

EE 503
Midterm #1
(Duration: 110 minutes)

(Solutions)

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1. (10 pts.) In the fair coin experiment, we define the process $x(t)$ as follows:

$$x(t) = \begin{cases} \sin(\pi t), & \text{if heads show} \\ 2t, & \text{if tails show} \end{cases}$$

- 3 (a) Find $E\{x(t)\}$.
3 (b) Find the first order distribution of $x(t)$ for $t = 1$.
4 (c) Find the joint distribution of $x(t)$ for $t = 1/2$ and $t = 1$.
2. (15 pts.) A real valued random process $y[n]$ is defined as follows

$$y[n] = x[n] - x[n-1]$$

Here $x[n]$ is i.i.d. and takes the values of $\{1, 0\}$ with probability p and $(1-p)$, respectively.

- 3 (a) Find the first order distribution of $y[n]$.
3 (b) Is the process $y[n]$ stationary?
3 (c) Are the processes $y[n]$ and $x[n]$ jointly stationary?
6 (d) What is the correlation between the samples, i.e. $E\{y[n]y[n-k]\}$?

3. (15 pts.) A real valued random process $x[n]$ has the power spectrum density of $S_x(e^{j\omega}) = 3 + 2 \cos(\omega)$. This process is filtered with $H(z) = 1 - \frac{1}{2}z^{-2}$.

- 3 (a) Is the output process stationary in any sense?
3 (b) What is the output autocorrelation?
6 (c) Is the process $z[n] = (x[n])^2$ stationary in any sense?
(d) Bonus (5 pts.): Find the first order density of $z[n] = (x[n])^2$ when $x[n]$ is a Gaussian process.

4. (20 pts.) The random variables x_1 and x_2 are defined as follows:

$$\begin{aligned} x_1 &= u + v + w \\ x_2 &= u - v + w \end{aligned}$$

Here u and v are zero mean Gaussian random variables with zero mean and variance of 1 and 2 respectively. The correlation coefficient of u and v is $1/\sqrt{2}$. The random variable w is Gaussian distributed with zero mean and variance of 2. The random variable w is independent from u and v .

- 5 (a) Write the joint pdf of x_1 and x_2 .
7 (b) Find a linear transformation \mathbf{A} , i.e.

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \mathbf{A} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix},$$

such that y_1 and y_2 are uncorrelated.

- 8 (c) Find a linear transformation of x_1 and x_2 , i.e. $z = \alpha_1 x_1 + \alpha_2 x_2$, such that z is independent from u .

5. (25 pts.) A real valued random process $x[n]$ is defined as follows:

$$x[n] = \rho x[n-1] + w[n], \quad n \geq 0$$

where $|\rho| < 1$ and $w[n]$ is white noise with zero mean and variance of σ_w^2 . The initial condition at $n = -1$ is given as zero, i.e. $x[-1] = 0$.

- 2 (a) Calculate the $\mu_x[n] = E\{x[n]\}$.
- 6 (b) Calculate $\text{var}\{x[0]\}$, $\text{var}\{x[1]\}$ explicitly and generalize to $\text{var}\{x[n]\}$. Is the process $x[n]$ stationary for $n \geq 0$?
- 8 (c) Calculate the autocorrelation of $x[n]$, i.e. $R_x[k_1, k_2] = E\{x[k_1]x[k_2]\}$ and check the consistency of the result with part (b). Is the process stationary?
- 9 (d) Assume that the initial condition of $x[-1]$ is also a random variable. The sample $x[-1]$ has zero mean and variance $\frac{\sigma_x^2}{1-\rho^2}$. Repeat part (b) for the random initialization. Comment on your results.
6. (15 pts.) The process $x(t)$ is zero mean WSS process.
Show that, if

$$s = \frac{1}{N} \sum_{k=1}^N x(kT)$$

then

$$E\{s^2\} = \frac{1}{2\pi N^2} \int_{-\infty}^{\infty} S_x(\omega) \frac{\sin^2(N\omega T/2)}{\sin^2(\omega T/2)} d\omega$$

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EE 503 MT#1

Solutions

Lastname, Name :

Date :

Question	Page	Grade
1		
2		
3		
4		
5		
6		
Total		

Instructions:

1. Number the pages and write your last name on each page.
2. On the first page, write the page number of each answer.
3. Start a new question on a new page.

Start writing below this line, please!

① a) $E\{x(t)\} = \frac{1}{2} \sin \pi t + \frac{1}{2} 2t$

b) $x(1) = \begin{cases} 0 & \text{heads} \\ 2 & \text{tails} \end{cases} \rightarrow f_{x(1)}(x_1) = \frac{1}{2} \delta(x) + \frac{1}{2} \delta(x-2)$

c)

$x(1)$	$x(1/2)$	1
0	H	H
2	T	T

$f_{x(1), x(1/2)}(x_1, x_{1/2}) = \delta(x_{1/2} - 1) \left[\frac{1}{2} \delta(x) + \frac{1}{2} \delta(x-2) \right]$

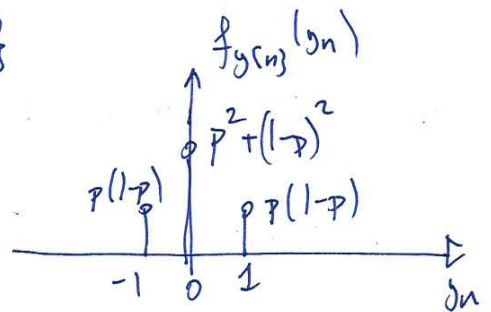
② $x[n] = \{1, 0\}$

a) $y[n] = x[n] - x[n-1] \rightarrow j[n] = \{1, 0, -1\}$

$P(y[n]=0) = p^2 + (1-p)^2$

$P(y[n]=1) = p(1-p)$

$P(y[n]=-1) = (1-p)p$



b) Yes, 1st order pdf stays the same for all shifts.

2nd order pdf $\rightarrow P(y[n], y[n-\Delta]] \rightarrow \Delta > 1$ $y[n], y[n-\Delta]$ are independent

3rd order pdf \rightarrow not a func. time $\rightarrow \Delta = 1$ not independent but not a func. of time.

$$c) f_{x[n], y[n]}(x_n, y_n) = f_{y[n]|x[n]}(y_n|x_n) f_{x_n}(x_n)$$

$$= \frac{1}{2} [\delta(x_n - 1) + \delta(x_n)] \cdot [p \delta(y_n - 1) + (1-p) \delta(y_n)]$$

clearly joint pdf not a func. of "n".

Stationary.

$$d) E\{y[n]y[n-k]\} = E\{(x[n] - x[n-1])(x[n-k] - x[n-k-1])\}$$

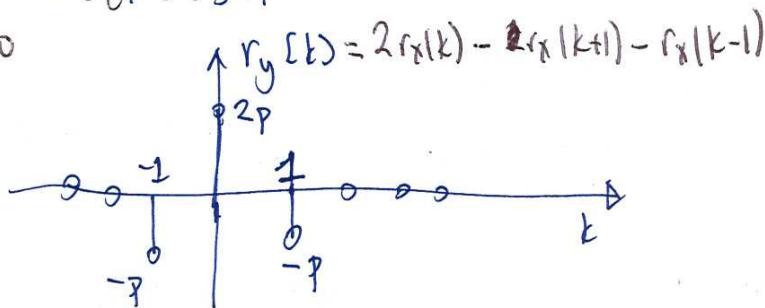
$$= r_x(k) - r_x(k+1) - r_x(k-1) + r_x(k)$$

$$r_x[k] = \delta[k]$$

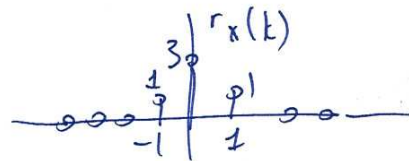
$$E\{x[n]^2\} = p$$

$$E\{x[n]x[n-k]\} = p^2 \quad k \neq 0$$

$$r_x(k) = \begin{cases} p & k=0 \\ p^2 & k \neq 0 \end{cases}$$



$$3) S_x(e^{j\omega}) = 3 + e^{j\omega} + e^{-j\omega} \rightarrow r_x(k) = 3\delta[k] + \delta[k-1] + \delta[k+1].$$



a) Yes, WSS process filtered with LTI system \rightarrow output: WSS.

$$b) r_y(k) = h(k) * h(-k) * r_x(k)$$

$$h(k) * h(-k) \rightarrow \left(\frac{1}{2} - \frac{1}{2}z^{-2}\right) \left(\frac{1}{2} - \frac{1}{2}z^2\right) = \frac{5}{4} - \frac{1}{2}z^{-2} - \frac{1}{2}z^2$$

$$h(k) * h(-k) * r_x(k) \rightarrow \left(\frac{5}{4} - \frac{1}{2}z^{-2} - \frac{1}{2}z^2\right) (3 + z^{-1} + z) = \left\{ \begin{array}{l} 15/4 + z^{-1} \left(\frac{5}{4} - \frac{1}{2} \right) + z^{-2} \left(-\frac{3}{2} \right) \\ z^1 + z^2 \left(\frac{5}{4} - \frac{1}{2} \right) + z^3 \left(-\frac{1}{2} \right) \end{array} \right.$$

$$r_y(k) = \begin{cases} 15/4 & k=0 \\ 3/4 & |k|=1 \\ -3/2 & |k|=2 \\ -1/2 & |k|=3 \\ 0 & \text{other} \end{cases}$$

c) We can not say that $z[n] = (y[n])^2$ is WSS ~~from~~; since

$$E\{z[n]z[n-k]\} = E\{y^2[n]y^2[n-k]\} \text{ depends}$$

on 4th order moments.

But if $x[n]$ is SSS; $z[n]$ is a memory less mapping

therefore $z[n]$ has density only depending on $x[n]$ itself;

therefore $z[n]$ has joint pdf not a function of " n ", i.e.

$z[n]$ is SSS if $x[n]$ is SSS.

d) $x[n] \sim N(\mu, \sigma_x^2 + \mu_x^2)$ $\leftarrow r_x(k) \downarrow k=0$

$$\text{Then } z[n] = (x[n])^2 \rightarrow P\{z[n] < z_n\} = P\{|x[n]| \leq \sqrt{z_n}\}$$

$$= F_x(\sqrt{z_n}) - F_x(-\sqrt{z_n})$$

$$f_{z(n)}(z_n) = \frac{1}{2\sqrt{z_n}} f_x(\sqrt{z_n}) + \frac{1}{2\sqrt{z_n}} f_x(-\sqrt{z_n})$$

$$(4) \underbrace{\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}}_x = \underbrace{\begin{bmatrix} 1 & 1 & 1 \\ 1 & -1 & 1 \end{bmatrix}}_{= T} \underbrace{\begin{bmatrix} v \\ v \\ v \end{bmatrix}}_v \rightarrow R_v = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix}$$

$$E\{v v^T\} = \int_{-\infty}^{\infty} v v^T p(v) dv = \frac{1}{\sqrt{2}} \cdot 1 \sqrt{2} = 1$$

$$a) \underbrace{R_x}_{=} = \underbrace{T}_{=} \underbrace{R_v}_{=} \underbrace{T^T}_{=} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & -1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 \\ 1 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & -1 \\ 1 & 1 \end{bmatrix} = \left[\begin{array}{c|c} 7 & 1 \\ \hline 1 & 3 \end{array} \right]$$

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \sim N \left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \underbrace{\begin{bmatrix} 7 & 1 \\ 1 & 3 \end{bmatrix}}_{C_x} \right) = \frac{1}{2\pi |C_x|^{1/2}} e^{-\frac{1}{2} [x_1 \ x_2] C_x^{-1} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}}$$

$$b) \underbrace{R_x}_{=} = \left[\begin{array}{c|c} 7 & 1 \\ \hline 1 & 3 \end{array} \right]$$

let's use LU decomp.

$$\begin{bmatrix} 1 & 0 \\ 0 & \frac{20}{7} \end{bmatrix} \underbrace{R_x}_{=} = \begin{bmatrix} 7 & 1 \\ 0 & \frac{20}{7} \end{bmatrix}$$

$$\underbrace{R_x}_{=} = \begin{bmatrix} 1 & 0 \\ 1/7 & 1 \end{bmatrix} \begin{bmatrix} 7 & 0 \\ 0 & 20/7 \end{bmatrix} \begin{bmatrix} 1 & 1/7 \\ 0 & 1 \end{bmatrix}$$

$$\text{Then let } \underbrace{A}_{=} = \begin{bmatrix} 1 & 0 \\ 1/7 & 1 \end{bmatrix}^{-1} = \begin{bmatrix} 1 & 0 \\ -1/7 & 1 \end{bmatrix}$$

$$\text{Then } \underbrace{A}_{=} \underbrace{R_x}_{=} \underbrace{A^T}_{=} = \begin{bmatrix} 7 & 0 \\ 0 & 20/7 \end{bmatrix}$$

$$c) z = [\alpha_1 \quad \alpha_2] \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$E\{z \neq 0\} = 0 \rightarrow E\left\{ [\alpha_1 \quad \alpha_2] \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} u \right\} = 0$$

$$\Rightarrow [\alpha_1 \quad \alpha_2] E\left\{ \begin{bmatrix} 1 & 1 & 1 \\ 1 & -1 & 1 \end{bmatrix} \begin{bmatrix} u \\ v \\ w \end{bmatrix} u \right\} = 0$$

$$[\alpha_1 \quad \alpha_2] \begin{bmatrix} 1 & 1 & 1 \\ 1 & -1 & 1 \end{bmatrix} E\left\{ \begin{bmatrix} u^2 \\ uv \\ vw \end{bmatrix} \right\} = 0$$

$$[\alpha_1 \quad \alpha_2] \begin{bmatrix} 1 & 1 & 1 \\ 1 & -1 & 1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} = 0$$

$$[\alpha_1 \quad \alpha_2] \begin{bmatrix} 2 \\ 0 \end{bmatrix} = 0$$

$2\alpha_1 = 0$ for independence.

Then $z = \alpha_2 x_2$ is independent of u for $\forall \alpha_2$.

$$5) x[n] = \rho x[n-1] + w[n], n \geq 0, x[-1] = 0$$

$$a) \mu_x[n] \rightarrow \mu_x[n] = \rho \mu_x[n-1] + 0, n \geq 0 \rightarrow \mu_x[n] = 0$$

$$\mu_x[-1] = 0 \quad n \geq -1$$

b)

$$\begin{aligned}
 b) \quad x[0] &= w[0] \\
 x[1] &= w[1] + \rho w[0] \\
 x[2] &= w[2] + \rho w[1] + \rho^2 w[0].
 \end{aligned}$$

$$\text{Var} \{x[0]\} = \sigma_w^2$$

$$\text{Var} \{x[1]\} = \sigma_w^2 (1 + \rho^2)$$

$$\text{Var} \{x[2]\} = \sigma_w^2 (1 + \rho^2 + \rho^4)$$

⋮

$$\text{Var} \{x[n]\} = \sigma_w^2 (1 + \rho^2 + \rho^4 + \dots + \rho^{2n})$$

The process is not stationary in any sense, since

$\text{Var} \{x[n]\}$ changes by n .

$$c) \quad x[n] = \underbrace{\rho^{n+1} x[-1]}_{\substack{\text{Zero-input} \\ \text{response} \\ \text{(due resp. to} \\ \text{initial} \\ \text{cond.)}}} + \underbrace{\sum_{k=0}^n \rho^k w[n-k]}_{\substack{\text{Zero-state resp.} \\ \text{(due resp. to} \\ \text{input)}}} \quad n \geq 0$$

~~$x[k_1] x[k_2]$~~

$$x[k_1] x[k_2] = \left(\rho^{k_1+1} x[-1] + \sum_{\ell_1=0}^{k_1} \rho^{\ell_1} w[k_1-\ell_1] \right) \left(\rho^{k_2+1} x[-1] + \sum_{\ell_2=0}^{k_2} \rho^{\ell_2} w[k_2-\ell_2] \right)$$

$$E \left\{ \cancel{x[k_1]} x[k_2] \right\} = \rho^{k_1+k_2+2} E \{ (x[-1])^2 \} + \sum_{\ell_1=0}^{k_1} \sum_{\ell_2=0}^{k_2} \rho^{\ell_1+\ell_2} \sigma_w^2 \delta[k_1-k_2+\ell_2-\ell_1]$$

$$R_x[k_1, k_2] = \sigma_w^2 \sum_{l_1=0}^{k_1} \sum_{\substack{l_2=0 \\ (l_1-l_2=k_1-k_2)}}^{k_2} \rho^{l_1+l_2}$$

Assume $k_1 > k_2$
without any loss
generality.

$$= \sigma_w^2 \sum_{l_1=k_1-k_2}^{k_1} \rho^{l_1+l_2} \quad \downarrow \quad l_2 = l_1 - (k_1 - k_2)$$

$$= \sigma_w^2 \sum_{l_1=k_1-k_2}^{k_1} \rho^{2l_1 - (k_1 - k_2)}$$

$$= \sigma_w^2 \rho^{-(k_1-k_2)} \sum_{l_1=k_1-k_2}^{k_1} \rho^{2l_1}$$

$$= \sigma_w^2 \sum_{l_1=0}^{k_2} \rho^{2l_1 + (k_1 - k_2)}$$

$$R_x[k_1, k_2] = \sigma_w^2 \rho^{k_1-k_2} \cdot \frac{1 - \rho^{2(k_2+1)}}{1 - \rho^2} \quad \text{for } k_1 > k_2.$$

Remember. $R_x[k_1, k_2] = R_x[k_2, k_1]$

Note Take $k_1 - k_2 = \Delta$ and make $k_2 \rightarrow \infty$

$$R_x[k_1, k_2] = \sigma_w^2 \frac{\rho^{\Delta}}{1 - \rho^2}$$

Stationary
depends only
on Δ .

This should be a familiar result from auto-corr. of
single pole system.

$$d) \quad x(0) = \rho x(-1) + w(0)$$

$$x(1) = \rho^2 x(-1) + w(1) + \rho w(0)$$

$$x(2) = \rho^3 x(-1) + w(2) + \rho w(1) + \rho^2 w(0)$$

$$\text{Var} \{ x(0) \} = \rho^2 \frac{\sigma_w^2}{1-\rho^2} + \sigma_w^2 = \sigma_w^2 \cdot \frac{1}{1-\rho^2}$$

$$\text{Var} \{ x(1) \} = \rho^4 \frac{\sigma_w^2}{1-\rho^2} + \sigma_w^2 (1 + \rho^2) = \sigma_w^2 \cdot \frac{1}{1-\rho^2}$$

$$\text{Var} \{ x(2) \} = \rho^6 \frac{\sigma_w^2}{1-\rho^2} + \sigma_w^2 \underbrace{(1 + \rho^2 + \rho^4)}_{\frac{1-\rho^6}{1-\rho^2}} = \sigma_w^2 \frac{1}{1-\rho^2}$$

$$\text{Var} \{ x(n) \} = \rho^{2(n+1)} \frac{\sigma_w^2}{1-\rho^2} + \sigma_w^2 \underbrace{(1 + \rho^2 + \rho^4 + \dots + \rho^{2n})}_{\frac{1-\rho^{2n+2}}{1-\rho^2}} = \sigma_w^2 \frac{1}{1-\rho^2}$$

The process with random initialization is initialized with the value that it should reach at "steady-state."

So for all $n \geq 0$, the process is stationary (WSS)

with the described random initialization.

$$b) \quad s = \frac{1}{N} \sum_{k=1}^N x(kT)$$

$$E\{s^2\} = \frac{1}{N^2} \sum_{k_1=1}^N \sum_{k_2=1}^N E\{x(k_1T)x(k_2T)\}$$

$$= \frac{1}{N^2} \sum_{k_1} \sum_{k_2} r_x((k_1 - k_2)T)$$

$$= \frac{1}{N^2} \sum_{k_1} \sum_{k_2} \frac{1}{2\pi} \int_{-\infty}^{\infty} S_x(j\omega) e^{j\omega(k_1 - k_2)T} d\omega$$

$$= \frac{1}{2\pi N^2} \int_{-\infty}^{\infty} S_x(e^{j\omega}) \left(\sum_{k_1=1}^N e^{j\omega k_1} \right) \left(\sum_{k_2=1}^N e^{-j\omega k_2} \right) d\omega$$

$$= \frac{1}{2\pi N^2} \int_{-\infty}^{\infty} S_x(e^{j\omega}) \left(\frac{1 - e^{j\omega TN}}{1 - e^{j\omega T}} \cdot e^{j\omega T} \right) \left(\frac{1 - e^{-j\omega TN}}{1 - e^{-j\omega T}} \cdot e^{-j\omega T} \right) d\omega$$

$$= \frac{1}{2\pi N^2} \int_{-\infty}^{\infty} S_x(e^{j\omega}) \left| \frac{1 - e^{j\omega TN}}{1 - e^{j\omega T}} \right|^2 d\omega$$

$$= \frac{1}{2\pi N^2} \int_{-\infty}^{\infty} S_x(e^{j\omega}) \left(\frac{\sin(N\omega T/2)}{\sin(\omega T/2)} \right)^2 d\omega$$

↖

Dirichlet function!