

$$\begin{aligned} \S 6.2 \quad \textcircled{1} \text{ a) First, for } x=1, 2, \dots, 6, \quad P(X=x) &= P(\min\{X_1, X_2\}=x) \\ &= P(X_1=x, X_2 \geq x) + P(X_1 > x, X_2=x) = \frac{1}{6} \cdot \frac{6-x+1}{6} + \frac{6-x}{6} \cdot \frac{1}{6} \\ &= \frac{12-2x+1}{36} = \frac{13-2x}{36} \end{aligned}$$

$$\begin{aligned} \text{Next, } P(Y=x|X=x) &= \frac{P(Y=x, X=x)}{P(X=x)} = \frac{P(X_1=x, X_2=x)}{P(X=x)} \\ &= \frac{\frac{1}{6} \times \frac{1}{6}}{\frac{13-2x}{36}} = \frac{1}{13-2x} = \frac{1}{13-2x} \end{aligned}$$

$$\begin{aligned} \text{And for } y=x+1, \dots, 6, \\ P(Y=y|X=x) &= \frac{P(Y=y, X=x)}{P(X=x)} \\ &= \frac{P(X_1=x, X_2=y) + P(X_1=y, X_2=x)}{P(X=x)} \\ &= \frac{2}{13-2x} \end{aligned}$$

Thus,

$$\begin{aligned} E(Y|X=x) &= \sum_{y=x}^6 y P(Y=y|X=x) \\ &= x P(Y=x|X=x) + \sum_{y=x+1}^6 y P(Y=y|X=x) \\ &= \frac{x}{13-2x} + \frac{2}{13-2x} \sum_{y=x+1}^6 y \\ &= \frac{x}{13-2x} + \frac{2}{13-2x} \left(21 - \sum_{y=1}^x y \right) \\ &= \frac{x}{13-2x} + \frac{2}{13-2x} \left(21 - \frac{x(x+1)}{2} \right) \\ &= \frac{x + 42 - x(x+1)}{13-2x} = \frac{42-x^2}{13-2x} \end{aligned}$$

b) $E(X|Y=y)$ is calculated in a similar way!

$$\S 6.2 \quad \textcircled{6} \quad \text{a) } P(\text{all } X\text{'s} \leq t \mid N=n) = P(X_1 \leq t) \cdots P(X_n \leq t) = t^n.$$

$$\begin{aligned} \text{b) } P(\text{all } X\text{'s} \leq t) &= \sum_{n=0}^{\infty} P(\text{all } X\text{'s} \leq t \mid N=n) P(N=n) \\ &= \sum_{n=0}^{\infty} t^n e^{-\mu} \frac{\mu^n}{n!} = \sum_{n=0}^{\infty} e^{-\mu} \frac{(\mu t)^n}{n!} = e^{-\mu} e^{\mu t} \\ &= e^{-\mu(1-t)} \end{aligned}$$

$$\text{c) } P(S_N=0) = P(N=0) = e^{-\mu}, \text{ since if } N \geq 1, \text{ then } S_N \geq S_1 = X_1 \text{ and } P(X_1 > 0) = 1.$$

$$\begin{aligned} \text{d) } E(S_N) &= \sum_{n=0}^{\infty} E(S_N \mid N=n) P(N=n) \\ &= \sum_{n=0}^{\infty} E(S_n \mid N=n) P(N=n) \\ &= \sum_{n=0}^{\infty} E(X_1 + \dots + X_n \mid N=n) P(N=n) \\ &\stackrel{(*)}{=} \sum_{n=0}^{\infty} E(X_1 + \dots + X_n) P(N=n) \\ &= \sum_{n=0}^{\infty} \frac{n}{2} P(N=n) = \frac{1}{2} E(N) = \frac{\mu}{2} \end{aligned}$$

(*): assuming that X_1, X_2, \dots are independent from N .

$$\S 6.3 \quad \textcircled{2} \quad \text{a) For } 0 < x < 1, f_X(x) = \int_0^1 f(x,y) dy = \int_0^1 (2x + 2y - 4xy) dy \\ = [2xy + y^2 - 2xy^2]_0^1 = 2x + 1 - 2x = 1 \quad (X \sim \text{uniform}(0,1)).$$

$$\text{For } 0 < y < 1, f_Y(y) = \int_0^1 (2x + 2y - 4xy) dx = 1 \quad (Y \sim \text{uniform}(0,1)).$$

$$\text{b) } f_Y(y \mid X = \frac{1}{4}) = \frac{f(\frac{1}{4}, y)}{f_X(\frac{1}{4})} = f(\frac{1}{4}, y) = \frac{1}{2} + 2y - y = y + \frac{1}{2}$$

$$\begin{aligned} \text{c) } E(Y \mid X = \frac{1}{4}) &= \int_0^1 y f_Y(y \mid X = \frac{1}{4}) dy = \int_0^1 y (y + \frac{1}{2}) dy \\ &= \int_0^1 (y^2 + \frac{1}{2}y) dy = \frac{1}{3} + \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{3} + \frac{1}{4} = \frac{7}{12}. \end{aligned}$$

(Remark: A joint density on $[0,1]^2$ of this type, with uniform marginal densities, is called a copula, and used to encode the dependence of two random variables with given marginal distributions.)

$$\S 6.3 \quad \textcircled{1} \quad P(A) = \int_0^1 P(A \mid X=x) f_X(x) dx = \int_0^1 x^2 \cdot 1 dx = \int_0^1 x^2 dx = \frac{1}{3}.$$