

# Half Logistic - Rayleigh Distribution: Properties

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**Abstract:** In this paper, we introduce a new two parameter Half Logistic Rayleigh Distribution (HL-RD). Which propound a more adjustable mode for modelling simulated data. The recommended distribution reveals increasing, decreasing and bath-shaped probability Density, Distribution and Hazard Rate Functions. Some distributional properties of new model are investigated which include the Density Function, Distribution Function (DF), Quantile Function(QF), Moments, Moment Generating Function (MGF), Cumulative Generating Function (CGF) . The parameters involved in the model are estimated using Maximum Likelihood Estimation (MLE) method.

**Keywords:** Half Logistic Rayleigh Distribution, Moments, MGF, CGF, QF, MLE.

## 1. Introduction:

The commonly used Lifetime distributions (Weibul, Logistic, Lomax, etc.,) have a restricted range of performance. Such type of distributions cannot give a better fit to model for all practical situations. Recently, several authors have developed a number of new families of statistical models by applying different techniques. Various techniques have been introduced in the literature to derive new flexible models as discussed by Lai (2013). Al-Awadhi and Ghitany (2001) introduced the discrete Poisson–Lomax distribution by using the Lomax distribution as a mixing distribution for the Poisson parameter. Beta-Pareto has been presented by Akinsete, Famoye and Lee (2008). Uniform Exponential Distribution (UED) and Exponential Pareto distribution (EPD) were introduced by Abed Al-Kadim and Abdalhussain Boshi(2013). Sharma and Shanker (2013) used a mixture of exponential (q) and gamma (2;q) to create a two-parameter Lindley distribution.

In this paper, we introduce new linear compound distribution as Half Logistic Rayleigh Distribution (HLRD) and discuss some of its properties, such as Distribution, Hazard Function, Quantile and Random Generation Moments, and its Moment Generating Function (MGF), Cumulative Generating Function (CGF). We estimate parameters by Maximum Likelihood method and also define Asymptotic Confidence bounds for Half Logistic Rayleigh Distribution. Finally we use Simulation study about the parameters estimation.

## 2. The Probability density and Distribution functions of the HLRD

### 2.1 HLRD Specifications

In this section, we define new Scale ( $\theta$ ) and Shape ( $\lambda$ ) two parameter distribution called Half Logistic Rayleigh Distribution with parameters  $\theta$  and  $\lambda$ . The Probability Density Function (PDF), Cumulative Distribution Function (CDF), Survival Function (SF) and Hazard Function (HF) of the new model HLRD are respectively defined as follows:

A random variable  $X \sim \text{HLRD}(\theta, \lambda)$  has Probability density function and is in the form

$$f(x; \theta, \lambda) = \frac{4\lambda(x-\theta)e^{-\lambda(x-\theta)^2}}{\left[1 + e^{-\lambda(x-\theta)^2}\right]^2}; x > \theta, \lambda > 0 \quad \dots(1)$$

Where  $x > \theta, \lambda > 0$

A random variable  $X \sim \text{HLRD}(\theta, \lambda)$  correspond Cumulative Distribution Function is in the form

$$F(x; \theta, \lambda) = \frac{1 - e^{-\lambda(x-\theta)^2}}{1 + e^{-\lambda(x-\theta)^2}}; x > \theta, \lambda > 0 \quad \dots(2)$$

Where  $x > \theta, \lambda > 0$

## 2.2 Limits of the HLRD function

The limit of the Probability Density Function is given by

$$\lim_{x \rightarrow 0} f(x; \theta, \lambda) = \frac{4\lambda(-\theta)e^{-\lambda(-\theta)^2}}{[1 + e^{-\lambda\theta^2}]^2} \quad \dots(3)$$

$$\lim_{x \rightarrow \infty} f(x; \theta, \lambda) = \frac{-4\lambda\theta e^{-\lambda\theta^2}}{[1 + e^{-\lambda\theta^2}]^2} = 0 \quad \dots(4)$$

## 3. Survival and Hazard Functions

### 3.1 Survival Function

If  $X \sim \text{HLRD}(\theta, \lambda)$ , then Survival Function of HLRD is given by

$$S(x, \theta, \lambda) = \frac{2e^{-\lambda(x-\theta)^2}}{1 + e^{-\lambda(x-\theta)^2}}; x > \theta, \lambda > 0 \quad \dots(5)$$

### 3.2 Hazard Function

If  $X \sim \text{HLRD}(\theta, \lambda)$ , then Hazard Function of HLRD is given by

$$h(x; \theta, \lambda) = \frac{2\lambda(x-\theta)}{[1 + e^{-\lambda(x-\theta)^2}]^2}; x > \theta, \lambda > 0 \quad \dots(6)$$

Here  $(\theta, \lambda)$  are Location and Scale parameters.

## 4. Quantile Function and Random Generation

### 4.1 Quantile Function

If  $x \sim \text{HLRD}(\theta, \lambda)$ , then Quantile Function is obtained as follows

$$Q(x; \theta, \lambda) = \theta + \frac{1}{\sqrt{\lambda}} \sqrt{\log p} \quad \dots(7)$$

$$\theta + \sqrt{\frac{\log p}{\lambda}}$$

### 4.2 Random Generation

Substitute  $F(x) = u$  in (2), we obtain

$$x = \theta + \frac{1}{\sqrt{\lambda}} [\log Y]^{1/2} \quad \dots(8)$$

Equation (8) can be used to simulate HLRvariable.

## 5 Statistical Properties of HLRD

### 5.1. Moments

The following theorem gives moments of HLRD

**Theorem:**

The  $r^{\text{th}}$  moment about the origin of  $X \sim \text{HLRD}(\theta, \lambda)$  is given by

$$\mu_r^1 = E(x^r) = 2 \sum_{k=0}^{\infty} \sum_{j=0}^{\infty} \binom{r}{k} \binom{-2}{j} \frac{\theta^{r-k}}{(\sqrt{\lambda})^k} \left[ \frac{\Gamma\left(\frac{k+2}{2}, 0\right)}{(-j-1)^{k/2} (j+1)} - \frac{\Gamma\left(\frac{k+2}{2}, (-j-1)\lambda\theta^2\right)}{(-j-1)^{k/2} (j+1)} \right]$$

**Proof:** We have

$$\begin{aligned} \mu_r^1 &= E(x^r) \\ &= \int_0^{\infty} x^r \cdot \frac{4\lambda(x-\theta)e^{-\lambda(x-\theta)^2}}{[1+e^{-\lambda(x-\theta)^2}]^2} dx \end{aligned} \quad \dots(9)$$

$$= 2 \int_{1+e^{-\lambda\theta^2}}^1 t^{-2} [\log(t-1)]^{k/2} dt \quad \dots(10)$$

$$= 2 \int_{\lambda\theta^2}^0 z^{k/2} \frac{e^z}{(1+e^z)^2} dz \quad \dots(11)$$

$$= 2 \sum_{k=0}^{\infty} \binom{r}{k} \frac{\theta^{r-k}}{(\sqrt{\lambda})^k} \int_{\lambda\theta^2}^0 z^{k/2} \frac{e^z}{(1+e^z)^2} dz \quad \dots(12)$$

$$(1+e^z)^{-2} = \sum_{j=0}^{\infty} \binom{-2}{j} (e^z)^j \quad \dots(13)$$

$$= 2 \sum_{k=0}^{\infty} \sum_{j=0}^{\infty} \binom{r}{k} \binom{-2}{j} \frac{\theta^{r-k}}{(\sqrt{\lambda})^k} \int_{\lambda\theta^2}^0 z^{k/2} e^{z(1+j)} dz \quad \dots(14)$$

Consider

$$= \int_{\lambda\theta^2}^0 z^{k/2} e^{z(1+j)} dz \quad \dots(15)$$

$$= \left[ \frac{\Gamma\left(\frac{k+2}{2}, 0\right)}{(-j-1)^{k/2} (j+1)} - \frac{\Gamma\left(\frac{k+2}{2}, (-j-1)\lambda\theta^2\right)}{(-j-1)^{k/2} (j+1)} \right] \quad \dots(16)$$

Substitute (16) in(14)

$$\mu_r^1 = 2 \sum_{k=0}^{\infty} \sum_{j=0}^{\infty} \binom{r}{k} \binom{-2}{j} \frac{\theta^{r-k}}{(\sqrt{\lambda})^k} \left[ \frac{\Gamma\left(\frac{k+2}{2}, 0\right)}{(-j-1)^{k/2} (j+1)} - \frac{\Gamma\left(\frac{k+2}{2}, (-j-1)\lambda\theta^2\right)}{(-j-1)^{k/2} (j+1)} \right]$$

### 5.2. Moment Generating Function

Theorem2:- prove that

$$M_x(t) = \sum_{i=0}^{\infty} \sum_{j=0}^{\infty} \frac{(t\theta)^j}{t^i (\sqrt{\lambda})^i i! j!} \left[ \left[ \frac{\Gamma\left(\frac{i}{2}+1\right), 0}{(-j-1)^{k/2} (j+1)} \right] - \left[ \frac{\Gamma\left(\frac{i}{2}+1\right), (-j-1)\lambda\theta^2}{(-j-1)^{k/2} (j+1)} \right] \right]$$

If  $X \sim \text{HLRD}(\theta, \lambda)$ , Moment Generating Function is given by

$$M_x(t) = E(e^{tx}) = \int_{x=0}^{\infty} e^{tx} \cdot f(x) dx \tag{17}$$

$$= \int_0^{\infty} e^{tx} \cdot \frac{4\lambda(x-\theta) \cdot e^{-\lambda(x-\theta)^2}}{[1+e^{-\lambda(x-\theta)^2}]^2} dx \tag{18}$$

$$= \int_0^{\infty} \frac{e^{tx} \cdot e^{-\lambda(x-\theta)^2}}{[1+e^{-\lambda(x-\theta)^2}]^2} dx \tag{19}$$

$$= \int_{1+e^{-\lambda\theta^2}}^1 e^{tx} \left( \frac{-dt}{t^2} \right) \tag{20}$$

$$= - \int_{1+e^{-\lambda\theta^2}}^1 t^{-2} \cdot e^{t\left[\theta + \frac{1}{\sqrt{\lambda}}\sqrt{\log(-t)}\right]} dt \tag{21}$$

$$= \int_1^{1+e^{-\lambda\theta^2}} t^{-2} \cdot e^{t\left[\theta + \frac{1}{\sqrt{\lambda}}\sqrt{\log(-t)}\right]} dt \tag{22}$$

$$e^{t\left[\frac{1}{\sqrt{\lambda}}\sqrt{\log(-t)}\right]} = \sum_{i=0}^{\infty} \frac{t^i \left(\frac{1}{\sqrt{\lambda}}\sqrt{\log(-t)}\right)^i}{i!} \tag{23}$$

$$e^{t\theta} = \sum_{j=0}^{\infty} \frac{(t\theta)^j}{j!} \tag{24}$$

Substitute (23) and(24) in(22)

$$= \int_1^{1+e^{-\lambda\theta^2}} t^{-2} \sum_{i=0}^{\infty} \frac{t^i \left(\frac{1}{\sqrt{\lambda}}\sqrt{\log(-t)}\right)^i}{i!} \sum_{j=0}^{\infty} \frac{(t\theta)^j}{j!} dt \tag{25}$$

$$= \sum_{i=0}^{\infty} \sum_{j=0}^{\infty} \frac{t^i \left(\frac{1}{\sqrt{\lambda}}\right)^i}{i!} \cdot \frac{(t\theta)^j}{j!} \int_{\lambda\theta^2}^0 t^{-2} \left(\sqrt{\log(-t)}\right)^i dt \tag{26}$$

$$= \sum_{i=0}^{\infty} \sum_{j=0}^{\infty} \frac{(t\theta)^j}{t^i (\sqrt{\lambda})^i i! j!} \left[ \left[ \frac{\Gamma\left(\frac{i}{2}+1\right), 0}{(-j-1)^{k/2} (j+1)} \right] - \left[ \frac{\Gamma\left(\frac{i}{2}+1\right), (-j-1)\lambda\theta^2}{(-j-1)^{k/2} (j+1)} \right] \right] \dots(27)$$

$$M_x(t) = \sum_{i=0}^{\infty} \sum_{j=0}^{\infty} \frac{(t\theta)^j}{t^i (\sqrt{\lambda})^i i! j!} \left[ \left[ \frac{\Gamma\left(\frac{i}{2}+1\right), 0}{(-j-1)^{k/2} (j+1)} \right] - \left[ \frac{\Gamma\left(\frac{i}{2}+1\right), (-j-1)\lambda\theta^2}{(-j-1)^{k/2} (j+1)} \right] \right]$$

**5.3. Cumulative Generating Function**

If  $X \sim \text{HLRD}(\theta, \lambda)$  Cumulative Generating Function is given by

$$K_x(t) = \log M_x(t)$$

$$= \log \left( \sum_{i=0}^{\infty} \sum_{j=0}^{\infty} \frac{(t\theta)^j}{t^i (\sqrt{\lambda})^i i! j!} \left[ \left[ \frac{\Gamma\left(\frac{i}{2}+1\right), 0}{(-j-1)^{k/2} (j+1)} \right] - \left[ \frac{\Gamma\left(\frac{i}{2}+1\right), (-j-1)\lambda\theta^2}{(-j-1)^{k/2} (j+1)} \right] \right] \right) \dots(28)$$

**6 Estimation of parameters of HLRD**

**6.1. Maximum Likelihood method of estimation**

Let  $x_1, x_2, \dots, x_n$  be a random sample of size  $n$  from  $\text{HLRD}(\theta, \lambda)$ , then the likelihood function  $L$  of this sample is defined as

$$\ln = \ln \left( \prod_{i=1}^n (x_i; \theta, \lambda) \right) \\ = \ln \prod_{i=1}^n \left[ \frac{4\lambda(x_i - \theta)e^{-\lambda(x_i - \theta)^2}}{\left[ 1 + e^{-\lambda(x_i - \theta)^2} \right]^2} \right] \dots(29)$$

$$\ln = n \ln 4\lambda + n \sum_{i=1}^n \ln(x_i - \theta) - \lambda \sum_{i=1}^n (x_i - \theta)^2 + 4\lambda \sum_{i=1}^n \ln(x_i - \theta)^2 \dots(30)$$

Calculating the 1<sup>st</sup> and 2<sup>nd</sup> order partial derivative of (31) with respect to  $(\theta, \lambda)$  and then 1<sup>st</sup> order partial derivatives equating to zero we get the following equations

$$\frac{\partial \ln}{\partial \theta} = 4\lambda \sum (x_i - \theta) - \sum \frac{1}{(x_i - \theta)} [8\lambda + n] \dots(31)$$

$$\frac{\partial \ln}{\partial \lambda} = \frac{n}{4\lambda} + 8 \sum_{i=1}^n \ln(x_i - \theta)^2 - \sum_{i=1}^n (x_i - \theta)^2 \dots(32)$$

$$\frac{\partial \ln}{\partial \lambda} = 0$$

$$\frac{n}{4\lambda} = \sum_{i=1}^n (x_i - \theta)^2 - 4 \sum_{i=1}^n \ln(x_i - \theta)^2$$

$$\hat{\lambda} = \frac{n}{4 \left[ \sum_{i=1}^n (x_i - \theta)^2 - 2 \sum_{i=1}^n \ln(x_i - \theta)^2 \right]} \quad \dots (33)$$

2<sup>nd</sup> order partial derivative is given by

$$\begin{aligned} \frac{\partial \ln}{\partial \theta \partial \lambda} &= 4 \sum (x_i - \theta) - \sum \frac{8}{(x_i - \theta)} \\ &= \frac{4 \left[ \sum (x_i - \theta) \right]^2 - 8 \sum (x_i - \theta)}{\sum (x_i - \theta)} \quad \dots (34) \end{aligned}$$

## 6.2. Asymptotic Confidence bounds:

Here we derive the asymptotic confidence bounds for unknown parameters  $\theta$ ,  $\lambda$  when  $\theta < x$ ,

$\lambda > 0$  The simplest large sample approach is to assume that the MLEs  $(\theta, \lambda)$  are approximately normal with mean  $(\theta, \lambda)$  and covariance matrix  $I_0^{-1}$ , where  $I_0^{-1}$  is the inverse of the observed information matrix which defined as follows

$$I_0^{-1} = \begin{bmatrix} \frac{d^2 \ln}{d\theta} & \frac{d^2 \ln}{d\theta d\lambda} \\ \frac{d^2 \ln}{d\theta d\lambda} & \frac{d^2 \ln}{d\lambda} \end{bmatrix}$$

$$\frac{\partial^2 \ln}{\partial \theta^2} = -4\lambda \sum (x_i - \theta) + \sum_{i=1}^n (x_i - \theta)^{-2} [8\lambda + n] \quad \dots (35)$$

$$\frac{\partial^2 \ln}{\partial \lambda^2} = \frac{-n}{4\lambda^2} \quad \dots (36)$$

## Conclusions:

In this paper, we introduced two parameter Half Logistic Rayleigh Distribution (HLRD). Some characteristics of new distribution, such as ordinary moments and generating functions are obtained.

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