

ECE600 Random Variables and Waveforms
Fall 2020

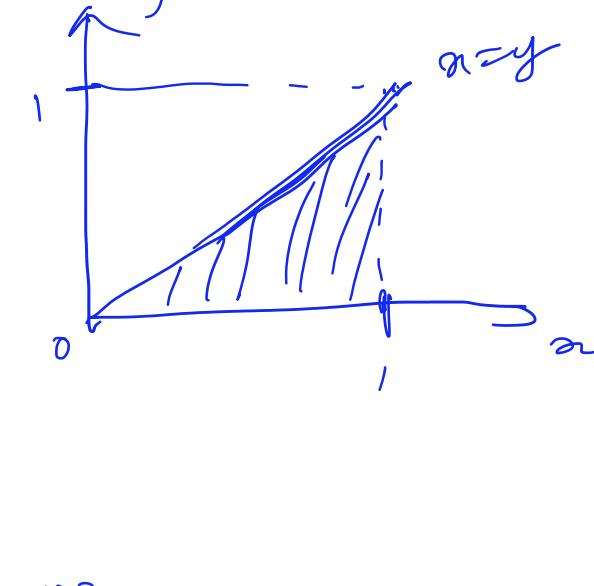
Midterm Exam #3
Session 25
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1	
2	
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8	
Total	

Q1

Given: $f_{XY}(x,y) = kx^2y \mathbb{1}_{[0,y]} \mathbb{1}_{[0,1]}$



(a) Finding k , For valid j-distⁿ pdf we have:

$$(i) kx^2y \geq 0, x \geq 0, y \geq 0 \\ \Rightarrow k \geq 0$$

$$(ii) \int_0^1 \int_0^y f_{XY}(x,y) dx dy = 1$$

$$\Rightarrow \int_0^1 \int_0^y kx^2y dx dy = \int_0^1 y \left[\int_0^y kx^2 dx \right] dy = 1$$

$$\Rightarrow \int_0^1 y \left[\frac{kx^3}{3} \right]_0^y dy = 1$$

$$\Rightarrow \frac{k}{3} \int_0^1 y^4 dy = \frac{k}{3 \times 5} y^5 \Big|_0^1 = 1$$

$$\Rightarrow \boxed{K = 15}$$



$$(b) f_Y(y) = \int_y^\infty f_{XY}(x,y) dx$$

$$= \int_y^\infty 15x^2y dx = 15y \left[\frac{x^3}{3} \right]_y^\infty$$

$$f_Y(y) = 5y(1-y^3) \mathbb{1}_{[0,1]}$$

$$(c) f_X(x | \{Y=y\}) = \frac{f_{XY}(x,y)}{f_Y(y)}$$

$$= \frac{15x^2y}{5y(1-y^3)} \mathbb{1}_{[0,y]} = \frac{3x^2}{(1-y^3)} \cdot \mathbb{1}_{[0,y]}$$



$$(d) \hat{X}_{\text{nms}}(y) = E_X [X | \{Y=y\}]$$

$$= \int_{-\infty}^{\infty} x \cdot f_X(x | \{Y=y\}) dx$$

$$= \int_0^y x \cdot \frac{3x^2}{(1-y^3)} dx$$

$$= \frac{3}{(1-y^3)} \left[\frac{x^4}{4} \right]_0^y$$

$$\Rightarrow \boxed{\hat{X}_{\text{nms}}(y) = \frac{3}{4} \frac{y^4}{(1-y^3)}}$$

$$(e) \hat{X}_{\text{MAP}}(y) = \underset{X}{\operatorname{argmax}} \{ f_X(x | \{Y=y\}) \}$$

$$= \underset{x}{\operatorname{argmax}} \left\{ \frac{3x^2}{(1-y^3)} \mathbb{1}_{[0,y]} \right\}$$



$$\Rightarrow \boxed{\hat{X}_{\text{MAP}}(y) = y}$$

$$\text{The function } \frac{3x^2}{(1-y^3)} \text{ (with } y \text{ as constant) is max at end points}$$

Q2

$$\text{Given: } \Phi_{XY}(w_1, w_2) = \frac{1}{(1-iw_1)} \cdot \frac{1}{(1-2iw_2)}$$

$$(a) E[X] = \left. \frac{\partial \Phi_{XY}(w_1, w_2)}{\partial(iw_1)} \right|_{w_1=0, w_2=0}$$

$$= \frac{1}{(1-2iw_2)} \cdot \frac{-1}{(1-iw_1)^2} \Big|_{w_1=0, w_2=0}$$

$$= \frac{1}{(-0)} \times \frac{1}{(-0)^2} = 1$$

$$\Rightarrow E[X] = 1$$

$$(b) E[Y] = \left. \frac{\partial \Phi_{XY}(w_1, w_2)}{\partial(iw_2)} \right|_{w_1=0, w_2=0}$$

$$= \frac{1}{(1-iw_1)} \cdot \frac{-1 \times -2}{(1-2iw_2)^2} \Big|_{w_1=0, w_2=0}$$

$$E[Y] = 2$$

$$(c) E[XY] = \left. \frac{\partial^2 \Phi_{XY}(w_1, w_2)}{\partial(iw_1) \partial(iw_2)} \right|_{\substack{w_1=0 \\ w_2=0}}$$

$$= \frac{-1 \times -1}{(1-iw_1)^2} \cdot \frac{-1 \times -2}{(1-2iw_2)^2} \Big|_{\substack{w_1=0 \\ w_2=0}}$$

$$E[XY] = 2$$

$$(d) E[X^j Y^k] = \left. \frac{\partial^{j+k}}{\partial(iw_1)^j \partial(iw_2)^k} \Phi_{XY}(w_1, w_2) \right|_{\substack{w_1=0 \\ w_2=0}}$$

$$= \frac{\partial^k}{\partial(iw_2)^k} \left(\frac{1}{(1-2iw_2)} \frac{\partial^j}{\partial(iw_1)^j} \frac{1}{(1-iw_1)} \right) \Big|_{\substack{w_1=0 \\ w_2=0}}$$

$$= \frac{2^k k!}{(1-2iw_2)^{k+1}} \cdot \frac{j!}{(1-iw_1)^{j+1}} \Big|_{\substack{w_1=0 \\ w_2=0}}$$

$$E[X^j Y^k] = 2^k k! j!$$

$$(e) \rho_{XY} = \frac{E[XY] - E[X]E[Y]}{\sigma_X \sigma_Y} = \frac{2 - 2 \times 1}{\sigma_X \sigma_Y} = 0$$

$$\boxed{\rho_{XY} = 0}$$

Q3

Given: N is a Poisson variable with mean λ

$$\therefore P_N(n) = e^{-\lambda} \frac{\lambda^n}{n!}, n=0, 1, \dots \infty$$

Probability of car exceeding speed limit: $P(\text{Independent } N)$

$M \sim \text{no. of cars exceeding speed limit}$

$$a) \quad \Phi_M(w) = E[e^{iwM}] = \sum_{k=0}^{\infty} e^{iwk} e^{-\lambda} \frac{\lambda^k}{k!}$$

$$= e^{-\lambda} \sum_{k=0}^{\infty} \underbrace{(e^{iw}\lambda)^k}_{k!}$$

$$\Phi_M(w) = \underline{e^{-\lambda}(1-e^{iw})}$$

$$b) \quad \Phi_M(w) = E[e^{iwM}] \\ = E_N \left[E_M \left[e^{iwM} \mid N \right] \right]$$

$$= \sum_{n=0}^{\infty} p_N(n) E_M \left[e^{iwM} \mid \{N=n\} \right]$$

We need to find $f_{M|N}(m \mid \{N=n\})$

which translates to, given there are n cars

on the road we need to find the probability

that m cars out of n are speeding

Clearly the conditional distribution is a Binomial Distribution with the success probability p

$$\therefore f_{M|N}(m \mid \{N=n\}) = \binom{n}{m} p^m (1-p)^{n-m}$$

$$\therefore E_M \left[e^{iwM} \mid \{N=n\} \right] = \sum_{m=0}^n e^{iw m} \binom{n}{m} p^m (1-p)^{n-m}$$

$$= \sum_{m=0}^n \binom{n}{m} (pe^{iw})^m (1-p)^{n-m}$$

$$= (pe^{iw} + 1-p)^n$$

$$\therefore \Phi_M(w) = \sum_{n=0}^{\infty} e^{-\lambda} \frac{\lambda^n}{n!} \cdot (pe^{iw} + 1-p)^n$$

$$= e^{-\lambda} \sum_{n=0}^{\infty} \underbrace{\left(\lambda (pe^{iw} + 1-p) \right)^n}_{n!}$$

$$= e^{-\lambda} (-\lambda + \lambda (pe^{iw} + 1-p))$$

$$\Phi_M(w) = e^{-\lambda} (-\lambda + pe^{iw} - 1)$$

$$= e^{-\lambda} p (1 - e^{iw})$$

$$+ e^{-\lambda} p e^{iw} \cdot (\lambda p) \cdot (-\lambda p (1 - e^{iw}))$$

$$= \lambda p + (\lambda p)^2$$

$$\therefore \sigma_M^2 = E[M^2] - (E[M])^2$$

$$E[M^2] = \frac{\partial^2}{\partial w^2} \Phi_M(w) \Big|_{w=0}$$

$$= \frac{\partial}{\partial w} (2pe^{iw} \cdot e^{-\lambda} p (1 - e^{iw})) \Big|_{w=0}$$

$$= 2pe^{iw} \cdot e^{-\lambda} p (1 - e^{iw})$$

$$+ 2pe^{iw} \cdot (\lambda p) \cdot e^{-\lambda} p (1 - e^{iw}) \Big|_{w=0}$$

$$= \lambda p + (\lambda p)^2$$

$$\therefore \sigma_M^2 = \lambda p + (\lambda p)^2 - (\lambda p)^2$$

$$\boxed{\sigma_M^2 = \lambda p}$$

Q4

Given: $X_1, X_2, X_3, \dots, X_n$ is a sequence of i.i.d.

exponential R.V. with mean μ

Another sequence $\{Y_n\}$ defined as:

$$Y_n = \min\{X_1, X_2, \dots, X_n\} \quad n=1, 2, 3, \dots$$

(a) As the sequence $\{X_n\}$ is i.i.d. exponential R.V. with mean μ

∴ Their joint dist' is given by:

$$f_{X_1, X_2, \dots, X_n}(x_1, x_2, \dots, x_n) = \prod_{i=1}^n f_{X_i}(x_i) = \prod_{i=1}^n \frac{1}{\mu} e^{-\frac{x_i}{\mu}} \quad x_i \geq 0, \mu > 0$$

$$= \frac{1}{\mu^n} \exp\left(-\frac{\sum_{i=1}^n x_i}{\mu}\right)$$

Also since $X_i \perp\!\!\! \perp X_j \forall i \neq j$ we can compute any marginal as:

$$f_{X_1, X_2, \dots, X_j}(x_1, x_2, \dots, x_j) = \frac{1}{\mu^j} \exp\left(-\frac{\sum_{i=1}^j x_i}{\mu}\right) \quad \text{--- Eq(1)}$$

For $n=1$

$Y_1 = \min\{X_1\} = X_1$ & Y_1 is exponential dist'

$$\therefore Y_1 \sim \frac{1}{\mu} e^{-\frac{y_1}{\mu}} \quad y \geq 0, \mu > 0$$

for any $n \geq 1$

$$Y_n = \min\{X_1, X_2, \dots, X_n\}$$

$$F_{Y_n}(y) = P(\{Y_n \leq y\} \cap \{\bigcup_{i=1}^n X_i = \min\{X_1, X_2, \dots, X_n\}\})$$

$$= \sum_{i=1}^n P(\{X_i \leq y\} \cap \{X_i = \min\{X_1, X_2, \dots, X_n\}\})$$

$$= 1 - P(\{Y_n > y\})$$

$$= 1 - \prod_{i=1}^n P(\{X_i > y\})$$

$$= 1 - \prod_{i=1}^n \left(1 - e^{-\frac{y}{\mu}}\right) = 1 - \left(1 - e^{-\frac{y}{\mu}}\right)^n$$

$$\therefore f_{Y_n}(y) = \frac{\partial}{\partial y} F_{Y_n}(y)$$

$$= n \left(1 - e^{-\frac{y}{\mu}}\right)^{n-1} \times \frac{1}{\mu}$$

$$= \frac{n}{\mu} \left(1 - e^{-\frac{y}{\mu}}\right)^{n-1} \cancel{\frac{1}{\{0\}^{(y)}}}$$

$$(b) P(|Y_n - Y| > \epsilon Y) \rightarrow 0 \text{ as } n \rightarrow \infty ?$$

$$E[Y_n^2] = \int_0^\infty y^2 \frac{n}{\mu} \left(1 - e^{-\frac{y}{\mu}}\right)^{n-1} dy$$

$$(c) E[|Y_n - Y|^2] \rightarrow 0 \text{ as } n \rightarrow \infty ?$$