



New Exponentiated Exponential- Rayleigh Distribution: Structure and Properties

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Abstract

In this research, we will introduce a new distribution, which is called the new exponentiated exponential-Rayleigh distribution, which is built by adding the shape parameter to the cumulative function of the exponential-Rayleigh distribution resulting from merging the cdf function of three parameters continuous distribution. Two of these parameters are scale parameters, and another is the shape parameter and this is to make this new continuous statistical distribution more flexible than other continuous statistical distributions. Besides, discuss the mathematical and statistical properties of this new distribution, including (the mode, the median, the moments about the origin, the coefficients of skewness and kurtosis, the characteristic function, the moment-generating function, the factorial moments-generating function, and the mean time to failure). The shape property of each of the pdf, cdf, and hazard functions was also studied. Multiple shapes were drawn with different parameter values for each of the PDF, CDF, survival, and hazard functions. An organisational table was also created for some of these properties.

Keywords: Exponential distribution, Rayleigh distribution, Exponential Rayleigh distribution, Moments about the origin, Quantile function.

1. Introduction

In order for statistical distributions to be important tools in various disciplines, these distributions had to be flexible distributions that help in understanding the data (1-3). Because the current distributions are restricted distributions and these restrictions make them insufficient in modelling diverse data, many scientists have resorted to integrating or adding a new parameter to these distributions in order to be more flexible (4-6). There are several methods to add a new parameter to statistical distributions, among these methods is adding a new parameter to the cumulative distribution function (CDF) (7, 8). There have been many scientists who have followed this method to add a new parameter to the distribution (9,10). In 2008, it was presented the exponentiated exponential distribution, a novel distribution that serves as a failure time distribution or the generalised exponential distribution (11). In 2010, it was presented a study on a mixture of the Pareto distribution and the Exponentiated



exponential distribution (12). Also in 2012, it was introduced the extended Exponentiated Weibull distribution with Gamma. (13), Also in 2012, it was presented the inverted Exponentiated exponential Weibull distribution (14). In 2013, it was introduced the class of generalised exponentiated exponential distributions, which is a new class of distributions that extends the exponentiated exponential type distributions. (15), Also in 2013, it was presented the extension of the family of Exponentiated exponential T-X distributions with some applications where it was noted Hence, the T-X distribution can be generated with an upper limit of $-\log(1-F(x))$, and it is obvious that alternative T-X distribution types can be generated with different higher limits. (16). Also in this year, it was presented the complementary exponentiated exponential geometric distribution of life span, where they proposed expanding the exponential distribution by fusing it with a geometric distribution to create a new distribution family (17), It was presented the modified exponential gamma distribution in 2014: Generalisation and extension of the Gamma distribution probabilities (18). In 2015, it was presented the exponentiated power Lindley distribution, which is a new distribution with three parameters called the exponential Lindley distribution [EPLD] (19). In 2017, it was presented the Exponentiated-Exponential Weibull distribution with applications, where they presented a novel four-parameter model that uses the (EW-G) distribution as a competitive extension of the Weibull distribution, which offers some new distributions and can be used to access several contemporary distributions (20). In 2018, it was was able to present the Exponentiated BURR X exponential distribution, which is the Burr X exponential distribution (BrxEE) and is a novel version of the (EE) model from which some of its statistical properties, including moments, incomplete moments, moment generating function, average deviations, and probability-weighted moments, were derived, and others (21). In the same year, it was introduced the Exponentiated Exponential Gamma distribution, where he defined the exponential distribution of Gamma and also introduced the Shannon entropy of (EGED) (22). In 2019, it was presented about the Exponentiated Weibull Rayleigh distribution (EWRD) (23). In 2020, it was presented the Exponentiated Exponential Weibull Distribution with mathematical properties and application and came up with the proposal of the Exponential Weibull Distribution (EWD) based on the idea of (24). Also in the same year, it was presented estimation techniques and applications to engineering data for the Odd Exponentiated half-logistic exponential distribution (25). In 2021, it was presented a new exponentiated generalised linear exponential distribution with properties and applications (26). In 2022, Dhungana G. and Kumar V. reported the Exponentiated Odd Lomax exponential distribution with application to COVID-19 death cases of Nepal (27). In 2023, it was presented an extension of the Exponentiated Rayleigh distribution with properties and applications, adding the shape parameter to improve and enhance the flexibility of this distribution (28). Finally, Lamyaa Khalid Hussein and Iden Hassan Hussein in 2023 represent a new distribution that is mixed between exponential and Rayleigh distributions (29). In this research, we will present a new mixture of continuous distributions, which we call the New Exponentiated Exponential-Rayleigh distribution properties and Applications by adding the shape parameter to the (CDF) function (30). The results of the document include the following: In Section 2, we introduced the fundamental statistical functions, including the probability density function, reciprocal density function, survival function, and hazard rate function, and explained how we extended the probability density function to obtain the Exponentiated Exponential-Rayleigh Distribution. The statistical and mathematical characteristics of the new distribution, including the mode, moments around the origin,

variance, skewness coefficient, characteristic function, moment generating function, and factorial moment generating function, are finally represented in Section 3.

2.The Exponentiated Exponential- Rayleigh Distribution:

$$F_{EER}(x) = (1 - e^{-(\delta x + \frac{\alpha}{2}x^2)})^\lambda \quad \delta, \alpha, \lambda > 0, x > 0 \quad (1)$$

$$f_{EER}(x; \delta, \alpha, \lambda) = \lambda (\delta + \alpha x) e^{-(\delta x + \frac{\alpha}{2}x^2)} (1 - e^{-(\delta x + \frac{\alpha}{2}x^2)})^{\lambda-1} \quad (2)$$

and survival (reliability) function defined as:

$$S_{EER}(x) = 1 - F(x) = 1 - (1 - e^{-(\delta x + \frac{\alpha}{2}x^2)})^\lambda \quad (3)$$

So the hazard rate function is:

$$h(x) = \frac{f(x)}{s(x)} = \frac{\lambda (\delta + \alpha x) e^{-(\delta x + \frac{\alpha}{2}x^2)} (1 - e^{-(\delta x + \frac{\alpha}{2}x^2)})^{\lambda-1}}{1 - (1 - e^{-(\delta x + \frac{\alpha}{2}x^2)})^\lambda} \quad (4)$$

$$f(x) \geq 0 \quad \forall \quad \delta, \alpha, \lambda > 0, x > 0$$

To proof $\int_0^\infty f(x) dx = 1$

$$\begin{aligned} \int_0^\infty \lambda (\delta + \alpha x) e^{-(\delta x + \frac{\alpha}{2}x^2)} (1 - e^{-(\delta x + \frac{\alpha}{2}x^2)})^{\lambda-1} dx &= \left[(1 - e^{-(\delta x + \frac{\alpha}{2}x^2)})^\lambda \right]_0^\infty \\ &= [(1 - e^{-(\infty)})^\lambda - (1 - e^0)^\lambda] = [(1 - 0)^\lambda - (1 - 1)^\lambda] = 1 - 0 = 1 \end{aligned}$$

2.1. Expand the Probability Density function:

This step has been explained in the previous steps.

$$f(x) = \lambda (\delta + \alpha x) e^{-(\delta x + \frac{\alpha}{2}x^2)} (1 - e^{-(\delta x + \frac{\alpha}{2}x^2)})^\lambda (1 - e^{-(\delta x + \frac{\alpha}{2}x^2)})^{-1}$$

$$f(x) = \lambda (\delta + \alpha x) e^{-(\delta x + \frac{\alpha}{2}x^2)} \sum_{k=0}^{\infty} (-1)^k \binom{\lambda}{k} e^{-k(\delta x + \frac{\alpha}{2}x^2)} \sum_{j=0}^{\infty} \frac{\Gamma(1+j)}{j! \Gamma(1)} e^{-j(\delta x + \frac{\alpha}{2}x^2)}$$

$$f(x) = \lambda (\delta + \alpha x) \sum_{k=0}^{\infty} \sum_{j=0}^{\infty} (-1)^k \binom{\lambda}{k} \frac{\Gamma(1+j)}{j!} e^{-(k+j+1)(\delta x + \frac{\alpha}{2}x^2)}$$

$$e^{-(k+j+1)(\frac{\alpha}{2}x^2)} = \sum_{b=0}^{\infty} \frac{(-1)^b (k+j+1)^b \alpha^b}{2^b b!} x^{2b}$$

$$\text{Let } \Psi_{k,j,b} = \lambda \sum_{k=0}^{\infty} \sum_{j=0}^{\infty} \sum_{b=0}^{\infty} \frac{(-1)^{k+b} (k+j+1)^b \alpha^b \Gamma(1+j)}{2^b b! j!}$$

$$f(x) = \Psi_{k,j,b} x^{2b} (\delta + \alpha x) e^{-(k+j+1)\delta x}$$

2.1.1 The shapes of the New Exponentiated Exponential- Rayleigh Distribution

Knowing the shape of the (SHWD) helps us understand the behavior and approach of distribution functions in dealing with data, to understand this mathematically, especially through the limit values of the probability density and hazard functions when $(x \rightarrow 0 \& x \rightarrow \infty)$.

$$\lim_{x \rightarrow 0} f(x) = \lim_{x \rightarrow 0} \left[\lambda (1 - e^{-(\delta x + \frac{\alpha}{2}x^2)})^{\lambda-1} e^{-(\delta x + \frac{\alpha}{2}x^2)} (\delta + \alpha x) \right] = \lim_{x \rightarrow 0} [\lambda (1 - e^0)^{\lambda-1} e^0 (\delta + 0)] = 0$$

$$\begin{aligned}
\lim_{x \rightarrow \infty} f(x) &= \lim_{x \rightarrow \infty} \left[\lambda (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{\lambda-1} e^{-(\delta x + \frac{\alpha}{2} x^2)} (\delta + \alpha x) \right] \\
&= \lim_{x \rightarrow \infty} \frac{\lambda (\delta + \alpha x) (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{\lambda}}{e^{(\delta x + \frac{\alpha}{2} x^2)} (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})} \\
&= \lim_{x \rightarrow \infty} \frac{\lambda (\delta + \alpha x)}{(e^{(\delta x + \frac{\alpha}{2} x^2)} - 1) (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{-\lambda}}
\end{aligned}$$

2.2. Applied L'Hospital's Rule:

$$\lim_{x \rightarrow \infty} f(x) = \lim_{x \rightarrow \infty} \frac{\lambda \alpha}{\infty} = 0$$

As we continue to derive for x the result of the numerator will be equal constant as considered an integer value for β and as contained to derive the denominator it always contains the exponential part which is equal to ∞ as $(x \rightarrow \infty)$. The final result of the limit is a constant divide by ∞ which equal to zero $\lim_{x \rightarrow \infty} f(x) = 0$.

$$\lim_{x \rightarrow 0} F(x) = \lim_{x \rightarrow 0} (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{\lambda} = 0$$

$$\lim_{x \rightarrow \infty} F(x) = \lim_{x \rightarrow \infty} (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{\lambda} = 1$$

The shape of hazard function can be defined as follows :-

$$h(x) = \frac{\lambda (\delta + \alpha x) e^{-(\delta x + \frac{\alpha}{2} x^2)} (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{\lambda-1}}{1 - (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{\lambda}}$$

$$\lim_{x \rightarrow 0} h(x) = \frac{\lambda \delta e^0 (1 - e^0)^{\lambda-1}}{1 - (1 - e^0)^{\lambda}} = \frac{0}{1} = 0$$

$$\lim_{x \rightarrow \infty} h(x) = \lim_{x \rightarrow \infty} \frac{\lambda (\delta + \alpha x) e^{-(\delta x + \frac{\alpha}{2} x^2)} (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{\lambda-1}}{[1 - (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{\lambda}]}$$

$$\lim_{x \rightarrow \infty} \frac{\lambda (\delta + \alpha x)}{e^{(\delta x + \frac{\alpha}{2} x^2)} (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)}) (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{-\lambda} [1 - (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{\lambda}]}$$

$$\lim_{x \rightarrow \infty} \frac{\lambda (\delta + \alpha x)}{(e^{(\delta x + \frac{\alpha}{2} x^2)} - 1) (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{-\lambda} [1 - (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{\lambda}]}$$

$$\lim_{x \rightarrow \infty} \frac{\lambda (\delta + \alpha x)}{(e^{(\delta x + \frac{\alpha}{2} x^2)} - 1) [(1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{-\lambda} - 1]} = \frac{\infty}{0}$$

By using L'Hopital's Rule:

$$\begin{aligned}
&= \lim_{x \rightarrow \infty} \frac{\lambda \alpha}{(e^{(\delta x + \frac{\alpha}{2} x^2)} - 1) [-\lambda (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{-(\lambda+1)} e^{-(\delta x + \frac{\alpha}{2} x^2)} (-\delta - \alpha x)] + [(1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{-\lambda} - 1] e^{(\delta x + \frac{\alpha}{2} x^2)} (\delta + \alpha x)} \\
&= \frac{\lambda \alpha}{(e^{\infty} - 1) [-\lambda (1 - e^{-\infty})^{-(\lambda+1)} e^{-\infty} (-\delta + \alpha \cdot \infty)] + [(1 - e^{-\infty})^{-\lambda} - 1] e^{\infty} (\delta + \alpha \cdot \infty)} \\
&= \frac{\lambda \alpha}{0 + \infty} = 0
\end{aligned}$$

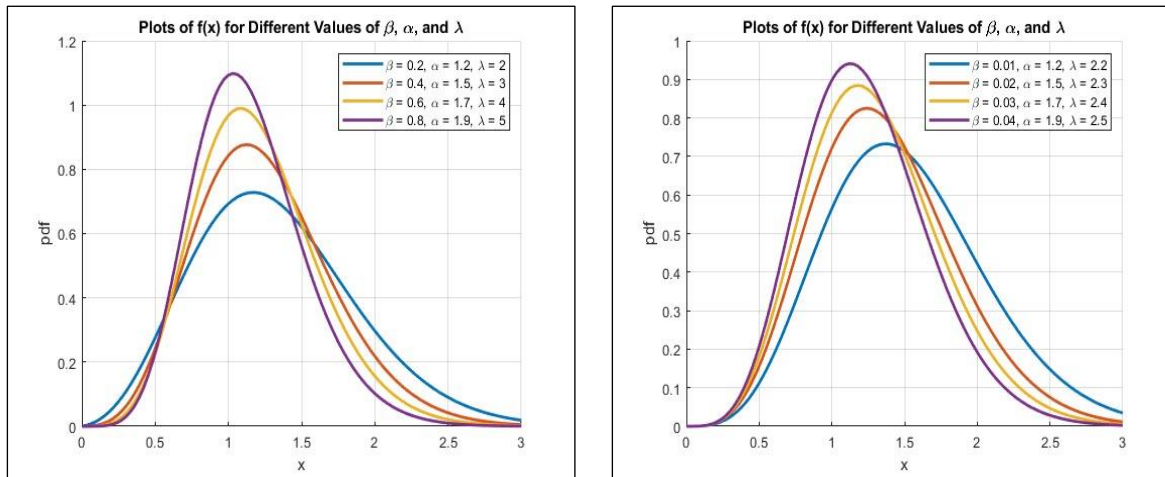


Figure 1. The shape of pdf for new Exponentiated Exponential- Rayleigh Distribution with different values of δ, α, λ .

This figure illustrates the graph of this function when $x \rightarrow \infty$ the PDF value is zero, and when $x \rightarrow 0$ the PDF value is also zero.

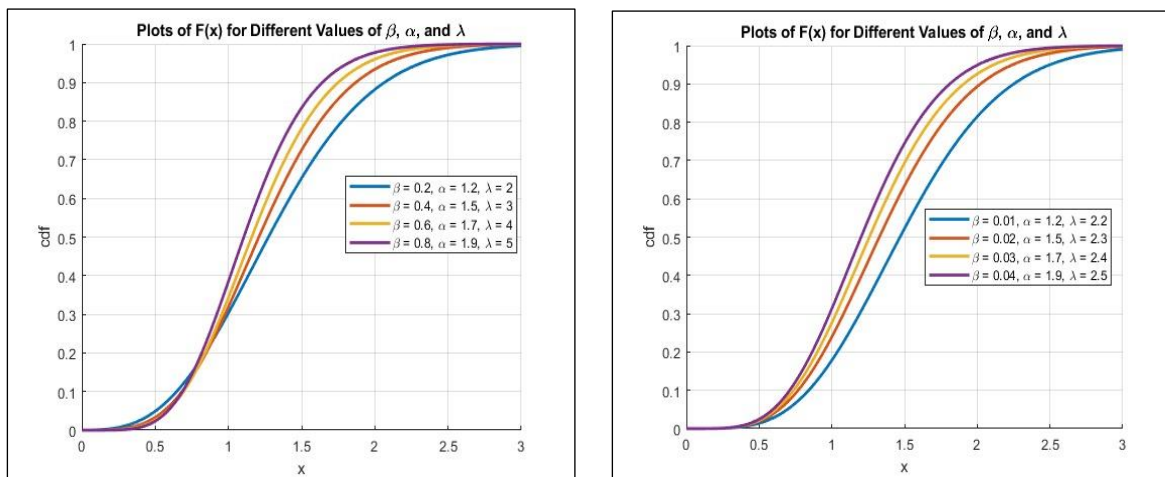


Figure 2. The shape of cdf for new Exponentiated Exponential- Rayleigh Distribution with different values of δ, α, λ .

This graph shows that the CDF function equals zero when $x \rightarrow 0$ and equals one when $x \rightarrow \infty$, which indicates that the CDF function is an increasing function.

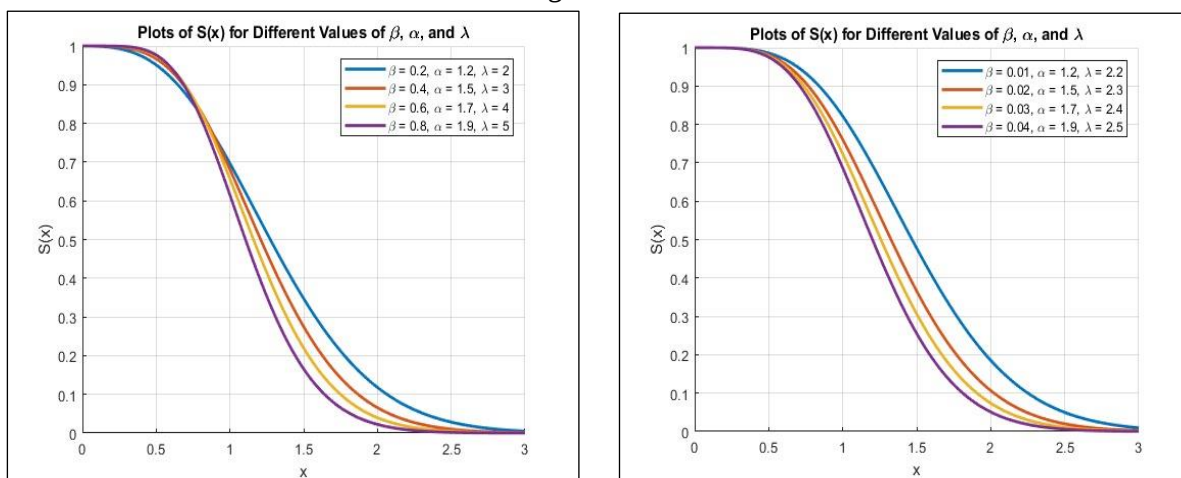


Figure 3. The shape of $S(x)$ for new Exponentiated Exponential- Rayleigh Distribution with different values of δ, α, λ .

This graph shows that the function $S(x)$ is a decreasing function.

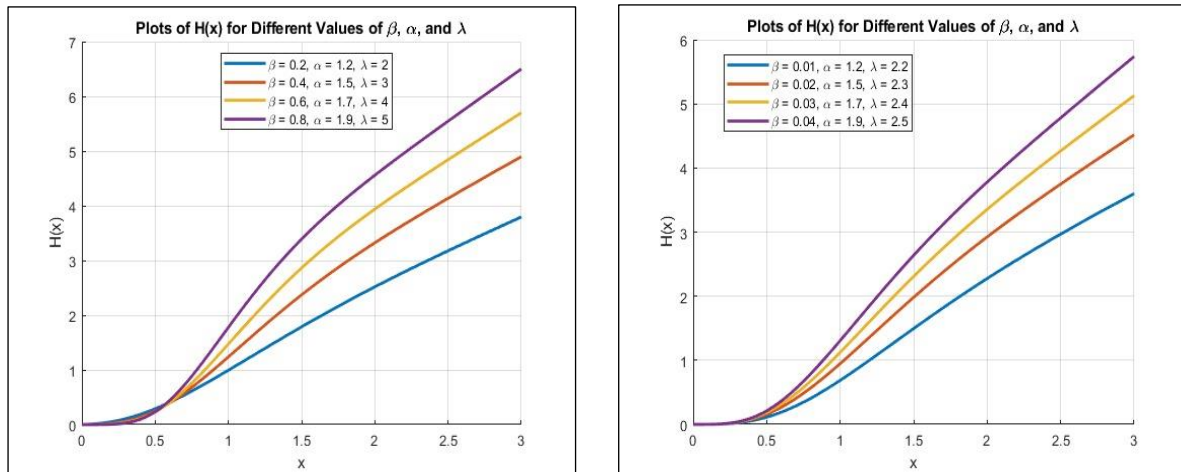


Figure 4. The shape of $h(x)$ for new Exponentiated Exponential- Rayleigh Distribution with different values of δ , α , and λ .

In the special case, if $\lambda = 1$; $\delta > 0$, then the distribution becomes the exponential Rayleigh distribution, also if $\lambda = 1$; $\delta = 0$, then the distribution becomes the Rayleigh distribution. Finally, if $\lambda = 1$; $\alpha = 0$, then the distribution becomes exponential distribution.

2.3. Quantile Function

This function defined by : $F(x) = u$; $u \sim u(0, 1)$; $x = F^{-1}(u)$

$$(1 - e^{-(\delta x + \frac{\alpha}{2}x^2)})^\lambda = u$$

$$1 - e^{-(\delta x + \frac{\alpha}{2}x^2)} = u^{\frac{1}{\lambda}}$$

$$1 - u^{\frac{1}{\lambda}} = e^{-(\delta x + \frac{\alpha}{2}x^2)}$$

$$\ln(1 - u^{\frac{1}{\lambda}}) = -(\delta x + \frac{\alpha}{2}x^2)$$

$$\frac{\alpha}{2}x^2 + \delta x + \ln(1 - u^{\frac{1}{\lambda}}) = 0$$

$$x = \frac{-\delta \pm \sqrt{\delta^2 - 2\alpha \ln(1 - u^{\frac{1}{\lambda}})}}{\alpha}$$

In spaiical case if $u = 0.5$ then the Medean is :

$$x = \frac{-\delta \pm \sqrt{\delta^2 - 2\alpha \ln(1 - (0.5)^{\frac{1}{\lambda}})}}{\alpha}$$

3. Mathematical and statistical properties of new (EERD)

3.1. The Mode

The mode is defined as the point at which the probability density function achieves its maximum value, the mode of (EER) is obtained as follows:

From **Equation 2** the mode of (EER) is defined by: $\frac{\partial f(x)}{\partial x} = 0$

$$\begin{aligned} \frac{\partial f(x)}{\partial x} = & \lambda(\delta + \alpha x) e^{-(\delta x + \frac{\alpha}{2}x^2)} \left((\lambda - 1)(1 - e^{-(\delta x + \frac{\alpha}{2}x^2)})^{\lambda-2} e^{-(\delta x + \frac{\alpha}{2}x^2)} (\delta + \alpha x) \right) \\ & + (1 - e^{-(\delta x + \frac{\alpha}{2}x^2)})^{\lambda-1} \left(-\lambda(\delta + \alpha x)^2 e^{-(\delta x + \frac{\alpha}{2}x^2)} + \lambda\alpha e^{-(\delta x + \frac{\alpha}{2}x^2)} \right) \end{aligned}$$

$$\begin{aligned} & \lambda(\lambda-1)(\delta+\alpha x)^2 e^{-2\left(\delta x+\frac{\alpha}{2}x^2\right)}(1-e^{-\left(\delta x+\frac{\alpha}{2}x^2\right)})^{\lambda-2} \\ & + \lambda e^{-\left(\delta x+\frac{\alpha}{2}x^2\right)}(1-e^{-\left(\delta x+\frac{\alpha}{2}x^2\right)})^{\lambda-1}(-(\delta+\alpha x)^2+\alpha)=0 \\ & \lambda e^{-\left(\delta x+\frac{\alpha}{2}x^2\right)}(1-e^{-\left(\delta x+\frac{\alpha}{2}x^2\right)})^{\lambda-2}\left((\lambda-1)(\delta+\alpha x)^2 e^{-\left(\delta x+\frac{\alpha}{2}x^2\right)}\right. \\ & \left.+(\alpha-(\delta+\alpha x)^2)(1-e^{-\left(\delta x+\frac{\alpha}{2}x^2\right)})\right)=0 \end{aligned}$$

since $\lambda > 0$, $\delta > 0$, $\alpha > 0$ and $x > 0$ then $\lambda e^{-\left(\delta x+\frac{\alpha}{2}x^2\right)}(1-e^{-\left(\delta x+\frac{\alpha}{2}x^2\right)})^{\lambda-2} \neq 0$

$$(\lambda-1)(\delta+\alpha x)^2 e^{-\left(\delta x+\frac{\alpha}{2}x^2\right)} + (1-e^{-\left(\delta x+\frac{\alpha}{2}x^2\right)})(\alpha-(\delta+\alpha x)^2) = 0$$

This equation is difficult to solve; it is solved using MATLAB programs.

3.2. Moments about the origin (31)

The r th moment about the origin can be obtained by:

$$M'_r(x) = \int_0^\infty x^r f(x) dx = \int_0^\infty x^r \left(\lambda (\delta + \alpha x) e^{-\left(\delta x + \frac{\alpha}{2}x^2\right)} (1 - e^{-\left(\delta x + \frac{\alpha}{2}x^2\right)})^{\lambda-1} \right) dx$$

$$(1 - e^{-\left(\delta x + \frac{\alpha}{2}x^2\right)})^{\lambda-1} = (1 - e^{-\left(\delta x + \frac{\alpha}{2}x^2\right)})^\lambda (1 - e^{-\left(\delta x + \frac{\alpha}{2}x^2\right)})^{-1}$$

$$(1 - e^{-\left(\delta x + \frac{\alpha}{2}x^2\right)})^\lambda = \sum_{k=0}^{\infty} (-1)^k \binom{\lambda}{k} e^{-k\left(\delta x + \frac{\alpha}{2}x^2\right)}$$

$$(1 - e^{-\left(\delta x + \frac{\alpha}{2}x^2\right)})^{-1} = \sum_{j=0}^{\infty} \frac{\Gamma(1+j)}{j!} e^{-j\left(\delta x + \frac{\alpha}{2}x^2\right)}$$

$$(1 - e^{-\left(\delta x + \frac{\alpha}{2}x^2\right)})^{\lambda-1} = \sum_{k=0}^{\infty} \sum_{j=0}^{\infty} \frac{(-1)^k \Gamma(1+j)}{j!} \binom{\lambda}{k} e^{-(k+j)\left(\delta x + \frac{\alpha}{2}x^2\right)}$$

$$M'_r(x) = \lambda \sum_{k=0}^{\infty} \sum_{j=0}^{\infty} \frac{(-1)^k \Gamma(1+j)}{j!} \binom{\lambda}{k} \int_0^\infty (\delta x^r + \alpha x^{r+1}) e^{-(k+j+1)\left(\delta x + \frac{\alpha}{2}x^2\right)} dx$$

$$\text{Since } e^{-(k+j+1)\left(\delta x + \frac{\alpha}{2}x^2\right)} = e^{-(k+j+1)\delta x} e^{-\frac{\alpha}{2}(k+j+1)x^2}$$

$$e^{-\frac{\alpha}{2}(k+j+1)x^2} = \sum_{b=0}^{\infty} \frac{(-1)^b (k+j+1)^b \alpha^b}{2^b b!} x^{2b}$$

$$\begin{aligned} M'_r(x) = & \lambda \sum_{k=0}^{\infty} \sum_{j=0}^{\infty} \sum_{b=0}^{\infty} \frac{(-1)^{j+b} (k+j+1)^b \alpha^b \Gamma(j+1)}{2^b j! b!} \int_0^\infty (\delta x^{r+2b} \\ & + \alpha x^{r+2b+1}) e^{-(k+j+1)\delta x} dx \end{aligned}$$

Suppose that

$$\lambda \sum_{k=0}^{\infty} \sum_{j=0}^{\infty} \sum_{b=0}^{\infty} \frac{(-1)^{j+b} (k+j+1)^b \alpha^b \Gamma(j+1)}{2^b j! b!} = \psi_{k,j,b}$$

$$M'_r(x) = \psi_{k,j,b} \left(\int_0^\infty \delta x^{r+2b} e^{-(k+j+1)\delta x} dx + \int_0^\infty \alpha x^{r+2b+1} e^{-(k+j+1)\delta x} dx \right)$$

$$\text{Let } y = (k+j+1)\delta x, \quad x = \frac{y}{\delta(k+j+1)}, \quad dx = \frac{dy}{\delta(k+j+1)}$$

$$M'_r(x) = \Psi_{k,j,b} \left(\int_0^{\infty} \delta \frac{y^{r+2b}}{(\delta(k+j+1))^{r+2b}} e^{-y} \frac{dy}{\delta(k+j+1)} + \int_0^{\infty} \alpha \frac{y^{r+2b+1}}{(\delta(k+j+1))^{r+2b+1}} e^{-y} \frac{dy}{\delta(k+j+1)} \right)$$

$$M'_r(x) = \Psi_{k,j,b} \frac{1}{\delta^{r+2b} (k+j+1)^{r+2b+1}} \Gamma(r+2b+1) + \Psi_{k,j,b} \frac{\alpha}{(\delta(k+j+1))^{r+2b+2}} \Gamma(r+2b+2)$$

As a direct result it Can be found

$E(x) = M'_1(x)$, $E(x^2) = M'_2(x)$ and $Var(x)$ as follows :

$$E(x) = M'_1(x) = \Psi_{k,j,b} \frac{1}{\delta^{2b+1} (k+j+1)^{2b+2}} \Gamma(2b+2) + \Psi_{k,j,b} \frac{\alpha}{(\delta(k+j+1))^{2b+3}} \Gamma(2b+3) \quad (5)$$

3.3. Mean Time to failure

We can find this property by find the expected : $E(x) = M'_1(x)$ as in **Equation 5**

$$E(x^2) = M'_2(x)$$

$$= \Psi_{k,j,b} \frac{1}{\delta^{2b+2} (k+j+1)^{2b+3}} \Gamma(2b+3) + \Psi_{k,j,b} \frac{\alpha}{(\delta(k+j+1))^{2b+4}} \Gamma(2b+4)$$

$$\text{var}(x) = M'_2(x) - (M'_1(x))^2$$

3.4. Coefficients of Skewness and Kurtosis (32)

Depending on the moment, the coefficients skewness ($C.S$) and kurtosis($C.K$) can be found through the following formulas (30):.

$$M'_3(x) = \Psi_{k,j,b} \frac{1}{\delta^{2b+3} (k+j+1)^{2b+4}} \Gamma(2b+4) + \Psi_{k,j,b} \frac{\alpha}{(\delta(k+j+1))^{2b+5}} \Gamma(2b+5)$$

$$C.S = \frac{M'_3(x)}{M'_2(x)^{\frac{3}{2}}}$$

$$M'_4(x) = \Psi_{k,j,b} \frac{1}{\delta^{2b+4} (k+j+1)^{2b+5}} \Gamma(2b+5) + \Psi_{k,j,b} \frac{\alpha}{(\delta(k+j+1))^{2b+6}} \Gamma(2b+6)$$

$$C.K = \frac{M'_4(x)}{(M'_2(x))^2} - 3$$

Table 1. The first - fourth moments, variance, skewness, and kurtosis for the distribution

δ	α	λ	μ'_1	μ'_2	μ'_3	μ'_4	K	S	var
2.5	2	1.5	0.4106	0.2538	0.2026	0.1946	0.0213	1.5849	0.0852
	1.5	1	0.3388	0.2039	0.1679	0.1708	1.1071	1.8228	0.0891
2	0.1	0.5	0.3005	0.2543	0.3407	0.6124	6.4725	2.6571	0.1640
	0.6	0.9	0.4168	0.3327	0.3747	0.5312	1.7988	1.9524	0.1590
1.5	2	0.7	0.3606	0.2448	0.2236	0.2474	1.1273	1.8457	0.1148
	2.8	0.5	0.2658	0.1536	0.1217	0.1173	1.9733	2.0217	0.0830

The distribution is skewed to the right based on the provided data values because the skewness values are positive. The values of the specified features determine the flatness. For numbers more than three, it is flattened; for values less than three, it is pointed.

3.5. Characteristic function (33)

$$\begin{aligned}
 \phi_x(it) &= E(e^{itx}) = \int_0^\infty e^{itx} f(x) dx = \\
 &= \int_0^\infty e^{itx} \left(\lambda (\delta + \alpha x) e^{-(\delta x + \frac{\alpha}{2} x^2)} (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^\lambda (1 - e^{-(\delta x + \frac{\alpha}{2} x^2)})^{-1} \right) dx \\
 \phi_x(it) &= \int_0^\infty e^{itx} \Psi_{k,j,b} x^{2b} (\delta + \alpha x) e^{-(k+j+b)\delta x} dx \\
 &= \Psi_{k,j,b} \left[\delta \int_0^\infty e^{itx} x^{2b} e^{-(k+j+b)\delta x} dx + \alpha \int_0^\infty e^{itx} x^{2b+1} e^{-(k+j+1)\delta x} dx \right] \\
 &= \Psi_{k,j,b} \delta \int_0^\infty x^{2b} e^{-((k+j+b)\delta - it)x} dx + \Psi_{k,j,b} \alpha \int_0^\infty x^{2b+1} e^{-((k+j+1)\delta - it)x} dx \\
 \text{let } ((k+j+b)\delta - it)x &= y \quad x = \frac{y}{(k+j+b)\delta - it} \quad dx = \frac{dy}{(k+j+b)\delta - it} \\
 &= \Psi_{k,j,b} \delta \int_0^\infty \left(\frac{y}{(k+j+b)\delta - it} \right)^{2b} e^{-y} \frac{dy}{(k+j+b)\delta - it} \\
 &\quad + \Psi_{k,j,b} \alpha \int_0^\infty \left(\frac{y}{(k+j+b)\delta - it} \right)^{2b+1} e^{-y} \frac{dy}{(k+j+b)\delta - it} \\
 &= \Psi_{k,j,b} \frac{\delta}{((k+j+b)\delta - it)^{2b+1}} \int_0^\infty y^{2b} e^{-y} dy \\
 &\quad + \Psi_{k,j,b} \frac{\alpha}{((k+j+b)\delta - it)^{2b+2}} \int_0^\infty y^{2b+1} e^{-y} dy \\
 \phi_x(it) &= \Psi_{k,j,b} \frac{\delta}{((k+j+b)\delta - it)^{2b+1}} \Gamma(2b+1) \\
 &\quad + \Psi_{k,j,b} \frac{\alpha}{((k+j+b)\delta - it)^{2b+2}} \Gamma(2b+2)
 \end{aligned}$$

3.6. Moment Generating Function (34)

$$\begin{aligned}
 M_x(t) &= E(e^{tx}) = \int_0^\infty e^{tx} f(x) dx \\
 M_x(t) &= \lambda \sum_{k=0}^\infty \sum_{j=0}^\infty \sum_{b=0}^\infty \frac{(-1)^{i+b} (i+j+1)^b (\alpha)^b \Gamma(j+1)}{2^b j! b!} \left[\int_0^\infty \delta x^{2b} e^{-((k+j+1)\delta - t)x} dx \right. \\
 &\quad \left. + \Psi_{k,j,b} \int_0^\infty \alpha e^{2b+1} e^{-((k+j+1)\delta - t)x} dx \right] \\
 \text{Let } y &= (k+j+1)\delta - tx \quad x = \frac{y}{(k+j+1)\delta - t} \quad dx = \frac{dy}{(k+j+1)\delta - t}
 \end{aligned}$$

$$\begin{aligned}
M_x(t) &= \Psi_{k,j,b} \frac{\delta}{((k+j+1)\delta - t)^{2b+1}} \int_0^\infty y^{2b} e^{-y} dy \\
&\quad + \Psi_{k,j,b} \frac{\alpha}{((k+j+1)\delta - t)^{2b+2}} \int_0^\infty y^{2b+1} e^{-y} dy \\
M_x(t) &= \Psi_{k,j,b} \frac{\delta}{((k+j+1)\delta - t)^{2b+1}} \Gamma(2b+1) + \Psi_{k,j,b} \frac{\alpha}{((k+j+1)\delta - t)^{2b+2}} \Gamma(2b+2)
\end{aligned}$$

3.7. Factorial Moments Generating Function (35)

$$\begin{aligned}
\mathcal{M}_x(t) &= E(e^{tx}) = \int_0^\infty t^x f(x) dx = \int_0^\infty e^{x \ln t} f(x) dx \\
\mathcal{M}_x(t) &= \int_0^\infty e^{x \ln t} \Psi_{k,j,b} x^{2b} (\delta + \alpha x) e^{-(k+j+1)\delta x} dx \\
\mathcal{M}_x(t) &= \Psi_{k,j,b} \int_0^\infty e^{x \ln t} x^{2b} (\delta + \alpha x) e^{-(k+j+1)\delta x} dx \\
\mathcal{M}_x(t) &= \Psi_{k,j,b} \left[\int_0^\infty x^{2b} \delta e^{-(k+j+1)\delta - \ln t} x dx + \int_0^\infty \alpha x^{2b+1} e^{-(k+j+1)\delta - \ln t} x dx \right] \\
\text{let } y &= [(k+j+1)\delta - \ln t]x \quad x = \frac{y}{(k+j+1)\delta - \ln t} \quad dx = \frac{dy}{(k+j+1)\delta - \ln t} \\
\mathcal{M}_x(t) &= \Psi_{k,j,b} \left[\int_0^\infty \delta \frac{y^{2b}}{[(k+j+1)\delta - \ln t]^{2b}} e^{-y} \frac{dy}{(k+j+1)\delta - \ln t} \right. \\
&\quad \left. + \int_0^\infty \alpha \frac{y^{2b+1}}{[(k+j+1)\delta - \ln t]^{2b+1}} e^{-y} \frac{dy}{(k+j+1)\delta - \ln t} \right] \\
\mathcal{M}_x(t) &= \Psi_{k,j,b} \frac{\delta}{[(k+j+1)\delta - \ln t]^{2b+1}} \int_0^\infty y^{2b} e^{-y} dy \\
&\quad + \frac{\alpha}{[(k+j+1)\delta - \ln t]^{2b+2}} \int_0^\infty y^{2b+1} e^{-y} dy \\
\mathcal{M}_x(t) &= \Psi_{k,j,b} \frac{\delta}{[(k+j+1)\delta - \ln t]^{2b+1}} \Gamma(2b+1) + \Psi_{k,j,b} \frac{\alpha}{[(k+j+1)\delta - \ln t]^{2b+2}} \Gamma(2b+2)
\end{aligned}$$

4. Conclusion

A new statistical lifetime distribution is represented by adding the shape parameter to the (cdf) function of the Exponentiated -Rayleigh Distribution called "Exponentiated Exponential-Rayleigh Distribution.". In addition, the shape of the probability density function, the cumulative function, and the hazard function are discussed. Also, the basic statistical and mathematical properties of the new distribution, such as mode, median, r-moments around the origin, and the moment-generating function, are varied. Finally, introduce the table of applications with different values of parameters.

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Conflict of Interest

No conflicts of interest.

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References

1. Mahdi G, Mohammed SF, Khan MKH. Enhanced support vector machine methods using stochastic gradient descent and its application to heart disease dataset. *Ibn Al-Haitham J Pure Appl Sci.* 2024;37(1):412-428. <https://doi.org/10.30526/37.1.3467>.
2. Jebur IG, Kalaf BA, Salman AN. An efficient shrinkage estimators for generalized inverse Rayleigh distribution based on bounded and series stress-strength models. *J Phys Conf Ser.* 2021;1897(1):012054. <https://doi.org/10.1088/1742-6596/1897/1/012054>.
3. Habeeb SB, Abdullah FK, Shalan RN, Hassan AS, Almetwally EM, Alghamdi FM, Alsheikh SM, Hossain MM. Comparison of some Bayesian estimation methods for type-I generalized extreme value distribution with simulation. *Alex Eng J.* 2024;98:356-363. <https://doi.org/10.1016/j.aej.2024.04.042>.
4. Shamran MA, Mohammed AA, Alkanani IH. Comparison between modified weighted Pareto distribution and many other distributions. *Baghdad Sci J.* 2023;20(3 Suppl):1108. <https://doi.org/10.21123/bsj.2023.8169>.
5. Aldraji ZA. Reliability function estimated for generalized exponential Rayleigh distribution under type-I censored data and fuzzy data. *Int J Neutrosophic Sci.* 2024;24(2). <https://doi.org/10.54216/IJNS.240202>.
6. Mohammed AT, Hussein IH. Nonparametric estimation for hazard rate function by wavelet procedures with simulation. *IOP Conf Ser Mater Sci Eng.* 2019;571(1):012013. <https://doi.org/10.1088/1757-899X/571/1/012013>.
7. Shalan RN, Alkanani IH. The simulation technique to estimate the parameters of generalized exponential Rayleigh model. *Iraqi J Sci.* 2024;65(3):1502-1520. <https://doi.org/10.24996/ij.2024.65.3.27>.
8. Mohammed AT, Mohammed MJ, Salman MD, Ibrahim RW. The inverse exponential Rayleigh distribution and related concepts. *Ital J Pure Appl Math.* 2022;47:852-861.
9. Raheem SH, Kalaf BA, Salman AN. Comparison of some estimation methods of stress-strength model: $R=$. *Baghdad Sci J.* 2021;18(1 Suppl):1103. [https://doi.org/10.21123/bsj.2021.18.1\(Suppl.\).1103](https://doi.org/10.21123/bsj.2021.18.1(Suppl.).1103).
10. Shalan RN. Compare some shrinkage Bayesian estimation method for Gumbel-Max distribution with simulation. *Iraqi J Sci.* 2024;65(4):2150-2159. <https://doi.org/10.24996/ij.2024.65.4.32>.
11. Algamal ZY. Exponentiated exponential distribution as a failure time distribution. *Iraqi J Stat Sci.* 2008;14:63-75.
12. Abuzinadah H. A study on mixture of exponentiated Pareto and exponential distributions. *J Appl Sci Res.* 2010;6(4).
13. Pinho LG, Cordeiro GM, Nobre JS. The gamma-exponentiated Weibull distribution. *J Stat Theory Appl.* 2012;11(4):379-395.
14. Flaih A, Elsalloukh H, Mendi E, Milanova M. The exponentiated inverted Weibull distribution. *Appl Math Inf Sci.* 2012;6(2):167-171.
15. Cordeiro GM, Ortega EM, da Cunha DC. The exponentiated generalized class of distributions. *J Data Sci.* 2013;11(1):1-27.
16. Alzaghal A, Famoye F, Lee C. Exponentiated TX family of distributions with some applications. *Int J Stat Probab.* 2013;2(3):31. <https://doi.org/10.5539/ijsp.v2n3p31>.
17. Louzada F, Marchi V, Carpenter J. The complementary exponentiated exponential geometric lifetime distribution. *J Probab Stat.* 2013;2013:502159. <https://doi.org/10.1155/2013/502159>.
18. Hussian MA. Transmuted exponentiated gamma distribution: A generalization of the

- exponentiated gamma probability distribution. Appl Math Sci. 2014;8(27):1297-1310. <https://doi.org/10.12988/ams.2014.42105>.
- 19.Ashour SK, Eltehiwy MA. Exponentiated power Lindley distribution. J Adv Res. 2015;6(6):895-905. <https://doi.org/10.1016/j.jare.2014.08.005>.
 - 20.Elgarhy M, Shakil M, Kibria G. Exponentiated Weibull-exponential distribution with applications. Appl Appl Math. 2017;12(2):5.
 - 21.Refaie MK. Burr X exponentiated exponential distribution. J Stat Appl. 2018;1(2):71-88.
 - 22.Jabeen S, Para B. Exponentiated gamma exponential distribution. Sohag J Math. 2018;5(3):79-84. <https://doi.org/10.18576/sjm/050301>.
 - 23.Elgarhy M. On the exponentiated Weibull Rayleigh distribution. Gazi Univ J Sci. 2019;32(3):1060-1081. <https://doi.org/10.35378/gujs.315832>.
 - 24.Al-Sulami D. Exponentiated exponential Weibull distribution: mathematical properties and application. Am J Appl Sci. 2020;17(1):188-195. <https://doi.org/10.3844/ajassp.2020.188.195>.
 - 25.Aldahlan MA, Afify AZ. The odd exponentiated half-logistic exponential distribution: estimation methods and application to engineering data. Mathematics. 2020;8(10):1684. <https://doi.org/10.3390/math8101684>.
 - 26.Poonia N, Azad S. A new exponentiated generalized linear exponential distribution: properties and application. Res Math Stat. 2021;8(1):1953233. <https://doi.org/10.1080/27658449.2021.1953233>.
 - 27.Dhungana GP, Kumar V. Exponentiated odd Lomax exponential distribution with application to COVID-19 death cases of Nepal. PLoS One. 2022;17(6):e0269450. <https://doi.org/10.1371/journal.pone.0269450>.
 - 28.Bhat AA, Ahmad SP. An extension of exponentiated Rayleigh distribution: properties and applications. Thailand Stat. 2023;21(1):209-227.
 - 29.Hussein LK, Rasheed HA, Hussein IH. A class of exponential Rayleigh distribution and new modified weighted exponential Rayleigh distribution with statistical properties. Ibn Al-Haitham J Pure Appl Sci. 2023;36(2):390-406. <https://doi.org/10.30526/36.2.3044>.
 - 30.Ghazal MG, Hasaballah HM. Exponentiated Rayleigh distribution: A Bayes study using MCMC approach based on unified hybrid censored data. J Adv Math. 2017;12(12):6863-6880.
 - 31.Hassan AS, Khaleel MA, Mohamd RE. An extension of exponentiated Lomax distribution with application to lifetime data. Thailand Stat. 2021;19(3):484-500.
 - 32.Ahmed MT, Khaleel MA, Oguntunde PE, Abdal-Hammed MK. A new version of the exponentiated Burr X distribution. J Phys Conf Ser. 2021;1818(1):012116. <https://doi.org/10.1088/1742-6596/1818/1/012116>.
 - 33.Ademola A, Adeyeye J, Khaleel M, Aako O. Exponentiated Gompertz exponential (EGOE) distribution: derivation, properties and applications. Istatistik J Turk Stat Assoc. 2021;13(1):12-28.
 - 34.Murat U, Özel G. Exponentiated Weibull-logistic distribution. Bilge Int J Sci Technol Res. 2020;4(2):55-62.
 - 35.Sridhar Babu M, Srinivasa Rao G, Rosaiah K. Double-acceptance sampling plan for exponentiated Fréchet distribution with known shape parameters. Math Probl Eng. 2021;2021:7308454. <https://doi.org/10.1155/2021/7308454>.