

Probability & Statistics (1)

# Jointly Distributed Random Variables (II)

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# Sums of Independent Random Variables

## Uniform Distribution

- 時常我們會需要用到兩個獨立的隨機變數相加的機率分配，譬如說現在有兩個隨機變數： $X$ 與 $Y$ 。我們需要計算 $X + Y$ 的*c.d.f.*就可以從它們各自的*p.d.f.* ( $f_X$  and  $f_Y$ ) 計算得知:

$$\begin{aligned} F_{X+Y}(a) &= P\{X + Y \leq a\} = \iint_{x+y \leq a} f_X(x) f_Y(y) dx dy \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{a-y} f_X(x) f_Y(y) dx dy = \int_{-\infty}^{\infty} \int_{-\infty}^{a-y} f_X(x) dx f_Y(y) dy \\ &= \int_{-\infty}^{\infty} F_X(a - y) f_Y(y) dy \end{aligned}$$

$F_{X+Y}$ 的*c.d.f.*其實也就是 $F_X$ 與 $F_Y$ 的convolution。

# Sums of Independent Random Variables

## Uniform Distribution

Differentiation both side,

$$\begin{aligned} f_{X+Y}(a) &= \frac{d}{da} \int_{-\infty}^{\infty} F_X(a-y) f_Y(y) dy = \int_{-\infty}^{\infty} \frac{d}{da} F_X(a-y) f_Y(y) dy \\ &= \int_{-\infty}^{\infty} f_X(a-y) f_Y(y) dy \end{aligned}$$

# Sums of Independent Random Variables

## Uniform Distribution

- 範例一

如果 $X$ 與 $Y$ 為獨立隨機變數且值域uniformly distributed on  $(0,1)$ ，試問 $X + Y$ 的*p.d.f.*。

**Solution:**

Let  $f_X$  and  $f_Y$  are the *p.d.f.* of  $X$  and  $Y$ , respectively. Let  $Z = X + Y$ .

$$f_Z(z) = \int_{-\infty}^{\infty} f_X(x)f_Y(z-x)dx, \text{ where } 0 < x < 1 \text{ and } 0 < z-x < 1$$

$f_Z(z) = 0$ , for  $z < 0$  and  $z \geq 2$ ; in other words,  $0 < z < 2$

# Sums of Independent Random Variables

## Uniform Distribution

(1) *Case 1:  $0 < z \leq 1, f_X(x)f_Y(z - x)$*

$$f_Z(z) = \int_{-\infty}^{\infty} f_X(x)f_Y(z - x)dx$$

To confirm  $f_Y(z - x) = 1$ , we need  $z - x \geq 0 \Rightarrow x \leq z$

$$\int_0^z 1dx = z; f_Z(z) = z, \text{ for } 0 < z \leq 1$$

(2) *Case 2:  $1 < z < 2$*

To confirm  $f_Y(z - x) = 1$ , we need  $z - x \leq 1 \Rightarrow x \geq z - 1$

$$\int_{z-1}^1 1dx = 2 - z; f_Z(z) = 2 - z, \text{ for } 1 < z < 2$$

# Sums of Independent Random Variables

## Uniform Distribution

As a result,

$$f_Z(z) = f_{X+Y}(z) = \begin{cases} z, & 0 \leq z \leq 1 \\ 2 - z, & 1 < z < 2 \\ 0, & \text{otherwise} \end{cases}$$

Because of the shape of *p. d. f.*, the random variable of  $X + Y$  is a *triangular distribution*.

# Sums of Independent Random Variables

## Uniform Distribution

假設  $X_1, X_2, \dots, X_n$  為 independent uniform random variable，使得

$$F_n(x) = P\{X_1 + \dots + X_n \leq x\}$$

雖然  $F_n(x)$  的 general form 非常難表示，不過  $x \leq 1$  的時候，就可以用 mathematical induction 得到

$$F_n(x) = \frac{x^n}{n!}, 0 \leq x \leq 1$$

已知  $n = 1$  是正確的，假設

$$F_{n-1}(x) = \frac{x^{n-1}}{(n-1)!}, 0 \leq x \leq 1$$



# Sums of Independent Random Variables

## Uniform Distribution

於是就可以表示成

$$\sum_{i=1}^n X_i = \sum_{i=1}^{n-1} X_i + X_n, X_i \text{ is nonnegative, where } 0 \leq x \leq 1$$

$$F_n(x) = \int_0^1 F_{n-1}(x-y) f_{X_n}(y) dy = \frac{1}{(n-1)!} \int_0^x (x-y)^{n-1} dy = \frac{x^n}{n!}$$

for  $n = 1$ ,  $X$  follows a uniform distribution

$$f_X(x) = \begin{cases} 1 & 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

for  $n = 2$ ,  $X$  follows a triangular distribution

$$f_X(x) = \begin{cases} x & 0 \leq x \leq 1 \\ 2-x & 1 \leq x \leq 2 \end{cases}$$

for  $n = 3$ ,

$$f_X(x) = \begin{cases} \frac{1}{2}x^2 & 0 \leq x \leq 1 \\ \frac{1}{2}(-2x^2 + 6x - 3) & 1 \leq x \leq 2 \\ \frac{1}{2}(3-x)^2 & 2 \leq x \leq 3 \end{cases}$$

# Sums of Independent Random Variables

## Uniform Distribution

for  $n = 4$ ,

$$f_X(x) = \begin{cases} \frac{1}{6}x^3 & 0 \leq x \leq 1 \\ \frac{1}{6}(-3x^3 + 12x^2 - 12x + 4) & 1 \leq x \leq 2 \\ \frac{1}{6}(3x^3 - 24x^2 + 60x - 44) & 2 \leq x \leq 3 \\ \frac{1}{6}(4-x)^3 & 3 \leq x \leq 4 \end{cases}$$

# Sums of Independent Random Variables

## Uniform Distribution

for  $n = 5$ ,

$$f_X(x) = \begin{cases} \frac{1}{24}x^4 & 0 \leq x \leq 1 \\ \frac{1}{24}(-4x^4 + 20x^3 - 30x^2 + 20x - 5) & 1 \leq x \leq 2 \\ \frac{1}{24}(6x^4 - 60x^3 + 210x^2 - 300x + 155) & 2 \leq x \leq 3 \\ \frac{1}{24}(-4x^4 + 60x^3 - 330x^2 + 780x - 655) & 3 \leq x \leq 4 \\ \frac{1}{24}(5-x)^4 & 4 \leq x \leq 5 \end{cases}$$

# Sums of Independent Random Variables

## Normal Distribution

### Proposition 1

If  $X_i$ ,  $i = 1, \dots, n$ , are independent random variables that are normally distributed with respective parameters  $\mu_i$ ,  $\sigma_i^2$ ,  $i = 1, \dots, n$ , then  $\sum_{i=1}^n X_i$  is normally distributed with parameters  $\sum_{i=1}^n \mu_i$  and  $\sum_{i=1}^n \sigma_i^2$ .

### Proof:

Let  $X$  and  $Y$  be independent normal random variables with  $X$  having mean 0 and variance  $\sigma^2$  and  $Y$  having mean 0 and variance 1. We will determine the density function of  $X + Y$  by utilizing  $f_{X+Y}(a) = \int_{-\infty}^{\infty} f_X(a - y)f_Y(y)dy$ .

# Sums of Independent Random Variables

## Normal Distribution

For independent random variables  $X$  and  $Y$ , the distribution  $f_Z$  of  $Z = X + Y$  equals the convolution of  $f_X$  and  $f_Y$ :

$$f_Z(z) = \int_{-\infty}^{\infty} f_Y(z - x) f_X(x) dx$$

Given that  $f_X$  and  $f_Y$  are normal densities,

$$f_X(x) = \mathcal{N}(x; \mu_X, \sigma_X^2) = \frac{1}{\sigma_X \sqrt{2\pi}} \exp \left\{ -\frac{(x - \mu_X)^2}{2\sigma_X^2} \right\}$$

$$f_Y(y) = \mathcal{N}(y; \mu_Y, \sigma_Y^2) = \frac{1}{\sigma_Y \sqrt{2\pi}} \exp \left\{ -\frac{(y - \mu_Y)^2}{2\sigma_Y^2} \right\}$$

# Sums of Independent Random Variables

## Normal Distribution

$$\begin{aligned}
 f_Z(z) &= \int_{-\infty}^{\infty} \frac{1}{\sigma_Y \sqrt{2\pi}} \exp\left[-\frac{(z-x-\mu_Y)^2}{2\sigma_Y^2}\right] \frac{1}{\sigma_X \sqrt{2\pi}} \exp\left[-\frac{(x-\mu_X)^2}{2\sigma_X^2}\right] dx \\
 &= \int_{-\infty}^{\infty} \frac{1}{\sigma_X \sqrt{2\pi} \sigma_Y \sqrt{2\pi}} \exp\left[-\frac{\sigma_X^2(z-x-\mu_Y)^2 + \sigma_Y^2(x-\mu_X)^2}{2\sigma_X^2 \sigma_Y^2}\right] dx \\
 &= \int_{-\infty}^{\infty} \frac{1}{\sigma_X \sqrt{2\pi} \sigma_Y \sqrt{2\pi}} \exp\left[-\frac{\sigma_X^2(z^2 + x^2 + \mu_Y^2 - 2xz - 2z\mu_Y + 2x\mu_Y) + \sigma_Y^2(x^2 + \mu_X^2 - 2x\mu_X)}{2\sigma_X^2 \sigma_Y^2}\right] dx \\
 &= \int_{-\infty}^{\infty} \frac{1}{\sigma_X \sqrt{2\pi} \sigma_Y \sqrt{2\pi}} \exp\left[-\frac{x^2(\sigma_X^2 + \sigma_Y^2) - 2x(\sigma_X^2(z - \mu_Y) + \sigma_Y^2\mu_X) + \sigma_X^2(z^2 + \mu_Y^2 - 2z\mu_Y) + \sigma_Y^2\mu_X^2}{2\sigma_X^2 \sigma_Y^2}\right] dx
 \end{aligned}$$

Let  $\sigma_Z = \sqrt{\sigma_X^2 + \sigma_Y^2}$

$$f_Z(z) = \int_{-\infty}^{\infty} \frac{1}{\sigma_Z \sqrt{2\pi}} \frac{1}{\frac{\sigma_X \sigma_Y}{\sigma_Z} \sqrt{2\pi}} \exp\left[-\frac{\frac{x^2(\sigma_Z^2)}{\sigma_Z^2} - 2x \frac{\sigma_X^2(z - \mu_Y) + \sigma_Y^2\mu_X}{\sigma_Z^2} + \frac{\sigma_X^2(z^2 + \mu_Y^2 - 2z\mu_Y) + \sigma_Y^2\mu_X^2}{\sigma_Z^2}}{2\left(\frac{\sigma_X \sigma_Y}{\sigma_Z}\right)^2}\right] dx$$

# Sums of Independent Random Variables

## Normal Distribution

$$\begin{aligned}
 f_Z(z) &= \int_{-\infty}^{\infty} \frac{1}{\sigma_Z \sqrt{2\pi}} \frac{1}{\frac{\sigma_X \sigma_Y}{\sigma_Z} \sqrt{2\pi}} \exp \left[ -\frac{x^2 - 2x \frac{\sigma_X^2(z - \mu_Y) + \sigma_Y^2 \mu_X}{\sigma_Z^2} + \frac{\sigma_X^2(z^2 + \mu_Y^2 - 2z\mu_Y) + \sigma_Y^2 \mu_X^2}{\sigma_Z^2}}{2 \left( \frac{\sigma_X \sigma_Y}{\sigma_Z} \right)^2} \right] dx \\
 &= \int_{-\infty}^{\infty} \frac{1}{\sigma_Z \sqrt{2\pi}} \frac{1}{\frac{\sigma_X \sigma_Y}{\sigma_Z} \sqrt{2\pi}} \exp \left[ -\frac{\left( x - \frac{\sigma_X^2(z - \mu_Y) + \sigma_Y^2 \mu_X}{\sigma_Z^2} \right)^2 - \left( \frac{\sigma_X^2(z - \mu_Y) + \sigma_Y^2 \mu_X}{\sigma_Z^2} \right)^2 + \frac{\sigma_X^2(z^2 + \mu_Y^2 - 2z\mu_Y) + \sigma_Y^2 \mu_X^2}{\sigma_Z^2}}{2 \left( \frac{\sigma_X \sigma_Y}{\sigma_Z} \right)^2} \right] dx \\
 &= \int_{-\infty}^{\infty} \frac{1}{\sigma_Z \sqrt{2\pi}} \exp \left[ -\frac{-(\sigma_X^2(z - \mu_Y) + \sigma_Y^2 \mu_X)^2 + \sigma_Z^2(\sigma_X^2(z^2 + \mu_Y^2 - 2z\mu_Y) + \sigma_Y^2 \mu_X^2)}{2\sigma_Z^2(\sigma_X \sigma_Y)^2} \right] \frac{1}{\frac{\sigma_X \sigma_Y}{\sigma_Z} \sqrt{2\pi}} \exp \left[ -\frac{\left( x - \frac{\sigma_X^2(z - \mu_Y) + \sigma_Y^2 \mu_X}{\sigma_Z^2} \right)^2}{2 \left( \frac{\sigma_X \sigma_Y}{\sigma_Z} \right)^2} \right] dx \\
 &= \frac{1}{\sigma_Z \sqrt{2\pi}} \exp \left[ -\frac{(z - (\mu_X + \mu_Y))^2}{2\sigma_Z^2} \right] \int_{-\infty}^{\infty} \frac{1}{\frac{\sigma_X \sigma_Y}{\sigma_Z} \sqrt{2\pi}} \exp \left[ -\frac{\left( x - \frac{\sigma_X^2(z - \mu_Y) + \sigma_Y^2 \mu_X}{\sigma_Z^2} \right)^2}{2 \left( \frac{\sigma_X \sigma_Y}{\sigma_Z} \right)^2} \right] dx
 \end{aligned}$$

# Sums of Independent Random Variables

## Normal Distribution

$$f_Z(z) = \frac{1}{\sigma_Z \sqrt{2\pi}} \exp \left[ -\frac{(z - (\mu_X + \mu_Y))^2}{2\sigma_Z^2} \right] \int_{-\infty}^{\infty} \frac{1}{\frac{\sigma_X \sigma_Y}{\sigma_Z} \sqrt{2\pi}} \exp \left[ -\frac{\left( x - \frac{\sigma_X^2(z - \mu_Y) + \sigma_Y^2 \mu_X}{\sigma_Z} \right)^2}{2 \left( \frac{\sigma_X \sigma_Y}{\sigma_Z} \right)^2} \right] dx$$

The expression in the integral is a normal density distribution on  $x$ , and so the integral evaluates to 1. Therefore, ...

$$f_Z(z) = \frac{1}{\sigma_Z \sqrt{2\pi}} \exp \left[ -\frac{(z - (\mu_X + \mu_Y))^2}{2\sigma_Z^2} \right]$$



# Sums of Independent Random Variables

## Normal Distribution

### • 範例二

假設籃球比賽一季有44場比賽: 26場與一級球隊比賽; 18場與二級球隊比賽。假設與一級球隊比賽的勝率為0.4, 而與二級球隊比賽的勝率為0.7, 而每一場比賽接為獨立事件, 則:

- (1) 某一球隊贏25場的機率為何?
- (2) 某一球隊贏一級球隊多於二級球隊的機率為何?

# Sums of Independent Random Variables

## Normal Distribution

**Solution:**

(1) 某一球隊贏25場的機率為何？

令  $X_A$  與  $X_B$  分別為贏一級球隊與二級球隊的比賽場數，且  $X_A$  與  $X_B$  為 independent binomial random variable。

$$E[X_A] = 26 \times 0.4 = 10.4; \text{Var}(X_A) = 26 \times 0.4 \times 0.6 = 6.24$$

$$E[X_B] = 18 \times 0.7 = 12.6; \text{Var}(X_B) = 18 \times 0.7 \times 0.3 = 3.78$$

此時，我們可以用 normal approximation 來逼近 binomial， $X_A$  與  $X_B$  會逼近 normal distribution 相同參數  $(\mu, \sigma^2)$  的分布。

依據 **Proposition 1** 可以得知， $X_A + X_B$  所逼近的常態分配參數為  $(\mu = 10.4 + 12.6 = 23, \sigma^2 = 6.24 + 3.78 = 10.02)$

# Sums of Independent Random Variables

## Normal Distribution

$$P\{X_A + X_B \geq 25\} = P\{X_A + X_B \geq 24.5\}$$

$$= P\left\{\frac{X_A + X_B - 23}{\sqrt{10.02}} \geq \frac{24.5 - 23}{\sqrt{10.02}}\right\} \approx P\left\{Z \geq \frac{1.5}{\sqrt{10.02}}\right\}$$

$$\approx 1 - P\{Z < 0.4739\} \approx 0.3178$$

# Sums of Independent Random Variables

## Normal Distribution

(2) 某一球隊贏一級球隊多於二級球隊的機率為何？

此提要求的就是  $P\{X_A - X_B \geq 1\}$ ，對於  $X_A - X_B$  來說，其 normal distribution 的參數為  $(\mu = 10.4 - 12.6 = -2.2, \sigma^2 = 6.24 + 3.78 = 10.02)$

$$\because \text{Var}(X - Y) = \text{Var}(X + (-Y)) = \text{Var}(X) + \text{Var}(-Y),$$

$$\text{since } \text{Var}(-Y) = \text{Var}(Y)$$

$$\therefore \text{Var}(X - Y) = \text{Var}(X) + \text{Var}(Y)$$

$$P\{X_A - X_B \geq 1\} = P\{X_A + X_B \geq 0.5\}$$

$$= P\left\{\frac{X_A - X_B - (-2.2)}{\sqrt{10.02}} \geq \frac{0.5 - (-2.2)}{\sqrt{10.02}}\right\} \approx P\left\{Z \geq \frac{2.7}{\sqrt{10.02}}\right\}$$

$$\approx 1 - P\{Z < 0.8530\} \approx 0.1968$$

# Sums of Independent Random Variables

## Lognormal Distribution

如果 $\log(Y)$ 為一個normal random variable  $(\mu, \sigma^2)$ ，則 $Y$ 就可以被定義為一個lognormal random variable  $(\mu, \sigma)$ 。  $Y$ 可以被表示為

$$Y = e^X$$

where  $X$  is a normal random variable

# Sums of Independent Random Variables

## Normal Distribution

### • 範例三

從某個固定時間開始，令  $S(n)$  表示在  $n$  週結束時某種證券的價格， $n \geq 1$ 。常見的價格預測方式為價格比  $S(n)/S(n-1)$ ， $n \geq 1$ ，為 independent and identical distributed lognormal random variables。假設其  $\mu = 0.0165$ ， $\sigma = 0.0730$ ，是求以下的機率：

- (1) 證券價格在未來兩周內都會增長
- (2) 兩週後的價格高於今天的價格

# Sums of Independent Random Variables

## Normal Distribution

**Solution:**

(1) 證券價格在未來兩周內都會增長

令  $Z$  為 standard normal random variable，為了解第一題，我們可以利用題目的線索，也就是持續增長，亦即  $\log(x)$  必須隨著  $x$  增加而增加，所以  $x$  必須大於 1，也可以表示為  $\log(x) > \log(1) \Rightarrow \log(x) - \log(1) > 0$ 。

$$\begin{aligned} P\left\{\frac{S(1)}{S(0)} > 1\right\} &= P\left\{\log\left(\frac{S(1)}{S(0)}\right) > 0\right\} = P\left\{Z > \frac{0 - 0.0165}{0.0730}\right\} \\ &= P\{Z < 0.2260\} = 0.5894; \text{ each of next two weeks} = (0.5894)^2 \\ &= 0.3474 \end{aligned}$$

# Sums of Independent Random Variables

## Normal Distribution

(2) 兩週後的價格高於今天的價格

$$P\left\{\frac{S(2)}{S(0)} > 1\right\} = P\left\{\frac{S(2)S(1)}{S(1)S(0)} > 1\right\} = P\left\{\log\left(\frac{S(2)}{S(1)}\right) + \log\left(\frac{S(1)}{S(0)}\right) > 0\right\}$$

這就可以視為兩個independent normal random variable相加，所以相加後的normal random variable ( $\mu = 0.0165 + 0.0165 = 0.0330$ ,  $\sigma^2 = 2 \times (0.0730)^2$ )

$$P\left\{\frac{S(2)}{S(0)} > 1\right\} = P\left\{Z > \frac{0 - 0.0330}{0.0730\sqrt{2}}\right\} = P\{Z < 0.31965\} = 0.6254$$



# Sums of Independent Random Variables

## Poisson and Binomial Distribution

### • 範例四

如果 $X$ 與 $Y$ 為independent Poisson random variable，其參數分別為 $\lambda_1$ 與 $\lambda_2$ ，是求出 $X + Y$ 的*p.m.f.*:

### **Solution:**

我們可以假設 $\{X + Y = n\}$ ，也可以拆成 $\{X = k, Y = n - k\}, 0 \leq k \leq n$ ，則

$$\begin{aligned} P\{X + Y = n\} &= \sum_{k=0}^n P\{X = k, Y = n - k\} = \sum_{k=0}^n P\{X = k\}P\{Y = n - k\} \\ &= \sum_{k=0}^n e^{-\lambda_1} \frac{\lambda_1^k}{k!} e^{-\lambda_2} \frac{\lambda_2^{n-k}}{(n-k)!} = e^{-(\lambda_1 + \lambda_2)} \sum_{k=0}^n \frac{\lambda_1^k \lambda_2^{n-k}}{k! (n-k)!} \end{aligned}$$

# Sums of Independent Random Variables

## Poisson and Binomial Distribution

$$\begin{aligned} &= \sum_{k=0}^n e^{-\lambda_1} \frac{\lambda_1^k}{k!} e^{-\lambda_2} \frac{\lambda_2^{n-k}}{(n-k)!} = e^{-(\lambda_1+\lambda_2)} \sum_{k=0}^n \frac{\lambda_1^k \lambda_2^{n-k}}{k! (n-k)!} \\ &= \frac{e^{-(\lambda_1+\lambda_2)}}{n!} \sum_{k=0}^n \frac{n!}{k! (n-k)!} \lambda_1^k \lambda_2^{n-k} = \frac{e^{-(\lambda_1+\lambda_2)}}{n!} \sum_{k=0}^n \binom{n}{k} \lambda_1^k \lambda_2^{n-k} \\ &= \frac{e^{-(\lambda_1+\lambda_2)}}{n!} (\lambda_1 + \lambda_2)^n \end{aligned}$$

故  $X + Y$  為一個 Poisson distribution  $(\lambda_1 + \lambda_2)$

# Sums of Independent Random Variables

## Poisson and Binomial Distribution

### • 範例五

如果 $X$ 與 $Y$ 為independent binomial random variables，其參數分別為 $(n, p)$ 與 $(m, p)$ ，試問 $X + Y$ 的*p.m.f.*為何？

### **Solution:**

$X + Y$ 的binomial random variable的參數為 $(n + m, p)$ ，故：

$$\begin{aligned} P\{X + Y = k\} &= \sum_{i=0}^k P\{X = i, Y = k - i\} = \sum_{i=0}^k P\{X = i\}P\{Y = k - i\} \\ &= \sum_{i=0}^k \binom{n}{i} p^i q^{n-i} \binom{m}{k-i} p^{k-i} q^{m-k+i}, \end{aligned}$$

# Sums of Independent Random Variables

## Poisson and Binomial Distribution

$$P\{X + Y\} = \sum_{i=0}^n \binom{n}{i} p^i q^{n-i} \binom{m}{k-i} p^{k-i} q^{m-k+i}$$
$$= p^{i+k-i} q^{n-i+m-k+i} \sum_{i=0}^n \binom{n}{i} \binom{m}{k-i} = p^k q^{n+m-k} \sum_{i=0}^n \binom{n}{i} \binom{m}{k-i}$$

$$\Rightarrow P\{X + Y\} = p^k q^{n+m-k} \sum_{i=0}^n \binom{n}{i} \binom{m}{k-i}$$

since  $P\{X + Y\} \sim \text{binomial}(n + m, k)$ , thus  $\Rightarrow$

$$\sum_{i=0}^n \binom{n}{i} \binom{m}{k-i} = \binom{n+m}{k}$$

# Sums of Independent Random Variables

## Geometric Distribution

### • 範例六

令  $X_1, \dots, X_n$  為 independent geometric random variables，且其機率為  $p_i$  for  $i = 1, \dots, n$ 。試問  $S_n = \sum_{i=1}^n X_i$  的 *p.m.f.* 為何？

### **Solution:**

我們可以試想一個投擲硬幣的情境， $X_n$  就可以想做有  $n$  枚硬幣，針對第  $i$  枚硬幣出現正面的機率為  $p_i$ ，於是乎  $S_n = \sum_{i=1}^n X_i$  就可以被當作投擲了  $n$  次正面的機率，所以就可以被視為 negative binomial random variable。

$$P\{S_n = k\} = \binom{k-1}{n-1} p^n (1-p)^{k-n}, k \geq n$$

# Sums of Independent Random Variables

## Geometric Distribution

$$P\{S_n = k\} = \binom{k-1}{n-1} p^n (1-p)^{k-n}, k \geq n$$

為了歸納其 $S_n$ 的*p.m.f.*，假設所有的 $p_i$ 都不一樣。我們先考慮 $n = 2$ 的情形，令 $q_j = 1 - p_j, j = 1, 2$ ，則

$$P\{S_2 = k\} = \sum_{j=1}^{k-1} P\{X_1 = j, X_2 = k - j\} = \sum_{j=1}^{k-1} P\{X_1 = j\}P\{X_2 = k - j\}$$

$$= \sum_{j=1}^{k-1} p_1 q_1^{j-1} p_2 q_2^{k-j-1} = p_1 p_2 q_2^{k-2} \sum_{j=1}^{k-1} \left(\frac{q_1}{q_2}\right)^{j-1}$$

$$= p_1 p_2 q_2^{k-2} \frac{1 - (q_1/q_2)^{k-1}}{1 - q_1/q_2} = p_1 p_2 q_2^{k-2} q_2 \frac{1 - (q_1/q_2)^{k-1}}{q_2 - q_1}$$

# Sums of Independent Random Variables

## Geometric Distribution

$$\begin{aligned}
 P\{S_2 = k\} &= p_1 p_2 q_2^{k-2} q_2 \frac{1 - (q_1/q_2)^{k-1}}{q_2 - q_1} = p_1 p_2 q_2^{k-1} \frac{1 - (q_1/q_2)^{k-1}}{q_2 - q_1} \\
 &= \frac{p_1 p_2 q_2^{k-1}}{q_2 - q_1} - \frac{p_1 p_2 q_2^{k-1} \left(\frac{q_1}{q_2}\right)^{k-1}}{q_2 - q_1} = \frac{p_1 p_2 q_2^{k-1}}{q_2 - q_1} - \frac{p_1 p_2 q_1^{k-1}}{q_2 - q_1} \\
 &= p_2 q_2^{k-1} \frac{p_1}{q_2 - q_1} - p_1 q_1^{k-1} \frac{p_2}{q_2 - q_1} \\
 &= p_2 q_2^{k-1} \frac{p_1}{p_1 - p_2} - p_1 q_1^{k-1} \frac{p_2}{p_1 - p_2} \\
 &= p_2 q_2^{k-1} \frac{p_1}{p_1 - p_2} + p_1 q_1^{k-1} \frac{p_2}{p_2 - p_1}
 \end{aligned}$$

# Sums of Independent Random Variables

## Geometric Distribution

那如果是  $n = 3$  的時候， $P\{S_3 = k\}$ ，

$$\begin{aligned} P\{S_3 = k\} &= \sum_{j=1}^{k-1} P\{S_2 = j, X_3 = k - j\} = \sum_{j=1}^{k-1} P\{S_2 = j\}P\{X_3 = k - j\} \\ &= p_1 q_1^{k-1} \frac{p_2}{p_2 - p_1} \frac{p_3}{p_3 - p_1} + p_2 q_2^{k-1} \frac{p_1}{p_1 - p_2} \frac{p_3}{p_3 - p_2} + p_3 q_3^{k-1} \frac{p_1}{p_1 - p_3} \frac{p_2}{p_2 - p_3} \end{aligned}$$



# Sums of Independent Random Variables

## Geometric Distribution

根據剛剛的規律，我們可以得出一個新的proposition。

### • Proposition 2

Let  $X_1, \dots, X_n$  為 independent geometric random variables, with  $X_i$  having parameters  $p_i$ , for  $i = 1, \dots, n$ . If all the  $p_i$  are distinct, then, for  $k \geq n$ ,

$$P\{S_n = k\} = \sum_{i=1}^n p_i q_i^{k-1} \prod_{j \neq i} \frac{p_j}{p_j - p_i}$$

# Conditional Distributions: Discrete Case

- 給定任何兩個事件  $E$  與  $F$ ，則  $F$  成立前提下  $E$  所發生的條件機率 (conditional probability) 為:

$$P(E|F) = \frac{P(EF)}{P(F)}, \text{ where } P(F) > 0$$

- 當今天有兩個 discrete random variable  $X$  與  $Y$ ，在  $Y$  成立前提下  $X$  發生的 *conditional probability mass function* 為

$$p_{X|Y}(x|y) = P\{X = x|Y = y\} = \frac{P\{X = x, Y = y\}}{P\{Y = y\}} = \frac{p(x, y)}{p_Y(y)},$$

where  $p_Y > 0$

# Conditional Distributions: Discrete Case

- 那計算在 $Y$ 成立前提下 $X$ 發生的*cumulative conditional p.m.f.*為

$$F_{X|Y}(x|y) = P\{X \leq x, Y \leq y\} = \sum_{a \leq x} p_{X|Y}(a|y)$$

- 如果 $X$ 與 $Y$ 相互獨立的話，

$$\begin{aligned} p_{X|Y}(x|y) &= P\{X = x|Y = y\} = \frac{P\{X = x, Y = y\}}{P\{Y = y\}} = \frac{P\{X = x\}P\{Y = y\}}{P\{Y = y\}} \\ &= P\{X = x\} \end{aligned}$$

# Conditional Distributions: Discrete Case

- 範例七

令  $p(x, y)$  為  $X$  與  $Y$  的 *joint p.m.f.* .

$$p(0,0) = 0.4; p(0,1) = 0.2; p(1,0) = 0.1; p(1,1) = 0.3$$

請問  $Y = 1$  的前提下  $X$  發生的 *conditional p.m.f.* 為何?

**Solution:**

$$p_Y(1) = \sum_x p(x, 1) = p(0,1) + p(1,1) = 0.2 + 0.3 = 0.5$$

$$p_{X|Y}(0|1) = \frac{p(0,1)}{p_Y(1)} = \frac{2}{5}; p_{X|Y}(1|1) = \frac{p(1,1)}{p_Y(1)} = \frac{3}{5}$$

# Conditional Distributions: Discrete Case

## • 範例八

假設 $X$ 與 $Y$ 皆為independent Poisson random variables，其參數分別為 $\lambda_1$ 與 $\lambda_2$ ，試問在 $X + Y = n$ 的前提下 $X$ 的conditional probability為何？

### **Solution:**

本題要計算 $X + Y = n$ 前提下 $X$ 的conditional probability mass function。

$$\begin{aligned} P\{X = k \mid X + Y = n\} &= \frac{P\{X = k, X + Y = n\}}{P\{X + Y = n\}} = \frac{P\{X = k, Y = n - k\}}{P\{X + Y = n\}} \\ &= \frac{P\{X = k\}P\{Y = n - k\}}{P\{X + Y = n\}} = \frac{e^{-\lambda_1} \lambda_1^k}{k!} \frac{e^{-\lambda_2} \lambda_2^{n-k}}{(n-k)!} \left[ \frac{e^{-(\lambda_1 + \lambda_2)} (\lambda_1 + \lambda_2)^n}{n!} \right]^{-1} \end{aligned}$$

# Conditional Distributions: Discrete Case

$$\begin{aligned} P\{X = k \mid X + Y = n\} &= \frac{e^{-\lambda_1} \lambda_1^k}{k!} \frac{e^{-\lambda_2} \lambda_2^{n-k}}{(n-k)!} \left[ \frac{e^{-(\lambda_1 + \lambda_2)} (\lambda_1 + \lambda_2)^n}{n!} \right]^{-1} \\ &= \frac{n!}{(n-k)! k!} \frac{\lambda_1^k \lambda_2^{n-k}}{(\lambda_1 + \lambda_2)^n} = \binom{n}{k} \left( \frac{\lambda_1}{\lambda_1 + \lambda_2} \right)^k \left( \frac{\lambda_2}{\lambda_1 + \lambda_2} \right)^{n-k} \end{aligned}$$

The conditional distribution of  $X$  given that  $X + Y = n$  is the binomial distribution with parameters  $n$  and  $\frac{\lambda_1}{\lambda_1 + \lambda_2}$

# Conditional Distributions: Continuous Case

- 如果 $X$ 與 $Y$ 的joint probability density function為 $f(x, y)$ ，則在給定 $Y = y$ 的前提下 $X$ 的conditional probability density function可以被定義，對於每一個 $y$ 其 $f_Y(y) > 0$ ，

$$\begin{aligned} f_{X+Y}(x|y)dx &= \frac{f(x, y)dxdy}{f_Y(y)dy} \approx \frac{P\{x \leq X \leq x + dx, y \leq Y \leq y + dy\}}{P\{y \leq y \leq y + dy\}} \\ &= P\{x \leq X \leq x + dx | y \leq Y \leq y + dy\} \end{aligned}$$

- 如果 $X$ 與 $Y$ 是jointly continuous random variable，則對於任何集合 $A$

$$P\{X \in A | Y = y\} = \int_A f_{X|Y}(x|y)dx$$

# Conditional Distributions: Continuous Case

- Let  $A = (-\infty, a]$ , 則在給定  $Y = y$  前提下  $X$  的 conditional cumulative distribution function 為

$$F_{X+Y}(a|y) \equiv P\{X \leq a|Y = y\} = \int_{-\infty}^a f_{X+Y}(x|y)dx$$



# Conditional Distributions: Continuous Case

- 範例九

$X$ 與 $Y$ 的joint density function為

$$f(x, y) = \begin{cases} \frac{12}{5}x(2 - x - y) & 0 < x < 1, 0 < y < 1 \\ 0 & \textit{otherwise} \end{cases}$$

計算 $Y = y$ 前提下 $X$ 的conditional density為何?

**Solution:**

$$\begin{aligned} f_{X+Y}(x|y) &= \frac{f(x, y)}{f_Y(y)} = \frac{f(x, y)}{\int_{-\infty}^{\infty} f(x, y)dx} = \frac{x(2 - x - y)}{\int_0^1 x(2 - x - y)dx} = \frac{x(2 - x - y)}{\frac{2}{3} - \frac{y}{2}} \\ &= \frac{6x(2 - x - y)}{4 - 3y} \end{aligned}$$

# Conditional Distributions: Continuous Case

## • 範例十

假設 $X$ 與 $Y$ 的joint density function為

$$f(x, y) = \begin{cases} \frac{e^{-\frac{x}{y}} e^{-y}}{y} & 0 < x < \infty, 0 < y < \infty \\ 0 & \text{otherwise} \end{cases}$$

試問 $P\{X > 1 | Y = y\}$ 。

**Solution:**

$$f_{X+Y}(x|y) = \frac{f(x, y)}{f_Y(y)} = \frac{\frac{e^{-\frac{x}{y}} e^{-y}}{y}}{e^{-y} \int_0^{\infty} \frac{1}{y} e^{-\frac{x}{y}} dx} = \frac{1}{y} e^{-\frac{x}{y}}$$

# Conditional Distributions: Continuous Case

$$f_{X+Y}(x|y) = \frac{f(x, y)}{f_Y(y)} = \frac{\frac{e^{-\frac{x}{y}} e^{-y}}{y}}{e^{-y} \int_0^{\infty} \frac{1}{y} e^{-\frac{x}{y}} dx} = \frac{1}{y} e^{-\frac{x}{y}}$$

Hence,

$$P\{X > 1 | Y = y\} = \int_1^{\infty} \frac{1}{y} e^{-\frac{x}{y}} dx = -e^{-\frac{x}{y}} \Big|_1^{\infty} = e^{-\frac{1}{y}}$$

# Conditional Distributions: Continuous Case

If  $X$  and  $Y$  are independent continuous random variables, the conditional density of  $X$  given that  $Y = y$  is just the unconditional density of  $X$ . This is so because, in the independent case,

$$f_{X+Y}(x|y) = \frac{f(x, y)}{f_Y(y)} = \frac{f_X(x)f_Y(y)}{f_Y(y)} = f_X(x)$$

# Order Statistics

- Let  $X_1, X_2, \dots, X_n$  be  $n$  independent and identically distributed continuous random variables having a common density  $f$  and distribution function  $F$ . Define

$$X_{(1)} = \text{smallest of } X_1, X_2, \dots, X_n$$

$$X_{(2)} = \text{second smallest of } X_1, X_2, \dots, X_n$$

$$\vdots$$

$$X_{(j)} = \text{jth smallest of } X_1, X_2, \dots, X_n$$

$$\vdots$$

$$X_{(n)} = \text{largest of } X_1, X_2, \dots, X_n$$

# Order Statistics

- The ordered values  $X_{(1)}, X_{(2)}, \dots, X_{(n)}$  are known as the order statistics corresponding to the random variables  $X_1, X_2, \dots, X_n$ . In other words,  $X_{(1)}, X_{(2)}, \dots, X_{(n)}$  are the ordered values of  $X_1, X_2, \dots, X_n$ .
- The joint density function of the ordered statistics is obtained by noting that the order statistics  $X_{(1)}, X_{(2)}, \dots, X_{(n)}$  will take on the values  $x_1 \leq x_2 \leq \dots \leq x_n$  if and only if, for some permutation  $(i_1, i_2, \dots, i_n)$  of  $(1, 2, \dots, n)$ ,
$$X_1 = x_{i_1}, X_2 = x_{i_2}, \dots, X_n = x_{i_n}$$

# Order Statistics

- Since, for any permutation  $(i_1, i_2, \dots, i_n)$  of  $(1, 2, \dots, n)$ ,

$$P \left\{ x_{i_1} - \frac{\varepsilon}{2} < X_1 < x_{i_1} + \frac{\varepsilon}{2}, \dots, x_{i_n} - \frac{\varepsilon}{2} < X_n < x_{i_n} + \frac{\varepsilon}{2} \right\} \\ \approx \varepsilon^n f_{X_1, \dots, X_n} (x_{i_1}, \dots, x_{i_n}) = \varepsilon^n f(x_{i_1}) \dots f(x_{i_n}) = \varepsilon^n f(x_{i_1}) \dots f(x_{i_n})$$

It follows that, for  $x_1 < x_2 < \dots < x_n$ ,

$$P \left\{ x_1 - \frac{\varepsilon}{2} < X_{(1)} < x_1 + \frac{\varepsilon}{2}, \dots, x_n - \frac{\varepsilon}{2} < X_{(n)} < x_n + \frac{\varepsilon}{2} \right\} \\ \approx n! \varepsilon^n f(x_1) \dots f(x_n)$$

Dividing by  $\varepsilon^n$  and letting  $\varepsilon \rightarrow 0$  yields

$$\varepsilon^n f_{X_{(1)}, \dots, X_{(n)}} (x_1, \dots, x_n) = n! \varepsilon^n f(x_1) \dots f(x_n)$$

$$f_{X_{(1)}, \dots, X_{(n)}} (x_1, \dots, x_n) = n! f(x_1) \dots f(x_n), x_1 < x_2 < \dots < x_n$$

# Order Statistics

- 範例十一

有3個人隨機位在(distributed at random)一條1公里的道路上，試問：不存在任兩人相距小於 $d$ 公里( $d < \frac{1}{2}$ )的機率。

**Solution:**

distributed at random這句話可以當作independent and uniformly distributed。令 $X_i$ 為第 $i$ 個人在這條道路上的位置，我們要求的機率就可以表示為 $P\{X_{(i)} > X_{(i-1)} + d, i = 2, 3\}$

$$f_{X_{(1)}, X_{(2)}, X_{(3)}}(x_1, x_2, x_3) = 3!, 0 < x_1 < x_2 < x_3 < 1$$



# Order Statistics

$$P\{X_{(i)} > X_{(i-1)} + d, i = 2, 3\} = \iiint_{x_i > x_{j-1} + d} f_{X_{(1)}, X_{(2)}, X_{(3)}}(x_1, x_2, x_3) dx_1 dx_2 dx_3$$

$$= 3! \int_0^{1-2d} \int_{x_1+d}^{1-d} \int_{x_2+d}^1 dx_3 dx_2 dx_1 = 6 \int_0^{1-2d} \int_{x_1+d}^{1-d} (1 - x_2 - d) dx_2 dx_1$$

let  $(1 - x_2 - d) = y_2$

$$= 6 \int_0^{1-2d} \int_0^{1-d-(x_1+d)=1-2d-x_1} y_2 dy_2 dx_1 = 3 \int_0^{1-2d} (1 - 2d - x_1)^2 dx_1$$

$$= 3 \int_0^{1-2d} y_1^2 dy_1 = (1 - 2d)^3 \Rightarrow \text{if there are } n \text{ people, ... } [1 - (n - 1)d]^n, \text{ where } d < \frac{1}{n - 1}$$

# Order Statistics

- The density function of the  $j$  –  $th$ -order statistics  $X_{(j)}$  can be obtained either by integrating the joint density function (pp.49) or by direct reasoning as follows: In order for  $X_{(j)}$  to equal  $x$ , it is necessary for  $j - 1$  of the  $n$  values  $X_1, \dots, X_n$  to be less than  $x$ ,  $n - j$  of them to be greater than  $x$ , and 1 of them to equal  $x$ .
- Now, the probability density that any given set of  $j - 1$  of the  $X_i$ 's are less than  $x$ , another given set of  $n - j$  are all greater than  $x$ , and the remaining value is equal to  $x$  equals.

$$[F(x)]^{j-1}[1 - F(x)^{n-j}]f(x)$$

Therefore,

$$\binom{n}{j-1, n-j, 1} = \frac{n!}{(n-j)!(j-1)!}$$

# Order Statistics

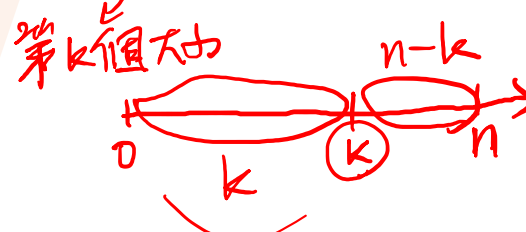
independent

$$F_n(y_n) = P\{\underline{Y}_n \leq y_n\} = P\{\underline{Y}_1 \leq y_n, \underline{Y}_2 \leq y_n, \dots, \underline{Y}_n \leq y_n\} = [F(y_n)]^n$$

differentiation

$$f_n(y_n) = \frac{\partial F_n(y_n)}{\partial y} = n [F(y_n)]^{n-1} f(y_n) \quad \hat{=} \quad y = y_n \text{ 某一個常數 } \geq Y_1 \sim Y_n$$

$$F_k(y) = P\{Y_k < y\} = \sum_{i=k}^n \binom{n}{i} (F(y))^i (1-F(y))^{n-i} = \sum_{i=k}^{n-1} \binom{n}{i} (F(y))^i (1-F(y))^{n-i} + \underline{(F(y))^n}$$



Differentiation  $f_k(y) = \frac{\partial F_k(y)}{\partial y} = \sum_{i=k}^{n-1} \binom{n}{i} i (F(y))^{i-1} f(y) (1-F(y))^{n-i} - \sum_{i=k}^{n-1} \binom{n}{i} (F(y))^i (n-i) (1-F(y))^{n-i-1} f(y) + n (F(y))^{n-1} f(y)$

$$\Rightarrow \sum_{i=k}^{n-1} \frac{n!}{i!(n-i)!} i (F(y))^{i-1} f(y) (1-F(y))^{n-i} - \sum_{i=k}^{n-1} \frac{n!}{i!(n-i)!} (F(y))^i (n-i) (1-F(y))^{n-i-1} f(y) + n (F(y))^{n-1} f(y)$$

# Order Statistics

$$\rightarrow \sum_{\bar{i}=k}^{n-1} \frac{n!}{(\bar{i}-1)!(n-\bar{i})!} (F(y))^{i-1} \cdot f(y) [1-F(y)]^{n-i} = \sum_{\bar{i}=k}^{n-1} \frac{n!}{\bar{i}!(n-\bar{i}-1)!} (F(y))^{\bar{i}} (1-F(y))^{n-\bar{i}-1} \cdot f(y) + n(F(n))^{n-1} \cdot f(y)$$

if  $\bar{i}=k$

$$\frac{n!}{(k-1)!(n-k)!} (F(y))^{k-1} \cdot f(y) [1-F(y)]^{n-k} = \frac{n!}{k!(n-k-1)!} (F(y))^k (1-F(y))^{n-k-1} \cdot f(y) + n(F(n))^{n-1} \cdot f(y)$$

$$\begin{aligned} \bar{i}=k+1 & \frac{n!}{k!(n-k-1)!} (F(y))^k \cdot f(y) [1-F(y)]^{n-k-1} = \frac{n!}{(k+1)!(n-k-2)!} (F(y))^{k+1} (1-F(y))^{n-k-2} \cdot f(y) + n(F(n))^{n-1} \cdot f(y) \\ & \vdots \\ \bar{i}=n-1 & \frac{n!}{(n-2)!(n-n+1)!} (F(y))^{n-2} \cdot f(y) [1-F(y)]^{n-n+1} = \frac{n!}{(n-1)!(n-(n-1)-1)!} (F(y))^{n-1} (1-F(y))^{n-1} \cdot f(y) \end{aligned}$$

$$\rightarrow \frac{n!}{(k-1)!(n-k)!} (F(y))^{k-1} \cdot f(y) [1-F(y)]^{n-k} = n \cdot (F(y))^{n-1} \cdot f(y) + n(F(n))^{n-1} \cdot f(y)$$

$$\rightarrow \frac{n!}{(k-1)!(n-k)!} (F(y))^{k-1} \cdot f(y) [1-F(y)]^{n-k}$$

# Order Statistics

- Different partitions of the  $n$  random variables  $X_1, \dots, X_n$  into the preceding three groups, it follows that the density function of  $X_{(j)}$  is given by

$$f_{X_{(j)}}(x) = \frac{n!}{(n-j)!(j-1)!} [F(x)]^{j-1} [1 - F(x)]^{n-j} f(x)$$

# Order Statistics

- 範例十二

如果今天有一個樣本含有  $2n + 1$  個 random variables ( $2n + 1$  independent and identically distributed random variables)，則第  $n + 1$  小的稱為 sample median。假設今天樣本大小為 3，且為 uniform distribution over  $(0,1)$ ，試問: sample median 落在  $\frac{1}{4}$  與  $\frac{3}{4}$  之間的機率為何？

# Order Statistics

**Solution:**

根據pp.51所導出來的公式得知 $X_{(2)}$

$$f_{X_{(2)}}(x) = \frac{3!}{1!1!} x(1-x), 0 < x < 1$$

Hence,

$$P\left\{\frac{1}{4} < X_{(2)} < \frac{3}{4}\right\} = 6 \int_{1/4}^{3/4} x(1-x) dx = 6 \left\{ \frac{x^2}{2} - \frac{x^3}{3} \right\} \Bigg|_{x=1/4}^{x=3/4} = \frac{11}{16}$$

# Order Statistics

The c.d.f. of  $X_{(j)}$  can be found by integrating pp.51 formula

$$F_{X_{(j)}}(y) = \frac{n!}{(n-j)!(j-1)!} \int_{-\infty}^y [F(x)]^{j-1} [1-F(x)]^{n-j} f(x) dx$$

$$F_{X_{(j)}}(y) = P\{X_{(j)} \leq y\} = P\{j \text{ or more of the } X_i \text{'s are } \leq y\}$$

$$= \sum_{k=j}^n \binom{n}{k} [F(y)]^k [1-F(y)]^{n-k}$$

We take  $F$  to be the uniform (0,1) distribution [that is,  $f(x) = 1, 0 < x < 1$ ], then we obtain the interesting analytical identity

$$\sum_{k=j}^n \binom{n}{k} y^k (1-y)^{n-k} = \frac{n!}{(n-j)!(j-1)!} \int_0^y x^{j-1} (1-x)^{n-j} dx, \text{ where } 0 \leq y \leq 1$$



# Order Statistics

Referring to pp.51

$$f_{X_{(j)}}(x) = \frac{n!}{(n-j)!(j-1)!} [F(x)]^{j-1} [1 - F(x)]^{n-j} f(x)$$

We can obtain...

$$f_{X_{(i)}, X_{(j)}}(x_i, x_j) = \frac{n!}{(i-1)!(j-i-1)!(n-j)!} [F(x_i)]^{i-1} \times [F(x_j) - F(x_i)]^{j-i-1} [1 - F(x_j)]^{n-j} f(x_i) f(x_j)$$

for all  $x_i < x_j$

# Joint Probability Distribution of Functions of Random Variables

- Let  $X_1$  and  $X_2$  be jointly continuous random variables with joint probability density function  $f_{X_1, X_2}$ . It is sometimes necessary to obtain the joint distribution of the random variables  $Y_1$  and  $Y_2$ , which arise as functions of  $X_1$  and  $X_2$ . Specifically, suppose that  $Y_1 = g_1(X_1, X_2)$  and  $Y_2 = g_2(X_1, X_2)$  for some functions  $g_1$  and  $g_2$ .

# Joint Probability Distribution of Functions of Random Variables

- Assume that the functions  $g_1$  and  $g_2$  satisfy the following conditions:
  1. The equations  $y_1 = g_1(x_1, x_2)$  and  $y_2 = g_2(x_1, x_2)$  can be uniquely solved for  $x_1$  and  $x_2$  in terms of  $y_1$  and  $y_2$ , with solutions given by, say,  $x_1 = h_1(y_1, y_2)$ ,  $x_2 = h_2(y_1, y_2)$ .
  2. The functions  $g_1$  and  $g_2$  have continuous partial derivatives at all points  $(x_1, x_2)$  and are such that the  $2 \times 2$  determinant.

# Joint Probability Distribution of Functions of Random Variables

$$J(x_1, x_2) = \begin{vmatrix} \frac{\partial g_1}{\partial x_1} & \frac{\partial g_1}{\partial x_2} \\ \frac{\partial g_2}{\partial x_1} & \frac{\partial g_2}{\partial x_2} \end{vmatrix} \equiv \frac{\partial g_1}{\partial x_1} \frac{\partial g_2}{\partial x_2} - \frac{\partial g_1}{\partial x_2} \frac{\partial g_2}{\partial x_1} \neq 0, \text{ at all point } (x_1, x_2)$$

- Under these two conditions, it can be shown that the random variables  $Y_1$  and  $Y_2$  are jointly continuous with joint density function given by  $f_{Y_1 Y_2}(y_1, y_2) = f_{X_1, X_2}(x_1, x_2) |J(x_1, x_2)|^{-1}$ , where  $x_1 = h_1(y_1, y_2)$ ,  $x_2 = h_2(y_1, y_2)$
- A proof of above equation would proceed along the following lines

$$P\{Y_1 \leq y_1, Y_2 \leq y_2\} = \iint_{\substack{(x_1, x_2): \\ g_1(x_1, x_2) \leq y_1 \\ g_2(x_1, x_2) \leq y_2}} f_{X_1, X_2}(x_1, x_2) dx_1 dx_2$$

# Joint Probability Distribution of Functions of Random Variables

- 範例十三

Let  $X_1$  and  $X_2$  be joint continuous random variables with probability density function  $f_{X_1, X_2}$ . Let  $Y_1 = X_1 + X_2$ ,  $Y_2 = X_1 - X_2$ . Find the joint density function of  $Y_1$  and  $Y_2$  in terms of  $f_{X_1, X_2}$ .

**Solution:**

Let  $g_1(x_1, x_2) = x_1 + x_2$  and  $g_2(x_1, x_2) = x_1 - x_2$ .

$$J(x_1, x_2) = \begin{vmatrix} 1 & 1 \\ 1 & -1 \end{vmatrix} = -2$$

# Joint Probability Distribution of Functions of Random Variables

Also, since the equations  $y_1 = x_1 + x_2$  and  $y_2 = x_1 - x_2$  have  $x_1 = (y_1 + y_2)/2, x_2 = (y_1 - y_2)/2$  as their solution, it follows from

$$f_{Y_1 Y_2}(y_1, y_2) = f_{X_1, X_2}(x_1, x_2) |J(x_1, x_2)|^{-1}, \text{ where } J(x_1, x_2) = -2$$

Then

$$f_{Y_1 Y_2}(y_1, y_2) = \frac{1}{2} f_{X_1, X_2} \left( \frac{y_1 + y_2}{2}, \frac{y_1 - y_2}{2} \right)$$

For instance, if  $X_1$  and  $X_2$  are independent uniform (0,1) random variables, then

$$f_{Y_1 Y_2}(y_1, y_2) = \begin{cases} \frac{1}{2} & 0 \leq y_1 + y_2 \leq 2, 0 \leq y_1 - y_2 \leq 2 \\ 0 & \text{otherwise} \end{cases}$$

# Joint Probability Distribution of Functions of Random Variables

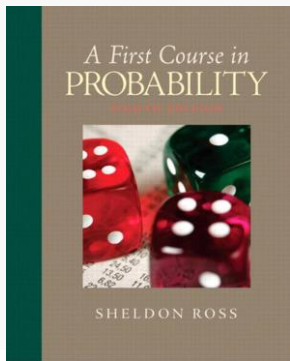
Or if  $X_1$  and  $X_2$  are independent **exponential random variables** with respective parameters  $\lambda_1$  and  $\lambda_2$ , then

$$f_{Y_1 Y_2}(y_1, y_2) = \begin{cases} \frac{\lambda_1 \lambda_2}{2} \exp\left\{-\lambda_1 \left(\frac{y_1 + y_2}{2}\right) - \lambda_2 \left(\frac{y_1 - y_2}{2}\right)\right\} & y_1 + y_2 \geq 0, y_1 - y_2 \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

Finally, if  $X_1$  and  $X_2$  are independent **standard normal random variables**, then

$$f_{Y_1 Y_2}(y_1, y_2) = \frac{1}{4\pi} e^{-\left[\frac{(y_1 + y_2)^2}{8} + \frac{(y_1 - y_2)^2}{8}\right]} = \frac{1}{4\pi} e^{-\frac{(y_1^2 + y_2^2)}{4}} = \frac{1}{\sqrt{4\pi}} e^{-\frac{y_1^2}{4}} \frac{1}{\sqrt{4\pi}} e^{-\frac{y_2^2}{4}}$$

# [#12] Assignment



- Selected Problems from Sheldon Ross Textbook [1].

**6.8.** The joint probability density function of  $X$  and  $Y$  is given by

$$f(x, y) = c(y^2 - x^2)e^{-y} \quad -y \leq x \leq y, 0 < y < \infty$$

- (a) Find  $c$ .
- (b) Find the marginal densities of  $X$  and  $Y$ .
- (c) Find  $E[X]$ .

**6.9.** The joint probability density function of  $X$  and  $Y$  is given by

$$f(x, y) = \frac{6}{7} \left( x^2 + \frac{xy}{2} \right) \quad 0 < x < 1, 0 < y < 2$$

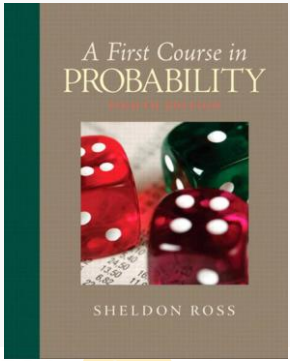
- (a) Verify that this is indeed a joint density function.
- (b) Compute the density function of  $X$ .
- (c) Find  $P\{X > Y\}$ .
- (d) Find  $P\{Y > \frac{1}{2} | X < \frac{1}{2}\}$ .
- (e) Find  $E[X]$ .
- (f) Find  $E[Y]$ .

**6.12.** The number of people that enter a drugstore in a given hour is a Poisson random variable with parameter  $\lambda = 10$ . Compute the conditional probability that at most 3 men entered the drugstore, given that 10 women entered in that hour. What assumptions have you made?

[1] Sheldon Ross. [A First of Course in Probability](#). 8th edition.



# [#12] Assignment



6.40. The joint probability mass function of  $X$  and  $Y$  is given by

$$\begin{aligned} p(1,1) &= \frac{1}{8} & p(1,2) &= \frac{1}{4} \\ p(2,1) &= \frac{1}{8} & p(2,2) &= \frac{1}{2} \end{aligned}$$

- (a) Compute the conditional mass function of  $X$  given  $Y = i, i = 1, 2$ .
- (b) Are  $X$  and  $Y$  independent?
- (c) Compute  $P\{XY \leq 3\}, P\{X + Y > 2\}, P\{X/Y > 1\}$ .

6.41. The joint density function of  $X$  and  $Y$  is given by

$$f(x,y) = xe^{-x(y+1)} \quad x > 0, y > 0$$

- (a) Find the conditional density of  $X$ , given  $Y = y$ , and that of  $Y$ , given  $X = x$ .
- (b) Find the density function of  $Z = XY$ .

# Reference

Ross, S. (2010). *A first course in probability*. Pearson.

# The End

*If you have any questions, please do not hesitate to ask me.*

*Thank you for your attention ))*