

Show all work clearly and in order. Circle or box your final answer but points will be awarded based on a correct solution. A solution should always justify the steps taken and explain the assumptions needed to reach a final answer (e.g. how do you know you are not dividing by zero in the last step?).

## Q1

Let  $X_1 \sim \text{Bernoulli}(1/2)$  independently of  $X_2 \sim \text{Bernoulli}(2/3)$  and define  $Y = 2X_1$  and  $Z = X_2 - X_1$ .

(a) Express the joint PMF of  $Y$  and  $Z$  in tabular form.

**First note that  $\text{supp}Y = \{0, 2\}$  and  $\text{supp}Z = \{-1, 0, 1\}$ . Then  $p(y, z)$  is tabular form is**

$Y \setminus Z$	-1	0	1
0	0	1/6	1/3
2	1/6	1/3	0

**If you let  $X_2 \sim \text{Bernoulli}(1/2)$ . Then  $p(y, z)$  is tabular form is**

$Y \setminus Z$	-1	0	1
0	0	1/4	1/4
2	1/4	1/4	0

(b) Find the marginal PMF of  $Y$ .

$$\begin{aligned} p_Y(0) &= 1/2 \\ p_Y(2) &= 1/2 \end{aligned}$$

(c) Find the conditional PMF of  $Z$ ?

$$\begin{aligned} p_{Z|Y}(z|y) &= p(y, z)/p(y) \\ p_{Z|Y}(z|0) &= \begin{cases} 0 & \text{if } z = -1 \\ 1/3 & \text{if } z = 0 \\ 2/3 & \text{if } z = 1 \end{cases} \\ p_{Z|Y}(z|2) &= \begin{cases} 1/3 & \text{if } z = -1 \\ 2/3 & \text{if } z = 0 \\ 0 & \text{if } z = 1 \end{cases} \end{aligned}$$

If instead, you used  $X_2 \sim \text{Bernoulli}(1/2)$ ,

$$p_{Z|Y}(z|0) = \begin{cases} 0 & \text{if } z = -1 \\ 1/2 & \text{if } z = 0 \\ 1/2 & \text{if } z = 1 \end{cases}$$

$$p_{Z|Y}(z|2) = \begin{cases} 1/2 & \text{if } z = -1 \\ 1/2 & \text{if } z = 0 \\ 0 & \text{if } z = 1 \end{cases}$$

(d) Calculate  $\text{Cov}(Y, Z)$ . Are  $Y$  and  $Z$  independent? What is the intuition?

$$\text{Cov}(Y, Z) = \mathbb{E}[(Y - \mathbb{E}[Y])(Z - \mathbb{E}[Z])] = \mathbb{E}[YZ] - \mathbb{E}[Z]\mathbb{E}[Y]$$

$$\mathbb{E}[Z] = 1/6 \cdot -1 + 1/2 \cdot 0 + 1 \cdot 1/3 = 1/6$$

$$\mathbb{E}[Y] = 1/2 \cdot 0 + 1/2 \cdot 1 = 1/2$$

$$\text{supp}(YZ) = \{-2, 0, 2\}$$

$$\mathbb{E}[YZ] = -2 \cdot 1/6 + 0 \cdot \Pr(YZ = 0) + 2 \cdot 0 = -1/3$$

$$\Rightarrow \text{Cov}(Y, Z) = \mathbb{E}[YZ] - \mathbb{E}[Z]\mathbb{E}[Y] = -1/3 - 1/6 = -1/2$$

$Y$  and  $Z$  are negatively correlated so they cannot be independent. Intuitively, both random variables depend on  $X_1$ , so knowing  $Y$  or  $Z$  tells us something about  $X_1$  which in turn tells us something about  $Z$  or  $Y$ , respectively.

If you used  $X_2 \sim \text{Bernoulli}(1/2)$ ,

$$\text{Cov}(Y, Z) = \mathbb{E}[(Y - \mathbb{E}[Y])(Z - \mathbb{E}[Z])] = \mathbb{E}[YZ] - \mathbb{E}[Z]\mathbb{E}[Y]$$

$$\mathbb{E}[Z] = 1/4 \cdot -1 + 1/2 \cdot 0 + 1 \cdot 1/4 = 0$$

$$\Rightarrow \text{Cov}(Y, Z) = \mathbb{E}[YZ]$$

$$\text{supp}(YZ) = \{-2, 0, 2\}$$

$$\mathbb{E}[YZ] = -2 \cdot 1/4 + 0 \cdot \Pr(YZ = 0) + 2 \cdot 0 = -1/2$$

$$\Rightarrow \text{Cov}(Y, Z) = \mathbb{E}[YZ] = -1/2$$

$Y$  and  $Z$  are negatively correlated so they cannot be independent. Intuitively, both random variables depend on  $X_1$ , so knowing  $Y$  or  $Z$  tells us something about  $X_1$  which in turn tells us something about  $Z$  or  $Y$ , respectively.

We can also see this directly

$$\text{Cov}(Y, Z) = \text{Cov}(2X_1, X_2 - X_1) = \underbrace{2\text{Cov}(X_1, X_2)}_{=0} - \text{Cov}(X_1, X_1) = -2\text{Var}(X_1).$$

We know the variance of a Bernoulli random variable is  $p(1-p) = 1/4$ , hence  $-2\text{Var}(X_1) = -1/2$ .

## Q2

Let  $X$  be a continuous random variable CDF  $F(x) = \log_c(x)$  where  $c \in \mathbb{R}_{++}$ .

(a) What is the support of  $X$ ?

Let the lower bound of the support be  $\hat{a}$  and the upper bound be  $\hat{b}$ . We know  $F(\hat{a}) = 0$  and  $F(\hat{b}) = 1$ . Hence,  $x \in [1, c]$ .

(b) What is the  $P(a \leq x \leq b)$  where  $a < c$  and  $b > c$ ?

We have to construct different cases where  $[a, b]$  overlaps with the support.

Case I:  $[a, b] \subseteq [1, c]$

$$P(a \leq x \leq b) = F(b) - F(a) = \log_c(b) - \log_c(a) = \log_c(b/a)$$

Case II:  $1 < c \leq a$

$$P(a \leq x \leq b) = 0$$

Case III:  $1 < a < c$  and  $b > c$

$$P(a \leq x \leq b) = F(c) - F(a) = 1 - \log_c(a) = \log_c(c/a)$$

Case IV:  $1 < b < c$  and  $a < 1$

$$P(a \leq x \leq b) = F(b) - F(0) = \log_c(b)$$

(c) Calculate  $\mathbb{E}[X]$ .

$$\begin{aligned}\mathbb{E}[X] &= \int_1^c x \cdot F'(x) dx \\ &= \int_1^c x \cdot \frac{1}{x \log(c)} dx \\ &= \int_1^c \frac{1}{\log(c)} dx \\ &= \frac{c-1}{\log(c)}\end{aligned}$$

(d) Calculate  $\mathbb{E}[X^2]$ .

$$\begin{aligned}\mathbb{E}[X^2] &= \int_1^c x^2 \cdot F'(x) dx \\ &= \int_1^c x^2 \cdot \frac{1}{x \log(c)} dx \\ &= \int_1^c \frac{x}{\log(c)} dx \\ &= \frac{1}{\log(c)} \int_1^c x dx \\ &= \frac{c^2 - 1}{2 \log(c)}\end{aligned}$$

Note that now we can calculate  $Var(X)$ .

$$Var(X) = \frac{c^2 - 1}{2 \log(c)} - \left( \frac{c - 1}{\log(c)} \right)^2$$