

## Constrained Probabilistic Economic Order Quantity Model under Varying Order Cost and Zero Lead Time Via Geometric Programming

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**Abstract: Problem statement:** In this study, we provide a simple method to determine the inventory policy of probabilistic single-item Economic Order Quantity (EOQ) model, that has varying order cost and zero lead time. The model is restricted to the expected holding cost and the expected available limited storage space. **Approach:** The annual expected total cost is composed of three components (expected purchase cost, expected ordering cost and expected holding cost. The problem is then solved using a modified Geometric Programming method (GP). **Results:** Using the annual expected total cost to determine the optimal solutions, number of periods, maximum inventory level and minimum expected total cost per period. A classical model is derived and numerical example is solved to confirm the model. **Conclusion/Recommendations:** The results indicated the total cost decreased with changes in optimal solutions. Possible future extension of this model was include continuous decreasing ordering function of the number of periods and introducing expected annual demand rate as a decision variable.

**Key words:** Inventory model, holding costs, storage area, lead time, geometric programming, Economic Order Quantity (EOQ), limited storage space, probabilistic single-item, varying order

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### INTRODUCTION

The simple EOQ model is the most fundamental of all inventory models. It is assumed that the expected order cost and demand rate are constants. Fabrycky and Banks (1967) studied some probabilistic models of the case, where both demand and procurement lead time are identically and independently random variables distributed. Discussed a simple method for determining order quantities in joint replenishment of deterministic demand. Unconstrained probabilistic inventory problem with constant cost units has been treated. Ben-Daya *et al.* (2006) presented integrated inventory control and inspection policies with deterministic demand. Also, Abou-El-Ata and Kotb (1997) developed a crisp inventory model under two restrictions. Teng and Yang (2007) studied deterministic inventory lot-size models with time-varying demand and cost under generalized holding costs. Other related studies are presented by Hadly and Whitin (1963); Cheng (1989); Jung and Klein (2001); Das *et al.* (2000) and Mandal *et al.* (2006). An optimal inventory policy for items having

linear demand and variable deterioration rate with trade credit has been discussed by Sarbjit and Raj (2010). Recently, EL-Sodany (2011) presented periodic review probabilistic inventory system with zero lead time under constraint and varying holding cost. Also, Kotb and Fergany (2011) discussed multi-item EOQ model with varying holding cost: a geometric programming approach.

In this study, we have proposed constrained probabilistic single-item EOQ model with varying order cost and zero lead time. The varying order cost is continually increasing function of number of periods per inventory cycle. The constraints are proposed to be the expected holding cost and the expected available limited storage space. The optimal number of periods, the optimal maximum inventory level and the minimum expected total cost per period are obtained using a modified geometric programming method. Finally, a numerical example is used to confirm the results.

**Assumptions and notations:** The following assumptions are made for developing the model:

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- Demand rate is random variable having a known probability distribution
- Lead time is zero
- Shortages are not allowed
- Review of the stock level is made every  $N$  period
- Ordering cost  $C_o(N) = \alpha + \beta N$ ,  $\alpha > 0, \beta \geq 0$  is continuous increasing function of the number of periods. Where  $\alpha$  and  $\beta$  are real constants selected to provide the best fit of the estimated cost function
- The minimization of the expected total cost is the objective

In addition, the following notation is adopted for developing the model:

- $C_h$  = Holding cost.
- $C_p$  = Purchase cost.
- $C_o$  = Ordering cost.
- $C_o(N)$  = Varying order cost per period.
- $\bar{D}$  = Expected annual demand rate.
- $K_1$  = Limitation on the expected holding cost.
- $K_2$  = Limitation on the storage area.
- $N$  = Number of periods.
- $N^*$  = Optimal number of periods.
- $Q_m$  = Maximum inventory level.
- $Q_m^*$  = Optimal maximum inventory level.
- $S$  = Available storage area.
- $\overline{TC}$  = Expected total cost.

**Model formulation and analysis:** The annual expected total cost is composed of three components (expected purchase cost, expected ordering cost and expected holding cost) according to the basic assumptions and notation of the EOQ model provided by Eq. 1 Fabrycky and Banks (1967):

$$\overline{TC} = C_p \bar{D} + \frac{C_o(N)}{N} + \frac{C_h \bar{D} [2v + N]}{2} \tag{1}$$

The restrictions on the expected holding cost and the expected storage area are the following two conditions Eq. 2:

$$\frac{C_h \bar{D} N}{2} \leq K_1 \text{ and } \bar{S} D N \leq K_2 \tag{2}$$

In order to solve this primal function which is a convex programming problem, it can be rewritten in the following form Eq. 3 and 4:

$$\min \overline{TC} = C_p \bar{D} + \frac{\alpha}{N} + \beta + C_h \bar{D} v + \frac{C_h \bar{D} N}{2} \tag{3}$$

Subject to:

$$\frac{C_h \bar{D} N}{2K_1} \leq 1 \text{ and } \frac{\bar{S} D N}{K_2} \leq 1 \tag{4}$$

The term  $C_p \bar{D} + \beta + C_h \bar{D} v$  is constant and hence can be ignored.

Applying Duffin *et al.* (1967) results of geometric programming technique on (3) and (4), the enlarged pre-dual function can be written in the form Eq. 5:

$$G(\underline{W}) = \left( \frac{\alpha}{N W_1} \right)^{w_1} \left( \frac{C_h \bar{D} N}{2 W_2} \right)^{w_2} \left( \frac{C_h \bar{D} N}{2 K_1 W_3} \right)^{w_3} \left( \frac{\bar{S} D N}{K_2 W_4} \right)^{w_4} \\ = \left( \frac{\alpha}{W_1} \right)^{w_1} \left( \frac{C_h \bar{D}}{2 W_2} \right)^{w_2} \left( \frac{C_h \bar{D}}{2 K_1 W_3} \right)^{w_3} \left( \frac{\bar{S} D}{K_2 W_4} \right)^{w_4} \\ \times N^{-w_1 + w_2 + w_3 + w_4}$$

where,  $\underline{W} = W_j, J = 1, 2, 3, 4 (0 < W_j < 1)$  are the weights and could be easily deduced from Equation 5 through the use of the following normal and orthogonal conditions Eq. 6:

$$\left. \begin{aligned} W_1 + W_2 &= 1 \\ \text{and} \\ -W_1 + W_2 + W_3 + W_4 &= 0 \end{aligned} \right\} \tag{6}$$

These are two linear equations in four unknowns having an infinite number of solutions. However, the problem is to select the optimal solution of the weights  $W_j^*, 0 < W_j^* < 1, J = 1, 2, 3, 4$ .

By solving Eq. 6, we have Eq. 7:

$$\left. \begin{aligned} W_1 &= \frac{1 + W_3 + W_4}{2} \\ \text{and} \\ W_2 &= \frac{1 - W_3 - W_4}{2} \end{aligned} \right\} \tag{7}$$

Substituting  $W_1$  and  $W_2$  in Eq. 5, then the dual function is Eq. 8:

$$g(W_3, W_4) = \left( \frac{2\alpha}{1 + W_3 + W_4} \right)^{\frac{1+W_3+W_4}{2}} \left( \frac{C_h \bar{D}}{1 - W_3 - W_4} \right)^{\frac{1+W_3+W_4}{2}} \quad (8)$$

$$\left( \frac{C_h \bar{D}}{2K_1 W_3} \right)^{W_3} \left( \frac{S\bar{D}}{K_2 W_4} \right)^{W_4}$$

In order to find the optimal  $W_3$  and  $W_4$  which maximize  $g(W_3, W_4)$ , the logarithm of both side of Eq. 8 and the partial derivatives were taken relative to  $W_3$  and  $W_4$ , respectively. Setting each of them to equal zero and simplifying, we get Eq. 9 and 10:

$$\left( \frac{2\alpha}{C_h \bar{D}} \right) \left( \frac{1 - W_3 - W_4}{1 + W_3 + W_4} \right) \left( \frac{C_h \bar{D}}{2eK_1 W_3} \right)^2 = 1 \quad (9)$$

and:

$$\left( \frac{2\alpha}{C_h \bar{D}} \right) \left( \frac{1 - W_3 - W_4}{1 + W_3 + W_4} \right) \left( \frac{S\bar{D}}{eK_2 W_4} \right)^2 = 1 \quad (10)$$

Multiplying relation (9) by the inverse of relation (10), we find Eq. 11:

$$\frac{W_3}{W_4} = \frac{C_h K_2}{2SK_1} \quad (11)$$

Substituting  $W_3$  and  $W_4$  into relations (9) and (10), respectively, we have Eq. 12:

$$f_i(W_j) = W_j^3 + C_i W_j^2 + B_i W_j - B_i C_i = 0, j=3,4, i=1,2 \quad (12)$$

Where:

$$B_1 = \frac{\alpha C_h \bar{D}}{2e^2 K_1^2}, B_2 = \frac{2\alpha S \bar{D}}{e^2 C_h K_2^2}$$

$$C_1 = \frac{C_h K_2}{C_h K_2 + 2SK_1} \quad C_2 = \frac{2SK_1}{C_h K_2 + 2SK_1}$$

It is clear that  $f_i(0) < 0$  and  $f_i(1) > 0, i=1,2$ , which means that there exists a root  $W_j \in (0,1), j = 3,4$ . The trial and error approach can be used to find these roots. However, we shall first verify any root  $W_j^*$ ,  $j = 3,4$  calculated from Eq. 12 to maximize  $f_i(W_j), i = 1,2, j = 3,4$ , respectively. This was confirmed by the second

derivative to  $\ln g(W_3, W_4)$  with respect to  $W_3$  and  $W_4$ , respectively, which is always negative.

Thus, the roots  $W_3^*$  and  $W_4^*$  calculated from Eq. 12 maximize the dual function  $g(W_3, W_4)$ . Hence, the optimal solutions are  $W_3^*$  and  $W_4^*$  of Eq. 12, respectively.  $W_1^*$  and  $W_2^*$  are evaluated by substituting the value of  $W_3^*$  and  $W_4^*$  in expression (7).

To find the optimal number of periods  $N^*$  and the optimal maximum inventory level  $Q_m^*$ , we applied the results of Duffin *et al.* (1967) for geometric programming as indicated below:

$$\frac{\alpha}{N^*} = W_1^* g(W_3^*, W_4^*) \quad \text{and} \quad \frac{C_{h\bar{D}N^*}}{2} = W_2^* g(W_3^*, W_4^*)$$

By solving these relations, the optimal number of periods is given by Eq. 13:

$$N^* = \sqrt{\frac{2\alpha W_2^*}{C_h \bar{D} W_1^*}} = \sqrt{\frac{2\alpha(1 - W_3^* - W_4^*)}{C_h \bar{D}(1 + W_3^* + W_4^*)}} \quad (13)$$

and the optimal maximum inventory level  $Q_m^*$  is Eq. 14:

$$Q_m^* = \bar{D} N^* g(N^*) = \bar{D} v + \sqrt{\frac{2\alpha \bar{D}(1 - W_3^* - W_4^*)}{C_h(1 + W_3^* + W_4^*)}} \quad (14)$$

By substituting the value of  $N^*$  in relation (3), we get the minimum expected total cost as Eq. 15:

$$\min \bar{TC} = \beta + (C_p + vC_h)\bar{D} + \sqrt{\frac{\alpha C_h \bar{D}}{2W_1^* W_2^*}} \quad (15)$$

As a special case, we assume  $\beta = 0$  and  $k_i \rightarrow \infty \Rightarrow C_o(N), i = 1, 2$ . This is the probabilistic single-item inventory model with constant order cost and without any restrictions, which is consistent with the results of Fabrycky and Banks (1967).

**An illustrative example:** The decision variables (the optimal number of periods  $N^*$  and the optimal maximum inventory level  $Q_m^*$ ) should be computed to minimize the annual relevant expected total cost. Assume the parameters of the inventory model as:

$$C_o = \$1 \text{ per procurement,}$$

$$C_h = \$0.05 \text{ per unit per period, } C_p = \$25 \text{ per unit}$$

Table 1: The optimal results of different values of  $\alpha$  and  $\beta$

$\beta$	$C_o(N^*)$					$\min \overline{TC}$					$N^*$	$Q_m^*$
	$\alpha$	0	10	20	50	100	0	10	20	50		
1	1	27.347	53.694	132.737	264.473	50.811	60.811	70.811	100.811	150.811	2.6340	11.269
2	2	34.095	66.190	162.477	322.954	51.083	61.083	71.083	101.083	151.083	3.2090	12.419
5	5	44.613	84.227	203.068	401.135	51.760	61.760	71.760	101.760	151.760	3.9610	13.922
8	8	51.066	94.133	223.332	438.665	52.372	62.372	72.372	102.372	152.372	4.3060	14.613
10	10	54.546	99.093	232.733	455.466	52.767	62.767	72.767	102.768	152.768	4.4540	14.909
15	15	61.925	108.851	249.629	484.257	53.731	63.731	73.731	103.731	153.731	4.6920	15.385
30	30	80.009	130.020	280.049	530.098	56.548	66.548	76.548	106.549	156.549	5.0009	16.002
50	50	101.546	153.092	307.731	565.462	60.258	70.257	80.257	110.258	160.258	5.1540	16.309
100	100	152.862	205.724	364.310	628.621	69.481	79.481	89.481	119.481	169.481	5.2860	16.572
200	200	253.585	307.171	467.927	735.853	87.891	97.891	107.892	137.892	187.892	5.3580	16.717
500	500	554.044	608.087	770.218	1040.440	143.088	153.088	163.088	193.088	243.088	5.4040	16.808

$\overline{D} = 2$  unit per period,  $S = 50$  cubicunit per item,  $v = 3$

$K_1 = \$1000$  per unit and  $K_2 = 200$  cubicunits of space

The Optimal results of different values of  $\alpha$  and  $\beta$  are shown in Table 1.

**CONCLUSION**

This work investigated how ordering cost function, two constraints and geometric programming approach affect the probabilistic EOQ model. Ordering cost function was assumed to depend on number of periods. In addition, the constraints were expected holding cost and expected available limited storage space. A geometric programming approach was devised to determine the optimal solution for probabilistic EOQ, number of periods, maximum inventory level and minimum expected total cost per period instead of the traditional Lagrangian method. Finally, a classical model is derived and numerical example is solved to confirm the model. The results indicated that the total cost decreased with changes in  $\alpha, \beta, N^*$  and  $Q_m^*$ . Possible future extension of this work was include continuous decreasing ordering function of the number of periods and introducing expected annual demand rate as a decision variable.

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