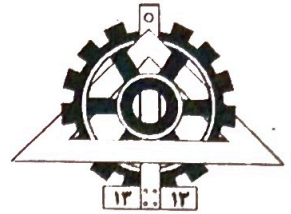


## Homework #5

Stochastic Process – Fall 2024

Instructor: Dr. Ali Olfat

Erfan Panahi (Student Number: 810103084)



**Problem 1.**  $X(t)$ : Markov stochastic process  $t_1 \leq t_2 \leq \dots \leq t_n$

Prove that:  $P_{X_1}(x_1; t_1 | X(t_2) = x_2, \dots, X(t_n) = x_n) = P_{X_1}(x_1; t_1 | X(t_2) = x_2)$

$$P_{X_1}(x_1; t_1 | X(t_2) = x_2, \dots, X(t_n) = x_n) = \frac{P_{X_1 \dots X_n}(X(t_1) = x_1, \dots, X(t_n) = x_n)}{P_{X_2 \dots X_n}(X(t_2) = x_2, \dots, X(t_n) = x_n)} \quad (I)$$

$$\text{Also we know: } P_{X_1 \dots X_n}(X(t_1) = x_1, \dots, X(t_n) = x_n) = P_{X_1}(x_1; t_1) P_{X_2}(x_2; t_2 | X(t_1) = x_1) \dots P_{X_n}(x_n; t_n | X(t_1) = x_1, \dots, X(t_{n-1}) = x_{n-1})$$

$$P_{X_2 \dots X_n}(X(t_2) = x_2, \dots, X(t_n) = x_n) = P_{X_2}(x_2; t_2) P_{X_3}(x_3; t_3 | X(t_2) = x_2) \dots P_{X_n}(x_n; t_n | X(t_2) = x_2, \dots, X(t_{n-1}) = x_{n-1})$$

$$\text{Markov SP Property} \begin{cases} P_{X_1 \dots X_n}(X(t_1) = x_1, \dots, X(t_n) = x_n) = P_{X_1}(x_1; t_1) P_{X_2}(x_2; t_2 | X(t_1) = x_1) \dots P_{X_n}(x_n; t_n | X(t_{n-1}) = x_{n-1}) \\ P_{X_2 \dots X_n}(X(t_2) = x_2, \dots, X(t_n) = x_n) = P_{X_2}(x_2; t_2) P_{X_3}(x_3; t_3 | X(t_2) = x_2) \dots P_{X_n}(x_n; t_n | X(t_{n-1}) = x_{n-1}) \end{cases} \quad (II)$$

According to (I) and (II) we can write:

$$\begin{aligned} P_{X_1}(x_1; t_1 | X(t_2) = x_2, \dots, X(t_n) = x_n) &= \frac{P_{X_1}(x_1; t_1) P_{X_2}(x_2; t_2 | X(t_1) = x_1)}{P_{X_2}(x_2; t_2)} \\ &= \frac{P_{X_1 X_2}(X(t_1) = x_1, X(t_2) = x_2)}{P_{X_2}(X(t_2) = x_2)} = P_{X_1}(x_1; t_1 | X(t_2) = x_2) \end{aligned}$$

✓ Proved ✓

**Problem 2.**  $X(t)$ : Poisson process with uniform density  $\lambda$

↳ jumps at Poisson points  $t_i \geq 0$  ( $i=1,2,\dots$ ) ;  $T_n = t_{i+n} - t_i$  ( $n \geq 1$ )

$$\text{Poisson process} \Rightarrow P(X(t) = k) = e^{-\lambda t} \frac{(\lambda t)^k}{k!}$$

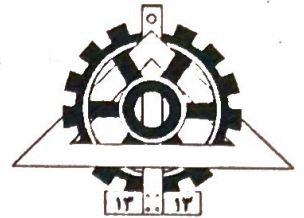
$$F_{T_n}(t) = P(T_n \leq t) = 1 - P(T_n > t) = 1 - P\left\{ \begin{array}{l} \text{zero points} \\ \text{in period} \\ (t_i, t_{i+n}) \end{array} \right\} = 1 - P(X(t) = k-1)$$

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$$\rightarrow F_{T_n}(t) = 1 - P\{X(t) = k-1\} = 1 - e^{-\lambda t} \frac{(\lambda t)^{k-1}}{(k-1)!}$$

$$\begin{aligned} \frac{\partial}{\partial t} F_{T_n}(t) &= \frac{\partial F_{T_n}(t)}{\partial t} = \lambda e^{-\lambda t} \frac{(\lambda t)^{k-1}}{(k-1)!} - (k-1) e^{-\lambda t} \frac{\lambda^{k-1} t^{k-2}}{(k-1)!} \\ &= e^{-\lambda t} \frac{(\lambda t)^{k-1}}{(k-1)!} \left[ \lambda - \frac{k-1}{t} \right] \end{aligned}$$

**Problem 4.**  $W(t)$ : Wiener process ;  $W(t) = \int_0^t X(\alpha) d\alpha \rightarrow \begin{cases} X(t): \text{zero-mean stationary white Gaussian process} \\ R_X(\tau) = N_0 \delta(\tau) \end{cases}$

$$Y(t) = W^2(t)$$

a. PDF of  $Y(t)$ :  $(f_Y(y; t))$

$$f_{W(t)}(w; t) = \frac{1}{\sqrt{2\pi N_0 t}} \exp\left(-\frac{w^2}{2N_0 t}\right) ; \quad Y = W^2 \rightarrow \begin{cases} W_1 = -\sqrt{Y} \\ W_2 = \sqrt{Y} \end{cases} ; \quad f_Y(y; t) = \frac{\sum_i f_{W_i}(w_i; t)}{|J|}$$

$$|J| = |2W| = 2\sqrt{y} \Rightarrow f_Y(y; t) = \frac{f_W(\sqrt{y}; t)}{2\sqrt{y}} + \frac{f_W(-\sqrt{y}; t)}{2\sqrt{y}}$$

$$\rightarrow f_Y(y; t) = \frac{1}{\sqrt{2\pi N_0 t y}} \exp\left(-\frac{y}{2N_0 t}\right) ; \quad y > 0, t > 0$$

b. Is  $Y(t)$  an IIP? IIP Def.  $t_1 \leq t_2 \leq \dots \leq t_n : Y(t_1) \perp Y(t_2) - Y(t_1) \perp \dots \perp Y(t_n) - Y(t_{n-1})$

$$\begin{aligned} f_{Y(t_1)Y(t_2)}(y_1, y_2; t_1, t_2) &= \frac{1}{4\sqrt{y_1 y_2}} \left[ f_W(\sqrt{y_1}, \sqrt{y_2}; t_1, t_2) + f_W(\sqrt{y_1}, -\sqrt{y_2}; t_1, t_2) + \right. \\ &\quad \left. f_W(-\sqrt{y_1}, \sqrt{y_2}; t_1, t_2) + f_W(-\sqrt{y_1}, -\sqrt{y_2}; t_1, t_2) \right] \\ &= \frac{1}{4\sqrt{y_1 y_2}} \left[ f_{W_1}(\sqrt{y_1}) f_{W_2}(\sqrt{y_2} + \sqrt{y_1}) + f_{W_1}(\sqrt{y_1}) f_{W_2}(\sqrt{y_1} - \sqrt{y_2}) + \right. \\ &\quad \left. f_{W_1}(-\sqrt{y_1}) f_{W_2}(\sqrt{y_2} - \sqrt{y_1}) + f_{W_1}(-\sqrt{y_1}) f_{W_2}(-\sqrt{y_1} - \sqrt{y_2}) \right] \\ &= \frac{\exp\left(-\frac{y_1}{2N_0 t_1}\right)}{4\pi N_0 \sqrt{y_1 y_2} t_1(t_2 - t_1)} \left[ \exp\left(-\frac{(\sqrt{y_1} + \sqrt{y_2})^2}{2N_0(t_2 - t_1)}\right) + \exp\left(-\frac{(\sqrt{y_1} - \sqrt{y_2})^2}{2N_0(t_2 - t_1)}\right) \right] \end{aligned}$$

$$\Rightarrow f_Y(y_1, y_2; t_1, t_2) \neq f_Y(y_2 - y_1; t_1, t_2) f_Y(y_1; t_1) \Rightarrow \underline{Y(t) \text{ is NOT IIP!}}$$

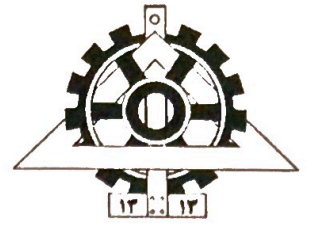
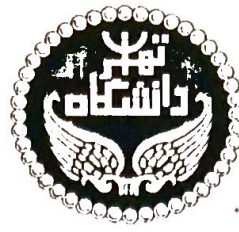


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**Problem 5.**  $X(t) = A \cos(2\pi Ft + \theta)$

$A$ : constant ;  $F \sim U[2, 4]$  ;  $\theta \sim U[0, 2\pi]$  ;  $F \perp \theta$

• Mean of  $X(t)$ :

$$\begin{aligned} E\{X(t)\} &= E\{A \cos(2\pi Ft + \theta)\} = A E\{\cos(2\pi Ft) \cos(\theta) - \sin(2\pi Ft) \sin(\theta)\} \\ &= A E\{\cos(2\pi Ft)\} E\{\cos(\theta)\} - A E\{\sin(2\pi Ft)\} E\{\sin(\theta)\} = 0 \quad \textcircled{I} \end{aligned}$$

• Autocorrelation function of  $X(t)$ :

$$\begin{aligned} R_X(t_1, t_2) &= E\{X(t_1) X^*(t_2)\} = E\{A^2 \cos(2\pi Ft_1 + \theta) \cos(2\pi Ft_2 + \theta)\} \\ &= \frac{A^2}{2} E\{\cos(2\pi F(t_1 + t_2) + 2\theta)\} + \frac{A^2}{2} E\{\cos(2\pi F(t_1 - t_2))\} \\ &= \frac{A^2}{2} E\{\cos(2\pi F(t_1 + t_2))\} E\{\cos(2\theta)\} - \frac{A^2}{2} E\{\cos(2\pi F(t_1 + t_2))\} E\{\sin(2\theta)\} \\ &\quad + \frac{A^2}{2} E\{\cos(2\pi F(t_1 - t_2))\} = \frac{A^2}{2} \int_2^4 \frac{1}{2} \cos(2\pi f(t_1 - t_2)) df \\ &= \frac{A^2}{8\pi(t_1 - t_2)} [\sin(8\pi(t_1 - t_2)) - \sin(4\pi(t_1 - t_2))] \\ \Rightarrow R_X(\tau) &= \frac{A^2}{8\pi\tau} [\sin(8\pi\tau) - \sin(4\pi\tau)] \quad \textcircled{II} \end{aligned}$$

$\textcircled{I}, \textcircled{II} \Rightarrow X(t)$  is a WSS Process

**Problem 6.**  $X(t)$ : zero-mean stationary Gaussian process ( $R_X(\tau)$ )

$Y = A e^{jX(t)}$  ;  $A$  is a Poisson RV :  $A \sim \text{Poisson}(a)$  ;  $A \perp X(t)$

• Mean of  $Y(t)$ :

$$E\{Y(t)\} = E\{A e^{jX(t)}\} = E\{A\} E\{e^{jX(t)}\} = a E\{e^{jX(t)}\} = a \Phi_{X(t)}(1)$$

• Normal RV  $\Rightarrow \Phi_X(\omega) = e^{j\omega m_X} e^{-\frac{1}{2} \sigma_X^2 \omega^2} \rightarrow m_X = 0, \sigma_X^2 = R_X(0)$

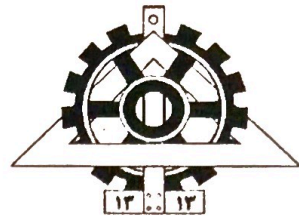
$$\Rightarrow E\{Y(t)\} = a e^{-\frac{R_X(0)}{2}} \quad \textcircled{I}$$

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### • Autocorrelation function of $Y(t)$ :

$$R_Y(t_1, t_2) = E\{Y(t_1) Y(t_2)^*\} = E\{A e^{jX(t_1)} \cdot A e^{-jX(t_2)}\} = E\{A^2\} E\{e^{j(X(t_1) - X(t_2))}\}$$

$$= (a + a^2) E\{e^{jU}\} = (a + a^2) \Phi_U(1)$$

$$U \sim \mathcal{N}(m_U, \sigma_U^2) \quad ; \quad \Phi_U(\omega) = e^{j\omega m_U} \cdot e^{-\frac{1}{2} \sigma_U^2 \omega^2} \rightarrow \Phi_U(1) = e^{j m_U} e^{-\frac{1}{2} \sigma_U^2}$$

$$m_U = E\{U\} = E\{X(t_1) - X(t_2)\} = E\{X(t_1)\} - E\{X(t_2)\} = 0$$

$$\sigma_U^2 = E\{U^2\} = E\{(X(t_1) - X(t_2))^2\} = E\{X^2(t_1) - 2X(t_1)X(t_2) + X^2(t_2)\}$$

$$= \underbrace{E\{X^2(t_1)\}}_{R_X(0)} - 2 \underbrace{E\{X(t_1)X(t_2)\}}_{R_X(t_1 - t_2)} + \underbrace{E\{X^2(t_2)\}}_{R_X(0)} = 2R_X(0) - 2R_X(t_1 - t_2)$$

$$\Rightarrow R_Y(t_1, t_2) = (a + a^2) e^{-(R_X(0) - R_X(t_1 - t_2))}$$

$$\rightarrow R_Y(\tau) = (a + a^2) e^{R_X(\tau) - R_X(0)} \quad \textcircled{II}$$

①, ②  $\Rightarrow Y(t)$  is a WSS process

**Problem 3.**  $W(t) = \int_0^t X(\alpha) d\alpha$  ;  $X(t)$ : WSS Gaussian process ;  $R_X(\tau) = N_0 \delta(\tau)$

■ We want to estimate  $W(2)$  given  $W(1)$  by MMSE criterion.

$$f_{W(t)}(\omega; t) = \frac{1}{\sqrt{2\pi N_0 t}} e^{-\frac{\omega^2}{2N_0 t}} \Rightarrow \begin{aligned} W(1): f_{W(1)}(\omega) &= \frac{1}{\sqrt{2\pi N_0}} \exp\left(-\frac{\omega^2}{2N_0}\right) \\ W(2): f_{W(2)}(\omega) &= \frac{1}{\sqrt{2\pi 2N_0}} \exp\left(-\frac{\omega^2}{4N_0}\right) \end{aligned}$$

$W(1), W(2) \Rightarrow$  Jointly Normal ( $\mu_{W(1)} = \mu_{W(2)} = 0$ ,  $\sigma_{W(1)}^2 = N_0$ ,  $\sigma_{W(2)}^2 = 2N_0$ ,  $\rho_{W(1), W(2)}$ )

$$\rho_{W(1), W(2)} = \frac{\text{cov}(W(1), W(2))}{\sigma_{W(1)} \sigma_{W(2)}} = \frac{E\{W(1)W(2)\}}{\sigma_{W(1)} \sigma_{W(2)}} = \frac{R_W(1, 2)}{1 \times \sqrt{2}} = \frac{\sqrt{2}}{2} N_0$$

according to the property of Jointly Normal dist.  $\Rightarrow E\{W(2) | W(1)\} = \frac{\rho_{W(1), W(2)}}{\sigma_{W(1)}} W(1) = \frac{\frac{\sqrt{2}}{2} N_0}{\sqrt{N_0}} W(1) = \frac{\sqrt{2N_0}}{2} W(1)$

$$\Rightarrow \hat{W}(2)_{\text{MMSE}} = E\{W(2) | W(1)\} = \frac{2}{2} N_0 \cdot \frac{\sqrt{2N_0}}{\sqrt{N_0}} W(1) = W(1)$$

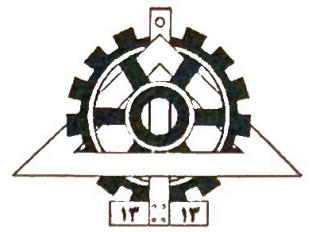
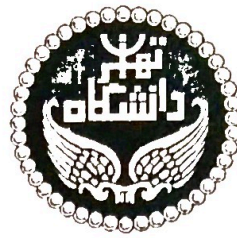


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$$\begin{aligned} \text{MSE} &= E\{(\omega(2) - \hat{\omega}(2))^2\} = E\{(\omega(2) - \omega(1))^2\} = E\{\omega^2(2) - 2\omega(1)\omega(2) + \omega^2(1)\} \\ &= \underbrace{E\{\omega^2(2)\}}_{2N_0 = \sigma_{\omega(2)}^2} - 2\underbrace{E\{\omega(1)\omega(2)\}}_{N_0 = R_{\omega(1,2)}} + \underbrace{E\{\omega^2(1)\}}_{N_0 = \sigma_{\omega(1)}^2} = 2N_0 - 2N_0 + N_0 = N_0 \end{aligned}$$

$$\text{Estimation} \Rightarrow \hat{\omega}(2)_{\text{MMSE}} = \omega(1) \quad ; \quad \text{MSE} = N_0$$

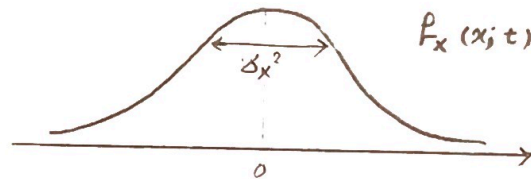
**Problem 8.**  $U(n)$  : iid sequence of Gaussian RVs  $\rightarrow \mathcal{N}(0, 1)$

$X(t) \Rightarrow$  obtained by linearly interpolating between  $U$ 's  $\rightarrow X(t) = U(t)$  for  $t = n \in \mathbb{Z}$

Part a.  $f_x(x; t)$   $\rightarrow X(t) = (1-\alpha)U(t) + \alpha U(t+1)$  ;  $t = n$

$$\rightarrow X(t) = \underbrace{(1-\alpha)U(n)}_{U_1} + \underbrace{\alpha U(n+1)}_{U_2} \quad ; \quad t = n \Rightarrow \sigma_x^2 = (1-\alpha)^2 + \alpha^2 = 2\alpha^2 - 2\alpha + 1$$

$$\rightarrow f_x(x; t) = \underbrace{f_{U_1}(u_1) * f_{U_2}(u_2)}$$



$f_{U_1}(u_1)$  and  $f_{U_2}(u_2)$  both have

a bell shape. So the convolution of them ( $f_x(x; t)$ ) will have a bell shape too!

$$t = n + \alpha \rightarrow \alpha = t - n \rightarrow \sigma_x^2 = 2(n+1-t)^2 + 2t - 2n - 1 = (n-t)^2$$

$$\Rightarrow f_x(x; t, n) = \frac{1}{\sqrt{2\pi(n-t)}} \exp\left(\frac{-x^2}{2(n-t)^2}\right)$$

Part b. Is the  $X(t)$  wide sense stationary?

• Mean :  $E\{X(t)\} = 0$

• Auto-correlation :  $R_x(t_1, t_2) = E\{X(t_1)X^*(t_2)\} = (1-\alpha_1)(1-\alpha_2)E\{U^2(n_1)\} + \alpha_1\alpha_2E\{U^2(n_1+1)\}$

$$\begin{aligned} \left. \begin{array}{l} \alpha_1 = t_1 - n_1 \\ \alpha_2 = t_2 - n_2 \end{array} \right\} \Rightarrow R_x(t_1, t_2) &= (n_1+1-t_1)(n_1+1-t_2) + (t_1-n_1)(t_2-n_1) \\ &= 2n_1^2 - 2n_1t_1 - 2n_1t_2 + 2t_1t_2 - t_1 - t_2 + 2n_1 + 1 \end{aligned}$$

$$\rightarrow R_x(t_1, t_2) \neq R_x(t_1 - t_2) \rightarrow X(t) \text{ is not WSS!}$$