

**EEL5542 – Fall 2001
HOMEWORK #5**

Solution of Problems

- 6.3 a)

...	$n = 0$	$n = 1$	$n = 2$...			
If $\xi = \text{Heads}$	X_n	...	1	-1	1	-1	...
If $\xi = \text{Tails}$	X_m	...	-1	1	-1	1	...
- b) n even $P[X_n = 1] = P[\text{Heads}] = \frac{1}{2}$
 n odd $P[X_m = 1] = P[\text{Tails}] = \frac{1}{2}$
- c) k even

$$P[X_n = 1, X_{n+k} = 1] = P[\text{Heads}] = \frac{1}{2}$$

$$P[X_n = -1, X_{n+k} = -1] = P[\text{Tails}] = \frac{1}{2}$$

$$P[X_n = \pm 1, X_{n+k} = \mp 1] = 0$$

k odd

$$P[X_n = 1, X_{n+k} = -1] = P[\text{Heads}] = \frac{1}{2}$$

$$P[X_n = -1, X_{n+k} = 1] = P[\text{Tails}] = \frac{1}{2}$$

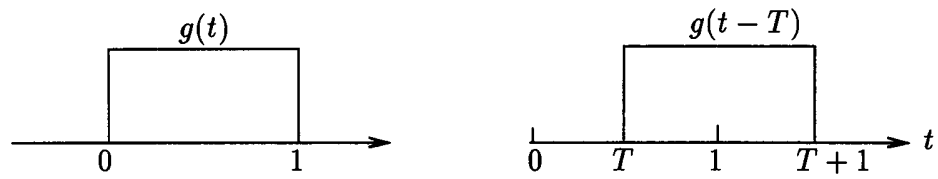
$$P[X_n = \pm 1, X_{n+k} = \pm 1] = 0$$

d) $\mathcal{E}[X_n] = 1\left(\frac{1}{2}\right) + (-1)\left(\frac{1}{2}\right) = 0$

k even $\mathcal{E}[X_n X_{n+k}] = (1)^2 \frac{1}{2} + (-1)^2 \frac{1}{2} = 1$

k odd $\mathcal{E}[X_n X_{n+k}] = (1)(-1)\frac{1}{2} + (-1)(1)\frac{1}{2} = -1$

6.6



$$\text{a) } P[Y(t) = 1] = P[g(t - T) = 1]$$

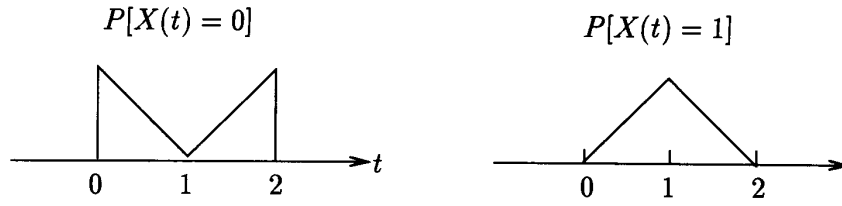
For $0 < t < 1$

$$P[Y(t) = 0] = P[t < T] = 1 - t = 1 - P[Y(t) = 1]$$

For $1 < t < 2$

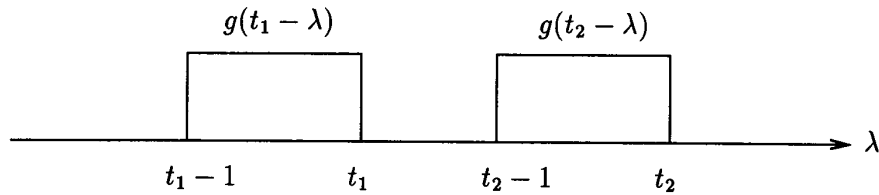
$$\begin{aligned} P[Y(t) = 1] &= P[t < T + 1] = P[T > t - 1] = 1 - (t - 1) \\ &= 2 - t \end{aligned}$$

$$P[Y(t) = 0] = 1 - P[Y(t) = 1] = t - 1$$



b) $E[Y(t)] = 1 \cdot P[Y(t) = 1] = P[Y(t) = 1]$

$$\begin{aligned}
 E[Y(t_1)Y(t_2)] &= \int_0^1 E[g(t_1 - T)g(t_2 - T)|T = \lambda]f_T(\lambda)d\lambda \\
 &= \int_0^1 g(t_1 - \lambda)g(t_2 - \lambda)d\lambda
 \end{aligned}$$



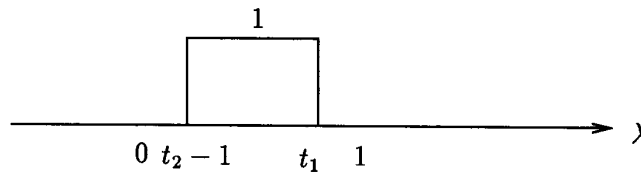
$$g(t_1 - \lambda)g(t_2 - \lambda) = 0 \quad \text{for } t_1 < t_2 - 1$$

$$\Rightarrow R_Y(t_1, t_2) = 0 \quad \text{for } t_2 - t_1 > 1$$

If $t_2 - 1 < t$, then

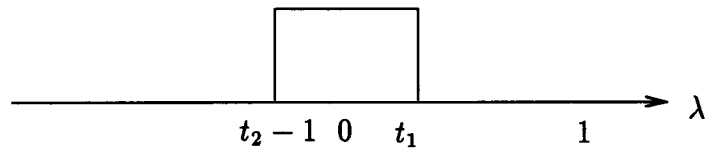
$$g(t_1 - \lambda)g(t_2 - \lambda) = \begin{cases} 1 & t_2 - 1 < \lambda < t_1 \\ 0 & \text{elsewhere} \end{cases}$$

Case 1



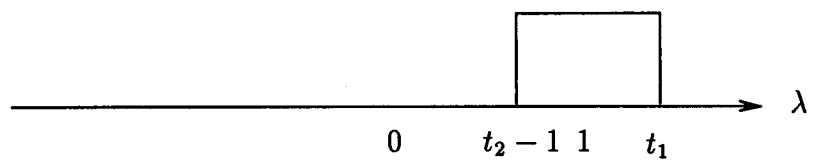
$$R_Y(t_1, t_2) = t_1 - (t_2 - 1) = 1 - (t_2 - t_1) \quad t_1 < 1, \quad 0 < t_2 - 1, \quad t_2 - t_1 < 1$$

Case 2



$$R_Y(t_1, t_2) = t_1 \quad t_1 < 1, \quad t_2 < 1, \quad t_2 - t_1 < 1$$

Case 3



$$R_Y(t_1, t_2) = 2 - t_2 \quad t_1 > 1, \quad t_2 < 2$$

$$6.10 \text{ a) } P[H(t) = 1] = P[X(t) \geq 0] = P[\xi \cos 2\pi t \geq 0] = \frac{1}{2} = P[H(t) = -1]$$

$$\begin{aligned} \mathcal{E}[H(t)] &= 1 \cdot P[H(t) = 1] + (-1)P[H(t) = -1] = 0 \\ C_H(t, t + \tau) &= \mathcal{E}[H(t)H(t + \tau)] \\ &= 1 \cdot P[\underbrace{H(t)H(t + \tau) = 1}_{\substack{H(t) \ \& \ H(t + \tau) \\ \text{same sign}}}] + (-1)P[\underbrace{H(t)H(t + \tau) = -1}_{\substack{H(t) \ \& \ H(t + \tau) \\ \text{opposite sign}}}] \end{aligned}$$

$$H(t)H(t + \tau) = 1 \Leftrightarrow \cos 2\pi t \text{ and } \cos 2\pi(t + \tau) \text{ have same sign}$$

$$H(t)H(t + \tau) = -1 \Leftrightarrow \cos 2\pi t \text{ and } \cos 2\pi(t + \tau) \text{ have different sign}$$

$$\therefore C_H(t, t + \tau) = \begin{cases} 1 & \text{for } t, \tau \text{ such that } \cos 2\pi t \cos 2\pi(t + \tau) = 1 \\ -1 & \text{for } t, \tau \text{ such that } \cos 2\pi t \cos 2\pi(t + \tau) = -1 \end{cases}$$

$$\text{b) } P[H(t) = 1] = P[X(t) \geq 0] = P[\cos(\omega t + \Theta) \geq 0] = \frac{1}{2} = P[H(t) = -1]$$

$$\begin{aligned} \mathcal{E}[H(t)] &= 1 \left(\frac{1}{2}\right) + (-1)\frac{1}{2} = 0 \\ \mathcal{E}[H(t)H(t + \tau)] &= 1 \cdot \underbrace{P[X(t)X(t + \tau) > 0]}_{1 - P[X(t)X(t + \tau) < 0]} + (-1)P[X(t)X(t + \tau) < 0] \\ &= 1 - 2P[X(t)X(t + \tau) < 0] \end{aligned}$$

$$\begin{aligned} P[X(t)X(t + \tau) < 0] &= P[\cos(\omega t + \Theta) \cos(\omega(t + \tau) + \Theta) < 0] \\ &= \left[\frac{1}{2} \cos \omega \tau + \frac{1}{2} \cos(2\omega t + \omega \tau + 2\Theta) < 0 \right] \\ &= P[\cos(2\omega t + \omega \tau + 2\Theta) < \cos \omega \tau] \\ &= 1 - \frac{\text{shaded region in figure}}{2\pi} \end{aligned}$$

$$\text{c) } P[H(t) = 1] = P[X(t) \geq 0] = 1 - F_{X(t)}(0^-) = 1 - P[H(t) = -1]$$

$$\begin{aligned}
\mathcal{E}[H(t)] &= 1 \cdot P[H(t) = 1] + (-1)P[H(t) = -1] \\
&= 1 - F_{X(t)}(0^-) - F_{X(t)}(0^-) \\
&= 1 - 2F_{X(t)}(0^-)
\end{aligned}$$

d) $\mathcal{E}[H(t)X(t)] = \mathcal{E}[|X(t)|]$

since

$$H(t)X(t) = \begin{cases} +X(t) & X(t) \geq 0 \\ -X(t) & X(t) < 0. \end{cases}$$

6.17 a) $\mathcal{E}[Y(t)] = \mathcal{E}[X(t) - aX(t+d)] = m_X(t) - am_X(t+d)$

$$\begin{aligned}
\mathcal{E}[Y(t_1)Y(t_2)] &= \mathcal{E}[(X(t_1) - aX(t_1+d))(X(t_2) - aX(t_2+d))] \\
&= \mathcal{E}[X(t_1)X(t_2)] - a\mathcal{E}[X(t_1+d)X(t_2)] \\
&\quad - a\mathcal{E}[X(t_1)X(t_2+d)] + a^2\mathcal{E}[X(t_1+d)X(t_2+d)] \\
C_Y(t_1, t_2) &= \mathcal{E}[Y(t_1)Y(t_2)] - m_Y(t_1)m_Y(t_2) \\
&= C_X(t_1, t_2) - aC_X(t_1, t_2+d) - aC_X(t_1+d, t_2) + a^2C_X(t_1+d, t_2+d)
\end{aligned}$$

b) Since $X(t)$ and $X(t+d)$ are jointly Gaussian RV's, $Y(t)$ is also Gaussian with mean $m_Y(t)$ and variance:

$$\sigma_{Y(t)}^2 = C_Y(t, t) = \sigma_X^2(t) - 2aC_X(t, t+d) + a^2\sigma_X^2(t+d)$$

$$f_{Y(t)}(y) = \frac{\exp\left\{-\frac{(y-m_X(t)+am_X(t+d))^2}{2\sigma_Y^2(t)}\right\}}{\sqrt{2\pi\sigma_Y(t)}}$$

c) $Y(t)$ and $Y(t+s)$ are jointly Gaussian RV's defined by the linear transformation

$$\begin{aligned}
Y(t) &= X(t) - aX(t+d) \\
Y(t+s) &= X(t+s) - aX(t+s+d)
\end{aligned}$$

$$\begin{aligned}
&f_{Y(t)Y(t+s)}(y_1, y_2) \\
&= \frac{\exp\left\{-\frac{\left(\frac{y_1-m_Y(t)}{\sigma_Y(t)}\right)^2 - \frac{2C_Y(t,t+s)}{\sigma_Y^2(t)\sigma_Y^2(t+s)}(y_1-m_Y(t))(y_2-m_Y(t+s)) + \left(\frac{y_2-m_Y(t+s)}{\sigma_Y(t+s)}\right)^2}{2(1-C_Y^2(t,t+s)/\sigma_Y^2(t)\sigma_Y^2(t+s))}\right\}}{2\pi\sigma_Y(t)\sigma_Y(t+s)\sqrt{1-C_Y^2(t,t+s)/\sigma_Y^2(t)\sigma_Y^2(t+s)}}
\end{aligned}$$

where $C_Y(t, t+s)$ is given in part a) and $\sigma_Y^2(t)$ is given in part b).

d) Any n time samples of $Y(t)$ are defined by a linear transformation of time samples of $X(t)$.

$$\mathbf{6.19} \quad \mathcal{E}[Y(t)] = \mathcal{E}[X^2(t)] = C_X(t, t)$$

$$\mathcal{E}[Y(t_1)Y(t_2)] = \mathcal{E}[X^2(t_1)X^2(t_2)]$$

To proceed further we need the result from Problem 4.69

$$\begin{aligned} \mathcal{E}[X^2(t_1)X^2(t_2)] &= \mathcal{E}[X^2(t_1)]\mathcal{E}[X^2(t_2)] + 2\mathcal{E}[X(t_1)X(t_2)]^2 \\ \Rightarrow \mathcal{E}[Y(t_1)Y(t_2)] &= C_X(t_1, t_1)C_X(t_2, t_2) + 2C_X^2(t_1, t_2) \end{aligned}$$

$$\begin{aligned} \therefore C_Y(t_1, t_2) &= \mathcal{E}[Y(t_1)Y(t_2)] - m_Y(t_1)m_Y(t_2) \\ &= 2C_X^2(t_1, t_2) \end{aligned}$$