Kumaraswamy Weighted Exponential Distribution

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Abstract:

In this paper, a new extension distribution with four parameters named Kumaraswamy weighted exponential has been introduced based on the family of Kumaraswamy generalized distribution. The new density function can be expressed as an infinite linear combination of weighted exponential densities. Some of the mathematical properties, special cases along with the maximum likelihood estimations of the parameters of new distribution have been discussed.

> توزيع كوماراسوامي الأسي الموزون نادية هاشم النور لمياء خالد حسين الجامعة المستنصرية – بغداد – العراق

المستخلص: في هذا البحث، تم تقديم توزيع كومار اسو امي الأسي الموزون كتوزيع جديد موسع بأربعة معلمات استنادا إلى عائلة التوزيع الموسع كومار اسو امي. يمكن صياغة دالة الكثافة الجديدة على أنها مزيج خطي غير منتهي من الكثافة الأسية الموزونة. تم مناقشة بعض الخصائص الرياضية ، الحالات الخاصة فضلا عن مقدر ات الامكان الاعظم لمعلمات التوزيع الجديد.

1. Introduction

The weighted exponential (*WE*) distribution has been introduced by Gupta and Kundu in (2009) [3]. The *WE* distribution has received appreciable usage in the fields of engineering and medicine [7]. The probability density function (pdf) of *WE* distribution is given by [3]:

$$f_{WE}(x;\alpha,\lambda) = \frac{\alpha+1}{\alpha} \lambda e^{-\lambda x} (1 - e^{-\alpha\lambda x}) ; x > 0 ; \alpha,\lambda > 0 \qquad \dots (1)$$

The corresponding cumulative distribution function of WE distribution is given by [1]:

$$F_{WE}(x;\alpha,\lambda) = 1 - \frac{1}{\alpha} e^{-\lambda x} \left(\alpha + 1 - e^{-\alpha\lambda x}\right) \qquad \dots (2)$$

The reliability and hazard functions of WE distribution at time (t), respectively, are given by [1]:

$$R_{WE}(t;\alpha,\lambda) = 1 - F(t;\alpha,\lambda) = \frac{1}{\alpha} e^{-\lambda t} \left(\alpha + 1 - e^{-\alpha\lambda t}\right) \qquad \dots (3)$$

$$\begin{split} h_{WE}(t;\alpha,\lambda) &= \frac{f(t;\alpha,\lambda)}{R(t;\alpha,\lambda)} = \frac{(\alpha+1)\lambda\left(1-e^{-\alpha\lambda t}\right)}{\alpha+1-e^{-\alpha\lambda t}} \qquad \dots (4) \\ \text{The } r^{th} \text{ moments about the origin is [4]:} \\ E(X^r) &= \frac{(\alpha+1)\Gamma(r+1)}{\alpha\lambda^r} \left(1-\frac{1}{(1+\alpha)^{r+1}}\right); \ r = 1,2,3,\dots \dots (5) \\ \text{The moment generating function, } M_X(t), \text{ for } -1 < t < 1 \text{ can be expressed by } \\ [5]: \\ M_X(t) &= E(e^{tX}) = \frac{(\alpha+1)\lambda}{\alpha} \left[\frac{1}{\lambda-t} - \frac{1}{\lambda-t+\alpha\lambda}\right] \qquad \dots (6) \end{split}$$

In particular [5]:

$$M'_X(0) = E(X) = \frac{\alpha + 2}{\lambda(\alpha + 1)} \qquad ...(7)$$

$$M_X''(0) = E(X^2) = \frac{2(\alpha + 3\alpha + 3)}{\lambda^2(\alpha + 1)^2} \qquad \dots (8)$$

Then,

$$\nu(X) = E(X^2) - [E(X)]^2 = \frac{1}{\lambda^2} \left[1 + \frac{1}{(\alpha + 1)^2} \right] \qquad \dots (9)$$

2. Kumaraswamy Weighted Exponential (KWE) Distribution

For any baseline cumulative distribution function G(x) of a random variable X with two additional shape parameters, a, b > 0, Cordeiro and de Castro in (2011) [2] proposed Kumaraswamy generalized (*KG*) distribution with cumulative distribution, probability density, reliability and hazard functions given, respectively, by:

$$F_{KG}(x; a, b) = 1 - \left[1 - \left(G(x)\right)^a\right]^b \qquad \dots (10)$$

$$f_{KG}(x;a,b) = abg(x)[G(x)]^{a-1} \left[1 - (G(x))^a\right]^{b-1} \qquad \dots (11)$$

$$R_{KG}(x; a, b) = \left[1 - (G(x))^{a}\right]^{b} \dots (12)$$

$$abg(x)(G(x))^{a-1} \dots (12)$$

$$h_{KG}(x; a, b) = \frac{abg(x)(G(x))}{1 - (G(x))^{a}} \dots (13)$$

where $g(x) = \frac{\partial G(x)}{\partial x}$. Hence, each new *KG* distribution can be generated from a specified cumulative distribution.

Now, suppose that G(x) represents the WE cumulative distribution as in equation (2), then (10) and (11) yields (*KWE*) cumulative distribution and probability density functions for x > 0, respectively, as:

$$F_{KWE}(x;\alpha,\lambda,a,b) = 1 - \left[1 - \left(1 - \frac{1}{\alpha}e^{-\lambda x}\left(\alpha + 1 - e^{-\lambda \alpha x}\right)\right)^{a}\right]^{b} \dots (14)$$

$$f_{KWE}(x;\alpha,\lambda,a,b) = ab\frac{\alpha+1}{\alpha}\lambda e^{-\lambda x}\left(1 - e^{-\lambda \alpha x}\right)\left[1 - \frac{1}{\alpha}e^{-\lambda x}\left(\alpha + 1 - e^{-\lambda \alpha x}\right)\right]^{a-1}\left[1 - \left(1 - \frac{1}{\alpha}e^{-\lambda x}\left(\alpha + 1 - e^{-\lambda \alpha x}\right)\right)^{a}\right]^{b-1} \dots (15)$$
where $\lambda \geq 0$ is the code corrector and $\alpha \in \lambda \geq 0$ are the chara representation of $\alpha \in \lambda > 0$.

where $\lambda > 0$ is the scale parameter and $\alpha, a, b > 0$ are the shape parameters. As special cases, when a = b = 1, the probability density function for *KWE* distribution will be the probability density function of WE and when b = 1, the probability density function for KWE distribution will be the probability density function of exponentiated WE distribution as in [6].

The reliability and hazard functions of KWE distribution at time (t) can be expressed as:

$$R_{KWE}(t;\alpha,\lambda,a,b) = \left[1 - \left(1 - \frac{1}{\alpha}e^{-\lambda t}\left(\alpha + 1 - e^{-\alpha\lambda t}\right)\right)^{a}\right]^{b} \dots (16)$$
$$h_{KWE}(t;\alpha,\lambda,a,b)$$
$$= \frac{ab\frac{\alpha+1}{\alpha}\lambda e^{-\lambda t}\left(1 - e^{-\alpha\lambda t}\right)\left(1 - \frac{1}{\alpha}e^{-\lambda t}\left(\alpha + 1 - e^{-\alpha\lambda t}\right)\right)^{a-1}}{1 - \left(1 - \frac{1}{\alpha}e^{-\lambda t}\left(\alpha + 1 - e^{-\alpha\lambda t}\right)\right)^{a}} \dots (17)$$

3. Expansions for Cumulative and Density Functions of *KWE* Distribution The cumulative and density functions of *KWE* distribution, equations (14) and (15), can be expansions according to the generalized binomial theorem, $(1+c)^{\nu} = \sum_{i=0}^{\infty} {\nu \choose i} c^{i}$, respectively as:

$$F_{KWE}(x;\alpha,\lambda,a,b) = 1 - \sum_{i=0}^{\infty} (-1)^{i} {b \choose i} \left[1 - \frac{1}{\alpha} e^{-\lambda x} \left(\alpha + 1 - e^{-\alpha \lambda x} \right) \right]^{ai}$$

$$f_{KWE}(x;\alpha,\lambda,a,b) = ab \frac{\alpha+1}{\alpha} \lambda e^{-\lambda x} \left(1 - e^{-\alpha \lambda x} \right) \sum_{i=0}^{\infty} (-1)^{i} {b-1 \choose i} \left(1 - \frac{1}{\alpha} e^{-\lambda x} \left(\alpha + 1 - e^{-\alpha \lambda x} \right) \right)^{a(i+1)-1}$$

Now, suppose that

Now, suppose that:

$$\eta_i = (-1)^i {b \choose i} \quad ; \quad {b \choose i} = \frac{b(b-1)\dots(b-i+1)}{i!}$$

and,

$$w_j = \frac{ab}{j+1} \sum_{i=0}^{\infty} (-1)^{i+j} {b-1 \choose i} {a(i+1)-1 \choose j}$$

Then the cumulative and density functions

Then the cumulative and density functions of *KWE* distribution can be expansions, respectively, as:

$$F_{KWE}(x;\alpha,\lambda,a,b) = 1 - \sum_{\substack{i=0\\\infty}} \eta_i \ K(x;\alpha,\lambda,ai) \qquad \dots (18)$$
$$f_{KWE}(x;\alpha,\lambda,a,b) = \sum_{\substack{j=0\\j=0}}^{\infty} w_j \ S(x;\alpha(j+1),\lambda) \qquad \dots (19)$$

where,

 $K(x; \alpha, \lambda, \theta) = \left[1 - \frac{1}{\alpha} e^{-\lambda x} (\alpha + 1 - e^{-\alpha \lambda x})\right]^{\theta}$ which can denotes the expansion *WE* cumulative distribution with parameters α, λ and $\theta = ai$. $S(x; \alpha(j+1), \lambda)$ denotes the expansion *WE* density function with parameters $\alpha(j+1)$ and λ and cumulative distribution as in (2). Thus, KWE density function can be expressed as an infinite linear combination of WE densities.

4. Moments and Moment Generating Function of KWE Distribution The r^{th} moments about the origin can be expressed by:

$$E(X^{r}) = \sum_{j=0}^{\infty} w_{j} \int_{0}^{\infty} x^{r} S(x; \alpha(j+1), \lambda) dx$$

$$= \sum_{j=0}^{\infty} w_{j} \int_{0}^{\infty} x^{r} \frac{\alpha(j+1)+1}{\alpha(j+1)} \lambda e^{-\lambda x} (1 - e^{-\alpha(j+1)\lambda x}) dx$$

$$\Rightarrow E(X^{r}) = \sum_{j=0}^{\infty} w_{j} \lambda \frac{\alpha(j+1)+1}{\alpha(j+1)} \left[\frac{\Gamma(r+1)}{\lambda^{r+1}} - \frac{\Gamma(r+1)}{[\lambda(\alpha(j+1))+1]^{r+1}} \right] \qquad \dots (20)$$

Now, setting r = 1 and r = 2, we get

$$E(X) = \sum_{j=0}^{\infty} w_j \lambda \frac{\alpha(j+1)+1}{\alpha(j+1)} \left[\frac{\Gamma(2)}{\lambda^2} - \frac{\Gamma(2)}{[\lambda(\alpha(j+1)+1)]^2]} \right]$$

$$E(X^2) = \sum_{j=0}^{\infty} w_j \lambda \frac{\alpha(j+1)+1}{\alpha(j+1)} \left[\frac{\Gamma(3)}{\lambda^3} - \frac{\Gamma(3)}{[\lambda(\alpha(j+1)+1)]^3]} \right]$$

Setting $a = b = 1$, we have,

because
$$u = b = 1$$
, we have,
 $w_j = \begin{cases} 1 & ; j = 0 \\ 0 & ; j \ge 1 \end{cases}$... (21)
Hence, the mean and variance of X are given by:
 $\Pi(U) = 2^{\alpha+1} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} = 2^{\alpha+2}$

$$\mu = E(X) = \lambda \frac{1}{\alpha} \left[\frac{1}{\lambda^2} - \frac{1}{[\lambda(\alpha+1)]^2} \right] = \frac{1}{\lambda(\alpha+1)}$$

$$\nu(X) = E(X^2) - [E(X)]^2 = \frac{2(\alpha^2 + 3\alpha + 3)}{\lambda^2(\alpha+1)^2} - \frac{(\alpha+2)^2}{\lambda^2(\alpha+1)^2} = \frac{1}{\lambda^2} \left[1 + \frac{1}{(\alpha+1)^2} \right]$$
which are precisely the mean and variance of *WE* distribution.

The moment generating function of *KWE* distribution for -1 < t < 1 can be expressed by:

$$M_{X}(t) = E(e^{xt}) = \sum_{j=0}^{\infty} w_{j} \int_{0}^{\infty} e^{xt} S(x; \alpha(j+1), \lambda) dx$$

$$= \sum_{j=0}^{\infty} w_{j} \int_{0}^{\infty} e^{xt} \frac{\alpha(j+1)+1}{\alpha(j+1)} \lambda e^{-\lambda x} (1 - e^{-\alpha(j+1)\lambda x}) dx$$

$$\Rightarrow M_{X}(t) = \sum_{j=0}^{\infty} w_{j} \lambda \frac{\alpha(j+1)+1}{\alpha(j+1)} \left[\frac{1}{\lambda - t} - \frac{1}{\lambda - t + \lambda \alpha(j+1)} \right] \qquad ...(22)$$

By (21),

'Y (2),

$$M_X(t) = \frac{(\alpha+1)\lambda}{\alpha} \left[\frac{1}{\lambda-t} - \frac{1}{\lambda-t+\lambda\alpha} \right]$$

which is precisely the moment generating function of WE distribution. 5. Likelihood Function and Estimation

The maximum likelihood estimations (MLEs) of α , λ , a and b are the solution of the first partial derivatives of the natural-log likelihood function ℓ_{KWE} with respect to that parameters where the likelihood and natural-log likelihood functions for equation (15) are defined, respectively, by:

$$L(\alpha, \lambda, a, b|\underline{x}) = a^{n} b^{n} \frac{(a+1)^{n}}{a^{n}} \lambda^{n} e^{-\lambda \sum_{i=1}^{n} x_{i}} \prod_{i=1}^{n} (1 - e^{-\lambda \alpha x_{i}}) \left[1 - \frac{1}{\alpha} e^{-\lambda x_{i}} (\alpha + 1 - e^{-\lambda \alpha x_{i}}) \right]^{a-1} \left[1 - \left(1 - \frac{1}{\alpha} e^{-\lambda x_{i}} (\alpha + 1 - e^{-\lambda \alpha x_{i}}) \right)^{a} \right]^{b-1} \qquad \dots (23)$$

$$\ell_{KWE} = \ln L(\alpha, \lambda, a, b|\underline{x}) = n \ln a + n \ln b + n \ln(\alpha + 1) - n \ln \alpha + n \ln \lambda - \lambda \sum_{i=1}^{n} x_{i} + \sum_{i=1}^{n} \ln(1 - e^{-\lambda \alpha x_{i}}) + (a - 1) \sum_{i=1}^{n} \ln \left[1 - \frac{1}{\alpha} e^{-\lambda x_{i}} (\alpha + 1 - e^{-\lambda \alpha x_{i}}) \right] + (b - 1) \sum_{i=1}^{n} \ln \left[1 - \frac{1}{\alpha} e^{-\lambda x_{i}} (\alpha + 1 - e^{-\lambda \alpha x_{i}}) \right] + (b - 1) \sum_{i=1}^{n} \ln \left[1 - \frac{1}{\alpha} e^{-\lambda x_{i}} (\alpha + 1 - e^{-\lambda \alpha x_{i}}) \right] + (b - 1) \sum_{i=1}^{n} \ln \left[1 - \frac{1}{\alpha} e^{-\lambda x_{i}} (\alpha + 1 - e^{-\lambda \alpha x_{i}}) \right] + (b - 1) \sum_{i=1}^{n} \ln \left[1 - \frac{1}{\alpha} e^{-\lambda x_{i}} (\alpha + 1 - e^{-\lambda \alpha x_{i}}) \right]^{a} \right] \qquad \dots (24)$$

Now, since there are no closed forms of the solutions, Newton-Raphson iterative technique, can be used to obtain the MLEs as,

$$\begin{bmatrix} \hat{\alpha} \\ \hat{\lambda} \\ \hat{\alpha} \\ \hat{b} \end{bmatrix}^{(h+1)} = \begin{bmatrix} \hat{\alpha} \\ \hat{\lambda} \\ \hat{\alpha} \\ \hat{b} \end{bmatrix}^{(h)} - \mathbf{J}_{(h)}^{-1} \begin{bmatrix} \frac{\partial \ell_{KWE}}{\partial \alpha} \\ \frac{\partial \ell_{KWE}}{\partial \lambda} \\ \frac{\partial \ell_{KWE}}{\partial a} \\ \frac{\partial \ell_{KWE}}{\partial b} \end{bmatrix}^{(h)} ; h = 0, 1, 2, \dots$$

where,

$$\begin{split} J_{(n)} &= \left[\frac{\partial^2 \ell_{KWE}}{\partial a^2} \frac{\partial^2 \ell_{KWE}}{\partial a \partial a} \frac{\partial^2 \ell_{KWE}}{\partial a a} \frac{\partial^2 \ell_{KWE}}{\partial a a} \frac{\partial^2 \ell_{KWE}}{\partial a} \frac{\partial^2$$

$$\begin{split} &\frac{\partial^{2}\ell_{KWE}}{\partial\lambda^{2}} = -\frac{n}{\lambda^{2}} - \sum_{i=1}^{n} \frac{a^{2} \lambda_{i}^{2} e^{-\lambda a x_{i}}}{1 - e^{-\lambda a x_{i}}} \left(1 + \frac{e^{-\lambda a x_{i}}}{1 - e^{-\lambda a x_{i}}}\right) - \sum_{i=1}^{n} \left(x_{i} \ e^{-\lambda x_{i}} \left(1 + \frac{1}{a} - e^{-\lambda a x_{i}} \left(1 + \frac{1}{a} - \frac{1}{a} + e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{2}\right)^{2} + \\ &\frac{a(b-1)\left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a-1}}{1 - \left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a-1}} \left(\frac{a\left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a-1}}{1 - \left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}} + \\ &\frac{a(b-1)\left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a-1}}{1 - \left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}} + \\ &\frac{a(b-1)\left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}}{1 - \left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}} + \\ &\frac{a(b-1)\left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}}{1 - \left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}} + \\ &\frac{a(b-1)\left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}}{1 - \left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}} \end{bmatrix} \\ &\frac{\partial^{2}\ell_{KWE}}{\partial a^{2}} = -\frac{n}{a^{2}} - \sum_{i=1}^{n} \frac{(b-1)\left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}}{1 - \left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}} \end{bmatrix} \\ &\frac{1 + \frac{\left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}}{1 - \left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}} \end{bmatrix}} \\ &\frac{1 + \frac{1}{a} \left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}}{1 - \left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}} \end{bmatrix} \\ &\frac{1 + \frac{1}{a} \left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}}{1 - \left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}} \end{bmatrix} \\ &\frac{1 + \frac{1}{a} \left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}}{1 - \left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}} \end{bmatrix} \\ &\frac{1 + \frac{1}{a} \left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{a}}{1 - \left(1 - \frac{1}{a} e^{-\lambda x_{i}} \left(a + 1 - e^{-\lambda a x_{i}}\right)\right)^{$$

When the convergence occurs between iteration (h + 1) and (h), i.e. the absolute difference between two successive iterations is less than pre-specified error tolerance, $\varepsilon > 0$, then the current $\hat{\alpha}^{(h+1)}$, $\hat{\lambda}^{(h+1)}$, $\hat{\alpha}^{(h+1)}$ and $\hat{b}^{(h+1)}$ represent the MLEs of α , λ , α and b via NR algorithm which we referred to as, $\hat{\alpha}_{ML}$, $\hat{\lambda}_{ML}$, $\hat{\alpha}_{ML}$ and \hat{b}_{ML} .

Then, according to an invariant property of the ML estimator, the estimate of reliability and hazard functions at mission time (t) can be obtained, respectively, by replacing α , λ , α and b in equations (16) and (17) by their ML estimates.

6. Concluding Remarks

The two parameter weighted exponential (WE) distribution introduced by Gupta and Kundu has been extension based on the family of Kumaraswamy generalized distribution introduced by Cordeiro and de Castro. Some of the mathematical properties along with the maximum likelihood estimations of the model parameters of new distribution named Kumaraswamy weighted

exponential (KWE) have been discussed. The *KWE* density function can be expressed as an infinite linear combination of *WE* densities and the *WE* distribution is a special case of *KWE* distribution when a = b = 1 and the exponentiated *WE* distribution is a special case of *KWE* distribution when b = 1.

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