An Analysis of Manufacturer Benefits under Vendor Managed Systems

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Vendor Managed Inventory (VMI) has attracted a lot of attention due to its benefits such as fewer stock-outs, higher sales and lower inventory levels at the retailers. Vendor Managed Availability (VMA) is an improvement that exploits the advantages beyond VMI. We complement VMA by analyzing the benefits beyond information sharing and by clearly assessing the motivation for the manufacturer (vendor) behind joining to such a program. We show that such vendor managed systems provide increased flexibility in manufacturer’s operations and may bring additional benefits. We analyze how the system parameters affect the profitability and determine the conditions that make the vendor managed system a viable strategy for the manufacturer.

Keywords: Vertical Collaboration; Vendor Managed Inventory; Capacity Management; Operational Flexibility; Consignment Stock

1 Introduction

Vendor Managed Inventory (VMI) is a collaborative process between a supplier/manufacturer and a manufacturer/retailer/distributor, where the manufacturer gains access to the demand and inventory information at the retailer and uses this information to “better” manage the retailer’s inventory. VMI started as a pilot program in retail industry between Procter&Gamble and WalMart in 80’s and resulted in significant benefits, such as lower inventory levels, fewer stock-outs and increased sales, and has been adopted by many other supply chains such as Dell’s, Barilla’s or Nestle’s. In many research and business articles, the benefits of VMI are attributed to information sharing between the manufacturer and the retailer (see Cachon and Fisher 1997; Schenck and McInerney 1998). However, there is more to VMI than just the information availability; there are benefits hidden in the increased flexibility of the manufacturer’s production operations. There exist limited analytical work in literature on how the manufacturer can translate this flexibility into benefit, and why the parties join to a VMI program. We believe that it is important to emphasize the benefits of VMI additional to information sharing, so that the motivation behind joining to a VMI program is better comprehended.

In a vendor managed setting, although the manufacturer takes control of inventory, it is the retailer that

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usually benefits from manufacturer managing the inventory (Dong and Xu 2002). The reason is, the retailer can always set the terms of the agreement such that the performance measures (such as number of stock-outs, average inventory level, etc.) will improve. Whether the manufacturer benefits from the vendor managed system on the other hand, depends on how well the manufacturer can take advantage of the increased flexibility. In the agreement, the retailer may reflect a required product availability on the shelf, or service level by imposing a lower bound on the inventory level. Similarly, due to shelf space constraints or to avoid high inventory levels, the retailer may limit the amount of replenishment from the manufacturer. Therefore a contract may consist of an upper and a lower bound on inventory level, where overshooting or undershooting by the manufacturer is penalized. While penalties compel the manufacturer to conform with the inventory limits, it is definitely a challenging task for the retailer to determine the penalties as well as to set the bounds on the inventory level that will result in the desired service level or inventory holding cost.

Our modeling of VMI is closer to “vendor managed availability” (Hausman 2003), where the vendor is more flexible in terms of replenishment operations than VMI, since in VMI, replenishments are more restricted due to the bounds on retailer’s inventory level. Vendor managed availability has been practiced by several major retailers such as J.C.Penney, or Costco. J.C.Penney supplies shirts from a Hong-Kong based shirt-maker where the supplier completely controls the inventory by monitoring J.C.Penney’s stock levels and making replenishments directly to the store, if necessary. To ensure availability, at times the supplier expedites the delivery by shipping through air (Kahn 2002; Hausman 2003). Similarly, Kimberly-Clark, a supplier of products such as diapers, tissues or paper towels for Costco in U.S., is very flexible in its replenishment operations. The company simply “keeps each [Costco] store’s inventory as low as possible without risking empty shelves” (Nelson and Zimmerman 2000). These examples describe more flexible agreement terms between the manufacturer and the retailer. To reflect this practice, in our vendor managed model we assume that the service level is the only constraint for the manufacturer, which results in an increased flexibility even compared to VMI. For instance, at times the manufacturer may not prefer to replenish retailer’s stock if the capacity can be used for a more profitable order. At other times when there is excess capacity, i.e., when the capacity is less valuable, several replenishments may enable an increased service level at the retailer. The retailer ends up with the same service level whereas the manufacturer effectively manages its production, capacity allocation and replenishment operations.

In this paper, we consider the notions introduced by VMA, an enhanced version of VMI. In the rest of the text, we use the terms VMA or vendor managed system to represent this enhanced version of VMI.

In this study, we model a supply chain consisting of a single manufacturer and a retailer. We first define the traditional system under which the manufacturer and the retailer operate, and then introduce the vendor managed system and compare the two systems. We assume that the retailer sets the terms of the contract
such that she is never worse off under the new (vendor managed) system. We make the analysis from the perspective of the manufacturer who carries most of the collaboration burden. The retailer faces stochastic demand and in the traditional system periodically places orders to the manufacturer. Manufacturer has limited capacity to meet the orders from the retailer, and a more expensive outsourcing option. To analyze benefits due to vendor managed system alone, our proposed model for the traditional system considers a manufacturer that has full information on end-demand distribution, demand realization and inventory levels at the retailer and hence revisits capacity planning aspects of operating a traditional manufacturing system. We assume that the parties do not share cost information. Furthermore, information on available capacity or end-of-period inventory level at the manufacturer is not shared with the retailer. Our focus is on the vertical collaboration process in the supply chain under this asymmetric and partially shared information setting.

In vendor managed systems the issue of who owns the inventory depends on the relationship between the manufacturer (supplier) and the retailer (manufacturer). If the manufacturer is very powerful (such as Dell) it may force the suppliers to own the inventory at the manufacturer’s site or at a supply hub nearby. On the other hand, if supplier is powerful then inventory may not be consigned. Intel, for instance, although has an agreement with Dell, does not operate through supply hub like other suppliers (Barnes et al. 2000). We consider two types of vendor managed agreements, consignment stock and no-consignment stock, and for each type analyze how the manufacturer may benefit from managing the retailer’s inventory. In our model there does not exist an upper and lower bound restriction at the retailer’s inventory level; however retailer explicitly specifies service level and average inventory level requirements. Given this setting we address the following questions: (i) Are there any benefits for the manufacturer in managing the retailer’s inventory apart from what is already achieved by sharing demand and inventory information? (ii) What are the conditions that make the manufacturer better off under the vendor managed system considered? (iii) Under the vendor managed system should the manufacturer consign the stock or not?

Our work contributes to the literature in several ways. Our work is one of the few studies that analyzes benefits due to vendor managed systems from the manufacturer’s perspective and that identifies the conditions to make the manufacturer willing to join such an agreement. Earlier studies either ignore the motivation behind vendor managed systems, or focus only on total supply chain benefits rather than the individuals’. Furthermore, we make a comparison of benefits under consignment stock and no-consignment stock models, to determine the type of agreement the manufacturer will benefit, whereas previous literature mostly assume centralized, consignment stock models.

The remainder of the paper is organized as follows. In §2 we review the previous work on vendor managed inventory systems. In §3 and §4 the model characteristics and structural properties are presented. In
§5 we make an experimental analysis and discuss the results, and based on these discussions we provide managerial insights. We present our conclusions in §6.

2 Literature Review

Majority of existing studies analyze the vendor managed system in a manufacturing-retailer setting, while a few consider a supplier-manufacturer setting (Choi, Dai and Song 2004). Inventory ownership is modeled either by totally consigned stock, or by the transfer of the title at the time of arrival. In most of the previous studies, the focus of the analysis is limited to designing an optimal operating policy for the vendor in a vendor managed system, and the motivation of the vendor in managing the inventory is not under consideration.

In the analysis of the vendor managed systems under a single manufacturer and multiple retailers, the focus is mainly on the savings in transportation due to better order consolidation or savings due to coordination of retailer replenishments. To analyze the benefit of VMI, Cetinkaya and Lee (2000) compare a VMI system with a traditional system. In the traditional system the manufacturer sends a shipment immediately when the demand arrives, while in VMI system shipments are consolidated. Authors determine the optimal dispatch quantity under VMI considering the inventory cost and the transportation cost incurred by the manufacturer, and conclude that when inventory holding cost and dispatching cost are low, VMI results in significant savings for the manufacturer. Kleywegt, Nori and Savelsberg (2002) study an inventory routing problem of a manufacturer who owns the inventory at the retailers. An approximation method is developed to find the minimum cost routing policy, however, there does not exist a discussion on whether the manufacturer is better off under the vendor managed system. Waller, Johnson and Davis (1999) also consider a multiple retailer setting and through a simulation analysis demonstrate the effects of VMI on the inventory levels at the retailers and on the capacity utilization at the manufacturer. VMI results in savings due a decrease in the inventory levels, which is a consequence of the increased frequency of retailer replenishments. Aviv and Federgruen (1998) consider a capacitated supplier with multiple retailers and analyze how coordination of retailer orders under VMI decrease the system-wide cost of operation. They explicitly model a traditional system with no information sharing and with full information sharing to assess the benefits of VMI beyond information sharing.

Fry, Kapuscinski and Olsen (2001) compare a VMI system with a traditional system in a single manufacturer, single retailer setting under full information sharing. The authors identify the optimal operating policies of both the manufacturer and the retailer in a stochastic setting. Under VMI the retailer determines the maximum inventory level and the vendor incurs a penalty if the inventory level is outside the limits. Authors find that VMI performs close to a centralized model in the presence of high demand.
variance and high cost of outsourcing. Several other papers study the optimal decisions of the manufacturer under VMI in a deterministic environment. Valentini and Zavanella (2003) and Shah and Goh (2006) consider a consignment stock system where the demand is deterministic with a constant rate. Jaruphongsa, Cetinkaya and Lee (2004) study a problem with delivery time windows and early shipment penalties under dynamic demand. The authors propose a dynamic programming algorithm to obtain the minimum cost under VMI.

Depending on the form of agreement between the retailers and the manufacturer, the system under vendor managed regime can be very close to a centralized system. A number of papers analyze the role of VMI as a channel coordinator. Bernstein, Chen and Federgruen (2006) study the constant wholesale price and quantity discount contracts that lead to perfect coordination in a supply chain with multiple competing retailers, and show how VMI helps achieve the coordination. Nagarajan and Rajagopalan (2008) show that simple contracts in VMI can improve the performance of the overall system under certain conditions. Dong and Xu (2002) analyze the benefits of VMI both in terms of total channel cost and vendor’s cost. In their model the retailers set the purchasing price in the contract and the supplier in turn determines the selling quantity. Authors determine the conditions under which the supplier benefits from VMI and conclude that VMI can always decrease the cost of channel as a whole. Fry, Kapuscinski and Olsen (2001) also discuss centralization of the supply chain.

There has been few work on the service level considerations in a VMI system. In most of the papers the service level is assumed implicit in the lower inventory level set by the lower-echelon. Choi, Dai and Song (2004) study the service level relationship between a supplier and a manufacturer in a VMI framework and show that high service levels at the supplier does not guarantee the desired service level at the manufacturer and that expected backorders should also be taken into account.

Our study is most closely related to Fry, Kapuscinski and Olsen (2001). We study a single manufacturer single retailer system and compare the vendor managed system with the traditional system to quantify the benefits beyond information sharing. However, we focus on the benefits to the manufacturer to determine the motivation to make an agreement. We furthermore consider capacity management as an important factor in determining the benefits of vendor managed system. Additionally, we study both consignment and no-consignment models to identify the conditions that make either model beneficial for the manufacturer. In our model, we do not necessarily regard the vendor managed system as a coordinated system. We propose a more realistic setting with asymmetric and partial information sharing and focus on the collaboration process. Since usually it is the manufacturer that is reluctant in these agreements, we analyze the problem from manufacturer’s perspective. Finally, we take service level considerations explicitly into account.

In summary, our model differs from the existing studies in the following aspects: (i) We look at manufac-
turer benefits in joining to the vendor managed system. (ii) We identify the benefits beyond information sharing to clearly assess the manufacturer’s motivation. (iii) We explicitly model the consignment and no-consignment systems and provide a comparison of these systems to determine which type of agreement is more beneficial to the manufacturer. In practice, if the lower echelon is more powerful, the stock is usually consigned by the manufacturer. Otherwise, if the manufacturer is powerful, the stock is not necessarily consigned. Therefore it is not apparent whether the manufacturer should consign the stock or not. (iv) Finally, we analyze how benefits under vendor managed system change with system parameters. Specifically, we measure the effect of capacity management and provide a detailed analysis of the benefits from production and transportation flexibility.

3 A Modeling Framework for the Manufacturer

We compare two settings; a traditional system where the retailer manages and owns the inventory, and a vendor managed system. In the vendor managed system we model two cases based on the ownership of stock. Under no-consignment stock model (VM-NC), the stock is managed by the manufacturer while owned by the retailer. Under consignment stock model (VM-C), the inventory is both managed and owned by the manufacturer. We assume the retailer accepts the agreement only if the performance measures are as good compared to the traditional case.

We consider a periodic-review model where the manufacturer has limited and non-stationary capacity, which is known by the manufacturer in advance. The non-stationarity in the capacity reflects an environment where the manufacturer has several customers and allocates some portion of the capacity to the retailer and the remaining to the other orders. We assume that the capacity allocated to the retailer may be 0 in some periods, i.e., the manufacturer produces for the retailer in every $T_p$ periods, and without loss of generality we assume non-negative capacity in the first period of $T_p$. We call the time span between two positive capacity levels as the production cycle. Note that cyclic production concept is a well-known and utilized idea in the literature. Maxwell and Muckstadt (1985) introduced the idea of consistent and realistic reorder intervals. Li and Wang (2007) mention cyclic structures within the supply chain as a coordination mechanism. Fry, Kapuscinski and Olsen (2001) consider a similar cyclic structure in their study. We further assume that the level of capacity may be non-stationary for the periods in which the manufacturer produces for the retailer. We assume this non-stationarity also shows a cyclic behaviour. In other words, in every $T_m$ periods the level of the capacity is the same and $T_m$ may consist of several $T_p$ cycles, each cycle with possibly a different capacity level (see Figure 1). We call this larger cycle as the capacity cycle. Similarly, due to scheduling practices the retailer places a replenishment order to the manufacturer in every $T_r$ periods. We call the retailer’s cycle as the replenishment cycle.
We assume the replenishment orders are quantized, where unit replenishment size $Q$ reflects economies of scale in manufacturing and transportation and is an agreed-upon quantity between the manufacturer and the retailer. Note that, this assumption implies that the manufacturer is expected to operate with this “bucket” size $Q$ with all of the customers. Hence, we can assume that the capacity at the manufacturer is a non-negative integer multiple of $Q$. This type of environment can be observed in practice. For example, DMC, a French thread company, lowered its shipment size from 24-unit cases to 12-unit cases after an agreement made with WalMart. Since switching to 12-unit case required significant investment now the company is shipping in 12-unit cases to all of its customers (Fishman 2006).

The end-item demand is stochastic and stationary. Holding cost is incurred based on end-of-period inventory level, and the retailer operates based on a service level constraint. Excess demand at the retailer can be backlogged (there is no cost associated), however the manufacturer (always) meets the retailer order either through regular stock or by subcontracting (for a similar usage of subcontracting option, see Gavirneni, Kapuscinski and Tayur 1999). Here, the term “subcontracting” actually corresponds to a variety of alternatives to meet the unsatisfied demand. The manufacturer can use an additional “set up” from the capacity of other products/customers, overtime production, expedite the supply, or let the retailer to take care of unmet demand but pay a(n implied) penalty. We assume that transportation time is negligible and hence the produced amount is delivered at the same period (overnight). Note that this is consistent with the JIT delivery concept.

We model the retailer’s and the manufacturer’s problem under the traditional system, and the manufacturer’s problem under the vendor managed system as a Markov Decision Process (MDP). We determine the optimal operating policy under each system. Model parameters, decision variables and state variables are presented in Table 1.

One of the objectives of this study is to quantify the benefits of the vendor managed system for the
manufacturer when demand and inventory information of the retailer is available. Specifically, we make
the following assumptions on information sharing:

1. The information of periodic demand realization, end-of-period inventory level at the retailer, and
retailer’s demand distribution is provided by the retailer to the manufacturer.

2. Information of unit inventory holding cost or any other cost information at the retailer is not
shared with the manufacturer. Similarly, cost information of the manufacturer is not shared with
the retailer. Cost information is mutually unavailable.

3. Information on capacity level and end-of-inventory level at the manufacturer is not shared with the
retailer. Therefore information sharing is asymmetric and partial.

Table 1: Notation for Traditional and Vendor Managed System Models

<table>
<thead>
<tr>
<th>Parameters</th>
<th>Description</th>
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<tbody>
<tr>
<td>$T_p$</td>
<td>length of the production cycle for the manufacturer</td>
</tr>
<tr>
<td>$T_m$</td>
<td>length of the capacity cycle for the manufacturer</td>
</tr>
<tr>
<td>$T_r$</td>
<td>length of the replenishment cycle for the retailer under traditional system</td>
</tr>
<tr>
<td>$D_i$</td>
<td>random variable denoting demand over $i$ periods, $i \in {1, \cdots, T_r}$</td>
</tr>
<tr>
<td>$P_i$</td>
<td>probability mass function for $D_i$</td>
</tr>
<tr>
<td>$Q$</td>
<td>batch order (dispatch) quantity</td>
</tr>
<tr>
<td>$c$</td>
<td>unit production cost</td>
</tr>
<tr>
<td>$w$</td>
<td>unit outsourcing cost</td>
</tr>
<tr>
<td>$h$</td>
<td>(manufacturer’s) unit holding cost</td>
</tr>
<tr>
<td>$1 - \beta$</td>
<td>service level at the retailer</td>
</tr>
<tr>
<td>$z$</td>
<td>the number of production cycles in a capacity cycle, $zT_p = T_m$</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Decision Variables</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>$R$</td>
<td>reorder level at the retailer</td>
</tr>
<tr>
<td>$p^n$</td>
<td>number of lots of $Q$ produced in period $n$</td>
</tr>
<tr>
<td>$d^n$</td>
<td>number of lots of $Q$ dispatched in period $n$</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>State Variables</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>$I^m_n$</td>
<td>number of lots on-hand at the manufacturer at the end of period $n - 1$, $I^m_n \in {0, 1, \cdots, \infty}$.</td>
</tr>
<tr>
<td>$I^r_n$</td>
<td>net inventory at the retailer at the end of period $n - 1$, $I^r_n \in {\infty, \cdots, \infty}$.</td>
</tr>
<tr>
<td>$t^m_m$</td>
<td>the relative position of period $n$ in capacity cycle, $t^m_m \in {1, \cdots, T_p, \cdots, 2T_p, \cdots, zT_p = T_m}$</td>
</tr>
<tr>
<td>$t^r_n$</td>
<td>the relative position of period $n$ in replenishment cycle, $t^r_n \in {1, \cdots, T_r}$</td>
</tr>
<tr>
<td>$K_n$</td>
<td>the capacity level in period $n$ (implied by $t^m_m$), $K_n \in {0, K_1, \cdots, K_z}$.</td>
</tr>
<tr>
<td>$S_T$</td>
<td>state under traditional system, $S_T = (I_m, I_r, t_m, t_r)$</td>
</tr>
<tr>
<td>$S_{NC}$</td>
<td>state under no-consignment vendor managed system, $S_{NC} = (I_m, I_r, t_m)$</td>
</tr>
<tr>
<td>$S_C$</td>
<td>state under consignment vendor managed system, $S_C = (I_r, t_m)$</td>
</tr>
</tbody>
</table>
3.1 Traditional System

In the traditional model, at the beginning of each period the manufacturer decides on how much to produce and/or to outsource. The manufacturer produces for the retailer in every $T_p$ periods, while the retailer places an order in every $T_r$ periods. $T_r$ is known by the manufacturer. We assume that fixed cost of transportation is zero under traditional and under vendor managed systems. We assume that the retailer places orders based on an $(R, nQ)$ type policy, where $R$ is the reorder point that guarantees a specified service level (Zheng and Chen 1992). Note that due to quantized shipments the analysis would not change under a fixed cost of transportation per batch. The sequence of events under traditional system is as follows:

1. At the beginning of a period, the manufacturer gives the decision of how many units to produce and/or to outsource, considering the allocated capacity (if allocated capacity is zero, there is no production). If an order is placed by the retailer in the last period of the replenishment cycle, a dispatch is made to the retailer in the first period of the following replenishment cycle. Production, outsourcing and dispatch lead times are negligible. Therefore the dispatched quantity is immediately ready at the retailer at the beginning of the replenishment cycle, before any demand is realized at the retailer.

2. Demand is realized at the retailer. If there is enough inventory in stock, the retailer fulfills the demand. If the retailer can not meet the demand completely, the unmet amount is backordered (at no explicit penalty). If it is the last period of the replenishment cycle, the retailer places an order at the manufacturer (if any), which is a non-negative integer multiple of $Q$. Otherwise, if it is not the last period, the retailer only passes the demand information to the manufacturer, and updates the inventory level.

3.1.1 Retailer’s problem under the traditional system

The problem of the retailer is to minimize the expected inventory level under a service level requirement (there is no explicit backorder cost for the retailer). We only consider the operating policies with $(R, nQ)$ structure. In the last period of the replenishment cycle, after the demand is realized, retailer places an order if the inventory level is equal to or less than the reorder point, $R$. The reorder point, $R$, is the decision variable and $Q$ is assumed to be a parameter.

First, consider the two measures for a given $R$ and $Q$: (i) Expected average inventory level ($\bar{I}$), and (ii) Average service level ($1 - \beta$).
The expected average inventory level is expressed as follows:

$$\bar{I} = \frac{1}{Q} \sum_{i=R+1}^{R+Q} \sum_{j=0}^{i} (i-j) \frac{P_1(j) + P_2(j) + \cdots + P_{T_r}(j)}{T_r}$$  \hspace{1cm} (1)$$

In (1), $P_1$ is the probability mass function of single period demand and $P_k$, $k \in \{1, \cdots, T_r\}$, is the $k$-convoluted probability mass function (i.e., probability mass function of $k$-period demand). Consider the replenishment cycle $T_r$. Under the quantized ordering policy, $(R, nQ)$, at the beginning of each cycle the inventory level at the retailer is $i$ with probability $\frac{1}{Q}$, where $i \in \{R + 1, \cdots, R + Q\}$. In the long-run, for the first period of the cycle, expected end-of-period inventory level is $\frac{1}{Q} \sum_{i=R+1}^{R+Q} \sum_{j=0}^{i} (i-j) P_1(j)$, Similarly, for the second period, expected end-of-period inventory level is $\frac{1}{Q} \sum_{i=R+1}^{R+Q} \sum_{j=0}^{i} (i-j) P_2(j)$, and so on. Since in the long-run, probability of being in any period in the replenishment cycle is equal to $\frac{1}{T_r}$, the time-averaged expected inventory level is expressed as in (1).

We define the average service level as $1 - \beta$, where $\beta$ is the expected average fraction of backordered demand per period. Let $\beta_i$, $i = 1, 2, \cdots, T_r$, denote the expected fraction of backordered demand in the $i^{th}$ period of the replenishment cycle. Then $\beta_i$ would be expressed as follows:

$$\beta_i = \sum_{I_i} P(I_i) \frac{E[(D_1 - I_i)^+]}{E[D_1]}$$

where $I_i$ is the beginning inventory level of the $i^{th}$ period, $P(I_i)$ is the probability that the beginning inventory level is $I_i$, and $D_1$ is the random variable denoting one-period demand. Then expected average fraction of backordered demand, $\beta$, is expressed as:

$$\beta = \frac{\beta_1 + \beta_2 + \cdots + \beta_{T_r}}{T_r}$$  \hspace{1cm} (2)$$

Equivalently, $\beta$ is expressed as follows:

$$\beta = \frac{1}{Q} \sum_{i=R+1}^{R+Q} \sum_{j=i+1}^{\infty} (j-i) \frac{P_{T_r}(j)}{T_r E[D_1]}.$$

We limit the operating policy of the retailer to the $(R, nQ)$ policy. Under this policy, to minimize the expected average inventory level in (1), the retailer simply chooses the minimum reorder point that guarantees the desired service level. However, as we analyze below, under quantized ordering $(R, nQ)$ type policy is not necessarily the optimal policy for the retailer. In other words, even if the optimal reorder point is chosen, expected inventory level may not be minimized. In Proposition 1 below, we identify the conditions under which the optimal policy is indeed an $(R, nQ)$ type policy for $T_r = 1$. We present the proofs in Appendix.

Each reorder point implies a service level $(1 - \beta)$, and an expected inventory level ($\bar{I}$). Let $S$ be the set of the $\beta$ values implied by all (integer and non-negative) reorder points (note that the elements of set $S$
vary with \( Q \). For \( \beta \in S \), let \( R(\beta) \) denote the reorder point that results in the service level of \( 1 - \beta \). (We assume there exists a unique \( R(\beta) \) for each \( \beta \in S \). Under \( T_r = 1 \) this is possible if \( R(\beta) + 1 \leq \max(D_1) \).

**Proposition 1** Suppose \( T_r = 1 \).

(i) For \( \beta \in S \), \((R(\beta), nQ)\) policy is the unique inventory level minimizing policy for the retailer.

(ii) For \( \beta \notin S \), there may exist more than one optimal ordering policy for the retailer, none of which is an \((R, nQ)\) policy.

Proposition 1 implies that for \( \beta \in S \) the only policy that achieves the minimum inventory level is \((R(\beta), nQ)\) policy. We use this result later in Section 4 when analyzing the manufacturer’s policy.

### 3.1.2 Manufacturer’s problem under the traditional system

We determine the optimal operating policy of the manufacturer under the traditional system. We model the manufacturer’s problem as a Markov Decision Process under average cost criteria as follows.

\[
g(s) = \min_{\delta} \lim_{N \to \infty} \frac{1}{N} E^\delta \left[ \sum_{n=1}^{N} r(s_n, a_n) \right]
\]  

(4)

where \( g(s) \) indicates the optimal average cost given that initial state is \( s \), \( \delta \) is any Markovian policy (note, the underlying chain is weakly communicating and under average cost criteria an optimal policy exists), \( s_n \) indicates the state in period \( n \), \( a_n \) indicates the action in period \( n \), and \( r(s_n, a_n) \) is the (immediate) cost of taking action \( a_n \) in state \( s_n \). We define the states under traditional model as, \( S_T = (I_m, I_r, t_m, t_r) \) where,

- \( I_m \) is the number of lots on-hand at the manufacturer at the end of the previous period, or at the beginning of the current period. Since capacity in every period is a non-negative integer multiple of \( Q \), without loss of optimality, \( I_m \) indicates a non-negative integer multiple of \( Q \), \( I_m = \{0, 1 \cdot \cdot \cdot , \infty \} \).

  - If \( I_m = 2 \) for instance, there exists \( 2Q \) units in inventory (see the discussion on action space).

- \( I_r \) is the net inventory at the retailer at the end of the previous period, \( I_r = \{-\infty, \cdot \cdot \cdot , \infty \} \).

- \( t_m \) denotes the relative position of a period in the capacity cycle, \( t_m \in \{1, \cdot \cdot \cdot , T_m\} \). We assume \( T_m \) implies the following capacity structure, \( (K_1, 0, 0, \cdot \cdot \cdot , 0, K_2, 0, 0, \cdot \cdot \cdot , 0, K_z, 0, 0, \cdot \cdot \cdot , 0) \)

- \( t_r \) denotes the relative position of a period in the replenishment cycle, \( t_r \in \{1, \cdot \cdot \cdot , T_r\} \).

In the traditional model, the action is defined only by the production quantity in period \( n \), \( p^n \). The quantity to be outsourced can already be inferred from the retailer’s order quantity at the end of the replenishment cycle. If the order quantity exceeds the amount in stock and the production capacity of
the manufacturer, then the remaining quantity should be outsourced. This implies outsourcing is not an independent decision. Note that outsourcing takes place only at the beginning of the replenishment cycle, since otherwise it will result in additional holding cost. The retailer orders in multiples of \( Q \), and capacity available is a multiple of \( Q \), therefore without loss of optimality, we limit the production quantity in every period to multiples of \( Q \) (this implies \( I_m \) is a multiple of \( Q \)). The action space in a period is denoted by \( p^n \in \{0, 1, \cdots, K^n\} \), where each value corresponds to the multiple of \( Q \). We assume that single period demand is characterized by a discrete probability distribution.

Next, we define the components of Equation (4) under the traditional model. We define \( r(s, a) \), where \( s = (I^n_m, I^n_r, t^n_m, t^n_r) \) and \( a = (p^n) \), as follows:

\[
    r(s, a) = \begin{cases} 
        cp^n + h(I^n_m + p^n - L)^+ + w(-I^n_m - p^n + L)^+ \Big)Q & \text{if } t^n_r = 1 \\
        (cp^n + h(I^n_m + p^n))Q & \text{if } t^n_r \neq 1 
    \end{cases}
\]

where \( L \) denotes the number of lots requested by the retailer at the end of replenishment cycle, i.e., at the end of period \( T_r \). The amount requested is dispatched by the manufacturer in the first period of the replenishment cycle, and is ready at the retailer before the demand is realized. Note that, the quantity \( L \) is deterministic and can be inferred from \( I_r \).

Transition probability \( P(j|s, a) \) denotes the probability that next state is \( j \) given current state is \( s \) and action taken is \( a \), where \( j = (I^{n+1}_m, I^{n+1}_r, t^{n+1}_m, t^{n+1}_r) \). We categorize all possible transitions under the traditional system as follows:

For \( t^n_r \neq 1 \),

\[
    P(j|s, a) = \begin{cases} 
        P_1(I^n_r - I^{n+1}_r) & \text{if } I^{n+1}_m = I^n_m + p^n, \\
        0, & \text{otherwise} 
    \end{cases}
\]

where \( 1_{\{t_m = T_m\}} \) takes value of 1 for the last period of the capacity cycle. \( P_1(I^n_r - I^{n+1}_r) \) is the probability that single period demand is \( I^n_r - I^{n+1}_r \).

For \( t^n_r = 1 \) there are two possibilities. The retailer’s order quantity does not exceed the available stock and production quantity, and therefore outsourcing is not necessary. When this is the case, \( I^{n+1}_m + p^n - I^{n+1}_m = L \). Otherwise, if outsourcing is necessary, then \( I^{n+1}_m = 0 \). We present the transition probability as follows.

\[
    P(j|s, a) = \begin{cases} 
        P_1(I^n_r - I^{n+1}_r + LQ), & \text{if } I^{n+1}_m = (I^n_m + p^n - L)^+, \\
        0, & \text{otherwise} 
    \end{cases}
\]
3.2 Vendor Managed System

Under the vendor managed system, we focus only on the manufacturer’s problem since the retailer does not make any decisions. Retailer only requires her performance measures to be as good as those under the traditional system. At the beginning of the production cycle the manufacturer decides on how much to produce, and in every period how much to outsource and to dispatch. The dispatched quantity immediately arrives at the retailer, i.e., lead time of transportation is zero. Note that due to the agreement there does not exist a replenishment cycle. The sequence of events is as follows:

1. At the beginning of a period, the manufacturer gives the decision of how many units to produce (if possible), to outsource and to dispatch. Inventory status of the manufacturer and the retailer are updated based on the dispatch quantity.

2. The demand is realized at the retailer’s site. Inventory status of the retailer is updated and end-of-period holding costs at the manufacturer and at the retailer are incurred.

We analyze the vendor managed setting under two cases; no-consignment stock and consignment stock.

3.2.1 No-consignment stock

Under no-consignment stock model (VM-NC) the ownership of the stock is transferred to the retailer once the dispatch arrives at the retailer. To be compatible with the traditional system, we assume that under the vendor managed system the retailer requires the average inventory investment to be as low as, and average service level to be as high as those levels under the traditional system. In other words, the retailer is indifferent between the traditional and the vendor managed system.

We determine the manufacturer’s optimal operating policy under the no-consignment system. We model the manufacturer’s problem as a Markov Decision Process under the average-cost criteria as follows:

\[
g(s) = \min_{\delta} \lim_{N \to \infty} \frac{1}{N} E_s \left[ \sum_{n=1}^{N} r(s_n, a_n) \right]
\]

s.t. Average Inventory Level at Retailer \( \leq \bar{I} \) \hspace{1cm} (7)

Service Level at Retailer \( \geq 1 - \beta \) \hspace{1cm} (8)

where \( \beta \) is defined as in (3), and \( \bar{I} \) as in (1). The constraint on inventory level in (7) reflects the case where the retailer is not willing to pay for inventory investment more than what it pays under the traditional system. In practice, the retailer may require that at least one of the performance measures described by Equation (7) or (8) are improved as a result of the agreement. Hence the service level specified by \( (1 - \beta) \)
can be regarded as a lower bound, and similarly the limit specified by $I$ on the average inventory level under vendor managed system can be regarded as an upper bound.

In all our analysis, right-hand-side (RHS) of Equation (7) or (8) is used as is, so that we have comparable cases. Note that it is through these constraints that the availability is ensured at the retailer at the right level of inventory. If instead, the retailer were to operate with min-max bounds on inventory, compared to the traditional system the retailer’s either the service level would be lower or average inventory level would be higher or both. Furthermore, the manufacturer’s benefits would decrease due to decreased operational flexibility.

The state is $S_{NC} = (I_m, I_r, t_m)$ where $I_m$, $I_r$, and $t_m$ are as defined in Table 1. At the beginning of the production cycle the manufacturer decides on how much to produce, $p^n$, and in every period how much to outsource, $y^n$, and to dispatch, $d^n$. The manufacturer produces, outsources and dispatches in multiples of $Q$, and capacity available at the beginning of the production cycle is a multiple of $Q$. The action space is denoted as, $p^n \in \{0, 1, \cdots, K^n\}$ and $d^n \in \{0, 1, \cdots, \infty\}$. Without loss of optimality, we limit the action space to multiples of $Q$. Note that the outsourced quantity at period $n$, $y^n$, is defined by $(d^n - I^n_m - p^n)^+$, and is not a(n independent) decision variable.

Next, we define the components of Equations (6) through (8). In (6) we define $r(s, a)$, where $s = (I^n_m, I^n_r, t^n_m)$ and $a = (p^n, d^n)$, as follows:

$$r(s, a) = \left(c_p^n + h(I^n_m + p^n - d^n)^+ + w(-I^n_m - p^n + d^n)^+\right)Q$$

(9)

Note that $L$ in Equation (5) is now a decision variable and is denoted with $d^n$. We define transition probabilities, $P(j|s, a)$ where $j = (I^n_{m+1}, I^n_{r+1}, t^n_{m+1})$, as follows:

$$P(j|s, a) = \begin{cases} P_1(I^n_r - I^{n+1}_r + d^n Q), & \text{if } I^{n+1}_m = (I^n_m + p^n - d^n)^+ \\
0, & \text{otherwise}
\end{cases}$$

Left-hand-side (LHS) of (7) reflects the expected inventory level per period at the retailer under the manufacturer’s optimal operating policy and is expressed as:

$$\sum_{I_m, t_m} \sum_{i > 0} i \pi_{NC}^M(I_m, I_r = i, t_m).$$

where $\pi_{NC}^M(I_m, I_r = i, t_m)$ is the fraction of time spent (or the steady-state probability) in state $(I_m, I_r = i, t_m)$ under the manufacturer’s optimal operating policy under no-consignment system. LHS of (8) reflects the average service level at the retailer under the manufacturer’s optimal operating policy:

$$1 - \sum_{I_m, t_m} \sum_{d(i)i+d(i)>0} \pi_{NC}^M(I_m, I_r = i, t_m) \frac{E[(D_1 - i - d(i))^+]}{E[D_1]} - \sum_{I_m, t_m} \sum_{d(i)i+d(i)\leq0} \pi_{NC}^M(I_m, I_r = i, t_m) \frac{E[D_1]}{E[D_1]}.$$
In (10) \( d(i) \) denotes the set of possible dispatch actions that can be taken at state \( i \). In the expression note that expected backordered demand is calculated differently if \( i + d(i) \leq 0 \). When \( i + d(i) > 0 \), the amount of available stock at the retailer before the demand is realized is positive. Then the expected backordered demand is \( E[(D_1 - i - d(i))^+] \). On the other hand, if \( i + d(i) \leq 0 \), then all demand occurred in that period should be backordered and expected backordered demand is \( E[D_1] \). For those periods service level is \( 1 - \frac{E[D_1]}{E[D_1]} = 0 \). Averaging over all periods gives the expression in (10).

Finally, we note that if under optimal dispatch policy the available stock at the retailer before the demand is realized is always positive, then service level is always positive in all of the periods. When this is the case, the service level expression in (10) can be replaced with the following expression:

\[
\sum_{I_m, t_m} \sum_{i < 0} \frac{|i|}{E[D_1]} \pi_{NC}(I_m, I_r = i, t_m).
\]  

(11)

### 3.2.2 Consignment Stock

In the consignment stock system (VM-C) the sequence of events is the same with no-consignment system except that the manufacturer owns and manages the inventory at the retailer’s site. We determine the manufacturer’s optimal operating policy under the consignment system. We model the manufacturer’s problem as a Markov Decision Process under average cost criteria as follows:

\[
g(s) = \min_{\delta} \lim_{N \to \infty} \frac{1}{N} E_s^\delta \left[ \sum_{n=1}^{N} r(s_n, a_n) \right]
\]

s.t. Service Level at Retailer \( \geq 1 - \beta \)

(12)

(13)

Note that since the stocking cost is incurred by the manufacturer there does not exist any constraint on the average inventory level. Furthermore, as we describe below, the reward function, \( r(s, a) \), now includes the holding cost at both the manufacturer and the retailer. Observe that Equation (13) is same as Equation (8).

In the consignment stock model, we assume that unit holding cost is the same at the manufacturer’s and the retailer’s site. The carrying charge of the inventory at a site is determined by the opportunity cost and risk level at the site. Since stocks at both echelons belong to the same firm (manufacturer), the opportunity costs of the tied up capital that could be used in some other investment is the same at both sites. Furthermore, the risk levels at both sites are the same, since the manufacturer has a single retailer. If there were multiple retailers, the manufacturer would prefer to keep stock at the upper echelon to minimize the risks and send the items to lower echelon only when necessary. Due to increased risks, the implied unit holding cost at the lower echelon would be higher. However, in this single retailer setting
keeping the items at the lower echelon rather than at both echelons does not affect the inventory holding cost while improving the service level.

Since unit holding cost is the same at manufacturer’s and retailer’s site, the manufacturer keeps inventory only at the retailer’s site and as a result immediately dispatches whatever it produces and outsources to the retailer’s site. Under consignment stock the state is defined as \( S_C = (I_r, t_m) \) where \( I_r \) and \( t_m \) are defined as before, and the actions are only how much to dispatch at the beginning of period \( n \), \( d^n \in \{0, 1, \cdots, \infty\} \), where each \( d^n \) value corresponds to the multiple of \( Q \).

We define \( r(s, a) \) where \( s = (I_r^n, t_m^n) \) and \( a = (d^n) \), as follows:

\[
 r(s, a) = \left( c. \min\{d^n, K^n\} + w. \max\{d^n - K^n, 0\}\right)Q + hE[(I_r^n + d^nQ - D_1)^+]. \tag{14}
\]

Note that in \( r(s, a) \) the holding cost at the manufacturer’s site is not expressed, since stock is kept only at the retailer’s site. The transition probabilities are expressed as follows:

\[
P(j|s, a) = \begin{cases} 
  P_1(I_r^n - I_r^{n+1} + d^nQ), & \text{if } t_{m+1}^n = t_m^n(1 - 1_{\{t_m = T_m\}}) + 1 \\
  0, & \text{otherwise}
\end{cases}
\]

The consignment and no-consignment models are different but related. Note that in the Markov Decision Process the reward functions and the constraints are different (see Equations (6)-(9) and (12)-(14)). However, the two systems are related in that there are parameter settings under which the actions taken under both systems are the same. Note that Equation (7) in the no-consignment model implies a unit holding cost. If the implied holding cost is equal to the manufacturer’s holding cost, \( h \), then consignment and no-consignment system can be regarded as equivalent in terms of the actions taken. For tighter or more relaxed inventory restrictions consignment and no-consignment systems are expected to result in different operating policies.

4 Analysis and comparison of traditional and vendor managed systems

In this part, we first provide an analysis on the structural properties of the optimal policy under traditional and no-consignment systems. Then we compare the cost under no-consignment system with the cost under traditional system and the cost under consignment system. In the remainder of the text, we denote traditional system with TRAD, no-consignment vendor managed system with VM-NC, and consignment system with VM-C.

4.1 Structural properties of the optimal policy

We analyze the structural properties of the optimal policy under traditional and under no-consignment systems. We show in Property 1 that the optimal policy under the traditional system is a modified
base-stock policy. For the no-consignment system, we discuss how the retailer’s optimal policy and the resultant inventory level and service level constraints affect the manufacturer’s optimal policy. In Property 2 we show that under certain conditions, the optimal policy under no-consignment system is also a modified base-stock policy.

**Property 1** Optimal policy of the manufacturer under traditional system is a modified base-stock policy.

Next, we discuss optimal operating policy of the manufacturer under VM-NC. Under VM-NC the manufacturer decides on how much to produce, outsource and dispatch to the retailer’s site. The dispatch policy is subject to the following two constraints: (i) Expected inventory level at the retailer can not exceed a certain level (as expressed in (7)), and (ii) Service level at the retailer should satisfy a minimum level (as expressed in (8)). These constraints make it difficult to characterize the optimal operating policy of the manufacturer. However, as we show in Property 2 under certain conditions the optimal policy of the manufacturer has the rather simple base-stock structure.

To define the manufacturer’s policy under VM-NC, we should focus on the retailer’s operating policy under the traditional system. In Proposition 1 we show that when $T_r = 1$, for $\beta \in S$ there exists a unique optimal policy which is $(R(\beta), nQ)$. This result leads to the following observation.

**Observation 1** For $T_r = 1$ and $\beta \in S$:

(i) Under VM-NC the manufacturer’s optimal operating policy is defined by a unique dispatch policy. This unique dispatch policy is the same policy as the retailer’s order policy under the traditional system, which is $(R(\beta), nQ)$.

(ii) Manufacturer’s optimal operating policy under VM-NC is independent of $\beta$.

Observation 1 states that for $\beta \in S$, under optimality the only possible dispatch policy of the manufacturer that satisfies constraints (7) and (8) is the retailer’s $(R(\beta), nQ)$ policy. In other words, in every period the manufacturer dispatches the minimum amount (in multiples of $Q$) to bring the retailer’s stock level above $R(\beta)$. Furthermore, the dispatch policy is the same for all $\beta \in S$. The reason is, discrete reorder points define $\beta$ and the value of the reorder point does not have an impact on the dispatch policy of the manufacturer (This structure resembles the one in a base-stock system where the order-up-to point does not affect the quantity ordered every period). For $\beta \notin S$, multiple dispatch policies may satisfy the constraints (7) and (8). Under optimality the manufacturer may select one of the eligible dispatch policies. Using Observation 1, in Property 2 we provide a characterization of the optimal policy of the manufacturer.

**Property 2** For $T_r = 1$ and $\beta \in S$, under VM-NC the manufacturer’s optimal policy is a modified base-stock policy.
4.2 Comparison of traditional and vendor managed systems

In this section we make two comparisons. First, we compare the no-consignment system with the traditional system. Using the structural results of the previous subsection, we show that the cost under the no-consignment system is always lower than or equal to the cost under the traditional system (Property 3). We then compare the no-consignment system with the consignment system. In Proposition 4 we show that under certain settings and under certain sufficient conditions the cost of the consignment system is lower than the cost of the no-consignment system.

4.2.1 Comparison of no-consignment and traditional systems

**Property 3** The cost of the manufacturer under VM-NC is always lower than or equal to the cost under TRAD.

Property 3 states that if under vendor managed system stock is not consigned, then vendor managed system results in lower cost than the traditional system, i.e., VM-NC is a no-risk case for the manufacturer.

4.2.2 Comparison of no-consignment and consignment systems

Although VM-NC is a no-risk case, the cost under VM-NC is not always lower than the cost under VM-C. As we show in the analysis below, under vendor managed system consigning the stock may be less costly than not consigning it. In the following, we introduce a specific instance. For this instance, we first obtain a lower bound on the optimal cost under VM-NC (Proposition 2), and an upper bound on the optimal cost under VM-C (Proposition 3). We then identify a set of sufficient conditions that make VM-C less costly than VM-NC (Proposition 4).

Assume that $Q = 1$ and that the (single-period) demand, $D$, has the following probability distribution:

$$P(D) = \begin{cases} 
1/2, & \text{if } D = \mu - 1 \\
1/2, & \text{if } D = \mu + 1 
\end{cases}$$

Assume that capacity per period is $E[D] = \mu$, and $T_r = 1$, i.e., under the traditional system retailer places orders in every period. In the analysis below, we focus on the cases where $\beta \in S$. In this setting since $Q = 1$, under the traditional system the retailer operates under the base-stock policy.

**Proposition 2** For the instance defined above, $LB(VM-NC) = \sqrt{(w - c)h} - \frac{h}{2} + c\mu$ is a lower bound on the manufacturer’s optimal cost under VM-NC.

In the following, we determine an upper bound on the optimal cost of the manufacturer under VM-C (Proposition 3). Under VM-C the manufacturer dispatches whatever he produces and outsources, and
the problem under consideration is how much to produce and outsource every period where the decisions are subject to the service level constraint. Below we propose two upper bounds on the optimal cost under VM-C.

**Proposition 3** For the instance under consideration,

(i) Suppose \( w \geq 3h + c \), and \( w \) is such that \( \sqrt{\frac{w-c}{h}} + 1 \in \mathbb{Z}^+ \). Then \( UB(VM-C) = h\left(\sqrt{\frac{w-c}{h}} + 1 - \frac{1}{2}\right) + c\mu \) is an upper bound on the optimal average cost under VM-C.

(ii) Suppose for \( k \geq 1 \) and \( k \in \mathbb{Z}^+ \), \( 1 - \beta \leq 1 - \frac{k^2 - k + 1}{2\mu\sqrt{\frac{w-c}{h} + (k^2 - k + 1)}} \), \( w \geq (5k + 3)h + c \) and \( w \) is such that \( \sqrt{\frac{w-c}{h}} + (k^2 - k + 1) \in \mathbb{Z}^+ \). Then \( UB(VM-C) = h\left(\sqrt{\frac{w-c}{h}} + (k^2 - k + 1) - (k + \frac{1}{2})\right) + c\mu \) is an upper bound on the optimal average cost under VM-C.

Proposition 3 suggests two upper bounds for the optimal cost under VM-C. Part(i) implies more relaxed sufficient conditions for the upper bound, and does not require any condition on the service level. When the service level is as high as 100%, the upper bound in part(i) is applicable. The upper bound in part(ii) requires tighter sufficient conditions, and in return gives a tighter upper bound. Note that \( UB(VM-C) \) in part (ii) is decreasing in the parameter \( k \). Parameter \( k \) denotes how low inventory level can be set at the retailer. As the service level requirement is lower (i.e., as \( \beta \) gets higher) \( k \) increases, and \( UB(VM-C) \) decreases.

Using Proposition 2 and Proposition 3, in Proposition 4 we present the main result of this subsection.

**Proposition 4** Suppose the following conditions are satisfied:

(i) \( \beta \geq \frac{k^2 - k + 1}{2\mu\sqrt{\frac{w-c}{h} + (k^2 - k + 1)}} \),

(ii) \( w > (5k + 3)h + c \),

where \( k \geq 2 \) and \( \sqrt{\frac{w-c}{h}} + (k^2 - k + 1) \in \mathbb{Z}^+ \). Then the optimal cost under VM-C is lower than the optimal cost under VM-NC.

Proposition 4 compares the no-consignment and consignment models under a deterministic reorder point at the retailer. In practice, firms prefer a fixed operating policy rather than a randomized one due to operational difficulties, even if a randomized policy may yield lower costs. The first condition in the proposition states that if service level requirement at the retailer is not high, then consignment stock is preferred. This result is in line with our experimental study where we observed that under 99% service level consignment stock is never preferred (see Section 5.1). The intuition behind this result is as follows. If the service level requirement of the retailer is low then this implies the expected inventory level requirement at the retailer is also low (i.e., RHS of the constraint in (7)). This corresponds to a high “implied unit holding cost” for the stock at the retailer’s site. If the implied cost is very high (i.e., if
expected inventory level is very low) then the manufacturer simply prefers owning the stock rather than trying to meet the requirement under no-consignment. In practice, for items with low implicit stock-out costs, the retailer may allow low service levels. Examples are the items for which the retailer also carry the substitutes, or products that are not competitive. For these items inventory requirement imposed by the retailer to the manufacturer would be low, and the manufacturer might prefer consigning the stock to no-consignment. The second condition states that if outsourcing cost is high, then consigning the stock is preferred. This result also supports our observations from the computational study. Under high outsourcing cost the manufacturer would prefer to keep high levels of inventory, which is allowed under the consignment stock model but not under no-consignment model.

Note that our construction assumes $T_r = 1$ and $\beta \in S$. Under these assumptions the cost and operating policies under TRAD and under VM-NC are the same. Therefore the intuition obtained from Proposition 4 could be extended to the comparison of the consignment system with the traditional system. We conclude that the manufacturer prefers VM-C to TRAD when the inventory level constraint is “tight”, i.e., when the operating policy of the retailer imposes an “inflexible” dispatch policy for the manufacturer.

5 Computational Analysis

We conduct experiments to analyze how the system parameters affect the manufacturer’s savings under the vendor managed system and identify the conditions under which manufacturer is willing to make an agreement. In designing the experiments we keep unit holding cost, unit production cost, expected demand per period as constant at $h = 1$, $c = 10$, and $E[D_1] = 20$. We assume the lot size is, $Q = 5$. We assume that the capacity cycle is two periods $T_m = 2$, production cycle is one period, $T_p = 1$, and replenishment cycle, $T_r$, can be one or two periods. Capacity levels in the capacity cycle are indicated with $K_1$ and $K_2$. We consider the effect of the following parameters on average cost per period:

1. **Total capacity.** We assumed the capacity levels are “tight”, “medium”, or “excessive”. Under tight capacity $\bar{K} = \frac{K_1 + K_2}{2} = E[D_1] = 20$, under medium capacity $\bar{K} = 25$, and under excessive capacity $\bar{K} = 30$.

2. **Outsourcing cost.** $w = 11$, 15, 20, and 30.

3. **Capacity non-stationarity.** $(K_1, K_2) =$ (40,0), (30,10), (20,20), (10,30), and (0,40).

4. **Replenishment cycle, $T_r = 1$, 2.** When $T_r = 2$, under the traditional system the retailer places orders in every two periods, whereas shares the demand and inventory level information in every period. Comparing the traditional system under $T_r = 2$ with the vendor managed system,
the manufacturer has a gain in terms of both dispatch quantity and dispatch time (i.e., dispatch frequency) flexibility. In §5.1.4 we quantify the benefit of flexibility.

5. **Demand coefficient of variation.** The demand faced by the retailer is modeled via a discrete distribution. The distributions considered and the corresponding values of the coefficient of variation, $cv$, are as follows: Uniform $[11, 29]$ ($cv = 0.28$), Truncated Normal ($\mu = 20, \sigma = 30$) ($cv = 0.57$), Beta ($0.3, 0.3$) ($cv = 0.80$).

6. **Service level at the retailer.** We assumed service levels are 90%, 95%, 99%. When determining the service level at the retailer, we only consider discrete reorder points, and we set the reorder point such that the service level is higher than 90% (or 95%, or 99%). For example, for uniform distributed demand, when $T_r = 1$ the retailer’s reorder point that gives a service level of at least 90% is $R = 18$ and the service level implied by this reorder point is 90.26%. We present the service levels in Table 2.

<table>
<thead>
<tr>
<th>T=1</th>
<th>T=2</th>
</tr>
</thead>
<tbody>
<tr>
<td>Uniform</td>
<td>Normal</td>
</tr>
<tr>
<td>90%</td>
<td>90.26%</td>
</tr>
<tr>
<td>95%</td>
<td>95.79%</td>
</tr>
<tr>
<td>99%</td>
<td>99.47%</td>
</tr>
</tbody>
</table>

In this section we observe the effect of system parameters on the benefits of vendor managed system. We have already shown that no-consignment system is a no-risk case for the manufacturer, and for $T_r = 1$, cost under TRAD and VM-NC are the same, so when making the observations we only compare TRAD and VM-C, unless otherwise stated. However, for certain cases, when we believe that comparison with VM-NC provides further insights we explicitly state this in the discussions.

In the following subsections, we present our results under two main titles: Analysis under Stationary Capacity and Analysis under Non-Stationary Capacity. We use linear programming (LP) model to solve (4) and (12) and use average cost per period criteria for the analysis. When LP is used to solve the corresponding MDP problem under VM-C we obtain at most a single randomized action as we have one additional constraint (corresponding to service level). We use the results as is. Number of variables of the LP model (cartesian space of states and actions of the MDP) are 55,000 for the traditional system, and 4,060 for the VM-C system. Total number of experiments carried out for the stationary capacity case is 216, and for the non-stationary case 288. Due to computational burden that Equation (10) brings, we use the expression in (11) as a surrogate of the service level at the retailer. Note that (10) and (11)
are equivalent if retailer’s beginning inventory level in every period is non-negative. In our experimental setting under vendor managed system we expect this to be the case, and believe that using (11) instead of (10) has a negligible effect in the results.

5.1 Analysis under Stationary Capacity

We first make a comparison of the vendor managed system with the traditional system under stationary capacity, when capacity per period is tight ($\bar{K} = 20$), medium ($\bar{K} = 25$), and excessive ($\bar{K} = 30$). Under stationary capacity, the capacity level over the periods is constant: $(K_1, K_2) = (20, 20), (25, 25)$, and $(30, 30)$. Based on the insights obtained in this section, we extend the analysis to the non-stationary capacity.

5.1.1 Effect of unit outsourcing cost on savings

We analyze the effect of outsourcing cost ($w$) on % savings under the vendor managed system

\[
\left( = \frac{\text{TRADCost} - \text{V.M.Scost}}{\text{TRADCost}} \right) \times 100
\]

for service levels 90%, 95%, and 99% (We refer to the service levels in Table 2, while we indicate the levels with 90%, 95%, and 99%). As unit outsourcing cost increases, under both traditional and vendor managed system, the average inventory level at the manufacturer increases while number of units outsourced decreases. However, although the number of outsourced units decreases, the total outsourcing cost increases under both systems. We observe that the increase in inventory level and decrease in number of outsourced units is more drastic under the traditional system compared to the vendor managed system. As a result of this, the number of units produced in-house increases significantly under traditional system. We conclude that, the traditional system is less robust to the changes in unit outsourcing cost compared to the vendor managed system. As a result of the changes the production cost, total outsourcing cost and inventory holding cost, the cost under TRAD increases at a steeper rate than the cost under VM-C. We observe that the savings under VM-C increase with unit outsourcing cost.

Experimental results for $T_r = 1$ support the conclusions derived in Proposition 4: we observe that when service level is 99% and demand coefficient of variance is high, cost under VM-C is always higher than the cost under TRAD or VM-NC. Otherwise, cost under VM-C can be lower especially if the outsourcing cost is high. For $T_r = 2$, the manufacturer keeps higher inventory compared to $T_r = 1$ under traditional system, and therefore VM-C can be more beneficial. Finally, as the demand coefficient of variance increases we observe that manufacturer’s savings under VM-NC increase while savings under VM-C decreases. We conclude that although higher coefficient of variation helps manufacturer to manage the operations more effectively, incurring the retailer’s inventory holding cost outweighs the savings.

In the overall setting, savings under VM-C can be as high as 5.37% (when outsourcing cost is high,
demand coefficient of variance is low, and service level requirement is low) and as low as \(-9.80\%\).

5.1.2 Effect of the vendor managed system on capacity utilization

We analyze how the capacity utilization change as system moves from traditional to vendor managed system. Capacity utilization is a measure of manufacturer’s ability to meet the demand through in-house production. The unmet demand is outsourced and in this respect outsourcing cost functions as a lost sales penalty, and capacity utilization reflects the service level provided by the manufacturer. Analysis indicates that capacity utilization is always higher under vendor managed system than the traditional system (see Figure 2). Also as unit outsourcing cost increase the capacity utilization increase. The capacity utilization increases at a higher rate under the traditional system as the unit outsourcing cost increase.

![Figure 2: Effect of outsourcing cost on utilization when total capacity is tight.](image)

We also analyze the effect of capacity level on the cost under TRAD and VM-C. An increase in capacity level from tight to medium or medium to excessive, decreases the cost under both traditional and vendor managed systems. We observe that how the two systems react to an increase in capacity level slightly differs with respect to the coefficient of variation in demand:

1. When the coefficient of variation of demand is low, under vendor managed system inventory burden on the manufacturer is low. The manufacturer already uses the capacity effectively, and therefore the benefit of additional capacity is relatively low. Beyond a certain threshold, the increase in capacity does not decrease the cost for vendor managed system. On the other hand, under traditional system, the additional capacity is more beneficial, since additional capacity will help the manufacturer to meet the retailer’s orders more effectively. For sufficiently high capacity, we expect that under traditional system in-house production will be equal to the demand, and no outsourcing cost or
holding cost will be incurred. This implies under sufficiently high capacity, cost under traditional system will be lower than the cost under vendor managed system, since under VM-C there always be the burden of inventory holding due to the retailer. Therefore when coefficient of variation is low, as the capacity level increases, benefit of vendor managed inventory decreases. Figure 3(a) shows the costs under two systems when coefficient of variation is low.

2. When coefficient of variation of demand is high, we observe that both vendor managed system and traditional system benefit an increase in the capacity level. When capacity level is excessively high, both systems reach to a stability in terms of cost and the cost does not decrease further with an increase in capacity level. We observe that under sufficiently high capacity the average cost under traditional system can be as low as the total in-house production cost (which is expressed as \( p.E[D] \) and which is the lowest level for the cost), whereas under VM-C the cost consists of the in-house production cost and the inventory cost at the retailer. We conclude that, under sufficiently high capacity, vendor managed system is not beneficial. In Figure 3(b) we show how the costs change with respect to the capacity level when coefficient of variation is high.

The amount of per period capacity necessary to attain 100% in-house production is higher under high coefficient of variation of demand compared to the case where coefficient of variation is low.

![Figure 3: Effect of capacity increase costs under traditional and vendor managed system](image)

5.1.3 Effect of vendor managed system on inventory levels

We compare the “expected total inventory level in the system (at the manufacturer and the retailer)” and the “expected inventory level at the retailer’s site” under traditional system and the vendor managed system. Experimental results show that the expected total inventory level in the system is lower under VM-C.
Property 4 \textit{Expected inventory level at the retailer’s site:}

(i) is higher under VM-C compared to the traditional system when $T_r = 1$,
(ii) may or may not be higher under VM-C compared to the traditional system when $T_r = 2$.

The inventory level at the retailer’s site may or may not be lower under the vendor managed system depending on how the retailer operates under the traditional system. If retailer already requires small and frequent replenishments under the traditional system (i.e., if $T_r = 1$), then inventory at retailer’s site increases under the vendor managed system. The reason is, under traditional system the retailer operates with the minimum inventory level for a given service level, since $(R, nQ)$ is the optimal operating policy (see Proposition 1). When the manufacturer manages the retailer’s inventory, due to capacity restrictions at the manufacturer, the vendor managed system corresponds to a constrained system compared to the traditional system. Therefore expected inventory level at the retailer is higher.

If under traditional system, the retailer places infrequent and lumpy orders (i.e., when $T_r = 2$), then under high outsourcing cost and low service levels the inventory level at retailer’s site is higher under VM-C. Under high service levels the inventory level at retailer’s site is lower under VM-C. We would like to note that the results may vary if we relax the assumption on same unit holding cost at both echelons under VM-C. In that case under both consignment and no-consignment stock systems, inventory would be held at both sites and the inventory kept at the retailer’s site would be lower. However, there will still be instances under low service levels and high outsourcing cost where the inventory level at retailer’s site is higher under VM-C. Under low outsourcing cost, we observe that inventory level at the retailer is always lower under VM-C compared to the inventory level under the traditional system.

5.1.4 \textbf{Quantifying dispatch time and dispatch quantity flexibility}

Under the vendor managed system the manufacturer decides on how much to dispatch in every period depending on the capacity level and demand. In some periods the manufacturer may not make any dispatches, other times may prefer more frequent dispatches. The vendor managed system implies a gain in “dispatch time” and “dispatch quantity” flexibility compared to the traditional system. In the following we quantify the benefits due to dispatch quantity flexibility and due to both dispatch time and dispatch quantity flexibility:

\textit{Measuring dispatch quantity flexibility.} To analyze benefits due to dispatch quantity flexibility only, we compare the following two settings: $T_r = 2$ under traditional system, and $T_r = 2$ under \textbf{no-consignment} vendor managed system. We restrict the dispatch time under vendor managed system to one dispatch in two periods. This implies compared to the traditional system there does not exist an increase in dispatch time flexibility under vendor managed system, but only an increase in dispatch quantity flexibility. (To obtain optimal cost of manufacturer under VM-NC with $T_r = 2$ restriction we use the constraint in (10)).
Measuring dispatch time and quantity flexibility. To analyze benefits under joint dispatch time and quantity flexibility, we relax the restriction on the dispatch time under vendor managed system, we simply compare the traditional system (with $T_r = 2$) with VM-NC (To obtain optimal cost of manufacturer under VM-NC with no dispatch time restriction we use the constraint in (10)).

Note that when quantifying the benefits of flexibility, we assume inventory is not consigned, to determine the benefits of flexibility.

$$\text{Percentage savings due to quantity and time flexibility}$$

![Figure 4: Effect of dispatch time and dispatch quantity flexibility on savings](image)

Figure 4 shows that savings due to dispatch quantity flexibility are high especially when unit outsourcing cost is high (for this analysis we assumed capacity is tight, $\tilde{K} = 20$, and demand has a simplified structure such that demand per period is either 15 or 25 each with probability $\frac{1}{2}$). Analysis show that, while dispatch quantity flexibility may or may not decrease the inventory level at the manufacturer, the additional flexibility due to dispatch frequency decreases the inventory level in the system and significantly increases the savings.

Finally, we note that under vendor managed system dispatch time flexibility not necessarily implies more frequent shipments. Analysis show that depending on capacity restrictions, under VM-C the manufacturer may prefer less frequent shipments compared to the traditional system which may result in lower inventory cost and lower total cost compared to traditional system.

### 5.2 Analysis under Non-Stationary Capacity

We now analyze how the costs differ under traditional and the vendor managed system as the capacity levels change throughout the periods. When replenishment cycle is one period, under both traditional and vendor managed system lowest cost is incurred when capacity is stationary at $(K_1, K_2) = (20, 20)$ (Figure 5.a). This is expected since the end-demand is stationary. In the two-period replenishment cycle, under traditional system as more capacity is allocated closer to the replenishment point (which is the first period of the replenishment cycle), total cost decreases. This is because the manufacturer can use the end-demand information available in the previous period and react accordingly in the replenishment
period (Figure 5.b).

![Graph](image)

**Figure 5:** Effect of capacity non-stationarity on savings

The analysis under non-stationary capacity provides insights on how a manufacturer should allocate the capacity under traditional system versus under the vendor managed system (of course if possible; i.e., if the manufacturer has the flexibility in shifting its capacity from one period to another). Under traditional system the manufacturer would schedule the orders from different customers so that bulk of production for a certain customer can be realized as the replenishment time for that customer approaches. This may result in erratic orders placed by the manufacturer to the upper echelons. On the other hand, under the vendor managed system the manufacturer prefers smoothing out the production and dispatch process by allocating the same capacity in every period. Allocation capacity uniformly would result in much less fluctuation in the orders placed by the manufacturer, therefore the vendor managed system would also benefit the players in the upper-echelons of the supply chain. Note that Lee, Padmanabhan and Whang (1997) specifies a similar manufacturing situation to show the bullwhip effect. In our case, we show that capacity management is a useful tool to reduce the bullwhip effect.

The analysis provides further insights on the type of settings that a manufacturer should or should not prefer the vendor managed system. If the orders from retailer are staggered, frequent orders with small lot sizes (in our case, when replenishment cycle is one period), then managing the retailer’s inventory would not bring much benefit to the manufacturer, and the cost of consigning the stock may outweigh the benefits (Figure 5.a).

On the other hand, if the retailer place the orders infrequently and in large lot sizes (in our case, when replenishment cycle is two periods) then the manufacturer may or may not benefit managing the retailer’s inventory depending on the flexibility in its operations, explained as follows. When the manufacturer does not have the flexibility in shifting the capacity, (if, for instance, the manufacturer has customers with strict delivery time requirements) experimental results indicate that savings are high (around 1.9% under consignment system and 11.6% under no-consignment system). The inflexible system is modeled with \((K_1, K_2) = (0, 40)\) under the vendor managed system and under the traditional system.
These figures imply that when the manufacturer does not have much control over allocating the capacity to respond to the orders effectively, dispatch quantity and dispatch time flexibility are most useful. However, if the manufacturer can shift the capacity to react to the demand patterns, then savings under the vendor managed system are rather limited. This is because it is already possible to operate the traditional system effectively. Analysis yield that on the average savings under consignment system are $-5.1\%$, and savings under no-consignment system are $4.8\%$ (here best performances are compared, i.e., averaged cost under $(40, 0)$ for the traditional system is compared with the cost under $(20, 20)$ for the vendor managed systems). Fully consigning inventory may not be a preferable option if manufacturer has sufficient flexibility in shifting its capacity.

## 5.3 Managerial Insights

In this section we present the highlights of our analysis. Several of the insights we obtained from the study support and build up on the previous findings in the literature, while others reveal new information.

**Insight 1** The benefits that the vendor managed system will bring to the manufacturer depends on the type of the vendor managed system relation. There may be benefits beyond sharing demand and inventory information. However, there are a number of cases where VM-C does not yield any additional savings over information sharing, as well. Hence, information sharing should be considered as a first step in the relation with a retailer before going into “risky” the vendor managed relations. However, the next level of relationship should not necessarily follow; requires careful evaluation of trade-offs.

Insight 1 complements the findings in the literature by assessing the benefit of vendor-managed system from the manufacturer’s perspective. Fry, Kapuscinski and Olsen (2001) quantify the effect of VMI on the system-wide cost given that information is already shared, and conclude that if certain parameters (such as dispatch quantity or penalty of violating inventory bounds) are not chosen properly, vendor managed setting can be more costly for the chain than the traditional setting. We show that if conditions or the terms of the agreement are tight for the manufacturer, then vendor-managed system does not bring additional benefit to information sharing.

**Insight 2** VM-NC constitutes a no-risk case for the manufacturer. However, there can be cases where VM-C is more profitable than the VM-NC for the manufacturer.

We explicitly compare the no-consignment stock system with the consignment stock system, and conclude that if the service level requirement at the retailer and the coefficient of variance for demand are low, and as a result inventory level has to be tightly kept at a low level, then consigning the stock might be
less costly for the manufacturer. In the literature either totally consigned stock or no-consignment stock models are studied and the question of whether the stock should be consigned is unaddressed.

**Insight 3** If capacity allocated is sufficiently high, then the manufacturer is less likely to benefit from the vendor managed system.

**Insight 4** Manufacturers usually have the practice of allocating capacity for a product or customer. It turns out that the way to allocate this level “optimally” is not very straightforward; whether the system operates with full information only, or under a certain type of the vendor managed system may lead to different capacity allocation schemes, significantly affecting the performance. If the manufacturer cannot easily change the capacity allocation, i.e., if operating in a rigid system, then it is most likely to benefit the vendor managed system.

We analyze the interaction of capacity management and the vendor managed system. Gavirneni, Kapuscinski and Tayur (1999) quantify the benefit of information sharing in a supply chain and show that as capacity level increases, the cost savings at manufacturer increases (with diminishing returns). We observe that the savings of the manufacturer due to vendor managed system decreases with the increase in capacity and there exist capacity levels where traditional system is less costly. We also analyze the impact of the operating strategy on the capacity allocation decisions. Under vendor-managed system, the manufacturer prefers smoothing the production decisions by allocating equal capacity in each period, whereas under traditional system if orders placed by the retailer are lumpy and intermittent, then capacity allocation is unbalanced and production amounts may fluctuate. Lee, Padmanabhan and Whang (1997) specify a similar manufacturing situation to show that lumpy orders increase the bullwhip effect. We complement the study by showing that vendor-managed system benefits the upper echelon through smooth production patterns. Our finding is also in line with Disney and Towill (2003). Our analysis on capacity management contributes to the literature by connecting the benefit of vendor managed system to capacity management decisions.

**Insight 5** The vendor managed system may or may not decrease the inventory level at the retailer.

**Insight 6** Main benefits of the vendor managed system can be described with “dispatch time” flexibility and “dispatch quantity” flexibility, with the former potentially leading to reduction in inventory investment in the system. Reduction in inventory investment may occur even if dispatches are less frequent.

Finally, we quantify the benefit of vendor managed system in terms of dispatch quantity flexibility and dispatch time flexibility. Cetinkaya and Lee (2000) contrasts a traditional system with frequent shipments
with a vendor-managed system where shipments are consolidated at the expense of increased inventory levels. Waller, Johnson and Davis (1999) shows through a simulation study that vendor-managed inventory may increase the frequency of dispatches to retailers which helps decrease the inventory level. We show that dispatch time flexibility may contribute significantly to the reduction of inventory in the system, and this is not necessarily achieved through increased frequency of the shipments.

6 Conclusions

In this study we analyze a vendor managed system for a supply chain consisting of a single manufacturer and a single retailer. We model the manufacturer effectively, so that benefits going under a vendor managed agreement can be analyzed. We assume that retailer demand information is fully available to the manufacturer and hence only study the benefits beyond information sharing. We study both the consignment stock and no-consignment stock systems under the vendor managed system. We consider the capacity limitation of the manufacturer in our analysis, which turned out to be a very important factor for the manufacturer and analyze the problem under different capacity allocation schemes to identify the effect of capacity management on benefits. Our main findings are the vendor managed system indeed brings benefit to the manufacturer beyond information sharing. The benefits are high especially under moderate or tight production capacity rather than excessive capacity, or under low service level requirements. Analyses indicate that if the inventory and service level requirements are too tight, rather than conforming with the specifications, owning the inventory might be less costly for the manufacturer.

Under the vendor managed system manufacturer can take a proactive approach in responding to retailer’s demand and thus can increase the capacity utilization. The vendor managed system provides the manufacturer with both dispatch time and dispatch quantity flexibility, and this flexibility is most valuable when under traditional system retailer requests irregular/large shipments rather than small and frequent shipments. We also compare the inventory levels under traditional and vendor managed systems. Total inventory level in the system is lower, however inventory level at the retailer may not be lower. We analyze the effect of end demand variability on the savings. Higher variability results in higher savings, but the savings are outweighed by the inventory holding costs under consignment stock model. Finally, we observe the savings under varying capacity allocation schemes. Under the vendor managed system the manufacturer prefers uniformly allocated capacity, thus helps reducing the bullwhip effect in the total chain, whereas under traditional system allocates most of the capacity towards the time of the replenishment. We conclude that consigning inventory under the vendor managed system may not be a preferable option if manufacturer has sufficient flexibility in allocating the resources. If manufacturer has limited or no flexibility, then the vendor managed system provides the highest benefit.
In the vendor managed system, we limit the analysis to the two extreme cases, totally consigned stock versus no-consignment stock. Between the two extreme cases, in general either the cost of stock at the retailer might be shared by the manufacturer and the retailer, or the ownership of inventory can be transferred from the manufacturer to the retailer after some time period between 0 and sales time. Future work includes analysis of a more general ownership model.

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7 Appendix

7.1 Proof of Proposition 1

Before starting the proof, we formally state the retailer’s problem of minimizing average inventory level as a Markov Decision Process under average-cost criteria and present the corresponding linear programming formulation as follows.

\[
\begin{align*}
\text{(P)} & \quad \min & \sum_{i > 0} i \pi_{i,a}^R \\
& \text{s.t.} & \sum_a \pi_{j,a}^R - \sum_{i,a} \pi_{i,a}^R P_{j|i,a} = 0 & \forall j \\
& & \sum_{i,a} \pi_{i,a}^R = 1 \\
& & \sum_{i < 0} |i| \pi_{i,a}^R \leq \beta \\
& & \pi_{i,a}^R \in \mathbb{R}^+ 
\end{align*}
\]

In (P),

- \( i \) is the end-of-period inventory level at the retailer
- \( \pi_{i,a}^R \) is steady-state probability that inventory level is \( i \) at the retailer
- \( a \) actions are (quantized) order quantities
- \( 1 - \beta \) is the required service level
- \( P_{j|i,a} \) is the transition probability of being in state \( j \) in the next period given that current state is \( i \) and action is \( a \).

Note that \( P_{j|i,a} = P_1(i + aQ - j) \), where \( P_1 \) is defined as in Table 1.

The objective function denotes the expected average end-of-period inventory level in steady-state. Equation (16) preserves the flow balance and Equation (17) ensures that sum of the steady-state probabilities do not exceed 1. Finally equation (18) ensures that the required service level is satisfied.
We now present the proof.

(i) We show that for $\beta \in S$ there is a unique solution to (P) which is an $(R, nQ)$ policy with reorder parameter $R(\beta)$.

Consider the lagrange relaxation of (P):

\[
L(\lambda) = \min \sum_{i>0} i\pi_{i,a} + \lambda \sum_{i<0} |i|\pi_{i,a} - \lambda \beta E[D_1]
\]
\[
s.t. \quad \sum_{a} \pi_{j,a}^R - \sum_{i,a} \pi_{i,a}^R P_{j|i,a} = 0 \quad \forall \ j
\]
\[
\sum_{i,a} \pi_{i,a}^R = 1
\]
\[
\pi_{i,a}^R \in \mathbb{R}^+
\]

where $\lambda > 0$ is the lagrange multiplier. Note that $L(\lambda)$ is equivalent to the periodic-review stochastic dynamic inventory problem with batch ordering where overage cost is linear with rate $h = 1$ and underage cost is linear with rate $\lambda$. For any $\lambda \in (0, \infty)$ optimal solution for $L(\lambda)$ is an $(R, nQ)$ policy (Veinott 1965). It is easy to see that $\exists \lambda$, namely $\lambda_\beta$, for which $(R(\beta), nQ)$ is the optimal inventory policy. Note that by definition, the value of the lagrangean relaxation problem (for any $\lambda$) is a lower bound on the optimal value of the main problem (P). Since the optimal solution of $L(\lambda_\beta)$ makes (18) an equality and is a feasible solution for (P), we conclude that $\lambda_\beta$ is the optimal lagrange multiplier for $L(\lambda)$ and $(R(\beta), nQ)$ is the optimal solution for (P). (Note that for $L(\lambda)$, $\lambda_\beta$ may not be unique.) Furthermore, $(R(\beta), nQ)$ is unique, since any other policy (i.e., if reorder point is $R(\beta) + 1$ or $R(\beta) - 1$) would either violate the service level constraint in (18) or increase the objective function value in (15).

(ii) Through an example, we show that for $\beta \notin S$, there may be more than one solution to (P).

Example 1. Let $Q = 1$. Under this assumption $(R, nQ)$ policy is a base-stock policy with order-up-to level $R + 1$. For some $\beta \notin S$, consider the optimal ordering policy. Suppose this policy could be a randomized or a deterministic policy. A policy simply states a set of order-up-to levels and the percentage of the time each order-up-to level is reached. Therefore any policy can be expressed as a convex combination of several order-up-to levels.

Consider the following instance. Let (single-period) demand take values 1, 2, or 3 with probability $1/3$. The optimal policy under $\beta = 0.1$ is a randomized policy with the following order-up-to levels: at states (i.e., beginning inventory level) -1 and 1 order up to 2; at state 0, 1/5 of the time order up to 2, 4/5 of the time order up to 3; at state 2, order up to 3. Steady-state probabilities under optimal policy are $\pi_{-1} = \frac{6}{30}$, $\pi_0 = \pi_1 = \frac{10}{30}$, $\pi_2 = \frac{4}{30}$. Therefore the optimal policy implies the following: 3/5 of the time order up to 2, 2/5 of the time order up to 3. There may be other policies that corresponds to the same scheme. For example, consider the policy: “At all states 3/5 of the time order up to 2, and 2/5 of the time order up to 3”. This policy also yields $\beta = 0.1$ and minimizes the inventory level. We conclude
that when $\beta \not\in S$, there may be multiple optimal policies which are randomized. Note that none of these optimal policies can be a non-randomized $(R, nQ)$ policy since in that case $\beta \in S$. □

7.2 Proof of Property 1

Under the traditional system, the retailer places orders to the manufacturer at the end of each replenishment cycle. At the beginning of every period (before dispatch) the manufacturer observes the retailer’s inventory level, $I_r$. If it is the first period of the replenishment cycle, the manufacturer dispatches the quantity required by the retailer, that is, dispatches the minimum amount that will bring the inventory level at the retailer above $R$. The dispatch quantity is simply the “demand” of the manufacturer for the first period of the replenishment cycle, which is “known” due to $I_r$. For the other periods “demand” is zero. Since the dispatch quantity is already implied by $I_r$ and there does not exist any uncertainty. This system is equivalent to a production-inventory system where the manufacturer’s demand is known with certainty in the current period. If total production quantity plus the available stock is not sufficient to meet the dispatch, the remaining amount is outsourced. Outsourcing is simply equivalent to a “lost sales” structure where per unit lost sales cost is $w$. Therefore manufacturer’s system is a periodic review single-echelon capacitated production-inventory system with markov-modulated demand, periodically changing capacity levels, and linear overage and “lost sales” costs. The optimal policy of the manufacturer is a modified base-stock policy (see Aviv and Federgruen 1997; Kapuscinski and Tayur 1998; Gavirneni, Kapuscinski and Tayur 1999). □

7.3 Proof of Property 2

To show why Property 2 holds, we argue that the manufacturer’s problem under VM-NC is a single-echelon capacitated production-inventory model with markov-modulated demand and periodically changing capacities. Under VM-NC the manufacturer determines the optimal production, outsourcing and dispatch policy. Based on Observation 1 the manufacturer’s dispatch policy is already determined as $(R(\beta), nQ)$. Specifically, the manufacturer observes the retailer’s end-of-period inventory level and dispatches exactly the minimum amount that brings the inventory level above $R(\beta)$ before demand in the current period is realized at the retailer. The dispatch quantity is simply the “demand” of the manufacturer. The dispatch quantity is inferred from retailer’s inventory level, $I_r$, and therefore does not involve any uncertainty. Similar to the traditional system, this system is equivalent to a production-inventory system where the manufacturer’s demand is known with certainty in the current period. Therefore manufacturer’s system is a periodic review single-echelon capacitated production-inventory system with markov-modulated demand, periodically changing capacity levels, and linear overage and “lost sales” costs. The optimal policy of the manufacturer is a modified base-stock policy.
Finally note that, for $\beta \not\in S$ the manufacturer’s policy is not necessarily a modified base-stock policy since there may exist more than one dispatch policy that result in the same $\beta$ and $\bar{I}(\beta)$. Manufacturer may prefer any of the dispatch policies to minimize the cost and the operating policy can be randomized.

7.4 Proof of Property 3

We make the comparison under two cases: (i) For $\beta \in S$ and $T_r = 1$, and (ii) For $\beta \not\in S$ or $T_r \neq 1$. For $\beta \in S$ and $T_r = 1$, Property 2 shows that the dispatch policy of the manufacturer can be characterized as $(R(\beta), nQ)$. Note that the dispatch policy of the manufacturer under TRAD is also $(R(\beta), nQ)$ as imposed by the retailer. From Property 1 and Property 2 both systems operate under modified base-stock policies and the policies are identical. Therefore the costs are equal.

For $\beta \not\in S$ or $T_r \neq 1$, the cost under no-consignment is lower than the cost under traditional system. The reason is, when $\beta \not\in S$, the manufacturer can now consider randomized policies and therefore is more flexible in terms of dispatch policies. For $T_r \neq 1$, under the traditional system dispatches are not allowed in any period except the first period of the replenishment cycle. In no-consignment system on the other hand, there is no restriction on the dispatch quantity in any period. This corresponds to a more flexible system and therefore the cost under no-consignment system is lower compared to the traditional system.

7.5 Proof of Proposition 2

We make the proof in two steps. In Step 1 for the simple scenario defined, we provide an exact characterization of the optimal markov-modulated modified base-stock policy under VM-NC (in Lemma 1). Using this characterization, in Step 2 we provide a lower bound on the optimal cost under VM-NC.

**Step 1.** We assumed $\beta \in S$ and $Q = 1$. This implies under optimality there exists a unique deterministic dispatch policy for the manufacturer, which is simply “dispatch the last period’s realized demand”. The states are $(I_m, I_r)$ where $I_m$ and $I_r$ are the inventory levels at the beginning of the period at the manufacturer and at the retailer, respectively. If retailer’s demand in last period is $\mu - 1$, $I_r =$ retailer’s base-stock level $-(\mu - 1)$. Retailer’s base-stock level is directly implied by the service level requirement and is irrelevant to the manufacturer’s optimal production and dispatch policy (see Observation 1). For notational simplicity, we ignore “retailer’s base-stock level” and simply indicate $I_r$ with $-(\mu - 1)$ or $-(\mu + 1)$. Lemma 1 characterizes optimal production policy of the manufacturer.

**Lemma 1** Under VM-NC the optimal production policy of the manufacturer is as follows (since in our model outsourcing and lost sales are equivalent, we use them interchangeably):

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(i) If \( I_m = 0 \) and retailer’s demand in the last period is \( \mu + 1 \) (i.e., if \( I_r = -\mu - 1 \)), produce \( \mu \) and outsource 1 (lose the sale of 1 unit),

(ii) If \( I_m = I_{\text{max}} \) and demand in the last period is \( \mu - 1 \) (i.e., if \( I_r = -\mu + 1 \)), produce \( \mu - 1 \),

(iii) Else produce \( \mu \) (i.e., for \( 0 < I_m \leq I_{\text{max}} \) and \( I_r = -\mu - 1 \), or for \( 0 \leq I_m < I_{\text{max}} \) and \( I_r = -\mu + 1 \) produce \( \mu \)),

where \( I_{\text{max}} \) indicates the maximum inventory level at the manufacturer (see Figure 6).

\[ \text{Figure 6: The transition diagram of the underlying Markov chain of the optimal policy under VM-NC.} \]

**Proof (Lemma 1).** Property 2 states that under VM-NC the optimal production policy is a modified base-stock policy. The modified base-stock policy states the following operating policy structure: For states \( I_m = 0, 1, \ldots, k \) produce \( \mu \), for \( I_m = k + 1 \) produce \( \mu - 1 \) and decrease the production quantity in unit increments as beginning stock level, \( I_m \), increases. For \( I_m = 0 \) producing less than \( \mu \) might as well be optimal. Note that \( k \) might be different for different \( I_r \) levels. In the following, for each state we determine the optimal production quantity (i.e., we determine \( k \) for the two \( I_r \) levels, \( I_r = -\mu + 1 \) and \( I_r = -\mu - 1 \)). We analyze the states \((I_m, -\mu + 1)\) and \((I_m, -\mu - 1)\) separately. The reason is, the problem has a markov-modulated structure and the base-stock levels might be different under \( I_r = -\mu + 1 \) and \( I_r = -\mu - 1 \). Note that \( I_m \geq 0 \).

The **optimal production policy under VM-NC** is as follows:

(i) For states \((I_m, -\mu + 1)\) dispatch quantity is \( \mu - 1 \). We start with the analysis of \( I_m = 0 \). For \((0, -\mu + 1)\) the manufacturer may prefer to produce \( \mu \) or \( \mu - 1 \). Note that producing \( \mu - 2 \) and outsourcing one unit, or outsourcing several units to accumulate stock are more costly actions. Therefore for \((0, -\mu + 1)\) only two actions, produce \( \mu \) or \( \mu - 1 \), are under consideration. Note furthermore that, outsourcing is not an optimal action in any state \((I_m, -\mu + 1)\).

For any \((I_m, -\mu + 1)\) if the production quantity is \( \mu \), then the next state is \((I_m + 1, -\mu + 1)\) or \((I_m + 1, -\mu - 1)\) (each with probability \( \frac{1}{2} \)). If, on the other hand, production quantity is \( \mu - 1 \), then next state is \((I_m, -\mu + 1)\) or \((I_m, -\mu - 1)\). In other words, the period where the manufacturer
decides to produce $\mu - 1$ under $(I_m, -\mu + 1)$ determines a candidate for the maximum stock level under optimal policy, say $I_{\text{max}}$. In (ii) below we show that $I_{\text{max}}$ is indeed the maximum stock level under optimal policy, and in Step 2 we determine the $I_{\text{max}}$ value in terms of problem parameters. Under the optimal policy, for $0 \leq I_m < I_{\text{max}}$ the manufacturer produces $\mu$ and for $I_m = I_{\text{max}}$ the manufacturer produces $\mu - 1$ in state $(I_m, -\mu + 1)$ (note, $I_{\text{max}}$ can be zero).

(ii) For states $(I_m, -\mu - 1)$ dispatch quantity is $\mu + 1$. We start with the analysis of $I_m = 0$. For $I_m = 0$ optimal production quantity is $\mu$ and the remaining one unit is outsourced. Note that producing less than $\mu$ and/or outsourcing more than one unit are more costly actions. Also for $I_m > 0$, outsourcing to accumulate stock is a more costly action. This implies that when the current state is $(I_m, -\mu - 1)$, in the next state the $I_m$ level decreases (for $I_m > 0$). In other words the candidate $I_{\text{max}}$ in (i) is indeed the maximum stock level.

Modified base-stock policy states that production quantity is $\mu$ for states $0 \leq I_m \leq k$ and then decreases in unit increments. We would like to determine the optimal value of $k$ (which might as well be zero). First observe that under $(I_m, -\mu - 1)$ if production quantity is $\mu$, then the next state is $(I_m - 1, -\mu + 1)$ or $(I_m - 1, -\mu + 1)$ (each with probability $\frac{1}{2}$), if production quantity is $\mu - 1$ then next state is $(I_m - 1, -\mu + 1)$ or $(I_m - 1, -\mu + 1)$, and so on. In the following, we show that for $0 \leq I_m \leq I_{\text{max}}$ optimal production quantity is $\mu$.

Define $i$, as the state where optimal production quantity is $\mu$ for $0 \leq I_m \leq i$ (where $i$ can be anything in $\{0, 1, \ldots, I_{\text{max}} - 1\}$). At state $(i + 1, -\mu - 1)$ possible actions are to produce $\mu$ or $\mu - 1$. The optimality equation (under average cost criteria) at state $(i + 1, -\mu - 1)$ is as follows:

$$v(i + 1, -\mu - 1) = -g + \min\{\text{produce } \mu, \text{produce } \mu - 1\},$$

$$v(i + 1, -\mu - 1) = -g + \min\{c\mu + (i)h + \frac{1}{2}v(i, -\mu + 1) + \frac{1}{2}v(i, -\mu - 1),$$

$$c(\mu - 1) + (i - 1)h + \frac{1}{2}v(i - 1, -\mu + 1) + \frac{1}{2}v(i - 1, -\mu - 1)\}, \hspace{1cm} (19)$$

where $v(i, j)$ is the optimal bias value of starting in state $(i, j)$, and $g$ is the optimal average cost in the Markov Decision Process (Puterman 1994).

Now consider the optimality equation at $(i - 1, -\mu + 1)$. Possible actions are produce $\mu$ or $\mu - 1$:

$$v(i - 1, -\mu + 1) = -g + \min\{\text{produce } \mu, \text{produce } \mu - 1\},$$

$$v(i - 1, -\mu + 1) = -g + \min\{c\mu + (i)h + \frac{1}{2}v(i, -\mu + 1) + \frac{1}{2}v(i, -\mu - 1),$$

$$c(\mu - 1) + (i - 1)h + \frac{1}{2}v(i - 1, -\mu + 1) + \frac{1}{2}v(i - 1, -\mu - 1)\}, \hspace{1cm} (20)$$
Note that right hand side of Equation (19) and Equation (20) are the same, therefore the optimal actions taken at states \((i + 1, -\mu - 1)\) and \((i - 1, -\mu + 1)\) must be the same. We have shown in (i) that for \(0 \leq I_m < I_{\text{max}}\) in state \((I_m, -\mu + 1)\) optimal action is to produce \(\mu\). We conclude that for \(0 \leq i \leq I_{\text{max}}\), optimal action at state \((i, -\mu - 1)\) is to produce \(\mu\). This equivalently implies that for \(I_r = -\mu - 1\), optimal value of \(k\) is \(I_{\text{max}}\).

Analysis in (i) and (ii) completes the proof of Lemma 1. □

**Step 2.** Given the manufacturer’s optimal policy structure, it is possible to determine the \(I_{\text{max}}\) value that minimizes the cost. Under the described optimal policy structure, the corresponding Markov chain implies that the steady-state probability of a state \((I_m, I_r)\) is:

\[
\pi_{\text{MC}}(I_m, I_r) = \frac{1}{2(I_{\text{max}}+1)}.
\]

The one-step reward at state \((I_m, I_r)\) is expressed as follows:

\[
r(I_m, I_r) = \begin{cases} 
  c\mu + w(1) & \text{if } I_m = 0, I_r = -\mu - 1 \\
  c(\mu - 1) + h(I_{\text{max}}) & \text{if } I_m = I_{\text{max}}, I_r = -\mu + 1 \\
  c\mu + h(I_m) & \text{if } 0 < I_m \leq I_{\text{max}} \text{ and } I_r = -\mu - 1, \text{ or,} \\
  & 0 \leq I_m \leq I_{\text{max}} \text{ and } I_r = -\mu + 1 
\end{cases}
\]

Based on the steady-state probabilities and the reward function, we express the average-cost under VM-NC as:

\[
\text{cost}(\text{VM-NC}) = \frac{w(1) + h(I_{\text{max}})(I_{\text{max}} + 1) - c(1)}{2(I_{\text{max}} + 1)} + c\mu,
\]

where \(I_{\text{max}}\) is integer. Note that \(\text{cost}(\text{VM-NC})\) is convex in \(I_{\text{max}}\). The \(I_{\text{max}}\) value that minimizes (21) is:

\[
I^*_{\text{max}} = \sqrt{\frac{w - c}{h}} - 1
\]

In (22) \(I^*_{\text{max}}\) can be a real number or an integer number. Replacing \(I_{\text{max}}\) in (21) with \(I^*_{\text{max}}\) gives a lower bound on the optimal average cost under VM-NC:

\[
LB(\text{VM-NC}) = \sqrt{(w - c)h - \frac{h}{2}} + c\mu
\]

We write \(LB(\text{VM-NC}) \leq \text{cost}(\text{VM-NC})\), since \(I^*_{\text{max}}\) in (22) is not necessarily an integer number. □

### 7.6 Proof of Proposition 3

We consider the following operating policy.

**Policy-VMC:**

(i) For \(I_r = I_r^{\text{min}}\), produce (and dispatch) \(\mu + 1\)

(ii) For \(I_r = I_r^{\text{max}}\), produce (and dispatch) \(\mu - 1\),

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(iii) Otherwise, for $I_r^\text{min} < I_r < I_r^\text{max}$, produce (and dispatch) $\mu$,

where we assume $I_r^\text{max} \geq I_r^\text{min} + 3$ (i.e., under Policy-VMC there exist at least four states). The proposed policy, Policy-VMC, is not necessarily the optimal policy under VM-C, therefore the cost implied by this policy will be an upper bound on the optimal cost under VM-C. Note that the transition probabilities for all $I_r$ are $\frac{1}{2}$. The transition diagram of the underlying Markov chain is presented in Figure 7.

![Figure 7: The transition diagram of the underlying Markov chain under Policy-VMC under VM-C.](image)

The steady-state probabilities, $\pi^M_C(I_r)$, implied by Policy-VMC are as follows:

$$
\pi^M_C(I_r) = \begin{cases} 
\frac{1}{2\Delta} & \text{for } I_r = I_r^\text{min}, I_r^\text{min} + 1, I_r^\text{max} - 1, I_r^\text{max} \\
\frac{1}{\Delta} & \text{for } I_r^\text{min} + 1 < I_r < I_r^\text{max} - 1 \\
0 & \text{otherwise}
\end{cases}
$$

where $\Delta = I_r^\text{max} - I_r^\text{min} - 1$. The reward in state $I_r$ is expressed as follows:

$$
r(I_r) = \begin{cases} 
c\mu + w(1) & \text{if } I_r = I_r^\text{min} \\
c(\mu - 1) + h(I_r^\text{max}) & \text{if } I_r = I_r^\text{max} \\
c\mu + h(I_r^\text{max})^+ & \text{if } I_r^\text{min} < I_r < I_r^\text{max}
\end{cases}
$$

Our aim is to determine an upper bound on the optimal average cost under VM-C. Note that in (24), $I_r^\text{max}$ is always positive, since otherwise $\beta \geq 1$. Furthermore, $I_r^\text{min}$ is non-positive (since it is already possible to attain 100% when $I_r^\text{min} = 0$). When determining $I_r^\text{min}$ and $\Delta$ we make sure that the service level constraint in (13), where $\beta \in S$, is satisfied. In other words the following constraint is satisfied:

$$
\frac{\sum_{i=0}^{i=I_r^\text{min}} |i| \pi^M_C(I_r = i)}{E[D]} \leq \beta
$$

Based on $r(I_r)$ and $\pi^M_C(I_r)$ we write an upper bound on optimal average cost, $UB(\text{VM-C})$, as follows:

$$
UB(\text{VM-C}) = r(I_r)\pi^M_C(I_r) = w \frac{1}{2\Delta} + h(\frac{1}{\Delta} + \frac{2}{\Delta} + \cdots + \frac{I_r^\text{max} - 2}{\Delta} + \frac{I_r^\text{max} - 1}{2\Delta} + \frac{I_r^\text{max}}{2\Delta} + \frac{I_r^\text{max} - 1}{2\Delta}) - c\frac{1}{2\Delta} + c\mu.
$$

The cost, $UB(\text{VM-C})$, consists of the cost of production and outsourcing in all states $\{I_r^\text{min}, \ldots, I_r^\text{max}\}$ and the cost of holding inventory in states $\{1, 2, \ldots, I_r^\text{max}\}$. Replacing $I_r^\text{max}$ with $\Delta - |I_r^\text{min}| + 1$:

$$
UB(\text{VM-C}) = w \frac{1}{2\Delta} + h(\frac{1}{\Delta} + \frac{2}{\Delta} + \cdots + \frac{\Delta - |I_r^\text{min}| - 1}{\Delta} + \frac{\Delta - |I_r^\text{min}|}{2\Delta} + \frac{\Delta - |I_r^\text{min}| + 1}{2\Delta}) - c\frac{1}{2\Delta} + c\mu,
$$

which is equivalent to,

$$
UB(\text{VM-C}) = \begin{cases} 
h(\Delta - |I_r^\text{min}| + 1/2)^2 + w - c + \frac{3}{2}h + c\mu & \text{if } I_r^\text{max} \geq 3 \\
w \frac{1}{2\Delta} + h(\frac{1}{2\Delta} + \frac{2}{\Delta} - c\frac{1}{2\Delta} + c\mu) & \text{if } I_r^\text{max} = 2 \\
w \frac{1}{2\Delta} + h(\frac{1}{2\Delta} - c\frac{1}{2\Delta} + c\mu) & \text{if } I_r^\text{max} = 1
\end{cases}
$$

To make the analysis simpler we focus on the case where $I_r^\text{max} \geq 3$. This is equivalent to $\Delta - |I_r^\text{min}| \geq 2$.

In summary we make the following assumptions when defining Policy-VMC.
A1. \( I_{r}^{\text{min}} \leq 0 \),

A2. \( \Delta - |I_{r}^{\text{min}}| \geq 2 \)

From second order condition, for \( I_{r}^{\text{max}} \geq 3 \) the expression in (26) is convex in \( \Delta \). From the first order condition, the minimizing \( \Delta \) is obtained as a function of \( I_{r}^{\text{min}} \) as follows:

\[
\Delta^* = \sqrt{\frac{w-c}{h}} + (|I_{r}^{\text{min}}|^2 - |I_{r}^{\text{min}}| + 1)
\]

In order to obtain a “reasonable” upper bound (i.e., an upper bound which is not very relaxed), we consider the \( \Delta \) that minimizes (26). To determine the values \( I_{r}^{\text{min}} \) could take, we make the following observation. **Observation.** Production and outsourcing cost of the manufacturer is independent of \( I_{r}^{\text{min}} \), and the inventory holding cost decreases as \( I_{r}^{\text{min}} \) decreases.

The observation implies that for a given \( \Delta \), as \( I_{r}^{\text{min}} \) decreases total cost decreases. If there were no service level constraint, to lower the cost one should lower the \( I_{r}^{\text{min}} \) value. However, there is a limit on the lowest value \( I_{r}^{\text{min}} \) can take when minimizing \( UB(\text{VM-C}) \), determined by the service level. Based on (25) and considering \( I_{r}^{\text{max}} \geq 3 \):

\[
\frac{|I_{r}^{\text{min}}|}{2\Delta} + \frac{|I_{r}^{\text{min}}| - 1}{2\Delta} + \frac{|I_{r}^{\text{min}}| - 2}{\Delta} + \cdots + \frac{1}{\Delta} \leq \beta \mu.
\]

In the expression above, for \( \beta = 0 \) (i.e, for 100% service level) \( I_{r}^{\text{min}} \) should be 0. Otherwise, to be able to set \( |I_{r}^{\text{min}}| \) as high as \( k \), \( \beta \) should satisfy:

\[
\beta \geq \frac{k^2 - k + 1}{2\mu \Delta} \text{ for } k = 1, \cdots
\]

(27)

For a required service level, if the \( |I_{r}^{\text{min}}| \) level is set to its highest possible value, then this would correspond to the lowest possible inventory holding cost under that service level. A limit on how high \( |I_{r}^{\text{min}}| \) can be set is obtained from Equation (27). For a given \( |I_{r}^{\text{min}}| \), \( \Delta^* \) minimizes the UB(VM-C) value. We place \( \Delta^* \) in equation (27) and we suggest two possible upper bound values for optimal average cost under VM-C: UB1 for \( I_{r}^{\text{min}} = 0 \), and a generalized upper bound for \( |I_{r}^{\text{min}}| > 0 \).

**Upper bound 1 (UB1).** We set \( I_{r}^{\text{min}} \) to the highest possible value, \( I_{r}^{\text{min}} = 0 \). Since as \( I_{r}^{\text{min}} \) decreases, \( UB(\text{VM-C}) \) decreases, upper bound obtained under \( I_{r}^{\text{min}} = 0 \) can be regarded as a “relaxed” upper bound.

For \( I_{r}^{\text{min}} = 0 \), we obtain \( \Delta^* = \sqrt{\frac{w-c}{h}} + 1 \). From Assumption A2, \( \Delta^* \) must be greater than or equal to 2. This is equivalent to, \( w \geq 3h + c \). Furthermore to guarantee that the expression in (26) is an upper bound, \( \Delta^* \) must be an integer. Equivalently, we say, \( w \) should be such that \( \Delta^* \) is integer-valued.

Placing \( I_{r}^{\text{min}} = 0 \) and \( \Delta^* \) in (26) we obtain the following.
Suppose that \( w \in W \), where \( W = \{ w | w \geq 3h + c, \sqrt{\frac{w-c}{h}} + 1 \in \mathbb{Z}^+ \} \). Then \( UB(VM-C) = h(\sqrt{\frac{w-c}{h}} + 1 + \frac{1}{2}) + cm \) is an upper bound on the optimal average cost under VM-C.

### A generalized upper bound

We construct a more general set of conditions for determining an upper bound. Let \( I_r^{\text{min}} = k \). For \( I_r^{\text{min}} = k \), we obtain \( \Delta^* = \sqrt{\frac{w-c}{h}} + (|I_r^{\text{min}}|^2 - |I_r^{\text{min}}| + 1) \). From Assumption A2, \( \Delta^* \) must be greater than or equal to \( |I_r^{\text{min}}| + 2 \). This is equivalent to, \( w \geq (5|I_r^{\text{min}}| + 3)h + c \). Also, \( \Delta^* \) must be an integer.

For a required service level \( 1 - \beta \), highest \( |I_r^{\text{min}}| \) level is obtained as follows:

\[
1 - \beta \leq 1 - \frac{|I_r^{\text{min}}|^2 - |I_r^{\text{min}}| + 1}{2\mu \sqrt{\frac{w-c}{h}} + |I_r^{\text{min}}|^2 - |I_r^{\text{min}}| + 1}
\]

(28)

Placing \( I_r^{\text{min}} \) and \( \Delta^* \) in (26) we obtain the following.

Suppose that \( w \in W \), where \( W = \{ w | w \geq (5k + 3)h + c, \sqrt{\frac{w-c}{h}} + (k^2 - k + 1) \in \mathbb{Z}^+ \} \) and \( SL \leq 1 - \frac{k^2 - k + 1}{2\mu \sqrt{\frac{w-c}{h}} + k^2 - k + 1} \). Then \( UB(VM-C) = h(\sqrt{\frac{w-c}{h}} + k^2 - k + 1 - (k - \frac{1}{2})) + cm \), where \( k \in \mathbb{Z}^+ \), is an upper bound on the optimal average cost under VM-C. \( \square \)

### 7.7 Proof of Proposition 4

We compare \( LB(VM-NC) \) and \( UB(VM-C) \) and obtain the (sufficient) conditions under which VM-NC results in higher cost than VM-C. In Proposition 3 we obtain two upper bounds for the optimal average cost of VM-C.

First we compare UB1 with \( LB(VM-NC) \). UB1 may be a reasonable upper bound when the service level requirement is very high. When \( I_r^{\text{min}} = 0 \), upper bound is expressed as: \( UB(VM-C) = h(\sqrt{\frac{w-c}{h}} + 1 + \frac{1}{2}) + cm \). Comparing this with \( LB(VM-NC) = \sqrt{(w-c)h} - \frac{h}{2} + cm \), we obtain that \( UB(VM-C) \) is always higher than the \( LB(VM-NC) \). Similarly, when \( I_r^{\text{min}} = 1 \), we again obtain that \( UB(VM-C) \) is higher than the \( LB(VM-NC) \). Therefore we conclude that when service level requirement at the retailer is high, it is less likely that VM-C results in lower cost.

For \( I_r^{\text{min}} \geq 2 \), under \( \Delta^* \) the upper bound in (26) is expressed as:

\[
UB(VM-C) = h(\sqrt{\frac{w-c}{h}} + |I_r^{\text{min}}|^2 - |I_r^{\text{min}}| + 1) - (|I_r^{\text{min}}| - \frac{1}{2})) + cm.
\]

Comparing this with \( LB(VM-NC) \), we obtain the following:

\[
UB(VM-C) < LB(VMNC)
\]

\[
h(\sqrt{\frac{w-c}{h}} + (|I_r^{\text{min}}|^2 - |I_r^{\text{min}}| + 1) - (|I_r^{\text{min}}| - \frac{1}{2})) < h(\sqrt{\frac{w-c}{h}} - \frac{1}{2})
\]

\[
|I_r^{\text{min}}| < 2\sqrt{\frac{w-c}{h}}(|I_r^{\text{min}}| - 1)
\]
As $|I^\text{min}_r|$ increases it is more likely that VM-C yields lower cost. For $I^\text{min}_r = k$, $w > \frac{1}{4}(\frac{k}{k-1})^2h + c$ is a sufficient condition for $UB(\text{VM-C}) < LB(\text{VM-NC})$. Note that, in Proposition 3 we obtained $w \geq (5k + 3)h + c$, for $k \geq 1$ as a tighter condition than $w > \frac{1}{4}(\frac{k}{k-1})^2h + c$. Therefore we only consider $w \geq (5k + 3)h + c$.

We summarize the sufficient conditions for preferring VM-C over VM-NC as follows:

1. $SL \leq 1 - \frac{k^2-k+1}{2\mu \sqrt{\frac{w-c}{h} + k^2-k+1}}$
2. $w \geq (5k + 3)h + c$
3. $w$ is such that $\sqrt{\frac{w-c}{h} + k^2-k+1}$ is integer-valued.
4. $k \geq 2$

We conclude that when the sufficient conditions are satisfied, the optimal cost under VM-C is lower than the optimal cost under VM-NC. ☐

References


