

## On Semicircular Extreme-value distribution

**R. Srinivas**

*Assistant Professor of Mathematics, Hindu College,  
Guntur, India*

**Y. Phani**

*Associate Professor of Mathematics, Shri Vishnu Engineering  
College for Women, Vishnupur, Bhimavaram, India.*

**S.V.S. Girija**

*Professor of Mathematics, Hindu College, Guntur, India*

### Abstract

This paper is aimed at construction of Semicircular Extreme-value distribution for modeling semicircular data by applying simple projection method on Extreme-value distribution. It is extended it to the  $l$ -axial Extreme-value distribution by a simple projection for modeling any arc of arbitrary length.

**Keywords:** characteristic function, semi circular models,  $l$ -axial data, projection, trigonometric moments.

**AMS Subject Classification:** 60E05, 62H11

### 1. INTRODUCTION

In some of the cases the directional / angular data does not require full circular models for modeling, this fact is noted in Guardiola (2004), Jones (1968) and Byoung et al(2008). For example, when a sea turtle emerges from the ocean in search of a nesting site on dry land, a random variable having values on a semicircle is well sufficient for modeling such data. Similarly, when an aircraft is lost but its departure and its initial headings are known, a semicircular random variable is sufficient for such angular data. And few more examples of semicircular data is available in Ugai et al (1977).

Guardiola (2004) obtained the semicircular normal distribution by using a simple projection and Byoung et al (2008) developed a family of the semicircular Laplace distributions for modeling semicircular data by simple projection, Phani et al (2013) constructed some semicircular distributions by applying inverse stereographic projection. In this paper the **Semicircular Extreme-value distribution (SCEVD)** is developed by projecting Extreme-value distribution over a semicircular segment. The graphs of the density and distribution functions for various values of parameters are plotted. The characteristic function is also derived. This model is extended to  $l$ -axial model.

$$g(\theta) = \frac{1}{\sigma} \sec^2(\theta) \exp\left(-\frac{(\tan(\theta) - \mu)}{\sigma}\right) \exp\left\{-\exp\left(-\frac{(\tan(\theta) - \mu)}{\sigma}\right)\right\},$$

$$\text{where, } \mu = \frac{\gamma}{v}, \sigma = \frac{\lambda}{v} > 0 \quad (2.3)$$

### 2. SEMICIRCULAR EXTREME-VALUE DISTRIBUTION

A random variable  $X$  on the real line is said to have Extreme-value Distribution with location parameter  $\gamma$  and scale parameter  $\lambda > 0$ , if the probability density function and cumulative distribution function of  $X$  for  $x, \gamma \in \mathbb{R}$  and  $\lambda > 0$  are given respectively by

$$f(x) = \frac{1}{\lambda} \exp\left(\frac{-(x-\gamma)}{\lambda}\right) \exp\left(-\exp\left(\frac{-(x-\gamma)}{\lambda}\right)\right),$$

where  $\lambda > 0$ , and  $\gamma, x \in \mathbb{R}$  (2.1)

$$F(x) = \exp\left(-\exp\left(\frac{-(x-\gamma)}{\lambda}\right)\right). \quad (2.2)$$

Simple projection is defined by

mapping  $x = v \tan(\theta)$  or  $\theta = \tan^{-1}\left(\frac{x}{v}\right)$ ,  $v > 0 \in \mathbb{R}$ .

Application of this simple projection on Extreme – value distribution results to a Semicircular Extreme-value distribution.

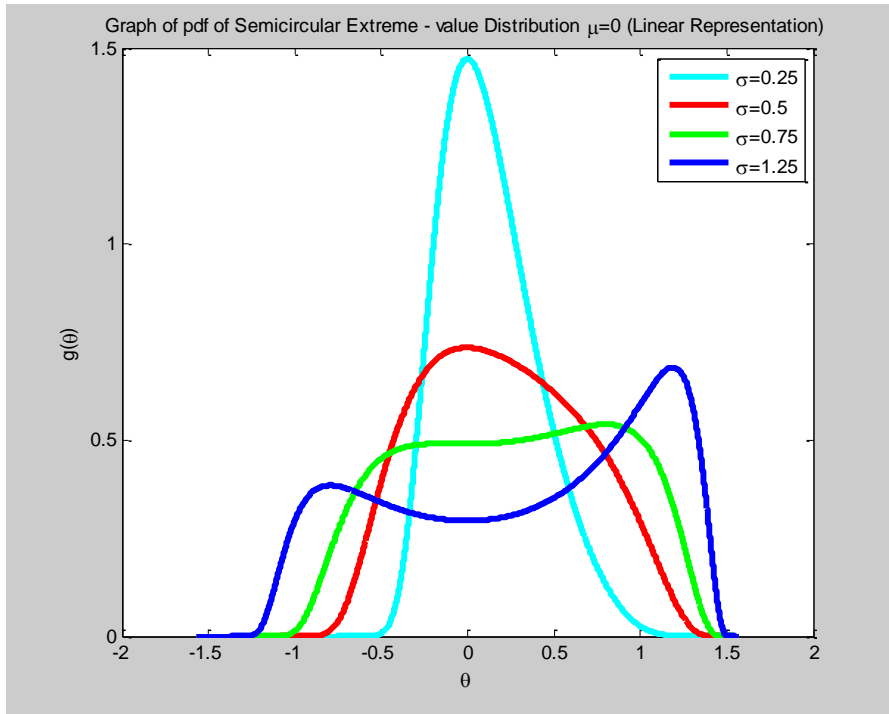
#### Definition:

A random variable  $X_{SC}$  on the Semicircle is said to have the Semicircular Extreme-value distribution with location parameter  $\mu$  scale parameter  $\sigma > 0$  denoted by **SCEVD**  $(\sigma, \mu)$ , if the probability density and the cumulative distribution functions are respectively given by

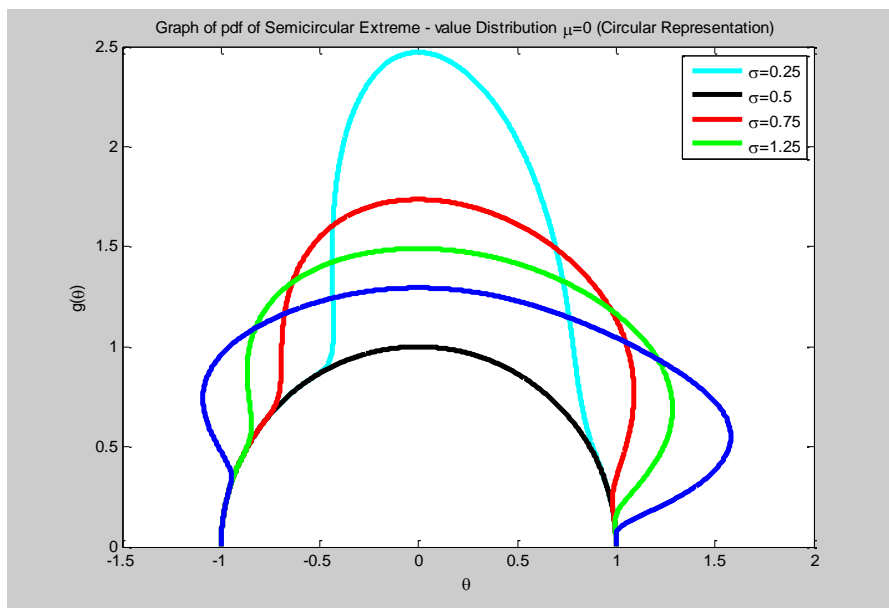
$$G(\theta) = \exp \left\{ - \exp \left( - \frac{(\tan(\theta) - \mu)}{\sigma} \right) \right\}, \text{ where } \sigma > 0, -\frac{\pi}{2} < \theta < \frac{\pi}{2} \quad (2.4)$$

Hence the proposed new model **SCED**( $\sigma, \mu$ ) is a Semicircular model.

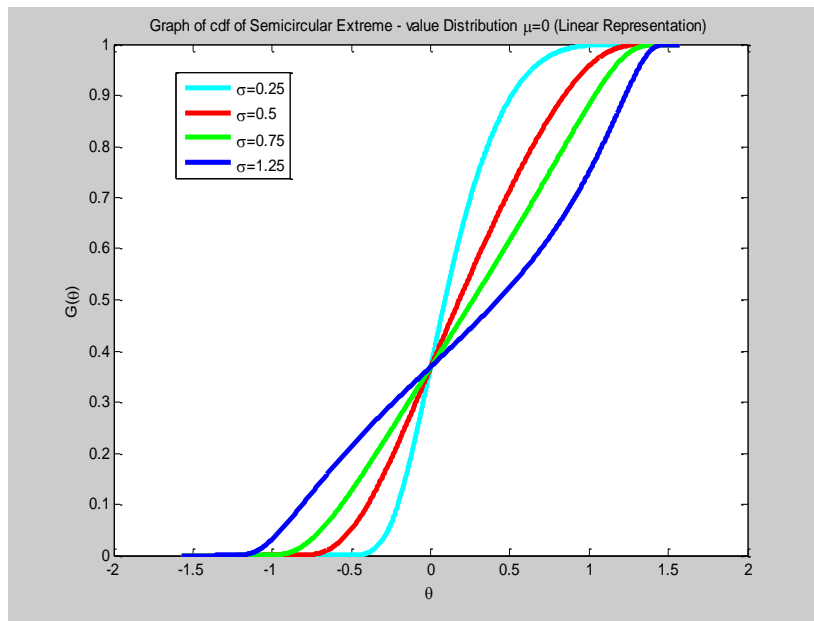
**Graphs of the probability density and cumulative distribution functions of the Semicircular Extreme-value distribution for various values of  $\sigma$  and  $\mu$**



**Fig1.** Graph of pdf of Semicircular Extreme – value Distribution (Linear Representation)



**Fig2.** Graph of pdf of Semicircular Extreme – value Distribution (Circular Representation)



**Fig3.** Graph of cdf of Semicircular Extreme – value Distribution

### 3. The Characteristic function of Semicircular Extreme-value distribution

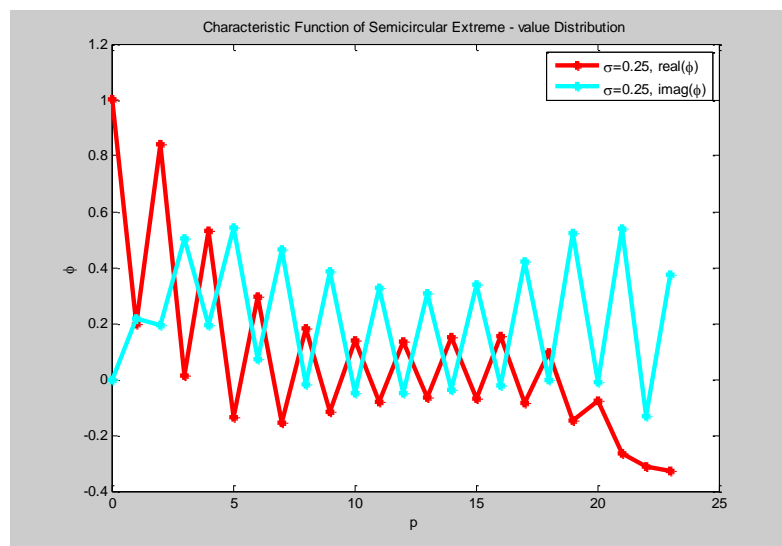
The characteristic function of a semicircular model with probability density function  $g(\theta)$  is defined as

$$\Phi_{X_{sc}}(p) = \int_{-\frac{\pi}{2}}^{\frac{\pi}{2}} e^{ip\theta} g(\theta) d\theta, p \in \mathbb{Z}.$$

$$\Phi_{X_{sc}}(p) = \frac{1}{\sigma} \int_{-\frac{\pi}{2}}^{\frac{\pi}{2}} e^{ip\theta} \sec^2(\theta) \exp\left(-\frac{(\tan(\theta)-\mu)}{\sigma}\right) \exp\left\{-\exp\left(-\frac{(\tan(\theta)-\mu)}{\sigma}\right)\right\} d\theta, p \in \mathbb{Z} \quad (3.1)$$

As the integral cannot be obtained analytically, MATLAB techniques are applied for the evaluation of the values of the characteristic function. Population characteristics are also studied using first two trigonometric moments.

Graphs of the characteristic function of Semicircular Extreme -value distribution for various values of parameter are plotted.



**Fig4.** Graph of Characteristic function of Semicircular Extreme – value Distribution

The expressions for mean direction, resultant length, circular variance, circular standard deviation, central trigonometric moments, skewness and kurtosis for circular distributions are available in Mardia and Jupp (2000). These characteristics for the Semicircular Extreme - value distribution are also based on their respective trigonometric moments and can be

expressed in terms of trigonometric moments  $\alpha_p$  and  $\beta_p$  which are presented here. From the population characteristics, it can be observed that with increasing value of scale parameter  $\sigma$ , the circular variance gradually decrease, the distribution is negatively skewed and remained platykurtic

**Table 3.1** Population Characteristics of Semicircular Extreme – value model

	$\sigma = 0.25$	$\sigma = 0.5$	$\sigma = 0.75$	$\sigma = 1.25$
<b>Trigonometric Moments</b>				
$\alpha_1$	0.1995	0.1074	-0.0183	-0.1525
$\alpha_2$	0.8426	0.6299	0.4812	0.1586
$\beta_1$	0.2176	0.3461	0.3727	0.2671
$\beta_2$	0.1926	0.2203	0.2488	0.3660
<b>Resultant Length</b>				
$\rho_1$	0.2952	0.3624	0.3732	0.3076
$\rho_2$	0.8643	0.6673	0.5417	0.3989
<b>Mean Direction <math>\mu_0</math></b>				
	0.8288	1.2700	-1.5218	-1.0518
<b>Circular Variance <math>\nu_0</math></b>				
	0.7048	0.6376	0.6268	0.6124
<b>Circular Standard Deviation</b>				
$\sigma_0$	1.5622	1.4249	1.4041	1.5356
	0.5400	0.8994	1.1074	1.3558
<b>Central Trigonometric Moments</b>				
$\alpha_1^*$	0.2952	0.3624	0.3732	0.3076
$\alpha_2^*$	0.1188	-0.3947	-0.5032	-0.3958
$\beta_1^*$	0	0	0	0
$\beta_2^*$	-0.8561	-0.5381	-0.2005	-0.0493
<b>Skewness <math>\gamma_1^0</math></b>				
	-1.4467	-1.0568	-0.4040	-0.0856
<b>Kurtosis <math>\gamma_2^0</math></b>				
	0.2239	-1.0131	-1.3299	-0.8442

**4 Extension to  $l$ -axial distribution**

The newly proposed model is extended to the  $l$ -axial distribution, which is applicable to any arc of arbitrary length say  $2\pi/l$  for  $l=1,2,\dots$ . So it is desirable to extend the Semicircular Extreme-value distribution to its  $l$ -axial model. To construct the  $l$ -axial Extreme-value distribution, the density function of Semicircular Extreme-value distribution is

considered and the transformation  $\phi = 2\theta/l$ ,  $l=1,2,\dots$ , is used. The probability density function of  $\phi$  is given by

$$g(\phi) = \frac{l}{2\sigma} \sec^2\left(\frac{l\phi}{2}\right) e^{-\frac{\tan\left(\frac{l\phi}{2}\right)}{\sigma}} e^{-e^{-\frac{\tan\left(\frac{l\phi}{2}\right)}{\sigma}}}$$

where  $-\frac{\pi}{l} < \phi < \frac{\pi}{l}$  (4.1)

**Case (1)** When  $l = 2$ , the probability density function (4.1) is the same as the probability density function of Semicircular **Extreme-value distribution**.

**Case (2)** When  $l = 1$ , the probability density function (4.1) is the same as that of **Stereographic Extreme-value Distribution** [Phani et al., 2012] which is circular distribution.

If  $X$  has Extreme-value with scale parameter  $\lambda$  and location parameter  $\gamma = 0$ , then  $Y = e^X$  has a **log-Extreme-value distribution** with the same parameters. So, using  $y = v \tan(\theta)$ , a probability density function is

$$g(\theta) = \frac{2}{\lambda \sin(2\theta)} e^{-\frac{1}{\lambda}(\log(v \tan(\theta)))} e^{-e^{-\frac{1}{\lambda}(\log(v \tan(\theta)))}} \quad 0 < \theta < \frac{\pi}{2}, \quad (4.2)$$

Suppose  $\phi = \frac{4\theta}{l}$ ,  $l=1,2,\dots$ , then  $\phi$  has a probability density function,

$$g(\phi) = \frac{1}{2\lambda \sin\left(\frac{l\phi}{2}\right)} e^{-\frac{1}{\lambda}\left(\log\left(v \tan\left(\frac{l\phi}{4}\right)\right)\right)} e^{-e^{-\frac{1}{\lambda}\left(\log\left(v \tan\left(\frac{l\phi}{4}\right)\right)\right)}}, \quad 0 < \theta < \frac{2\pi}{l} \quad (4.3)$$

It is said that  $\phi$  follows the  $l$ -axial **log-Extreme-value distribution**.

**Case (1):** When  $l=1$ ,

$$g(\phi) = \frac{1}{2\lambda \sin\left(\frac{\phi}{2}\right)} e^{-\frac{1}{\lambda}\left(\log\left(v \tan\left(\frac{\phi}{4}\right)\right)\right)} e^{-e^{-\frac{1}{\lambda}\left(\log\left(v \tan\left(\frac{\phi}{4}\right)\right)\right)}}, \quad 0 < \theta < 2\pi \quad (4.4)$$

It is called as a **Circular log-Extreme-value distribution**.

**Case (2):** When  $l = 2$ ,

$$g(\phi) = \frac{1}{2\lambda \sin(\phi)} e^{-\frac{1}{\lambda}\left(\log\left(v \tan\left(\frac{\phi}{2}\right)\right)\right)} e^{-e^{-\frac{1}{\lambda}\left(\log\left(v \tan\left(\frac{\phi}{2}\right)\right)\right)}}, \quad 0 < \theta < \pi \quad (4.5)$$

It is called as a **Semicircular log-Extreme-value distribution**.

**Case (3):** When  $l = 4$ ,

$$g(\phi) = \frac{1}{2\lambda \sin(2\phi)} e^{-\frac{1}{\lambda}(\log(v \tan(\theta)))} e^{-e^{-\frac{1}{\lambda}(\log(v \tan(\theta)))}}, \quad 0 < \theta < \frac{\pi}{2} \quad (4.6)$$

## REFERENCES

- [1] Abramowitz, M. and Stegun, I.A. (1965). Handbook of Mathematical Functions, *Dover, New York*.
- [2] Byoung, J.A and Hyoun M.K.(2008). A New Family of Semicircular Models: The Semicircular Laplace Distributions, *Communications of the Korean Statistical Society* Vol.15, pp. 775-781.
- [3] Gradshteyn and Ryzhik (2007). Table of Integrals, series and products, 7th edition, Academic Press.
- [4] Guardiola, J.H. (2004). The Semicircular Normal Distribution, Ph.D Dissertation, Baylor University, Institute of Statistics.
- [5] Jammalamadaka S. Rao and Sen Gupta, A. (2001). Topics in Circular Statistics, *World Scientific Press, Singapore*.
- [6] Jones, T.A. (1968). Statistical Analysis of orientation data, *Journal of Sedimentary Petrology*, 38, 61-67.
- [7] Kim,H.M.(2008). New family of the t distributions for modelling semicircular data, *Communications of the Korean Statistical Society*.
- [8] Mardia, K.V. and Jupp, P.E. (2000), *Directional Statistics, John Wiley, Chichester*.
- [9] Ugai.S.,K., Nishijima, M. and Kan,T.(1977), Characteristics of raindrop size and raindrop shape, *Open symposium URSI Commission F*, 225-230.
- [10] Phani Y, Girija S.V.S., and Dattatreya Rao A.V. (2013). On Construction of Stereographic Semicircular Models, *Journal of Applied Probability and Statistics*. Vol 8, No. 1, pp. 75-90.