<u>Problem Set 9 – Wiener Filter</u>

Problem 1:

Two independent WSS Gaussian random processes X[n], Z[n], are given. The two processes have zero expectation and the following Power Spectral Densities:

$$S_{X}\left(e^{j\omega}\right) = \frac{|\omega|}{\pi}, -\pi < \omega \leq \pi$$

$$S_{Z}\left(e^{j\omega}\right) = 1 - \frac{|\omega|}{\pi}, -\pi < \omega \leq \pi$$

$$S_{Z}\left(e^{j\omega}\right)$$

$$S_{Z}\left(e^{j\omega}\right)$$

$$S_{Z}\left(e^{j\omega}\right)$$

We are interested in calculating X[n] out of Y[n] = X[n] + Z[n].

- 1. Are X[n] and Y[n] JWSS?
- 2. Calculate the MMSE estimator of the process X[n] from the process Y[n].
- 3. What is the mean squared error (MSE) of the estimator from section 2?

We define the processes $Y'[n] = (-1)^n X[n] + Z[n]$ and $Y''[n] = (-1)^n Y'[n]$

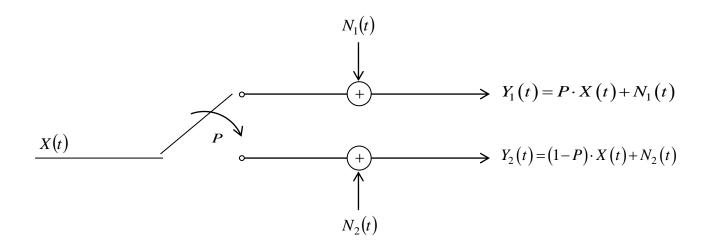
- 4. Are X[n] and Y'[n] JWSS?
- 5. What is the MMSE (not necessarily linear) estimator of the process X[n] from the process Y'[n]?
- 6. Are X[n] and Y[n] JWSS?
- 7. What is the MMSE (not necessarily linear) estimator of the process X[n] from the process Y[n]?

Problem 2:

Given are three independent random processes X(t), $N_1(t)$ and $N_2(t)$.

The three processes are stationary, with expectation zero and spectrums $S_X(\omega)$, $S_1(\omega)$ and $S_2(\omega)$, respectively. Also a random variable P is given, independent of the three processes X(t), $N_1(t)$ and $N_2(t)$, with the following distribution:

$$P = \begin{cases} 1 & w.p \ p \\ 0 & w.p \ 1-p \end{cases} \qquad 0 \le p \le 1$$



- 1. Show that the three processes X(t), $Y_1(t)$ and $Y_2(t)$ are JWSS in pairs (meaning that each pair of them is JWSS).
- 2. Calculate $\hat{X}_1(t)$, which is the optimal linear MMSE estimator of the process X(t) from the process $Y_1(t)$.
- 3. Calculate $\hat{X}_2(t)$, which is the optimal linear MMSE estimator of the process X(t) from the process $Y_2(t)$.

We define the estimator $\hat{X}(t) = \hat{X}_1(t) + \hat{X}_2(t)$ and estimation error $\varepsilon(t) = X(t) - \hat{X}(t)$

- 4. Show that the cross-correlation between $\varepsilon(t)$ and any one of the two processes $Y_1(t)$, $Y_2(t)$ is zero.
- 5. Deduce that $\hat{X}(t)$ is the optimal linear estimator of the process X(t) from both processes $\{Y_1(t), Y_2(t)\}_{t=-\infty}^{t=\infty}$.

Problem 3:

Two JWSS random processes $N_1(t), N_2(t)$ are given, each with expectation zero, autocorrelation and cross-correlation functions $R_{N_1}(\tau), R_{N_2}(\tau), R_{N_1,N_2}(\tau)$ and spectrums $S_{N_1}(\omega), S_{N_2}(\omega), S_{N_1,N_2}(\omega)$, respectively.

1. Find the optimal linear estimator $H(\omega)$ that uses all the sample signal $N_2(t)$:

$$\hat{N}_1(t_0) = \int_{-\infty}^{\infty} h(t_0 - t) \cdot N_2(t) \cdot dt$$

Now, it is given that $N_1(t) = V(t) * g_1(t) N_2(t) = V(t) * g_2(t)$, where V(t) is a WSS process with expectation zero and spectrum:

$$S_{V}(\omega) = \begin{cases} a & |\omega| \le 2\\ 0 & |\omega| > 2 \end{cases}$$

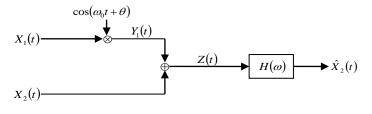
The frequency responses of the filters are given by:

$$G_{1}(\omega) = \begin{cases} 1 & |\omega| \leq \omega_{1} \\ 0 & |\omega| > \omega_{1} \end{cases}, G_{2}(\omega) = \begin{cases} 0 & |\omega| \leq \omega_{2} \\ 1 & |\omega| > \omega_{2} \end{cases}$$

- 2. Explain why indeed $N_1(t), N_2(t)$ are JWSS with expectation zero and calculate $S_{N_1}(\omega), S_{N_2}(\omega), S_{N_1,N_2}(\omega)$.
- 3. Calculate the estimator you found in section 1 and its mean squared error.
- 4. For what values of the parameters ω_1, ω_2 is the estimation error zero?
- 5. For what values of the parameters ω_1, ω_2 is the estimation error maximal? What is the estimator in this case and what is the estimation error?

Problem 4:

Consider the following system:



The system in the figure represents a communication channel designated to transmit two signals, $X_1(t)$ and $X_2(t)$, at the same time. It is assumed that $X_1(t)$ and $X_2(t)$ are independent random processes, wide-sense stationary with power spectrum:

$$S_{X_2}(\omega) = e^{-|\omega|}$$
 $S_{X_1}(\omega) = \begin{cases} P & |\omega| \le 1 \\ 0 & o.w. \end{cases}$

Assume that $\omega_0 = 1.5$ and that θ is a random variable that distributes uniformly in the section $(-\pi,\pi)$ and is independent of $X_1(t)$ and $X_2(t)$.

In order to reproduce the signal $X_2(t)$, an LPF is used in the receiver, whose transfer function is:

$$H(\omega) = \begin{cases} 1 & |\omega| < B \\ 0 & O.W. \end{cases}$$

This creates the signal $\hat{X}_2(t)$.

- 1. Calculate and draw the power spectrum of Z(t).
- 2. Find the power spectrum of $\hat{X}_2(t)$, using the parameter B (the bandwidth of the filter in the receiver).

- 3. Show that $S_{X,\hat{X}_2}(\omega) = S_{X_2}(\omega)H(\omega)$.
- 4. Draw the power spectrum of the reproduction error process $e(t) = X_2(t) \hat{X}_2(t)$ for each of the cases: B = 0.25, B = 2, B = 3.
- 5. Assuming $R_{X_1}(0) = \frac{4}{\pi e}$, find B such that the mean squared error in the reproduction, $E\{e^2(t)\}$, is minimal.
- 6. Find the Wiener filter for the estimation of $X_2(t)$ from Z(t).

Problem 5:

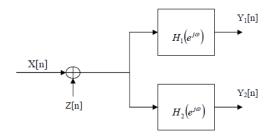
Given is the linear AR random process X[n], which is WSS:

$$X[n] = 0.5X[n-1] + W[n]$$

where the innovations process W[n] is i.i.d with expectation 0 and variance 1. An i.i.d noise Z[n], independent of W[n], with expectation 0 and variance 1 is added to X[n]. After the noise addition, the signal is passed in parallel through two filters that satisfy:

$$h_1[n] = -0.5\delta[n] + \delta[n-1]$$

 $H_2(e^{jw}) = \frac{1}{H_1(e^{jw})}$



- 1. Find $S_{Y_2}\left(e^{j\omega}\right)$, the spectrum of the process $Y_2[n]$, and the cross-spectrum between $Y_2[n]$ and the innovations process W[n], $S_{Y_2W}\left(e^{j\omega}\right)$ (Hint: represent the process X[n] as a transition of the process W[n] through an LTI system which will be marked as $H_0\left(e^{j\omega}\right)$.)
- 2. Calculate the optimal linear estimator of the process W[n] from the $Y_2[n]$.
- 3. Prove that the expectation of the squared error of the optimal linear estimator of the process W[n] from the process $Y_1[n]$ is identical to the expectation of the squared error of the estimator found in section 2.

(**Hint:** it can be proved without finding the estimator explicitly.)