

NOTE ON A CHARACTERIZATION OF α -MODIFIED
AND TRUNCATED POISSON DISTRIBUTION

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Abstract: Poisson, binomial and negative binomial distributions as elements of a family power-series distributions were characterized by relating their first and second-order moments (see [5]). We give similar characterizations of α -modified and truncated Poisson distributions.

AMS Subject Classification: 60E05, 62E10

Key Words: power-series distribution, moments, characterization, α -modified and truncated Poisson distribution

1. Introduction

Recall that a discrete random variables X has power-series distribution (PSD) if its probability function (pf) is given by

$$P[X = x] = \frac{a(x)\theta^x}{f(\theta)}, \quad x \in T \subseteq \mathbb{N}_o, \quad (1)$$

where $\mathbb{N}_o = \mathbb{N} \cup \{0\}$, $f(\theta) = \sum_{x \in T} a(x)\theta^x$, θ is an unknown parameter taking

Received: December 3, 2010

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values in some open interval Θ , $a(x)\theta^x > 0$, coefficients $a(x)$ are nonnegative and independent of θ (see [6], [7]).

The characterizations of Poisson, binomial and negative binomial distributions are as follows (cf. [5]).

Theorem 1. *Suppose that a random variable X is distributed according to (1) and has finite expectation. Then X follows a Poisson, binomial or negative binomial distribution if and only if*

$$\frac{E_{\theta}(X(X-1))}{E_{\theta}^2 X} = c \quad \text{for all } \theta \in \Theta$$

for $c = 1$, $c = (n-1)/n$ for some positive integer $n \neq 1$, or $c = (n-1)/n$ for some negative integer n , respectively, where $\Theta = (0, k)$ for some constant k .

The aim of this note is to give counterparts of these characterizations for a α -modified Poisson distribution and for a truncated at 0 Poisson distribution.

2. Characterizations

We start with a α -modified Poisson distribution. Note that α -modified distributions and their applications were studied in [1], [2], and [3]. We are interested in a α -modified Poisson distribution with *pf*:

$$P[X = x] = \frac{\lambda^x}{x!} D_x (1 - \lambda) e^{\lambda}, \quad x = 0, 1, 2, \dots; \quad 0 < \lambda < 1, \quad (2)$$

and $D_x = (-1 + \alpha)^x = \Delta^x 0! = x! \sum_{i=0}^k \frac{(-1)^i}{i!}$, where α is Riordan's combinatorial symbol defined by $\alpha_k \equiv \alpha^k = k!$ (see [8]). One can see that this distribution is a member of power-series distribution with

$$a(x) = \frac{D_x}{x!}, \quad \theta = \lambda \quad \text{and} \quad f(\theta) = f(\lambda) = (1 - \lambda)^{-1} e^{-\lambda}.$$

A characterization of a α -modified Poisson distribution is contained in the following theorem.

Theorem 2. *Suppose that the random variable X is distributed according to (1) with $0 < \theta = \lambda < 1$, and has finite expectation. Then X follows a α -modified Poisson distribution if and only if*

$$\frac{\lambda^2 E_{\lambda}(X(X-1)) - E_{\lambda}^2 X}{\lambda^2 E_{\lambda}^2 X} = 1, \quad \text{for all } \lambda \in (0, 1). \quad (3)$$

Proof. The “only if” part. Note that the probability generating function of the distribution in (2) is

$$G_X(s) = \frac{1 - \lambda}{1 - s\lambda} e^{\lambda(1-s)}.$$

Hence

$$E_\lambda X = G'(1) = \frac{\lambda^2}{1 - \lambda}$$

and

$$E_\lambda(X(X - 1)) = G''(1) = \frac{\lambda^2(1 + \lambda^2)}{(1 - \lambda)^2}.$$

Therefore

$$\frac{E_\lambda(X(X - 1))}{(E_\lambda X)^2} = 1 + \frac{1}{\lambda^2}$$

which proves (3).

The same result one can get taking into account that for X having PSD we have

$$E_\lambda X = \lambda \frac{f'(\lambda)}{f(\lambda)} \quad \text{and} \quad E_\lambda X^2 = \lambda \frac{f'(\lambda)}{f(\lambda)} + \lambda^2 \frac{f''(\lambda)}{f(\lambda)} \tag{4}$$

(see [4]), and

$$\frac{E_\lambda(X(X - 1))}{(E_\lambda X)^2} = \frac{f''(\lambda)f(\lambda)}{(f'(\lambda))^2}.$$

Hence for a α -modified Poisson distribution where $f(\lambda) = (1 - \lambda)^{-1}e^{-\lambda}$ we get (3).

The “if” part. From the assumption (3) we have

$$\frac{E_\lambda(X(X - 1))}{(E_\lambda X)^2} = 1 + \frac{1}{\lambda^2}, \quad \text{for all } \lambda \in (0, 1),$$

and from the assumption that X belongs to a family PSD, by (4), we obtain

$$\frac{E_\lambda(X(X - 1))}{(E_\lambda X)^2} = \frac{f''(\lambda)}{f(\lambda)} / \left[\frac{f'(\lambda)}{f(\lambda)} \right]^2.$$

Therefore we have

$$\frac{f''(\lambda)}{f(\lambda)} / \left[\frac{f'(\lambda)}{f(\lambda)} \right]^2 = 1 + \frac{1}{\lambda^2},$$

or

$$\frac{f''(\lambda)}{f(\lambda)} = \left(1 + \frac{1}{\lambda^2}\right) \left[\frac{f'(\lambda)}{f(\lambda)}\right]^2. \quad (5)$$

But

$$\left[\frac{f'(\lambda)}{f(\lambda)}\right]' = \frac{f''(\lambda)}{f(\lambda)} - \left[\frac{f'(\lambda)}{f(\lambda)}\right]^2$$

or

$$\frac{f''(\lambda)}{f(\lambda)} = \left[\frac{f'(\lambda)}{f(\lambda)}\right]' + \left[\frac{f'(\lambda)}{f(\lambda)}\right]^2 \quad (6)$$

and by (5) we get

$$\left[\frac{f'(\lambda)}{f(\lambda)}\right]' + \left[\frac{f'(\lambda)}{f(\lambda)}\right]^2 = \left(1 + \frac{1}{\lambda^2}\right) \left[\frac{f'(\lambda)}{f(\lambda)}\right]^2$$

which leads to

$$\left[\frac{f'(\lambda)}{f(\lambda)}\right]' = \frac{1}{\lambda^2} \left[\frac{f'(\lambda)}{f(\lambda)}\right]^2.$$

Hence

$$\frac{f'(\lambda)}{f(\lambda)} = \frac{\lambda}{1 - c_1\lambda}$$

which gives

$$f(\lambda) = \frac{c_2 e^{-\lambda/c_1}}{(1 - c_1\lambda)^{1/c_1^2}}, \quad (7)$$

where c_1 and c_2 are arbitrary allowing to use the theory of power series in (7). With $c_1 = c_2 = 1$ we have

$$f(\lambda) = (1 - \lambda)^{-1} e^{-\lambda},$$

which ends the proof. \square

As a complement of the moments discussed in Theorem 2 we give a list of moments ($m_r := EX^r$, $\mu_r := E(X - EX)^r$) of a α -modified Poisson distribution

which can be useful in statistical inferences:

$$\begin{aligned}
 m_1 &= \frac{\lambda^2}{1-\lambda}, & m_2 &= \frac{\lambda^4 - \lambda^3 + 2\lambda^2}{(1-\lambda)^2}, \\
 m_3 &= \frac{\lambda^6 - 3\lambda^5 + 7\lambda^4 - 3\lambda^3 + 4\lambda^2}{(1-\lambda)^3}, \\
 m_4 &= \frac{\lambda^8 - 6\lambda^7 + 19\lambda^6 - 25\lambda^5 + 32\lambda^4 - 5\lambda^3 + 8\lambda^2}{(1-\lambda)^4}, \\
 \mu_2 &= \frac{\lambda^2(2-\lambda)}{(1-\lambda)^2}, & \mu_3 &= \frac{\lambda^4 - 3\lambda^3 + 4\lambda^2}{(1-\lambda)^3}, \\
 \mu_4 &= \frac{3\lambda^6 - 13\lambda^5 + 16\lambda^4 - 5\lambda^3 + 8\lambda^2}{(1-\lambda)^4}.
 \end{aligned}$$

Skewness coefficient and kurtosis are as follows:

$$\gamma = \frac{\lambda^2 - 3\lambda + 4}{\lambda(2-\lambda)^{3/2}}, \quad \kappa = \frac{3\lambda^4 - 13\lambda^3 + 16\lambda^2 - 5\lambda + 8}{\lambda^2(2-\lambda)^2}.$$

Now we present a similar characterization of a Poisson distribution truncated at 0, i.e. the distribution with the probability function

$$P[X = x] = \frac{\lambda^x}{x!} \frac{e^{-\lambda}}{1 - e^{-\lambda}}, \quad x \in T = \mathbb{N}.$$

We see that this distribution is a member of GPSD (generalized power series distribution, cf. [7]) with

$$a(x) = \frac{1}{x!}, \quad \theta = \lambda, \quad f(\theta) = f(\lambda) = e^\lambda - 1.$$

Theorem 3. *Suppose that a random variable X is distributed according to (1) with $T = \mathbb{N}$, and has finite expectation. Then X follows a Poisson distribution truncated at 0 if and only if*

$$\frac{e^\lambda E_\lambda(X(X-1)) + E_\lambda^2 X}{e^\lambda E_\lambda^2 X} = 1, \quad \lambda \in (0, +\infty). \tag{8}$$

Proof. The “only if” part. From the assumption that X belongs to a family PSD we have

$$E_\lambda X = \lambda \frac{f'(\lambda)}{f(\lambda)} \quad \text{and} \quad E_\lambda X^2 = \lambda \frac{f'(\lambda)}{f(\lambda)} + \lambda^2 \frac{f''(\lambda)}{f(\lambda)}.$$

Hence

$$\frac{E_\lambda(X(X-1))}{(E_\lambda X)^2} = \frac{f''(\lambda)f(\lambda)}{(f'(\lambda))^2}.$$

Therefore for a Poisson distribution truncated at 0 with $f(\lambda) = e^\lambda - 1$ we have

$$\frac{E_\lambda(X(X-1))}{(E_\lambda X)^2} = 1 - e^{-\lambda},$$

which proves (8).

The “if” part. From the assumption (8) and from the assumption that X belongs to a family PSD we have

$$\frac{E_\lambda(X(X-1))}{(E_\lambda X)^2} = \frac{f''(\lambda)}{f(\lambda)} / \left[\frac{f'(\lambda)}{f(\lambda)} \right]^2.$$

Hence

$$\frac{f''(\lambda)}{f(\lambda)} = (1 - e^{-\lambda}) \left[\frac{f'(\lambda)}{f(\lambda)} \right]^2$$

and by (6)

$$\left[\frac{f'(\lambda)}{f(\lambda)} \right]' = -e^{-\lambda} \left[\frac{f'(\lambda)}{f(\lambda)} \right]^2$$

which leads to

$$f(\lambda) = c_2(c_1 e^\lambda - 1)^{1/c_1} \tag{9}$$

for arbitrary c_1 and c_2 allowing to use the theory of power series in (9). With $c_1 = c_2 = 1$ we have

$$f(\lambda) = e^\lambda - 1$$

which ends the proof. \square

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