

**Note:** Problems (or parts of problems) marked with a star (★) are required for graduate students to receive 4 credit hours; undergraduate students who solve these problems will receive extra credit points.

**Submission:** Write your name, netid, and u for undergrad/G for grad in the upper right-hand corner of the first page of your written solutions. Typewritten solutions will receive 5 extra credit points.

### Problems to be handed in

1 Let  $X = (X_t)_{t \in \mathbb{R}}$  be a weakly stationary stochastic signal with zero mean and autocorrelation function  $R_X(\tau) = \sigma^2 e^{-|\tau|}$ . Let  $Y$  be the stochastic signal obtained from  $X$  via

$$Y_t = \frac{1}{2t_0} \int_{t-t_0}^{t+t_0} X_t dt,$$

where  $t_0 > 0$  is a fixed window size.

- Prove that  $Y$  is also weakly stationary.
- Find the mean  $m_Y$  and the power spectral density  $S_Y$ .

2 In this problem, we look at the Doppler effect due to a signal source moving with random velocity. For simplicity, we assume that the motion takes place in one dimension. Consider an observer located at the origin and a signal source initially located at some fixed point at distance  $r$  from the origin. The source emits a complex exponential signal with amplitude  $a$  and angular frequency  $\omega_0$ ; the stochastic signal received by the observer can be represented by the complex exponential

$$X_t = a \exp \left[ i \omega_0 \left( t - \frac{r + Vt}{c} \right) \right],$$

where  $c > 0$  is the speed of signal propagation. Here, the velocity  $V$  is a random variable with pdf  $f_V$ . The autocorrelation function for a complex-valued stochastic signal  $X$  is defined as  $R_X(s, t) \triangleq \mathbf{E}[X_s^* X_t]$ , where the star denotes complex conjugation. Prove  $R_X(t, t + \tau)$  depends only on  $\tau$  and compute its “power spectral density”  $S_X$ . The quotes around “power spectral density” are necessary because the mean of  $X_t$  changes with  $t$ , so  $X$  is nonstationary. Nevertheless, since in this case  $R_X(t, t + \tau) = R_X(\tau)$ , the Fourier inversion formula gives

$$\mathbf{E}[|X_t|^2] = R_X(0) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) d\omega$$

for any  $t$ , so it still makes sense to think about  $S_X$  as the power spectral density of  $X$ .

*Hint:* The formula for  $S_X$  should involve  $a$ ,  $c$ ,  $\omega_0$ , and  $f_V$ .

3 In this problem, we will explore some properties of jointly Gaussian random variables.

(a) Recall that the characteristic function of a scalar random variable  $X$  is given by

$$\Phi_X(u) = \mathbf{E}[e^{iuX}], \quad u \in \mathbb{R}.$$

and that the joint characteristic function of a random vector  $X = (X_t)_{t \in \{1, \dots, n\}}$  is given by

$$\Phi_{X_1, \dots, X_n}(u_1, \dots, u_n) = \mathbf{E}[e^{i(u_1 X_1 + \dots + u_n X_n)}], \quad u_1, \dots, u_n \in \mathbb{R}.$$

Assume that  $X_1, \dots, X_n$  have a joint pdf  $f_{X_1, \dots, X_n}$ . We say that  $X_1, \dots, X_n$  are independent random variables if

$$f_{X_1, \dots, X_n}(x_1, \dots, x_n) = f_{X_1}(x_1)f_{X_2}(x_2)\dots f_{X_n}(x_n),$$

where  $f_{X_i}$  denotes the marginal pdf of  $X_i$ . Prove that  $X_1, \dots, X_n$  are independent if and only if

$$\Phi_{X_1, \dots, X_n}(u_1, \dots, u_n) = \Phi_{X_1}(u_1)\Phi_{X_2}(u_2)\dots \Phi_{X_n}(u_n)$$

for all  $u_1, \dots, u_n \in \mathbb{R}$ .

(b) We say that  $X_1, \dots, X_n$  are uncorrelated if the covariance matrix  $C_X$  is diagonal, i.e., if  $C_X(s, t) = \mathbf{E}[X_s X_t] - \mathbf{E}[X_s]\mathbf{E}[X_t] = 0$  for  $s \neq t$ . In general, uncorrelated random variables can still be dependent. Use the result from part (a) to prove that if  $X_1, \dots, X_n$  are uncorrelated and jointly Gaussian, then they are independent.

(c) Let  $X$  be a Gaussian random vector. In class, we have proved that the projection  $a^T X$  of  $X$  onto any deterministic vector  $a = (a_1, \dots, a_n)^T \in \mathbb{R}^n$  is Gaussian. Now consider an  $m \times n$  matrix  $A = (A_{ij})_{i,j \in \{1, \dots, n\}}$  and form the random vector  $Y = AX$ . Prove that  $Y$  is also a Gaussian random vector.

(d) (★) Let  $X$  be a Gaussian random variable with mean 0 and variance  $\sigma^2$ . Let  $U$  be a Rademacher random variable (i.e.,  $\mathbf{P}[U = \pm 1] = \frac{1}{2}$ ) independent of  $X$ . Prove that  $Y = UX$  is also Gaussian with mean 0 and variance  $\sigma^2$ , but  $X$  and  $Y$  are *not* jointly Gaussian.

*Hint:* Consider the sum  $X + Y$ .

4 Let  $X$  be a zero-mean stationary Gaussian stochastic signal. Compute the crosscorrelation function  $R_{XY}(\tau)$  between  $X$  and  $Y$ , where  $Y_t = g(X_t)$  with the following choices for  $g$ :

- The full-wave rectifier  $g(x) = |x|$ .
- The power-law detector  $g(x) = x^p$  for  $p \in \mathbb{N}$ .
- The gating function  $g(x) = u(t+1) - u(t-1)$ .