# AN INVENTORY MODEL AND ITS OPTIMIZATION 

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An inventory model with constant demand of rate $\mu(\mu>0)$ is considered. The inventory is replenished up to $\beta$ by a deliveryman who arrives according to a Poisson process of rate $\lambda$, only if the stock does not exceed a threshold $\alpha(0 \leq \alpha \leq \beta)$. The distribution function of $X(t)$, the stock at time $t$, is deduced from a partial differential equation, two interesting characteristics, the first passage time to state 0 and the probability that the stock exceeds a certain level during a given interval, are considered, the stationary distribution is obtained more explicitly, and an optimal policy with respect to the threshold $\alpha$ is studied.

## 1. Introduction

In this paper, an inventory model is introduced. Consider an inventory whose stock is initially $\beta$, thereafter decreases linearly at rate $\mu, \mu>0$, and remains at 0 if the inventory becomes empty. The inventory is replenished by a deliveryman who arrives at the inventory according to a Poisson process of rate $\lambda$. If the level of the inventory exceeds a threshold $\alpha$, $0 \leq \alpha \leq \beta$, he does nothing, otherwise he instantaneously increases the level of the inventory up to $\beta$.

Baxter and Lee [1] introduced a similar inventory model where the size of a delivery is a random variable $Y$ such that $Y \geq \alpha$ almost surely. In the paper, they derived a Laplace-Stieltjes transform of the distribution function of the level of the inventory at time $t$ and considered the stationary case where the distribution function does not depend on time $t$.

Since the inventory is replenished up to $\beta$ in our model rather than by a random amount ( $\beta$ may be considered as the capacity of the inventory),

[^0]the points where the restockings occur form a renewal process, and this fact enables us to obtain the distribution function of the level of the inventory at time $t$ directly and to study the stationary case more explicitly. We further consider two interesting characteristics of the model, the first passage time to state 0 and the probability that the stock exceeds a certain level during a given interval. We also show that there exists a unique optimal policy with respect to the threshold $\alpha$, after assigning costs to the inventory.

## 2. The Distribution Function

Let $X(t)$ be the level of the inventory at time $t$ and $F(x, t)$ be the distribution function of $X(t)$. We can have the following three mutually exclusive events during the small interval $(t, t+\delta t)$ :
(a) The deliveryman does not come, then

$$
X(t+\delta t)= \begin{cases}X(t)-\mu \delta t & \text { almost surely if } X(t)>\mu \delta t \\ 0 & \text { almost surely if } X(t) \leq \mu \delta t\end{cases}
$$

(b) The deliveryman comes but does nothing since $X(t)>\alpha$, then

$$
X(t+\delta t)=X(t)-\mu \delta t \quad \text { almost surely } .
$$

(c) The deliveryman comes and makes a delivery since $X(t) \leq \alpha$ then

$$
X(t+\delta t)=\beta-\mu \delta t \quad \text { almost surely }
$$

Thus, for $0 \leq x<\beta$,

$$
\begin{aligned}
F(x, t+\delta t)= & (1-\lambda \delta t) F(x+\mu \delta t, t)+\lambda \delta t P\{X(t) \leq x+\mu \delta t, \\
& X(t)>\alpha\}+\lambda \delta t I\{x \geq \beta-\mu \delta t\} F(\alpha, t)+o(\delta t),
\end{aligned}
$$

where $I_{A}$ denotes the indicator of event $A$. Now

$$
F(x+\mu \delta t, t)=F(x, t)+\mu \delta t \frac{\partial}{\partial x} F(x, t)+o(\delta t)
$$

on performing a Taylor series expansion, assuming that $\frac{\partial}{\partial x} F(x, t)$ exists. Substituting this expression into the above equation, subtracting $F(x, t)$ from each side of the equation, dividing by $\delta t$, and letting $\delta t \rightarrow 0$, we have the following partial differential equation:

$$
\begin{equation*}
\frac{\partial}{\partial t} F(x, t)=\mu \frac{\partial}{\partial x} F(x, t)-\lambda F(x \wedge \alpha, t), \quad \text { for } \quad 0 \leq x<\beta \tag{2.1}
\end{equation*}
$$

Since the level of the inventory cannot exceed $\beta, F(\beta, t)=1$ for $t>0$.
Before we solve the equation (2.1), we first derive a formula for $F(\alpha, t)$, which can be used as a boundary condition.

Lemma 2.1. If we ignore the first passage time to $\alpha$, i.e. $\frac{\beta-\alpha}{\mu}$, then

$$
F(\alpha, t)=e^{-\lambda t}+\int_{0}^{t} e^{-\lambda(t-u)} h(u) d u
$$

where $h(t)=\sum_{n=1}^{\infty} g^{(n)}(t)$ and $g(t)=\lambda e^{-\lambda\left(t-\frac{\beta-\alpha}{u}\right)}$.
Proof. Notice that the points where the stock of the inventory reaches $\alpha$ from an embedded renewal process. Let $T^{*}$ be the time between successive renewals. Then

$$
T^{*}=T+\frac{\beta-\alpha}{\mu},
$$

where $T$ is an exponential random variable with parameter $\lambda$. The probability density function of $T^{*}, g(t)$ say, is given by

$$
g(t)=\lambda e^{-\lambda\left(t-\frac{\beta-\alpha}{\mu}\right)}, \quad \text { for } \quad t>\frac{\beta-\alpha}{\mu} .
$$

Let $h(t)$ denote the renewal density function of the embedded renewal process, that is, $h(t)=\sum_{n=1}^{\infty} g^{(n)}(t)$, where the superscript denotes $n$-fold recursive convolution.

Now, notice that $F(\alpha, t)=1$ if the deliveryman has not arrived until time $t$ or if there is a renewal in the embedded renewal process at $u \in(0, t]$ and the deliveryman does not arrive in the interval $[u, t]$. Hence

$$
F(\alpha, t)=e^{-\lambda t}+\int_{0}^{t} e^{-\lambda(t-u)} h(u) d u .
$$

Now, the equation (2.1) can be divided into the following two equations:

$$
\frac{\partial}{\partial t} F(x, t)=\mu \frac{\partial}{\partial x} F(x, t)-\lambda F(x, t), \quad \text { for } \quad 0 \leq x<\alpha
$$

and

$$
\frac{\partial}{\partial t} F(x, t)=\mu \frac{\partial}{\partial x} F(x, t)-\lambda F(\alpha, t), \quad \text { for } \quad \alpha \leq x<\beta .
$$

Applying $F(\alpha, t)$ obtained in Lemma 2.1 to both equations as a boundary condition and solving the partial differential equations for $F(x, t)$ by an
argument similar to that of Colton [2, p.6-11], we see that
$F(x, t)=F\left(\alpha, t+\frac{x-\alpha}{\mu}\right) e^{\lambda(x-\alpha) / \mu}, \quad$ for $\quad 0 \leq x<\alpha, \quad$ and
$F(x, t)=F\left(\alpha, t+\frac{x-\alpha}{\mu}\right)+\frac{\lambda}{\mu} \int_{\alpha}^{x} F\left(\alpha, t+\frac{x-u}{\mu}\right) d u, \quad$ for $\quad \alpha \leq x<\beta$, where $F(\alpha, t)$ is given in Lemma 2.1.

## 3. The First Passage Time to State 0

Define $T_{0}=\inf \{t \mid X(t)=0\}$, the first passage time to state 0 . Let $Y_{1}, Y_{2}, \cdots, Y_{N}$ be the sequence of the amounts of the deliveries made by the deliveryman, before the stock reaches state 0 , then

$$
Y_{i} \stackrel{\underline{\mathcal{D}}}{=} \beta-\alpha+\mu T, \quad i=1,2, \cdots, N,
$$

under the condition that $T$, the exponential random variable with parameter $\lambda$, is less than $\frac{\alpha}{\mu}$, and so the distribution function of $Y_{i}, D(y)$ say, is given by

$$
\begin{aligned}
D(y) & =P\left(\beta-\alpha+\mu T \leq y \left\lvert\, T<\frac{\alpha}{\mu}\right.\right) \\
& = \begin{cases}0, & \text { for } y \leq \beta-\alpha \\
\frac{1-e^{-\lambda(y-\beta+\alpha) / \mu}}{1-e^{-\alpha \lambda / \mu}}, & \text { for } \beta-\alpha<y \leq \beta \\
1, & \text { for } y>\beta .\end{cases}
\end{aligned}
$$

Further,

$$
P(N=n)=e^{-\alpha \lambda / \mu}\left(1-e^{-\alpha \lambda / \mu}\right)^{n}, \quad n=0,1,2, \cdots .
$$

Now, observed that $T_{0}$ satisfies the following relation :

$$
T_{0} \frac{\mathcal{D}}{=} \frac{1}{\mu}\left(\beta+\sum_{i=0}^{N} Y_{i}\right)
$$

and hence the distribution function of $T_{0}, L(t)$ say, is given by

$$
L(t)=\sum_{n=0}^{\infty} D^{(n)}(\mu t-\beta) e^{-\alpha \lambda / \mu}\left(1-e^{-\alpha \lambda / \mu}\right)^{n},
$$

where $D^{(n)}$ is the $n$-fold recursive Stieltjes convolution of $D, D^{(0)}$ being the Heaviside function. It can be also shown that

$$
E\left(T_{0}\right)=\frac{\beta-\alpha}{\mu} e^{\alpha \lambda / \mu}+\frac{1}{\lambda}\left(e^{\alpha \lambda / \mu}-1\right) .
$$

## 4. The Probability That the Stock Exceeds a Given Level

We now derive an expression for $\pi_{x}\left(t_{1}, t_{2}\right)=P\{X(t)>x$, for all $\left.t \in\left[t_{1}, t_{2}\right]\right\}$. Since the result is trivial if $x \geq \alpha$, we consider only the case when $x<\alpha$. Observe that $X(t)>x$ for all $t \in\left[t_{1}, t_{2}\right]$ if and only if $X\left(t_{1}\right)>x$ and the first passage time from $X\left(t_{1}\right)$ to $x$ is greater than $t_{2}-t_{1}$. Let $S_{y-x}$ denote the first passage time from state $y$ to state $x$, then

$$
\begin{aligned}
\pi_{x}\left(t_{1}, t_{2}\right) & =P\left\{X\left(t_{1}\right)>x, S_{X\left(t_{1}\right)}-x>t_{2}-t_{1}\right\} \\
& =\int_{x}^{\beta} P\left\{S_{y-x}>t_{2}-t_{1} \mid X\left(t_{1}\right)=y\right\} d F\left(y, t_{1}\right)
\end{aligned}
$$

by conditioning on $X\left(t_{1}\right)$. Let $L_{x}^{y}(t)$ denote the distribution function of $S_{y-x}$. By an argument similar to that of the previous section, it can be shown that

$$
\begin{aligned}
L_{x}^{y}(t)= & D^{(0)}(\mu t+x-y) e^{-\lambda((\alpha \wedge y)-x) / \mu} \\
& +\sum_{n=1}^{\infty} D_{x}^{(n)}(\mu t+x-y)\left(1-e^{-\lambda((\alpha \wedge y)-x) / \mu}\right) \\
& e^{-\lambda(\alpha-x) / \mu}\left(1-e^{-\lambda(\alpha-x) / \mu}\right)^{n-1}
\end{aligned}
$$

where

$$
D_{x}(y)= \begin{cases}0, & \text { for } y<\beta-\alpha \\ \frac{1-e^{-\lambda(y-\beta+\alpha) / \mu}}{1-e^{-\lambda(\alpha-x) / \mu},} & \text { for } \beta-\alpha<y \leq \beta-x \\ 1, & \text { for } y>\beta-x\end{cases}
$$

Summarizing the foregoing, we see that

$$
\pi_{x}\left(t_{1}, t_{2}\right)=\int_{x}^{\beta} L_{x}^{y}\left(t_{2}-t_{1}\right) d F\left(y, t_{1}\right) .
$$

## 5. The Stationary Case

In this section, we consider the case where the distribution function of $X(t)$ does not depend on time $t$, that is, $\partial F(x, t) / \partial t=0$. Notice that this stationary distribution is the same as the equilibrium distribution $F(x)=\lim _{t \rightarrow \infty} F(x, t)$ (c.f. Baxter and Lee [1]).

From the equation (2.1), it follows that

$$
\begin{array}{lrl}
\mu \frac{d}{d x} F(x)-\lambda F(x)=0, & \text { for } & 0 \leq x<\alpha \\
\mu \frac{d}{d x} F(x)-\lambda F(\alpha)=0, & \text { for } & \alpha \leq x<\beta \tag{5.2}
\end{array}
$$

Applying the key renewal theorem to $F(\alpha, t)$ obtained in Lemma 2.1, we see that

$$
\begin{equation*}
F(\alpha)=\frac{\mu}{\mu+\lambda(\beta-\alpha)} \tag{5.3}
\end{equation*}
$$

Hence, solving the equations (5.1) and (5.2) with the boundary condition given by the equation (5.3), we obtain

$$
\begin{aligned}
& F(x)=\frac{\mu e^{\lambda(x-\alpha) / \mu}}{\mu+\lambda(\beta-\alpha)}, \quad \text { for } \quad 0 \leq x<\alpha, \quad \text { and } \\
& F(x)=\frac{\mu-\alpha \lambda+\lambda x}{\mu+\lambda(\beta-\alpha)}, \quad \text { for } \quad \alpha \leq x<\beta
\end{aligned}
$$

From the above stationary distribution, it can be also shown that the average level of the inventory over an infinite horizon is given by

$$
\frac{1}{\mu+\lambda(\beta-\alpha)}\left[\alpha \mu+\frac{\lambda\left(\beta^{2}-\alpha^{2}\right)}{2}-\mu^{2}\left(1-e^{-\alpha \lambda / \mu}\right) / \lambda\right] .
$$

## 6. The Optimal Policy with Respect to $\alpha$

In this section, we show that there exists a unique $\alpha$ which minimizes the average cost per unit time over an infinite horizon, after assigning costs to the inventory, the cost per unit time of the inventory being empty, $C_{1}$ say, and the cost of keeping a unit per unit time, $C_{2}$ say.

To calculate $C(\alpha)$, the average cost per unit time over an infinite horizon, we define as a cycle the interval between two successive points where the inventory is replenished up to $\beta$. Notice again that the sequence of such points forms an embedded renewal process. The duration of a generic interval is denoted $T^{*}$. It can be shown that the total cost during a cycle is given by

$$
\begin{aligned}
& C_{2} \int_{0}^{(\beta-\alpha) / \mu+T}(\beta-\mu x) d x, \quad \text { if } T<\alpha / \mu, \\
& C_{1}(T-\alpha / \mu)+C_{2} \frac{\beta^{2}}{2 \mu}, \quad \text { otherwise, }
\end{aligned}
$$

where $T$ is an exponential random variable with parameter $\lambda$. Hence, the expected total cost in a cycle can be obtained by conditioning on $T$,

$$
\begin{aligned}
\hat{C}(\alpha)= & C_{1} \int_{\alpha / \mu}^{\infty}(t-\alpha / \mu) \lambda e^{-\lambda t} d t \\
& +C_{2} \int_{0}^{\alpha / \mu} \int_{0}^{(\beta-\alpha) / \mu+t}(\beta-\mu x) d x \lambda e^{-\lambda t} d t+C_{2} \int_{\alpha / \mu}^{\beta} \frac{\beta^{2}}{2 \mu} \lambda e^{-\lambda t} d t \\
= & C_{1} e^{-\alpha \lambda / \mu} / \lambda+C_{2}\left(\frac{\beta^{2}}{2 \mu}-\frac{\alpha^{2}}{2 \mu}+\frac{\alpha}{\lambda}+\frac{\dot{\mu}}{\lambda^{2}} e^{-\alpha \lambda / \mu}-\frac{\mu}{\lambda^{2}}\right) .
\end{aligned}
$$

Since $C(\alpha)=\hat{C}(\alpha) / E\left(T^{*}\right)$ and $E\left(T^{*}\right)=((\beta-\alpha) / \mu+1 / \lambda)$, it follows that

$$
C(\alpha)=\frac{1}{(\beta-\alpha) \lambda+\mu}\left[C_{1} \mu e^{-\alpha \lambda / \mu}+C_{2}\left(\beta^{2} \lambda^{2}-\alpha^{2} \lambda^{2}+2 \alpha \mu \lambda+2 \mu^{2} e^{-\alpha \lambda / \mu}-2 \mu^{2}\right) / 2 \lambda\right] .
$$

Theorem 6.1. If $C_{1} \leq C_{2} \beta / 2$, then $C(\alpha)$ is minimized at $\alpha=0$, if $C_{1} \geq C_{2} \mu\left(e^{\lambda \beta / \mu}-1\right) / \lambda$, then $C(\alpha)$ is minimized at $\alpha=\beta$, otherwise, there exists a unique $\alpha^{*}, 0<\alpha^{*}<\beta$, which minimized $C(\alpha)$.
Proof. First, $C^{\prime}(\alpha)$ is given by

$$
\begin{aligned}
C^{\prime}(\alpha) & =\frac{-\lambda(\beta-\alpha)}{((\beta-\alpha) \lambda+\mu)^{2}}\left[\left(C_{1} \lambda+\mu C_{2}\right) e^{-\alpha \lambda / \mu}+C_{2} \alpha \lambda / 2-C_{2} \beta \lambda / 2-C_{2} \mu\right] \\
& =\frac{-\lambda(\beta-\alpha)}{((\beta-\alpha) \lambda+\mu)^{2}}\left[A_{1}(\alpha)-A_{2}(\alpha)\right],
\end{aligned}
$$

$$
\text { where } \begin{aligned}
A_{1}(\alpha) & =\left(C_{1} \lambda+C_{2} \mu\right) e^{-\alpha \lambda / \mu} \quad \text { and } \\
A_{2}(\alpha) & =-C_{2} \alpha \lambda / 2+C_{2} \beta \lambda / 2+C_{2} \mu .
\end{aligned}
$$

Notice that $A_{1}(\alpha)$ is an exponential function of $\alpha$ and $A_{2}(\alpha)$ is a linear function of $\alpha$. There are three cases to consider :
(i) when $C_{1} \leq C_{2} \beta / 2$.

Since $A_{1}(0) \leq A_{2}(0)$ and $A_{1}(\beta) \leq A_{2}(\beta), A_{1}(\alpha) \leq A_{2}(\alpha)$, for all $0 \leq \alpha \leq \beta$. Thus $C^{\prime}(\alpha) \geq 0$, for all $0 \leq \alpha \leq \beta$.
(ii) when $C_{2} \beta / 2<C_{1}<C_{2} \mu\left(e^{\lambda \beta / \mu}-1\right) / \lambda$.

Since $A_{1}(0)>A_{2}(0)$ and $A_{1}(\beta)<A_{2}(\beta)$, there exists a unique $\alpha^{*}, 0<$ $\alpha^{*}<\beta$, which satisfies that $C^{\prime}(\alpha)=0$, and $C(\alpha)$ is minimized at this $\alpha^{*}$. (iii) when $C_{1} \geq C_{2} \mu\left(e^{\lambda \beta / \mu}-1\right) / \lambda$.

For any $0 \leq \alpha \leq \beta$,

$$
\begin{aligned}
A_{1}(\alpha)= & \left(\lambda C_{1}+C_{2} \mu\right) e^{-\alpha \lambda / \mu} \\
\geq & C_{2} \mu e^{\lambda(\beta-\alpha) / \mu} \\
& \text { from the condition that } C_{1} \geq C_{2} \mu\left(e^{\lambda \beta / \mu}-1\right) / \lambda \\
\geq & C_{2} \mu(\lambda(\beta-\alpha) / \mu+1) \\
& \text { Since } e^{x} \geq x+1 \text { for } x \in R \\
= & -C_{2} \alpha \lambda+C_{2} \beta \lambda+C_{2} \mu \\
\geq & -C_{2} \alpha \lambda / 2+C_{2} \beta \lambda / 2+C_{2} \mu \\
= & A_{2}(\alpha) .
\end{aligned}
$$

Thus $C^{\prime}(\alpha) \leq 0$, for all $0 \leq \alpha \leq \beta$.

## References

[1] Baxter, L. A. and Lee, E. Y.(1987), An Inventory with Constant Demand and Poisson Restocking, Probability in the Engineering and Informational Sciences, 1, 203-210.
[2] Colton, D. (1988). Partial Differential Equations, New York: Random House.

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