

# CS561 HW4

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# 1. Problem 1: Unfair dice, convolution.

Prove or disprove if the following statements are true in general:

- (a) Consider two independent dice  $X \sim p(x)$  and  $Y \sim q(y)$  where

$$p(x) \begin{cases} p_i & x = i, i = 1, 2, 3, 4, 5, 6 \\ 0 & \text{otherwise} \end{cases}$$

and

$$q(y) \begin{cases} q_i & y = i, i = 1, 2, 3, 4, 5, 6 \\ 0 & \text{otherwise} \end{cases}$$

then we can compute the PMF of  $Z = X + Y$  as such

- $\mathbb{P}(Z = 2) = \mathbb{P}(X = 1, Y = 1) = p_1 q_1$
- $\mathbb{P}(Z = 3) = \mathbb{P}(X = 1, Y = 2) + \mathbb{P}(X = 2, Y = 1) = p_1 q_2 + p_2 q_1$
- $\mathbb{P}(Z = 4) = \mathbb{P}(X = 1, Y = 3) + \mathbb{P}(X = 2, Y = 2) + \mathbb{P}(X = 3, Y = 1) = p_1 q_3 + p_2 q_2 + p_3 q_1$
- $\mathbb{P}(Z = 5) = \mathbb{P}(X = 1, Y = 4) + \mathbb{P}(X = 2, Y = 3) + \mathbb{P}(X = 3, Y = 2) + \mathbb{P}(X = 4, Y = 1) = p_1 q_4 + p_2 q_3 + p_3 q_2 + p_4 q_1$
- $\mathbb{P}(Z = 6) = \mathbb{P}(X = 1, Y = 5) + \mathbb{P}(X = 2, Y = 4) + \mathbb{P}(X = 3, Y = 3) + \mathbb{P}(X = 4, Y = 2) + \mathbb{P}(X = 5, Y = 1) = p_1 q_5 + p_2 q_4 + p_3 q_3 + p_4 q_2 + p_5 q_1$
- $\mathbb{P}(Z = 7) = \mathbb{P}(X = 1, Y = 6) + \mathbb{P}(X = 2, Y = 5) + \mathbb{P}(X = 3, Y = 4) + \mathbb{P}(X = 4, Y = 3) + \mathbb{P}(X = 5, Y = 2) + \mathbb{P}(X = 6, Y = 1) = p_1 q_6 + p_2 q_5 + p_3 q_4 + p_4 q_3 + p_5 q_2 + p_6 q_1$
- $\mathbb{P}(Z = 8) = \mathbb{P}(X = 1, Y = 7) + \mathbb{P}(X = 2, Y = 6) + \mathbb{P}(X = 3, Y = 5) + \mathbb{P}(X = 4, Y = 4) + \mathbb{P}(X = 5, Y = 3) + \mathbb{P}(X = 6, Y = 2) + \mathbb{P}(X = 7, Y = 1) = p_1 q_7 + p_2 q_6 + p_3 q_5 + p_4 q_4 + p_5 q_3 + p_6 q_2 + p_7 q_1$
- $\mathbb{P}(Z = 9) = \mathbb{P}(X = 1, Y = 8) + \mathbb{P}(X = 2, Y = 7) + \mathbb{P}(X = 3, Y = 6) + \mathbb{P}(X = 4, Y = 5) + \mathbb{P}(X = 5, Y = 4) + \mathbb{P}(X = 6, Y = 3) + \mathbb{P}(X = 7, Y = 2) + \mathbb{P}(X = 8, Y = 1) = p_1 q_8 + p_2 q_7 + p_3 q_6 + p_4 q_5 + p_5 q_4 + p_6 q_3 + p_7 q_2 + p_8 q_1$
- $\mathbb{P}(Z = 10) = \mathbb{P}(X = 1, Y = 9) + \mathbb{P}(X = 2, Y = 8) + \mathbb{P}(X = 3, Y = 7) + \mathbb{P}(X = 4, Y = 6) + \mathbb{P}(X = 5, Y = 5) + \mathbb{P}(X = 6, Y = 4) + \mathbb{P}(X = 7, Y = 3) + \mathbb{P}(X = 8, Y = 2) + \mathbb{P}(X = 9, Y = 1) = p_1 q_9 + p_2 q_8 + p_3 q_7 + p_4 q_6 + p_5 q_5 + p_6 q_4 + p_7 q_3 + p_8 q_2 + p_9 q_1$
- $\mathbb{P}(Z = 11) = \mathbb{P}(X = 1, Y = 10) + \mathbb{P}(X = 2, Y = 9) + \mathbb{P}(X = 3, Y = 8) + \mathbb{P}(X = 4, Y = 7) + \mathbb{P}(X = 5, Y = 6) + \mathbb{P}(X = 6, Y = 5) + \mathbb{P}(X = 7, Y = 4) + \mathbb{P}(X = 8, Y = 3) + \mathbb{P}(X = 9, Y = 2) + \mathbb{P}(X = 10, Y = 1) = p_1 q_{10} + p_2 q_9 + p_3 q_8 + p_4 q_7 + p_5 q_6 + p_6 q_5 + p_7 q_4 + p_8 q_3 + p_9 q_2 + p_{10} q_1$
- $\mathbb{P}(Z = 12) = \mathbb{P}(X = 1, Y = 11) + \mathbb{P}(X = 2, Y = 10) + \mathbb{P}(X = 3, Y = 9) + \mathbb{P}(X = 4, Y = 8) + \mathbb{P}(X = 5, Y = 7) + \mathbb{P}(X = 6, Y = 6) + \mathbb{P}(X = 7, Y = 5) + \mathbb{P}(X = 8, Y = 4) + \mathbb{P}(X = 9, Y = 3) + \mathbb{P}(X = 10, Y = 2) + \mathbb{P}(X = 11, Y = 1) = p_1 q_{11} + p_2 q_{10} + p_3 q_9 + p_4 q_8 + p_5 q_7 + p_6 q_6 + p_7 q_5 + p_8 q_4 + p_9 q_3 + p_{10} q_2 + p_{11} q_1$

Or in simpler terms, for  $z = 2, 3, \dots, 12$

$$p_Z(z) = \sum_{i=\max(1, z-6)}^{\min(6, z-1)} p_i q_{z-i}$$

- (b) Let  $X$  and  $Y$  be random variables and  $Z = X + Y$ . For any value  $n$  that  $Z$  may take, we can compute (by the partition rule)

$$\begin{aligned} \mathbb{P}(Z = n) &= \sum_i \mathbb{P}(X = i, Z = n) \\ &= \sum_i \mathbb{P}(X = i, X + Y = n) \\ &= \sum_i \mathbb{P}(X = i, Y = n - i) \\ &= \sum_i \mathbb{P}(X = i) \mathbb{P}(Y = n - i) \\ p_Z(n) &= \sum_i p_X(i) p_Y(n - i) \end{aligned}$$

## 2. Problem 2: Minimal Expectation

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Let  $X_i \sim U[0, 1]$  be i.i.d. random variables and define  $Y = \min_{i=1,2,\dots,n} X_i$

- (a) Find an expression for the pdf of  $Y$ . First, we compute the cdf, which is the chance in which  $Y \leq k$  for  $k \in [0, 1]$ .

$$\begin{aligned}
 P(Y \leq k) &= 1 - P(Y > k) \\
 &= 1 - P\left(\min_{i=1,2,\dots,n} X_i > k\right) \\
 &= 1 - P(X_i > k : i = 1, 2, \dots, n) \\
 &= 1 - \prod_{i=1}^k P(X_i > k) \\
 &= 1 - (1 - k)^n
 \end{aligned}$$

and therefore, the cdf  $F$  follows that

$$F(k) = \begin{cases} 0 & k < 0 \\ 1 - (1 - k)^n & k \in [0, 1] \\ 1 & k > 1 \end{cases}$$

we can find the pdf by taking the derivative, resulting in

$$f(k) = \frac{d}{dk} F(k) = \begin{cases} n(1 - k)^{n-1} & k \in [0, 1] \\ 0 & \text{otherwise} \end{cases}$$

- (b) We can compute the expectation  $\mathbb{E}[Y]$  as follows

$$\begin{aligned}
 \mathbb{E}[y] &= \int_{-\infty}^{\infty} y f(y) dy \\
 &= \int_0^1 y n (1 - y)^{n-1} dy \\
 &= \left. \frac{(1 - x)^n (nx + 1)}{n + 1} \right|_0^1 \\
 &= \frac{1}{n + 1}
 \end{aligned}$$

- (c) The plot looks convincing and close to our analytical solution, since we can compute  $\log(\mathbb{E}[y]) = \log(1/(n + 1)) = -\log(n + 1)$  which matches the plot from the first hw.

### 3. Problem 3: Expectation Basics

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Let  $\mathbb{P}(X = 1) = 1/2$ ,  $\mathbb{P}(X = 2) = 1/4$ , and  $\mathbb{P}(X = 3) = 1/4$ .

(a) Compute  $\mathbb{E}[X]$  :

$$\begin{aligned}\mathbb{E}[X] &= \sum_i i\mathbb{P}(X = i) \\ &= 1\mathbb{P}(X = 1) + 2\mathbb{P}(X = 2) + 3\mathbb{P}(X = 3) \\ &= 1(1/2) + 2(1/4) + 3(1/4) \\ &= 7/4\end{aligned}$$

(b) Compute  $\mathbb{E}[g(X)]$  when  $g(x) = x^2$ :

$$\begin{aligned}\mathbb{E}[g(X)] &= \sum_i g(i)\mathbb{P}(X = i) \\ &= 1^2\mathbb{P}(X = 1) + 2^2\mathbb{P}(X = 2) + 3^2\mathbb{P}(X = 3) \\ &= 1(1/2) + 4(1/4) + 9(1/4) \\ &= 15/4\end{aligned}$$

(c) Compute  $\text{Var}(X) = \mathbb{E}[(X - \mathbb{E}[X])^2]$  :

$$\begin{aligned}\text{Var}(X) &= \mathbb{E}[(X - \mathbb{E}[X])^2] \\ &= \mathbb{E}[X^2 - 2X\mathbb{E}[X] + \mathbb{E}[X]^2] \\ &= \mathbb{E}[X^2] - \mathbb{E}[2X\mathbb{E}[X]] + \mathbb{E}[\mathbb{E}[X]^2] \\ &= \mathbb{E}[X^2] - \mathbb{E}[X]\mathbb{E}[2X] + \mathbb{E}[X]^2 \\ &= \mathbb{E}[X^2] - \mathbb{E}[X]^2 \\ &= 15/4 - (7/4)^2 \\ &= 11/16\end{aligned}$$

(d) Express  $\mathbb{E}[-\log(g(X))]$ :

$$\mathbb{E}[-\log(g(X))] = \frac{-\log(g(1))}{2} + \frac{-\log(g(2))}{4} + \frac{-\log(g(3))}{4}$$

(e) Compute the entropy  $H := \mathbb{E}[-\log_s(p(X))]$ :

$$H = \mathbb{E}[-\log_2(p(X))] = \frac{-\log_2(1/2)}{2} + \frac{-\log_2(1/4)}{4} + \frac{-\log_2(1/4)}{4} = 3/2$$

## 4. Problem 4: Bivariate Random Variables

The joint pmf  $p(x, y)$  is defined as follows:

- $\mathbb{P}(X = 1, Y = 1) = 1/8$
- $\mathbb{P}(X = 1, Y = 2) = 1/8$
- $\mathbb{P}(X = 2, Y = 1) = 1/2$
- $\mathbb{P}(X = 2, Y = 2) = 1/4$

(a) Are  $X$  and  $Y$  dependent? We can compute the marginal probabilities:

- $\mathbb{P}(X = 1) = \mathbb{P}(X = 1, Y = 1) + \mathbb{P}(X = 1, Y = 2) = 1/4$
- $\mathbb{P}(X = 2) = \mathbb{P}(X = 2, Y = 1) + \mathbb{P}(X = 2, Y = 2) = 3/4$
- $\mathbb{P}(Y = 1) = \mathbb{P}(X = 1, Y = 1) + \mathbb{P}(X = 2, Y = 1) = 5/8$
- $\mathbb{P}(Y = 2) = \mathbb{P}(X = 1, Y = 2) + \mathbb{P}(X = 2, Y = 2) = 3/8$

In which we can see that  $\mathbb{P}(X = 1)\mathbb{P}(Y = 1) = 5/32 \neq 1/8 = \mathbb{P}(X = 1, Y = 1)$ . Therefore,  $X$  and  $Y$  are not independent.

(b) We can compute the expectations

- $\mathbb{E}[X] = 1\mathbb{P}(X = 1) + 2\mathbb{P}(X = 2) = 1/4 + 6/4 = 7/4$
- $\mathbb{E}[Y] = 1\mathbb{P}(Y = 1) + 2\mathbb{P}(Y = 2) = 5/8 + 6/8 = 11/8$
- $\mathbb{E}[XY] = 1\mathbb{P}(XY = 2) + 2\mathbb{P}(XY = 3) + 4\mathbb{P}(XY = 4) = 1(1/8) + 2(5/8) + 4(1/4) = 19/8$

We can compute the covariance as follows:

$$\begin{aligned}\text{Cov}(X, Y) &= \mathbb{E}[(X - \mathbb{E}[X])(Y - \mathbb{E}[Y])] \\ &= \mathbb{E}[(XY - X\mathbb{E}[Y] - Y\mathbb{E}[X] + \mathbb{E}[X]\mathbb{E}[Y])] \\ &= \mathbb{E}[XY] - \mathbb{E}[X\mathbb{E}[Y]] - \mathbb{E}[Y\mathbb{E}[X]] + \mathbb{E}[\mathbb{E}[X]\mathbb{E}[Y]] \\ &= \mathbb{E}[XY] - \mathbb{E}[X]\mathbb{E}[Y] - \mathbb{E}[X]\mathbb{E}[Y] + \mathbb{E}[X]\mathbb{E}[Y] \\ &= \mathbb{E}[XY] - \mathbb{E}[X]\mathbb{E}[Y] \\ &= 19/8 - 77/32 \\ &= -1/32\end{aligned}$$

(c) Define a new random variable  $Z = X + Y$ , then

- $\mathbb{P}(Z = 2) = \mathbb{P}(X = 1, Y = 1) = 1/8$
- $\mathbb{P}(Z = 3) = \mathbb{P}(X = 1, Y = 2) + \mathbb{P}(X = 2, Y = 1) = 1/8 + 1/2 = 5/8$
- $\mathbb{P}(Z = 4) = \mathbb{P}(X = 2, Y = 2) = 1/4$
- $Z$  only takes values 2, 3, 4 since  $X$  and  $Y$  only takes value 1 and 2.

(d) A function that generates  $X, Y$  given the above pmf

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```
import random
def sample():
    return random.choices(
        population=[[1,1],[1,2],[2,1],[2,2]],
        weights=[1/8,1/8,1/2,1/4],
    )[0]
# Randomly Sample X and Y
X, Y = sample()
```

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## 5. Problem 5: Minimal Expectation

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Let  $X$  and  $Y$  be discrete random variables. Conditional expectation is defined as

$$\mathbb{E}[X|Y] = \sum_x x\mathbb{P}(X = x|Y)$$

for discrete random variables. Then, we can show that

$$\begin{aligned}\mathbb{E}[\mathbb{E}[X|Y]] &= \mathbb{E}\left[\sum_x x\mathbb{P}(X = x|Y)\right] \\&= \sum_y \left(\sum_x x\mathbb{P}(X = x|Y = y)\right) \mathbb{P}(Y = y) \\&= \sum_y \sum_x x\mathbb{P}(Y = y)\mathbb{P}(X = x|Y = y) \\&= \sum_x x \sum_y \mathbb{P}(Y = y)\mathbb{P}(X = x|Y = y) \\&= \sum_x x \sum_y \mathbb{P}(Y = y)\mathbb{P}(X = x, Y = y)/\mathbb{P}(Y = y) \\&= \sum_x x \sum_y \mathbb{P}(X = x, Y = y) \\&= \sum_x x\mathbb{P}(X = x) \\&= \mathbb{E}[X]\end{aligned}$$

as desired.