ECE 673-Random signal analysis I Final - Dec. 20th 2006

Q1. The random processes X[n] and Y[n] are both WSS with zero mean and autocorrelation sequences $r_X[k]$ and $r_Y[k]$. Moreover, every sample of X[n] is independent from every sample of Y[n]. Is Z[n] = X[n] + Y[n] WSS? If so, find mean and autocorrelation sequence of Z[n].

Sol.: Let us calculate mean and autocorrelation in order to check if the process is WSS:

$$E[Z[n]] = E[X[n]] + E[Y[n]] = 0$$

$$E[Z[n]Z[n+k]] = E[(X[n] + Y[n])(X[n+k] + Y[n+k])] =$$

= $E[X[n]X[n+k]] + E[Y[n]Y[n+k]] =$
= $r_X[k] + r_Y[k] = r_Z[k],$

where we have used the fact that $E[X[n]Y[n+k]] = E[X[n]] \cdot E[Y[n+k]] = 0$ since X[n] and Y[n] are independent and zero mean. We can conclude that the process is in fact WSS.

Q2. Consider the same random process X[n] and Y[n] of the previous point. Is Z[n] = X[n]Y[n] WSS? If so, find mean and autocorrelation sequence of Z[n].

Sol.: Let us calculate mean and autocorrelation in order to check if the process is WSS. We have

$$E[Z[n]] = E[X[n]] \cdot E[Y[n]] = 0,$$

since the processes are independent and zero mean, and

$$\begin{split} E[Z[n]Z[n+k]] &= E[(X[n]Y[n]) \cdot (X[n+k]Y[n+k])] = \\ &= E[X[n]X[n+k]] \cdot E[Y[n]Y[n+k]] = \\ &= r_X[k] \cdot r_Y[k] = r_Z[k], \end{split}$$

where we have used the fact that X[n]X[n+k] and Y[n]Y[n+k] are independent. We can conclude that the process is in fact WSS.

Q3. You are given a WSS process U[n] with power spectrum density $P_U(f) = 1/(2 - 2\cos(4\pi f))$ and you are asked to design a LTI with impulse response h[n] such that the output of the LTI with input U[n] is white noise. Find the impulse response h[n].

Sol.: The output of the LTI with impulse response h[n] and input U[n] is a WSS process with power spectral density

$$P_X(f) = |H(f)|^2 P_U(f) = \frac{|H(f)|^2}{(2 - 2\cos(4\pi f))}$$

Therefore, in order for the output X[n] to be white noise, we should set $|H(f)|^2 = (2 - 2\cos(4\pi f))$, possibly multiplied by an arbitrary constant. This implies that $H(f) = 1 - 2\cos(4\pi f)$.

 $\exp(-j4\pi f)$ (in fact: $|H(f)|^2 = H(f)H^*(f) = (1 - \exp(-j4\pi f))(1 - \exp(j4\pi f)) = 1 + 1 - 2\cos(4\pi f))$ and thus $h[n] = \delta[n] - \delta[n-2]$.

Q4. A WSS random process X[n] is characterized by mean $\mu_X = 1$ and autocorrelation function $r_X[k] = (1/2)^{|k|}$. Find the covariance cov(X[1], X[2]).

Sol.: We have

$$cov(X[1], X[2]) = E[X[1]X[2]] - E[X[1]]E[X[2]] =$$

= $r_X[1] - \mu_X^2 = 1/2 - 1 = -1/2.$

P1. Consider a Linear Time-Invariant system (LTI) with input U[n] being a white Gaussian noise with power $\sigma^2 = 1/2$ and impulse response $h[n] = \delta[n] + \delta[n-1]$.

(i) Is the output process X[n] WSS? Is it a Gaussian process? Is it stationary? If it is WSS, find mean and autocorrelation of the output.

Sol.: Yes, X[n] is WSS and Gaussian, since it is the output of a LTI system with a WSS Gaussian process at the input. Let us evaluate mean and autocorrelation

$$\mu_X = \mu_U \sum_n h[n] = 0,$$

$$r_X[k] = \frac{1}{2} r_h[k] = \frac{1}{2} (2\delta[n] + \delta[n-1] + \delta[n+1]),$$

where the latter result can be easily shown by using the graphical approach.

(ii) Find and sketch the joint PDF of X[3] and X[4].

Sol.: From the previous point, X[3] and X[4] are jointly Gaussian with mean and covariance matrix as $(\mathbf{X} = [X[3] \ X[4]]^T)$

$$\mathbf{X} \sim \mathcal{N}\left(\left[\begin{array}{c}0\\0\end{array}\right], \left[\begin{array}{c}r_X[0] & r_X[1]\\r_X[-1] & r_X[0]\end{array}\right] = \left[\begin{array}{c}1 & 0.5\\0.5 & 1\end{array}\right]\right).$$

Therefore, X[3] and X[4] are standard jointly Gaussian variables with $\rho = 0.5$.

(*iii*) Find the conditional PDF of X[4] given X[3].

Sol.: Since X[3] and X[4] are standard jointly Gaussian variables, we have that

$$(X[4] \mid X[3] = x) \sim \mathcal{N}(\rho x = 0.5x, 1 - \rho^2 = 0.75).$$

(iv) Evaluate the probability that X[3] - X[4] > 1.

Sol.: Linear combinations of jointly Gaussian random variables are Gaussian as well. In particular, it is zero-mean and with variance

$$var(X[3] - X[4]) = var(X[3]) + var(-X[4]) + 2 \cdot cov(X[3], -X[4]) = = var(X[3]) + var(X[4]) - 2 \cdot cov(X[3], X[4]) = = 2 \cdot 1 - 2 \cdot 0.5 = 1$$

so that $X[3] - X[4] \sim \mathcal{N}(0, 1)$ and

$$P[X[3] - X[4] > 1] = Q(1).$$

P2. A process is described by the difference equation

$$X[n] = \frac{1}{2}X[n-2] + U[n],$$

where U[n] is white noise with variance one.

(i) Interpret the difference equation as a LTI by finding the corresponding impulse response h[n].

Sol.: Let us find the impulse response corresponding to the difference equation at hand. We can do so by letting $U[n] = \delta[n]$ and evaluating the output h[n] time instant by time instant (notice that h[n] = 0 for n < 0)

$$h[0] = \delta[0] = 1$$

$$h[1] = \delta[1] = 0$$

$$h[2] = \frac{1}{2}h[0] = \frac{1}{2}$$

$$h[3] = \frac{1}{2}h[1] = 0$$

$$h[4] = \frac{1}{2}h[2] = \frac{1}{4}$$

...

We can then conclude that

$$h[n] = \begin{cases} 0 & n \text{ odd} \\ (\frac{1}{2})^{n/2} & n \text{ even} \end{cases}$$

(ii) Evaluate and plot the power spectral density of X[n].

Sol.: In order to find the power spectra density it is convenient to use the z-transform (system function) approach starting from the difference equation:

$$X[n] = \frac{1}{2}X[n-2] + U[n]$$

$$\longleftrightarrow \quad X(z) = \frac{1}{2}X(z)z^{-2} + U(z),$$

which easily implies

$$H(z) = \frac{1}{1 - 1/2 \cdot z^{-2}}$$

Therefore, the frequency response reads

$$H(f) = \frac{1}{1 - 1/2 \cdot \exp(-4\pi f)}.$$



Figure 1:

The power spectral density of the output is then

$$P_X(f) = |H(f)|^2 = \frac{1}{(1 - 1/2 \cdot \exp(-4\pi f))(1 - 1/2 \cdot \exp(4\pi f))} = \frac{1}{1 + 1/4 - \cos(4\pi f)} = \frac{1}{5/4 - \cos(4\pi f)}.$$

P3. We are interested in studying the sum-process

$$X[n] = \sum_{i=0}^{n} U[i],$$

where U[n] is an IID process with $U[n] \sim Ber(0.6)$.

(i) Is X[n] stationary? Is it WSS?

Sol.: No, the process is not stationary and not even WSS. In fact, the average is dependent on time n:

$$E[X[n]] = \sum_{i=0}^{n} E[U[n]] = (n+1)E[U[n]] = (n+1) \cdot 0.6.$$

(ii) Evaluate the probabilities P[X[0] = 1] and P[X[0] = 1, X[1] = 2].

Sol.: We have

$$P[X[0] = 1] = P[U[0] = 1] = 0.6$$

and

$$\begin{split} P[X[0] &= 1, X[1] = 2] = P[U[0] = 1, U[1] = 1] = \\ P[U[0] &= 1] \cdot P[U[1] = 1] = (0.6)^2 = 0.36. \end{split}$$

(iii) Find the PMF of X[n] for a given n.

Sol.: Since X[n] is a sum of n + 1 independent Bernoulli random variables, $X[n] \sim bin(n + 1, 0.6)$.