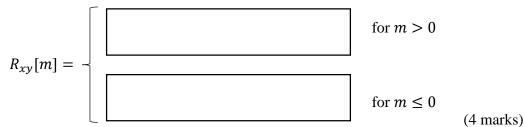
## EE7401 Homework 3

## **Instructions:**

- Please write your final solutions in the boxes shown below, then scan (if needed) and upload via NTULearn/Assignments/Homework3 by 23:59, 14 November 2023.
- Please do not copy from others, or let others copy your solution.
- Penalties may apply for late submission.
- Homework 3 carries 10% overall marks.
- **Q1**) When did you start your simulations: Consider the simulation of a first order autoregressive, or AR(1), process. A wide-sense stationary real white noise process  $\mathbf{x}[n]$  with autocorrelation  $R_{xx}[m] = 5\delta[m]$  is passed through the AR(1) filter with a = 0.5, such that the output is  $\mathbf{y}[n] = \mathbf{x}[n] + 0.5\mathbf{y}[n-1]$ .

$$\mathbf{x}[n] \longrightarrow H(z) = \frac{1}{1 - 0.5z^{-1}} \longrightarrow \mathbf{y}[n]$$

**1.1**) (1.1.a) Assume the above filter operates at all times. Then  $\mathbf{y}[n]$  is jointly wide-sense stationary. Find the cross-correlation between  $\mathbf{x}[n]$  and  $\mathbf{y}[n]$ ,  $R_{xy}[m] = E\{\mathbf{x}[n_1]\mathbf{y}[n_2]\}$ , where  $m = n_1 - n_2$ . [Hint: There are many ways to solve this problem. You may use the frequency domain or the z domain approach, finding  $S_{xy}(\omega)$  or  $S_{xy}(z)$  first, and then taking the inverse Fourier or z transform. You may express  $\mathbf{y}[n]$  as a convolution of  $\mathbf{x}[n]$  with h[n] (impulse response), multiply the equation by  $\mathbf{x}[n_1]$ , and take the expectation.]



(1.1.b) Find the autocorrelation of the AR(1) process,  $R_{yy}[m]$ .

$$\mathbf{x}[n] \longrightarrow H(z) = \frac{1}{1 - 0.5z^{-1}} \longrightarrow \mathbf{y}[n]$$

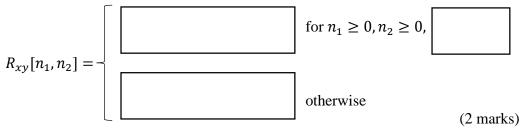
- 1.2) Part (1.1) results are true if the simulation is started at  $n = -\infty$ . However, real-life simulations start at finite times. Consider a real-life simulation of the same AR(1) process as above, where the simulation starts at n = 0.  $\mathbf{x}[n]$  is still the same wide-sense stationary with the same autocorrelation. However, the AR(1) filter starts at n = 0 (meaning, there was no filter before n = 0, or the output was zero). Therefore, the output becomes  $\mathbf{y}[n] = \begin{cases} 0 & n < 0 \\ \mathbf{x}[n] + 0.5\mathbf{y}[n-1] & n \geq 0 \end{cases}$ . Note that  $\mathbf{y}[n]$  is no longer stationary. Therefore, the auto/cross-correlations involving  $\mathbf{y}[n]$  no longer depend on the time difference m but depend on both times, like  $R_{yy}[n_1, n_2]$ . As a result, the power spectrums of  $\mathbf{y}[n]$  do not exist, and the power spectrum based approaches can no longer be used to find the auto/cross-correlations. The time-domain approach may still be used.
- (1.2.a) Express  $\mathbf{y}[0]$  using only the input  $\mathbf{x}[0]$ . Express  $\mathbf{y}[1]$  as a sum of only input terms of the form  $\mathbf{x}[k]$ . There should not be any past output term such as  $\mathbf{y}[n-1]$ . Continuing as above, express  $\mathbf{y}[n_2]$  for any  $n_2 \ge 0$  as a sum of only input terms of the form  $\mathbf{x}[k]$ . There should not be any past output term such as  $\mathbf{y}[n_2-1]$ .

$$\mathbf{y}[n_2] = \boxed{ \text{for } n_2 \ge 0 \dots \text{eq.}(1) }$$
(3 marks)

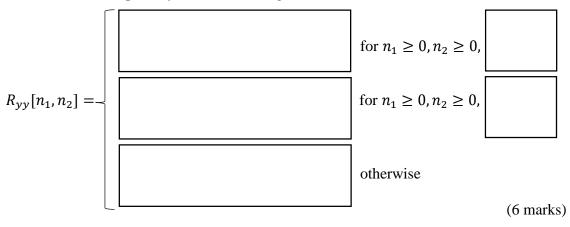
(1.2.b) Find the cross-correlation between  $\mathbf{x}[n]$  and  $\mathbf{y}[n]$ ,  $R_{xy}[n_1, n_2] = E\{\mathbf{x}[n_1]\mathbf{y}[n_2]\}$ , for all  $n_1, n_2 \ge 0$  by multiplying eq.(1) by  $\mathbf{x}[n_1]$ , and taking the expectation of both sides. [Hint: Find 2 cases: write answers in 2 left boxes, and write the cases in 2 right boxes.]

$$R_{xy}[n_1,n_2] = \left\{ \begin{array}{|c|c|c|c|} & \text{for } n_1 \geq 0, n_2 \geq 0, \\ & & \text{for } n_1 \geq 0, n_2 \geq 0, \\ & & & \text{(4 marks)} \end{array} \right.$$

(1.2.c) Extend your result of (1.2.b) to all possible  $n_1$ ,  $n_2$  values. [Hint: Find 2 cases: write answers in the 2 left boxes, and write the condition of the primary case in the right box.]



(1.2.d) Find the autocorrelation of the AR(1) process,  $R_{yy}[n_1, n_2]$  for all possible  $n_1, n_2$ values. [Hint: Replace both  $y[n_1]$  and  $y[n_2]$  in  $E\{y[n_1]y[n_2]\}$  by eq.(1) twice, then take the expectation of this double summation. You should find 3 cases (it is possible to combine 2 primary cases into a single case): write answers in the 3 left boxes, and write the conditions of 2 primary cases in the 2 right boxes.]



(1.2.e) For all non-negative  $n_1$  and  $n_2$ , is  $R_{xy}[n_1, n_2]$  of (1.2.c) equal to  $R_{xy}[m]$  of (1.1.a)? If not, when are they equal?

For all non-negative  $n_1$  and  $n_2$ , is  $R_{yy}[n_1, n_2]$  of (1.2.d) equal to  $R_{yy}[m]$  of (1.1.b)? If not,

when are they equal?	
	(4 marks

- **Q2**) Bandlimited process does change with time: In lecture we have upper bounded the change in value of a bandlimited process over a small time  $\tau$ . Here we obtain a lower bound (and a new upper bound). First, two intermediate results are obtained as below.
- **2.1**) Assume that the time is bounded by  $|\tau| < (\pi/\sigma)$  and that the frequency is bounded by  $|\omega| \le \sigma$ . Then, using the fact that if  $0 < \varphi < (\pi/2)$ , then  $(2\varphi/\pi) < \sin \varphi < \varphi$ , find a lower bound and an upper bound on  $\sin^2(\omega \tau/2)$ :

$$\leq \sin^2(\omega\tau/2) \leq \tag{4 marks}$$

**2.2**) Let the power spectrum of  $\mathbf{x}(t)$  be  $S_{xx}(\omega)$ . If  $\mathbf{x}(t)$  is passed through a differentiator (frequency response  $H(\omega) = j\omega$ ), then you have already found  $S_{x'x'}(\omega)$ , the power spectrum of the output  $\mathbf{x}'(t)$ , in terms of  $S_{xx}(\omega)$ , in Homework 2 Q(2.b). Copy this  $S_{x'x'}(\omega)$  below to obtain the autocorrelation of the output  $\mathbf{x}'(t)$  as

$$R_{x'x'}(\tau) = \int_{-\infty}^{\infty} S_{xx}(\omega)e^{j\omega\tau}d\omega/2\pi$$

Now, putting  $\tau = 0$ , the average power of the output  $\mathbf{x}'(t)$  is:

$$E\{|\mathbf{x}'(t)|^2\} = \int_{-\infty}^{\infty} S_{xx}(\omega)d\omega/2\pi$$
(2 marks)

**2.3**) (2.3.a) Express the expectation of the square of the change in  $\mathbf{x}(t)$  over time  $\tau$ ,  $E\{|\mathbf{x}(t+\tau)-\mathbf{x}(t)|^2\}$ , using its autocorrelation:

$$E\{|\mathbf{x}(t+\tau) - \mathbf{x}(t)|^2\} = 2R_{xx}(\boxed{) - R_{xx}(\boxed{)} - R_{xx}(\boxed{)}$$
(5 marks)

(2.3.b) Use (2.3.a),  $R_{xx}(\tau) = \int_{-\infty}^{\infty} S_{xx}(\omega) e^{j\omega\tau} \frac{d\omega}{2\pi}$ , and  $1 - \cos\theta = 2\sin^2(\theta/2)$ , to get:

$$E\{|\mathbf{x}(t+\tau) - \mathbf{x}(t)|^2\} = \int_{-\infty}^{\infty} S_{xx}(\omega)$$

$$(4 \text{ marks})$$

(2.3.c) Let us say the integral you obtained in part (2.3.b) is  $\int_{-\infty}^{\infty} S_{xx}(\omega)g(\omega)d\omega/2\pi$  for some function  $g(\omega)$  that you wrote inside the box. Now, since  $\mathbf{x}(t)$  is bandlimited, its power spectrum  $S_{xx}(\omega) = 0$  for  $|\omega| > \sigma$ . Therefore, this integral's limits may be changed,  $E\{|\mathbf{x}(t+\tau) - \mathbf{x}(t)|^2\} = \int_{-\infty}^{\infty} S_{xx}(\omega)g(\omega)d\omega/2\pi = \int_{-\sigma}^{\sigma} S_{xx}(\omega)g(\omega)d\omega/2\pi$ . Apply the lower and upper bounds on  $\sin^2(\omega\tau/2)$  from part (2.1) to obtain the lower and upper bounds on the expectation:

$$\int_{-\sigma}^{\sigma} S_{xx}(\omega) \, \omega^2 \boxed{\frac{d\omega}{2\pi}} \le E\{|\mathbf{x}(t+\tau) - \mathbf{x}(t)|^2\} \le \int_{-\sigma}^{\sigma} S_{xx}(\omega) \, \omega^2 \boxed{\frac{d\omega}{2\pi}}$$
(4 marks)

(2.3.d) The difficulty is that, unlike in the lecture, we can't replace  $\omega$  by  $\sigma$  in the lower bound. Therefore, we need to evaluate  $\int_{-\sigma}^{\sigma} S_{xx}(\omega) \omega^2 d\omega/2\pi$ . This has already been done in part (2.2) using the differentiated process  $\mathbf{x}'(t)$ . Use the result of (2.2) on the lower and upper bounds of (2.3.c) to obtain the final result:

$$E\{|\mathbf{x}'(t)|^2\} \le E\{|\mathbf{x}(t+\tau) - \mathbf{x}(t)|^2\} \le E\{|\mathbf{x}'(t)|^2\}$$
(2 marks)

(2.3.e) Is the upper bound of (2.3.d) smaller, or larger, than the upper bound found in the class?

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(4 marks)