



On the distribution-free newsboy problem with some non-skewed demands



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ABSTRACT

We consider three cases of the risk-neutral newsboy problem in which the probability distribution of random demand is only known to be non-skewed with given support, mean and variance. In particular, we derive some closed form formulas for the worst-case and best-case order quantities when this distribution is symmetric, and symmetric and unimodal. Extensions of our results are indicated.

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1. Introduction

The classic single-period, single-item inventory problem with random demand, commonly referred to as the newsboy or newsvendor problem, plays a central role at the conceptual foundations of stochastic inventory theory with vast applications in revenue management and supply chain management [18,4,19]. The problem, formulated and solved by Arrow et al. [2] and Morse and Kimball [14], is as follows. Each day the proverbial newsboy has to decide how many newspapers to stock before observing demand. He purchases them from a publisher at a unit cost c and sells them at a price p to customers whose uncertain demand is described by a random variable X . Any unsold items are recycled with a unit salvage value s ; to avoid trivialities, $p > c > s$ is assumed. The problem is to find the order (purchase) quantity that maximizes the expected profit. Thus, the decision maker is assumed to be risk-neutral; the models in which he/she is risk-averse, risk-seeking, or uses a maximum entropy approach are proposed in [22,8,1]. Numerous extensions of the newsboy problem were reviewed in [19,13].

Since the demand distribution can be hardly known in practice, Scarf [20] was the first to address the distribution-free newsboy problem, that is, the newsboy problem under incomplete probabilistic information. He assumed that merely the mean $\mu = E(X)$ and the variance $\sigma^2 = \text{Var}(X)$ are known, and proved that

whenever $r > \frac{\sigma^2}{\mu^2 + \sigma^2}$, where $r = \frac{p-c}{p-s}$, the order quantity $q^* = \mu + \frac{(2r-1)\sigma}{2\sqrt{r(1-r)}}$ maximizes the minimum expected profit over all distributions with given μ and σ^2 . Thus, Scarf's quantity is the worst-case order quantity. The contribution of Scarf, who is one of the pioneers in inventory theory, was recognized in the 50th anniversary issue of the journal *Operations Research* [21]. The proof of Scarf's formula was simplified and its economic interpretations provided [6]. It was also empirically demonstrated that if the demand distribution is characterized by μ and σ^2 , then Scarf's order quantity performs quite well whenever the demand distribution is assumed to be approximately normal [6,7].

The problem of maximizing the expected profit under the best-case demand scenario has been much less examined [6,23,9]. However, it is more trivial because the best-case order quantity is often $\bar{q}^* = \mu$.

Recently, closed form formulas for the worst-case and best-case order quantities were found when the demand distribution has known support $[a, b]$, mean μ , and variance σ^2 [11]. The main purpose of this paper is to derive these two quantities under the assumption that the distribution is non-skewed, symmetric, or symmetric and unimodal, with given support, mean, and variance.

The paper is organized as follows. In Section 2 we formulate the problem under study, and in Section 3 we present the theoretical background for seeking the order quantities under the worst-case and best case scenarios. These quantities are listed in Section 4. Final remarks, including future research and some extensions of the obtained results, are made in Section 5.

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2. Problem formulation

If q is an order quantity and X denotes the random demand, $\min(X, q)$ represents the demand that is met and $q - \min(X, q)$ is the salvage amount. Consequently, the expected profit is expressed by

$$\begin{aligned} \pi(q) &= pE[\min(X, q)] + sE[q - \min(X, q)] - cq \\ &= (p - s)E[\min(X, q)] - (c - s)q \\ &= (p - s) \int_{-\infty}^{\infty} \min(x, q) dF(x) - (c - s)q, \end{aligned}$$

where F is the cdf (cumulative distribution function) of X .

Since for every cdf F (defined as a right-continuous function) with a finite mean

$$\begin{aligned} \frac{\partial_{\pm}}{\partial q} \left[\int_{-\infty}^{\infty} \min(x, q) dF(x) \right] &= \frac{\partial_{\pm}}{\partial q} \left[q - \int_{-\infty}^q F(x) dx \right] \\ &= 1 - F(q), \end{aligned}$$

the first right derivative of $\pi(q)$ is $\frac{\partial_{+}}{\partial q} [\pi(q)] = (p - c) - (p - s)F(q)$, and the optimal order quantity q^* is typically defined as the smallest q such that $F(q) \geq r$. Note that $q^* = F^{-1}(r)$ whenever X is a continuous random variable.

Suppose only a partial information about the cdf F of X is available in the sense that $F \in \mathcal{F}$, where \mathcal{F} is a non-empty family of cdfs representing some distributions bounded on $[a, b]$. For every $q \in [a, b]$, let $L(q)$ and $U(q)$ be sharp lower and upper bounds on the expected met demand $E[\min(X, q)]$, that is, there exist $\underline{F}_q, \bar{F}_q \in \mathcal{F}$ such that

$$\begin{aligned} L(q) &= \int_a^b \min(x, q) d\underline{F}_q(x) = \min_{F \in \mathcal{F}} \int_a^b \min(x, q) dF(x), \\ U(q) &= \int_a^b \min(x, q) d\bar{F}_q(x) = \max_{F \in \mathcal{F}} \int_a^b \min(x, q) dF(x). \end{aligned}$$

The bounds $L(q)$ and $U(q)$ lead to the following sharp lower and upper bounds on the expected profit $\pi(q)$:

$$\begin{aligned} \underline{\pi}(q) &= (p - s)L(q) - (c - s)q \quad \text{and} \\ \bar{\pi}(q) &= (p - s)U(q) - (c - s)q. \end{aligned}$$

When $\underline{\pi}(q)$ and $\bar{\pi}(q)$ are maximized over q , one can find the worst-case and best-case order quantities denoted by \underline{q}^* and \bar{q}^* , respectively. Consequently, for any $F \in \mathcal{F}$, the corresponding maximum expected profit $\pi(q^*)$ satisfies $\underline{\pi}(\underline{q}^*) \leq \pi(q^*) \leq \bar{\pi}(\bar{q}^*)$.

In this paper we assume that the mean demand is $\mu = (a+b)/2$, that is, the allowable distributions are on the interval $[\mu - d, \mu + d]$, where $0 < d \leq \mu$. Under this assumption, we consider the cases when these distributions are additionally non-skewed, symmetric, and symmetric and unimodal.

3. Theoretical background

As it was observed in the previous section, for a given non-empty family \mathcal{F} of cdfs on $[a, b]$, to find the worst-case and best-case order quantities, \underline{q}^* and \bar{q}^* , it suffices to identify sharp bounds, $L(q)$ and $U(q)$, on the expected met demand $E[\min(X, q)]$. If $L(q)$ ($U(q)$) is concave, then $\underline{F}(q) = 1 - \frac{\partial_{+}L(q)}{\partial q}$ ($\bar{F}(q) = 1 - \frac{\partial_{+}U(q)}{\partial q}$) is the infimum (supremum) of \mathcal{F} with respect to the increasing concave order, and \underline{q}^* (\bar{q}^*) is the smallest q such that $\underline{F}(q) \geq r$ ($\bar{F}(q) \geq r$) [11]. Recall here that F is said to be smaller than G in the sense of this order, written $F \preceq_{icv} G$, if for every non-decreasing and concave function $\varphi(x)$, $\int_a^b \varphi(x) dF(x) \leq \int_a^b \varphi(x) dG(x)$. Furthermore,

$F \preceq_{icv} G$ is equivalent to $\int_a^b \min(x, q) dF(x) \leq \int_a^b \min(x, q) dG(x)$ for every $q \in [a, b]$. A cdf \underline{F} (\bar{F}) is called the infimum (supremum) of \mathcal{F} with respect to \preceq_{icv} if \underline{F} (\bar{F}) is the greatest (smallest) cdf, not necessarily in \mathcal{F} , such that $\underline{F} \preceq_{icv} F$ ($F \preceq_{icv} \bar{F}$) for all $F \in \mathcal{F}$ [15].

Let δ_x and $U_{\mu \mp x}$ denote the cdfs of the one-point (degenerate) distribution at x , and the uniform distribution on $[\mu - x, \mu + x]$, respectively. It is well-known that if \mathcal{F} represents all distributions (unimodal distributions) on $[\mu - d, \mu + d]$, then $\frac{1}{2}\delta_{\mu-d} + \frac{1}{2}\delta_{\mu+d}$ ($U_{\mu \mp d}$) and δ_{μ} are the minimum and maximum of \mathcal{F} . Since $\frac{1}{2}\delta_{\mu-d} + \frac{1}{2}\delta_{\mu+d}$ and $U_{\mu \mp d}$ have variances d^2 and $d^2/3$, respectively, the following is true; see e.g. [5,3].

Lemma 1. *Let X have a distribution on $[\mu - d, \mu + d]$. Then its variance σ^2 satisfies $\sigma^2 \leq d^2$, and the bound d^2 remains sharp when the distribution is symmetric. If it is also unimodal, then $\sigma^2 \leq d^2/3$.*

Lemma 1 will clarify the assumptions imposed on the variance in the remainder of this section.

First, we assume that the demand distributions on $[\mu - d, \mu + d]$ (with mean μ) are non-skewed, that is, $E[(X - \mu)^3] = 0$. The next lemma can be deduced from Theorem 2.1 in the excellent monograph of Karlin and Studden [12, p. 472]; see also Theorem 3.1 [17] and Theorem 2 [11].

Lemma 2. *Let X have a non-skewed distribution on $[\mu - d, \mu + d]$ with variance $\sigma^2 < d^2$. Then for every $q \in [\mu - d, \mu + d]$, the sharp lower and upper bounds, $L(q)$ and $U(q)$, on $E[\min(X, q)]$ can be defined as follows:*

$$\begin{aligned} L(q) &= \max_{c_i} [c_0 + \mu c_1 + (\mu^2 + \sigma^2) c_2 + (\mu^3 + 3\mu\sigma^2) c_3] \\ \text{s.t. } c_0 + c_1x + c_2x^2 + c_3x^3 &\leq \min(x, q) \quad \text{for } x \in [\mu - d, \mu + d]; \\ U(q) &= \min_{c_i} [c_0 + \mu c_1 + (\mu^2 + \sigma^2) c_2 + (\mu^3 + 3\mu\sigma^2) c_3] \\ \text{s.t. } c_0 + c_1x + c_2x^2 + c_3x^3 &\geq \min(x, q) \quad \text{for } x \in [\mu - d, \mu + d]. \end{aligned}$$

Theorem 1. *Let the cdf of X belong to the family \mathcal{F} representing the non-skewed distributions on $[\mu - d, \mu + d]$ with variance $\sigma^2 < d^2$. Then the sharp lower and upper bounds on $E[\min(X, q)]$ are:*

$$L(q) = \begin{cases} \min(\mu, q) - \frac{(d - |\mu - q|)\sigma^2}{2d^2} & \text{if } \frac{3d}{5} \leq |\mu - q| \leq d, \\ \min(\mu, q) - \frac{(z - |\mu - q|)(d^2 - \sigma^2)\sigma^2}{(z + d)(dz^2 - 2\sigma^2z + d\sigma^2)} & \\ \text{if } \frac{\sigma^2}{d + 2\mu} \leq |\mu - q| \leq \frac{3d}{5}, & \\ \frac{q + \mu - \sigma}{2} & \text{if } |\mu - q| \leq \frac{\sigma^2}{d + 2\sigma}, \end{cases}$$

where $z = y^* - d$,

$$y^* = \begin{cases} \frac{1}{6d} \left[B_1 + B_2 + 2(B_1 - B_2) \cos\left(\frac{1}{3} \arccos(1 + R)\right) \right] & \\ \text{if } |\mu - q| \geq e - d, & \\ \frac{1}{6d} \left[B_1 + B_2 + 2(B_1 - B_2) \cosh\left(\frac{1}{3} \operatorname{arcosh}(1 + R)\right) \right] & \\ \text{otherwise,} & \end{cases}$$

$$B_1 = 3d(|\mu - q| + d), \quad B_2 = 2(d^2 + \sigma^2),$$

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