# Stochastic Processes

Homework #2

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**Q.** We define  $X_0 = Y_0$  and  $X_n = \lambda X_n 1 + Y_n$  for  $n \ge 1$ . Suppose  $Y_0, Y_1, ...$  are uncorrelated random variables with  $\mathrm{E}[Y_n] = 0$  and  $\mathrm{Var}(Y_n) = \sigma^2/(1-\lambda^2)$  if n = 0 and  $\sigma^2$  otherwise, where  $\lambda < 1$ . Find  $Cov(X_n, X_{n+m})$ . Is this process WSS? why?

**A.** If we expand  $X_n$  we will have:

$$X_n = \lambda X_{n-1} + Y_n$$

$$= \lambda(\lambda X_{n-2} + Y_{n-1}) + Y_n$$
...
$$= \lambda^n Y_0 + \lambda^{n-1} Y_1 + \dots + Y_n$$

$$= \sum_{i=0}^n \lambda^{n-i} Y_i$$

For expectation we can write:

$$E[X_n] = E[\sum_{i=0}^n \lambda^{n-i} Y_i]$$

$$= \sum_{i=0}^n \lambda^{n-i} E[Y_i]$$

$$= \sum_{i=0}^n \lambda^{n-i} \times 0$$

$$= 0$$

For covariance we can write:

$$Cov(X_{n}, X_{n+m}) = C_{XX}(n, n+m)$$

$$= R_{XX}(n, n+m) - E[X]^{2}$$

$$= R_{XX}(n, n+m)$$

$$= E[X_{n}X_{n+m}]$$

$$= E[(\sum_{i=0}^{n} \lambda^{n-i}Y_{i})(\sum_{i=0}^{n+m} \lambda^{n+m-i}Y_{i})]$$

$$= E[\sum_{i=0}^{n} \sum_{j=0}^{n+m} \lambda^{n-i}Y_{i}\lambda^{n+m-j}Y_{j}]$$

$$= \sum_{i=0}^{n} \sum_{j=0}^{n+m} \lambda^{2n+m-i-j}E[Y_{i}Y_{j}]$$

Since  $Y_i$  are correlated we have  $E[Y_iY_j] = E[Y_i]E[Y_j] = 0$  for  $i \neq j$ . So:

$$\operatorname{Cov}(X_n, X_{n+m}) = \sum_{i=0}^n \lambda^{2n+m-2i} \operatorname{E}[Y_i Y_i]$$

$$= \lambda^{2n+m} \frac{\sigma^2}{(1-\lambda^2)} + \sum_{i=1}^n \lambda^{2n+m-2i} \sigma^2$$

$$= \lambda^{2n+m} \sigma^2 \left[ \frac{1}{(1-\lambda^2)} + \sum_{i=1}^n \lambda^{-2i} \right]$$

$$= \lambda^{2n+m} \sigma^2 \frac{1}{(1-\lambda^2)(\lambda^{2n})}$$

$$= \frac{\lambda^m \sigma^2}{(1-\lambda^2)}$$

The autocorrelation function  $R_{XX}$  only depends on the difference of indexes, m. Also, as we see in previous part,  $E[X_n]$  is a constant. Thus, this process is wide-sense stationary.

**Q.** Prove that Var(X(t+s)X(t)) = 2RX(0)2RX(s). Let X(t) be a WSS.

**A.** We have:

$$\begin{split} \mathrm{E}[(X(t+s)-X(t))^2] &= \mathrm{E}[X(t+s)^2 + X(t)^2 - 2X(t+s)X(t)] \\ &= \mathrm{E}[X(t+s)^2] + \mathrm{E}[X(t)^2] - 2\mathrm{E}[X(t+s)X(t)] \\ &= R_X(0) + R_X(0) - 2R_X(s) \\ &= 2R_X(0) - 2R_X(s) \\ \\ E[X(t+s) - X(t)]^2 &= \left(\mathrm{E}[X(t+s)] - \mathrm{E}[X(t)]\right)^2 \\ &= (\mu_X - \mu_X)^2 \\ &= 0 \end{split}$$

So:

$$Var(X(t+s) - X(t)) = E[(X(t+s) - X(t))^{2}] - E[X(t+s) - X(t)]^{2}$$
$$= 2R_{X}(0) - 2R_{X}(s)$$

**Q.** Let  $X_i (i \in Z)$  be a process in which the  $X_i$ s are i.i.d. Supposing CDF  $F_{X_i}(x) = F(x)$ , prove that this process is SSS.

**A.** We will calculate joint CDF function and show that it is independent of indexes.

$$F(x_1, x_2, ..., x_n; k_1, ..., k_n) = P(X_{k_1} \le x_1, ..., X_{k_n} \le x_n)$$

Since  $X_i$ s are i.i.d we can write:

$$F(x_1, x_2, ..., x_n; k_1, ..., k_n) = P(X_{k_1} \le x_1, ..., X_{k_n} \le x_n)$$

$$= P(X_{k_1} \le x_1) P(X_{k_2} \le x_2) ... P(X_{k_n} \le x_n)$$

$$= \prod_{i=1}^{i=n} P(X_{k_i} \le x_i)$$

$$= \prod_{i=1}^{i=n} F_{X_{k_i}}(x_i)$$

$$= \prod_{i=1}^{i=n} F(x_i)$$

So the joint CDF function is independent of indexes, and thus, this process is strict-sense stationary.

**Q.** Consider the process X(t) = Y(t+T) in which  $T \sim U(0,T_0)$  and Y(t) is a periodic function with period  $T_0$ . Is this process SSS? Give reasons for your answer.

A. Consider the CDF of n-th order distribution function:

$$F(x_1, ..., x_n; t_1, ..., t_n) = P(X(t_1) \le x_1, ..., X(t_n) \le x_n)$$
  
=  $P(Y(t_1 + T) \le x_1, ..., Y(t_n + T) \le x_n)$ 

However, since Y(t) is periodic and  $T \sim U(0, T_0)$ , the space that  $Y(t_i + T) \leq x_i$  represents is the same as that of  $Y(T) \leq x_i$ ; Because for every  $\theta$  as a value of T there exists a  $0 \leq \theta' \leq T_0$  where  $Y(t_i + \theta) = Y(\theta')$ . Specifically,  $\theta' = (t_i + \theta) \% T_0$ .

An illustration of this is shown below:

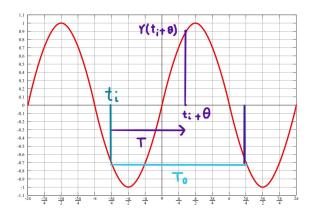


Figure 1:  $Y(t_i + T)$ 

Figure 2: Y(T)

Hence, we can write:

$$F(x_1,...,x_n;t_1,...,t_n) = P(Y(T) \le x_1,...,Y(T) \le x_n)$$

Thus, the n-th order distribution function is independent of indexes and this process is strict-sense stationary.

**Q.** Consider the process X(t) = Y + Zt in wich Y and Z are normal N(1,1) and independent random variables. For this process find the correlation function and covariance function.

A.

$$\begin{split} R_{XX}(t,t+s) &= \mathrm{E}[X(t+s)X(t)] \\ &= \mathrm{E}[\left(Y+Zt\right)\left(Y+Z(t+s)\right)] \\ &= \mathrm{E}[Y^2+YZ(t+s)+YZt+Z^2t(t+s)] \\ &= \mathrm{E}[Y^2] + \mathrm{E}[YZ(2t+s)] + \mathrm{E}[Z^2t(t+s)] \\ &= \mathrm{E}[Y^2] + (2t+s)\mathrm{E}[Y]\mathrm{E}[Z] + t(t+s)\mathrm{E}[Z^2] \\ &= (\sigma^2+\mu^2) + (2t+s)(\mu^2) + t(t+s)(\sigma^2+\mu^2) \\ &= 2 + (2t+s) + 2t(t+s) \\ &= 2t^2 + 2st + 2t + s + 2 \end{split}$$

We can also write the above result in following form if we assume  $t_1 = t$  and  $t_2 = t + s$ :

$$R_{XX}(t,t+s) = (\sigma^2 + \mu^2) + (t_1 + t_2)(\mu^2) + t_1 t_2(\sigma^2 + \mu^2)$$
$$= 2t_1 t_2 + t_1 + t_2 + 2$$

$$E[X(t)] = E[(Y + Zt)]$$

$$= E[Y] + tE[Z]$$

$$= (1 + t)\mu$$

$$= t + 1$$

$$C_{XX}(t,t+s) = R_{XX}(t,t+s) - E[X(t)]E[X(t+s)]$$

$$= 2t^2 + 2st + 2t + s + 2 - (t+1)(t+s+1)$$

$$= 2t^2 + 2st + 2t + s + 2 - t^2 - ts - 2t - s - 1$$

$$= t^2 + ts + 1$$

We can also write the above result in following form if we assume  $t_1 = t$  and  $t_2 = t + s$ :

$$C_{XX}(t, t + s) = t_1 t_2 + 1$$

**Q.** Define  $X(t) = A \cos(wt) + B \sin(wt)$ , where A and B are independent unit normal random variables and w is constant . Show that X(t) is a WSS.

**A.** First we consider expected value:

$$E[X(t)] = E[A\cos(wt) + B\sin(wt)]$$
$$= \cos(wt)E[A] + \sin(wt)E[B]$$
$$= 0$$

So first-order moment is a constant and independent of t.

For second-order moment, we first calculate  $E[A^2]$ ,  $E[B^2]$ , and E[AB]:

$$E[B^2] = E[A^2] = var(A) + E[A]^2 = 1$$
  
 $E[AB] = E[A]E[B] = 0$ 

Then we calculate auto-correlation:

$$\begin{split} R_{XX}[t,t+s] = & \operatorname{E}[X(t)X(t+s)] \\ = & \operatorname{E}\Big[\Big(A\cos(wt) + B\sin(wt)\Big)\Big(A\cos(w(t+s)) + B\sin(w(t+s))\Big)\Big] \\ = & \operatorname{E}[A^2(wt)\cos(w(t+s)) + B^2\sin(wt)\sin(w(t+s)) + AB\cos(wt)\sin(w(t+s)) + AB\sin(wt)\cos(w(t+s))] \\ = & \cos(wt)\cos(w(t+s))\operatorname{E}[A^2] + \sin(wt)\sin(w(t+s))\operatorname{E}[B^2] \\ & + \Big(\cos(wt)\sin(w(t+s)) + \sin(wt)\cos(w(t+s))\Big)\operatorname{E}[AB] \\ = & \cos(wt)\cos(w(t+s)) + \sin(wt)\sin(w(t+s)) \\ = & \cos(wt - w(t+s)) \\ = & \cos(ws) \end{split}$$

Second-order moment is a function of the difference of two indexes. Thus, this process is wide-sense stationary

**Q.** Suppose  $X_1, X_2, ...$  are i.i.d with  $E[X_i] = 0$  and  $Var(X_i) = 4$ . For the process  $y(n) = X_1 + X_2 + ... + X_n (n \in N)$  find the mean and covariance.

#### Α.

Mean:

$$E[Y_n] = E[X_1 + \dots + X_n]$$
$$= \sum_{i=1}^n E[X_i]$$
$$= 0$$

Covariance:

$$R_{YY}(n, n + m) = E[Y_n Y_{n+m}]$$

$$= E\Big[\Big(\sum_{i=1}^n X_i\Big)\Big(\sum_{j=1}^{n+m} X_j\Big)\Big]$$

$$= E\Big[\sum_{i=1}^n \sum_{j=1}^{n+m} X_i X_j\Big]$$

$$= \sum_{i=1}^n E[X_i^2] + \sum_{i=1}^n \sum_{j=1, i \neq j}^{n+m} E[X_i X_j]$$

$$= \sum_{i=1}^n E[X_i^2] + \sum_{i=1}^n \sum_{j=1, i \neq j}^{n+m} E[X_i] E[X_j]$$

$$= \sum_{i=1}^n E[X_i^2]$$

$$= \sum_{i=1}^n (Var[X_i] + E[X_i]^2)$$

$$= \sum_{i=1}^n 4$$

$$= 4n$$