

EE 3025 S2005 Final Exam Review Questions

1. Airline A and Airline B are the only two airlines serving the city of Metropolis. An Airline A outgoing flight has a late departure with probability 0.25, whereas an Airline B outgoing flight has a late departure with probability 0.40. On each of Lois Lane's trips out of Metropolis, she flips her lucky coin; if the coin comes up heads, she takes Airline A, and if the coin comes up tails, she takes Airline B.

- (a) After traveling for many years, Lois concludes that she has a late departure from Metropolis with probability $1/3$. Is her lucky coin a fair coin? What is the probability that her lucky coin will come up heads?

Solution. The array of "forward conditional probabilities" is

$$\begin{array}{c} \textit{on time} \quad \textit{late} \\ A \begin{pmatrix} .75 & .25 \\ .6 & .4 \end{pmatrix} \\ B \end{array}$$

$$1/3 = P(\textit{late}) = [P(A) \ P(B)] \cdot [.25 \ 4] = .25P(H) + .4(1 - P(H)).$$

Solving, we get

$$P(H) = 4/9.$$

The coin is not fair.

- (b) Given that Lois's flight leaves Metropolis on time, what is the probability she flew on Airline B?

Solution.

$$P(B|\textit{on time}) = \frac{P(B \cap \{\textit{on time}\})}{P(\textit{on time})} = \frac{P(B)P(\textit{on time}|B)}{2/3} = \frac{(5/9)(0.6)}{2/3} = 1/2.$$

2. Use the table on page 123 of your textbook to answer the following questions concerning Gaussian RV's.

- (a) A Gaussian RV X satisfies

$$P(X \leq 5.65) = 0.7939$$

$$P(X \leq 7.4) = 0.9357$$

Compute the mean μ and the standard deviation σ of X .

Solution. Let Z be the standard Gaussian RV

$$Z = \frac{X - \mu}{\sigma}.$$

From the table, since

$$P(Z \leq \frac{5.65 - \mu}{\sigma}) = 0.7939,$$

we must have

$$\frac{5.65 - \mu}{\sigma} = 0.82 \tag{1}$$

Also, since

$$P\left(Z \leq \frac{7.4 - \mu}{\sigma}\right) = 0.9357,$$

we must have

$$\frac{7.4 - \mu}{\sigma} = 1.52 \tag{2}$$

Solving equations (1)-(2) simultaneously, you'd be able to compute the values of μ and σ .

- (b) A random sample of size 100 is taken from a Gaussian distribution with mean 0 and standard deviation 6. Let \bar{X} be the sample mean of this random sample. Compute $P(-0.5 \leq \bar{X} \leq 0.5)$.

Solution. \bar{X} is Gaussian with mean 0 and standard deviation $6/\sqrt{100} = 0.6$. Therefore,

$$Z = \frac{\bar{X}}{.6}$$

is standard Gaussian.

$$P(-0.5 \leq \bar{X} \leq 0.5) = P(-5/6 \leq Z \leq 5/6) = 2\Phi(5/6) - 1 = 0.596.$$

3. 6 coins are tossed. Coins 1,2,3 are fair, but coins 4,5,6 are unfair, each satisfying $P(H) = 2/3$. Let X be the number of heads you obtain among the three fair coins. Let Y be the number of heads you obtain among the three unfair coins. Using the binomial RV's X, Y , answer the following questions.

- (a) Compute $P(\text{all 6 coins are heads})$.

Solution.

$$P(\text{all 6 coins are heads}) = P(X = 3)P(Y = 3) = [(1/2)^3] * [(2/3)^3].$$

- (b) Compute $P(\text{at least one head})$.

Solution. This is the same thing as 1 minus the prob of no heads.

$$1 - P(\text{no heads}) = 1 - P(X = 0)P(Y = 0) = 1 - [(1/2)^3] * [(1/3)^3].$$

- (c) Compute $P(\text{exactly 3 heads})$.

Solution. This is

$$P(X = 0)P(Y = 3) + P(X = 1)P(Y = 2) + P(X = 2)P(Y = 1) + P(X = 3)P(Y = 0),$$

which reduces to

$$[(1/2)^3] * [(2/3)^3] + [3(1/2)^3] * [3(2/3)^2(1/3)] + [3(1/2)^3] * [3(2/3)(1/3)^2] + [(1/2)^3] * [(1/3)^3].$$

4. Let $X(t)$ be WSS with $R_X(\tau) = \exp(-|\tau|)$ be passed through a filter with $h(t) = \exp(-2t)u(t)$. Compute power P_Y generated by the filter output process $Y(t)$.

Solution. The frequency response of the filter is

$$H(f) = \frac{1}{j2\pi f + 2}$$

and therefore

$$|H(f)|^2 = \frac{1}{(2\pi f)^2 + 4}$$

We also have

$$S_X(f) = \frac{2}{(2\pi f)^2 + 1}$$

The output power is therefore

$$\begin{aligned} P_Y &= \int_{-\infty}^{\infty} S_Y(f) df \\ &= \int_{-\infty}^{\infty} S_X(f) |H(f)|^2 df \\ &= \int_{-\infty}^{\infty} \frac{2}{((2\pi f)^2 + 1)((2\pi f)^2 + 4)} df \\ &= \int_{-\infty}^{\infty} \left[\frac{2/3}{(2\pi f)^2 + 1} + \frac{-2/3}{(2\pi f)^2 + 4} \right] df \\ &= [1/3 - 1/6] = 1/6 \end{aligned}$$

5. Let $X(t)$ and $Y(t)$ be independent WSS processes with the following power spectral densities:

$$S_X(f) = \begin{cases} 3, & -5 \leq f \leq 5 \\ 0, & \text{elsewhere} \end{cases}$$

$$S_Y(f) = \begin{cases} 4, & -5 \leq f \leq 5 \\ 0, & \text{elsewhere} \end{cases}$$

Let $Q(t)$ be the random signal

$$Q(t) = X(t)Y(t).$$

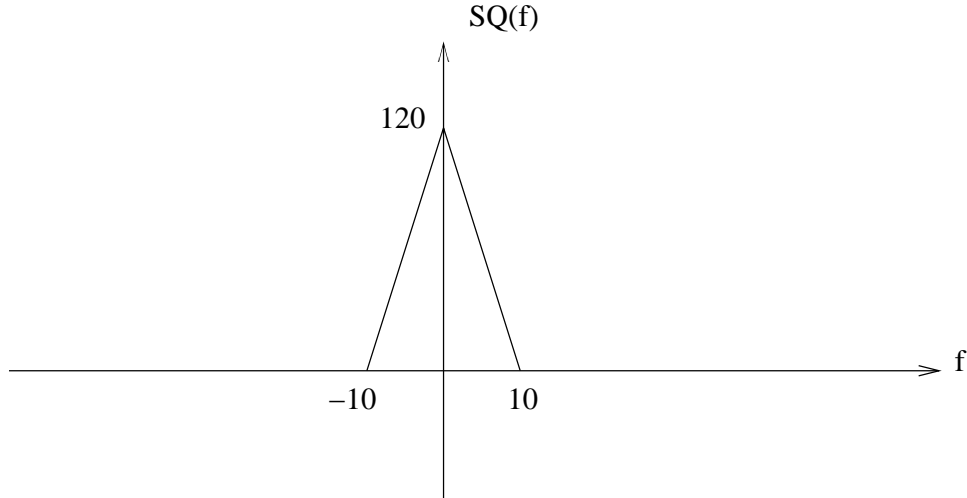
Find $S_Q(f)$ and P_Q .

Solution. We know from earlier work that

$$R_Q(\tau) = R_X(\tau)R_Y(\tau). \tag{3}$$

Plugging in $\tau = 0$, we get

$$P_Q = R_Q(0) = R_X(0)R_Y(0) = P_X P_Y = \int S_X(f) df \int S_Y(f) df = 30 * 40 = 1200.$$



Fourier transforming equation (3), we obtain

$$S_Q(f) = S_X(f) * S_Y(f).$$

We know from EE 3015 that $S_Q(f)$ automatically will be a symmetric triangular pulse extending from $f = -10$ to $f = 10$. The height of the triangle must be 120 in order for the area under it to be 1200. So the plot of $S_Q(f)$ is the the plot given above.

6. Let X_n (“the signal”) and Z_n (“the noise”) be WSS processes satisfying

$$R_X(\tau) = \begin{cases} 1, & \tau = 0 \\ 1/2, & \tau = \pm 1 \\ 0, & \text{elsewhere} \end{cases}$$

$$R_Z(\tau) = \begin{cases} 1, & \tau = 0 \\ 0, & \text{elsewhere} \end{cases}$$

We filter as follows:

$$X_n + Z_n \rightarrow \boxed{h[n]} \rightarrow X_n^0 + Z_n^0$$

where the filter impulse response is

$$h[n] = \begin{cases} 1, & n = 0, 1 \\ 0, & \text{elsewhere} \end{cases}$$

(X_n^0 and Z_n^0 are the filter responses to X_n and Z_n , respectively.)

(a) Compute the filter input SNR.

Solution.

$$\text{filter input SNR} = \frac{P_X}{P_Z} = \frac{R_X(0)}{R_Z(0)} = 1.$$

(b) Compute the filter output SNR.

Solution.

$$\text{filter output SNR} = \frac{E[(X_n^0)^2]}{E[(Z_n^0)^2]}.$$

By Theorem 5.13 of your textbook,

$$\begin{aligned} E[(X_n^0)^2] &= \begin{bmatrix} h[0] & h[1] \end{bmatrix} \begin{bmatrix} R_X(0) & R_X(1) \\ R_X(1) & R_X(0) \end{bmatrix} \begin{bmatrix} h[0] \\ h[1] \end{bmatrix} \\ &= \begin{bmatrix} 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1/2 \\ 1/2 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = 3 \end{aligned}$$

$E[(Z_n^0)^2]$ is a similar calculation, with the “matrix in the middle” equal to

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}.$$

You get

$$E[(Z_n^0)^2] = [1 \ 1] \cdot [1 \ 1] = 2.$$

The output SNR is therefore 3/2.

7. WSS process X_n is filtered as follows:

$$X_n \rightarrow \boxed{H(z)} \rightarrow Y_n$$

The filter transfer function is

$$H(z) = \frac{2 + z^{-1}}{1 + z^{-1} + 4z^{-2}}.$$

(a) Compute $S_Y(f)$ assuming that X_n is white noise with unit variance.

Solution. We have

$$S_X(f) = 1,$$

and

$$H(z)H(z^{-1}) = \frac{5 + 2(z + z^{-1})}{18 + 5(z + z^{-1}) + 4(z^2 + z^{-2})}.$$

Substituting in $z = \exp(j2\pi f)$, we obtain

$$S_Y(f) = S_X(f)H(f)H(-f) = \frac{5 + 4 \cos(2\pi f)}{18 + 10 \cos(2\pi f) + 8 \cos(4\pi f)}.$$

(b) What is μ_Y if $\mu_X = 4$?

Solution. We have

$$\mu_Y = \mu_X H(z)_{z=1} = 4 * (3/6) = 2.$$

8. A continuous-time process X is WSS and ergodic. One of its realizations $x(t)$ is periodic with period 6 and is defined over one period (extending from $t = -3$ to $t = 3$) as follows:

$$x(t) = \begin{cases} 0, & -3 \leq t < -2 \\ 6, & -2 \leq t \leq 2 \\ 0, & 2 < t \leq 3 \end{cases}$$

(All the other realizations are either forward or backward translations of $x(t)$.)

- (a) Compute the process mean μ_X via the time-averaging technique.

Solution.

$$\mu_X = \frac{1}{6} \int_{-3}^3 x(t) dt = 4.$$

- (b) Compute $R_X(0)$ via the time-averaging technique.

Solution.

$$R_X(0) = \frac{1}{6} \int_{-3}^3 x(t)^2 dt = 24.$$

- (c) Compute $R_X(2)$ via the time-averaging technique.

Solution.

$$R_X(2) = \frac{1}{6} \int_{-3}^3 x(t)x(t-2) dt = 12.$$

- (d) Compute $Var[X(52) + X(50)]$.

Solution.

$$\begin{aligned} Var[X(52) + X(50)] &= Var[X(52)] + Var[X(50)] + 2Cov(X(52), X(50)) \\ &= 2(24 - 16) + 2 * (12 - 16) = 8. \end{aligned}$$

9. A discrete-time process X is WSS but nonergodic. It has exactly four realizations $x_1[n]$, $x_2[n]$, $x_3[n]$, $x_4[n]$ (occurring with prob 0.25 each) defined as follows:

$$\begin{aligned} x_1[n] &= 1 \text{ for even } n \text{ and } x_1[n] = 2 \text{ for odd } n \\ x_2[n] &= 2 \text{ for even } n \text{ and } x_2[n] = 1 \text{ for odd } n \\ x_3[n] &= 3 \text{ for even } n \text{ and } x_3[n] = 6 \text{ for odd } n \\ x_4[n] &= 6 \text{ for even } n \text{ and } x_4[n] = 3 \text{ for odd } n \end{aligned}$$

- (a) Compute the process mean μ_X via the space-averaging technique.

Solution.

$$\mu_X = \frac{x_1[n] + x_2[n] + x_3[n] + x_4[n]}{4} = 3.$$

- (b) Compute the process variance σ_X^2 via the space-averaging technique.

Solution.

$$R_X(0) = E[X[n]^2] = \frac{x_1[n]^2 + x_2[n]^2 + x_3[n]^2 + x_4[n]^2}{4} = 12.5.$$

$$\sigma_X^2 = R_X(0) - \mu_X^2 = 3.5.$$

(c) Compute $R_X(1)$ via the space-averaging technique.

Solution.

$$R_X(1) = \frac{x_1[n]x_1[n+1] + x_2[n]x_2[n+1] + x_3[n]x_3[n+1] + x_4[n]x_4[n+1]}{4} = 10.$$

10. A continuous-time WSS process X has autocorrelation function

$$R_X(\tau) = 20 \cos(\pi|\tau|) + 100 \exp(-0.5|\tau|).$$

(a) A first order predictor

$$\hat{X}(t + 2.5) = AX(t)$$

is to be designed, in which the constant A is chosen to minimize the mean square prediction error $E[(X(t + 2.5) - \hat{X}(t + 2.5))^2]$. Using the orthogonality principle, compute the value of A .

Solution. The prediction error $X(t + 2.5) - \hat{X}(t + 2.5)$ must be orthogonal to $X(t)$. This tells us that

$$E[X(t)X(t + 2.5)] = E[X(t)\hat{X}(t + 2.5)],$$

which reduces to

$$R_X(2.5) = AR_X(0).$$

It follows that

$$A = R_X(2.5)/R_X(0) = (5/6)e^{-1.25}.$$

(b) A second order predictor

$$\hat{X}(t + 2.5) = BX(t) + CX(t - 1.5)$$

is to be designed, in which the constants B, C are chosen to minimize the prediction error $E[(X(t + 2.5) - \hat{X}(t + 2.5))^2]$. The system of equations you solve to obtain B, C takes the matrix form

$$\begin{bmatrix} a_1 & a_2 \\ a_3 & a_4 \end{bmatrix} \begin{bmatrix} B \\ C \end{bmatrix} = \begin{bmatrix} a_5 \\ a_6 \end{bmatrix},$$

where each a_i is a certain constant. Compute the values of the 6 constants $a_1, a_2, a_3, a_4, a_5, a_6$. DO NOT SOLVE FOR B and C .

Solution. The orthogonality relations tell us that

$$\begin{aligned} E[X(t)\hat{X}(t + 2.5)] &= E[X(t)X(t + 2.5)] \\ E[X(t - 1.5)\hat{X}(t + 2.5)] &= E[X(t - 1.5)X(t + 2.5)] \end{aligned}$$

These simplify to

$$\begin{aligned} BR_X(0) + CR_X(1.5) &= R_X(2.5) \\ BR_X(1.5) + CR_X(0) &= R_X(4) \end{aligned}$$

11. A continuous-time process X has power spectral density

$$S_X(f) = \cos\left(\frac{\pi f}{60}\right), \quad -30 \leq f \leq 30,$$

and equal to zero for all other f .

(a) Compute the power generated by the process X .

Solution.

$$P_X = \int_{-30}^{30} \cos(\pi f/60) df = 120/\pi.$$

(b) Process X is passed through an ideal low-pass filter with frequency response function

$$H(f) = 1, \quad -B \leq f \leq B,$$

and equal to zero for all other f . Compute the value of the bandwidth B (correct to one decimal place) so that the filter output process will generate exactly 95% of the power generated by X .

Solution. The filter output power is

$$\int_{-B}^B \cos(\pi f/60) df = \frac{120 \sin(B\pi/60)}{\pi}.$$

Setting this equal to 0.95 times $120/\pi$ leads us to the equation

$$\sin(B\pi/60) = 0.95,$$

from which we see that

$$B = \frac{60 * \text{Sin}^{-1}(0.95)}{\pi}.$$

12. In each part of this problem, the discrete-time linear time-invariant filter with impulse response

$$h[n] = 2\delta[n] - 3\delta[n - 1] + 4\delta[n - 2]$$

is used to filter a random input signal.

(a) The input X to the filter is the discrete-time white noise process with zero mean and unit variance. The output process Y has power spectral density of the form

$$S_Y(f) = A + B \cos(2\pi f) + C \cos(4\pi f).$$

Compute the constants A, B, C .

Solution. First, determine that

$$h[n] * h[-n] = 29\delta[n] - 18(\delta[n - 1] + \delta[n + 1]) + 8(\delta[n - 2] + \delta[n + 2]). \quad (4)$$

It follows easily from this that

$$A = 29, \quad B = -36, \quad C = 16.$$

- (b) The input X to the filter is a discrete-time WSS process with mean $\mu_X = 7$. Compute the mean μ_Y of the filter output process Y .

Solution.

$$\mu_Y = \mu_X \left[\sum h[n] \right] = 7 * 3 = 21.$$

- (c) The input X to the filter is a discrete-time WSS process with

$$R_X(\tau) = 8\delta[\tau] + 4(\delta[\tau - 1] + \delta[\tau + 1]).$$

Compute the power generated by the filter output process.

Solution. The output power is the coefficient of the $\delta[n]$ component of $R_X(n)$ convoluted with $h[n] * h[-n]$. Use superposition: (1) convoluting $\delta[n]$ with $h[n] * h[-n]$ gives $\delta[n]$ coefficient of 29, by (4); (2) convoluting $\delta[n - 1]$ with $h[n] * h[-n]$ gives coeff of $\delta[n]$ equal to -18 , the coeff of $\delta[n + 1]$ in $h[n] * h[-n]$; and (3) convoluting $\delta[n + 1]$ with $h[n] * h[-n]$ gives coefficient of $\delta[n]$ equal to coeff of $\delta[n - 1]$ in $h[n] * h[-n]$, which is -18 . The answer is therefore

$$[8, 4, 4] \cdot [29, -18, -18] = 88.$$

13. Let Θ_1 and Θ_2 be independent RV's, each uniformly distributed in $[0, 2\pi]$. Define continuous-time processes $X(t)$ and $Y(t)$ by:

$$X(t) = 8 \cos(t + \Theta_1) + 5, \quad -\infty < t < \infty$$

$$Y(t) = 8 \cos(t + \Theta_2) + 5, \quad -\infty < t < \infty$$

- (a) Compute $E[X(t)Y(t)]$.

Solution. $X(t)$ and $Y(t)$ are WSS, each having mean equal to 5. Therefore,

$$E[X(t)Y(t)] = E[X(t)]E[Y(t)] = \mu_X \mu_Y = 25.$$

- (b) Compute $E[(X(t) + Y(t))^2]$.

Solution. The autocorrelation functions of $X(t)$ and $Y(t)$ are

$$R_X(\tau) = 32 \cos(\tau) + 25$$

$$R_Y(\tau) = 32 \cos(\tau) + 25$$

$$E[(X(t) + Y(t))^2] = R_X(0) + R_Y(0) + 2\mu_X \mu_Y = 164.$$

- (c) Compute $E[(X(t)Y(t))^2]$.

Solution.

$$E[(X(t)Y(t))^2] = E[X(t)^2]E[Y(t)^2] = R_X(0)R_Y(0) = 3249.$$

14. Continuous-time WSS process $X(t)$ has power spectral density as follows:

$$S_X(f) = \begin{cases} 50, & -5 \leq f \leq -3 \\ 50, & 3 \leq f \leq 5 \\ 0, & \text{elsewhere} \end{cases}$$

(a) Compute $E \left[\left(\int_{-\infty}^t X(u) du \right)^2 \right]$.

Solution. You are passing $X(t)$ through a LTI system (an integrator) with frequency response function

$$H(f) = \frac{1}{j2\pi f}.$$

Letting $Y(t)$ be the output, you are computing P_Y .

$$P_Y = \int S_X(f) |H(f)|^2 df = 2 \int_3^5 \frac{50}{(2\pi f)^2} df = \frac{10}{3\pi^2}.$$

(b) Compute $E \left[\left(\frac{dX(t)}{dt} \right)^2 \right]$.

Solution. You are passing $X(t)$ through a LTI system (a differentiator) with frequency response function

$$H(f) = j2\pi f.$$

Letting $Y(t)$ be the output, you are computing P_Y .

$$P_Y = \int S_X(f) |H(f)|^2 df = 2 \int_3^5 50(2\pi f)^2 df = \frac{39200\pi^2}{3}.$$

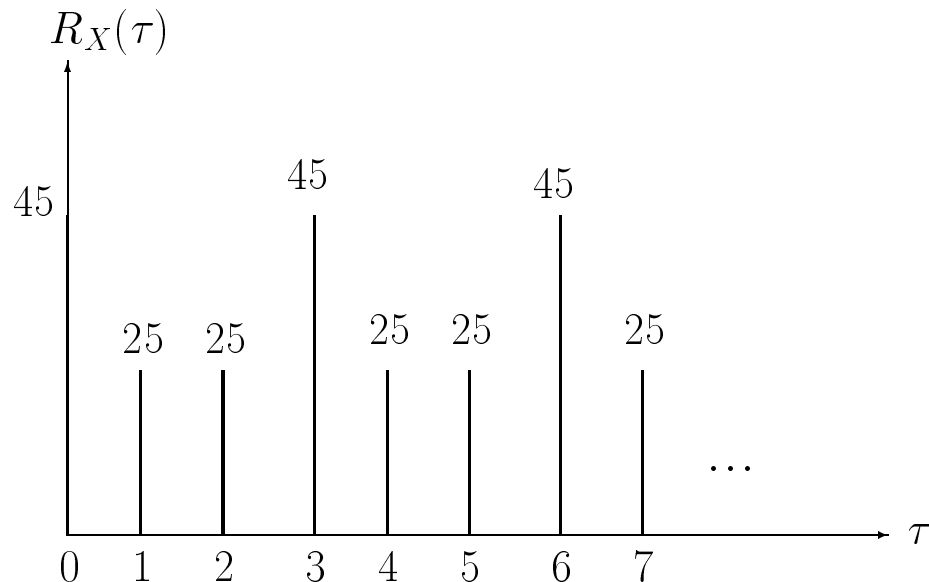
(c) Compute $E \left[(X(t) - X(t-1))^2 \right]$.

Solution. The freq response is now

$$H(f) = 1 - \exp(-j2\pi f).$$

$$P_Y = \int S_X(f) |H(f)|^2 df = 2 \int_3^5 50[2 - 2\cos(2\pi f)] df = 400.$$

15. A discrete-time WSS process X_n has autocorrelation function $R_X(\tau)$ which is periodic with period 3, and is plotted below for $\tau \geq 0$:



- (a) X_2 is to be predicted from X_0 via the first order linear predictor of the form $\hat{X}_2 = AX_0$. Determine the value of the predictor coefficient A .

Solution.

$$A = \frac{R_X(2)}{R_X(0)} = 25/45.$$

- (b) Determine the values of $\tau > 0$ for which X_τ can be perfectly predicted from X_0 using a first order linear predictor.

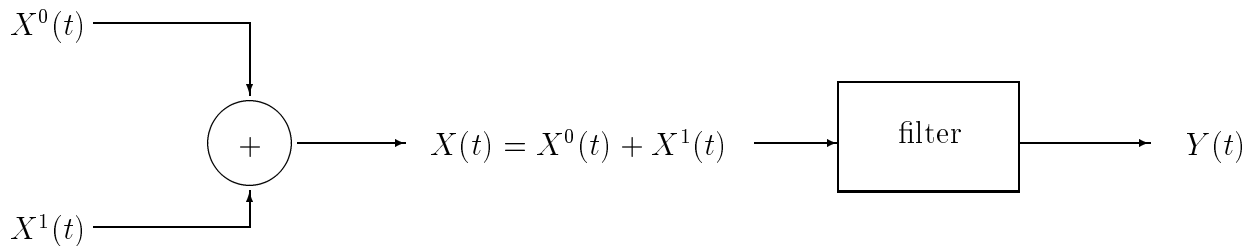
Solution. Identify the τ values where $|R_X(\tau)| = R_X(0)$. These are

$$\tau = 3, 6, 9, 12, \dots$$

- (c) Given that $\mu_X^2 = \langle R_X(\tau) \rangle$, compute μ_X^2 and σ_X^2 .

Solution.

$$\begin{aligned} \mu_X^2 &= (1/3)(45 + 25 + 25) = 95/3 \\ \sigma_X^2 &= R_X(0) - \mu_X^2 = 45 - 95/3 = 40/3 \end{aligned}$$



16. In the diagram above, $X^0(t), X^1(t)$ are 0-mean uncorrelated continuous-time random processes with

$$R_{X^0}(\tau) = 2.4e^{-10|\tau|}, \quad R_{X^1}(\tau) = \cos(20\tau).$$

The filter frequency response function is

$$H(f) = \frac{100}{j2\pi f + 20}.$$

- (a) Compute $S_X(f)$.

Solution.

$$S_X(f) = S_{X^0}(f) + S_{X^1}(f) = \frac{48}{(2\pi f)^2 + 100} + (0.5)\delta(f - 10/\pi) + (0.5)\delta(f + 10/\pi).$$

- (b) Compute P_X .

Solution.

$$P_X = R_X(\tau) = R_{X^0}(0) + R_{X^1}(0) = 3.4.$$

(c) Compute $S_Y(f)$.

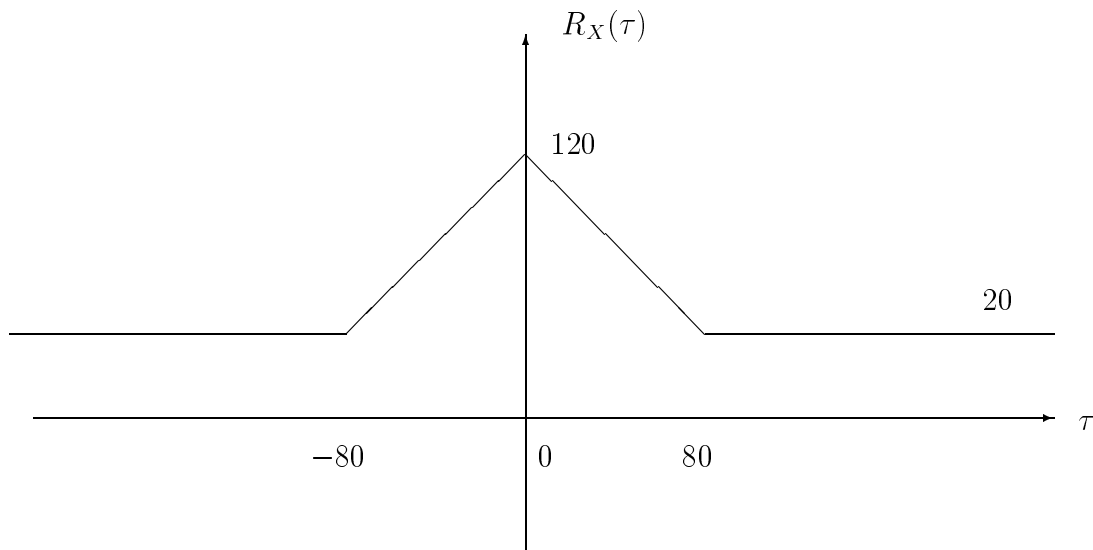
Solution.

$$\begin{aligned} S_Y(f) &= |H(f)|^2 S_X(f) \\ &= \frac{10000}{(2\pi f)^2 + 400} S_X(f) \\ &= \frac{480000}{(2\pi f)^2 + 100} \delta(f - 10/\pi) + 6.25\delta(f + 10/\pi) \end{aligned}$$

(d) Compute P_Y .

Solution.

$$\begin{aligned} P_Y &= \int_{-\infty}^{\infty} S_Y(f) df \\ &= 12.5 + \int_{-\infty}^{\infty} \left[\frac{1600}{(2\pi f)^2 + 100} - \frac{1600}{(2\pi f)^2 + 400} \right] df \\ &= 12.5 + 80 - 40 = 52.5 \end{aligned}$$



17. $X(t)$ is a Gaussian WSS ergodic process whose autocorrelation function $R_X(\tau)$ is plotted above.

(a) $E[X(500)]E[X(540)] = ?$

Solution. The answer is $\mu_X^2 = 20$.

(b) $E[X(500)X(540)] = ?$

Solution. The answer is $R_X(40) = 70$.

(c) Standard deviation of $X(500) = ?$

Solution.

$$\sigma_X^2 = R_X(0) - \mu_X^2 = 120 - 20 = 100.$$

Therefore, the standard deviation is 10.

(d) $Cov(X(500), X(540)) = ?$

Solution.

$$Cov(X(500), X(540)) = R_X(40) - \mu_X^2 = 70 - 20 = 50.$$

(e) Find smallest $t \geq 0$ for which $Cov(X(500), X(t)) = 0$.

Solution. The t we are looking for is the time t for which $R_X(t - 500) = 20$. Looking at the plot of $R_X(\tau)$, this clearly takes place at $t - 500 = 80$. Hence, $t = 580$.

(f) Find the standard deviation of $X(500) - X(540)$.

Solution.

$$Var(X(500) - X(540)) = 2\sigma_X^2 - 2Cov(X(500), X(540)) = 100.$$

Therefore, the standard deviation is 10.

(g) Assuming $\mu_X > 0$, find the constant D for which

$$P(X(500) > 15) = \int_D^\infty \frac{e^{-z^2/2}}{\sqrt{2\pi}} dz.$$

Solution. $X(500)$ has mean $\sqrt{20}$ and standard deviation 10. Subtracting this mean and dividing by this standard deviation in order to convert to standard Gaussian, we see that

$$D = \frac{15 - \sqrt{20}}{10} = 1.053.$$

(h) Find the constant C for which

$$Cov(X(500), X(500) - CX(540)) = 0.$$

Solution.

$$\begin{aligned} Cov(X(500), X(500) - CX(540)) &= \sigma_X^2 - C * Cov(X(500), X(540)) \\ &= 100 - C * 50 \end{aligned}$$

It follows that $C = 2$.

18. $X(t)$ is a Gaussian WSS process satisfying $\mu_X = 1$ and $R_X(\tau) = \delta(\tau) + 1$. $Y(t)$ is the process defined by

$$Y(t) = \int_0^1 (s+t)X(s)ds, \quad -\infty < t < \infty$$

(a) Compute $E[Y(1)]$.

Solution. Since $E[X(s)] = 1$, we have

$$E[Y(1)] = \int_0^1 (s+1)ds = 3/2$$

(b) Compute $E[Y(1)^2]$.

Solution. By the “double integral trick”,

$$\begin{aligned} E[Y(1)^2] &= \int_0^1 \int_0^1 (s_1 + 1)(s_2 + 1)(1 + \delta(s_1 - s_2)) ds_1 ds_2 \\ &= \left[\int_0^1 (s + 1) ds \right]^2 + \int_0^1 (s + 1)^2 ds = 55/12 \end{aligned}$$

(c) Determine the PDF $f(y)$ of $Y(1)$.

Solution. $Y(1)$ is Gaussian with mean 0 and variance is $55/12 - (3/2)^2 = 7/3$. This gives us

$$f(y) = \frac{1}{\sqrt{14\pi/3}} \exp(-3[y - 3/2]^2/14)$$

19. A discrete-time WSS process X has power spectral density

$$S_X(f) = 2\delta(f+1/6) + 2\delta(f-1/6) + (1.5)\delta(f+3/8) + (1.5)\delta(f-3/8), \quad -1/2 \leq f \leq 1/2.$$

Let Y be the discrete-time WSS process obtained by passing X through the discrete-time LTI filter with frequency response function

$$H(f) = \frac{5}{2 + \exp(-j2\pi f)}$$

(a) Compute the power P_X generated by process X .

Solution. Integrating $S_X(f)$ from $-1/2$ to $1/2$ gives $P_X = 7$.

(b) Compute the power P_Y generated by process Y .

Solution.

$$|H(f)|^2 = \frac{25}{5 + 4 \cos(2\pi f)}$$

$$|H(\pm 1/6)|^2 = 25/7.$$

$$|H(\pm 3/8)|^2 = 25/(5 - 2\sqrt{2}) = 11.51$$

$$P_Y = 2(25/7) + 2(25/7) + (1.5)(11.51) + (1.5)(11.51) = 48.8.$$

20. Let A, B be independent RV's each having mean 0 and variance 1. Let $X(t)$ be the continuous-time random process

$$X(t) = At + (2t - 1)B, \quad -\infty < t < \infty$$

(a) Find the value of t for which $\text{Var}[X(t)]$ is a minimum.

Solution.

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

Setting the first derivative equal to zero, one easily determines that $t = 2/5$.

(b) Find the value of t for which $E[X(t)X(t-1)]$ is a minimum.

Solution.

$$X(t)X(t-1) = \{At + (2t - 1)B\}\{A(t-1) + (2t-3)B\}$$

$$E[X(t)X(t-1)] = t(t-1) + (2t-1)(2t-3)$$

Setting the derivative equal to zero yields $t = 9/10$.