

# Poisson Approximation to the Beta-Negative Binomial Distribution

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## Abstract

The  $w$ -function and the Stein-Chen identity are used to determine two formulas of uniform and non-uniform upper bounds on Poisson approximation to the beta-negative binomial distribution.

**Mathematics Subject Classification:** Primary 60F05

**Keywords:** Beta-negative binomial distribution; Poisson approximation; Stein-Chen identity;  $w$ -function

## 1 Introduction

The beta-negative binomial distribution is a negative binomial distribution whose probability of success parameter  $p$  follows a beta distribution with shape parameters  $\alpha$  and  $\beta$ . In other words, if the probability of success parameter  $p$  of a negative binomial distribution (with parameters  $n$  and  $p$ ) has a beta distribution with shape parameters  $\alpha$  and  $\beta$ , then the resulting distribution is referred to as the beta-negative binomial distribution with parameters  $\alpha$ ,  $\beta$  and  $n$ , denoted by  $\mathcal{BN}(\alpha, \beta, n)$ . For a standard negative binomial distribution,  $p$  is usually assumed to be fixed for successive trials, but the value of  $p$  changes for each trial for the beta-negative binomial distribution. The beta-negative binomial distribution is sometimes referred to as *inverse Markov-Pólya distribution*.

Let  $X$  be the beta-negative binomial random variable. Then, for  $k \in \mathbb{N} \cup \{0\}$ , its probability distribution (Johnson et al. [4]) is defined by

$$p(k) = \frac{\binom{-\beta}{k} \binom{\beta+\alpha-1}{-n-k}}{\binom{\alpha-1}{-n}} = \frac{\Gamma(n+\alpha)\Gamma(k+\beta)\Gamma(n+k)\Gamma(\alpha+\beta)}{\Gamma(n+k+\alpha+\beta)\Gamma(n)\Gamma(k+1)\Gamma(\alpha)\Gamma(\beta)}, \quad (1.1)$$

where  $\alpha$ ,  $\beta$  and  $n$  are all positive real numbers,  $\Gamma$  gamma function, and mean and variance of  $X$  are  $\frac{n\beta}{\alpha-1}$  and  $\frac{n\beta(n+\alpha-1)(\alpha+\beta-1)}{(\alpha-2)(\alpha-1)^2}$ , respectively.

Immediately from (1.1), we obtain the recurrence relation of probability function as follows:

$$\frac{p(k-1)}{p(k)} = \frac{k(n+k+\alpha+\beta-1)}{(k+\beta-1)(n+k-1)}, \quad k = 1, 2, \dots, \quad (1.2)$$

$$\text{where } p(0) = \frac{(n+\alpha-1)(\alpha+\beta-1)!}{(n+\alpha+\beta-1)(\alpha-1)!}.$$

**Remarks.** 1. In the case of  $n = 1$ , the beta-negative binomial distribution is the beta-geometric distribution with parameters  $\alpha$  and  $\beta$ .

2. If  $\beta = a$  and  $\alpha = c - a$ , then the distribution is the so-called *generalized Waring distribution*, see Johnson et al. [4] on pp. 257.

It is well-known that the negative binomial distribution with parameters  $n$  and  $p$  can be approximated by the Poisson distribution with mean  $\lambda$ , denoted by  $\mathcal{P}(\lambda)$ , under some conditions concerning parameters  $n$ ,  $p$  and  $\lambda$ . As mentioned above and the beta-negative binomial distribution is obtained from a negative binomial distribution, we expect that the beta-negative binomial distribution can also be approximated by the Poisson distribution. In this paper, we use the  $w$ -function associated with the random variable  $X$  together with the Stein-Chen identity to determine two formulas of uniform and non-uniform upper bounds for approximating the beta-negative binomial distribution by the Poisson distribution.

## 2 Main results

We will prove our main results by using the  $w$ -function associated with the beta-negative binomial random variable  $X$  and the Stein-Chen identity. For the  $w$ -function, Majsnerowska [5] adapted the relation of  $w$ -function associated with a non-negative integer-valued random variable  $X$  (Cacoullos and Papathanasiou [2]) to be the recurrence relation of  $w$ -function in the form of

$$w(k+1) = \frac{p(k)}{p(k+1)}w(k) - \frac{\mu - (k+1)}{\sigma^2} \geq 0, \quad k = 0, 1, \dots, \quad (2.1)$$

where  $w(0) = \frac{\mu}{\sigma^2}$  and  $\mu$  and  $\sigma^2$  are mean and variance of  $X$ .

**Proposition 2.1.** *Let  $w(X)$  be the  $w$ -function associated with the beta-negative binomial random variable  $X$  and  $p(k) > 0$  for every  $k \in \mathbb{N} \cup \{0\}$ . Then we have*

$$w(k) = \frac{(n+k)(\beta+k)}{(\alpha-1)\sigma^2}, \quad k = 0, 1, \dots, \quad (2.2)$$

where  $\sigma^2 = \frac{n\beta(n + \alpha - 1)(\alpha + \beta - 1)}{(\alpha - 2)(\alpha - 1)^2}$  and  $\alpha > 2$ .

**Proof.** Following (1.2) and (2.1), we have  $w(0) = \frac{n\beta}{(\alpha - 1)\sigma^2}$  and

$$w(k) = \frac{n\beta}{(\alpha - 1)\sigma^2} + w(k - 1) \frac{k(n + k + \alpha + \beta - 1)}{(k + \beta - 1)(n + k - 1)} - \frac{k}{\sigma^2}, \quad k = 1, 2, \dots,$$

Therefore,

$$w(1) = \frac{(n + 1)(\beta + 1)}{(\alpha - 1)\sigma^2}, w(2) = \frac{(n + 2)(\beta + 2)}{(\alpha - 1)\sigma^2}, w(3) = \frac{(n + 3)(\beta + 3)}{(\alpha - 1)\sigma^2}, \dots,$$

which gives (2.2).  $\square$

For the Stein-Chen identity, Chen [3] adapted and applied idea of normal case to the Poisson setting. The Stein-Chen identity or the Stein identity for Poisson distribution with a parameter  $\lambda$  which, given  $h$ , is defined by

$$\lambda f(x + 1) - x f(x) = h(x) - \mathcal{P}_\lambda(h), \tag{2.3}$$

where  $\mathcal{P}_\lambda(h) = e^{-\lambda} \sum_{k=0}^{\infty} h(k) \frac{\lambda^k}{k!}$  and  $f$  and  $h$  are bounded real valued functions defined on  $\mathbb{N} \cup \{0\}$ .

For  $A \subseteq \mathbb{N} \cup \{0\}$ , let  $h_A$  be defined by

$$h_A(x) = \begin{cases} 1 & \text{if } x \in A, \\ 0 & \text{if } x \notin A. \end{cases} \tag{2.4}$$

Follows from Barbour et al. [1], the solution  $f_A(x)$ , when  $h = h_A$ , of (2.3) is of the form

$$f_A(x) = \begin{cases} (x - 1)! \lambda^{-x} e^\lambda [\mathcal{P}_\lambda(h_{A \cap C_{x-1}}) - \mathcal{P}_\lambda(h_A) \mathcal{P}_\lambda(h_{C_{x-1}})] & \text{if } x \geq 1, \\ 0 & \text{if } x = 0, \end{cases} \tag{2.5}$$

where  $C_x = \{0, \dots, x\}$ , and, for  $x_0 \in \mathbb{N} \cup \{0\}$ , the solutions  $f_{C_{x_0}}$ , when  $h = h_{C_{x_0}}$ , of (2.3) can be expressed in the form of

$$f_{C_{x_0}}(x) = \begin{cases} (x - 1)! \lambda^{-x} e^\lambda [\mathcal{P}_\lambda(h_{C_{x_0}}) \mathcal{P}_\lambda(1 - h_{C_{x-1}})] & \text{if } x_0 < x, \\ (x - 1)! \lambda^{-x} e^\lambda [\mathcal{P}_\lambda(h_{C_{x-1}}) \mathcal{P}_\lambda(1 - h_{C_{x_0}})] & \text{if } x_0 \geq x, \\ 0 & \text{if } x = 0, \end{cases} \tag{2.6}$$

The following proposition is established to prove the main results.

**Proposition 2.2.** *Let  $x \in \mathbb{N}$  and  $\Delta f_A(x) = f_A(x + 1) - f_A(x)$ . Then, for  $A \subseteq \mathbb{N} \cup \{0\}$ ,*

$$\sup_{x \geq 1} |\Delta f_A(x)| \leq \lambda^{-1}(1 - e^{-\lambda}) \tag{2.7}$$

and, for  $x_0 \in \mathbb{N} \cup \{0\}$ ,

$$\sup_{x \geq 1} |\Delta f_{C_{x_0}}(x)| \leq \frac{\lambda^{-1}(e^\lambda - 1)}{x_0 + 1}. \tag{2.8}$$

**Proof.** (2.7) follows from Barbour et al. [1] and, by Lemma 2.1 (3) of Teerapabolarn and Neammanee [7], (2.8) is valid.  $\square$

For constructing the two theorems below, we first mention the relation of  $w$ -function associated with a non-negative integer-valued random variable  $X$ , which was stated by Cacoullos and Papathanasiou [2]. *If a function  $g$  satisfies  $E|w(X)\Delta g(X)| < \infty$  and  $E|(X - \mu)g(X)| < \infty$ , then*

$$Cov(X, g(X)) = \sigma^2 E[w(X)\Delta g(X)], \tag{2.9}$$

where  $\Delta g(X) = g(X + 1) - g(X)$ . Note that, follows from (2.9),  $E[w(X)] = 1$ .

**Theorem 2.1.** *Let  $X$  be defined as above,  $\lambda = \frac{n\beta}{\alpha - 1}$  and  $\alpha > 2$ . Then, for  $A \subseteq \mathbb{N} \cup \{0\}$ ,*

$$d_{TV}(\mathcal{BN}(\alpha, \beta, n), \mathcal{P}(\lambda)) \leq (1 - e^{-\lambda}) \frac{\lambda + n + \beta + 1}{\alpha - 2}, \tag{2.10}$$

where  $d_{TV}(\mathcal{BN}(\alpha, \beta, n), \mathcal{P}(\lambda)) = \sup_A |\mathcal{BN}(\alpha, \beta, n)(A) - \mathcal{P}(\lambda)(A)|$ .

**Proof.** From (2.3), when  $h = h_A$  and  $f = f_A$ , we have

$$\begin{aligned} d_{TV}(\mathcal{BN}(\alpha, \beta, n), \mathcal{P}(\lambda)) &= |\lambda E[f(X + 1)] - E[Xf(X)]| \\ &= |\lambda E[f(X + 1)] - Cov(X, f(X)) - \mu E[f(X)]| \\ &= |\lambda E[\Delta f(X)] - Cov(X, f(X))| \\ &= |\lambda E[\Delta f(X)] - E[\sigma^2 w(X)\Delta f(X)]| \quad (\text{by (2.9)}) \\ &\leq E[|\lambda - \sigma^2 w(X)|\Delta f(X)|] \\ &\leq \sup_{x \geq 1} |\Delta f(x)| E|\lambda - \sigma^2 w(X)| \\ &\leq \lambda^{-1}(1 - e^{-\lambda}) E|\lambda - \sigma^2 w(X)| \quad (\text{by (2.7)}) \\ &= \lambda^{-1}(1 - e^{-\lambda}) E[\sigma^2 w(X) - \lambda] \quad (\text{by (2.2)}) \\ &= \lambda^{-1}(1 - e^{-\lambda})(\sigma^2 - \lambda), \end{aligned} \tag{2.11}$$

this implies (2.10).  $\square$

**Theorem 2.2.** For  $x_0 \in \mathbb{N} \cup \{0\}$ ,  $\alpha > 2$  and  $\lambda = \frac{n\beta}{\alpha - 1}$ ,

$$\left| \sum_{k=0}^{x_0} p(k) - \sum_{k=0}^{x_0} \frac{e^{-\lambda} \lambda^k}{k!} \right| \leq \frac{(e^\lambda - 1)(\lambda + n + \beta + 1)}{(\alpha - 2)(x_0 + 1)}. \quad (2.12)$$

**Proof.** From (2.3), when  $h = h_{C_{x_0}}$  and  $f = f_{C_{x_0}}$  and using the same argument detailed as in the proof of Theorem 2.1 together with (2.8), the theorem is also obtained.  $\square$

**Remarks.** 1. The upper bounds in (2.10) and (2.12) are small when  $\alpha$  is large, or  $\beta$  and  $n$  are small, i.e. each result of the theorems yields a good Poisson approximation whenever  $\alpha$  is large and/or  $\beta$  and  $n$  are small.

2. In the case of  $x_0 = 0$ , by applying Lemma 2.1 (3) of Teerapabolarn and Neammanee [6], we get the zero-point probability approximation as the following

$$|p(0) - e^{-\lambda}| \leq \frac{\lambda^{-1}(\lambda + e^{-\lambda} - 1)(\lambda + n + \beta + 1)}{(\alpha - 2)}. \quad (2.13)$$

## References

- [1] A.D. Barbour, L. Holst, S. Janson, Poisson approximation, Oxford Studies in Probability 2, Clarendon Press, Oxford, 1992.
- [2] T. Cacoullos, V. Papathanasiou, Characterization of distributions by variance bounds, Statist. Probab. Lett., **7** (1989), 351-356.
- [3] L.H.Y. Chen, Poisson approximation for dependent trials, Ann. Prob., **3** (1975), 534-545.
- [4] N.L. Johnson, S. Kotz, A.W. Kemp, Univariate Discrete Distributions, 3rd edition, Wiley, New York, 2005.
- [5] M. Majsnerowska, A note on Poisson approximation by  $w$ -functions, Appl. Math., **25** (1998), 387-392.
- [6] K. Teerapabolarn, K. Neammanee, A non-uniform bound on Poisson approximation in somatic cell hybrid model, Math. BioSc., **195** (2005), 56-64.
- [7] K. Teerapabolarn, K. Neammanee, Poisson approximation for sums of dependent Bernoulli random variables, Acta Math. Acad. Paedagogicae Nyiregyhaziensis, **22** (2006), 87-99.

**Received: October 24, 2007**