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Cooperation and game-theoretic cost allocation in stochastic inventory models with continuous review



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ABSTRACT

We study cooperation strategies for companies that continuously review their inventories and face Poisson demand. Our main goal is to analyze stable cost allocations of the joint costs. These are such that any group of companies has lower costs than the individual companies. If such allocations exist they provide an incentive for the companies to cooperate.

We consider two natural cooperation strategies: (i) the companies jointly place an order for replenishment if their joint inventory position reaches a certain reorder level, and (ii) the companies reorder as soon as one of them reaches its reorder level. Numerical experiments for two companies show that the second strategy has the lowest joint costs. Under this strategy, the game-theoretical Shapley value and the distribution rule—a cost allocation in which the companies share the procurement cost and each pays its own holding cost—are shown to be stable cost allocations. These results also hold for situations with three companies.

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1. Introduction

Several companies or business units may have the same item on stock to meet the demands of their customers. Instead of working on their own, the companies may jointly place an order for replenishment of their stocks and save on procurement costs. Consider, for example, the inventory of band-aids in a hospital with several departments. Each department has its own warehouse where it stocks the band-aids. There is central purchasing of replenishments, as is common practice in many Dutch hospitals. The ordered goods arrive in the central warehouse, from which they are distributed to the departments. A major concern is how to organize the central purchasing. Hospital management has several options. First, they could base the purchase decision on the joint inventory of band-aids in all departments. If this joint inventory falls below a certain threshold, a replenishment order for all departments is placed. Second, they could base it on the individual stocks of the departments; if one department runs out of stock, a replenishment order for all departments is placed. Finally, one could decide to allow decentralized replenishment of stock by the individual departments.

The main questions in such joint replenishment problems are how much to order, when to order and how much money can be

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saved. The distribution of these cost savings or of the joint costs among the companies is also of importance. If a company believes it pays too much of the costs, then it may not be willing to cooperate in the first place. Therefore, our main goal is to study allocations of the costs.

This paper studies joint replenishments for multiple companies that continuously review their inventories and face Poisson demand. We consider two natural cooperation strategies. Under the sum constraint strategy the companies jointly place an order for replenishment if their joint inventory position reaches a certain reorder level. Under the individual constraints strategy the companies reorder as soon as one of them reaches its reorder level. We compare the resulting expected costs of these strategies with the costs of non-cooperation. Numerical experiments show that the individual constraints strategy is better, that is, it has the lowest joint costs. Hence, this strategy saves costs compared to non-cooperation. Now, a natural question that arises is how to allocate the saved costs among the companies.

Cooperative game theory is a very suitable tool for studying cost allocations. A natural requirement for such allocations is that they are *stable*. Stability means that the cost is allocated to the companies such that any group of companies pays at most its own cost; the group has no incentive to disagree with the cost allocation. This requirement is represented by the core—a stability concept from cooperative game theory. The core is the set of all stable cost allocations. Often, there are many stable allocations. Then the

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companies should select one of these. Under the individual constraints strategy, we study existence of stable cost allocations in our model.

Two specific cost allocations are the Shapley value (Shapley, 1953), and the distribution rule (Meca et al., 2004). The Shapley value is a cost allocation that distributes marginal contributions to the costs equally among the companies. The distribution rule consists of two parts: (i) the joint procurement costs are allocated in a proportional way to the companies, and (ii) each company pays its individual inventory holding costs under cooperation. Our numerical experiments show that both these cost allocations are stable for the joint replenishment problem with continuous review and Poisson demand under the individual constraints strategy.

There is a large literature on joint replenishment problems. The papers by Arshinder and Deshmukh (2008) and Khouja and Goyal (2008) are excellent surveys on this research subject. Initiated by Balintfy (1964) and Silver (1965), nowadays complex stochastic inventory control problems are studied. Since it is very difficult to determine the optimal policies of these problems, the main focus is on studying the performance of certain classes of replenishment policies. Federgruen et al. (1984) provides an efficient heuristic algorithm to search for an optimal can-order policy in case the companies face compound stochastic demand. Viswanathan (1997) and Chiou et al. (2007) study classes of replenishment policies that lead to low joint costs. More recent, Tanrikulu et al. (2010) introduce a new policy for stochastic joint replenishment problems under continuous review. This policy performs well in case of large backup and order costs. Kiesmüller (2010) considers inventory control when full truckloads are required. Two policies are compared, one depends on the aggregate inventory position, and the other on the individual inventory positions of the items. Most of these papers focus on showing that a certain class of policies leads to lower costs than other well-known policies. However, none of these papers study the allocation of the joint costs to the companies.

There are a few papers that analyse cost allocations for joint replenishment problems by means of cooperative game theory. Hartman and Dror (1996) study allocations of joint costs in inventory models under continuous review. A central warehouse is set up to store the goods and to meet the demand of the customers. In the case of three companies the authors show that there exists a cost allocation that is stable, justifiable (the allocated cost is in line with the cost savings) and computable in polynomial time. More general, Dror et al. (2012) study the computation of stable cost allocations in joint replenishment problems, and the sensitivity of stable cost allocations to the cost parameter values. A review on general game theoretical applications in supply chain management may be found in Leng and Parlar (2005).

In Meca et al. (2004) cooperation in an inventory system with stationary deterministic demand is studied. The companies face procurement and holding costs. Further, the authors introduce the distribution rule, which is a cost allocation designed for joint replenishment problems. The authors show that this cost allocation is stable. In this paper, we extend their model to inventory situations with stochastic (Poisson) demand and integer order quantities, and show that their result also holds for our model. Also related is Chessa (2009), which contains an initial study on our model with some other types of cost allocations.

The outline of this paper is as follows. In Section 2 we introduce our model. Section 3 analyses the costs of non-cooperating companies. Joint replenishment is studied in Section 4, where we consider two natural cooperation strategies. We compare their costs with the cost of non-cooperation. Section 5 studies cost allocations. Finally, Section 6 concludes. All proofs are in Appendix A.

2. Model

We consider the inventory control problem of a single product for multiple companies under continuous review. Our model is an extension of the model of Meca et al. (2004) to Poisson demand. Let N denote the set of companies; we will mainly consider situations with two companies, $N = \{1, 2\}$. The demand for the product at the different companies occurs in discrete units, and the demand processes are independent Poisson processes with rate λ_i for company i, $i \in N$.

To meet their demands the companies place orders for replenishment of their stocks. We assume that the lead time of an order is zero time units and that backorders are not allowed. The replenishment policy for company i is to place an order for Q_i items when its inventory position falls below r_i items. Hence, such a policy is defined by the reorder level r_i and the order quantity Q_i . The state of company i is determined by the inventory position Z_i^t at time t, and let Z_i denote the steady state random variable.

To evaluate the benefits of cooperation, let us specify the cost structure of the companies. We identify procurement costs and inventory holding costs. The procurement costs are the costs associated with procuring (replenishing) the units stocked. We assume that each replenishment order (either by a single company or by multiple cooperating companies) incurs the fixed procurement cost A. There are no minor ordering costs for individual companies. The inventory holding costs are the costs of carrying the items in inventory. Let company i have holding cost h_i per unit in stock per time unit. Then the inventory holding costs equal $h_i Z_i^t$ per time unit when the inventory at time t equals Z_i^t . Since backorders cannot occur, the inventory position of a company is equal to its inventory level.

3. Non-cooperating companies

In this section we consider non-cooperating companies that place their orders independently. The lead time is zero, so that company i uses the following replenishment policy: place an order for an amount Q_i each time the inventory level drops to 0. It is obvious that the inventory level processes $\{Z_i^{nc,t}\}_{i\in N}$ of the companies are independent processes, and that process $Z_i^{nc,t}$ has state space $S_i = \{n_i: 1 \le n_i \le Q_i\}$. For completeness, and to support the more complicated expressions for cooperating companies, we review below the results for a single company.

The marginal inventory level equilibrium distribution is (see e.g., Hadley and Whitin, 1963, p. 183)

$$v_i(j) = \lim_{t \to \infty} P(Z_i^t = j) = \frac{1}{Q_i}, \quad j = 1, \dots, Q_i,$$

$$(1)$$

with expectation $\mathbb{E}Z_i^{nc} = \frac{1}{2}(Q_i + 1)$.

The expected procurement cost per unit time can be obtained from a renewal argument. The inventory level process Z_i^t forms a renewal process that regenerates each time an order is placed. Thus, the long run average procurement costs are

$$\lim_{t \to \infty} \frac{1}{t} A N_i(t) = A \frac{\lambda_i}{O_i},\tag{2}$$

where $N_i(t)$ is the number of replenishment orders in the time interval (0,t]. By the renewal property $\lim_{t\to\infty}N_i(t)/t=1/\mathbb{E}T_i$ with probability 1, where T_i is the cycle time for company i, so that

$$\mathbb{E}T_i = Q_i/\lambda_i. \tag{3}$$

Observe that $v_i(Q_i)^{-1}$ is the mean recurrence time to state Q_i of the Markov jump chain on S_i , so that $\lambda_i^{-1} v_i(Q_i)^{-1}$ is the mean time for the inventory level process to return to state Q_i .

The total expected cost rate $K_i^{nc}(Q_i)$ for company *i* is

$$K_i^{nc}(Q_i) = A \frac{\lambda_i}{Q_i} + h_i \mathbb{E} Z_i^{nc} = \frac{A\lambda_i}{Q_i} + \frac{1}{2} h_i(Q_i + 1). \tag{4}$$

This is a convex function in Q_i . Company i will select an integer quantity that minimizes this expected cost per time unit. This quantity Q_i^{nc} is called the optimal replenishment quantity for company i, and it equals |x| or |x| with $x = \sqrt{2A\lambda_i/h_i}$.

4. Joint replenishment

In this section we consider cooperating companies. Cooperation means that the companies join their orders for replenishment of their inventories; this way they save on procurement costs. We introduce and study two cooperation strategies. We compare their costs, also with the costs of non-cooperation. We first consider two companies, and then extend our results to multiple companies.

Note that the companies may want to keep information about their inventory levels private or they may not communicate with each other. Then joint replenishment may be implemented via an intermediary, who acts as a central purchasing agent. The companies inform the intermediary about their individual inventory levels. Using this information, the intermediary can keep track of the joint inventory level, issue an order when the reorder level is reached, and assign the ordered items to the individual companies.

4.1. Sum constraint strategy

A natural candidate for a replenishment strategy for two companies is that the companies jointly order up to some desired inventory levels (Q_1,Q_2) as soon as their total inventory level falls below a certain level. This happens when the aggregate demand reaches a target level. Without loss of generality assume that $Q_1 \geqslant Q_2$. Let n_i denote the inventory level of company i, and let (n_1,n_2) be the state of the system. A joint order for replenishment is placed when the total inventory level $n_1 + n_2$ drops to Q_1 units, that is, when the joint demand reaches Q_2 units. Note that this ensures nonnegative inventory levels for both companies.

A replenishment order is issued when the total inventory level drops to Q_1 , so that the state space \widetilde{S} is

$$\widetilde{S} = \{(n_1, n_2): n_i \leqslant Q_i, i = 1, 2; n_1 + n_2 \geqslant Q_1 + 1\}.$$
 (5)

We name this strategy the sum constraint strategy as it involves a constraint on the sum of the inventory levels. Notice that this policy requires very little information; only the total inventory level is needed.

The joint inventory level (n_1,n_2) evolves as a continuous time Markov chain at the state space \widetilde{S} . Let $\widetilde{\pi}(n_1,n_2)$ denote the equilibrium probability for state (n_1,n_2) . The flow balance equations ("rate out equals rate in") are

$$\begin{split} &(\lambda_1+\lambda_2)\tilde{\pi}(n_1,n_2)\\ &= \begin{cases} \lambda_1\tilde{\pi}(n_1+1,n_2) + \lambda_2\tilde{\pi}(n_1,n_2+1), & n_1 < Q_1,n_2 < Q_2\\ \lambda_1\tilde{\pi}(n_1+1,n_2), & n_1 < Q_1,n_2 = Q_2,\\ \lambda_2\tilde{\pi}(n_1,n_2+1), & n_1 = Q_1,n_2 < Q_2,\\ &(\lambda_1+\lambda_2)\sum_{n_2'=1}^{Q_2}\tilde{\pi}\big(Q_1+1-n_2',n_2'\big), & n_1 = Q_1,n_2 = Q_2. \end{cases} \end{split} \tag{6}$$

The equilibrium distribution $\tilde{\pi}$ is a truncated binomial distribution:

$$\tilde{\pi}(n_1,n_2) = \frac{1}{Q_2} \binom{Q_1 - n_1 + Q_2 - n_2}{Q_1 - n_1} p^{Q_1 - n_1} (1 - p)^{Q_2 - n_2}, \tag{7}$$

for all $(n_1, n_2) \in \widetilde{S}$, with $p = \lambda_1/(\lambda_1 + \lambda_2)$ the proportion of demand for company 1. This distribution follows directly from substitution into the balance Eq. (6).

Let \widetilde{T} be the joint cycle time under cooperation. The probability that the cycle did not end by time t is $P(\widetilde{T} > t)$. In that case, the total demand is smaller than Q_2 units. Since the joint demand is Poisson with rate $\lambda_1 + \lambda_2$, the expected length of a cycle is

$$\mathbb{E}\widetilde{T} = \int_0^\infty P(\widetilde{T} > t) dt = \int_0^\infty \sum_{k=0}^{Q_2 - 1} \frac{\left((\lambda_1 + \lambda_2) t \right)^k}{k!} e^{-(\lambda_1 + \lambda_2)t} dt = \frac{Q_2}{(\lambda_1 + \lambda_2)}.$$
(8)

The expression for the expected cycle time is natural since it is the expected time until replenishment, that is, until Q_2 demands have occurred. Standard Markov chain theory yields that $\mathbb{E}\widetilde{T}=[(\lambda_1+\lambda_2)\widetilde{\pi}(Q_1,Q_2)]^{-1}$, where $\widetilde{\pi}(Q_1,Q_2)^{-1}$ is the mean recurrence time to state (Q_1,Q_2) of the Markov jump chain with transition probabilities p and 1-p, and $\lambda_1+\lambda_2$ the rate at which jumps occur.

The lemma below gives the joint cost rate for the companies if they use the cooperation strategy under the sum constraint. The proof of this lemma is in Appendix A.

Lemma 1. Consider the cooperation strategy under the sum constraint. The expected joint costs per time unit given order quantities (Q_1, Q_2) equals

$$\begin{split} \widetilde{K}(Q_1,Q_2) &= A \frac{\lambda_1 + \lambda_2}{Q_2} + \frac{1}{2} h_1 (2Q_1 - p(Q_2 - 1)) \\ &+ \frac{1}{2} h_2 (Q_2 + 1 + p(Q_2 - 1)). \end{split} \tag{9}$$

The results can readily be extended to multiple companies. Then all companies jointly reorder when their total inventory level falls below a certain level. Without loss of generality assume that the order quantities are non-increasing: $Q_1 \geqslant Q_2 \geqslant \cdots \geqslant Q_{|N|}$, where |N| denotes the cardinality of N. Let

$$\widetilde{S} = \left\{ n = (n_i)_{i \in N} : \ n_i \leqslant Q_i, \ i \in N; \ \sum_{i \in N} n_i \geqslant \sum_{j=1}^{|N|-1} Q_j + 1 \right\}.$$
 (10)

be the state space. The equilibrium distribution $\tilde{\pi}$ is a truncated multinomial distribution:

$$\tilde{\pi}(n) = \frac{1}{Q_{|N|}} \frac{\left(\sum_{i \in N} (Q_i - n_i)\right)!}{\prod_{i \in N} (Q_i - n_i)!} \prod_{i \in N} p_i^{Q_i - n_i}, \tag{11}$$

where $n \in \widetilde{S}$ and $p_i = \lambda_i / \sum_{j \in N} \lambda_j$ is the proportion of demand for company i. The expected cycle time is

$$\mathbb{E}\widetilde{T} = Q_{|N|} / \sum_{i \in N} \lambda_i, \tag{12}$$

and the average joint cost is

$$\widetilde{K}(Q) = A \frac{\sum_{i \in N} \lambda_i}{Q_{|N|}} + \sum_{i \in N} \frac{h_i}{2} (2Q_i - p_i(Q_{|N|} - 1)), \tag{13}$$

where $Q = (Q_i)_{i \in N}$.

4.2. Individual constraints strategy

In this section another natural replenishment strategy is studied. Namely, both companies reorder as soon as one of the companies reaches its individual reorder level, $n_i = 0$. Hence, this strategy is named joint replenishment under individual constraints. The state space corresponding to this strategy is

$$S = \{(n_1, n_2): 1 \leqslant n_i \leqslant Q_i, i = 1, 2\}. \tag{14}$$

A joint order for replenishment is placed as soon as a company runs out of inventory. The global balance equations are analogous to (6), except for the case $n_1 = Q_1$, $n_2 = Q_2$, where the right-hand side is

 $\lambda_1 \sum_{n'_2=1}^{Q_2} \tilde{\pi}(1, n'_2) + \lambda_2 \sum_{n'_1=1}^{Q_1} \tilde{\pi}(n'_1, 1)$. The corresponding equilibrium distribution π is a truncated binomial distribution:

$$\pi(n_1, n_2) = \frac{1}{G(Q_1, Q_2)} \binom{Q_1 - n_1 + Q_2 - n_2}{Q_1 - n_1} p^{Q_1 - n_1} (1 - p)^{Q_2 - n_2},$$
(15)

for states $(n_1, n_2) \in S$ with normalizing constant

$$G(Q_1, Q_2) = \sum_{z_1 = 0}^{Q_1 - 1Q_2 - 1} \sum_{z_2 = 0}^{Z_1 - 1Q_2 - 1} {z_1 \choose z_1} p^{z_1} (1 - p)^{z_2},$$
(16)

where $z_i = Q_i - n_i$. This is shown by substitution of the equilibrium distribution into the balance equations.

Let us now consider the cycle time T. Let T_1 and T_2 denote the time from a replenishment until a reorder level for companies 1 and 2 if the companies would operate on their own. T_1 and T_2 are independent random variables. If $T_i > t$ then company i did not reach its reorder level by time t, thus, the total demand for company i so far during this cycle is less than Q_i . The expected cycle time is

$$\mathbb{E}T = \int_{0}^{\infty} P(T > t) dt = \int_{0}^{\infty} P(T_{1} > t) P(T_{2} > t) dt$$

$$= \int_{0}^{\infty} \sum_{k=0}^{Q_{1}-1} \frac{(\lambda_{1}t)^{k}}{k!} e^{-\lambda_{1}t} \sum_{\ell=0}^{Q_{2}-1} \frac{(\lambda_{2}t)^{\ell}}{\ell!} e^{-\lambda_{2}t} dt = \frac{G(Q_{1}, Q_{2})}{\lambda_{1} + \lambda_{2}}.$$
(17)

Standard Markov chain theory yields that $\mathbb{E} T = [(\lambda_1 + \lambda_2)\pi(Q_1, Q_2)]^{-1}$, where $\pi(Q_1, Q_2)^{-1}$ is the mean recurrence time to state (Q_1, Q_2) of the Markov jump chain with transition probabilities p and 1-p, and $\lambda_1 + \lambda_2$ the rate at which jumps occur.

Define $P(n_1, n_2)$ as the probability that the system reaches the state (n_1, n_2) in a cycle. Clearly, the regeneration point is reached for sure– $P(Q_1, Q_2)$ = 1—and

$$P(n_1, n_2) = pP(n_1 + 1, n_2) + (1 - p)P(n_1, n_2 + 1).$$
(18)

Hence

$$P(n_1, n_2) = {Q_1 - n_1 + Q_2 - n_2 \choose Q_1 - n_1} p^{Q_1 - n_1} (1 - p)^{Q_2 - n_2},$$
 (19)

for all states $(n_1,n_2) \in S$. Further, define the probability $P(0,n_2)$ as the probability that the cycle terminates from state $(1,n_2)$ due to an arrival of a demand for company 1; $P(0,n_2) = pP(1,n_2)$. Similarly, define $P(n_1,0) = (1-p)P(n_1,1)$. The probability that firm 1 ends a cycle is

$$\sum_{n_1=1}^{Q_2} P(0, n_2) = \sum_{z_2=0}^{Q_2-1} {Q_1 - 1 + z_2 \choose z_2} p^{Q_1} (1-p)^{z_2} = I_p(Q_1, Q_2), \quad (20)$$

where $I_q(a,b) = \sum_{s=0}^{b-1} {s+a-1 \choose s} q^a (1-q)^s$ is the generalized incomplete beta function (Abramowitz and Stegun, 1972, section 26.5). Further,

$$\sum_{n=1}^{Q_1} P(n_1, 0) = \sum_{n=0}^{Q_1-1} {z_1 + Q_2 - 1 \choose z_1} p^{z_1} (1-p)^{Q_2} = I_{1-p}(Q_2, Q_1)$$
 (21)

is the probability that firm 2 ends the cycle. The cycle terminates via a demand for company 1 or 2 and therefore these probabilities sum to 1; this is directly verified from the property $I_q(a,b) + I_{1-q}(b,a) = 1$ of the generalized incomplete beta function.

The probabilities $P(n_1, n_2)$ allow for an alternative formulation of the expected cycle time and the normalizing constant of the equilibrium distribution. The cycle ends because of a demand for company 1 or company 2. If company 2 ends the cycle then the expected cycle length is the expected time it takes for $Q_1 - n_1 + Q_2$ demands to occur times the probability that company 2 triggers the replenishment order while company 1 still has n_1

items on stock. A similar interpretation holds for the case that company 1 ends the cycle. This leads to

$$\mathbb{E}T = \sum_{n_1=1}^{Q_1} \frac{Q_1 - n_1 + Q_2}{\lambda_1 + \lambda_2} P(n_1, 0) + \sum_{n_2=1}^{Q_2} \frac{Q_1 + Q_2 - n_2}{\lambda_1 + \lambda_2} P(0, n_2)$$

$$= \frac{Q_2}{\lambda_2} I_{1-p}(Q_2 + 1, Q_1) + \frac{Q_1}{\lambda_1} I_p(Q_1 + 1, Q_2). \tag{22}$$

By (17)

$$G(Q_1, Q_2) = \frac{Q_2}{1 - p} I_{1-p}(Q_2 + 1, Q_1) + \frac{Q_1}{p} I_p(Q_1 + 1, Q_2)$$
 (23)

is an alternative expression for the normalizing constant. The joint costs for the companies are as follows.

Lemma 2. In case of cooperation under individual constraints, the expected joint costs per time unit given order quantities (Q_1, Q_2) equal

$$\begin{split} K(Q_{1},Q_{2}) &= A(\lambda_{1}+\lambda_{2})/G(Q_{1},Q_{2}) \\ &+ \frac{Q_{2}}{G(Q_{1},Q_{2})} \sum_{z_{1}=0}^{Q_{1}-1} \left[h_{1}(Q_{1}-z_{1}/2) \right. \\ &+ h_{2}(Q_{2}+1)/2 \left[\binom{z_{1}+Q_{2}}{z_{1}} p^{z_{1}} (1-p)^{Q_{2}} \right. \\ &+ \frac{Q_{1}}{G(Q_{1},Q_{2})} \sum_{z_{2}=0}^{Q_{2}-1} \left[h_{1}(Q_{1}+1)/2 \right. \\ &+ h_{2}(Q_{2}-z_{2}/2) \left[\binom{Q_{1}+z_{2}}{z_{2}} p^{Q_{1}} (1-p)^{z_{2}} \right]. \end{split} \tag{24}$$

The companies minimize the costs K by selecting a pair $\left(Q_1^N,Q_2^N\right)$ of optimal integer order quantities, $N=\{1,2\}$.

The result can readily be extended to multiple companies. Now all companies reorder as soon as one of them reaches its re-order level. Let

$$S = \{ n = (n_i)_{i \in \mathbb{N}} : 1 \leqslant n_i \leqslant Q_i, \ i \in \mathbb{N} \}, \tag{25}$$

be the state space. The equilibrium distribution π is a truncated multinomial distribution:

$$\pi(n) = \frac{1}{G(Q)} \frac{\left(\sum_{i \in N} (Q_i - n_i)\right)!}{\prod_{i \in N} (Q_i - n_i)!} \prod_{i \in N} p_i^{Q_i - n_i}, \tag{26}$$

where $n \in S$, and

$$G(Q) = \sum_{n \in S} \frac{\left(\sum_{i \in N} (Q_i - n_i)\right)!}{\prod_{i \in N} (Q_i - n_i)!} \prod_{i \in N} p_i^{Q_i - n_i}.$$
 (27)

The expected cycle time is

$$\mathbb{E}T = G(Q) / \sum_{j \in \mathbb{N}} \lambda_j, \tag{28}$$

and the average joint cost is

$$K_{N}(Q) = A \frac{\sum_{j \in N} \lambda_{j}}{G(Q)} + \sum_{i \in N} \frac{Q_{i}}{G(Q)}$$

$$\times \sum_{j \neq i} \sum_{z_{j}=0}^{Q_{j}-1} \left[\sum_{j \neq i} h_{j}(Q_{j} - z_{j}/2) + h_{i}(Q_{i} + 1)/2 \right]$$

$$\times \frac{(Q_{i} + \sum_{j \neq i} z_{j})!}{Q_{j}! \prod_{i \neq i} z_{i}!} p_{i}^{Q_{i}} \prod_{i} p_{j}^{z_{j}}.$$
(29)

4.3. Comparison of strategies

We now study the costs of the three strategies: non-cooperation, sum constraint and individual constraints. We first consider

two companies and show that the sum constraint strategy is not optimal compared to both independent companies and the individual constraints strategy. Then we numerically investigate the cost structure for two and three companies.

A strategy is said to be better than another one if it has lower expected joint costs.

Theorem 1. We have the following ordering:

- Non-cooperation is better than cooperation under the sum constraint.
- The individual constraints strategy is better than the sum-constraint strategy.
- The optimal order quantity under cooperation with the individual constraints does not exceed the individual optimal order quantity, Q_iⁿ ≤ Q_i^{nc} for any company i.

The first and second item in this theorem state that the sumconstraint strategy does not perform well. This follows from the observation that under the sum-constraint strategy replenishment usually occurs when both companies have an inventory level larger than 1 unit. This implies larger holding costs for the companies than under, for example, non-cooperation.

The cost structure (24) prohibits an analytical comparison of these costs with those for non-cooperation. For identical companies that have equal parameters for holding costs, demand rates, and therefore also for the order quantities, the costs structure simplifies to

$$K(Q,Q) = \frac{A\lambda/Q + hQ}{1 - {2Q \choose Q} {1 \choose \frac{1}{2}}^{2Q}},$$
(30)

where h, λ and Q, denote respectively the common holding costs, demand rates and order quantities. The derivation of this expression is in the appendix.

Invoking Stirling's approximation $\binom{2Q}{Q} \left(\frac{1}{2}\right)^{2Q} \approx \frac{1}{\sqrt{\pi Q}}$, an approximate expression for the joint cost is

$$K(Q,Q) \approx (A\lambda/Q + hQ)/(1 - 1/\sqrt{\pi Q}).$$
 (31)

Taking the derivative with respect to *Q* we obtain that the optimal order quantity is approximated by the solution of

$$2h\sqrt{\pi}Q^{5/2} - 3hQ^2 - 2A\lambda\sqrt{\pi}Q^{1/2} + A\lambda = 0.$$
 (32)

It is straightforward to compare these costs (31) with those under non-cooperation. To this end, note that if the procurement cost A decreases, then replenishing inventory becomes cheaper. Hence, the companies replenish more often and the order quantity per replenishment decreases for both individual firms and cooperating firms. If the procurement cost A is low enough then the order quantities are equal to one unit, $Q^* = 1$ and $Q^{nc} = 1$. In this case, cooperation leads to the same cost as no cooperation: $K(1,1) = 2(A\lambda + h) = K_1(1) + K_2(1)$. This is also true for even lower values of A. Therefore, we determine the largest value of the procurement cost A such that the cost of cooperation is equal to the total individual cost.

Theorem 2. For two identical companies cooperation under individual constraints is better than non-cooperation if and only if $A > h/\lambda$.

Fig. 1 gives the joint costs K(Q,Q) and the total individual costs $K_1(Q) + K_2(Q)$ for parameter values A = 20, $\lambda = 60$, and h = 6. Both cost functions are convex in the order quantity Q. The optimal joint costs (198.7) are lower than the individual optimal costs (246.0); the same relation holds for the optimal order quantities (15 and 20, respectively).

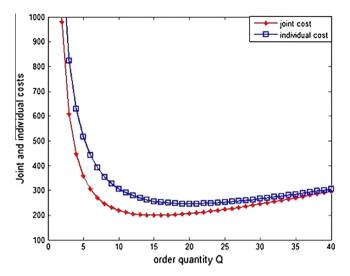


Fig. 1. The joint cost K(Q,Q) and total individual cost $K_1(Q) + K_2(Q)$ as a function of the order quantity Q for A = 20, $\lambda = 60$, and h = 6.

For non-identical companies, the cost structure, as in (24), (29), prohibits an analytical comparison of the joint costs and the total individual costs. Note that in general the optimal strategy may be rather complicated, see e.g., Ignall (1969). The explicit expression for the costs is amenable for numerical comparison. This is a common approach in literature, see e.g., Federgruen et al. (1984). For numerical comparison, we randomly select the problem parameters $(A, (\lambda_i, h_i)_{i \in N})$ from the following ranges similar to those used in the numerical tests in Viswanathan (1997)

$$A \in \{50, 100, \dots, 250\}, \quad \lambda_i \in \{20, 25, \dots, 40\}, \text{ and } h_i \in \{2, 6, 10\}, i = 1, 2.$$

We consider the cost effectiveness, that is the optimal joint cost divided by the total optimal cost under non-cooperation. For two companies (Table 1) and three companies (Table 2) we observe that cooperation under the individual constraints outperforms non-cooperation.

Hence, in all the test instances comparing the numerical optimum of the joint $\cos K$ with the analytical optimum of the individual total $\cos K$ reveals that cooperation outperforms

Table 1The effect of cooperation for two companies

Α	Cost effectivene	ess		
	Average	Minimum	Maximum	
50	0.87	0.81	0.94	
100	0.87	0.80	0.96	
150	0.86	0.80	0.96	
200	0.86	0.79	0.96	
250	0.87	0.79	0.96	

Table 2The effect of cooperation for three companies.

Α	Cost effectivene	ess		
	Average	Minimum	Maximum	
50	0.72	0.69	0.74	
100	0.70	0.66	0.73	
150	0.69	0.65	0.72	
200	0.68	0.65	0.72	
250	0.68	0.65	0.71	

non-cooperation. We further observe that cooperation is more beneficial for more companies. Some further observations from the experiments are as follows.

- (i) The cost function *K* under the individual constraints strategy appears to be convex. Hence, it has a unique minimum.
- (ii) Cooperation under the optimal quantities for non-cooperation yields lower costs: $K(Q_1^{nc}, Q_2^{nc}) \leq K_1(Q_1^{nc}) + K_2(Q_2^{nc})$.
- (iii) Cooperation results in shorter expected cycle times. That is, on average a joint order for replenishment is placed more often than any individual order.
- (iv) Under cooperation the procurement costs per time unit are smaller than under individual optimization. Under cooperation the firms pay A per cycle instead of 2A. On the other hand, the expected cycle time is smaller. Apparently the reduction in procurement costs dominates the decrease in cycle time.
- (v) If the optimal order quantity under cooperation of firm j is the same as the individual optimal order quantity, $Q_j^N = Q_j^{nc}$, then the holding cost of this firm is larger under cooperation than under individual optimization. This has two causes. Under cooperation for each sample path of demands firm j has a weakly larger inventory level with probability 1. This causes a larger average inventory for firm j, so larger holding costs. On the other hand, with positive probability company $k \neq j$ ends the cycle. Then the cycle time is lower than under individual optimization.

5. Cost allocation

We have seen that the individual constraints strategy saves costs. Thus the companies are willing to cooperate. In this section we investigate how to allocate the joint costs among the companies. For this, we use cooperative game theory as a tool. We start with a description of the game theoretic concepts and then proceed with a numerical investigation.

The replenishment game is a cooperative cost game (N,c), see e.g., Peters (2008). N is the player set consisting of the companies. A coalition U of players is a nonempty subset of N. The cost function c assigns to any coalition of players a cost. In this game, the cost of firm i is the minimal cost of cost function $K_i^{nc}(Q_i)$, as defined in (4); $c(\{i\}) = K_i^{nc}(Q_i^{nc})$ for $i \in N$. Let $\left\{Q_i^U\right\}_{i \in U}$ be the optimal order quantities that minimize the cost $K_U(\{Q_i\}_{i \in U})$. The cost of coalition U is the minimal cost of the joint cost function K_U as defined in (29), $K_U\left(\left\{Q_i^U\right\}_{i \in U}\right)$. A game (N,c) is called *concave* if $c(U_1 \cup U_2) + c(U_1 \cap U_2) \leqslant c(U_1) + c(U_2)$ for any coalitions U_1 , U_2 .

An allocation of the joint cost c(N) should be in the core C(N,c) of the game,

$$C(N,c) = \left\{ x \in \mathbb{R}^N : \sum_{i \in N} x_i = c(N), \sum_{i \in T} x_i \leqslant c(U) \text{ for all } U \right\}, \quad (33)$$

if this set is nonempty. Then all coalitions U pay a quantity that is at most equal to their cost c(U). Hence, no coalition wants to deviate from the cooperation within coalition N. Such an allocation is called a *stable* allocation. Since the core often contains more than one allocation, a natural question that arises is which core element to select

We consider two specific cost allocations and check if they belong to the core. The first one is the Shapley value (Shapley, 1953). It is defined as follows. Let σ be a permutation of the players, with player $\sigma(k)$ in position k. The marginal vector $m^{\sigma}(c)$ is a vector that assigns to each player its marginal contribution to the cost for the permutation σ :

$$m^{\sigma}_{\sigma(i)}(c) \ = \begin{cases} c(\{\sigma(i)\}), & i=1, \\ c(\{\sigma(1),\sigma(2),...,\sigma(i)\}) - c(\{\sigma(1),\sigma(2),...,\sigma(i-1)\}), & i>1. \end{cases} \eqno(34)$$

The *Shapley value* $\phi(c)$ is an allocation of the joint cost c(N) such that each player pays its average marginal contribution to the costs:

$$\phi(c) = \frac{1}{|N|!} \sum_{\sigma} m^{\sigma}(c). \tag{35}$$

The distribution rule δ is a cost allocation designed for inventory cost games. This rule is an extension of the distribution rule for deterministic inventory cost games (Meca et al., 2004), which was shown to be stable. The distribution rule consists of two parts. The first part is the distribution of the joint average order costs $A\left(\sum_{j\in N}\lambda_j\right)/G(Q^N)$ among the firms proportional to the square of the individual optimal order costs $\left(A\lambda_i/Q_i^{\rm nc}\right)^2$. The second part is the individual holding cost of each firm as experienced under cooperation; recall (29). The distribution rule allocates

$$\begin{split} \delta_{i} &= \frac{\left(A\lambda_{i}/Q_{i}^{nc}\right)^{2}}{\sum_{j \in N} \left(A\lambda_{j}/Q_{j}^{nc}\right)^{2}} \frac{A\left(\sum_{j \in N}\lambda_{j}\right)}{G(Q^{N})} \\ &+ \sum_{j \neq i} \frac{Q_{j}^{N}}{G(Q^{N})} \sum_{z_{i} = 0}^{Q_{i}^{N} - 1} h_{i}\left(Q_{i}^{N} - z_{i}/2\right) \sum_{k \neq \{i,j\}} \sum_{z_{k} = 0}^{Q_{k}^{N} - 1} \frac{\left(Q_{j}^{N} + \sum_{k \neq j} z_{k}\right)!}{Q_{j}^{N}! \prod_{k \neq j} z_{k}!} p_{j}^{Q_{j}^{N}} \prod_{k \neq j} p_{k}^{z_{k}} \\ &+ \frac{Q_{i}^{N}}{G(Q^{N})} \sum_{j \neq i} \sum_{z_{j} = 0}^{Q_{j}^{N} - 1} h_{i}\left(Q_{i}^{N} + 1\right)/2 \frac{\left(Q_{i}^{N} + \sum_{j \neq i} Z_{j}\right)!}{Q_{i}^{N}! \prod_{j \neq i} Z_{j}!} p_{i}^{Q_{i}^{N}} \prod_{j \neq i} p_{j}^{z_{j}}, \end{split}$$

$$(36)$$

to company i.

Notice that the Shapley value is a general solution for cooperative games, while the distribution rule is tailor-made for inventory situations. Neither is better than the other. A cost allocation is chosen based on its properties.

For two companies, cooperation is better than non-cooperation, $c(N) < c(\{1\}) + c(\{2\})$, as seen in Table 1. Consequently, the two-firm replenishment game is concave and the core

$$C(N,c) = \{x \in \mathbb{R}^2: \ x_1 + x_2 = c(N), \ x_1 \leqslant c(\{1\}), \ x_2 \leqslant c(\{2\})\}$$
 (37)

is a nonempty set. Then the Shapley value is a stable allocation because it belongs to the core of the game (Shapley, 1971). The distribution rule belongs to the core of the game in all test instances, and is therefore stable; see Table 3 for an example.

For three companies, the corresponding replenishment game is a concave game in all test instances, as summarized in Table 2. Further, the Shapley value and the distribution rule are both stable allocations. See Table 4 for an example. Our numerical results indicate the following conjecture.

Conjecture 1. The Shapley value $\phi(c)$ and the distribution rule δ are stable cost allocations for replenishment situations.

These stability results extend those for deterministic inventory situations (Meca et al., 2004) to Poisson demand.

Table 3 The optimal cost, the Shapley value and the distribution rule for two companies with problem parameters A = 200, $\lambda_1 = 20$, $\lambda_2 = 40$, $\lambda_1 = 10$, $\lambda_2 = 10$.

Companies	1	2	1, 2
Optimal cost	287.86	405.00	549.95
Shapley value	216.40	333.55	
Distribution rule	197.98	351.97	

Table 4 The optimal cost, the Shapley value and the distribution rule for three companies with problem parameters A = 250, $\lambda_1 = 25$, $\lambda_2 = 30$, $\lambda_3 = 25$, $h_1 = 10$, $h_2 = 2$, $h_3 = 6$.

Companies	1	2	3	1, 2	1, 3	2, 3	1, 2, 3
Optimal cost	358.57	174.21	276.87	424.78	497.58	350.95	553.26
Shapley value	265.51	100.01	187.74				
Distribution rule	291.30	79.23	182.73				

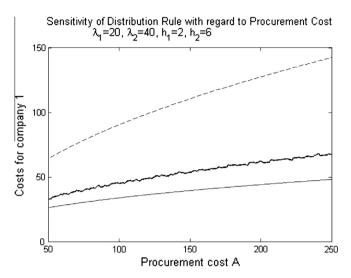


Fig. 2. The cost δ_1 by the distribution rule as a function of the procurement cost A, is represented by the solid line. The dashed lines indicate the upper bound $c(\{1\})$ and lower bound $c(N) - c(\{2\})$ in the core for the cost allocated to company 1.

We also investigate how sensitive the cost allocations are to changes in the parameters in a two-company setting. First, we consider the Shapley value. In a two-company game this value equals $\phi(c) = ((c(\{1\}) - c(\{2\}) + c(N))/2, (c(\{2\}) - c(\{1\}) + c(N))/2)$ by definition. Since the core is the line segment between the two allocations $(c(\{1\}), c(N) - c(\{1\}))$ and $(c(N) - c(\{2\}), c(\{2\}))$, the

Shapley value always lies in the center of the core. This means that the cost savings due to cooperation are shared equally among the companies. Thus, both companies have the same strong incentive to cooperate if the cost is allocated according to the Shapley value.

Next, we analyse the sensitivity of the distribution rule δ . If both companies are identical, then they will order identical quantities. By definition, the distribution rule allocates the same cost to both companies. Hence, the cost allocation lies in the center of the core, and coincides with the Shapley value. This means that the cost savings are shared equally among the companies, so they always have the same strong incentive to cooperate.

Fig. 2 illustrates the sensitivity with regard to the procurement cost *A* if the companies are not identical. In this situation, the allocated cost for company 1 shows an increasing trend, although not monotonically. The reasons for this are as follows. First, the companies can only order integer quantities, hence a discretisation effect occurs. Second, due to the increase in procurement cost, both companies will order larger quantities. However, company 2 has a larger demand rate and his increase in order quantity is larger than for company 1. These effects result in the non-smooth cost allocation for company 1.

Finally, we consider the sensitivity with regard to changes in demand rate and holding cost. In Fig. 3 one sees that the allocation to company 1 by the distribution rule, increases with the demand rate and the holding cost. Also, the position in the core increases slightly from the bottom half of the core to the top half. In all cases, the cost allocation does not get near the bounds of the core. Hence, both companies receive reasonable large parts from the cost savings. Therefore, the incentives for cooperation remain strong.

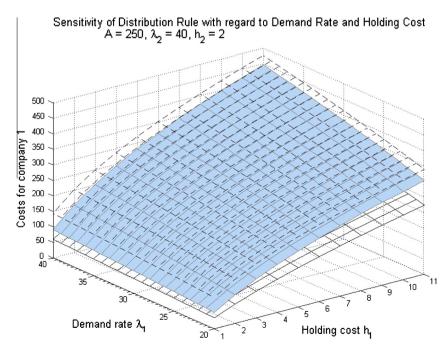


Fig. 3. The cost δ_1 by the distribution rule as a function of the demand rate d_1 and holding cost h_1 , is represented by the solid plane. The plane indicated by the dashed lines illustrates the upper bound $c(\{1\})$ in the core for the cost allocated to company 1. The plane indicated by the solid lines illustrates the lower bound in the core.

We conclude that changes in the parameters have a very small effect on the incentives of the companies to cooperate if the distribution rule is used. These incentives remain strong. In case of cost allocation via the Shapley value, the incentives even remain the same.

6. Conclusions

In this paper we study stable cost allocations for companies that jointly control their inventories. These inventories are reviewed continuously, and the companies face Poisson demand. The paper starts with the analysis of two natural cooperation strategies. First, the sum constraint strategy prescribes the companies to place a joint order for replenishment of their stocks when their joint inventory position reaches a certain reorder level. Second, the individual constraints strategy prescribes the firms to reorder as soon as a company reaches its reorder level. We obtain explicit expressions of the joint costs under cooperation, and of the individual costs.

Comparison of the costs of these two strategies, also with the cost of non-cooperation, shows that the individual constraints strategy has the lowest costs for situations with two or three companies. For two identical companies we characterize when cooperation is beneficial. The numerical experiments show that stable cost allocations of the joint cost exist, namely the distribution rule and the Shapley value are stable allocations for two or three companies. Further, when the parameters change, both cost allocations still provide strong incentives to cooperate for two companies.

In future research we like to extend our model to include minor ordering costs, positive lead time, backorders, batch arrivals at multiple companies, and to study other suitable cost allocations.

Appendix A. Proofs

Proof of Lemma 1. We start by calculating the probability that a demand for one of the companies ends the cycle. Let $P(n_1, n_2)$ be the probability that the system reaches state (n_1, n_2) in a cycle. The regeneration point is reached for sure— $P(Q_1, Q_2) = 1$ —and $P(n_1, -(n_1, n_2) = pP(n_1 + 1, n_2) + (1 - p)P(n_1, n_2 + 1)$. Notice that this equation resembles the first case in (6), hence the solution is

$$P(n_1, n_2) = \binom{Q_1 + Q_2 - n_1 - n_2}{Q_1 - n_1} p^{Q_1 - n_1} (1 - p)^{Q_2 - n_2}. \tag{A.1}$$

Next define the probability $P_1(Q_1 - x, x)$ that a cycle terminates due to a demand for company 1 in state $(Q_1 - x + 1, x)$. Then

$$P_1(Q_1-x,x) = pP(Q_1-x+1,x) = \binom{Q_2-1}{x-1}p^x(1-p)^{Q_2-x},$$
 (A.2)

for $x = 1, ..., Q_2$. Similarly, the probability $P_2(Q_1 - y, y)$ that a cycle terminates due to a demand for company 2 at state $(Q_1 - y, y + 1)$ is

$$P_2(Q_1-y,y) = (1-p)P(Q_1-y,y+1) = \binom{Q_2-1}{y}p^y(1-p)^{Q_2-y}, \tag{A.3}$$

for $y = 0, ..., Q_2 - 1$. The expected holding cost incurred during a cycle that is terminated by company 1 in state $(Q_1 - x, x)$, $x = 1, ..., Q_2$, is the expected cost times the cycle length:

$$[h_1(2Q_1 - x + 1)/2 + h_2(Q_2 + x)/2] \frac{Q_2}{\lambda_1 + \lambda_2}.$$
 (A.4)

Similarly, the expected holding cost during a cycle that is terminated by company 2 in state $(Q_1 - y, y)$, $y = 0, ..., Q_2 - 1$, is

$$[h_1(2Q_1 - y)/2 + h_2(Q_2 + y + 1)/2] \frac{Q_2}{\lambda_1 + \lambda_2}.$$
 (A.5)

Thus the expected holding cost during a cycle terminated by company 1 is

$$\sum_{x=1}^{Q_2} [h_1(2Q_1 - x + 1)/2 + h_2(Q_2 + x)/2] \frac{Q_2}{\lambda_1 + \lambda_2} P_1(Q_1 - x, x)$$
 (A.6)

$$= \frac{Q_2}{\lambda_1 + \lambda_2} \sum_{x=1}^{Q_2} \left[h_1 \left(Q_1 - \frac{x-1}{2} \right) + h_2 \frac{Q_2 + x}{2} \right] {Q_2 - 1 \choose x-1} p^x (1-p)^{Q_2 - x}$$
(A.7)

$$= \frac{Q_2}{\lambda_1 + \lambda_2} p \sum_{y=0}^{Q_2 - 1} \left[h_1 \left(Q_1 - \frac{y}{2} \right) + h_2 \frac{Q_2 + y + 1}{2} \right] {Q_2 - 1 \choose y} p^y (1 - p)^{Q_2 - 1 - y}$$
(A.8)

Let *Y* be the random variable with realisations *y* in this expression. This variable *Y* is binomial distributed with $Q_2 - 1$ trials and probability *p* of success. Therefore we proceed:

$$= \frac{Q_2}{\lambda_1 + \lambda_2} p \mathbb{E} \bigg[h_1 \bigg(Q_1 - \frac{1}{2} Y \bigg) + h_2 \bigg(\frac{1}{2} (Q_2 + 1) + \frac{1}{2} Y \bigg) \bigg] \tag{A.9}$$

$$= \frac{Q_2}{\lambda_1 + \lambda_2} p \left[h_1 \left(Q_1 - \frac{1}{2} p(Q_2 - 1) \right) + h_2 \left(\frac{1}{2} (Q_2 + 1) + \frac{1}{2} p(Q_2 - 1) \right) \right]. \tag{A.10}$$

In a similar fashion we derive the expected holding cost during a cycle terminated by company 2.

$$\sum_{y=0}^{Q_2-1} [h_1(2Q_1-y)/2 + h_2(Q_2+y+1)/2] \frac{Q_2}{\lambda_1+\lambda_2} P_2(Q_1-y,y)$$
(A.11)

$$= \frac{Q_2}{\lambda_1 + \lambda_2} (1 - p) \left[h_1 \left(Q_1 - \frac{1}{2} p(Q_2 - 1) \right) + h_2 \left(\frac{1}{2} (Q_2 + 1) + \frac{1}{2} p(Q_2 - 1) \right) \right]. \tag{A.12}$$

Summarizing, the expected procurement and holding cost per cycle

$$\begin{split} A + & \frac{Q_2}{\lambda_1 + \lambda_2} p \bigg[h_1 \bigg(Q_1 - \frac{1}{2} p(Q_2 - 1) \bigg) + h_2 \bigg(\frac{1}{2} (Q_2 + 1) + \frac{1}{2} p(Q_2 - 1) \bigg) \bigg] \\ + & \frac{Q_2}{\lambda_1 + \lambda_2} (1 - p) \bigg[h_1 \bigg(Q_1 - \frac{1}{2} p(Q_2 - 1) \bigg) + h_2 \bigg(\frac{1}{2} (Q_2 + 1) + \frac{1}{2} p(Q_2 - 1) \bigg) \bigg] \\ = & A + \frac{Q_2}{\lambda_1 + \lambda_2} \bigg[h_1 \bigg(Q_1 - \frac{1}{2} p(Q_2 - 1) \bigg) + h_2 \bigg(\frac{1}{2} (Q_2 + 1) + \frac{1}{2} p(Q_2 - 1) \bigg) \bigg] . \end{split} \tag{A.13}$$

Dividing this by the expected cycle time results in the expected joint cost per time unit. $\ \square$

Proof of Lemma 2. The expected holding cost incurred during a cycle are the holding cost per time unit times the cycle length

$$[h_1(Q_1+n_1)/2+h_2(Q_2+1)/2]\frac{Q_1-n_1+Q_2}{\lambda_1+\lambda_2} \eqno(A.14)$$

if firm 2 ends the cycle while the inventory position of firm 1 is n_1 , and

$$[h_1(Q_1+1)/2 + h_2(Q_2+n_2)/2] \frac{Q_1+Q_2-n_2}{\lambda_1+\lambda_2}$$
 (A.15)

if firm 1 ends the cycle while the inventory position of firm 2 is n_2 . Then the expected joint procurement and holding costs per cycle under cooperation are

$$\begin{split} A + \sum_{n_1=1}^{Q_1} [h_1(Q_1 + n_1)/2 + h_2(Q_2 + 1)/2] \\ \times \frac{Q_1 - n_1 + Q_2}{\lambda_1 + \lambda_2} P(n_1, 0) \\ + \sum_{n_2=1}^{Q_2} [h_1(Q_1 + 1)/2 + h_2(Q_2 + n_2)/2] \frac{Q_1 + Q_2 - n_2}{\lambda_1 + \lambda_2} P(0, n_2) \end{split} \tag{A.16}$$

$$\begin{split} &= A + \frac{Q_2}{\lambda_1 + \lambda_2} \sum_{z_1=0}^{Q_1-1} [h_1(Q_1 - z_1/2) \\ &\quad + h_2(Q_2 + 1)/2] \binom{z_1 + Q_2}{z_1} p^{z_1} (1 - p)^{Q_2} \\ &\quad + \frac{Q_1}{\lambda_1 + \lambda_2} \sum_{z_2=0}^{Q_2-1} [h_1(Q_1 + 1)/2 \\ &\quad + h_2(Q_2 - z_2/2)] \binom{Q_1 + z_2}{z_2} p^{Q_1} (1 - p)^{z_2} \end{split} \tag{A.17}$$

The average cost per time unit K is obtained by dividing this cost by the expected cycle time (17). \Box

Proof of Theorem 1. Let (Q_1^s,Q_2^s) be the optimal order quantities for cooperation under sum constraint. The expected joint cost per time unit (9) and the assumption $Q_1 \geqslant Q_2$ imply that $Q_1^s = Q_2^s$. Let Q^s be this optimal order quantity. Notice that

$$\begin{split} K_{1}(Q^{s}) + K_{2}(Q^{s}) \\ &= A \frac{\lambda_{1} + \lambda_{2}}{Q^{s}} + \frac{1}{2}(h_{1} + h_{2})(Q^{s} + 1) \\ &\leq A \frac{\lambda_{1} + \lambda_{2}}{Q^{s}} + \frac{1}{2}h_{1}((2 - p)Q^{s} + p) + \frac{1}{2}h_{2}((1 + p)Q^{s} + 1 - p)(A.19) \end{split}$$

Together with $K_i(Q_i^{nc}) \le K_i(Q^s)$ for i = 1, 2 this proves that non-cooperation is better than cooperation under the sum constraint.

For the second statement, consider a joint inventory situation with parameters A, λ_1 , λ_2 , h_1 and h_2 . We show that $K(Q^s,Q^s) \leqslant \widetilde{K}(Q^s,Q^s)$, the cost of cooperation under the individual constraints while using the optimal quantities of the sum constraint strategy do not exceed the optimal cost of cooperation under the sum constraint. Then the proof is finished because the strategy (Q^s,Q^s) need not be optimal for cooperation under the individual constraints, $K\left(Q_1^N,Q_2^N\right) \leqslant K(Q^s,Q^s)$.

First, when cooperating under the individual constraints the joint inventory position ranges from $2Q^s$ down to $n_i + 1$ with n_i between 1 and Q^s , for some company i. When the companies cooperate under the sum constraint then the joint inventory position ranges from $2Q^s$ down to $Q^s + 1$. This lower bound is larger than for cooperation under the individual constraints. Therefore, the average inventory position of both companies are larger than for cooperation under the sum constraints.

Second, in case of individual constraints inventory is replenished when the inventory position of one of the firms—say firm i—drops to 0; the joint accumulated demand equals $2Q^s - n_j$, $j \neq i$. In case of the sum constraint, replenishment occurs when the joint inventory position reaches Q^s . Then, the joint accumulated demand equals Q^s . This is lower than in case of individual constraints, $Q^s \leq 2Q^s - n_j$. Therefore, the cycle time is lower than for cooperation under the individual constraints. Together with the first result this implies that both the holding cost per time unit and the order cost per time unit are lower under cooperation with individual constraints. This strategy is better than cooperation under the sum constraint.

For the third statement, without loss of generality consider i = 1. Assume that the order quantity of company 2, Q_2 , is fixed. Then under cooperation there is a positive probability that firm 2 ends the cycle and initiates a new joint order. In that case, the cycle ends before company 1 has reached its reorder level; its inventory position is rather high, leading to rather large holding costs. A lower order quantity would decrease these costs. If under cooperation firm 1 ends the cycle, then the situation is the same as under individual optimization. Hence, the optimum order quantity for company 1 under cooperation does not exceed the individual optimal quantity Q_1^{nc} . \square

Derivation of equation (30). According to (23), $G(Q,Q) = 4QI_{\underline{1}}(Q+1,Q)$. By Lemma 2,

$$\begin{split} K(Q,Q) &= \frac{A\lambda/Q}{2I_{\frac{1}{2}}(Q+1,Q)} + \frac{1}{2I_{\frac{1}{2}}(Q+1,Q)} \\ &\times \sum_{z=0}^{Q-1} [h(3Q+1)/2 - hz/2] \binom{z+Q}{z} \left(\frac{1}{2}\right)^{z+Q}. \end{split} \tag{A.21}$$

The summation in this expression reduces to:

$$\sum_{z=0}^{Q-1} [h(3Q+1)/2 - hz/2] \binom{z+Q}{z} \left(\frac{1}{2}\right)^{z+Q}$$
 (A.22)

$$=h(3Q+1)\sum_{z=0}^{Q-1}\binom{z+Q}{z}\left(\frac{1}{2}\right)^{z+Q+1}$$

$$-h(Q+1)\sum_{z=1}^{Q-1} {z+Q \choose Q+1} \left(\frac{1}{2}\right)^{z+Q+1}$$
 (A.23)

$$=h(3Q+1)I_{\frac{1}{2}}(Q+1,Q)-h(Q+1)I_{\frac{1}{2}}(Q+2,Q-1). \tag{A.24}$$

Therefore,

$$\begin{split} K(Q,Q) &= \frac{A\lambda/Q}{2I_{\frac{1}{2}}(Q+1,Q)} + \frac{h}{2}(3Q+1) - h(Q+1) \\ &\times \frac{I_{\frac{1}{2}}(Q+2,Q-1)}{2I_{\frac{1}{2}}(Q+1,Q)}. \end{split} \tag{A.25}$$

Using the equality $I_{\frac{1}{2}}(Q+2,Q-1) = I_{\frac{1}{2}}(Q+1,Q) - \binom{2Q}{Q+1}\binom{1}{2}^{2Q}$ (Abramowitz and Stegun, 1972, (26.5.15)), we obtain

$$K(Q,Q) = \frac{A\lambda/Q}{2I_{\frac{1}{2}}(Q+1,Q)} + hQ + h(Q+1)\frac{\binom{2Q}{Q+1}^{\binom{1}{2}^{2Q}}}{2I_{\frac{1}{2}}(Q+1,Q)}. \tag{A.26}$$

Finally, because
$$h(Q+1) \binom{2Q}{Q+1} (\frac{1}{2})^{2Q} = hQ \binom{2Q-1}{Q} (\frac{1}{2})^{2Q-1}$$
, and

$$I_{\frac{1}{2}}(Q+1,Q)=1-I_{\frac{1}{2}}(Q,Q+1)=\frac{1}{2}-\left(rac{2Q}{Q}
ight)\!\left(\!rac{1}{2}\!
ight)^{2Q+1}$$
, we conclude

$$K(Q,Q) = \frac{A\lambda/Q}{2I_{\frac{1}{2}}(Q+1,Q)} + hQ\left(1 + \frac{\binom{2Q-1}{Q}^{\left(\frac{1}{2}\right)^{2Q-1}}}{2I_{\frac{1}{2}}(Q+1,Q)}\right)$$
$$= \frac{A\lambda/Q + hQ}{1 - \binom{2Q}{Q}^{\left(\frac{1}{2}\right)^{2Q}}}.$$
 (A.27)

Proof of Theorem 2. By definition of the switch value, cooperation under the individual constraints has lower cost than individual optimization if and only if $A > \overline{A}$. To determine the switch value, we observe the following. If the procurement cost A is slightly larger than \overline{A} , then the optimal order quantities increase to 2

(because they are integer valued). In other words, the switch value is the smallest value of A such that the optimal order quantities are equal to 2, and the costs of cooperation and non-cooperation are the same. By Eq. (30), $K(2,2) = \frac{4}{5}(A\lambda + 4h)$. Also, $K_1(2) + K_2(2) = A\lambda + 3h$. The solution of $K(2,2) = K_1(2) + K_2(2)$ is $A = h/\lambda$. \square

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