

# Estimation of Stress-Strength Reliability Model Using Finite Mixture of M-Transformed Exponential Distributions

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## Abstract

*In this paper Stress –Strength reliability is studied where various cases have been considered for stress (Y) and strength (X) variables viz., the strength follows finite mixture of M-Transformed Exponential distributions and stress follows exponential, Lindley and M-Transformed Exponential distributions. The reliability of a system and the parameters are obtained by the method of maximum likelihood method. At the end results are illustrated with the help of numerical evaluations and real life data.*

**Key words:** Reliability, M-Transformed Exponential Distribution, Stress, Strength.

## 1. Introduction

The term "stress-strength reliability" specify the quantity  $R=P(Y < X)$ , where X is a random strength and Y random stress such that the system fails if the stress Y exceeds the strength Y. Church and Harris imported the term stress-strength for the first time. The stress-strength reliability and its problems for considerable distributions have been discussed by Church and Harris [1970], Woodward and Kelley [1977], Beg and Singh [1979], Awad and Gharraf [1986], Surles and Padgett [1998, 2001], Raqab and Kundu [2005], Mokhlis [2005], and Saraçoğlu et al. [2011]. Kotz et al. [2003] have presented an inspection of all approaches and results on the stress-strength reliability in the last four decades. Adil H. Khan and T.R Jan [2014] have studied the stress-strength reliability for two parameter Lindley distribution, Exponential distribution and Gamma distribution. And have considered different conditions for stress and strength variables.

In the present paper, we discuss stress strength reliability and have considered that the strength variable follows finite mixture of M-Transformed Exponential distribution and stress variables follows finite mixture of exponential or Lindley or M-Transformed Exponential distributions. The structural properties and uses of M-Transformed Exponential distribution as a lifetime distribution are studied by Dinesh Kumar et al. [2017]. We debate the estimation method for finite mixture of M-Transformed Exponential distribution by the method of maximum likelihood estimation. The M-Transformed Exponential distribution with parameters  $\alpha$  is defined by its probability density function (p.d.f.)

$$f(x; \alpha) = \frac{2e^{-\frac{x}{\alpha}}}{\alpha \left(2 - e^{-\frac{x}{\alpha}}\right)^2} ; \alpha, x > 0 \quad (1)$$

In the present paper, we consider three cases

- 1) Stress follows exponential distribution and strength follows finite mixture of M-Transformed Exponential distribution.
- 2) Stress follows Lindley distribution and strength follows finite mixture of M-Transformed Exponential distribution.
- 3) Stress follows finite mixture of Lindley distributions and strength follows finite mixture of M-Transformed Exponential distributions.
- 4) Stress and strength both follows finite mixture of M-Transformed Exponential distribution.

## 2. Statistical Model

In this model we assume that random variables  $X$  (strength) and  $Y$  (stress) are independent and the values of  $X$  and  $Y$  are non-negative. The reliability of a component with strength  $X$  and stress  $Y$  imposed on it is given by

$$R = P(X > Y) = \int_0^{\infty} \left[ \int_0^x g(y) dy \right] f(x) dx \quad (2.1)$$

where  $f(x)$  and  $g(y)$  are pdf of strength and stress respectively.

A finite mixture of M-Transformed Exponential distributions with  $v$  components can be expressed as

$$f(x) = p_1 f_1(x) + p_2 f_2(x) + \dots + p_v f_v(x); p_i > 0, i = 1, 2, \dots, v, \sum_{i=1}^v p_i = 1 \quad (2.2)$$

## 3. Reliability Computations

Let  $X$  be the strength of the  $v$ -components with probability density functions  $f_i(x); i = 1, 2, \dots, v$ . The pdf of  $X$  which follows finite mixture of M-Transformed Exponential distributions is

$$f_i(x) = p_i \frac{2e^{-\frac{x}{\alpha_i}}}{\alpha_i \left( 2 - e^{-\frac{x}{\alpha_i}} \right)^2}; x > 0, \alpha_i > 0, p_i > 0, i = 1, 2, \dots, v, \sum_{i=1}^v p_i = 1$$

### Case I: The stress $Y$ follows exponential distribution

As  $Y$  follows exponential distribution, pdf of  $Y$  is given by

$$g(y) = \lambda e^{-\lambda y}, \lambda > 0, y > 0$$

For two components  $v = 2$ , and

$$f(x) = p_1 \frac{2e^{-\frac{x}{\alpha_1}}}{\alpha_1 \left( 2 - e^{-\frac{x}{\alpha_1}} \right)^2} + p_2 \frac{2e^{-\frac{x}{\alpha_2}}}{\alpha_2 \left( 2 - e^{-\frac{x}{\alpha_2}} \right)^2}; p_1 + p_2 = 1, \alpha_1, \alpha_2, x > 0$$

As  $X$  and  $Y$  are independent then from (2), Reliability function  $R_2$  is

$$\begin{aligned} R_2 &= \int_0^{\infty} \int_0^x \lambda e^{-\lambda y} \left[ p_1 \frac{2e^{-\frac{x}{\alpha_1}}}{\alpha_1 \left( 2 - e^{-\frac{x}{\alpha_1}} \right)^2} + p_2 \frac{2e^{-\frac{x}{\alpha_2}}}{\alpha_2 \left( 2 - e^{-\frac{x}{\alpha_2}} \right)^2} \right] dx dy \\ &= \int_0^{\infty} \int_0^x \lambda e^{-\lambda y} \left[ p_1 \sum_{m_1=1}^{\infty} \frac{m_1}{2^{m_1} \alpha_1} e^{-\frac{m_1 x}{\alpha_1}} + p_2 \sum_{m_2=1}^{\infty} \frac{m_2}{2^{m_2} \alpha_2} e^{-\frac{m_2 x}{\alpha_2}} \right] dx dy \\ R_2 &= \int_0^{\infty} (1 - e^{-\lambda x}) \sum_{i=1}^2 \sum_{m_i=1}^{\infty} \frac{p_i m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} dx \\ &= 1 - \sum_{i=1}^2 \sum_{m_i=1}^{\infty} \frac{p_i m_i}{2^{m_i} (\lambda \alpha_i + m_i)}; \sum_{i=1}^2 p_i = 1 \end{aligned} \quad (3.1)$$

For three components  $v = 3$ , we have

$$\begin{aligned}
 f(x) &= p_1 \frac{2e^{-\frac{x}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{x}{\alpha_1}}\right)^2} + p_2 \frac{2e^{-\frac{x}{\alpha_2}}}{\alpha_2 \left(2 - e^{-\frac{x}{\alpha_2}}\right)^2} + p_3 \frac{2e^{-\frac{x}{\alpha_3}}}{\alpha_3 \left(2 - e^{-\frac{x}{\alpha_3}}\right)^2} \\
 & \qquad \qquad \qquad ; p_1 + p_2 + p_3 = 1, \alpha_1, \alpha_2, \alpha_3, x > 0 \\
 R_3 &= \int_0^\infty \int_0^x \lambda e^{-\lambda y} \left[ p_1 \frac{2e^{-\frac{x}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{x}{\alpha_1}}\right)^2} + p_2 \frac{2e^{-\frac{x}{\alpha_2}}}{\alpha_2 \left(2 - e^{-\frac{x}{\alpha_2}}\right)^2} + p_3 \frac{2e^{-\frac{x}{\alpha_3}}}{\alpha_3 \left(2 - e^{-\frac{x}{\alpha_3}}\right)^2} \right] dx dy \\
 R_3 &= \int_0^\infty \int_0^x \lambda e^{-\lambda y} \left[ p_1 \sum_{m_1=1}^\infty \frac{m_1}{2^{m_1} \alpha_1} e^{-\frac{m_1 x}{\alpha_1}} + p_2 \sum_{m_2=1}^\infty \frac{m_2}{2^{m_2} \alpha_2} e^{-\frac{m_2 x}{\alpha_2}} + p_3 \sum_{m_3=1}^\infty \frac{m_3}{2^{m_3} \alpha_3} e^{-\frac{m_3 x}{\alpha_3}} \right] dx dy \\
 R_3 &= \int_0^\infty (1 - e^{-\lambda x}) \sum_{i=1}^3 \sum_{m_i=1}^\infty \frac{p_i m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} dx \\
 R_3 &= 1 - \sum_{i=1}^3 \sum_{m_i=1}^\infty \frac{p_i m_i}{2^{m_i} (\lambda \alpha_i + m_i)} \quad ; \quad \sum_{i=1}^3 p_i = 1 \tag{3.2}
 \end{aligned}$$

In general for  $v$ -components,  $f(x) = p_1 f_1(x) + p_2 f_2(x) + \dots + p_v f_v(x)$ ;  $\sum_{i=1}^v p_i = 1$ , we have

$$R_v = 1 - \sum_{i=1}^v \sum_{m_i=1}^\infty \frac{p_i m_i}{2^{m_i} (\lambda \alpha_i + m_i)} \quad ; \quad \sum_{i=1}^v p_i = 1 \tag{3.3}$$

**Case II: The stress Y follows two parameter Lindley distribution**

As Y follows Lindley distribution, pdf of Y is given by

$$g(y) = \frac{\theta^2}{\theta + \lambda} (1 + \lambda y) e^{-\theta y} \quad ; \quad y > 0, \theta > 0, \alpha > -\theta$$

For two components  $v = 2$

$$f(x) = p_1 \frac{2e^{-\frac{x}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{x}{\alpha_1}}\right)^2} + p_2 \frac{2e^{-\frac{x}{\alpha_2}}}{\alpha_2 \left(2 - e^{-\frac{x}{\alpha_2}}\right)^2} \quad ; \quad p_1 + p_2 = 1, \alpha_1, \alpha_2, x > 0$$

As X and Y are independent then from (2), Reliability function  $R_2$  is

$$\begin{aligned}
 R_2 &= \int_0^\infty \int_0^x \frac{\theta^2}{\theta + \lambda} (1 + \lambda y) e^{-\theta y} \left( p_1 \frac{2e^{-\frac{x}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{x}{\alpha_1}}\right)^2} + p_2 \frac{2e^{-\frac{x}{\alpha_2}}}{\alpha_2 \left(2 - e^{-\frac{x}{\alpha_2}}\right)^2} \right) dx dy \\
 R_2 &= \int_0^\infty \int_0^x \frac{\theta^2}{\theta + \lambda} (1 + \lambda y) e^{-\theta y} \left( p_1 \sum_{m_1=1}^\infty \frac{m_1}{2^{m_1} \alpha_1} e^{-\frac{m_1 x}{\alpha_1}} + p_2 \sum_{m_2=1}^\infty \frac{m_2}{2^{m_2} \alpha_2} e^{-\frac{m_2 x}{\alpha_2}} \right) dx dy \\
 R_2 &= \int_0^\infty \left( 1 - \frac{\theta + \lambda + \lambda \theta x}{\theta + \lambda} e^{-\theta x} \right) \left( \sum_{i=1}^2 \sum_{m_i=1}^\infty p_i \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} \right) dx \\
 R_2 &= 1 - \sum_{i=1}^2 \sum_{m_i=1}^\infty p_i \frac{m_i}{2^{m_i} \alpha_i} \left( \int_0^\infty e^{-(\theta + \frac{m_i}{\alpha_i})x} dx + \frac{\lambda \theta}{\theta + \lambda} \int_0^\infty x e^{-(\theta + \frac{m_i}{\alpha_i})x} dx \right) \\
 R_2 &= 1 - \sum_{i=1}^2 \sum_{m_i=1}^\infty p_i \frac{m_i}{2^{m_i}} \left( \frac{1}{m_i + \theta \alpha_i} + \frac{\lambda \theta \alpha_i}{(\theta + \lambda)(m_i + \theta \alpha_i)^2} \right) \tag{3.4}
 \end{aligned}$$

For three components  $v = 3$ , we have

$$\begin{aligned}
 f(x) &= f(x) = p_1 \frac{2e^{-\frac{x}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{x}{\alpha_1}}\right)^2} + p_2 \frac{2e^{-\frac{x}{\alpha_2}}}{\alpha_2 \left(2 - e^{-\frac{x}{\alpha_2}}\right)^2} + p_3 \frac{2e^{-\frac{x}{\alpha_3}}}{\alpha_3 \left(2 - e^{-\frac{x}{\alpha_3}}\right)^2} \\
 & \qquad \qquad \qquad ; p_1 + p_2 + p_3 = 1, \alpha_1, \alpha_2, \alpha_3, x > 0
 \end{aligned}$$

$$\begin{aligned}
 R_3 &= \int_0^\infty \int_0^x \frac{\theta^2}{\theta + \lambda} (1 + \lambda y) e^{-\theta y} \left[ p_1 \frac{2e^{-\frac{x}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{x}{\alpha_1}}\right)^2} + p_2 \frac{2e^{-\frac{x}{\alpha_2}}}{\alpha_2 \left(2 - e^{-\frac{x}{\alpha_2}}\right)^2} + p_3 \frac{2e^{-\frac{x}{\alpha_3}}}{\alpha_3 \left(2 - e^{-\frac{x}{\alpha_3}}\right)^2} \right] dx dy \\
 R_3 &= \int_0^\infty \left(1 - \frac{\theta + \lambda + \lambda \theta x}{\theta + \lambda} e^{-\theta x}\right) \left(\sum_{i=1}^3 \sum_{m_i=1}^\infty p_i \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}}\right) dx \\
 R_3 &= 1 - \sum_{i=1}^3 \sum_{m_i=1}^\infty p_i \frac{m_i}{2^{m_i} \alpha_i} \int_0^\infty \left(e^{-\left(\frac{m_i}{\alpha_i} + \theta\right)x} + \frac{\lambda \theta}{\theta + \lambda} x e^{-\left(\frac{m_i}{\alpha_i} + \theta\right)x}\right) dx \\
 R_3 &= 1 - \sum_{i=1}^3 \sum_{m_i=1}^\infty p_i \frac{m_i}{2^{m_i}} \left(\frac{1}{m_i + \theta \alpha_i} + \frac{\lambda \theta \alpha_i}{(\theta + \lambda)(m_i + \theta \alpha_i)^2}\right) \tag{3.5}
 \end{aligned}$$

In general for  $v$ -components,  $f(x) = p_1 f_1(x) + p_2 f_2(x) + \dots + p_v f_v(x)$ ;  $\sum_{i=1}^v p_i = 1$ , we have

$$R_v = 1 - \sum_{i=1}^v \sum_{m_i=1}^\infty p_i \frac{m_i}{2^{m_i}} \left(\frac{1}{m_i + \theta \alpha_i} + \frac{\lambda \theta \alpha_i}{(\theta + \lambda)(m_i + \theta \alpha_i)^2}\right) \tag{3.6}$$

**Special case**

When  $\lambda = 1$ , two parameter Lindley distribution reduces to one parameter Lindley distribution and then the reliability function is given as

$$R_v = 1 - \sum_{i=1}^v \sum_{m_i=1}^\infty p_i \frac{m_i}{2^{m_i}} \left(\frac{1}{m_i + \theta \alpha_i} + \frac{\theta \alpha_i}{(\theta + 1)(m_i + \theta \alpha_i)^2}\right) \tag{3.7}$$

**Case III: The stress Y follows mixture of two parameter Lindley distributions**

As Y follows mixture of M-Transformed Exponential distributions, pdf of X and Y is given by

For two components,  $v = 2$

$$\begin{aligned}
 f(x) &= p_1 \frac{2e^{-\frac{x}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{x}{\alpha_1}}\right)^2} + p_2 \frac{2e^{-\frac{x}{\alpha_2}}}{\alpha_2 \left(2 - e^{-\frac{x}{\alpha_2}}\right)^2}; \quad p_1 + p_2 = 1, \alpha_1, \alpha_2, x > 0 \\
 g(y) &= p_3 \frac{\theta_3^2}{\theta_3 + \lambda_3} (1 + \lambda_3 y) e^{-\lambda_3 y} + p_4 \frac{\theta_4^2}{\theta_4 + \lambda_4} (1 + \lambda_4 y) e^{-\theta_4 y}; \quad p_3 + p_4 = 1,
 \end{aligned}$$

As X and Y are independent then from (2), Reliability function  $R_2$  is

$$\begin{aligned}
 R_2 &= \int_0^\infty \int_0^x \left( p_1 \frac{2e^{-\frac{x}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{x}{\alpha_1}}\right)^2} + p_2 \frac{2e^{-\frac{x}{\alpha_2}}}{\alpha_2 \left(2 - e^{-\frac{x}{\alpha_2}}\right)^2} \right) \left( p_3 \frac{\theta_3^2}{\theta_3 + \lambda_3} (1 + \lambda_3 y) e^{-\lambda_3 y} \right. \\
 &\quad \left. + p_4 \frac{\theta_4^2}{\theta_4 + \lambda_4} (1 + \lambda_4 y) e^{-\theta_4 y} \right) dx dy \\
 R_2 &= \sum_{j=1+2}^4 \sum_{i=1}^2 \int_0^\infty \int_0^x p_i \sum_{m_i=1}^\infty \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} p_j \frac{\theta_j^2}{\theta_j + \lambda_j} (1 + \lambda_j y) e^{-\lambda_j y} dx dy \\
 R_2 &= \sum_{j=1+2}^4 \sum_{i=1}^2 \int_0^\infty p_i p_j \sum_{m_i=1}^\infty \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} \left(1 - \frac{\theta_j + \lambda_j + \lambda_j \theta_j x}{\theta_j + \lambda_j} e^{-\theta_j x}\right) dx \\
 R_2 &= 1 - \sum_{j=1+2}^4 \sum_{i=1}^2 \sum_{m_i=1}^\infty \frac{p_i p_j m_i}{2^{m_i} \alpha_i} \int_0^\infty \left(e^{-\left(\frac{m_i}{\alpha_i} + \theta_j\right)x} + \frac{\lambda_j \theta_j}{\theta_j + \lambda_j} x e^{-\left(\frac{m_i}{\alpha_i} + \theta_j\right)x}\right) dx \\
 R_2 &= 1 - \sum_{j=1+2}^4 \sum_{i=1}^2 \sum_{m_i=1}^\infty \frac{p_i p_j m_i}{2^{m_i}} \left[ \frac{1}{m_i + \theta_j \alpha_i} + \frac{\lambda_j \theta_j \alpha_i}{(\theta_j + \lambda_j)(m_i + \theta_j \alpha_i)^2} \right] \tag{3.8}
 \end{aligned}$$

For three components  $v = 3$ , we have

$$\begin{aligned}
 f(x) &= p_1 \frac{2e^{-\frac{x}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{x}{\alpha_1}}\right)^2} + p_2 \frac{2e^{-\frac{x}{\alpha_2}}}{\alpha_2 \left(2 - e^{-\frac{x}{\alpha_2}}\right)^2} + p_3 \frac{2e^{-\frac{x}{\alpha_3}}}{\alpha_3 \left(2 - e^{-\frac{x}{\alpha_3}}\right)^2} \\
 &\quad ; \quad p_1 + p_2 + p_3 = 1, \alpha_1, \alpha_2, \alpha_3, x > 0
 \end{aligned}$$

$$g(y) = p_4 \frac{\lambda_4^2}{\lambda_4 + \alpha_4} (1 + \alpha_4 y) e^{-\lambda_4 y} + p_5 \frac{\lambda_5^2}{\lambda_5 + \alpha_5} (1 + \alpha_5 y) e^{-\lambda_5 y} + p_3 \frac{\lambda_6^2}{\lambda_6 + \alpha_6} (1 + \alpha_6 y) e^{-\lambda_6 y}$$

;  $p_4 + p_5 + p_6 = 1, \lambda_4, \lambda_5, \lambda_6, y > 0$

X and Y are independent then from (I), Reliability function R is

$$R_3 = \int_0^\infty \int_0^x \left( p_1 \frac{2e^{-\frac{x}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{x}{\alpha_1}}\right)^2} + p_2 \frac{2e^{-\frac{x}{\alpha_2}}}{\alpha_2 \left(2 - e^{-\frac{x}{\alpha_2}}\right)^2} + p_3 \frac{2e^{-\frac{x}{\alpha_3}}}{\alpha_3 \left(2 - e^{-\frac{x}{\alpha_3}}\right)^2} \right) \times \left( \frac{p_4 \lambda_4^2}{\lambda_4 + \alpha_4} (1 + \alpha_4 y) e^{-\lambda_4 y} + \frac{p_5 \lambda_5^2}{\lambda_5 + \alpha_5} (1 + \alpha_5 y) e^{-\lambda_5 y} + \frac{p_3 \lambda_6^2}{\lambda_6 + \alpha_6} (1 + \alpha_6 y) e^{-\lambda_6 y} \right) dx dy$$

$$R_3 = \sum_{j=1+3}^6 \sum_{i=1}^3 \int_0^\infty \int_0^x p_i \sum_{m_i=1}^\infty \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} p_j \frac{\theta_j^2}{\theta_j + \lambda_j} (1 + \lambda_j y) e^{-\lambda_j y} dx dy$$

$$R_3 = \sum_{j=1+3}^6 \sum_{i=1}^3 \int_0^\infty p_i p_j \sum_{m_i=1}^\infty \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} \left( 1 - \frac{\theta_j + \lambda_j + \lambda_j \theta_j x}{\theta_j + \lambda_j} e^{-\theta_j x} \right) dx$$

$$R_3 = 1 - \sum_{j=1+3}^6 \sum_{i=1}^3 \sum_{m_i=1}^\infty \frac{p_i p_j m_i}{2^{m_i} \alpha_i} \int_0^\infty \left( e^{-\left(\frac{m_i}{\alpha_i} + \theta_j\right)x} + \frac{\lambda_j \theta_j}{\theta_j + \lambda_j} x e^{-\left(\frac{m_i}{\alpha_i} + \theta_j\right)x} \right) dx$$

$$R_3 = 1 - \sum_{j=1+3}^6 \sum_{i=1}^3 \sum_{m_i=1}^\infty \frac{p_i p_j m_i}{2^{m_i}} \left[ \frac{1}{m_i + \theta_j \alpha_i} + \frac{\lambda_j \theta_j \alpha_i}{(\theta_j + \lambda_j)(m_i + \theta_j \alpha_i)^2} \right] \tag{3.9}$$

In general for  $v$ -components,

$$f(x) = p_1 f_1(x) + p_2 f_2(x) + \dots + p_v f_v(x); \quad \sum_{i=1}^v p_i = 1$$

$$g(y) = p_{v+1} f_{v+1}(y) + p_{v+2} f_{v+2}(y) + \dots + p_{2v} f_{2v}(y); \quad \sum_{i=v+1}^{2v} p_i = 1$$

$$R_v = \int_0^\infty \int_0^x (p_1 f_1(x) + p_2 f_2(x) + \dots + p_k f_k(x)) (p_{v+1} f_{v+1}(y) + p_{v+2} f_{v+2}(y) + \dots + p_{2v} f_{2v}(y)) dx dy$$

$$R_v = \sum_{j=1+v}^{2v} \sum_{i=1}^v \int_0^\infty \int_0^x p_i \sum_{m_i=1}^\infty \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} p_j \frac{\theta_j^2}{\theta_j + \lambda_j} (1 + \lambda_j y) e^{-\lambda_j y} dx dy$$

$$R_v = \sum_{j=1+v}^{2v} \sum_{i=1}^v \int_0^\infty p_i p_j \sum_{m_i=1}^\infty \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} \left( 1 - \frac{\theta_j + \lambda_j + \lambda_j \theta_j x}{\theta_j + \lambda_j} e^{-\theta_j x} \right) dx$$

$$R_v = 1 - \sum_{j=1+v}^{2v} \sum_{i=1}^v \sum_{m_i=1}^\infty \frac{p_i p_j m_i}{2^{m_i} \alpha_i} \int_0^\infty \left( e^{-\left(\frac{m_i}{\alpha_i} + \theta_j\right)x} + \frac{\lambda_j \theta_j}{\theta_j + \lambda_j} x e^{-\left(\frac{m_i}{\alpha_i} + \theta_j\right)x} \right) dx$$

$$R_v = 1 - \sum_{j=1+v}^{2v} \sum_{i=1}^v \sum_{m_i=1}^\infty \frac{p_i p_j m_i}{2^{m_i}} \left[ \frac{1}{m_i + \theta_j \alpha_i} + \frac{\lambda_j \theta_j \alpha_i}{(\theta_j + \lambda_j)(m_i + \theta_j \alpha_i)^2} \right] \tag{3.10}$$

**Special case**

When  $\lambda_j = 1$ , two parameter Lindley distribution reduces to one parameter Lindley distribution and then  $R_k$  is the reliability function when X follows mixture of one parameter Lindley distribution and Y follows mixture of one M-Transformed Exponential distribution and is given as

$$R_v = 1 - \sum_{j=1+v}^{2v} \sum_{i=1}^v \sum_{m_i} \frac{p_i p_j m_i}{2^{m_i}} \left[ \frac{1}{m_i + \theta_j \alpha_i} + \frac{\theta_j \alpha_i}{(\theta_j + 1)(m_i + \theta_j \alpha_i)^2} \right] \tag{3.11}$$

**Case IV: The stress Y follows mixture of M-Transformed Exponential distributions**

As Y follows mixture of M-Transformed Exponential distributions, pdf of X and Y is given by

For two components,  $v = 2$

$$f(x) = p_1 \frac{2e^{-\frac{x}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{x}{\alpha_1}}\right)^2} + p_2 \frac{2e^{-\frac{x}{\alpha_2}}}{\alpha_2 \left(2 - e^{-\frac{x}{\alpha_2}}\right)^2}; \quad p_1 + p_2 = 1, \alpha_1, \alpha_2, x > 0$$

$$g(y) = p_3 \frac{2e^{-\frac{y}{\alpha_3}}}{\alpha_3 \left(2 - e^{-\frac{y}{\alpha_3}}\right)^2} + p_4 \frac{2e^{-\frac{y}{\alpha_4}}}{\alpha_4 \left(2 - e^{-\frac{y}{\alpha_4}}\right)^2}; \quad p_3 + p_4 = 1,$$

As X and Y are independent then from (2), Reliability function  $R_2$  is

$$R_2 = \int_0^\infty \int_0^x \left( p_1 \frac{2e^{-\frac{x}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{x}{\alpha_1}}\right)^2} + p_2 \frac{2e^{-\frac{x}{\alpha_2}}}{\alpha_2 \left(2 - e^{-\frac{x}{\alpha_2}}\right)^2} \right) \left( p_3 \frac{2e^{-\frac{y}{\alpha_3}}}{\alpha_3 \left(2 - e^{-\frac{y}{\alpha_3}}\right)^2} + p_4 \frac{2e^{-\frac{y}{\alpha_4}}}{\alpha_4 \left(2 - e^{-\frac{y}{\alpha_4}}\right)^2} \right) dx dy$$

$$R_2 = \sum_{j=i+2}^4 \sum_{i=1}^2 \int_0^\infty \int_0^x p_i \sum_{m_i=1}^\infty \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} p_j \sum_{m_j=1}^\infty \frac{m_j}{2^{m_j} \alpha_j} e^{-\frac{m_j x}{\alpha_j}} dx dy$$

$$R_2 = \sum_{j=i+2}^4 \sum_{i=1}^2 \int_0^\infty p_i p_j \sum_{m_i=1}^\infty \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} \left( 1 - \sum_{m_j=1}^\infty \frac{1}{2^{m_j}} e^{-\frac{m_j x}{\alpha_j}} \right) dx$$

$$R_2 = 1 - \sum_{j=i+2}^4 \sum_{i=1}^2 p_i p_j \int_0^\infty \left( \sum_{m_i=1}^\infty \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} \right) \left( \sum_{m_j=1}^\infty \frac{1}{2^{m_j}} e^{-\frac{m_j x}{\alpha_j}} \right) dx$$

$$R_2 = 1 - \sum_{j=i+2}^4 \sum_{i=1}^2 p_i p_j \left( \frac{\alpha_j}{\alpha_i + \alpha_j} \right) \tag{3.12}$$

For three components,  $v = 3$

$$f(x) = p_1 \frac{2e^{-\frac{x}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{x}{\alpha_1}}\right)^2} + p_2 \frac{2e^{-\frac{x}{\alpha_2}}}{\alpha_2 \left(2 - e^{-\frac{x}{\alpha_2}}\right)^2} + p_3 \frac{2e^{-\frac{x}{\alpha_3}}}{\alpha_3 \left(2 - e^{-\frac{x}{\alpha_3}}\right)^2}; \quad p_1 + p_2 + p_3 = 1$$

$$g(y) = p_4 \frac{2e^{-\frac{y}{\alpha_4}}}{\alpha_4 \left(2 - e^{-\frac{y}{\alpha_4}}\right)^2} + p_5 \frac{2e^{-\frac{y}{\alpha_5}}}{\alpha_5 \left(2 - e^{-\frac{y}{\alpha_5}}\right)^2} + p_6 \frac{2e^{-\frac{y}{\alpha_6}}}{\alpha_6 \left(2 - e^{-\frac{y}{\alpha_6}}\right)^2}; \quad p_4 + p_5 + p_6 = 1,$$

As X and Y are independent then from (2), Reliability function  $R_2$  is

$$R_3 = \int_0^\infty \int_0^x \left( p_1 \frac{2e^{-\frac{x}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{x}{\alpha_1}}\right)^2} + p_2 \frac{2e^{-\frac{x}{\alpha_2}}}{\alpha_2 \left(2 - e^{-\frac{x}{\alpha_2}}\right)^2} + p_3 \frac{2e^{-\frac{x}{\alpha_3}}}{\alpha_3 \left(2 - e^{-\frac{x}{\alpha_3}}\right)^2} \right) \left( p_4 \frac{2e^{-\frac{y}{\alpha_4}}}{\alpha_4 \left(2 - e^{-\frac{y}{\alpha_4}}\right)^2} \right. \\ \left. + p_5 \frac{2e^{-\frac{y}{\alpha_5}}}{\alpha_5 \left(2 - e^{-\frac{y}{\alpha_5}}\right)^2} + p_6 \frac{2e^{-\frac{y}{\alpha_6}}}{\alpha_6 \left(2 - e^{-\frac{y}{\alpha_6}}\right)^2} \right) dx dy$$

$$R_3 = \sum_{j=i+3}^6 \sum_{i=1}^3 \int_0^\infty \int_0^x p_i \sum_{m_i=1}^\infty \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} p_j \sum_{m_j=1}^\infty \frac{m_j}{2^{m_j} \alpha_j} e^{-\frac{m_j x}{\alpha_j}} dx dy$$

$$R_3 = \sum_{j=i+3}^6 \sum_{i=1}^3 \int_0^\infty p_i p_j \sum_{m_i=1}^\infty \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} \left( 1 - \sum_{m_j=1}^\infty \frac{1}{2^{m_j}} e^{-\frac{m_j x}{\alpha_j}} \right) dx$$

$$R_3 = 1 - \sum_{j=i+3}^6 \sum_{i=1}^3 p_i p_j \int_0^\infty \left( \sum_{m_i=1}^\infty \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} \right) \left( \sum_{m_j=1}^\infty \frac{1}{2^{m_j}} e^{-\frac{m_j x}{\alpha_j}} \right) dx$$

$$R_3 = 1 - \sum_{j=i+3}^6 \sum_{i=1}^3 p_i p_j \left( \frac{\alpha_j}{\alpha_i + \alpha_j} \right) \tag{3.13}$$

In general for  $v$ -components,

$$\begin{aligned}
 f(x) &= p_1 f_1(x) + p_2 f_2(x) + \dots + p_v f_v(x); \quad \sum_{i=1}^v p_i = 1 \\
 g(y) &= p_{v+1} f_{v+1}(y) + p_{v+2} f_{v+2}(y) + \dots + p_{2v} f_{2v}(y); \quad \sum_{i=v+1}^{2v} p_i = 1 \\
 R_v &= \int_0^{\infty} \int_0^x (p_1 f_1(x) + p_2 f_2(x) + \dots + p_v f_v(x)) (p_{v+1} f_{v+1}(y) + p_{v+2} f_{v+2}(y) + \dots + p_{2v} f_{2v}(y)) dx dy \\
 R_v &= \sum_{j=i+v}^{2v} \sum_{i=1}^v \int_0^{\infty} \int_0^x p_i \sum_{m_i=1}^{\infty} \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} p_j \sum_{m_j=1}^{\infty} \frac{m_j}{2^{m_j} \alpha_j} e^{-\frac{m_j y}{\alpha_j}} dx dy \\
 R_v &= \sum_{j=i+v}^{2v} \sum_{i=1}^v \int_0^{\infty} p_i p_j \sum_{m_i=1}^{\infty} \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} \left( 1 - \sum_{m_j=1}^{\infty} \frac{1}{2^{m_j}} e^{-\frac{m_j}{\alpha_j} x} \right) dx \\
 R_v &= 1 - \sum_{j=i+2v}^{2v} \sum_{i=1}^v p_i p_j \int_0^{\infty} \left( \sum_{m_i=1}^{\infty} \frac{m_i}{2^{m_i} \alpha_i} e^{-\frac{m_i x}{\alpha_i}} \right) \left( \sum_{m_j=1}^{\infty} \frac{1}{2^{m_j}} e^{-\frac{m_j}{\alpha_j} x} \right) dx \\
 R_v &= 1 - \sum_{j=i+v}^{2v} \sum_{i=1}^v p_i p_j \left( \frac{\alpha_j}{\alpha_i + \alpha_j} \right) \tag{3.14}
 \end{aligned}$$

#### 4. Maximum Likelihood Estimation

$$\begin{aligned}
 L(\alpha_1, \alpha_2 p_1 / \hat{y}) &= \prod_{j=1}^n \left[ p_1 \frac{2e^{-\frac{x_i}{\alpha_1}}}{\alpha_1 \left( 2 - e^{-\frac{x_i}{\alpha_1}} \right)^2} + p_2 \frac{2e^{-\frac{x_i}{\alpha_2}}}{\alpha_2 \left( 2 - e^{-\frac{x_i}{\alpha_2}} \right)^2} \right] \\
 L(\alpha_1, \alpha_2 p_1 / \hat{y}) &= \frac{2^n n!}{n_1! n_2!} \left( \frac{p_1}{\alpha_1} \right)^{n_1} \left( \frac{p_2}{\alpha_2} \right)^{n_2} \prod_{i=1}^{n_1} \frac{e^{-\frac{x_i}{\alpha_1}}}{\left( 2 - e^{-\frac{x_i}{\alpha_1}} \right)^2} \prod_{i=1}^{n_2} \frac{e^{-\frac{x_i}{\alpha_2}}}{\left( 2 - e^{-\frac{x_i}{\alpha_2}} \right)^2} \\
 \log L(\alpha_1, \alpha_2 p_1 / \hat{y}) &= \log \left( \frac{2^n n!}{n_1! n_2!} \right) + n_1 \log p_1 - n_1 \log \alpha_1 + n_2 \log (1 - p_1) - n_2 \log \alpha_2 \\
 &\quad + \sum_{i=1}^{n_1} \left[ -\frac{x_i}{\alpha_1} - 2 \log \left( 2 - e^{-\frac{x_i}{\alpha_1}} \right) \right] + \sum_{i=1}^{n_2} \left[ -\frac{x_i}{\alpha_2} - 2 \log \left( 2 - e^{-\frac{x_i}{\alpha_2}} \right) \right]
 \end{aligned}$$

Now,

$$\frac{\partial \log L}{\partial \alpha_1} = 0$$

$$\begin{aligned}
 -\frac{n_1}{\alpha_1} + \sum_{i=1}^{n_1} \left[ \frac{x_i}{\alpha_1^2} + \frac{2x_i e^{-\frac{x_i}{\alpha_1}}}{\alpha_1^2 \left( 2 - e^{-\frac{x_i}{\alpha_1}} \right)} \right] &= 0 \\
 \alpha_1 &= \frac{1}{n_1} \sum_{i=1}^{n_1} \left[ x_i + \frac{2x_i e^{-\frac{x_i}{\alpha_1}}}{\left( 2 - e^{-\frac{x_i}{\alpha_1}} \right)} \right] \tag{4.1}
 \end{aligned}$$

$$\frac{\partial \log L}{\partial \alpha_2} = 0$$

$$\begin{aligned}
 -\frac{n_2}{\alpha_2} + \sum_{i=1}^{n_2} \left[ \frac{x_i}{\alpha_2^2} + \frac{2x_i e^{-\frac{x_i}{\alpha_2}}}{\alpha_2^2 \left( 2 - e^{-\frac{x_i}{\alpha_2}} \right)} \right] &= 0 \\
 \alpha_2 &= \frac{1}{n_2} \sum_{i=1}^{n_2} \left[ x_i + \frac{2x_i e^{-\frac{x_i}{\alpha_2}}}{\left( 2 - e^{-\frac{x_i}{\alpha_2}} \right)} \right] \tag{4.2}
 \end{aligned}$$

$$\frac{\partial \log L}{\partial p_1} = 0$$

$$\frac{n_1}{p_1} - \frac{n_2}{1-p_1} = 0$$

$$\hat{p}_1 = \frac{n_1}{n}; n_1 + n_2 + 1 \tag{4.3}$$

Generalizing the above results for k-components we get

$$\alpha_i = \frac{1}{n_i} \sum_{j=1}^{n_i} \left[ x_j + \frac{2x_i e^{-\frac{x_j}{\alpha_i}}}{\left(2 - e^{-\frac{x_j}{\alpha_i}}\right)} \right] \text{ and } \hat{p}_i = \frac{n_i}{n}; \quad n = \sum_{i=1}^v n_i \tag{4.4}$$

It is obvious that first equation is nonlinear in  $\alpha_j$  and is not easily solvable, therefore for obtaining their estimates we propose to use numerical iteration method. Therefore  $\alpha_j$  can be obtained by using the nonlinear equation

$$h(\alpha_i) = \alpha_i \tag{4.5}$$

where

$$h(\alpha_i) = \frac{1}{n_i} \sum_{j=1}^{n_i} \left[ x_j + \frac{2x_i e^{-\frac{x_j}{\alpha_i}}}{\left(2 - e^{-\frac{x_j}{\alpha_i}}\right)} \right] \tag{4.6}$$

since  $\hat{\alpha}_i$  is a fixed point solution of  $h(\alpha_i) = \alpha_i$ , therefore it can be obtained by using simple iteration procedure as  $h(\alpha_{i(k)}) = \alpha_{i(k+1)}$ , where  $\alpha_{i(k)}$  and  $\alpha_{i(k+1)}$  are  $k^{\text{th}}$  and  $(k+1)^{\text{th}}$  iterations of  $\hat{\alpha}_i$ . When the difference between  $\alpha_{i(k)}$  and  $\alpha_{i(k+1)}$  is very small we stop the iteration and once we obtain the estimates  $\hat{\alpha}_i$  and  $\hat{p}_i$  the M.L.E of reliabilities for different cases are given as

1. M.L.E of reliability function when stress follows exponential distribution with known parameter  $\lambda$  and strength follows finite mixture of M-Transformed Exponential distribution with parameter  $\alpha_i$  is

$$\hat{R}_v = 1 - \sum_{i=1}^v \sum_{m_i=1}^{\infty} \frac{\hat{p}_i m_i}{2^{m_i} (\lambda \hat{\alpha}_i + m_i)}$$

2. M.L.E of reliability function when stress follows Lindley distribution with known parameters  $\lambda$  and  $\theta$  and strength follows finite mixture of M-Transformed Exponential distributions with parameter  $\alpha_i$  is

$$\hat{R}_v = 1 - \sum_{i=1}^v \sum_{m_i=1}^{\infty} \hat{p}_i \frac{m_i}{2^{m_i}} \left( \frac{1}{m_i + \theta \hat{\alpha}_i} + \frac{\lambda \theta \hat{\alpha}_i}{(\theta + \lambda)(m_i + \theta \hat{\alpha}_i)^2} \right)$$

3. M.L.E of reliability function when stress follows finite mixture of Lindley distributions with known parameters  $\lambda_j$  and  $\theta_j$  and strength follows finite mixture of M-Transformed Exponential distributions with parameter  $\alpha_i$  is

$$\hat{R}_v = 1 - \sum_{j=1+v}^{2v} \sum_{i=1}^v \sum_{m_i=1}^{\infty} \hat{p}_i \hat{p}_j m_i \left[ \frac{1}{m_i + \theta_j \hat{\alpha}_i} + \frac{\lambda_j \theta_j \hat{\alpha}_i}{(\theta_j + \lambda_j)(m_i + \theta_j \hat{\alpha}_i)^2} \right]$$

4. M.L.E of reliability function when stress and strength both follows finite mixture of M-Transformed Exponential distributions with parameters  $\alpha_i$  and  $\alpha_j$  is

$$\hat{R}_v = 1 - \sum_{j=1+v}^{2v} \sum_{i=1}^v \hat{p}_i \hat{p}_j \left( \frac{\hat{\alpha}_j}{\hat{\alpha}_i + \hat{\alpha}_j} \right)$$



### 5. Numerical Evaluation

In this section, we have evaluated reliability of a system with two components for specific values of a parameters involved in the reliability expressions in section 3 using graphical approach. Here we will discuss how reliability of two component system behaves when stress, strength or probability parameters are changed.

**Case I:** Stress follows exponential distribution and strength follows finite mixture of M-Transformed Exponential distribution

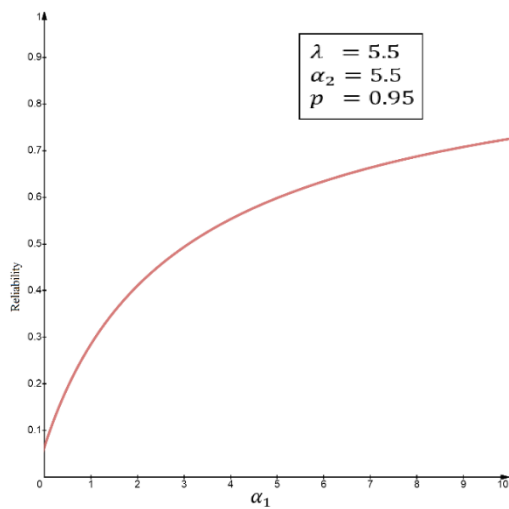


Figure 1

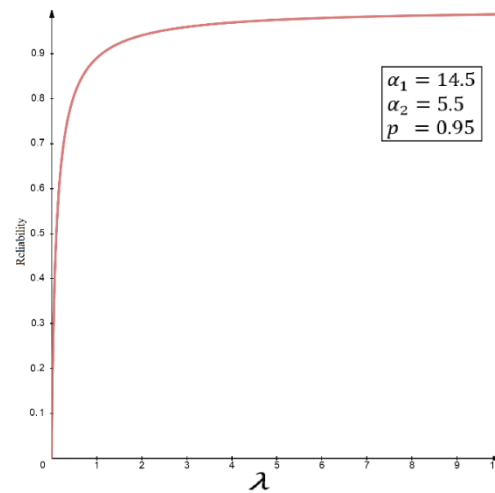


Figure 2

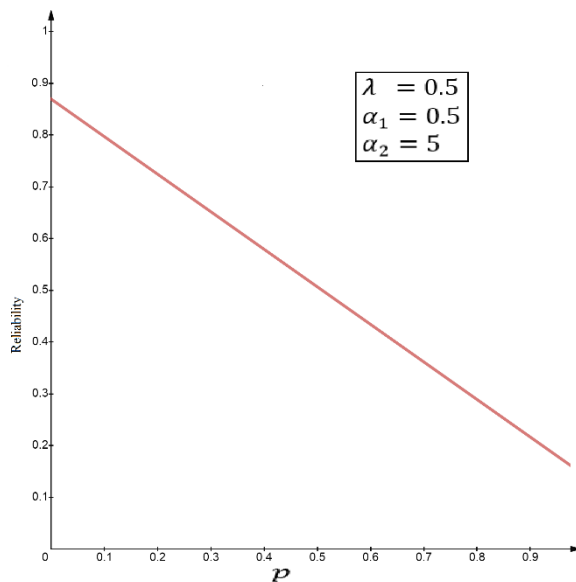


Figure 3

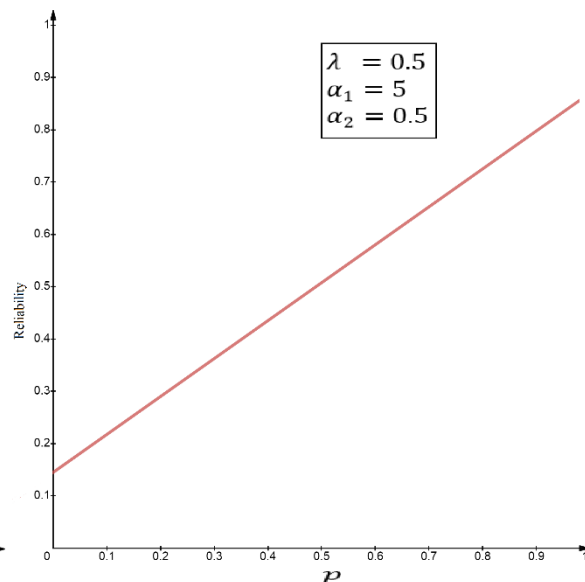


Figure 4

**Case II:** Stress follows Lindley distribution and strength follows finite mixture of M-Transformed Exponential distribution

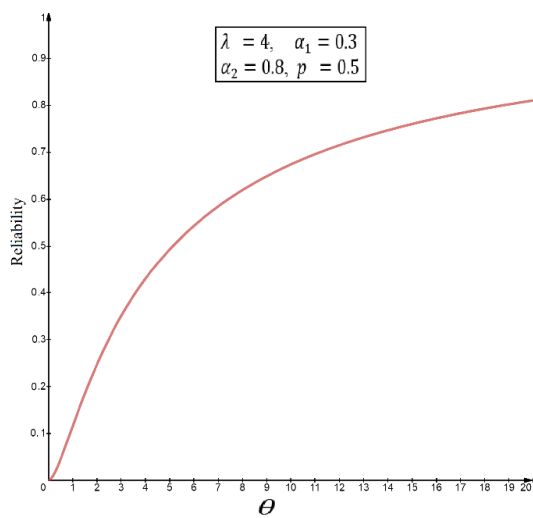


Figure 5

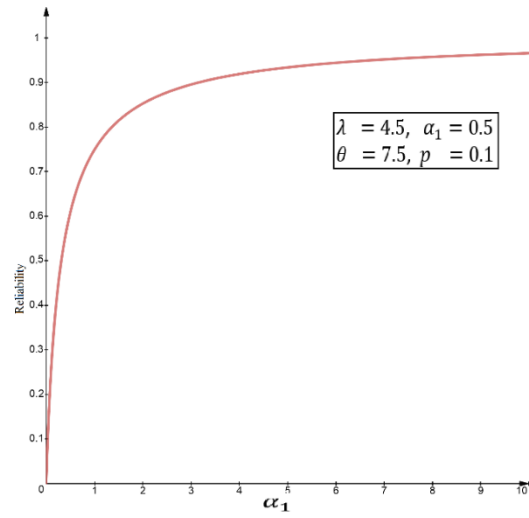


Figure 6

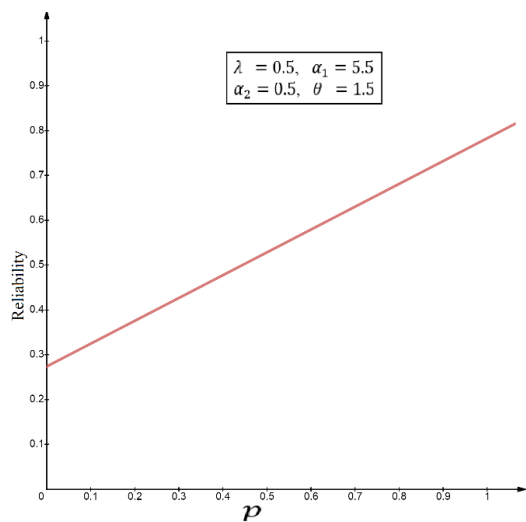


Figure 7

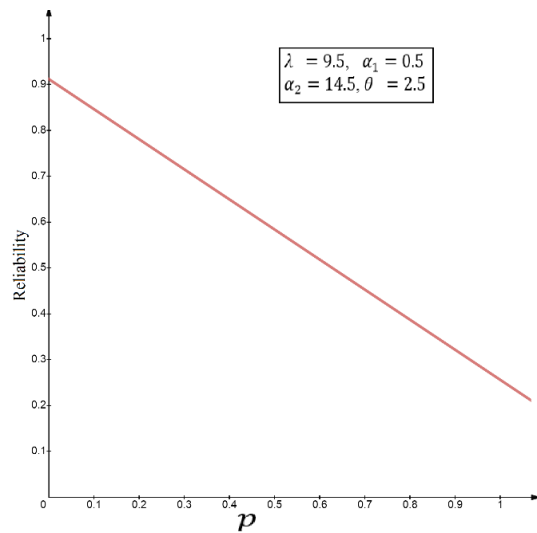


Figure 8

**Case III:** Stress follows finite mixture of Lindley distributions and strength follows finite mixture of M-Transformed Exponential distributions

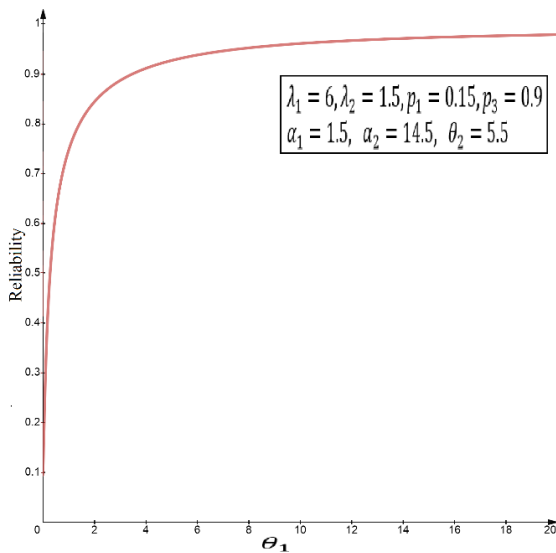


Figure 9

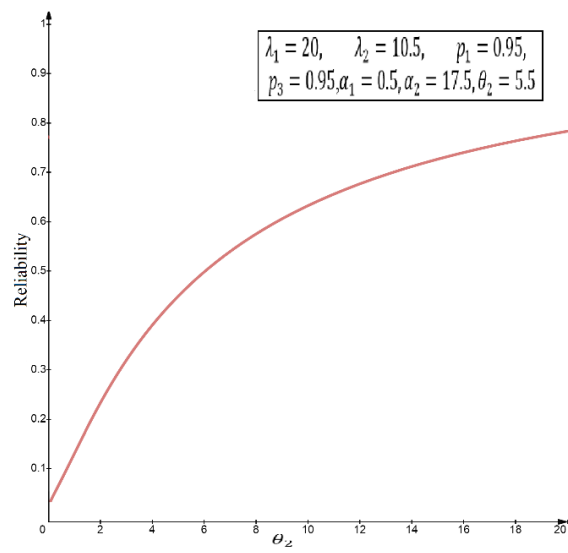


Figure 10

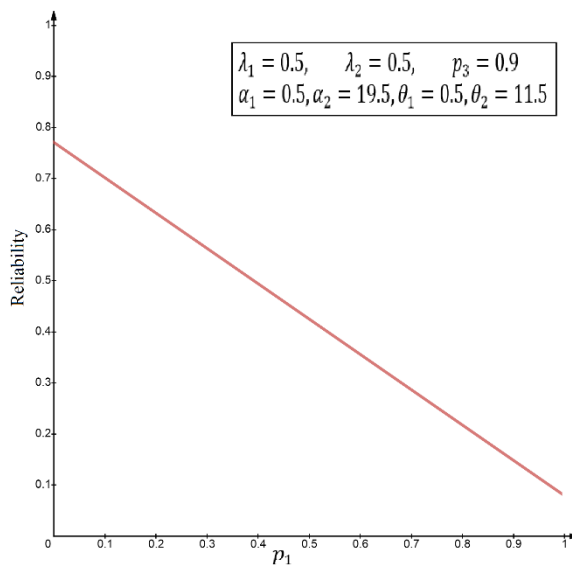


Figure 11

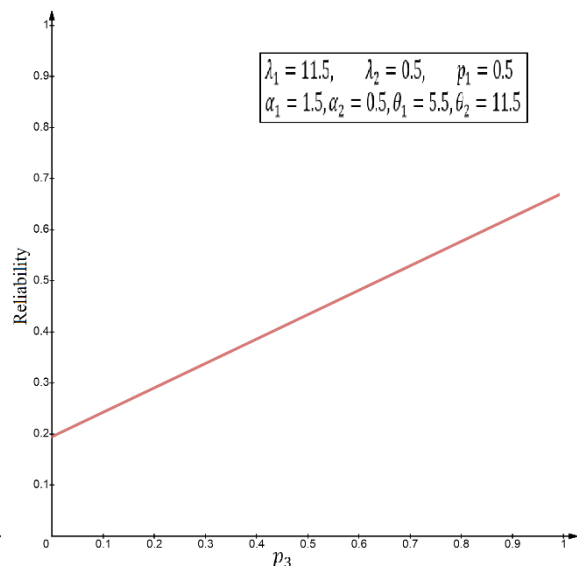


Figure 12

**Case IV:** Stress and strength both follows finite mixture of M-Transformed Exponential distributions

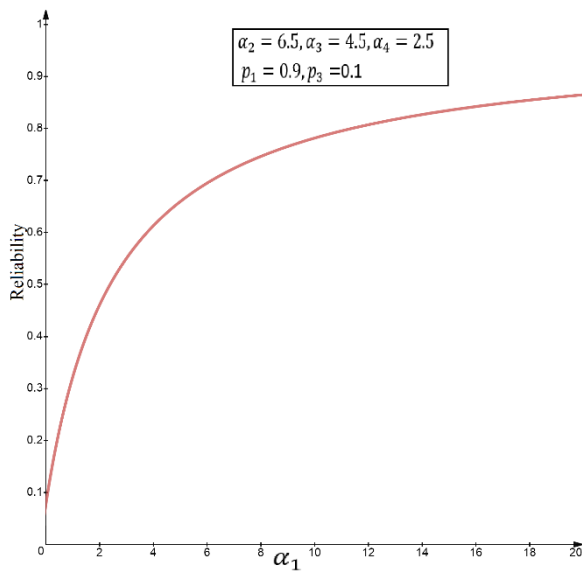


Figure 13

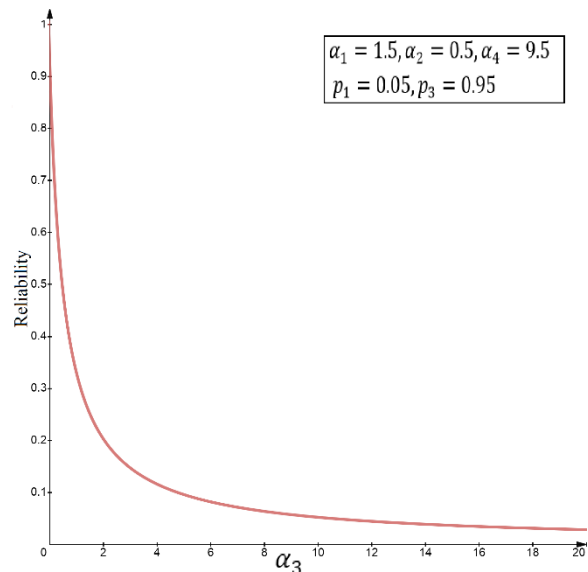


Figure 14

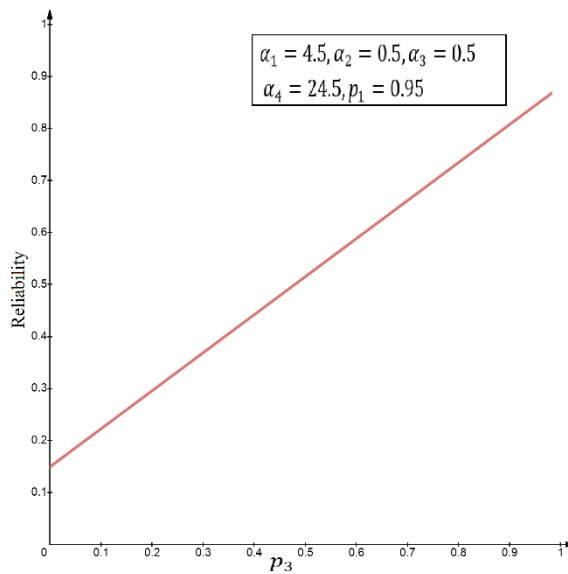


Figure 15

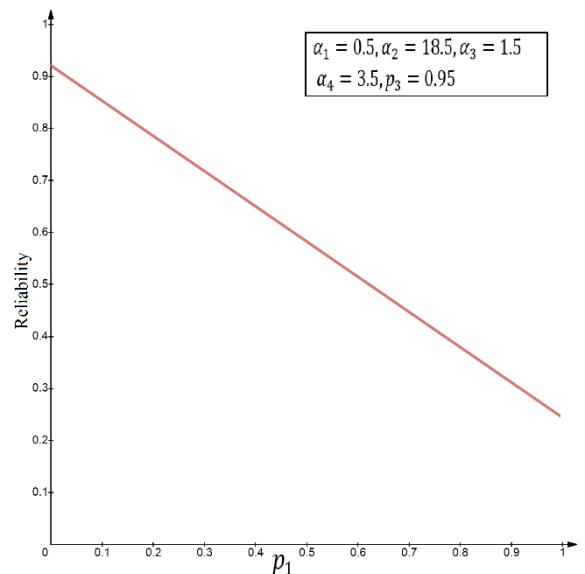


Figure 16

## 6. Real Data Analysis

In this sub section we analyze 33 leukaemia patients with two causes of death AG positive (presence of Auer rods and/or significant granulation of the leukaemic cells) or AG negative (both Auer rods and granulation are absent). The survival times in weeks are given

AG positive patients: 65, 156, 100, 134, 16, 108, 121, 4, 39, 143, 56, 26, 22, 1, 1, 5, 65

AG negative patients: 56, 65, 17, 7, 16, 22, 3, 4, 2, 3, 8, 4, 3, 30, 4, 43

We use the iterative procedure to obtain MLE of the parameters  $\alpha_1$  and  $\alpha_2$  using equation (4.5) and the final value of estimates are  $\hat{\alpha}_1 = 90.65061, \hat{\alpha}_2 = 25.44067$  and  $\hat{p}_1 = 0.5151$ . Based on these estimates the MLE of the reliability are given below

Stress and strength both follows finite mixture of M-Transformed Exponential distributions  $\hat{R}_2 = 0.68$

Table 1: Stress follows exponential distribution and strength follows finite mixture of M-Transformed Exponential distribution

R	0.9161	0.9547	0.9689	0.9763	0.9809	0.9840	0.9862	0.9880	0.9892	0.9903
$\lambda$	0.5	1	1.5	2	2.5	3	3.5	4	4.5	5

Table 2: Stress follows Lindley distribution and strength follows finite mixture of M-Transformed Exponential distribution

$\lambda$ $\theta$	0.5	1	1.5	2	2.5	3	3.5	4	4.5	5
0.5	0.9317	0.9264	0.9238	0.9223	0.9212	0.9205	0.9200	0.9195	0.9191	0.9189
1	0.9409	0.9340	0.9298	0.9270	0.9250	0.9235	0.9224	0.9215	0.9207	0.9200
1.5	0.9450	0.9375	0.9327	0.9292	0.9265	0.9245	0.9228	0.9215	0.9204	0.9195
2	0.9473	0.9399	0.9346	0.9306	0.9275	0.9250	0.9230	0.9214	0.9200	0.9187
2.5	0.9490	0.9415	0.9360	0.9317	0.9283	0.9255	0.9232	0.9212	0.9195	0.9181
3	0.9500	0.9427	0.9371	0.9326	0.9289	0.9258	0.9233	0.9211	0.9191	0.9174
3.5	0.9507	0.9436	0.9380	0.9333	0.9294	0.9262	0.9234	0.9209	0.9188	0.9169
4	0.9514	0.9444	0.9387	0.9340	0.9299	0.9264	0.9234	0.9208	0.9185	0.9164
4.5	0.9520	0.9450	0.9393	0.9344	0.9303	0.9266	0.9235	0.9207	0.9182	0.9160
5	0.9523	0.9455	0.9398	0.9348	0.9305	0.9268	0.9235	0.9206	0.9180	0.9156

Table 3: Stress follows finite mixture of Lindley distributions and strength follows finite mixture of M-Transformed Exponential distributions,  $p_3 = 0.35, \lambda_2 = 6.5, \theta_2 = 5.5$

$\lambda_1$ $\theta_1$	1	2	3	4	5	6	7	8	9	10
1	0.9680	0.9656	0.9644	0.9637	0.9632	0.9629	0.9626	0.9624	0.9622	0.9621
2	0.9784	0.9762	0.9748	0.9739	0.9733	0.9728	0.9725	0.9722	0.9719	0.9717
3	0.9818	0.9794	0.9780	0.9768	0.9760	0.9754	0.9749	0.9744	0.9741	0.9738
4	0.9833	0.9810	0.9792	0.9779	0.9770	0.9761	0.9754	0.9750	0.9744	0.9740
5	0.9842	0.9817	0.9798	0.9784	0.9771	0.9762	0.9754	0.9747	0.9741	0.9736
6	0.9848	0.9822	0.9801	0.9785	0.9772	0.9761	0.9751	0.9743	0.9736	0.9730
7	0.9852	0.9825	0.9803	0.9786	0.9771	0.9758	0.9747	0.9738	0.9730	0.9723
8	0.9855	0.9827	0.9804	0.9786	0.9769	0.9755	0.9743	0.9733	0.9723	0.9715
9	0.9858	0.9829	0.9805	0.9785	0.9768	0.9753	0.9740	0.9728	0.9717	0.9708
10	0.9860	0.9830	0.9806	0.9784	0.9766	0.9750	0.9735	0.9723	0.9711	0.9701

## 7. Conclusion

In this paper we have obtained stress strength reliability of a system with  $v$ -components using different distributions for stress variables viz. exponential distribution, M-Transformed exponential distribution and for strength variable we have used M-Transformed exponential distribution. In the last section we showed that reliability of 2-component system can be monotonically increasing and monotonically decreasing for specific values of parameter. Thus by proper choice of parameters leads to high reliability. The maximum likelihood estimates of parameters involved in the reliability function of  $v$ -components system are also obtained using iteration method.

## References

1. Awad, A.M., Gharraf, M.K., 1986, Estimation of  $P(Y < X)$  in the Burr case: A comparative study. *Commun. Statist. Simul. Comp.*, 15(2), 389-403.
2. Beg, M.A., Singh, N., 1978, Estimation of  $P(Y < X)$  for the pareto distribution. *IEEE Trans. Reliab.*, 28(5), 411-414.
3. Church, J.D., Harris, B., 1970, The estimation of reliability from stress strength relationships. *Technometrics*, 12, 49-54.
4. Khan, A.H., Jan, T.R., 2014, Estimation of Multi Component Systems Reliability in Stress-Strength Models. *Journal of Modern Applied Statistical Methods*, 13 (2), 389-398.
5. Khan, A.H., Jan, T.R., 2014, Reliability Estimates of Generalized Poisson Distribution and Generalized Geometric Series Distribution. *Journal of Modern Applied Statistical Methods*, 13 (2), 379-388.
6. Kotz, S., Lumelskii, Y., Pensky, M., 2003, *The Stress-Strength Model and its Generalizations: Theory and Applications*, World Scientific Publishing, Singapore.
7. Kumar, D., Singh, U., Singh, S.K, Mukherjee, S., 2017, The New Probability Distribution: An Aspect to Lifetime Distribution. *Math. Sci. Lett.* 6(1), 35-42.
8. Mokhlis, N.A., 2005, Reliability of a stress-strength model with Burr Type III distributions. *Commun. Statist.Theory Meth.* 34(7). 1643-1657.
9. Raqab, M. Z., Kunda, D., 2005, Comparison of Different Estimates of  $P(Y < X)$  for a Scaled Burr Type X distribution, *Communication in Statistics-Computations and Simulations*, 34(2), 465-483.
10. Saraçoğlu, B., Kınacı, İ., Kundu, D., 2011, On estimation of  $P(Y < X)$  for exponential distribution under progressive type-II censoring. *J. Stat. Comput. Simul.* 82(5), 729-744.
11. Surles, J. G., Pudgett, W. J., 1998, Inference for  $P(Y < X)$  in the Burr X Model, *Journal of Applied Statistical Sciences*, 225-238.
12. Surles, J. G., Pudgett, W. J., 2001, Inference for Reliability and Stress-Strength for Scald Burr X distribution, *Lifetime Data Analysis*, 7, 187-200.
13. Woodward, W.A., Kelley, G.D., 1977, Minimum variance unbiased estimation of  $P(Y < X)$  in the normal case. *Technometrics*, 19, 95-98.