

HOMEWORK SOLUTIONSCOMPUTING PROBS. & EXP. BY CONDITIONING

1. Three coins;  $P(\text{heads}) = 0.3, 0.5, 0.7$ , respectively. A coin is randomly selected and flipped twice.

$N = \#$  of heads on the two flips,  
Let  $X =$  coin selected (1, 2, or 3).

Want (a)  $P(N=n)$  for  $n=0, 1, 2$   
(b)  $E(N)$ .

In both parts, condition on  $X$ , the coin selected.

$$(a) P(N=n) = \sum_{i=1}^3 P(N=n | X=i) P(X=i)$$

But conditional on  $X=i$ ,  $N$  is a binomial RV with 2 trials and probability of success 0.3 if  $X=1$ , 0.5 if  $X=2$ , and 0.7 if  $X=3$ .

Thus

$$P(N=n) = \binom{2}{n} (0.3)^n (0.7)^{2-n} \cdot \left(\frac{1}{3}\right) \quad \text{coin 1} \\ + \binom{2}{n} (0.5)^n (0.5)^{2-n} \cdot \left(\frac{1}{3}\right) \quad \text{coin 2} \\ + \binom{2}{n} (0.7)^n (0.3)^{2-n} \cdot \left(\frac{1}{3}\right) \quad \text{coin 3}$$

for  $n=0, 1, 2$ .

Calculating these for each value of  $N$ , we obtain

$$P(N=0) = \frac{1}{3} \binom{2}{0} \left[ (0.3)^0 (0.7)^2 + (0.5)^0 (0.5)^2 + (0.7)^0 (0.3)^2 \right] \\ = 0.27667$$

$$P(N=1) = \frac{1}{3} \binom{2}{1} \left[ (0.3)^1 (0.7)^1 + (0.5)^1 (0.5)^1 + (0.7)^1 (0.3)^1 \right] \\ = 0.44667$$

$$P(N=2) = \frac{1}{3} \binom{2}{2} \left[ (0.3)^2 (0.7)^0 + (0.5)^2 (0.5)^0 + (0.7)^2 (0.3)^0 \right] \\ = 0.27667$$

(2)

(2)  $E(N) = E[E(N|X)]$  by the double expectation theorem. The outer expectation is with respect to  $X$ , the coin selected, so

$$E[E(N|X)] = \sum_{i=1}^3 E(N|X=i) P(X=i)$$

Since  $P(X=i) = \frac{1}{3}$  for  $i=1, 2, 3$ , and conditional on  $X=i$ ,  $N$  is a binomial RV with 2 trials and  $P(\text{success}) = 0.3, 0.5, 0.7$  for  $X=1, 2, 3$ , respectively, we get

$$\begin{aligned} E[E(N|X)] &= E(N|X=1) P(X=1) && \text{coin 1} \\ &+ E(N|X=2) P(X=2) && \text{coin 2} \\ &+ E(N|X=3) P(X=3) && \text{coin 3} \\ &= \left[ 2(0.3) \cdot \frac{1}{3} \right] + \left[ 2(0.5) \cdot \frac{1}{3} \right] + \left[ 2(0.7) \cdot \frac{1}{3} \right] \\ &= 1 \end{aligned}$$

(3)

2.  $U \sim U(0, 1)$ . We are told that  $n$  trials are to be performed and, conditional on the value,  $u$ , of  $U$ , the trials are independent with common  $P(\text{success}) = u$ . Thus, conditional on  $U = u$ , the number,  $X$ , of successes on the  $n$  trials is a binomial random variable,  $(X | U = u) \sim \text{bin}(n, u)$ . Thus

$$E[X | U = u] = nu$$

$$V[X | U = u] = nu(1-u).$$

We are asked for  $E[X]$  and  $V[X]$ . Condition on  $U = u$  and use the Double Expectation Theorem and the Conditional Variance Theorem:

$$E[X] = E[E(X|U)] = E[nU] = nE[U]$$

$$= n \cdot \left(\frac{1}{2}\right) \text{ since } U \sim U(0, 1) \Rightarrow E(U) = \frac{1}{2}$$

$$= n/2$$

$$V[X] = V[E(X|U)] + E[V(X|U)]$$

$$= V[nU] + E[nU(1-U)]$$

$$= n^2 V[U] + nE[U - U^2]$$

$$= n^2 V[U] + nE[U] - nE[U^2]$$

(4)

Since  $U \sim U(0,1)$ ,

$$E[U] = \frac{1}{2}$$

$$V[U] = \frac{1}{12}$$

and  $E[U^2] = V[U] + (E[U])^2 = \frac{1}{12} + \left(\frac{1}{2}\right)^2 = \frac{1}{3}$

so

$$V[X] = n^2 V[U] + n E[U] - n E[U^2]$$

$$= n^2 \left(\frac{1}{12}\right) + n \left(\frac{1}{2}\right) - n \left(\frac{1}{3}\right)$$

$$= \frac{1}{12} n^2 + \frac{1}{6} n.$$

(5)

3. Let  $N = \#$  of customers entering the store on the given day. Then  $N \sim \text{Poisson}(\lambda = 10)$ . Let  $A_i$  be the amount spent by customer  $i$ ,  $1 \leq i \leq N$ . Then  $A_i \sim U(0, 100)$ . We are asked for the mean and variance of the amount of money taken in by the store in one day.

Let  $X = \sum_{i=1}^N A_i$ . Then  $X$  is the amount taken in by the store in one day. We want  $E(X)$  and  $V(X)$ . Assuming that the  $A_i$  are independent of each other and of  $N$ , we can compute them by conditioning on  $N$ :

$$\begin{aligned}
 E[X | N=n] &= E\left[\sum_{i=1}^n A_i \mid N=n\right] \\
 &= E\left[\sum_{i=1}^n A_i \mid N=n\right] \\
 &= E\left[\sum_{i=1}^n A_i\right] \quad \text{since } A_i \text{ indep of } N \\
 &= \sum_{i=1}^n E(A_i) \quad \text{by properties of } E \\
 &= \sum_{i=1}^n \frac{100}{2} \quad \text{since } A_i \sim U(0, 100) \\
 &= 50n
 \end{aligned}$$

(6)

$$\begin{aligned}
\text{also, } V[X|N=n] &= V\left[\sum_{i=1}^N A_i \mid N=n\right] \\
&= V\left[\sum_{i=1}^n A_i \mid N=n\right] \\
&= V\left[\sum_{i=1}^n A_i\right] \text{ since } A_i \text{ indep of } N \\
&= \sum_{i=1}^n V(A_i) \text{ since the } A_i \text{ are indep} \\
&= \sum_{i=1}^n \frac{(100-0)^2}{12} \text{ since } A_i \sim U(0, 100) \\
&= \frac{10,000}{12} n.
\end{aligned}$$

thus,

$$\begin{aligned}
E[X] &= E[E[X|N]] = E[50N] = 50 E[N] \\
&= 50 \lambda = 50(10) = 500 \text{ as } N \sim \text{Poisson}(10)
\end{aligned}$$

and

$$\begin{aligned}
V(X) &= V(E[X|N]) + E[V(X|N)] \\
&= V(50N) + E\left[\frac{10,000}{12} N\right] \\
&= 50^2 V(N) + \left(\frac{10,000}{12}\right) E[N] \\
&= 2500 \lambda + \frac{10,000}{12} \lambda \\
&= (2500)(10) + \left(\frac{10,000}{12}\right)(10) \text{ as } N \sim \text{Poisson}(10) \\
&= \frac{100,000}{3}
\end{aligned}$$

Alternatively, since  $X = \sum_{i=1}^N A_i$  is a sum of a random number,  $N$ , of random variables,  $A_i$ , having common distribution  $U(0, 100)$ , we may use the formulas

$$E(X) = E(N) E(A_i), \text{ and}$$

$$V(X) = E(N) V(A_i) + V(N) [E(A_i)]^2.$$

Since  $N \sim \text{Poisson}(10)$ ,  $E(N) = V(N) = 10$  and since  $A_i \sim U(0, 100)$

$$E(A_i) = \frac{100+0}{2} = 50, \text{ and}$$

$$V(A_i) = \left[ \frac{(100-0)^2}{12} \right] = \frac{10,000}{12} = \frac{2500}{3}$$

so

$$E(X) = E(N) E(A_i)$$

$$= (10)(50)$$

$$= 500$$

and

$$V(X) = E(N) V(A_i) + V(N) [E(A_i)]^2$$

$$= (10) \left( \frac{2500}{3} \right) + (10) (50)^2$$

$$= \frac{25,000}{3} + 25,000$$

$$= \frac{100,000}{3}.$$

4.  $X \sim \text{Poisson}(\lambda)$  and  
 $\lambda \sim \text{Exponential}$  with mean 1.  
 We want to show for  $n = 0, 1, 2, \dots$ ,

$$P(X=n) = \left(\frac{1}{2}\right)^{n+1}.$$

Condition on the value of  $\lambda$ :

$$P(X=n) = \int_{-\infty}^{\infty} P(X=n|\lambda) f(\lambda) d\lambda.$$

Since  $X \sim \text{Poisson}(\lambda)$ , for  $\lambda \geq 0$ ,

$$P(X=n|\lambda) = \frac{\lambda^n e^{-\lambda}}{n!}, \text{ and since}$$

$$\lambda \sim \text{Exp}(1), \quad f(\lambda) = \begin{cases} e^{-\lambda}, & \lambda > 0 \\ 0, & \lambda \leq 0. \end{cases}$$

Thus,

$$P(X=n) = \int_0^{\infty} \frac{\lambda^n e^{-\lambda}}{n!} \cdot e^{-\lambda} d\lambda$$

$$= \frac{1}{n!} \int_0^{\infty} \lambda^n e^{-2\lambda} d\lambda.$$

variable part of a gamma density with (in WMS notation)  $\alpha = n+1, \beta = \frac{1}{2}$

$$P(X=n) = \frac{1}{n!} \Gamma(n+1) \left(\frac{1}{2}\right)^{n+1} \int_0^{\infty} \frac{\lambda^n e^{-2\lambda}}{\Gamma(n+1) \left(\frac{1}{2}\right)^{n+1}} d\lambda$$

$$= \frac{1}{n!} (n!) \left(\frac{1}{2}\right)^{n+1} \cdot 1$$

$$= \left(\frac{1}{2}\right)^{n+1}.$$

5.  $X =$  number of traffic accidents tomorrow.  
 Let  $Y = \begin{cases} 0, & \text{if it does not rain tomorrow} \\ 1, & \text{if it does rain tomorrow.} \end{cases}$   
 Then we are told:

$$P_Y(y) = \begin{cases} 0.4, & y=0 \\ 0.6, & y=1 \\ 0, & \text{elsewhere} \end{cases}$$

$$(X | Y=0) \sim \text{Poisson}(3)$$

$$(X | Y=1) \sim \text{Poisson}(9),$$

In all three parts, Condition on the value of  $Y$  (or on rain/no rain).

$$\begin{aligned} \text{(a)} \quad E(X) &= E[E[X|Y]] \\ &= E[X|Y=0]P_Y(0) + E[X|Y=1]P_Y(1) \\ &= (3)(0.4) + (9)(0.6) \\ &= 6.6 \end{aligned}$$

$$\begin{aligned} \text{(b)} \quad P(X=0) &= P(X=0|Y=0)P_Y(0) + P(X=0|Y=1)P_Y(1) \\ &= \left(\frac{3^0 e^{-3}}{0!}\right)(0.4) + \left(\frac{9^0 e^{-9}}{0!}\right)(0.6) \\ &= 0.4 e^{-3} + 0.6 e^{-9} \\ &= 0.01999 \end{aligned}$$

where the second equality follows since  
 $(X|Y=0) \sim \text{Poisson}(3) \Rightarrow P(X=x|Y=0) = \frac{3^x e^{-3}}{x!}$   
 and  
 $(X|Y=1) \sim \text{Poisson}(9) \Rightarrow P(X=x|Y=1) = \frac{9^x e^{-9}}{x!}$

5. (c) To compute  $V(X)$ , we must compute  $E(X)$  and  $E(X^2)$  by conditioning on  $Y$  and then use the computing formula for the variance.

From (a),  $E(X) = 6.6$ .

$$E(X^2) = E[E(X^2 | Y)]$$

$$= E(X^2 | Y=0)P_Y(0) + E(X^2 | Y=1)P_Y(1).$$

Since  $(X | Y=0) \sim \text{Poisson}(\lambda=3)$ ,

$$E(X^2 | Y=0) = V(X | Y=0) + [E(X | Y=0)]^2$$

$$= \lambda + \lambda^2$$

$$= 3 + 3^2$$

$$= 12.$$

Similarly,  $(X | Y=1) \sim \text{Poisson}(\lambda=9) \Rightarrow$

$$E(X^2 | Y=1) = 9 + 9^2 = 90.$$

Thus

$$E(X^2) = E(X^2 | Y=0)P_Y(0) + E(X^2 | Y=1)P_Y(1)$$

$$= (12)(0.4) + (90)(0.6)$$

$$= 58.8$$

Thus,

$$V(X) = E(X^2) - [E(X)]^2$$

$$= 58.8 - (6.6)^2$$

$$= 15.24$$

6. Let  $X = \#$  of storms this season.  
We are told that:

$$(X | \Lambda = \lambda) \sim \text{Poisson}(\lambda)$$

$$\Lambda \sim U(0, 5).$$

Thus,

$$P(X=x | \Lambda=\lambda) = \begin{cases} \frac{\lambda^x e^{-\lambda}}{x!}, & x=0, 1, 2, \dots \\ 0, & \text{elsewhere} \end{cases}$$

and

$$f_{\Lambda}(\lambda) = \begin{cases} \frac{1}{5}, & 0 < \lambda < 5 \\ 0, & \text{elsewhere.} \end{cases}$$

We are asked for  $P(X \geq 3) = 1 - P(X < 3)$   
 $= 1 - P(X \leq 2).$

Condition on the value of  $\Lambda$ :

$$P(X \geq 3) = 1 - P(X \leq 2)$$

$$= 1 - \int_{-\infty}^{\infty} P(X \leq 2 | \Lambda = \lambda) f_{\Lambda}(\lambda) d\lambda$$

$$= 1 - \int_0^5 \left[ \frac{\lambda^0 e^{-\lambda}}{0!} + \frac{\lambda^1 e^{-\lambda}}{1!} + \frac{\lambda^2 e^{-\lambda}}{2!} \right] \cdot \frac{1}{5} d\lambda$$

$$= 1 - \int_0^5 (e^{-\lambda} + \lambda e^{-\lambda} + \frac{1}{2} \lambda^2 e^{-\lambda}) \cdot \frac{1}{5} d\lambda$$

$$= 1 - \frac{1}{5} \int_0^5 e^{-\lambda} d\lambda - \frac{1}{5} \int_0^5 \lambda e^{-\lambda} d\lambda - \frac{1}{10} \int_0^5 \lambda^2 e^{-\lambda} d\lambda.$$

The latter two integrals can be calculated using integration by parts.

$$\int e^{-\lambda} d\lambda = -e^{-\lambda} + C, \text{ so}$$

$$\int_0^5 e^{-\lambda} d\lambda = \left[-e^{-\lambda}\right]_{\lambda=0}^{\lambda=5} = 1 - e^{-5}$$

$$\int \lambda e^{-\lambda} d\lambda = uv - \int v du$$

$$u = \lambda, dv = e^{-\lambda} d\lambda \\ du = d\lambda, v = -e^{-\lambda}$$

$$= -\lambda e^{-\lambda} - \int (-e^{-\lambda}) d\lambda$$

$$= -\lambda e^{-\lambda} + \int e^{-\lambda} d\lambda$$

$$= -\lambda e^{-\lambda} - e^{-\lambda} + C, \text{ so}$$

$$\int_0^5 \lambda e^{-\lambda} d\lambda = \left[-\lambda e^{-\lambda} - e^{-\lambda}\right]_{\lambda=0}^{\lambda=5}$$

$$= 1 - 5e^{-5} - e^{-5} = 1 - 6e^{-5}$$

$$\int \lambda^2 e^{-\lambda} d\lambda = uv - \int v du$$

$$u = \lambda^2, dv = e^{-\lambda} d\lambda \\ du = 2\lambda d\lambda, v = -e^{-\lambda}$$

$$= -\lambda^2 e^{-\lambda} - \int 2\lambda (-e^{-\lambda}) d\lambda$$

$$= -\lambda^2 e^{-\lambda} + 2 \int \lambda e^{-\lambda} d\lambda$$

$$= -\lambda^2 e^{-\lambda} + 2[-\lambda e^{-\lambda} - e^{-\lambda}] + C$$

$$= -\lambda^2 e^{-\lambda} - 2\lambda e^{-\lambda} - 2e^{-\lambda} + C, \text{ so}$$

$$\int_0^5 \lambda^2 e^{-\lambda} d\lambda = \left[-\lambda^2 e^{-\lambda} - 2\lambda e^{-\lambda} - 2e^{-\lambda}\right]_{\lambda=0}^{\lambda=5}$$

$$= 2 - 25e^{-5} - 10e^{-5} - 2e^{-5}$$

$$= 2 - 37e^{-5}$$

Thus,

$$\begin{aligned}P(X \geq 3) &= 1 - \frac{1}{5} \int_0^5 e^{-\lambda} d\lambda - \frac{1}{5} \int_0^5 \lambda e^{-\lambda} d\lambda - \frac{1}{10} \int_0^5 \lambda^2 e^{-\lambda} d\lambda \\&= 1 - \frac{1}{5}(1 - e^{-5}) - \frac{1}{5}(1 - 6e^{-5}) \\&\quad - \frac{1}{10}(2 - 37e^{-5}) \\&= 1 - \frac{1}{10}(6 - 51)e^{-5} \\&= 0.434\end{aligned}$$