

## 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  $\mu$  and  $\sigma^2$ .  
ie  $\ln(X) \sim N(\mu, \sigma^2)$

If  $X$  has lognormal distribution

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

Note:  $\mu$  and  $\sigma^2$  mean and Variance of  $\ln(X)$

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

② If  $X$  has lognormal distribution

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

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

③ Can use standard normal table to calculate probabilities:

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

### 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

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

Note: ① Called Beta distribution because the Beta function

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

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

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

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

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

Notice that both  $\mu_X$  and  $\sigma_X^2$  are in  $(0, 1)$ .

③ Depending on values of  $\alpha, \beta \rightarrow$  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 ( $\alpha = 1, \beta = 2$ )
- 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  $\theta$  if the pdf of  $X$  is

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

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

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).

$$f(x) = \frac{1}{\pi} \cdot \frac{1}{1+x^2} \quad E(X) = \int_{-\infty}^{\infty} x f(x) dx = \int_{-\infty}^{\infty} \frac{1}{\pi} \cdot \frac{x}{1+x^2} dx = 0 \quad \text{odd function}$$

## Lognormal Distribution

$Y$  has lognormal  $\rightarrow$  if  $\ln(Y)$  has the distribution:  $N(\mu, \sigma^2)$   
normal dist with parameters  $\mu$  and  $\sigma^2$   
ie  $\ln(Y) \sim N(\mu, \sigma^2)$

What is the pdf of  $Y$ ?

$$F_Y(y) = P(Y \leq y) = P(\ln(Y) < \ln(y))$$
$$= \int_{-\infty}^{\ln(y)} \frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\mu)^2/2\sigma^2} dx.$$

to get pdf of  $Y$

$$\frac{d}{dy}(F_Y(y)) = \frac{d}{dy} \left( \int_{-\infty}^{\ln(y)} \frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\mu)^2/2\sigma^2} dx \right).$$

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Plot various density functions of  $Y \sim \text{logNormal}(\mu, \sigma^2)$ .

