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Substitutable Inventory Systems with Coordinated Reorder Levels

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Abstract: This paper considers a two commodity continuous review inventory system with Markovian demands. The two commodities are assumed to be both way substitutable. That is, if the inventory level of one commodity reaches zero, then a demand for this commodity will be satisfied by an item of the other commodity. A joint order is placed when the inventory level reaches to any one of the reorder levels in the set of reorder levels with some prefixed probability. The limiting probability distribution for the joint inventory levels is computed. Various operational characteristics and the expression for the long run total expected cost rate are derived.

Keywords: Two-commodity, substitutable items, Poisson demand, continuous review, coordinated reorder levels.

1 Introduction

With the advent of advanced computing systems, many organizations have increasingly use multi commodity inventory systems. Further, models were proposed with independently established reorder points. When several products compete for limited storage space, or share the same transport facility, or items are produced on (procured from) the same equipment (supplier), the above strategy overlooks the potential savings associated with the joint replenishment and cost reduction in the ordering and setup costs.

In continuous review inventory systems, Balintfy and Silver [1,2] have considered a coordinated reordering policy which is represented by the triplet of vectors ($\mathbf{S}, \mathbf{c}, \mathbf{s}$), where the parameters S_i, c_i and s_i (the components of \mathbf{S}, \mathbf{c} and \mathbf{s} respectively) are specified for each item *i* with $s_i \leq c_i \leq S_i$. In this policy, if the level of commodity *i* at any time is below s_i , an order is placed for $S_i - s_i$ items and at the same time, for any other item $j \neq i$) with available inventory at or below its can-order level c_j , an order is placed so as to bring its level back to its capacity S_j . Subsequently many articles have appeared with models involving the above policy. Another article of interest is due to Federgruen [3] et al., which deals with the general case of compound Poisson demands and non-zero lead times. A review of inventory models under joint replenishment is provided by Goyal and Satir[4].

Kalpakam and Arivarignan[5] have introduced an (s, S) policy with a single reorder level s defined in terms of the total number of items in the stock. This policy avoids separate ordering for each commodity and hence a single processing of orders for both commodities has some advantages in situation where the procurement is made from the same suppliers, items are produced on the same machine, or items have to be supplied by the same transport facility.

Krishnamoorthy [6] et al. have considered a two commodity continuous review inventory system without lead time. In their model, each demand is for one unit of the first commodity or one unit of second commodity or one unit of each commodity 1 and 2, with prefixed probabilities.Krishnamoorthy and Varghese[7] have considered a two commodity inventory problem without lead time and with Markov shift in demand for the type of commodity namely "commodity-1", "commodity-2" or "both commodities", using the direct Markov renewal theoretical results. For the

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same problem, Sivasamy and Pandiyan[8] have derived various results by applying filtering techniques.

A natural extension of (s, S) policy to two-commodity inventory system is to have two reorder levels and to place orders for each commodity independent of the other. But this policy will increase the total cost as separate processing of two orders is required.

Anbazhagan and Arivarignan [9] have considered a two commodity inventory system with independent reorder levels where a joint order for both commodities is placed only when the levels of both commodities fall below their respective reorder levels. The demand points for each commodity form independent Poisson processes and the lead time is distributed as negative exponential. They have also assumed unit demands for both commodities. Yadavalli [10, 11] et al. have analyzed two commodity inventory system under various ordering policies. Sivakumar[12] et al. have considered a two commodity coordinated and individual ordering policies with renewal demands. Anbazhagan and Vigneshwaran[13] have considered a two commodity markovian inventory system with joint reorder levels.

In this article we considered a two commodity continuous review inventory system with independent reorder levels $s_i - k, k = 0, 1, \dots, r$ where a joint order for both commodities is placed only when the levels of both commodities fall below their respective reorder levels. It is assumed that the demand for commodity *i* is of unit size and the time points of demand occurrences form a Poisson process with parameter λ_i (*i* = 1,2). The two commodities are assumed to be substitutable. That is, if the inventory level of one commodity reaches zero, then any demand for this commodity will be satisfied by an item of the other commodity. The lead time is assumed to be distributed as negative exponential with parameter μ_k , $k = 0, 1, \dots, r$. The joint probability distribution of the two inventory levels is obtained in the steady state case. Various measures of systems performance in the steady state are also derived.

2 Problem formulation

Consider a two commodity inventory system with capacity S_i units for commodity i, (i = 1, 2). It is assumed that the demands for *i*-th commodity are of unit size and having Poisson distribution with parameter λ_i (i = 1, 2). The demand process of the two commodities are further assumed to be independent. The two commodities are assumed to be substitutable. That is, if the inventory level of one commodity reaches zero, then any demand for this commodity will be satisfied by the item of the other commodity. The reordering policy is to place order for both commodities when both inventory levels are less than or equal to their respective reorder levels $s_i - k, k = 0, 1, \dots, r$, with probability p_k , $\sum p_k = 1$, $(0 \le r \le \min \{s_1 - 2, s_2 - 2\}$ and $S_i - s_i + k > s_i + 1, i = 1, 2)$. The ordering quantity is $Q_{s_i-k}^i (= S_i - s_i + k), i = 1, 2$. The lead time initiated at level $s_i - k$ is assumed to be distributed as exponential with parameter $\mu_k(>0)$. The demands that occur during stock out periods are lost.

Let $L_i(t)$ denote the net inventory level of commodity *i* at time *t*. Then the process

 $L = \{(L_1(t), L_2(t)), t \ge 0\}$ has the state space

 $E = \{0, 1, \cdots, S_1\} \ge \{0, 1, \cdots, S_2\}.$

The space of inventory levels of commodity 1 and 2 is shown in Figure 1.

Notations

0 : zero matrix

 $1'_N$: $(1, 1, \cdots, 1)_{1 \times N}$

$$\mathbf{e}^T \qquad : (1,1,\ldots,1).$$



- I_N : identity matrix of order N
- δ_{ij} : Kronecker delta

 $\sum_{j=0}^{i} a^{j} = \begin{cases} a^{0} + a^{1} + \dots + a^{i}, & \text{if } i \text{ is nonnegative integer} \\ 0, & \text{otherwise} \end{cases}$

 $[A]_{ii}$: (i, j)-th element of the matrix A

$$H(x) = \begin{cases} 1 & \text{if } x \ge 0\\ 0 & \text{if } x < 0 \end{cases}$$

 $\underset{i=j}{\overset{k}{\Omega}} c_i \quad = \begin{cases} c_j c_{j-1} \cdots c_k & \text{if } j \ge k \\ 1 & \text{if } j < k \end{cases}$

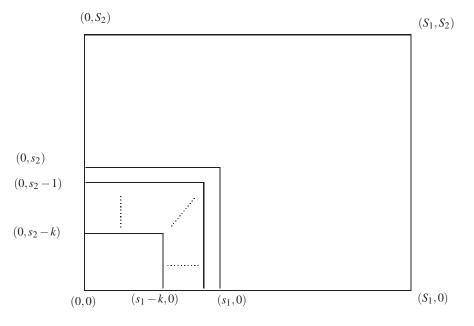


Fig. 1: Space of inventory levels of commodity 1 and 2

From the assumptions made on demand and the replenishment processes it follows that *L* is a Markov process. To determine the infinitesimal generator $\tilde{A} = ((a((i, j); (k, l)))), (i, j), (k, l) \in E$, of this process. **Theorem 1:** The infinitesimal generator of this Markov process a((i, j), (k, l)) is given by,

$$\begin{cases} \lambda_1 + \delta_{0j}\lambda_2, & k = i - 1, & l = j, \\ i = 1, 2, \cdots, S_1, & j = 0, 1, \cdots, S_2 \\ \delta_{io}\lambda_1 + \lambda_2, & k = i, & l = j - 1, \\ i = 0, 1, \cdots, S_1, & j = 1, 2, \cdots, S_2 \\ -(\lambda_1 + \lambda_2), & k = i, & l = j, \\ i = s_1 + 1, \cdots, S_1, & j = 0, 1, \cdots, S_2 \\ -(\lambda_1 + \lambda_2), & k = i, & l = j, \\ i = 0, 1, \cdots, s_1, & j = s_2 + 1, \cdots, S_2 \\ -(\lambda_1 + \lambda_2 + H(s_1 - i - s_2 + j - 1)\sum_{n=0}^{s_2 - j} \mu_n p_n + H(s_2 - j - s_1 + i)\sum_{n=0}^{s_2 - j} \mu_n p_n), & k = i, & l = j \\ i = s_1 - r, \cdots, s_1 & j = 0, 1 \cdots, s_2 \\ -((1 - \delta_{0i}\delta_{0j})(\lambda_1 + \lambda_2) + H(s_1 - i - s_2 + j - 1)\sum_{n=0}^{s_2 - j} \mu_n p_n + H(s_2 - j - s_1 + i)\sum_{n=0}^{s_2 - j} \mu_n p_n, & k = i, & l = j, \\ i = 0, 1, \cdots, s_1 - r - 1 & j = 0, 1, \cdots, s_2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 - m} P_{s_2 - m} & k = i + Q_{m-s_2 + s_1}^1, & l = j + Q_m^2 \\ \mu_{s_2 -$$

Proof:

The infinitesimal generator a((i, j), (k, l)) of this process can be obtained using the following arguments:

(i)Let i > 0 and j > 0. A demand takes the inventory level (i, j) to (i - 1, j) with intensity λ_1 the demand being for the first commodity or to (i, j - 1) with intensity λ_2 the demand being for the second commodity.

(ii)From the state (i, j), $(\leq (s_1 - k, s_2 - k))$ a replenishment takes the joint inventory level to $(i + Q_{s_1-k}^1, j + Q_{s_2-k}^2)$ and the intensity of transition is given by μ_k , $k = 0, 1, \dots, r$.

(iii)We observe that no transition other than the above is possible except $(i, j) \neq (k, l)$. (iv)Finally the value of a((i, j), (i, j)) is obtained by

$$a((i,j),(i,j)) = -\sum_{\substack{k \ l \\ (k,l) \neq (i,j)}} \sum_{a((i,j),(k,l))} a((i,j),(k,l))$$

Hence we get the infinitesimal generator a((i, j), (k, l)).

In order to write the infinitesimal generator \tilde{A} in matrix form, we arrange the states in lexicographic order and group $S_2 + 1$ states as



$$i = ((i,0), (i,1), \cdots, (i,S_2)), i = 0, 1, \cdots, S_1.$$

Then the rate matrix \tilde{A} has a block partitioned form with the following sub matrix $[\tilde{A}]_{ij}$ at the *i*-th row and *j*-th column position.

$$[\tilde{A}]_{ij} = \begin{cases} B & \text{if} \quad j = i - 1, & i = S_1, S_1 - 1, \cdots, 1 \\ A & \text{if} \quad j = i, & i = S_1, S_1 - 1, \cdots, s_1 + 1 \\ A_{s_1 + 1 - i} & \text{if} \quad j = i, & i = s_1, s_1 - 1, \cdots, s_1 + 1 \\ M_{[j - i - Q_{s_1}^1]} & \text{if} \quad j = S_1, S_1 - 1, \cdots, S_1 - (s_1 - i), & i = s_1, s_1 - 1, \cdots, s_1 - r \\ M_{[j - i - Q_{s_1}^1]} & \text{if} \quad j = S_1 - (s_1 - i) + r, \cdots, S_1 - (s_1 - i), i = s_1 - r - 1, \cdots, 1, 0 \\ 0 & \text{otherwise.} \end{cases}$$

where

$$[B]_{ab} = \begin{cases} \lambda_1 & \text{if } b = a, \ a = S_2, S_2 - 1, \cdots, 1\\ \lambda_1 + \lambda_2 & \text{if } b = a, \ a = 0\\ 0 & \text{otherwise.} \end{cases}$$

$$[A]_{ab} = \begin{cases} \lambda_2 & \text{if } b = a - 1, \, a = S_2, S_2 - 1, \cdots, 1\\ -(\lambda_1 + \lambda_2) & \text{if } b = a, \quad a = S_2, S_2 - 1, \cdots, 0\\ 0 & \text{otherwise.} \end{cases}$$

$$[M_i]_{ab} = \begin{cases} p_i \mu_i \text{ if } b = Q_2 + i + a \ a = s_2 - i, \cdots, 1, 0, \\ 0 \text{ otherwise.} \end{cases}$$

with $i = 0, 1, \cdots, r$

$$[A_i]_{ab} = \begin{cases} \lambda_2 & \text{if } b = a - 1, a = S_2, S_2 - 1, \cdots, 1 \\ -(\lambda_1 + \lambda_2) & \text{if } b = a, a = S_2, S_2 - 1, \cdots, s_2 + 1 \\ -(\lambda_1 + \lambda_2 + B_1) & \text{if } b = a, a = S_2, S_2 - 1, \cdots, s_2 + 1 \\ H(i - s_2 + b - 2) \sum_{k=0}^{s_2 - a} p_k \mu_k + B_1 & \text{if } b = a, a = s_2, s_2 - 1, \cdots, 1, 0 \\ 0 & \text{otherwise.} \end{cases}$$
with $i = 1, 2, \cdots, r + 1$

$$[A_i]_{ab} = \begin{cases} \lambda_2 & \text{if } b = a - 1, a = S_2, S_2 - 1, \cdots, 1 \\ -(\lambda_1 + \lambda_2) & \text{if } b = a, a = S_2, S_2 - 1, \cdots, s_2 + 1 \\ -(\lambda_1 + \lambda_2 + K_2) & \text{if } b = a, a = S_2, S_2 - 1, \cdots, s_2 + 1 \\ H(i - s_2 + b - 2) \sum_{\substack{k=0 \ r}{r}}^{s_2 - a} p_k \mu_k + K_2 & \text{if } b = a, a = s_2, s_2 - 1, \cdots, 1, 0 \\ 0 & \text{otherwise.} \end{cases}$$

with $i = r+2, \cdots, s_1$

$$[A_{s_1+1}]_{ab} = \begin{cases} \lambda_1 + \lambda_2 & \text{if } b = a - 1, a = S_2, S_2 - 1, \cdots, 1 \\ -(\lambda_1 + \lambda_2) & \text{if } b = a, a = S_2, S_2 - 1, \cdots, s_2 + 1 \\ -(\lambda_1 + \lambda_2 + B_1) \sum_{k=0}^{s_2 - a} p_k \mu_k + B_1 \\ H(s_1 - s_2 + b - 1) \sum_{k=0}^r p_k \mu_k & \text{if } b = a, a = s_2, s_2 - 1, \cdots, 1 \\ H(s_2 - s_1 - b) \sum_{k=0}^r p_k \mu_k & \text{if } b = a, a = 0 \\ 0 & \text{otherwise.} \end{cases}$$

3 Steady state Results

It can be seen from the structure of A that the homogeneous Markov process $\{(L_1(t), L_2(t)), t \ge 0\}$ on the state space E is irreducible. Hence the limiting distribution

$$\Phi = (\phi^{(S_1)}, \phi^{(S_1-1)}, \dots, \phi^{(0)})$$

with $\phi^{(q)} = (\phi^{(q,S_2)}, \phi^{(q,S_2-1)}, \dots, \phi^{(q,0)})$, where $\phi^{(i,j)}$ denotes the steady state probability for the state (i,j) of the inventory level process, exists and is given by

$$\Phi \tilde{A} = 0 \quad \text{and} \sum_{(i,j)\in E} \sum \phi^{(i,j)} = 1 \tag{1}$$

Theorem 2: The steady state probability Φ is given by



$$\begin{split} \phi^{(i)} &= (-1)^{i} \phi^{(0)} \sum_{m=s_{1}+1}^{s_{1}+2-i} A_{m} B^{-1}, \qquad i = 1, 2, \cdots, s_{1}+1 \\ \phi^{(i)} &= (-1)^{i} \phi^{(0)} \left(\sum_{m=s_{1}+1}^{1} A_{m} B^{-1} \right) (AB^{-1})^{i-s_{1}-1}, \qquad i = s_{1}+2, \cdots, Q_{s_{1}}^{1} \\ \phi^{(i)} &= \phi^{(0)} \left\{ (-1)^{i} \left(\sum_{m=s_{1}+1}^{1} A_{m} B^{-1} \right) (AB^{-1})^{i-s_{1}-1} + \right. \\ & \left. \sum_{l=Q_{s_{1}}^{1}+1}^{i} \sum_{k=0}^{l-Q_{s_{1}}^{1}-1} (-1)^{i-l+1} [\delta_{k0} + (1-\delta_{k0})(-1)^{k} \sum_{m=s_{1}+1}^{s_{1}+2-k} A_{m} B^{-1}] \right. \\ & M_{[l-Q_{s_{1}}^{1}-1-k]} B^{-1} (AB^{-1})^{i-l} \right\} \qquad i = Q_{s_{1}}^{1} + 1, \cdots, Q_{s_{1}}^{1} + r + 1 \\ & \phi^{(i)} &= \phi^{(0)} \left\{ (-1)^{i} \left(\sum_{m=s_{1}+1}^{1} A_{m} B^{-1} \right) (AB^{-1})^{i-s_{1}-1} + \right. \\ & \left. \sum_{l=Q_{s_{1}}^{1}+1}^{Q_{s_{1}}^{1}-1} (-1)^{i-l+1} [\delta_{k0} + (1-\delta_{k0})(-1)^{k} \sum_{m=s_{1}+1}^{s_{1}+2-k} A_{m} B^{-1}] \right. \\ & M_{[l-Q_{s_{1}}^{1}-1-k]} B^{-1} (AB^{-1})^{i-l} + \\ & \left. \sum_{l=Q_{s_{1}}^{1}+r+2} \sum_{k=l-Q_{s_{1}}^{1}-r-1}^{l-l} (-1)^{i-l+k+1} [\sum_{m=s_{1}+1}^{s_{1}+2-k} A_{m} B^{-1}] \right. \\ & M_{[l-Q_{s_{1}}^{1}-1-k]} B^{-1} (AB^{-1})^{i-l} \right\} \qquad i = Q_{s_{1}}^{1} + r + 2, \cdots, S_{1} \end{split}$$

The value of $\phi^{(0)}$ can be obtained from the relation $\sum_{(i,j)\in E} \phi^{(i,j)} = 1$, as

$$\begin{split} \phi^{(0)} &= \left\{ I + \sum_{i=1}^{s_1+1} (-1)^i \left(\sum_{m=s_1+1}^{s_1+2-i} A_m B^{-1} \right) + \sum_{i=s_1+2}^{Q_{s_1}^1} (-1)^i \left(\left(\prod_{m=s_1+1}^{1} A_m B^{-1} \right) (AB^{-1})^{i-s_1-1} \right) + \right. \\ &\left. \sum_{i=Q_{s_1}^1+1}^{Q_{s_1}^1+r+1} \left((-1)^i \left(\prod_{m=s_1+1}^{1} A_m B^{-1} \right) (AB^{-1})^{i-s_1-1} + \right. \right. \\ &\left. \sum_{l=Q_{s_1}^1+1}^{i} \sum_{k=0}^{l-Q_{s_1}^1-1} (-1)^{i-l+1} [\delta_{k0} + (1-\delta_{k0})(-1)^k \sum_{m=s_1+1}^{s_1+2-k} A_m B^{-1}] M_{[l-Q_{s_1}^1-1-k]} B^{-1} (AB^{-1})^{i-l} \right) + \right. \\ &\left. \sum_{i=Q_{s_1}^1+r+2}^{S_1} \left((-1)^i \left(\prod_{m=s_1+1}^{1} A_m B^{-1} \right) (AB^{-1})^{i-s_1-1} + \right. \right. \\ &\left. \sum_{l=Q_{s_1}^1+r+2}^{S_1} \sum_{k=0}^{l-Q_{s_1}^1-1} (-1)^{i-l+1} [\delta_{k0} + (1-\delta_{k0})(-1)^k \sum_{m=s_1+1}^{s_1+2-k} A_m B^{-1}] M_{[l-Q_{s_1}^1-1-k]} B^{-1} (AB^{-1})^{i-l} + \right. \\ &\left. \right. \\ &\left.$$

$$\sum_{l=\mathcal{Q}_{s_{1}}^{l}+r+2}^{i} \sum_{k=l-\mathcal{Q}_{s_{1}}^{l}-r-1}^{l-\mathcal{Q}_{s_{1}}^{l}-1} (-1)^{i-l+k+1} [\prod_{m=s_{1}+1}^{s_{1}+2-k} A_{m}B^{-1}] M_{[l-\mathcal{Q}_{s_{1}}^{l}-1-k]}B^{-1}(AB^{-1})^{i-l} \Bigg) \Bigg\}^{-1}$$

Proof:

The first equation of (1) yields the following set of equations:

$$\begin{split} \phi^{(i)}B + \phi^{(i-1)}A_{s_1-i+2} &= 0, \qquad i = 1, 2, \cdots, s_1 + 1 \\ \phi^{(i)}B + \phi^{(i-1)}A &= 0, \qquad i = s_1 + 2, s_1 + 3, \cdots, Q_{s_1}^1 \\ \phi^{(i)}B + \phi^{(i-1)}A + \sum_{k=0}^{i-Q_{s_1}^1 - 1} \phi^{(k)}M_{[i-Q_{s_1}^1 - 1-k]} &= 0, \qquad i = Q_{s_1}^1 + 1, \cdots, Q_{s_1}^1 + r + 1 \\ \phi^{(i)}B + \phi^{(i-1)}A + \sum_{k=i-Q_{s_1}^1 - 1}^{i-Q_{s_1}^1 - 1} \phi^{(k)}M_{[i-Q_{s_1}^1 - 1-k]} &= 0, \qquad i = Q_{s_1}^1 + r + 2, \cdots, S_1 \\ \text{and} \qquad \phi^{(S_1)}A + \sum_{k=s_1-r}^{s_1} \phi^{(k)}M_{[s_1-k]} &= 0, \end{split}$$

Solving the above set of equations we get the required result.

4 System Performance Measures

In this section, some performance measures of the system are derived under consideration.

4.1 Mean Inventory Level

Let β_1 denote the average inventory level of the commodity 1 in the steady state. Then we have

$$\beta_1 = \sum_{i=1}^{S_1} i \left(\sum_{j=0}^{S_2} \phi^{(i,j)} \right).$$
(2)

Let β_2 denote the average inventory level of the commodity 2 in the steady state. Then we have

$$\beta_2 = \sum_{j=1}^{S_2} j \left(\sum_{i=0}^{S_1} \phi^{(i,j)} \right).$$
(3)

4.2 Mean Reorder Rate

Let β_3 denote the mean reorder rate then we have

$$\beta_{3} = \sum_{k=0}^{r} p_{k} \left(\sum_{i=0}^{s_{1}-k} (\delta_{i0}\lambda_{1} + \lambda_{2})\phi^{(i,s_{2}-k+1)} + \sum_{i=0}^{s_{2}-k} (\lambda_{1} + \delta_{0j}\lambda_{2})\phi^{(s_{1}-k+1,i)} \right)$$
(4)



4.3 Mean Shortage level

Let β_4 denote the mean shortage level, then we have

$$\beta_4 = (\lambda_1 + \lambda_2)\phi^{(0,0)}.$$

(5)

5 Cost Analysis

We assume a specified cost structure for the proposed inventory system as follows:

- k : ordering cost per order.
- h_i : holding cost for the commodity *i* per unit item per unit time.
- c : shortage cost per unit item.

Under the above cost structure, the expected total cost per unit time(expected total cost rate) in the steady state for this model is defined to be

$$TC(S_1, S_2, s_1, s_2, r) = h_1\beta_1 + h_2\beta_2 + k\beta_3 + c\beta_4.$$

Substituting the values for β_i 's we can compute the value of $TC(S_1, S_2, s_1, s_2, r)$.

6 Numerical Illustration

As the expected total cost rate is obtained in a complex form, the convexity of the expected total cost rate cannot be studied analytically. Hence, numerical search procedures are used to find the local optimal values for (S_1, S_2) with fixed (s_1, s_2, r) , s_1 with fixed (S_1, S_2, r) , s_2 with fixed (S_1, S_2, r) and r with fixed (S_1, S_2, s_1, s_2) . With a large number of numerical examples it is found that the expected total cost rate in the long run is either a convex function of both S_1 and S_2 or any one of the variables r and (s_1, s_2) .

Table 1 gives the expected total cost rate as a function of S_1 and S_2 by fixing constant values for the other variables and costs. After obtaining the local optima, S_1^* and S_2^* , the sensitivity analysis are carried out to see how the changes in S_1 and S_2 affect the expected total cost rate(see figure 2). For this the values of

$$\frac{TC(S_1, S_2, 8, 7, 5)}{TC(S_1^*, S_2^*, 8, 7, 5)}$$

by fixing the parameters and costs as $\lambda_1 = 1.5; \lambda_2 = 1.5; p_i = (0.6)(0.4)^i, i = 0, 1, \dots, r-1; p_r = 1 - \sum_{i=0}^{r-1} p_i; \mu_i = 5.2 + i(0.2), i = 0, 1, \dots, r; h_1 = 3.85; h_2 = 3.0; k = 1400; c = 13.2, are computed.$

| | S_2 | 21 | 22 | 23 | 24 | 25 |
|-------|-------|----------|----------|----------|----------|----------|
| S_1 | | | | | | |
| 31 | | 1.004901 | 1.003075 | 1.002683 | 1.003793 | 1.006230 |
| 32 | | 1.003084 | 1.001339 | 1.000775 | 1.001521 | 1.003599 |
| 33 | | 1.002129 | 1.000584 | 1.000000 | 1.000514 | 1.002226 |
| 34 | | 1.001902 | 1.000635 | 1.000150 | 1.000569 | 1.002011 |
| 35 | | 1.002304 | 1.001361 | 1.001060 | 1.001499 | 1.002789 |

Table 1: Sensitivity of S_1 and S_2 on expected total cost rate

Here $S_1^* = 33$ and $S_2^* = 23$ and TC(33, 23, 8, 7, 5) = 181.71435. It appears that the expected total cost rate is more sensitive to the changes in S_2 than S_1 .

Fixing all parameters and other cost values except s_1 and s_2 , the expected total cost rates are computed as shown in tables 2 and 3 respectively. The four curves in figures 3 and 4 correspond to $(S_1, S_2) = (38, 38)$, $(S_1, S_2) = (38, 40)$, $(S_1, S_2) = (40, 40)$ and $(S_1, S_2) = (40, 38)$ represent the different convex functions of s_1 and s_2 respectively.

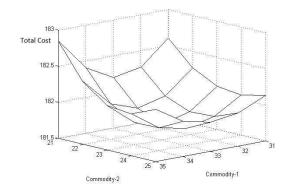


Fig. 2: Effect of S_1 and S_2 on total expected cost rate

| (S_1, S_2) | (38,38) | (38,40) | (40,40) | (40,38) |
|-----------------------|-----------|-----------|-----------|-----------|
| <i>s</i> ₁ | | | | |
| 10 | 97.994613 | 96.139090 | 96.431995 | 98.081569 |
| 11 | 97.994360 | 96.023943 | 96.289144 | 98.065952 |
| 12 | 98.000400 | 95.923640 | 96.151807 | 98.035658 |
| 13 | 98.025336 | 95.843428 | 96.028996 | 98.012370 |
| 14 | 98.075970 | 95.786353 | 95.925645 | 98.008380 |
| 15 | 98.155708 | 95.754412 | 95.844185 | 98.028728 |
| 16 | 98.267247 | 95.749442 | 95.786557 | 98.077060 |
| 17 | 98.412948 | 95.773273 | 95.754432 | 98.155900 |
| 18 | 98.595169 | 95.827835 | 95.749426 | 98.267220 |
| 19 | 98.816480 | 95.915211 | 95.773260 | 98.412886 |

Table 2: Effect of s_1 values on Expected total cost rate

| $(S_1$ | $, S_{2})$ | (38,38) | (38,40) | (40,40) | (40,38) |
|-----------------------|------------|------------|------------|------------|------------|
| <i>s</i> ₂ | | | | | |
| 7 | | 102.231805 | 102.480491 | 101.029476 | 100.907859 |
| 8 | | 102.085301 | 102.077340 | 100.840682 | 100.870128 |
| 9 | | 102.109983 | 101.909026 | 100.778083 | 100.924047 |
| 10 | | 102.253535 | 101.908423 | 100.804334 | 101.044312 |
| 11 | | 102.480068 | 102.023224 | 100.894234 | 101.215964 |
| 12 | | 102.768004 | 102.216954 | 101.032716 | 101.431622 |
| 13 | | 103.106375 | 102.466501 | 101.212020 | 101.688819 |
| 14 | | 103.492655 | 102.763268 | 101.430368 | 101.988307 |
| 15 | | 103.927836 | 103.104920 | 101.688470 | 102.332150 |

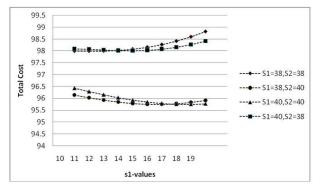
Table 3: Effect of s₂ values on Expected total cost rate

In Table 4 the expected total cost rates by fixing constant values for all variables and costs except *r* are presented. The four curves in figure 5 correspond to $(S_1, S_2) = (45, 47)$, $(S_1, S_2) = (45, 45)$, $(S_1, S_2) = (47, 45)$ and $(S_1, S_2) = (47, 47)$ represent different convex functions of r.

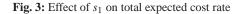
Next the impact of the holding costs h_1 and h_2 on the optimal values (S_1^*, S_2^*) and the corresponding expected total cost rate are studied. For the parameters and the probability distribution $s_1 = 8$; $s_2 = 7r = 5$; $\lambda_1 = 1.5$; $\lambda_2 = 1.5$; $p_i = (0.6)(0.4)^i$, $i = 0, 1, \dots, r-1$; $p_r = 1 - \sum_{i=0}^{r-1} p_i$; $\mu_i = 5.2 + i(0.2)$, $i = 0, 1, \dots, r$; k = 1400; c = 13.2, the total cost rate increases when h_1 and h_2 increase(see table 5). The impact of ordering cost per order and the shortage cost per unit item on the optimal values $(S_1^*, S2^*)$ and the corresponding expected total cost rate are studied by fixing the parameters and the probability distribution: $s_1 = 8$; $s_2 = 7$; r = 5; $\lambda_1 = 1.5$; $\lambda_2 = 1.5$; $p_i = (0.6)(0.4)^i$, $i = 0, 1, \dots, r-1$; $p_r = 1 - \sum_{i=0}^{r-1} p_i$; $\mu_i = 5.2 + i(0.2)$, $i = 0, 1, \dots, r-1$; $p_r = 1 - \sum_{i=0}^{r-1} p_i$; $\mu_i = 5.2 + i(0.2)$, $i = 0, 1, \dots, r-1$; $p_r = 1 - \sum_{i=0}^{r-1} p_i$; $\mu_i = 5.2 + i(0.2)$, $i = 0, 1, \dots, r; h_1 = 3.85$; $h_2 = 3$. Table 6 shows that the total cost rate increases when k and c increase.

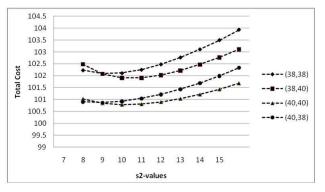
230





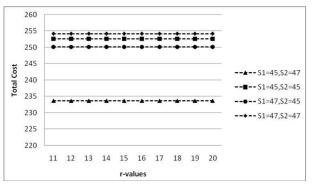
 $\begin{array}{l} r=5;\,s2=12;\,\lambda_{1}=1.9;\,\lambda_{2}=1;\,\,p_{i}=(0.6)(0.4)\overset{i}{,}\underbrace{i=0,1,\ldots}r-1;\,h1=1.5;\\ h2=1;\,p_{r}=1-\sum_{i}^{r-1}p_{i}\;;\,\mu_{i}=5.2+\underbrace{i}(0.2),\,\underbrace{i=0,1,\ldots}r;\,\,k=1800;\,c=13.2 \end{array}$





r=5; s1=12; λ_1 =1; λ_2 =1.9; p_i = (0.6)(0.4); <u>i</u>=0,1,...r-1; h1 = 1; h2 = 1.5; p_r = 1 - $\sum_{r=1}^{r-1} p_r$; μ_i = 5.2 + <u>i</u> (0.2), <u>i</u>=0,1,...r; k = 1800; c = 13.2

Fig. 4: Effect of s₂ on total expected cost rate



s1=22; s2=22; λ_1 =1.5; λ_2 =1.5; $\mathbf{p}_i = (0.6)(0.4)^i$, $\underline{i}=0,1,...$ r-1; h1 = 3.85; h2 = 3; $\mathbf{p}_r = 1 - \sum_{i=1}^{r-1} p_i$; $\mu_i = 5.2 + \underline{i}(0.2)$, $\underline{i}=0,1,...$ r; $\mathbf{k} = 1400$; $\mathbf{c} = 13.2$

Fig. 5: Effect of r on total expected cost rate



| (S_1, S_2) | (45,47) | (45,45) | (47, 45) | (47, 47) |
|--------------|---------------------|---------------------|---------------------|---------------------|
| r | | | | |
| 11 | 233.665264632532280 | 252.592956907586910 | 250.136940405272130 | 254.204713535347680 |
| 12 | 233.665264624049230 | 252.592956903543780 | 250.136940396980090 | 254.204713531245800 |
| 13 | 233.665264622816750 | 252.592956902972840 | 250.136940395784110 | 254.204713530661730 |
| 14 | 233.665264622637580 | 252.592956902892440 | 250.136940395612160 | 254.204713530578770 |
| 15 | 233.665264622611800 | 252.592956902881010 | 250.136940395587940 | 254.204713530567120 |
| 16 | 233.665264622608160 | 252.592956902879620 | 250.136940395584360 | 254.204713530565530 |
| 17 | 233.665264622607680 | 252.592956902879450 | 250.136940395584080 | 254.204713530565240 |
| 18 | 233.665264622607540 | 252.592956902879370 | 250.136940395583990 | 254.204713530565160 |
| 19 | 233.665264622607590 | 252.592956902879420 | 250.136940395584020 | 254.204713530565160 |
| 20 | 233.665264622607590 | 252.592956902879420 | 250.136940395584020 | 254.204713530565270 |

 Table 4: Effect of r on Expected total cost rate

| h_2 | 2.9 | | 3 | .0 | 3 | .1 | 3 | .2 | 3 | .3 |
|-------|----------|----|------|------|------|------|------|------|------|------|
| h_1 | | | | | | | | | | |
| 3.75 | 178.7545 | | 179. | 5042 | 180. | 2539 | 180. | 9377 | 181. | 5953 |
| | 34 2 | 3 | 34 | 23 | 34 | 23 | 34 | 22 | 35 | 21 |
| 3.8 | 179.85 | 34 | 180. | 6229 | 181. | 3726 | 182. | 0617 | 182. | 7397 |
| | 33 2 | 3 | 34 | 23 | 34 | 23 | 34 | 22 | 34 | 22 |
| 3.85 | 180.94 | 00 | 181. | 7144 | 182. | 4887 | 183. | 1857 | 183. | 8637 |
| | 33 2 | 3 | 33 | 23 | 33 | 23 | 34 | 22 | 34 | 22 |
| 3.90 | 182.02 | 66 | 182. | 8010 | 183. | 5753 | 184. | 3097 | 184. | 9877 |
| | 33 2 | 3 | 33 | 23 | 33 | 23 | 34 | 22 | 34 | 22 |
| 3.95 | 183.11 | 32 | 183. | 8876 | 184. | 6620 | 185. | 4032 | 186. | 1021 |
| | 33 2 | .4 | 33 | 23 | 33 | 23 | 33 | 22 | 33 | 22 |

Table 5: Effect of holding costs h_1 and h_2 on optimal values

| С | 13.0 | 13.1 | 13.2 | 13.3 | 13.4 | |
|------|------------|------------|------------|------------|------------|--|
| k | | | | | | |
| 1300 | 176.370606 | 176.370608 | 176.370610 | 176.370612 | 176.370613 | |
| | 33 23 | 33 23 | 33 23 | 33 23 | 33 23 | |
| 1350 | 179.042477 | 179.042479 | 179.042481 | 179.042483 | 179.042485 | |
| | 33 23 | 33 23 | 33 23 | 33 23 | 33 23 | |
| 1400 | 181.714348 | 181.714350 | 181.714352 | 181.714354 | 181.714356 | |
| | 33 23 | 33 23 | 33 23 | 33 23 | 33 23 | |
| 1450 | 184.386219 | 184.386221 | 184.386224 | 184.386226 | 184.386228 | |
| | 33 23 | 33 23 | 33 23 | 33 23 | 33 23 | |
| 1500 | 187.058091 | 187.058093 | 187.058095 | 187.058097 | 187.058099 | |
| | 33 23 | 33 23 | 33 23 | 33 23 | 33 23 | |

Table 6: Effect of ordering cost and shortage cost on optimal values

7 Conclusion

In this paper, a substitutable inventory system of two commodities with reorder levels of band width r has been studied. The joint probability distribution of the inventory levels in the steady state and the stationary measures of system performances have been derived. An example has also been provided to prove the existence of local optima when the total cost function is treated as a function of two variables S_1 and S_2 or a single variable s_1 or s_2 or r. Future work will consider the demand process as renewal type.

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