

Quantitative methods, Assignment Chapter 6

Problem 1 (30 points)

Consider the process  $\{X(t), t \geq 0\}$  defined by

$$X(t) = \begin{cases} 1 & \text{if } t \leq Y \\ 0 & \text{if } t > Y \end{cases}$$

where  $Y$  is a uniformly distributed random variable on the interval  $(0, 1)$

1. (5 points) Compute, for  $t \in [0, 1]$ , the first-order probability mass of  $X$ ,  $f(t, x)$ .

→ Using C.D.F. (of) uniform distribution.

$$P(Y \geq t) = \frac{1-t}{1-0} = 1-t$$

$$P(Y < t) = \frac{t-0}{1-0} = t$$

∴ Probability mass of  $X$

$$f(t, x) = \begin{cases} 1-t & \text{when } x=1 \\ t & \text{when } x=0 \\ 0 & \text{otherwise} \end{cases}$$

2. What is first-order probability mass of  $X$  for

→ for  $t > 1 \rightarrow$  implies  $Y < t \rightarrow$  always the probability mass of  $X$

$$\text{is } X(t) = \begin{cases} 1 & p=0 \\ 0 & p=1 \end{cases}$$

3. (5 points). Compute the expectation and the variance of  $X$  at time  $t$ , for  $t \in (0, 1)$ .

$$\rightarrow E[X(t)] = \sum x_i p_i$$

$$= 0 \cdot (1-t) + 1 \cdot (1-t) + 0 \cdot t$$

$$E[X(t)] = 1 - t$$

$$\text{var}[X(t)] = E[X^2(t)] - (E[X(t)])^2$$

$$E[X^2(t)] = \sum x_i^2 p_i$$

$$= 0^2 \cdot (1-t) + 1^2 \cdot (1-t) + 0^2 \cdot t$$

$$= 1 - t - t = 1 - 2t + t^2$$

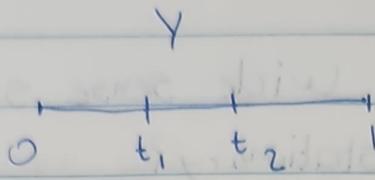
$$\therefore \text{var}[X(t)] = 1 - t - (1-t)^2$$

$$= 1 - t - (1 - 2t + t^2)$$

$$= t - t^2$$

4. (5 points) Compute the autocovariance function of  $X$ ,  $C_X(t_1, t_2)$  for  $t_1, t_2 \in (0, 1)$  with  $t_1 \leq t_2$ .

$$C_X(t_1, t_2) = E[X(t_1)X(t_2)] - E[X(t_1)]E[X(t_2)]$$



When  $X(t_2) = 1 \Rightarrow Y > t_2 \geq t_1 \Rightarrow X(t_1) = 1$   
 $\therefore P = 1 - t_2 \quad \& \quad X(t_2) \cdot X(t_1) = 1 \cdot 1 = 1$ .

When  $X(t_1) = 0 \Rightarrow Y \leq t_1 \leq t_2 \Rightarrow X(t_2) = 0$   
 $\therefore P = t_1 \quad \& \quad X(t_2) \cdot X(t_1) = 0 \cdot 0 = 0$

When  $t_1 \leq Y < t_2 \Rightarrow X(t_2) = 0 \quad \& \quad X(t_1) = 1$   
 $\therefore P = \frac{t_2 - t_1}{1 - 0} = t_2 - t_1 \quad \& \quad X(t_2) \cdot X(t_1) = 0 \cdot 1 = 0$

$\therefore$  PMF of  $X(t_1) X(t_2)$  is

$$\begin{aligned} & \text{if } t_1 \leq Y < t_2 \Rightarrow X(t_2) = 0 \quad \& \quad P = t_2 - t_1 \\ & \text{if } Y \geq t_2 \Rightarrow X(t_2) = 1 \quad \& \quad P = 1 - t_2 \\ & \text{if } Y \leq t_1 \Rightarrow X(t_1) = 0 \quad \& \quad P = t_1 \\ & \therefore E[X(t_1) X(t_2)] = 1 \cdot (1 - t_2) + 0 \cdot t_2 \\ & \quad \quad \quad = 1 - t_2 \end{aligned}$$

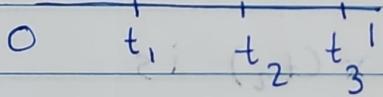
$$\begin{aligned} & \text{if } X(t_1, t_2) = 1 - t_2 - ((1 - t_1)(1 - t_2)) \\ & \quad \quad \quad = 1 - t_2 - [1 - t_2 - t_1 + t_1 t_2] \\ & \quad \quad \quad = t_1 - t_1 t_2 \\ & \therefore C_{X(t_1, t_2)} = t_1 - t_1 t_2 \end{aligned}$$

5. (5 points) Is  $X$  wide sense stationary?  
is  $X$  strict sense stationary?

As  $E[X(t)] = 1-t$  does depend on  $t$

Hence the process is not wide sense stationary, which in turn means it is not strict sense stationary.

6. (5 points) What is the distribution of an increment of  $X$ ? Does  $X$  have stationary increments?



$$\text{When } Y \in [t_2, 1] : X(t_2) - X(t_1) = 1 - 1 = 0, P = 1 - t_2$$

$$\text{When } Y \in [0, t_1] : X(t_2) - X(t_1) = 0 - 0 = 0, P = t_1$$

$$Y \in (t_1, t_2) : X(t_2) - X(t_1) = 0 - 1 = -1, P = t_2 - t_1$$

The distribution of an increment from  $t_1$  to  $t_2$  is

$$X(t_2) - X(t_1) = \begin{cases} 0 & , P = 1 - t_2 + t_1 \\ -1 & , P = t_2 - t_1 \end{cases}$$

Similarly from  $t_2$  to  $t_3$  is

$$X(t_3) - X(t_2) = \begin{cases} 0 & , P = 1 - t_3 + t_2 \\ -1 & , P = t_3 - t_2 \end{cases}$$

As  $P$  changes, increments do not follow same distribution. Hence

$x$  does not have stationary increments.

If time steps are equally spaced  
i.e., uniform increments of  $\tau$  seconds,

$$t_2 - t_1 = t_3 - t_2 \text{ uniformly}$$

then the pr. increments of  $x$  follow are  
stationary and independently distributed

the set  $\{x(t) \}_{t \geq 0}$  forms a st.

$$\begin{aligned} & X \sim N(0, \sigma^2) \\ \text{opposite} & \quad 0 \left\{ \begin{array}{l} \dots \\ \dots \end{array} \right. \\ \text{is} & \quad P \left\{ \begin{array}{l} \dots \\ \dots \end{array} \right. \\ \text{not} & \quad 1 \end{aligned}$$

$$(x, x') \sim x \leftarrow (1, 0) \otimes \mathbb{A}$$

minimum of  $\mathbb{A}$  in  $\mathbb{A}$

Exercise

$$(x \otimes x) \otimes = (x) \otimes$$

$$(x \otimes x') \otimes =$$

$$(x \otimes x \otimes x) \otimes =$$

$$(x \otimes x \otimes x') \otimes =$$

$$(x \otimes x \otimes x \otimes x) \otimes =$$

$$\left[ (x \otimes x) \otimes \right] \otimes =$$

$$0 = 1$$

$$x \otimes x + 1 = (x) \otimes$$

## Problem 2 (30 points)

Consider the stochastic process

$$X(t) = e^{-Yt}, \text{ for } t > 0$$

where  $Y$  is a random variable with uniform distribution on the interval  $(0, 1)$ .

1. (5 points) Calculate the first-order cumulative distribution function  $F(t, x)$  of the process  $\{X(t), t > 0\}$  for  $x \in (0, +\infty)$



CDF of  $Y$

$$F_Y(y) = \begin{cases} 0 & \text{for } y \leq 0 \\ y & 0 < y < 1 \\ 1 & \text{for } y \geq 1 \end{cases}$$

$$\text{As } Y \in (0, 1) \Rightarrow X \in (\bar{e}^{-t}, 1)$$

As  $\bar{e}^t$  is monotonically decreasing

$$\begin{aligned} F_X(x) &= P(X \leq x) \\ &= P(\bar{e}^{-Yt} \leq x) \\ &= P(-Yt \leq \ln x) \\ &= P(Y \geq -\frac{1}{t} \ln x) \\ &= 1 - P(Y < -\frac{1}{t} \ln x) \\ &= 1 - \left[ \frac{-\frac{1}{t} \ln x - 0}{1 - 0} \right] \end{aligned}$$

$$F_X(x) = 1 + \frac{1}{t} \ln x$$

$$F(t, x) = F_x(x) = \begin{cases} 0 & \text{for } x \leq e^{-t} \\ 1 + \frac{1}{t} \ln x & \text{for } e^{-t} < x < 1 \\ 1 & \text{for } x \geq 1 \end{cases}$$

with  $x > 0$ ,  $t > 0$   
 working out for  $(x_1, x_2, t_1, t_2)$  if  $F(x_2, t_2) - F(x_1, t_1) = P(x_1 < X(t_1) < x_2 \text{ and } t_1 < T < t_2)$

2. (5 points) Deduce the first-order density function of the process  $\{X(t), t > 0\}$  on  $(0, +\infty)$

→  $f_x(x) = \text{first order density function} = F_x'(x)$

$$\therefore f_x(x) = \begin{cases} 0 & \text{for } x \leq e^{-t} \\ \frac{1}{xt} & \text{for } e^{-t} < x < 1 \\ 0 & \text{for } x > 1 \end{cases}$$

3. (5 points) Compute  $E[X(t)]$ , for  $t > 0$

→  $E[X(t)] = E[e^{-Yt}]$  PDF of uniform distribution  
 $= \int_0^1 e^{-yt} \cdot p(y) \cdot dy$  bet  $(0, 1)$

$$E[X(t)] = \left[ \frac{-e^{-yt}}{t} \right]_0^1 = -\frac{1}{t} [e^{-t} - e^0] = -\frac{1}{t} (e^{-t} - 1)$$

$$\therefore E[x(t)] = \frac{1 - e^{-t}}{t}$$

4. (5 points) Compute the second-order cumulative function  $F(t_1, t_2, x_1, x_2)$  of the process  $\{x(t), t > 0\}$  for  $(x_1, x_2) \in (0, +\infty)^2$

*(fuerst obwohl es nicht mit subtil schwingt)*

$$f_{x_1}(x_1) = \frac{1}{x_1 t_1} e^{-t_1/x_1}$$

$$f_{x_2}(x_2) = \frac{1}{x_2 t_2} e^{-t_2/x_2}$$

$$F(t_1, t_2, x_1, x_2) = \int_{x_2}^{x_1} \frac{1}{x_1 t_1} \frac{1}{x_2 t_2} dx_1 dx_2$$

$$= \int_{x_2}^{x_1} \frac{1}{x_2 t_2} \frac{1}{t_1} \left[ \ln x_1 \right] dx_2$$

$$= \int_{e^{-t_2}}^{e^{-t_1}} \frac{1}{t_1 t_2} \frac{1}{x_2} \left[ \ln x_1 + t_1 \right] dx_2$$

some about  $X$  coming out at time 2)

$$F(t_1, t_2, x_1, x_2) = \int_{x_2}^{\infty} \frac{1}{x_2 t_2} dx_2 (t_1 + \ln x_1)$$

so we have

$$= \frac{(t_1 + \ln x_1)}{t_1 t_2} [ \ln x_2 ] e^{-t_2}$$

so for  $x_1$

$$F(t_1, t_2, x_1, x_2) = \frac{(t_1 + \ln x_1)(t_2 + \ln x_2)}{t_1 t_2}$$

compute some time for  $x_1$  (answer)

5. (5 points) Compute the process  $\{X(t), t \geq 0\}$  defined autocovariance function

$C_X(t, t+s)$  for  $(s, t) > 0$

→  $C_X(t, t+s) = E[X(t) X(t+s)] - E[X(t)] E[X(t+s)]$

$$\begin{aligned} X(t) \cdot X(t+s) &= e^{-Yt} \cdot e^{-Y(t+s)} \\ &= e^{-Y(2t+s)} \end{aligned}$$

$$\therefore E[e^{-Y(2t+s)}] = \frac{1 - e^{-(2t+s)}}{2t+s} = E[X(t) X(t+s)]$$

$$E[X(t)] = \frac{1 - e^{-t}}{t}, \quad E[X(t+s)] = \frac{1 - e^{-(t+s)}}{t+s}$$

$$\therefore C_X(t, t+s) = 1 - e^{-2t+s} - (1 - e^{-t})(1 - e^{-(t+s)})$$

6. (5 points) Is the process  $X$  wide sense stationary, strict sense stationary?

→ As

$$E[X(t)] = 1 - e^{-t} \text{ depends on } t$$

Hence the process is not wide sense stationary.

Consequently it is not strict sense stationary.

Problem 3 (40 points)

Consider the process  $\{X(t), t \geq 0\}$

defined by

$$X(t) = N(t) - \lambda t,$$

where  $N$  is a Poisson process

with rate  $\lambda > 0$

1. (5 points) Compute, for  $t_1, t_2, n_1, n_2 \geq 0$ ,  
the second-order probability mass of  
 $N, G(t_1, t_2; n_1, n_2)$

→ As  $N(t)$  is a poisson process

&  $N(0) = 0$ ,  $N(t_1) - N(0)$  will follow  
poisson distribution with mean  $\lambda t_1$ ,

$$\therefore P[N(t_1) - N(0)] = P[N(t_1)] = \frac{e^{-\lambda t_1} (\lambda t_1)^{n_1}}{n_1!}$$

~~$\therefore P[N(t_1) - N(0)] = \frac{e^{-\lambda t_1} (\lambda t_1)^{n_1}}{n_1!}$~~

$$\forall n_1 = 0, 1, \dots$$

Similarly

$$P[N(t_2)] = \frac{e^{-\lambda t_2} (\lambda t_2)^{n_2}}{n_2!}, \forall n_2 = 0, 1, \dots$$

∴ Second order probability mass of  $N$ ,

$$G(t_1, t_2, n_1, n_2) = \sum_{n_2=0}^{n_2} \sum_{n_1=0}^{n_1} \frac{e^{-\lambda t_1} (\lambda t_1)^{n_1}}{n_1!} \frac{e^{-\lambda t_2} (\lambda t_2)^{n_2}}{n_2!}$$

~~$$\therefore G(t_1, t_2, n_1, n_2) = \frac{e^{-\lambda(t_1+t_2)} (\lambda(t_1+t_2))^{n_1+n_2}}{n_1! n_2!}$$~~

$$G(t_1, t_2, n_1, n_2) = \sum_{n_2=0}^{n_2} \sum_{n_1=0}^{n_1} \frac{e^{-\lambda(t_1+t_2)}}{\lambda^{\frac{n_1+n_2}{2}}} \cdot \frac{\lambda^{n_1} t_1^{n_1}}{n_1!} \cdot \frac{\lambda^{n_2} t_2^{n_2}}{n_2!}$$

2. (5 points) Give the expectation and the variance of  $X(t)$ .

a)

$$E[X(t)] = E[N(t)] - \lambda t$$

$$\therefore E[X(t)] = E[N(t)] - \lambda t$$

$$= E[N(t) - N(0)] - \lambda t$$

As  $N(t)$  for is a poisson process

$$E[N(t) - N(0)] = \lambda t$$

$$\therefore E[X(t)] = \lambda t - \lambda t = 0$$

$$\therefore E[X(t)] = 0$$

$$b) \quad \text{Var}[X(t)] = \text{Var}[N(t) - \lambda t]$$

$$= \text{Var}[N(t)] \text{ as } \lambda t \text{ is deterministic}$$

$$\text{As } \text{Var}[N(t) - N(0)] = \lambda t$$

$$\therefore \text{Var}[X(t)] = \text{Var}[N(t) - N(0)] = \lambda t$$

$$\therefore \text{Var}[X(t)] = \lambda t$$

3. (5 points) Compute  $E[X(t_1)X(t_2)]$ , for

$$0 < t_1 \leq t_2$$



$$X(t_1)X(t_2) = [N(t_1) - \lambda t_1][N(t_2) - \lambda t_2]$$

$$= N(t_1)N(t_2) - \lambda t_2 N(t_1)$$

$$- \lambda t_1 N(t_2) + \lambda^2 t_1 t_2$$

$$\begin{aligned} \text{As } N(0) &= 0 \\ E[N(t_1)N(t_2)] &= E[(N(t_1) - N(0))(N(t_2) - N(t_1) + N(t_1))] \\ &= E[(N(t_1) - N(0))(N(t_2) - N(t_1))] \\ &\quad + E[(N(t_1) - N(0))(N(t_1))] \end{aligned}$$

As  $N(t_1) - N(0)$  &  $N(t_2) - N(t_1)$  are non overlapping inter increments, by definition they are independent.

$$\therefore E[(N(t_1) - N(0))(N(t_2) - N(t_1))] = E[N(t_1) - N(0)] E[N(t_2) - N(t_1)] \quad (1)$$

$$\text{Also } E[(N(t_1) - N(0))N(t_1)] = E[N^2(t_1)]$$

$$\text{As } \text{Var}[N(t_1)] = E[N^2(t_1)] - (E[N(t_1)])^2$$

$$\text{Simplifying, } \Delta t_1 = E[N^2(t_1)] - (E[N(t_1)])^2$$

$$\therefore E[N^2(t_1)] = \Delta t_1 (1 + \Delta t_1) \quad (2)$$

from 1 & 2 &  $E[N(t_2) - N(t_1)] = \lambda(t_2 - t_1)$  &

$$E[N(t_1) - N(0)] = \Delta t_1$$

$$E[N(t_1)N(t_2)] = E[N(t_1) - N(0)] E[N(t_2) - N(t_1)] + E[N^2(t_1)]$$

$$\therefore E[N(t_1)N(t_2)] = \Delta t_1 \cdot \lambda(t_2 - t_1)$$

$$+ \Delta t_1 (1 + \Delta t_1)$$

$$= \Delta t_1 [\Delta t_2 - \Delta t_1 + 1 + \Delta t_1]$$

$$\therefore E[N(t_1)N(t_2)] = \lambda t_1(1 + \lambda t_2)$$

$$\text{As } E[x(t_1)x(t_2)] = E[N(t_1)N(t_2)]$$

$$= (\lambda t_1)(\lambda t_2) - \lambda t_2 E[N(t_1)] - \lambda t_1 E[N(t_2)] \\ + \lambda^2 t_1 t_2$$

$$\therefore E[x(t_1)x(t_2)] = \lambda t_1(1 + \lambda t_2)$$

$$= \lambda t_1 + \lambda^2 t_1 t_2 - \lambda t_2 \lambda t_1 - \lambda t_1 \lambda t_2 \\ + \lambda^2 t_1 t_2$$

$$= [\lambda t_1 - \lambda t_2](\lambda t_1 - \lambda t_2) + \lambda^2 t_1 t_2 - \lambda^2 t_1 t_2$$

$$= -\lambda^2 t_1 t_2 + \lambda^2 t_1 t_2$$

$$\therefore E[x(t_1)x(t_2)] = \lambda t_1$$

4. (5 points) Is  $x$  wide sense stationary?

is  $x$  strict sense stationary?

As  $E[x(t)] = 0$ , it does not depend on time

$$R_x(t, t+h) = E[x(t)x(t+h)] = \lambda t$$

As  $R_x$  depends on  $t$ ,  $x$  is

not wide sense stationary. Which in turn means  $x$  is not strict sense

Stationary either.

5. (5 points) is  $X$  a martingale? Justify your answer.

→ As  $\mathbb{E}[X(0) - N(0)] = \mathbb{E}[0] - 0 = 0$

$\mathbb{E}[X(t) + X(0)] = \mathbb{E}[0] + \mathbb{E}[X(0)] \cdot \mathbb{P}[X(t) = \infty | X(0) = 0]$

As  $\mathbb{E}[X(t) - N(t)] = \mathbb{E}[(\mathbb{E}[X]) - (\mathbb{E}[N])] = 0$

$\mathbb{E}[\sum x(t) \cdot \mathbb{P}[X(t) = x]]$

Suppose  $t_1 < t_2$

$$X(t_2) - X(t_1) = N(t_2) - N(t_1) - \mathbb{E}(t_2 - t_1)$$

&  $X(t_1) = N(t_1) - \mathbb{E}t_1$

$$\mathbb{E}[X(t_2) | X(t_1)] = \mathbb{E}[N(t_2) - \mathbb{E}t_2 | N(t_1) - \mathbb{E}t_1]$$

$$= \mathbb{E}[N(t_2) - N(t_1) + N(t_1) - \mathbb{E}t_2 | N(t_1) - \mathbb{E}t_1]$$

$$= \mathbb{E}[N(t_2) - N(t_1) | N(t_1) - \mathbb{E}t_1]$$

$$+ \mathbb{E}[N(t_1) - \mathbb{E}t_2 | N(t_1) - \mathbb{E}t_1]$$

We can drop  $\mathbb{E}t_1$  in the condition as it is deterministic.

$$\therefore E[X(t_2) | X(t_1)] = E[N(t_2) - N(t_1) | N(t_1) - N(0)]$$

+  $E[N(t_1) - \lambda t_1 | N(t_1)]$

by def<sup>n</sup> of poisson process  $N(t_2) - N(t_1)$  is independent

$$\text{of } N(t_1) - N(0)$$

$$\therefore E[X(t_2) | X(t_1)] = E[N(t_2) - N(t_1)]$$

$$= N(t_1) - \lambda t_1$$

$$= E[N(t_2)] - E[N(t_1)] + N(t_1) - \lambda t_1$$

$$E[X(t_2) | X(t_1)] = \lambda t_2 - \lambda t_1 + N(t_1) - \lambda t_2$$

$$= N(t_1) - \lambda t_1$$

$$\text{but } X(t_1) = N(t_1) - \lambda t_1$$

$$\therefore E[X(t_2) | X(t_1)] = X(t_1)$$

Hence  $X$  is a martingale.

6. (15 points) Is  $t \rightarrow \underline{x}(t)$  mean ergodic? Justify your answer rigorously.

$$\rightarrow Y = \frac{\underline{x}(t)}{t} = \frac{N(t)}{t} - \lambda$$

$$\langle Y \rangle_0 = \frac{1}{M+1} \sum_{n=0}^M Y$$

$$\begin{aligned} E[\langle Y \rangle_0] &= \frac{1}{M+1} \sum_{n=0}^M E[Y] \\ &= \frac{1}{M+1} \sum_{n=0}^M E\left[\frac{N(t)}{t} - \lambda\right] \end{aligned}$$

$$\begin{aligned} E\left[\frac{N(t)}{t} - \lambda\right] &= \frac{1}{t} E[N(t)] - \lambda \\ &= \frac{\lambda t}{t} - \lambda = \lambda - \lambda = 0 \end{aligned}$$

$$\therefore E[\langle Y \rangle_0] = \frac{1}{M+1} \sum_{n=0}^M 0 = 0$$

but the process  $Y$  has mean

$$E[Y] = E\left[\frac{X(t)}{t}\right] = \frac{1}{t} E[X(t)] = 0$$

$$\therefore E[\langle Y \rangle_0] = E[Y] = 0$$

Hence temporal mean converges to mean of the process in mean squares.

therefore the process is mean ergodic.

$$E - \{(\phi)Y\} = (\phi)X = Y$$

$$E[X] = \sum_{n=0}^N \frac{1}{1+H}$$

$$E[X] = \sum_{n=0}^N \frac{1}{1+H}$$

$$E[\phi(Y)] = \sum_{n=0}^N \frac{1}{1+H}$$

$$E - \{(\phi)Y\} = E - \{(\phi)X\}$$

$$\phi = h \cdot h^{-1} \cdot h = \frac{h}{h}$$

$$E - \{(\phi)Y\} = E - \{(\phi)X\}$$

now consider X moving left and

$$\phi = (\text{shift}) + (\text{mix}) = \text{ID}$$

$$\phi(X) = E - \{(\phi)X\}$$

as we move from left to right small

changes in ID do not change the mean