

On Weighted Nwikpe Distribution: Properties and Applications

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Abstract: In this article, two-parameter continuous distribution has been introduced. The proposed distribution is obtained by using a weight technique and is referred to as weighted Nwikpe distribution. This distribution is a generalization of baseline distribution that is Nwikpe distribution. Some structural properties of the distribution has been derived and studied. These are density function, distribution function, and reliability function, hazard rate function, moments, moment generating function, entropies, order statistics, Bonferroni and Lorenz curves. The method of maximum likelihood technique has been established to investigate the parameters of the model. The behaviour of the parameters of the distribution has been examined by a simulation study. Real data set is used to determine whether the weighted Nwikpe distribution is better than other well-known distributions in modeling data or not.

Keywords: Weighted distribution, Nwikpe distribution, Maximum likelihood estimator, Entropies, Order statistics, Bonferroni and Lorenz curves.

1 Introduction

The theory of weighted distributions provides a collective access to the problems of model specification and data interpretation. Weighted distributions provide a technique for fitting the models to the weight functions when the samples are taken from original distribution and developed distributions. The idea of weighted distributions was first given by Fisher [1] to model the ascertainment bias. Rao [2] introduced and formulated the method of ascertainment for modeling statistical data when the standard distributions were not appropriate to record these observations with equal probabilities. Weighted distributions can be applied in various research areas related to biomedicine, reliability, and ecology and branching processes. Researchers have applied the concept of weighted distributions for different purposes. Rao [3] extended the concept of ascertainment upon the estimation of frequencies and introduced the concept of weighted

distributions. Lappi and Bailey [4] used weighted distributions to analyse the HPS diameter increment data.

Castillo and Casany [5] developed weighted Poisson distribution by considering the Poisson distribution with parameter λ and weight

$$w(k) = (k + \alpha)^r.$$

The pdf of the weighted Poisson distribution is defined as

$$P_w(k, \lambda, r, \alpha) = \frac{(k + \alpha)^r \lambda^r e^{-\lambda}}{(E(x + \alpha))^r k!},$$

$$k = 0, 1, 2, \dots, \alpha \geq 0, \lambda > 0, r \in \mathfrak{R}$$

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Gupta and Kundu [6] have introduced a new class of weighted exponential distributions by applying the method proposed by Azzalini [7] to the exponential distribution. The two-parameter weighted exponential distribution introduced by Gupta and Kundu has pdf defined as

$$f_w(x, \alpha, \beta) = \frac{\alpha + 1}{\alpha!} \beta (1 - \exp(-(\alpha\beta x))) \exp(-(\alpha\beta x))$$

Kersey [8] introduced the weighted inverse Weibull distribution. The proposed distribution namely weighted inverse distribution was obtained by using $w(x) = x$ as a weigh function and pdf of the inverse Weibull distribution. The pdf of the newly generated distribution is defined as

$$f_w(x, \alpha, \beta) = \frac{\beta \alpha^{1-\beta}}{\Gamma\left(1 - \frac{1}{\beta}\right)} x^{-\beta} \exp(-(x\alpha))^{-\beta}, x > 0$$

Ye et al. [9] introduced a weighted generalized beta distribution of the second kind. The proposed distribution namely weighted generalized beta distribution was obtained by using $w(x) = x^k$ as a weigh function and pdf of the inverse Weibull distribution. The pdf of the newly generated distribution defined as

$$f_w(x, \alpha, \beta, p, q, k) = \frac{\alpha x^{(\alpha p - 1)}}{\beta^{\alpha p + k} B(p + k / \alpha, q - k / \alpha) \left(1 + \left(\frac{x}{\beta}\right)^\alpha\right)^{p+q}}, x > 0$$

Aleem et al. [10] developed the class of modified Weibull distribution and its properties. The distribution was generated by considering the cumulative distribution function as weigh and pdf of the Weibull distribution in the weighted model. The pdf of the modified weighted Weibull distribution is defined as

$$f_w(x) = \beta_\gamma (c\theta^r + 1) x^{\gamma-1} e^{-\beta(c\theta^r + 1)x^\gamma}, x > 0$$

Bashir and Rahul [11] introduced the weighted Lindley distribution. The pdf of the weighted Lindley distribution is given by

$$f_w(x, \theta, \Phi) = \frac{(\theta - \varphi)^2}{\theta - \varphi + 1} (1 + x) \exp(-(\theta - \varphi))x, x > 0$$

Asgharzadeh et al. [12] introduced a new weighted Lindley distribution with application to survival analysis. The pdf of a new weighted Lindley distribution is given by

$$f_w(x, \lambda, \alpha) = \frac{\lambda^2 (1 + \alpha)^2}{\alpha \lambda (1 + \alpha) + \alpha (2 + \alpha)} (1 + x) (1 - \exp(-\alpha \lambda x) \exp(-\lambda x)), x > 0$$

Para and Jan [13] introduced the Weighted Pareto type-II distribution as a new model for handling medical science data and studied its statistical properties and applications. In addition, further work on the weighted model has been done by the various researchers like Shakhatreh [14] proposed two-parameter weighted exponential distribution, Sherina et al. [15] introduced Weighted Weibull distribution, Nasiru [16] obtained weighted Weibull distribution, Domma et al. [17] introduced a new generalized weighted Weibull distribution, Ghitany [18] et al. obtained two-parameter weighted Lindley distribution, Dey et al [19] obtained weighted Weibull distribution, Badmus and Bamiduro et al. [20] proposed Lehmann Type II weighted Weibull, Alqallaf et al. [21] introduced Weighted exponential, Alsmarian [22] developed weighted Suja, Mizaal [23] discuss the mathematical study of weighted two-parameters exponential distribution. Saghir et al. [24] discussed the brief perspective and characterizations of the weighted distributions. Mahdavi [25] introduced the two weighted distributions generated by the exponential distribution. The author(s) has obtained the two distributions by incorporating the exponential distribution in Azzalini's model to get the weighted gamma-exponential model and the weighted generalized exponential-exponential model. Shanker and Shukla [26] obtained weighted Akash distribution. Dey and Perk [27] proposed weighted exponential distribution and its properties with the application.

A class of bivariate modified weighted distributions are obtained by Hiba Zeyada [28] based on the Marshall and Olkin concept. The statistical properties of the model were obtained including joint pdf, joint cdf, joint survival function, product moments, marginal conditional density are obtained in compact forms. Shoaee [29] introduced a new class of bivariate survival distributions based on the model of dependent lives and its generalization. A new class of distributions using a functional mixture was proposed by Bouali, Chesneau et al. [30].

The weighted exponential family of distributions are introduced by Zubair et al. [31]. The general expressions for the mathematical properties are studied. The length biased weighted exponentiated inverted Weibull distribution was obtained by Tazeem, Ishfaq et al. [32]. The properties of the distribution have been discussed along with parameter estimation.

Nwike, et al. [33] proposed a one-parameter distribution called as Nwike distribution. The proposed model aims to model the strength of the aircraft window data set. The distribution is a mixture model consists of the gamma distribution, and exponential distribution with θ parameter. They studied the properties of the proposed model. Parameters of the proposed distribution are presented and examined with the method of maximum likelihood estimation. The probability density function and cumulative distribution function of the Nwike distribution is given by

$$f(x, \theta) = \left(\frac{\theta^3}{12(\theta^2 + 10)} \right) (\theta^3 x^5 + 12) e^{-\theta x} \quad x > 0, \theta > 0 \quad (1)$$

$$F(x, \theta) = 1 - \left(1 + \frac{\theta^5 x^5 + 5\theta^3 x^3 (4 + x) + 60\theta x(\theta x + 2)}{12(\theta^2 + 10)} \right) e^{-\theta x} \quad (2)$$

2 Weighted Nwike Distributions: Definition and Properties

2.1 Weighted Nwike Distribution

Let X be a random variable with probability density function $f(x)$. Let $w(x)$ be the non-negative weight function, then the probability density function of the weighted random variable X_w is given by

$$f_w(x) = \frac{w(x)f(x)}{E(w(x))}; \quad x > 0$$

Where, $w(x)$ is a non-negative weight function and

$$E(w(x)) = \int w(x)f(x) < \infty, \quad 0 < E(w(x)) < \infty$$

For different weighted models, we can have different choices of the weight function $w(x)$. In this paper, we

study the weighted version of Nwike distribution, taking $w(x) = x^c$, in order to get the weighted Nwike distribution, the probability density function (pdf) of weighted version is given by

$$f_w(x) = \frac{x^c f(x)}{E(X^c)} \quad (3)$$

Where,

$$E(X^c) = \int_0^\infty x^c f(x) dx$$

$$E(X^c) = \frac{(c+5)! + 12\theta^2 c!}{(12(\theta^2 + 10))\theta^c} \quad (4)$$

Substituting the values of equation (1) and (4) in equation (3), we get the probability density function (pdf) of weighted Nwike distribution.

$$f_w(x, c, \theta) = \left(\frac{\theta^{c+3} x^c (\theta^3 x^5 + 12) e^{-\theta x}}{12\theta^2 c! + (c+5)!} \right), \quad x > 0, \theta > 0, c > 0 \quad (5)$$

The cumulative distribution function (cdf) of the weighted Nwike distribution is obtained as

$$F_w(x, c, \theta) = \int_0^x f_w(x, c, \theta) dx$$

$$F_w(x, c, \theta) = \int_0^x \left(\frac{\theta^{c+3} x^c (\theta^3 x^5 + 12) e^{-\theta x}}{12\theta^2 c! + (c+5)!} \right) dx \quad (6)$$

After simplification, using an incomplete gamma function

$$\gamma((c+1), \theta x) = \int_0^{\theta x} t^{(c+1)-1} e^{-t} dt$$

to equation (6). We get the cumulative distribution function of weighted Nwike distribution as

$$F_w(x, c, \theta) = \frac{\gamma(c+6; \theta x) + 12\theta^2 \gamma(c+1; \theta x)}{12\theta^2 c! + (c+5)!} \quad (7)$$

Where, $\gamma(s; x) = \int_0^x t^{s-1} e^{-t} dt$ is an incomplete gamma

function.

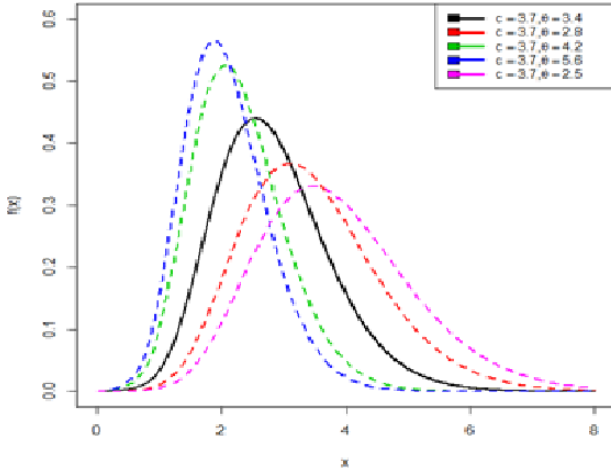


Fig.1: pdf plot of weighted Nwike distribution

2.2 Reliability Function and Hazard Rate

The reliability function of the weighted Nwike distribution can be obtained as

$$R_w(x, c, \theta) = 1 - F_w(x, c, \theta)$$

$$R_w(x, c, \theta) = 1 - \frac{\gamma(c+6; \theta x) + 12\theta^2 \gamma(c+1; \theta x)}{12\theta^2 c! + (c+5)!}$$

$$; x > 0, \theta > 0, c > 0$$

The hazard function or failure rate of weighted Nwike distribution is given by

$$h_w(x, c, \theta) = \frac{f_w(x, c, \theta)}{R_w(x, c, \theta)}$$

$$h_w(x, \theta, c) = \frac{\theta^{c+3} x^c (\theta^3 x^5 + 12) e^{-\theta x}}{12\theta^2 c! + (c+5)! - \gamma(c+6; \theta x) + 12\theta^2 \gamma(c+1; \theta x)}$$

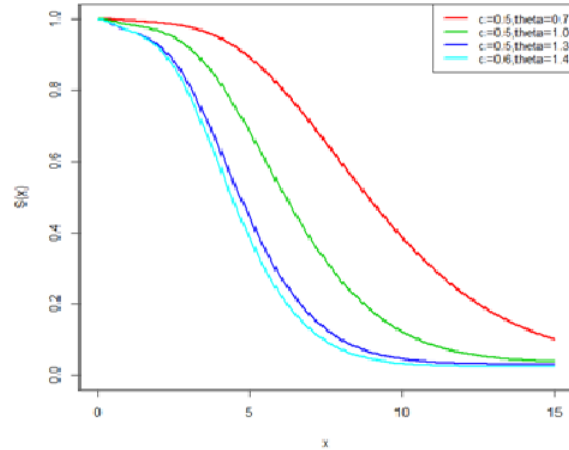


Fig.3: Reliability plot of weighted Nwike distribution

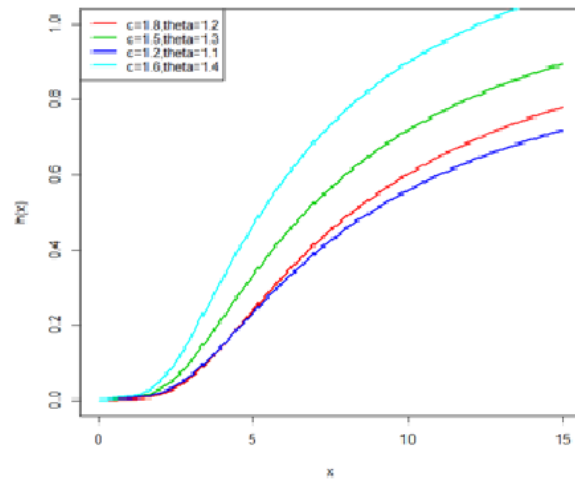


Fig.4: Hazard plot of weighted Nwike distribution

The figure 4 depicts that the hazard rate of the weighted distribution is increasing for the parametric value $\alpha > 1$ and $\theta > 1$.

2.3 Moments

Let X_w denote the random variable following weighted Nwike distribution then r^{th} order moment $E(X^r)$ is obtained as

$$E(X_w^r) = \mu_r' = \int_0^\infty x^r f_w(x, c, \theta) dx$$

$$E(X_w^r) = \left(\frac{\theta^3 (c+r+5)! + 12\theta^5 (c+r)!}{\theta^r (12\theta^2 c! + (c+5)!)} \right) \quad (8)$$

Letting $r = 1$ in equation (8), the mean of weighted Nwike distribution is obtained as

$$\mu_1' = \left(\frac{\theta^3 (c+6)! + 12\theta^5 (c+1)!}{\theta (12\theta^2 c! + (c+5)!)} \right)$$

$$\mu_2' = \left(\frac{\theta^3 (c+7)! + 12\theta^5 (c+2)!}{\theta^2 (12\theta^2 c! + (c+5)!)} \right)$$

$$\mu_3' = \left(\frac{\theta^3 (c+8)! + 12\theta^5 (c+3)!}{\theta^3 (12\theta^2 c! + (c+5)!)} \right)$$

$$\mu_4' = \left(\frac{\theta^3 (c+9)! + 12\theta^5 (c+4)!}{\theta^4 (12\theta^2 c! + (c+5)!)} \right)$$

$$\text{Variance} = \mu_2' - (\mu_1')^2$$

$$\mu_2 = \left(\frac{\theta^3 (c+7)! + 12\theta^5 (c+2)!}{\theta^2 (12\theta^2 c! + (c+5)!)} \right) - \left(\frac{\theta^3 (c+6)! + 12\theta^5 (c+1)!}{\theta (12\theta^2 c! + (c+5)!)} \right)^2$$

$$S.D(\sigma) = \sqrt{\left(\frac{(\theta^3 (c+7)! + 12\theta^5 (c+2)!)(12\theta^2 c! + (c+5)!)}{(\theta^3 (c+6)! + 12\theta^5 (c+1)!)^2} - \frac{(\theta^3 (c+6)! + 12\theta^5 (c+1)!)^2}{\theta^2 (12\theta^2 c! + (c+5)!)} \right)}$$

2.4 Moment Generating Function and Characteristic Function

Let X_w follows Weighted Nwike distribution, then the moment generating function (mgf) of X is obtained as

$$M_{X_w}(t) = E(e^{tx}) = \int_0^\infty e^{tx} f_w(x, c, \theta) dx$$

$$M_{X_w}(t) = \int_0^\infty \sum_{j=0}^\infty \frac{(tx)^j}{j!} f_w(x, c, \theta) dx$$

$$M_{X_w}(t) = \sum_{j=0}^\infty \frac{(t)^j}{j!} \int_0^\infty x^j f_w(x, c, \theta) dx$$

$$M_{X_w}(t) = \sum_{j=0}^\infty \frac{t^j}{j!} \left(\frac{\theta^3 (c+r+5)! + 12\theta^5 (c+r)!}{\theta^r (12\theta^2 c! + (c+5)!)} \right)$$

Similarly, we can get the characteristic function of weighted Nwike distribution as

$$\phi_{X_w}(t) = M_{X_w}(it)$$

$$\phi_{X_w}(t) = \sum_{j=0}^\infty \frac{(it)^j}{j!} \mu_j'$$

$$\phi_{X_w}(t) = \sum_{j=0}^\infty \frac{(it)^j}{j!} \left(\frac{\theta^3 (c+r+5)! + 12\theta^5 (c+r)!}{\theta^r (12\theta^2 c! + (c+5)!)} \right)$$

3 Order Statistics

Let $x_{(1)}, x_{(2)}, x_{(3)}, \dots, x_{(n)}$ be the order statistics of a random sample $x_1, x_2, x_3, \dots, x_n$ drawn from the continuous population with probability density function $f_X(x)$ and cumulative distribution function with $F_X(x)$ then the pdf of r^{th} order statistics $X_{(r)}$ is given by

$$f_{X_{(r)}}(x) = \frac{n!}{(r-1)!(n-r)!} f_X(x) [F_X(x)]^{r-1} [1 - F_X(x)]^{n-r} \quad (9)$$

Inserting equations (5) and (7) in equation (9), the probability density function of r^{th} order statistic $X_{(r)}$ of Weighted Nwike distribution is given by

$$f_{Xw(r)}(x) = \frac{n!}{(r-1)!(n-1)!} \left(\frac{\theta^{c+3} x^c (\theta^3 x^5 + 12) e^{-\theta x}}{12 \theta^2 c! + (c+5)!} \right)$$

$$\left(\frac{\gamma(c+6; \theta x) + 12 \theta^2 \gamma(c+1; \theta x)}{12 \theta^2 c! + (c+5)!} \right)$$

$$\left(1 - \frac{\gamma(c+6; \theta x) + 12 \theta^2 \gamma(c+1; \theta x)}{12 \theta^2 c! + (c+5)!} \right)^{n-r}$$

Therefore, the probability density function of highest order statistics $X_{w(n)}$ can be obtained as

$$f_{Xw(n)}(x) = n \left(\frac{\theta^{c+3} x^c (\theta^3 x^5 + 12) e^{-\theta x}}{12 \theta^2 c! + (c+5)!} \right)$$

$$\left(\frac{\gamma(c+6; \theta x) + 12 \theta^2 \gamma(c+1; \theta x)}{12 \theta^2 c! + (c+5)!} \right)^{n-1}$$

and pdf of first order statistic $X_{w(1)}$ can be obtained as

$$f_{Xw(1)}(x) = \left(\frac{\theta^{c+3} x^c (\theta^3 x^5 + 12) e^{-\theta x}}{12 \theta^2 c! + (c+5)!} \right)$$

$$\left(1 - \frac{\gamma(c+6; \theta x) + 12 \theta^2 \gamma(c+1; \theta x)}{12 \theta^2 c! + (c+5)!} \right)^{n-1}$$

4 Entropies

4.1 Renyi Entropy

Renyi entropy [34] of order δ can be obtained as

$$R(\delta) = \frac{1}{1-\delta} \log \left(\int f_w^\delta(x, c, \theta) dx \right)$$

$$R(\delta) = \frac{1}{1-\delta} \log \int_0^\infty \left(\frac{\theta^{c+3} x^c (\theta^3 x^5 + 12) e^{-\theta x}}{12 \theta^2 c! + (c+5)!} \right)^\delta dx \quad (10)$$

Using Binomial expansion to equation (10), we get the renyi entropy of weighted Nwike distribution

$$R(\delta) = \frac{1}{1-\delta} \log \left(\frac{\theta^{c+3}}{12 \theta^2 c! + (c+5)!} \right)^\delta \sum_{i=0}^{\infty} \binom{\delta}{i} \frac{\Gamma(\delta c + 5i + 1)}{(\theta \delta)^{(\delta c + 5i + 1)}}$$

4.2 Tsallis Entropy

Tsallis entropy [35] of order λ is given by

$$S_\lambda = \frac{1}{1-\lambda} \left(1 - \int f_w^\lambda(x, c, \theta) dx \right)$$

$$S_\lambda = \frac{1}{1-\lambda} \left(1 - \int_0^\infty \left(\frac{\theta^{c+3} x^c (\theta^3 x^5 + 12) e^{-\theta x}}{12 \theta^2 c! + (c+5)!} \right)^\lambda dx \right) \quad (11)$$

Using Binomial expansion to equation (11), we get Tsallis entropy of weighted Nwike distribution.

$$S(\lambda) = \frac{1}{1-\lambda} \left(1 - \left(\frac{\theta^{c+3}}{12 \theta^2 c! + (c+5)!} \right)^\lambda \sum_{i=0}^{\infty} \binom{\lambda}{i} \frac{\Gamma(\lambda c + 5i + 1)}{(\theta \lambda)^{(\lambda c + 5i + 1)}} \right)$$

5 Bonferroni and Lorenz Curves

The Bonferroni [36] and Lorenz [37] curves can be obtained as

$$B(p) = \frac{1}{p\mu_1^q} \int_0^q x f(x, c, \theta) dx$$

and

$$L(p) = \frac{1}{\mu_1^q} \int_0^q x f(x, c, \theta) dx$$

Where,

$$\mu_1' = \left(\frac{\theta^3 (c+6)! + 12 \theta^5 (c+1)!}{\theta (12 \theta^2 c! + (c+5)!)} \right)$$

$$q = F^{-1}(p)$$

On simplification, we get,

$$B(p) = \left(\frac{\gamma((c+7), \theta q) + 12\theta^2 \gamma((c+2), \theta q)}{p(12\theta^5(c+1)! + \theta^2(c+6)!)} \right)$$

Where, $q = F^{-1}(p)$

$$L(p) = p B(p)$$

$$L(p) = \left(\frac{\gamma((c+7), \theta q) + 12\theta^2 \gamma((c+2), \theta q)}{(12\theta^5(c+1)! + \theta^2(c+6)!)} \right)$$

6 Estimation of Parameters

6.1 Maximum Likelihood Estimation

Consider $x_1, x_2, x_3, \dots, x_n$ be a random sample of size n from weighted Nwike distribution with parameters θ, α , the Likelihood function is defined as

$$L(x; c, \theta) = \left(\left(\frac{\theta^{c+3}}{12\theta^2 c! + (c+5)!} \right) \right)^n \prod_{i=1}^n x_i^c (\theta^3 x_i^5 + 12) e^{-\theta x_i} \quad (12)$$

The log-likelihood function is

$$\begin{aligned} \log L &= n(c+3) \log \theta - n \log (12\theta^2 c! + (c+5)!) \\ &\quad - \theta \sum_{i=1}^n x_i + \sum_{i=1}^n \log x_i^c (\theta^3 x_i^5 + 12) \end{aligned} \quad (13)$$

The maximum likelihood estimates of θ and c can be obtained by differentiating equation (13) with respect to θ and c and must satisfy the normal equations.

$$\frac{\partial \log L}{\partial \theta} = \frac{n(c+3)}{\theta} - \frac{24n\theta c!}{12\theta^2 c! + (c+5)!} + \quad (1)$$

$$\sum_{i=1}^n \left(\frac{3\theta^2 x_i^5}{\theta^3 x_i^5 + 12} \right) - \sum_{i=1}^n x_i = 0$$

$$\frac{\partial \log L}{\partial c} = n \log \theta - n(\psi(c) + \psi(c+5)) + \sum_{i=1}^n \log x_i = 0 \quad (15)$$

The solution of equations (14) and (15) gives the maximum likelihood estimates of the parameters for the weighted Nwike distribution. However the equations cannot be solved analytically, thus we solved numerically using R programming [38] with some data set.

6.2 Simulation Study of ML Estimators

In this subsection, we study the performance of ML estimators for different sample sizes. However, we set the sample size $N = 25, 50, 75, 100, 250$, and 400 . We have employed the inverse CDF technique for data simulation for weighted Nwike distribution using R software. Bias, Variance and MSE for the weighted Nwike distribution is observed. We found that, as the sample size increase, the values of MSE are getting smaller for the parameter estimate.

7 Likelihood Ratio Test

Let $x_1, x_2, x_3, \dots, x_n$ be a random sample from the weighted Nwike Distribution.

To test the hypothesis we have,

$$H_0 : f(x) = f(x, \theta) \text{ against } H_1 : f(x) = f_w(x, c, \theta)$$

In order to test whether the random sample of size n comes from the Nwike distribution or weighted Nwike distribution, the following test statistic is used

$$\Delta = \frac{L_1}{L_0} = \prod_{i=1}^n \frac{f_w(x_i, c, \theta)}{f(x_i, \theta)}$$

8.1 Data Set 1

$$\Delta = \prod_{i=1}^n \left(\frac{\theta^c (12\theta^2 + 10)}{12\theta^2 c! + (c+5)!} \right) x_i^c$$

We reject the null hypothesis if

$$\Delta = \left(\frac{\theta^c (12\theta^2 + 10)}{12\theta^2 c! + (c+5)!} \right)^n \prod_{i=1}^n x_i^c > k$$

$$\Delta = \prod_{i=1}^n x_i^c > k \left(\frac{12\theta^2 c! + (c+5)!}{\theta^c (12\theta^2 + 10)} \right)^n$$

$$\Delta^* = \prod_{i=1}^n x_i^c > k^*,$$

$$\text{Where, } k^* = k \left(\frac{12\theta^2 c! + (c+5)!}{\theta^c (12\theta^2 + 10)} \right)^n$$

For large sample size n , $2 \log \Delta$ is distributed as chi-square variate with one degree freedom. Thus we reject the null hypothesis, when the probability value is given by

$P(\Delta^* > \alpha^*)$, where $\alpha^* = \prod_{i=1}^n x_i^c$ is less than level of significance and $\prod_{i=1}^n x_i^c$ is the observed value of the statistic Δ^* .

8 Illustration

In this section, we determine the flexibility of the generalization of the Nwikpe distribution by fitting the model to two data sets over the Nwikpe, Exponential, Shanker, Akash, Amarendra, and Lindley distribution. The results and the performance of the weighted Nwikpe distribution are determined by using Akaike Information Criterion, Bayesian Information Criterion, Akaike Information Criterion Corrected and $-2\ln L$. The distribution with the lowest AIC, BIC, AICC and $-2\ln L$ is considered as the most flexible distribution for a given data set.

The following data represent the tensile strength, measured in GPA, of 69 carbon fibers tested under tension at gauge lengths of 20mm, Bader and Priest [39].

1.312, 1.314, 1.479, 1.552, 1.700, 1.803, 1.861, 1.865, 1.944, 1.958, 1.966, 1.997, 2.006, 2.021, 2.027, 2.055, 2.063, 2.098, 2.14, 2.179, 2.224, 2.240, 2.253, 2.270, 2.272, 2.274, 2.301, 2.301, 2.359, 2.382, 2.382, 2.426, 2.434, 2.435, 2.478, 2.490, 2.511, 2.514, 2.535, 2.554, 2.566, 2.57, 2.586, 2.629, 2.633, 2.642, 2.648, 2.684, 2.697, 2.726, 2.770, 2.773, 2.800, 2.809, 2.818, 2.821, 2.848, 2.88, 2.954, 3.012, 3.067, 3.084, 3.090, 3.096, 3.128, 3.233, 3.433, 3.585, 3.585.

8.2 Data Set 2

The data set represents the strength of glass of the aircraft windows given by Fuller et al. [40].

18.83, 20.80, 21.657, 23.03, 23.23, 24.05, 24.321, 25.50, 25.52, 25.80, 26.69, 26.77, 26.78, 27.05, 27.67, 29.90, 31.11, 33.20, 33.73, 33.76, 33.89, 34.76, 35.75, 35.91, 36.98, 37.08, 37.09, 39.58, 44.045, 45.29, 45.381.

It can be easily seen from table 2 and 3 that the weighted Nwikpe distribution gives better fit than all the considered distributions and hence it can be considered as an important two-parameter lifetime distribution for modeling lifetime data.

Table 1: Bias, Variance and ML Estimators for Different Sample Sizes.

| Sample Size | $\theta = 1.2$ | | | $c = 0.3$ | | |
|-------------|----------------|----------|----------|-----------|----------|----------|
| | Bias | Variance | MSE | Bias | Variance | MSE |
| 25 | 20.43670 | 257.7957 | 675.4567 | 18.79052 | 364.8161 | 717.8996 |
| 50 | 17.16548 | 94.49843 | 389.152 | 14.8353 | 129.2258 | 349.3110 |
| 75 | 10.85597 | 31.54542 | 149.3978 | 7.58014 | 43.3548 | 100.8133 |
| 100 | 16.14025 | 58.48277 | 318.9883 | 13.6211 | 79.0009 | 264.5340 |
| 250 | 11.62651 | 5.990382 | 141.1662 | 8.30251 | 7.27722 | 76.20384 |
| 400 | 10.13205 | 2.411737 | 105.0701 | 6.61653 | 3.384882 | 47.16331 |
| | $\theta = 0.3$ | | | $c = 0.5$ | | |
| 25 | 78.13097 | 11.13845 | 8.184896 | 33.49206 | 260.9222 | 1382.640 |
| 50 | 8.17184 | 7.249752 | 74.02872 | 33.02830 | 178.8993 | 1269.768 |
| 75 | 8.019381 | 2.337856 | 66.64833 | 32.42125 | 56.44626 | 1107.584 |
| 100 | 7.305664 | 5.308559 | 58.68128 | 28.84660 | 60.7805 | 954.9066 |
| 250 | 6.978366 | 2.820293 | 51.51788 | 27.53613 | 65.42063 | 823.6592 |
| 400 | 6.843113 | 1.098669 | 47.92686 | 26.56396 | 25.49759 | 731.1413 |
| | $\theta = 2$ | | | $c = 1$ | | |
| 25 | 18.71878 | 157.2793 | 507.6720 | 36.78858 | 329.0645 | 1682.464 |
| 50 | 54.52185 | 599.4488 | 3572.081 | 31.68343 | 306.4980 | 1310.338 |
| 75 | 37.15250 | 256.7097 | 1637.018 | 19.57602 | 123.1012 | 506.3216 |
| 100 | 38.81826 | 87.44561 | 1594.303 | 20.53530 | 41.38000 | 463.0787 |
| 250 | 40.65167 | 84.95280 | 1737.511 | 21.95189 | 38.94486 | 520.8301 |
| 400 | 37.46984 | 32.12828 | 1436.117 | 19.74412 | 16.22667 | 406.0568 |

Table 2: MLEs AIC, BIC AICc and -2 log L of the fitted distribution for the given data set 1

| Distribution | ML Estimates | -2 log L | AIC | BIC | AICc |
|-----------------|--|----------|----------|----------|----------|
| Weighted Nwikpe | $\hat{\theta} = 9.5402$ $\hat{c} = 17.3879$ | 100.0697 | 104.0697 | 108.5379 | 104.4389 |
| Nwikpe | $\hat{\theta} = 2.1555$ | 198.8053 | 200.8053 | 200.6442 | 200.8621 |
| Exponential | $\hat{\theta} = 0.4079$ | 261.7432 | 263.7411 | 265.9655 | 263.8011 |
| Shanker | $\hat{\theta} = 0.6582$ | 233.0054 | 235.0054 | 237.2376 | 235.0135 |
| Amarendra | $\hat{\theta} = 1.2445$ | 207.947 | 209.947 | 209.7750 | 210.0121 |
| Sujatha | $\hat{\theta} = 0.9363$ | 221.6088 | 223.6088 | 225.8355 | 223.6688 |
| Akash | $\hat{\theta} = 0.9647$ | 224.2798 | 226.2797 | 228.5132 | 226.3424 |
| Lindley | $\hat{\theta} = 0.6590$ | 238.3667 | 240.3659 | 242.6134 | 240.4400 |

Table 3: MLEs AIC, BIC AICc and -2 log L of the fitted distribution for the given data set 2

| Distribution | ML Estimates | -2 log L | AIC | BIC | AICc |
|-----------------|--|----------|----------|----------|----------|
| Weighted Nwikpe | $\hat{\theta} = 0.6145$ $\hat{c} = 12.9342$ | 208.2313 | 212.2313 | 215.0992 | 213.1203 |
| Nwikpe | $\hat{\theta} = 0.1945$ | 223.3051 | 225.3051 | 224.7964 | 225.4402 |
| Exponential | $\hat{\theta} = 0.1640$ | 454.9130 | 456.9100 | 459.3044 | 456.959 |
| Shanker | $\hat{\theta} = 0.3084$ | 408.9220 | 410.9216 | 410.8300 | 410.9722 |
| Amarendra | $\hat{\theta} = 0.6015$ | 373.9708 | 375.9707 | 375.8792 | 376.0212 |
| Sujatha | $\hat{\theta} = 0.4403$ | 392.3864 | 394.3863 | 394.2949 | 394.4369 |
| Akash | $\hat{\theta} = 0.4605$ | 388.6078 | 390.6073 | 390.5162 | 390.5668 |
| Lindley | $\hat{\theta} = 0.2910$ | 418.5780 | 420.5781 | 420.4782 | 420.6286 |

9 Conclusion

In this article, we have introduced a new generalization of the two- parameter continuous distribution termed as Weighted Nwike distribution. The new distribution is generated by using the weighting technique. The method of maximum likelihood estimation is used for estimating the parameters of the model. Some statistical properties along with reliability measures has been discussed. The simulation study is carried out to know the performance of the parameters of the model. The result observed from table 1 depicts that the bias, variance and MSE decrease when the sample size tends to increase. The applicability of proposed model is demonstrated by using a goodness of fit criterion such as AIC, BIC, AICC and $-2\ln L$. It has been noted from table 2 and 3 that the weighted Nwike distribution provides a better fit than competing models Nwike distribution, Exponential, Shanker, Amarendra, Sujatha, Akash, Lindley distributions

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Competing interests

The authors declare that they have no competing interests.

Reference

- [1] Fisher, R. A. (1934). The effects of methods of ascertainment upon the estimation of frequencies. *Annals of Eugenics*, **6**(3), 13-25 (1934).
- [2] Rao, C. R. (1965). On discrete distributions arising out of methods of ascertainment, in *Classical and Contagious Discrete Distribution*. Patil, G. P. ed. Pergamon Press and Statistical Publishing Society, Calcutta, 320-332 (1965).
- [3] Patil, G. P. & Rao, G. R. (1978). Weighted distributions and size biased sampling with applications to wildlife populations and human families. *Biometrika*, **(34)**, 179-189 (1978).
- [4] Lappi, J. & Bailey, R. L. (1987). Estimation of diameter increment function or other tree relations using angle-count samples. *Forest Science*, **(33)**, 725-739 (1987).
- [5] Castillo, J. D. & Casany, M. P. (1998). Weighted Poison Distribution for over dispersion and under dispersion Solutions. *Annals of Institute of Statistical Mathematics*, **(50)**, 567-585 (1998).
- [6] Gupta, R. D. & Kundu, D. A. (2009). New class of weighted exponential distribution. *Statistics*, **(43)**, 621-634 (2009).
- [7] Azzalini, A. (1985). A class of distribution which includes the normal ones. *Scandinavian Journal of Statistics*, **(12)**, 171-178 (1985).
- [8] Kersey, J. X. (2010). Weighted Inverse Weibull and Beta-Inverse Weibull distribution. M.Sc. Thesis, university of Georgia Southern, (2010).
- [9] Ye, Y., Oluyede, B. O. & Pararai, M. (2012). Weighted Generalized Beta Distribution of the Second Kind and Related Distributions. *Journal of Statistical and Econometric Methods*, **(1)**, 13-31 (2012).
- [10] Aleem, M., Sufyan, M. & Khan, N. S. (2013). A Class of Modified Weighted Weibull Distribution and its properties. *American Review of Mathematics and Statistics*, **(1)**, 29-37 (2013).
- [11] Bashir, S. & Rahul, M. (2015). Some Properties of the Weighted Lindley distribution. *International Journal of Economics and Business Review.*, **(3)**, 11-17 (2015).
- [12] Asgharzadeh, A., Hassan S. Bakouch, Nadarajah, S. & Sharafi, F. (2016). A new weighted Lindley distribution with application. , *Brazilian Journal of Probability and Statistics*, **30**(1), 1-27 (2016).
- [13] Para, B. A. & Jan, T. R. (2018). On Three Parameter Weighted Pareto Type II Distribution: Properties and Applications in Medical Sciences. *Applied Mathematics & Information Sciences Letters*, **6**(1), 13-26 (2018)
- [14] Shakhathreh, M. K. (2012). A two parameter of weighted exponential distributions. *Statistics and Probability Letters*, **82**(2), 252-261 (2012).
- [15] Sherina, V. & Oluyede, B. O. (2014). Weighted Inverse Weibull Distribution: Statistical Properties and Applications. *Theoretical Mathematics and Applications*, **(4)**, 1-30 (2014).
- [16] Nasiru, S. (2015). Another Weighted Weibull Distribution from Azzalini's Family. *European Scientific Journal*, **(11)**, 134-144 (2015).
- [17] Domma, F., Condino, F. & Popovic, B. V. (2016). A new generalized weighted Weibull distribution with

- decreasing, increasing, upside-down bathtub, N-shape and M-shape hazard rate. *Journal of Applied Statistics*, (2016).
- [18] Ghitany, M. E, Al-Mutairi, D. K. & Husain, H. A. (2011). A two-parameter weighted Lindley distribution and its applications to survival data. *Mathematics and Computers in Simulation*, **(81)**, 1190-1201 (2011).
- [19] Dey, S. & Anis, M. Z. (2015). Weighted Weibull distribution: Properties and Estimation. *Journal of Statistical Theory and Practice*, **(9)**, 250-265 (2015).
- [20] Bamus, I. N., Bamiduro, A.T. & Ogunobi, G. S. (2014). Lehmann Type II Weighted Weibull distribution. *Journal of Physical Science*, **9(4)**, 71-78 (2014).
- [21] Alqallaf, F., Ghitany, M. E. & Agostinelli, C. (2015). Weighted Exponential distribution: Different methods of Estimations. *Applied Mathematics and Information Sciences*, **(9)**, 1167- 1173 (2015).
- [22] Al-Omari, I.A. & Alsmarian, I. K. (2020). Weighted Suja distribution with applicaton to ball bearing. *Life Cycle Reliability and Safety Engineering*, **(9)**, 195-211 (2020). DOI: [10.1007/s41872-019-00106-y](https://doi.org/10.1007/s41872-019-00106-y)
- [23] Mizaal, R. A. (2015). Mathematical Study of Weighted Two Parameters Exponential Distribution. *Statistics*, 1-12 (2015).
- [24] Saghir, A., Hamedani, G. G., Tazeem, S. & Khadim, A. (2017). Weighted Distribution: a breif review, pererspective and characterization. *IJSP*, **6(3)**, 109-131 (2017).
- [25] Mahdavi, A. (2015). Two weighted distributions generated by exponential distribution. *Journal of Mathematical Extension*, **(9)**, 1-12 (2015).
- [26] Shanker, R. & Shukla, K. K. (2016). Weighted Akash Distribution and Its Application to model lifetime data. *International Journal of Statistics*, **39(2)**, 1138-1147 (2016).
- [27] Dey S., Ali, S. & Perk, C. (2015), On Weighted Exponential Distribution. *Journal of Satistical Computational and Simulation*, (2015).
- [28] Muhammad, Z., H. (2021). A Class of Bivariate Modified weighted distributions: Properties and Applications. *Annals of Data Science*, (2021). <https://doi.org/10.1007/s40745-021-00346-9>
- [29] Shoae, S. (2020). on a new class of Bivariate Survival Distributions Based on the Model of Dependent lives and its Generalization, *International Journal of Applications and Applied Mathematics*, **15(2)**, 801-829, (2020).
- [30] Bouali, L.D., Chesneau, C. & Sharma, K. V. (2021). A new class of distributions as a finite functional mixture using functional weights. *Anais da Academia Brasileira de Ciências*, **93(2)**, (2021).
- [31] Zubair, A., Elgarhy, M. & Hamedani, G. (2019). Weighted Exponentiated Family of Distributions: Properties, Applications and Characterizations. *Journal of Iranian Statistical Society*, **19(1)**, 209-228 (2019).
- [32] Saghir, A. Ishfaq & Tazeem, S. (2017). The length biased weighted exponentiated inverted Weibull distribution. *Cognent Mathematics*. **3(1)**, (2017).
- [33] Barinaadaa John Nwike, Isaac, Didi Essi, Amos Emeka, (2021). Nwike Probability Distribution: Statistical Properties and Goodness of fit. *Asian Journal of Probability and Statistics*, 52-61 (2021).
- [34] Renyi, A. (1961). On Measures of Entropy and Information, *Proceedings of fourth Berkeley Symposium Mathematics Statistics and Probability*. University of California Press, Berkeley, **(1)**, 547–561 (1961).
- [35] Tsallis, C. (1988). Possible generalization of boltzmann-gibbs Statistics. *Journal of Statistical Physics*, **(52)**, 479-487 (1988).
- [36] Bonferroni, C. E. (1930). *Elementi di Statistica Generale*, Libreria Seeber, Firenze, (1930).
- [37] Lorenz, M.O (1905). Methods of measuring the concentration of wealth. *American Statistical Association*, **(9)**, 209–219 (1905).
- [38] R Core Team, R. version 3.6.2, A Language and environment for statistical computing. R Foundation for Statistical Computing, Vienna, Austria, 2019. <https://www.R-project.org/>
- [39] Bader M.G, Priest A.M. (1982). Statistical aspects of fiber and bundle strength in hybrid composites, In; hayashi, T., Kawata, K. Umekawa, S. (Eds), *Progressin Science in engineering Composites, ICCM-IV*, Tokyo, 1129–1136 (1982).
- [40] Fuller, E. (1994). *Proceedings of SPIE: The International Society for Optical Engineering*, **(2286)**, 419-430 (1994).