remains a fundamental, albeit emergent aspect. In the words of Elmaghraby (1968), “It is well to remember the dictum that nobody solves the problem. Rather, everybody solves the model that she [or she] has constructed of the problem”.

According to Law and Kelton (2000: 265,275), one must always keep in mind with any simulation study that the simulation model itself can only be an approximation to an actual system, no matter how much effort is spent on model building. In fact,

Increasing the validity of a model beyond a certain level might be quite expensive, since extensive data collection may be required, but might not lead to significantly better insight or decisions... There are several techniques for increasing a simulation model's validity and credibility, which include building the model based on existing theory, relevant results from similar simulation studies, and the experience and intuition of the simulation modeler.

Regarding the experience and intuition of the modeler, as Law and Kelton point out, the United States Department of Defense was an early adopter and innovator of simulation techniques. This included strategic analyses of the complex interactions of men, technologies, and tactics in future battles against enemies with technologies and tactics yet to be developed. It is therefore often necessary to build hypotheses based on the researchers experience or intuition, to be substantiated later in a simulation study.

The whole process of simulation modeling research has elements of art as well as science, as articulated by the title of one of the first published text books on simulation *The Art of Simulation* (Tocher 1963). According to Balci (1989),

Simulation modeling is an art. Give a set of objectives, if ten economists are asked to build a simulation model of the U.S. economy, each one will come up with a model which will produce a different set of results. The differences in the results are considered normal and as expected under the paradigm of the art of modeling... Modeling is an artful balancing of opposites. On the one hand, a model should not exclude the essential elements of the system, and on the other hand, it should not include unnecessary details. Missing an essential element may invalidate model representation. Inclusion of unnecessary details would only make the model unnecessarily complex and difficult to analyze. A model is an abstraction of the reality and it is built for a specific purpose. The level of representativeness of a model must be judged with respect to that purpose.

While there are no perfect solutions to V&V, there is increasing agreement on the broader terminology. V&V in the operations research literature has come to be defined by one of the classic simulation textbooks by Law and Kelton (2000:299) now in its third edition since 1983: "Verification is determining that a simulation computer program performs as intended, i.e. debugging the computer program... Validation is concerned

with determining whether the conceptual simulation model (as opposed to the computer program) is an accurate representation of the system under study.” Validation determines whether the conceptual simulation model, as opposed to the computer program, is an accurate representation of the system under study. Model formulation is the process by which a conceptual model is envisioned to represent the system under study (Balci 1989).

The conceptual model is the model which is formulated in the mind of the modeler (Nance 1981). In a similar vein, Sargent (1991:38) states that:

The conceptual model is the mathematical/logical/verbal representation (mimic) of the problem entity developed for a particular study, and the computerized model is the conceptual model implemented on a computer. The conceptual model is developed through analysis and modeling phase, the computerized model is developed through a computer programming and implementation phase and inferences about the problem entity are obtained by conducting computer experiments on the computerized model in the experimentation phases.

Kelton and others (2002: 42-43) point out that while no simulation study will follow a cut-and-dried formula, there are few key aspects that need to be addressed:

- *Understand the system.* Whether it exists or not, the simulation analyst must have a down-to-earth feel for what is going on.
- *Be clear about your goals.* Understand what can be learned from the study and expect no more.
- *Formulate the model representation.* What level of detail is appropriate to the goals of the study.
- *Verify the model.* A faithful representation of the conceptual model.
- *Validate the model.* While statistical tests can be carried out here, a good dose of common sense is also valuable.
- *Document what you have done.*

Kleijnen (1995) surveys the literature from several fields using simulation methodologies in research in order to document various approaches to V&V. There are several techniques commonly used in verification of simulation models, and – while individually none is perfect – the author points out that general good programming

practice can go a long way in verifying the computer model performs as intended.

Borrowing largely from software engineering research, the author suggests: (1) modular programming, (2) checking intermediate simulation outputs through tracing and statistical testing, (3) comparing final simulation results with analytical results, (4) animation.

Assessment of accuracy (i.e. verification and validation) must be done right after completing each phase of a simulation study (Balci 1989). One of the powers of ARENA is that it has the built-in diagnostic capability for verification that can be applied at each stage of the model development. Through this capability, this study applies several of the verification techniques described above to ensure the simulation program performs as intended. Modular programming is used to model this research, utilizing an existing ARENA model as the baseline for this study. In addition, ARENA has the built-in diagnostic capability to self-check intermediate simulation outputs ensuring that all modules are performing as intended.

According to Kelton and others (2002), one easy verification method using ARENA is to allow only a single entity to enter the system and follow that entity to be sure that the model logic and data are correct. This is easily accomplished using the 'step' feature found as a 'run' option on the tool bar. This allows the simulation analyst to advance step-by-step through the simulation modules. The authors also suggest what they term 'performance estimation'. Borrowing the logic of the engineering slide rule, which in an era before calculators and computers represented a relatively quick way for the engineer to check the reasonability of estimates, performance estimation essentially entails logically predicting what the model will do and comparing this to how one entity advances through the system.

Multifaceted and multidisciplinary knowledge and experience are required for a successful simulation study (Balci 1989). Kleijnen makes special note of the power of animation in debugging the program. Face validation is useful as a preliminary approach to validation (Baci 1989). The simulation analyst is most familiar with the corresponding conceptual model, which means they can often easily detect programming errors through dynamic displays. Because ARENA has a very powerful and easy to use animation capability, animation is used to debug the simulation module by module and throughout the simulation assuring face validation. Intermediate simulation outputs are also checked through computer tracing, manual calculation, and statistical testing, module by module, to ensure that orders are created exactly as planned, and that the flow of orders through the shop is correct. When this is done, actual utilization rates are equal to the calculated utilization rates.

Validation of a simulation model poses a unique problem for many researchers. Since a model is an abstraction of reality, we cannot talk about absolute accuracy (Balci 1989). Validation can never be assumed to result in a perfect model, since the perfect model would be the real system itself and, by definition, any model is a simplification of reality (Kleijnen 1995). Kelton and others (2002:119) emphasize that this may be nearly impossible even in instances where a real system exists for comparison.

Although verification can be very difficult, complete validation (the next activity) can sometimes be almost impossible. That's because validation implies that the simulation is behaving just like the real-world system, which may not even exist. And even if the system does exist, you have to have output performance data from it, as well as convince yourself and other nonbelievers that your model can really capture and predict the events of the real system.

The clear advantages of using a simulation model are that the normal transience present in a real manufacturing environment can be eliminated along with exogenous

events and that the same data collection and analysis procedures can be used for each replication. Holding all exogenous inputs constant, several replications of a stochastic model are made to determine the amount of stochastic variability in the model. The unexplained variance between these replications would provide a measure of internal validity (Hermann 1967, Balci 1989). Therefore the internal validity of the simulation model can be assured without further testing.

Hypothesis and construct validation is achieved through careful literature review and analysis to ensure that the treatment effect being measured is caused by experimental factors. Kleijnen (1991) reminds the simulation analyst that the model need only be 'good enough' as determined by the goals of the model. "For example," the author states, "some applications need only relative, not absolute, simulation responses corresponding to different scenarios." Again, based on the goals of this research to directly compare different MAS treatments in the same manufacturing environment, a relatively simple simulation model will be adequate for obtaining meaningful data.

According to Balci (1989) sometimes due to the lack of data, we may not be able to characterize and input variable or parameter. In this case, a heuristic procedure such as one based on a triangular or beta probability distribution (Law & Kelton 2000) may be used. Kleijnen (1991) agrees, as the system concept implies that the analyst must subjectively decide on the boundary of that system and the attributes to be quantified in the model. He points out that sometimes it is difficult, impossible, or simply undesirable to obtain relevant data, such as in simulation studies of nuclear war, quite fortunately. In such cases, the analyst may show logically that that exact values of input data are not

critical. Moreover, in some cases the marginal value of available ‘real world’ data is *de minimus* in terms of the goals of the study.

Borrowing from all fields of scientific research, Karplus (1983) classifies mathematical models (including simulation models) as ranging from black box (non-causal) model in the social sciences through grey box models in ecology to white box (causal) models in physical sciences. Balci (1989) classifies simulation models broadly as either self-driven or trace-driven. A self-driven (distribution-driven or probabilistic) simulation model is the one which is driven by input values obtained via sampling from probability distributions using random numbers. A trace-driven (or retrospective) simulation model, on the other hand, is driven by input sequences derived from trace data obtained through measurement of the real system.

Again, the uniqueness of this study is that it takes a causal (normative) approach in proposing and testing the impact of a new time-based MAS alternative and then tests the hypothesized relationships with a self-driven simulation model. The author points out that a typical aspect of many white box simulation studies is that the conceptual models are based on common sense and understand of real systems. Two additional techniques proposed by Kleijnen (1991) in validating causal simulation models are animation for face validity (discussed earlier also as a model verification technique) and sensitivity analysis for construct validity.

Models and sub models with unobservable inputs can be subjected to sensitivity analysis in order to determine whether the model’s behavior agrees with the judgment and knowledge of the simulation analyst (Kleijnen 1991). Unexpected effects may reveal invalidity (Balci 1989). Sensitivity analysis is a systematic investigation of the reaction

of model outputs to drastic changes in model inputs and model structure. An often ignore aspect of simulation studies, the magnitudes of the sensitivity estimates show which inputs are important (Law & Kelton 2000:144). The input/output behavior of the model may be approximated through the meta model:

$$y_i = \beta_0 + \sum_{k=1}^K \beta_k x_{ik} + \sum_{k=1}^{K-1} \sum_{k'=k+1}^K \beta_{kk'} x_{ik} x_{ik'} + e_i$$

Where:

- y_i - denotes the simulation response in replication i
- i - the number of simulation replications, $i = 1, \dots, n$
- k - simulation input, $k = 1, \dots, K$. of the K simulation
- x_{ik} - the value of simulation input k in combination with i
- β_k - the main or first order effect of input k
- $\beta_{kk'}$ - the interaction between inputs k and k'
- e_i - the approximation error in run i

So this simulation model is valid within a certain area of its inputs only; the area defined as a K -dimensional hypercube formed by the K input ranges. Within that area the simulation model's input/output behavior may vary. The first order meta model presented above is a good approximation of the input/output behavior.

Statistical conclusion validity determines whether the sample size is large enough to detect a treatment effect, and whether a desired alpha level is obtained (Cook & Campbell 1979). As discussed earlier, if the simulation model is carefully programmed and controlled, all treatments and replications are implemented in the same way and independently, the sample size is computed to be large enough to detect treatment effect, then the statistical conclusion validity of the simulation model can be obtained. As concluded by Kleijnen (1991), regardless of developments vis-à-vis V&V techniques, simulation will remains both an art as well as a science.

Chapter Five

Results

This chapter presents the results and statistical analyses of the data collected in the ARENA simulation experiment described previously in Chapter 3. The initial data were downloaded into Excel and then uploaded into SPSS for statistical analysis. After screening the data for missing data and outliers, Multivariate Analysis of Variance (MANOVA) was performed to determine whether or not a factor and/or its interaction is statistically significant in determining overall performance. The results were further analyzed using a more detailed Univariate Analysis of Variance (ANOVA) post-hoc tests. The remainder of this chapter is organized as follows: the first section presents the raw data collection and descriptive statistics, the second section presents the assumption testing for MANOVA, the third section presents the results of MANOVA and individual ANOVA, the fourth section presents a discussion of the hypothesis testing, and the fifth section discusses the results by experimental factor and its implications for management accounting research and practice.

Raw Data and Descriptive Statistics

As demonstrated in Appendix B, the product costs were first determined in the simulation model by using different management accounting alternatives: traditional costing

systems, activities-based costing, and throughput accounting. The product cost data were then input into the integer linear programming (ILP) model to determine the optimal product mix, which was then input into the simulation model for use in the product mix decision. Average performance data were collected for 60 replications of 30 days each for 27 experimental condition groups, representing three different management accounting systems (traditional costing systems, activities-based costing, and time-based accounting), three levels of manufacturing overhead (low, medium, high), and three levels of product mix complexity (low, medium, high) for a total of 1620 data points. Figure 5.1 below show the number of observations by experimental factor.

Figure 5.1: Total Number Between-Subjects Factors

		N
MAS	1	540
	2	540
	3	540
MOH	1	540
	2	540
	3	540
MIX	1	540
	2	540
	3	540

Data Screening and Assumption Tests

Prior to the actual multivariate statistical analysis, the data were screened and its quality assessed. According to Mertler and Vannatta (2002: 25), there are four main purposes for screening data prior to conducting a multivariate analysis. The first of these deals with the accuracy of the data collected, the second deals with missing data and the pattern of missing data, the third deals with assessing the effect of extreme values, i.e. outliers, and

finally the fit between the data and the assumptions of the specific procedure must be assessed. Because the data were generated through an ARENA simulation model and manually entered into an Excel spreadsheet for sorting and financial calculations uploading into SPSS, the possibility of researcher error in transferring the data exists. The raw data uploaded into SPSS can be seen in Appendix C. Figure 5.2 below shows that there were in fact no missing data at the time of the initial upload into SPSS. For each dependent variable there are exactly 1620 observations.

Figure 5.2: Case Processing Summary

	Cases					
	Valid		Missing		Total	
	N	Percent	N	Percent	N	Percent
DFR_2	1620	100.0%	0	.0%	1620	100.0%
CT_2	1620	100.0%	0	.0%	1620	100.0%
NOI_2	1620	100.0%	0	.0%	1620	100.0%

A visual review of the data prior to uploading into SPSS revealed no missing data, and any unrealistic values were checked against the original ARENA data reports and corrected as necessary. With regards to the accounting calculation for net operating income (NOI), there were occasional offsetting extreme values between replications (accounting periods) due simply to timing differences. In these instances, which numbered no more than three instances per experimental condition group, the extreme values were replaced with the average net value of the two points.

Multivariate outliers consist of unusual combinations of scores on two or more variables and are often subtle and more difficult to detect than univariate outliers. Therefore, the univariate outliers were identified for each group using box plots and stem and leaf plots. Univariate outliers are defined as cases with unusual or extreme values one or both ends of the a sample distribution. There are three fundamental causes for

outliers: 1) data entry errors were made by the research, 2) the subject is not a member of the population for which the sample is intended, or 3) the subject is simply different from the remainder of the sample (Tabachnick & Fidell 1996).

It is important to note that both ANOVA and MANOVA are robust to moderate violations of normality, provided the violation is created by skewness and not by outliers (Tabachnick & Fidell 1996). The real danger of outliers is that they can significantly distort the results of statistical tests, due to the fact that many statistical procedures rely on squared deviations from the mean (Aron & Aron 1997). Therefore, an observation falling far from the rest of the distribution mean could potentially exert a great deal of influence on the results of the statistical test. A single outlier, if extreme enough, according to Mertler and Vannatta (2002: 27), could influence a false significance or insignificance as well as seriously affect the values of correlation coefficients.

Appendix B presents the results of the univariate outlier screening for each dependent variable within each group. Univariate outliers can be detected by means of graphical methods (Tabachnick & Fidell 1996). Since the number of outlying cases for each variable in each group was fairly small, i.e. less than 5 in all groups, and the sample size is relatively large, i.e. 60 replications, the outliers can either be deleted or altered to a value that is within the extreme value of the tail of the accepted distribution (Mertler & Vannatta 2002: 40). In order to ensure the equality of sample size between experimental condition groups, and robustness to minor violations of normality and homoscedasticity, the latter option was chosen.

There are three general assumptions of multivariate statistical testing. The first of these assumptions is that of a normal sample distribution. Prior to examining

multivariate normality, one should first assess univariate normality (Mertler & Vannatta 2002: 30). Multivariate normality is a difficult concept to describe, much less to test. Suffice to say, normality on each of the variable separately is a necessary condition for multivariate normality to hold (Stevens 1996: 245). According to Mertler & Vannatta (2002: 31) characteristics of multivariate normality include:

1. Each of the individual variables is normally distributed;
2. Any linear combination of the variables is normally distributed; and
3. All subsets of the set of variables, i.e. every pairwise combination, have a multivariate normal distribution (bivariate normality).

Moreover, because data were collected for 60 replications for each of the 27 experimental groups (3 experiment factors with 3 levels each), there are a total of 1,620 data points utilized for this analysis. With equal or unequal samples sizes and only a few DVs, a sample size of 20 in the smallest cell should be sufficient to ensure robustness to violations of univariate and multivariate normality (Mertler & Vannatta 2002: 124). Therefore, given equal sample sizes of 60 in each group, normality may be assumed under the central limit theorem.

Univariate normality refers to the extent to which all observations in the sample for a given variable in a given group are distributed normally. Among the non-graphical test that can be used are the chi-square goodness of fit and the Kolmogorov-Smirnov test. The chi-square test suffers from the defect of depending on the number of intervals used for the grouping. Therefore, the Kolmogorov-Smirnov statistic with Lilliefos significance level was utilized to test univariate normality for each dependent variable in each group. The Kolmogorov-Smirnov statistic tests the null hypothesis the population is normally distributed and an associated significance level serves as an indication that the variable is not normally distributed (Mertler & Vannatta 2002: 30). Appendix C presents

the Kolmogorov-Smirnov test statistics for each variable in each experimental condition group, with insignificance in all cases indicating normality of distributions.

The second assumption, linearity, presupposes that there is a straight line relationship between any two variables. It is a critical assumption in multivariate analyses due to the fact that many of the techniques are based on linear combinations of the variables. The Pearson correlation coefficient (r) is the most commonly used bivariate correlation technique, measuring the association between two quantitative variables. Figure 5.3 below shows significance of this measure for all bivariate combinations of the dependent variables, indicating a significant linear relationship.

Figure 5.3: Correlations

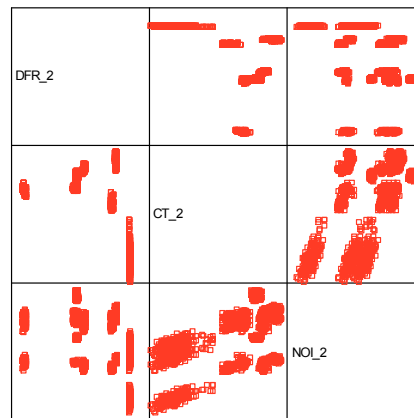
		DFR_2	CT_2	NOI_2
DFR_2	Pearson Correlation	1	-.578**	-.406**
	Sig. (2-tailed)	.	.000	.000
	N	1620	1620	1620
CT_2	Pearson Correlation	-.578**	1	.590**
	Sig. (2-tailed)	.000	.	.000
	N	1620	1620	1620
NOI_2	Pearson Correlation	-.406**	.590**	1
	Sig. (2-tailed)	.000	.000	.
	N	1620	1620	1620

** . Correlation is significant at the 0.01 level (2-tailed).

The final assumption of homoscedasticity is that the variability in scores for one continuous variable will be roughly the same across all values of another continuous variable. This concept is analogous to the univariate assumption of homogeneity of variance. Homoscedasticity is closely related to the assumption of normality, because if the assumption of multivariate normality is met, two variables must be homoscedastic (Tabachnic & Fidell 1996). Although subjective in nature, homoscedasticity is sometimes best assessed through the examination of bivariate scatterplots.

Figure 5.4 below presents the bivariate scatterplots for the three dependent variables. The output for the three dependent variables indicates a non-elliptical shapes between DFR_2 and the other two variables CT_2 and NOI_2. The bivariate scatterplots between CT_2 and NOI_2, on the other hand, show a somewhat elliptical pattern. Since the use of bivariate scatterplots is fairly subjective in examining linearity (Mertler & Vannatta 2002: 55), we will not place reliance on this test. However, reliance can be placed on the Pearson's correlation coefficients above, indicating that a linear relationship does indeed exist.

Figure 5.4: Bivariate Scatterplots



In multivariate cases, homoscedasticity may be assessed statistically using Box's M test for equality of variance-covariance matrices. This test evaluates the hypothesis that covariance matrices are equal, and if the observed significance level for the Box's M test is small, i.e. $p < .05$, one should reject H_0 . Highly sensitive to violations of normality, Box's Test should be interpreted with caution (Mertler & Vannatta 2002: 125).

It is important to note that violations of homoscedasticity, similar to a violation of homogeneity in univariate analysis, will not prove fatal to the analysis so long as the linear relationship between the variables can still be established (Tabachnick & Fidell

1996, Kennedy & Bush 1985). Wilk's Lambda is the most commonly reported MANOVA statistic; however, Pillai's Trace statistic is most often used in instances where homogeneity of variance-covariance is in question (Mertler & Vannatta 2002: 125). Box's test in figure 5.5 below is significant, so Pillai's Trace statistic will be used in evaluating the multivariate tests.

Figure 5.5: Box's Test of Equality of Covariance Matrices

Box's M	5136.282
F	32.294
df1	156
df2	1173890
Sig.	.000

Tests the null hypothesis that the observed covariance matrices of the dependent variables are equal across groups.

a. Design: Intercept+MAS+MOH+MIX+MAS * MOH+MAS * MIX+MOH * MIX+MAS * MOH * MIX

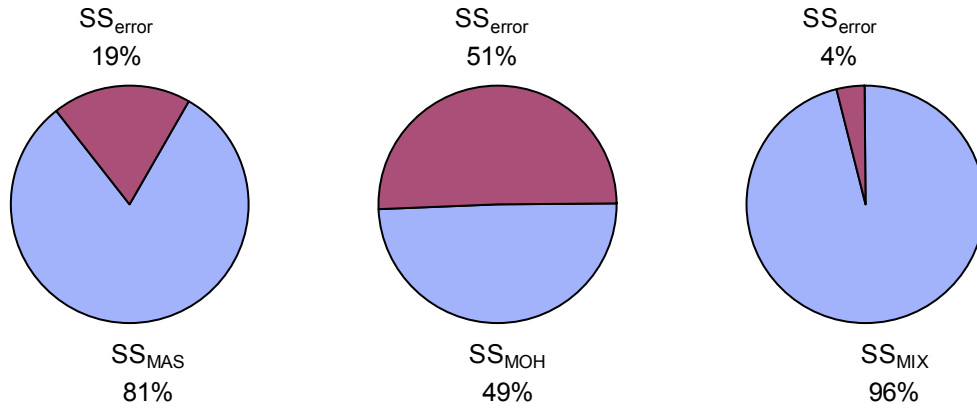
MANOVA Results

The collected experimental data were first analyzed using a factorial MANOVA procedure. This analysis is meant to determine if the combination of dependent variables – the performance measures discussed in Chapter 3: demand fulfillment rate (DFR_2), average cycle time (CT_2), and net operating income (NOI_2) – is significantly effected by the independent variables. The experimental factors, discussed in Chapter 3, include management accounting system (MAS), product mix complexity (MIX), and manufacturing overhead levels (MOH). As shown in figure 24 below, the treatment effects are all significant as are all the bivariate interactions. Moreover, the effect sizes are generally very high.

Measures of effect size in MANOVA and ANOVA are measures of the degree of association between the effect, either the main effect or any interactions, and the dependent variable(s). It is the proportion of variance in the dependent variable that is attributable to each effect. There are several commonly used measures for effect size, the most common being Eta Squared (η^2) and Partial Eta Squared (η_p^2). One of the problems with η^2 is that the values of each effect are dependent upon the number of other effects and the magnitude of those effects. Partial Eta Squared presents an alternative computation of Eta Squared for each individual effect (Tabachnick & Fidell 1989). Partial Eta Squared is defined as: $\eta_p^2 = SS_{\text{effect}} / (SS_{\text{effect}} + SS_{\text{error}})$, and is a standard output in SPSS.

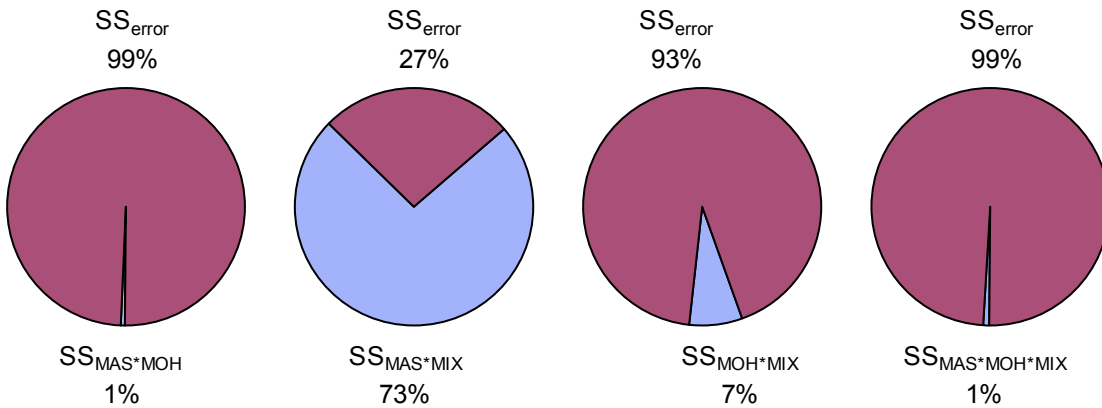
It should be noted that sums of η_p^2 values are not additive, i.e. they do not sum the amount of dependent variable variance accounted for by the independent variables, and therefore it is possible for the sum of η_p^2 values to be greater than one. The η_p^2 values presented below in figure 5.6 clearly show high effect size for all three experimental factors (main effects), especially for management accounting system and product mix complexity, which explains 81% and 96% of the variability in the dependent variable combination respectively. Manufacturing overhead level was associated with 49% of the variability in the dependent variable combination. Although it is low when compared with the other two main effects, it still shows a high relationship.

Figure 5.6: Partial Eta Squared Values for MAS, MOH, and MIX Effects



The η_p^2 values presented below in figure 5.7 clearly show high effect size for the two-way interaction of management accounting system and product mix complexity and a significant, albeit it rather low, effect size for manufacturing overhead level and product mix complexity. The amount of variance in the dependent variable combination explained by these interactions was 73% and 7% respectively. The two-way combination of management accounting system and manufacturing overhead level as well as the three-way interaction of management accounting system, manufacturing overhead level, and product mix complexity was insignificant with less than 1% in effect size.

Figure 5.7: Partial Eta Squared Values for MAS, MOH, and MIX Interaction Effects



MANOVA results in figure 5.8 below indicate that management accounting system (Pillai's Trace=1.62, $F(6, 3184)=2268.712$, $p=.000$, $\eta_p^2=.810$), manufacturing overhead level (Pillai's Trace=.984, $F(6, 3184)=514.306$, $p=.000$, $\eta_p^2=.492$), and product mix complexity (Pillai's Trace=1.925, $F(6, 3184)=13603.070$, $p=.000$, $\eta_p^2=.962$) significantly affect the combined DV of demand fulfillment rate, average cycle time, and net operating income. In addition, the bivariate combinations of management accounting system and manufacturing overhead levels (Pillai's Trace=0.019, $F(12, 4779)=2.52$, $p=.000$, $\eta_p^2=.006$), management accounting system and product mix complexity (Pillai's Trace=2.20, $F(12, 4779)=1095.489$, $p=.000$, $\eta_p^2=.733$), and manufacturing overhead level and product mix complexity (Pillai's Trace=0.220, $F(12, 4779)=31.495$, $p=.000$, $\eta_p^2=.073$) are all found to significantly affect the combined DV of demand fulfillment rate, average cycle time, and net operating income. However, multivariate effect sizes are small for the combinations of management accounting system and manufacturing overhead level as well as the combination of manufacturing overhead level and product mix complexity. The three-way interaction of management accounting system, manufacturing overhead level, and product mix complexity were not found to have a significant affect on the combined DV of demand fulfillment rate, average cycle time, and net operating income.

Figure 5.8: Multivariate Tests

Effect		Value	F	Hypothesis df	Error df	Sig.	Partial Eta Squared
Intercept	Pillai's Trace	1.000	23118783.276 ^a	3.000	1591.000	.000	1.000
	Wilks' Lambda	.000	23118783.276 ^a	3.000	1591.000	.000	1.000
	Hotelling's Trace	43592.929	23118783.276 ^a	3.000	1591.000	.000	1.000
	Roy's Largest Root	43592.929	23118783.276 ^a	3.000	1591.000	.000	1.000
MAS	Pillai's Trace	1.621	2268.712	6.000	3184.000	.000	.810
	Wilks' Lambda	.001	16101.286 ^a	6.000	3182.000	.000	.968
	Hotelling's Trace	370.873	98281.457	6.000	3180.000	.000	.995
	Roy's Largest Root	369.217	195931.096 ^b	3.000	1592.000	.000	.997
MOH	Pillai's Trace	.984	514.306	6.000	3184.000	.000	.492
	Wilks' Lambda	.016	3703.096 ^a	6.000	3182.000	.000	.875
	Hotelling's Trace	62.719	16620.609	6.000	3180.000	.000	.969
	Roy's Largest Root	62.719	33283.011 ^b	3.000	1592.000	.000	.984
MIX	Pillai's Trace	1.925	13603.070	6.000	3184.000	.000	.962
	Wilks' Lambda	.000	83260.044 ^a	6.000	3182.000	.000	.994
	Hotelling's Trace	1872.499	496212.261	6.000	3180.000	.000	.999
	Roy's Largest Root	1860.086	987085.721 ^b	3.000	1592.000	.000	.999
MAS * MOH	Pillai's Trace	.019	2.552	12.000	4779.000	.002	.006
	Wilks' Lambda	.981	2.566	12.000	4209.682	.002	.006
	Hotelling's Trace	.019	2.579	12.000	4769.000	.002	.006
	Roy's Largest Root	.019	7.702 ^b	4.000	1593.000	.000	.019
MAS * MIX	Pillai's Trace	2.200	1095.489	12.000	4779.000	.000	.733
	Wilks' Lambda	.001	5231.569	12.000	4209.682	.000	.913
	Hotelling's Trace	213.949	28342.331	12.000	4769.000	.000	.986
	Roy's Largest Root	210.259	83735.577 ^b	4.000	1593.000	.000	.995
MOH * MIX	Pillai's Trace	.220	31.495	12.000	4779.000	.000	.073
	Wilks' Lambda	.780	34.513	12.000	4209.682	.000	.079
	Hotelling's Trace	.282	37.330	12.000	4769.000	.000	.086
	Roy's Largest Root	.282	112.208 ^b	4.000	1593.000	.000	.220
MAS * MOH * MIX	Pillai's Trace	.020	1.343	24.000	4779.000	.122	.007
	Wilks' Lambda	.980	1.350	24.000	4614.985	.118	.007
	Hotelling's Trace	.021	1.358	24.000	4769.000	.114	.007
	Roy's Largest Root	.020	4.049 ^b	8.000	1593.000	.000	.020

a. Exact statistic

b. The statistic is an upper bound on F that yields a lower bound on the significance level.

c. Design: Intercept+MAS+MOH+MIX+MAS * MOH+MAS * MIX+MOH * MIX+MAS * MOH * MIX

Univariate ANOVA and Scheffé post hoc tests were conducted as follow-up tests.

ANOVA results indicate that demand fulfillment rate differs significantly for management accounting system ($F(2, 1593)=290159.67, p=.000, \eta_p^2=.997$), product mix complexity ($F(2, 1593)=1471806.2, p=.000, \eta_p^2=.999$), and the two-way interaction of management accounting system and product mix complexity ($F(2, 1593)=82837.12, p=.000, \eta_p^2=.995$). Average cycle-time differs significantly for management accounting system ($F(2, 1593)=960.591, p=.000, \eta_p^2=.547$), product mix complexity ($F(2,$

1593)=20756.710, $p=.000$, $\eta_p^2=.963$), and the two-way interaction of management accounting system and product mix complexity ($F(2, 1593)=591.132$, $p=.000$, $\eta_p^2=.597$).

Net operating income differs significantly for management accounting system ($F(2, 1593)=1704.381$, $p=.000$, $\eta_p^2=.682$), manufacturing overhead level ($F(2, 1593)=31768.716$, $p=.000$, $\eta_p^2=.976$), and product mix complexity ($F(2, 1593)=20449.024$, $p=.000$, $\eta_p^2=.963$); the two-way interactions of management accounting system and manufacturing overhead level ($F(2, 1593)=5.061$, $p=.000$, $\eta_p^2=.013$), management accounting system and product mix complexity ($F(2, 1593)=679.384$, $p=.000$, $\eta_p^2=.630$), and manufacturing overhead level and product mix complexity ($F(2, 1593)=71.264$, $p=.000$, $\eta_p^2=.152$); and moderately in the three-way interaction of management accounting system, manufacturing overhead level, and product mix complexity ($F(2, 1593)=2.49$, $p=.011$, $\eta_p^2=.012$).

As expected, manufacturing overhead level had an amplification effect and only significantly affected the performance measure of net operating income. As shown in figure 5.9 below, post-hoc Scheffé tests show significant differences between the three levels of manufacturing overhead and net operating income. This effect presents some interesting implications for management accounting practice, which will be discussed in greater detail in the final section of this chapter. As well, the amplification effect of manufacturing overhead level can be seen on the charts of cumulative net operating income at the very end of this chapter.

Figure 5.9: Net Operating Income Post-Hoc Test

Scheffe^{a,b,c}

MOH	N	Subset		
		1	2	3
3	540	68.81071		
2	540		92.52141	
1	540			98.55130
Sig.		1.000	1.000	1.000

Means for groups in homogeneous subsets are displayed.

Based on Type III Sum of Squares

The error term is Mean Square(Error) = 4.201.

- a. Uses Harmonic Mean Sample Size = 540.000.
- b. The group sizes are unequal. The harmonic mean of the group sizes is used. Type I error levels are not guaranteed.
- c. Alpha = .05.

Manufacturing overhead level did not have a significant impact on demand fulfillment rate ($F(2, 1593)=.038, p=.962, \eta_p^2=.000$) or average cycle-time ($F(2, 1593)=.014, p=.986, \eta_p^2=.000$), nor do any of its interactions significantly affect demand fulfillment rate or average cycle-time. The two-way interactions of management accounting system and manufacturing overhead level have an insignificant impact on demand fulfillment rate ($F(2, 1593)=.056, p=.994, \eta_p^2=.000$) and average cycle-time ($F(2, 1593)=.006, p=1.000, \eta_p^2=.000$). The interactions of manufacturing overhead level and product mix complexity also have an insignificant effect on demand fulfillment rate ($F(2, 1593)=.012, p=1.00, \eta_p^2=.000$) and average cycle-time ($F(2, 1593)=.005, p=1.000, \eta_p^2=.000$). Finally, the three-way interactions of management accounting system, manufacturing overhead level, and product mix complexity had an insignificant affect on demand fulfillment rate ($F(2, 1593)=.057, p=1.00, \eta_p^2=.000$) and average cycle-time ($F(2, 1593)=.008, p=1.000, \eta_p^2=.000$). Figure 5.10 below presents the summary of the between-subjects effects for this model.

Figure 5.10: Test of Between Subjects Effects

Source	Dependent Variable	Type III Sum of Squares	df	Mean Square	F	Sig.	Partial Eta Squared
Corrected Model	DFR_2	62.838 ^a	26	2.417	148280.04	.000	1.000
	CT_2	45262856.302 ^b	26	1740879.089	1761.510	.000	.966
	NOI_2	465888.411 ^c	26	17918.785	4264.884	.000	.986
Intercept	DFR_2	1053.052	1	1053.052	64607878	.000	1.000
	CT_2	466080178.4	1	466080178.361	471603.69	.000	.997
	NOI_2	12157090.348	1	12157090.348	2893532.4	.000	.999
MAS	DFR_2	9.459	2	4.729	290159.66	.000	.997
	CT_2	1898679.810	2	949339.905	960.591	.000	.547
	NOI_2	14321.816	2	7160.908	1704.381	.000	.682
MOH	DFR_2	1.254E-06	2	6.272E-07	.038	.962	.000
	CT_2	27.157	2	13.578	.014	.986	.000
	NOI_2	266950.628	2	133475.314	31768.716	.000	.976
MIX	DFR_2	47.978	2	23.989	1471806.2	.000	.999
	CT_2	41027206.507	2	20513603.254	20756.710	.000	.963
	NOI_2	171831.926	2	85915.963	20449.024	.000	.963
MAS * MOH	DFR_2	3.661E-06	4	9.152E-07	.056	.994	.000
	CT_2	23.795	4	5.949	.006	1.000	.000
	NOI_2	85.047	4	21.262	5.061	.000	.013
MAS * MIX	DFR_2	5.401	4	1.350	82837.117	.000	.995
	CT_2	2336834.240	4	584208.560	591.132	.000	.597
	NOI_2	11417.652	4	2854.413	679.384	.000	.630
MOH * MIX	DFR_2	7.801E-07	4	1.950E-07	.012	1.000	.000
	CT_2	21.054	4	5.264	.005	1.000	.000
	NOI_2	1197.652	4	299.413	71.264	.000	.152
MAS * MOH * MIX	DFR_2	7.380E-06	8	9.225E-07	.057	1.000	.000
	CT_2	63.738	8	7.967	.008	1.000	.000
	NOI_2	83.690	8	10.461	2.490	.011	.012
Error	DFR_2	2.596E-02	1593	1.630E-05			
	CT_2	1574342.494	1593	988.288			
	NOI_2	6692.942	1593	4.201			
Total	DFR_2	1115.916	1620				
	CT_2	512917377.2	1620				
	NOI_2	12629671.701	1620				
Corrected Total	DFR_2	62.864	1619				
	CT_2	46837198.796	1619				
	NOI_2	472581.353	1619				

a. R Squared = 1.000 (Adjusted R Squared = 1.000)

b. R Squared = .966 (Adjusted R Squared = .966)

c. R Squared = .986 (Adjusted R Squared = .986)

Hypothesis Testing

The specific research questions and corresponding statistical hypotheses were posited previously in Chapter 3. In this section, the results of the univariate testing above are further summarized in relation to these statistical hypotheses. These hypotheses along were as follows:

H₀₁: The use of different managerial accounting systems has no effect on average throughput-time, demand fill rate, and/or net operating income.

$$\text{MAS}_a = 0$$

As shown in figure 5.10 above, the main factor for the management accounting system was found to significantly affect all three manufacturing performance measures. Therefore it is necessary to reject **H₀₁**.

H₀₂: Manufacturing overhead level has no effect on average throughput-time, demand fill rate, and/or net operating.

$$\text{MOH}_o = 0$$

As shown in figure 5.10 above, the main factor for the manufacturing overhead level was found to significantly affect net operating income, but not the other two manufacturing performance measures. Therefore it is necessary to accept **H₀₁** as it relates to demand fulfillment rate and average cycle time and to reject **H₀₁** in regards to net operating income.

H₀₃: The use of different managerial accounting systems has no effect on average throughput-time, demand fill rate, and/or net operating income for a given level of manufacturing overhead.

$$\mathbf{MAS}_a * \mathbf{MOH}_o = \mathbf{0}$$

As shown in figure 5.10 above, the interaction for the management accounting systems and manufacturing overhead level was found to significantly affect net operating income, but not the other two manufacturing performance measures. Therefore it is necessary to accept **H₀₁** as it relates to demand fulfillment rate and average cycle time and to reject **H₀₁** in regards to net operating income.

H₀₄: Product mix complexity has no effect on average throughput-time, demand fill rate, and/or net operating income.

$$\mathbf{MIX}_m = \mathbf{0}$$

As shown in figure 5.10 above, the main factor for product mix complexity was found to significantly affect all three manufacturing performance measures. Therefore it is necessary to reject **H₀₁**.

H₀₅: The use of different managerial accounting systems has no effect on average throughput-times, demand fill rates, and/or net operating income for a given level of product mix complexity.

$$\mathbf{MAS}_a * \mathbf{MIX}_m = \mathbf{0}$$

As shown in figure 5.10 above, the interaction for the management accounting systems and product mix complexity was found to significantly affect all three manufacturing performance measures. Therefore it is necessary to reject **H₀₁**.

H₀₆: Manufacturing overhead level has no effect on average throughput-time, demand fill rate, and/or net operating income for a given product mix complexity.

$$\mathbf{MOH}_o * \mathbf{MIX}_m = \mathbf{0}$$

As shown in figure 5.10 above, the interaction for the manufacturing overhead level and product mix complexity was found to significantly affect net operating income, but not the

other two manufacturing performance measures. Therefore it is necessary to accept **H₀₁** as it relates to demand fulfillment rate and average cycle time and to reject **H₀₁** in regards to net operating income.

H₀₇: The use of different managerial accounting systems has no effect on average throughput- time, demand fill rate, and/or net operating income for a given level of product mix complexity and a given manufacturing overhead level.

$$\mathbf{MAS_a * MOH_o * MIX_m = 0}$$

As shown in figure 5.10 above, the interaction for the management accounting system, manufacturing overhead level, and product mix complexity was found to significantly affect net operating income, but not the other two manufacturing performance measures. Therefore it is necessary to accept **H₀₁** as it relates to demand fulfillment rate and average cycle time and to reject **H₀₁** in regards to net operating income.

Practical Implications

Because the primary focus of this study is to examine the impact of different management accounting systems – traditional costing systems (TCS), activities-based costing (ABC), and time-based accounting (TBA) – on manufacturing performance in the context of a time-based competitive environment, it is necessary to take a more detailed look at this impact on each individual performance measure. The three performance measures were chosen because they represent both internal and external and financial and non-financial measures of performance. Figure 5.11 below presents a summary of the results in performance measures by management accounting system alternative.

Figure 5.11: Multiple Comparisons by MAS

Scheffe

Dependent Variable	(I) MAS	(J) MAS	Mean Difference (I-J)	Std. Error	Sig.	95% Confidence Interval	
						Lower Bound	Upper Bound
DFR_2	1	2	-.16799907*	.000245697	.000	-.16860104	-.16739710
		3	-.15545861*	.000245697	.000	-.15606058	-.15485664
	2	1	.16799907*	.000245697	.000	.16739710	.16860104
		3	.01254046*	.000245697	.000	.01193849	.01314244
	3	1	.15545861*	.000245697	.000	.15485664	.15606058
		2	-.01254046*	.000245697	.000	-.01314244	-.01193849
CT_2	1	2	-29.211410*	1.91319765	.000	-33.89883996	-24.52397938
		3	53.468745*	1.91319765	.000	48.78131487	58.15617545
	2	1	29.211410*	1.91319765	.000	24.52397938	33.89883996
		3	82.680155*	1.91319765	.000	77.99272454	87.36758512
	3	1	-53.468745*	1.91319765	.000	-58.15617545	-48.78131487
		2	-82.680155*	1.91319765	.000	-87.36758512	-77.99272454
NOI_2	1	2	-6.3592155*	.124743740	.000	-6.66484388	-6.05358703
		3	-.10501750	.124743740	.702	-.41064593	.20061092
	2	1	6.35921545*	.124743740	.000	6.05358703	6.66484388
		3	6.25419795*	.124743740	.000	5.94856952	6.55982637
	3	1	.10501750	.124743740	.702	-.20061092	.41064593
		2	-6.2541979*	.124743740	.000	-6.55982637	-5.94856952

Based on observed means.

*. The mean difference is significant at the .05 level.

Demand fulfillment rate represents an external (market) non-financial measure of manufacturing performance. It represents the percentage of demand that is ultimately fulfilled by the production system. As presented in figure 5.12 below, the highest performance in terms of this measure was activities-based costing (MAS_2) with a rate of 86.6% of demand filled and time-based accounting (MAS_3) with 85.4% of demand filled. The worst performance was traditional costing systems (MAS_1) with 69.8% of demand filled. Although the difference between ABC and TBA in terms of demand fulfillment rate was statistically significant, from a practical perspective, this difference may not justify the high cost of implementing an ABC system.

Figure 5.12: Demand Fulfillment Rate by MAS

Scheffe^{a,b,c}

MAS	N	Subset		
		1	2	3
1	540	.69842651		
3	540		.85388512	
2	540			.86642558
Sig.		1.000	1.000	1.000

Means for groups in homogeneous subsets are displayed.

Based on Type III Sum of Squares

The error term is Mean Square(Error) = 1.630E-05.

- a. Uses Harmonic Mean Sample Size = 540.000.
- b. The group sizes are unequal. The harmonic mean of the group sizes is used. Type I error levels are not guaranteed.
- c. Alpha = .05.

As discussed at length in Chapters 1 and 2, the primary non-financial measure of success for a time-based manufacturer is cycle-time, or the total time from receipt of an order to the shipment of the product to the customer. Reducing cycle-time is the primary focus of time-based competition, and is therefore a key internal measure of success. As presented in figure 5.13 below, the best performance was TBA with an average cycle time of 491.00 minute. The second best performance along this key measure was TCS with an average cycle-time of 544.47 minutes, followed by ABC with an average cycle-time of 573.68 minutes.

Figure 5.13: Cycle-Time by MAS

Scheffe^{a,b,c}

MAS	N	Subset		
		1	2	3
3	540	490.9973		
1	540		544.4661	
2	540			573.6775
Sig.		1.000	1.000	1.000

Means for groups in homogeneous subsets are displayed.

Based on Type III Sum of Squares

The error term is Mean Square(Error) = 988.288.

- a. Uses Harmonic Mean Sample Size = 540.000.
- b. The group sizes are unequal. The harmonic mean of the group sizes is used. Type I error levels are not guaranteed.
- c. Alpha = .05.

Net operating income represents the primary internal measure of financial performance. As presented in Figure 5.14 below, the best performance in terms of this performance measure was ABC with an average net operating income of 90.83 (thousands) per accounting period (replication). It should be noted that TBA performed slightly better than TCS over the long run, 84.58 and 84.47 respectively, but the difference was not statistically significant.

Figure 5.14: Net Operating Income by MAS

Scheffe^{a,b,c}

MAS	N	Subset	
		1	2
1	540	84.47306	
3	540	84.57808	
2	540		90.83228
Sig.		.702	1.000

Means for groups in homogeneous subsets are displayed.

Based on Type III Sum of Squares

The error term is Mean Square(Error) = 4.201.

- a. Uses Harmonic Mean Sample Size = 540.000.
- b. The group sizes are unequal. The harmonic mean of the group sizes is used. Type I error levels are not guaranteed.
- c. Alpha = .05.

As shown in the figure 5.15 (Tests of Between-Subjects Effects) above, product mix complexity and its combination with management accounting system has a significant affect on all three of the performance measures. As summarized below in figure 31, product mix complexity has a significant impact on all three performance measures.

Figure 5.15: Multiple Comparisons by Product Mix Complexity

Scheffe

Dependent Variable	(I) MIX	(J) MIX	Mean Difference (I-J)	Std. Error	Sig.	95% Confidence Interval	
						Lower Bound	Upper Bound
DFR_2	1	2	.15082359*	.000245697	.000	.15022161	.15142556
		3	.41631103*	.000245697	.000	.41570906	.41691300
	2	1	-.15082359*	.000245697	.000	-.15142556	-.15022161
		3	.26548745*	.000245697	.000	.26488548	.26608942
	3	1	-.41631103*	.000245697	.000	-.41691300	-.41570906
		2	-.26548745*	.000245697	.000	-.26608942	-.26488548
CT_2	1	2	-352.30321*	1.9131977	.000	-356.990639	-347.615779
		3	-320.63793*	1.9131977	.000	-325.325364	-315.950503
	2	1	352.30321*	1.9131977	.000	347.6157787	356.9906393
		3	31.665276*	1.9131977	.000	26.97784547	36.35270605
	3	1	320.63793*	1.9131977	.000	315.9505030	325.3253636
		2	-31.665276*	1.9131977	.000	-36.35270605	-26.97784547
NOI_2	1	2	-18.783746*	.124743740	.000	-19.08937396	-18.47811711
		3	-23.975737*	.124743740	.000	-24.28136503	-23.67010817
	2	1	18.783746*	.124743740	.000	18.47811711	19.08937396
		3	-5.1919911*	.124743740	.000	-5.49761949	-4.88636264
	3	1	23.975737*	.124743740	.000	23.67010817	24.28136503
		2	5.19199106*	.124743740	.000	4.88636264	5.49761949

Based on observed means.

*. The mean difference is significant at the .05 level.

As presented in figure 5.16 below, product mix complexity has a significant affect on the demand fulfillment rate measure. Average demand fulfillment rate was 99.5% under a low level of product mix complexity and drops to 84.5% under medium level and 57.9% under a high level of product mix complexity.

Figure 5.16: Demand Fulfillment Rate by MIX Level

Scheffe^{a,b,c}

MIX	N	Subset		
		1	2	3
3	540	.57897957		
2	540		.84446702	
1	540			.99529061
Sig.		1.000	1.000	1.000

Means for groups in homogeneous subsets are displayed.

Based on Type III Sum of Squares

The error term is Mean Square(Error) = 1.630E-05.

- a. Uses Harmonic Mean Sample Size = 540.000.
- b. The group sizes are unequal. The harmonic mean of the group sizes is used. Type I error levels are not guaranteed.
- c. Alpha = .05.

As presented in figure 5.17 below, product mix complexity has a significant affect on the average cycle-time measure. Average cycle-time was 312.1 minutes under a low level of product mix complexity and increases to 632.7 minutes under medium level and 664.4 minutes under a high level of product mix complexity.

Figure 5.17: Cycle-Time by MIX Level

Scheffe^{a,b,c}

MIX	N	Subset		
		1	2	3
1	540	312.0666		
3	540		632.7045	
2	540			664.3698
Sig.		1.000	1.000	1.000

Means for groups in homogeneous subsets are displayed.

Based on Type III Sum of Squares

The error term is Mean Square(Error) = 988.288.

- a. Uses Harmonic Mean Sample Size = 540.000.
- b. The group sizes are unequal. The harmonic mean of the group sizes is used. Type I error levels are not guaranteed.
- c. Alpha = .05.

As presented in figure 5.18 below, product mix complexity has a significant affect on net operating income. Average net operating income was 72.37 (thousands) under a low level of product mix complexity and increases to 91.16 under medium level and 96.35 under a high level of product mix complexity.

Figure 5.18: Net Operating Income by MIX Level

Scheffe^{a,b,c}

MIX	N	Subset		
		1	2	3
1	540	72.37465		
2	540		91.15839	
3	540			96.35038
Sig.		1.000	1.000	1.000

Means for groups in homogeneous subsets are displayed.
Based on Type III Sum of Squares
The error term is Mean Square(Error) = 4.201.

- a. Uses Harmonic Mean Sample Size = 540.000.
- b. The group sizes are unequal. The harmonic mean of the group sizes is used. Type I error levels are not guaranteed.
- c. Alpha = .05.

Summary of Research Results

This study applied a simulation modeling methodology to examine the impact of different management accounting system alternatives, manufacturing overhead levels, and product mix complexity levels on manufacturing performance measures. The manufacturing performance measures examined included both internal and external as well as financial and non-financial measures of success. These measures were demand fulfillment rate, cycle time, and net operating income. Figure 5.19 below summarizes the results of this study in terms of these three manufacturing performance measures by management accounting system alternative and combined weighted score.

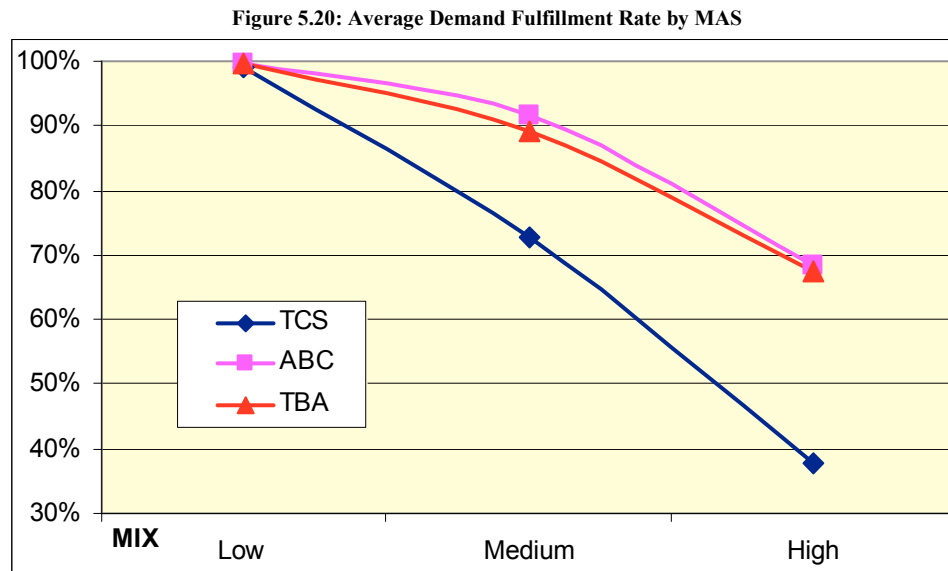
Figure 5.19: Summary of MAS Performance by Experimental Condition Group

MOH Level	MIX Level	Performance Measure											
		Demand Fulfillment Rate			Cycle Time			Net Operating Income			Combined Weighted Score (Maxium 6)		
Low	Low	1	ABC	99.8%	1	TBA	304.91	1	ABC	86.188	1	ABC	5
		2	TBA	99.6%	2	ABC	305.13	2	TCS	85.660	2	TBA	3
		3	TCS	99.2%	3	TCS	326.38	3	TBA	85.603	3	TCS	2
	Medium	1	ABC	91.6%	1	TBA	549.88	1	ABC	105.922	1	ABC	4
		2	TBA	89.1%	2	TCS	698.46	2	TCS	101.416	2	TBA	3
		3	TCS	72.6%	3	ABC	745.55	3	TBA	101.405	3	TCS	2
	High	1	ABC	68.5%	1	TCS	608.89	1	ABC	115.412	1	ABC	4
		2	TBA	67.5%	2	TBA	619.20	2	TBA	103.579	2	TBA	3
		3	TCS	37.7%	3	ABC	670.13	3	TCS	101.771	3	TCS	2
Medium	Low	1	ABC	99.8%	1	TBA	304.91	1	ABC	78.087	1	ABC	5
		2	TBA	99.6%	2	ABC	305.13	2	TCS	77.803	2	TBA	3
		3	TCS	99.2%	3	TCS	325.38	3	TBA	77.480	3	TCS	1
	Medium	1	ABC	91.6%	1	TBA	548.21	1	ABC	100.462	1	ABC	4
		2	TBA	89.1%	2	TCS	698.46	2	TCS	95.799	2	TBA	3
		3	TCS	72.6%	3	ABC	745.55	3	TBA	95.319	3	TCS	2
	High	1	ABC	68.5%	1	TCS	608.89	1	ABC	112.319	1	ABC	4
		2	TBA	67.5%	2	TBA	619.15	2	TBA	98.462	2	TBA	3
		3	TCS	37.7%	3	ABC	670.13	3	TCS	96.620	3	TCS	2
High	Low	1	ABC	99.8%	1	TBA	304.91	1	ABC	53.781	1	ABC	5
		2	TBA	99.6%	2	ABC	305.46	2	TCS	53.507	2	TBA	3
		3	TCS	99.2%	3	TCS	326.38	3	TBA	53.258	3	TCS	1
	Medium	1	ABC	91.6%	1	TBA	548.88	1	ABC	76.283	1	ABC	4
		2	TBA	89.1%	2	TCS	698.46	2	TCS	72.467	2	TBA	3
		3	TCS	72.6%	3	ABC	745.89	3	TBA	71.352	3	TCS	2
	High	1	ABC	68.5%	1	TCS	608.89	1	ABC	89.038	1	ABC	4
		2	TBA	67.5%	2	TBA	618.94	2	TCS	74.866	2	TCS	3
		3	TCS	37.7%	3	ABC	670.13	3	TBA	74.744	3	TBA	2

The combined weighted score is a composite measure of the three primary manufacturing performance measures, whereby two points are assigned to the best performing management accounting system, one point to the second best performance, no points to the worst performance. Therefore a perfect score of 6 would indicate that the management accounting system scored the highest along all three manufacturing performance measures. As can be seen in figure 5.19 above, no single management

accounting system excelled across all three measures, indicating that each alternative has its own limitations in terms of performance that must be considered in decision making. This is an important point to note, especially for management accounting practitioners.

As can be seen in figure 5.20 below, all three management accounting alternatives performed nearly equally well when the product mix complexity (MIX) was low. As product mix complexity increased, all three saw a decrease in demand fulfillment rate. However, the falloff in demand fulfillment rate occurred at a far greater rate under traditional costing systems as compared to the two other management accounting system alternatives. Although activities-based costing performed the best across all levels of product mix complexity, time-based accounting performed nearly as well along this crucial customer service measure.

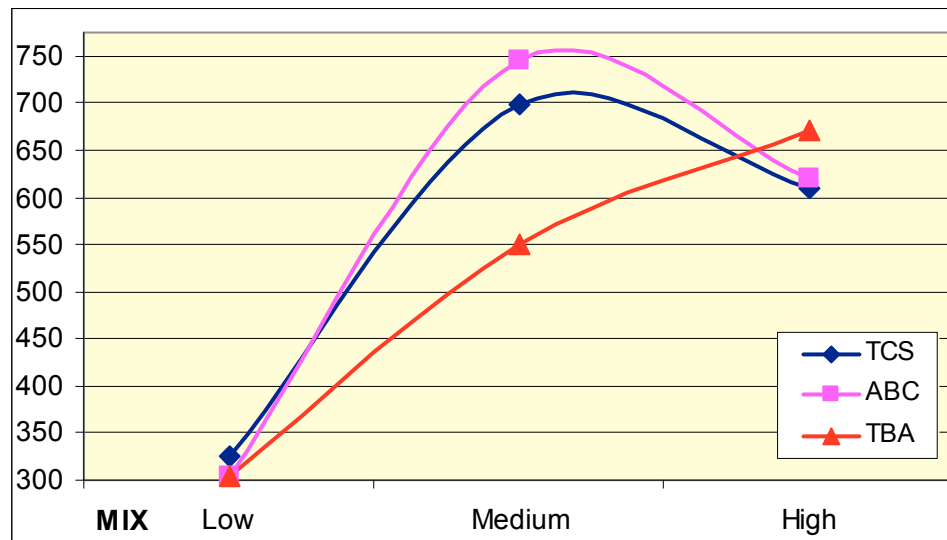


Because a major focus of this study was to examine the impact of management accounting system alternatives within the context of today's increasingly time-based competitive environment, the internal manufacturing performance measure of cycle time is of primary importance. As discussed in previous chapters, cycle-time is the primary

success measure for a time-based competitor. In terms of this strategic measure, time-based accounting performed the best at nearly all setting of product mix complexity.

Time-based accounting drove a product mix decision that better balanced the manufacturing line and resulted in the lowest average cycle-times for all products. It is interesting to note that activities-based costing, which generally outperformed vis-à-vis the other two manufacturing performance measures, was least effective in terms of cycle-times. It is important to note that the variability of cycle-times across the various levels of product mix complexity was much less than the variability under the traditional costing and activities-based costing systems. This may have important implications for the time-based manufacturer that is concerned with consistently delivering faster cycle-times under varying levels of product mix complexity demanded by the market.

Figure 5.21: Average Cycle-Time (Minutes) by MAS



Net operating income is the only financial measure of manufacturing success included in this study, and an argument could certainly be made that it is the bottom line and the most important measure. Figures 5.22 through 5.24 present the average net operating income measures for the various management accounting system alternatives

under differing levels of product mix complexity demand and differing levels of manufacturing overhead. Activities-based costing clearly outperformed the two other management accounting system alternatives along this measure. Traditional costing systems and time-based accounting performed nearly equally well under low and medium demand settings for product mix complexity. As the product mix complexity increases; however, traditional costing systems begin to fall behind time-based accounting.

Figure 5.22: Average Net Operating Income by MAS
(Low Manufacturing Overhead Level)

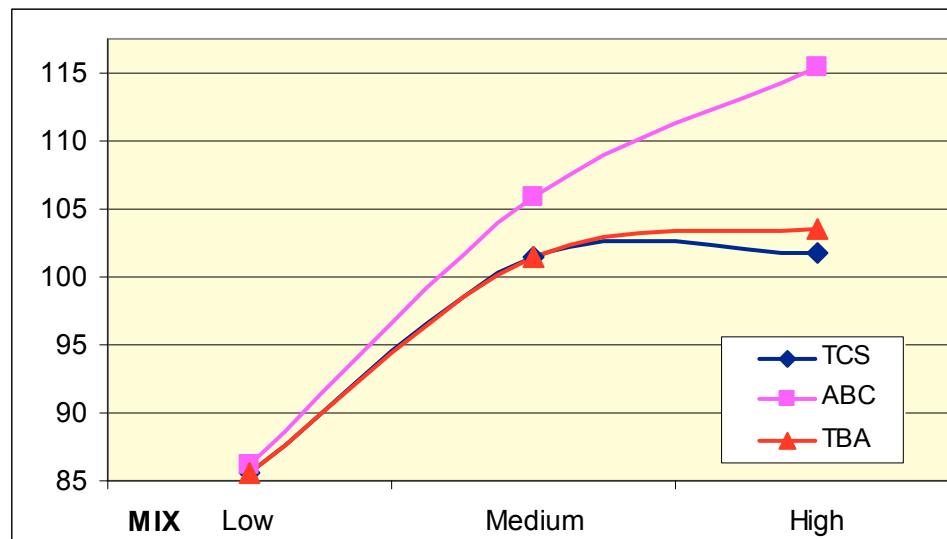


Figure 5.23 below shows essentially the same results, with activities-based costing clearly outperforming the other two management accounting system alternatives. The difference between traditional costing systems and time-based accounting is not as great under medium levels of product mix complexity but increases with high levels of product mix complexity.

Figure 5.23: Average Net Operating Income by MAS
(Medium Manufacturing Overhead Level)

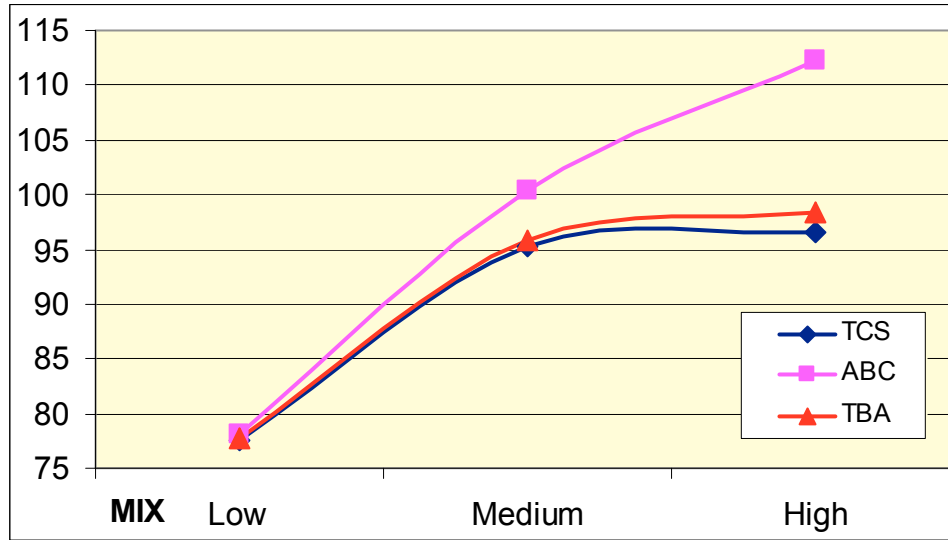
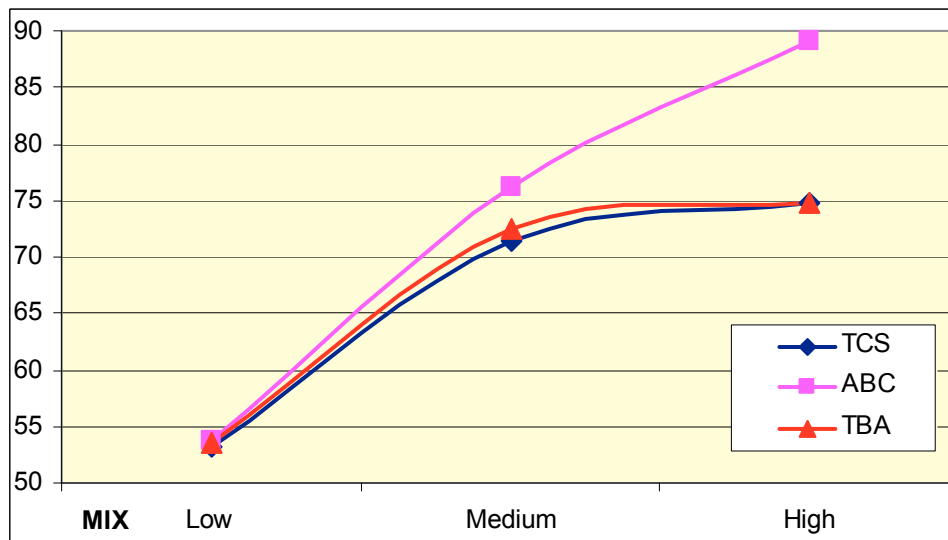


Figure 5.24 again shows very similar results, with activities-based costing clearly outperforming the other two management accounting system alternatives. Overall, average net operating income is at its lowest given the higher levels of manufacturing overhead. The difference between traditional costing systems and time-based accounting again is not as great under medium levels of product mix complexity but increases with high levels of product mix complexity.

Figure 5.24: Average Net Operating Income by MAS
(High Manufacturing Overhead Level)



The results in the figures above present particularly interesting implications for management accounting practice. The increase of demand for more complex and higher priced products presents an opportunity for increased revenues. However, as discussed in Chapter 2, it often presents the paradox as these products may also drive higher overall manufacturing costs. Higher levels of manufacturing overhead had no significant effect on the product mix decision; however, total costs and differences between the various management accounting system alternatives are amplified. As the manufacturing overhead level setting increases, the slope of the cumulative net operating income curve decreases. The implication for both management accounting researchers and practitioners is that the choice of management accounting system alternative becomes increasingly important as product mix complexity increases and may be amplified as manufacturing overhead levels increase.

As can be seen in the following figures (D.1-D.9) in Appendix D, higher levels of product mix complexity drive increasing long-term variances in cumulative net operating income. Review of management accounting system performance under the nine experimental conditions (three levels of manufacturing overhead by three levels of product mix complexity) shows no significant difference in cumulative net operating income when product mix complexity is low. Activities-based costing begins to significantly outperform the other two management accounting system alternatives at a medium demand setting for product mix complexity. This difference becomes more pronounced as product mix complexity is set at a high level. At this high setting, time-based accounting begins to slowly outperform traditional costing systems.

Chapter Six

Conclusions

The primary focus of this study was to evaluate the direct impact of different managerial accounting system alternatives on product mix decisions and ultimately on different manufacturing performance measures under differing levels of manufacturing overhead with differing levels of product mix complexity. The first chapter had posed the research questions and presented an experimental research framework in which to answer these questions. In the second chapter, the relevant literature on managerial accounting and operations management was reviewed as well as a select group of simulation studies that had included managerial accounting as a variable of interest. The third chapter provided the justification for utilizing a simulation research methodology and further developed the variables to be used in the research model. The fourth chapter outlined the specific details of the simulation model itself, while the fifth chapter provided the statistical analyses (MANOVA) of the simulation data as well as a summary of the research findings. This is followed by a discussion of implications for both managerial accounting research and practice. This chapter will conclude by illuminating some of the limitations of this particular study and how these can be addressed in future studies. The final section presents several opportunities to both to advance and extend the findings of this study in future research.

Limitations of Current Study

It is important to remember that every research methodology has its own unique set of strengths and corresponding limitations, and simulation modeling is no exception to this rule. Probably the greatest strength of simulation modeling is the model itself is virtually endlessly reconfigurable and therefore may be relatively easily extended and improved to incorporate more detail. The principal limitation is that no simulation model can possibly capture the infinite number of extraneous variables that exist within any real system.

This study represents an initial step towards better understanding the impact and interactions of different managerial accounting system alternatives, manufacturing overhead levels, and product mix complexity on manufacturing performance. Although it provides a number of interesting results for managerial accounting researchers to consider, it is important to remember that this study presents only a single operating environment. Therefore, the results of the study may not necessarily be generalizable across all manufacturing environments. As has been stated before in the Chapter 3 discussion of methodologies, the results of any simulation study are greatly impacted by the assumptions built into the model and must be interpreted with caution. However, the benefit of being able to observe the behavior of the performance measures under the same environmental settings is the major benefit of simulation modeling, and may provide insight and guidance for future research.

As mentioned above, one specific limitation of this study was that it considered only one particular simulated manufacturing environment, albeit under differing manufacturing overhead levels and with differing demand levels of product mix complexity. Traditional costing systems were used initially in the experiment. Given the

three product structures used in this study, the average percentage of each product's cost content varied by product and experimental condition group. Product A had direct material of between 20% and 60%, direct labor from 16% to 26%, and manufacturing overhead from 15% to 63% of total product cost. Product B had direct material of between 60% and 78%, direct labor from 16% to 24%, and manufacturing overhead from 6% to 15% of total product cost. Product C had direct material of between 28% and 71%, direct labor from 16% to 24%, and manufacturing overhead from 6% to 56% of total product cost. Different industries will have differing cost structures that may have a significant impact on performance measures and the selection of managerial accounting system alternative. Future experiments should be conducted in a variety of operating environments to enhance the generalizability of the findings.

Any form of analytical modeling, be it mathematical or simulation, must make assumptions in order to arrive at a solution. This particular study assumed that there were no changes in products demand over the entire five year period. Another assumption, with great implications for competitive markets, is that lost demand had no effect on future demand distributions. Future studies may address this issue by building in feedback loops and dynamic learning of the market, i.e. the inability to fulfill a given market demand for a particular product will affect the future market demand for that same product. This is a particularly important point for the time-based manufacturer and the proposed time-based accounting system as the literature suggests that delivering products with greater speed presents opportunities for premium pricing and with increased demands (Blackburn 1991).

This study had only one external measure of manufacturing performance in terms of customer service levels defined as demand fulfillment rate. In reality this is certainly not the only measure of customer service typically used, and future studies may consider using measures of on-time delivery as well as other quantifiable factors such as contract commitments. It is possible that a particular manufacturing firm may have vastly different performance results for the various managerial accounting system alternatives when different measures of customer service are used. Future studies may want to incorporate multiple measures for customer service as well as differing levels of customer sensitivity to these various measures.

One particularly important limitation for practice is the lack of a measure of return-on-investment. While this study utilized net operating income (NOI) as the primary financial measure of performance, an established practice in the managerial accounting literature, this measure only makes sense under the assumption that all three managerial accounting alternatives require the same initial and ongoing cost of investment. This is particularly an important point when comparing existing traditional costing systems (TCS) to time-based accounting (TBA) and particularly to activities-based-costing (ABC). Many of the failures of ABC in the literature are linked to the high cost of implementation and ongoing costs of maintaining such a detailed accounting system. While ABC was shown to outperform TCS and TBA in many cases and along several of the performance measures, any rational manager would need to weigh the potential benefits against the potentially higher cost of implementation and maintenance.

Suggestions for Future Research

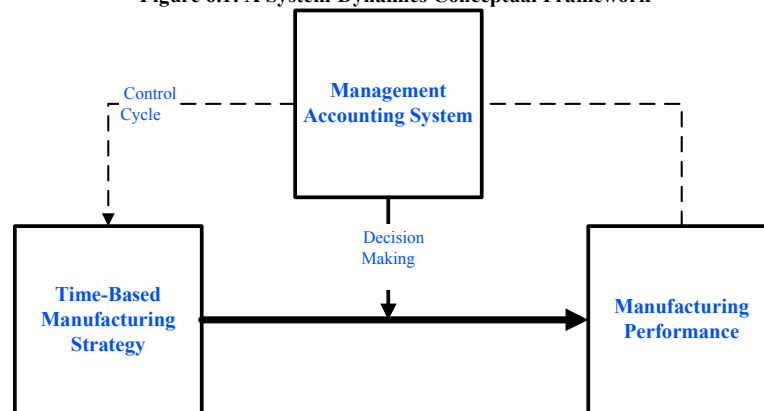
This research represents a first step towards integrating managerial accounting and operations management research. Because the scope of this study is somewhat limited, as outlined in the preceding discussion, further research will be needed to develop a complete understanding of the impact of managerial accounting systems on manufacturing performance. The following discussion proposes some possibilities for both advancing and extending this research.

As was mentioned above, one of the greatest strengths of simulation modeling is the malleability of the model itself, especially with newer software packages such as ARENA. The simulation model can be endlessly reconfigured to increase complexity and to incorporate additional realism. One suggestion is to take a systems dynamic approach building learning into the simulation model itself over the length of the simulation run. System dynamics is an approach to understanding the behavior of complex systems over time, and is increasingly finding application in management disciplines. System dynamics deals with internal feedback loops and time delays that affect the behavior of the entire system, and computer software is often used to simulate system dynamics models. System dynamics is very similar to systems thinking and constructs the same causal loop diagrams of systems with feedback. However, system dynamics typically goes further and utilizes simulation to study the behavior of systems and the impact of alternative policies. This type of systems dynamic model learning could be incorporated both on the supply process and demand sides.

Based on this literary evidence, a conceptual framework is presented in figure 6.1 to better illustrate the problem (Hutchinson 2007). In this framework, the MAS has a

moderating effect, via the decision making process, on the relationship between time-based strategy and manufacturing performance. Manufacturing managers, knowing the accounting measures by which their performance is evaluated, quite rationally game the system in order to maximize their personal utility, i.e. maximize their personal performance reward or to consolidate their power through obtaining greater allocations of corporate capital. Legacy cost systems have the tendency to encourage behavior that is incongruent with TBM strategy, e.g. manufacturing managers frequently attempt to lower average costs through longer production runs and producing to stock. A strategically congruent MAS, with a clear linkage between throughput time and product cost, will create incentive for manufacturing managers to focus on time compression in production, e.g. produce to market demands only.

Figure 6.1: A System-Dynamics Conceptual Framework



This is a system-dynamics framework, with a feedback loop from manufacturing performance to TBM strategy. System dynamics is an approach to studying complex systems, through the use of feedback loops (Forrester 1961). Stocks and flows are the basic building blocks, connected by feedback loops which create the nonlinearity found so frequently in modern day problems. In this example, manufacturing performance impacts time-based manufacturing strategy through the MAS, i.e. the control cycle. The

use of an appropriate MAS, which best reflects the time-based competitive reality, will reinforce the practices of TBM strategy over time. Conversely, the choice of inappropriate MAS, which does not reflect the importance of throughput-time, will undermine TBM strategy and may prove a fetter to its advancement.

On the supply side, it would be interesting to develop a product mix determination using dynamic integer goal programming as opposed to simply integer linear programming in a static environment. Manufacturing performance measures as driven by the various managerial accounting system alternatives can be fed into the goal program through a feedback loop, thereby continually driving change in product cost and product mix decision. The choice of managerial accounting system affects the product cost, which in turn affects the product mix decision, which affect manufacturing performance measures, which is fed back into the managerial accounting system itself.

From a demand perspective, incorporating learning into the different product demand distributions would add an additional level of realism. Given a competitive market, it is quite likely that any demand lost to the market may be permanently lost, i.e. a particular customer may never order again from a particular supplier. Breaking down product demands into individual customer demands, with differing levels of customer service requirements and differing sensitivities to stock outs and price increases would add a great deal of complexity and realism to the simulation model.

Future studies may also look at additional measures of financial and operational performance, such as return-on-investment capacity utilization rate. As mentioned above in the limitations, the decision to implement an alternative managerial accounting system will certainly be weighed against the cost of implementation and maintenance. Therefore

measures of return-on-investment become increasingly important as the fixed costs of implementing such control systems increases.

In terms of operational performance measures, capacity utilization has become increasingly important in industries with a heavy overhead burden. As is often said in industries with high fixed investments, such as automotive or aircraft, “capacity utilization is king”. This performance metric is easily calculated within ARENA and other simulation software programs as well.

For each resource, ARENA reports two utilization statistics, called scheduled utilization and simply utilization. To understand what these are and how they differ requires some mathematical notation. Let $B(t)$ be the number of units of a particular resource that are available at time t , and let $M(t)$ be the number of units of that resource that are available (busy or not) at time t . If the resource has a fixed capacity, then $M(t)$ is a fixed constant for all t , but if the resource capacity follows a variable schedule, then $M(t)$ will vary with t . Of course, $0 \leq B(t) \leq M(t)$ at all times t . If the resource is not available at time t , e.g. it has failed, then $M(t) = 0$, which then forces $B(t) = 0$. Let $U(t) = B(t)/M(t)$ whenever $M(t) > 0$; thus $0 \leq U(t) \leq 1$, and $U(t)$ represents what might be called instantaneous utilization of the resource.

What ARENA calls simply utilization of the resource is just the (time) average of this instantaneous utilization over the simulation where T is the length of the simulation:

$$\text{Utilization } U(t) = \frac{\int_0^T U(t) dt}{T} = \frac{1}{T} \int_0^T \frac{B(t)}{M(t)} dt$$

In words, utilization is the time average of the ratio of the number busy to the number available. ARENA also reports scheduled utilization, which is a ratio of the average number busy to the average number available:

$$\text{Scheduled Utilization} = \frac{\int_0^T B(t) dt/T}{\int_0^T M(t) dt/T} = \frac{\int_0^T B(t) dt}{\int_0^T M(t) dt}$$

While the preceding suggestions could rather easily be built into any simulation modeling study, another interesting extension of this research would be to model an actual manufacturing facility and apply the findings *post hoc* to the actual manufacturing system. This would provide a unique opportunity to combine positivist and normative research techniques within a single study, something that is rarely, if ever, seen within managerial accounting research. This would give the opportunity not only to collect real data for developing demand and process distributions, but would also add a new dimension to the kind of case study methodology often employed in managerial accounting research. The ability to suggest normal solutions to real accounting issues through simulation modeling would also help to increase the standing of managerial accounting research and further the development and dissemination of innovative managerial accounting practices. Of course crossing roles between managerial accounting research and consultant has the potential to bring up a unique host of issues to be addressed, but would still present an excellent opportunity if an actual company were willing to participate in such a project.

In the interest of furthering normative accounting research, there are many possibilities for incorporating other managerial accounting alternatives into future studies. With increasing recognition of the deficiencies of traditional costing systems,

and newly emerging competitive pressures from the market, new managerial accounting system alternatives are being proposed on many fronts. The focus of some industries on capacity utilization has driven many companies to adopt Resource Consumption Accounting, which has proven popular in Germany. Environmental pressures may drive the need to develop environmentally focused systems such as life-cycle costing. The growth and proliferation of enterprise information systems has brought the practice of back-flush costing into discussion as well. Regardless of what is driving changes in the market, this research has shown that managerial accounting systems can play a significant role in manufacturing success. However, this research has only partially addressed the knowledge gap at the interface of managerial accounting and manufacturing management; continued research is needed.

###

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Appendix A

Direct and Variable Costing Example

Example 1

Assume the following (per unit)

Direct Materials	2.5 lbs @ \$4.00	\$10.00
Direct Labor	.5 hr @ \$16.00	\$ 8.00
VOH	.5 hr @ \$4.00	\$ 2.00
FOH	\$40,000	\$ 2.50
Actual Output	16,000 units	
Variable S&A	\$6.00 per unit	
Fixed S&A	\$60,000	
Selling price	\$40	

What do the income statements look like if actual sales equal 16,000 units?

<u>Absorption Costing</u>		<u>Direct Costing</u>	
Revenue (40)(16000)	640,000	Revenue (40)(16000)	640,000
Cogs (22.50)(16000)	<u>360,000</u>	Vbl Mfg (20)(16000)	320,000
GM (17.50)(16000)	280,000	Vbl S+A (6)(16000)	<u>96,000</u>
Vbl S+A (6)(16000)	96,000	CM	224,000
Fx S+A	<u>60,000</u>	Fx Mfg	40,000
Profit	<u>124,000</u>	Fx S+A	<u>60,000</u>
		Profit	<u>124,000</u>

- Note: When sales equals production, profit under absorption costing and direct costing are equal.

Example 2

Assume sales of 12,000 units. What is the profit under each costing method?

<u>Absorption Costing</u>		<u>Direct Costing</u>	
Revenue (40)(12000)	480,000	Revenue (40)(12000)	480,000
Cogs (22.50)(12000)	<u>270,000</u>	Vbl Mfg (20)(12000)	240,000
GM (17.50)(12000)	210,000	Vbl S+A (6)(12000)	<u>72,000</u>
Vbl S+A (6)(12000)	72,000	CM (14)(12000)	168,000
Fx S+A	<u>60,000</u>	Fx Mfg	40,000
Profit	<u>78,000</u>	Fx S+A	<u>60,000</u>
		Profit	<u>68,000</u>

- Note: When production exceeds sales, absorption profit exceeds direct profit.

Example 3

Assume sales of 18,000 units. What is the profit under each costing method?

<u>Absorption Costing</u>		<u>Direct Costing</u>	
Revenue (40)(18000)	720,000	Revenue (40)(18000)	720,000
Cogs (22.50)(18000)	<u>405,000</u>	Vbl Mfg (20)(18000)	360,000
GM (17.50)(18000)	315,000	Vbl S+A (6)(18000)	<u>108,000</u>
Vbl S+A (6)(18000)	108,000	CM (14)(18000)	252,000
Fx S+A	<u>60,000</u>	Fx Mfg	40,000
Profit	<u>147,000</u>	Fx S+A	<u>60,000</u>
		Profit	<u>152,000</u>

- Note: When sales exceed production, direct profit exceeds absorption profit.

Appendix B

Time-Based Accounting Overhead Allocation Example

This appendix presents a simple deterministic example of the application of a time-based accounting system for allocating manufacturing overheads (MOH). Based on the actual production data, total cycle time units are calculated for all products.

Product	A	B	C	All Products
Standard Production	1,400	300	100	1,800
Average Cycle Time (minutes)	300	500	800	361
Total Cycle Time Units	420,000	150,000	80,000	650,000

The total cycle time units will then be used to calculate the manufacturing overhead burden per cycle time unit. The total budgeted manufacturing overhead burden used in this study, along with the MOH/Cycle-Time Unit calculation, is as follows:

Depreciation	\$ 24,000
Data Entry	8,800
Cost Analysis	9,600
Production Engineering	6,400
Scheduling	4,400
Quality Control	1,653
Production Supervision	8,400
General Utilities	9,000
Miscellaneous	1,000
Total MOH	\$ 73,253
MOH/Cycle-Time Unit	\$ 0.113

With the cost of direct material and direct labor given, the allocation of manufacturing overhead burden is calculated by simply multiplying the average cycle

times of the individual products by the MOH/Cycle Time Unit (\$0.113). The following is a breakdown of total product cost by product under the time-based accounting system:

Product	A	B	C
Direct Material	\$ 20	\$ 40	\$ 120
Direct Labor	15	20	60
Manufacturing Overhead Allocation	34	56	90
Manufacturing Cost/Unit	<u>\$ 69</u>	<u>\$ 116</u>	<u>\$ 270</u>
Sales Price/Unit	\$ 100	\$ 160	\$ 360
Gross Margin	31	44	90
Gross Margin Percentage	31%	27%	25%

As shown below, the time-based accounting system, in keeping with US Generally Accepted Accounting Principles (US GAAP), fully absorbs all manufacturing overheads into finished goods inventory. The total manufacturing overhead burden allocated to all three products is \$73,253.

Product	A	B	C	All Products
Standard Production	1,400	300	100	1,800
Total MOH Allocated	47,333	16,905	9,016	73,253

Appendix C

SIMAN Code

This appendix presents the actual SIMAN language code for the simulation model used in this experiment.

```

;
;
;   Model statements for module: Create 1
;
94$   CREATE,   1,HoursToBaseTime(0.0),Part
A:HoursToBaseTime(EXPO(.5)):NEXT(95$);

95$   ASSIGN:   Part A Order Arrival.NumberOut=Part A Order
Arrival.NumberOut + 1:NEXT(88$);

;
;
;   Model statements for module: Record 29
;
88$   COUNT:   Record Total Part A Demand,1:NEXT(0$);

;
;
;   Model statements for module: Assign 1
;
0$   ASSIGN:   Rework Time=TRIA(20,60,80):
            Unload Time=TRIA(0.5,1.5,1.75):
            Load Time=TRIA(1.5,2,2.5):
            Transport Velocity=UNIF(25,35):
            Picture=Picture.Blue Ball:
            Sealer Time=TRIA(16, 18, 20):
            Prep Time=TRIA(14,18,22):NEXT(Part A);
```

Part A QUEUE, Part A Order.Queue,16,21\$:MARK(Arrive
Time):DETACH;

;
;

; Model statements for module: Record 7

21\$ COUNT: Record Part A Demand Unfilled,1:NEXT(91\$);

;
;

; Model statements for module: Record 32

91\$ TALLY: Record DFR Part A,1 - (NC(Record Part A Demand
Unfilled) / NC(Record Total Part A Demand)),1
:NEXT(22\$);

;
;

; Model statements for module: Dispose 6

22\$ ASSIGN: Part A Demand Unfilled.NumberOut=Part A Demand
Unfilled.NumberOut + 1;
98\$ DISPOSE: Yes;

;
;

; Model statements for module: Create 2

99\$ CREATE, 1,HoursToBaseTime(0.0),Part
B:HoursToBaseTime(EXPO(2)):NEXT(100\$);

100\$ ASSIGN: Part B Order Arrival.NumberOut=Part B Order
Arrival.NumberOut + 1:NEXT(89\$);

;
;

; Model statements for module: Record 30

89\$ COUNT: Record Total Part B Demand,1:NEXT(1\$);

```

;
;
;
; Model statements for module: Assign 2
;
1$    ASSIGN:    Rework Time=TRIA(40,80,100):
          Unload Time=TRIA(0.5,1.5,2):
          Load Time=TRIA(1.5,2,3):
          Transport Velocity=UNIF(20,30):
          Picture=Picture.Yellow Ball:
          Prep Time=TRIA(12,18,24):
          Sealer Time=TRIA(18,20,22):NEXT(Part B);

Part B    QUEUE,    Part B Order.Queue,5,23$:DETACH;

```

```

;
;
;
; Model statements for module: Record 8
;
23$    COUNT:    Record Part B Demand Unfilled,1:NEXT(92$);

```

```

;
;
;
; Model statements for module: Record 33
;
92$    TALLY:    Record DFR Part B,1 - ( NC(Record Part B Demand
Unfilled) / NC(Record Total Part B Demand) ),1
          :NEXT(24$);

```

```

;
;
;
; Model statements for module: Dispose 7
;
24$    ASSIGN:    Part B Demand Unfilled.NumberOut=Part B Demand
Unfilled.NumberOut + 1;
103$   DISPOSE:    Yes;

```

```

;
;
;
; Model statements for module: Create 4
;

```

104\$ CREATE, 1,HoursToBaseTime(0.0),Part
C:HoursToBaseTime(EXPO(8)):NEXT(105\$);

105\$ ASSIGN: Part C Order Arrival.NumberOut=Part C Order
Arrival.NumberOut + 1:NEXT(90\$);

;
;
;
; Model statements for module: Record 31
;

90\$ COUNT: Record Total Part C Demand,1:NEXT(10\$);

;
;
;
; Model statements for module: Assign 4
;

10\$ ASSIGN: Rework Time=TRIA(120,180,300):
Unload Time=TRIA(1,2.5,5):
Load Time=TRIA(2,3,5):
Transport Velocity=UNIF(10,30):
Picture=Picture.Green Ball:
Sealer Time=TRIA(68,72,78):
Prep Time=TRIA(64,90,124):NEXT(Part C);

Part C QUEUE, Part C Order.Queue,1,25\$:DETACH;

;
;
;
; Model statements for module: Record 9
;

25\$ COUNT: Record Part C Demand Unfilled,1:NEXT(93\$);

;
;
;
; Model statements for module: Record 34
;

93\$ TALLY: Record DFR Part C,1 - (NC(Record Part C Demand
Unfilled) / NC(Record Total Part C Demand),1
:NEXT(26\$);

;
;

```

; Model statements for module: Dispose 8
;
26$ ASSIGN: Part C Demand Unfilled.NumberOut=Part C Demand
Unfilled.NumberOut + 1;
108$ DISPOSE: Yes;

;
;
; Model statements for module: Enter 1
;
;
11$ STATION, Prep Arrival.Station;
109$ DELAY: Unload Time,,Transfer:NEXT(111$);

111$ FREE: Prep Cart:NEXT(2$);

;
;
; Model statements for module: Process 2
;
;
2$ ASSIGN: Prep Process.NumberIn=Prep Process.NumberIn + 1:
Prep Process.WIP=Prep Process.WIP+1;
149$ STACK, 1:Save:NEXT(123$);

123$ QUEUE, Prep Process.Queue;
122$ SEIZE, 1,VA:
Part Prep,1:NEXT(121$);

121$ DELAY: Prep Time,,VA:NEXT(164$);

164$ ASSIGN: Prep Process.WaitTime=Prep Process.WaitTime +
Diff.WaitTime;
128$ TALLY: Prep Process.WaitTimePerEntity,Diff.WaitTime,1;
130$ TALLY: Prep Process.TotalTimePerEntity,Diff.StartTime,1;
154$ ASSIGN: Prep Process.VATime=Prep Process.VATime +
Diff.VATime;
155$ TALLY: Prep Process.VATimePerEntity,Diff.VATime,1;
120$ RELEASE: Part Prep,1;
169$ STACK, 1:Destroy:NEXT(168$);

168$ ASSIGN: Prep Process.NumberOut=Prep Process.NumberOut + 1:
Prep Process.WIP=Prep Process.WIP-1:NEXT(27$);

```

```

;
;
;
;   Model statements for module: Station 12
;
;
27$      STATION,    Prep Station;
173$     DELAY:      0.0,,VA:NEXT(41$);

;
;
;
;   Model statements for module: Request 1
;
;
41$      QUEUE,      Request Cart 1 to Prep Station.Queue;
          REQUEST,    1:Sealer Cart(SDS),50:NEXT(44$);

;
;
;
;   Model statements for module: Delay 2
;
;
44$      DELAY:      Load Time,,Transfer:NEXT(43$);

;
;
;
;   Model statements for module: Transport 2
;
;
43$      TRANSPORT:  Sealer Cart,Sealer Arrival.Station,Transport Velocity;

;
;
;
;   Model statements for module: Enter 2
;
;
12$      STATION,    Sealer Arrival.Station;
175$     DELAY:      Unload Time,,Transfer:NEXT(177$);

177$     FREE:       Sealer Cart:NEXT(3$);

;
;
;
;   Model statements for module: Process 3
;
;

```

```

3$      ASSIGN:    Sealer Process.NumberIn=Sealer Process.NumberIn + 1:
                Sealer Process.WIP=Sealer Process.WIP+1;
215$    STACK,    1:Save:NEXT(189$);

189$    QUEUE,    Sealer Process.Queue;
188$    SEIZE,    1,VA:
                Sealer,1:NEXT(187$);

187$    DELAY:    Sealer Time,,VA:NEXT(230$);

230$    ASSIGN:    Sealer Process.WaitTime=Sealer Process.WaitTime +
Diff.WaitTime;
194$    TALLY:    Sealer Process.WaitTimePerEntity,Diff.WaitTime,1;
196$    TALLY:    Sealer Process.TotalTimePerEntity,Diff.StartTime,1;
220$    ASSIGN:    Sealer Process.VATime=Sealer Process.VATime +
Diff.VATime;
221$    TALLY:    Sealer Process.VATimePerEntity,Diff.VATime,1;
186$    RELEASE:  Sealer,1;
235$    STACK,    1:Destroy:NEXT(234$);

234$    ASSIGN:    Sealer Process.NumberOut=Sealer Process.NumberOut + 1:
                Sealer Process.WIP=Sealer Process.WIP-1:NEXT(4$);

;
;
;   Model statements for module: Decide 1
;
4$      BRANCH,    1:
                If,Entity.Type==Part A,13$,Yes:
                If,Entity.Type==Part B,14$,Yes:
                If,Entity.Type==Part C,15$,Yes:
                Else,20$,Yes;

;
;
;   Model statements for module: Dispose 5
;
20$     ASSIGN:    Dispose of Sealer Exceptions.NumberOut=Dispose of
Sealer Exceptions.NumberOut + 1;
239$    DISPOSE:   Yes;

;
;
;   Model statements for module: Decide 3

```

```

;
13$    BRANCH,    1:
        With,4/100,240$,Yes:
        Else,241$,Yes;
240$    ASSIGN:    Failed Sealer Inspection Part A.NumberOut True=Failed
Sealer Inspection Part A.NumberOut True + 1
        :NEXT(28$);

241$    ASSIGN:    Failed Sealer Inspection Part A.NumberOut False=Failed
Sealer Inspection Part A.NumberOut False + 1
        :NEXT(29$);

```

```

;
;
;
;    Model statements for module: Station 13
;
;

```

```

28$    STATION,    Failed Parts Station;
244$    DELAY:    0.0,,VA:NEXT(47$);

```

```

;
;
;
;    Model statements for module: Request 2
;
;

```

```

47$    QUEUE,    Request Cart to Failed Parts.Queue;
        REQUEST,    1:Cart 2(SDS),50:NEXT(45$);

```

```

;
;
;
;    Model statements for module: Delay 3
;
;

```

```

45$    DELAY:    Load Time,,Transfer:NEXT(46$);

```

```

;
;
;
;    Model statements for module: Transport 3
;
;

```

```

46$    TRANSPORT:    Cart 2,Rework Arrival.Station,Transport Velocity;

```

```

;
;
;

```



```
; Model statements for module: Station 14
;
```

```
29$ STATION, Good Parts Station;
248$ DELAY: 0.0,,VA:NEXT(50$);
```

```
;
;
; Model statements for module: Request 3
;
```

```
50$ QUEUE, Request Cart to Good Parts.Queue;
REQUEST, 1:Cart 2(SDS),50:NEXT(52$);
```

```
;
;
; Model statements for module: Delay 4
;
```

```
52$ DELAY: Load Time,,Transfer:NEXT(49$);
```

```
;
;
; Model statements for module: Transport 4
;
```

```
49$ TRANSPORT: Cart 2,Shipped Parts Arrival.Station,Transport Velocity;
```

```
;
;
; Model statements for module: Decide 4
;
```

```
14$ BRANCH, 1:
With,6/100,250$,Yes:
Else,251$,Yes;
```

```
250$ ASSIGN: Failed Sealer Inspection Part B.NumberOut True=Failed
Sealer Inspection Part B.NumberOut True + 1
:NEXT(28$);
```

```
251$ ASSIGN: Failed Sealer Inspection Part B.NumberOut False=Failed
Sealer Inspection Part B.NumberOut False + 1
:NEXT(29$);
```

```
;
```

```

;
; Model statements for module: Decide 5
;
15$ BRANCH, 1:
      With,10/100,252$,Yes:
      Else,253$,Yes;
252$ ASSIGN: Failed Sealer Inspection Part C.NumberOut True=Failed
Sealer Inspection Part C.NumberOut True + 1
      :NEXT(28$);

253$ ASSIGN: Failed Sealer Inspection Part C.NumberOut False=Failed
Sealer Inspection Part C.NumberOut False + 1
      :NEXT(29$);

      QPICK, POR:

$TCSS$
      Part C:
      Part B:
      Part A;
;
;

$ABCS$
      Part B:
      Part A:
      Part C;

;
;

$TBAS$
      Part A:
      Part B:
      Part C;

36$ ALLOCATE, 1:Prep Cart,Order Release Station:MARK(Arrive
Time):NEXT(37$);

;
;

```

```

; Model statements for module: Station 17
;
37$ STATION, Order Release Station;
256$ DELAY: 0.0,,VA:NEXT(65$);

;
;
; Model statements for module: Assign 5
;
65$ ASSIGN: Arrival Time=TNOW:NEXT(38$);

;
;
; Model statements for module: Move 1
;
38$ MOVE: Prep Cart,Prep Arrival.Station,50:NEXT(40$);

;
;
; Model statements for module: Delay 1
;
40$ DELAY: Load Time,,Transfer:NEXT(64$);

;
;
; Model statements for module: Decide 9
;
64$ BRANCH, 1:
      If,NQ(Prep Process.Queue) <= 16 && NQ(Sealer Process.Queue)
<= 16,257$,Yes:
      Else,258$,Yes;
257$ ASSIGN: Check Prep and Sealer Queue Availability.NumberOut
True=
      Check Prep and Sealer Queue Availability.NumberOut True +
1:NEXT(39$);

258$ ASSIGN: Check Prep and Sealer Queue Availability.NumberOut
False=
      Check Prep and Sealer Queue Availability.NumberOut False +
1:NEXT(63$);

```

```

;
;
;
;   Model statements for module: Transport 1
;
39$   TRANSPORT:   Prep Cart,Prep Arrival.Station,Transport Velocity;

;
;
;
;   Model statements for module: Delay 7
;
63$   DELAY:      EXPO( .5 ),,Wait:NEXT(64$);

;
;
;
;   Model statements for module: Enter 4
;

30$   STATION,    Rework Arrival.Station;
259$  DELAY:      Unload Time,,Transfer:NEXT(261$);

261$  FREE:       Cart 2:NEXT(5$);

;
;
;
;   Model statements for module: Process 4
;
5$    ASSIGN:     Rework Process.NumberIn=Rework Process.NumberIn + 1;
                Rework Process.WIP=Rework Process.WIP+1;
299$  STACK,      1:Save:NEXT(273$);

273$  QUEUE,      Rework Process.Queue;
272$  SEIZE,      1,Other:
                Rework,1:NEXT(271$);

271$  DELAY:      Rework Time,,Other:NEXT(314$);

314$  ASSIGN:     Rework Process.WaitTime=Rework Process.WaitTime +
Diff.WaitTime;
278$  TALLY:      Rework Process.WaitTimePerEntity,Diff.WaitTime,1;
280$  TALLY:      Rework Process.TotalTimePerEntity,Diff.StartTime,1;
304$  ASSIGN:     Rework Process.OtherTime=Rework Process.OtherTime +
Diff.OtherTime;

```

```

305$    TALLY:      Rework Process.OtherTimePerEntity,Diff.OtherTime,1;
270$    RELEASE:    Rework,1;
319$    STACK,      1:Destroy:NEXT(318$);

318$    ASSIGN:     Rework Process.NumberOut=Rework Process.NumberOut
+ 1:
                Rework Process.WIP=Rework Process.WIP-1:NEXT(16$);

;
;
;
;   Model statements for module: Decide 6
;
;
16$     BRANCH,     1:
                If,Entity.Type==Part A,6$,Yes:
                If,Entity.Type==Part B,17$,Yes:
                If,Entity.Type==Part C,18$,Yes:
                Else,19$,Yes;

;
;
;
;   Model statements for module: Dispose 4
;
;
19$     ASSIGN:     Dispose of Rework Exceptions.NumberOut=Dispose of
Rework Exceptions.NumberOut + 1;
323$    DISPOSE:    Yes;

;
;
;
;   Model statements for module: Decide 2
;
;
6$      BRANCH,     1:
                With,19/100,324$,Yes:
                Else,325$,Yes;
324$    ASSIGN:     Failed Rework Inspection Part A.NumberOut True=Failed
Rework Inspection Part A.NumberOut True + 1
                :NEXT(32$);

325$    ASSIGN:     Failed Rework Inspection Part A.NumberOut False=Failed
Rework Inspection Part A.NumberOut False + 1
                :NEXT(34$);

;
;
;

```

```

; Model statements for module: Station 15
;
32$ STATION, Failed Rework Parts Station;
328$ DELAY: 0.0,,VA:NEXT(53$);

;
;
; Model statements for module: Allocate 1
;
55$ QUEUE, Request Cart to Scrapped Parts.Queue;
ALLOCATE, 2:Cart 2(SDS):NEXT(55$);

;
;
; Model statements for module: Move 2
;
55$ MOVE: Cart 2,Failed Rework Parts Station,50:NEXT(56$);

;
;
; Model statements for module: Delay 5
;
56$ DELAY: Load Time,,Transfer:NEXT(57$);

;
;
; Model statements for module: Transport 5
;
57$ TRANSPORT: Cart 2,Scrapped Parts Arrival.Station,Transport
Velocity;

;
;
; Model statements for module: Station 16
;
34$ STATION, Salvaged Parts Station;
332$ DELAY: 0.0,,VA:NEXT(58$);

```

```

;
;
;
;   Model statements for module: Request 4
;
;
58$   QUEUE,      Request Cart to Salvaged Parts.Queue;
      REQUEST,    1:Cart 2(SDS),50:NEXT(60$);

```

```

;
;
;
;   Model statements for module: Move 3
;
;
60$   MOVE:      Cart 2,Salvaged Parts Station,50:NEXT(62$);

```

```

;
;
;
;   Model statements for module: Delay 6
;
;
62$   DELAY:    Load Time,,Transfer:NEXT(61$);

```

```

;
;
;
;   Model statements for module: Transport 6
;
;
61$   TRANSPORT: Cart 2,Salvaged Parts Arrival.Station,Transport
Velocity;

```

```

;
;
;
;   Model statements for module: Decide 7
;
;
17$   BRANCH,    1:
          With,27/100,334$,Yes:
          Else,335$,Yes;
334$   ASSIGN:   Failed Rework Inspection Part B.NumberOut True=Failed
Rework Inspection Part B.NumberOut True + 1
          :NEXT(32$);

335$   ASSIGN:   Failed Rework Inspection Part B.NumberOut False=Failed
Rework Inspection Part B.NumberOut False + 1
          :NEXT(34$);

```

```

;
;
;
;   Model statements for module: Decide 8
;
;
18$   BRANCH,    1:
        With,40/100,336$,Yes:
        Else,337$,Yes;
336$   ASSIGN:   Failed Rework Inspection Part C.NumberOut True=Failed
Rework Inspection Part C.NumberOut True + 1
        :NEXT(32$);

337$   ASSIGN:   Failed Rework Inspection Part C.NumberOut False=Failed
Rework Inspection Part C.NumberOut False + 1
        :NEXT(34$);

;
;
;
;   Model statements for module: Enter 5
;
;
31$   STATION,   Shipped Parts Arrival.Station;
338$   DELAY:    Unload Time,,Transfer:NEXT(340$);

340$   FREE:     Cart 2:NEXT(70$);

;
;
;
;   Model statements for module: Decide 11
;
;
70$   BRANCH,    1:
        If,Entity.Type==Part A,72$,Yes:
        If,Entity.Type==Part B,71$,Yes:
        If,Entity.Type==Part C,73$,Yes:
        Else,74$,Yes;

;
;
;
;   Model statements for module: Dispose 10
;
;
74$   ASSIGN:    Dispose 10.NumberOut=Dispose 10.NumberOut + 1;
351$   DISPOSE:  Yes;

;

```



```
;
;
;   Model statements for module: Record 16
;
;
72$   TALLY:   Record Part A Shipped Time,INT(Arrival
Time),1:NEXT(75$);
```

```
;
;
;
;   Model statements for module: Record 18
;
;
75$   COUNT:   Record Part A Shipped,1:NEXT(78$);
```

```
;
;
;
;   Model statements for module: Record 21
;
;
78$   COUNT:   Record Total Shipped,1:NEXT(9$);
```

```
;
;
;
;   Model statements for module: Dispose 3
;
;
9$    ASSIGN:   Shipped.NumberOut=Shipped.NumberOut + 1;
352$  DISPOSE:  Yes;
```

```
;
;
;
;   Model statements for module: Record 15
;
;
71$   TALLY:   Record Part B Shipped Time,INT(Arrive
Time),1:NEXT(76$);
```

```
;
;
;
;   Model statements for module: Record 19
;
;
76$   COUNT:   Record Part B Shipped,1:NEXT(78$);
```

```
;
;
;
```

```

; Model statements for module: Record 17
;
73$ TALLY: Record Part C Shipped Time,INT(Arrival
Time),1:NEXT(77$);

;
;
; Model statements for module: Record 20
;
77$ COUNT: Record Part C Shipped,1:NEXT(78$);

;
;
; Model statements for module: Enter 6
;
33$ STATION, Scrapped Parts Arrival.Station;
353$ DELAY: Unload Time,,Transfer:NEXT(355$);

355$ FREE: Cart 2:NEXT(79$);

;
;
; Model statements for module: Decide 12
;
79$ BRANCH, 1:
      If,Entity.Type==Part A,81$,Yes:
      If,Entity.Type==Part B,80$,Yes:
      If,Entity.Type==Part C,82$,Yes:
      Else,83$,Yes;

;
;
; Model statements for module: Dispose 11
;
83$ ASSIGN: Dispose 11.NumberOut=Dispose 11.NumberOut + 1;
366$ DISPOSE: Yes;

;
;
; Model statements for module: Record 23
;

```

81\$ TALLY: Record Part A Scrapped Time,INT(Arrival
Time),1:NEXT(84\$);

;
;
;
; Model statements for module: Record 25
;

84\$ COUNT: Record Part A Scrapped,1:NEXT(87\$);

;
;
;
; Model statements for module: Record 28
;

87\$ COUNT: Record Total Scrapped,1:NEXT(8\$);

;
;
;
; Model statements for module: Dispose 1
;

8\$ ASSIGN: Scrapped.NumberOut=Scrapped.NumberOut + 1;
367\$ DISPOSE: Yes;

;
;
;
; Model statements for module: Record 22
;

80\$ TALLY: Record Part B Scrapped Time,INT(Arrive
Time),1:NEXT(85\$);

;
;
;
; Model statements for module: Record 26
;

85\$ COUNT: Record Part B Scrapped,1:NEXT(87\$);

;
;
;
; Model statements for module: Record 24
;

82\$ TALLY: Record Part C Scrapped Time,INT(Arrival
Time),1:NEXT(86\$);

;
;
;
;

Model statements for module: Record 27

86\$ COUNT: Record Part C Scrapped,1:NEXT(87\$);

;
;
;
;

Model statements for module: Enter 7

35\$ STATION, Salvaged Parts Arrival.Station;
368\$ DELAY: Unload Time,,Transfer:NEXT(370\$);

370\$ FREE: Cart 2:NEXT(66\$);

;
;
;
;

Model statements for module: Decide 10

66\$ BRANCH, 1:
If,Entity.Type==Part A,68\$,Yes:
If,Entity.Type==Part B,7\$,Yes:
If,Entity.Type==Part C,69\$,Yes:
Else,67\$,Yes;

;
;
;
;

Model statements for module: Dispose 9

67\$ ASSIGN: Dispose 9.NumberOut=Dispose 9.NumberOut + 1;
381\$ DISPOSE: Yes;

;
;
;
;

Model statements for module: Record 12

68\$ TALLY: Record Part A Salvaged Time,INT(Arrival
Time),1:NEXT(75\$);

```
.  
;  
;  
;  
;   Model statements for module: Record 2  
;  
7$   TALLY:   Record Part B Salvaged Time,INT(Arrive  
Time),1:NEXT(76$);
```

```
.  
;  
;  
;  
;   Model statements for module: Record 14  
;  
69$   TALLY:   Record Part C Salvaged Time,INT(Arrival  
Time),1:NEXT(77$);
```

Appendix D

Cumulative NOI

This appendix presents the cumulative NOI under the various experimental conditions. Review of management accounting system performance under the nine experimental conditions (three levels of manufacturing overhead by three levels of product mix complexity) shows no significant difference in cumulative net operating income when product mix complexity is low. Activities-based costing begins to significantly outperform the other two management accounting system alternatives at a medium demand setting for product mix complexity. This difference becomes more pronounced as product mix complexity is set at a high level. At this high setting, time-based accounting begins to slowly outperform traditional costing systems.

Figure D.1: Cumulative Net Operating Income by MAS
(Experimental Condition Group 1)

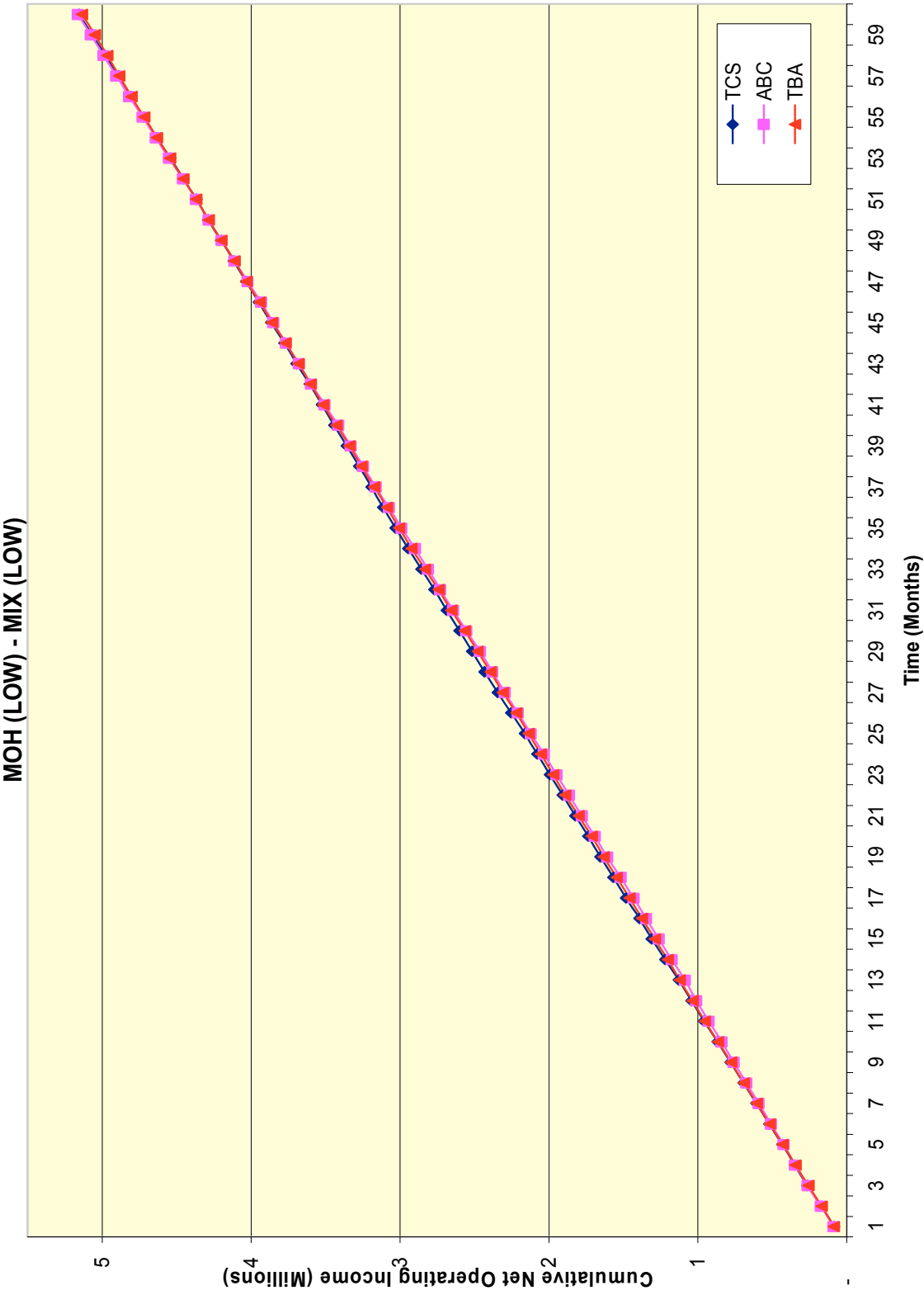


Figure D.2: Cumulative Net Operating Income by MAS
(Experimental Condition Group 2)

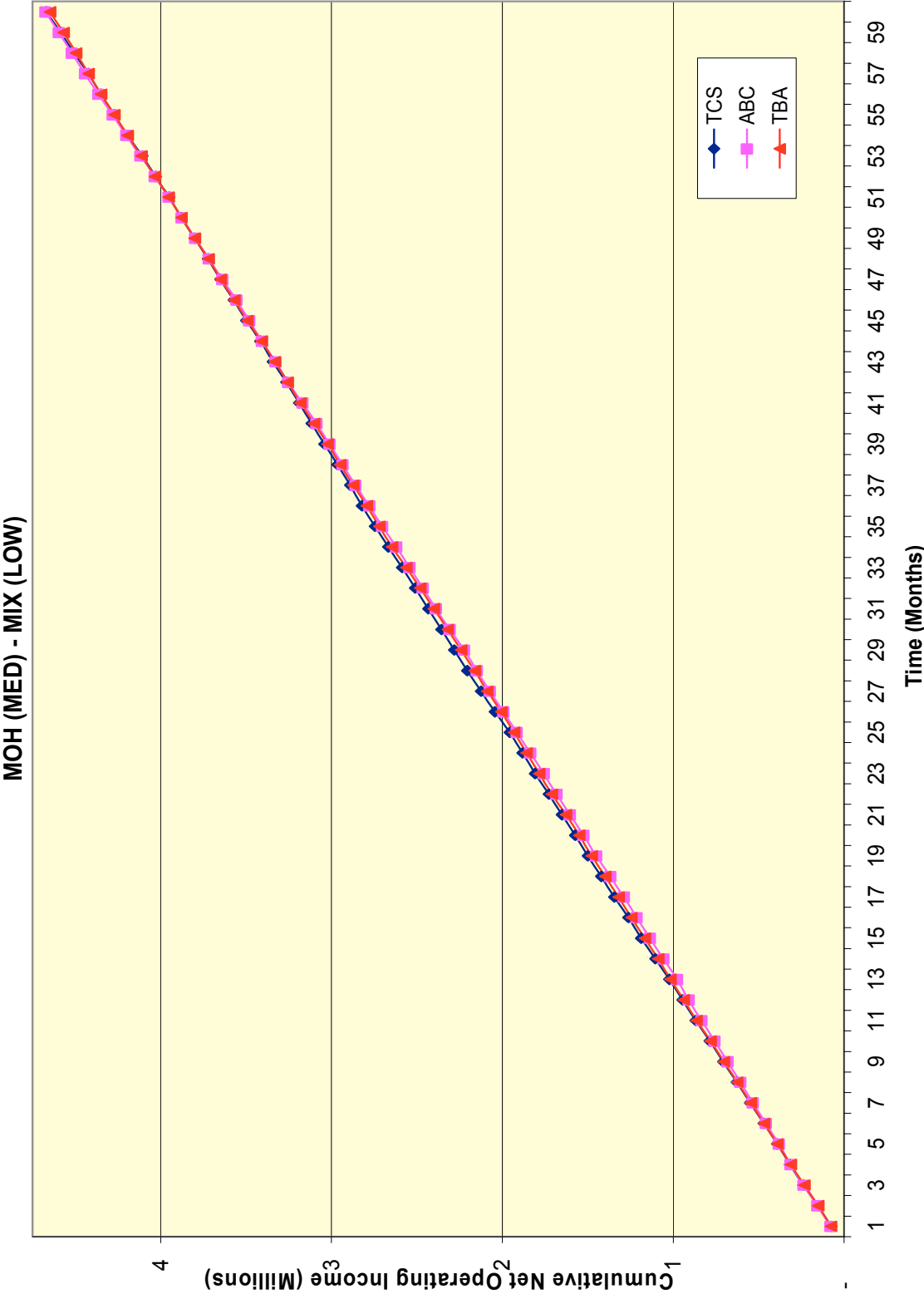


Figure D.3: Cumulative Net Operating Income by MAS
(Experimental Condition Group 3)

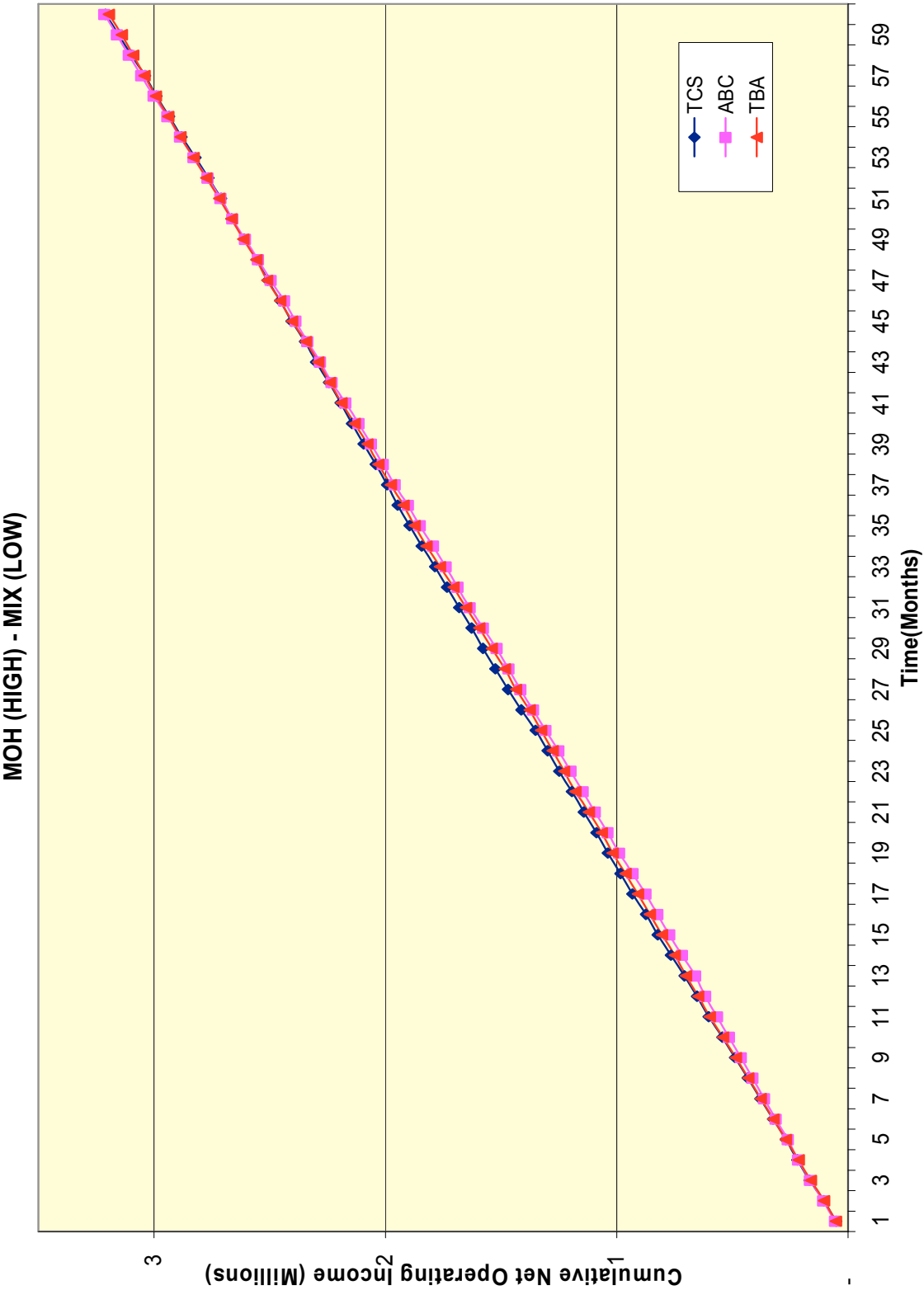


Figure D.4: Cumulative Net Operating Income by MAS
(Experimental Condition Group 4)

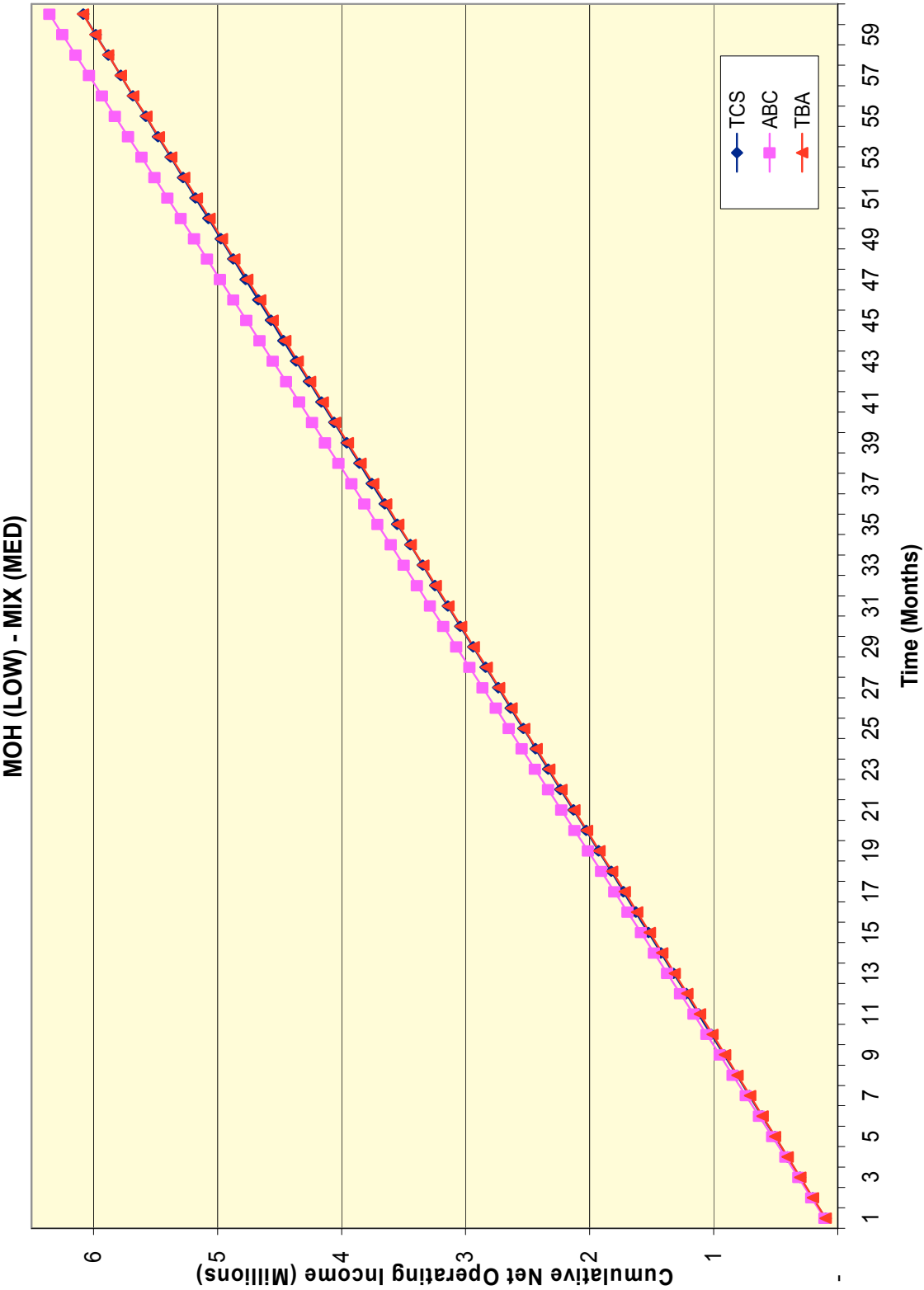


Figure D.5: Cumulative Net Operating Income by MAS
(Experimental Condition Group 5)

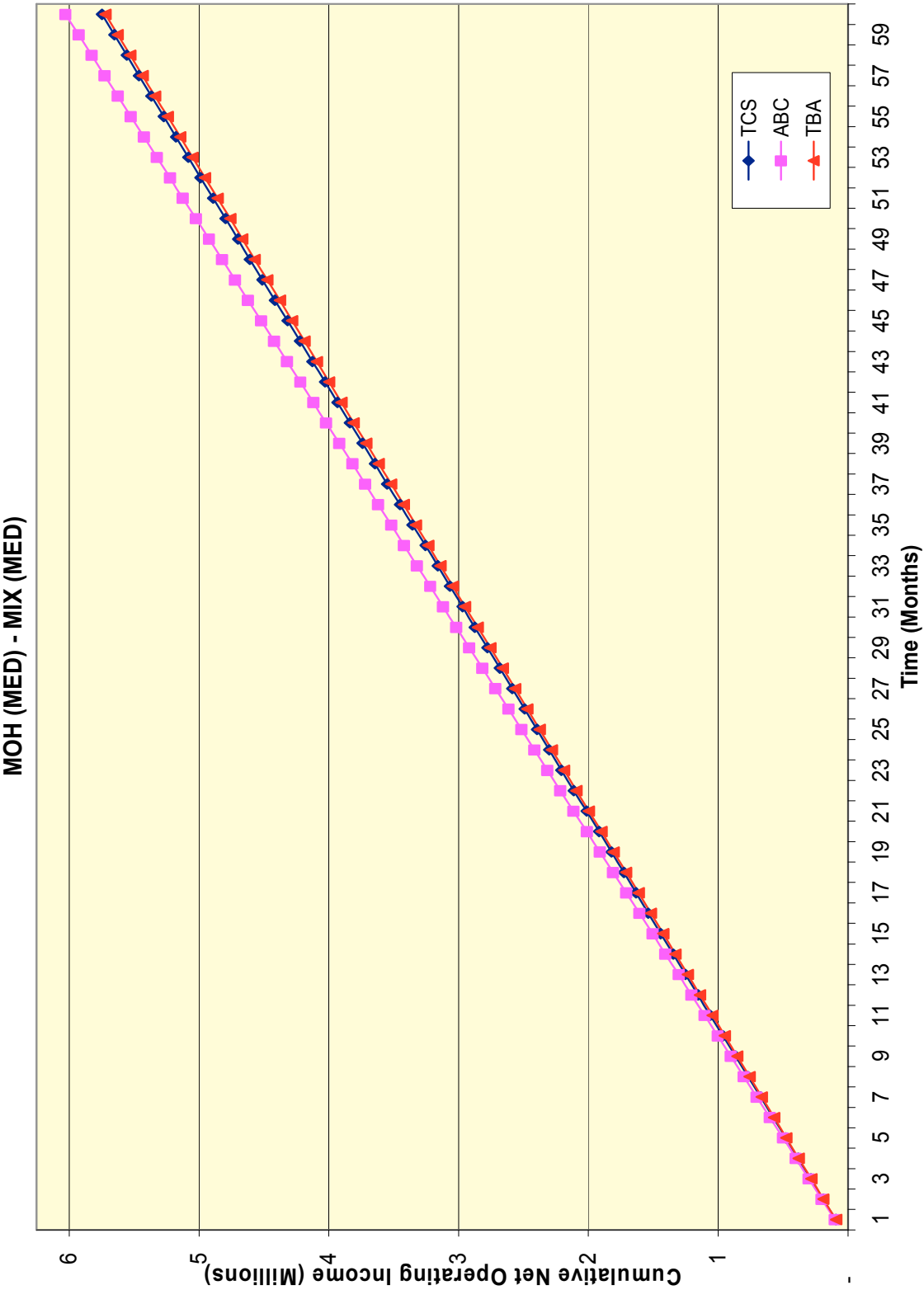


Figure D.6: Cumulative Net Operating Income by MAS
(Experimental Condition Group 6)

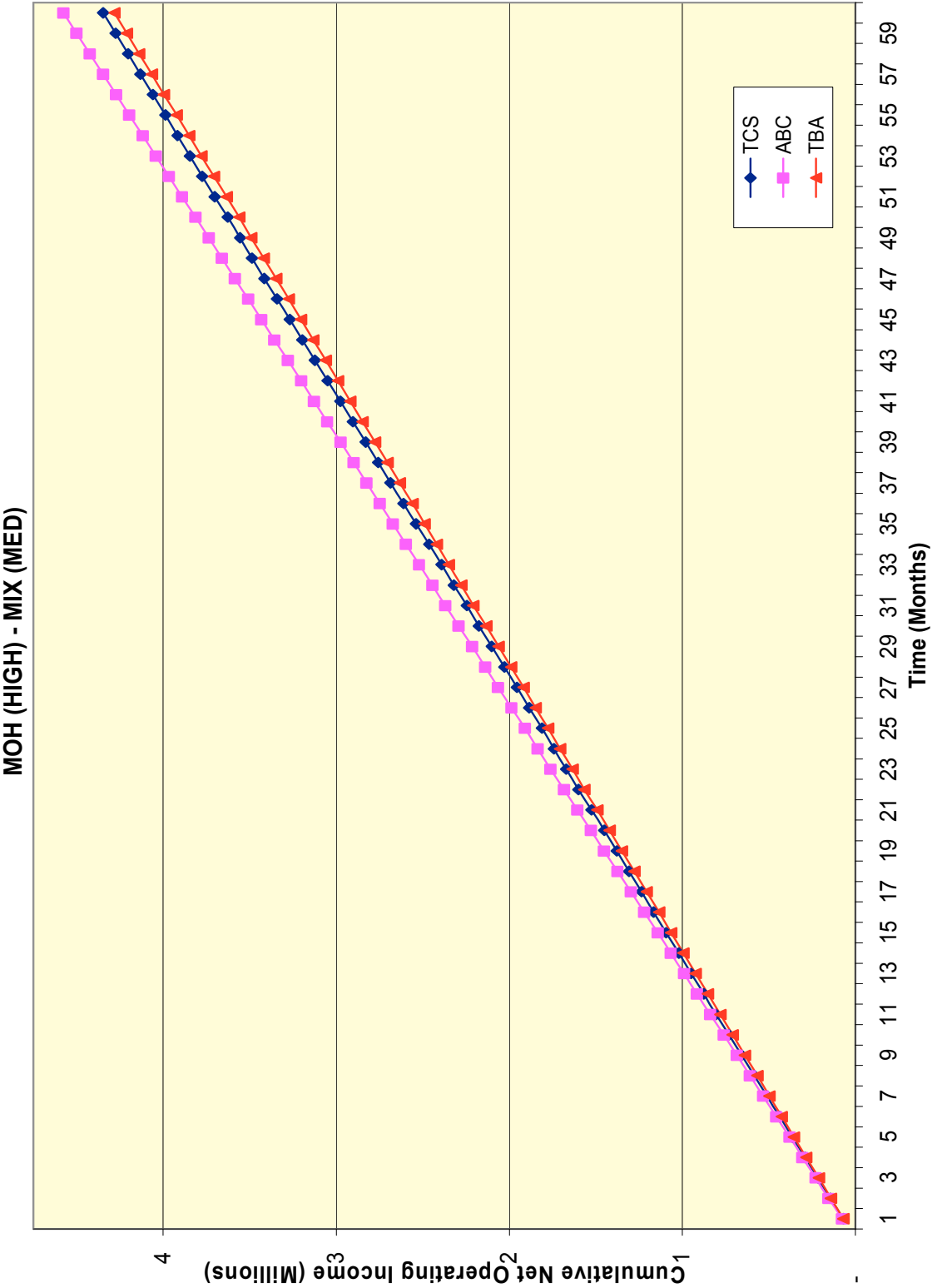


Figure D.7: Cumulative Net Operating Income by MAS
(Experimental Condition Group 7)

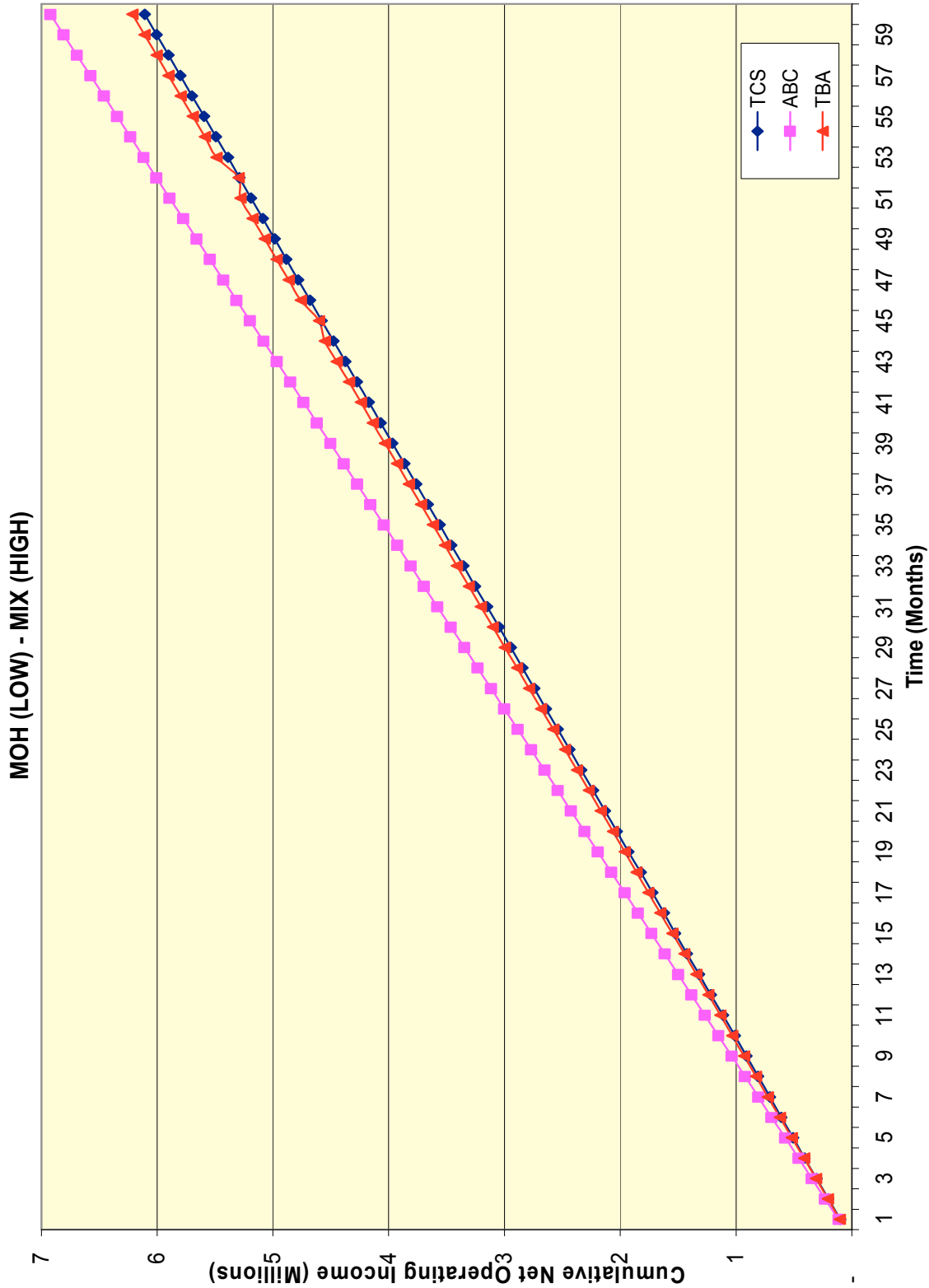


Figure D.8: Cumulative Net Operating Income by MAS
(Experimental Condition Group 8)

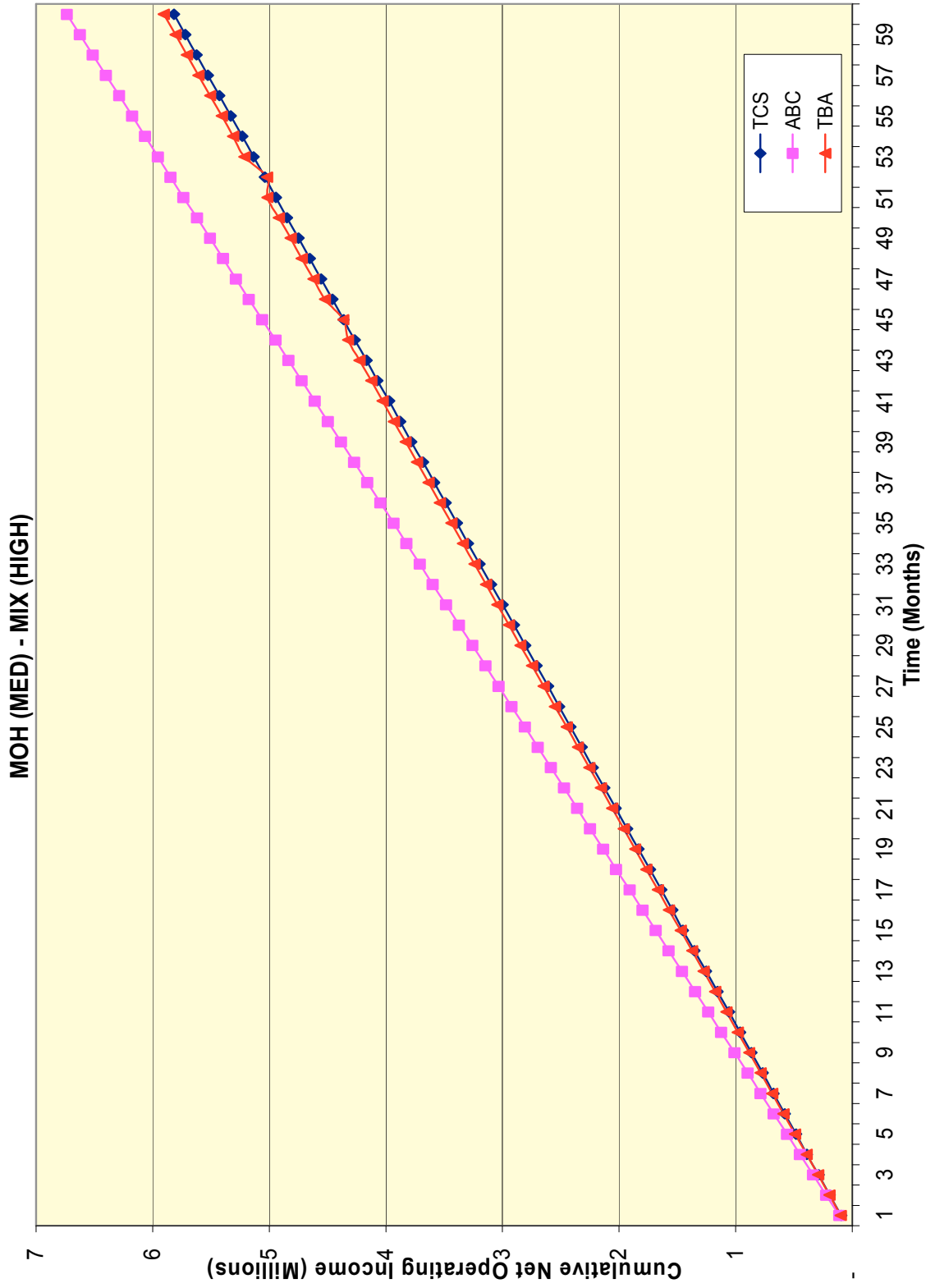
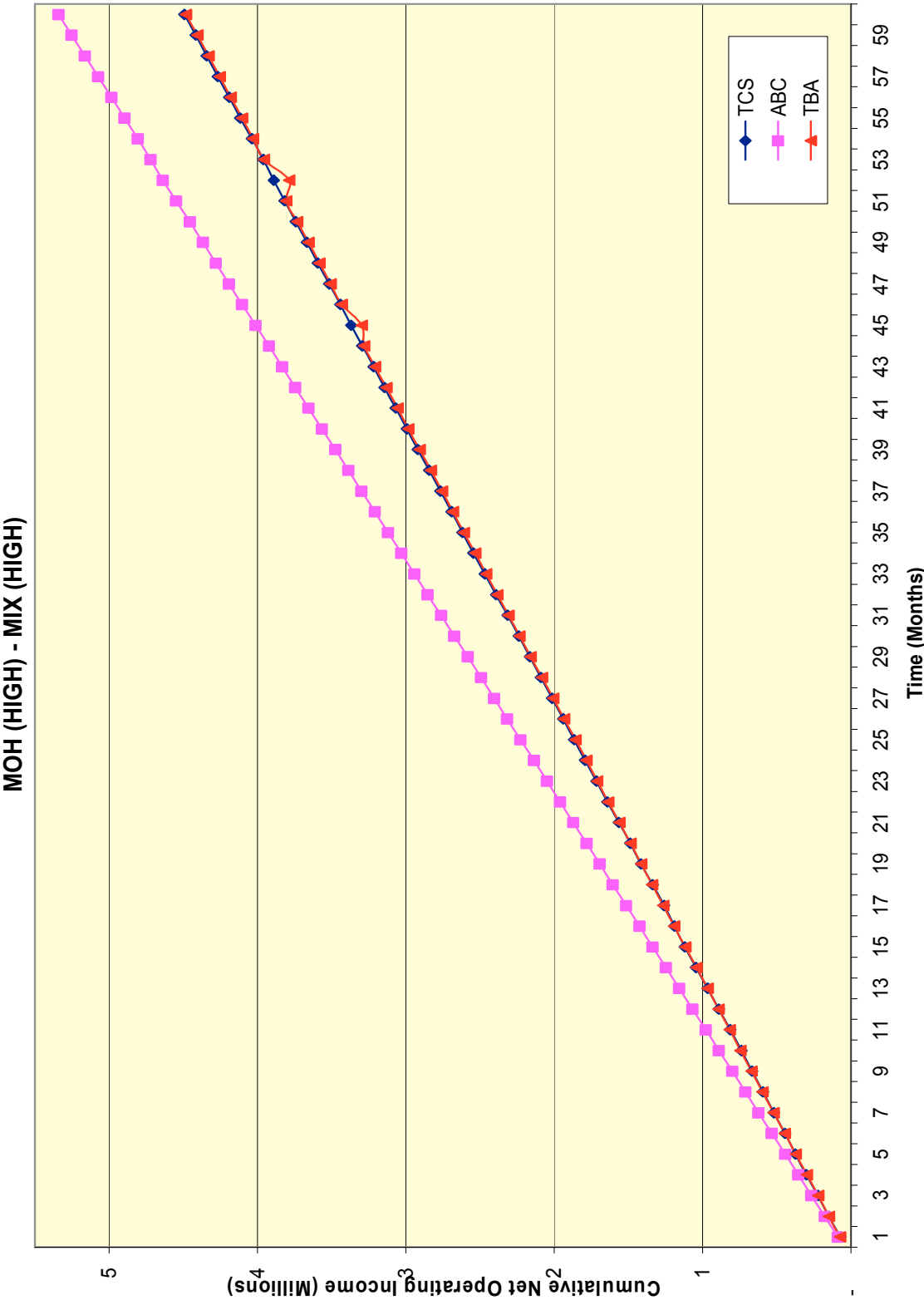


Figure D.9: Cumulative Net Operating Income by MAS
(Experimental Condition Group 9)



Appendix E

Raw Data

This appendix presents the raw data generated from the ARENA simulation model. These data were then loaded into an Excel spreadsheet and sorted for uploading into SPSS for further statistical analyses. What is shown in the following 18 pages are the data in the Excel spreadsheet format.

Replication Number	Independent Variables			Dependent Variables																				
	M&S	MOH	MIX	Product A				Product B				Product C												
				Period Demand	Period Shipped	Total Shipped	Period DFR	Period Time	Total Cycle Time	Period Demand	Period Shipped	Total Shipped	Period DFR	Period Time	Total Cycle Time	Period Demand	Period Shipped	Total Shipped	Period DFR	Period Time	Total Cycle Time			
1	3	1	1	1422	1422	1419	100	100	268.95	276.85	359	359	862	100	100	264.83	264.83	86	86	82	0.86	0.87	363.25	
2	3	1	1	1421	2645	1404	100	100	255.05	274.86	342	701	653	100	100	303.33	253.46	95	181	52	164	0.90	0.97	366.41
3	3	1	1	1504	4547	1488	100	100	320.94	280.63	463	1104	392	100	100	335.43	302.43	87	268	236	236	0.91	0.95	404.84
4	3	1	1	1431	5778	1421	100	100	289.50	280.50	359	1463	350	100	100	312.88	302.02	84	352	81	317	0.95	0.95	385.64
5	3	1	1	1491	7269	1466	100	100	321.50	286.78	342	1895	327	100	100	349.38	314.28	98	460	91	408	0.95	0.95	385.68
6	3	1	1	1508	8777	1513	100	100	323.18	301.37	338	2143	327	100	100	328.01	314.67	86	536	76	484	0.95	0.95	421.44
7	3	1	1	1447	10224	1486	100	100	333.21	305.88	356	2499	349	100	100	351.15	319.81	91	627	83	697	0.95	0.95	399.88
8	3	1	1	1447	11671	1486	100	100	295.76	304.63	356	2855	349	100	100	304.70	318.00	91	719	83	650	0.95	0.95	381.25
9	3	1	1	1447	13118	1486	100	100	298.94	304.00	356	3210	349	100	100	309.44	317.05	91	810	83	733	0.95	0.95	384.47
10	3	1	1	1447	14665	1486	100	100	297.68	303.37	356	3666	349	100	100	307.53	316.09	91	901	83	816	0.95	0.95	383.14
11	3	1	1	1488	16063	1483	100	100	301.68	308.79	351	3917	337	100	100	307.41	322.75	110	1011	88	904	0.95	0.95	449.33
12	3	1	1	1417	17470	1466	100	100	274.89	306.04	346	4263	330	100	100	277.07	319.11	85	1066	81	865	0.95	0.95	405.87
13	3	1	1	1463	18933	1461	100	100	312.51	306.54	337	4600	337	100	100	309.27	321.38	86	1182	87	1133	0.95	0.95	437.13
14	3	1	1	1461	20394	1449	100	100	244.12	302.07	346	4846	334	100	100	246.31	316.10	75	1257	67	1333	0.95	0.95	369.69
15	3	1	1	1397	21791	1385	100	100	306.66	302.30	358	5304	356	100	100	309.36	315.88	98	1355	93	1226	0.95	0.95	404.80
16	3	1	1	1465	23256	1464	100	100	277.23	300.72	353	5657	349	100	100	275.99	313.45	81	1436	77	1303	0.95	0.95	401.17
17	3	1	1	1414	24670	1393	100	100	291.94	300.22	366	6023	350	100	100	293.27	312.88	88	1524	82	1477	0.95	0.95	419.94
18	3	1	1	1379	26049	1382	100	100	270.11	298.81	357	6380	352	100	100	270.33	314.23	97	1713	89	1656	0.95	0.95	419.94
19	3	1	1	1449	27468	1438	100	100	323.28	299.91	349	6729	334	100	100	320.03	310.34	78	1791	72	1638	0.95	0.95	331.20
20	3	1	1	1395	28893	1382	100	100	213.00	295.72	354	7083	349	100	100	217.52	306.77	94	1885	92	1730	0.95	0.95	337.91
21	3	1	1	1417	30310	1411	100	100	247.33	293.46	346	7429	342	100	100	245.90	310.43	111	1966	93	1823	0.95	0.95	425.14
22	3	1	1	1410	31720	1400	100	100	342.24	295.82	360	7819	380	100	100	340.20	310.43	111	1966	93	1823	0.95	0.95	425.14
23	3	1	1	1439	33169	1429	100	100	204.89	291.88	354	8203	376	100	100	211.92	310.43	63	2059	61	1894	0.95	0.95	315.43
24	3	1	1	1404	34693	1385	100	100	290.19	291.82	339	8642	331	100	100	298.24	305.42	88	2143	84	1968	0.95	0.95	402.90
25	3	1	1	1432	35995	1413	100	100	298.08	291.40	341	8983	330	100	100	295.42	305.42	87	2234	80	2048	0.95	0.95	304.68
26	3	1	1	1424	37419	1424	100	100	343.64	289.63	348	9231	346	100	100	343.64	305.42	87	2304	80	2048	0.95	0.95	304.68
27	3	1	1	1443	38862	1427	100	100	345.57	291.63	357	9686	355	100	100	345.57	305.42	105	2414	96	2216	0.95	0.95	401.64
28	3	1	1	1430	40232	1421	100	100	247.36	289.86	333	9931	327	100	100	246.30	304.27	82	2486	77	2282	0.95	0.95	360.66
29	3	1	1	1408	41338	1402	100	100	303.86	291.16	379	10388	388	100	100	303.86	305.70	97	2563	86	2377	0.95	0.95	380.57
30	3	1	1	1469	44967	1448	100	100	377.34	293.98	363	11051	346	100	100	377.34	308.63	109	2797	91	2594	0.95	0.95	441.47
31	3	1	1	1429	46026	1414	100	100	321.75	294.84	362	11413	358	100	100	321.75	308.95	91	2888	86	2726	0.95	0.95	415.90
32	3	1	1	1480	47508	1472	100	100	364.01	297.00	356	11769	353	100	100	364.01	311.34	94	2982	86	2726	0.95	0.95	460.27
33	3	1	1	1609	49015	1607	100	100	339.61	298.32	346	12115	339	100	100	339.61	313.84	82	3074	88	2814	0.95	0.95	433.82
34	3	1	1	1436	50461	1430	100	100	295.17	298.23	364	12479	354	100	100	295.17	313.84	80	3154	76	2880	0.95	0.95	400.39
35	3	1	1	1399	51850	1386	100	100	242.17	296.72	352	12831	348	100	100	242.17	311.84	77	3231	72	2962	0.95	0.95	331.61
36	3	1	1	1474	53324	1464	100	100	296.99	296.70	397	13228	369	100	100	296.99	311.44	79	3310	75	3037	0.95	0.95	397.89
37	3	1	1	1474	54324	1464	100	100	337.75	297.75	359	13687	358	100	100	337.75	312.79	89	3399	85	3122	0.95	0.95	427.25
38	3	1	1	1396	54720	1389	100	100	234.05	296.12	381	13948	353	100	100	234.05	312.73	70	3469	66	3188	0.95	0.95	325.69
39	3	1	1	1437	56157	1426	100	100	346.00	297.41	359	14307	353	100	100	346.00	312.73	108	3569	91	3279	0.95	0.95	449.92
40	3	1	1	1459	57566	1421	100	100	334.14	298.29	342	14803	370	100	100	334.14	312.80	108	3677	93	3372	0.95	0.95	396.42
41	3	1	1	1395	58995	1403	100	100	272.53	297.39	369	15035	331	100	100	272.53	312.80	96	3753	88	3460	0.95	0.95	370.80
42	3	1	1	1400	60390	1370	100	100	298.73	297.97	362	15404	361	100	100	298.73	312.80	79	3822	76	3536	0.95	0.95	312.92
43	3	1	1	1398	61888	1381	100	100	239.73	296.12	368	15772	359	100	100	239.73	310.75	88	3940	85	3621	0.95	0.95	325.29
44	3	1	1	1469	64847	1443	100	100	361.02	297.68	379	16151	373	100	100	361.02	312.24	100	4040	92	3713	0.95	0.95	416.42
45	3	1	1	1468	67599	1456	100	100	321.18	298.10	353	16504	362	100	100	321.18	313.13	86	4126	73	3786	0.95	0.95	394.09
46	3	1	1	1464	69105	1449	100	100	362.90	299.51	367	16871	345	100	100	362.90	314.66	103	4229	86	3872	0.95	0.95	466.35
47	3	1	1	1381	69650	1375	100	100	261.69	298.75	369	17240	346	100	100	261.69	313.46	78	4307	72	3944	0.95	0.95	367.42
48	3	1	1	1394	70394	1434	100	100	304.11	298.86	362	17622	368	100	100	304.11	313.46	90	4397	84	4028	0.95	0.95	390.83
49	3	1	1	1388	71772	1378	100	100	295.42	298.80	368	18005	377	100	100	295.42	312.88	84	4481	80	4108	0.95	0.95	412.18
50	3	1	1	1467	73225	1447	100	100	315.56	298.63	346	18354	370	100	100	315.56	313.27	86	4567	80	4188	0.95	0.95	432.81
51	3	1	1	1463	74712	1467	100	100	315.56	299.22	346	18724	359	100	100	315.56	313.00	86	4653	79	4267	0.95	0.95	407.20
52	3	1	1	1388	76080	1388	100	100	246.66	298.12	381	19105	359	100	100	246.66	311.53	108	4706	96	4385	0.95	0.95	374.50
53	3	1	1	1439	77619	1432	100	100	340.63	300.06	376	19481	362	100	100	340.63	314.66	96	4862	86	4461	0.95	0.95	461.54
54	3	1	1	1392	78911	1385	100	100	296.92	299.81	349	19830	340	100	100	296.92	314.75	95	4957	91	4642	0.95	0.95	364.25
55	3	1	1	1448	80357	1437	100	100	305.36	299.81	353	20183	349	100	100	305.36	314.75	85	5042	82	4624	0.		

Replication Number	Independent Variables			Product A												Product B												Product C											
	MAS	MDR	MIX	Period Demand	Total Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle Time	Total Cycle Time	Period Demand	Total Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle Time	Total Cycle Time	Period Demand	Total Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle Time	Total Cycle Time												
	1	2	3	1463	1466	1469	1472	1475	1478	1481	1484	1487	1490	1493	1496	1499	1502	1505	1508	1511	1514	1517	1520	1523	1526	1529	1532	1535											
1	3	1	3	1463	1466	1469	1472	1475	1478	1481	1484	1487	1490	1493	1496	1499	1502	1505	1508	1511	1514	1517	1520	1523	1526	1529	1532	1535											
2	3	1	3	1462	2905	1431	2897	1000	1000	717.38	710.62	1000	1000	1000	705.05	708.73	1000	1000	1000	705.05	708.73	1000	1000	1000	705.05	708.73	1000	1000	705.05	708.73									
3	3	1	3	1503	4408	1488	4385	1000	1000	721.89	711.63	1000	1000	1000	711.63	711.63	1000	1000	1000	711.63	711.63	1000	1000	1000	711.63	711.63	1000	1000	711.63	711.63									
4	5	3	1	1438	5846	1409	5794	1000	1000	723.99	705.24	1000	1000	1000	705.24	705.24	1000	1000	1000	705.24	705.24	1000	1000	1000	705.24	705.24	1000	1000	705.24	705.24									
5	4	3	1	1436	7302	1445	7239	1000	1000	678.42	705.24	1000	1000	1000	705.24	705.24	1000	1000	1000	705.24	705.24	1000	1000	1000	705.24	705.24	1000	1000	705.24	705.24									
6	3	1	3	1470	8772	1454	8693	1000	1000	705.49	704.78	1000	1000	1000	704.78	704.78	1000	1000	1000	704.78	704.78	1000	1000	1000	704.78	704.78	1000	1000	704.78	704.78									
7	6	3	1	1473	10345	1449	10142	1000	1000	705.11	705.11	1000	1000	1000	705.11	705.11	1000	1000	1000	705.11	705.11	1000	1000	1000	705.11	705.11	1000	1000	705.11	705.11									
8	3	1	3	1473	11177	1449	11041	1000	1000	707.77	705.95	1000	1000	1000	705.95	705.95	1000	1000	1000	705.95	705.95	1000	1000	1000	705.95	705.95	1000	1000	705.95	705.95									
9	3	1	3	1473	13190	1449	13046	1000	1000	708.44	706.78	1000	1000	1000	706.78	706.78	1000	1000	1000	706.78	706.78	1000	1000	1000	706.78	706.78	1000	1000	706.78	706.78									
10	3	1	3	1473	14652	1449	14489	1000	1000	710.80	706.86	1000	1000	1000	706.86	706.86	1000	1000	1000	706.86	706.86	1000	1000	1000	706.86	706.86	1000	1000	706.86	706.86									
11	3	1	3	1530	16192	1526	16025	1000	1000	710.80	706.86	1000	1000	1000	706.86	706.86	1000	1000	1000	706.86	706.86	1000	1000	1000	706.86	706.86	1000	1000	706.86	706.86									
12	3	1	3	1482	17684	1488	17513	1000	1000	723.39	707.89	1000	1000	1000	707.89	707.89	1000	1000	1000	707.89	707.89	1000	1000	1000	707.89	707.89	1000	1000	707.89	707.89									
13	3	1	3	2220	18904	1385	18908	1000	1000	707.85	707.89	1000	1000	1000	707.89	707.89	1000	1000	1000	707.89	707.89	1000	1000	1000	707.89	707.89	1000	1000	707.89	707.89									
14	3	1	3	851	20555	1464	20362	1000	1000	710.82	708.19	1000	1000	1000	708.19	708.19	1000	1000	1000	708.19	708.19	1000	1000	1000	708.19	708.19	1000	1000	708.19	708.19									
15	3	1	3	1483	22048	1476	21838	1000	1000	713.96	708.63	1000	1000	1000	708.63	708.63	1000	1000	1000	708.63	708.63	1000	1000	1000	708.63	708.63	1000	1000	708.63	708.63									
16	3	1	3	1414	23462	1385	23223	1000	1000	717.80	709.13	1000	1000	1000	709.13	709.13	1000	1000	1000	709.13	709.13	1000	1000	1000	709.13	709.13	1000	1000	709.13	709.13									
17	3	1	3	1506	24968	1485	24708	1000	1000	706.64	708.62	1000	1000	1000	708.62	708.62	1000	1000	1000	708.62	708.62	1000	1000	1000	708.62	708.62	1000	1000	708.62	708.62									
18	3	1	3	1442	26410	1437	26145	1000	1000	716.11	709.26	1000	1000	1000	709.26	709.26	1000	1000	1000	709.26	709.26	1000	1000	1000	709.26	709.26	1000	1000	709.26	709.26									
19	3	1	3	1552	27962	1509	27654	1000	1000	705.05	709.03	1000	1000	1000	709.03	709.03	1000	1000	1000	709.03	709.03	1000	1000	1000	709.03	709.03	1000	1000	709.03	709.03									
20	3	1	3	1469	29431	1449	29103	1000	1000	701.20	708.64	1000	1000	1000	708.64	708.64	1000	1000	1000	708.64	708.64	1000	1000	1000	708.64	708.64	1000	1000	708.64	708.64									
21	3	1	3	1443	30874	1426	30529	1000	1000	714.85	708.63	1000	1000	1000	708.63	708.63	1000	1000	1000	708.63	708.63	1000	1000	1000	708.63	708.63	1000	1000	708.63	708.63									
22	3	1	3	1436	32310	1431	31960	1000	1000	717.64	709.32	1000	1000	1000	709.32	709.32	1000	1000	1000	709.32	709.32	1000	1000	1000	709.32	709.32	1000	1000	709.32	709.32									
23	3	1	3	1483	33793	1433	33393	1000	1000	708.62	709.29	1000	1000	1000	709.29	709.29	1000	1000	1000	709.29	709.29	1000	1000	1000	709.29	709.29	1000	1000	709.29	709.29									
24	3	1	3	1509	35302	1503	34996	1000	1000	707.20	709.20	1000	1000	1000	709.20	709.20	1000	1000	1000	709.20	709.20	1000	1000	1000	709.20	709.20	1000	1000	709.20	709.20									
25	3	1	3	1472	36774	1468	36564	1000	1000	715.19	708.82	1000	1000	1000	708.82	708.82	1000	1000	1000	708.82	708.82	1000	1000	1000	708.82	708.82	1000	1000	708.82	708.82									
26	3	1	3	1423	38197	1413	37767	1000	1000	707.72	709.26	1000	1000	1000	709.26	709.26	1000	1000	1000	709.26	709.26	1000	1000	1000	709.26	709.26	1000	1000	709.26	709.26									
27	3	1	3	1460	39647	1443	39210	1000	1000	712.79	709.36	1000	1000	1000	709.36	709.36	1000	1000	1000	709.36	709.36	1000	1000	1000	709.36	709.36	1000	1000	709.36	709.36									
28	3	1	3	1465	41102	1441	40651	1000	1000	710.98	709.34	1000	1000	1000	709.34	709.34	1000	1000	1000	709.34	709.34	1000	1000	1000	709.34	709.34	1000	1000	709.34	709.34									
29	3	1	3	1352	42464	1327	41978	1000	1000	716.46	709.86	1000	1000	1000	709.86	709.86	1000	1000	1000	709.86	709.86	1000	1000	1000	709.86	709.86	1000	1000	709.86	709.86									
30	3	1	3	1438	43892	1434	43412	1000	1000	711.76	709.82	1000	1000	1000	709.82	709.82	1000	1000	1000	709.82	709.82	1000	1000	1000	709.82	709.82	1000	1000	709.82	709.82									
31	3	1	3	1438	46330	1416	44828	1000	1000	711.76	709.82	1000	1000	1000	709.82	709.82	1000	1000	1000	709.82	709.82	1000	1000	1000	709.82	709.82	1000	1000	709.82	709.82									
32	3	1	3	1464	46794	1445	46273	1000	1000	707.68	709.85	1000	1000	1000	709.85	709.85	1000	1000	1000	709.85	709.85	1000	1000	1000	709.85	709.85	1000	1000	709.85	709.85									
33	3	1	3	1435	48229	1419	47692	1000	1000	706.14	709.71	1000	1000	1000	709.71	709.71	1000	1000	1000	709.71	709.71	1000	1000	1000	709.71	709.71	1000	1000	709.71	709.71									
34	3	1	3	1463	48692	1440	48132	1000	1000	716.88	709.62	1000	1000	1000	709.62	709.62	1000	1000	1000	709.62	709.62	1000	1000	1000	709.62	709.62	1000	1000	709.62	709.62									
35	3	1	3	1411	51093	1407	50539	1000	1000	706.69	709.63	1000	1000	1000	709.63	709.63	1000	1000	1000	709.63	709.63	1000	1000	1000	709.63	709.63	1000	1000	709.63	709.63									
36	3	1	3	1463	52546	1431	51970	1000	1000	716.73	710.02	1000	1000	1000	710.02	710.02	1000	1000	1000	710.02	710.02	1000	1000	1000	710.02	710.02	1000	1000	710.02	710.02									
37	3	1																																					

Replication Number	Independent Variables			Product A												Product B												Product C											
	MAS	MDR	MIX	Period Demand	Total Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle Time	Total Cycle Time	Period Demand	Total Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle Time	Total Cycle Time	Period Demand	Total Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle Time	Total Cycle Time												
	1	2	1	1484	1464	1460	1460	1.00	1.00	311.75	311.75	391	386	338	336	1.00	1.00	303.18	303.18	81	81	77	77	0.97	0.97	360.87	360.87												
1	1	1	1	1484	1464	1460	1460	1.00	1.00	311.75	311.75	391	386	338	336	1.00	1.00	303.18	303.18	81	81	77	77	0.97	0.97	360.87	360.87												
2	1	1	1	1484	1464	1460	1460	1.00	1.00	299.22	299.22	338	336	338	336	1.00	1.00	304.24	304.24	89	89	170	170	0.95	0.95	430.13	430.13												
3	2	1	1	1446	2901	1435	2885	1.00	1.00	325.06	312.03	354	343	343	343	1.00	1.00	323.94	310.60	91	91	271	271	0.93	0.93	393.01	394.99												
4	2	1	1	1374	6713	1329	6665	1.00	1.00	248.16	297.02	369	1462	360	1427	1.00	1.00	250.95	296.69	69	368	96	346	0.95	0.95	360.83	382.83												
5	4	2	1	1374	7087	1378	7033	1.00	1.00	233.22	294.52	345	1797	343	1470	1.00	1.00	220.83	281.90	69	436	66	411	0.95	0.95	340.84	377.50												
6	2	1	1	1420	8507	1398	8431	1.00	1.00	319.80	290.37	363	2147	338	2048	1.00	1.00	322.03	288.41	97	533	66	487	0.95	0.95	424.77	395.68												
7	2	1	1	1424	9031	1417	8948	1.00	1.00	228.91	281.34	343	2480	338	2444	1.00	1.00	248.86	282.91	85	618	80	577	0.95	0.95	379.28	394.79												
8	2	1	1	1424	11356	1417	11285	1.00	1.00	258.12	278.86	343	2833	337	2780	1.00	1.00	264.89	280.73	85	704	80	659	1.00	0.96	378.80	393.79												
9	2	1	1	1424	12780	1417	12681	1.00	1.00	246.80	274.82	343	3175	337	3116	1.00	1.00	260.48	278.54	85	789	80	738	0.96	0.96	374.60	382.79												
10	2	1	1	1424	14204	1417	14088	1.00	1.00	302.86	277.73	343	3518	336	3462	1.00	1.00	280.15	279.87	85	874	80	818	0.96	0.96	371.48	381.68												
11	2	1	1	1432	16336	1415	16513	1.00	1.00	251.97	275.38	353	3871	348	3800	1.00	1.00	270.93	278.87	87	961	82	900	0.96	0.96	346.46	378.38												
12	2	1	1	1439	17075	1452	16946	1.00	1.00	237.51	272.18	337	4208	329	4129	1.00	1.00	246.11	276.26	84	1046	81	981	0.96	0.96	333.81	374.70												
13	2	1	1	1387	18462	1370	18315	1.00	1.00	208.75	267.51	358	4656	352	4481	1.00	1.00	222.87	272.05	88	1113	63	1044	0.96	0.96	334.80	372.28												
14	2	1	1	1502	18964	1484	18769	1.00	1.00	332.22	272.36	351	4917	346	4827	1.00	1.00	332.18	276.36	86	1209	84	1128	0.96	0.96	372.55	372.30												
15	2	1	1	1478	21442	1478	21277	1.00	1.00	290.50	273.62	364	5281	355	5182	1.00	1.00	289.57	277.95	79	1288	79	1207	0.96	0.96	377.46	372.84												
16	2	1	1	1460	22892	1429	22706	1.00	1.00	297.77	275.14	329	5610	321	5503	1.00	1.00	308.12	279.71	83	1381	85	1292	0.96	0.96	393.77	374.03												
17	2	1	1	1407	24289	1384	24100	1.00	1.00	257.33	274.11	368	5978	367	6870	1.00	1.00	255.24	278.18	84	1465	77	1369	1.00	0.97	368.56	373.16												
18	2	1	1	1421	25720	1418	25518	1.00	1.00	306.14	275.89	367	6346	362	6232	1.00	1.00	293.67	279.08	83	1568	89	1468	0.97	0.97	385.12	373.89												
19	2	1	1	1483	27203	1480	26968	1.00	1.00	340.28	279.42	379	6724	376	6608	1.00	1.00	348.89	283.61	80	1648	81	1539	0.97	0.97	395.55	375.03												
20	2	1	1	1391	28594	1386	28384	1.00	1.00	224.33	276.73	350	7074	343	6951	1.00	1.00	229.30	280.93	84	1732	82	1621	0.97	0.97	390.90	371.72												
21	2	1	1	1448	30042	1436	29820	1.00	1.00	295.34	277.63	381	7465	375	7326	1.00	1.00	295.00	281.65	81	1813	79	1700	0.97	0.97	401.42	375.10												
22	2	1	1	1462	31494	1430	31260	1.00	1.00	323.42	279.63	372	7827	367	7683	1.00	1.00	320.17	283.44	100	1913	82	1782	0.97	0.97	416.91	375.07												
23	2	1	1	1410	32904	1398	32648	1.00	1.00	303.68	280.86	343	8170	346	8037	1.00	1.00	289.47	283.70	91	2004	84	1896	0.97	0.97	395.73	375.55												
24	2	1	1	1495	34699	1429	34077	1.00	1.00	345.00	279.60	375	8545	368	8397	1.00	1.00	347.04	282.07	86	2000	78	1944	0.97	0.97	325.46	373.54												
25	2	1	1	1389	35786	1425	35592	1.00	1.00	368.20	282.88	352	8897	344	8741	1.00	1.00	365.13	285.81	101	2181	89	2033	0.97	0.97	478.30	378.17												
26	2	1	1	1375	37143	1359	36871	1.00	1.00	306.67	283.83	389	9226	379	9120	1.00	1.00	312.26	286.91	90	2287	89	2122	0.97	0.97	483.44	378.23												
27	2	1	1	1481	38634	1475	38346	1.00	1.00	400.66	288.42	362	9646	354	9474	1.00	1.00	405.20	290.93	97	2384	78	2200	0.97	0.97	489.95	383.51												
28	2	1	1	1407	40041	1407	39753	1.00	1.00	281.07	288.16	393	10094	384	9898	1.00	1.00	282.09	290.97	76	2480	70	2270	0.97	0.97	413.02	384.42												
29	2	1	1	1481	41622	1469	41212	1.00	1.00	290.98	288.26	351	10492	347	10205	1.00	1.00	306.85	291.51	84	2544	72	2342	0.97	0.97	376.31	384.14												
30	2	1	1	1441	42963	1441	42563	1.00	1.00	367.88	290.95	381	10753	363	10588	1.00	1.00	367.31	293.77	110	2654	95	2437	0.97	0.97	486.21	386.17												
31	2	1	1	1447	44410	1430	44063	1.00	1.00	266.68	290.26	385	11138	379	10947	1.00	1.00	278.75	293.25	87	2741	85	2622	0.97	0.97	372.52	385.71												
32	2	1	1	1407	46817	1383	46476	1.00	1.00	313.11	290.96	387	11625	376	11323	1.00	1.00	318.55	294.09	89	2830	82	2604	0.97	0.97	414.28	386.61												
33	2	1	1	1485	47302	1472	46948	1.00	1.00	295.11	291.09	342	11887	338	11659	1.00	1.00	283.40	294.07	82	2912	72	2676	0.97	0.97	380.86	386.45												
34	2	1	1	1490	48792	1480	48428	1.00	1.00	329.70	292.27	352	12219	350	12009	1.00	1.00	317.06	294.74	88	3000	78	2754	0.97	0.97	413.64	387.22												
35	2	1	1	1465	50247	1436	49864	1.00	1.00	298.17	292.44	353	12572	346	12354	1.00	1.00	316.87	295.33	102	3102	96	2850	0.97	0.97	364.06	386.44												
36	2	1	1	1440	51687	1425	51289	1.00	1.00	316.46	293.08	309	12881	306	12660	1.00	1.00	335.46	296.30	99	3201	89	2939	0.97	0.97	416.83	387.33												
37	2	1	1	1481	53168	1470	52769	1.00	1.00	316.56	293.79	353	13234	342	13002	1.00	1.00	317.59	296.86	96	3297	90	3029	0.97	0.97	378.92	387.08												
38	2	1	1	1428	54596	1416	54175	1.00	1.00	278.10	293.38	345	13579	336	13338	1.00	1.00	291.53	295.97	85	3382	82	3111	0.97	0.97	378.35	386.95												
39	2	1	1	1406	56002	1407	55592	1.00	1.00	298.91	293.62	386	13965	385	13723	1.00	1.00	295.97	296.97	76	3468	72	3183	0.51	0.66	389.50	386.91												
40	2	1	1	1470	57472	1454	57036	1.00	1.00	270.38	293.83	330	14295	326	14049	1.00	1.00	272.70	295.43	85	3543	83	3266	0.96	0.96	379.04	386.71												
41	2	1	1	1448	58920	1432	58488	1.00	1.00	329.68	293.83	336	14831	329	14678	1.00	1.00	310.10	296.88	104	3647	100	3366	0.96	0.96	431.81	386.05												
42	2	1	1	1430	60350	1414	59882	1.00	1.00	367.35	295.93	376	15007	370	14848	1.00	1.00	360.86	297.51	744269	38355	75	3559	0.96	0.96	421.83	389.00												
43	2	1	1	1415	61765	1400	61282	1.00	1.00	252.45	294.35	366	15375	363	15111	1.00	1.00	244.83	296.23	(14259)	38355	75	3559	0.96	0.96	366.56	386.52												
44	2	1	1	1447	63212	1436	62738	1.00	1.00	365.46	296.00	366	15743	361	15472	1.00	1.00	363.95	297.81	82	3827	81	3620	0.96	0.96	448.75	389.89												
45	2	1	1	1405	64817	1384	64122	1.00	1.00	246.21	294.98	368	16111	362	15834	1.00	1.00	260.82	296.69	87	4014	79	3699	0.96	0.96	374.01	388.58												
46	2	1	1	1387	66004	1383	65505	1.00	1.00	266.10	294.38	344	16465	340	16174	1.00	1.00	260.54	295.93	90	4104	83	3782	0.96	0.96	465.92	390.28												
47	2	1	1	1420	67424	1417	66922	1.00	1.00	427.55																													

Replication Number	Independent Variables			Product A												Product B												Product C											
	MAS	MDR	MIK	Period Demand	Total Shipped	Period Shipped	Period			Total			Period Demand	Total Shipped	Period Shipped	Period			Total			Period Demand	Total Shipped	Period Shipped	Period			Total											
							DFR	Total	Cycle Time	DFR	Total	Cycle Time				DFR	Total	Cycle Time	DFR	Total	Cycle Time				DFR	Total	Cycle Time	DFR	Total	Cycle Time									
1	2	1	2	1411	1333	1333	0.97	0.97	803.14	786	786	744	744	744	0.97	0.97	673.38	673.38	673.38	182	182	182	182	182	182	0.37	0.37	332.46	332.46	332.46									
2	2	1	2	1442	2853	1334	0.96	0.96	788.69	756	1642	731	1475	1475	0.97	0.97	667.67	670.55	670.55	214	406	57	214	406	57	0.29	0.33	349.82	341.62	341.62									
3	2	1	2	1443	4296	1395	0.96	0.96	784.41	716	2258	689	1194	1194	0.97	0.97	671.27	670.78	670.78	148	562	49	148	562	49	0.29	0.32	408.81	392.59	392.59									
4	2	1	2	1396	5902	1373	0.96	0.96	808.03	707	3945	676	2840	2840	0.97	0.97	667.76	667.68	667.68	182	734	56	182	734	56	0.22	0.32	404.01	373.48	373.48									
5	2	1	2	1402	7184	1433	0.96	0.96	800.63	680	3945	669	3509	3509	0.97	0.97	667.87	666.81	666.81	187	901	47	187	901	47	0.20	0.31	417.68	381.47	381.47									
6	2	1	2	1430	8514	1396	0.96	0.96	800.63	727	4372	700	4020	4020	1.00	0.98	674.89	667.32	667.32	173	1074	59	173	1074	59	0.31	0.31	425.64	399.62	399.62									
7	2	1	2	1449	10063	1382	0.96	0.96	801.88	727	5026	703	4912	4912	0.98	0.98	674.31	669.32	669.32	172	1246	49	172	1246	49	0.31	0.31	403.70	391.50	391.50									
8	2	1	2	1448	11611	1382	0.96	0.96	803.69	727	5828	703	6614	6614	0.98	0.98	676.31	669.32	669.32	172	1418	48	172	1418	48	0.31	0.31	407.46	393.29	393.29									
9	2	1	2	1446	12860	1382	0.96	0.96	805.51	727	6653	703	8317	8317	0.98	0.98	678.31	670.32	670.32	172	1589	48	172	1589	48	0.31	0.31	411.23	395.27	395.27									
10	2	1	2	1446	14408	1382	0.96	0.96	807.32	727	7480	703	9199	9199	0.98	0.98	680.31	673.32	673.32	172	1761	48	172	1761	48	0.31	0.31	414.89	397.16	397.16									
11	2	1	2	1489	16897	1419	0.96	0.96	817.37	711	7991	700	7719	7719	0.98	0.98	688.29	673.86	673.86	187	1948	46	187	1948	46	0.31	0.31	380.97	393.37	393.37									
12	2	1	2	1424	17321	1334	0.96	0.96	823.42	740	8731	710	8429	8429	0.98	0.98	688.29	673.86	673.86	174	2122	62	174	2122	62	0.31	0.31	465.84	400.58	400.58									
13	2	1	2	1630	18851	1446	0.96	0.96	810.42	689	9420	662	9081	9081	0.98	0.98	682.43	676.90	676.90	175	2297	43	175	2297	43	0.31	0.31	384.81	398.57	398.57									
14	2	1	2	1423	20274	1366	0.96	0.96	779.02	723	10143	707	9798	9798	0.98	0.98	682.43	676.90	676.90	181	2478	56	181	2478	56	0.31	0.31	372.20	387.45	387.45									
15	2	1	2	1427	21701	1361	0.96	0.96	806.23	731	10874	697	10495	10495	0.98	0.98	683.08	677.48	677.48	188	2648	52	188	2648	52	0.31	0.31	400.87	400.37	400.37									
16	2	1	2	1342	23043	1280	0.96	0.96	806.23	718	11592	701	11196	11196	0.98	0.98	683.08	677.88	677.88	184	2840	76	184	2840	76	0.31	0.31	446.84	404.47	404.47									
17	2	1	2	1433	24476	1352	0.96	0.96	801.63	729	12321	714	11910	11910	0.98	0.98	680.05	678.01	678.01	174	3014	97	174	3014	97	0.31	0.31	414.87	405.13	405.13									
18	2	1	2	1423	25899	1356	0.96	0.96	790.64	696	13017	672	12582	12582	0.98	0.98	688.46	677.50	677.50	167	3171	66	167	3171	66	0.31	0.31	457.76	408.70	408.70									
19	2	1	2	1477	27369	1401	0.96	0.96	832.41	763	13780	731	13313	13313	0.98	0.98	678.59	677.56	677.56	181	3352	41	181	3352	41	0.31	0.31	326.34	405.37	405.37									
20	2	1	2	1477	28846	1412	0.96	0.96	816.55	690	14470	667	13980	13980	0.98	0.98	678.59	677.56	677.56	180	3542	56	180	3542	56	0.31	0.31	402.70	405.23	405.23									
21	2	1	2	1384	30230	1302	0.96	0.96	794.78	759	15229	747	14727	14727	0.98	0.98	681.26	677.52	677.52	180	3691	80	180	3691	80	0.31	0.31	470.86	408.72	408.72									
22	2	1	2	1394	31824	1300	0.96	0.96	789.46	736	15965	703	16430	16430	0.98	0.98	681.26	677.52	677.52	182	3873	58	182	3873	58	0.31	0.31	397.86	408.18	408.18									
23	2	1	2	1377	33001	1348	0.96	0.96	789.46	714	16679	695	16125	16125	0.98	0.98	683.27	674.34	674.34	187	4060	80	187	4060	80	0.31	0.31	416.87	409.54	409.54									
24	2	1	2	1601	34602	1422	0.96	0.96	789.46	701	17390	675	16920	16920	0.98	0.98	683.27	674.34	674.34	188	4208	48	188	4208	48	0.31	0.31	427.44	409.24	409.24									
25	2	1	2	1463	36865	1416	0.96	0.96	801.82	672	18052	651	17451	17451	0.98	0.98	688.15	673.25	673.25	165	4371	64	165	4371	64	0.31	0.31	427.46	409.87	409.87									
26	2	1	2	1413	37378	1353	0.96	0.96	786.00	759	18811	721	18172	18172	0.98	0.98	688.15	673.25	673.25	166	4569	51	166	4569	51	0.31	0.31	380.63	407.81	407.81									
27	2	1	2	1416	38794	1354	0.96	0.96	816.63	769	19663	732	18904	18904	0.98	0.98	688.15	673.25	673.25	166	4761	51	166	4761	51	0.31	0.31	380.76	408.86	408.86									
28	2	1	2	1442	40238	1381	0.96	0.96	811.83	769	20332	724	19628	19628	0.98	0.98	673.92	673.86	673.86	186	4837	41	186	4837	41	0.31	0.31	389.19	405.58	405.58									
29	2	1	2	1414	41650	1378	0.96	0.96	748.33	684	21016	670	20288	20288	0.98	0.98	685.23	672.13	672.13	187	5119	60	187	5119	60	0.31	0.31	389.19	405.58	405.58									
30	2	1	2	1481	43131	1380	0.96	0.96	828.98	733	21749	697	20995	20995	0.98	0.98	685.23	672.13	672.13	187	5306	51	187	5306	51	0.31	0.31	435.40	405.46	405.46									
31	2	1	2	1393	44624	1335	0.96	0.96	789.05	788	22182	715	21710	21710	0.98	0.98	685.23	672.13	672.13	183	5489	59	183	5489	59	0.31	0.31	401.05	405.46	405.46									
32	2	1	2	1471	46995	1414	0.96	0.96	788.67	731	22880	694	22374	22374	0.98	0.98	685.23	672.13	672.13	185	5672	51	185	5672	51	0.31	0.31	389.80	404.93	404.93									
33	2	1	2	1405	47400	1359	0.96	0.96	792.29	788	23188	702	22883	22883	0.98	0.98	685.23	672.13	672.13	187	5854	62	187	5854	62	0.31	0.31	415.63	405.30	405.30									
34	2	1	2	1417	48817	1359	0.96	0.96	775.12	724	24607	694	23788	23788	0.98	0.98	681.16	670.90	670.90	180	6031	56	180	6031	56	0.31	0.31	374.45	404.36	404.36									
35	2	1	2	1426	50243	1321	0.96	0.96	730.64	718	25325	698	24466	24466	0.98	0.98	637.61	609.95	609.95	182	6213	60	182	6213	60	0.31	0.31	394.80	404.06	404.06									
36	2	1	2	1358	51601	1337	0.96	0.96	800.20	700	26025	688	25144	25144	0.98	0.98	688.15	687.11	687.11	180	6379	60	180	6379	60	0.31	0.31	401.78	403.99	403.99									
37	2	1	2	1464	53095	1409	0.96	0.96	806.74	707	26732	688	26532	26532	0.98	0.98	689.36	687.17	687.17	188	6567	49	188	6567	49	0.31	0.31	365.60	403.05	403.05									
38	2	1	2	1447	54502	1380	0.96	0.96	828.43	742	27474	694	26526	26526	0.98	0.98	687.43	687.70	687.70	180	6747	49	180	6747	49	0.31	0.31	366.26	402.89	402.89									
39	2	1	2	1424	56926	1394	0.96	0.96	787.80	680	28154	661	27187	27187	0.98	0.98	684.08	687.15	687.15	179	6926	61	179	6926	61	0.31	0.31	402.89	402.89	402.89									
40	2	1	2	1412	57338	1361	0.96	0.96	806.33	715	28869	704	27891	27891	0.98	0.98	680.92	687.75	687.75	177	7103	55	177	7103	55	0.31	0.31	418.25	403.22	403.22									
41	2	1	2	1437	58815	1428	0.96	0.96	795.97	700	29599	679	28570	28570	0.98	0.98	685.24	688.48	688.48	180	7283	46	180	7283	46	0.31	0.31	400.90	403.15	403.15									
42	2	1	2	1387	60302	1346	0.96	0.96	821.94	742	30311	714	29294	29294	0.98	0.98	685.24	688.48	688.48	179	7462	60	179	7462	60	0.31	0.31	400.90	403.15	403.15									
43	2	1	2	1386	61686	1347	0.96	0.96	771.57	664	30875	647																											

Replication Number	Independent Variables			Product A												Product B												Product C											
	MAS	MDR	MIX	Period Demand	Total Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle Time	Total Cycle Time	Period Demand	Total Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle Time	Total Cycle Time	Period Demand	Total Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle Time	Total Cycle Time												
				1	1415	1416	1417	1418	1419	1420	1421	1422	1423	1424	1425	1426	1427	1428	1429	1430	1431	1432	1433	1434	1435	1436	1437	1438	1439	1440									
1	1	1	1	1415	1416	1417	1418	1419	1420	1421	1422	1423	1424	1425	1426	1427	1428	1429	1430	1431	1432	1433	1434	1435	1436	1437	1438	1439	1440										
2	1	1	1	1465	2870	1419	2820	0.99	0.99	296.63	295.44	343	339	1000	1000	289.71	286.71	81	81	85	85	1000	1000	430.02	430.02														
3	1	1	1	1465	2870	1419	2820	0.99	0.99	296.63	295.44	343	339	1000	1000	289.71	286.71	81	81	85	85	1000	1000	430.02	430.02														
4	1	1	1	1465	2870	1419	2820	0.99	0.99	296.63	295.44	343	339	1000	1000	289.71	286.71	81	81	85	85	1000	1000	430.02	430.02														
5	1	1	1	1374	6075	1362	6003	0.99	0.99	290.49	362.33	364	348	1000	1000	309.46	331.26	87	87	86	86	1000	1000	362.40	362.40														
6	1	1	1	1432	8542	1419	8425	0.99	0.99	290.06	334.41	389	373	1000	1000	321.32	322.46	76	76	541	541	1000	1000	362.76	362.76														
7	1	1	1	1440	9682	1419	9644	0.99	0.99	367.63	337.70	373	365	1000	1000	331.16	323.76	89	89	630	630	1000	1000	424.70	424.70														
8	1	1	1	1440	11423	1419	11262	0.99	0.99	364.58	341.16	373	365	1000	1000	333.70	325.03	89	89	719	719	1000	1000	426.36	426.36														
9	1	1	1	1440	12863	1419	12651	0.99	0.99	371.33	344.64	373	365	1000	1000	338.73	328.28	89	89	808	808	1000	1000	427.92	427.92														
10	1	1	1	1440	14303	1419	14089	0.99	0.99	375.05	347.91	373	365	1000	1000	342.35	332.56	89	89	897	897	1000	1000	429.48	429.48														
11	1	1	1	1441	15744	1409	15508	0.99	0.99	437.17	350.02	347	340	1000	1000	389.35	333.54	110	107	104	104	1000	1000	468.41	468.41														
12	1	1	1	1416	17160	1407	16915	0.99	0.99	295.67	351.00	378	368	1000	1000	291.06	329.87	716	1083	74	1039	0.99	0.99	466.35	427.74														
13	1	1	1	1430	18590	1431	18346	0.99	0.99	296.87	346.70	373	365	1000	1000	307.79	328.14	83	1166	79	1118	0.99	0.99	376.51	424.12														
14	1	1	1	1486	20076	1431	19777	0.99	0.99	385.95	346.54	377	373	1000	1000	384.88	332.22	100	1266	92	1210	0.99	0.99	439.38	423.28														
15	1	1	1	1402	21478	1407	21184	0.99	0.99	407.36	353.38	334	334	1000	1000	377.26	335.03	109	1375	105	1316	0.99	0.99	469.47	428.01														
16	1	1	1	1467	22945	1466	22640	0.99	0.99	333.32	352.09	291	275	1000	1000	321.05	334.33	83	1468	88	1403	0.99	0.99	469.64	429.95														
17	1	1	1	1446	24391	1367	24027	0.99	0.99	546.59	363.26	368	361	1000	1000	466.75	342.28	103	1571	101	1504	0.99	0.99	536.12	437.08														
18	1	1	1	1364	25755	1362	25389	0.99	0.99	274.34	358.49	374	360	1000	1000	270.39	338.06	89	1680	85	1609	0.99	0.99	349.59	432.40														
19	1	1	1	1391	27146	1353	26772	0.99	0.99	230.34	351.67	387	368	1000	1000	229.24	331.95	86	1746	84	1673	0.99	0.99	318.08	426.66														
20	1	1	1	1412	28558	1399	28171	0.99	0.99	233.67	346.00	338	336	1000	1000	240.92	328.07	86	1832	79	1752	0.99	0.99	335.51	422.55														
21	1	1	1	1429	29897	1412	29583	0.99	0.99	309.55	344.26	387	383	1000	1000	305.20	326.90	92	1911	78	1830	0.99	0.99	383.84	420.60														
22	1	1	1	1425	31412	1405	30968	0.99	0.99	331.91	343.70	345	345	1000	1000	332.34	325.83	92	2003	86	1916	0.99	0.99	393.94	419.69														
23	1	1	1	1443	32855	1428	32416	0.99	0.99	337.27	346.06	363	353	1000	1000	373.05	327.84	92	2095	87	2003	0.99	0.99	455.01	421.25														
24	1	1	1	1436	34300	1420	33846	0.99	0.99	349.18	341.07	362	352	1000	1000	252.81	324.76	75	2170	73	2075	0.99	0.99	351.01	415.73														
25	1	1	1	1379	35669	1363	35286	0.99	0.99	286.89	340.07	375	363	1000	1000	287.11	323.18	87	2257	82	2168	0.99	0.99	354.57	416.34														
26	1	1	1	1438	37107	1357	36665	0.99	0.99	461.32	344.67	467	446	1000	1000	413.25	327.00	111	2366	108	2266	0.99	0.99	467.72	416.76														
27	1	1	1	1485	38602	1507	38072	0.99	0.99	352.15	344.67	365	360	1000	1000	348.92	327.74	88	2462	80	2346	0.99	0.99	362.66	417.87														
28	1	1	1	1442	40034	1428	39500	0.99	0.99	364.23	346.57	376	370	1000	1000	368.52	327.93	88	2540	82	2428	0.99	0.99	421.13	417.98														
29	1	1	1	1448	41484	1446	40946	0.99	0.99	268.83	342.86	349	345	1000	1000	269.90	326.93	83	2623	80	2508	0.99	0.99	347.76	415.74														
30	1	1	1	1395	42838	1342	42288	0.99	0.99	304.73	341.63	371	369	1000	1000	321.42	324.31	84	2707	83	2591	0.99	0.99	390.24	414.92														
31	1	1	1	1421	43709	1421	43709	0.99	0.99	347.19	341.63	308	303	1000	1000	340.32	324.75	89	2806	87	2688	0.99	0.99	386.54	414.21														
32	1	1	1	1421	46996	1404	46113	0.99	0.99	277.57	339.63	348	339	1000	1000	274.13	323.24	92	2888	87	2775	0.99	0.99	364.46	412.85														
33	1	1	1	1465	47171	1465	46668	0.99	0.99	320.95	339.24	362	352	1000	1000	293.86	322.35	76	2974	72	2847	0.99	0.99	381.41	411.86														
34	1	1	1	1421	48592	1405	47973	0.99	0.99	322.17	338.74	328	323	1000	1000	321.68	322.34	110	3084	106	2953	0.99	0.99	424.12	412.30														
35	1	1	1	1413	50005	1374	49347	0.99	0.99	378.25	339.84	392	383	1000	1000	342.45	322.96	88	3172	83	3036	0.99	0.99	468.26	413.83														
36	1	1	1	1411	51416	1369	50746	0.99	0.99	257.14	337.56	341	332	1000	1000	265.32	321.46	90	3262	84	3120	0.99	0.99	363.32	412.47														
37	1	1	1	1359	52775	1356	52102	0.99	0.99	218.06	334.45	360	350	1000	1000	213.64	318.46	70	3332	69	3189	0.99	0.99	337.60	410.85														
38	1	1	1	1395	54160	1369	53471	0.99	0.99	303.59	333.66	331	326	1000	1000	291.68	317.84	91	3423	87	3276	0.99	0.99	395.03	410.43														
39	1	1	1	1404	55694	1380	54851	0.99	0.99	264.90	331.63	362	352	1000	1000	272.63	316.70	91	3514	87	3363	0.99	0.99	360.95	406.15														
40	1	1	1	1407	56971	1415	56266	0.99	0.99	354.09	331.63	343	344	1000	1000	344.19	317.35	91	3605	83	3446	0.99	0.99	476.82	410.78														
41	1	1	1	1424	58395	1393	57659	0.99	0.99	310.99	331.96	342	342	1000	1000	297.06	318.64	83	3688	78	3524	0.99	0.99	395.48	410.22														
42	1	1	1	1388	59753	1355	59014	0.99	0.99	258.23	331.96	362	350	1000	1000	278.34	315.71	90	3778	83	3607	0.99	0.99	405.01	410.10														
43	1	1	1	1433	61216	1421	60485	0.99	0.99	258.23	330.13	366	354	1000	1000	265.22	314.50	87	3866	86	3693	0.99	0.99	341.82	405.51														
44	1	1	1	1431	62647	1422	61867	0.99	0.99	251.07	328.30	347	340	1000	1000	259.70	313.30	75	3940	69	3762	0.99	0.99	360.53	407.63														
45	1	1	1	1489	64146	1467	63324	0.99	0.99	476.94	331.72	371	361	1000	1000	421.43	315.77	84	4034	87	3849	0.99	0.99	467.00	409.65														
46	1	1	1	1371	65517	1364	64888	0.99	0.99	308.01	331.22	350	340	1000	1000	300.93	315.46	89	4122	86	3835	0.99	0.99	468.28	409.62														
47	1	1	1	1486	67003	1462	66140	0.99	0.99	297.67	330.47	371	368	1000	1000	280.28	314.68	84	4206	73	4008	0.99	0.99	364.80	408.80														
48	1	1	1	1368	68371	1354	67464	0.99	0.99	278.65	329.49	378	374	1000	1000	280.99	313.30	78	4284	76	4084	0.99	0.99	368.82	407.87														
49	1	1	1	1416	71240	1381	70319	0.99	0.99	253.85	328.16	408	402	1000	1000	264.48	312.17	72	4356	69	4163	0.99	0.99	372.96	407.29														
50	1	1	1	1416	72420	1381	71391	0.99	0.99	253.85	326.69	370	365	1000	1000	257.26	311.07	87	4463	90	4243	0.99	0.99	347.42	406.02														
51	1																																						

Replication Number	Independent Variables			Product A										Product B										Product C									
	MAS	MDR	MIK	Period Demand	Total Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle Time	Total Cycle Time	Period Demand	Total Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle Time	Total Cycle Time	Period Demand	Total Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle Time	Total Cycle Time						
	1	1	1	1477	1477	826	826	0.60	0.80	754.99	754.99	895	895	871	871	0.98	0.98	651.90	651.90	185	185	162	162	0.84	0.84	669.35	669.35						
2	1	1	1	1465	2942	811	1637	0.56	0.68	736.14	746.65	730	1463	737	1476	0.98	0.98	639.31	648.12	175	350	161	323	0.96	0.90	666.88	667.97						
3	1	1	1	1367	4309	808	2446	0.56	0.68	788.10	759.68	730	2193	718	1436	0.98	0.98	653.63	648.12	175	535	155	478	0.93	0.91	664.08	666.71						
4	1	1	1	1469	5768	773	3218	0.54	0.57	754.54	753.94	720	2900	741	2897	0.98	0.98	660.21	660.26	184	729	161	639	0.95	0.92	618.63	664.57						
5	1	1	1	1387	7165	770	3988	0.57	0.57	760.77	760.17	720	3870	696	3593	1.00	0.98	666.85	660.26	184	915	170	809	0.92	0.92	676.66	660.19						
6	1	1	1	1422	8577	842	4830	0.57	0.57	748.25	758.40	732	5117	711	4978	0.99	0.99	638.87	648.38	174	1089	161	970	0.92	0.92	647.14	667.19						
7	1	1	1	1406	10073	843	6673	0.57	0.57	744.71	758.63	732	5848	711	6689	0.99	0.99	642.37	647.05	178	1404	166	1287	0.92	0.92	635.68	664.20						
8	1	1	1	1486	11668	843	7358	0.57	0.57	741.17	754.86	732	6580	711	6400	0.99	0.98	641.04	646.35	178	1623	166	1487	0.92	0.92	623.82	664.23						
9	1	1	1	1486	14659	843	8200	0.57	0.57	737.63	753.08	732	7311	711	6400	0.99	0.98	639.70	646.71	178	1801	166	1593	0.92	0.92	617.84	646.23						
10	1	1	1	1366	16925	749	8949	0.57	0.57	746.56	752.46	764	8075	737	7848	0.99	0.98	669.55	647.01	183	1984	168	1761	0.92	0.92	646.54	646.26						
11	1	1	1	1529	17464	869	9818	0.57	0.57	746.13	751.90	702	8777	690	8538	0.99	0.98	642.18	646.62	186	2160	149	1910	0.92	0.92	677.31	647.76						
12	1	1	1	1432	18886	770	10568	0.57	0.57	744.75	751.38	758	9635	729	9287	0.99	0.98	666.79	647.42	181	2341	163	2073	0.92	0.92	661.50	648.84						
13	1	1	1	1470	20396	752	11340	0.57	0.57	724.69	749.61	716	10251	696	9903	0.99	0.98	657.44	648.12	184	2538	172	2246	0.92	0.92	661.50	648.81						
14	1	1	1	1464	21810	863	12303	0.57	0.57	763.02	750.66	705	10956	689	10652	0.99	0.98	653.23	646.51	184	2834	172	2246	0.92	0.92	641.23	648.31						
15	1	1	1	1400	23210	783	13066	0.57	0.57	763.36	751.42	764	11720	746	11367	0.99	0.98	647.12	646.55	178	2888	157	2541	0.92	0.92	660.28	648.37						
16	1	1	1	1465	24675	940	14026	0.57	0.57	734.72	750.36	729	12438	690	12087	0.99	0.98	652.20	646.16	183	3026	135	2676	0.92	0.92	628.75	648.33						
17	1	1	1	1306	26071	852	14878	0.57	0.57	766.86	751.37	746	13913	716	13508	0.99	0.98	629.50	643.61	184	3100	149	2825	0.92	0.92	666.34	648.28						
18	1	1	1	1418	27489	859	16737	0.57	0.57	745.58	750.96	724	14637	683	14191	0.99	0.98	662.52	644.52	201	3367	171	3144	0.92	0.92	669.86	646.82						
19	1	1	1	1426	28915	770	16507	0.57	0.57	756.58	751.22	724	15388	727	14918	0.99	0.98	638.98	644.25	175	3742	180	3304	0.92	0.92	620.18	647.30						
20	1	1	1	1463	30368	816	17307	0.57	0.57	746.45	750.96	725	16113	711	16029	0.99	0.98	645.79	644.32	180	3932	168	3472	0.92	0.92	633.46	646.63						
21	1	1	1	1406	31824	816	18123	0.57	0.57	765.71	751.73	716	16829	689	16318	0.99	0.98	654.36	645.47	189	4128	170	3642	0.92	0.92	642.56	646.44						
22	1	1	1	1407	33321	755	18378	0.57	0.57	755.05	751.73	707	17639	679	16937	0.99	0.98	654.36	645.47	189	4327	163	3905	0.92	0.92	611.80	644.06						
23	1	1	1	1431	34662	806	19694	0.57	0.57	755.05	751.73	689	18226	677	17674	0.98	0.88	642.98	645.34	171	4469	147	3862	0.92	0.92	633.87	644.54						
24	1	1	1	1448	36110	880	20664	0.57	0.57	755.05	751.73	689	18226	677	17674	0.98	0.88	642.98	645.34	171	4469	147	3862	0.92	0.92	633.87	644.54						
25	1	1	1	1472	37662	888	21462	0.57	0.57	766.26	752.89	717	18942	687	18371	0.99	0.98	642.70	646.24	169	4607	153	4105	0.92	0.92	612.88	645.36						
26	1	1	1	1421	39003	808	22800	0.57	0.57	772.76	752.89	716	19658	688	19099	0.99	0.98	660.23	646.42	167	4864	162	4267	0.92	0.92	667.85	645.91						
27	1	1	1	1413	40416	839	23969	0.57	0.57	758.62	752.84	748	20404	717	19776	0.99	0.98	665.21	646.05	177	5041	153	4420	0.92	0.92	628.02	645.36						
28	1	1	1	1412	41828	834	23969	0.57	0.57	758.62	752.84	748	20404	717	19776	0.99	0.98	665.21	646.05	177	5041	153	4420	0.92	0.92	628.02	645.36						
29	1	1	1	1515	43343	762	24625	0.57	0.57	724.55	751.48	724	21876	719	21213	0.99	0.98	661.74	646.79	188	5412	169	4755	0.92	0.92	663.62	646.29						
30	1	1	1	1441	44784	756	25381	0.57	0.57	770.78	752.05	705	22681	682	21895	0.99	0.98	669.50	646.79	188	5610	172	4927	0.92	0.92	663.62	646.29						
31	1	1	1	1419	46203	840	26221	0.57	0.57	757.66	752.22	736	23317	713	22608	0.99	0.98	635.38	646.43	184	5784	168	5085	0.92	0.92	650.44	646.07						
32	1	1	1	1375	47578	819	27040	0.57	0.57	727.36	751.45	753	24070	733	23341	0.99	0.98	646.79	646.43	184	5968	148	5233	0.92	0.92	630.16	646.52						
33	1	1	1	1442	49020	749	27789	0.57	0.57	743.96	751.33	738	25646	711	24062	0.99	0.98	641.77	647.28	183	6161	165	5398	0.92	0.92	663.67	646.78						
34	1	1	1	1510	50530	830	28619	0.57	0.57	784.33	752.26	726	26270	698	25469	0.99	0.98	648.94	647.17	181	6341	164	5692	0.92	0.92	640.00	646.58						
35	1	1	1	1379	51909	830	29449	0.57	0.57	744.03	752.26	768	27038	746	26215	0.99	0.98	630.52	646.69	189	6522	162	5724	0.92	0.92	629.87	646.11						
36	1	1	1	1483	53392	847	30296	0.57	0.57	736.80	751.66	739	27777	715	26920	0.99	0.98	607.52	646.65	148	6600	132	5673	0.92	0.92	632.31	646.76						
37	1	1	1	1636	54928	945	31241	0.57	0.57	754.14	751.72	739	28516	723	27653	0.99	0.98	643.36	646.59	189	6838	164	6169	0.92	0.92	624.38	646.29						
38	1	1	1	1439	56397	778	32164	0.57	0.57	780.42	752.33	893	29209	672	26325	0.99	0.98	678.89	646.38	204	7027	163	6362	0.92	0.92	663.33	646.28						
39	1	1	1	1410	57796	737	32764	0.57	0.57	760.42	752.33	724	30633	702	29027	0.99	0.98	639.84	646.98	174	7405	162	6594	0.92	0.92	626.97	646.36						
40	1	1	1	1600	60706	891	34389	0.57	0.57	730.62	751.75	724	30857	711	29738	0.99	0.98	633.01	646.67	170	7676	148	6962	0.92	0.92	649.41	646.45						
41	1	1	1	1388	62104	797	35285	0.57	0.57	760.09	752.39	746	31405	728	30497	0.98	0.88	642.74	646.60	162	7767	162	6914	0.92	0.92	663.44	646.54						
42	1	1	1	1483	63697	877	36172	0.57	0.57	738.60	752.08	722	32127	704	31171	0.99	0.98	631.43	646.28	172	7929	162	6960	0.92	0.92	627.77	646.25						
43	1	1	1	1463	65040	936	37108	0.57	0.57	756.42	752.24	695	32822	673	31844	0.99	0.98	627.30	644.90	166	8095	146	7112	0.92	0.92	611.64	644.56						
44	1	1	1	1385	66435	749	37657	0.57	0.57	764.26	752.28	747	33669	718	32662	0.99	0.98	661.23	646.26	189	8284	172	7284	0.92	0.92	681.40	646.43						
45	1	1	1	162221	678696	759	38616	0.57	0.57	746.21	752.20	769	34338	740	33302	0.99	0.98	660.99	646.38	187	8471	162	7446	0.92	0.92	662.32	646.58						
46	1	1	1	8100255	693711	874	39460	0.57	0.57	744.52	752.03	699	35037	690	33992	0.99	0.98	669.47	646.26	174	8646	164	7600	0.92	0.92	636.22	646.37						
47	1	1	1	1362	70763	885	40375	0.57	0.57	756.59	752.13	674	36711	636	34628	0.99	0.98	633.83	646.05	174	8819	163	7753	0.92	0.92	667.67	646.81						

Replication Number	Independent Variables			Product A												Product B												Product C											
	M&S	MOR	MFX	Period 1				Period 2				Period 3				Period 4				Period 5				Period 6				Period 7				Period 8							
				Total Demand	Period Shipped	Total Shipped	Period Cycle Time	Total Cycle Time	Period DFR	Total DFR	Period Demand	Period Shipped	Total Shipped	Period Cycle Time	Total Cycle Time	Period DFR	Total DFR	Period Demand	Period Shipped	Total Shipped	Period Cycle Time	Total Cycle Time	Period DFR	Total DFR	Period Demand	Period Shipped	Total Shipped	Period Cycle Time	Total Cycle Time	Period DFR	Total DFR	Period Demand	Period Shipped	Total Shipped	Period Cycle Time	Total Cycle Time	Period DFR	Total DFR	
1	1	1	3	1423	1423	36	0.04	0.04	292.06	292.06	1424	1424	848	848	0.81	0.81	588.00	588.00	350	350	274	274	0.77	0.77	730.25	730.25													
2	1	1	3	1474	2897	49	85	0.02	0.03	613.79	477.63	1466	2890	927	1775	0.63	0.62	586.35	587.14	363	363	268	268	0.79	0.78	701.00	701.00												
3	1	1	3	1391	4288	38	123	0.03	0.03	467.92	474.86	1390	4270	830	2605	0.82	0.82	596.63	590.26	379	1082	823	823	0.81	0.79	708.41	703.63												
4	1	1	3	1439	5727	41	169	0.03	0.03	515.20	484.72	1383	5653	868	347	0.62	0.62	610.27	595.26	347	1419	1107	1107	0.79	0.79	747.15	714.72												
5	1	1	3	1446	7173	45	204	0.03	0.03	498.51	467.76	1434	7073	881	4334	0.82	0.82	596.37	595.48	373	1812	272	272	0.79	0.79	691.75	710.19												
6	1	1	3	1468	8641	34	243	0.03	0.03	488.19	467.76	1434	8507	839	6173	0.82	0.82	591.42	593.20	368	2180	1668	1668	0.79	0.79	718.27	711.95												
7	1	1	3	1497	10098	30	273	0.03	0.03	479.76	466.89	1421	9928	849	6022	0.82	0.82	597.55	593.81	375	2455	279	279	0.79	0.79	700.05	710												
8	1	1	3	1467	11654	30	303	0.03	0.03	479.02	466.02	1421	11348	849	6871	0.82	0.82	600.00	595.04	375	3304	279	279	0.79	0.79	696.74	708												
9	1	1	3	1467	14467	30	332	0.03	0.03	474.53	464.27	1421	14189	849	6568	0.82	0.82	601.22	595.65	375	3879	279	279	0.79	0.79	680.12	704.93												
10	1	1	3	1422	15889	41	403	0.03	0.03	546.49	480.00	1432	15621	781	6348	0.82	0.82	597.70	592.46	401	4680	296	296	0.79	0.79	724.42	708.81												
11	1	1	3	1489	17388	31	434	0.03	0.03	541.28	484.22	1473	17094	886	10235	0.82	0.82	587.63	592.06	384	4444	281	281	0.79	0.79	665.86	705.90												
12	1	1	3	1468	18856	55	489	0.03	0.03	445.12	488.36	1408	18502	823	11118	0.82	0.82	596.34	592.40	375	4819	273	273	0.79	0.79	686.75	702.95												
13	1	1	3	1454	20310	58	547	0.03	0.03	491.00	488.64	1555	20057	924	12042	0.82	0.82	594.08	598.46	340	5159	273	273	0.79	0.79	708.62	703.34												
14	1	1	3	1418	21728	33	560	0.03	0.03	428.89	485.07	1399	21466	891	12933	0.82	0.82	611.52	590.98	381	5520	271	271	0.79	0.79	708.57	703.68												
15	1	1	3	1484	23222	42	622	0.03	0.03	597.71	480.65	1460	22906	784	13717	0.82	0.82	595.64	588.96	382	5902	283	283	0.79	0.79	743.78	706.23												
16	1	1	3	1468	24960	86	708	0.03	0.03	519.92	483.72	1384	24280	879	14596	0.82	0.82	596.43	589.47	346	8247	255	255	0.79	0.79	689.07	705.30												
17	1	1	3	1410	26100	64	772	0.03	0.03	522.67	486.12	1427	25717	838	15434	0.82	0.82	593.15	589.01	347	8604	282	282	0.79	0.79	738.02	707.15												
18	1	1	3	1397	27497	53	825	0.03	0.03	475.26	484.78	1467	27174	901	16395	0.82	0.82	597.77	589.68	332	8936	262	262	0.79	0.79	675.50	705.57												
19	1	1	3	1441	28938	51	876	0.03	0.03	509.04	485.61	1442	28616	890	17295	0.82	0.82	591.52	589.68	348	9284	265	265	0.79	0.79	675.50	705.05												
20	1	1	3	1406	30404	73	949	0.03	0.03	513.38	486.90	1404	30020	888	18273	0.82	0.82	617.79	591.20	331	7615	246	246	0.79	0.79	642.76	700.42												
21	1	1	3	1459	31863	69	1018	0.03	0.03	505.31	487.47	1481	31601	883	19256	0.82	0.82	592.57	591.27	329	7944	246	246	0.79	0.79	640.18	700.22												
22	1	1	3	1447	33310	48	1066	0.03	0.03	490.08	483.31	1460	32961	838	20094	0.82	0.82	597.05	590.26	383	8307	284	284	0.79	0.79	722.36	701.22												
23	1	1	3	1446	34756	48	1116	0.03	0.03	489.68	483.69	1386	34347	832	20976	0.82	0.82	600.72	590.91	366	8693	272	272	0.79	0.79	714.26	701.76												
24	1	1	3	1414	36170	43	1158	0.03	0.03	509.75	484.19	1386	35715	820	21796	0.82	0.82	602.34	591.34	373	8936	266	266	0.79	0.79	712.89	702.20												
25	1	1	3	1449	37619	20	1178	0.03	0.03	471.61	483.51	1379	37094	813	22809	0.82	0.82	610.25	592.02	373	9409	263	263	0.79	0.79	732.66	702.46												
26	1	1	3	1367	38956	91	1269	0.03	0.03	595.89	501.13	1412	38906	658	23547	0.82	0.82	597.64	592.24	330	9736	248	248	0.79	0.79	670.05	702.96												
27	1	1	3	1467	40463	61	1330	0.03	0.03	464.06	489.43	1460	39896	914	24481	0.82	0.82	608.33	592.41	347	10086	281	281	0.79	0.79	670.05	701.47												
28	1	1	3	1441	41894	68	1388	0.03	0.03	491.62	489.43	1441	41407	957	25418	0.82	0.82	600.33	592.41	343	10429	248	248	0.79	0.79	688.56	699.48												
29	1	1	3	1466	43350	31	1429	0.03	0.03	412.39	487.17	1448	42855	903	26321	0.82	0.82	594.16	592.47	370	10799	271	271	0.79	0.79	671.75	696.56												
30	1	1	3	1468	44808	43	1472	0.03	0.03	557.08	488.62	1461	44306	868	27189	0.82	0.82	596.21	592.27	355	11154	277	277	0.79	0.79	731.49	699.64												
31	1	1	3	1464	46262	84	1566	0.03	0.03	517.63	489.63	1463	45759	930	28119	0.82	0.82	585.32	592.04	339	11483	283	283	0.79	0.79	677.78	696.98												
32	1	1	3	1485	47747	45	1601	0.03	0.03	433.40	488.06	1404	47163	940	29059	0.82	0.82	611.82	592.68	332	11830	259	259	0.79	0.79	687.25	696.93												
33	1	1	3	1447	49194	51	1652	0.03	0.03	470.20	487.20	1486	48649	937	29966	0.82	0.82	587.88	592.53	337	12162	261	261	0.79	0.79	695.04	696.82												
34	1	1	3	1460	50644	35	1667	0.03	0.03	558.41	488.47	1524	50173	858	30854	0.82	0.82	590.46	591.36	388	12560	274	274	0.79	0.79	675.24	696.14												
35	1	1	3	1477	52121	49	1736	0.03	0.03	480.76	487.97	1493	51666	920	31774	0.82	0.82	590.85	591.05	334	12884	265	265	0.79	0.79	713.82	696.56												
36	1	1	3	1349	53470	69	1805	0.03	0.03	530.93	489.23	1429	53095	913	32697	0.82	0.82	599.88	591.02	351	13235	261	261	0.79	0.79	685.50	696.22												
37	1	1	3	1413	54883	54	1859	0.03	0.03	536.75	500.32	1418	54613	877	33564	0.82	0.82	596.76	591.17	368	13603	271	271	0.79	0.79	712.28	696.59												
38	1	1	3	1400	56283	40	1899	0.03	0.03	497.27	488.36	1479	55992	904	34468	0.82	0.82	579.35	590.86	396	13969	275	275	0.79	0.79	650.91	697.35												
39	1	1	3	1408	57891	35	1934	0.03	0.03	498.29	486.73	1425	57417	901	35369	0.82	0.82	608.13	591.30	383	14332	285	285	0.79	0.79	689.82	697.21												
40	1	1	3	1400	59091	34	1968	0.03	0.03	639.28	489.02	1435	58652	864	36323	0.82	0.82	590.88	591.30	373	14705	275	275	0.79	0.79	689.82	697.03												
41	1	1	3	1347	60438	61	2029	0.03	0.03	577.52	501.38	1480	60332	930	37163	0.82	0.82	575.51	590.89	357	15062	266	266	0.79	0.79	648.50	696.90												
42	1	1	3	1460	61888	37	2066	0.03	0.03	596.06	489.63	1496	61766	829	37962	0.82	0.82	574.85	590.54	381	15443	280	280	0.79	0.79	704.23	696.10												
43	1	1	3	1460	63246	74	2140	0.03	0.03	597.37	501.53	1437	63205	952	38944	0.82	0.82	599.64	590.76	333	15746	266	266	0.79	0.79	662.81	696.17												
44	1	1	3	1462	64608	20	2196	0.03	0.03	139.14	487.18	1489	64703	829	39073	0.82	0.82	598.14	599.08	399	16175	285	285	0.79	0.79	699.89	696.21												
45	1	1	3	1404	66212	85	2251	0.03	0.03	652.10	503.03	1441	66144	896	40776	0.82	0.82	578.81	599.63	349	16521	280	280	0.79	0.79	695.21	696.21												
46	1	1	3	1403	67615	68	2319	0.03	0.03	482.46	502.72	1419	67663	890	41009	0.82	0.82	596.63	599.23	354	16875	279	279	0.79	0.79	733.12	696.04												
47	1	1	3	1408	69021	28	2347	0.03	0.03	339.75	500.74	1481	69044	830	42339	0.82																							

Replication Number	Independent Variables			All Products										LOW			MED			HIGH		
	MAS	MDH	MX	Period Demand	Total Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle	Period Time	Period NDI	Cumulative NDI	Period NDI	Cumulative NDI	Period NDI	Cumulative NDI	Period NDI	Cumulative NDI			
1	3	1	1	1867	1867	1863	1853	0.998	0.998	270.37	270.37	85.46441	85.46441	77.33349	77.33349	53.11183	53.11183	50.28044	50.28044			
2	3	1	1	1868	3725	1817	3670	0.999	0.999	291.38	291.38	82.86222	168.36222	81.86628	161.81677	79.64343	103.37207	76.94343	103.37207			
3	3	1	1	1869	5592	1862	5622	0.999	0.999	296.80	296.80	85.46121	288.86121	81.86628	288.86121	79.64343	288.86121	76.94343	288.86121			
4	3	1	1	1870	7563	1862	7593	0.998	0.998	296.58	296.58	85.46121	342.91105	80.63629	310.61734	76.94343	310.61734	74.14343	310.61734			
5	3	1	1	1871	9524	1874	9348	0.998	0.998	305.44	305.44	86.07451	430.98556	80.05658	390.07602	76.94343	390.07602	74.14343	390.07602			
6	3	1	1	1872	11460	1916	11264	0.996	0.996	327.90	327.90	86.52011	517.51487	78.49608	469.17411	74.14343	469.17411	71.94628	469.17411			
7	3	1	1	1873	13350	1868	13132	0.996	0.996	339.52	339.52	86.39706	603.90173	78.32738	547.50149	74.14343	547.50149	71.94628	547.50149			
8	3	1	1	1874	15244	1868	14999	0.996	0.996	304.70	304.70	86.39706	690.29879	78.32738	626.62888	74.14343	626.62888	71.94628	626.62888			
9	3	1	1	1875	17138	1868	16887	0.996	0.996	303.32	303.32	86.39706	776.67585	78.32738	704.16626	74.14343	704.16626	71.94628	704.16626			
10	3	1	1	1876	19032	1868	18724	0.995	0.995	315.14	315.14	86.43741	862.50032	81.65646	864.08013	77.33349	864.08013	74.14343	864.08013			
11	3	1	1	1877	20926	1868	20642	0.995	0.995	315.14	315.14	82.01622	1,035.11654	74.17929	1,010.15110	71.94628	1,010.15110	69.73349	1,010.15110			
12	3	1	1	1878	22820	1817	22469	0.995	0.995	324.54	324.54	85.98861	1,120.71415	77.46768	1,088.22294	74.17929	1,088.22294	71.94628	1,088.22294			
13	3	1	1	1879	24715	1869	24528	0.996	0.996	308.77	308.77	85.98861	1,207.69036	72.50589	1,167.53167	71.94628	1,167.53167	69.73349	1,167.53167			
14	3	1	1	1880	26597	1850	26178	0.996	0.996	312.37	312.37	85.98861	1,299.03058	73.30689	1,244.99356	71.94628	1,244.99356	69.73349	1,244.99356			
15	3	1	1	1881	28460	1834	28012	0.995	0.995	280.42	280.42	85.47481	1,374.57839	75.63189	1,320.62546	71.94628	1,320.62546	69.73349	1,320.62546			
16	3	1	1	1882	30349	1880	29892	0.996	0.996	299.14	299.14	85.47481	1,468.49120	75.63189	1,390.05753	71.94628	1,390.05753	69.73349	1,390.05753			
17	3	1	1	1883	32217	1825	31717	0.996	0.996	279.84	279.84	85.47481	1,546.03922	78.48209	1,478.07762	71.94628	1,478.07762	69.73349	1,478.07762			
18	3	1	1	1884	34046	1826	33543	0.996	0.996	334.50	334.50	87.09601	1,632.13624	79.02009	1,549.72311	74.43384	1,549.72311	71.94628	1,549.72311			
19	3	1	1	1885	35940	1861	35404	0.996	0.996	302.54	302.54	87.09601	1,709.57646	79.17789	1,628.90100	74.43384	1,628.90100	71.94628	1,628.90100			
20	3	1	1	1886	37767	1803	37207	0.995	0.995	253.25	253.25	87.09601	1,789.74387	82.44248	1,711.34348	76.22083	1,711.34348	71.94628	1,711.34348			
21	3	1	1	1887	39524	1845	39025	0.995	0.995	209.92	209.92	81.60801	1,971.35188	73.34709	1,794.69057	76.22083	1,794.69057	71.94628	1,794.69057			
22	3	1	1	1888	41325	1873	40825	0.996	0.996	298.63	298.63	82.40782	2,063.84970	74.02689	1,868.11746	76.22083	1,868.11746	71.94628	1,868.11746			
23	3	1	1	1889	43142	1866	42791	0.996	0.996	298.63	298.63	82.40782	2,156.43011	74.26149	1,932.97894	76.22083	1,932.97894	71.94628	1,932.97894			
24	3	1	1	1890	44971	1823	44514	0.996	0.996	294.35	294.35	82.86241	2,248.63013	73.68208	2,006.46103	76.22083	2,006.46103	71.94628	2,006.46103			
25	3	1	1	1891	46825	1841	46255	0.996	0.996	298.40	298.40	81.65646	2,340.81654	73.68208	2,089.06552	76.22083	2,089.06552	71.94628	2,089.06552			
26	3	1	1	1892	48694	1878	48133	0.995	0.995	298.56	298.56	81.65646	2,432.95395	73.68208	2,162.06440	76.22083	2,162.06440	71.94628	2,162.06440			
27	3	1	1	1893	50574	1825	50133	0.995	0.995	252.07	252.07	81.65646	2,525.09348	73.68208	2,244.26440	76.22083	2,244.26440	71.94628	2,244.26440			
28	3	1	1	1894	52476	1825	51995	0.995	0.995	297.82	297.82	82.47826	2,617.28399	80.10728	2,326.38177	76.22083	2,326.38177	71.94628	2,326.38177			
29	3	1	1	1895	54392	1877	53835	0.995	0.995	297.82	297.82	82.47826	2,709.46836	80.10728	2,408.50666	76.22083	2,408.50666	71.94628	2,408.50666			
30	3	1	1	1896	56314	1848	55881	0.995	0.995	300.33	300.33	82.86241	2,801.65283	81.38288	2,490.89103	76.22083	2,490.89103	71.94628	2,490.89103			
31	3	1	1	1897	58244	1885	57696	0.995	0.995	304.94	304.94	82.86241	2,893.83630	81.38288	2,573.27541	76.22083	2,573.27541	71.94628	2,573.27541			
32	3	1	1	1898	60174	1868	59424	0.996	0.996	303.92	303.92	82.86241	2,986.01981	81.38288	2,655.65998	76.22083	2,655.65998	71.94628	2,655.65998			
33	3	1	1	1899	62104	1911	61335	0.996	0.996	303.92	303.92	82.86241	3,078.20332	81.38288	2,738.04357	76.22083	2,738.04357	71.94628	2,738.04357			
34	3	1	1	1900	64024	1934	63269	0.996	0.996	306.32	306.32	84.36681	3,170.38713	83.18168	2,820.22815	76.22083	2,820.22815	71.94628	2,820.22815			
35	3	1	1	1901	65944	1860	65120	0.996	0.996	304.20	304.20	84.36681	3,262.57094	83.18168	2,902.41174	76.22083	2,902.41174	71.94628	2,902.41174			
36	3	1	1	1902	67864	1807	66936	0.996	0.996	303.77	303.77	80.40422	3,354.75475	81.87929	2,984.59544	76.22083	2,984.59544	71.94628	2,984.59544			
37	3	1	1	1903	69784	1869	68934	0.996	0.996	300.45	300.45	86.62141	3,446.93856	78.60648	3,066.77915	76.22083	3,066.77915	71.94628	3,066.77915			
38	3	1	1	1904	71704	1832	70966	0.996	0.996	346.52	346.52	82.56921	3,539.12237	77.12629	3,148.96286	76.22083	3,148.96286	71.94628	3,148.96286			
39	3	1	1	1905	73624	1846	72511	0.996	0.996	303.15	303.15	81.03781	3,631.30618	72.63189	3,231.14667	76.22083	3,231.14667	71.94628	3,231.14667			
40	3	1	1	1906	75544	1849	74560	0.995	0.995	307.58	307.58	82.05461	3,723.48999	78.95389	3,313.33148	76.22083	3,313.33148	71.94628	3,313.33148			
41	3	1	1	1907	77464	1866	76528	0.995	0.995	305.32	305.32	80.49801	3,815.67380	81.67508	3,395.51629	76.22083	3,395.51629	71.94628	3,395.51629			
42	3	1	1	1908	79384	1769	78915	0.996	0.996	299.67	299.67	82.05461	3,907.85761	74.45489	3,477.70109	76.22083	3,477.70109	71.94628	3,477.70109			
43	3	1	1	1909	81304	1835	80650	0.996	0.996	281.30	281.30	83.27381	3,999.04142	75.00789	3,559.88590	76.22083	3,559.88590	71.94628	3,559.88590			
44	3	1	1	1910	83224	1825	81675	0.996	0.996	244.96	244.96	81.04781	4,091.22523	75.76689	3,642.07019	76.22083	3,642.07019	71.94628	3,642.07019			
45	3	1	1	1911	85144	1908	83583	0.995	0.995	306.14	306.14	83.99681	4,183.40904	83.99681	3,724.25400	76.22083	3,724.25400	71.94628	3,724.25400			
46	3	1	1	1912	87064	1870	85493	0.995	0.995	300.09	300.09	84.10081	4,275.59285	83.99681	3,806.43877	76.22083	3,806.43877	71.94628	3,806.43877			
47	3	1	1	1913	88984	1860	87343	0.995	0.995	318.81	318.81	85.22381	4,367.77666	80.30288	3,888.62354	76.22083	3,888.62354	71.94628	3,888.62354			
48	3	1	1	1914	90904	1763	89136	0.996	0.996	264.94	264.94	87.05641	4,459.96047	80.30288	3,970.80833	76.22083	3,970.80833	71.94628	3,970.80833			
49	3	1	1	1915	92824	1884	91020	0.996	0.996	307.34	307.34	84.10081	4,552.14428	80.30288	4,053.00210	76.22083	4,053.00210	71.94628	4,053.00210			
50	3	1	1	1916	94744	1835	92855	0.996	0.996	291.78	291.78	84.10081	4,644.32809	80.30288	4,135.18590	76.22083	4,135.18590	71.94628	4,135.18590			
51	3	1	1	1917	96664	1867	94712	0.996	0.996	306.14	306.14	84.10081	4,736.51190	80.30288	4,217.37071	76.22083	4,217.37071	71.94628	4,217.37071			
52	3	1	1	1918	98584	1815	96432	0.995	0.995	309.05	309.05	87.05641	4,828.69571	80.30288	4,299.55452	76.22083	4,299.55452	71.94628	4,299.55452			
53	3	1	1	1919	100504	1880	100312	0.996	0.996	307.04	307.04	85.21881	4,920.87952	80.30288	4,381.73833	76.22083	4,381.73833	71.94628	4,381.73833			
54	3	1	1	1920	102424	1816	102128	0.996	0.996	295.12	295.12	85.21881	5,013.06333	77.15109	4,463.92214	76.22083	4,463.92214	71.94628	4,463.9			

Replication Number	Independent Variables			All Products										LOW			MED			HIGH		
	MAS	MDH	MX	Period Demand	Total Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle Time	Cycle Time	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	
1	3	1	2	2323	2323	2033	2033	0.890	0.890	644.50	644.50	99,741.46	99,741.46	93,407.93	93,407.93	66,440.80	66,440.80	66,440.80	66,440.80	66,440.80	66,440.80	
2	3	1	2	2286	4619	2065	4086	0.893	0.893	556.84	556.84	102,344.91	102,344.91	96,814.91	96,814.91	72,647.81	72,647.81	72,647.81	72,647.81	72,647.81	72,647.81	
3	3	1	2	2316	6635	2020	6124	0.900	0.900	542.81	542.81	302,758.19	302,758.19	284,082.65	284,082.65	212,181.56	212,181.56	212,181.56	212,181.56	212,181.56	212,181.56	
4	3	1	2	2388	8323	2101	8225	0.893	0.893	548.69	548.69	405,558.04	405,558.04	380,973.98	380,973.98	285,105.86	285,105.86	285,105.86	285,105.86	285,105.86	285,105.86	
5	3	1	2	2323	11646	2067	10392	0.900	0.900	546.71	546.71	101,465.95	101,465.95	97,412.13	97,412.13	71,446.10	71,446.10	71,446.10	71,446.10	71,446.10	71,446.10	
6	3	1	2	2424	14070	2092	12384	0.895	0.894	559.36	559.36	600,624.19	600,624.19	569,828.10	569,828.10	420,025.91	420,025.91	420,025.91	420,025.91	420,025.91	420,025.91	
7	3	1	2	2386	18466	2082	16446	0.894	0.894	559.67	559.67	100,954.26	100,954.26	94,883.25	94,883.25	69,416.22	69,416.22	69,416.22	69,416.22	69,416.22	69,416.22	
8	3	1	2	2386	18843	2082	16547	0.894	0.894	559.67	559.67	100,954.26	100,954.26	94,883.25	94,883.25	69,416.22	69,416.22	69,416.22	69,416.22	69,416.22	69,416.22	
9	3	1	2	2386	21229	2082	18629	0.897	0.891	561.45	561.45	100,954.26	100,954.26	94,883.25	94,883.25	69,416.22	69,416.22	69,416.22	69,416.22	69,416.22	69,416.22	
10	3	1	2	2386	23816	2082	20710	0.891	0.891	564.87	564.87	100,954.26	100,954.26	94,883.25	94,883.25	69,416.22	69,416.22	69,416.22	69,416.22	69,416.22	69,416.22	
11	3	1	2	2263	26878	2019	22729	0.889	0.889	553.83	553.83	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
12	3	1	2	2316	28184	2074	24803	0.891	0.891	561.63	561.63	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
13	3	1	2	2388	30562	2076	26879	0.888	0.891	567.14	567.14	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
14	3	1	2	2325	32887	2037	28916	0.887	0.891	545.07	545.07	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
15	3	1	2	2390	35186	2024	30940	0.887	0.890	534.44	534.44	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
16	3	1	2	2329	37515	2066	33006	0.895	0.891	538.42	538.42	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
17	3	1	2	2285	39810	2014	36024	0.886	0.890	536.60	536.60	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
18	3	1	2	2333	42143	2064	37084	0.888	0.890	546.57	546.57	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
19	3	1	2	2334	44477	2048	39132	0.888	0.890	534.24	534.24	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
20	3	1	2	2388	46746	1968	41130	0.896	0.890	533.68	533.68	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
21	3	1	2	2309	49054	2063	43103	0.892	0.890	546.30	546.30	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
22	3	1	2	2265	51309	2029	46222	0.886	0.890	546.88	546.88	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
23	3	1	2	2374	53663	2024	49246	0.889	0.890	520.28	520.28	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
24	3	1	2	2338	56021	2049	49295	0.888	0.890	533.46	533.46	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
25	3	1	2	2365	58376	2064	51569	0.893	0.890	533.68	533.68	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
26	3	1	2	2427	60863	2082	53961	0.887	0.890	535.74	535.74	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
27	3	1	2	2241	63044	2049	56600	0.887	0.890	534.04	534.04	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
28	3	1	2	2468	65902	2077	57377	0.896	0.890	554.71	554.71	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
29	3	1	2	2424	67926	2105	59882	0.891	0.890	547.03	547.03	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
30	3	1	2	2413	70339	2097	61779	0.891	0.890	547.03	547.03	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
31	3	1	2	2223	72562	2016	63795	0.889	0.890	564.62	564.62	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
32	3	1	2	2325	74887	2073	67896	0.892	0.890	574.85	574.85	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
33	3	1	2	2270	77157	2028	71657	0.890	0.890	536.32	536.32	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
34	3	1	2	2327	79484	2078	74984	0.894	0.890	541.39	541.39	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
35	3	1	2	2379	81863	2043	72017	0.889	0.890	526.00	526.00	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
36	3	1	2	2283	84146	2027	74944	0.891	0.890	546.59	546.59	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
37	3	1	2	2361	86607	2087	77404	0.888	0.890	546.59	546.59	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
38	3	1	2	2376	88883	2075	78206	0.891	0.890	566.85	566.85	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
39	3	1	2	2368	91241	2083	80298	0.890	0.890	561.84	561.84	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
40	3	1	2	2348	93589	2082	82361	0.917	0.891	569.84	569.84	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
41	3	1	2	2414	96063	2081	84462	0.888	0.891	561.73	561.73	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
42	3	1	2	2286	98269	2013	86462	0.887	0.891	546.93	546.93	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
43	3	1	2	2233	100502	2032	89497	0.895	0.891	543.12	543.12	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
44	3	1	2	2286	102789	2019	90516	0.890	0.891	525.77	525.77	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
45	3	1	2	2278	105036	2010	92526	0.889	0.891	529.02	529.02	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
46	3	1	2	2342	107378	2082	94808	0.894	0.891	559.58	559.58	1,110,988.37	1,110,988.37	943,035.61	943,035.61	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	780,027.36	
47	3	1	2	2274	109762	210																

Replication Number	Independent Variables			All Products										LOW			MED			HIGH		
	MAS	MOH	MIX	Period Demand	Total Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle Time	Cycle Time	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	
1	3	1	3	3215	3215	2191	2191	0.668	0.668	0.671	0.671	102,781.92	102,781.92	97,559.64	97,559.64	102,781.92	102,781.92	97,559.64	97,559.64	102,781.92	102,781.92	
2	3	1	3	3296	6511	2164	4955	0.673	0.673	0.674	0.674	104,217.67	206,994.40	99,020.30	196,070.94	104,217.67	308,142.84	99,020.30	196,070.94	104,217.67	412,312.12	
3	3	1	3	3241	9752	2172	6627	0.679	0.679	0.680	0.680	101,143.46	308,142.84	95,651.18	202,312.12	101,143.46	412,312.12	95,651.18	202,312.12	101,143.46	513,455.58	
4	3	1	3	3215	12967	2169	8996	0.670	0.670	0.671	0.671	104,441.75	412,312.12	99,364.47	301,595.59	104,441.75	513,455.58	99,364.47	301,595.59	104,441.75	612,771.14	
5	3	1	3	3283	16250	2172	10868	0.691	0.676	0.682	0.676	104,686.46	513,455.58	99,509.18	401,104.77	104,686.46	612,771.14	99,509.18	401,104.77	104,686.46	711,957.35	
6	3	1	3	3286	19510	2195	13063	0.679	0.677	0.681	0.677	103,715.86	612,771.14	98,603.68	500,708.36	103,715.86	711,957.35	98,603.68	500,708.36	103,715.86	811,133.33	
7	3	1	3	3266	22776	2188	15251	0.678	0.677	0.681	0.678	103,342.09	711,957.35	98,184.82	600,893.17	103,342.09	811,133.33	98,184.82	600,893.17	103,342.09	910,470.27	
8	3	1	3	3266	26042	2188	17438	0.678	0.677	0.681	0.678	103,342.09	811,133.33	98,184.82	700,077.99	103,342.09	910,470.27	98,184.82	700,077.99	103,342.09	1,010,811.33	
9	3	1	3	3266	29307	2188	19626	0.678	0.677	0.681	0.678	103,342.09	910,470.27	98,184.82	800,263.81	103,342.09	1,010,811.33	98,184.82	800,263.81	103,342.09	1,111,153.33	
10	3	1	3	3266	32573	2188	21813	0.678	0.677	0.681	0.678	103,342.09	1,010,811.33	98,184.82	900,447.63	103,342.09	1,111,153.33	98,184.82	900,447.63	103,342.09	1,211,306.67	
11	3	1	3	3286	35869	2208	24021	0.690	0.678	0.682	0.678	101,176.91	1,111,153.33	95,734.64	1,078,182.27	101,176.91	1,211,306.67	95,734.64	1,078,182.27	101,176.91	1,311,522.23	
12	3	1	3	3347	39216	2210	26231	0.677	0.678	0.682	0.677	102,895.38	1,211,306.67	97,653.11	1,175,835.38	102,895.38	1,311,522.23	97,653.11	1,175,835.38	102,895.38	1,411,650.78	
13	3	1	3	4061	43277	2168	28398	0.733	0.683	0.683	0.683	103,685.51	1,311,522.23	98,616.24	1,274,453.61	103,685.51	1,411,650.78	98,616.24	1,274,453.61	103,685.51	1,511,779.23	
14	3	1	3	2462	46759	2198	30598	0.673	0.677	0.681	0.673	104,056.80	1,411,650.78	98,684.52	1,373,438.14	104,056.80	1,511,779.23	98,684.52	1,373,438.14	104,056.80	1,611,903.67	
15	3	1	3	3255	49014	2195	32739	0.665	0.678	0.683	0.665	103,254.86	1,511,779.23	98,042.58	1,471,480.72	103,254.86	1,611,903.67	98,042.58	1,471,480.72	103,254.86	1,711,615.15	
16	3	1	3	3307	52321	2162	34855	0.665	0.677	0.681	0.665	105,002.51	1,611,903.67	98,536.23	1,571,504.55	105,002.51	1,711,615.15	98,536.23	1,571,504.55	105,002.51	1,811,730.10	
17	3	1	3	3309	55630	2185	37140	0.662	0.678	0.681	0.662	103,781.92	1,711,615.15	98,634.23	1,668,039.78	103,781.92	1,811,730.10	98,634.23	1,668,039.78	103,781.92	1,911,855.05	
18	3	1	3	3174	58804	2187	39327	0.680	0.678	0.682	0.678	105,750.47	1,811,730.10	99,230.46	1,766,274.48	105,750.47	1,911,855.05	99,230.46	1,766,274.48	105,750.47	2,011,980.00	
19	3	1	3	3381	62195	2203	41530	0.682	0.678	0.681	0.682	102,254.46	1,911,855.05	97,057.17	1,863,321.27	102,254.46	1,963,452.72	97,057.17	1,863,321.27	102,254.46	2,063,500.00	
20	3	1	3	3281	65476	2189	43719	0.679	0.678	0.682	0.679	103,752.80	2,011,980.00	98,665.59	1,961,016.12	103,752.80	2,112,132.88	98,665.59	1,961,016.12	103,752.80	2,212,258.83	
21	3	1	3	3295	68771	2178	46097	0.699	0.677	0.682	0.699	103,752.80	2,112,132.88	98,665.59	2,059,681.77	103,752.80	2,160,834.68	98,665.59	2,059,681.77	103,752.80	2,260,016.57	
22	3	1	3	3273	72044	2182	48079	0.699	0.677	0.682	0.699	103,752.80	2,212,258.83	98,665.59	2,158,347.27	103,752.80	2,257,104.72	98,665.59	2,158,347.27	103,752.80	2,356,351.50	
23	3	1	3	3302	75346	2170	50049	0.679	0.677	0.681	0.679	103,119.88	2,311,480.72	97,902.71	2,255,255.94	103,119.88	2,411,650.78	97,902.71	2,255,255.94	103,119.88	2,511,820.73	
24	3	1	3	3300	78626	2216	52465	0.683	0.677	0.681	0.683	103,323.70	2,411,650.78	98,146.52	2,353,402.46	103,323.70	2,511,820.73	98,146.52	2,353,402.46	103,323.70	2,611,995.19	
25	3	1	3	3213	81839	2168	54833	0.679	0.676	0.682	0.679	101,511.51	2,511,820.73	96,119.24	2,449,521.68	101,511.51	2,611,995.19	96,119.24	2,449,521.68	101,511.51	2,712,166.63	
26	3	1	3	3200	85046	2163	56966	0.676	0.677	0.681	0.676	105,132.34	2,611,995.19	100,030.00	2,549,551.68	105,132.34	2,712,166.63	100,030.00	2,549,551.68	105,132.34	2,812,338.18	
27	3	1	3	3297	88342	2162	59098	0.672	0.677	0.682	0.672	105,233.10	2,712,166.63	99,990.30	2,649,546.98	105,233.10	2,812,338.18	99,990.30	2,649,546.98	105,233.10	2,912,510.68	
28	3	1	3	3231	91673	2183	61141	0.681	0.677	0.682	0.681	104,887.28	2,812,338.18	99,990.30	2,749,539.38	104,887.28	2,912,510.68	99,990.30	2,749,539.38	104,887.28	3,012,682.18	
29	3	1	3	3144	94717	2147	63268	0.688	0.677	0.683	0.688	103,598.04	2,912,510.68	99,990.30	2,849,532.68	103,598.04	3,012,682.18	99,990.30	2,849,532.68	103,598.04	3,112,854.68	
30	3	1	3	3262	97981	2175	65463	0.699	0.677	0.683	0.699	103,752.80	3,012,682.18	99,990.30	2,949,526.98	103,752.80	3,112,854.68	99,990.30	2,949,526.98	103,752.80	3,213,027.18	
31	3	1	3	3264	101197	2174	67637	0.674	0.677	0.682	0.674	103,752.80	3,112,854.68	99,990.30	3,049,521.28	103,752.80	3,213,027.18	99,990.30	3,049,521.28	103,752.80	3,313,199.68	
32	3	1	3	3347	104544	2186	69823	0.667	0.676	0.682	0.667	104,027.63	3,213,027.18	99,990.30	3,149,516.58	104,027.63	3,313,199.68	99,990.30	3,149,516.58	104,027.63	3,413,372.18	
33	3	1	3	3168	107712	2180	71993	0.681	0.677	0.682	0.681	104,027.63	3,313,199.68	99,990.30	3,249,511.88	104,027.63	3,413,372.18	99,990.30	3,249,511.88	104,027.63	3,513,544.68	
34	3	1	3	3376	111088	2188	74172	0.669	0.676	0.681	0.669	104,538.46	3,413,372.18	99,990.30	3,349,507.18	104,538.46	3,513,544.68	99,990.30	3,349,507.18	104,538.46	3,613,717.18	
35	3	1	3	3244	114332	2172	76344	0.665	0.676	0.681	0.665	104,538.46	3,513,544.68	99,990.30	3,449,502.48	104,538.46	3,613,717.18	99,990.30	3,449,502.48	104,538.46	3,713,889.68	
36	3	1	3	3244	117576	2168	78512	0.675	0.676	0.682	0.675	102,922.09	3,613,717.18	99,990.30	3,549,497.78	102,922.09	3,713,889.68	99,990.30	3,549,497.78	102,922.09	3,814,062.18	
37	3	1	3	3301	120877	2196	80708	0.684	0.676	0.682	0.684	104,383.41	3,713,889.68	99,990.30	3,649,493.08	104,383.41	3,814,062.18	99,990.30	3,649,493.08	104,383.41	3,914,234.68	
38	3	1	3	3121	123968	2125	82833	0.670	0.676	0.681	0.670	104,383.41	3,814,062.18	99,990.30	3,749,488.38	104,383.41	3,914,234.68	99,990.30	3,749,488.38	104,383.41	4,014,407.18	
39	3	1	3	3313	127311	2184	85017	0.677	0.676	0.681	0.677	103,720.27	3,914,234.68	99,990.30	3,849,483.68	103,720.27	4,014,407.18	99,990.30	3,849,483.68	103,720.27	4,114,579.68	
40	3	1	3	3182	130493	2173	87190	0.674	0.676	0.682	0.674	103,871.69	4,014,407.18	99,990.30	3,949,479.00	103,871.69	4,114,579.68	99,990.30	3,949,479.00	103,871.69	4,214,752.18	
41	3	1	3	3276	133769	2164	89344	0.668	0.676	0.682	0.668	102,632.22	4,114,579.68	99,990.30	4,049,474.30	102,632.22	4,214,752.18	99,990.30	4,049,474.30	102,632.22	4,314,926.68	
42	3	1	3	3165	136634	2162	91486	0.678	0.676	0.682	0.678	103,341.75	4,214,752.18	99,990.30	4,149,469.60	103,341.75	4,314,926.68	99,990.30	4,149,469.60	103,341.75	4,415,101.18	
43	3	1	3	3151	140066	2156	93652	0.680	0.676	0.682	0.680	103,341.75	4,314,926.68	99,990.30	4,249,464.90	103,341.75	4,415,101.18	99,990.30	4,249,464.90	103,341.75	4,515,275.68	
44	3	1	3	3201	143266	2174	95826	0.680	0.676	0.682	0.680	103,341.75	4,415,101.18	99,990.30	4,349,460.20	103,341.75	4,515,275.68	99,990.30	4,349,460.20	103,341.75	4,615,450.18	
45	3	1	3	3232	146469	2147	97993	0.673	0.676	0.682	0.673	103,341.75	4,515,275.68	99,990.30	4,449,455.50	103,341.75	4,615,450.18	99,990.30	4,449,455.50	103,341.75	4,715,624.68	
46	3	1	3	3201	149699	2117	100140	0.670	0.675	0.681	0.670	103,341.75	4,615,450.18	99,990.30	4,549,450.80	103,341.75	4,715,624.68					

Replication Number	Independent Variables			All Products										LOW			MED			HIGH		
	MAS	MDH	MIX	Period Demand	Period Shipped	Total Shipped	Total DFR	Period DFR	Total DFR	Period Cycle Time	Cycle Times	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	
1	2	1	1	1936	1936	1913	0.999	0.999	0.999	312.00	88,165.29	88,165.29	80,375.30	80,375.30	80,375.30	80,375.30	80,375.30	80,375.30	80,375.30	80,375.30	80,375.30	80,375.30
2	2	1	1	1884	3800	1863	0.998	0.998	0.998	308.95	84,243.27	172,408.57	75,065.29	156,333.59	156,333.59	156,333.59	156,333.59	156,333.59	156,333.59	156,333.59	156,333.59	156,333.59
3	2	1	1	1901	5701	1877	0.996	0.996	0.996	329.23	315.36	86,960.28	281,358.85	81,005.30	237,338.99	237,338.99	237,338.99	237,338.99	237,338.99	237,338.99	237,338.99	237,338.99
4	2	1	1	1832	7533	1784	0.997	0.997	0.997	300.94	95,036.77	346,396.62	76,741.79	314,080.68	314,080.68	314,080.68	314,080.68	314,080.68	314,080.68	314,080.68	314,080.68	314,080.68
5	2	1	1	1787	9320	1787	0.998	0.998	0.998	295.34	288.18	77,361.27	423,756.89	88,561.29	382,641.97	382,641.97	382,641.97	382,641.97	382,641.97	382,641.97	382,641.97	382,641.97
6	2	1	1	1887	11187	1822	0.997	0.997	0.997	325.17	294.29	84,107.11	507,864.00	75,757.12	468,399.08	468,399.08	468,399.08	468,399.08	468,399.08	468,399.08	468,399.08	468,399.08
7	2	1	1	1852	13039	1835	0.998	0.998	0.998	337.56	289.20	83,307.19	591,171.20	74,830.96	533,330.05	533,330.05	533,330.05	533,330.05	533,330.05	533,330.05	533,330.05	533,330.05
8	2	1	1	1852	14892	1834	0.998	0.998	0.998	357.22	289.42	83,223.03	767,659.34	74,883.88	608,213.93	608,213.93	608,213.93	608,213.93	608,213.93	608,213.93	608,213.93	608,213.93
9	2	1	1	1852	16596	1833	0.988	0.988	0.988	305.55	282.72	83,160.94	840,840.28	74,788.71	757,540.43	757,540.43	757,540.43	757,540.43	757,540.43	757,540.43	757,540.43	757,540.43
10	2	1	1	1872	20468	1846	0.988	0.988	0.988	299.70	289.62	84,631.95	925,472.23	76,386.96	894,227.38	894,227.38	894,227.38	894,227.38	894,227.38	894,227.38	894,227.38	894,227.38
11	2	1	1	1880	22328	1842	0.988	0.988	0.988	277.50	83,631.56	1,009,103.67	76,261.46	939,488.85	939,488.85	939,488.85	939,488.85	939,488.85	939,488.85	939,488.85	939,488.85	939,488.85
12	2	1	1	1813	24141	1785	0.988	0.988	0.988	216.70	272.95	1,085,881.61	67,967.95	977,466.80	977,466.80	977,466.80	977,466.80	977,466.80	977,466.80	977,466.80	977,466.80	977,466.80
13	2	1	1	1948	26090	1914	0.988	0.988	0.988	333.98	277.49	88,742.46	1,174,624.06	80,862.47	1,088,319.27	1,088,319.27	1,088,319.27	1,088,319.27	1,088,319.27	1,088,319.27	1,088,319.27	1,088,319.27
14	2	1	1	1821	28011	1912	0.988	0.988	0.988	278.75	87,695.12	1,262,279.18	79,735.13	1,138,054.40	1,138,054.40	1,138,054.40	1,138,054.40	1,138,054.40	1,138,054.40	1,138,054.40	1,138,054.40	1,138,054.40
15	2	1	1	1872	29883	1835	0.988	0.988	0.988	304.03	280.32	84,019.28	1,346,288.46	75,834.29	1,213,688.66	1,213,688.66	1,213,688.66	1,213,688.66	1,213,688.66	1,213,688.66	1,213,688.66	1,213,688.66
16	2	1	1	1859	31742	1838	0.990	0.990	0.990	261.16	278.67	83,588.78	1,429,887.24	75,328.79	1,289,017.48	1,289,017.48	1,289,017.48	1,289,017.48	1,289,017.48	1,289,017.48	1,289,017.48	1,289,017.48
17	2	1	1	1881	33623	1869	0.999	0.999	0.999	307.49	280.79	88,105.95	1,517,963.18	80,155.96	1,369,173.46	1,369,173.46	1,369,173.46	1,369,173.46	1,369,173.46	1,369,173.46	1,369,173.46	1,369,173.46
18	2	1	1	1952	35575	1937	0.999	0.999	0.999	346.17	284.39	90,139.29	1,608,132.47	82,479.30	1,461,652.75	1,461,652.75	1,461,652.75	1,461,652.75	1,461,652.75	1,461,652.75	1,461,652.75	1,461,652.75
19	2	1	1	1825	37400	1811	0.999	0.999	0.999	229.13	281.69	82,660.28	1,690,801.75	74,229.29	1,525,892.04	1,525,892.04	1,525,892.04	1,525,892.04	1,525,892.04	1,525,892.04	1,525,892.04	1,525,892.04
20	2	1	1	1910	39310	1890	0.999	0.999	0.999	298.19	282.49	87,105.46	1,777,907.20	79,175.46	1,605,057.51	1,605,057.51	1,605,057.51	1,605,057.51	1,605,057.51	1,605,057.51	1,605,057.51	1,605,057.51
21	2	1	1	1924	41234	1899	0.999	0.999	0.999	328.89	284.53	86,186.95	1,864,094.15	78,106.96	1,683,164.47	1,683,164.47	1,683,164.47	1,683,164.47	1,683,164.47	1,683,164.47	1,683,164.47	1,683,164.47
22	2	1	1	1844	43078	1828	0.999	0.999	0.999	304.76	285.40	84,170.11	1,948,264.26	75,860.13	1,759,034.59	1,759,034.59	1,759,034.59	1,759,034.59	1,759,034.59	1,759,034.59	1,759,034.59	1,759,034.59
23	2	1	1	1896	45024	1875	0.999	0.999	0.999	269.77	284.27	85,638.95	2,034,103.20	77,785.96	1,836,909.66	1,836,909.66	1,836,909.66	1,836,909.66	1,836,909.66	1,836,909.66	1,836,909.66	1,836,909.66
24	2	1	1	1822	46956	1856	0.988	0.988	0.988	369.57	297.70	86,863.11	2,121,086.52	78,686.13	1,915,678.88	1,915,678.88	1,915,678.88	1,915,678.88	1,915,678.88	1,915,678.88	1,915,678.88	1,915,678.88
25	2	1	1	1860	48716	1837	0.988	0.988	0.988	318.93	286.70	86,931.01	2,207,967.81	78,906.63	1,994,893.31	1,994,893.31	1,994,893.31	1,994,893.31	1,994,893.31	1,994,893.31	1,994,893.31	1,994,893.31
26	2	1	1	1900	50666	1907	0.999	0.999	0.999	285.25	293.15	87,106.28	2,278,878.83	76,409.63	2,073,724.61	2,073,724.61	2,073,724.61	2,073,724.61	2,073,724.61	2,073,724.61	2,073,724.61	2,073,724.61
27	2	1	1	1876	52542	1881	0.999	0.999	0.999	289.25	292.52	83,574.61	2,359,747.21	79,141.30	2,249,734.70	2,249,734.70	2,249,734.70	2,249,734.70	2,249,734.70	2,249,734.70	2,249,734.70	2,249,734.70
28	2	1	1	1916	54469	1878	0.999	0.999	0.999	297.15	293.05	83,835.46	2,442,514.28	76,600.46	2,324,734.70	2,324,734.70	2,324,734.70	2,324,734.70	2,324,734.70	2,324,734.70	2,324,734.70	2,324,734.70
29	2	1	1	1912	56370	1899	0.998	0.998	0.998	369.28	291.25	91,243.95	2,527,768.23	83,638.97	2,399,427.80	2,399,427.80	2,399,427.80	2,399,427.80	2,399,427.80	2,399,427.80	2,399,427.80	2,399,427.80
30	2	1	1	1919	58289	1894	0.999	0.999	0.999	296.11	296.01	88,964.12	2,612,712.34	81,154.13	2,480,427.80	2,480,427.80	2,480,427.80	2,480,427.80	2,480,427.80	2,480,427.80	2,480,427.80	2,480,427.80
31	2	1	1	1883	60172	1851	0.999	0.999	0.999	318.70	296.75	85,818.95	2,702,531.29	77,743.96	2,497,171.76	2,497,171.76	2,497,171.76	2,497,171.76	2,497,171.76	2,497,171.76	2,497,171.76	2,497,171.76
32	2	1	1	1909	62081	1880	0.999	0.999	0.999	298.08	296.82	83,600.80	2,812,132.07	75,320.80	2,542,492.56	2,542,492.56	2,542,492.56	2,542,492.56	2,542,492.56	2,542,492.56	2,542,492.56	2,542,492.56
33	2	1	1	1930	64011	1908	0.999	0.999	0.999	330.81	298.88	87,035.45	2,899,167.52	80,656.47	2,621,548.02	2,621,548.02	2,621,548.02	2,621,548.02	2,621,548.02	2,621,548.02	2,621,548.02	2,621,548.02
34	2	1	1	1910	65921	1877	0.999	0.999	0.999	304.80	297.11	83,764.78	2,988,935.81	81,879.30	2,703,426.32	2,703,426.32	2,703,426.32	2,703,426.32	2,703,426.32	2,703,426.32	2,703,426.32	2,703,426.32
35	2	1	1	1948	67769	1820	0.998	0.998	0.998	323.74	297.83	83,072,700.58	75,289.79	2,778,716.11	2,778,716.11	2,778,716.11	2,778,716.11	2,778,716.11	2,778,716.11	2,778,716.11	2,778,716.11	2,778,716.11
36	2	1	1	1930	69599	1902	0.999	0.999	0.999	321.24	298.48	89,402.46	3,162,193.03	81,622.47	2,860,338.58	2,860,338.58	2,860,338.58	2,860,338.58	2,860,338.58	2,860,338.58	2,860,338.58	2,860,338.58
37	2	1	1	1888	71557	1834	0.999	0.999	0.999	279.55	297.99	83,675.11	3,246,868.14	75,315.13	2,935,653.70	2,935,653.70	2,935,653.70	2,935,653.70	2,935,653.70	2,935,653.70	2,935,653.70	2,935,653.70
38	2	1	1	1898	73425	1864	0.998	0.998	0.998	301.80	298.08	84,269.11	3,330,139.27	76,154.13	3,011,807.83	3,011,807.83	3,011,807.83	3,011,807.83	3,011,807.83	3,011,807.83	3,011,807.83	3,011,807.83
39	2	1	1	1885	75310	1863	0.998	0.998	0.998	327.62	297.52	85,158.95	3,415,296.21	76,905.96	3,088,716.79	3,088,716.79	3,088,716.79	3,088,716.79	3,088,716.79	3,088,716.79	3,088,716.79	3,088,716.79
40	2	1	1	1888	77198	1861	0.998	0.998	0.998	298.38	89,439.61	3,504,735.52	81,439.63	3,170,156.42	3,170,156.42	3,170,156.42	3,170,156.42	3,170,156.42	3,170,156.42	3,170,156.42	3,170,156.42	3,170,156.42
41	2	1	1	1823	79061	1862	0.989	0.989	0.989	301.39	298.90	91,313.12	3,590,038.93	83,603.13	3,253,759.56	3,253,759.56	3,253,759.56	3,253,759.56	3,253,759.56	3,253,759.56	3,253,759.56	3,253,759.56
42	2	1	1	1907	80975	1838	0.988	0.988	0.988	285.46	288.87	82,960.78	3,678,999.71	74,660.79	3,328,410.39	3,328,410.39	3,328,410.39	3,328,410.39	3,328,410.39	3,328,410.39	3,328,410.39	3,328,410.39
43	2	1	1	1907	82882	1898	0.998															

Replication Number	Independent Variables			All Products										LOW			MED			HIGH																																																																																																																	
	MAS	MOH	MIX	Period Demand	Total Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle Time	Total Cycle Time	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI																																																																																																																		
1	1	1	1	1840	1840	1825	1825	0.992	0.992	300.34	300.34	84,212.16	84,212.16	75,884.38	75,884.38	51,588.82	51,588.82	59,447.19	105,036.01	61,425.56	106,461.57	59,057.31	219,557.31	46,888.64	269,446.95	56,289.05	324,735.00	54,727.03	379,462.04	484,189.07	488,916.10	54,727.03	543,643.14	602,156.63	653,622.53	51,466.60	653,622.53	707,544.41	771,771.77	765,316.18	823,379.18	50,480.02	873,869.18	68,047.20	931,916.38	52,710.52	984,626.91	54,076.00	1,038,702.91	46,239.73	1,087,942.64	54,032.48	1,142,035.22	52,334.51	1,194,369.73	54,654.59	1,249,024.32	50,346.80	1,309,373.81	51,416.82	1,369,790.74	51,991.08	1,412,381.63	57,156.63	1,469,538.76	55,475.67	1,525,014.43	52,890.08	1,577,874.52	50,146.54	1,628,020.06	53,264.42	1,681,284.48	52,454.87	1,733,697.78	52,413.31	1,783,697.78	52,005.47	1,834,703.26	57,048.81	1,894,751.88	59,510.31	1,966,262.18	50,488.82	1,946,751.01	46,842.95	1,992,593.95	46,156.55	2,041,750.50	51,771.73	2,093,522.23	51,281.51	2,144,803.74	48,848.33	2,193,652.07	50,242.14	2,243,894.21	55,458.38	2,299,322.98	48,282.40	2,347,604.98	57,950.95	2,405,355.95	48,794.24	2,465,120.19	53,379.56	2,508,499.78	48,587.04	2,567,066.82	54,232.76	2,611,299.55	54,338.81	2,661,299.55	54,338.81	2,710,805.04	56,300.48	2,766,105.58	54,328.27	2,820,433.85	61,329.27	2,881,763.12	59,949.20	2,936,712.32	53,683.88	2,989,376.19	49,986.50	3,039,342.70	55,109.85	3,097,801.55	55,109.85	3,152,901.65	57,910.88	3,210,462.27
2	1	1	1	1900	3765	1859	3684	0.992	0.992	300.16	300.16	85,860.63	170,662.60	77,742.75	153,627.13	69,447.19	105,036.01	61,425.56	106,461.57	59,057.31	219,557.31	46,888.64	269,446.95	56,289.05	324,735.00	54,727.03	379,462.04	484,189.07	488,916.10	54,727.03	543,643.14	602,156.63	653,622.53	51,466.60	653,622.53	707,544.41	771,771.77	765,316.18	823,379.18	50,480.02	873,869.18	68,047.20	931,916.38	52,710.52	984,626.91	54,076.00	1,038,702.91	46,239.73	1,087,942.64	54,032.48	1,142,035.22	52,334.51	1,194,369.73	54,654.59	1,249,024.32	50,346.80	1,309,373.81	51,416.82	1,369,790.74	51,991.08	1,412,381.63	57,156.63	1,469,538.76	55,475.67	1,525,014.43	52,890.08	1,577,874.52	50,146.54	1,628,020.06	53,264.42	1,681,284.48	52,454.87	1,733,697.78	52,413.31	1,783,697.78	52,005.47	1,834,703.26	57,048.81	1,894,751.88	59,510.31	1,966,262.18	50,488.82	1,946,751.01	46,842.95	1,992,593.95	46,156.55	2,041,750.50	51,771.73	2,093,522.23	51,281.51	2,144,803.74	48,848.33	2,193,652.07	50,242.14	2,243,894.21	55,458.38	2,299,322.98	48,282.40	2,347,604.98	57,950.95	2,405,355.95	48,794.24	2,465,120.19	53,379.56	2,508,499.78	48,587.04	2,567,066.82	54,232.76	2,611,299.55	54,338.81	2,661,299.55	54,338.81	2,710,805.04	56,300.48	2,766,105.58	54,328.27	2,820,433.85	61,329.27	2,881,763.12	59,949.20	2,936,712.32	53,683.88	2,989,376.19	49,986.50	3,039,342.70	55,109.85	3,097,801.55	55,109.85	3,152,901.65	57,910.88	3,210,462.27		
3	1	1	1	1907	5665	1893	5677	0.992	0.992	300.16	300.16	85,860.63	170,662.60	77,742.75	153,627.13	69,447.19	105,036.01	61,425.56	106,461.57	59,057.31	219,557.31	46,888.64	269,446.95	56,289.05	324,735.00	54,727.03	379,462.04	484,189.07	488,916.10	54,727.03	543,643.14	602,156.63	653,622.53	51,466.60	653,622.53	707,544.41	771,771.77	765,316.18	823,379.18	50,480.02	873,869.18	68,047.20	931,916.38	52,710.52	984,626.91	54,076.00	1,038,702.91	46,239.73	1,087,942.64	54,032.48	1,142,035.22	52,334.51	1,194,369.73	54,654.59	1,249,024.32	50,346.80	1,309,373.81	51,416.82	1,369,790.74	51,991.08	1,412,381.63	57,156.63	1,469,538.76	55,475.67	1,525,014.43	52,890.08	1,577,874.52	50,146.54	1,628,020.06	53,264.42	1,681,284.48	52,454.87	1,733,697.78	52,413.31	1,783,697.78	52,005.47	1,834,703.26	57,048.81	1,894,751.88	59,510.31	1,966,262.18	50,488.82	1,946,751.01	46,842.95	1,992,593.95	46,156.55	2,041,750.50	51,771.73	2,093,522.23	51,281.51	2,144,803.74	48,848.33	2,193,652.07	50,242.14	2,243,894.21	55,458.38	2,299,322.98	48,282.40	2,347,604.98	57,950.95	2,405,355.95	48,794.24	2,465,120.19	53,379.56	2,508,499.78	48,587.04	2,567,066.82	54,232.76	2,611,299.55	54,338.81	2,661,299.55	54,338.81	2,710,805.04	56,300.48	2,766,105.58	54,328.27	2,820,433.85	61,329.27	2,881,763.12	59,949.20	2,936,712.32	53,683.88	2,989,376.19	49,986.50	3,039,342.70	55,109.85	3,097,801.55	55,109.85	3,152,901.65	57,910.88	3,210,462.27		
4	1	1	1	1824	7489	1804	7381	0.992	0.992	303.34	303.34	84,212.16	84,212.16	75,884.38	75,884.38	51,588.82	51,588.82	59,447.19	105,036.01	61,425.56	106,461.57	59,057.31	219,557.31	46,888.64	269,446.95	56,289.05	324,735.00	54,727.03	379,462.04	484,189.07	488,916.10	54,727.03	543,643.14	602,156.63	653,622.53	51,466.60	653,622.53	707,544.41	771,771.77	765,316.18	823,379.18	50,480.02	873,869.18	68,047.20	931,916.38	52,710.52	984,626.91	54,076.00	1,038,702.91	46,239.73	1,087,942.64	54,032.48	1,142,035.22	52,334.51	1,194,369.73	54,654.59	1,249,024.32	50,346.80	1,309,373.81	51,416.82	1,369,790.74	51,991.08	1,412,381.63	57,156.63	1,469,538.76	55,475.67	1,525,014.43	52,890.08	1,577,874.52	50,146.54	1,628,020.06	53,264.42	1,681,284.48	52,454.87	1,733,697.78	52,413.31	1,783,697.78	52,005.47	1,834,703.26	57,048.81	1,894,751.88	59,510.31	1,966,262.18	50,488.82	1,946,751.01	46,842.95	1,992,593.95	46,156.55	2,041,750.50	51,771.73	2,093,522.23	51,281.51	2,144,803.74	48,848.33	2,193,652.07	50,242.14	2,243,894.21	55,458.38	2,299,322.98	48,282.40	2,347,604.98	57,950.95	2,405,355.95	48,794.24	2,465,120.19	53,379.56	2,508,499.78	48,587.04	2,567,066.82	54,232.76	2,611,299.55	54,338.81	2,661,299.55	54,338.81	2,710,805.04	56,300.48	2,766,105.58	54,328.27	2,820,433.85	61,329.27	2,881,763.12	59,949.20	2,936,712.32	53,683.88	2,989,376.19	49,986.50	3,039,342.70	55,109.85	3,097,801.55	55,109.85	3,152,901.65	57,910.88	3,210,462.27
5	1	1	1	1824	7489	1804	7381	0.992	0.992	303.34	303.34	84,212.16	84,212.16	75,884.38	75,884.38	51,588.82	51,588.82	59,447.19	105,036.01	61,425.56	106,461.57	59,057.31	219,557.31	46,888.64	269,446.95	56,289.05	324,735.00	54,727.03	379,462.04	484,189.07	488,916.10	54,727.03	543,643.14	602,156.63	653,622.53	51,466.60	653,622.53	707,544.41	771,771.77	765,316.18	823,379.18	50,480.02	873,869.18	68,047.20	931,916.38	52,710.52	984,626.91	54,076.00	1,038,702.91	46,239.73	1,087,942.64	54,032.48	1,142,035.22	52,334.51	1,194,369.73	54,654.59	1,249,024.32	50,346.80	1,309,373.81	51,416.82	1,369,790.74	51,991.08	1,412,381.63	57,156.63	1,469,538.76	55,475.67	1,525,014.43	52,890.08	1,577,874.52	50,146.54	1,628,020.06	53,264.42	1,681,284.48	52,454.87	1,733,697.78	52,413.31	1,783,697.78	52,005.47	1,834,703.26	57,048.81	1,894,751.88	59,510.31	1,966,262.18	50,488.82	1,946,751.01	46,842.95	1,992,593.95	46,156.55	2,041,750.50	51,771.73	2,093,522.23	51,281.51	2,144,803.74	48,848.33	2,193,652.07	50,242.14	2,243,894.21	55,458.38	2,299,322.98	48,282.40	2,347,604.98	57,950.95	2,405,355.95	48,794.24	2,465,120.19	53,379.56	2,508,499.78	48,587.04	2,567,066.82	54,232.76	2,611,299.55	54,338.81	2,661,299.55	54,338.81	2,710,805.04	56,300.48	2,766,105.58	54,328.27	2,820,433.85	61,329.27	2,881,763.12	59,949.20	2,936,712.32	53,683.88	2,989,376.19	49,986.50	3,039,342.70	55,109.85	3,097,801.55	55,109.85	3,152,901.65	57,910.88	3,210,462.27
6	1	1	1	1897	11252	1898	11082	0.992	0.992	303.34	303.34	84,212.16	84,212.16	75,884.38	75,884.38	51,588.82	51,588.82	59,447.19	105,036.01	61,425.56	106,461.57	59,057.31	219,557.31	46,888.64	269,446.95	56,289.05	324,735.00	54,727.03	379,462.04	484,189.07	488,916.10	54,727.03	543,643.14	602,156.63	653,622.53	51,466.60	653,622.53	707,544.41	771,771.77	765,316.18	823,379.18	50,480.02	873,869.18	68,047.20	931,916.38	52,710.52	984,626.91	54,076.00	1,038,702.91	46,239.73	1,087,942.64	54,032.48	1,142,035.22	52,334.51	1,194,369.73	54,654.59	1,249,024.32	50,346.80	1,309,373.81	51,416.82	1,369,790.74	51,991.08	1,412,381.63	57,156.63	1,469,538.76	55,475.67	1,525,014.43	52,890.08	1,577,874.52	50,146.54	1,628,020.06	53,264.42	1,681,284.48	52,454.87	1,733,697.78	52,413.31	1,783,697.78	52,005.47	1,834,703.26	57,048.81	1,894,751.88	59,510.31	1,966,262.18	50,488.82	1,946,751.01	46,842.95	1,992,593.95	46,156.55	2,041,750.50	51,771.73	2,093,522.23	51,281.51	2,144,803.74	48,848.33	2,193,652.07	50,242.14	2,243,894.21	55,458.38	2,299,322.98	48,282.40	2,347,604.98	57,950.95	2,405,355.95	48,794.24	2,465,120.19	53,379.56	2,508,499.78	48,587.04	2,567,066.82	54,232.76	2,611,299.55	54,338.81	2,661,299.55	54,338.81	2,710,805.04	56,300.48	2,766,105.58	54,328.27	2,820,433.85	61,329.27	2,881,763.12	59,949.20	2,936,712.32	53,683.88	2,989,376.19	49,986.50	3,039,342.70	55,109.85	3,097,801.55	55,109.85	3,152,901.65	57,910.88	3,210,462.27
7	1	1	1	1902	16056	1898	14817	0.992	0.992	303.34	303.34	84,212.16	84,212.16	75,884.38	75,884.38	51,588.82	51,588.82	59,447.19	105,036.01	61,425.56	106,461.57	59,057.31	219,557.31	46,888.64	269,446.95	56,289.05	324,735.00	54,727.03	379,462.04	484,189.																																																																																																							

Replication Number	Independent Variables			All Products										LOW			MED			HIGH		
	MAS	MOH	MIX	Period Demand	Period Shipped	Total Shipped	Period DFR	Total DFR	Period Cycle Time	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI	Period NOI	Cumulative NOI			
1	1	1	2	2357	2357	1659	0.731	0.731	704.93	99.824.01	99.824.01	93.964.60	93.964.60	70.635.80	70.635.80	70.635.80	70.635.80	70.635.80	70.635.80			
2	1	1	2	2408	4766	1709	0.723	0.723	698.25	104.125.32	203.949.33	98.830.92	192.795.52	76.502.11	146.137.91	146.137.91	146.137.91	146.137.91	146.137.91			
3	1	1	2	2272	7037	1681	0.736	0.736	719.23	703.90	100.620.23	304.569.62	95.000.82	287.795.32	217.809.92	217.809.92	217.809.92	217.809.92	217.809.92			
4	1	1	2	2410	9447	1675	0.711	0.711	686.08	699.46	102.544.35	407.113.91	97.009.95	384.896.29	73.771.14	291.581.06	291.581.06	291.581.06	291.581.06			
5	1	1	2	2263	11740	1636	0.733	0.733	709.77	701.48	101.461.64	508.572.24	97.732.24	480.628.63	72.403.43	393.684.46	393.684.46	393.684.46	393.684.46			
6	1	1	2	2315	16451	1707	0.726	0.726	709.64	702.86	103.002.03	611.577.59	97.532.63	578.161.16	74.203.82	438.188.31	438.188.31	438.188.31	438.188.31			
7	1	1	2	2405	16466	1709	0.724	0.724	694.50	701.35	102.008.86	713.686.46	96.506.96	674.688.12	73.178.15	511.366.46	511.366.46	511.366.46	511.366.46			
8	1	1	2	2405	16881	1709	0.724	0.724	691.65	700.38	102.008.86	815.694.31	96.506.96	771.175.07	73.178.15	584.544.80	584.544.80	584.544.80	584.544.80			
9	1	1	2	2405	21266	1709	0.724	0.724	688.81	102.008.86	102.008.86	917.694.17	96.506.96	897.682.03	73.178.15	674.544.80	674.544.80	674.544.80	674.544.80			
10	1	1	2	2405	23571	1709	0.724	0.724	688.86	102.008.86	102.008.86	1019.013.03	96.506.96	894.168.98	73.178.15	730.900.96	730.900.96	730.900.96	730.900.96			
11	1	1	2	2323	25894	1654	0.737	0.737	697.08	99.879.69	1.122.744.31	97.666.87	1.061.959.86	74.338.06	895.238.98	895.238.98	895.238.98	895.238.98				
12	1	1	2	2397	28391	1708	0.717	0.717	688.14	99.877.73	1.222.338.49	93.879.77	1.155.735.63	70.550.97	875.789.63	875.789.63	875.789.63	875.789.63				
13	1	1	2	2371	30762	1662	0.731	0.731	688.00	99.775	1.023.04.46	1.324.362.83	96.485.04	1.252.220.68	73.195.23	948.946.16	948.946.16	948.946.16	948.946.16			
14	1	1	2	2380	31142	1620	0.725	0.725	689.09	99.716	1.01.163.30	1.425.525.24	96.383.90	1.347.804.57	72.055.09	1.021.001.25	1.021.001.25	1.021.001.25	1.021.001.25			
15	1	1	2	2314	35466	1791	0.721	0.721	689.79	99.730	1.01.205.32	1.526.731.56	96.750.82	1.443.355.46	72.422.11	1.093.423.35	1.093.423.35	1.093.423.35	1.093.423.35			
16	1	1	2	2342	37786	1685	0.734	0.734	697.60	102.161.81	1.628.893.37	96.727.41	1.540.882.90	73.388.80	1.166.821.96	1.166.821.96	1.166.821.96	1.166.821.96	1.166.821.96			
17	1	1	2	2341	40139	1765	0.729	0.729	686.65	99.542.52	1.727.817.89	93.285.11	1.633.368.01	69.966.31	1.236.778.27	1.236.778.27	1.236.778.27	1.236.778.27	1.236.778.27			
18	1	1	2	2289	42428	1706	0.727	0.727	686.04	99.542.52	1.827.776.77	94.309.48	1.727.877.50	70.980.87	1.307.768.94	1.307.768.94	1.307.768.94	1.307.768.94	1.307.768.94			
19	1	1	2	2340	44768	1723	0.730	0.730	686.54	99.542.52	1.928.677.15	95.375.97	1.823.053.46	72.047.16	1.379.806.10	1.379.806.10	1.379.806.10	1.379.806.10	1.379.806.10			
20	1	1	2	2351	47119	1624	0.727	0.727	702.43	99.832	1.00.708.89	2.029.386.03	94.869.48	1.917.922.94	71.540.67	1.451.346.77	1.451.346.77	1.451.346.77	1.451.346.77			
21	1	1	2	2379	49498	1687	0.728	0.728	692.97	102.400.11	2.131.846.14	95.900.70	2.014.913.65	73.681.89	1.525.008.67	1.525.008.67	1.525.008.67	1.525.008.67	1.525.008.67			
22	1	1	2	2371	51899	1695	0.726	0.726	698.46	104.381.28	2.236.227.42	96.991.87	2.113.905.62	75.683.06	1.600.671.73	1.600.671.73	1.600.671.73	1.600.671.73	1.600.671.73			
23	1	1	2	2371	54188	1614	0.727	0.727	709.81	100.141.42	2.336.368.94	94.267.02	2.208.172.54	70.938.21	1.671.609.94	1.671.609.94	1.671.609.94	1.671.609.94	1.671.609.94			
24	1	1	2	2337	56525	1648	0.727	0.727	699.84	99.708	1.00.141.42	2.436.141.24	93.915.90	2.302.885.64	70.584.19	1.742.194.13	1.742.194.13	1.742.194.13	1.742.194.13			
25	1	1	2	2368	58933	1704	0.721	0.721	689.67	98.486.69	2.534.627.63	92.657.19	2.394.743.72	69.328.38	1.811.622.51	1.811.622.51	1.811.622.51	1.811.622.51	1.811.622.51			
26	1	1	2	2368	61191	1738	0.727	0.727	704.74	99.474	2.639.951.40	96.654.16	2.491.956.88	71.925.35	1.885.047.88	1.885.047.88	1.885.047.88	1.885.047.88	1.885.047.88			
27	1	1	2	2334	63525	1688	0.728	0.728	694.47	100.300.96	2.737.252.26	94.621.46	2.586.116.38	72.646.11	1.962.240.51	1.962.240.51	1.962.240.51	1.962.240.51	1.962.240.51			
28	1	1	2	2336	65881	1708	0.731	0.731	702.10	99.542.52	2.836.791.09	95.974.92	2.682.093.27	72.986.11	2.028.888.02	2.028.888.02	2.028.888.02	2.028.888.02	2.028.888.02			
29	1	1	2	2343	68204	1648	0.731	0.731	697.64	99.757.40	2.940.478.99	96.117.99	2.778.211.28	72.789.19	2.101.675.81	2.101.675.81	2.101.675.81	2.101.675.81	2.101.675.81			
30	1	1	2	2427	70631	1650	0.722	0.722	689.84	99.300	3.043.624.89	93.043.88	2.875.276.56	73.736.48	2.175.412.91	2.175.412.91	2.175.412.91	2.175.412.91	2.175.412.91			
31	1	1	2	2344	72975	1610	0.720	0.720	708.58	99.758	3.143.331.52	94.328.44	2.969.604.98	70.999.63	2.246.411.91	2.246.411.91	2.246.411.91	2.246.411.91	2.246.411.91			
32	1	1	2	2329	75304	1711	0.729	0.729	703.24	100.227.84	3.246.057.14	97.276.21	3.068.881.20	73.947.40	2.320.359.32	2.320.359.32	2.320.359.32	2.320.359.32	2.320.359.32			
33	1	1	2	2312	77616	1700	0.735	0.735	697.78	100.274.01	3.346.331.14	94.719.60	3.161.800.80	71.900.79	2.391.750.11	2.391.750.11	2.391.750.11	2.391.750.11	2.391.750.11			
34	1	1	2	2371	79987	1625	0.729	0.729	700.37	99.758	3.446.446.18	94.330.63	3.255.931.43	71.001.82	2.462.751.93	2.462.751.93	2.462.751.93	2.462.751.93	2.462.751.93			
35	1	1	2	2428	82415	1713	0.727	0.727	691.12	99.758	3.550.970.09	99.204.51	3.355.135.93	75.875.70	2.538.627.63	2.538.627.63	2.538.627.63	2.538.627.63	2.538.627.63			
36	1	1	2	2286	84701	1690	0.731	0.731	713.62	99.810	3.653.165.63	95.646.14	3.461.182.07	73.317.33	2.611.944.96	2.611.944.96	2.611.944.96	2.611.944.96	2.611.944.96			
37	1	1	2	2419	87120	1742	0.727	0.727	685.77	103.035.15	3.756.220.78	97.770.75	3.549.552.82	74.441.94	2.686.386.90	2.686.386.90	2.686.386.90	2.686.386.90	2.686.386.90			
38	1	1	2	2423	89543	1792	0.719	0.719	697.22	100.297.47	3.856.518.25	94.873.06	3.644.525.88	71.544.26	2.757.931.16	2.757.931.16	2.757.931.16	2.757.931.16	2.757.931.16			
39	1	1	2	2367	91800	1683	0.730	0.730	697.17	102.140.59	3.958.658.95	96.591.19	3.741.017.08	73.282.38	2.831.183.54	2.831.183.54	2.831.183.54	2.831.183.54	2.831.183.54			
40	1	1	2	2336	94236	1692	0.731	0.731	708.14	99.757.40	4.060.446.65	95.912.80	3.836.039.88	72.583.99	2.903.777.53	2.903.777.53	2.903.777.53	2.903.777.53	2.903.777.53			
41	1	1	2	2308	96544	1717	0.728	0.728	697.62	101.169.59	4.161.619.65	94.605.09	3.932.539.97	72.380.38	2.976.057.91	2.976.057.91	2.976.057.91	2.976.057.91	2.976.057.91			
42	1	1	2	2394	98938	1740	0.722	0.722	685.32	99.758	4.263.217.41	96.135.46	4.028.872.45	72.894.65	3.048.862.46	3.048.862.46	3.048.862.46	3.048.862.46	3.048.862.46			
43	1	1	2	2328	101266	1688	0.732	0.732	708.62	103.076.25	4.366.263.66	97.651.86	4.123.324.27	74.333.04	3.123.185.60	3.123.185.60	3.123.185.60	3.123.185.60	3.123.185.60			
44	1	1	2	2377	103643	1733	0.723	0.723	685.95	99.727	4.468.331.00	96.565.43	4.222.887.70	73.234.62	3.196.420.12	3.196.420.12	3.196.420.12	3.196.420.12	3.196.420.12			
45	1	1	2	2314	105957	1756	0.721	0.721	686.93	100.668.05	4.568.969.95	95.058.65	4.317.946.39	71.729.84	3.268.149.96	3.268.149.96	3.268.149.96	3.268.149.96	3.268.149.96			
46	1	1	2	2331	108288	1639	0.733	0.733	697.42	102.988.08	4.671.797.63	97.222.68	4.415.170.03	73.894.87	3.342.044.83	3.342.044.83	3.342.044.83	3.342.044.83	3.342.044.83			
47	1	1	2	2311	110628	1639	0.727	0.727	705.86	99.727	4.773.860.83	96.556.90	4.511.289.93	73.230.09	3.416.274.92	3.416.274.92	3.416.274.92	3.416.274.92	3.416.274.92			
48	1	1	2	2240	115293	1674	0.726	0.726	692.63	99.727	4.875.216.88	96.766.24	4.607.495.17	72.437.43	3.487.712.35	3.487.712.35	3.487.712.35	3.487.712.35	3.487.712.35			
49	1	1	2	2240	117637	1724	0.726	0.726	701.82	99.727	4.972.439.78	96.766.24	4.698.623.97	67.999.99	3.555.612.34	3.555.612.34	3.555.612.34	3.555.612.34	3.555.612.34			
50	1	1	2	2344	119657	1769	0.727	0.727	689.61	99.727	5.073.563.30</											

