

# Deteriorating Inventory Model When Demand Depends on Advertisement and Stock Display

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**Abstract**—In this study, a mathematical model is developed to obtain optimal ordering policy of time dependent deteriorating item when demand rate is dependent on displayed stock level and frequency of advertisement through media. Shortages are not allowed. The objective is to minimize total cost. The significant features and the results are studied with the help of a numerical example. The effect of changes in the demand parameter, deterioration rate ( $\alpha$ -constant deterioration,  $\beta$ - time dependent deterioration), rate of frequency of advertisements, stock dependent parameter and salvage parameter for deteriorated items on total cycle time, total cost and on procurement quantity is studied numerically.

**Keyword**—Lot-size, Time dependent deterioration, Advertisement frequency, Procurement quantity and total cost.

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## 1. INTRODUCTION

Classical inventory Economic order quantity (EOQ) model is based on the assumption that an item in stock remains to its 100 % efficiency for infinite time. Actually items like volatile liquids, blood, X – ray plates, medicines, electronic components, fashion goods, fruits and vegetable etc. loses their utility after some time. The most of the most of the articles were based on the fact that these deteriorated units are complete loss to the inventory system. See review articles by Raafat (1991), Shah and Shah (2000), Goyal and Giri (2001).

These days' advertisements and display of products plays a vital role in attracting mass customer. The advertisement through electronic media, news paper, internet, using innovative ways of display of product in the show room is the best tool for the promotion of a product. This attracted researcher to analyze the inventory problem when demand depends on stock displayed. Refer to Baker and Urban (1988), Mandal and Phaujdar (1989), Datta and Pal (1990), Padmanabhan and Vrat (1996), Giri et al. (1996), Sarkar et al. (1997) etc. The effect of advertisement on the demand of the product is studied by Bhunia and Maiti (1997), Goyal and Gunasekaran (1995), Pal et al. (1996, 2006) developed an inventory model for deteriorating items by taking into account the impact of marketing strategies viz pricing, advertisement and the displayed stock level of the demand rate of the system. These articles optimize total cost or net profit per time unit of an inventory system.

Misra (1979 a) studied the inventory decisions under inflationary conditions for EOQ model. Misra (1979 b) derived an inflation model for the EOQ, in which the time value of money and different inflation rates were considered. Gurunani (1983) gave the economic analysis of inventory systems and claimed that the discounting effects on EOQ were substantial. Related articles are by Queyranne (1985), Roundy (1986), Federgruen et al. (1992), Shah et al. (2003, 2004)

In this paper, a mathematical inventory model for time dependent deteriorating items is developed by considering demand to be function of advertisement and the displayed stock level. Shortages are not allowed. The storage capacity of the inventory system is finite. The objective is to minimize total cost.

## 2. ASSUMPTIONS AND NOTATIONS

The Mathematical model is derived using the following assumptions and notations

### 2.1 Notations

$T$  : Cycle time (decision variable)

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$t_0$  : time point when the stock level is  $S_1$

$t_1$  : time point when the stock level is  $S_0$  ( $S_1 > S_0$ )

$C$  : purchase cost per unit item

$b$  : inventory holding cost per unit per unit time

$G$  : cost of advertisement

$O$  : ordering cost per order

$\lambda$  : rate of change of frequency of advertisement

$\gamma$  : salvage parameter for deteriorated units

$A$  : frequency of advertisement in the cycle

$a$  : fixed demand

$b$  : rate of change of demand

$q(t)$  : inventory level at any instant of time  $t$  during cycle time

$\theta(t)$  :  $\theta(t) = \alpha\beta t^{\beta-1}$  ( Weibull distribution )

where  $\alpha > 0$  ,  $\beta > 0$  . Here  $\beta > 1$  is considered which means deterioration increases with time.

$S$  : order quantity per cycle

$K$  : total cost per time unit.

## 2.2 Assumptions

1. The inventory system deals with a single item.
2. Replenishment rate is infinite.
3. Shortages are not allowed and lead time is zero or negligible.
4. The planning horizon is finite.
5. The deterioration rate of units in inventory follows the Weibull distribution function given by  $\theta(t) = \alpha\beta t^{\beta-1}$  where  $\alpha > 0$  ,  $\beta > 0$  . Here  $\beta > 1$  is considered which means deterioration increases with time.
6. Deteriorated units can neither be repaired nor replaced during the cycle time.
7. The demand rate  $R(A, q)$  is a function of the frequency of advertisement  $A$  and the displayed inventory level in the super mall functional form as

$$\begin{aligned} R(A, q) &= A^\lambda (a + bS_1) \text{ for } q > S_1 \\ &= A^\lambda (a + bq(t)) \text{ for } S_0 < q \leq S_1 \\ &= A^\lambda (a + bS_0) \text{ for } q > S_0 \end{aligned}$$

where  $a, b > 0$  ,  $a > b$  .

## 3. MATHEMATICAL MODEL

In this model, the cycle starts with on hand inventory level  $S$  at  $t = 0$  after clearing shortages. Then inventory level depletes continuously and reaches to zero at  $t = T$  due to demand and constant rate of deterioration of units. Meanwhile inventory reaches to  $S_1$  at some time  $t_0$  and it reaches to  $S_0$  at some time  $t_1$  as shown in the Figure 1.

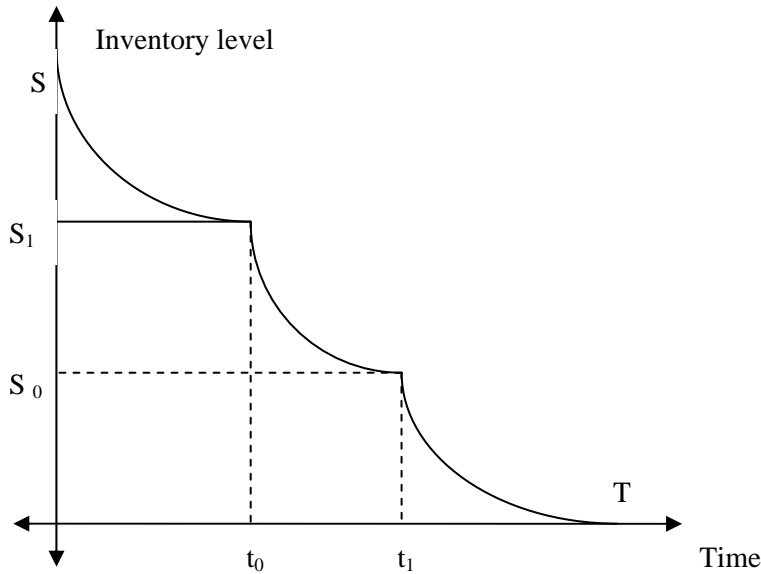


Figure 1. Time – inventory status.

The following differential equations represent the inventory level  $q(t)$  at any instant of time  $t$ .

$$\frac{d q_1(t)}{dt} + \alpha \beta t^{\beta-1} q_1(t) = -A^\lambda (a + bS_1), \quad 0 \leq t \leq t_0 \quad (1)$$

$$\frac{d q_2(t)}{dt} + \alpha \beta t^{\beta-1} q_2(t) = -A^\lambda (a + bq_2(t)), \quad t_0 \leq t \leq t_1 \quad (2)$$

$$\frac{d q_3(t)}{dt} + \alpha \beta t^{\beta-1} q_3(t) = -A^\lambda (a + bS_0), \quad t_1 \leq t \leq T \quad (3)$$

$$\text{with boundary conditions } q_1(0) = S, q_2(t_0) = S_1, q_3(t_1) = S_0, q_3(T) = 0 \quad (4)$$

The solution of differential Eq (1) – (3) using (4) is

$$q_1(t) = S(1 - \alpha t^\beta) - A^\lambda (a + bS_1) \left( t - \frac{\alpha \beta t^{\beta+1}}{\beta+1} \right) \quad (5)$$

$$q_2(t) = S_1 \left( 1 - \alpha (t^\beta - t_0^\beta) \right) - bA^\lambda (t - t_0) - aA^\lambda \left( t - t_0 - \frac{\alpha \beta}{\beta+1} (t^{\beta+1} - t_0^{\beta+1}) + \alpha t_0^\beta (t - 1) + \alpha t_0 (t^\beta - t_0^\beta) + bA^\lambda \left( t t_0 - \frac{t^2}{2} - \frac{t_0^2}{2} \right) \right) \quad (6)$$

$$q_3(t) = S(1 - \alpha (t^\beta - t_1^\beta)) - A^\lambda (a + bS_0) \left( t - t_1 + \frac{\alpha}{\beta+1} (t^{\beta+1} - t_1^{\beta+1}) - \alpha t^\beta (t - t_1) \right) \quad (7)$$

Using boundary condition (4), we have

$$S = S_1(1 + \alpha t_0^\beta) + (a + bS_1) \left( t_0 + \frac{\alpha t_0^{\beta+1}}{\beta+1} \right) \quad (8)$$

Now total inventory from (0, T) is

$$IT = \int_0^{t_0} q_1 dt + \int_{t_0}^{t_1} q_2 dt + \int_{t_1}^T q_3 dt \quad (9)$$

Number of deteriorated during positive inventory time interval is

$$DU = S = A\lambda a - A\lambda b(S_1 t_0 + S_0(T - t_1)) - A^\lambda b \int_{t_0}^{t_1} q_2(t) dt \quad (10)$$

Hence, total cost per time unit  $T$  is

$K(T) = (\text{Inventory holding cost} + \text{Advertisement cost} + \text{ordering cost} + \text{cost due to Deterioration} - \text{salvage value of the deteriorated units}) / T$ .

$$= \frac{1}{T}(b \times IT + A \times G + OC + C \times DU - C \times \gamma \times DU)$$

The necessary condition for  $K(T)$  to be minimum is  $\frac{\partial K(T)}{\partial T} = 0$  and solving it for  $T$  by a suitable mathematical

software For obtained  $T$ ,  $K(T)$  is minimum only if  $\frac{\partial^2 K}{\partial T^2} > 0$  for all  $T > 0$ .

#### 4. COMPUTATIONAL ALGORITHM

The following steps are to be completed for the optimal solution

Step 1: start with  $A = 1$ .

Step 2: compute  $t_0, t_1, T, K(T)$ .

Step 3: Increment  $A$  by 1.

Step 4: Perform step 2 until  $K(A-1, t_0, t_1, T) \geq K(A, t_0, t_1, T) \leq K(A+1, t_0, t_1, T)$ .

Step 5: stop.

#### 5. NUMERICAL EXAMPLE

Consider an inventory system with following parameters in proper units :

$[O, b, C, G, \lambda, a, b, \gamma, \alpha, \beta, S_0, S_1] = [100, 1, 10, 100, 1.3, 200, 0.3, 0.20, 0.10, 3.5, 40, 150]$ .

The optimum value of frequency  $A$  is 3,  $t_0 = 1.8980$  years,  $t_1 = 2.5002$  years,  $T = 3.6685$  years,  $K(T) = \$ 6072.22$ ,  $S = 4479.04$  units (see Figure 2). The sensitivity analysis is carried out to study the effect of deterioration of units;  $\alpha$  and  $\beta$ , stock dependent parameter;  $b$ , rate of change of frequency;  $\lambda$ , and salvage parameter;  $\gamma$  of deteriorated units on the objective function are exhibited in following tables:

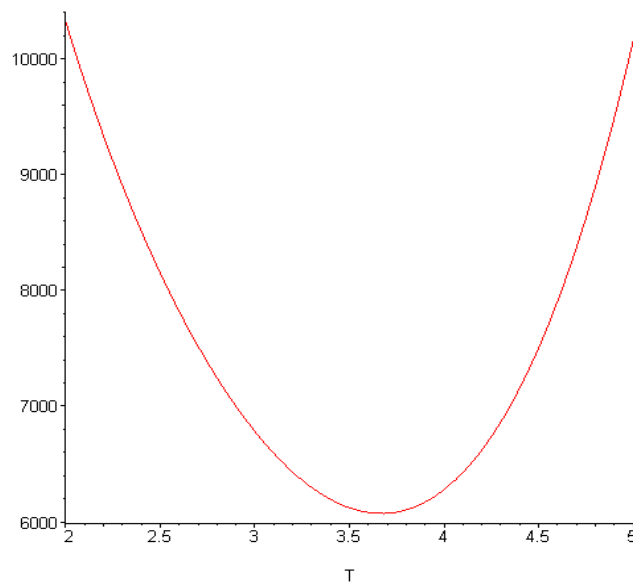


Figure 2. Convexity of  $K(T)$ .

Table 1. Effect of  $\lambda$ 

$\lambda$	1.30	1.32	1.34	1.36
$T$	3.6685	3.6571	3.6460	3.6350
$t_0$	1.8980	1.8916	1.8854	1.8792
$t_1$	2.5002	2.4860	2.4720	2.4579
$K$	6072.22	6107.90	6144.48	6181.97
$S$	4479.04	4508.36	4538.68	4570.05

Table 2. Effect of  $\alpha$ 

$\alpha$	0.10	0.12	0.18	0.20
$T$	3.6684	3.5444	3.2935	3.2335
$t_0$	1.8898	1.8423	1.7310	1.7048
$t_1$	2.500	2.4286	2.280	2.2434
$K$	6072.22	6276.40	6289.49	6997.95
$S$	4479.09	4511.09	4630.82	4672.89

Table 3. Effect of  $\beta$ 

$\beta$	3.3	3.5	3.7	4.0
$T$	3.8777	3.6685	3.4887	3.2623
$t_0$	1.9590	1.8980	1.8448	1.7767
$t_1$	2.5808	2.5002	2.4299	2.3396
$K$	5873.83	6072.22	6270.42	6571.11
$S$	4531.17	4479.04	4445.010	4452.64

Table 4. Effect of  $\gamma$ 

$\gamma$	0.10	0.20	0.30	0.40
$T$	3.7067	3.6684	3.6280	3.5855
$t_0$	1.8930	1.8980	1.9044	1.9128
$t_1$	2.4937	2.5002	2.5083	2.5185
$K$	6562.48	6072.21	5576.38	5074.64
$S$	4452.36	4479.04	4512.10	4554.08

Table 5. Effect of  $b$ 

$b$	0.1	0.15	0.2
$T$	4.0012	3.8264	3.6684
$t_0$	2.2152	2.0419	1.8980
$t_1$	3.0462	2.8499	2.5002
$K$	8411.55	7211.87	6182.03
$S$	6617	5678	4236

Table 6. Effect of  $\alpha$  &  $\gamma$  on decision variables

$\gamma \backslash \alpha$		1.30	1.32	1.34
0.10	$t_0$	2.0940	2.0875	2.0810
	$t_1$	2.8579	2.8487	2.8394
	$T$	3.9900	3.9801	3.9035
	$K(T)$	5367.27	5367.27	5398.99
	$S$	3977.77	3988.51	3999.24
0.12	$t_0$	2.0371	2.0306	2.0242
	$t_1$	2.7665	2.7580	2.7493
	$T$	3.8263	3.8527	3.8430
	$K(T)$	5566.32	5583.09	5599.77
	$S$	4019.99	4031.17	4042.20
0.14	$t_0$	1.9921	1.9851	1.9787
	$t_1$	2.6913	2.6833	2.6752
	$T$	3.7594	3.7500	3.7405
	$K(T)$	5747.17	5764.93	5782.58
	$S$	4060.30	4071.89	4083.43

Table 7. Effect of  $\beta$  &  $\lambda$  on decision variables.

$\lambda \backslash \beta$		1.30	1.32	1.34
3.3	$t_0$	2.1640	2.1572	2.1505
	$t_1$	2.9605	2.9505	2.9405
	$T$	4.2170	4.2066	4.1963
	$K(T)$	5120.77	5137.61	5154.07
	$S$	3636.27	3979.37	4025.38
3.5	$t_0$	2.0940	2.0874	2.0810
	$t_1$	2.8579	2.8487	2.8394
	$T$	3.9900	3.9801	3.9703
	$K(T)$	5367.27	5383.14	5398.98
	$S$	3977.78	3988.51	3999.24
3.7	$t_0$	2.0333	2.0269	2.0207
	$t_1$	2.7674	2.7589	2.7503
	$T$	3.7950	3.7856	3.7763
	$K(T)$	5620.41	5635.24	5650.00
	$S$	4003.05	4012.70	4022.33

Table 8. Effect of  $\lambda$  &  $\gamma$  on decision variables

$\gamma \backslash \lambda$		1.10	1.20	1.30
1.30	$t_0$	2.0856	2.0940	2.1055
	$t_1$	2.8638	2.8579	2.8510
	$T$	4.0352	3.9900	3.9425
	$K(T)$	5851.71	5367.27	4876.15
	$S$	3999.82	3977.78	3952.49
1.32	$t_0$	2.0874	2.0792	2.0986
	$t_1$	2.8486	2.8541	2.8424
	$T$	3.9801	4.0251	3.9329
	$K(T)$	5866.88	5383.14	4892.69
	$S$	3988.51	5866.88	3965.31
1.34	$t_0$	2.0729	2.0810	2.1005
	$t_1$	2.8443	2.8394	2.8190
	$T$	4.0151	3.9703	3.8649
	$K(T)$	5882.06	5398.99	4909.18
	$S$	4017.60	3999.24	3923.97

Table 9. Effect of  $b$  &  $\lambda$  on decision variables

$b \backslash \lambda$		0.2	0.3	0.4
1.30	$t_0$	2.2179	2.0904	2.0395
	$t_1$	3.0407	2.8579	2.7270
	$T$	4.1316	3.9900	3.9116
	$K(T)$	6113.98	5367.27	4909.85
	$S$	4483.36	3977.78	3697.30
1.32	$t_0$	2.2086	2.0875	2.0340
	$t_1$	3.0328	2.8487	2.7174
	$T$	4.1206	3.9801	3.9024
	$K(T)$	6149.49	5383.14	4922.50
	$S$	4500.18	3988.50	3706.72
1.34	$t_0$	2.1995	2.0810	2.0284
	$t_1$	3.0248	2.8394	2.7078
	$T$	4.1099	3.9703	3.8931
	$K(T)$	6164.79	5398.99	4935.24
	$S$	4516.88	3999.24	3716.27

Table 10. Effect of  $\beta$  &  $b$  on decision variables

$\beta \backslash b$		0.2	0.3	0.4
3.3	$t_0$	2.2951	2.1640	2.1061
	$t_1$	3.1631	2.9605	2.8184
	$T$	4.3707	4.2170	4.1328
	$K(T)$	5835.49	5120.78	4689.10
	$S$	4474.55	3967.44	3694.35
3.5	$t_0$	2.2179	2.0940	2.0395
	$t_1$	3.0407	2.8579	2.7270
	$T$	4.1316	3.9900	3.9116
	$K(T)$	6113.98	5367.27	4909.85
	$S$	4483.36	3977.78	3679.30
3.7	$t_0$	2.1508	2.0333	1.9818
	$t_1$	2.9326	2.7674	2.6467
	$T$	3.9262	3.7950	3.7218
	$K(T)$	6398.40	5620.41	5136.53
	$S$	4506.91	4003.05	3715.25

Table 11. Effect of  $\alpha$  &  $\beta$  on decision variables

$\beta \backslash \alpha$		3.3	3.4	3.5
0.10	$t_0$	2.1640	2.0940	2.0331
	$t_1$	2.9605	2.8579	2.7674
	$T$	4.2170	3.9900	3.7950
	$K(T)$	5120.77	5367.27	5620.41
	$S$	3967.44	3977.78	4003.05
0.12	$t_0$	2.1004	2.0375	1.9821
	$t_1$	2.8509	2.7665	2.6829
	$T$	4.0732	3.8624	3.6808
	$K(T)$	5305.13	5566.32	5834.22
	$S$	3999.36	4019.99	4054.26
0.14	$t_0$	2.0495	1.9916	1.9410
	$t_1$	2.7792	2.6913	2.6131
	$T$	3.9574	3.7594	3.5884
	$K(T)$	5473.37	5747.17	6027.63
	$S$	4031.54	4060.30	4101.65



Table 12. Effect of  $\alpha$  &  $\gamma$  on decision variables

$\alpha \backslash \gamma$		0.10	0.20	0.30
0.10	$t_0$	2.0856	2.0940	2.1055
	$t_1$	2.8638	2.8579	2.8511
	$T$	4.0352	3.9900	3.9425
	$K(T)$	5851.82	3977.78	3952.49
	$S$	3999.82	3977.78	3952.49
0.12	$t_0$	2.0293	2.0371	2.0477
	$t_1$	2.7752	2.7665	2.7563
	$T$	3.9081	3.8624	3.8142
	$K(T)$	6088.03	5566.32	5037.93
	$S$	4054.83	4019.99	3980.22
0.14	$t_0$	1.9843	1.9916	2.0015
	$t_1$	2.7023	2.6913	2.6786
	$T$	3.8057	3.7549	3.7107
	$K(T)$	6303.66	5747.17	5184.24
	$S$	4106.71	4060.30	4007.61

Table 13. Effect of  $\alpha$  &  $b$  on decision variables

$\alpha \backslash b$		0.2	0.3	0.4
0.10	$t_0$	2.2179	2.0940	2.0395
	$t_1$	3.0407	2.8579	2.7270
	$T$	4.1316	3.990	3.9116
	$K(T)$	6113.98	5369.27	4909.85
	$S$	4483.36	3977.78	3697.30
0.12	$t_0$	2.1605	2.0371	1.9823
	$t_1$	2.9338	2.7664	2.5541
	$T$	3.9992	3.862	3.7860
	$K(T)$	6303.55	5566.32	5104.19
	$S$	4506.28	4019.99	3742.84
0.14	$t_0$	2.1144	1.9916	1.9367
	$t_1$	2.8460	2.6913	2.5772
	$T$	3.8920	3.7594	3.6848
	$K(T)$	6472.74	5747.17	5283.22
	$S$	4528.60	4060.30	3783.57

Table 14. Effect of  $\beta$  &  $\gamma$  on decision variables

$\beta \backslash \gamma$		0.10	0.20	0.30
3.3	$t_0$	2.1551	2.1640	2.1761
	$t_1$	2.9646	2.96050	2.9558
	$T$	4.2652	4.2170	2.9558
	$K(T)$	5571.69	5120.76	4663.26
	$S$	3981.24	3967.44	3951.56
3.5	$t_0$	2.0856	2.0940	2.1055
	$t_1$	2.8638	2.8579	2.8510
	$T$	4.0351	3.9900	3.9425
	$K(T)$	5851.71	5367.27	4876.15
	$S$	3999.82	3977.78	3952.49
3.7	$t_0$	2.0254	2.0330	2.0441
	$t_1$	2.7750	2.7674	2.758
	$T$	3.8375	3.7950	3.7501
	$K(T)$	6140.53	5620.41	5093.71
	$S$	3389.32	4003.05	3967.87

Table 15. Effect of  $\gamma$  &  $b$  on decision variables

$\gamma \backslash b$		0.2	0.3	0.4
0.10	$t_0$	2.1997	2.0856	2.0338
	$t_1$	3.0561	2.8638	2.7279
	$T$	4.1771	4.0352	3.9554
	$K(T)$	6716.63	5851.71	5332.03
	$S$	4548.55	3999.82	3700.35
0.20	$t_0$	2.2179	2.0940	2.0395
	$t_1$	3.0407	2.8579	2.7270
	$T$	4.1316	3.9900	3.9116
	$K(T)$	6113.98	5367.27	4909.85
	$S$	4483.36	3977.78	5498.80
0.30	$t_0$	2.2438	2.1055	2.0471
	$t_1$	3.0233	2.8510	2.7260
	$T$	4.0860	3.9425	3.8653
	$K(T)$	5504.87	4876.15	4481.65
	$S$	4411.22	3952.49	3693.83

## Observations

As deterioration parameter  $\alpha$  and  $\beta$  increases total cycle time decreases whereas total cost increases. Total cost and cycle time both decreases as salvage parameter  $\gamma$  increases. This is because there is some savings by selling deteriorated items instead of just throwing it away.

As stock displayed parameter  $b$  increases total cost and cycle time both decreases. The retailer will have to put orders frequently resulting increase in the total cost. Cycle time decreases whereas total cost increases as rate at which advertisement is displayed increases. All these are critical factors in deciding optimal ordering policy to minimize the objective function total cost.

This sensitivity analysis done here shows that stock dependent demand increases total cost decreases significantly. Frequent advertisement is going to increase in total cost but if salvage parameter increases significantly increment in total cost can be checked.

## 6. CONCLUSIONS

In market, the units deteriorate due to vaporization; damages while loading and unloading, perishability and many other factors. So the deterioration effect should not be ignored while computing inventory cost of the retailer. For products like food grains, blood components, vegetables and fruits, pharmaceuticals etc. utility decreases with passage of time and hence retailer needs to find trade off among inventory carrying cost and deterioration cost. On the other hand, retailer uses media for the sale of the product through advertisement and displayed stock to attract the customers. This concept encourage author to develop propose model. This model can be extended to the incorporate selling price and advertisement dependent demand.

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