10.1.6 Solved Problems

Problem 1

Let Y_1, Y_2, Y_3, \cdots be a sequence of i.i.d. random variables with mean $EY_i = 0$ and $Var(Y_i) = 4$. Define the discrete-time random process $\{X(n), n \in \mathbb{N}\}$ as

$$X(n)=Y_1+Y_2+\cdots+Y_n,\quad ext{for all }n\in\mathbb{N}.$$

Find $\mu_X(n)$ and $R_X(m,n)$, for all $n,m\in\mathbb{N}$.

Solution

We have

$$\mu_X(n) = E[X(n)]$$

$$= E[Y_1 + Y_2 + \dots + Y_n]$$

$$= E[Y_1] + E[Y_2] + \dots + E[Y_n]$$

$$= 0.$$

Let $m \leq n$, then

$$egin{aligned} R_X(m,n) &= E[X(m)X(n)] \ &= E\left[X(m)ig(X(m) + Y_{m+1} + Y_{m+2} + \dots + Y_nig)
ight] \ &= E[X(m)^2] + E[X(m)]E[Y_{m+1} + Y_{m+2} + \dots + Y_n] \ &= E[X(m)^2] + 0 \ &= \mathrm{Var}ig(X(m)ig) \ &= \mathrm{Var}ig(Y(m)ig) + \mathrm{Var}ig(Y(m)ig) + \dots + \mathrm{Var}ig(Y(m)ig) \ &= 4m. \end{aligned}$$

Similarly, for $m \ge n$, we have

$$R_X(m,n) = E[X(m)X(n)]$$

= $4n$.

We conclude

$$R_X(m,n) = 4\min(m,n).$$

Problem 2

For any $k \in \mathbb{Z}$, define the function $g_k(t)$ as

$$g_k(t) = \left\{ egin{array}{ll} 1 & & k < t \leq k+1 \ & & \ 0 & & ext{otherwise} \end{array}
ight.$$

Now, consider the continuous-time random process $\left\{X(t), t \in \mathbb{R} \right\}$ defined as

$$X(t) = \sum_{k=-\infty}^{+\infty} A_k g_k(t),$$

where $A_1,\,A_2,\,\cdots$ are i.i.d. random variables with $EA_k=1$ and ${\rm Var}(A_k)=1$. Find $\mu_X(t)$, , $R_X(s,t)$, and $C_X(s,t)$ for all $s,t\in\mathbb{R}$.

Solution

Note that, for any $k \in \mathbb{Z}$, g(t) = 0 outside of the interval (k, k+1]. Thus, if $k < t \le k+1$, we can write

$$X(t) = A_k$$
.

Thus,

$$\mu_X(t) = E[X(t)]$$

= $E[A_k] = 1$.

So, $\mu_X(t)=1$ for all $t\in\mathbb{R}.$ Now consider two real numbers s and t. If for some $k\in\mathbb{Z},$ we have

$$k < s, t < k + 1,$$

then

$$R_X(s,t) = E[X(s)X(t)] \ = E[A_k^2] = 1 + 1 = 2.$$

On the other hand, if s and t are in two different subintervals of \mathbb{R} , that is if

$$k < s \le k+1$$
, and $l < t \le l+1$,

where k and l are two different integers, then

$$R_X(s,t) = E[X(s)X(t)]$$

= $E[A_kA_l] = E[A_k]E[A_l] = 1.$

To find $C_X(s,t)$, note that if \

$$k < s, t \le k + 1,$$

then

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

= 2 - 1 \cdot 1 = 1.

On the other hand, if

$$k < s \le k+1$$
, and $l < t \le l+1$,

where k and l are two different integers, then

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

= 1 - 1 \cdot 1 = 0.

Problem 3

Let X(t) be a continuous-time WSS process with mean $\mu_X = 1$ and

$$R_X(au) = \left\{ egin{array}{ll} 3 - | au| & -2 \leq au \leq 2 \ \ 1 & ext{otherwise} \end{array}
ight.$$

a. Find the expected power in X(t).

b. Find
$$E\left[\left(X(1)+X(2)+X(3)\right)^2\right]$$
.

Solution

a. The expected power in X(t) at time t is $E[X(t)^2]$, which is given by

$$R_X(0) = 3.$$

b. We have

$$E\left[\left(X(1) + X(2) + X(3)\right)^{2}\right] = E\left[X(1)^{2} + X(2)^{2} + X(3)^{2} + 2X(1)X(2) + 2X(1)X(3) + 2X(2)X(3)\right]$$

$$= 3R_{X}(0) + 2R_{X}(-1) + 2R_{X}(-2) + 2R_{X}(-1)$$

$$= 3 \cdot 3 + 2 \cdot 2 + 2 \cdot 1 + 2 \cdot 2$$

$$= 19.$$

Problem 4

Let X(t) be a continuous-time WSS process with mean $\mu_X=0$ and

$$R_X(\tau) = \delta(\tau),$$

where $\delta(\tau)$ is the Dirac delta function. We define the random process Y(t) as

$$Y(t) = \int_{t-2}^t X(u) du.$$

- a. Find $\mu_Y(t) = E[Y(t)]$.
- b. Find $R_{XY}(t_1, t_2)$.

Solution

a. We have

$$\mu_Y(t) = E\left[\int_{t-2}^t X(u)du
ight]$$

$$= \int_{t-2}^t E[X(u)] du$$

$$= \int_{t-2}^t 0 du$$

$$= 0.$$

b. We have

$$egin{aligned} R_{XY}(t_1,t_2) &= E\left[X(t_1)\int_{t_2-2}^{t_2}X(u)du
ight] \ &= E\left[\int_{t_2-2}^{t_2}X(t_1)X(u)du
ight] \ &= \int_{t_2-2}^{t_2}R_X(t_1-u)\;du \ &= \int_{t_2-2}^{t_2}\delta(t_1-u)\;du \ &= \begin{cases} 1 & t_2-2 < t_1 < t_2 \ 0 & ext{otherwise} \end{aligned}$$

Problem 5

Let X(t) be a Gaussian process with $\mu_X(t)=t$, and $R_X(t_1,t_2)=1+2t_1t_2$, for all $t,t_1,t_2\in\mathbb{R}$. Find $P\big(2X(1)+X(2)<3\big)$.

Solution

Let Y = 2X(1) + X(2). Then, Y is a normal random variable. We have

$$\begin{split} EY &= 2E[X(1)] + E[X(2)] \\ &= 2 \cdot 1 + 2 = 4. \end{split}$$

$$\mathrm{Var}(Y) &= 4\mathrm{Var}\big(X(1)\big) + \mathrm{Var}\big(X(2)\big) + 4\mathrm{Cov}\big(X(1), X(2)\big). \end{split}$$

Note that

$$\operatorname{Var}ig(X(1)ig) = E[X(1)^2] - E[X(1)]^2$$
 $= R_X(1,1) - \mu_X(1)^2$
 $= 1 + 2 \cdot 1 \cdot 1 - 1 = 2.$
 $\operatorname{Var}ig(X(2)ig) = E[X(2)^2] - E[X(2)]^2$
 $= R_X(2,2) - \mu_X(2)^2$
 $= 1 + 2 \cdot 2 \cdot 2 - 4 = 5.$
 $\operatorname{Cov}ig(X(1),X(2)ig) = E[X(1)X(2)] - E[X(1)]E[X(2)]$
 $= R_X(1,2) - \mu_X(1)\mu_X(2)$
 $= 1 + 2 \cdot 1 \cdot 2 - 1 \cdot 2 = 3.$

Therefore,

$$Var(Y) = 4 \cdot 2 + 5 + 4 \cdot 3 = 25.$$

We conclude $Y \sim N(4,25)$. Thus,

$$egin{split} Pig(Y < 3ig) &= \Phi\left(rac{3-4}{5}
ight) \ &= \Phi(-0.2) pprox 0.42 \end{split}$$