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## Properties of new derived power series distribution

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### Abstract

Jayasree and Bhatra charyulu (2017) [2] provided a new derived power series using the power series distributions of Kulasekera and Tonkyn (1992) [5] and Geometric distribution. In this paper, the new derived power series distribution moment estimators and maximum likelihood estimators are obtained and also shown that it belongs to the two-parameter exponential family of distributions.

**Keywords:** Derived Power Series Distribution (DPSD), JS-NB Distribution, Maximum likelihood estimator

### 1. Introduction

Suppose that  $\underline{a} = (a_0, a_1, a_2, \dots)$  is a sequence of nonnegative real numbers. The partial sum of order  $n \in \mathbb{N}$  is  $g_n(\theta) = \sum_{k=0}^n a_k \theta^k$ ,  $k=0, 1, \dots, n$  for all  $\theta \in \mathbb{R}$ . The power series is then defined by  $g(\theta) = \lim_{n \rightarrow \infty} g_n(\theta)$  for  $\theta \in \mathbb{R}$  for which the limit exists, and is denoted  $g_n(\theta) = \sum_{k=0}^n a_k \theta^k$ ,  $k=0, 1, \dots, \forall \theta \in \mathbb{R}$ . A random variable  $n$  with values in  $\mathbb{N}$  has the power series distribution associated with the function  $g$  (or equivalently with the sequence  $a$ ), with parameter  $\theta \in [0, r)$ , if  $N$  has discrete probability density function  $f_\theta(n) = a_n \theta^n / g(\theta)$ ,  $n \in \mathbb{N}$ .

Let  $P_1(s)$  and  $P_2(s)$ ;  $|s| \leq 1$  are the probability generating functions of two power series distributions. Jayasree and Swamy (2006) [2] defined a family of new power series distributions with the convolution of  $P_1(s)$  and  $[P_2(s)]^{-1}$  called Derived Power Series Distributions (DPSD). Consider a DPSD for which the probability mass function is given by

$$P[X=x] = [ \{a_0 g(\theta_2)\} / \{b_0 f(\theta_1)\} ] d_x ; x=1,2, \dots \dots (1.1)$$

where  $d_x = a'_x - \sum_{j=1}^x b'_j d_x$ ;  $x=1, 2, \dots$  with  $a'_x = [a_x(\theta_1)^x] / a_0$  and  $b'_j = [b_j(\theta_2)^j] / b_0$ ;  $\theta_1, \theta_2 > 0$

The mean and variance of derived power series distributions are  $\mu = \mu_1 - \mu_2$  and  $\sigma^2 = \sigma_1^2 - \sigma_2^2$  where  $\mu_1, \sigma_1^2$  are the mean and variance of  $P_1(s)$  and  $\mu_2, \sigma_2^2$  are the mean and variance of  $P_2(s)$  power series distributions. A power series distribution derived by Kulasekera and Tonkyn (1992) [5] is presented in definition 1.1.

**Definition 1.1:** A random variable  $X$  is said be Kulasekera and Tonkyn probability distribution if it satisfies the probability law  $P [X = x] = x^\alpha q^\alpha / (\sum x^\alpha q^\alpha)$ ;  $x = 1, 2, \dots ; \infty < \alpha < \infty$ ;  $0 < q < 1$ ;  $p+q = 1$ . For fixed value of  $\alpha$ , the distribution belongs to the family of power series distribution. The mean and variance of the distribution are  $(1+q) p^{-1}$  and  $2qp^{-2}$ .

A power series distribution derived by Jayasree and Bhatra Charyulu (2017) [2] were presented is like as follows. Let  $X_1$  be the random variable follows a power series probability distributions of Kulasekera and Tonkyn (1992) [5], and  $X_2$  be the random variable follows Geometric distribution. Let  $P_1(s)$  and  $P_2(s)$  are the probability generating functions of the two power series distributions in  $s$  and convergent for  $|s| \leq 1$ ,

$$P_1(s) = b^{-1} \sum_{x=1}^{\infty} x q_1^x s^x \text{ and } P_2(s) = p_2 \sum_{y=1}^{\infty} q_2^{y-1} s^y ; |s| \leq 1 \quad (1.2)$$

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Where  $P [X_1 = x_1] = x_1^\alpha q_1^{x_1} / (\sum_{x=1}^{\infty} x^\alpha q_1^x)$ ; and  $P [X_2 = x_2] = p_2 q_2^{x_2-1}$ ;  $x_2 = 1, 2, \dots$ . The convolution of  $P_1(s)$  and  $[P_2(s)]^{-1}$  provides a power series distribution with random variable  $X$  satisfies the probability law

$$P[X=x] = q_1(bp_2)^{-1} [(x+1)(q_1)^x - x(q_1)^{x-1}q_2]; \quad (1.3)$$

$0 < p_1 < p_2 < 1$ , such that  $p_1 + q_1 = 1$  and  $p_2 + q_2 = 1$ .

Under appropriate conditions for the coefficients of the power series  $P(s)$  to be non-negative, one can consider this as the probability generating function of a random variable, where from, the p.m.f. can be obtained by identifying the coefficients in  $P(s)$ . The  $P(s)$  is a proper Power series probability distribution for specified vector of probabilities with The mean  $(1+q_1) p_1^{-1} - p_2^{-1}$  and variance  $\sigma^2 = 2q_1p_1^{-2} - q_2p_2^{-2}$ .

**2. Properties of the Power Series Distribution**

Some properties of new derived power series distribution of Jayasree and Bhatra Charyulu (2017) [2] are studied and presented below.

**Theorem 2.1:** If  $u_1(u_1-1)^{-1} > c^2$  then the moment estimators of  $p_1$  and  $p_2$  are given by

$$\hat{p}_1 = 2 [u_1 + u_2 \{-3 u_1 + u_3\}^{-1}] \quad (2.1)$$

and  $\hat{p}_2 = u_2^{-1} \{-3u_1 + u_3\}$  (2.2)

where  $u_1 = \mu + 1$ ,  $u_2 = 2(u_1^2 - 2u_1 - 2\sigma^2)$ ,  $u_3 = (13u_1^2 - 8u_1 - 8\sigma^2)^{1/2}$ , and  $c = \sigma\mu^{-1}$  which is the coefficient of variation of the distribution.

**Proof:** The moment equations for the estimators are given by

$$\mu = 2p_1^{-1} - 1 - p_2^{-1} \quad (2.3)$$

$$\sigma^2 = p_2^{-1}(1 - p_2^{-1}) - 2 p_1^{-1} (1 - p_1^{-1}) \quad (2.4)$$

The solution of these equations yields  $\hat{p}_2 = u_2^{-1} \{-3u_1 + u_3\}$ . Substituting  $\hat{p}_2$  in (2.3), one can obtain (2.4).

Further  $\hat{p}_2$  is real if  $13u_1^2 - 8u_1 - 8\sigma^2 > 0$ . Equivalently, if  $\mu + \mu^2 > \sigma^2$ . In terms of  $u_1$  and  $c$ , the moment estimators are given by (2.3) and (2.4) if  $u_1(u_1-1)^{-1} > c^2$ .

**Theorem 2.2:** The derived power series distribution belongs to the two-parameter exponential family of distributions, if the  $k^{th}$  power of  $\omega(x; q_1, q_2)$  is negligible for  $k \geq 3$ , where  $\omega(x; q_1, q_2) = |x^{-1} - q_1^{-1} q_2|$ .

**Proof:** The probability mass function of the new Derived Power series distribution specified in (1.1) can be expressed as  $P_x = \exp[x \log_e q_1 + \log_e x + \log_e \{ (1-q_1)^2 (1-q_2)^{-1} \} + \log_e \{ 1+(x^{-1}-q_2q_1^{-1}) \}]$ ,  $x=1,2,\dots$  (2.5)

When the  $k$  th power of  $\omega(x; q_1, q_2) = |x^{-1} - q_1^{-1} q_2|$  is negligible for  $k \geq 3$ , considering  $\log_e \{ 1+(x^{-1}-q_2q_1^{-1}) \}$ , one has  $P_x = \exp[x \log_e q_1 + \log_e x + q_2 (xq_1)^{-1+x^{-1}} - (2x^2)^{-1} + \lambda(q_1, q_2)]$  (2.6)

where  $\lambda(q_1, q_2) = \log_e \{ (1-q_1)^2 (1-q_2)^{-1} \} - q_2 q_1^{-1} - q_2^2 (2q_1^2)^{-1}$

Let  $\theta_1 = \log_e q_1$ ;  $\theta_2 = q_2 q_1^{-1}$ , so that  $q_1 = e^{\theta_1}$  and  $q_2 = \theta_2 e^{\theta_1}$  then by the re-parameterization, one has,  $P_x = (1 - \theta_2 e^{\theta_1})^{-1} x (1 - e^{\theta_1})^2 \exp [\theta_1 x + \theta_2 x^{-1+x^{-1}} - (2x^2)^{-1} - \theta_2 - 2^{-1} \theta_2^2]$ ;  $x=1,2,\dots$  (2.7)

From (2.7), we can obtain maximum likelihood estimators for the parameters  $\theta_1$  and  $\theta_2$ .

**Theorem 2.3:** If the  $\max_{1 \leq r \leq n} \omega^k(x_r; q_1, q_2)$  is negligible for  $k \geq 3$ , the maximum likelihood equation estimators of  $q_1$  and  $q_2$  are given by the solutions of the equations

$$A = [ e^{\theta_1} \{ (1-\theta_2) + (1-\theta_2 e^{\theta_1}) \} ] [ (1-e^{\theta_1}) (1-\theta_2 e^{\theta_1}) ]^{-1} \text{ and } H^{-1} = (1+\theta_2) - \{ e^{\theta_1} (1-\theta_2 e^{\theta_1})^{-1} \} \quad (2.8)$$

**Proof** From (1.1) the likelihood based on the sample  $\mathbf{x} = (x_1, x_2, \dots, x_n)$  is given by

$$L(x; \theta_1, \theta_2) = \exp [\theta_1 \sum_{i=1}^n x_i + \theta_2 \sum_{i=1}^n x_i^{-1} + \sum_{i=1}^n H(x_i) + n \lambda(\theta_1, \theta_2)]$$

where  $H(x) = \log x + x^{-1} - (2x^2)^{-1}$  and  $\lambda(\theta_1, \theta_2) = 2\log(1 - e^{\theta_1}) - \log(1 - \theta_2 e^{\theta_1}) - \theta_2 - (\theta_2^2)/2$

Thus, the maximum likelihood equations are given by

$$A = [ e^{\theta_1} \{ (1-\theta_2) + (1-\theta_2 e^{\theta_1}) \} ] [ (1 - e^{\theta_1}) (1 - \theta_2 e^{\theta_1}) ]^{-1}$$

$$\text{and } H^{-1} = (1+\theta_2) - \{ e^{\theta_1} (1 - \theta_2 e^{\theta_1})^{-1} \}$$

where,  $A$  is the arithmetic mean and  $H$  is the Harmonic mean of the sample. Since the solution of these equations seem to be difficult, one can adopt an iterative procedure like the Newton-Raphson, to obtain the same.

**Example 2.1:** Consider distribution of spiders among patches of habitat or prey data presented in table 2.1. The fitting of derived distribution is presented below.

**Table 2.1:** Frequency Distribution

$x$	0	1	2	3	4	5
$f(x)$	85	53	30	16	8	8

When a distribution is fit, using the method of moments for estimating the parameters one can obtain the estimates of  $\hat{p}_1 = 0.6277$ ,  $\hat{p}_2 = 0.9792$ ,  $\hat{q}_1 = 0.3723$  and  $\hat{q}_2 = 0.0208$  and the expected frequencies are obtained as 0.1975, 0.4310, 0.125, 0, 0.1428 and 0.6666, so that the observed value of  $\chi^2$  is given by  $\chi^2 = 1.5629$ .

Note: It can be noted that the derived power series distribution of Swamy and Jayasree (2006) SJS-4 is also an equally good fit to the same data.

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