Solution to Problems for Week 4

Yunwen Lei

School of Computer Science University of Birmingham

Alex and Bob each flips a fair coin twice. Denote "1" as head, and "0" as tail. Let X be the maximum of the two numbers Alex gets, and let Y be the minimum of the two numbers Bob gets.

- **1** Find and sketch the joint PMF $P_{X,Y}(x,y)$
- ② Find the marginal PMF $P_X(x)$ and $P_Y(y)$
- **3** Find the conditional PMF $P_{X|Y}(x|y)$. Does $P_{X|Y}(x|y) = P_X(x)$? Why

• We know $X(\Omega)=\{0,1\}, Y(\Omega)=\{0,1\}$. Note X=0 if Alex gets two 0, which happens with probability 1/4. Y=0 if Bob gets either (0,0),(0,1) or (1,0), which happens with probability 3/4

$$P_{X,Y}(0,0) = \mathbb{P}(\{X=0,Y=0\}) = \mathbb{P}(\{X=0\})\mathbb{P}(\{Y=0\}) = \frac{1}{4}\frac{3}{4} = \frac{3}{16}$$

$$P_{X,Y}(1,0) = \mathbb{P}(\{X=1,Y=0\}) = \mathbb{P}(\{X=1\})\mathbb{P}(\{Y=0\}) = \frac{3}{4}\frac{3}{4} = \frac{9}{16}$$

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The marginal PMF is

$$P_X(0) = P_{X,Y}(0,0) + P_{X,Y}(0,1) = \frac{3}{16} + \frac{1}{16} = \frac{4}{16}$$

$$P_X(1) = P_{X,Y}(1,0) + P_{X,Y}(1,1) = \frac{9}{16} + \frac{3}{16} = \frac{12}{16}$$

$$P_Y(0) = P_{X,Y}(0,0) + P_{X,Y}(1,0) = \frac{3}{16} + \frac{9}{16} = \frac{12}{16}$$

① By $P_{X|Y}(x|y) = \frac{P_{X,Y}(x,y)}{P_Y(y)}$, we know

$$P_{X|Y}(0|0) = \frac{P_{X,Y}(0,0)}{P_Y(0)} = \frac{3/16}{12/16} = \frac{1}{4}$$

$$P_{X|Y}(1|0) = \frac{P_{X,Y}(1,0)}{P_Y(0)} = \frac{9/16}{12/16} = \frac{3}{4}$$

$$P_{X|Y}(0|1) = \frac{P_{X,Y}(0,1)}{P_Y(1)} = \frac{1/16}{4/16} = \frac{1}{4}$$

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Then it is clear that

$$P(X|Y)(x|y) = P_X(x) \quad \forall x, y \in \{0,1\}.$$

Therefore, X and Y are independent.

Find the marginal CDFs $F_X(x)$ and $F_Y(y)$ and determine whether or not X and Y are independent, if

$$F_{X,Y}(x,y) = \begin{cases} x - 1 - \frac{e^{-y} - e^{-xy}}{y}, & \text{if } 1 \le x \le 2, y \ge 0 \\ 1 - \frac{e^{-y} - e^{-2y}}{y}, & \text{if } x > 2, y \ge 0, \\ 0, & \text{otherwise.} \end{cases}$$

Note $e^{-\infty} = 0$. Therefore

$$F_X(x) = F_{X,Y}(x,\infty) = \begin{cases} x - 1, & \text{if } 1 \le x \le 2\\ 1, & \text{if } x > 2\\ 0, & \text{otherwise.} \end{cases}$$

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No they are not independent because

$$F_{X,Y}(x,y) \neq F_X(x)F_Y(y) \text{ if } 1 \leq x \leq 2, y \geq 0.$$

Let X and Y have a joint PDF

$$f_{X,Y}(x,y) = \begin{cases} c(x+y), & \text{if } x \in [0,1], y \in [0,1] \\ 0, & \text{otherwise.} \end{cases}$$

- Find $c, f_Y(y)$ and $\mathbb{E}[Y]$
- **3** Find $\mathbb{P}(Y > X | X > 1/2)$.

① We know $\int_{[0,1]\times[0,1]} f_{X,Y}(x,y) dxdy = 1$. Then $(\int_0^1 x dx = \int_0^1 dx^2/2 = \frac{1}{2}(x^2|_0^1) = 1/2)$ $\int_0^1 \int_0^1 c(x+y) dxdy = c \int_0^1 x dx + c \int_0^1 y dy = c(1/2+1/2) = 1 \Longrightarrow c = 1$

$$f_X(x) = \int_{\mathbb{R}} f_{X,Y}(x,y) dy = \int_0^1 (x+y) dy = x + 1/2$$
$$f_Y(y) = \int_{\mathbb{R}} f_{X,Y}(x,y) dx = \int_0^1 (x+y) dx = y + 1/2$$
$$\mathbb{E}[Y] = \int_{\mathbb{R}} y f_Y(y) dy = \int_0^1 y (y+1/2) dy = 1/3 + 1/4 = 7/12$$

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2 By the definition of conditional probability we know

$$f_{Y|X}(y|x) = \frac{f_{X,Y}(x,y)}{f_X(x)} = \frac{x+y}{x+1/2}$$

1

$$\mathbb{P}(\{Y > X | X > 1/2\}) = \frac{\mathbb{P}(\{Y > X\} \cap \{X > 1/2\})}{\mathbb{P}(\{X > 1/2\})}$$
$$= \frac{\mathbb{P}(\{1/2 < X < Y\})}{1 - F_X(1/2)}$$
$$= \frac{\int_{\frac{1}{2}}^{1} \int_{\frac{1}{2}}^{Y} (x + y) dx dy}{1 - \int_{0}^{\frac{1}{2}} (x + 1/2) dx} = \frac{3}{10}$$

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Here are more detailed calculations (optional)

$$\int_{1/2}^{y} (x+y) dx = \int_{1/2}^{y} d(x^2/2 + xy) = (x^2/2 + xy)_{1/2}^{y} = \frac{3y^2}{2} - \frac{1}{8} - \frac{y}{2}.$$

$$\int_{1/2}^{1} \left(\frac{3y^2}{2} - \frac{1}{8} - \frac{y}{2} \right) dy = \int_{1/2}^{1} \frac{1}{2} dy^3 - \frac{1}{8} \int_{1/2}^{1} dy - \frac{1}{4} \int_{1/2}^{1} dy^2 = \frac{7}{16} - \frac{1}{16} - \frac{3}{16} = \frac{3}{16}$$
$$\int_{0}^{1/2} (x + 1/2) dx = \int_{0}^{1/2} d(x^2/2 + x/2) = (x^2/2 + x/2)_{0}^{1/2} = \frac{1}{8} + \frac{1}{4} = \frac{3}{8}.$$

Suppose that X and Y are independent and both have the same density

$$f(x) = \begin{cases} 2x, & \text{if } 0 \le x \le 1\\ 0, & \text{otherwise.} \end{cases}$$

Let us find $\mathbb{P}(X + Y \leq 1)$.

Using independence, the joint density is

$$f(x,y) = f_X(x)f_Y(y) = \begin{cases} 4xy, & \text{if } x \in [0,1], y \in [0,1] \\ 0, & \text{otherwise.} \end{cases}$$

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Now (note $\int_0^{1-x} y dy = \frac{1}{2} \int_0^{1-x} dy^2 = \frac{1}{2} y^2 \Big|_0^{1-x} = \frac{(1-x)^2}{2}$) $\mathbb{P}(X+Y \le 1) = \iint_{x+y \le 1} f(x,y) dy dx$ $= 4 \int_0^1 x \int_0^{1-x} y dy dx = 4 \int_0^1 \frac{x(1-x)^2}{2} dx$ $= 4 \int_0^1 \frac{x+x^3-2x^2}{2} dx = 2\left(\frac{x^2}{2}\Big|_0^1 + \frac{x^4}{4}\Big|_0^1 - \frac{2x^3}{3}\Big|_0^1\right)$ $= \frac{1}{-}$

Show that

$$\mathsf{MSE}(\hat{\Theta}) = \mathbb{E}[(\hat{\Theta} - \theta)^2] = \mathsf{Var}(\hat{\Theta}) + B(\hat{\Theta})^2 \tag{1}$$

$$\begin{split} \mathbb{E}[(\hat{\Theta} - \theta)^2] &= \mathbb{E}\big[\big(\hat{\Theta} - \mathbb{E}[\hat{\Theta}] + \mathbb{E}[\hat{\Theta}] - \theta\big)^2\big] \\ &= \underbrace{\mathbb{E}\big[\big(\hat{\Theta} - \mathbb{E}[\hat{\Theta}]\big)^2\big]}_{:= \mathsf{Var}(\hat{\Theta})} + \mathbb{E}\big[\big(\underbrace{\mathbb{E}[\hat{\Theta}] - \theta}_{:= B(\hat{\Theta})}\big)^2\big] + 2\mathbb{E}\big[\big(\hat{\Theta} - \mathbb{E}[\hat{\Theta}]\big)\big(\mathbb{E}[\hat{\Theta}] - \theta\big)\big] \end{split}$$

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It is clear

$$\mathbb{E}\big[\big(\hat{\Theta} - \mathbb{E}[\hat{\Theta}]\big)\big(\mathbb{E}[\hat{\Theta}] - \theta\big)\big] = \big(\mathbb{E}[\hat{\Theta}] - \theta\big)\underbrace{\mathbb{E}\big[\big(\hat{\Theta} - \mathbb{E}[\hat{\Theta}]\big)\big]}_{=0}.$$

We can combine the above two inequalities to derive the stated bound.