

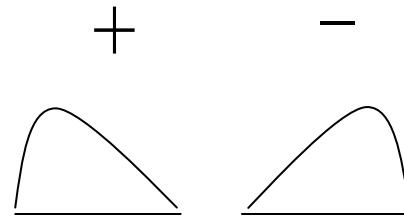
# Chapter 3

## Review of Statistics

- Histogram → probability density  
→ cumulative probability
- Mean, Standard deviation, Skew, and Kurtosis  
(p.27~p.37)

- Normal Distribution

{ Mean  
s.d.  
skew = 0 (偏態)  
kurtosis = 3 (峰態)

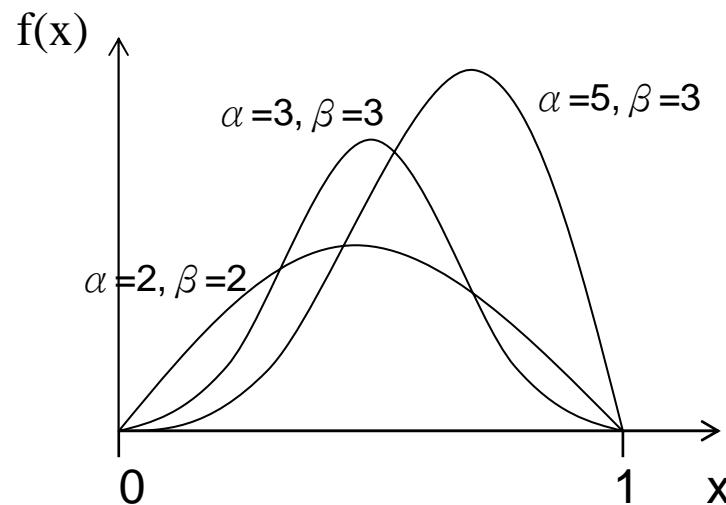


- Log Normal Distribution ( $X > 0$  才能取log)
  - If  $X \sim \text{LogNormal}$ , then  $\text{Log}(X) \sim \text{Normal}$
  - If  $Z \sim \text{Normal}$ , then  $e^z \sim \text{LogNormal}$

- Beta Distribution (常用來描述credit risk)

$$f(x) = \begin{cases} \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha) + \Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1} & , 0 < x < 1 \\ 0 & , 0 / w \end{cases}$$

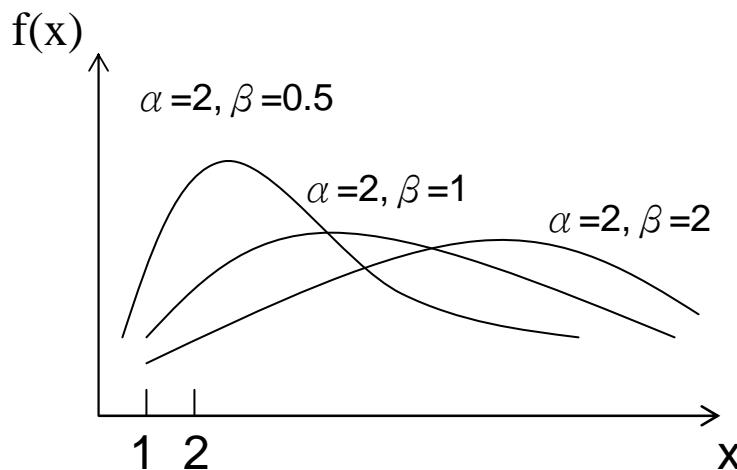
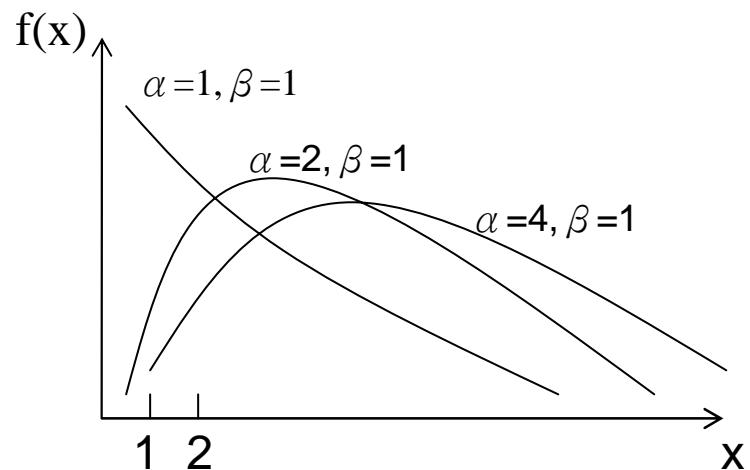
其中  $\alpha > 0, \beta > 0$ , 且  $\int_{-\infty}^{\infty} f(x) dx = \int_0^1 f(x) dx = 1$



## ● Gamma distribution

$$f(x) = \begin{cases} \frac{1}{\Gamma(\alpha) \cdot \beta^\alpha} x^{\alpha-1} \cdot e^{-\frac{x}{\beta}} & x > 0 \\ 0 & x \leq 0 \end{cases}$$

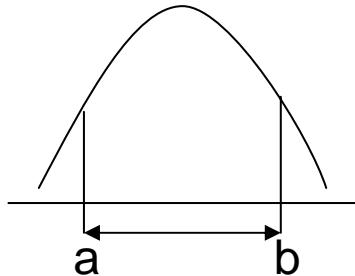
其中  $\alpha > 0$ ,  $\beta > 0$ , 且  $\Gamma(\alpha) = \int_0^\infty y^{\alpha-1} e^{-y} dy$ , which is known as the Gamma function



⊗  $\Gamma(\alpha) = (\alpha - 1)!$

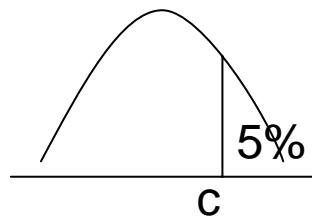
⊗ By integrating by parts,  $\Gamma(\alpha) = (\alpha - 1) \cdot \Gamma(\alpha - 1)$  <sup>5</sup>

- Confidence Intervals (信賴區間)



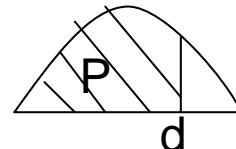
- Confidence Levels (信賴水準)

(given 5%, find c, c為95%之confidence level (有95%不會超過c) )



- Percentile

(given p, 算出d, 則d為p%ile )



- Correlation, Covariance

If  $y = \sum_{i=1}^n x_i$ ,  $\bar{y} = \sum_{i=1}^n \bar{x}_i$  and  $\sigma_y^2 = \sum_{i=1}^n \sum_{j=1}^n \rho_{i,j} \sigma_{x_i} \sigma_{x_j}$

$$*\text{if } \rho_{i,j} = 0 \Rightarrow \sigma_y^2 = n\sigma_x^2 \Rightarrow \sigma_y = \sqrt{n}\sigma_x$$

- Matrix Representation of Standard Deviation

$$S = [\sigma_{x_1} \quad \sigma_{x_2} \quad \dots \dots \quad \sigma_{x_n}]$$

$$R = \begin{bmatrix} 1 & \rho_{1,2} & \rho_{1,3} & \dots & \rho_{1,n} \\ \rho_{2,1} & 1 & & & \vdots \\ \rho_{3,1} & & \ddots & & \vdots \\ \vdots & & & \ddots & \vdots \\ \rho_{n,1} & \dots & \dots & \dots & 1 \end{bmatrix} \Rightarrow \sigma_y^2 = \underset{1 \times n}{S} \underset{n \times n}{R} \underset{n \times 1}{S^T}$$

---

- Random Process (Stochastic Process)

- For Stock Price

$$S_{t+\Delta t} = S_t + \Delta S_t$$

(1)  $\Delta S_t = \sigma \sqrt{\Delta t} z_t$ , where  $z_t \sim N(0,1)$

(2)  $\Delta S_t = S_t \sigma \sqrt{\Delta t} z_t$ , where  $z_t \sim N(0,1)$

(3)  $\Delta S_t = S_t \mu \Delta t + S_t \sigma \sqrt{\Delta t} z_t$ , where  $z_t \sim N(0,1)$

- For Interest Rate (Cox-Ingersoll-Ross (CIR) model)

$$r_{t+\Delta t} = r_t + c(r_\mu - r_t) + \sigma_r \sqrt{r_t} z_t, \text{ where } z_t \sim N(0,1)$$