

Formulae for Bayesian A/B Testing: Walkthrough

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[Original material](#)

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A/B testing: binary outcomes

For a binary-outcome test (e.g. a test of conversion rates), the probability that B will beat A in the long run is given by:

$$\Pr(p_B > p_A) = \sum_{i=0}^{\alpha_B-1} \frac{B(\alpha_A + i, \beta_B + \beta_A)}{(\beta_B + i)B(1 + i, \beta_B)B(\alpha_A, \beta_A)}$$

Where:

- $\alpha_A = 1 + S_A$
- $\beta_A = 1 + F_A$
- $\alpha_B = 1 + S_B$
- $\beta_B = 1 + F_B$
- B is the beta function. For arbitrary $\alpha > 0$ and $\beta > 0$, beta function is defines as:

$$B(\alpha, \beta) = \frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha + \beta)}.$$

Derivation

Suppose we have two independent experimental branches (A and B) and have a Bayesian belief for each one:

$$\begin{aligned} p_A &\sim \text{Beta}(\alpha_A, \beta_A) \\ p_B &\sim \text{Beta}(\alpha_B, \beta_B) \end{aligned}$$

Using the pdf of the beta distribution, we can get the total probability that p_B is greater than p_A by integrating the joint distribution over all values for which $p_B > p_A$:

$$\begin{aligned} \Pr(p_B > p_A) &= \int_0^1 \Pr(p_B > p_A | p_A) \Pr(p_A) dp_A = \int_0^1 \int_{p_A}^1 \Pr(p_B | p_A) dp_B \Pr(p_A) dp_A \\ &= \int_0^1 \int_{p_A}^1 \frac{p_A^{\alpha_A-1} (1-p_A)^{\beta_A-1}}{B(\alpha_A, \beta_A)} \frac{p_B^{\alpha_B-1} (1-p_B)^{\beta_B-1}}{B(\alpha_B, \beta_B)} dp_B dp_A \end{aligned} \quad (1)$$

Define $B(x; a, b) = \int_0^x t^{a-1} (1-t)^{b-1} dt$ and $I_x(a, b) = \frac{B(x; a, b)}{B(a, b)}$. $I_x(a, b)$ is the [regularized incomplete beta function](#) and

note that $I_x(1, b) = \frac{B(x; 1, b)}{B(1, b)} = b \cdot \int_0^x (1-t)^{b-1} dt = b \left\{ \frac{1}{b} - \frac{(1-x)^b}{b} \right\} = 1 - (1-x)^b$.

With these facts, evaluate the inner integral:

$$\int_{p_A}^1 \frac{p_B^{\alpha_B-1} (1-p_B)^{\beta_B-1}}{B(\alpha_B, \beta_B)} dp_B = 1 - \int_0^{p_A} \frac{p_B^{\alpha_B-1} (1-p_B)^{\beta_B-1}}{B(\alpha_B, \beta_B)} dp_B = 1 - \frac{B(p_A; \alpha_B, \beta_B)}{B(\alpha_B, \beta_B)} = 1 - I_{p_A}(\alpha_B, \beta_B). \quad (2)$$

So the equation becomes:

$$\Pr(p_B > p_A) = 1 - \int_0^1 \frac{p_A^{\alpha_A-1}(1-p_A)^{\beta_A-1}}{B(\alpha_A, \beta_A)} I_{p_A}(\alpha_B, \beta_B) dp_A. \quad (3)$$

Now, there is a recursive relationship

$$I_x(a, b) = I_x(a-1, b) - \frac{x^{a-1}(1-x)^b}{(a-1)B(a-1, b)} \quad (4)$$

because if we denote $f(a, b, x) = I_x(a, b) - I_x(a-1, b) + \frac{x^{a-1}(1-x)^b}{(a-1)B(a-1, b)}$, $f(a, b, 0) = 0$ and

$$\begin{aligned} f'(a, b, x) &= I'_x(a, b) - I'_x(a-1, b) + \frac{x^{a-2}(1-x)^{b-1}\{(a-1)(1-x) - bx\}}{(a-1)B(a-1, b)} \\ &= \frac{x^{a-1}(1-x)^{b-1}}{B(a, b)} + \frac{x^{a-2}(1-x)^{b-1}}{B(a-1, b)} + \frac{x^{a-2}(1-x)^{b-1}\{(a-1)(1-x) - bx\}}{(a-1)B(a-1, b)} \end{aligned} \quad (5)$$

so $f'(a, b, 0) = 0$ and $f(a, b, x) = 0$.

Using this relationship and the fact that α and β are integers, we can express I_x as:

$$\begin{aligned} I_x(a, b) &= I_x(a-1, b) - \frac{x^{a-1}(1-x)^b}{(a-1)B(a-1, b)} \\ &= I_x(a-2, b) - \frac{x^{a-2}(1-x)^b}{(a-2)B(a-2, b)} - \frac{x^{a-1}(1-x)^b}{(a-1)B(a-1, b)} = \dots = 1 - (1-x)^b - \sum_{j=1}^{a-1} \frac{x^{a-j}(1-x)^b}{(a-j)B(a-j, b)} \end{aligned} \quad (6)$$

Or equivalently:

$$I_x(a, b) = 1 - \sum_{i=0}^{a-1} \frac{x^i(1-x)^b}{(b+i)B(1+i, b)} \quad (7)$$

The probability integral (3) can therefore be written:

$$\begin{aligned} \Pr(p_B > p_A) &= 1 - \int_0^1 \frac{p_A^{\alpha_A-1}(1-p_A)^{\beta_A-1}}{B(\alpha_A, \beta_A)} \left(1 - \sum_{i=0}^{\alpha_B-1} \frac{p_A^i(1-p_A)^{\beta_B}}{(\beta_B+i)B(1+i, \beta_B)} \right) dp_A \\ &= 1 - 1 + \int_0^1 \frac{p_A^{\alpha_A-1}(1-p_A)^{\beta_A-1}}{B(\alpha_A, \beta_A)} \sum_{i=0}^{\alpha_B-1} \frac{p_A^i(1-p_A)^{\beta_B}}{(\beta_B+i)B(1+i, \beta_B)} dp_A \\ &= \int_0^1 \sum_{i=0}^{\alpha_B-1} \frac{p_A^{\alpha_A-1+i}(1-p_A)^{\beta_A+\beta_B-1}}{(\beta_B+i)B(\alpha_A, \beta_A)B(1+i, \beta_B)} dp_A \\ &= \sum_{i=0}^{\alpha_B-1} \int_0^1 \frac{p_A^{\alpha_A-1+i}(1-p_A)^{\beta_A+\beta_B-1}}{(\beta_B+i)B(\alpha_A, \beta_A)B(1+i, \beta_B)} dp_A \\ &= \sum_{i=0}^{\alpha_B-1} \frac{B(\alpha_A+i, \beta_A+\beta_B)}{(\beta_B+i)B(\alpha_A, \beta_A)B(1+i, \beta_B)} \int_0^1 \frac{p_A^{\alpha_A-1+i}(1-p_A)^{\beta_A+\beta_B-1}}{B(\alpha_A+i, \beta_A+\beta_B)} dp_A \end{aligned}$$

Finally:

$$\Pr(p_B > p_A) = \sum_{i=0}^{\alpha_B-1} \frac{B(\alpha_A+i, \beta_A+\beta_B)}{(\beta_B+i)B(1+i, \beta_B)B(\alpha_A, \beta_A)} \quad (8)$$

Equivalent Formulas

It's possible to derive similar formulas that sum over the other three parameters:

$$\Pr(p_B > p_A) = 1 - \sum_{i=0}^{\alpha_A-1} \frac{B(\alpha_B+i, \beta_B+\beta_A)}{(\beta_A+i)B(1+i, \beta_A)B(\alpha_B, \beta_B)} \quad (9)$$

$$\Pr(p_B > p_A) = \sum_{i=0}^{\beta_A-1} \frac{B(\beta_B+i, \alpha_A+\alpha_B)}{(\alpha_A+i)B(1+i, \alpha_A)B(\alpha_B, \beta_B)} \quad (10)$$

$$\Pr(p_B > p_A) = 1 - \sum_{i=0}^{\beta_B-1} \frac{B(\beta_A+i, \alpha_A+\alpha_B)}{(\alpha_B+i)B(1+i, \alpha_B)B(\alpha_A, \beta_A)} \quad (11)$$

The above formulas can be found with symmetry arguments.

A/B/C testing: binary outcomes

It is possible to extend the binary-outcome formula to three test groups, call them A, B, and C. The probability that C will beat both A and B in the long run is:

$$\Pr(p_C > \max\{p_A, p_B\}) = 1 - \Pr(p_A > p_C) - \Pr(p_B > p_C) + \sum_{i=0}^{\alpha_A-1} \sum_{j=0}^{\alpha_B-1} \frac{B(i+j+\alpha_C, \beta_A+\beta_B+\beta_C)}{(\beta_A+i)B(1+i, \beta_A)(\beta_B+j)B(1+j, \beta_B)B(\alpha_C, \beta_C)}$$

Where:

- α_X is one plus the number of successes for $X \in \{A, B, C\}$
- β_X is one plus the number of failures for $X \in \{A, B, C\}$
- $\Pr(p_X > p_C)$ is the formula for the two-group case, given by (8)

Note that this formula can be computed in $O(\alpha_A \alpha_B)$ time (see the [implementation](#) section below).

Derivation

Start with a Bayesian belief for each of three experimental branches (A, B, and C):

$$\begin{aligned} p_A &\sim \text{Beta}(\alpha_A, \beta_A) \\ p_B &\sim \text{Beta}(\alpha_B, \beta_B) \\ p_C &\sim \text{Beta}(\alpha_C, \beta_C) \end{aligned}$$

Calling the pdf of the beta distribution $f(p|\alpha, \beta) = f(p)$, we can get the total probability that p_C is greater than both p_A and p_B by integrating the joint distribution over all values for which $p_C > p_A$ and $p_C > p_B$:

$$\begin{aligned} \Pr(\max\{p_A, p_B\} < p_C) &= \int_0^1 \Pr(p_C) \Pr(p_C | \max\{p_A, p_B\} < p_C) dp_C = \int_0^1 \Pr(p_C) \Pr(p_A < p_C) \Pr(p_B < p_C) dp_C \\ &= \int_0^1 \int_0^{p_C} \int_0^{p_C} f(p_A) f(p_B) f(p_C) dp_A dp_B dp_C \end{aligned}$$

Evaluating the inner two integrals, the equation becomes:

$$\Pr(p_C > \max\{p_A, p_B\}) = \int_0^1 I_{p_C}(\alpha_A, \beta_A) I_{p_C}(\alpha_B, \beta_B) f(p_C) dp_C \quad (12)$$

Using the identity for I_X (7), we have:

$$\Pr(p_C > \max\{p_A, p_B\}) = \int_0^1 \left(1 - \sum_{i=0}^{\alpha_A-1} \frac{p_C^i (1-p_C)^{\beta_A}}{(\beta_A+i)B(1+i, \beta_A)} \right) \left(1 - \sum_{i=0}^{\alpha_B-1} \frac{p_C^i (1-p_C)^{\beta_B}}{(\beta_B+i)B(1+i, \beta_B)} \right) f(p_C) dp_C$$

Multiplying out the parenthetical terms and integrating them separately:

$$\begin{aligned} \Pr(p_C > \max\{p_A, p_B\}) &= 1 - \int_0^1 \sum_{i=0}^{\alpha_A-1} \frac{p_C^i (1-p_C)^{\beta_A}}{(\beta_A+i)B(1+i, \beta_A)} f(p_C) dp_C - \int_0^1 \sum_{i=0}^{\alpha_B-1} \frac{p_C^i (1-p_C)^{\beta_B}}{(\beta_B+i)B(1+i, \beta_B)} f(p_C) dp_C \\ &\quad + \int_0^1 \sum_{i=0}^{\alpha_A-1} \frac{p_C^i (1-p_C)^{\beta_A}}{(\beta_A+i)B(1+i, \beta_A)} \sum_{i=0}^{\alpha_B-1} \frac{p_C^i (1-p_C)^{\beta_B}}{(\beta_B+i)B(1+i, \beta_B)} f(p_C) dp_C \end{aligned}$$

From the previous derivation, we can rewrite the first two integrals as $\Pr(p_A > p_C)$ and $\Pr(p_B > p_C)$, and consolidate the terms inside the third integral:

$$\begin{aligned} \Pr(p_C > \max\{p_A, p_B\}) &= 1 - \Pr(p_A > p_C) - \Pr(p_B > p_C) \\ &\quad + \int_0^1 \sum_{i=0}^{\alpha_A-1} \sum_{j=0}^{\alpha_B-1} \frac{p_C^{i+j} (1-p_C)^{\beta_A+\beta_B}}{(\beta_A+i)(\beta_B+j)B(1+i, \beta_A)B(1+j, \beta_B)} \frac{p_C^{\alpha_C-1} (1-p_C)^{\beta_C-1}}{B(\alpha_C, \beta_C)} dp_C \\ &= 1 - \Pr(p_A > p_C) - \Pr(p_B > p_C) \\ &\quad + \int_0^1 \sum_{i=0}^{\alpha_A-1} \sum_{j=0}^{\alpha_B-1} \frac{p_C^{i+j+\alpha_C-1} (1-p_C)^{\beta_A+\beta_B+\beta_C-1}}{(\beta_A+i)(\beta_B+j)B(1+i, \beta_A)B(1+j, \beta_B)B(\alpha_C, \beta_C)} dp_C \end{aligned}$$

Finally, evaluating the integral we have:

$$\begin{aligned} \Pr(p_C > \max\{p_A, p_B\}) &= 1 - \Pr(p_A > p_C) - \Pr(p_B > p_C) \\ &+ \sum_{i=0}^{\alpha_A-1} \sum_{j=0}^{\alpha_B-1} \frac{B(i+j+\alpha_C, \beta_A+\beta_B+\beta_C)}{(\beta_A+i)(\beta_B+j)B(1+i, \beta_A)B(1+j, \beta_B)B(\alpha_C, \beta_C)} \end{aligned}$$