

$$1. a) V[X] = E[X^2] - (E[X])^2$$

variance is always non-negative, so $V[X] \geq 0$ and

$$E[X^2] - (E[X])^2 \geq 0$$

$$E[X^2] \geq (E[X])^2$$

So the statement is true

$$b) P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{P(\emptyset)}{P(B)} = 0$$

We are given that A has positive probability, so $P(A) > 0$.

$P(A|B) = 0 \neq P(A)$, so A and B are not independent and the statement is false

$$c) E[X] = E[Y] = 0$$

$$\begin{aligned} V[XY] &= E[(XY)^2] - (E[XY])^2 && \text{since } X \perp\!\!\!\perp Y, \\ &= E[X^2Y^2] - (E[X]E[Y])^2 && \text{since } E[X] = E[Y] = 0, \\ &= E[X^2Y^2] \\ &= E[X^2]E[Y^2] \\ &= (E[X^2] - (E[X])^2)(E[Y^2] - (E[Y])^2) \\ &= V[X]V[Y] \\ &= \sigma_x^2 \sigma_y^2 \end{aligned}$$

The statement is true

$$d) E[E[X|Y]] = E[X] \quad E[E[Y|X]] = E[Y]$$

(It is not true that $E[X] = E[Y]$ for any two random variables X and Y, so $E[E[X|Y]] \neq E[E[Y|X]]$ and the statement is false)

$$e) P(a \leq X \leq b) = \int_a^b f(x) dx$$

$$P(a \leq X \leq a) = \int_a^a f(x) dx = 0$$

So the statement is true

$$\begin{aligned} 2. \quad P(\text{spam} | \text{free}) &= \frac{P(\text{free} | \text{spam}) P(\text{spam})}{P(\text{free})} \\ &= \frac{0.9 \cdot 0.4}{0.9(0.4) + 0.05(0.5) + 0.05(0.1)} \\ &= 0.923 \end{aligned}$$

3. Let Y_n be the number of left jumps after n jumps. Since a left jump happens w/ probability p , we have that $Y_n \sim \text{Bin}(n, p)$

The number of right jumps after n jumps is $n - Y_n$, so

$$X_n = \text{right jumps} - \text{left jumps} = n - Y_n - Y_n = n - 2Y_n$$

$$\begin{aligned} E[X_n] &= E[n - 2Y_n] \\ &= n - 2E[Y_n] \\ &= n - 2np \\ &= n(1 - 2p) \end{aligned}$$

taking out constants,
since $Y_n \sim \text{Bin}(n, p)$,

$$\begin{aligned} V[X_n] &= V[n - 2Y_n] \\ &= V[-2Y_n] \\ &= 4V[Y_n] \\ &= 4np(1-p) \end{aligned}$$

taking out constants,
since $Y_n \sim \text{Bin}(n, p)$,

$$4. f_{X,Y}(x,y) = \frac{e^{-x/y} e^{-y}}{y} \quad 0 < x < \infty \quad 0 < y < \infty$$

$$E[X|Y=y] = \int_0^{\infty} x f_{X|Y}(x|y) dx$$

$$= \int_0^{\infty} x \frac{f_{X,Y}(x,y)}{f_Y(y)} dx$$

$$f_Y(y) = \int_0^{\infty} f_{X,Y}(x,y) dx = \int_0^{\infty} \frac{e^{-x/y} e^{-y}}{y} dx$$

$$= \frac{e^{-y}}{y} \int_0^{\infty} e^{-x/y} dx$$

$$= \frac{e^{-y}}{y} \cdot -ye^{-x/y} \Big|_0^{\infty} = -e^{-y} e^{-x/y} \Big|_0^{\infty}$$

$$= (-e^{-y}(0)) - (-e^{-y}) = e^{-y}$$

$$E[X|Y=y] = \int_0^{\infty} x \frac{e^{-x/y} e^{-y}}{y} \cdot \frac{1}{e^{-y}} dx = \frac{1}{y} \int_0^{\infty} x e^{-x/y} dx$$

$$u = x \quad dv = e^{-x/y} dx$$

$$du = dx \quad v = -ye^{-x/y}$$

$$= \frac{1}{y} \left[-xye^{-x/y} - \int -ye^{-x/y} dx \right]$$

$$= \frac{1}{y} \left[-xye^{-x/y} + y \int e^{-x/y} dx \right]$$

$$= \frac{1}{y} \left[-xye^{-x/y} \Big|_0^{\infty} + y(-ye^{-x/y}) \Big|_0^{\infty} \right]$$

$$= \frac{1}{y} \left[0 + (y^2) \right] = y$$

$$5. \quad M_{X,Y}(s,t) = \exp(1 + s + s^2 - \cos t + s^{2023} t)$$

$$\begin{aligned} M_{X,Y}(s,0) &= E[\exp(sX + Y(0))] \\ &= E[\exp(sX)] \\ &= M_X(s) \end{aligned}$$

$$\begin{aligned} M_X(s) = M_{X,Y}(s,0) &= \exp(1 + s + s^2 - 1) \\ &= \exp(s + s^2) \\ &= \exp((1)s + 2s^2/2) \Rightarrow \mu = 1, \sigma^2 = 2 \end{aligned}$$

Therefore, $X \sim N(1, 2)$

$$\begin{aligned}
 6. \quad P(2 < X < 8) &= P(-3 < X - 5 < 3) \\
 &= P(|X - 5| < 3) \\
 &= P(|X - \mu| < 3) \\
 &= 1 - P(|X - \mu| \geq 3) \\
 &\geq 1 - \frac{\sigma^2}{3^2} \quad \text{by Chebyshev's Inequality} \\
 &= 1 - \frac{3}{9} = \frac{2}{3}
 \end{aligned}$$

$$P(2 < X < 8) \geq \frac{2}{3}$$

$$\begin{aligned}
 \tau. \quad & P\left(\sum_{i=1}^n T_i > 2E\left[\sum_{i=1}^n T_i\right]\right) = P\left(\sum_{i=1}^n T_i > 2\sum_{i=1}^n E[T_i]\right) \\
 & = P(n\bar{T}_n > 2n\mu) = P(\bar{T}_n > 2\mu) = 1 - P(\bar{T}_n \leq 2\mu) \\
 & = 1 - P(\bar{T}_n - \mu \leq \mu) \\
 & = 1 - P\left(\frac{\bar{T}_n - \mu}{\sigma/\sqrt{n}} \leq \frac{\mu}{\sigma/\sqrt{n}}\right)
 \end{aligned}$$

By CLT,

$$\begin{aligned}
 & = 1 - \Phi\left(\frac{\mu}{\sigma/\sqrt{n}}\right) \\
 & = \Phi\left(-\frac{\mu}{\sigma/\sqrt{n}}\right)
 \end{aligned}$$

$$8. a) E[X] = E[E[X | \text{raining}]]$$

$$= E[X | \text{raining}] P(\text{raining}) + E[X | \text{dry}] P(\text{dry})$$

$$= 9(0.1) + 3(0.9)$$

$$= 3.6$$

$$b) P(X=0) = P(X=0 | \text{raining}) P(\text{raining}) + P(X=0 | \text{dry}) P(\text{dry})$$

$$= e^{-\lambda_r} \frac{\lambda_r^0}{0!} (0.1) + e^{-\lambda_d} \frac{\lambda_d^0}{0!} (0.9)$$

$$= e^{-9} (0.1) + e^{-3} (0.9)$$

$$= 0.0448$$

$$c) V[X] = E[X^2] - (E[X])^2$$

$$= E[X^2 | \text{raining}] P(\text{raining}) + E[X^2 | \text{dry}] P(\text{dry}) - 3.6^2$$

$$= (\lambda_r^2 + \lambda_r) (0.1) + (\lambda_d^2 + \lambda_d) (0.9) - 3.6^2$$

$$= (90)(0.1) + (12)(0.9) - 3.6^2$$

$$= 6.84$$