

## IRAQI STATISTICIANS JOURNAL

https://isj.edu.iq/index.php/rjes

**ISSN:** (3007-1658) Online



## Sine Type II Topp-Leone Burr XII Distribution: Mathematical Properties and Application to Biomedical and Environmental Data

A. M. Isa<sup>1</sup>, S. I. Doguwa<sup>2</sup>, B. B. Alhaji<sup>3</sup> and H. G. Dikko<sup>4</sup>

## ARTICLE INFO

#### **ABSTRACT**

## Article history:

Received 01 September 2024 Revised 02 September 2024 Accepted 17 September 2024 Available online 25 September 2024

## Keywords:

Sine G Family
Type II Topp-Leone G
Maximum Likelihood Estimate
Moment
Burr XII Distribution

This paper introduces a novel three-parameter lifetime model known as the Sine Type II Top-Leone Burr XII distribution. We derive various mathematical properties of this distribution, including its survival function, hazard rate function, quantile and generating functions, moments, Rényi entropy, moment generating function, and order statistics. The parameters of the proposed model were estimated using the maximum likelihood method. A simulation study examines the consistency of these estimates across both small and large sample sizes. To demonstrate the flexibility and significance of the distribution, we apply it to model Biomedical and Environmental data sets.

#### 1. Introduction

Understanding and accurately modelling real-life data is crucial for advancing knowledge across diverse fields such as finance, engineering, environmental science. biomedical sciences, and geophysics. A pivotal aspect of this endeavour is the selection of an statistical distribution, appropriate directly influences the quality of statistical the development analysis. Despite application of numerous extended distributions to model survival data, significant challenges persist, particularly when dealing with complex real-world data.

In this study, we focus on the Burr XII (BXII) distribution, originally proposed by [1].

This distribution has proven versatile, with numerous applications including failure time modelling, reliability analysis, and use in engineering and medical science. For instance, utilized the three-parameter distribution to model extreme events like flood frequency, while [3] explored the BXII model and its related distributions such as the loglogistic, compound Weibull gamma, Pareto II (Lomax). and Weibull exponential distributions.

Motivated by the extensive use and proven utility of the BXII distribution, several generalizations have been proposed by [3] to [12] to enhance its flexibility. These generalizations were obtained using different types of distribution families. Notably, the Sine

<sup>\*</sup> Corresponding author: E-mail address: alhajimoduisa@bosu.edu.ng https://doi.org/10.62933/tga3z730



<sup>&</sup>lt;sup>1</sup>Department of Mathematics and Computer Science, Borno State University, Maiduguri, Nigeria

<sup>&</sup>lt;sup>2,4</sup>Department of Statistics, Ahmadu Bello University Zaria, Kaduna State, Nigeria

<sup>&</sup>lt;sup>3</sup>Nigerian Defense Academy, Kaduna State, Nigeria

Type II Topp-Leone family, a recent innovation by [13], incorporates a shape parameter and a trigonometric function, offering a versatile tool for enhancing traditional distributions.

In this paper, we propose a novel combination of the Sine Type II Topp-Leone distribution with the BXII distribution, aiming to create a more robust and flexible model.

This new distribution is designed to address unresolved challenges in modelling real data, providing a powerful tool for statistical analysis across various fields. Our objective is extend the Burr XII distribution's applicability, thereby contributing to advancement of statistical modelling techniques and their practical applications.

## 2. Methodology

The cumulative distribution function (CDF) of the Sine Type II Topp-Leone generator as defined by [14] is given by:

$$F(x) = \sin\left\{\frac{\pi}{2} \left[1 - (1 - (G(x))^2)^{\alpha}\right]\right\}. \tag{1}$$

and the corresponding probability density function (PDF) is given as

$$f(x) = \frac{\pi}{2} 2\alpha g(x)G(x) \left[ 1 - (G(x))^2 \right]^{\alpha - 1} \cos \left\{ \frac{\pi}{2} \left[ 1 - \left( 1 - (G(x))^2 \right)^{\alpha} \right] \right\}. \tag{2}$$

where  $\theta$  is the shape parameter, g(x) and G(x) are the PDF and CDF of any baseline distribution. And the CDF and PDF of the baseline (Burr XII) distribution are given in (3) and (4) below:

$$G(x) = 1 - \frac{1}{(1+x^{\alpha})^{\beta}}, x > 0$$
(3)

$$g(x) = \frac{\alpha \beta x^{\alpha - 1}}{(1 + x^{\alpha})^{\beta + 1}},\tag{4}$$

Where  $\alpha > 0$  and  $\beta > 0$  are shape parameters. Substituting (3) and (4) into (1) and (2), we have the CDF of the Sine Type II Topp-Leone Burr XII (STIITLBXII) distribution as follows:

$$F(x) = \sin\left\{\frac{\pi}{2}\left[1 - \left(1 - \left(1 - \frac{1}{(1 + x^{\alpha})^{\beta}}\right)^{2}\right)^{\theta}\right]\right\}, x \alpha, \theta > 0$$

$$(5)$$

and the corresponding PDF is given by:

$$f(x) = \frac{\pi}{2} \frac{2\alpha\beta\theta x^{\alpha-1}}{(1+x^{\alpha})^{\beta+1}} \left( 1 - \frac{1}{(1+x^{\alpha})^{\beta}} \right) \left[ 1 - \left( 1 - \frac{1}{(1+x^{\alpha})^{\beta}} \right)^{2} \right]^{\theta-1} \cos \left\{ \frac{\pi}{2} \left[ 1 - \left( 1 - \frac{1}{(1+x^{\alpha})^{\beta}} \right)^{2} \right]^{\theta} \right] \right\}$$
(6)

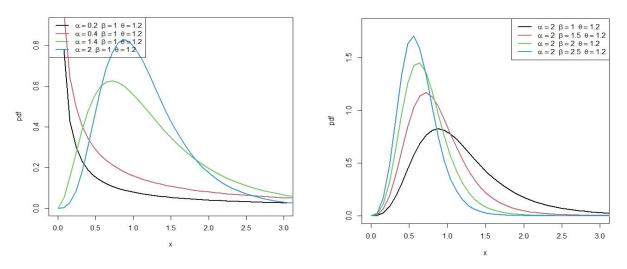


Figure 1: pdf plots of STIITL-BXII distribution for varying parameter values of  $\alpha$ ,  $\beta$  and  $\theta$ 

The survival function S(x), hazard function h(x), reverse hazard function r(x) and the quantile function Q(u) are given by equation (7) to equation (10).

$$S(x) = 1 - \sin\left\{\frac{\pi}{2} \left[1 - \left(1 - \left(1 - \frac{1}{(1 + x^{\alpha})^{\beta}}\right)^{2}\right]^{\alpha}\right]\right\}.$$
 (7)

$$h(x) = \frac{\frac{\pi}{2} \frac{2\alpha\beta\theta x^{\alpha-1}}{(1+x^{\alpha})^{\beta+1}} \left(1 - \frac{1}{(1+x^{\alpha})^{\beta}}\right) \left[1 - \left(1 - \frac{1}{(1+x^{\alpha})^{\beta}}\right)^{2}\right]^{\alpha-1} \cos\left\{\frac{\pi}{2} \left[1 - \left(1 - \left(1 - \frac{1}{(1+x^{\alpha})^{\beta}}\right)^{2}\right)^{\alpha}\right]\right\} \\ 1 - \sin\left\{\frac{\pi}{2} \left[1 - \left(1 - \left(1 - \frac{1}{(1+x^{\alpha})^{\beta}}\right)^{2}\right]^{\alpha}\right]\right\}$$
(8)

$$r(x) = \frac{\pi}{2} \frac{2\alpha\beta\theta x^{\alpha-1}}{(1+x^{\alpha})^{\beta+1}} \left( 1 - \frac{1}{(1+x^{\alpha})^{\beta}} \right) \left[ 1 - \left( 1 - \frac{1}{(1+x^{\alpha})^{\beta}} \right)^{2} \right]^{\alpha-1} \cot \left\{ \frac{\pi}{2} \left[ 1 - \left( 1 - \frac{1}{(1+x^{\alpha})^{\beta}} \right)^{2} \right]^{\alpha} \right] \right\}. \tag{9}$$

$$Q(u) = \left(-\frac{2}{\theta} \ln \left(1 - \left(1 - \left(\frac{\sin^{-1}(u)}{\frac{\pi}{2}}\right)\right)^{\frac{1}{\alpha}}\right)^{\frac{1}{2}}\right)^{\frac{1}{2}}.$$

$$(10)$$

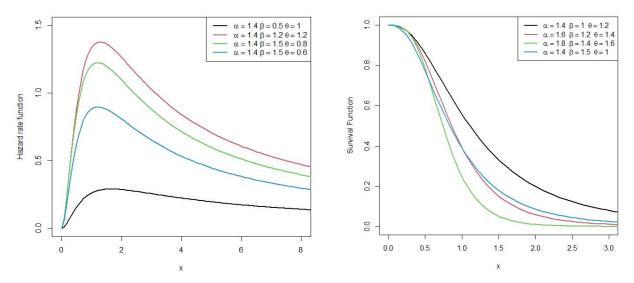


Figure 2: Hazard and Survival plots of STIITL-BXII distribution

## 3. Some Statistical Properties

## 3.1 Linear Expansion of the PDF

The PDF and the CDF of the Sine Type II Topp-Leone Rayleigh distribution in equation (5) and (6) can be expanded using power series expansion as follows:

The Taylor's series expansion of Sine and Cosine functions can be expressed as follows:

$$\cos(x) = \sum_{i=0}^{\infty} \frac{(-1)^i x^{2i}}{2i!},\tag{11}$$

$$\sin(x) = \sum_{i=0}^{\infty} \frac{(-1)^i x^{2i+1}}{(2i+1)!} \,. \tag{12}$$

Then, employing expansion (10) in the last term of PDF in equation (6) follows:

$$\cos\left\{\frac{\pi}{2}\left[1 - \left(1 - \left(1 - \frac{1}{(1+x^{\alpha})^{\beta}}\right)^{2}\right]^{\alpha}\right]\right\} = \sum_{i=0}^{\infty} \frac{(-1)^{i}}{(2i)!} \frac{\pi^{2i}}{2^{2i}} \left[1 - \left(1 - \left(1 - \frac{1}{(1+x^{\alpha})^{\beta}}\right)^{2}\right)^{\alpha}\right]^{2i}.$$
 (13)

Again, using binomial expansion in (13) gives

$$f(x) = \frac{\pi}{2} \frac{2\alpha\beta\theta x^{\alpha-1}}{(1+x^{\alpha})^{\beta+1}} \left(1 - \frac{1}{(1+x^{\alpha})^{\beta}}\right) \left[1 - \left(1 - \frac{1}{(1+x^{\alpha})^{\beta}}\right)^{2}\right]^{\alpha-1}$$

$$\times \sum_{i=0}^{\infty} \frac{(-1)^{i}}{(2i)!} \frac{\pi^{2i}}{2^{2i}} \left[ 1 - \left( 1 - \left( 1 - \frac{1}{(1+x^{\alpha})^{\beta}} \right)^{2} \right)^{\alpha} \right]^{2i}$$

Also, 
$$\left[ 1 - \left( 1 - \left( 1 - \frac{1}{(1 + x^{\alpha})^{\beta}} \right)^{2} \right)^{\alpha} \right]^{2i} = \sum_{j=0}^{\infty} (-1)^{j} {2i \choose j} \left( 1 - \left( 1 - \frac{1}{(1 + x^{\alpha})^{\beta}} \right)^{2} \right)^{\alpha j},$$
 (14)

Then, inserting equation Therefore, equation (6) becomes;

$$f(x) = \frac{\pi}{2} \frac{2\alpha\beta\theta x^{\alpha-1}}{(1+x^{\alpha})^{\beta+1}} \left(1 - \frac{1}{(1+x^{\alpha})^{\beta}}\right) \left[1 - \left(1 - \frac{1}{(1+x^{\alpha})^{\beta}}\right)^{2}\right]^{\alpha-1}$$

$$\times \sum_{i=0}^{\infty} \sum_{j=0}^{2i} \frac{(-1)^{i+j}}{(2i)!} \frac{\pi^{2i}}{2^{2i}} {2i \choose j} \left( 1 - \left( 1 - \frac{1}{(1+x^{\alpha})^{\beta}} \right)^{2} \right)^{\alpha j}$$
 (15)

Again, using binomial expansion,

$$\left(1 - \left(1 - \frac{1}{(1+x^{\alpha})^{\beta}}\right)^{2}\right)^{\alpha j + \alpha - 1} = \sum_{k=0}^{\infty} (-1)^{k} \left(1 - \frac{1}{(1+x^{\alpha})^{\beta}}\right)^{2k}.$$

(16)

 $\left(1-\frac{1}{(1+x^{\alpha})^{\beta}}\right)^2$  can further be simplified using binomial series expansion as follows:

$$\left(1 - \frac{1}{(1+x^{\alpha})^{\beta}}\right)^{2k+1} = \sum_{l=0}^{2k+1} (-1)^{l} {2k+1 \choose l} (1+x^{\alpha})^{-\beta l},$$

$$f(x) = \sum_{i,k=0}^{\infty} \sum_{j=0}^{2i} \sum_{l=0}^{2k+1} \frac{(-1)^{i+j+k+l}}{(2i)!} \frac{\pi^{2i+1}}{2^{2i}} {2i \choose j} {\alpha j + \alpha - 1 \choose k} {2k+1 \choose l} \frac{2\alpha\beta\theta x^{\alpha-1}}{(1+x^{\alpha})^{\beta+1}}$$

Therefore, the PDF can be expressed as

$$f(x) = \sum_{i,k=0}^{\infty} \Phi_{i,j,k,l} \alpha \theta x^{\alpha-1} (1 + x^{\alpha})^{-(\beta l + \beta + 1)} , \qquad (17)$$

Where

$$\Phi_{i,j,k,l} = \sum_{j=0}^{2i} \sum_{l=0}^{2k+1} \frac{(-1)^{i+j+k+l}}{(2i)!} \frac{\pi^{2i+1}}{2^{2i}} \binom{2i}{j} \binom{\alpha \, j + \alpha - 1}{k} \binom{2k+1}{l}.$$

The cdf of the TIITLR distribution in equation (5) can be expanded as follows:

$$\sin\left(\frac{\pi}{2}\left[1-\left(1-\left(1-\frac{1}{(1+x^{\alpha})^{\beta}}\right)^{2}\right)^{\theta}\right]\right) = \sum_{n=0}^{\infty} \frac{(-1)^{n} \pi^{2n+1}}{(2n+1)! 2^{2n+1}} \left[1-\left(1-\frac{1}{(1+x^{\alpha})^{\beta}}\right)^{2}\right)^{\theta}\right]^{2n+1},$$

Also.

$$\left[1 - \left(1 - \left(1 - \frac{1}{(1 + x^{\alpha})^{\beta}}\right)^{2}\right)^{\theta}\right]^{2n+1} = \sum_{m=0}^{2n+1} (-1)^{m} {2n+1 \choose m} \left(1 - \left(1 - \frac{1}{(1 + x^{\alpha})^{\beta}}\right)^{2}\right)^{\alpha m},$$

New, equation (5) becomes:

$$F(x) = \sum_{n=0}^{\infty} \sum_{m=0}^{2n+1} \frac{(-1)^{n+m} \pi^{2n+1}}{(2n+1)! 2^{2n+1}} {2n+1 \choose m} \left(1 - \left(1 - \frac{1}{(1+x^{\alpha})^{\beta}}\right)^{2}\right)^{\theta m},$$

Expanding  $\left(1 - \left(1 - \frac{1}{(1 + x^{\alpha})^{\beta}}\right)^{2}\right)^{\theta m}$  using binomial expansion,

$$\left(1 - \left(1 - \frac{1}{(1 + x^{\alpha})^{\beta}}\right)^{2}\right)^{\theta m} = \sum_{p=0}^{\infty} (-1)^{p} {\alpha m \choose p} \left(1 - \frac{1}{(1 + x^{\alpha})^{\beta}}\right)^{2p},$$

Equation (5) can be written as:

$$F(x) = \sum_{n=0}^{\infty} \sum_{m=0}^{2n+1} \sum_{p=0}^{\infty} \frac{(-1)^{n+m} \pi^{2n+1}}{(2n+1)! 2^{2n+1}} {2n+1 \choose m} {\alpha m \choose p} \left(1 - \frac{1}{(1+x^{\alpha})^{\beta}}\right)^{2p},$$

 $\left(1 - \frac{1}{(1 + e^{\alpha})^{\beta}}\right)^{2p}$  can further be expanded using binomial expansion as follows;

$$\left(1 - \frac{1}{(1+x^{\alpha})^{\beta}}\right)^{2p} = \sum_{q=0}^{2p} (-1)^q \binom{2p}{q} (1+x^{\alpha})^{-\beta q},$$

Therefore the cdf in equation (5) becomes;

$$F(x) = \sum_{n,m,p,q=0}^{\infty} \Psi_{m,n,p,q} (1 + x^{\alpha})^{-\beta q}.$$
 (18) where

$$\Psi_{m,n,p,q} = \sum_{m=0}^{2n} \sum_{q=0}^{2p} \frac{(-1)^{n+m+p+1}}{(2n+1)!} \frac{\pi^{2n+1}}{2^{2n+1}} \binom{2n+1}{m} \binom{\alpha m}{p} \binom{2p}{q}.$$

## 3.2 Moment

Let X be a random sample from STIITL-BXII distribution, the  $r^{th}$  moment of X is obtained as follow:

$$\mu_r = E(x^r) = \int_{-\infty}^{\infty} x^r f(x) dx, \tag{19}$$

inserting the PDF in (17) gives;

$$\mu_r = \sum_{i,k=0}^{\infty} \Phi_{i,j,k,l} \alpha \theta \beta \int_0^{\infty} x^{r+\alpha-1} (1+x^{\alpha})^{-(\beta l+\beta+1)} dx, \tag{20}$$

By integrating with respect to x, equation (20) can be expressed as:

$$\mu_{r} = \sum_{i,k=0}^{\infty} \Phi_{i,j,k,l} \theta \beta \frac{\Gamma\left(\frac{r}{\alpha} + 1\right) \Gamma\left(\beta q + \beta - \frac{r}{\alpha}\right)}{\Gamma\left(\beta q + \beta - 1\right)},$$
(21)

Therefore, the first, second, third and forth moments can be obtained by substituting r = 1, 2, 3 and 4 respectively.

$$\mu_{1} = \sum_{i,k=0}^{\infty} \Phi_{i,j,k,l} \theta \beta \frac{\Gamma\left(\frac{1}{\alpha} + 1\right) \Gamma\left(\beta q + \beta - \frac{1}{\alpha}\right)}{\Gamma\left(\beta q + \beta - 1\right)}$$

$$\mu_{2} = \sum_{i,k=0}^{\infty} \Phi_{i,j,k,l} \theta \beta \frac{\Gamma\left(\frac{2}{\alpha} + 1\right) \Gamma\left(\beta q + \beta - \frac{2}{\alpha}\right)}{\Gamma\left(\beta q + \beta - 1\right)}$$

$$\mu_{3} = \sum_{i,k=0}^{\infty} \Phi_{i,j,k,l} \theta \beta \frac{\Gamma\left(\frac{3}{\alpha} + 1\right) \Gamma\left(\beta q + \beta - \frac{3}{\alpha}\right)}{\Gamma\left(\beta q + \beta - 1\right)}$$

$$\mu_4 = \sum_{i,k=0}^{\infty} \Phi_{i,j,k,l} \theta \beta \frac{\Gamma\bigg(\frac{4}{\alpha} + 1\bigg) \Gamma\bigg(\beta q + \beta - \frac{4}{\alpha}\bigg)}{\Gamma\bigg(\beta q + \beta - 1\bigg)}$$

## 3.3 Renyi Entropy

Let X be an arbitrary variable with parameter  $\theta$  that follows the STIITL-BXII distribution. The Renyi entropy of X with  $\varphi = (\theta)$  by is given as:

$$I_{\lambda}(x) = \frac{1}{1-\lambda} \int_{-\infty}^{\infty} f(x)^{\lambda} dx. \tag{22}$$

For the Sine Type II Topp-Leone Burr XII distribution is given by:

$$f(x)^{\lambda} = \left(\sum_{i,j,k,l=0}^{\infty} \Phi_{i,j,k,l} \alpha \beta \theta x^{\alpha-1} (1+x^{\alpha})^{-(\beta l+\beta+1)}\right)^{\lambda},$$

$$f(x;\theta)^{\lambda} = \left(\sum_{i,j,k,l=0}^{\infty} \Phi_{i,j,k,l} \alpha \beta \theta\right)^{\lambda} \left(x^{\alpha-1} (1+x^{\alpha})^{-(\beta l+\beta+1)}\right)^{\lambda},$$

$$I_{\delta}(x,\theta) = \frac{1}{1-\lambda} \left(\sum_{i,j,k,l=0}^{\infty} \Phi_{i,j,k,l} \alpha \beta \theta\right)^{\lambda} \int_{0}^{\infty} x^{\lambda(\alpha-1)} (1+x^{\alpha})^{-\lambda(\beta l+\beta+1)} dx,$$
(23)

By integrating the last part equation (23), the Renyi Entropy is given by:

$$I_{\lambda}(x) = \frac{1}{(1-\lambda)} \left( \sum_{i,j,k,l=0}^{\infty} \Phi_{i,j,k,l} \beta \theta \right)^{\lambda} \frac{\Gamma(\lambda) \Gamma(\alpha \beta \lambda (l+1) + \lambda r + \lambda - 1)}{\Gamma(\lambda (\beta l + \beta + 1))}$$
(24)

Where  $\lambda \neq 1$ , and  $\lambda > 0$ 

## 4.0 Parameter Estimation

## 4.1 Maximum Likelihood Method

The maximum likelihood method was employed to estimate the parameters of the proposed STIITL-BXII distribution. The likelihood function of the PDF in equation (6) is given by:

$$l = n \log \left(\frac{\pi}{2}\right) + n \log 2 + n \log \alpha + n \log \beta + n \log \theta + (\alpha - 1) \sum_{i=1}^{n} \log x_{i} - (\beta + 1) \sum_{i=1}^{n} \left(1 + x^{\alpha}\right) + \sum_{i=1}^{n} \left(1 - \frac{1}{\left(1 + x^{\alpha}\right)^{\beta}}\right) + (\theta - 1) \sum_{i=1}^{n} \log \left[1 - \left(1 - \frac{1}{\left(1 + x^{\alpha}\right)^{\beta}}\right)^{2}\right] + \sum_{i=1}^{n} \log \left[\cos \left\{\frac{\pi}{2}\left[1 - \left(1 - \frac{1}{\left(1 + x^{\alpha}\right)^{\beta}}\right)^{2}\right]^{\theta}\right]\right).$$
(25)

Differentiating equation (25) partially with respect to  $\alpha$  yields the following expressing.

$$\frac{\partial l}{\partial \alpha} = \frac{n}{\alpha} + \sum_{i=1}^{n} \log x - (\beta - 1) \sum_{i=1}^{n} \log \frac{x^{\alpha} \log x}{(1 + x^{\alpha})} + \sum_{i=1}^{n} \frac{x^{\alpha} \log x}{(1 + x^{\alpha})} \left[ 1 - \frac{1}{\left(1 + x^{\alpha}\right)^{\beta}} \right]^{-1} + 2(\theta - 1) \sum_{i=1}^{n} \frac{x^{\alpha} \log x}{(1 + x^{\alpha})} \left[ 1 - \frac{1}{\left(1 + x^{\alpha}\right)^{\beta}} \right]^{-1} \tan \left\{ \frac{\pi}{2} \left[ 1 - \left[ 1 - \left(1 - \frac{1}{\left(1 + x^{\alpha}\right)^{\beta}}\right)^{2} \right]^{\alpha} \right] \right\}.$$
(26)

Differentiating equation (25) partially with respect to  $\theta$  will gives us the following expressing.

$$\frac{\partial l}{\partial \theta} = \frac{n}{\theta} - \frac{1}{2} \sum_{i=1}^{n} x_{i}^{2} + \sum_{i=1}^{n} \frac{x_{i}^{2} e^{-\left(\frac{\theta}{2}x_{i}^{2}\right)}}{2\left(1 - e^{-\left(\frac{\theta}{2}x_{i}^{2}\right)}\right)} - (\alpha - 1) \sum_{i=1}^{n} x_{i}^{2} e^{-\left(\frac{\theta}{2}x_{i}^{2}\right)} \left(1 - e^{-\left(\frac{\theta}{2}x_{i}^{2}\right)}\right) \left[1 - \left(1 - e^{-\left(\frac{\theta}{2}x_{i}^{2}\right)}\right)^{2}\right] \\
- \sum_{i=1}^{n} \frac{\pi}{2} \alpha x_{i}^{2} e^{-\left(\frac{\theta}{2}x_{i}^{2}\right)} \left(1 - e^{-\left(\frac{\theta}{2}x_{i}^{2}\right)}\right) \left[1 - \left(1 - e^{-\left(\frac{\theta}{2}x_{i}^{2}\right)}\right)^{2}\right]^{\alpha - 1} \tan \left\{\frac{\pi}{2} \left[1 - \left(1 - e^{-\left(\frac{\theta}{2}x_{i}^{2}\right)}\right)^{2}\right]^{\alpha}\right\}.$$
(27)

Equations (26) and (27) are the maximum likelihood estimates of the parameters  $\alpha$  and  $\theta$  respectively.

# 4.2 Assessing the Consistency of the Parameter Estimates

To evaluate the effectiveness of the proposed Sine-Type II Tope Leone-Burr XII distribution, simulation was carried out using the Monte Carlo Simulation method. This aimed to determine the accuracy of parameter estimates derived from maximum likelihood estimates, by computing their mean, bias, and root mean square error. Simulated datasets were generated using the quantile function outlined in equation (10), varying sample sizes from n = 20, 50, 70, 100, 150, 200, 300, 500 and 1000, each replicated 1000 times. Parameter combinations  $\alpha = 0.5$ ,  $\beta = 5.0$  and  $\theta = 1.0$  and  $\alpha = 0.1$ ,  $\beta = 3.3$  and  $\theta = 1.1$  were tested for each sample size. The estimation results, along with bias and root mean square error, are summarized in Table 1.

Table 4.1: Simulation of Sine Type II Topp-Leone Burr XII distribution								
n	<b>Properties</b>	$\alpha = 0.1$	$\beta = 5$	$oldsymbol{ heta}=1$	$\alpha = 0.1$	$\beta = 3.3$	$\theta = 1.1$	
20	Est.	0.6146	5.5305	0.0486	0.5340	3.3175	0.0631	
	Bias	0.5146	0.5305	-0.9514	0.4340	0.0175	-1.0369	
	<b>RMSE</b>	0.5485	0.6545	0.9519	0.4737	0.6014	1.0376	
50	Est.	0.5932	5.6089	0.0490	0.5128	3.2778	0.0653	
	Bias	0.4932	0.6089	-0.9510	0.4128	-0.0222	-1.0347	
	<b>RMSE</b>	0.5080	0.6539	0.9515	0.4487	0.5669	1.0353	
70	Est.	0.5880	5.6285	0.0491	0.4947	3.2756	0.0665	
	Bias	0.4880	0.6285	-0.9509	0.3947	-0.0244	-1.0335	
	<b>RMSE</b>	0.4995	0.6535	0.9514	0.4285	0.5517	1.0341	
100	Est.	0.5705	5.6366	0.0510	0.4859	3.2492	0.0680	
	Bias	0.4705	0.6366	-0.9490	0.3859	-0.0508	-1.0320	
	<b>RMSE</b>	0.4807	0.6554	0.9495	0.4173	0.5208	1.0326	
150	Est.	0.5399	5.6280	0.0533	0.4760	3.2427	0.0688	
	Bias	0.4399	0.6280	-0.9467	0.3760	-0.0573	-1.0312	
	RMSE	0.4492	0.6486	0.9472	0.4044	0.4911	1.0318	

200	Est.	0.5210	5.6165	0.0547	0.4619	3.2476	0.0699
	Bias	0.4210	0.6165	-0.9453	0.3619	-0.0524	-1.0301
	<b>RMSE</b>	0.4297	0.6381	0.9458	0.3880	0.4864	1.0307
300	Est.	0.4939	5.6135	0.0575	0.4634	3.2065	0.0706
	Bias	0.3939	0.6135	-0.9425	0.3634	-0.0935	-1.0294
	<b>RMSE</b>	0.4013	0.6406	0.9431	0.3885	0.4725	1.0300
500	Est.	0.4613	5.6130	0.0609	0.4416	3.2066	0.0738
	Bias	0.3613	0.6130	-0.9391	0.3416	-0.0934	-1.0262
	<b>RMSE</b>	0.3676	0.6517	0.9397	0.3660	0.4753	1.0268
1000	Est.	0.4157	5.5906	0.0659	0.4170	3.1885	0.0772
	Bias	0.3157	0.5906	-0.9341	0.3170	-0.1115	-1.0228
	<b>RMSE</b>	0.3206	0.6718	0.9346	0.3399	0.4962	1.0234

Table 2 presents a simulation study of the Sine Type II Topp-Leone Burr XII distribution, showing the estimated parameters, bias, and root mean square error (RMSE) for different sample sizes (n) and parameter values  $\alpha, \beta$  and  $\theta$ . The results are reported for two sets of parameter combinations:  $\alpha = 0.1$ ,  $\beta = 5$ ,  $\theta = 1$  and  $\alpha = 0.1$ ,  $\beta = 3.3$ ,  $\theta = 1.1$ . As the sample size increases from 20 to 1000, the estimated values for each parameter generally get closer to the true indicating parameter values, improved estimation accuracy. Bias and RMSE decrease with larger sample sizes, showing more precise and reliable estimates.

## 5. Application

The first data set is the relief times of twenty patients receiving an analysesic is by [15]. The data sets are as follows:

1.1, 1.4, 1.3, 1.7, 1.9, 1.8, 1.6, 2.2, 1.7, 2.7, 4.1, 1.8, 1.5, 1.2, 1.4, 3, 1.7, 2.3, 1.6, 2.

The second dataset has thirty consecutive values of precipitation (in inches) in the month of March in Minneapolis, as provided by [16] and recently used by [17]. The data are as follows:

0.77, 1.74, 0.81, 1.2, 1.95, 1.2, 0.47, 1.43, 3.37, 2.2, 3, 3.09, 1.51, 2.1, 0.52, 1.62, 1.31, 0.32, 0.59, 0.81, 2.81, 1.87, 1.18, 1.35, 4.75, 2.48, 0.96, 1.89, 0.9, 2.05

Table 3: MLE, AIC, BIC, CAIC and HQIC of the two datasets

MODEL	a	ι β	θ	λ	-LL	AIC	CAIC	BIC	HQIC
STIITLBXII	8.55	559 0.06	620 6.4606	-	-15.8815	37.7631	39.2601	40.7503	38.3462
TLEBXII	6.73	344 0.32	255 2.1806	1.9714	-15.4249	38.8499	41.5166	42.8329	39.6274
SBXII	13.0	0.00	)13 -	-	-32.6757	69.3115	70.0574	71.3429	69.7403
BXII	8.1698	2.6948	-	-	-31.7233	67.447	68.1525	69.4381	67.8354
MODEL						ATO	GATO	DIC	TIOIC
MODEL	α	β	$oldsymbol{ heta}$	λ	-LL	AIC	CAIC	BIC	HQIC
STIITLBXII	1.6053	0.2743	6.4751	-	-38.510	82.8676	84.4676	85.5848	83.4507
TLEBXII	1.4798	0.0302	0.1416	13.353	-78.639	151.277	150.354	147.974	149.933
SBXII	8.8319	0.0013	-	-	-46.281	96.5617	97.0061	99.3641	97.4582
BXII	5.0722	14.938	-	-	-44.371	92.7429	93.1873	95.5453	93.6394

Based on the model selection criteria, that is, the AIC, CAIC, BIC, and HQIC, the STIITLBXII model is the best fit for both datasets, as it consistently has the lowest values across all indices, indicating a superior balance between model complexity and fit to the data. This suggests that STIITLBXII provides the most efficient representation with minimal information loss.

#### 6.0 Conclusion

The application of the STIITLBXII model to two distinct datasets: one involving relief times for patients receiving analysis and the other consisting of precipitation data in Minneapolis demonstrates the superior performance of the proposed model across various model selection criteria, including AIC, CAIC, BIC, and HQIC.

#### References

- [1] Burr, I. W. (1942). Cumulative frequency functions. *The Annals of mathematical statistics*, 13(2), 215-232.
- [2] Shao, Q. (2004). Notes on maximum likelihood estimation for the three-parameter Burr XII distribution. *Computational statistics & data analysis*, 45(3), 675-687.
- [3] Tadikamalla, P. R. (1980). A look at the Burr and related distributions. *International Statistical Review*, 337-344.
- [4] Afify, A., & Abdellatif, A. (2020). The extended Burr XII distribution: properties and applications. *J. of Nonlinear Sci. and Appl, 13*, 133-146.
- [5] Guerra, R. R., Peña-Ramírez, F. A., & Cordeiro, G. M. (2017). The gamma Burr XII distribution: Theory and application. *Journal of Data Science*, 15(3), 467-494.
- [6] Handique, L., & Chakraborty, S. (2018). A new four-parameter extension of Burr-XII distribution: Its properties and applications. *Japanese Journal of Statistics and Data Science*, 1, 271-296.
- [7] Reyad, H. M., & Othman, S. A. (2017). The Topp-Leone Burr-XII distribution: properties and applications. *British Journal of Mathematics & Computer Science*, 21(5), 1-15.
- [8] Ghosh, I., & Bourguinon, M. (2017). A new extended Burr XII distribution. *Austrian Journal of Statistics*, 46(1), 33-39.
- [9] Kamal, R. M., & Ismail, M. A. (2020). The flexible Weibull extension-burr XII

For both datasets, the STIITLBXII model achieved the lowest values for these criteria, indicating it is the most efficient model in fitting the two datasets. Furthermore, the Monte Carlo simulation studies validate the consistency and accuracy of the parameter estimates of the STIITLBXII distribution across various sample sizes, demonstrating that as the sample size increases, the parameter estimates become more accurate, and both bias and RMSE decrease.

- distribution: model, properties and applications. *Pakistan Journal of Statistics and Operation Research*, 447-460.
- [10] Paranaíba, P. F., Ortega, E. M., Cordeiro, G. M., & Pascoa, M. A. D. (2013). The Kumaraswamy Burr XII distribution: theory and practice. *Journal of Statistical Computation and Simulation*, 83(11), 2117-2143.
- [11] Santos, M. C., & Pescim, R. R. (2024). A new extension of the Burr XII distribution generated by odd log-logistic random variables. *Communications in Statistics-Theory and Methods*, *53*(14), 5003-5017.
- [12] Anafo, A. Y., Ocloo, S. K., & Nasiru, S. (2024). New Weighted Burr XII Distribution: Statistical Properties, Applications, and Regression. International Journal of Mathematics and Mathematical Sciences, 2024(1), 4098771.
- [13] Isa, A. M., Doguwa, S. I., Alhaji, B. B., & Dikko, H. G. (2023). Sine Type II Topp-Leone G Family of Probability Distribution: Mathematical Properties and Application. *Arid Zone Journal of Basic and Applied Research*, 2(4), 124-138.
- [14] Gross, A. J., & Clark, V. A. (1975). Survival distributions: Reliability applications in the biomedical sciences John Wiley and Sons. *New York*.
- [15] Hinkley, D. (1977). On quick choice of power transformation. *Journal of the Royal Statistical*

- Society: Series C (Applied Statistics), 26(1), 67-69.
- [16] Yusuf, A., Mikail, B. B., Aliyu, A. I., & Sulaiman, A. L. (2016). The inverse Burr negative binomial distribution with application to real data. *Stat. J. Theor. Appl. Stat*, *5*, 53-65.