

# Estimation of Mean Time to Recruitment in a Two Graded Manpower System with Depletion and Inter-Decision Times are Independent and Non - Identically Distributed Random Variables

S.Jenita<sup>#1</sup> & S.Sendhamizh Selvi<sup>#2</sup>

<sup>#1</sup> Research Scholar, PG & Research Department of Mathematics, Government Arts College, Trichy-22,TN, India.

<sup>#2</sup> Assistant Professor, PG & Research Department of Mathematics, Government Arts College, Trichy-22,TN, India.

## Abstract

*In this paper, the problem of time to recruitment prevailing in an organization with two grades when it is subjected to loss of manpower due to the policy decisions taken by the organization is studied. Two mathematical models are constructed by considering two types of thresholds in each grade namely optional and mandatory thresholds for the loss of man hours in these grades. Based on shock model approach in reliability theory, two cases are constructed using an appropriate univariate policy of recruitment is suggested in both the grades. Performance measures namely mean and variance of the time to recruitment are obtained for model I when (i)The loss of manpower form a sequence of independent and non-identically distributed exponential random variables (ii)The inter-decision times are independent and non-identically distributed exponential random variables and for model II when (i)The loss of manpower form a sequence of independent and non-identically distributed exponential random variables (ii)The inter-decision times are independent and identically distributed exponential random variables. The mean time to recruitment is obtained for these models when the optional and the mandatory thresholds follows extended exponential distribution.*

## Keywords:

*Manpower planning, Shock models, Univariate recruitment policy, Extended Exponential distribution, Hypo-exponential distribution.*

## 1.INTRODUCTION

Exodus of personnel is a common phenomenon in any marketing organization whenever the organization announces revised policies regarding sales target, revision of wages, incentives and perquisites. This in turn produces loss in man hours, which adversely affects

the sales turnover of the organization. Frequent recruitment is not advisable as it will be expensive due to cost of recruitment. As the loss of man hours is unpredictable, a suitable recruitment policy has to be designed to overcome this loss. In [1] authors have stated a replacement policy for a device, which is exposed to shocks. One univariate recruitment policy based on this replacement policy under shock model approach in reliability theory suggests that recruitment is done whenever the cumulative loss of man hours crosses a particular level known as threshold or breakdown point, otherwise, the organization reaches an uneconomic status. A detailed account of the application of stochastic processes in manpower planning models can be seen from [2], [3]. The problem of finding the time to recruitment is studied for a single grade and multigrade system by several authors under different conditions.

For a single grade system, in [4] authors have suggested a new univariate recruitment policy involving two thresholds for the loss of manpower in which one threshold is optional and the other is mandatory. According to this recruitment policy, if the total loss of man hours crosses optional threshold level the organization may or may not go for recruitment, but if the total loss of manhours crosses the mandatory threshold recruitment is necessary. In this work they have obtained the mean and variance of time to recruitment when (i) the loss of manpower form a sequence of independent and identically distributed exponential random variables (ii) the inter-decision times are independent and identically distributed exponential random variables (iii) the thresholds are exponential random variables. In [5] and [7] the authors have studied this work when the distribution of the thresholds have SCBZ property and the thresholds are geometric random variables respectively, In [6] authors have also studied their work derived in [4] when (i) the distribution of optional threshold is exponential and the distribution of mandatory threshold has SCBZ property

and vice versa and (ii) the inter-decision times are correlated. In [8] authors using a bivariate recruitment policy, have studied the problem of time to recruitment in a single grade organization when (i) the loss of manpower form a sequence of independent and identically distributed exponential random variables (ii) the inter-decision times are independent and identically distributed exponential random variables (iii) the thresholds are exponential random variables. In [9] authors have extended their work obtained in [8] for correlated inter-decision times. In [10],[11],[12],[13] authors have extended the results of [4] for a two grade system according as the thresholds are exponential random variables or geometric random variables or SCBZ property possessing random variables or extended exponential random variables. In [14] authors have also extended the results of [6] for a two graded system. In [15] authors have also extended the results of [6] for the two grade system according as the thresholds for the first grade have SCBZ property and those for second grade are extended exponential random variables. In [16], the author has obtained mean time to recruitment for a two graded manpower system using the univariate cumulative recruitment policy considering optional and mandatory thresholds for both the grades.

The objective of the present paper is to obtain the mean time to recruitment for a two graded system using the univariate cumulative recruitment policy considering optional and mandatory thresholds for both the grades. The present paper convert the results of V.Vasudevan and A. Srinivasan [16] for a two graded manpower system when the loss of manpower and inter decision times are independent and non-identical distributed random variables. The distribution of the optional and the mandatory thresholds are extended exponential random variables.

**I. MODEL DESCRIPTION FOR MODEL - I**

- i. Consider an organization having two grades 1 & 2 with univariate policy of recruitment which takes decisions at random epoch in  $(0, \infty)$ .
- ii. At every decision making epoch a random number of persons quit the organization.
- iii. There is an associated loss of man hours to the organization if a person quits.
- iv. The loss of man hours and inter-decision times are at any decision forms a sequence of independent and non-identically distributed random variables.
- v. The loss of man hours process and the process of inter-decision times are statistically independent.
- vi. The loss of man hours are linear and cumulative.

**III. NOTATIONS**

$X_i$  : The loss of manpowers due to the  $i^{th}$  decision epoch  $i=1,2,3,\dots$  forming a sequence of independent and non-identically distributed exponential random variables with parameter  $\alpha_i (\alpha_i > 0)$ .

$G_i(.)$  : The Distribution function of  $X_i$

$g_i(.)$  : The probability density function of  $X_i$  with mean  $\frac{1}{\alpha_i} (\alpha_i > 0)$

$S_k$  : Cumulative loss of manpower in the first  $k$ -decisions ( $k=1,2,3,\dots$ )

That is 
$$S_k = \sum_{i=0}^k X_i$$

$G_k(.)$  : The Distribution function of sum of  $k$  independent and non-identically distributed exponential random variables.

$g_k(.)$  : The probability density function of  $S_k$

we note that 
$$G_k(t) = \sum_{i=1}^k c_i (1 - e^{-\alpha_i t}) ,$$

$$g_k(t) = \sum_{i=1}^k c_i \alpha_i e^{-\alpha_i t} , \quad g_k^*(s) = \sum_{i=1}^k c_i \frac{\alpha_i}{\alpha_i + s}$$

Where 
$$c_i = \prod_{\substack{j=1 \\ j \neq i}}^k \frac{\alpha_j}{\alpha_j - \alpha_i} , i = 1,2,3..k$$

$U_i$  : The inter-decision times are independent and non-identically distributed exponential random variables between  $(i-1)$  and  $i^{th}$  decisions with parameter  $\beta_i (\beta_i > 0)$

$F_i(.)$  : The Distribution function of  $U_i$

$f_i(.)$  : The probability density function of  $U_i$  with mean  $\frac{1}{\beta_i} (\beta_i > 0)$

$R_k$  : The waiting time upto  $k$  decisions

$F_k(.)$  : The Distribution function of  $R_k$

$f_k(.)$  : The probability density function of  $R_k$

we note that 
$$F_k(t) = \sum_{i=1}^k b_i (1 - e^{-\beta_i t}) ,$$

$$f_k(t) = \sum_{i=1}^k b_i \beta_i e^{-\beta_i t} , \quad f_k^*(s) = \sum_{i=1}^k b_i \frac{\beta_i}{\beta_i + s}$$

Where 
$$b_i = \prod_{\substack{j=1 \\ j \neq i}}^k \frac{\beta_j}{\beta_j - \beta_i} , i = 1,2,3..k$$

$Y_1, Y_2$  : The continuous random variables denoting the optional thresholds levels for the grade 1 and grade 2 follows extended exponential distribution with parameters  $\lambda_1$  &  $\lambda_2$  respectively.

$Z_1, Z_2$  : The continuous random variables denoting

the mandatory thresholds levels for the grade 1 and grade 2 follows extended exponential distribution with parameters  $\mu_1$  &  $\mu_2$  respectively.

W : The continuous random variable denoting the time to recruitment in the organization.

p : The probability that the organization is not going for recruitment whenever the total loss of manpower crosses the optional threshold Y.

$V_k(t)$  : The probability that exactly k decisions are taken in  $[0,t)$ .

$L(\cdot)$  : Distribution function of W

$l(\cdot)$  : The probability density function of W

$l^*(\cdot)$  : The laplace transform of  $l(\cdot)$

$E(W)$  : The expected 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 the optional threshold.**

**IV.MAIN RESULTS**

The survival function of W is given by

$P(W > t) = \sum_{k=0}^{\infty}$  [Probability that exactly k-decisions are taken in  $[0,t)$ ,  $k = 0,1,2,\dots$ ] [Probability that the total number of exits in these k-decisions does not cross the optional level Y or the total number of exits in these k- decisions crosses the optional level Y but lies below the mandatory level Z and the organization is not making recruitment]

$$P(W>t) = \sum_{k=0}^{\infty} V_k(t) P(S_k < Y) + \sum_{k=0}^{\infty} V_k(t) P(S_k \geq Y) P(S_k < Z) p \quad (1)$$

**Case (i):**

For maximum case, Consider  $P(S_k < Y)$ , Conditioning upon  $S_k$  and using the law of total probability ,We get

$$\begin{aligned} P(S_k < Y) &= \int_0^{\infty} P(S_k < Y | S_k = x) g_k(x) dx \\ &= \int_0^{\infty} P(Y = \max(Y_1, Y_2) > x) g_k(x) dx \\ &= \int_0^{\infty} [1 - P(Y \leq x)] g_k(x) dx \\ &= \int_0^{\infty} [1 - P(Y_1 \leq x) P(Y_2 \leq x)] g_k(x) dx \\ &= \int_0^{\infty} [1 - (1 - e^{-\lambda_1 x})^2 (1 - e^{-\lambda_2 x})^2] g_k(x) dx \\ &= 2 \int_0^{\infty} e^{-\lambda_1 x} g_k(x) dx + 2 \int_0^{\infty} e^{-\lambda_2 x} g_k(x) dx \\ &\quad - 4 \int_0^{\infty} e^{-(\lambda_1 + \lambda_2)x} g_k(x) dx + 2 \int_0^{\infty} e^{-(2\lambda_1 + 2\lambda_2)x} g_k(x) dx \\ &\quad + 2 \int_0^{\infty} e^{-(\lambda_1 + 2\lambda_2)x} g_k(x) dx \end{aligned}$$

$$\begin{aligned} & - \int_0^{\infty} e^{-2\lambda_1 x} g_k(x) dx - \int_0^{\infty} e^{-2\lambda_2 x} g_k(x) dx \\ & - \int_0^{\infty} e^{-(2\lambda_1 + 2\lambda_2)x} g_k(x) dx \end{aligned}$$

$$\begin{aligned} P(S_k < Y) &= 2g_k^*(\lambda_1) + 2g_k^*(\lambda_2) - 4g_k^*(\lambda_1 + \lambda_2) \\ & + 2g_k^*(2\lambda_1 + \lambda_2) + 2g_k^*(\lambda_1 + 2\lambda_2) \\ & - g_k^*(2\lambda_1) - g_k^*(2\lambda_2) - g_k^*(2\lambda_1 + 2\lambda_2) \end{aligned} \quad (2)$$

$$P(S_k < Y) = 2D_1 + 2D_2 - 4D_3 + 2D_4 + 2D_5 - D_6 - D_7 - D_8 \quad (3)$$

Where  $D_1 = g_k^*(\lambda_1)$ ,  $D_2 = g_k^*(\lambda_2)$ ,  $D_3 = g_k^*(\lambda_1 + \lambda_2)$ ,  
 $D_4 = g_k^*(2\lambda_1 + \lambda_2)$ ,  $D_5 = g_k^*(\lambda_1 + 2\lambda_2)$ ,  
 $D_6 = g_k^*(2\lambda_1)$ ,  $D_7 = g_k^*(2\lambda_2)$ ,  
 $D_8 = g_k^*(2\lambda_1 + 2\lambda_2)$ ,

Similarly

$$\begin{aligned} (S_k < Z) &= 2 \int_0^{\infty} e^{-\mu_1 x} g_k(x) dx + 2 \int_0^{\infty} e^{-\mu_2 x} g_k(x) dx \\ & - 4 \int_0^{\infty} e^{-(\mu_1 + \mu_2)x} g_k(x) dx \\ & + 2 \int_0^{\infty} e^{-(2\mu_1 + \mu_2)x} g_k(x) dx \\ & + 2 \int_0^{\infty} e^{-(\mu_1 + 2\mu_2)x} g_k(x) dx \\ & - \int_0^{\infty} e^{-2\mu_1 x} g_k(x) dx - \int_0^{\infty} e^{-2\mu_2 x} g_k(x) dx \\ & - \int_0^{\infty} e^{-(2\mu_1 + 2\mu_2)x} g_k(x) dx \end{aligned}$$

$$\begin{aligned} P(S_k < Z) &= 2g_k^*(\mu_1) + 2g_k^*(\mu_2) - 4g_k^*(\mu_1 + \mu_2) \\ & + 2g_k^*(2\mu_1 + \mu_2) + 2g_k^*(\mu_1 + 2\mu_2) \\ & - g_k^*(2\mu_1) - g_k^*(2\mu_2) - g_k^*(2\mu_1 + 2\mu_2) \end{aligned} \quad (4)$$

$$P(S_k < Z) = 2D_9 + 2D_{10} - 4D_{11} + 2D_{12} + 2D_{13} - D_{14} - D_{15} - D_{16} \quad (5)$$

Where  $D_9 = g_k^*(\mu_1)$ ,  $D_{10} = g_k^*(\mu_2)$ ,  $D_{11} = g_k^*(\mu_1 + \mu_2)$ ,  
 $D_{12} = g_k^*(2\mu_1 + \mu_2)$ ,  $D_{13} = g_k^*(\mu_1 + 2\mu_2)$ ,  $D_{14} = g_k^*(2\mu_1)$ ,  
 $D_{15} = g_k^*(2\mu_2)$ ,  $D_{16} = g_k^*(2\mu_1 + 2\mu_2)$ ,

Substituting (3) & (5) in (1), we get

$$\begin{aligned} P(W > t) &= \sum_{k=0}^{\infty} V_k(t) \{ (2D_1 + 2D_2 - 4D_3 + 2D_4 + 2D_5 - D_6 - D_7 - D_8) \\ & + p[1 - (2D_1 + 2D_2 - 4D_3 + 2D_4 + 2D_5 - D_6 - D_7 - D_8)] \\ & [2D_9 + 2D_{10} - 4D_{11} + 2D_{12} + 2D_{13} - D_{14} - D_{15} - D_{16}] \} \end{aligned}$$

$$\begin{aligned} P(W > t) &= \sum_{k=0}^{\infty} V_k(t) \{ (2D_1 + 2D_2 - 4D_3 + 2D_4 + 2D_5 - D_6 - D_7 - D_8) \\ & [1 - p(2D_9 + 2D_{10} - 4D_{11} + 2D_{12} + 2D_{13} - D_{14} - D_{15} - D_{16})] \\ & + p(2D_9 + 2D_{10} - 4D_{11} + 2D_{12} + 2D_{13} - D_{14} - D_{15} - D_{16}) \} \end{aligned} \quad (6)$$

$$P(W > t) = \sum_{k=0}^{\infty} V_k(t) \{ B_k(1 - pC_k) + pC_k \} \quad (7)$$

Where  $B_k = 2D_1 + 2D_2 - 4D_3 + 2D_4 + 2D_5 - D_6 - D_7 - D_8$ ,  $C_k =$

$$C_k = 2D_9+2D_{10}-4D_{11}+2D_{12}+2D_{13}-D_{14}-D_{15}-D_{16}$$

$$P(W>t) = \sum_{k=0}^{\infty} V_k(t) A_k, \text{ where } A_k=B_k(1-pC_k)+pC_k \quad (8)$$

From renewal theory (Medhi),

$$P(W>t) = \sum_{k=0}^{\infty} [F_k(t) - F_{k+1}(t)]A_k \quad (9)$$

$$P(W>t) = \sum_{k=0}^{\infty} F_k(t)A_k - \sum_{k=0}^{\infty} F_{k+1}(t)A_k$$

Since  $L(t) = 1 - P(W>t)$  and  $l(t) = \frac{d}{dt}L(t)$ , we get

$$L(t) = 1 - \sum_{k=0}^{\infty} F_k(t)A_k + \sum_{k=0}^{\infty} F_{k+1}(t)A_k$$

$$l(t) = - \sum_{k=0}^{\infty} f_k(t)A_k + \sum_{k=0}^{\infty} f_{k+1}(t)A_k \quad (10)$$

Taking Laplace Transform on both sides, we get

$$l^*(s) = \sum_{k=0}^{\infty} f_{k+1}^*(s) A_k - \sum_{k=0}^{\infty} f_k^*(s) A_k \quad (11)$$

Where  $f_k^*(s) = \sum_{i=1}^k \frac{b_i \beta_i}{s + \beta_i}$

The mean time to recruitment, is known that

$$E(W) = - \left[ \frac{d}{ds} l^*(s) \right]_{s=0} \quad (12)$$

Now

$$- \frac{d}{ds} f_k^*(s)|_{s=0} = E[U_1+U_2+U_3+\dots+U_k] = \sum_{i=1}^k \frac{1}{\beta_i} \quad (13)$$

From the equations (11),(12) & (13), we get

$$E(W) = \sum_{k=0}^{\infty} \sum_{i=1}^{k+1} \frac{1}{\beta_i} A_k - \sum_{k=0}^{\infty} \sum_{i=1}^k \frac{1}{\beta_i} A_k$$

$$E(W) = \sum_{k=0}^{\infty} \frac{1}{\beta_{k+1}} A_k \quad (14)$$

Equation (14) gives the mean time to recruitment for maximum case in Model I.

Where  $A_k = (2D_1+2D_2-4D_3+2D_4+2D_5-D_6-D_7-D_8)$   
 $[1-p(2D_9+2D_{10}-4D_{11}+2D_{12}+2D_{13}-D_{14}-D_{15}-D_{16})]$   
 $+ p(2D_9+2D_{10}-4D_{11}-D_{15}-D_{16})]$   
 $+ 2D_{12}+2D_{13}-D_{14}-D_{15}-D_{16})$  and

$$D_1 = g_k^*(\lambda_1), D_2 = g_k^*(\lambda_2), D_3 = g_k^*(\lambda_1 + \lambda_2),$$

$$D_4 = g_k^*(2\lambda_1 + \lambda_2), D_5 = g_k^*(\lambda_1 + 2\lambda_2), D_6 = g_k^*(2\lambda_1)$$

$$D_7 = g_k^*(2\lambda_2), D_8 = g_k^*(2\lambda_1 + 2\lambda_2), D_9 = g_k^*(\mu_1),$$

$$D_{10} = g_k^*(\mu_2), D_{11} = g_k^*(\mu_1 + \mu_2), D_{12} = g_k^*(2\mu_1 + \mu_2),$$

$$D_{13} = g_k^*(\mu_1 + 2\mu_2), D_{14} = g_k^*(2\mu_1), D_{15} = g_k^*(2\mu_2),$$

$$D_{16} = g_k^*(2\mu_1 + 2\mu_2).$$

**Case (ii):**

For minimum case, the optional and mandatory thresholds for the organization are given by

$Y = \min(Y_1, Y_2)$  and  $Z = \min(Z_1, Z_2)$ , with same assumptions and notations of case(i).

As in case(i),

$$P(S_k < Y) = \int_0^{\infty} P(S_k < Y | S_k = x) g_k(x) dx$$

$$= \int_0^{\infty} P(Y = \min(Y_1, Y_2) > x) g_k(x) dx$$

$$= \int_0^{\infty} P(Y_1 > x) P(Y_2 > x) g_k(x) dx$$

$$= \int_0^{\infty} (1 - P(Y_1 \leq x))(1 - P(Y_2 \leq x)) g_k(x) dx$$

$$= \int_0^{\infty} [1 - (1 - e^{-\lambda_1 x})^2][1 - (1 - e^{-\lambda_2 x})^2] g_k(x) dx$$

$$= 4 \int_0^{\infty} e^{-(\lambda_1 + \lambda_2)x} g_k(x) dx$$

$$- 2 \int_0^{\infty} e^{-(2\lambda_1 + \lambda_2)x} g_k(x) dx$$

$$- 2 \int_0^{\infty} e^{-(\lambda_1 + 2\lambda_2)x} g_k(x) dx$$

$$+ \int_0^{\infty} e^{-(2\lambda_1 + 2\lambda_2)x} g_k(x) dx$$

$$P(S_k < Y) = 4g_k^*(\lambda_1 + \lambda_2) - 2g_k^*(2\lambda_1 + \lambda_2) - 2g_k^*(\lambda_1 + 2\lambda_2) + g_k^*(2\lambda_1 + 2\lambda_2) \quad (15)$$

$$P(S_k < Y) = 4D_3 - 2D_4 - 2D_5 + D_8 \quad (16)$$

Where  $D_3 = g_k^*(\mu_1 + \mu_2)$ ,  $D_4 = g_k^*(2\mu_1 + \mu_2)$ ,  
 $D_5 = g_k^*(\mu_1 + 2\mu_2)$ ,  $D_8 = g_k^*(2\mu_1 + 2\mu_2)$

Similarly

$$P(S_k < Z) = 4g_k^*(\mu_1 + \mu_2) - 2g_k^*(2\mu_1 + \mu_2) - 2g_k^*(\mu_1 + 2\mu_2) + g_k^*(2\mu_1 + 2\mu_2) \quad (17)$$

$$(ie) P(S_k < Z) = 4D_{11} - 2D_{12} - 2D_{13} + D_{16} \quad (18)$$

Where  $D_{11} = g_k^*(\mu_1 + \mu_2)$ ,  $D_{12} = g_k^*(2\mu_1 + \mu_2)$ ,  
 $D_{13} = g_k^*(\mu_1 + 2\mu_2)$ ,  $D_{16} = g_k^*(2\mu_1 + 2\mu_2)$

$$P(W>t) = \sum_{k=0}^{\infty} V_k(t) \{ (4D_3 - 2D_4 - 2D_5 + D_8) [1 - p(4D_{11} - 2D_{12} - 2D_{13} + D_{16})] + p(4D_{11} - 2D_{12} - 2D_{13} + D_{16}) \} \quad (19)$$

From Renewal Theory,

$$P(W>t) = \sum_{k=0}^{\infty} V_k(t) \{ B_k(1-pC_k) + pC_k \} \quad (20)$$

Where  $B_k = 4D_3 - 2D_4 - 2D_5 + D_8$ ,

$$C_k = 4D_{11} - 2D_{12} - 2D_{13} + D_{16}$$

$$P(W>t) = \sum_{k=0}^{\infty} V_k(t) A_k, \text{ where } A_k = B_k(1-pC_k) + pC_k \quad (21)$$

As we found in case (i)

$$E(W) = \sum_{k=0}^{\infty} \frac{1}{\beta_{k+1}} A_k \quad (22)$$

Equation (22) gives the mean time to recruitment for minimum case in Model I.

Where  $A_k = (4D_3 - 2D_4 - 2D_5 + D_8)[1 - p(4D_{11} - 2D_{12}$

$$- 2D_{13} + D_{16})] + p(4D_{11} - 2D_{12} - 2D_{13} + D_{16}) \ \&$$

$$D_3 = g_k^*(\mu_1 + \mu_2), \ D_4 = g_k^*(2\mu_1 + \mu_2),$$

$$D_5 = g_k^*(\mu_1 + 2\mu_2), \ D_8 = g_k^*(2\mu_1 + 2\mu_2),$$

$$D_{11} = g_k^*(\mu_1 + \mu_2), \ D_{12} = g_k^*(2\mu_1 + \mu_2),$$

$$D_{13} = g_k^*(\mu_1 + 2\mu_2), \ D_{16} = g_k^*(2\mu_1 + 2\mu_2)$$

**Model II:**

Description of model II is similar to that of model I except the condition on inter-decision times.

The analytical results for the survival function of time to recruitment when the loss of manpower is independent and non-identically distributed exponential random variables with  $\alpha_i$  and inter-decision times are independent and identically distributed exponential random variables with parameter  $\beta$ . The thresholds optional and mandatory follows extended exponential distribution with parameters  $\lambda_1$  &  $\lambda_2$  and  $\mu_1$  &  $\mu_2$ .

**Case (i):**

As in model I, for maximum case,

We define  $A_k = [2D_1 + 2D_2 - 4D_3 + 2D_4 + 2D_5 - D_6 - D_7 - D_8]$

$$[1 - p(2D_9 + 2D_{10} - 4D_{11} + 2D_{12} + 2D_{13} - D_{14} - D_{15} - D_{16})]$$

$$+ p[2D_9 + 2D_{10} - 4D_{11} + 2D_{12} + 2D_{13} - D_{14} - D_{15} - D_{16}]$$

By definition  $V_k(t) = P[N(t) = k]$ , where  $N(t)$  is the number of decisions taken in  $[0, t)$ .

From (8), we get

$$P(W>t) = \sum_{k=0}^{\infty} V_k(t) \{ [2D_1 + 2D_2 - 4D_3 + 2D_4 + 2D_5 - D_6 - D_7 - D_8][1 - p(2D_9 + 2D_{10} - 4D_{11} + 2D_{12} + 2D_{13} - D_{14} - D_{15} - D_{16})] + p[2D_9 + 2D_{10} - 4D_{11} + 2D_{12} + 2D_{13} - D_{14} - D_{15} - D_{16}] \}$$

$$\begin{aligned} \text{(ie)} P(W>t) &= \sum_{k=0}^{\infty} V_k(t) \{ [2g_k^*(\lambda_1) 2g_k^*(\lambda_2) - 4g_k^*(\lambda_1 + \lambda_2) \\ &+ 2g_k^*(2\lambda_1 + \lambda_2) - g_k^*(2\lambda_1) - g_k^*(2\lambda_2) - g_k^*(2\lambda_1 + 2\lambda_2)] \\ &[1 - p(2g_k^*(\mu_1) + 2g_k^*(\mu_2) - 4g_k^*(\mu_1 + \mu_2) + 2g_k^*(2\mu_1 + \mu_2) \\ &- g_k^*(2\mu_1) - g_k^*(2\mu_2) - g_k^*(2\mu_1 + 2\mu_2))] + p[2g_k^*(\mu_1) \\ &+ 2g_k^*(\mu_2) - 4g_k^*(\mu_1 + \mu_2) + 2g_k^*(2\mu_1 + \mu_2) \\ &- g_k^*(2\mu_1) - g_k^*(2\mu_2) - g_k^*(2\mu_1 + 2\mu_2)] \} \\ &= \sum_{k=0}^{\infty} V_k(t) A_k(\lambda_1, \lambda_2, \mu_1, \mu_2) \end{aligned}$$

Now,

$$P(W>t) = \sum_{k=0}^{\infty} P[N(t)=k] \{ A_k(\lambda_1, \lambda_2, \mu_1, \mu_2) \} \quad (23)$$

Since  $W$  is a non-negative continuous random variable, its known from [Medhi] that

$$E[W] = \int_0^{\infty} P(W>t) dt$$

Therefore from the equation (23), we get

$$E[W] = \int_0^{\infty} \sum_{k=0}^{\infty} P[N(t)=k] \{ A_k(\lambda_1, \lambda_2, \mu_1, \mu_2) \} dt \quad (24)$$

Since  $N(t)$  is a Poisson process with rate  $\beta$  by hypothesis, we can write the equation (24) as

$$\begin{aligned} E[W] &= \sum_{k=0}^{\infty} A_k(\lambda_1, \lambda_2, \mu_1, \mu_2) \int_0^{\infty} \frac{e^{-\beta t} (\beta t)^k}{k!} dt \\ &= \sum_{k=0}^{\infty} A_k(\lambda_1, \lambda_2, \mu_1, \mu_2) \frac{\beta^k}{k!} \int_0^{\infty} t^k e^{-\beta t} dt \\ &= \sum_{k=0}^{\infty} A_k(\lambda_1, \lambda_2, \mu_1, \mu_2) \frac{\beta^k}{k!} \frac{k!}{\beta^{k+1}} \end{aligned}$$

$$E[W] = \frac{1}{\beta} \sum_{k=0}^{\infty} A_k(\lambda_1, \lambda_2, \mu_1, \mu_2) \quad (25)$$

Equation (25) gives the mean time to recruitment for maximum case in model II

**Case (ii):**

For minimum case,

We define  $A_k = [4D_3 - 2D_4 - 2D_5 + D_8]$

$$[1 - p(4D_{11} - 2D_{12} - 2D_{13} + D_{16})] + p[4D_{11} - 2D_{12} - 2D_{13} + D_{16}]$$

As in case (i),

$$P(W>t) = \sum_{k=0}^{\infty} V_k(t) \{ [4D_3 - 2D_4 - 2D_5 + D_8]$$

$$\begin{aligned}
 & [1-p(4D_{11}-2D_{12}-2D_{13}+D_{16}) \\
 & + p[4D_{11}-2D_{12}-2D_{13}+D_{16}]] \\
 & = \sum_{k=0}^{\infty} V_k(t) \{ [4g_k^*(\lambda_1 + \lambda_2) - 2g_k^*(2\lambda_1 + \lambda_2) \\
 & - 2g_k^*(\lambda_1 + 2\lambda_2) + g_k^*(2\lambda_1 + 2\lambda_2)] \\
 & [1 - p(4g_k^*(\mu_1 + \mu_2) - 2g_k^*(2\mu_1 + \mu_2) \\
 & - 2g_k^*(\mu_1 + 2\mu_2) + g_k^*(2\mu_1 + 2\mu_2))] \\
 & + p[4g_k^*(\mu_1 + \mu_2) - 2g_k^*(2\mu_1 + \mu_2) \\
 & - 2g_k^*(\mu_1 + 2\mu_2) + g_k^*(2\mu_1 + 2\mu_2)] \} \\
 & = \sum_{k=0}^{\infty} V_k(t) A_k(\lambda_1, \lambda_2, \mu_1, \mu_2)
 \end{aligned}$$

$$P(W>t) = \sum_{k=0}^{\infty} P[N(t) = k] \{ A_k(\lambda_1, \lambda_2, \mu_1, \mu_2) \}$$

$$\text{and } E[W] = \int_0^{\infty} \sum_{k=0}^{\infty} P[N(t) = k] \{ A_k(\lambda_1, \lambda_2, \mu_1, \mu_2) \} dt$$

$$\begin{aligned}
 \text{(ie) } E[W] &= \sum_{k=0}^{\infty} A_k(\lambda_1, \lambda_2, \mu_1, \mu_2) \int_0^{\infty} \frac{e^{-\beta t}(\beta t)^k}{k!} dt \\
 &= \sum_{k=0}^{\infty} A_k(\lambda_1, \lambda_2, \mu_1, \mu_2) \frac{\beta^k}{k!} \int_0^{\infty} t^k e^{-\beta t} dt \\
 &= \sum_{k=0}^{\infty} A_k(\lambda_1, \lambda_2, \mu_1, \mu_2) \frac{\beta^k}{k!} \frac{k!}{\beta^{k+1}}
 \end{aligned}$$

$$E[W] = \frac{1}{\beta} \sum_{k=0}^{\infty} A_k(\lambda_1, \lambda_2, \mu_1, \mu_2) \quad (26)$$

Equation (26) gives the mean time to recruitment for minimum case in model II

**SPECIAL CASES :**

**Case (i) :**

The optional threshold  $Y_1, Y_2$  follows exponential distribution with parameters  $(\lambda_1, \lambda_2)$  and the mandatory threshold  $Z_1, Z_2$  follows extended exponential distribution with parameters  $(\mu_1, \mu_2)$ , the loss of manpower and the inter-decision times are independent and non-identically distributed exponential random variables with parameters  $\alpha_i, \beta_i$

In maximum case,

$$P(S_k < Y) = \int_0^{\infty} [1 - (1 - e^{-\lambda_1 x})(1 - e^{-\lambda_2 x})] g_k(x) dx$$

$$\begin{aligned}
 & = \int_0^{\infty} e^{-\lambda_1 x} g_k(x) dx + \int_0^{\infty} e^{-\lambda_2 x} g_k(x) dx \\
 & - \int_0^{\infty} e^{-(\lambda_1 + \lambda_2)x} g_k(x) dx
 \end{aligned}$$

$$P(S_k < Y) = g_k^*(\lambda_1) + g_k^*(\lambda_2) - g_k^*(\lambda_1 + \lambda_2) \quad (27)$$

$$P(S_k < Y) = D_1 + D_2 - D_3 \quad (28)$$

Where  $D_1 = g_k^*(\lambda_1), D_2 = g_k^*(\lambda_2),$

$$D_3 = g_k^*(\lambda_1 + \lambda_2)$$

From model I,

$$\begin{aligned}
 P(S_k < Z) &= 2g_k^*(\mu_1) + 2g_k^*(\mu_2) - 4g_k^*(\mu_1 + \mu_2) \\
 & + 2g_k^*(2\mu_1 + \mu_2) + 2g_k^*(\mu_1 + 2\mu_2) \\
 & - g_k^*(2\mu_1) - g_k^*(2\mu_2) - g_k^*(2\mu_1 + 2\mu_2) \quad (29)
 \end{aligned}$$

$$\begin{aligned}
 P(S_k < Z) &= 2D_9 + 2D_{10} - 4D_{11} + 2D_{12} \\
 & + 2D_{13} - D_{14} - D_{15} - D_{16} \quad (30)
 \end{aligned}$$

Where  $D_9 = g_k^*(\mu_1), D_{10} = g_k^*(\mu_2), D_{11} = g_k^*(\mu_1 + \mu_2),$

$$D_{12} = g_k^*(2\mu_1 + \mu_2), D_{13} = g_k^*(\mu_1 + 2\mu_2)$$

$$D_{14} = g_k^*(2\mu_1), D_{15} = g_k^*(2\mu_2),$$

$$D_{16} = g_k^*(2\mu_1 + 2\mu_2)$$

Using the equations (28) and (30) in (1), we get

$$\begin{aligned}
 P(W>t) &= \sum_{k=0}^{\infty} V_k(t) \{ (D_1 + D_2 - D_3) + p(1 - (D_1 + D_2 - D_3)) \\
 & (2D_9 + 2D_{10} - 4D_{11} + 2D_{12} + 2D_{13} - D_{14} - D_{15} - D_{16}) \}
 \end{aligned}$$

$$\begin{aligned}
 P(W>t) &= \sum_{k=0}^{\infty} V_k(t) \{ [D_1 + D_2 - D_3] [1 - p(2D_9 + 2D_{10} \\
 & - 4D_{11} + 2D_{12} + 2D_{13} - D_{14} - D_{15} - D_{16})] \\
 & + p[2D_9 + 2D_{10} - 4D_{11} + 2D_{12} + 2D_{13} - D_{14} - D_{15} - D_{16}] \}
 \end{aligned}$$

$$P(W>t) = \sum_{k=0}^{\infty} V_k(t) A_k,$$

$$E(W) = \sum_{k=0}^{\infty} \frac{1}{\beta_{k+1}} A_k \quad (31)$$

$$\begin{aligned}
 \text{Where } A_k &= [g_k^*(\lambda_1) + g_k^*(\lambda_2) - g_k^*(\lambda_1 + \lambda_2)] \\
 & [1 - p(2g_k^*(\mu_1) + 2g_k^*(\mu_2) - 4g_k^*(\mu_1 + \mu_2) + 2g_k^*(2\mu_1 + \mu_2) \\
 & + 2g_k^*(\mu_1 + 2\mu_2) - g_k^*(2\mu_1) - g_k^*(2\mu_2) - g_k^*(2\mu_1 + 2\mu_2))] \\
 & + p[2g_k^*(\mu_1) + 2g_k^*(\mu_2) - 4g_k^*(\mu_1 + \mu_2) + 2g_k^*(2\mu_1 + \mu_2) \\
 & + 2g_k^*(\mu_1 + 2\mu_2) - g_k^*(2\mu_1) - g_k^*(2\mu_2) - g_k^*(2\mu_1 + 2\mu_2)]
 \end{aligned}$$

In minimum case,

$$P(S_k < Y) = \int_0^{\infty} e^{-(\lambda_1 + \lambda_2)x} g_k(x) dx$$

$$P(S_k < Y) = g_k^*(\lambda_1 + \lambda_2)$$

$$P(S_k < Y) = D_3, \text{ Where } D_3 = g_k^*(\lambda_1 + \lambda_2) \quad (32)$$

As in model I,

$$P(S_k < Z) = 4g_k^*(\mu_1 + \mu_2) - 2g_k^*(2\mu_1 + \mu_2) - 2g_k^*(\mu_1 + 2\mu_2) + g_k^*(2\mu_1 + 2\mu_2)$$

$$P(S_k < Z) = 4D_{11} - 2D_{12} - 2D_{13} + D_{16} \quad (33)$$

Where  $D_{11} = g_k^*(\mu_1 + \mu_2)$ ,

$D_{12} = g_k^*(2\mu_1 + \mu_2), D_{13} = g_k^*(\mu_1 + 2\mu_2)$ ,

$D_{16} = g_k^*(2\mu_1 + 2\mu_2)$

$$P(W > t) = \sum_{k=0}^{\infty} V_k(t) \{ [D_3][1 - p(4D_{11} - 2D_{12} - 2D_{13} + D_{16})] + p[4D_{11} - 2D_{12} - 2D_{13} + D_{16}] \}$$

$$P(W > t) = \sum_{k=0}^{\infty} V_k(t) A_k$$

$$E(W) = \sum_{k=0}^{\infty} \frac{1}{\beta_{k+1}} A_k \quad (34)$$

Where  $A_k = [g_k^*(\lambda_1 + \lambda_2)]$

$$[1 - p(4g_k^*(\mu_1 + \mu_2) - 2g_k^*(2\mu_1 + \mu_2) - 2g_k^*(\mu_1 + 2\mu_2) + g_k^*(2\mu_1 + 2\mu_2))] + p[4g_k^*(\mu_1 + \mu_2) - 2g_k^*(2\mu_1 + \mu_2) - 2g_k^*(\mu_1 + 2\mu_2) + g_k^*(2\mu_1 + 2\mu_2)]$$

**Note : 1**

The optional threshold  $Y_1, Y_2$  follows exponential distribution with parameters  $(\lambda_1, \lambda_2)$  and the mandatory threshold  $Z_1, Z_2$  follows extended exponential distribution with parameters  $(\mu_1, \mu_2)$ , the loss of manpower are independent and non-identically distributed exponential random variables with parameters  $\alpha_i$  and the inter-decision times are independent and identically distributed exponential random variables with parameters  $\beta$ .

In maximum case,

$$E(W) = \frac{1}{\beta} \sum_{k=0}^{\infty} A_k \quad (35)$$

Where

$$A_k = [g_k^*(\lambda_1) + g_k^*(\lambda_2) - g_k^*(\lambda_1 + \lambda_2)]$$

$$[1 - p(2g_k^*(\mu_1) + 2g_k^*(\mu_2) - 4g_k^*(\mu_1 + \mu_2) + 2g_k^*(2\mu_1 + \mu_2) + 2g_k^*(\mu_1 + 2\mu_2) - g_k^*(2\mu_1) - g_k^*(2\mu_2) - g_k^*(2\mu_1 + 2\mu_2))] + p[2g_k^*(\mu_1) + 2g_k^*(\mu_2) - 4g_k^*(\mu_1 + \mu_2) + 2g_k^*(2\mu_1 + \mu_2) + 2g_k^*(\mu_1 + 2\mu_2) - g_k^*(2\mu_1) - g_k^*(2\mu_2) - g_k^*(2\mu_1 + 2\mu_2)]$$

In minimum case, the same results of maximum case but

$$A_k = [g_k^*(\lambda_1 + \lambda_2)][1 - p(4g_k^*(\mu_1 + \mu_2) - 2g_k^*(2\mu_1 + \mu_2) - 2g_k^*(\mu_1 + 2\mu_2) + g_k^*(2\mu_1 + 2\mu_2))] + p[4g_k^*(\mu_1 + \mu_2) - 2g_k^*(2\mu_1 + \mu_2) - 2g_k^*(\mu_1 + 2\mu_2) + g_k^*(2\mu_1 + 2\mu_2)]$$

**Case (ii) :**

If we interchange the distributions, we get the same results as in case (i) with different parameters.

In maximum case,

$$E(W) = \sum_{k=0}^{\infty} \frac{1}{\beta_{k+1}} A_k$$

(36)

$$\text{Where } A_k = [2g_k^*(\lambda_1) + 2g_k^*(\lambda_2) - 4g_k^*(\lambda_1 + \lambda_2) + 2g_k^*(2\lambda_1 + \lambda_2) + 2g_k^*(\lambda_1 + 2\lambda_2) - g_k^*(2\lambda_1) - g_k^*(2\lambda_2) - g_k^*(2\lambda_1 + 2\lambda_2)] [1 - p(g_k^*(\mu_1) + g_k^*(\mu_2) - g_k^*(\mu_1 + \mu_2))] + p[g_k^*(\mu_1) + g_k^*(\mu_2) - g_k^*(\mu_1 + \mu_2)]$$

In minimum case,

$$E(W) = \sum_{k=0}^{\infty} \frac{1}{\beta_{k+1}} A_k$$

Where

$$A_k = [4g_k^*(\lambda_1 + \lambda_2) - 2g_k^*(2\lambda_1 + \lambda_2) - 2g_k^*(\lambda_1 + 2\lambda_2) + g_k^*(2\lambda_1 + 2\lambda_2)] [1 - p(g_k^*(\mu_1 + \mu_2))] + p[g_k^*(\mu_1 + \mu_2)]$$

**Note:2**

The optional threshold  $Y_1, Y_2$  follows extended exponential distribution with parameters  $(\lambda_1, \lambda_2)$  and the mandatory threshold  $Z_1, Z_2$  follows exponential distribution with parameters  $(\mu_1, \mu_2)$ , the loss of manpower are independent and non-identically distributed exponential random variables with parameters  $\alpha_i$  and the inter-decision times are independent and identically distributed exponential random variables with parameters  $\beta$ .

In maximum case,

$$E(W) = \frac{1}{\beta} \sum_{k=0}^{\infty} A_k \quad (38)$$

Where

$$A_k = [2g_k^*(\lambda_1) + 2g_k^*(\lambda_2) - 4g_k^*(\lambda_1 + \lambda_2) + 2g_k^*(2\lambda_1 + \lambda_2) + 2g_k^*(\lambda_1 + 2\lambda_2) - g_k^*(2\lambda_1) - g_k^*(2\lambda_2)]$$

$$-g_k^*(2\lambda_1+2\lambda_2)] [1-p(g_k^*(\mu_1) + g_k^*(\mu_2) - g_k^*(\mu_1 + \mu_2))] + p[g_k^*(\mu_1) + g_k^*(\mu_2) - g_k^*(\mu_1 + \mu_2)]$$

In minimum case, the same results of maximum case but

$$A_k = [4g_k^*(\lambda_1+\lambda_2) - 2g_k^*(2\lambda_1+\lambda_2) - 2g_k^*(\lambda_1+2\lambda_2) + g_k^*(2\lambda_1+2\lambda_2)] [1-p(g_k^*(\mu_1+\mu_2))] + p[g_k^*(\mu_1+\mu_2)]$$

## V.CONCLUSION

The manpower planning model developed in this work is more general compared to earlier work in the context to non-identical nature of loss of manpower, Inter-decision times. The stochastic models developed in this work can be used to plan for the adequate provision of manpower for the organization at graduate, professional and management levels in the context of attrition. There is a scope for studying the applicability of the designed models using simulation. Further, by collecting relevant data, one can test the goodness of fit for the distributions assumed in this work. The findings given in this work enable one to estimate manpower gap in future, thereby facilitating the assessment of manpower profile in predicting future manpower development not only on industry but in a wider domain. The present work can be studied for a two sources of depletion.

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