

# CS 5291: Stochastic Processes for Networking

## HW2

1. A fair coin is tossed until two tails occur successively. Find the expected number of the tosses required.

Hint: Let

$$X = \begin{cases} 0 & \text{if the first toss results in tails} \\ 1 & \text{if the first toss results in heads,} \end{cases}$$

and condition on  $X$ .

Let  $N$  be the number of tosses required;

$$\text{Then } E[N] = E[E[N|X]] = E[X|X=1] P\{X=1\} + E[N|X=0] P\{X=0\}$$

$$= (1 + E[N]) \times \frac{1}{2} + \left(2 \times \frac{1}{2} + (2 + E[N]) \times \frac{1}{2}\right) \times \frac{1}{2} = \frac{3}{4}E[N] + \frac{3}{2}$$

$$\rightarrow E[N] = 6$$

2. A deck of 52 playing cards, containing all 4 aces, is randomly divided into 4 piles of 13 cards each. Define events  $E_1$ ,  $E_2$ ,  $E_3$ , and  $E_4$  as follows:

$$E_1 = \{\text{the first pile has exactly 1 ace}\},$$

$$E_2 = \{\text{the second pile has exactly 1 ace}\},$$

$$E_3 = \{\text{the third pile has exactly 1 ace}\},$$

$$E_4 = \{\text{the fourth pile has exactly 1 ace}\}.$$

Find  $P\{E_1, E_2, E_3, E_4\}$ , the probability that each pile has an ace.

Hint:

$$P\{E_1 E_2 \dots E_n\} = P\{E_1\} P\{E_2|E_1\} P\{E_3|E_1 E_2\} \dots P\{E_n|E_1 \dots E_{n-1}\}$$

$$P\{E_1, E_2, E_3, E_4\} = \frac{C_1^4 C_{12}^{48}}{C_{13}^{52}} \times \frac{C_1^3 C_{12}^{36}}{C_{13}^{39}} \times \frac{C_1^2 C_{12}^{24}}{C_{13}^{26}} \times 1$$

3. Two unbiased six-sided dice are thrown.

(a) What is the probability that at least one lands on six?

Let  $E$  be the event at least one lands on six.

$$P\{E\} = \frac{\text{number of ways to get } E}{\text{number of sample points}} = \frac{11}{36}$$

(b) If the two dice land on the different values, what is the probability that at least one lands on six?

Let  $D$  be the event two dice land on the different values.

$$P\{D\} = 1 - \text{Prob}\{\text{two dice land on the same value}\}$$

$$= 1 - \frac{6}{36} = \frac{5}{6}$$

$$P\{E|D\} = \frac{P\{ED\}}{P\{D\}} = \frac{\frac{10}{36}}{\frac{5}{6}} = \frac{1}{3}$$

4. Let  $Y_1, Y_2, \dots, Y_n$  be independent random variables, each having a uniform distribution over  $(0,1)$ . Let  $X = \text{maximum}(Y_1, Y_2, \dots, Y_n)$ .

- (a) Show that the cumulative distribution function of  $n$ ,  $F_X(\cdot)$ , is given by

$$F_X(x) = x^n, \quad 0 \leq x \leq 1$$

If  $x$  is a real number between 0 and 1, then  $X < x$  if and only if  $Y_k < x$  for all  $k$ , and  $P\{Y_k < x\} = x$  since  $x$  is uniform on  $(0,1)$ . (The PDF is equal to 1, so the CDF is  $\int_0^x 1 dx = x$ )

Because  $Y_k$  are independent,  $P\{X < x\} = P\{Y_1 < x\} * P\{Y_2 < x\} * \dots * P\{Y_n < x\} = x^n$ .

- (b) What is the probability density function of  $X$ ?

The PDF of  $X$  is obtained by differentiating,

$$f_X(x) = \frac{F_X(x)}{dx} = nx^{n-1}, \quad 0 \leq x \leq 1$$

5. Derive the moment generating functions for the following random variables. Then, derive the expected value, second moment, and variance of each random variable.

- (a) Uniform distribution

$$\text{Uniform distribution: } f_X(x) = \frac{1}{b-a}$$

$$\phi(t) = E[e^{tX}]$$

$$\begin{aligned} &= \int_a^b e^{tx} \frac{1}{b-a} dx = \frac{1}{b-a} \frac{1}{t} \int_a^b t e^{tx} dx \\ &= \frac{1}{t(b-a)} [e^{tx}]_a^b = \frac{e^{tb} - e^{ta}}{t(b-a)} \end{aligned}$$

$$\phi'(t) = \frac{d\phi(t)}{dt} = \frac{1}{b-a} \left( \frac{be^{tb} - ae^{ta}}{t} - \frac{e^{tb} - e^{ta}}{t^2} \right)$$

$$\phi''(t) = \frac{1}{b-a} \left( \frac{2(e^{tb} - e^{ta})}{t^3} - \frac{2(be^{tb} - ae^{ta})}{t^2} + \frac{b^2e^{tb} - a^2e^{ta}}{t} \right)$$

$$\begin{aligned} E[X] = \phi'(0) &= \lim_{s \rightarrow 0} \frac{1}{b-a} \left( \frac{be^{tb} - ae^{ta}}{t} - \frac{e^{tb} - e^{ta}}{t^2} \right) \\ &= \frac{a+b}{2} \text{ (by L'Hospital's rule)} \end{aligned}$$

$$E[X^2] = \phi'(0) = \frac{a^2 + ab + b^2}{3}$$

$$Var(X) = E[X^2] - E[X]^2 = \frac{(b-a)^2}{12}$$

(b) Exponential distribution

Exponential distribution:  $f_X(x) = \lambda e^{-\lambda x}$

$$\begin{aligned} \phi(t) = E[e^{tX}] &= \int_0^\infty e^{tx} \lambda e^{-\lambda x} dx = \lambda \int_0^\infty e^{-(\lambda-t)x} dx \\ &= \lambda \frac{1}{\lambda-t} \int_0^\infty (\lambda-t) e^{-(\lambda-t)x} dx = \frac{\lambda}{\lambda-t} \end{aligned}$$

$$\phi'(t) = \frac{\lambda}{(\lambda-t)^2}$$

$$\phi''(t) = \frac{2\lambda}{(\lambda-t)^3}$$

$$E[X] = \phi'(0) = \frac{1}{\lambda}$$

$$E[X^2] = \phi''(0) = \frac{2}{\lambda^2}$$

$$Var(X) = E[X^2] - E[X]^2 = \frac{1}{\lambda^2}$$

6. Derive the tightest Chernoff's Bound for the Poisson random variable  $X$  with

PMF  $P_X(n) = \frac{e^{-\lambda} \lambda^n}{n!}, n = 0, 1, 2, 3, \dots$

$$f_X(x) = \frac{e^{-\lambda} \lambda^x}{x!}$$

$$M_X(t) = \sum_{x=0}^{\infty} e^{tx} \frac{e^{-\lambda} \lambda^x}{x!} = e^{-\lambda} \sum_{x=0}^{\infty} \frac{(e^t \lambda)^x}{x!} = e^{-\lambda} e^{e^t \lambda} = e^{-\lambda(1-e^t)}$$

$$P\{X \geq a\} \leq \min_t (e^{-ta} M_X(t)) = \min_t (e^{-ta - \lambda(1-e^t)})$$

$$\text{Let } g(t) = e^{-ta - \lambda(1-e^t)}$$

To find the minimum of  $g(t)$ , we need to solve  $\frac{dg(t)}{dt} = 0$ .

$$\frac{dg(t)}{dt} = 0 \Rightarrow (\lambda e^t - a) e^{-ta - \lambda(1-e^t)} = 0 \Rightarrow t = \ln\left(\frac{a}{\lambda}\right)$$

$$P\{X \geq a\} \leq g\left(\ln\left(\frac{a}{\lambda}\right)\right) = e^{-a \ln\left(\frac{a}{\lambda}\right) - \lambda + a} = e^{-\lambda} e^{\ln\left(\left(\frac{\lambda}{a}\right)^a\right)} e^a = e^{-\lambda} \left(\frac{\lambda e}{a}\right)^a$$

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

$$K'(t) = \frac{E[X e^{tX}]}{E[e^{tX}]}$$

$$K''(t) = \frac{E[e^{tX}] E[X^2 e^{tX}] - E^2[X e^{tX}]}{E^2[e^{tX}]}$$

Hence,

$$K'(0) = E[X],$$

$$K''(0) = E[X^2] - E^2[X] = \text{Var}(X)$$