The T-R \{Generalized Lambda V\} Families of Distributions

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Abstract

The four-parameter generalized lambda distribution (GLD) was proposed in [1]. We say the GLD is of type V, if the quantile function corresponds to case (v) in [2], that is,

\[ Q(u; a, b) = \frac{1 - (1 - u)^v}{b} \]

where \( u \in (0, 1) \) and \( a, b \in (0, 1) \). In this short note, we introduce the T-R \{Generalized Lambda V\} Families of Distributions and show a sub-model of this class of distributions is significant in modeling real life data, in particular the Wheaton river data, [2]. We conjecture the new class of distributions can be used to fit biological and health data.

Keywords and Phrases: Generalized Lambda Distribution; Exponential Distribution; Weibull Distribution

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The T – R \{Y\} Family of Distributions

This family of distributions was proposed in [3]. In particular, let \( T, R, Y \) be random variables with CDF’s \( FT(x) = P(T \leq x) \), \( FR(x) = P(R \leq x) \), and \( FY(x) = P(Y \leq x) \), respectively. Let the corresponding quantile functions be denoted by \( QT(p) \), \( QR(p) \), and \( QY(p) \), respectively. Also, if the densities exist, let the corresponding PDF’s be denoted by \( fT(x) \), \( fR(x) \), and \( fY(x) \), respectively. Following this notation, the, the CDF of the T – R \{Y\} is given by

\[ F_X(x) = \int_{a}^{Q_Y(F_R(x))} f_T(t)dt = F_T(Q_Y(F_R(x))) \]

and the PDF of the T-R \{Y\} family is given by

\[ f_X(x) = \frac{f_T(x)}{f_T(Q_Y(F_R(x)))} f_Y(Q_Y(F_R(x))) \]

The New Distribution

Theorem: The CDF of the T-R \{Generalized Lambda V\} Families of Distributions is given by

\[ F_X(x) = \frac{1 - (1 - F_R(x))^v}{b} \]

where the random variable \( R \) has CDF \( FR(x) \), the random variable \( T \) with support \((0,1)\) has CDF \( FT \), and \( a, b \in (0, 1) \). Proof.

Consider the integral

\[ F_X(x) = \int_{a}^{Q_Y(F_R(x))} f_T(t)dt = F_T(Q_Y(F_R(x))) \]

and let \( Y \) follow the generalized lambda class of distributions of type \( V \), where the quantile is as stated in the abstract

Remark: the PDF can be obtained by differentiating the CDF

Practical Significance

In this section, we show a sub-model of the new distribution defined in the previous section is significant in modeling real life data. We assume \( T \) is standard exponential so that \( FT(t) = 1 - e^{-t}, t > 0 \) and \( R \) follows the two-parameter Weibull distribution, so that
\[ F_R(x) = 1 - e^{-\left(\frac{x}{c}\right)^d}, \quad x \geq 0; c, d > 0 \]

Now from Theorem 2.1, we have the following

**Theorem:** The CDF of the Standard Exponential-Weibull (Generalized Lambda V) Families of Distributions is given by

\[ F_{a,b,c,d}(x) = 1 - e^{-\left(\frac{x}{c}\right)^d} \]

where \(c, d > 0\) and \(a, b \in (-1, 0)\)

By differentiating the CDF, we obtain the following

**Theorem:** the PDF of the Standard Exponential-Weibull (Generalized Lambda V) Families of Distributions is given by

\[ f_{a,b,c,d}(x) = \frac{ace}{bd} \left(\frac{x}{c}\right)^{a-1} \left(\frac{e^{-\left(\frac{x}{c}\right)^d}}{d}\right)^{-1} \]

**Remark:** If a random variable \(B\) follows the Standard Exponential-Weibull (Generalized Lambda V) Families of Distributions write \(B \sim \text{SEWGLV}(a, b, c, d)\)

**Open Problem**

**Conjecture:** The new class of distributions can be used in forecasting and modeling of biological and health data. Related to the above conjecture is the following

**Question:** Is there a sub-model of the T-R (Generalized Lambda V) Families of Distributions that can fit? [3] (Appendix 1) and (Figures 1–3).

**Appendix 1:**

Data 11 = \{1.7, 1.4, 0.6, 9.0, 5.6, 1.5, 2.2, 18.7, 2.2, 1.7, 30.8, 2.5, 14.4, 8.5, 39.0, 7.0, 13.3, 27.4, 1.1, 25.5, 0.3, 20.1, 4.2, 1.0, 0.4, 11.6, 15.0, 0.4, 25.5, 27.1, 20.6, 14.1, 11.0, 2.8, 3.4, 20.2, 5.3, 22.1, 7.3, 14.1, 11.9, 16.8, 0.7, 1.1, 22.9, 9.9, 21.5, 5.3, 1.9, 2.5, 1.7, 10.4, 27.6, 9.7, 13.0, 14.4, 0.1, 10.7, 36.4, 27.5, 12.0, 1.7, 1.1, 30.0, 2.7, 2.5, 9.3, 37.6, 0.6, 3.6, 64.0, 27.0\}

DataQ1 = Flatten Data11; Length DataQ1 72

Min DataQ1 0.1; Max DataQ1 64.

AX1 = Empirical Distribution DataQ1

Data Distribution «Empirical», { 58 }

K1 = Discrete Plot [CDF [AX1, x], \{x, 0, 65, (65-0)/58\}, Plot Style c {Black, Thick}, Plot Markers c {Automatic, Small}, Filling c None, Plot Range c All]

(Figure 1)

\[ F_1 x_, a_, b: = 1 - 1 - x ^ ab \]

CDF Weibull Distribution c, d, x

\[ \{10 - \frac{x}{c} X > 0 \quad \text{True} \]

\[ M x_, c_,d := 1 - E^x \cdot d \cdot c \cdot M x, c, d \]

\[ 1 - \frac{|\frac{X}{d}| ^ a}{b} \]

1. **Weibull. nb**

\[ D 1 - E^ - F1 M x, c, d, a, b, x \]

\[ ac & - \frac{1 - |\frac{X}{d}| ^ a}{b} (\frac{X}{d}) ^{\alpha ^{-1} c} \]

\[ MQ x_a b_, c_, d := a c E^ - 1 - E^x d ^ c ^ a b E^x d ^ c ^ a x d ^ - 1 + c d b d M Q x, a, b, c, d \]
$$\frac{1 - \frac{x^{a}}{b} \left( 1 - \frac{x}{d} \right)^{-c}}{bd}$$

PLK1 = Sum Log MQ DataQ1, a, b, c, d, i, i, 1, Length DataQ1;
JK=D [PLK, a];
JK1=D [PLK, b];
JK2=D [PLK, c];
JK3=D [PLK, d];
Find Root JK, JK1, JK2, JK3, a, -0.11, b, -0.12, c, 0.9, d, 11.6
ac = 7.2577, bc = 0.395297, cc 0.776499, dc 657.998
RR = Plot 1 - E^ - F1 Mx, 0.776499, 657.998, - 7.2577, - 0.395297, x, 0, 65, Plot Style c Thick, Blue, Plot Range c All
(Figure 2)
II. Weibull. nb
CMU = Show K1, RR, Plot Range c All
(Figure 3)
Export "CMU.jpg", CMU CMU.jpg
References

