

# A New Two-Parameter Poisson-Sujatha Distribution

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**Abstract** A new two-parameter Poisson-Sujatha distribution, a Poisson mixture of a new two-parameter Sujatha distribution, which includes Poisson-Sujatha distribution and Poisson-Akash distribution as particular cases, has been introduced. Its moments based statistical measures including coefficient of variation, skewness, kurtosis and index of dispersion have been obtained. Maximum likelihood estimation has been explained for estimating its parameters. Goodness of fit of the proposed distribution has been explained with six over-dispersed count datasets and fit has been compared with Poisson-Sujatha distribution and other generalizations of Poisson-Sujatha distributions.

**Keywords** Sujatha distribution, Poisson-Sujatha distribution, A New two-parameter Sujatha distribution, Moments based measures, Maximum likelihood estimation, Applications

## 1. Introduction

Poisson distribution is the common distribution for modeling count data when the mean and the variance of the data are the same (equi-dispersed). However, the unique feature of equality of equi-dispersion of Poisson distribution makes it unsuitable for modeling count data which are under-dispersed (mean greater than variance) or over-dispersed (mean less than variance). In recent years, several researchers have proposed Poisson mixture of lifetime distributions which are useful for over-dispersed or under-dispersed. Some over-dispersed Poisson mixed distributions are Poisson-Lindley distribution (PLD), a Poisson mixture of Lindley distribution of Lindley (1958) proposed by Sankaran (1970), Poisson-Sujatha distribution (PSD), a Poisson mixture of Sujatha distribution of Shanker (2016 a) introduced by Shanker (2016 b), Poisson-Akash distribution of Shanker (2017), some among others.

The probability density function (pdf) of Sujatha distribution having scale parameter  $\theta$  and introduced by Shanker (2016a) is

$$f_1(x; \theta) = \frac{\theta^3}{\theta^2 + \theta + 2} (1 + x + x^2) e^{-\theta x} \quad ; x > 0, \theta > 0 \quad (1.1)$$

Statistical properties including shapes of the density, moments and moments based measures, skewness, kurtosis, index of dispersion, hazard rate function, mean residual life

function, stochastic ordering, mean deviation, stress-strength reliability, along with the estimation of parameter and applications for modeling lifetime data from biomedical science and engineering of Sujatha distribution are available in Shanker (2016 a). Kaliraja and Perarasan (2019) studied a stochastic model on the generalization of Sujatha distribution for the effects of two types of exercise on plasma growth hormone.

Shanker (2016 b) obtained Poisson-Sujatha distribution (PSD) by compounding Poisson distribution with Sujatha distribution. The PSD is defined by its probability mass function (pmf)

$$P_1(x; \theta) = \frac{\theta^3}{\theta^2 + \theta + 2} \frac{x^2 + (\theta + 4)x + (\theta^2 + 3\theta + 4)}{(\theta + 1)^{x+3}} \quad ; \quad (1.2)$$

$$x = 0, 1, 2, \dots, \theta > 0$$

Statistical properties including shapes of pmf, moments and moments based measures, over-dispersion, unimodality and increasing hazard rate, estimation of parameters and applications of model over-dispersed data have been discussed by Shanker (2016 b). Wesley et al (2018) proposed a zero-modified Poisson-Sujatha distribution to model over-dispersed count data and discussed its several important properties and applications.

Shanker *et al* (2017) have introduced a generalization of Sujatha distribution (AGSD) having pdf and cdf given by

$$f_2(x; \theta, \alpha) = \frac{\theta^3}{\theta^2 + \theta + 2\alpha} (1 + x + \alpha x^2) e^{-\theta x} \quad ; \quad (1.3)$$

$$x > 0, \theta > 0, \alpha > 0$$

Shanker *et al* (2017) have discussed important statistical properties including shapes of the density, moments and

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moments based measures, hazard rate function, mean residual life function, stochastic ordering, mean deviations, Bonferroni and Lorenz curves, stress-strength reliability, along with estimation of parameters using maximum likelihood estimation and applications of AGSD for modeling lifetime data from engineering and medical sciences. It can be easily verified that at  $\alpha = 1$ , the pdf of AGSD reduces to the corresponding pdf of Sujatha distribution. Also, at  $\alpha = 0$ , the pdf of AGSD reduces to Lindley distribution introduced by Lindley (1958).

Shanker and Shukla (2019) introduced a generalization of Poisson-Sujatha distribution (AGPSD) by compounding Poisson distribution with AGSD (1.3) and obtained the pmf in the form

$$P_2(x; \theta, \alpha) = \frac{\theta^3}{\theta^2 + \theta + 2\alpha} \frac{\alpha x^2 + (\theta + 3\alpha + 1)x + (\theta^2 + 3\theta + 2\alpha + 2)}{(\theta + 1)^{x+3}}, \quad (1.4)$$

$$x = 0, 1, 2, \dots, \theta > 0, \alpha > 0$$

Various properties, estimation of parameters, and applications of AGPSD have been discussed by Shanker and Shukla (2019). Further, Poisson-Lindley distribution (PLD) of Sankaran (1970) and Poisson-Sujatha distribution of Shanker (2016 b) are particular cases of AGPSD.

Shanker and Shukla (2020 a) proposed a two-parameter Poisson-Sujatha distribution (TPPSD) defined by its pmf

$$P_3(x; \theta, \alpha) = \frac{\theta^3}{(\alpha\theta^2 + \theta + 2)} \frac{x^2 + (\theta + 4)x + \{\alpha\theta^2 + (2\alpha + 1)\theta + (\alpha + 3)\}}{(\theta + 1)^{x+3}}, \quad (1.5)$$

$$x = 0, 1, 2, \dots, (\theta, \alpha) > 0.$$

Important statistical properties, estimation of parameters using maximum likelihood estimation, applications and the superiority of TPPSD over other one parameter and two-parameter discrete distributions have been discussed by Shanker and Shukla (2020 a). It should be noted that TPPSD is a Poisson mixture of a two-parameter Sujatha distribution (TPSD) introduced by Mussie and Shanker (2018) and defined by its pdf

$$f_3(x; \theta, \alpha) = \frac{\theta^3}{\alpha\theta^2 + \theta + 2} (\alpha + x + x^2) e^{-\theta x}; \quad (1.6)$$

$$x > 0, \theta > 0, \alpha \geq 0$$

It can be easily verified that one parameter Sujatha distribution is a particular case of NTPSD for  $\alpha = 1$ . Its moments and moments based measures including skewness, kurtosis, index of dispersion; hazard rate function, mean residual life function, stochastic ordering, mean deviations, Bonferroni and Lorenz curves, stress-strength reliability the estimation of the parameters using methods of moments and method of maximum likelihood and superiority over exponential, Lindley, Akash and Sujatha distributions have been explained in Mussie and Shanker (2018).

Recently, Mussie and Shanker (2019) proposed another

two-parameter Sujatha distribution (ATPSD) defined by its pdf

$$f_4(x; \theta, \alpha) = \frac{\theta^3}{\theta^2 + \alpha\theta + 2\alpha} (1 + \alpha x + \alpha x^2) e^{-\theta x}; \quad (1.6)$$

$$x > 0, \theta > 0, \alpha \geq 0$$

where  $\theta$  is a scale parameter and  $\alpha$  is a shape parameter. It can be easily verified that (1.3) reduces to exponential distribution and Sujatha distribution for  $\alpha = 0$  and  $\alpha = 1$  respectively. Its moments and moments based measures including skewness, kurtosis, index of dispersion; hazard rate function, mean residual life function, stochastic ordering, mean deviations, Bonferroni and Lorenz curves, stress-strength reliability the estimation of the parameters using methods of moments and method of maximum likelihood and superiority over exponential, Lindley and Sujatha distributions have been explained in Mussie and Shanker (2019). Shanker and Shukla (2020 b) suggested another two-parameter Poisson-Sujatha distribution (ATPPSD) by compounding Poisson distribution with another two-parameter Poisson-Sujatha distribution

$$P_4(x; \theta, \alpha) = \frac{\theta^3}{(\theta^2 + \alpha\theta + 2\alpha)} \frac{\alpha x^2 + \alpha(\theta + 4)x + \{\theta^2 + (\alpha + 2)\theta + (3\alpha + 1)\}}{(\theta + 1)^{x+3}}, \quad (1.7)$$

$$x = 0, 1, 2, \dots, \theta > 0, \alpha \geq 0$$

Important statistical properties, estimation of parameters using maximum likelihood estimation, applications and the superiority of ATPPSD over other one parameter and two-parameter discrete distributions have been discussed by Shanker and Shukla (2020 b).

Mussie and Shanker (2018) proposed a new two-parameter Sujatha distribution (NTPSD) defined by its pdf

$$f_5(x; \theta, \alpha) = \frac{\theta^3}{\theta^2 + \alpha\theta + 2} (1 + \alpha x + x^2) e^{-\theta x}; \quad (1.8)$$

$$x > 0, \theta > 0, \alpha \geq 0$$

It can be easily verified that one parameter Akash distribution of Shanker (2015) and Sujatha distribution are particular cases of NTPSD for  $\alpha = 0$  and  $\alpha = 1$ , respectively. Its moments and moments based measures including skewness, kurtosis, index of dispersion; hazard rate function, mean residual life function, stochastic ordering, mean deviations, Bonferroni and Lorenz curves, stress-strength reliability the estimation of the parameters using methods of moments and method of maximum likelihood and superiority over exponential, Lindley, Akash and Sujatha distributions have been explained in Mussie and Shanker (2018).

The main motivation for proposing ATPPSD are (i) Sujatha distribution is a better model than both exponential and Lindley distribution for modeling lifetime data, and PSD being a Poisson mixture of Sujatha distribution gives better fit than both Poisson and Poisson-Lindley distribution (PLD),

(ii) NTPSD gives much better fit than exponential, Lindley and Sujatha distribution, NTPPSD being a Poisson mixture of NTPSD provides better fit over PSD and other discrete distributions, and (iii) have a comparative study of NTPPSD with other two-parameter generalizations of Poisson-Sujatha distributions including AGPSD, TPPSD, and ATPPSD. Keeping these points in mind, a new two-parameter Poisson-Sujatha distribution (NTPPSD), a Poisson mixture of NTPSD has been proposed and its moments and moments based measures have been obtained and their behaviors have been studied. Maximum likelihood estimation of NTPPSD has been discussed for the estimation its parameters and its applications have been discussed with six examples of observed count datasets from various fields of knowledge.

## 2. A New Two-Parameter Poisson-Sujatha Distribution

A random variable  $X$  is said to follows a new two-parameter Poisson-Sujatha distribution (NTPPSD) if

$$X | \lambda \sim P(\lambda) \text{ and } \lambda | \theta, \alpha \sim NTPSD(\theta, \alpha).$$

That is,

$$P(X = x | \lambda) = \frac{e^{-\lambda} \lambda^x}{x!}; x = 0, 1, 2, \dots, \lambda > 0, \text{ and}$$

$$f(\lambda | \theta, \alpha) = \frac{\theta^3}{\theta^2 + \alpha\theta + 2} (1 + \alpha\lambda + \lambda^2) e^{-\theta\lambda};$$

$$\lambda > 0, \theta > 0, \alpha \geq 0$$

The pmf of unconditional random variable  $X$  can be obtained as

$$P_5(x; \theta, \alpha) = P(X = x) = \int_0^{\infty} P(X = x | \lambda) f(\lambda | \theta, \alpha) d\lambda$$

$$= \int_0^{\infty} \frac{e^{-\lambda} \lambda^x}{x!} \frac{\theta^3}{\theta^2 + \alpha\theta + 2} (1 + \alpha\lambda + \lambda^2) e^{-\theta\lambda} d\lambda \quad (2.1)$$

$$= \frac{\theta^3}{(\theta^2 + \alpha\theta + 2)} \int_0^{\infty} e^{-(\theta+1)\lambda} (\lambda^x + \alpha\lambda^{x+1} + \lambda^{x+2}) d\lambda$$

$$= \frac{\theta^3}{(\theta^2 + \alpha\theta + 2)} \left[ \frac{\Gamma(x+1)}{(\theta+1)^{x+1}} + \frac{\alpha\Gamma(x+2)}{(\theta+1)^{x+2}} + \frac{\Gamma(x+3)}{(\theta+1)^{x+3}} \right]$$

$$= \frac{\theta^3}{(\theta^2 + \alpha\theta + 2)} \frac{x^2 + (\alpha\theta + \alpha + 3)x + \{\theta^2 + (\alpha+2)\theta + (\alpha+3)\}}{(\theta+1)^{x+3}} \quad (2.2)$$

$$x = 0, 1, 2, \dots, \theta > 0, \alpha \geq 0$$

We would call this a new two-parameter Poisson-Sujatha distribution (NTPPSD) because for  $\alpha = 1$ , it reduces to one parameter PSD given in (1.2). Also at  $\alpha = 0$ , it reduces to Poisson-Akash distribution.

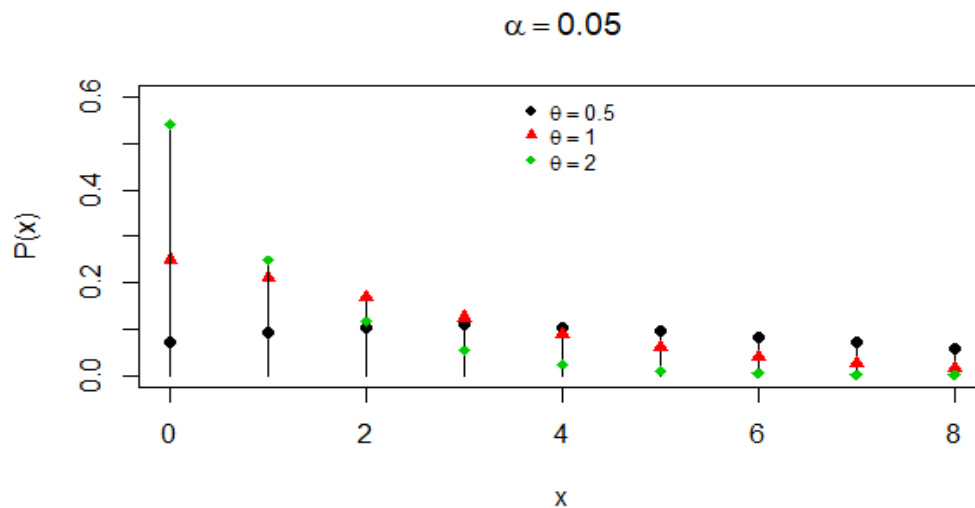
It can be easily shown that NTPPSD is unimodal and has increasing hazard rate. Since

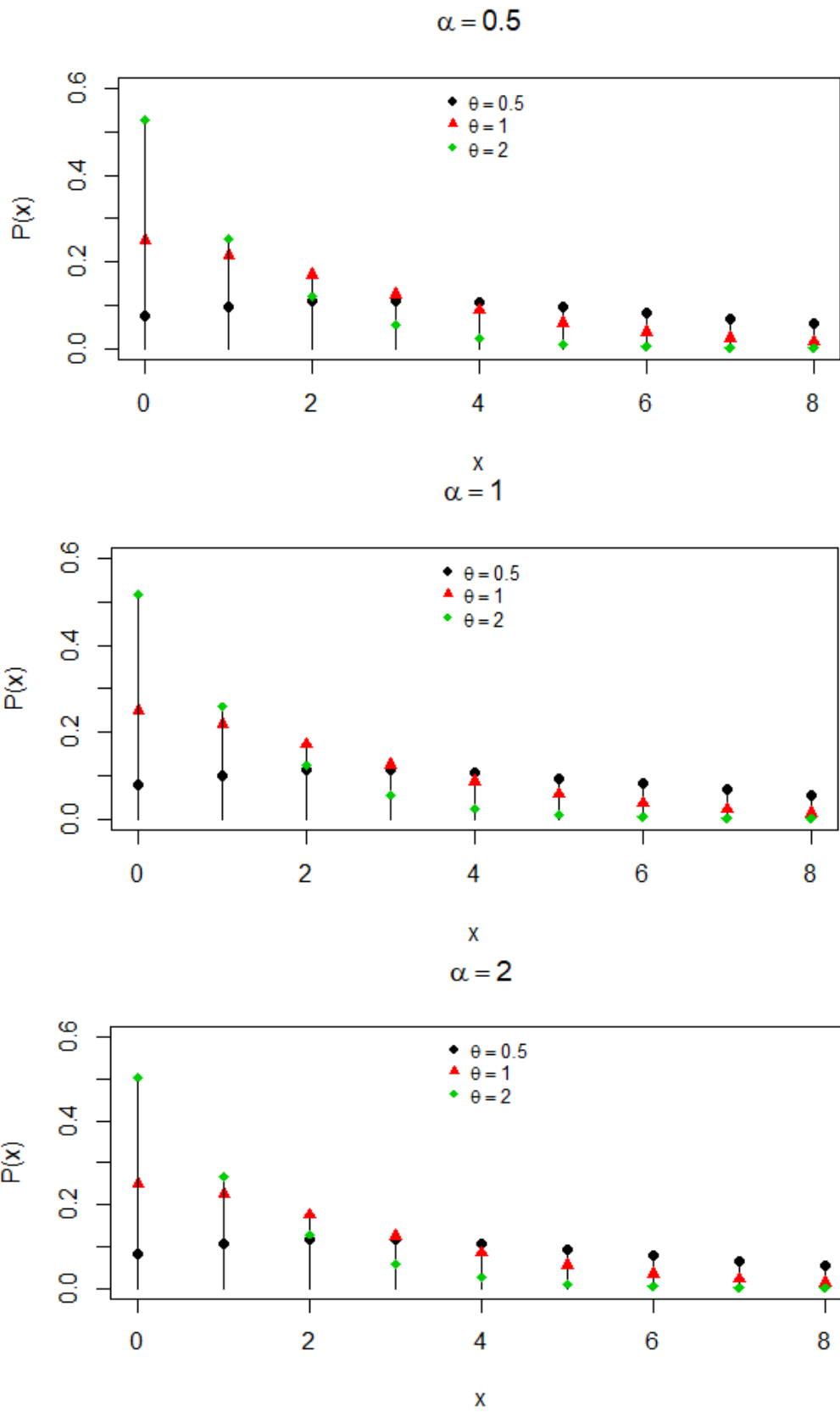
$$\frac{P_5(x+1; \theta, \alpha)}{P_5(x; \theta, \alpha)} = \frac{1}{\theta+1} \left[ 1 + \frac{2x + \alpha\theta + \alpha + 4}{x^2 + (\alpha\theta + \alpha + 3)x + \{\theta^2 + (\alpha+2)\theta + (\alpha+3)\}} \right]$$

is decreasing function in  $x$ ,  $P_5(x; \theta, \alpha)$  is log-concave.

Now using the results of relationship between log-concavity, unimodality and increasing hazard rate (IHR) of discrete distributions available in Grandell (1997), it can concluded that NTPPSD has an increasing hazard rate and unimodal.

The behavior of the pmf of NTPPSD for varying values of parameters  $\theta$  and  $\alpha$  are shown in figure 1.





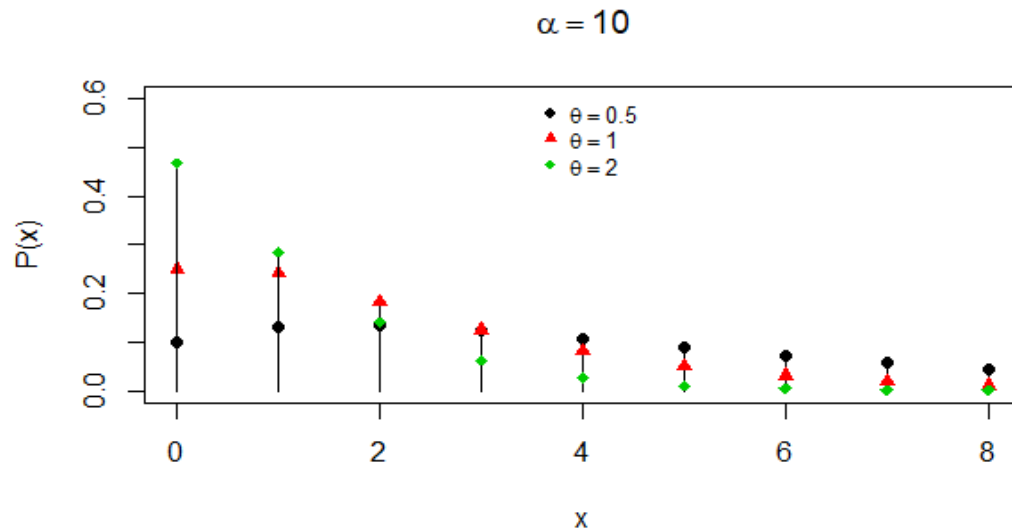


Figure 1. Behaviour of pmf of NTTPSD for varying values of parameters  $\theta$  and  $\alpha$

### 3. Moments Based Measures

The  $r$ th factorial moment about origin  $\mu_{(r)}'$  of NTTPSD can be obtained as

$\mu_{(r)}' = E\left[E\left(X^{(r)} \mid \lambda\right)\right]$ , where  $X^{(r)} = X(X-1)(X-2)\dots(X-r+1)$ . Using (2.1), the  $r$ th factorial moment about origin  $\mu_{(r)}'$  of NTTPSD can be obtained as

$$\begin{aligned}\mu_{(r)}' &= \frac{\theta^3}{\theta^2 + \alpha\theta + 2} \int_0^\infty \left[ \sum_{x=0}^\infty x^{(r)} \frac{e^{-\lambda} \lambda^x}{x!} \right] (1 + \alpha\lambda + \lambda^2) e^{-\theta\lambda} d\lambda \\ &= \frac{\theta^3}{\theta^2 + \alpha\theta + 2} \int_0^\infty \left[ \lambda^r \sum_{x=0}^\infty \frac{e^{-\lambda} \lambda^{x-r}}{(x-r)!} \right] (1 + \alpha\lambda + \lambda^2) e^{-\theta\lambda} d\lambda\end{aligned}$$

Taking  $x-r=y$  within the bracket, we get

$$\begin{aligned}\mu_{(r)}' &= \frac{\theta^3}{\theta^2 + \alpha\theta + 2} \int_0^\infty \left[ \lambda^r \sum_{y=0}^\infty \frac{e^{-\lambda} \lambda^y}{y!} \right] (1 + \alpha\lambda + \lambda^2) e^{-\theta\lambda} d\lambda \\ &= \frac{\theta^3}{\theta^2 + \alpha\theta + 2} \int_0^\infty \lambda^r (1 + \alpha\lambda + \lambda^2) e^{-\theta\lambda} d\lambda\end{aligned}$$

After some tedious algebraic simplification, a general expression for the  $r$ th factorial moment about origin  $\mu_{(r)}'$  of NTTPSD can be expressed as

$$\mu_{(r)}' = \frac{r! \left\{ \theta^2 + \alpha(r+1)\theta + (r+1)(r+2) \right\}}{\theta^r (\theta^2 + \alpha\theta + 2)}; r = 1, 2, 3, \dots \quad (3.1)$$

It can be easily verified that at  $\alpha = 0$  and  $\alpha = 1$ , the expression (3.1) reduces to the corresponding expression of PAD and PSD. Substituting  $r = 1, 2, 3$ , and 4 in (3.1), the first four factorial moments about origin of NTTPSD can be obtained as

$$\mu_{(1)}' = \frac{\theta^2 + 2\alpha\theta + 6}{\theta(\theta^2 + \alpha\theta + 2)}, \quad \mu_{(2)}' = \frac{2(\theta^2 + 3\alpha\theta + 12)}{\theta^2(\theta^2 + \alpha\theta + 2)}$$

$$\mu_{(3)}' = \frac{6(\theta^2 + 4\alpha\theta + 20)}{\theta^3(\theta^2 + \alpha\theta + 2)}, \mu_{(3)}' = \frac{24(\theta^2 + 5\alpha\theta + 30)}{\theta^4(\theta^2 + \alpha\theta + 2)}.$$

Now using the relationship between factorial moments about origin and moments about origin, the first four moment about origin of the NTPPSD are obtained as

$$\mu_1' = \frac{\theta^2 + 2\alpha\theta + 6}{\theta(\theta^2 + \alpha\theta + 2)}$$

$$\mu_2' = \frac{\theta^3 + (2\alpha + 2)\theta + (6\alpha + 6)\theta + 24}{\theta^2(\theta^2 + \alpha\theta + 2)}$$

$$\mu_3' = \frac{\theta^4 + (2\alpha + 6)\theta^3 + (18\alpha + 12)\theta^2 + (24\alpha + 72)\theta + 120}{\theta^3(\theta^2 + \alpha\theta + 2)}$$

$$\mu_4' = \frac{\theta^5 + (2\alpha + 14)\theta^4 + (42\alpha + 42)\theta^3 + (144\alpha + 192)\theta^2 + (120\alpha + 720)\theta + 720}{\theta^4(\theta^2 + \alpha\theta + 2)}$$

Using the relationship between moments about mean and the moments about origin, the moments about mean of NTPPSD are obtained as

$$\mu_2 = \frac{\theta^5 + (3\alpha + 1)\theta^4 + (2\alpha^2 + 4\alpha + 8)\theta^3 + (2\alpha^2 + 10\alpha + 16)\theta^2 + (12\alpha + 12)\theta + 12}{\theta^2(\theta^2 + \alpha\theta + 2)^2}$$

$$\mu_3 = \frac{\left\{ \begin{aligned} &\theta^8 + (4\alpha + 3)\theta^7 + (5\alpha^2 + 15\alpha + 12)\theta^6 + (2\alpha^3 + 18\alpha^2 + 36\alpha + 54)\theta^5 \\ &+ (6\alpha^3 + 26\alpha^2 + 108\alpha + 88)\theta^4 + (4\alpha^3 + 48\alpha^2 + 116\alpha + 132)\theta^3 \\ &+ (36\alpha^2 + 108\alpha + 96)\theta^2 + (72\alpha + 72)\theta + 48 \end{aligned} \right\}}{\theta^3(\theta^2 + \alpha\theta + 2)^3}$$

$$\mu_4 = \frac{\left\{ \begin{aligned} &\theta^{11} + (5\alpha + 10)\theta^{10} + (9\alpha^2 + 60\alpha + 30)\theta^9 + (7\alpha^3 + 116\alpha^2 + 168\alpha + 197)\theta^8 \\ &+ (2\alpha^4 + 92\alpha^3 + 288\alpha^2 + 724\alpha + 576)\theta^7 + (26\alpha^4 + 198\alpha^3 + 856\alpha^2 + 1728\alpha + 1208)\theta^6 \\ &+ (48\alpha^4 + 356\alpha^3 + 3096\alpha^2 + 2792\alpha + 2144)\theta^5 \\ &+ (24\alpha^4 + 528\alpha^3 + 1880\alpha^2 + 4000\alpha + 2584)\theta^4 + (288\alpha^3 + 1872\alpha^2 + 3696\alpha + 2928)\theta^3 \\ &+ (1008\alpha^2 + 2736\alpha + 2496)\theta^2 + (1440\alpha + 1440)\theta + 720 \end{aligned} \right\}}{\theta^4(\theta^2 + \alpha\theta + 2)^4}$$

The coefficient of variation ( $C.V$ ), coefficient of Skewness ( $\sqrt{\beta_1}$ ), coefficient of Kurtosis ( $\beta_2$ ), and index of dispersion ( $\gamma$ ) of NTPPSD are thus given by

$$C.V = \frac{\sigma}{\mu_1'} = \frac{\sqrt{\theta^5 + (3\alpha + 1)\theta^4 + (2\alpha^2 + 4\alpha + 8)\theta^3 + (2\alpha^2 + 10\alpha + 16)\theta^2 + (12\alpha + 12)\theta + 12}}{\theta^2 + 2\alpha\theta + 6}$$

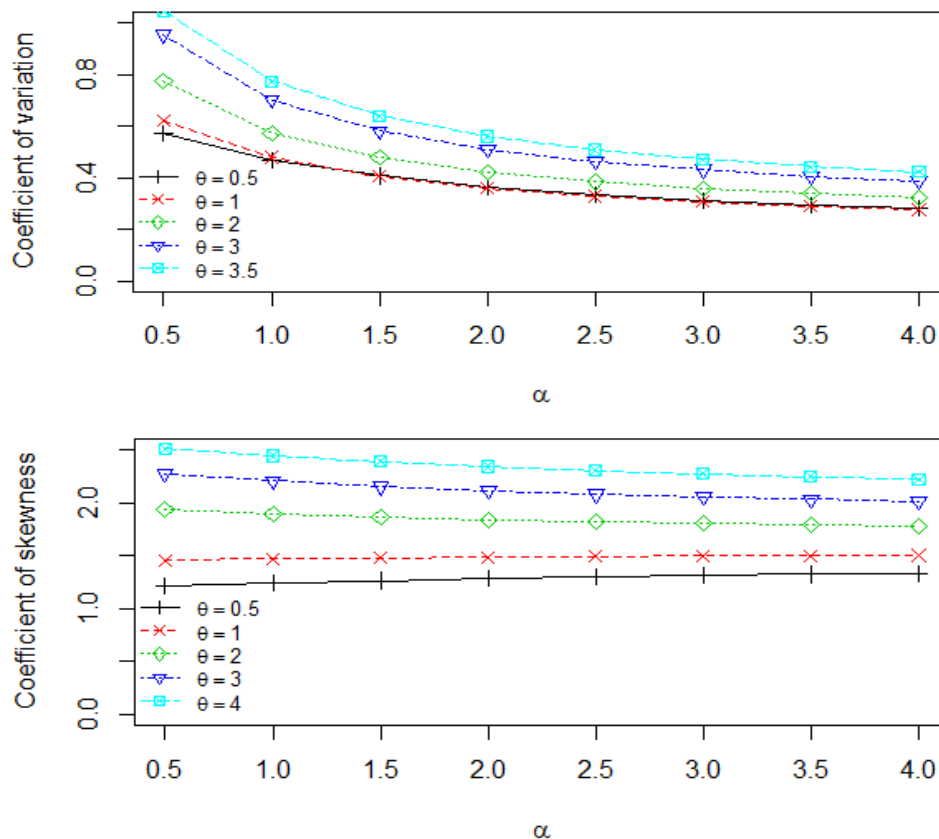
$$\sqrt{\beta_1} = \frac{\mu_3}{\mu_2^{3/2}} = \frac{\left\{ \begin{aligned} &\theta^8 + (4\alpha + 3)\theta^7 + (5\alpha^2 + 15\alpha + 12)\theta^6 + (2\alpha^3 + 18\alpha^2 + 36\alpha + 54)\theta^5 \\ &+ (6\alpha^3 + 26\alpha^2 + 108\alpha + 88)\theta^4 + (4\alpha^3 + 48\alpha^2 + 116\alpha + 132)\theta^3 \\ &+ (36\alpha^2 + 108\alpha + 96)\theta^2 + (72\alpha + 72)\theta + 48 \end{aligned} \right\}}{\left\{ \theta^5 + (3\alpha + 1)\theta^4 + (2\alpha^2 + 4\alpha + 8)\theta^3 + (2\alpha^2 + 10\alpha + 16)\theta^2 + (12\alpha + 12)\theta + 12 \right\}^{3/2}}$$

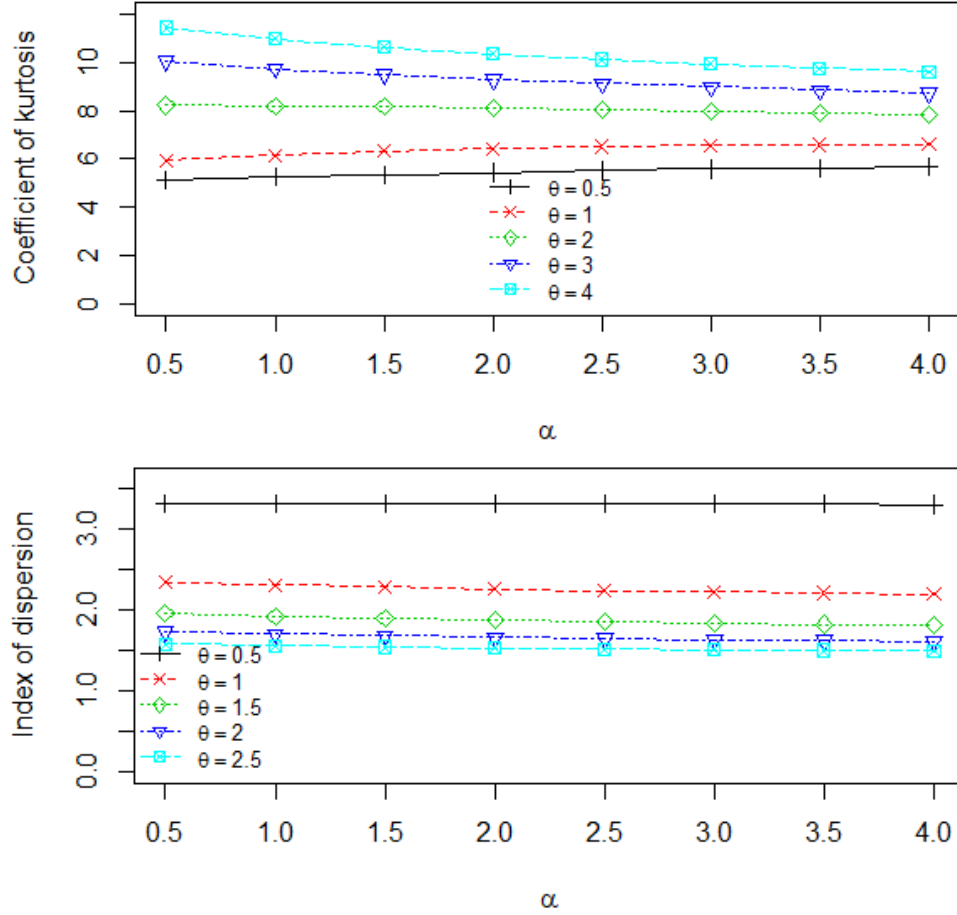
$$\beta_2 = \frac{\mu_4}{\mu_2^2} = \frac{\left\{ \begin{aligned} &\theta^{11} + (5\alpha + 10)\theta^{10} + (9\alpha^2 + 60\alpha + 30)\theta^9 + (7\alpha^3 + 116\alpha^2 + 168\alpha + 197)\theta^8 \\ &+ (2\alpha^4 + 92\alpha^3 + 288\alpha^2 + 724\alpha + 576)\theta^7 + (26\alpha^4 + 198\alpha^3 + 856\alpha^2 + 1728\alpha + 1208)\theta^6 \\ &+ (48\alpha^4 + 356\alpha^3 + 3096\alpha^2 + 2792\alpha + 2144)\theta^5 \\ &+ (24\alpha^4 + 528\alpha^3 + 1880\alpha^2 + 4000\alpha + 2584)\theta^4 + (288\alpha^3 + 1872\alpha^2 + 3696\alpha + 2928)\theta^3 \\ &+ (1008\alpha^2 + 2736\alpha + 2496)\theta^2 + (1440\alpha + 1440)\theta + 720 \end{aligned} \right\}}{\left\{ \theta^5 + (3\alpha + 1)\theta^4 + (2\alpha^2 + 4\alpha + 8)\theta^3 + (2\alpha^2 + 10\alpha + 16)\theta^2 + (12\alpha + 12)\theta + 12 \right\}^2}$$

$$\gamma = \frac{\sigma^2}{\mu_1'} = \frac{\theta^5 + (3\alpha + 1)\theta^4 + (2\alpha^2 + 4\alpha + 8)\theta^3 + (2\alpha^2 + 10\alpha + 16)\theta^2 + (12\alpha + 12)\theta + 12}{\theta(\theta^2 + \alpha\theta + 2)(\theta^2 + 2\alpha\theta + 6)}.$$

It can be easily verified that at  $\alpha = 0$  and  $\alpha = 1$  expressions of these statistical constants of NTPPSD reduce to the corresponding expressions for PAD and PSD.

The behaviors of coefficient of variation (C.V), coefficient of skewness (C.S), coefficient of kurtosis (C.K) and index of dispersion (I.D) of NTPPSD for varying values of parameters  $\theta$  and  $\alpha$  have been explained through graphs and presented in figure 2.





**Figure 2.** Behaviors of coefficient of variation (C.V), coefficient of skewness (C.S), coefficient of kurtosis (C.K) and index of dispersion (I.D) of NTTPSD for varying values of parameters  $\theta$  and  $\alpha$

#### 4. Maximum Likelihood Estimation of Parameters

Suppose  $(x_1, x_2, \dots, x_n)$  be a random sample of size  $n$  from NTTPSD and  $f_x$  be the observed frequency in the sample corresponding to  $X = x$  ( $x = 1, 2, 3, \dots, k$ ) such that  $\sum_{x=1}^k f_x = n$ , where  $k$  is the largest observed value having non-zero frequency. The likelihood function  $L$  of NTTPSD is given by

$$L = \left( \frac{\theta^3}{\theta^2 + \alpha\theta + 2} \right)^n \frac{1}{(\theta + 1)^{\sum_{x=1}^k (x+3)f_x}} \prod_{x=1}^k \left[ x^2 + (\alpha\theta + \alpha + 3)x + \left\{ \theta^2 + (\alpha + 2)\theta + (\alpha + 3) \right\} \right]^{f_x}$$

The log likelihood function is thus obtained as

$$\begin{aligned} \log L = n \left\{ 3 \log \theta - \log (\theta^2 + \alpha\theta + 2) \right\} &- \sum_{x=1}^k (x+3) f_x \log (\theta + 1) \\ &+ \sum_{x=1}^k f_x \log \left[ x^2 + (\alpha\theta + \alpha + 3)x + \left\{ \theta^2 + (\alpha + 2)\theta + (\alpha + 3) \right\} \right] \end{aligned}$$

The maximum likelihood estimates  $(\hat{\theta}, \hat{\alpha})$  of  $(\theta, \alpha)$  of NTTPSD is the solutions of the following log likelihood equations



$$\frac{\partial \log L}{\partial \theta} = \frac{3n}{\theta} - \frac{n(2\theta + \alpha)}{\theta^2 + \alpha\theta + 2} - \frac{n(\bar{x} + 3)}{\theta + 1} + \sum_{x=1}^k \left[ \frac{(\alpha x + 2\theta + \alpha + 2)f_x}{x^2 + (\alpha\theta + \alpha + 3)x + \{\theta^2 + (\alpha + 2)\theta + (\alpha + 3)\}} \right] = 0$$

$$\frac{\partial \log L}{\partial \alpha} = -\frac{-n\theta}{\theta^2 + \alpha\theta + 2} + \sum_{x=1}^k \left[ \frac{\{(\theta + 1)x + (\theta + 1)\}f_x}{x^2 + (\alpha\theta + \alpha + 3)x + \{\theta^2 + (\alpha + 2)\theta + (\alpha + 3)\}} \right] = 0,$$

where  $\bar{x}$  is the sample mean. These two log likelihood equations do not seem to be solved directly because they do not have closed forms. Therefore, to find the maximum likelihood estimates of parameters an iterative method such as Fisher Scoring method, Bisection method, Regula Falsi method or Newton-Raphson method can be used. In this paper Newton-Raphson method has been used using R-software.

**Table 1.** Observed and expected number of European corn-borer available in Gosset (1908)

Number of yeast cells per square	Observed frequency	Expected frequency				
		PSD	AGPSD	TPPSD	ATPPSD	NTPPSD
0	213	233.2	219.9	238.5	221.0	223.0
1	128	99.6	115.3	95.8	114.4	112.1
2	37	41.0	44.3	38.7	43.5	43.1
3	18	16.3	14.3	15.7	14.5	14.7
4	3	6.7	1.2	6.4	4.5	4.7
5	1	2.3	0.4	2.6	1.3	1.4
6	0	0.9	4.7	2.3	0.8	1.0
Total	400	400.0	400.0	400.0	400.0	400.0
ML Estimates		$\hat{\theta} = 2.3731$	$\hat{\theta} = 3.9945$ $\hat{\alpha} = 56.988$	$\hat{\theta} = 1.5484$ $\hat{\alpha} = 40.948$	$\hat{\theta} = 3.4626$ $\hat{\alpha} = 4983.24$	$\hat{\theta} = 2.8819$ $\hat{\alpha} = 65.80$
$\chi^2$		10.86	2.91	14.5	2.91	3.56
d.f.		2	1	1	1	1
p-value		0.0044	0.0880	0.00	0.0880	0.0591
$-2\log L$		904.88	893.21	909.64	894.65	895.63
AIC		906.88	897.21	913.64	898.65	899.63

**Table 2.** Observed and expected number of European corn-borer available in Mc Guire *et al* (1957)

Number of Corn-borer per plant	Observed frequency	Expected frequency				
		PSD	AGPSD	TPPSD	ATPPSD	NTPPSD
0	188	193.6	187.0	186.8	186.8	186.8
1	83	79.6	87.2	87.8	87.8	87.9
2	36	31.6	33.5	33.1	33.1	32.9
3	14	12.1	11.3	11.2	11.2	11.1
4	2	4.5	3.5	3.5	3.5	3.5
5	1	2.6	1.5	1.6	1.6	1.8
Total	324	324.0	324.0	324.0	324.0	324.0
ML Estimates		$\hat{\theta} = 2.4717$	$\hat{\theta} = 3.6824$ $\hat{\alpha} = 16.4716$	$\hat{\theta} = 3.31692$ $\hat{\alpha} = 0.094800$	$\hat{\theta} = 3.31671$ $\hat{\alpha} = 10.53961$	$\hat{\theta} = 2.9284$ $\hat{\alpha} = 18.4236$
$\chi^2$		1.16	0.41	0.55	0.55	0.59
d.f.		2	1	1	1	1
p-value		0.5599	0.5220	0.4583	0.4583	0.4424
$-2\log L$		713.83	711.38	711.65	711.65	711.79
AIC		715.83	715.38	715.65	715.65	715.79

**Table 3.** Accidents to 647 women working on high explosive shells in 5 weeks, available in Sankaran (1970)

No of Accidents	Observed Frequency	Expected frequency				
		PSD	AGPSD	TPPSD	ATPPSD	NTPPSD
0	447	439.8	439.8	442.2	442.2	426.7
1	132	142.1	142.1	139.3	139.3	159.6
2	42	45.0	45.0	44.4	44.4	45.4
3	21	13.9	13.9	14.2	14.2	11.5
4	3	4.2	4.2	4.5	4.5	2.7
$\geq 5$	2	2.0	2.0	2.4	2.4	1.1
Total	647	647.0	647.0	647.0	647.0	647.0
ML Estimates		$\hat{\theta} = 3.168063$	$\hat{\theta} = 3.1047$ $\hat{\alpha} = 0.81286$	$\hat{\theta} = 2.6496$ $\hat{\alpha} = 3.4025$	$\hat{\theta} = 2.6570$ $\hat{\alpha} = 0.3004$	$\hat{\theta} = 4.2303$ $\hat{\alpha} = 105.0791$
$\chi^2$		2.75	2.76	4.34	4.34	14.47
d.f.		3	1	2	2	1
p-value		0.4318	0.0966	0.1141	0.1141	0.000
$-2\log L$		1185.21	1185.21	1184.78	1184.78	1193.99
AIC		1187.21	1189.21	1188.78	1188.78	1197.99

**Table 4.** Observed and Expected number of European red mites on Apple leaves, available in Bliss (1953)

Number of Red mites per leaf	Observed frequency	Expected frequency				
		PSD	AGPSD	TPPSD	ATPPSD	NTPPSD
0	70	66.4	67.3	69.1	69.1	60.9
1	38	39.2	38.7	37.4	37.4	43.9
2	17	21.8	21.2	20.3	20.3	24.0
3	10	11.4	11.2	11.0	11.0	11.7
4	9	5.7	5.7	5.8	5.8	5.3
5	3	2.8	2.9	3.0	3.0	2.3
6	2	1.3	1.4	1.6	1.6	1.0
7	1	0.6	0.7	0.8	0.8	0.4
8	0	0.8	0.9	1	1	0.5
Total	150	150.0	150.0	150.0	150.0	150.0
ML estimates		$\hat{\theta} = 1.6533$	$\hat{\theta} = 1.4043$ $\hat{\alpha} = 0.2316$	$\hat{\theta} = 1.3640$ $\hat{\alpha} = 3.2989$	$\hat{\theta} = 1.3640$ $\hat{\alpha} = 0.3031$	$\hat{\theta} = 1.7391$ $\hat{\alpha} = 74.9996$
$\chi^2$		3.41	2.99	2.43	2.43	7.62
d.f.		4	3	3	3	2
p-value		0.4916	0.3931	0.4880	0.4880	0.0221
$-2\log L$		445.27	444.95	444.53	444.53	449.44
AIC		447.27	448.95	448.53	448.53	453.44

**Table 5.** Distribution of epileptic seizure counts due to Chakraborty (2010)

Number of epileptic seizure	Observed frequency	Expected frequency				
		PSD	AGPSD	TPPSD	ATPPSD	NTPPSD
0	126	124.3	121.5	122.3	122.3	124.5
1	80	88.3	89.7	89.5	89.6	88.1
2	59	57.6	59.3	58.8	58.8	57.5
3	42	35.3	36.1	35.8	35.8	35.3
4	24	20.6	20.7	20.6	20.6	20.6
5	8	11.5	11.3	11.4	11.4	11.6
6	5	6.3	6.0	6.1	6.1	6.3
7	4	3.3	3.1	3.2	3.2	3.3
8	3	3.8	3.3	3.3	3.2	3.8

Number of epileptic seizure	Observed frequency	Expected frequency				
		PSD	AGPSD	TPPSD	ATPPSD	NTPPSD
Total	351	351.0	351.0	351.0	351.0	351.0
ML estimates		$\hat{\theta} = 1.3241$	$\hat{\theta} = 1.4285$ $\hat{\alpha} = 1.7314$	$\hat{\theta} = 1.3715$ $\hat{\alpha} = 0.7601$	$\hat{\theta} = 1.3716$ $\hat{\alpha} = 1.3155$	$\hat{\theta} = 1.3243$ $\hat{\alpha} = 0.9331$
$\chi^2$		4.00	3.89	4.00	4.04	4.02
d.f		6	5	5	5	5
p-value		0.6766	0.5653	0.5494	0.5436	0.5465
$-2\log L$		1188.23	1187.91	1188.09	1188.09	1188.23
AIC		1190.23	1191.91	1192.09	1192.09	1192.23

**Table 6.** Observed and Expected number of households according to the number of male migrants aged 15 years and above, available in Shukla and Yadav (2006)

Number of migrants	Observed frequency	Expected frequency				
		PSD	AGPSD	TPPSD	ATPPSD	NTPPSD
0	242	239.9	239.9	240.8	240.8	241.6
1	97	98.7	98.8	97.8	97.8	96.8
2	35	39.3	39.3	39.0	39.0	38.8
3	19	15.1	15.1	15.1	15.1	15.3
4	6	5.6	5.6	5.7	5.7	5.8
5	3	2.0	2.0	2.1	2.1	2.2
6	0	0.7	0.7	0.7	0.7	0.8
7	0	0.2	0.2	0.3	0.3	0.3
8	0	0.5	0.4	0.5	0.5	0.4
Total	402	402.0	402.0	402.0	402.0	402.0
ML estimates		$\hat{\theta} = 2.4665$	$\hat{\theta} = 2.4835$ $\hat{\alpha} = 1.0589$	$\hat{\theta} = 2.3668$ $\hat{\alpha} = 1.2916$	$\hat{\theta} = 2.3665$ $\hat{\alpha} = 0.7737$	$\hat{\theta} = 2.4000$ $\hat{\alpha} = 0.4223$
$\chi^2$		1.52	1.53	1.43	1.43	1.29
d.f		3	2	2	2	2
p-value		0.6776	0.4653	0.4891	0.4891	0.5246
$-2\log L$		892.25	892.25	892.22	892.22	892.15
AIC		894.25	896.25	896.22	896.22	896.15

## 5. Applications

To examine the goodness of fit of NTPPSD over PSD, AGPSD, TPPSD and ATPPSD, six count datasets which are over-dispersed have been considered. The goodness of fit of all these distributions are based on maximum likelihood estimation. The first dataset is regarding the number of European corn-borer available in Gosset (1908), the second dataset is regarding the number of European corn-borer available in Mc Guire *et al* (1957), the third dataset is regarding the accidents to 647 women working on high explosive shells in 5 weeks, available in Sankaran (1970), the fourth dataset is regarding the number of European red mites on Apple leaves, available in Bliss (1953), the fifth dataset is regarding the distribution of epileptic seizure counts available in Chakraborty (2010) and the six dataset is regarding the observed number of households according to the number of male migrants aged 15 years and above, available in Shukla and Yadav (2006). The maximum

likelihood estimates, chi-squares, value of  $-2\log L$  and Akaike information criterion (AIC) for the considered distributions for the given datasets have been computed and presented in the respective table. The AIC has been calculated using the formula  $AIC = -2\log L + 2k$ , where  $k$  is the number of parameters involved in the distribution. In table 1, AGPSD and ATPPSD give almost the same fit. In table 2, AGPSD gives the best fit, whereas TPPSD and ATPPSD gives the second best fit. In table 3, PSD and AGPSD give almost identical and best fit whereas TPPSD and ATPPSD gives the second best fit. TPPSD and ATPPSD gives the same fit in table 4, whereas in table 5 AGPSD gives the best fit. Finally in table 6, NTPPSD gives the best fit. Therefore, we can say that NTPPSD is competing well with AGPSD, TPPSD and ATPPSD for count datasets. Therefore, it is obvious from these goodness of fit of distributions that each distribution has some advantages and disadvantages for modeling count data due to its theoretical or applied point of view. It is also true that the nature of the count data regarding

degree of over-dispersion is different and hence we can not say with confidence that a particular distribution will give best fit in every datasets.

## 6. Concluding Remarks

In this paper, a new two-parameter Poisson Sujatha distribution (NTPPSD) which includes Poisson-Akash distribution (PAD) introduced by Shanker (2017) and Poisson-Sujatha distribution (PSD) proposed by Shanker (2016 b) as a special case has been proposed.. Its unimodality, increasing hazard rate, moments and moments based measures including coefficients of variation, skewness, kurtosis and index of dispersion has been obtained and their behaviors have been explained graphically for varying values of parameters. The method of maximum likelihood estimation has been discussed. The applications of the proposed distribution has been explained through two examples of count data from ecology and the goodness of fit of the distribution has been found quite satisfactory over PSD, AGPSD, TPPSD and ATPPSD.

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