

## Assignment 4

15-359 Probability and Computing

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Section: B

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### Problem 2: Shipping error

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Let  $G$  be the event that you are shipped a working server, and let  $B$  be the event that you are shipped a faulty server. Then, if  $T$  is a random variable denoting the time in days until the server crashes,  $(T|G) \sim \text{Geometric}(\frac{1}{100})$ , and  $(T|B) \sim \text{Geometric}(\frac{5}{100})$ . Thus, as shown in class,

$$E[T^2|G] = \frac{2 - \frac{1}{100}}{(\frac{1}{100})^2} = 19900, \quad E[T|G] = \frac{1}{1/100} = 100,$$

$$E[T^2|B] = \frac{2 - \frac{5}{100}}{(\frac{5}{100})^2} = 780, \quad E[T|B] = \frac{1}{1/20} = 20.$$

Then, since  $E[T^2] = E[T^2|G]P(G) + E[T^2|B]P(B)$  and  $E[T] = E[T|G]P(G) + E[T|B]P(B)$ ,  $\text{Var}(T) = E[T^2] - E[T]^2 = \boxed{\frac{73340}{9}}$ .

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### Problem 3: Skew and skewer

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For purposes of this question, abbreviate Skewer as  $S$ . Note that, for all random variables  $X$ ,

$$S(X) = E[(X - E[X])^3] = E[X^3] - 3E[X^2]E[X] + 2E[X]^3.$$

Let  $X$  and  $Y$  be independent random variables, noting that this implies that  $E[X]E[Y] = E[XY]$ . Thus,

$$\begin{aligned} S(X + Y) &= E[(X + Y)^3] - 3E[X^2]E[X] + 2E[X]^3 \\ &= E[X^3] + 3E[X^2Y] + 3E[XY^2] + E[Y^3] \\ &\quad - 3(E[X^2]E[X] - 2E[XY]E[X] + E[XY^2]) \\ &\quad - 3(E[X^2Y] - 2E[XY][EY] + E[Y^2]E[Y]) \\ &\quad + 2E[X + Y]^3 \\ &= E[X^3] + E[Y^3] \\ &\quad - 3(E[X^2]E[X] - 2E[XY]E[X]) \\ &\quad - 3(-2E[XY][EY] + E[Y^2]E[Y]) \\ &\quad + 2E[X + Y]^3 \\ &= E[X^3] - 3E[X^2]E[X] + 2E[X]^3 \\ &\quad + E[Y^3] - 3E[Y^2]E[Y] + 2E[Y]^3 \end{aligned}$$

$$= S(X) + S(Y)$$

Thus,  $S(X + Y) = S(X) + S(Y)$ . ■

#### Problem 4: Tails

Let  $X$  be a non-negative, discrete, integer-valued random variable.  
Let  $S = \{(i, j) \in \mathbb{N} \times \mathbb{N} | 1 \leq j \leq i\}$ . Then,

$$\begin{aligned} \sum_{j=1}^{\infty} jP(X \geq j) &= \sum_{j=1}^{\infty} j \sum_{i=j}^{\infty} P(X = i) \\ &= \sum_{j=1}^{\infty} \sum_{i=j}^{\infty} jP(X = i) \\ &= \sum_{(i,j) \in S} jP(X = i) \\ &= \sum_{i=1}^{\infty} \sum_{j=1}^i jP(X = i) \\ &= \frac{\sum_{i=1}^{\infty} (i + i^2)P(X = i)}{2} \\ &= \boxed{\frac{E[X^2] + E[X]}{2}} \end{aligned}$$

#### Problem 5: Coffee-theorem revisited

Let  $W$  be the event that the student starts off working, let  $C$  be the event that the student starts off at the coffee shop, and let  $X$  be a random variable denoting the time at which the student returns home. Note that, as computed in the solution to Problem 6 on Assignment 2,  $E[X|W] = 9$  and  $E[X|C] = 12$ . Furthermore, conditioning on what the student does at the end of any given hour,  $P(X = i|W) = \frac{2}{3}P(X = i - 1|C)$ , and  $P(X = i|C) = \frac{2}{3}P(X = i - 1|C) + \frac{1}{3}P(X = i - 1|W)$ . Thus, by definition of expected value, since  $\sum_{i=1}^{\infty} P(X = i|C) = \sum_{i=1}^{\infty} P(X = i|W) = 1$ ,

$$\begin{aligned} E[X^2|C] &= \sum_{i=2}^{\infty} i^2 P(X = i|C) \\ &= \frac{1}{3} \left( \sum_{i=2}^{\infty} i^2 2P(X = i - 1|C) + P(X = i - 1|W) \right) \\ &= \frac{1}{3} \left( \sum_{i=1}^{\infty} (i + 1)^2 2P(X = i|C) + P(X = i|W) \right) \\ &= \frac{1}{3} (2(E[X^2|C] + 2E[X|C] + 1) + (E[X^2|W] + 2E[X|W] + 1)) \end{aligned}$$

$$\begin{aligned}
&= \frac{1}{3} (2(E[X^2|C] + 48 + 1) + (E[X^2|W] + 18 + 1)) \\
&= \frac{1}{3} (2E[X^2|C] + E[X^2|W] + 69),
\end{aligned}$$

giving  $E[X^2|C] = E[X^2|W] + 69$ . A similar derivation gives  $3E[X^2|W] = 51 + 2E[X^2|C]$ . Solving this system of two linear equations in two variables gives  $E[X^2|W] = 189$ ,  $E[X^2|C] = 258$ . Thus, since the quantity  $\sigma^2$  in question is the variance of  $X$  given  $W$ ,  $\sigma^2 = E[X^2|W] - E[X|W]^2 = 189 - 81 = \boxed{108}$ .

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### Problem 6: Counterexamples

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A. Let  $X$  be a random variable such that  $\forall$  positive integers  $i$ ,  $P(X = i) = \frac{6}{(\pi i)^2}$ . Then,

$$\sum_{i=1}^{\infty} P(X = i) = \frac{6}{\pi^2} \sum_{i=1}^{\infty} \frac{1}{i^2} = 1,$$

but

$$E[X] = \sum_{i=1}^{\infty} i \cdot P(X = i) = \frac{6}{\pi^2} \sum_{i=1}^{\infty} \frac{1}{i} = \infty. \quad \blacksquare$$

B. Let  $X, Y : \{0, 1\} \rightarrow [0, 1]$  be random variables with the joint distribution below:

$P(X = i, Y = j)$	$i = 0$	$i = 1$
$j = 0$	1/6	1/2
$j = 1$	1/12	1/4

Clearly,  $X$  and  $Y$  are not independent. However,  $E[XY] = \frac{1}{4} = \left(\frac{1}{12} + \frac{1}{4}\right) \left(\frac{1}{2} + \frac{1}{4}\right) = E[X]E[Y]$ .

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### Problem 7: Some Chebyshev on the side

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A. Suppose that, for some  $a, b \in \mathbb{R}$ , a random variable  $X$  takes values  $a$  and  $b$  so that, for some  $p \in [0, 1]$ ,  $P(X = a) = p$  and  $P(X = b) = 1 - p$ . Then, for  $k = \frac{1}{\sqrt{1-p}}$ ,  $\frac{1}{1+k^2} = p$ , so that, for  $\mu = E[X]$ ,  $\sigma^2 = \text{Var}(X)$ ,  $P(X - \mu \geq k\sigma) = P(X \leq \mu) = p = \frac{1}{1+k^2}$ , so that equality holds in the one-sided Chebyshev Inequality.  $\blacksquare$

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### Problem 8: See what I mean

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Let  $X_1, X_2, \dots$  be an infinite sequence of independent, identically distributed random variables with finite mean  $\mu$  and finite variance  $\sigma^2$ .  $\forall n \in \mathbb{N}$ , let

$$S_n = \frac{X_1 + X_2 + \dots + X_n}{n},$$

and let  $\epsilon > 0$ . Since  $X_1, X_2, \dots$  are independent, and expectation is linear,  $\forall n \in \mathbb{N}$ ,

$$\begin{aligned} \text{Var}(S_n) &= E \left[ \left( \frac{X_1 + X_2 + \dots + X_n}{n} \right)^2 \right] - E \left[ \frac{X_1 + X_2 + \dots + X_n}{n} \right]^2 \\ &= E \left[ \frac{(X_1 + X_2 + \dots + X_n)^2}{n^2} \right] - E \left[ \frac{X_1 + X_2 + \dots + X_n}{n} \right]^2 \\ &= \frac{1}{n^2} \left( E \left[ (X_1 + X_2 + \dots + X_n)^2 \right] - E \left[ X_1 + X_2 + \dots + X_n \right]^2 \right) \\ &= \text{Var} \left( \frac{X_1 + X_2 + \dots + X_n}{n} \right) \\ &= \frac{1}{n^2} (\text{Var}(X_1) + \text{Var}(X_2) + \dots + \text{Var}(X_n)) \\ &= \frac{1}{n^2} n \sigma^2 = \frac{\sigma^2}{n}. \end{aligned}$$

Also, by Linearity of Expectation,  $E[S_n] = \mu$ . Thus, by Chebyshev's Inequality,  $\forall \epsilon_2 > 0$ , for

$$n = \left( \frac{\epsilon \sigma}{\epsilon_2} \right)^2 + 1, \quad k = \frac{\epsilon \sqrt{n}}{\sigma},$$

$$\begin{aligned} P(|S_n - \mu| > \epsilon) &\leq P(|S_n - \mu| \geq \epsilon) \\ &= P(|S_n - \mu| \geq k \frac{\sigma}{\sqrt{n}}) \\ &\leq \frac{1}{k^2} = \frac{\sigma^2}{\epsilon^2 n} < \epsilon_2. \end{aligned}$$

Thus,  $\lim_{n \rightarrow \infty} P(|S_n - \mu| > \epsilon) = 0$ . ■