

Continuous Random Variables (IV)

Goal: Study Lognormal and Beta distributions

① Lognormal Distribution

We say X has a Lognormal Distribution if $\ln(X)$ has normal distribution with parameters μ and σ^2

$$\text{ie } \ln(X) \sim N(\mu, \sigma^2)$$

If X has lognormal distribution

$$\text{pdf } f(x, \mu, \sigma) = \begin{cases} \frac{1}{\sqrt{2\pi}} \sigma^{-1} e^{-\frac{(\ln(x)-\mu)^2}{2\sigma^2}} & \text{if } x > 0 \\ 0 & \text{otherwise.} \end{cases}$$

Note: μ and σ^2 \rightarrow mean and Variance of $\ln(X)$

$$\text{ie } \begin{cases} E(\ln(X)) = \mu \\ V(\ln(X)) = \sigma^2 \end{cases}$$

② If X has lognormal distribution

$$E(X) = \mu_X = e^{\mu + \frac{\sigma^2}{2}}$$

$$V(X) = \sigma_X^2 = e^{(2\mu + \sigma^2)} (e^{\sigma^2} - 1)$$

③ Can use standard normal table to calculate probabilities:

$$\begin{aligned} F_X(x; \mu, \sigma) &= P(X \leq x) = P(\ln(X) \leq \ln(x)) \\ &= P\left(Z \leq \frac{\ln(x) - \mu}{\sigma}\right) \quad (\because \ln(X) \sim N(\mu, \sigma^2)) \\ &= \Phi\left(\frac{\ln(x) - \mu}{\sigma}\right) \quad \text{for } x > 0. \end{aligned}$$

② Beta Distribution

takes values in a finite interval
good to model proportions \rightarrow naturally lie in $(0, 1)$.

X is said to have Beta Distribution with parameters: $\alpha, \beta > 0$ if pdf of X is

$$f(x; \alpha, \beta) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} \cdot x^{\alpha-1} (1-x)^{\beta-1} \quad x \in (0, 1).$$

Note: ① Called Beta distribution because the Beta function

$$B(\alpha, \beta) = \frac{\Gamma(\alpha) \cdot \Gamma(\beta)}{\Gamma(\alpha + \beta)} \quad \text{for } \alpha, \beta > 0$$

$$\text{so that } f(x; \alpha, \beta) = \frac{1}{B(\alpha, \beta)} \cdot x^{\alpha-1} (1-x)^{\beta-1} \quad \text{for } x \in (0, 1).$$

② If $X \sim \text{Beta}(\alpha, \beta)$

$$E(X) = \mu_X = \frac{\alpha}{\alpha + \beta}$$

\rightarrow Notice that both μ_X and σ_X^2 are in $(0, 1)$.

$$V(X) = \sigma_X^2 = \frac{\alpha\beta}{(\alpha + \beta)^2 (\alpha + \beta + 1)}$$

③ Depending on values of α, β \rightarrow the pdf $f(x; \alpha, \beta)$ has different shapes.

- If
- a) $\alpha > 1, \beta = 1 \rightarrow$ strictly increasing
 - b) $\alpha = 1, \beta > 1 \rightarrow$ strictly decreasing
 - c) $\alpha < 1, \beta < 1 \rightarrow$ U-shaped
 - d) $\alpha = \beta \rightarrow$ symmetric about $\frac{1}{2}$, with $\mu_X = \frac{1}{2}$ and $\sigma_X^2 = \frac{1}{4(2\alpha + 1)}$
 - e) $\alpha = \beta = 1 \rightarrow$ get uniform distribution on $(0, 1)$.

Cauchy Distribution

X is said to have Cauchy Distribution with parameter: θ if the pdf of X is

$$f(x; \theta) = \frac{1}{\pi} \cdot \frac{1}{1 + (x - \theta)^2} \quad \begin{matrix} x \in (-\infty, \infty) \\ \theta \in (-\infty, \infty) \end{matrix}$$

Note: ① $E(X)$ and $V(X)$ do not exist if X has Cauchy distribution.

② The graph of $f(x; \theta)$ is bell shaped \rightarrow like the Normal Density. But has heavier tails (than Normal density).