

$$\textcircled{1} \quad \text{Let } P(H) = p \\ P(T) = q$$

Define  $R = \{ \text{A run of 5 heads before a run of 5 tails} \}$

Let  $H_1 = \{ \text{first flip is a head} \}$

Then,

$$P(R) = P(R|H_1)P(H_1) + P(R|T_1)P(T_1) \\ P(R) = p P(R|H_1) + q P(R|T_1) \quad \text{--- (1)}$$

Given  $H_1$ , we can obtain a run of 5 heads if the next 4 flips (flips 2-5) are all heads.

$$A = \{ \text{flips 2-5 are } H_s \}$$

Therefore,

$$P(R|H_1) = \underbrace{P(R|H_1 \cap A)P(A|H_1)}_{\substack{=1 \\ \text{(sequence of 5H is completed)}}} + P(R|H_1 \cap A^c)P(A^c|H_1) \quad \text{--- (2)}$$

Also,  $P(R|H_1 \cap A^c) = P(R|T_1)$  because if  $H_1 \cap A^c$  occurs, a tail occurs at some point in the next 4 trials & we have

to start over again. The situation would be exactly as if we started out with a T.

Also, due to independence,

$$P(A|H_1) = P(A) = p^4$$

$$(2) \Rightarrow P(R|H_1) = p^4 + P(R|T_1)(1 - p^4) \quad \text{--- (3)}$$

Now, to obtain a similar expression for  $P(R|T_1)$ , define

$$B = \{2-5 \text{ are tails}\}$$

$$P(R|T_1) = \underbrace{P(R|T_1 \cap B)}_{=0} \underbrace{P(B|T_1)}_{q^4} + \underbrace{P(R|T_1 \cap B^c)}_{P(R|H_1)} \underbrace{P(B^c|T_1)}_{1-q^4}$$

$$P(R|T_1) = (1 - q^4) P(R|H_1) \quad \text{--- (4)}$$

Now, solve (3), (4):

$$P(R|H_1) = p^4 + (1 - q^4) P(R|H_1) (1 - p^4)$$

$$\Rightarrow P(R|H_1) = \frac{p^4}{1 - (1 - p^4)(1 - q^4)}$$

$$P(R|T_1) = \frac{p^4 (1 - q^4)}{1 - (1 - p^4)(1 - q^4)}$$

(1)

⇒

$$P(R) = \frac{p^4 + p^4 q (1 - q^4)}{1 - (1 - p^4)(1 - q^4)}$$

(2) We proved this in class.

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(3) Let  $\{X_n\}$  be i.i.d such that  $X_n \sim \text{Poisson}(1)$ .

$$\text{Then } E(X_n) = 1, \quad V(X_n) = 1,$$

$$\text{Therefore, } E\left(\sum_{i=1}^n X_i\right) = n$$

$$V\left(\sum_{i=1}^n X_i\right) = n$$

$$\& \sum_{i=1}^n X_i \sim \text{Poisson}(n) \quad (\text{because of independence})$$

By CLT,

$$\frac{\sum_{i=1}^n X_i - n}{\sqrt{n}} \longrightarrow N(0, 1)$$

or,

$$\frac{\sum_{i=1}^n X_i - n}{\sqrt{n}} \longrightarrow N(0, 1).$$

Note that

$$P\left(\sum_{i=1}^n X_i \leq n\right) = \sum_{k=0}^n \frac{e^{-n} \cdot n^k}{k!}$$

By CLT,

$$P\left[\frac{\sum_{i=1}^n X_i - n}{\sqrt{n}} \leq \frac{n-n}{\sqrt{n}}\right] = P\left[\frac{\sum_{i=1}^n X_i - n}{\sqrt{n}} \leq 0\right] \longrightarrow \Phi(0) = \frac{1}{2}$$

Therefore,  $\sum_{k=0}^n \frac{e^{-n} \cdot n^k}{k!} \longrightarrow \frac{1}{2}$

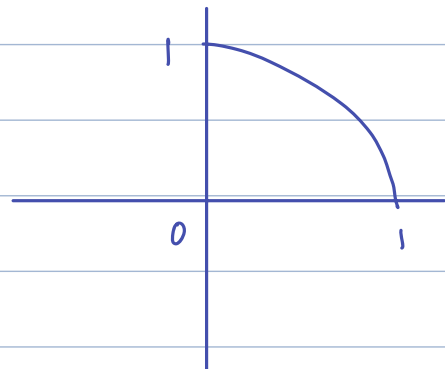
(4) See lecture 25 page 2

(5)

$$f(x, y) = \begin{cases} cx^3y & ; x, y \geq 0, x^2 + y^2 \leq 1 \\ 0 & ; \text{otherwise} \end{cases}$$

$$1 = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dy dx$$

$$= \int_{x=0}^1 \int_{y=0}^{\sqrt{1-x^2}} cx^3y dy dx$$



$$= \int_{x=0}^{x=1} c x^3 (1-x^2) dx$$

$$= c \left[ \frac{x^4}{4} - \frac{x^6}{6} \right]_0^1 = c \left[ \frac{1}{4} - \frac{1}{6} \right] = \frac{c}{12} \Rightarrow c=12$$

$$f_Y(y) = \int_{x=0}^{x=\sqrt{1-y^2}} 12x^3 y dx = 3y \cdot x^4 \Big|_0^{\sqrt{1-y^2}} \\ = 3y (1-y^2)^2 ; \quad 0 < y < 1$$

$$f_{X|Y}(x|y) = \frac{12x^3 y}{3y(1-y^2)^2} = \frac{4x^3}{(1-y^2)^2} ; \quad \begin{matrix} x, y \geq 0 \\ x^2 + y^2 \leq 1 \end{matrix}$$

$$E[X|Y=y] = \int_{x=0}^{x=\sqrt{1-y^2}} x \cdot \frac{4x^3}{(1-y^2)^2} dx \\ = \frac{4}{(1-y^2)^2} \cdot \frac{x^5}{5} \Big|_0^{\sqrt{1-y^2}}$$

$$E[X|Y=y] = \frac{4}{5} \sqrt{1-y^2} ; \quad 0 < y < 1.$$

$$(6) \quad X|Y=y \sim U(-y, y), \quad Y \sim U(0, 1)$$

$$\text{Recall: } W \sim U(a, b) \Rightarrow E(W) = \frac{a+b}{2}$$

$$V(W) = \frac{(b-a)^2}{12}$$

$$E[X|Y=y] = \frac{-y+y}{2} = 0$$

$$E[X^2|Y=y] = \frac{(y - (-y))^2}{12} = \frac{y^2}{3}$$

$$E[X^2] = E[E[X^2|Y]] = E\left[\frac{Y^2}{3}\right] = \frac{1}{3} E[Y^2]$$

$$= \frac{1}{3} \int_0^1 y^2 dy = \frac{1}{9}$$

$$V(X) = E(X^2) - (E(X))^2 = \boxed{\frac{1}{9}}$$

$$(7) \quad \text{Let } A_i = \{i\text{th coin is selected}\}$$

$$B_n = \{\text{first } n \text{ flips are heads}\}$$

$$H_{n+1} = \{(n+1)\text{ flip is a head}\}$$

$$P(H_{n+1} | F_n) = \sum_{i=0}^{99} P(H_{n+1} | B_n \cap A_i) P(A_i | B_n)$$

Due to independence,

$$P(H_{n+1} | A_i \cap B_n) = P(H_{n+1} | A_i) = \frac{i}{99}$$

By Bayes' rule

$$P(A_i | B_n) = \frac{P(B_n | A_i) P(A_i)}{\sum_{j=0}^{99} P(B_n | A_j) P(A_j)}$$

$$= \frac{\left(\frac{i}{99}\right)^n \cdot \cancel{\frac{1}{100}}}{\sum_{j=0}^{99} \left(\frac{j}{99}\right)^n \cdot \cancel{\frac{1}{100}}}$$

$$= \boxed{\frac{\left(\frac{i}{99}\right)^n}{\sum_{j=0}^{99} \left(\frac{j}{99}\right)^n}}$$

(8) Let  $X, Y \sim U(0, 1)$

Notice that

$$E\left|\left|x - \frac{1}{2}\right|\right| = \int_0^{\frac{1}{2}} -(x - \frac{1}{2}) dx + \int_{\frac{1}{2}}^1 (x - \frac{1}{2}) dx = \frac{1}{4}$$

Similarly,

$$E\left|\left|y - \frac{1}{2}\right|\right| = \frac{1}{4}$$

$$\text{Since } |x - y| \leq \left|x - \frac{1}{2}\right| + \left|y - \frac{1}{2}\right|$$

$$\Rightarrow E|x - y| \leq \frac{1}{2} \quad \square$$

(9) See lecture 24, pages 7-9

$$(10) \begin{pmatrix} x & y & z \end{pmatrix} \begin{pmatrix} 0.5 & 0 & 0.5 \\ 0.5 & 0.1 & 0.4 \\ 0 & 1 & 0 \end{pmatrix} = \begin{pmatrix} x & y & z \end{pmatrix}$$

$$0.5x + 0.5y = x \Rightarrow x = y$$

$$0.1y + z = y \Rightarrow z = 0.9y$$

$$x + y + z = 1 \Rightarrow 2.9y = 1 \Rightarrow y = \frac{1}{2.9} = x, \quad z = \frac{0.9}{2.9}$$



$$\pi = \begin{pmatrix} \frac{1}{2.9} & \frac{1}{2.9} & \frac{0.9}{2.9} \end{pmatrix}$$

(11) See lecture 21, page 4

(12) See lecture 23, page 1

(13) See lecture 24, page 2-3

(14) Lecture 21, page 7

(15)  $P(X_n = 1) = a_n$   
 $P(X_n = 0) = 1 - a_n$

Suppose  $\lim_{n \rightarrow \infty} a_n = 0$

Let  $\varepsilon > 0$ .

$$P(|X_n| \geq \varepsilon) \leq \frac{E|X_n|}{\varepsilon} = \frac{a_n}{\varepsilon} \rightarrow 0 \text{ as } n \rightarrow \infty.$$

$$\Rightarrow X_n \xrightarrow{P} 0 \quad \square$$

(16) (a)  $P =$

	1	2	3	4
1	0.5	0.5	0	0
2	0.25	0.75	0	0
3	0	0	0.25	0.75
4	0	0	0.75	0.25

(b) If the chain begins in 1 or 2, then it never enters 3, 4. Therefore, if  $X_0 = 1$  or  $X_0 = 2$ ,

$$(a \ b) \begin{pmatrix} 0.5 & 0.5 \\ 0.25 & 0.75 \end{pmatrix} = (a \ b)$$

$$0.5a + 0.25b = a \Rightarrow b = 2a$$

$$a + b = 1 \Rightarrow a = \frac{1}{3}, \ b = \frac{2}{3}$$

$$\Rightarrow \pi = \left( \frac{1}{3} \ \frac{2}{3} \ 0 \ 0 \right).$$

Similarly, if  $X_0 = 3$  or  $X_0 = 4$ ,

$$(c \ d) \begin{pmatrix} 0.25 & 0.75 \\ 0.75 & 0.25 \end{pmatrix} = (c \ d)$$

$$0.25c + 0.75d = c \Rightarrow c = d$$

$$c + d = 1 \Rightarrow c = d = \frac{1}{2}$$

Therefore,  $\pi = \left( 0 \ 0 \ \frac{1}{2} \ \frac{1}{2} \right)$ .

(8) Let  $S = \{1, 2\}$

$$P = \begin{matrix} & \begin{matrix} 1 & 2 \end{matrix} \\ \begin{matrix} 1 \\ 2 \end{matrix} & \begin{pmatrix} 0.7 & 0.3 \\ 0.6 & 0.4 \end{pmatrix} \end{matrix}$$

We are given that  $P(X_0 = 1) = P(X_0 = 2) = \frac{1}{2}$

$$P^3 \approx \begin{bmatrix} 0.67 & 0.33 \\ 0.67 & 0.33 \end{bmatrix}$$

$$\left(\frac{1}{2} \quad \frac{1}{2}\right) P^3 \approx (0.67 \quad 0.33)$$

$$\Rightarrow \boxed{P(X_3 = 1) \approx 0.67}$$

$$P(X_4 = 1 \mid X_0 = 1) = P_{11}^{(4)}$$

$$\text{Since } P^4 \approx \begin{pmatrix} 0.67 & 0.33 \\ 0.67 & 0.33 \end{pmatrix}, \quad P_{11}^{(4)} \approx 0.67$$

(19) See lecture 18, page 5

(20) Let  $\varepsilon > 0$ . Then  $\exists \delta > 0$  s.t

$$|x - k| < \delta \Rightarrow |g(x) - g(k)| < \varepsilon.$$

or  $|g(x) - g(k)| \geq \varepsilon \Rightarrow |x - k| \geq \delta$

$$\Rightarrow P(|g(X_n) - g(k)| \geq \varepsilon) \leq P(|X_n - k| \geq \delta) \rightarrow 0$$

$$\Rightarrow g(X_n) \xrightarrow{P} g(k) \quad \square$$

(21) (a)  $E[X|\lambda] = \lambda$

(b)  $P(E[X|\lambda] > 1.5)$

$$= P(\lambda > 1.5) = \int_{1.5}^2 \frac{1}{2-1} = \boxed{0.5}$$

(22)

$$\begin{aligned} P(A|B \cap C) P(C|B) &= \frac{P(A \cap B \cap C)}{P(B \cap C)} \cdot \frac{P(B \cap C)}{P(B)} \\ &= \frac{P(A \cap B \cap C)}{P(B)} \quad \text{--- (1)} \end{aligned}$$

$$\begin{aligned} P(A|B \cap C^c) P(C^c|B) &= \frac{P(A \cap B \cap C^c)}{P(B \cap C^c)} \cdot \frac{P(B \cap C^c)}{P(B)} \\ &= \frac{P(A \cap B \cap C^c)}{P(B)} \quad \text{--- (2)} \end{aligned}$$

(1) + (2)  $\Rightarrow$

$$\begin{aligned} &P(A|B \cap C) P(C|B) + P(A|B \cap C^c) P(C^c|B) \\ &= \frac{P(A \cap B \cap C) + P(A \cap B \cap C^c)}{P(B)} \\ &= \frac{P(A \cap B)}{P(B)} = P(A|B) \end{aligned}$$

(23)

$$E[\max\{Z - c, 0\}] = \int_c^{\infty} (z - c) \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz$$

$$= \underbrace{\int_c^{\infty} z \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz}_{\text{Let } u = \frac{z^2}{2}} - c \cdot \underbrace{\int_c^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz}_{1 - \Phi(c)}$$

$$= \frac{1}{\sqrt{2\pi}} \int_{\frac{c^2}{2}}^{\infty} e^{-u} du + c(1 - \Phi(c))$$

$$= \frac{1}{\sqrt{2\pi}} e^{-\frac{c^2}{2}} + c(1 - \Phi(c))$$

(24) When  $X \geq 0$ ,

$$E[X] = \int_0^{\infty} x \cdot f(x) dx$$

Note that

$$\int_0^{\infty} P(X > t) dt = \int_{t=0}^{\infty} \int_{y=t}^{\infty} f(y) dy dt$$

$$= \int_{y=0}^{\infty} \int_{t=0}^{t=y} f(y) dt dy = \int_{y=0}^{\infty} y f(y) dy = E(X) \quad \square$$

②⑤ See lecture 11, page 3

②⑥ - ②⑦ } see lecture 19