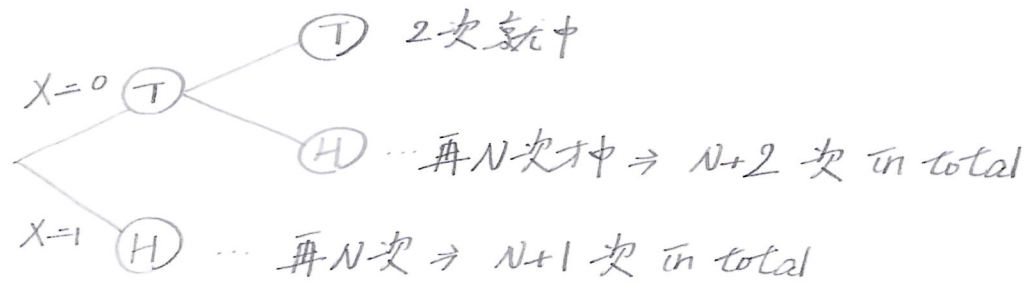


1. 試問 預期至少需要擲多少次公正硬幣才會連續出現 2 次數字?  
(expected  $N$  tosses required) (fair coin) (tails)

$$X = \begin{cases} 0, & 1^{\text{st}} \text{ toss is tails} \\ 1, & 1^{\text{st}} \text{ toss is head} \end{cases}$$

利用 law of total expectation  $E(Y) = \sum_x E(Y|X=x) \cdot P(X=x)$

sample space  $S = \{(r_1, r_2, \dots, r_n), n \geq 2\}, r_i \in \{\text{heads, tails}\}$



又知道 1<sup>st</sup> 結果

$$\begin{aligned} \Rightarrow E(N) &= E(E(N|X)) = \sum_x E(N|X=x) \cdot P(X=x) \\ &= E(N|X=0) \cdot P(X=0) + E(N|X=1) \cdot P(X=1) \end{aligned}$$

再拆成 2 項

$$\begin{aligned} &= \left[ E(2) \cdot \frac{1}{2} + E(N+2) \cdot \frac{1}{2} \right] \cdot \frac{1}{2} + E(N+1) \cdot \frac{1}{2} \\ &= 2 \cdot \frac{1}{4} + E(N) \cdot \frac{1}{4} + 2 \cdot \frac{1}{4} + E(N) \cdot \frac{1}{2} + 1 \cdot \frac{1}{2} \end{aligned}$$

$$\Rightarrow E(N) = \frac{3}{4} E(N) + \frac{3}{2}$$

$$\Rightarrow \frac{1}{4} E(N) = \frac{3}{2}$$

$$\Rightarrow E(N) = 6 \quad \therefore \text{the expected number of the tosses required is } 6.$$

\*

2. 一副牌 52 張, 有 4 張 Ace, 隨機分 4 个 pile, 1 个 pile 13 張

定義 4 个 event  $E_1 = \{1^{st} \text{ pile 有剛好 1 張 Ace}\}$

$$E_2 = \{2^{nd} \text{ "}\}$$

$$E_3 = \{3^{rd} \text{ "}\}$$

$$E_4 = \{4^{th} \text{ "}\}$$

試求  $P\{E_1 E_2 E_3 E_4\}$ ?

$$\Rightarrow P(E_1 E_2 E_3 E_4) = P(E_1) P(E_2|E_1) P(E_3|E_1 E_2) P(E_4|E_1 E_2 E_3)$$

$$P(E_1) = \frac{C_1^4 \cdot C_{12}^{48}}{C_{13}^{52}} = \frac{(4 \text{ 張 Ace 任選 1}) \cdot (\text{扣除 Ace 的 48 張任選 2})}{(52 \text{ 張任選 13})}$$

$$P(E_2|E_1) = \frac{C_1^3 \cdot C_{12}^{36}}{C_{13}^{49}}$$

$$P(E_3|E_1 E_2) = \frac{C_1^2 \cdot C_{12}^{24}}{C_{13}^{26}}$$

$$P(E_4|E_1 E_2 E_3) = \frac{C_1^1 \cdot C_{12}^{12}}{C_{13}^{13}} = 1$$

$$\Rightarrow P(E_1 E_2 E_3 E_4) = \frac{C_1^4 C_{12}^{48}}{C_{13}^{52}} \cdot \frac{C_1^3 C_{12}^{36}}{C_{13}^{49}} \cdot \frac{C_1^2 C_{12}^{24}}{C_{13}^{26}} \cdot 1$$

$$= \frac{4 \cdot \frac{48!}{12! \cdot 36!} \cdot 3 \cdot \frac{36!}{12! \cdot 24!} \cdot 2 \cdot \frac{24!}{12! \cdot 12!}}{\frac{52!}{13! \cdot 39!} \cdot \frac{49!}{13! \cdot 36!} \cdot \frac{26!}{13! \cdot 13!}} = \frac{4! \cdot 48! \cdot (13!)^4}{52! \cdot (12!)^4}$$

$$= \frac{24 \cdot 13^4}{49 \cdot 50 \cdot 51 \cdot 52} = \frac{2197}{20825}$$

$$\approx 0.1055$$

3. 擲2顆公正6面骰子, 試問 (a) 至少1个6 的机率? (b) 2面值不同的情況下, 至少1个6 的机率?

$$(a) P_{\text{至少1个6}} = 1 - P_{\text{都没6}} = 1 - \frac{5}{6} \cdot \frac{5}{6} = \frac{11}{36} \quad *$$

$$(b) P(\text{至少1个6} | \text{2面不同}) = \frac{P(A \cap B)}{P(B)} = \frac{\frac{1}{6} \cdot \frac{5}{6} \cdot 2!}{1 - \frac{6}{36}} = \frac{1}{3} \quad *$$

4.  $Y_1, Y_2, \dots, Y_n$  iid r.v.  $\sim U[0, 1]$ .  $X = \max\{Y_1, \dots, Y_n\}$

(a) 証証 cdf  $F_X(x) = x^n$ ,  $0 \leq x \leq 1$

$$\text{由定義 } F_X(x) \triangleq P\{X \leq x\} = P\{\max\{Y_1, Y_2, \dots, Y_n\} \leq x\}$$

$$= P\{Y_1 \leq x\} \cdot P\{Y_2 \leq x\} \cdots P\{Y_n \leq x\} \quad (\because \text{iid})$$

$$= F_{Y_1}(x) \cdots F_{Y_n}(x) = \int_0^x \frac{1}{1-0} dx \cdots \int_0^x \frac{1}{1-0} dx$$

$$= x \cdots x = x^n \quad \text{証証} \quad *$$

(b) 試問 pdf  $f_X(x) = ?$

$$f_X(x) = \frac{d}{dx} F_X(x) = \frac{d}{dx} x^n = nx^{n-1} \quad *$$

Note.  $X \sim U[a, b]$  is continuous r.v.

$$\text{pdf } f_X(x) = \begin{cases} \frac{1}{b-a}, & a \leq x \leq b \\ 0, & \text{elsewhere} \end{cases}$$

$$\text{cdf } F_X(x) = \begin{cases} 0, & x < a \\ \frac{x-a}{b-a}, & a \leq x < b \\ 1, & x \geq b \end{cases}$$

$$\text{mgf } M_X(t) = E[e^{tX}] = \frac{e^{tb} - e^{ta}}{t(b-a)}$$

$$\text{mean } E(X) = \frac{a+b}{2}$$

$$\text{variance } \text{var}(X) = \frac{(b-a)^2}{12}$$

$$P(c \leq X \leq d) = F(d) - F(c)$$

$$= \int_c^d \frac{1}{b-a} dx = \frac{d-c}{b-a}$$

5. 推導 (a) Uniform distribution 的 mgf  $M_X(t)$ , 並利用 mgf 算出 mean, 2<sup>nd</sup> moment, variance.

(a) consider a continuous r.v.  $X \sim U[a, b]$

$$\begin{aligned} 1^\circ \phi(t) &= E(e^{tX}) = \int_{-\infty}^{\infty} e^{tx} \cdot f_X(x) dx = \int_a^b e^{tx} \frac{1}{b-a} dx \\ &= \frac{1}{b-a} \int_a^b e^{tx} dx = \frac{1}{b-a} \cdot \frac{1}{t} e^{tx} \Big|_a^b = \frac{e^{tb} - e^{ta}}{(b-a)t} \\ &= \frac{1}{b-a} \cdot t^{-1} (e^{tb} - e^{ta}), \quad \# \end{aligned}$$

$$\begin{aligned} 2^\circ \phi'(t) &= \frac{1}{b-a} \left[ (-t^{-2})(e^{tb} - e^{ta}) + t^{-1}(be^{tb} - ae^{ta}) \right] \\ \phi''(t) &= \frac{1}{b-a} \left[ 2t^{-3}(e^{tb} - e^{ta}) - t^{-2}(be^{tb} - ae^{ta}) \right. \\ &\quad \left. - t^{-2}(be^{tb} - ae^{ta}) + t^{-1}(b^2e^{tb} - a^2e^{ta}) \right] \quad \# \end{aligned}$$

$$\begin{aligned} 3^\circ E(X) &= \phi'(0) \\ E(X^2) &= \phi''(0) \end{aligned}$$

$$4^\circ \text{var}(X) = E(X^2) - (E(X))^2 = \phi''(0) - [\phi'(0)]^2 \quad \#$$

(b)  $X \sim \text{Exp}(\lambda)$ ,  $f_X(x) = \lambda e^{-\lambda x}$ ,  $x \geq 0$

$$\begin{aligned} 1^\circ \phi(t) &= \int_{-\infty}^{\infty} e^{tx} f_X(x) dx = \int_0^{\infty} e^{tx} \cdot \lambda e^{-\lambda x} dx \\ &= \int_0^{\infty} \lambda e^{-(\lambda-t)x} dx \\ &= \frac{\lambda e^{-(\lambda-t)x}}{-(\lambda-t)} \Big|_0^{\infty} \\ &= \frac{\lambda(0-1)}{-(\lambda-t)} \\ &= \frac{\lambda}{\lambda-t}, \quad \lambda > t \quad \# \end{aligned}$$

$$\begin{aligned} 2^\circ \phi'(t) &= \frac{0 - \lambda(-1)}{(\lambda-t)^2} = \frac{\lambda}{(\lambda-t)^2} \\ \phi''(t) &= -2\lambda(\lambda-t)^{-3}(-1) = 2\lambda(\lambda-t)^{-3} \quad \# \\ 3^\circ E(X) &= \phi'(0) = \lambda^{-1}, E(X^2) = \phi''(0) = 2\lambda^{-2} \\ 4^\circ \text{var}(X) &= E(X^2) - (E(X))^2 = \frac{2}{\lambda^2} - \left(\frac{1}{\lambda}\right)^2 = \lambda^{-2} = \frac{1}{\lambda^2} \quad \# \end{aligned}$$

b. derive the tightest Chernoff's bound for the Poisson r.v.  $X$  with PMF  $p_X(n) = \frac{e^{-\lambda} \lambda^n}{n!}$ ,  $n=0, 1, 2, \dots$

Chernoff's bound  $P\{X \geq a\} \leq e^{-ta} M_X(t)$   $\forall t > 0$   
 $\triangleq g(t) \triangleq e^{-ta} M_X(t)$ ,  $t > 0$  EP tightest bound =  $\min_{t>0} g(t)$

$$1^\circ \quad M_X(t) = \sum_x e^{tx} p_X(x) = \sum_{n=0}^{\infty} e^{tn} \frac{e^{-\lambda} \lambda^n}{n!}$$

$$= e^{-\lambda} \sum_{n=0}^{\infty} \frac{e^{tn} \lambda^n}{n!} = e^{-\lambda} \sum_{n=0}^{\infty} \frac{(e^t \lambda)^n}{n!} \quad (\because e^x = \sum_{k=0}^{\infty} \frac{x^k}{k!})$$

$$= e^{-\lambda} \cdot e^{(e^t \lambda)} = e^{\lambda(e^t - 1)}$$

$$2^\circ \quad \Rightarrow g(t) = e^{-ta} M_X(t) = e^{-ta} e^{\lambda(e^t - 1)} = e^{\lambda e^t - \lambda - ta}, \quad t > 0$$

極值發生在 - 階導數為 0  $\rightarrow g'(t) = (\lambda e^t - a) e^{\lambda e^t - \lambda - ta}$ ,  $t > 0$

$$\therefore \lambda e^t - a = 0 \quad e^t = \frac{a}{\lambda} \quad t = \ln \frac{a}{\lambda} \text{ 等 } \xrightarrow{\text{恆正}} g(t) \text{ 有極值}$$

3°

$t = \ln \frac{a}{\lambda}$  代入  $g(t)$  求解

$$g(\ln \frac{a}{\lambda}) = e^{\lambda e^{\ln \frac{a}{\lambda}} - \lambda - a \ln \frac{a}{\lambda}} = e^{\lambda \frac{a}{\lambda} - \lambda - (\ln \frac{a}{\lambda})^a}$$

$$= e^{a - \lambda - (\ln \frac{a}{\lambda})^a} = e^a \cdot e^{-\lambda} \cdot e^{-(\ln \frac{a}{\lambda})^a}$$

$$= e^a \cdot e^{-\lambda \left(\frac{\lambda}{a}\right)^a}$$

$$\hookrightarrow e^{(\ln \frac{a}{\lambda})^{-a}} = \left(\frac{a}{\lambda}\right)^{-a} = \left(\frac{\lambda}{a}\right)^a$$

$$= e^{-\lambda \left(\frac{\lambda e}{a}\right)^a}$$

$$\therefore \text{tightest bound} = e^{-\lambda \left(\frac{\lambda e}{a}\right)^a}$$

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7. with  $K(t) = \ln[Ee^{tX}]$ , show that  $K'(0) = E(X)$   
 $K''(0) = \text{var}(X)$

$$M_X(t) = Ee^{tX}, \quad e^x = \sum_{k=0}^{\infty} \frac{x^k}{k!}$$

$$e^{tX} = \sum_{n=0}^{\infty} \frac{(tX)^n}{n!} = 1 + tX + \frac{t^2 X^2}{2!} + \dots + \frac{t^n X^n}{n!}$$

$$\Rightarrow M_X(t) = E\left(\sum_{n=0}^{\infty} \frac{t^n X^n}{n!}\right) = 1 + t \cdot E(X) + \frac{t^2}{2!} E(X^2) + \dots + \frac{t^n}{n!} E(X^n)$$

$$= 1 + t \cdot m_1 + \frac{t^2}{2!} m_2 + \dots + \frac{t^n}{n!} m_n = 1 + \sum_{n=1}^{\infty} \frac{m_n t^n}{n!}$$

$n^{\text{th}}$  moment  $m_n = E(X^n) = M_X^{(n)}(0) = \left. \frac{d^n}{dt^n} M_X(t) \right|_{t=0}$

1<sup>st</sup>  $m_1 = M_X'(0) = 0 + m_1 + 0 + \dots = m_1$

2<sup>nd</sup>  $m_2 = M_X''(0) = 0 + 0 + \frac{2 \cdot 1}{2!} m_2 + 0 + \dots = m_2$

$$\Rightarrow M_X(t) = 1 + \sum_{n=1}^{\infty} \frac{m_n t^n}{n!} = e^{K(t)} \quad (\because \sum_{n=1}^{\infty} K(t) = \ln M_X(t))$$

$$1 + \sum_{n=1}^{\infty} \frac{m_n t^n}{n!} = 1 + K(t) + \frac{K(t)^2}{2!} + \dots + \frac{K(t)^n}{n!}$$

$$K(t) = \ln(M_X(t)) \quad K'(t) = \frac{d}{dt} K(t) = \frac{d}{dt} \ln(M_X(t)) = \frac{1}{M_X(t)} \cdot \frac{d}{dt} M_X(t)$$

$$\Rightarrow K'(0) = \left. \frac{d}{dt} K(t) \right|_{t=0} = \frac{M_X'(0)}{M_X(0)} = \frac{m_1}{1} = m_1 = E(X)$$

$$K''(t) = \left. \frac{d}{dt} \left( \frac{\frac{d}{dt} M_X(t)}{M_X(t)} \right) \right|_{t=0} = \frac{\left( \frac{d^2}{dt^2} M_X(t) \right) \cdot M_X(t) - \frac{d}{dt} M_X(t) \cdot \frac{d}{dt} M_X(t)}{(M_X(t))^2} \Big|_{t=0}$$

$$\Rightarrow K''(0) = \frac{M_X''(0) \cdot M_X(0) - (M_X'(0))^2}{(M_X(0))^2} = \frac{m_2 \cdot 1 - (m_1)^2}{1^2} = m_2 - m_1^2$$

$$= E(X^2) - (E(X))^2 = \text{var}(X)$$