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?).

$\mathbf{Q}\mathbf{1}$

Let $X_1 \sim Bernoulli(1/2)$ independently of $X_2 \sim Bernoulli(2/3)$ 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 $suppY = \{0,2\}$ and $suppZ = \{-1,0,1\}$. Then p(y,z) is tabular form is

(b) Find the marginal PMF of Y.

$$p_Y(0) = 1/2$$

 $p_Y(2) = 1/2$

(c) Find the conditional PMF of Z?

$$p_{Z|Y}(z|y) = p(y, z)/p(y)$$

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

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

(d) Calculate Cov(Y, Z). Are Y and Z independent? What is the intuition?

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

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

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

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

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

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

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

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

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

$\mathbf{Q2}$

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 $P(a \le x \le b)$ where a < e and b > e?

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

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

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

Case II: $1 < c \le a$

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

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

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

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

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

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

$$\mathbb{E}[X] = \int_{1}^{c} x \cdot F'(x) dx$$
$$= \int_{1}^{c} x \cdot \frac{1}{x \log(c)} dx$$
$$= \int_{1}^{c} \frac{1}{\log(c)} dx$$
$$= \frac{c - 1}{\log(c)}$$

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

$$\mathbb{E}[X^2] = \int_1^c x^2 \cdot F'(x) dx$$

$$= \int_1^c x^2 \cdot \frac{1}{x \log(c)} dx$$

$$= \int_1^c \frac{x}{\log(c)} dx$$

$$= \frac{1}{\log(c)} \int_1^c x dx$$

$$= \frac{c^2 - 1}{2 \log(c)}$$

Note that now we can calculate Var(X).

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