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## ON THE USE OF DIFFERENCE OF TWO PROPORTIONS

*Abstract.* Differences of two *proportions* occur most frequently in biomedical research. However, as far as published work is concerned, only approximations have been used to study the distribution of such differences. In this note, we derive the exact probability distribution of the difference of two proportions for seven flexible beta type distributions. The expressions involve several well known special functions. The use of these results with respect to known approximations is illustrated.

**1. Introduction.** Differences of two *proportions* arise in many areas of the sciences, engineering and medicine. The need for the probability distribution of the difference of two proportions arises especially when one is interested in comparing the performances of two *entities*. Two example are:

- Suppose that there are two drugs, say A and B, and that one wishes to compare their efficiencies. The drugs are tested on a group of patients. Let  $X_1$  denote the proportion of effectiveness for drug A. Let  $X_2$  denote the proportion of effectiveness for drug B. Then the probability distribution of  $D = X_1 - X_2$  can be used to study whether drug A is more effective than drug B or vice versa.
- Suppose that there are two pesticides, say A and B, and that one wishes to compare their efficiencies. The pesticides are tested on a field of crops. Let  $X_1$  denote the proportion of effectiveness for pesticide A. Let  $X_2$  denote the proportion of effectiveness for pesticide B. Then the probability distribution of  $D = X_1 - X_2$  can be used to study whether pesticide A is more effective than pesticide B or vice versa.

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The exact probability distribution of the difference of two proportions does not appear to have been known in the literature. Various approximations have been used for this distribution. We refer the readers to Radhakrishna et al. (1992) and Newcombe (1998) for reviews of known approximation methods. Two of the most commonly used approximations are the *normal approximation* and the *beta approximation*. The normal approximation suggests that  $D = X_1 - X_2$  follows a truncated normal distribution when  $X_1$  and  $X_2$  are independent beta type random variables. The beta approximation suggests that  $D = X_1 - X_2$  follows a beta distribution.

In this note, we derive the exact probability distribution of the difference of two proportions, i.e. the difference of two beta type random variables  $X_1$  and  $X_2$ . We consider seven of the most known beta type distributions for  $X_1$  and  $X_2$ . For each distribution, we derive an exact analytical expression for the probability density function (pdf) of the difference  $D = X_1 - X_2$ . These derivations are given in Section 2. The exact results are compared to the normal and beta approximations in Section 3. It turns out that these approximations perform poorly.

**2. Exact distributions of  $D = X_1 - X_2$ .** In this section, we provide seven theorems for the derivation of the pdf of  $D = X_1 - X_2$ . Theorem 1 considers the case that  $X_1$  and  $X_2$  have the standard beta distribution. The remaining theorems consider various generalized beta distributions. Theorems 2 and 3 consider the generalized beta distributions due to Libby and Novick (1982) and McDonald (1984), respectively. Theorems 4 and 5 consider the non-central and doubly non-central beta distributions. Theorems 6 and 7 consider the generalized beta distributions due to Gordy (1998) and Armero and Bayarri (1994), respectively. The details of the derivation of the pdf of  $D = X_1 - X_2$  are given for Theorem 1, but omitted for the other theorems. The full details can be obtained from the author.

**THEOREM 1.** *Suppose  $X_1$  and  $X_2$  are independent beta random variables given by the pdfs*

$$(1) \quad f_1(x_1) = \frac{x_1^{a_1-1}(1-x_1)^{b_1-1}}{B(a_1, b_1)}$$

and

$$(2) \quad f_2(x_2) = \frac{x_2^{a_2-1}(1-x_2)^{b_2-1}}{B(a_2, b_2)},$$

respectively, for  $0 < x_1, x_2 < 1$  and  $a_1, a_2, b_1, b_2 > 0$ , where  $B(\cdot, \cdot)$  denotes the beta function defined by

$$B(a, b) = \int_0^1 w^{a-1}(1-w)^{b-1} dw.$$

Then the pdf of the difference  $D = X_1 - X_2$  can be expressed as

$$f_D(d) = \Gamma(a_1 + b_1)\Gamma(a_2 + b_2) \times \begin{cases} \frac{\Gamma(a_1 + b_1)\Gamma(a_2 + b_2)(-d)^{a_2-1}(1+d)^{a_1+b_2-1}}{\Gamma(b_1)\Gamma(a_2)\Gamma(a_1 + b_2)} \\ \quad \times F_1\left(a_1, 1 - b_1, 1 - a_1, a_1 + b_2; 1 + d, \frac{1+d}{d}\right) & \text{if } -1 \leq d \leq 0, \\ \frac{\Gamma(a_1 + b_1)\Gamma(a_2 + b_2)d^{a_1-1}(1-d)^{a_2+b_1-1}}{\Gamma(b_2)\Gamma(a_1)\Gamma(a_2 + b_1)} \\ \quad \times F_1\left(a_2, 1 - b_2, 1 - a_1, a_2 + b_1; 1 - d, \frac{d-1}{d}\right) & \text{if } 0 \leq d \leq 1 \end{cases}$$

for  $-1 \leq d \leq 1$ , where  $F_1$  denotes the Appell hypergeometric function of the first kind defined by

$$F_1(a, b, b', c; z, \xi) = \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \frac{(a)_{k+l}(b)_k(b')_l z^k \xi^l}{(c)_{k+l} k! l!},$$

where  $(c)_k = c(c + 1) \cdots (c + k - 1)$  denotes the ascending factorial.

*Proof.* Since

$$f_D(d) = \begin{cases} \int_0^{1+d} f_1(t)f_2(t-d) dt & \text{if } -1 \leq d \leq 0, \\ \int_0^{1-d} f_1(d+t)f_2(t) dt & \text{if } 0 \leq d \leq 1, \end{cases}$$

one can write

$$(3) \quad f_D(d) = \frac{\Gamma(a_1 + b_1)\Gamma(a_2 + b_2)}{\Gamma(a_1)\Gamma(a_2)\Gamma(b_1)\Gamma(b_2)} J(d)$$

if  $-1 \leq d \leq 0$ , where

$$(4) \quad J(d) = \int_0^{1+d} t^{a_1-1}(1-t)^{b_1-1}(t-d)^{a_2-1}(1+d-t)^{b_2-1} dt.$$

Setting  $w = t/(1+d)$ , (4) can be rewritten as

$$(5) \quad J(d) = (-d)^{a_2-1}(1+d)^{a_1+b_2-1} K(d),$$

where

$$(6) \quad K(d) = \int_0^1 w^{a_1-1}(1-w)^{b_2-1} \{1 - (1+d)w\}^{b_1-1} \left\{1 - \frac{(1+d)w}{d}\right\}^{a_2-1} dw.$$

Using equation (3.211) in Gradshteyn and Ryzhik (2000), one can express (6) as

$$(7) \quad K(d) = B(a_1, b_2)F_1\left(a_1, 1 - b_1, 1 - a_1, a_1 + b_2; 1 + d, \frac{1 + d}{d}\right).$$

Combining (3), (5) and (7) yields (3) for  $-1 \leq d \leq 0$ . The result for  $0 \leq d \leq 1$  can be established similarly. ■

**THEOREM 2.** *If  $X_1$  and  $X_2$  are independent random variables given by the pdfs*

$$f_1(x_1) = \frac{p_1 x_1^{a_1 p_1 - 1} \{1 - x_1^{p_1}\}^{b_1 - 1}}{B(a_1, b_1)}$$

and

$$f_2(x_2) = \frac{p_2 x_2^{a_2 p_2 - 1} \{1 - x_2^{p_2}\}^{b_2 - 1}}{B(a_2, b_2)},$$

respectively, for  $0 < x_1, x_2 < 1$  and  $a_1, a_2, b_1, b_2, p_1, p_2 > 0$ , then the pdf of  $D = X_1 - X_2$  can be expressed as

$$f_D(d) = C d^{a_1 p_1 + a_2 p_2 - 1} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} A(k, l) d^{p_1 k + p_2 l} I_{k,l}(d)$$

for  $-1 \leq d \leq 1$ , where

$$I_{k,l}(d) = \begin{cases} B_{1/2}(1 - a_1 p_1 - p_1 k - a_2 p_2 - p_2 l, a_2 p_2 + p_2 l) \\ \quad - B_{d/(1+3d)}(1 - a_1 p_1 - p_1 k - a_2 p_2 - p_2 l, a_2 p_2 + p_2 l) & \text{if } -1 \leq d \leq 0, \\ B_{1-d}(1 - a_1 p_1 - p_1 k - a_2 p_2 - p_2 l, a_2 p_2 + p_2 l) & \text{if } 0 \leq d \leq 1, \end{cases}$$

$$A(k, l) = \frac{(1 - b_1)_k (1 - b_2)_l}{k! l!}, \quad C = \frac{p_1 p_2}{B(a_1, b_1) B(a_2, b_2)},$$

where  $B_x(\cdot, \cdot)$  denotes the incomplete beta function defined by

$$B_x(\alpha, \beta) = \int_0^x t^{\alpha-1} (1 - t)^{\beta-1} dt.$$

**THEOREM 3.** *If  $X_1$  and  $X_2$  are independent random variables given by the pdfs*

$$f_1(x_1) = \frac{\lambda_1^{a_1} x_1^{a_1 - 1} (1 - x_1)^{b_1 - 1}}{B(a_1, b_1) \{1 - (1 - \lambda_1)x_1\}^{a_1 + b_1}}$$

and

$$f_2(x_2) = \frac{\lambda_2^{a_2} x_2^{a_2-1} (1-x_2)^{b_2-1}}{B(a_2, b_2) \{1 - (1-\lambda_2)x_2\}^{a_2+b_2}},$$

respectively, for  $0 < x_1, x_2 < 1$  and  $a_1, a_2, b_1, b_2, \lambda_1, \lambda_2 > 0$ , then the pdf of  $D = X_1 - X_2$  can be expressed as

$$f_D(d) = C \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} A(k, l) I_{k,l}(d)$$

for  $-1 \leq d \leq 1$ , where

$$I_{k,l}(d) = \begin{cases} B(b_2, a_1 + k) (-d)^{a_2+l-1} (1+d)^{a_1+b_2+k-1} \\ \quad \times F_1\left(a_1 + k, 1 - b_1, 1 - a_2 - l, a_1 + b_2 + k; 1 + d, \frac{1+d}{d}\right) & \text{if } -1 \leq d \leq 0, \\ B(b_1, a_2 + l) d^{a_1+k-1} (1-d)^{a_2+b_1+l-1} \\ \quad \times F_1\left(a_2 + l, 1 - b_2, 1 - a_1 - k, a_2 + b_1 + l; 1 - d, \frac{d-1}{d}\right) & \text{if } 0 \leq d \leq 1, \end{cases}$$

$$A(k, l) = \frac{(a_1 + b_1)_k (a_2 + b_2)_l (1 - \lambda_1)^k (1 - \lambda_2)^l}{k!l!},$$

$$C = \frac{\lambda_1^{a_1} \lambda_2^{a_2}}{B(a_1, b_1) B(a_2, b_2)}.$$

**THEOREM 4.** If  $X_1$  and  $X_2$  are independent random variables given by the pdfs

$$f_1(x_1) = \frac{x_1^{a_1-1} (1-x_1)^{b_1-1}}{\Gamma(b_1)} \exp\left(-\frac{\lambda_1}{2}\right) \sum_{k=0}^{\infty} \frac{\Gamma(a_1 + b_1 + k) (\lambda_1 x_1)^k}{\Gamma(a_1 + k) 2^k k!}$$

and

$$f_2(x_2) = \frac{x_2^{a_2-1} (1-x_2)^{b_2-1}}{\Gamma(b_2)} \exp\left(-\frac{\lambda_2}{2}\right) \sum_{k=0}^{\infty} \frac{\Gamma(a_2 + b_2 + k) (\lambda_2 x_2)^k}{\Gamma(a_2 + k) 2^k k!},$$

respectively, for  $0 < x_1, x_2 < 1$  and  $a_1, a_2, b_1, b_2, \lambda_1, \lambda_2 > 0$ , then the pdf of  $D = X_1 - X_2$  can be expressed as

$$f_D(d) = \exp\left(-\frac{\lambda_1 + \lambda_2}{2}\right) \times \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \Gamma(a_1 + b_1 + k) \Gamma(a_2 + b_2 + l) \frac{\lambda_1^k \lambda_2^l}{2^{k+l} k! l!} I_{k,l}(d)$$

for  $-1 \leq d \leq 1$ , where

$$I_{k,l}(d) = \begin{cases} \frac{(-d)^{a_2+l-1}(1+d)^{a_1+b_2+k-1}}{\Gamma(b_1)\Gamma(a_2+l)\Gamma(a_1+b_2+k)} \\ \quad \times F_1\left(a_1+k, 1-b_1, 1-a_1-l, a_1+b_2+k; 1+d, \frac{1+d}{d}\right) & \text{if } -1 \leq d \leq 0, \\ \frac{d^{a_1+k-1}(1-d)^{a_2+b_1+l-1}}{\Gamma(b_2)\Gamma(a_1+k)\Gamma(a_2+b_1+l)} \\ \quad \times F_1\left(a_2+l, 1-b_2, 1-a_1-k, a_2+b_1+l; 1-d, \frac{d-1}{d}\right) & \text{if } 0 \leq d \leq 1. \end{cases}$$

THEOREM 5. If  $X_1$  and  $X_2$  are independent random variables given by the pdfs

$$f_1(x_1) = x_1^{a_1-1}(1-x_1)^{b_1-1} \exp\left(-\frac{\lambda_1 + \mu_1}{2}\right) \\ \times \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \frac{\Gamma(a_1+b_1+k+l)(\lambda_1 x_1)^k \{\mu_1(1-x_1)\}^l}{\Gamma(a_1+k)\Gamma(b_1+l)2^{k+l}k!l!}$$

and

$$f_2(x_2) = x_2^{a_2-1}(1-x_2)^{b_2-1} \exp\left(-\frac{\lambda_2 + \mu_2}{2}\right) \\ \times \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \frac{\Gamma(a_2+b_2+k+l)(\lambda_2 x_2)^k \{\mu_2(1-x_2)\}^l}{\Gamma(a_2+k)\Gamma(b_2+l)2^{k+l}k!l!},$$

respectively, for  $0 < x_1, x_2 < 1$  and  $a_1, a_2, b_1, b_2, \lambda_1, \lambda_2, \mu_1 > 0, \mu_2 > 0$ , then the pdf of  $D = X_1 - X_2$  can be expressed as

$$f_D(d) = C \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{m=0}^{\infty} \sum_{n=0}^{\infty} A(k, l, m, n) J_{k,l,m,n}(d)$$

for  $-1 \leq d \leq 1$ , where

$$J_{k,l,m,n}(d) = \begin{cases} \frac{(-d)^{a_2+m-1}(1+d)^{a_2+b_2+k+n-1}}{\Gamma(a_1+b_2+k+n)\Gamma(a_2+m)\Gamma(b_1+l)} \\ \quad \times F_1\left(a_1+k, 1-b_1-l, 1-a_2-m, a_1+b_2+k+n; 1+d, \frac{1+d}{d}\right) & \text{if } -1 \leq d \leq 0, \\ \frac{d^{a_1+k-1}(1-d)^{a_2+b_1+l+m-1}}{\Gamma(a_2+b_1+l+m)\Gamma(a_1+k)\Gamma(b_2+n)} \\ \quad \times F_1\left(a_2+m, 1-b_2-n, 1-a_1-k, a_2+b_1+l+m; 1-d, \frac{d-1}{d}\right), & \text{if } 0 \leq d \leq 1, \end{cases}$$

$$A(k, l, m, n) = \frac{\lambda_1^k \mu_1^l \lambda_2^m \mu_2^n \Gamma(a_1 + b_1 + k + l) \Gamma(a_2 + b_2 + m + n)}{2^{k+l+m+n} k! l! m! n!},$$

$$C = \exp\left(-\frac{\lambda_1 + \mu_1 + \lambda_2 + \mu_2}{2}\right).$$

THEOREM 6. If  $X_1$  and  $X_2$  are independent random variables given by the pdfs

$$f_1(x_1) = \frac{x_1^{a_1-1} (1-x_1)^{b_1-1} \exp(-\gamma_1 x_1)}{B(a_1, b_1) {}_1F_1(a_1; a_1 + b_1; -\gamma_1)}$$

and

$$f_2(x_2) = \frac{x_2^{a_2-1} (1-x_2)^{b_2-1} \exp(-\gamma_2 x_2)}{B(a_2, b_2) {}_1F_1(a_2; a_2 + b_2; -\gamma_2)},$$

respectively, for  $0 < x_1, x_2 < 1$  and  $a_1, a_2, b_1, b_2, \gamma_1, \gamma_2 > 0$ , where  ${}_1F_1$  denotes the confluent hypergeometric function defined by

$${}_1F_1(\alpha; \beta; x) = \sum_{k=0}^{\infty} \frac{(\alpha)_k x^k}{(\beta)_k k!},$$

then the pdf of  $D = X_1 - X_2$  can be expressed as

$$f_D(d) = \begin{cases} C(-d)^{a_2-1} (1+d)^{a_1+b_2-1} \exp(\gamma_2 d) \sum_{k=0}^{\infty} \frac{(1-b_1)_k}{k!} (1+d)^k \\ \quad \times B(a_1+k, b_2) \Phi_1\left(a_1+k, 1-a_2, a_1+b_2+k; \frac{1+d}{d}, -(\gamma_1+\gamma_2)(1+d)\right) & \text{if } -1 \leq d \leq 0, \\ C d^{a_1-1} (1-d)^{b_1+a_2-1} \exp(-\gamma_1 d) \sum_{k=0}^{\infty} \frac{(1-b_2)_k}{k!} (1-d)^k \\ \quad \times B(b_1, a_2+k) \Phi_1\left(a_2+k, 1-a_1, b_1+a_2+k; \frac{d-1}{d}, -(\gamma_1+\gamma_2)(1-d)\right) & \text{if } 0 \leq d \leq 1 \end{cases}$$

for  $-1 \leq d \leq 1$ , where

$$1/C = B(a_1, b_1) {}_1F_1(a_1; a_1 + b_1; -\gamma_1) B(a_2, b_2) {}_1F_1(a_2; a_2 + b_2; -\gamma_2)$$

and  $\Phi_1$  denotes the Appell hypergeometric series of the first kind defined by

$$\Phi_1(\alpha, \beta, \gamma; x, y) = \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \frac{(\alpha)_{k+l} (\beta)_k x^k y^l}{(\gamma)_{k+l} k! l!}.$$

THEOREM 7. If  $X_1$  and  $X_2$  are independent random variables given by the pdfs

$$f_1(x_1) = \frac{x_1^{a_1-1} (1-x_1)^{b_1-1} / (1+z_1 x_1)^{\gamma_1}}{B(a_1, b_1) {}_2F_1(\gamma_1, a_1; a_1 + b_1; -z_1)}$$

and

$$f_2(x_2) = \frac{x_2^{a_2-1}(1-x_2)^{b_2-1}/(1+z_2x_2)^{\gamma_2}}{B(a_2, b_2) {}_2F_1(\gamma_2, a_2; a_2 + b_2; -z_2)},$$

respectively, for  $0 < x_1, x_2 < 1$  and  $a_1, a_2, b_1, b_2, \gamma_1, \gamma_2, z_1, z_2 > 0$ , where  ${}_2F_1$  denotes the Gauss hypergeometric function defined by

$${}_2F_1(\alpha, \beta; \gamma; x) = \sum_{k=0}^{\infty} \frac{(\alpha)_k(\beta)_k}{(\gamma)_k} \frac{x^k}{k!},$$

then the pdf of  $D = X_1 - X_2$  can be expressed as

$$(8) \quad f_D(d) = C \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \frac{(-1)^{k+l}(\gamma_1)_k(\gamma_2)_l z_1^k z_2^l}{k!l!} I_{k,l}(d)$$

for  $-1 \leq d \leq 1$ , where

$$I_{k,l}(d) = \begin{cases} B(a_1 + k, b_2)(-d)^{a_2+l-1}(1+d)^{a_1+b_2+k-1} \\ \quad \times F_1\left(a_1 + k, 1 - b_1, 1 - a_2 - l, a_1 + b_2 + k; 1 + d, \frac{1+d}{d}\right) & \text{if } -1 \leq d \leq 0, \\ B(b_1, a_2 + l)d^{a_1+k-1}(1-d)^{a_2+b_1+l-1} \\ \quad \times F_1\left(a_2 + l, 1 - b_2, 1 - a_1 - k, a_2 + b_1 + l; 1 - d, \frac{d-1}{d}\right) & \text{if } 0 \leq d \leq 1, \end{cases}$$

$$1/C = B(a_1, b_1) {}_2F_1(\gamma_1, a_1; a_1 + b_1; -z_1)B(a_2, b_2) {}_2F_1(\gamma_2, a_2; a_2 + b_2; -z_2).$$

**3. Discussion.** We have derived analytical expressions for the pdf of the difference of two proportions by considering seven flexible beta type distributions. The expressions involve the Appell hypergeometric function of the first kind, the Appell hypergeometric series of the first kind, the confluent hypergeometric function, the Gauss hypergeometric function, and the incomplete beta function. These functions are well known and well established in the mathematics literature; see Lebedev (1972), Erdélyi et al. (1981), Prudnikov et al. (1986) and Gradshteyn and Ryzhik (2000) for their detailed properties. In-built numerical routines for computing them are available in most mathematical packages, e.g. Maple, Mathematica and Matlab.

We feel that the results in Section 2 can be of help in deriving an exact approach for the various modelling problems that involve differences of two proportions. Their use can be illustrated, for example, by comparing the derived pdfs with the approximations mentioned in Section 1. Suppose  $X_1$  and  $X_2$  are independent beta random variables with the pdfs specified by (1) and (2), respectively. The beta approximation for  $D = X_1 - X_2$  assumes that  $D \approx 2X - 1$  with  $X$  taken to be a beta random variable with



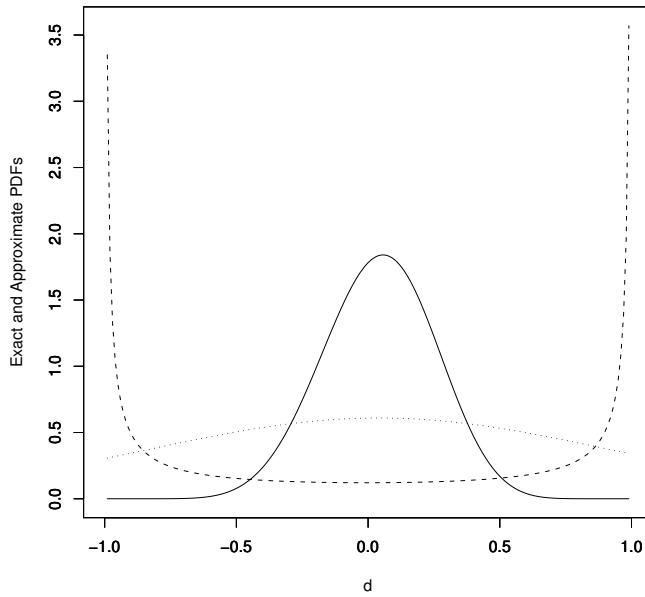


Fig. 1. The exact (solid curve), beta approximated (dashed curve) and normal approximated (dotted curve) pdfs of  $D = X_1 - X_2$  for  $a_1 = 0.9$ ,  $b_1 = 0.8$ ,  $a_2 = 0.7$  and  $b_2 = 0.5$ .

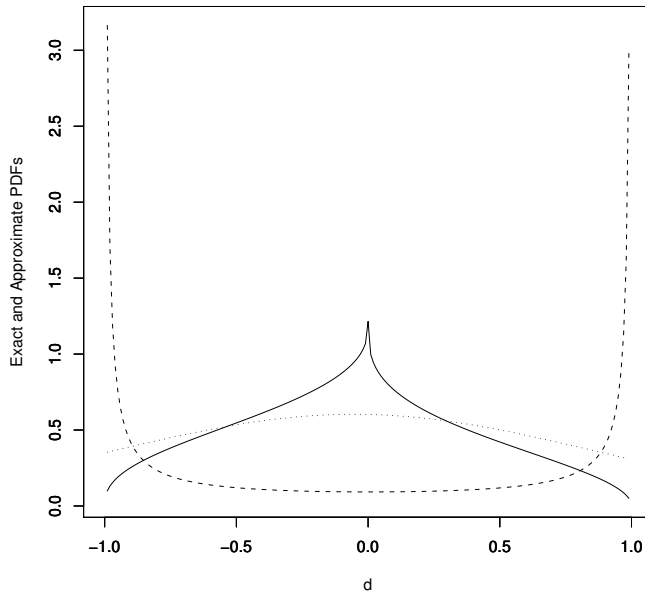


Fig. 2. The exact (solid curve), beta approximated (dashed curve) and normal approximated (dotted curve) pdfs of  $D = X_1 - X_2$  for  $a_1 = 0.9$ ,  $b_1 = 0.8$ ,  $a_2 = 0.7$  and  $b_2 = 0.5$ .

parameters, say,  $a$  and  $b$ . These parameters can be determined by setting  $E(X_1) - E(X_2) = 2E(X) - 1$  and  $\text{Var}(X_1) + \text{Var}(X_2) = 4 \text{Var}(X)$ , yielding

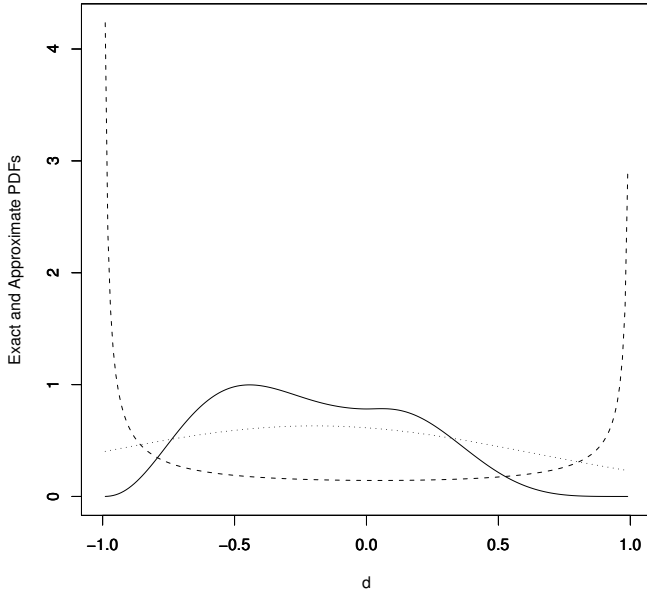


Fig. 3. The exact (solid curve), beta approximated (dashed curve) and normal approximated (dotted curve) pdfs of  $D = X_1 - X_2$  for  $a_1 = 0.6$ ,  $b_1 = 0.8$ ,  $a_2 = 5$  and  $b_2 = 3$ .

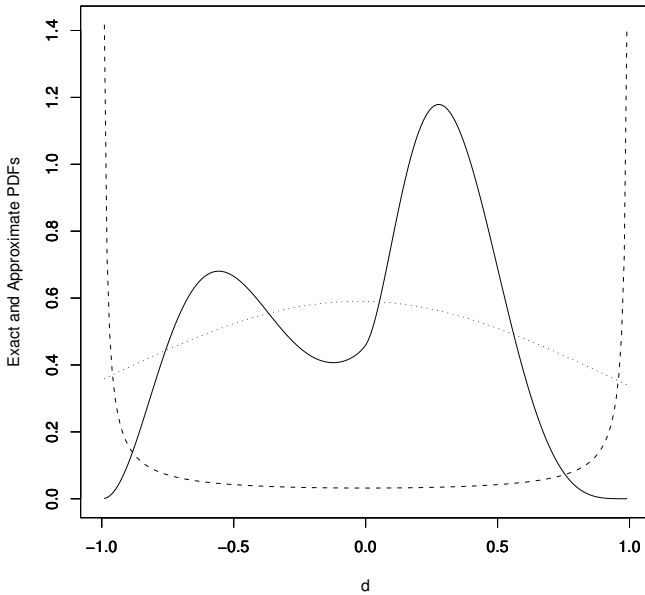


Fig. 4. The exact (solid curve), beta approximated (dashed curve) and normal approximated (dotted curve) pdfs of  $D = X_1 - X_2$  for  $a_1 = 0.3$ ,  $b_1 = 0.2$ ,  $a_2 = 5$  and  $b_2 = 3$ .

the solutions

$$a = D_1 \left\{ \frac{D_1(1 - D_1)}{D_2} - 1 \right\}, \quad b = (1 - D_1) \left\{ \frac{D_1(1 - D_1)}{D_2} - 1 \right\},$$

where

$$D_1 = \frac{1}{2} \left\{ \frac{a_1}{a_1 + b_1} - \frac{a_2}{a_2 + b_2} + 1 \right\},$$

$$D_2 = \frac{1}{4} \left\{ \frac{a_1 b_1}{(a_1 + b_1)^2 (a_1 + b_1 + 1)} + \frac{a_2 b_2}{(a_2 + b_2)^2 (a_2 + b_2 + 1)} \right\}.$$

The normal approximation for  $D = X_1 - X_2$  assumes that  $D \approx N(\mu, \sigma^2)$  truncated over the interval  $[-1, 1]$ . The parameters  $\mu$  and  $\sigma$  can be obtained as the simultaneous solutions of the equations

$$\frac{1}{\Phi((1 - \mu)/\sigma) - \Phi(-(1 + \mu)/\sigma)} \int_{-1}^1 x \phi\left(\frac{x - \mu}{\sigma}\right) dx = D_3,$$

$$\frac{1}{\Phi((1 - \mu)/\sigma) - \Phi(-(1 + \mu)/\sigma)} \int_{-1}^1 x^2 \phi\left(\frac{x - \mu}{\sigma}\right) dx = D_4,$$

where

$$\phi(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right),$$

$$\Phi(x) = \int_{-\infty}^x \phi(y) dy,$$

$$D_3 = \frac{a_1}{a_1 + b_1} - \frac{a_2}{a_2 + b_2},$$

$$D_4 = \frac{a_1(a_1 + 1)}{a_1 + b_1} + \frac{a_2(a_2 + 1)}{a_2 + b_2} - \frac{2a_1 a_2}{(a_1 + b_1)(a_2 + b_2)}.$$

The exact pdf of  $D = X_1 - X_2$  is given by (3) of Theorem 1. Figures 1, 2, 3 and 4 show how the exact and approximate pdfs compare for some chosen values of  $a_1, b_1, a_2$  and  $b_2$ . It is clear that both the approximations are very poor. The normal approximation appears to perform better than the beta approximation.

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