

53. If X is uniform over $(0, 1)$, calculate $E[X^n]$ and $\text{Var}(X^n)$.

55. Suppose that the joint probability mass function of X and Y is

$$P(X = i, Y = j) = \binom{j}{i} e^{-2\lambda} \lambda^j / j!, \quad 0 \leq i \leq j$$

- (a) Find the probability mass function of Y .
- (b) Find the probability mass function of X .
- (c) Find the probability mass function of $Y - X$.

61. Let X and W be the working and subsequent repair times of a certain machine. Let $Y = X + W$ and suppose that the joint probability density of X and Y is

$$f_{X,Y}(x, y) = \lambda^2 e^{-\lambda y}, \quad 0 < x < y < \infty$$

- (a) Find the density of X .
- (b) Find the density of Y .
- (c) Find the joint density of X and W .
- (d) Find the density of W .

76. Let X and Y be independent random variables with means μ_x and μ_y and variances σ_x^2 and σ_y^2 . Show that

$$\text{Var}(XY) = \sigma_x^2 \sigma_y^2 + \mu_y^2 \sigma_x^2 + \mu_x^2 \sigma_y^2$$

79. With $K(t) = \log(E[e^{tX}])$, show that

$$K'(0) = E[X], \quad K''(0) = \text{Var}(X)$$

$$53. E[X^n] = \int_0^1 x^n \cdot 1 \cdot dx = \left. \frac{x^{n+1}}{n+1} \right|_0^1 = \underline{\underline{\frac{1}{n+1}}}$$

$$E[X^{2n}] = \int_0^1 x^{2n} dx = \left. \frac{x^{2n+1}}{2n+1} \right|_0^1 = \frac{1}{2n+1}$$

$$\begin{aligned} V[X] &= E[X^{2n}] - (E[X^n])^2 = \frac{1}{2n+1} - \frac{1}{(n+1)^2} \\ &= \frac{n^2 + 2n + 1 - (2n+1)}{(n+1)^2(2n+1)} \\ &= \underline{\underline{\frac{n^2}{(n+1)^2(2n+1)}}} \end{aligned}$$

$$\begin{aligned} 55. a) P(Y=j) &= \sum_{i=0}^j P(X=i, Y=j) = \sum_{i=0}^j \binom{j}{i} e^{-2\lambda} \frac{\lambda^j}{j!} \\ &= e^{-2\lambda} \frac{\lambda^j}{j!} \underbrace{\sum_{i=0}^j \binom{j}{i}}_{2^j} = \frac{e^{-2\lambda} 2^j \lambda^j}{j!}, j \geq 0 \\ &\quad \text{Poisson}(2\lambda) \end{aligned}$$

$$\begin{aligned} b) P(X=i) &= \sum_{j=i}^{\infty} P(X=i, Y=j) = \sum_{j=i}^{\infty} \binom{j}{i} e^{-2\lambda} \frac{\lambda^j}{j!} \\ &\quad \text{let } k=j-i \\ &= \sum_{k=0}^{\infty} \binom{k+i}{i} e^{-2\lambda} \frac{\lambda^{k+i}}{(k+i)!} \end{aligned}$$

$$= e^{-2\lambda} \sum_{k=0}^{\infty} \frac{\cancel{(k+i)!}}{i! k!} \frac{\lambda^{k+i}}{\cancel{(k+i)!}} = \frac{e^{-2\lambda} \lambda^i}{i!} \underbrace{\sum_{k=0}^{\infty} \frac{\lambda^k}{k!}}_{e^{\lambda}}$$

$$= \frac{e^{-\lambda} \lambda^i}{i!}, \quad i \geq 0$$

Poisson(λ)

$$\begin{aligned} \text{c) } P(Y-X=k) &= \sum_{i=0}^{\infty} P(X=i \cap Y=i+k) \\ &= \sum_{i=0}^{\infty} \binom{k+i}{i} e^{-2\lambda} \frac{\lambda^{k+i}}{(k+i)!} = \frac{e^{-\lambda} \lambda^k}{k!} \quad (\text{as in (b)}) \end{aligned}$$

Poisson(λ)

$$\begin{aligned} \text{6) a) } f_X(x) &= \int_x^{\infty} \lambda^2 e^{-\lambda y} dy = \lambda^2 \frac{e^{-\lambda y}}{(-\lambda)} \Big|_{y=x}^{\infty} = 0 - (-\lambda e^{-\lambda x}) \\ &= \lambda e^{-\lambda x}, \quad x > 0 \end{aligned}$$

(Exponential(λ))

$$\begin{aligned} \text{b) } f_Y(y) &= \int_0^y \lambda^2 e^{-\lambda x} dx = \lambda^2 e^{-\lambda x} x \Big|_{x=0}^y \\ &= \lambda^2 y e^{-\lambda y} \end{aligned}$$

Gamma($\alpha=2, \lambda$)

c) let $V = X$ $X = V$ $J^* = \begin{vmatrix} 1 & 0 \\ 1 & 1 \end{vmatrix} = 1$
 $W = Y - X$ $Y = V + W$

$$g(v, w) = f(x, y) \cdot 1 = \lambda^2 e^{-\lambda(v+w)}, \quad 0 < v < v+w < \infty,$$

or $v > 0$
 $w > 0$

d) Note that $g(v, w) = \lambda e^{-\lambda v} \cdot \lambda e^{-\lambda w}$

So $f_W(w) = \lambda e^{-\lambda w}, w > 0$
 $\text{Exp}(\lambda)$

76. $V(XY) = E[X^2 Y^2] - (E[XY])^2$
 $= E[X^2] E[Y^2] - (E[X] E[Y])^2$ by independence
 $= (\sigma_x^2 + \mu_x^2)(\sigma_y^2 + \mu_y^2) - \mu_x^2 \mu_y^2$
 $= \sigma_x^2 \sigma_y^2 + \sigma_x^2 \mu_y^2 + \sigma_y^2 \mu_x^2$

77. $K'(t) = \frac{1}{E[e^{tx}]} \cdot E[X e^{tx}] \Rightarrow K'(0) = E[X]$

$$K''(t) = \frac{E[e^{tx}] E[X^2 e^{tx}] - E[X e^{tx}] E[X e^{tx}]}{(E[e^{tx}])^2} \Rightarrow K''(0) = \frac{E[X^2] - (E[X])^2}{1} = V[X]$$