

Gamma and Beta Distribution

Chao Cheng

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1 Introduction

The Gamma function is defined as

$$\Gamma(\alpha) = \int_0^{\infty} t^{\alpha-1} e^{-t} dt. \quad (1)$$

And it's easy to verify that

$$\Gamma(\alpha + 1) = \alpha \Gamma(\alpha), \quad \Gamma(n) = (n-1)!, \quad \Gamma(0.5) = \sqrt{\pi}.$$

Let $B(\alpha, \beta)$ denote the Beta function

$$B(\alpha, \beta) = \int_0^1 x^{\alpha-1} (1-x)^{\beta-1} dx. \quad (2)$$

Here we point out that $B(\alpha, \beta)$ is related to Gamma function via

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

2 Gamma distribution

The pdf for a Gamma distribution is

$$f(x|\alpha, \beta) = \frac{1}{\Gamma(\alpha) \beta^\alpha} x^{\alpha-1} \exp(-x/\beta), \quad 0 \leq x < \infty, \quad \alpha, \beta > 0, \quad (3)$$

with

$$EX = \alpha\beta, \quad \text{Var}X = \alpha\beta^2,$$

and m.g.f.

$$M_X(t) = \left(\frac{1}{1 - \beta t} \right)^\alpha, \quad t < \frac{1}{\beta}.$$

3 Beta distribution

The pdf for a Beta distribution is

$$f(x|\alpha, \beta) = \frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1}, \quad 0 \leq x \leq 1, \quad \alpha, \beta > 0, \quad (4)$$

with

$$EX = \frac{\alpha}{\alpha + \beta}, \quad \text{Var}X = \frac{\alpha\beta}{(\alpha + \beta)^2 (\alpha + \beta + 1)}$$

and m.g.f.

$$M_X(t) = 1 + \sum_{k=1}^{\infty} \left(\prod_{r=0}^{k-1} \frac{\alpha + r}{\alpha + \beta + r} \right) \frac{t^k}{k!}.$$

4 Relationship with other distribution

- Gamma distribution with $\alpha = 1$ is just exponential distribution.

$$\text{gamma}(1, \beta) = \exp(-\beta).$$

- Gamma distribution with $\beta = 2$ is just χ^2 distribution with degree of freedom $p = 2\alpha$.

$$\text{gamma}(\alpha, 2) = \chi^2(2\alpha).$$

- If X follows $\text{gamma}(\alpha, \beta)$ where α is an integer. Then for any x

$$P(X \leq x) = P(Y \geq \alpha).$$

where Y follows $\text{Poisson}(x/\beta)$.

- If

$$X \sim \text{gamma}(\alpha_1, \beta), \quad Y \sim \text{gamma}(\alpha_2, \beta), \quad X \perp Y.$$

Then

$$X + Y \sim \text{gamma}(\alpha_1 + \alpha_2, \beta), \quad \frac{X}{X + Y} \sim \text{beta}(\alpha_1, \alpha_2), \quad (X + Y) \perp \frac{X}{X + Y}.$$

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$$\text{beta}(1, 1) = \text{Unif}(0, 1).$$

- $\text{beta}(\frac{1}{2}, \frac{1}{2})$ is the **non-informative Jeffreys prior** for binomial rate test.

- Let $U_1, \dots, U_n \stackrel{\text{i.i.d.}}{\sim} \text{Unif}(0, 1)$. Then the order statistics

$$U_{(k)} \sim \text{beta}(k, n + 1 - k), \quad 1 \leq k \leq n.$$

And

$$U_{(k)} - U_{(j)} \sim \text{beta}(k - j, n - (k - j) + 1), \quad 1 \leq j < k \leq n.$$

- If X follows a $Binomial(n, p)$, then the c.d.f of X satisfies

$$\begin{aligned}
& P(X \leq k) \\
&= \sum_{i=0}^k C_n^i p^i (1-p)^{n-i} \\
&= \frac{\Gamma(n+1)}{\Gamma(n-k)\Gamma(k+1)} \int_0^{1-p} t^{n-k-1} (1-t)^k dt = pbeta(1-p; n-k, k+1) \\
&= P(Y \leq 1-p),
\end{aligned}$$

where Y follows $beta(n-k, k+1)$. This can be verified using integration by parts.

References