# Variance of Time to Recruitment for a Two Graded Manpower System with Inter-Decision Times Having Independent and Non-Identically Distributed Random Variables under Geometric Process Wastage 

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#### Abstract

In this paper, an organization subjected to a random exit of personnel due to policy decisions taken by the organization is considered; there is an associated loss of manpower if a person quits the organization. As the exit of personnel is unpredictable, a recruitment policy involving two thresholds, optional and mandatory is suggested to enable the organization to plan its decision on appropriate univariate policy of recruitment. Based on shock model approach, two mathematical models are constructed using an appropriate univariate policy of recruitment. The analytical expressions for mean and variance of time to recruitment is obtained wheni) loss of manpower forms a geometric process ii) inter-decision times forms a sequence of independent and non-identically distributed exponential random variables iii) optional and mandatory thresholds having exponential distribution.


Keywords: Manpower planning, shock models, univariate recruitment policy, hypo-exponential distribution, geometric process.

## I. INTRODUCTION

Many models have been discussed using different kinds of wastage and different types of distributions for the threshold. Such models could be seen in [1],[2],[3]. The problem of time to recruitment is studied by several authors both for a single and multigraded system for different types of thresholds, according as the inter-decision times are independent and identically distributed random variables (or) correlated random variables. For a multi-graded system ,in [7] the author has obtained the performance measures namely mean and variance of the time to recruitment for a two graded system, when (i) loss of manpower and the threshold for the loss of manpower in each grades are exponential random variables (ii) inter-decision times are independent and identically distributed exponential random variables forming the same renewal process for both grades and (iii) thresholds for the organization is the max(min) of the thresholds for the two grades $\max (\mathrm{min})$ model using the above cited univariate cumulative policy of recruitment.

In [8] the author has extended the results for geometric thresholds when the inter-decision times for the two grades form two different renewal processes. In [9] the authors have studied the results in [8] when the threshold for the organization is the sum of the thresholds for the grades .In [8] the work in [9] is studied when the loss of manpower and thresholds are geometric random variables according as the inter-decision times for the two grades are correlated random variables or forming two different renewal processes. This author has also obtained the mean time to recruitment for constant combined thresholds using a univariate max policy of recruitment. In all the earlier research works the monotonicity of inter-decision times which do exists in reality is not taken into account. In [9] the above limitation is removed and the authors have obtained the mean time to recruitment for a single grade manpower system by assuming that (i) inter-decision times form a geometric process in which the monotonicity is inbuilt in the process itself (ii) loss of manpower is a sequence of independent and identically distributed exponential random variables and (iii) distribution of the thresholds for the loss of manpower in the organization is exponential. In [11] the authors have obtained the mean time to recruitment for a two graded manpower system with a univariate policy of recruitment involving combined thresholds using geometric process for inter-decision times. In [12] the authors have obtained mean and variance of time to recruitment for a two graded manpower system with a univariate policy of recruitment involving (i) loss of manpower and inter-decisions times are independent and non-identically distributed exponential random variables (ii) thresholds optional and mandatory follows exponential random variables. Recently in [13], the authors have obtained mean and variance of time to recruitment for a two graded manpower system with a univariate policy of recruitment when i) loss of manpower forms a geometric process ii) inter-decision times forms a sequence of independent and non-identically distributed exponential random variables and iii) optional and mandatory thresholds having exponential distribution.

The objective of the present paper is to study the problem of time to recruitment for a two graded manpower systems and to obtain the mean and variance of time to recruitment using CUM univariate recruitment policy for exponential thresholds with loss of manpower forms a geometric process, inter-decision times forms a sequence of independent and non-identically distributed exponential random variables and optional and mandatory thresholds having exponential distribution. The analytical results obtained and are numerically illustrated and the influence of nodal parameters on the mean and variable of time to recruitment is studied.

## II. Notations

$\mathrm{X}_{1} \quad: \quad$ Exponential distribution with parameter $\theta$.
$\mathrm{S}_{\mathrm{k}} \quad$ :
$\mathrm{J}_{\mathrm{i}}($. Cumulative loss of man hours in the first $\mathrm{k}^{\text {th }}$ decisions $\mathrm{k}=1,2, \ldots \mathrm{~S}_{\mathrm{k}}=\sum_{i=1}^{k} \mathrm{X}_{\mathrm{i}}$
Probability distribution function of $\mathrm{X}_{\mathrm{i}}$, the $\mathrm{i}^{\text {th }}$ term of geometric process.
$\mathrm{j}_{\mathrm{i}}() \quad:$.$\quad probability density function of \mathrm{S}_{\mathrm{k}}$.
$\mathrm{g}_{\mathrm{k}}{ }^{*}($.$) : Laplace transform of convolution of density function of \mathrm{S}_{\mathrm{k}}$.
$\mathrm{U}_{\mathrm{k}} \quad: \quad$ Inter-decision times are independent and non-identically distributed exponential random variables
between $(\mathrm{k}-1)^{\mathrm{th}}$ and $\mathrm{k}^{\text {th }}$ decisions with parameters $\beta_{i}\left(\beta_{i}>0\right)$.
$\mathrm{F}_{\mathrm{k}}($.$) : Distribution function of \mathrm{U}_{\mathrm{k}}$.
$\mathrm{f}_{\mathrm{k}}() \quad:$.$\quad Probability density function of \mathrm{U}_{\mathrm{k}}$ with mean $\frac{1}{\beta_{i}}\left(\beta_{i}>0\right)$.
$\mathrm{Y}_{1}, \mathrm{Y}_{2}$ : Continuous random variables denoting the optional threshold for grade1 and 2 respectively following an exponential distribution with parameters $\lambda_{1}$ and $\lambda_{2}$ respectively.
$\mathrm{Z}_{1}, \mathrm{Z}_{2}$ : Continuous random variables denoting the mandatory threshold for grade 1 and 2 respectively following an exponential distribution with parameters $\mu_{1}$ and $\mu_{2}$ respectively.
W : Continuous random variable denoting the time to recruitment in the organization.
$\mathrm{p} \quad: \quad$ Probability that the organization is not going for recruitment whenever the total loss of manpower crosses Y.
$\mathrm{V}_{\mathrm{k}}(\mathrm{t})$ : $\quad$ Probability that exactly k decisions are taken in $[0, \mathrm{t})$.
$\mathrm{L}() \quad:$.$\quad Distribution function of \mathrm{W}, \mathrm{l}($.$) : The probability density function of \mathrm{W}$.
$l^{*}() \quad:$.$\quad Laplace transform of 1($.$) .$
E(W) : Expected time to recruitment.
V(W) : Variance of time to recruitment.
CUM policy: Recruitment is done whenever the cumulative loss of manpower crosses the mandatory threshold.
The organization may or may not go for recruitment if the cumulative loss of manpower crosses optional threshold.

## III. Main Result

Analytical results for the above cited measures related to time to recruitment, we are derived for the present model.
The survival function of W is given by
$\mathrm{P}(\mathrm{W}>\mathrm{t})=\sum_{k=0}^{\infty} \mathrm{V}_{\mathrm{k}}(\mathrm{t}) \mathrm{P}\left(\mathrm{S}_{\mathrm{k}}<\mathrm{Y}\right)+\sum_{k=0}^{\infty} \mathrm{V}_{\mathrm{k}}(\mathrm{t}) \mathrm{P}\left(\mathrm{S}_{\mathrm{k}} \geq \mathrm{Y}\right) \mathrm{P}\left(\mathrm{S}_{\mathrm{k}}<\mathrm{Z}\right) \mathrm{p}$

### 3.1 Model I : Maximum Model $\mathbf{Y}=\operatorname{Max}\left(\mathbf{Y}_{1}, \mathbf{Y}_{2}\right), \mathbf{Z}=\operatorname{Max}\left(\mathbf{Z}_{1}, \mathbf{Z}_{2}\right)$.

Conditioning upon $\mathrm{S}_{\mathrm{k}}$ and using the law of total probability in equation (3), it can be shown that
$\mathrm{P}\left(\mathrm{S}_{\mathrm{k}}<\mathrm{Y}\right) \quad=\int_{0}^{\infty} P\left(S_{k}<Y \mid S_{k}=x\right) g_{k}(x) d x$

$$
\begin{equation*}
=g_{k}^{*}\left(\lambda_{1}\right)+g_{k}^{*}\left(\lambda_{2}\right)-g_{k}^{*}\left(\lambda_{1}+\lambda_{2}\right) \tag{2}
\end{equation*}
$$

ie. $\mathrm{P}\left(\mathrm{S}_{\mathrm{k}}<\mathrm{Y}\right)=\mathrm{D}_{1}+\mathrm{D}_{2}-\mathrm{D}_{3}$
$\mathrm{P}\left(\mathrm{S}_{\mathrm{k}}<\mathrm{Z}\right) \quad=g_{k}^{*}\left(\mu_{1}\right)+g_{k}^{*}\left(\mu_{2}\right)-g_{k}^{*}\left(\mu_{1}+\mu_{2}\right)$
$\mathrm{P}\left(\mathrm{S}_{\mathrm{k}}<\mathrm{Y}\right) \quad=\mathrm{D}_{4}+\mathrm{D}_{5}-\mathrm{D}_{6}$
Using (2) and (3) in (1) and on simplification, the tail distribution of time to recruitment is given by
$\mathrm{P}(\mathrm{W}>\mathrm{t}) \quad=\sum_{k=0}^{\infty} V_{k}(t) A_{k}$
Where $\quad A_{k}=\left[\left(D_{1}+D_{2}-D_{3}\right)\left(1-p\left(\left(D_{4}+D_{5}-D_{6}\right)\right)+\mathrm{p}\left(D_{4}+D_{5}-D_{6}\right)\right]\right.$
and
$\mathrm{D}_{1}=g_{k}^{*}\left(\lambda_{1}\right), \mathrm{D}_{2}=g_{k}^{*}\left(\lambda_{2}\right), \mathrm{D}_{3}=g_{k}^{*}\left(\lambda_{1}+\lambda_{2}\right), \mathrm{D}_{4}=g_{k}^{*}\left(\mu_{1}\right), \mathrm{D}_{5}=g_{k}^{*}\left(\mu_{1}\right), \mathrm{D}_{6}=g_{k}^{*}\left(\mu_{1}+\mu_{2}\right)$.
From renewal theory Medhi (2009), it is known that
$\mathrm{V}_{\mathrm{k}}(\mathrm{t})=\mathrm{F}_{\mathrm{k}}(\mathrm{t})-\mathrm{F}_{\mathrm{k}+1}(\mathrm{t})$ with $\mathrm{F}_{0}(\mathrm{t})=1$.
Since $\mathrm{l}(\mathrm{t})=-\frac{d}{d t} P(\mathrm{~W}>\mathrm{t})$, from (7) it can be shown that
$l^{*}(\mathrm{~s})=\sum_{k=0}^{\infty}\left[f_{k+1}^{*}(s)-f_{k}^{*}(s)\right] A_{k}$
Since
$\mathrm{E}(\mathrm{W})=-\left[\frac{d}{d s} l^{*}(\mathrm{~s})\right]_{\mathrm{s}=0}$
$\mathrm{E}\left(\mathrm{W}^{2}\right)=\left[\frac{d^{2}}{d s^{2}} l^{*}(\mathrm{~s})\right]_{\mathrm{s}=0}$
$\mathrm{V}(\mathrm{W})=\mathrm{E}\left(\mathrm{W}^{2}\right)-(E(W))^{2}$
Since $\left\{\mathrm{X}_{\mathrm{i}}\right\}$ is a geometric process then distribution and density function of $\left\{\mathrm{X}_{\mathrm{i}}\right\}$ are given below
$\mathrm{J}_{\mathrm{i}}(\mathrm{x})=\mathrm{P}\left(\mathrm{X}_{\mathrm{i}} \leq x\right)=\mathrm{G}\left(\alpha^{k-1} x\right), \mathrm{k}=1,2,3, \ldots, \mathrm{~J}_{\mathrm{i}}(\mathrm{x})=\alpha^{k-1} \mathrm{~g}\left(\alpha^{k-1} x\right)$
$\mathrm{J}_{\mathrm{i}}^{*}(\mathrm{~s})=\alpha^{k-1} g^{*}\left(\alpha^{k-1} s\right)=g^{*}\left(\frac{s}{\alpha^{k-1}}\right)$
Since $\mathrm{g}_{\mathrm{k}}{ }^{*}(\mathrm{~s})=\prod_{r=1}^{k} g^{*}\left(\frac{s}{\alpha^{r-1}}\right), \mathrm{k}=1,2,3 \ldots$.
From (10), let us consider $\mathrm{g}_{\mathrm{k}}{ }^{*}\left(\lambda_{1}\right)=\prod_{r=1}^{k} g^{*}\left(\frac{\lambda_{1}}{\alpha^{r-1}}\right)$
$X_{1}$ forms an exponential distribution with parameter $\theta, \mathrm{g}(\mathrm{x})=\theta e^{-\theta x}$
$g^{*}\left(\lambda_{1}\right)=\prod_{r=1}^{k} \int_{0}^{\infty} e^{-\left(\frac{\lambda_{1}}{\alpha^{r-1}}\right) x} \theta e^{-\theta x} \mathrm{dx}$
$g^{*}\left(\lambda_{1}\right)=\prod_{r=1}^{k}\left[\frac{\theta \alpha^{r-1}}{\left(\alpha^{r-1}+\lambda_{1}\right)}\right]=G_{1}$
Similarly
$g^{*}\left(\lambda_{2}\right)=\prod_{r=1}^{k}\left[\frac{\theta \alpha^{r-1}}{\left(\alpha^{r-1}+\lambda_{2}\right)}\right]=G_{2}, g^{*}\left(\lambda_{1}+\lambda_{2}\right)=\prod_{r=1}^{k}\left[\frac{\theta \alpha^{r-1}}{\alpha^{r-1}+\left(\lambda_{1}+\lambda_{2}\right)}\right]=G_{3}$
$g^{*}\left(\mu_{1}\right)=\prod_{r=1}^{k}\left[\frac{\theta \alpha^{r-1}}{\left(\alpha^{r-1}+\mu_{1}\right)}\right]=G_{4}, g^{*}\left(\mu_{2}\right)=\prod_{r=1}^{k}\left[\frac{\theta \alpha^{r-1}}{\left(\alpha^{r-1}+\mu_{2}\right)}\right]=G_{5}, g^{*}\left(\mu_{1}+\mu_{2}\right)=\prod_{r=1}^{k}\left[\frac{\theta \alpha^{r-1}}{\alpha^{r-1}+\left(\mu_{1}+\mu_{2}\right)}\right]=G_{6}$
Using the above results in (5), it can be shown that
$l^{*}(\mathrm{~s})=\sum_{k=0}^{\infty}\left[f_{k+1}^{*}(s)-f_{k}^{*}(s)\right]\left[\left(G_{1}+G_{2}-G_{3}\right)\left(1-p\left(\left(G_{4}+G_{5}-G_{6}\right)\right)+\mathrm{p}\left(G_{4}+G_{5}-G_{6}\right)\right]\right.$
Inter-decision times form independent and non-identically distributed exponential random variables, so it follows hypo-exponential distribution with parameters $\beta_{i}$
The probability density function is $\mathrm{f}_{\mathrm{k}}(\mathrm{t})=\sum_{i=1}^{k} b_{i} \beta_{i} e^{-\beta_{i} t}$ and the Laplace transform is
$f_{k}^{*}(s)=\sum_{i=1}^{k} b_{i} \frac{\beta_{i}}{\beta_{i}+s}$, where $b_{i}=\prod_{\substack{j=1 \\ j \neq i}}^{k} \frac{\beta_{j}}{\beta_{j}-\beta_{i}}$
Using (13) in (12), it can be shown that
$l^{*}(\mathrm{~s})=\sum_{k=0}^{\infty} \sum_{i=1}^{k+1}\left(b_{i} \frac{\beta_{i}}{\beta_{i}+s}\right)\left[\left(G_{1}+G_{2}-G_{3}\right)\left(1-p\left(\left(G_{4}+G_{5}-G_{6}\right)\right)+\mathrm{p}\left(G_{4}+G_{5}-G_{6}\right)\right]\right.$
From (7) and (8), we get
$\mathrm{E}(\mathrm{W})=\sum_{k=0}^{\infty} \frac{1}{\beta_{k+1}}\left[\left(G_{1}+G_{2}-G_{3}\right)\left(1-p\left(\left(G_{4}+G_{5}-G_{6}\right)\right)+\mathrm{p}\left(G_{4}+G_{5}-G_{6}\right)\right]\right.$
Equation (15) gives meantime to recruitment for maximum model.
$\mathrm{E}\left(\mathrm{W}^{2}\right)=\left[\frac{d^{2}}{d s^{2}} \psi^{*}(s)\right]_{\mathrm{s}=0}=2 \sum_{k=0}^{\infty} \frac{1}{\beta_{k+1}{ }^{2}}\left[\left(G_{1}+G_{2}-G_{3}\right)\left(1-p\left(\left(G_{4}+G_{5}-G_{6}\right)\right)+\mathrm{p}\left(G_{4}+G_{5}-G_{6}\right)\right]\right.$
Using (15) and (16) in (9), we get variance of time to recruitment for maximum model.

### 3.1.1 Numerical Illustrations

The analytical expressions for the performance measures namely mean and variance of time to recruitment are analyzed numerically by varying one parameter and keeping other parameters fixed. The effect of nodal parameter ' $\alpha$ ' on the performance measures are shown in the following tables.

## Effect of $\boldsymbol{\alpha}$ on performance measures

$\beta_{2}=0.006, \beta_{3}=0.007, \beta_{4}=0.008, \beta_{5}=0.009, \beta_{6}=0.010, \lambda_{1}=0.01, \lambda_{2}=0.02, \mu_{1}=0.03, \mu_{2}=0.04, p=0.06, \theta=0.05$, For any value of $\beta_{1}, \mathrm{~A}_{0}=0$.

| $\boldsymbol{\alpha}>\mathbf{1}$ | $\mathbf{1 . 5}$ | $\mathbf{1 . 8}$ | $\mathbf{2 . 5}$ | $\mathbf{3 . 5}$ | $\mathbf{4 . 5}$ |
| :--- | :--- | :--- | :--- | :--- | :--- |
| $\mathbf{E}(\mathbf{W})$ | 16.6751 | 16.6762 | 16.6779 | 16.6793 | 16.6804 |
| $\mathbf{V}(\mathbf{W})$ | $3.7752 \times 10^{3}$ | $3.7753 \times 10^{3}$ | $3.7755 \times 10^{3}$ | $3.7756 \times 10^{3}$ | $3.7757 \times 10^{3}$ |
| $\boldsymbol{\alpha}<\mathbf{1}$ | $\mathbf{0 . 9 5 7}$ | $\mathbf{0 . 8 2 5}$ | $\mathbf{0 . 7 3 5}$ | $\mathbf{0 . 6 5 5}$ | $\mathbf{0 . 5 6 3}$ |
| $\mathbf{E}(\mathbf{W})$ | 16.6706 | 16.6683 | 16.6661 | 16.6634 | 16.6588 |
| $\mathbf{V}(\mathbf{W})$ | $3.7747 \times 10^{3}$ | $3.7745 \times 10^{3}$ | $3.7742 \times 10^{3}$ | $3.7740 \times 10^{3}$ | $3.7735 \times 10^{3}$ |

## Findings:

1. As $\alpha>1$, the $X_{k}$ 's, $k=1,2,3 \ldots \ldots . . n$ form a increasing sequence and hence the loss of man hours between decision epochs will increase. Consequently, the expected time to recruitment and the variance of the time to recruitment increase.
2. As $\alpha<1$, the $\mathrm{X}_{\mathrm{k}}$ 's, $\mathrm{k}=1,2,3 \ldots \ldots . . \mathrm{n}$ form a decreasing sequence and hence the loss of man hours between decision epochs will decrease. Consequently, the expected time to recruitment and the variance of the time to recruitment decrease.

### 3.2Model II : Minimum Model $\mathbf{Y}=\operatorname{Min}\left(\mathbf{Y}_{1}, \mathbf{Y}_{2}\right), \mathbf{Z}=\operatorname{Min}\left(\mathbf{Z}_{1}, \mathbf{Z}_{2}\right)$.

Proceeding as in model I it can be shown for the present model that
$\mathrm{P}\left(\mathrm{S}_{\mathrm{k}}<\mathrm{Y}\right)=g_{k}^{*}\left(\lambda_{1}+\lambda_{2}\right)=\mathrm{D}_{3}$
$\mathrm{P}\left(\mathrm{S}_{\mathrm{k}}<\mathrm{Z}\right) \quad=g_{k}^{*}\left(\mu_{1}+\mu_{2}\right)=D_{6}$
$\mathrm{P}(\mathrm{W}>\mathrm{t})=\sum_{k=0}^{\infty} V_{k}(t)\left[D_{3}\left(1-p D_{6}\right)+\mathrm{p} D_{6}\right]$
$\left.\left.\mathrm{E}(\mathrm{W})=\sum_{k=0}^{\infty} \frac{1}{\beta_{k+1}}\left[G_{3}\right)\left(1-p G_{6}\right)+\mathrm{p} G_{6}\right)\right]$
Equation (20) gives meantime to recruitment for minimum model.
$\mathrm{E}\left(\mathrm{W}^{2}\right)=2 \sum_{k=0}^{\infty} \frac{1}{\beta_{k+1}{ }^{2}}\left[G_{3}\left(1-p G_{6}\right)+\mathrm{p} G_{6}\right]$
Using (20) and (21) in (9), we get variance of time to recruitment for minimum model.

### 3.2.1 Numerical Illustrations

The analytical expressions for the performance measures namely mean and variance of time to recruitment are analyzed numerically by varying one parameter and keeping other parameters fixed. The effect of nodal parameter ' $\alpha$ ' on the performance measures are shown in the following tables.

## Effect of $\boldsymbol{\alpha}$ on performance measures

$\beta_{2}=0.006, \beta_{3}=0.007, \beta_{4}=0.008, \beta_{5}=0.009, \beta_{6}=0.010, \lambda_{1}=0.01, \lambda_{2}=0.02, \mu_{1}=0.03, \mu_{2}=0.04, \mathrm{p}=0.06, \theta=0.05$

| $\boldsymbol{\alpha}>\mathbf{1}$ | $\mathbf{1 . 5}$ | $\mathbf{1 . 8}$ | $\mathbf{2 . 5}$ | $\mathbf{3 . 5}$ | $\mathbf{4 . 5}$ |
| :--- | :--- | :--- | :--- | :--- | :--- |
| $\mathbf{E}(\mathbf{W})$ | 16.0062 | 16.0337 | 16.0724 | 16.1012 | 16.1174 |
| $\mathbf{V}(\mathbf{W})$ | $3.6506 \times 10^{3}$ | $3.6537 \times 10^{3}$ | $3.6579 \times 10^{3}$ | $3.6610 \times 10^{3}$ | $3.6628 \times 10^{3}$ |
| $\boldsymbol{\alpha}<\mathbf{1}$ | $\mathbf{0 . 9 5 7}$ | $\mathbf{0 . 8 2 5}$ | $\mathbf{0 . 7 3 5}$ | $\mathbf{0 . 6 5 5}$ | $\mathbf{0 . 5 6 3}$ |
| $\mathbf{E}(\mathbf{W})$ | 15.9125 | 15.8712 | 15.8346 | 15.7937 | 15.7325 |
| $\mathbf{V}(\mathbf{W})$ | $3.6404 \times 10^{3}$ | $3.6359 \times 10^{3}$ | $3.6320 \times 10^{3}$ | $3.6275 \times 10^{3}$ | $3.6209 \times 10^{3}$ |

## Findings:

1. As $\alpha>1$, the $X_{k}$ 's, $k=1,2,3 \ldots \ldots . . n$ form a increasing sequence and hence the loss of man hours between decision epochs will increase. Consequently, the expected time to recruitment and the variance of the time to recruitment increase.
2. As $\alpha<1$, the $X_{k}$ 's, $k=1,2,3 \ldots \ldots . . n$ form a decreasing sequence and hence the loss of man hours between decision epochs will decrease. Consequently, the expected time to recruitment and the variance of the time to recruitment decrease.

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