



# Stress-Strength Reliability Quantification using M-Transformed Exponential Distributions

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## Authors' contributions

*This work was carried out in collaboration between both authors. Author AHK designed the study, performed the statistical analysis, wrote the protocol and wrote the first draft of the manuscript and Authors TRJ managed the analyses of the study and the literature searches. Both authors read and approved the final manuscript.*

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## ABSTRACT

The term "Stress-Strength reliability" means  $P(X > Y)$ , where a system or an equipment with random strength  $X$  is subjected to random stress  $Y$  in a way that system breaks down, if the stress surpasses the strength. In this paper, a system is considered with standby redundancy, and it is presumed that the distinct components in the system for both stress and strength variables are independent and have different probability distributions viz. M- Transformed Exponential, Exponential, Gamma and Lindley. The expressions for the marginal reliabilities  $R(1)$ ,  $R(2)$ ,  $R(3)$  etc. based on its stress-strength models are obtained.

*Keywords: Reliability; M-transformed exponential distribution; stress; strength.*

## 1. INTRODUCTION

The reliability of a system is defined as the probability that the system will perform its

assigned function successfully under desired environmental conditions. The collocation 'Stress-Strength' was first naturalized by Church and Harris. In 'stress-strength models', strength

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X of the system or equipment and stress Y enforced on it by its operational environments are considered as random variables. The expression  $R = P(Y < X)$  is the reliability of the system and is defined as the probability that the system or the equipment is strong enough to surpass the stress enforced on it. In a standby system, there are a lot of connected units out of which only one unit works at a time and the rest of units remains standby. When the strike of stress crosses the strength of working unit, it fails and the substitute unit among standbys is activated (if available) and faces the impacting force of stresses of the operating system. The comprehensive system fails when all the units including standbys have failed. The history of this problem is very deep starting with the creative work of Z. W. Birnbaum and R. C. McCarty [1] in which they assumed a problem of procuring a one-sided distribution free confidence interval for  $P = P(Y < X)$ . The collocation 'Stress-Strength' was first naturalized by Church and Harris [2]. A lot of valuable investigative work has been done since then, both from a parametric and non-parametric viewpoint. Kakati and Sriwastav [3] works with a redundant Stress-Strength Model. The excellent monograph "The Stress-Strength Model and its Generalizations" by Samuel Kotz, Yan Lumelskii, Marianna Pensky [4], offers a complete study of the various Stress-Strength simulations up to 2001. The practical and theoretical findings on the theory and the implementation of Stress-Strength relationships in economic and industrial systems are summarized and published in this book. For the first time, the findings and assumptions that have been spread in the literature over the past forty years have been addressed in a unified way accessible to applied and theoretical statisticians. The estimation of  $R = P(Y < X)$  where X and Y are two independent scaled Burr type X distributions with same scale parameters is considered by Raqab and Kundu ([5]). Kundu and Gupta ([6] and [7]) contrasts diverse perspectives of estimating  $R = P(Y < X)$  using generalized exponential distribution and Weibull distribution. The Stress vs. Strength problem concerning multi-component devices such as standby redundancy is explored by Gogio and Borah [8]. Adil H. Khan and T.R Jan ([9] and [10]) in 2014 obtained expressions for stress-strength reliability for various distributions and also obtained the reliability function of Generalized Poisson distribution and Generalized Geometric distribution.

## 2. MODEL

Consider a  $v$ -standby configuration in which there are  $v$  components originally, where only one performs during stress impact and the ones left ( $v - 1$ ) are stand-by. However, when the working component fails, other takes its spot from the standby and is exposed to stress impacts and the system works. When the entire part fails, the system fails.

Let  $X_1, X_2, \dots, X_v$  be a set of  $v$  independent random variables chosen to represent the strengths of  $v$  components organized in the process in order of activation and let  $Y_1, Y_2, \dots, Y_v$  be just another set of independent random variables chosen to represent the stresses on the  $v$  components respectively, then the system reliability  $R_v$  of the system is given by

$$R_v = r(1) + R(2) + \dots + R(v) \tag{2.1}$$

In which the marginal reliability  $R(v)$  is the contribution to the reliability of the system by the  $r^{\text{th}}$  component and is defined as

$$R(r) = \Pr[X_1 < Y_1, X_2 < Y_2, \dots, X_{r-1} < Y_{r-1} \geq Y_r]$$

and if  $f_i(x)$  and  $g_i(y)$  are the probability density functions of  $X_i$  and  $Y_i$ ;  $i = 1, 2, \dots, v$  respectively then

$$R(r) = \left[ \int_{-\infty}^{\infty} F_1(y)h_1(y)dy \right] \times \left[ \int_{-\infty}^{\infty} F_2(y)h_2(y)dy \right] \times \dots \times \left[ \int_{-\infty}^{\infty} F_{r-1}(y)h_{r-1}(y)dy \right] \left[ \int_{-\infty}^{\infty} \bar{F}_r(y)h_r(y)dy \right] \tag{2.2}$$

In which  $F_i(y)$  is the commutative distribution function of  $X_i$  and

$$\bar{F}_i(y) = 1 - F(x)$$

We have assumed strength and stress obey various distributions in this paper, the following instances are regarded.

- 1) Exponential strength and M-Transformed Exponential stress.
- 2) Lindley strength and M-Transformed Exponential stress.
- 3) M-Transformed Exponential strength and M-Transformed Exponential stress.
- 4) M-Transformed Exponential Strength and Two Parameter Gamma Stress.

**Case I: Exponential strength and M-Transformed Exponential stress**

Let  $f_i(x)$  be the exponential strength with parameter  $\lambda_i$  and  $g_i(y)$  be the M-Transformed exponential stress with parameter  $\alpha_i$ ;  $i = 1, 2, \dots, v$ , then

$$f_i(x, \lambda_i) = \lambda_i e^{-\lambda_i x} \quad ; x, \lambda_i > 0$$

$$\bar{F}_i(x) = 1 - F_i(x) = e^{-\lambda_i x} \quad ; x, \lambda_i > 0$$

And

$$g_i(y; \alpha_i) = \frac{2e^{-\frac{y}{\alpha_i}}}{\alpha_i \left(2 - e^{-\frac{y}{\alpha_i}}\right)^2} \quad ; y, \alpha_i > 0$$

Then from equation (2.2)

$$R(1) = \int_0^{\infty} \bar{F}_1(y) g_1(y) dy$$

$$R(1) = \int_0^{\infty} e^{-\lambda_1 y} \frac{2e^{-\frac{y}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{y}{\alpha_1}}\right)^2} dy$$

$$R(1) = \sum_{n=1}^{\infty} \frac{n}{2^n \alpha_1} \int_0^{\infty} e^{-(\lambda_1 + \frac{n}{\alpha_1})y} dy$$

$$R(1) = \sum_{n=1}^{\infty} \frac{n}{2^n (n + \alpha_1 \lambda_1)}$$

$$R(2) = \left[ \int_0^{\infty} F_1(y) g_1(y) dy \right] \left[ \int_0^{\infty} \bar{F}_2(y) g_2(y) dy \right]$$

$$R(2) = \left[ \int_0^{\infty} (1 - e^{-\lambda_1 y}) \frac{2e^{-\frac{y}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{y}{\alpha_1}}\right)^2} dy \right] \left[ \int_0^{\infty} e^{-\lambda_2 y} \frac{2e^{-\frac{y}{\alpha_2}}}{\alpha_2 \left(2 - e^{-\frac{y}{\alpha_2}}\right)^2} dy \right]$$

$$R(2) = \left[ 1 - \sum_{n=1}^{\infty} \frac{n}{2^n (n + \alpha_1 \lambda_1)} \right] \left[ \sum_{n=1}^{\infty} \frac{n}{2^n (n + \alpha_2 \lambda_2)} \right]$$

$$R(3) = \left[ \int_0^{\infty} F_1(y) g_1(y) dx \right] \left[ \int_0^{\infty} F_2(y) g_2(y) dy \right] \left[ \int_0^{\infty} \bar{F}_3(y) g_3(y) dy \right]$$

$$R(3) = \left[ 1 - \sum_{n=1}^{\infty} \frac{n}{2^n (n + \alpha_1 \lambda_1)} \right] \left[ 1 - \sum_{n=1}^{\infty} \frac{n}{2^n (n + \alpha_2 \lambda_2)} \right] \left[ \sum_{n=1}^{\infty} \frac{n}{2^n (n + \alpha_3 \lambda_3)} \right]$$

$$R(v) = \left[ 1 - \sum_{n=1}^{\infty} \frac{n}{2^n (n + \alpha_1 \lambda_1)} \right] \left[ 1 - \sum_{n=1}^{\infty} \frac{n}{2^n (n + \alpha_2 \lambda_2)} \right] \dots$$

$$\times \left[ 1 - \sum_{n=1}^{\infty} \frac{n}{2^n (n + \alpha_{v-1} \lambda_{v-1})} \right] \left[ \sum_{n=1}^{\infty} \frac{n}{2^n (n + \alpha_v \lambda_v)} \right]$$

$$R(v) = \left[ \sum_{n=1}^{\infty} \frac{n}{2^n (n + \alpha_v \lambda_v)} \right] \prod_{i=1}^v \left[ 1 - \sum_{n=1}^{\infty} \frac{n}{2^n (n + \alpha_{i-1} \lambda_{i-1})} \right]$$

**Case II: Lindley strength and M-Transformed Exponential stress**

$$f_i(x, \lambda_i) = \frac{\lambda_i^2}{1 + \lambda_i} (1 + x) e^{-\lambda_i x} \quad ; x, \lambda_i > 0$$

$$\bar{F}_i(x, \lambda_i) = 1 - F_i(x, \lambda_i) = \frac{1 + \lambda_i + \lambda_i x}{1 + \lambda_i} e^{-\lambda_i x} \quad ; x, \lambda_i > 0$$

And

$$g_i(y; \alpha_i) = \frac{2e^{-\frac{y}{\alpha_i}}}{\alpha_i \left(2 - e^{-\frac{y}{\alpha_i}}\right)^2} ; y, \alpha_i > 0$$

Then from equation (2.2)

$$R(1) = \int_0^{\infty} \bar{F}_1(y)g_1(y)dy$$

$$R(1) = \int_0^{\infty} \frac{1 + \lambda_1 + \lambda_1 y}{1 + \lambda_1} e^{-\lambda_1 y} \frac{2e^{-\frac{y}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{y}{\alpha_1}}\right)^2} dy$$

$$R(1) = \sum_{n=1}^{\infty} \frac{n}{2^n \alpha_1} \left[ \int_0^{\infty} e^{-(\lambda_1 + \frac{n}{\alpha_1})y} dy + \frac{\lambda_1}{1 + \lambda_1} \int_0^{\infty} y^2 e^{-(\lambda_1 + \frac{n}{\alpha_1})y} dy \right]$$

$$R(1) = \sum_{n=1}^{\infty} \frac{n}{2^n} \left[ \frac{1}{n + \alpha_1 \lambda_1} + \frac{\lambda_1 \alpha_1}{(1 + \lambda_1)(n + \alpha_1 \lambda_1)^2} \right]$$

$$R(2) = \left[ \int_0^{\infty} F_1(y)g_1(y)dy \right] \left[ \int_0^{\infty} \bar{F}_2(y)g_2(y)dy \right]$$

$$R(2) = \left[ \int_0^{\infty} \left(1 - \frac{1 + \lambda_1 + \lambda_1 y}{1 + \lambda_1} e^{-\lambda_1 y}\right) \frac{2e^{-\frac{y}{\alpha_1}}}{\alpha_1 \left(2 - e^{-\frac{y}{\alpha_1}}\right)^2} dy \right] \left[ \int_0^{\infty} \frac{1 + \lambda_2 + \lambda_2 y}{1 + \lambda_2} e^{-\lambda_2 y} \frac{2e^{-\frac{y}{\alpha_2}}}{\alpha_2 \left(2 - e^{-\frac{y}{\alpha_2}}\right)^2} dy \right]$$

$$R(2) = \left[ 1 - \sum_{n=1}^{\infty} \frac{n}{2^n} \left( \frac{1}{n + \alpha_1 \lambda_1} + \frac{\lambda_1 \alpha_1}{(1 + \lambda_1)(n + \alpha_1 \lambda_1)^2} \right) \right] \left[ \sum_{n=1}^{\infty} \frac{n}{2^n} \left( \frac{1}{n + \alpha_2 \lambda_2} + \frac{\lambda_2 \alpha_2}{(1 + \lambda_2)(n + \alpha_2 \lambda_2)^2} \right) \right]$$

$$R(3) = \left[ \int_0^{\infty} F_1(y)g_1(y)dy \right] \left[ \int_0^{\infty} F_2(y)g_2(y)dy \right] \left[ \int_0^{\infty} \bar{F}_3(y)g_3(y)dy \right]$$

$$R(3) = \left[ 1 - \sum_{n=1}^{\infty} \frac{n}{2^n} \left( \frac{1}{n + \alpha_1 \lambda_1} + \frac{\lambda_1 \alpha_1}{(1 + \lambda_1)(n + \alpha_1 \lambda_1)^2} \right) \right]$$

$$\times \left[ 1 - \sum_{n=1}^{\infty} \frac{n}{2^n} \left( \frac{1}{n + \alpha_2 \lambda_2} + \frac{\lambda_2 \alpha_2}{(1 + \lambda_2)(n + \alpha_2 \lambda_2)^2} \right) \right] \left[ \sum_{n=1}^{\infty} \frac{n}{2^n} \left( \frac{1}{n + \alpha_3 \lambda_3} + \frac{\lambda_3 \alpha_3}{(1 + \lambda_3)(n + \alpha_3 \lambda_3)^2} \right) \right]$$

$$R(v) = \left[ 1 - \sum_{n=1}^{\infty} \frac{n}{2^n} \left( \frac{1}{n + \alpha_1 \lambda_1} + \frac{\lambda_1 \alpha_1}{(1 + \lambda_1)(n + \alpha_1 \lambda_1)^2} \right) \right] \left[ 1 - \sum_{n=1}^{\infty} \frac{n}{2^n} \left( \frac{1}{n + \alpha_2 \lambda_2} + \frac{\lambda_2 \alpha_2}{(1 + \lambda_2)(n + \alpha_2 \lambda_2)^2} \right) \right] \dots$$

$$\times \left[ 1 - \sum_{n=1}^{\infty} \frac{n}{2^n} \left( \frac{1}{n + \alpha_{v-1} \lambda_{v-1}} + \frac{\lambda_{v-1} \alpha_{v-1}}{(1 + \lambda_{v-1})(n + \alpha_{v-1} \lambda_{v-1})^2} \right) \right] \left[ \sum_{n=1}^{\infty} \frac{n}{2^n} \left( \frac{1}{n + \alpha_v \lambda_v} + \frac{\lambda_v \alpha_v}{(1 + \lambda_v)(n + \alpha_v \lambda_v)^2} \right) \right]$$

$$R(v) = \left[ \sum_{n=1}^{\infty} \frac{n}{2^n} \left( \frac{1}{n + \alpha_v \lambda_v} + \frac{\lambda_v \alpha_v}{(1 + \lambda_v)(n + \alpha_v \lambda_v)^2} \right) \right] \\ \times \prod_{i=1}^v \left[ 1 - \sum_{n=1}^{\infty} \frac{n}{2^n} \left( \frac{1}{n + \alpha_{i-1} \lambda_{i-1}} + \frac{\lambda_{i-1} \alpha_{i-1}}{(1 + \lambda_{i-1})(n + \alpha_{i-1} \lambda_{i-1})^2} \right) \right]$$

**Case III: M-Transformed Exponential strength and M-Transformed Exponential stress**

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

$$\bar{F}_i(x; \alpha_i) = 1 - F_i(x; \alpha_i) = 1 - \frac{2 \left( 1 - e^{-\frac{x}{\alpha_i}} \right)}{2 - e^{-\frac{x}{\alpha_i}}} ; x, \alpha_i > 0$$

And

$$g_i(y; \alpha_i) = \frac{2e^{-\frac{y}{\alpha_i}}}{\alpha \left( 2 - e^{-\frac{y}{\alpha_i}} \right)^2} ; y, \alpha_i > 0$$

Then from equation (2.2)

$$R(1) = \int_0^{\infty} \bar{F}_1(y) g_1(y) dy \\ R(1) = \int_0^{\infty} \left[ 1 - \frac{2 \left( 1 - e^{-\frac{y}{\alpha_1}} \right)}{2 - e^{-\frac{y}{\alpha_1}}} \right] \left[ \frac{2e^{-\frac{y}{\beta_1}}}{\beta_1 \left( 2 - e^{-\frac{y}{\beta_1}} \right)^2} \right] dy \\ R(1) = \int_0^{\infty} \left[ \sum_{n=1}^{\infty} \frac{1}{2^n} e^{-\frac{ny}{\alpha_1}} \right] \left[ \sum_{m=1}^{\infty} \frac{m}{2^m \beta_1} e^{-\frac{my}{\beta_1}} \right] dy \\ R(1) = \frac{\alpha_1}{\alpha_1 + \beta_1}$$

Special Case: If  $\alpha_1 = \beta_1$ , then  $R(1) = \frac{1}{2}$

$$R(2) = \left[ \int_0^{\infty} F_1(y) g_1(y) dy \right] \left[ \int_0^{\infty} \bar{F}_2(y) g_2(y) dy \right] \\ R(2) = \left[ \int_0^{\infty} \left( 1 - \sum_{n=1}^{\infty} \frac{1}{2^n} e^{-\frac{ny}{\alpha_1}} \right) \left( \sum_{m=1}^{\infty} \frac{m}{2^m \beta_1} e^{-\frac{my}{\beta_1}} \right) dx \right] \left[ \int_0^{\infty} \left( \sum_{n=1}^{\infty} \frac{1}{2^n} e^{-\frac{ny}{\alpha_2}} \right) \left( \sum_{m=1}^{\infty} \frac{m}{2^m \beta_2} e^{-\frac{my}{\beta_2}} \right) dx \right] \\ R(2) = \left( \frac{\beta_1}{\alpha_1 + \beta_1} \right) \left( \frac{\alpha_2}{\alpha_2 + \beta_2} \right)$$

Special Case: If  $\alpha_i = \beta_i ; i = 1,2$  then  $R(2) = \frac{1}{2^2}$

$$R(3) = \left[ \int_0^\infty F_1(y)g_1(y)dy \right] \left[ \int_0^\infty F_2(y)g_2(y)dy \right] \left[ \int_0^\infty \bar{F}_3(y)g_3(y)dy \right]$$

$$R(3) = \left[ \int_0^\infty \left( 1 - \sum_{n=1}^\infty \frac{1}{2^n} e^{-\frac{ny}{\alpha_1}} \right) \left( \sum_{m=1}^\infty \frac{m}{2^m \beta_1} e^{-\frac{my}{\beta_1}} \right) dx \right] \left[ \int_0^\infty \left( 1 - \sum_{n=1}^\infty \frac{1}{2^n} e^{-\frac{ny}{\alpha_2}} \right) \left( \sum_{m=1}^\infty \frac{m}{2^m \beta_2} e^{-\frac{my}{\beta_2}} \right) dx \right]$$

$$\times \left[ \int_0^\infty \left( \sum_{n=1}^\infty \frac{1}{2^n} e^{-\frac{ny}{\alpha_3}} \right) \left( \sum_{m=1}^\infty \frac{m}{2^m \beta_3} e^{-\frac{my}{\beta_3}} \right) dx \right]$$

$$R(2) = \left( \frac{\beta_1}{\alpha_1 + \beta_1} \right) \left( \frac{\beta_2}{\alpha_2 + \beta_2} \right) \left( \frac{\alpha_3}{\alpha_3 + \beta_3} \right)$$

Special Case: If  $\alpha_i = \beta_i ; i = 1,2,3$  then  $R(2) = \frac{1}{2^3}$

$$R(v) = \left( \frac{\alpha_v}{\alpha_v + \beta_v} \right) \left( \frac{\beta_2}{\alpha_2 + \beta_2} \right) \left( \frac{\beta_3}{\alpha_3 + \beta_3} \right) \dots \left( \frac{\alpha_v}{\alpha_v + \beta_v} \right)$$

$$R(v) = \left( \frac{\alpha_v}{\alpha_v + \beta_v} \right) \prod_{i=1}^v \left( \frac{\alpha_v}{\alpha_{i-1} + \beta_{i-1}} \right)$$

Special Case: If  $\alpha_i = \beta_i ; i = 1,2, \dots, v$  then  $R(2) = \frac{1}{2^v}$

**Case IV: M-Transformed Exponential Strength and Two Parameter Gamma Stress**

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

$$\bar{F}_i(x; \alpha_i) = 1 - F_i(x; \alpha_i) = 1 - \frac{2 \left( 1 - e^{-\frac{x}{\alpha_i}} \right)}{2 - e^{-\frac{x}{\alpha_i}}} ; x, \alpha_i > 0$$

And

$$g_i(x; \beta, \theta) = \frac{\theta \beta_i x^{\beta_i - 1} e^{-\theta x}}{\Gamma(\beta_i)} ; x, \beta_i, \theta_i > 0$$

Then from equation (2.2)

$$R(1) = \int_0^\infty \bar{F}_1(y)g_1(y)dy$$

$$R(1) = \int_0^\infty \left[ 1 - \frac{2 \left( 1 - e^{-\frac{y}{\alpha_1}} \right)}{2 - e^{-\frac{y}{\alpha_1}}} \right] \left[ \frac{\theta_1^{\beta_1} y^{\beta_1 - 1} e^{-\theta_1 y}}{\Gamma(\beta_1)} \right] dy$$

$$R(1) = \frac{\theta_1^{\beta_1}}{\alpha_1 \Gamma(\beta_1)} \sum_{n=1}^{\infty} \frac{n}{2^n} \int_0^{\infty} y^{\beta_1-1} e^{-(\frac{n}{\alpha_1} + \theta_1)y} dy$$

$$R(1) = \sum_{n=1}^{\infty} \frac{n \theta_1^{\beta_1} \alpha_1^{\beta_1-1}}{2^n (n + \alpha_1 \theta_1)^{\beta_1}}$$

$$R(2) = \left[ \int_0^{\infty} F_1(y) g_1(y) dy \right] \left[ \int_0^{\infty} \bar{F}_2(y) g_2(y) dy \right]$$

$$R(2) = \left[ \int_0^{\infty} \left( \frac{2(1 - e^{-\frac{y}{\alpha_1}})}{2 - e^{-\frac{y}{\alpha_1}}} \right) \left( \frac{\theta_1^{\beta_1} y^{\beta_1-1} e^{-\theta_1 y}}{\Gamma(\beta_1)} \right) dy \right] \left[ \int_0^{\infty} \left( 1 - \frac{2(1 - e^{-\frac{y}{\alpha_2}})}{2 - e^{-\frac{y}{\alpha_2}}} \right) \left( \frac{\theta_2^{\beta_2} y^{\beta_2-1} e^{-\theta_2 y}}{\Gamma(\beta_2)} \right) dy \right]$$

$$R(2) = \left[ 1 - \sum_{n=1}^{\infty} \frac{n \theta_1^{\beta_1} \alpha_1^{\beta_1-1}}{2^n (n + \alpha_1 \theta_1)^{\beta_1}} \right] \left[ \sum_{n=1}^{\infty} \frac{n \theta_2^{\beta_2} \alpha_2^{\beta_2-1}}{2^n (n + \alpha_2 \theta_2)^{\beta_2}} \right]$$

$$R(3) = \left[ \int_0^{\infty} F_1(y) g_1(y) dy \right] \left[ \int_0^{\infty} F_2(y) g_2(y) dy \right] \left[ \int_0^{\infty} \bar{F}_3(y) g_3(y) dy \right]$$

$$R(3) = \left[ \int_0^{\infty} \left( \frac{2(1 - e^{-\frac{y}{\alpha_1}})}{2 - e^{-\frac{y}{\alpha_1}}} \right) \left( \frac{\theta_1^{\beta_1} y^{\beta_1-1} e^{-\theta_1 y}}{\Gamma(\beta_1)} \right) dx \right] \left[ \int_0^{\infty} \left( \frac{2(1 - e^{-\frac{y}{\alpha_2}})}{2 - e^{-\frac{y}{\alpha_2}}} \right) \left( \frac{\theta_2^{\beta_2} y^{\beta_2-1} e^{-\theta_2 y}}{\Gamma(\beta_2)} \right) dy \right] \\ \times \left[ \int_0^{\infty} \left( 1 - \frac{2(1 - e^{-\frac{y}{\alpha_3}})}{2 - e^{-\frac{y}{\alpha_3}}} \right) \left( \frac{\theta_3^{\beta_3} y^{\beta_3-1} e^{-\theta_3 y}}{\Gamma(\beta_3)} \right) dy \right]$$

$$R(3) = \left[ 1 - \sum_{n=1}^{\infty} \frac{n \theta_1^{\beta_1} \alpha_1^{\beta_1-1}}{2^n (n + \alpha_1 \theta_1)^{\beta_1}} \right] \left[ 1 - \sum_{n=1}^{\infty} \frac{n \theta_2^{\beta_2} \alpha_2^{\beta_2-1}}{2^n (n + \alpha_2 \theta_2)^{\beta_2}} \right] \left[ \sum_{n=1}^{\infty} \frac{n \theta_3^{\beta_3} \alpha_3^{\beta_3-1}}{2^n (n + \alpha_3 \theta_3)^{\beta_3}} \right]$$

$$R(v) = \left[ 1 - \sum_{n=1}^{\infty} \frac{n \theta_1^{\beta_1} \alpha_1^{\beta_1-1}}{2^n (n + \alpha_1 \theta_1)^{\beta_1}} \right] \left[ 1 - \sum_{n=1}^{\infty} \frac{n \theta_2^{\beta_2} \alpha_2^{\beta_2-1}}{2^n (n + \alpha_2 \theta_2)^{\beta_2}} \right] \dots \\ \times \left[ 1 - \sum_{n=1}^{\infty} \frac{n \theta_{v-1}^{\beta_{v-1}} \alpha_{v-1}^{\beta_{v-1}-1}}{2^n (n + \alpha_{v-1} \theta_{v-1})^{\beta_{v-1}}} \right] \left[ \sum_{n=1}^{\infty} \frac{n \theta_v^{\beta_v} \alpha_v^{\beta_v-1}}{2^n (n + \alpha_v \theta_v)^{\beta_v}} \right]$$

$$R(v) = \left[ \sum_{n=1}^{\infty} \frac{n \theta_v^{\beta_v} \alpha_v^{\beta_v-1}}{2^n (n + \alpha_v \theta_v)^{\beta_v}} \right] \prod_{i=1}^v \left[ 1 - \sum_{n=1}^{\infty} \frac{n \theta_{i-1}^{\beta_{i-1}} \alpha_{i-1}^{\beta_{i-1}-1}}{2^n (n + \alpha_{i-1} \theta_{i-1})^{\beta_{i-1}}} \right]$$

### 3. NUMERICAL EVALUATION

For different cases of stress strength distributions viz. M-Transformed Exponential, Lindley, Two Parameter Gamma and Exponential distributions the marginal reliabilities

$R(1), R(2), R(2)$  and the system reliability  $R_3$  have been evaluated using R for some distinct values of the parameters included in the expressions of  $R(1), R(2), R(2)$ .

**Table1. Exponential strength and M-transformed exponential stress**

$\alpha_1$	$\alpha_2$	$\alpha_3$	$\lambda_1$	$\lambda_2$	$\lambda_3$	R(1)	R(2)	R(3)	$R_3$
1	1	1	1	1	1	0.6137	0.2371	0.0916	0.9424
1	1	1	2	2	2	0.4548	0.2480	0.1352	0.8380
1	1	1	3	3	3	0.3645	0.2316	0.1472	0.7433
2	2	2	1	1	1	0.4548	0.2480	0.1352	0.8380
2	2	2	3	3	3	0.2315	0.1778	0.1367	0.5460
1	1	0.3	1	1	0.3	0.6137	0.2371	0.1406	0.9914
1	1	0.2	2	2	0.2	0.4548	0.2480	0.2893	0.9921
1	1	0.1	3	3	0.1	0.3645	0.2316	0.4012	0.9973
2	2	0.3	1	1	0.3	0.4548	0.2480	0.2800	0.9828
2	2	0.2	2	2	0.2	0.3052	0.2120	0.4698	0.9870
2	2	0.1	3	3	0.1	0.2315	0.1778	0.5865	0.9958

From Table 1 it is clear,  $R_3$  decreases by increasing the value of  $\alpha_1, \alpha_2, \lambda_1$  and  $\lambda_2$ . But if the value of  $\alpha_3$  and  $\lambda_3$  is decreased  $R_3$  increases. For example if  $\alpha_1, \alpha_2 = 1, \lambda_1, \lambda_2 = 1, 2, 3$  and for varying values of  $\alpha_3, \lambda_3$  decreases  $R_3$  increases from 0.7433 to 0.9973. Particularly, when  $\alpha_1, \alpha_2 = 1, \lambda_1, \lambda_2 = 3$  and  $\alpha_3, \lambda_3 = 0.3$ , the system reliability is maximum

**Table 2. Lindley strength and M-transformed exponential stress**

$\alpha_1$	$\alpha_2$	$\alpha_3$	$\lambda_1$	$\lambda_1$	$\lambda_3$	R(1)	R(2)	R(3)	$R_3$
1	1	1	1	1	1	0.7246	0.1996	0.0549	0.9791
1	1	1	2	2	2	0.5313	0.2490	0.1167	0.8970
1	1	1	3	3	3	0.4180	0.2432	0.1416	0.8028
2	2	2	1	1	1	0.5695	0.2452	0.1055	0.9202
2	2	2	3	3	3	0.2731	0.1985	0.1443	0.6159
1	1	0.3	1	1	0.3	0.7246	0.1996	0.0746	0.9988
1	1	0.2	2	2	0.2	0.5313	0.2490	0.2185	0.9988
1	1	0.1	3	3	0.1	0.4180	0.2432	0.3385	0.9997
2	2	0.3	1	1	0.3	0.5695	0.2452	0.1823	0.9970
2	2	0.2	2	2	0.2	0.3708	0.2333	0.3938	0.9979
2	2	0.1	3	3	0.1	0.2731	0.1985	0.5280	0.9996

From Table 2 it is clear,  $R_3$  decreases by increasing the value of  $\lambda_1, \lambda_2$  and  $\lambda_3$ . But if the value of  $\lambda_1, \lambda_2$  and  $\lambda_3$  is decreased  $R_3$  increases. For example if  $\alpha_1, \alpha_2, \alpha_3 = 1$  and  $\lambda_1, \lambda_2, \lambda_3 = 1, 2, 3$  then  $R_3$  decreases from 0.9791 to 0.8028. Also for some proper values of the parameters system reliability can be attained very close to one

**Table 3. M-transformed exponential strength and M-transformed exponential stress**

$\alpha_1$	$\alpha_2$	$\alpha_3$	$\beta_1$	$\beta_2$	$\beta_3$	R(1)	R(2)	R(3)	$R_3$
1	1	1	1	1	1	0.5000	0.2500	0.1250	0.8750
1	1	1	2	2	2	0.3333	0.2222	0.1481	0.7036
1	1	1	3	3	3	0.2500	0.1875	0.1406	0.5781
2	2	2	1	1	1	0.3333	0.1111	0.0740	0.5184
2	2	2	3	3	3	0.2000	0.2400	0.1440	0.5840
1	1	0.3	1	1	0.3	0.5000	0.2500	0.1250	0.8750
1	1	0.2	2	2	0.2	0.3333	0.2222	0.2222	0.7777
1	1	0.1	3	3	0.1	0.2500	0.1875	0.2812	0.7187
2	2	0.3	1	1	0.3	0.6666	0.2222	0.0555	0.9443
2	2	0.2	2	2	0.2	0.5000	0.2500	0.1250	0.8750
2	2	0.1	3	3	0.1	0.4000	0.2400	0.1800	0.8200

Clearly from Table 3, the system reliability can be approached closer to one by taking particular values of the parameters. For example  $\alpha_1 = \alpha_2 = \beta_1 = \beta_2 = 2$  and  $\alpha_3 = \beta_3 = 0.3$ , the system reliability is 0.9443. As  $\beta_i, i = 1, 2, 3$  the system reliability decreases and it increases when  $\alpha_3$  and  $\beta_3$  is decreased



**Table 4. M-transformed exponential strength and two parameter gamma stress**

$\alpha_1$	$\alpha_2$	$\alpha_3$	$\beta_1$	$\beta_2$	$\beta_3$	$\theta_1$	$\theta_2$	$\theta_3$	R(1)	R(2)	R(3)	$R_3$
1	1	1	2	2	2	1	1	1	0.2218	0.0492	0.0109	0.2819
1	1	1	2	2	2	2	2	2	0.4586	0.2483	0.1344	0.8413
1	1	1	2	2	2	3	3	3	0.6426	0.2296	0.0821	0.9543
1	1	1	2	2	2	6	6	6	0.9981	0.0018	0.0000	0.9999
2	2	2	1	1	1	2	2	2	0.6105	0.2378	0.0926	0.9409
3	3	3	1	1	1	2	2	2	0.4630	0.2486	0.1335	0.8451
4	4	4	1	1	1	2	2	2	0.3739	0.2342	0.1467	0.7548
2	2	2	2	2	2	1	1	1	0.2293	0.1767	0.1362	0.5422
2	2	2	3	3	3	1	1	1	0.1250	0.1094	0.0957	0.3301
2	2	2	4	4	4	1	1	1	0.0722	0.0670	0.0621	0.2013
2	2	2	1	1	1	4	4	4	0.7478	0.1886	0.0475	0.9839

From Table 4 it is clear,  $R_3$  increases by increasing the value of  $\theta_1, \theta_2$  and  $\theta_3$ . But if the value of  $\theta_1, \theta_2$  and  $\theta_3$  is decreased  $R_3$  decreases. For example if  $\alpha_1, \alpha_2, \alpha_3 = 1$  and  $\beta_1 = \beta_2 = \beta_3 = 2$  then  $R_3$  increases from 0.2819 to 0.9999 by increasing the value of  $\theta_1, \theta_2$  and  $\theta_3$ . Also for some proper values of the parameters system reliability can be attained very close to one

**4. CONCLUSION**

In this paper, a system is considered with standby redundancy and have presumed that the distinct components in the system for both stress and strength variables are independent and have different probability distributions viz. M-Transformed Exponential, Exponential, Gamma and Lindley. In the last section, for different cases of stress strength distributions, the marginal reliabilities R(1),R(2),R(2) and the system reliability  $R_3$  have been evaluated for some distinct values of the parameters included in the expressions of R(1),R(2),R(2) and have showed that reliability of system can be monotonically increasing and monotonically decreasing for specific values of parameter. Thus, by proper choice of parameters leads to high reliability.

**COMPETING INTERESTS**

Authors have declared that no competing interests exist.

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