MVE136 Random Signals Analysis

Written exam Sunday 26 January 2014 1 – 5 pm

TEACHER AND JOUR: Patrik Albin.

AIDS: Beta <u>or</u> 2 sheets (4 pages) of hand-written notes (computer print-outs and/or xerox-copies are not allowed), but not both these aids.

GRADES: 12 (40%), 18 (60%) and 24 (80%) points for grade 3, 4 and 5, respectively. GOOD LUCK!

Task 1. Let *A* be a random variable with finite mean and finite variance that is independent of another random variable Θ that is uniformly distributed over the interval $[0, 2\pi]$. Consider the discrete time random process $\{X_n, n \in \mathbb{Z}\}$ given by $X_n = A \cos(\Theta n)$. Show that the process X_n is not wide-sense stationary (WSS). (Hint: The formula $\cos(x)\cos(y) = \frac{1}{2}\cos(x+y) + \frac{1}{2}\cos(y-x)$ can be useful.) (5 points)

Task 2. A Wiener process (also called Brownian motion) is a zero-mean continuous time Gaussian process $\{W(t), t \ge 0\}$ with autocorrelation function $R_{WW}(s,t) = \min(s,t)$. [Hence the autocorrelation agrees with the autocovariance function of a Poisson processes with arrival rate (/intensity) 1.] Prove that the Wiener process has independent increments, which is to say, that $W(t_4) - W(t_3)$ and $W(t_2) - W(t_1)$ are independent random variables for $0 \le t_1 \le t_2 \le t_3 \le t_4$. **(5 points)**

Task 3. Find the probability P(X(t) + X(2t) = 5) when $\{X(t), t \ge 0\}$ is a Poisson process with arrival rate (/intensity) $\lambda > 0$. (5 points)

Task 4. Consider a so called discrete time random walk process $\{X[n], n \ge 0\}$ on the (both positive and negative) integers that starts at the origin (that is, X[0] = 0). At each time instant, the process either increases by 1 with probability p or decreases by 1 with probability 1 - p. Explain why this process is a Markov chain and find its transition probabilities $p_{i,j}$. Are the states of this chain periodic or aperiodic? If they are period, what is the period? (5 points)

Task 5. A wide sense stationary (WSS) continuous time signal process Z(t) with power spectral density (PSD) $S_{ZZ}(f)$ is sent on a noisy channel with an additive WSS noise process N(t) that is independent of Z(t) and has PSD $S_{NN}(f)$. Hence the recived noise disturbed signal process X(t) is given by X(t) = Z(t) + N(t). We send X(t) through a so called Wiener filter with output process Y(t) that is designed to minimize the mean square error $E[(Z(t) - Y(t))^2]$. What is the PSD of Y(t)? (5 points)

Task 6. Explain what weakness of the periodogram that the modified periodogram tries to remove. What is the difference (/are the differences) between the periodogram and the modified periodogram? (5 points)

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Solutions to written exam Sunday 26 January 2014

Task 1. We have $E(X_m X_n) = \frac{1}{2} E(A^2) E(\cos(\Theta(m+n)) + \frac{1}{2} E(A^2) E(\cos(\Theta(m-n))) = \frac{1}{2} E(A^2) \delta(m+n) + \frac{1}{2} E(A^2) \delta(m-n)$, which is not a function of m-n only, so that X_n is not WSS.

Task 2. As $W(t_4) - W(t_3)$ and $W(t_2) - W(t_1)$ are (jointly) Gaussian random variables they are independent if they are uncorrelated. That their correlation is zero, in turn, follows from noting that $E[(W(t_4) - W(t_3)) (W(t_2) - W(t_1))] = C_{WW}(t_4, t_2) - C_{WW}(t_3, t_2) - C_{WW}(t_4, t_1) + C_{WW}(t_3, t_1) = R_{WW}(t_4, t_2) - R_{WW}(t_3, t_2) - R_{WW}(t_4, t_1) + R_{WW}(t_3, t_1) = min(t_4, t_2) - min(t_3, t_2) - min(t_4, t_1) + min(t_2, t_1) = t_2 - t_2 - t_1 + t_1 = 0.$

Task 3. As X(t) and X(2t) - X(t) are independent Poisson distributed with parameter λt it follows that $P(X(t) + X(2t) = 5) = P(X(2t) - X(t) = 5) P(X(t) = 0) + P(X(2t) - X(t) = 3) P(X(t) = 1) + P(X(2t) - X(t) = 1) P(X(t) = 2) = \frac{(\lambda t)^5}{5!} e^{-\lambda t} \frac{(\lambda t)^0}{0!} e^{-\lambda t} + \frac{(\lambda t)^3}{3!} e^{-\lambda t} \frac{(\lambda t)^1}{1!} e^{-\lambda t} + \frac{(\lambda t)^1}{1!} e^{-\lambda t} \frac{(\lambda t)^2}{2!} e^{-\lambda t} = (\frac{(\lambda t)^5}{120} + \frac{(\lambda t)^4}{6} + \frac{(\lambda t)^3}{2}) e^{-2\lambda t}.$

Task 4. This problem is solved in Example 9.11 in the book by Miller and Childers.

Task 5. As the Wiener filter has transfer function $H(f) = S_{ZZ}(f)/(S_{ZZ}(f)+S_{NN}(f))$ we have $S_{YY}(f) = |H(f)|^2 S_{XX}(f) = H(f)^2 (S_{ZZ}(f)+S_{NN}(f)) = S_{ZZ}(f)^2/(S_{ZZ}(f)+S_{NN}(f))$.

Task 6. The periodogram has limited resolution whenever the sampled (observed) sequence has finite length, N. Perhaps the most serious problem is that weak frequency components can be hidden/masked in the sidelobes of a strong frequency component. The modified periodogram tries to limit this "frequency masking". Another way to put it, is that the periodogram has a bias for finite N and that the modified periodogram tries to reduce the bias.

The difference between the periodogram and the modified periodogram can be explained as follows. The original periodogram is the DTFT

$$\hat{P}_{per}(e^{j\omega}) = \sum_{n=-\infty}^{\infty} w_R[n]x[n] e^{j\omega n},$$

where

$$w_R[n] = \begin{cases} 1 & \text{for } 0 \le n \le N-1 \\ 0 & \text{otherwise.} \end{cases}$$

This rectangular window is introduced as one way to describe that we have only observed x[n] for n = 0, 1, ..., N - 1. One problem with $w_R[n]$ is that its Fourier transform, $W_R(e^{j\omega})$, has very large sidelobes which give rise to the frequency masking mentioned above. The modified periodogram reduces the sidelobes by replacing $w_R[n]$ with a different window function that has lower sidelobes (but unfortunately the alternative windows always have wider mainlobes).