

A Bound on the Binomial Approximation to the Beta Binomial Distribution

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Abstract

We use the w -function and the Stein identity to give a result on the binomial approximation to the beta-binomial distribution in terms of the total variation distance between the beta-binomial and binomial distributions and its upper bound.

Mathematics Subject Classification: Primary 60F05

Keywords: Beta-binomial distribution; binomial distribution; binomial approximation; Stein identity; w -function

1 Introduction

It is well known that the binomial distribution is the distribution of the number of successes that occur in n independent trials with the probability of success in each trial is p . The binomial distribution with parameters $n > 0$ and $0 < p < 1$, written $\mathcal{B}(n, p)$, has probabilities

$$p(k) = \binom{n}{k} p^k (1-p)^{n-k}, k = 0, 1, \dots, n.$$

The beta binomial distribution is a binomial distribution whose probability of success p follows a beta distribution with shape parameters α and β . In other words, if the probability of success p of a binomial distribution has a beta distribution with shape parameters α and β , then the resulting distribution is referred to as the beta binomial distribution with parameters α , β and n . For a standard binomial distribution, p is usually assumed to be fixed for successive trials. For the beta binomial distribution the value of p changes for each trial.

Let X be the beta binomial random variable with distribution written by

$$\mathcal{BB}(\alpha, \beta, n) = \left\{ p(k) = \binom{n}{k} \frac{B(k + \alpha, n - k + \beta)}{B(\alpha, \beta)}, k = 0, 1, \dots, n \right\}, \quad (1.1)$$

where B is the complete beta function, α, β are positive real numbers and n is a positive integer. It's mean and variance are $\mu = \frac{n\alpha}{\alpha + \beta}$ and $\sigma^2 = \frac{n\alpha\beta(n + \alpha + \beta)}{(\alpha + \beta)^2(1 + \alpha + \beta)}$, respectively, see Weisstein [5].

Note that the beta binomial distribution is obtained from a binomial distribution as mentioned above, thus it is natural to speculate that the beta binomial distribution is approximately the binomial distribution. In this paper, we use the w -function associated with the random variable X together with the Stein identity to give an upper bound for the total variation distance between the beta binomial and binomial distributions.

2 Main result

This section we use the w -function associated with the beta binomial random variable X and the Stein identity for binomial distribution to give the main result of the binomial approximation to the beta binomial distribution. For the w -function, Majsnierowska [4] adapted the relation of w -function associated with the random variable X (Cacoullos and Papathanasiou [2]) to be the recurrence relation of w -function in the form of

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

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

The following proposition is an important property of the w -function associated with the beta binomial distribution.

Proposition 2.1. *Let $w(X)$ be the w -function associated with the beta binomial random variable X and $p(k) > 0$ for every $0 \leq k \leq n$. Then we have*

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

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

Proof. Following (1.1), we have

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

Using (2.1), the recurrence relation of the w -function can be expressed in the form

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

where $w(0) = \frac{n\alpha}{\sigma^2(\alpha + \beta)}$.

Therefore, we have

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

which gives (2.2). \square

For the Stein identity, we can apply the Stein identity in Barbour et al. [1] on pp. 188-189, i.e. for fixed parameters $n \geq 1$ and $p = 1 - q \in (0, 1)$, every subset A of $\{0, \dots, n\}$ and the bounded real valued function $f = f_A : \mathbb{N} \cup \{0\} \rightarrow R$ (defined as in [1]) the Stein identity for binomial case is given by

$$\mathcal{BB}(\alpha, \beta, n)(A) - \mathcal{B}(n, p)(A) = E[(n - X)pf(X + 1) - qXf(X)]. \quad (2.3)$$

For any subset A of $\{0, \dots, n\}$, Ehm [3] showed that

$$\sup_{k,A} |\Delta f(k)| = \sup_{k,A} |f(k + 1) - f(k)| \leq \frac{(1 - p^{n+1} - q^{n+1})}{(n + 1)pq}. \quad (2.4)$$

The theorem below gives an upper bound on the binomial approximation to the beta binomial distribution.

Theorem 2.1. For $A \subseteq \mathbb{N} \cup \{0\}$, if $p = \frac{\alpha}{\alpha + \beta}$, then

$$d_{TV}(\mathcal{BB}(\alpha, \beta, n), \mathcal{B}(n, p)) \leq (1 - p^{n+1} - q^{n+1}) \frac{n(n - 1)}{(n + 1)(1 + \alpha + \beta)}, \quad (2.5)$$

where $d_{TV}(\mathcal{BB}(\alpha, \beta, n), \mathcal{B}(n, p)) = \sup_A |\mathcal{BB}(\alpha, \beta, n)(A) - \mathcal{B}(n, p)(A)|$.

Proof. It can be seen that

$$\begin{aligned} E[(n - X)pf(X + 1) - qXf(X)] &= E[npf(X + 1) - pX\Delta f(X) - Xf(X)] \\ &= E[\mu f(X + 1)] - pE[X\Delta f(X)] \\ &\quad - E[Xf(X)] \\ &= E[\mu f(X + 1)] - pE[X\Delta f(X)] \\ &\quad - Cov(X, f(X)) - E[\mu f(X)] \\ &= E[\mu\Delta f(X)] - pE[X\Delta f(X)] \\ &\quad - \sigma^2 E[w(X)\Delta f(X)] \\ &\quad \text{(by Cacoullos and Papathanasiou [2])} \\ &= E\{[\mu - pX - \sigma^2 w(X)]\Delta f(X)\} \end{aligned}$$

and, by (2.3) and (2.4), we have

$$d_{TV}(\mathcal{BB}(\alpha, \beta, n), \mathcal{B}(n, p)) \leq \frac{(1 - p^{n+1} - q^{n+1})}{(n+1)pq} E|\mu - pX - \sigma^2 w(X)|. \quad (2.6)$$

Using Proposition 2.1, we get

$$\begin{aligned} E|\mu - pX - \sigma^2 w(X)| &= \sum_{x=0}^n |\mu - px - \sigma^2 w(x)| p(x) \\ &= \sum_{x=0}^n \left| \frac{n\alpha}{\alpha + \beta} - \frac{\alpha x}{\alpha + \beta} - \frac{(n-x)(\alpha+x)}{\alpha + \beta} \right| p(x) \\ &= \sum_{x=0}^n \frac{(n-x)x}{\alpha + \beta} p(x) \\ &= \frac{n^2\alpha}{(\alpha + \beta)^2} - \frac{n\alpha[n(1+\alpha) + \beta]}{(\alpha + \beta)^2(1 + \alpha + \beta)} \\ &= \frac{n(n-1)pq}{1 + \alpha + \beta} \end{aligned} \quad (2.7)$$

Hence, by (2.6) and (2.7), the theorem is proved. \square

Remark. Note that the result gives a good binomial approximation if $\frac{\alpha}{\alpha + \beta}$ or $\frac{n}{\beta}$ is small.

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Received: March 31, 2008