



Sharif University of Technology

CE Department

Course: Stochastic Processes

PS. 1

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**1****(a)**

Let  $A$  be the event that 5 coin tosses appear heads. by the law of total probability:

$$\mathbb{P}[A] = \int_0^1 \mathbb{P}[A|p]\mathbb{P}[p] dp = \int_0^1 p^5 \times 1 dp = \left. \frac{p^6}{6} \right|_0^1 = \frac{1}{6}$$

**(b)**

Let  $A$  be the event that 4 first coin tosses are heads and let  $B$  be the event that the fifth toss is also a head. we want to calculate:

$$\mathbb{P}[B|A] = \frac{\mathbb{P}[B \cap A]}{\mathbb{P}[A]}$$

Again by using the law of total probability:

$$\mathbb{P}[A \cap B] = \int_0^1 \mathbb{P}[A \cap B|p]\mathbb{P}[p] dp = \int_0^1 p^5 dp = \left. \frac{p^6}{6} \right|_0^1 = \frac{1}{6}$$

$$\mathbb{P}[A] = \int_0^1 \mathbb{P}[A|p]\mathbb{P}[p] dp = \int_0^1 p^4 dp = \left. \frac{p^5}{5} \right|_0^1 = \frac{1}{5}$$

thus,

$$\mathbb{P}[B|A] = \frac{5}{6}$$

**2**

First we need to find  $f_{X_1+X_2}$ , this function can be derived by applying convolution.

$$Z = X_1 + X_2 \quad , \quad f_Z(z) = \int_{-\infty}^{\infty} f_{X_1, X_2}(\tau, z - \tau) d\tau = \lambda_1 \lambda_2 e^{-\lambda_2 z} \int_0^z e^{(\lambda_2 - \lambda_1)\tau} d\tau$$

Now we should solve for two cases.

**(1)**

Suppose  $\lambda_1 = \lambda_2 = \lambda$  then:

$$\begin{aligned} f_Z(z) &= \lambda^2 e^{-\lambda z} z \sim \text{Gamma}(2, \lambda) \\ f_{X_1|Z}(x|z) &= \frac{f_{X_1 Z}(x, z)}{f_Z(z)} = \frac{f_{X_1 X_2}(x, z - x)}{f_Z(z)} = \frac{\lambda^2 e^{-\lambda x} e^{-\lambda(z-x)}}{\lambda^2 e^{-\lambda z} z} \\ f_{X_1|Z=2}(x) &= \frac{\lambda^2 e^{-\lambda x} e^{-\lambda(2-x)}}{\lambda^2 e^{-\lambda 2} 2} = \frac{1}{2} \end{aligned}$$

which is equal to:

$$\lim_{\lambda_1 - \lambda_2 \rightarrow 0} \frac{(\lambda_1 - \lambda_2) e^{-(\lambda_1 - \lambda_2)x}}{1 - e^{-2(\lambda_1 - \lambda_2)}}$$


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**(2)**

Now suppose that  $\lambda_1 \neq \lambda_2$ :

$$\begin{aligned} f_Z(z) &= \frac{\lambda_1 \lambda_2}{\lambda_2 - \lambda_1} (e^{-\lambda_1 z} - e^{-\lambda_2 z}) \\ f_{X_1|Z}(x|z) &= \frac{f_{X_1 Z}(x, z)}{f_Z(z)} = \frac{f_{X_1 X_2}(x, z - x)}{f_Z(z)} = \frac{(\lambda_1 - \lambda_2) e^{-(\lambda_1 - \lambda_2)x}}{1 - e^{-z(\lambda_1 - \lambda_2)}} \\ f_{X_1|Z=2}(x) &= \frac{(\lambda_1 - \lambda_2) e^{-(\lambda_1 - \lambda_2)x}}{1 - e^{-2(\lambda_1 - \lambda_2)}} \end{aligned}$$

now we can calculate the expected value:

$$\mathbb{E}[X_1|Z=2] = \frac{(\lambda_1 - \lambda_2)}{1 - e^{-2(\lambda_1 - \lambda_2)}} \int_0^2 x e^{-(\lambda_1 - \lambda_2)x} dx = \frac{(2\lambda_2 - 2\lambda_1 - 1) e^{2\lambda_2} + e^{2\lambda_1}}{(\lambda_2 - \lambda_1) (e^{2\lambda_2} - e^{2\lambda_1})}$$

**3**

Let  $X$  be the number of tosses until 2 heads or 2 tails appear. and also  $\mathbb{P}[Heads] = p$  then:

$$\mathbb{P}[X = k] = \begin{cases} 0 & k < 2 \\ p^{n+1}(1-p)^{n-1} + (1-p)^{n+1}p^{n-1} & k = 2n \\ p^{n+1}(1-p)^n + (1-p)^{n+1}p^n & k = 2n + 1 \end{cases}$$

then  $\mathbb{E}[X]$  can be derived by:

$$\begin{aligned} \mathbb{E}[X] &= \sum_{n=1}^{\infty} (2n)(p^{n+1}(1-p)^{n-1} + (1-p)^{n+1}p^{n-1}) + \sum_{n=1}^{\infty} (2n+1)(p^{n+1}(1-p)^n + (1-p)^{n+1}p^n) \\ &= 2 \frac{p^2 + q^2}{(1-pq)^2} + 2 \frac{p^2q + q^2p}{(1-pq)^2} + \frac{p^2q + q^2p}{1-pq} \end{aligned}$$

**(a)**

this part doesn't make any sense.

**(b)**

by Markov's inequality:

$$\mathbb{P}[X \leq 9] \geq 1 - \frac{\mathbb{E}[X]}{9} = 2 \frac{p^2 + q^2}{9(1-pq)^2} + 2 \frac{p^2q + q^2p}{9(1-pq)^2} + \frac{p^2q + q^2p}{9(1-pq)}$$

for special case  $p = \frac{1}{2}$ :

$$\mathbb{P}[X \leq 9] \geq 1 - \frac{3}{9} = \frac{2}{3}$$


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**4**

$$\lambda = 1 \quad , \quad X_i \sim \text{Pois}(1) \quad , \quad Z = \sum_{i=1}^n X_i \sim \mathcal{N}(n, n)$$

Also sum of independent poissonns is also a poisson.

$$Z = \sum_{i=1}^n X_i \sim \text{Pois}(n)$$

Next we calculate  $\mathbb{P}[Z \leq n]$  in two different ways.

$$Z \sim \mathcal{N}(n, n) \rightarrow \mathbb{P}[Z \leq n] = 1/2 \quad \mu \text{ is normal r.v's median}$$

$$Z \sim \text{Pois}(n) \rightarrow \mathbb{P}[Z \leq n] = \sum_{i=0}^n \frac{e^{-n} n^i}{i!}$$

thus,

$$\lim_{n \rightarrow \infty} \sum_{i=0}^n \frac{e^{-n} n^i}{i!} = \frac{1}{2}$$

**5**

$$\begin{aligned} \text{Var}(Z|Y) &= \mathbb{E}[Z^2|Y] - \mathbb{E}[Z|Y]^2 = \mathbb{E}[(XY)^2|Y] - \mathbb{E}[XY|Y]^2 \\ \mathbb{E}[(XY)^2|Y] &= \int_{-\infty}^{\infty} x^2 y^2 \frac{f_{XY}(x, y)}{f_Y(y)} dx = y^2 \int_{-\infty}^{\infty} x^2 \frac{f_{XY}(x, y)}{f_Y(y)} dx = Y^2 \mathbb{E}[X^2|Y] \\ \mathbb{E}[XY|Y] &= \int_{-\infty}^{\infty} xy \frac{f_{XY}(x, y)}{f_Y(y)} dx = y \int_{-\infty}^{\infty} x \frac{f_{XY}(x, y)}{f_Y(y)} dx = Y \mathbb{E}[X|Y] \\ \rightarrow \text{Var}(Z|Y) &= Y^2 \mathbb{E}[X^2|Y] - Y^2 \mathbb{E}[X|Y]^2 = Y^2 (\mathbb{E}[X^2|Y] - \mathbb{E}[X|Y]^2) = Y^2 \text{Var}(X|Y) \end{aligned}$$

**6**

(a)

$$\text{Cov}(Y, X) = \text{Cov}(X^2, X) = \mathbb{E}[X^3] - \mathbb{E}[X^2]\mathbb{E}[X]$$

note that  $\mathbb{E}[X^3] = \mathbb{E}[X] = 0$  because of negative values of  $x$ :

$$\mathbb{E}[X^3] = \int_{-1}^1 \frac{x^3}{2} = 0 \quad \mathbb{E}[X] = \int_{-1}^1 \frac{x}{2} = 0$$

$$\text{Cov}(Y, X) = 0$$

(b)

There is no conflict.  $\text{Cov}(X, Y) = 0$  implies that there is no linear correlation between  $X$  and  $Y$ . But independency is a stronger condition. So in conclusion  $\text{Cov}(X, Y) = 0$  doesn't always imply independency.

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## 7

Let  $X_i$  be the number of beads taken starting from  $(i-1)^{th}$  unique number seen until the  $i^{th}$  unique number is seen. In this way total number of beads taken ( $X$ ) is given by:

$$X = \sum_{i=1}^n X_i \quad , \quad X_i \sim Geo\left(\frac{n-i+1}{n}\right)$$

$$\mathbb{E}[X] = \sum_{i=1}^n \mathbb{E}[X_i] = \sum_{i=1}^n \frac{n}{n-i+1} = n\left(\frac{1}{n} + \cdots + \frac{1}{1}\right)$$

## 8

(a)

By Chebyshev's inequality:

$$\mathbb{P}[|X - \mu| \geq a] \leq \frac{\sigma^2}{a^2}$$

Take  $X$  to be the mean of i.i.d random variables with defined mean and variance  $\mu$  and  $\sigma^2$ .

$$X = \frac{1}{n} \times \sum_{i=1}^n X_i \sim \mathcal{N}\left(\mu, \frac{\sigma^2}{n}\right)$$

now for every  $a > 0$  we can write:

$$\lim_{n \rightarrow \infty} \mathbb{P}[|X - \mu| \geq a] \leq \lim_{n \rightarrow \infty} \frac{\sigma^2}{na^2} = 0 \rightarrow \lim_{n \rightarrow \infty} \mathbb{P}[|X - \mu| \geq a] = 0$$

(b)

Resulting from the weak law of large numbers, sample mean is the interval  $[\mu - a, \mu + a]$  with probability 1.

In other words

$$\lim_{n \rightarrow \infty} \mathbb{P}[\mu - a \leq X \leq \mu + a] = 1$$

Now by letting  $a$  take smaller and smaller values we have

$$\lim_{n \rightarrow \infty, a \rightarrow \infty} \mathbb{P}[\mu - a \leq X \leq \mu + a] = 1$$

$$\lim_{n \rightarrow \infty} X \rightarrow \mu$$


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