ON A VOLUME FLEXIBLE STOCK-DEPENDENT INVENTORY MODEL

S. SANA ¶ AND K. S. CHAUDHURI ¶¶

ABSTRACT. An inventory model of a *volume flexible manufacturing system* for a deteriorating item is developed, taking a stock -dependent demand rate. It is assumed that the demand rate remains stock-dependent for an initial period after which a uniform demand rate follows as the stock comes down to a certain level. The unit production cost is taken to be a function of the finite production rate which is treated to be a decision variable. The mathematical expression for the *average profit function* is derived and it is maximized subject to the different constraints of the system using **Zoutendijk's Method** of constrained optimization, the algorithm of which is given. The solution procedure is illustrated with the help of a numerical example. Using the result of the example, sensitivity of the near optimal solution to changes in the values of the parameters of the system is analyzed.

Key words and phrases : volume flexible manufacturing system, deteriorating item, stock-dependent demand rate, constrained optimization, sensitivity anlysis.

^{¶¶} Professor, Dept. of Maths. Jadavpur University, Cal- 32, India.

 $[\]P$ Dept. of Maths. Bhangar Mahavidyalaya, University of Calcutta, India.

Introduction

In the Classical Economic Production Lot Size(**EPLS**) model, the production rate of a machine is regarded to be pre-determinded and inflexible¹. Alder and Nanda², Sule^{3,4}, Axsater and Elmaghraby⁵, Muth and Spearmann⁶ extended the **EPLS** model to situations where *learning effects* would induce an increase in the production rate. Proteus⁷, Rosenblat and Lee⁸ and Cheng⁹ considered the **EPLS** model in an imperfect production process in which the demand would exceed the supply.

Schweitzer and Seidmann¹⁰ adopted, for the first time, the concept of flexibility in the machine production rate and discussed optimization of processing rates for a **FMS** (*flexible manufacturing system*). Obviously, the machine production rate is a decision variable in the case of a **FMS** and then the unit production cost becomes a function of the production rate. Khouja and Mehrez¹¹ and Khouja¹² extended the **EPLS** model to an imperfect production process with a flexible production rate. Silver¹³, Moon, Gallego and Simchi-Levi¹⁴ discussed the effects of slowing down production in the context of a manufacturing equipment of a family of items, assuming a common cycle for all the items. Gallego¹⁵ extended this model by removing the stipulation of a common cycle for all the items.

But the above studies did not consider the demand rate to be variable. It is a common belief that large piles of goods displayed in a supermarket lead the customers to buy more. Silver and Peterson¹⁶ and Silver¹⁷ have also noted that sales at the retail level tend to be proportional to the inventory displayed. Baker and Urban¹⁸ and Urban¹⁹ considered an inventory system in which the demand rate of the product is a function of the on-hand inventory. Goh²⁰ discussed the model of Baker and Urban¹⁸ relaxing the assumption of a constant holding cost. Mandal and Phaujder²¹ then extended this model to the case of deteriorating items with a constant production rate. Datta and Pal²² presented an inventory model in which the demand rate of an item is dependent on the on-hand inventory level until a given inventory level is achieved, after which the demand rate becomes constant. Giri, Pal , Goswami and Chaudhuri²⁴ extended the model of Urban[19] to the case of items deteriorating overtime. Ray and Chaudhuri²⁵ discussed an **EOQ** (economic order quantity) model with stock-dependent demand, shortage, inflation and time discounting of different costs and prices associated with the system. Ray, Goswami and Chaudhuri²⁶ studied the inventory problem

with a stock-dependent demand rate and two levels of storage, *rented ware*house (**RW**) and own warehouse (**OW**). Giri and Chaudhuri²⁷ extended the model of Goh^{20} to cover an inventory of a deteriorating item and discussed both the case of nonlinear time-dependent and stock-dependent holding costs.

Volume flexibility is a major component in a FMS. The manufacturing flexibility which is capable of adjusting the production rate with the variability in the market demand is known as volume flexibility²⁸. In the present paper, we consider a volume flexible manufacturing system for a deteriorating item with an inventory-level-dependent demand rate. In reality, the demand rate remains stock-dependent for some time and then becomes a constant after the stock falls down to a certain level. Several factors like limited number of potential customers and their goodwill, price and quality of the goods, locality of shop, etc. can be accounted for the change in the demand pattern.

Fundamental Assumptions and Notations

Assumptions:

- 1. The inventory system involves only one item and is a self-production system.
- 2. Lead time is zero.
- 3. No shortages are permitted.
- 4. The time-horizon is infinite.
- 5. The production cost per unit item is a function of the production rate.
- 6. The production rate is considered to be a decision variable.

Notations:

- *P* The production rate per unit time ;
- I(t) On-hand inventory at time 't' ≥ 0 ;
- R(I) Demand-rate function varying with I(t);

- S_0 The stock-level, beyond which the demand rate becomes constant;
- θ Constant deterioration rate of the On-hand inventory, $0 < \theta < 1$;
- C_h Holding cost per unit per unit time;
- C_s Setup cost per production run;
- $\eta(P)$ The production cost per unit item ;
- S_p Salvage cost per unit item;
- T The duration of the production cycle;
- ∇ Gradient operator.

Formulation of the Model

We consider a self-manufacturing system in which the items are manufactured in a machine and the market demand is filled by these manufactured items. The demand rate is dependent on the on-hand inventory down to a level S_0 , beyond which it is assumed to be a constant, i.e.,

$$R(I) = D + \gamma I(t), \ I > S_0$$

= $D + \gamma S_0, \ 0 \le I \le S_0,$

where D and γ are non-negative constants and D < P . The production cost per unit is

$$\eta(P) = r + \frac{g}{P} + \alpha P$$

where r, g, α are all positive constants. This cost is based on the following factors:

- 1. The material cost r per unit item is fixed.
- 2. As the production rate increases, some costs like labour and energy costs are equally distributed over a large number of units. Hence the production cost per unit(g/P) decreases as the production rate(P) increases.

3. The third term(αP), associated with tool/die costs, is proportional to the production rate.

The production cycle begins with zero stock. Production starts at t = 0and the stock reaches a level S_0 at time $t = t_1$ after meeting demands. The demand rate in the interval $(0, t_1)$ is $(D+\gamma S_0)$. In the interval (t_1, t_2) , production continues uninterruptedly and the demand rate depends on the instantaneous stock level. Production is stopped at time $t = t_2$. The demand rate continues to depend on the instanteneous inventory level until t = t_3 when the stock falls down to the level S_0 again. The inventory falls to the zero level at the end t = T of the production cycle. This cycle of production is repeated over and over again. Therefore, the governing equations of this model are

$$\frac{dI(t)}{dt} + \theta I(t) = P - (D + \gamma S_0), \ 0 \le t \le t_1, \ I \le S_0;$$
(1)

with
$$I(0) = 0$$
 and $I(t_1) = S_0;$

$$\frac{dI(t)}{dt} + (\theta + \gamma)I(t) = P - D, \ t_1 \le t \le t_2, \ I > S_0;$$
(2)

$$\frac{dI(t)}{dt} + (\theta + \gamma) I(t) = -D, \ t_2 \le t \le t_3, \ I > S_0$$
(3)
with $I(t_3) = S_0;$

$$\frac{dI(t)}{dt} + \theta I(t) = -(D + \gamma S_0), \ t_3 \le t \le T, \ I \le S_0$$
(4)
with $I(T) = 0.$

From equation (1), we have

i.e.

$$\int_0^t d(I \ e^{\theta s}) = (P - D - \gamma \ S_0) \ \int_0^t e^{\theta s} \ ds,$$
$$I(t) = \left(\begin{array}{c} \frac{P - D - \gamma \ S_0}{\theta} \end{array} \right) (1 - e^{-\theta t}), \ 0 \le t \le t_1.$$
(5)

Now the condition $I(t_1) = S_0$ gives us

$$\frac{P - D - \gamma S_0}{\theta} \left(1 - e^{-\theta t_1}\right) = S_0$$

$$\Rightarrow e^{-\theta t_1} = 1 - \frac{\theta S_0}{P - D - \gamma S_0},$$

i.e.,

$$t_1 = -\frac{1}{\theta} \ln \left[1 - \frac{\theta S_0}{P - D - \gamma S_0} \right], \tag{6}$$

where $P > D + (\theta + \gamma)S_0$.

Also from equation (2) , we have

$$\int_{t_1}^t d[I \ e^{(\theta + \gamma)s}] = (P - D) \ \int_{t_1}^t \ e^{(\theta + \gamma)s} \ ds \ ,$$

i.e.,

$$I(t) = \left(S_0 - \frac{P-D}{\theta+\gamma}\right) e^{(\theta+\gamma)(t_1-t)} + \left(\frac{P-D}{\theta+\gamma}\right), t_1 \le t \le t_2.$$
(7)

Therefore,

$$I(t_2) = \left(S_0 - \frac{P-D}{\theta+\gamma}\right) e^{(\theta+\gamma)(t_1-t_2)} + \left(\frac{P-D}{\theta+\gamma}\right).$$
(8)

Similarly, the equation (3) gives us

$$\begin{split} \int_{t_2}^t d[Ie^{(\theta+\gamma)s}] &= -D \int_{t_2}^t e^{(\theta+\gamma)s} ds \\ \Rightarrow I(t)e^{(\theta+\gamma)t} - I(t_2)e^{(\theta+\gamma)t_2} &= -\frac{D}{\theta+\gamma}[e^{(\theta+\gamma)t} - e^{(\theta+\gamma)t_2}] \\ \Rightarrow I(t) &= I(t_2)e^{(\theta+\gamma)(t_2-t)} - \frac{D}{\theta+\gamma}[1 - e^{(\theta+\gamma)(t_2-t)}], \end{split}$$

i.e.,

$$I(t) = \left(I(t_2) + \frac{D}{\theta + \gamma} \right) e^{(\theta + \gamma)(t_2 - t)} - \frac{D}{\theta + \gamma}, t_2 \leq t \leq t_3 \quad (9)$$

Since $I(t_3) = S_0$, we have

$$S_0 = [I(t_2) + \frac{D}{\theta + \gamma}] e^{(\theta + \gamma)(t_2 - t_3)} - \frac{D}{\theta + \gamma}$$

$$\Rightarrow e^{(\theta + \gamma)(t_2 - t_3)} = [\frac{S_0 + D/(\theta + \gamma)}{I(t_2) + D/(\theta + \gamma)}]$$

$$\Rightarrow t_2 - t_3 = \frac{1}{\theta + \gamma} ln [\frac{S_0 + D/(\theta + \gamma)}{I(t_2) + D/(\theta + \gamma)}],$$

i.e.,

$$t_3 = t_2 - \frac{1}{\theta + \gamma} \ln \left[\frac{S_0 + D/(\theta + \gamma)}{I(t_2) + D/(\theta + \gamma)} \right] , \qquad (10)$$

where $I(t_2) > S_0$.

Again the equation (4) becomes

$$\begin{aligned} \int_{t_3}^t dI(Ie^{\theta s}) &= - (D + \gamma S_0) \int_{t_3}^t e^{\theta s} ds \\ \Rightarrow I(t) e^{\theta t} - S_0 e^{\theta t_3} &= - (\frac{D + \gamma S_0}{\theta}) (e^{\theta t} - e^{\theta t_3}) \\ \Rightarrow I(t) &= S_0 e^{\theta (t_3 - t)} - \frac{D + \gamma S_0}{\theta} \left\{ 1 - e^{\theta (t_3 - t)} \right\}, \end{aligned}$$

i.e.,

$$I(t) = \left(S_0 + \frac{D + \gamma S_0}{\theta} \right) e^{\theta(t_3 - t)} - \left(\frac{D + \gamma S_0}{\theta} \right), \ t_3 \le t \le T.$$
(11)

Now I(T) = 0 implies

$$\begin{array}{ll} \left(S_0 + \frac{D + \gamma S_0}{\theta}\right) \, e^{\theta(t_3 - T)} - \frac{D + \gamma S_0}{\theta} &= 0 \\ \Rightarrow e^{\theta(t_3 - T)} &= \frac{D + \gamma S_0}{D + (\gamma + \theta) S_0} \ , \end{array}$$

i.e.,

$$T = t_3 - \frac{1}{\theta} \ln \left[\frac{D + \gamma S_0}{D + (\gamma + \theta) S_0} \right].$$
(12)

Let Inv_1 , Inv_2 , Inv_3 , Inv_4 be the total inventories in the intervals $0 \le t \le t_1$, $t_1 \le t \le t_2$, $t_2 \le t \le t_3$, $t_3 \le t \le T$ respectively. Then

$$Inv_{1} = \int_{0}^{t_{1}} I(t) dt = \int_{0}^{t_{1}} \left(\frac{P - D - \gamma S_{0}}{\theta}\right) \left(1 - e^{-\theta t}\right) dt = \left(\frac{P - D - \gamma S_{0}}{\theta^{2}}\right) \left(\theta t_{1} + e^{-\theta t_{1}} - 1\right);$$

$$Inv_{2} = \int_{t_{1}}^{t_{2}} I(t) dt = \{ S_{0} - \frac{P-D}{\theta+\gamma} \} e^{(\theta+\gamma)t_{1}} \int_{t_{1}}^{t_{2}} e^{-(\theta+\gamma)t} dt + (\frac{P-D}{\theta+\gamma}) \int_{t_{1}}^{t_{2}} dt = \frac{1}{(\theta+\gamma)} \{ S_{0} - \frac{P-D}{\theta+\gamma} \} \{ 1 - e^{-(\theta+\gamma)(t_{2}-t_{1})} \} + (\frac{P-D}{\theta+\gamma})(t_{2}-t_{1});$$

$$Inv_{3} = \int_{t_{2}}^{t_{3}} I(t) dt = [I(t_{2}) + \frac{D}{\theta + \gamma}] e^{(\theta + \gamma)t_{2}} \int_{t_{2}}^{t_{3}} e^{-(\theta + \gamma)t} dt - (\frac{D}{\theta + \gamma}) \int_{t_{2}}^{t_{3}} dt = \frac{1}{(\theta + \gamma)} \{I(t_{2}) + \frac{D}{\theta + \gamma}\} \{1 - e^{(\theta + \gamma)(t_{2} - t_{3})}\} - (\frac{D}{\theta + \gamma})(t_{3} - t_{2});$$

$$Inv_{4} = \int_{t_{3}}^{T} I(t)dt = (S_{0} + \frac{D + \gamma S_{0}}{\theta})e^{\theta t_{3}} \int_{t_{3}}^{T} e^{-\theta t} dt - (\frac{D + \gamma S_{0}}{\theta})\int_{t_{3}}^{T} dt = \frac{1}{\theta^{2}}(\theta S_{0} + D + \gamma S_{0})(1 - e^{\theta(t_{3} - T)}) - (\frac{D + \gamma S_{0}}{\theta})(T - t_{3})$$

$$= \frac{S_0}{\theta} + \frac{D + \gamma S_0}{\theta^2} \ln \left[\frac{D + \gamma S_0}{D + (\gamma + \theta) S_0} \right], \text{ by (12)}.$$

The values of θ , S_0 and $(D + \gamma S_0)$ must be such that $Inv_4 > 0$ is satisfied. Now the total deteriorated item (I_d) is

$$I_d = \theta \{ \int_0^{t_1} I(t) dt + \int_{t_1}^{t_2} I(t) dt + \int_{t_2}^{t_3} I(t) dt + \int_{t_3}^{T} I(t) dt \} = \theta (Inv_1 + Inv_2 + Inv_3 + Inv_4).$$

Therefore, the total demand in (0,T) becomes $D_T = (Pt_2 - I_d)$. Then the average profit during time (0,T) takes the form

$$\pi(P, t_2) = \frac{1}{T} [(Pt_2 - I_d)S_p - \{C_s + C_h(Inv_1 + Inv_2 + Inv_3 + Inv_4) + (r + \frac{g}{P} + \alpha P)Pt_2\}]$$
(13)

Therefore, we have to Maximize $\pi(P, t_2)$;

subject to the constraints :

$$\begin{array}{rcl} D+(\theta+\gamma)S_0-P<0\;,\\ &-Inv_1\;<\;0\;,\\ &-Inv_2\;<\;0\;,\\ &-Inv_3\;<\;0\;,\\ &-I(t_2)+S_0\;<\;0\;,\\ &-t_2+t_1\;<\;0\;. \end{array}$$

The condition $D + (\theta + \gamma)S_0 - P < 0 \Rightarrow 0 < \frac{\theta S_0}{P - D - \gamma S_0} < 1$ which is necessary for the value of t_1 in eqn. (6) to be real.

The three conditions $-Inv_1 < 0$, $-Inv_2 < 0$ and $-Inv_3 < 0$ ensure that Inv_1 , Inv_2 and Inv_3 must be positive.

The condition $-I(t_2) + S_0 < 0$ implies that $I(t_2)$, the inventory level at time t_2 , is higher than S_0 .

The condition $-t_2 + t_1 < 0$ ensures that t_2 is greater than t_1 .

This problem can be solved by **Zoutendijk's Method²³** whose Algorithm is discussed below.

General Problem : Minimize $\{-\pi(\bar{X})\}$ subject to the constraints : $G_j(\bar{X}) < 0$, where $\bar{X} \in \mathbb{R}^n$, $j = 1, 2, \dots, m$.

Algorithm:

- 1. Start with an initial feasible point \bar{X}_1 , evaluate $\pi(\bar{X}_1)$ and $G_j(\bar{X}_1)$, $j = 1, 2, \dots, m$. Set the iteration number as i = 1.
- 2. If $G_j(\bar{X}_i) < 0$, j = 1, 2, ..., m. (i.e., \bar{X}_i is an interior feasible point), set the current search direction as $\bar{S}_i = -\nabla \pi(\bar{X}_i)$. Normalize \bar{S}_i in a suitable manner.
- 3. Find a suitable step length λ_i along the direction \bar{S}_i and obtain a new point \bar{X}_{i+1} as $\bar{X}_{i+1} = \bar{X}_i + \lambda_i \bar{S}_i$.
- 4. Evaluate the objective function $\pi(\bar{X}_{i+1})$.
- 5. Test for the convergence of the method. If $|\frac{\pi(\bar{X}_i)-\pi(\bar{X}_{i+1})}{\pi(\bar{X}_i)}| \leq \epsilon$ where ϵ is a preassigned small positive quantity, terminate the iteration by taking $\bar{X}_{opt} \approx \bar{X}_{i+1}$. Otherwise, go to next step.
- 6. Set the new iteration number as i = i + 1, and repeat from step 2 onwards.

Numerical Example

We take the parameter values as D = 50, $\theta = 0.05$, $\gamma = 0.1$, $S_0 = 100$, $C_s = 300$, $C_h = 0.1$, $S_p = 6.0$, r = 1.0, g = 250, $\alpha = 0.01$ in appropriate units. We obtain the optimum results $t_1^* = 1.258883$, $t_2^* = 6.696204$, $t_3^* = 10.07596$, $T^* = 11.67682$, $P^* = 141.9617$, $\pi^* = 41.93613$.

Sensitivity Analysis

Using the same numerical example, the sensitivity of each variable t_1^* , t_2^* , t_3^* , T^* , P^* and π^* to changes in the values of each of the parameters θ , γ , S_0 , C_s , C_h , S_p , r, g, α is examined which is shown in Table 1. In the proposed

model, t_1^* changes by 10.28 %, 03.27 %, 06.64 %, -12.67 % respectively with 50 %, 20 %, -20 %, -50 % changes in the value of θ . t_2^* changes by 1787.29 % , 44.52 % , -24.76 % , -37.03 % respectively with 50 % , 20 % , -20 % , -50 % changes in the value of γ . t_3^* changes by -14.14 % , -06.30 % , 00.46 % , 11.70 % respectively with 50 % , 20 % , -20 % , -50 % changes in the value of S_0 . T^* changes by 39.74 % , 14.36 % , -14.63 % , -36.63 % respectively with 50%, 20%, -20%, -50% changes in the value of C_s . π^* changes by -92.74% , -38.01 % , 39.89 % , 106.62 % , respectively with 50 % , 20 % , -20 % , -50 % changes in the value of 'r'. Similarly P^* changes by (152.99 %, 11.64 %, -04.35 %, -04.46 %); (23.40 %, 09.38 %, -10.98 %, -04.01 %); (-22.42 % , -11.60 % , 13.83 % , 64.83 %) for 50 % , 20 % , -20 % , -50 % changes in the value of S_p , g and α respectively. In a similar manner, the changes in the solution variables for changes in other parameters can be computed. It is seen that $t_1^*, t_2^*, t_3^*, T^*, P^*$ and π^* are moderately sensitive to changes in the parameters $\bar{\theta}$, γ , S_0 , C_s , C_h , S_p , r, g, α . P^* is insensitive to changes in S_0 ; but P^* is fairly sensitive to the changes in the parameters S_p , g and α .

 Table 1: Sensitivity Analysis

| Changing | t_1^* | t_2^* | t_3^* | T^* | P^* | π^* |
|-------------------------|---------|----------|----------|----------|----------|-----------|
| parameter | | | | | | |
| $\operatorname{in}(\%)$ | | | | | | |
| θ +50 | 1.3883 | 4.8833 | 7.0585 | 8.6290 | 135.8456 | 24.6350 |
| +20 | 1.3001 | 5.6528 | 8.4405 | 10.0290 | 139.9558 | 34.3048 |
| -20 | 1.3424 | 9.5264 | 13.6156 | 15.2291 | 136.5070 | 50.7171 |
| -50 | 1.0993 | 123.4375 | 130.5588 | 132.1917 | 152.2191 | 74.5473 |
| γ +50 | 1.1375 | 126.3767 | 130.3652 | 131.8474 | 155.4316 | 73.7782 |
| +20 | 1.2217 | 9.6771 | 13.4629 | 15.0141 | 146.3745 | 50.7842 |
| -20 | 1.2478 | 5.0385 | 8.0524 | 9.7062 | 140.6663 | 34.5639 |
| -50 | 1.2790 | 4.2163 | 7.0012 | 8.7415 | 135.7072 | 25.1930 |
| $S_0 + 50$ | 2.0531 | 6.2040 | 8.6511 | 10.8351 | 141.8735 | 44.6257 |
| +20 | 1.5592 | 6.4515 | 9.4408 | 11.2882 | 141.9970 | 42.8400 |
| -20 | 0.9817 | 6.5000 | 10.1227 | 11.4565 | 141.5000 | 41.1946 |
| -50 | 0.5808 | 6.9485 | 11.2549 | 12.1440 | 142.3439 | 40.5015 |
| C_s +50 | 1.3704 | 10.7304 | 14.7161 | 16.3170 | 135.4992 | 30.9589 |
| +20 | 1.2917 | 8.0744 | 11.7522 | 13.3530 | 139.9434 | 37.1254 |
| -20 | 1.2130 | 5.3606 | 8.3671 | 9.9679 | 144.9651 | 47.4783 |
| -50 | 1.2120 | 3.6595 | 5.7987 | 7.3996 | 145.0331 | 57.8428 |
| C_h +50 | 1.2618 | 5.2080 | 8.0330 | 9.6339 | 141.9800 | 32.8058 |
| +20 | 1.2744 | 6.0152 | 9.1276 | 10.7284 | 140.9916 | 38.0860 |
| -20 | 1.2554 | 7.7621 | 11.4545 | 13.0554 | 142.1793 | 46.1549 |
| -50 | 1.2076 | 10.2054 | 14.5382 | 16.1390 | 145.3308 | 53.5354 |
| $S_p + 50$ | 0.6465 | 291.8469 | 299.8896 | 301.4904 | 217.1918 | 394.8986 |
| +20 | 1.0419 | 270.1158 | 276.0579 | 277.6588 | 158.4912 | 147.9354 |
| -20 | 1.3651 | 4.5496 | 6.8570 | 8.4579 | 135.7813 | -39.8138 |
| -50 | 1.3678 | 3.2075 | 4.7591 | 6.3600 | 135.6360 | -156.0006 |
| r + 50 | 1.3253 | 4.8981 | 7.4475 | 9.0483 | 137.9794 | 3.0430 |
| +20 | 1.2890 | 5.8232 | 8.8309 | 10.4318 | 140.1007 | 25.9964 |
| -20 | 1.2364 | 8.2314 | 12.0896 | 13.6905 | 143.4028 | 58.6627 |
| -50 | 1.1649 | 15.8558 | 20.9309 | 22.5317 | 148.3684 | 86.6486 |

| Changing | t_1^* | t_2^* | t_3^* | T^* | P^* | π^* |
|--------------|---------|----------|----------|----------|----------|----------|
| paramete | r | | | | | |
| in(%) | | | | | | |
| g + 50 | 0.8876 | 3.2432 | 5.96640 | 7.5672 | 175.1753 | -18.9558 |
| +20 | 1.0781 | 4.6696 | 7.71267 | 9.3135 | 155.2737 | 15.3291 |
| -20 | 1.5664 | 13.6505 | 17.5322 | 19.1331 | 126.3721 | 73.6078 |
| -50 | 1.3561 | 81.9617 | 86.8966 | 88.4974 | 136.2690 | 137.1216 |
| α +50 | 2.1010 | 6.5581 | 8.5022 | 10.1030 | 110.1384 | -5.0713 |
| +20 | 1.5884 | 6.4549 | 9.0780 | 10.6788 | 125.4862 | 20.7892 |
| -20 | 1.0093 | 8.1195 | 12.6550 | 14.2559 | 161.5913 | 68.1260 |
| -50 | 0.5831 | 292.7949 | 301.3344 | 302.9353 | 233.9999 | 153.9163 |

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