

1. Poisson Process: Direct Calculation [5 points each]

Consider a Poisson Process with intensity λ . Let Y_K denote the time to see the K th arrival. Let us use the following direct approach to calculate the joint PDF of Y_1, Y_2, \dots, Y_K .

- (a) Consider any two positive numbers $y_1 < y_2$. Observe that $(X(y_2) - X(y_1))$ is the number of arrivals in the interval $(y_1, y_2]$, and hence $X(y_2) - X(y_1)$ is independent of $X(y_1)$. Using this show

$$P(X(y_1) < 1, X(y_2) < 2) = e^{-\lambda y_2} (1 + \lambda(y_2 - y_1))$$

$$(a) P(N=k) = \frac{\lambda^k e^{-\lambda}}{k!} \quad P(X(y_1)=0, X(y_2)-X(y_1)<2) = P(X(y_1)=0, X(y_2)-X(y_1)<2)$$

For $P(X(y_1)=0)$: $k=0 \quad \tilde{\lambda} = \lambda y_1 \quad P(X(y_1)=0) = \frac{(\lambda y_1)^0 e^{-\lambda y_1}}{0!} = e^{-\lambda y_1}$

For $P(X(y_2)-X(y_1)=1)$: $k=1 \quad \tilde{\lambda} = \lambda(y_2-y_1) \quad P(X(y_2)-X(y_1)=1) = \frac{e^{-\lambda(y_2-y_1)} \cdot \lambda(y_2-y_1)}{1!}$

$P(X(y_2)-X(y_1)=0)$: $k=0 \quad \tilde{\lambda} = \lambda(y_2-y_1) \quad P(X(y_2)-X(y_1)=0) = e^{-\lambda(y_2-y_1)}$

$$P(X(y_1)=0, X(y_2)-X(y_1)<2) = e^{-\lambda y_1} \cdot e^{-\lambda(y_2-y_1)} (\lambda(y_2-y_1)+1) = e^{-\lambda y_2} (\lambda(y_2-y_1)+1)$$

- (b) Use part (a) to find $P(Y_1 > y_1, Y_2 > y_2)$, and then find the joint PDF of Y_1 and Y_2 using the inclusion-exclusion principle.

$$(b) P(Y_1 > y_1, Y_2 > y_2) = P(X(y_1) < 1, X(y_2) < 2) = e^{-\lambda y_2} (1 + \lambda(y_2 - y_1))$$

$$P(Y_1 \leq y_1, Y_2 \leq y_2) = 1 - P(Y_1 > y_1 \cup Y_2 > y_2) = 1 - P(Y_1 > y_1) - P(Y_2 > y_2) + P(Y_1 > y_1, Y_2 > y_2)$$

$$F_{Y_1, Y_2}(y_1, y_2) = 1 - P(X(y_1) < 1) - P(X(y_2) < 2) + P(X(y_1) < 1, X(y_2) < 2)$$

$$e^{-\lambda y_2} (1 + \lambda(y_2 - y_1)) \quad dy_1$$

- (c) To extend this idea to the general case, first prove the following result (again using the inclusion-exclusion principle) for any K -dimensional random vector \mathbf{Z} :

$$f_{\mathbf{Z}}(\mathbf{z}) = (-1)^K \frac{\partial^K}{\partial \mathbf{z}} P(\mathbf{Z} > \mathbf{z}).$$

(d) Using ideas from (a), (b) and (c), find the joint PDF of Y_1, Y_2, \dots, Y_K .

Hint: Use the inclusion-exclusion principle one more time and evaluate only that term in $P(\mathbf{Y} \geq \mathbf{y})$ that contains all variables y_1, y_2, \dots, y_K . The rest of the terms will be canceled by the partial derivatives.

2. Wide-Sense Stationary [15 points]

Consider a random process $\{X_t\}$ such that

$$X_t = A \sin(t + \Theta),$$

where A is a Bernoulli random variable with mean $1/4$, Θ is uniformly distributed over $[0, 2\pi]$, and A and Θ are independent.

- Is $\{X_t\}$ WSS?

$$A \sin(t + \Theta) = A \sin t \cos \Theta + A \cos t \sin \Theta \quad \mu[X_t] = E[A] \cdot (E[\cos \Theta] \cdot \sin t + E[\sin \Theta] \cdot \cos t)$$

Since Θ is uniformly distributed over $[0, 2\pi]$: $E[\cos \Theta] = E[\sin \Theta] = 0$

$$\mu_{X_t} = 0$$

$$\begin{aligned} R_{s, s+t} &= E[A^2] E[\sin(s+\Theta) \sin(s+t+\Theta)] \\ &= E[A^2] \cdot \frac{1}{2} \cdot E[\cos(s+\Theta - s-t-\Theta) - \cos(s+\Theta + s+t+\Theta)] \\ &= E[A^2] \cdot \frac{1}{2} \cdot E[\cos(-t) - \cos(2s+t+2\Theta)] \\ &= \frac{E[A^2]}{2} \cos(t) - \frac{E[A^2]}{2} E[\cos(s+s+t+2\Theta)] \\ &= \frac{E[A^2]}{2} \cos(t) - \frac{E[A^2]}{2} \int_0^{2\pi} \cos(2s+t+2\Theta) d\Theta \cdot \frac{1}{2\pi} \\ &= \frac{E[A^2]}{2} \cos(t) - \frac{E[A^2]}{2 \times 2\pi} \left[\frac{\sin(2s+t+2\Theta)}{2} \right] \Big|_0^{2\pi} \\ &= \frac{E[A^2]}{2} \cos(t) = \frac{1}{8} \cos(t) \end{aligned}$$

Therefore, $\{X_t\}$ is WSS.

3. Wide-Sense Stationary 2 [15 points]

Consider a random process $\{X_t\}$ such that

$$X_t = A \sin(t + \Theta) + B,$$

where A is a Bernoulli random variable with mean $1/4$, B is a Bernoulli random variable with mean $1/2$, Θ is uniformly distributed over $[0, 2\pi]$, and A , B , and Θ are independent from each other.

- Is $\{X_t\}$ WSS?

$$X_t = A \sin(t + \Theta) + B = A (\sin t \cos \Theta + \cos t \sin \Theta) + B$$

$$\mu[X_t] = E[A] E(\sin t \cos \Theta + \cos t \sin \Theta) + E[B] \quad \text{Based on the above question 2:}$$

$$\mu[X_t] = 0 + E[B] = \frac{1}{2} \quad (\text{a Constant})$$

$$\begin{aligned} R_{s, s+t} &= E[X_s X_{s+t}] = E[(A \sin(s + \Theta) + B)(A \sin(s + t + \Theta) + B)] \\ &= E[A^2 \sin(s + \Theta) \sin(s + t + \Theta) + AB \sin(s + \Theta) + AB \sin(s + t + \Theta) + B^2] \end{aligned}$$

Based on $\sin A \cdot \sin B = \frac{1}{2} [\cos(A - B) - \cos(A + B)]$, besides, A, B are Bernoulli distri, $E(A^2) = \frac{1}{4}$ $E(B^2) = \frac{1}{2}$

$$\begin{aligned} R_{s, s+t} &= E[A^2 \cdot \frac{1}{2} [\cos t - \cos(s + s + t + 2\Theta)] + AB \sin(s + \Theta) + AB \sin(s + t + \Theta) + B^2] \\ &= \frac{E[A^2]}{2} \cdot E[\cos t - \cos(s + s + t + 2\Theta)] + E[AB \sin(s + \Theta)] + E[AB \sin(s + t + \Theta)] + E[B^2] \\ &= \frac{E[A^2]}{2} \cdot \cos t + E[B^2] = \frac{1}{8} \cos t + \frac{1}{2} \end{aligned}$$

Therefore $\{X_t\}$ is WSS