

**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 Consider the RC circuit shown in Figure 1 of the lecture on the Johnson–Nyquist noise. Suppose that the noisy voltage source  $E$  is bandlimited, i.e., its power spectral density has the form

$$S_E(\omega) = \begin{cases} 2kTR, & |\omega| \leq \omega_0 \\ 0, & \text{otherwise} \end{cases},$$

where  $\omega_0$  is the bandwidth in rad/s.

- Compute the average power  $\mathbf{E}[V_t^2]$  of the voltage across the capacitor.
- Let  $\beta$  denote the ratio of the noise bandwidth  $\omega_0$  to the RC circuit bandwidth  $1/RC$ , i.e.,  $\beta = \omega_0 RC$ . How does the expression from part (a) behave in the limit  $\beta \rightarrow \infty$ ?

2 In the lecture on  $1/f$  noise, we encountered a continuous-time stochastic signal  $X = (X_t)_{t \in \mathbb{R}}$  taking values in  $X = \{-1, +1\}$ , with the probabilities  $p_t(-) := \mathbf{P}[X_t = -1]$  and  $p_t(+1) := \mathbf{P}[X_t = +1]$  obeying the differential equation

$$\frac{d}{dt} \begin{pmatrix} p_t(-) & p_t(+1) \end{pmatrix} = \begin{pmatrix} p_t(-) & p_t(+1) \end{pmatrix} \begin{pmatrix} -\alpha & \alpha \\ \alpha & -\alpha \end{pmatrix}.$$

By diagonalizing the matrix on the right, show that, for any  $t \in \mathbb{R}$  and any  $\tau \geq 0$ , the solution is given by

$$\begin{pmatrix} p_{t+\tau}(-) & p_{t+\tau}(+) \end{pmatrix} = \begin{pmatrix} p_t(-) & p_t(+1) \end{pmatrix} \begin{pmatrix} \frac{1+e^{-2\alpha\tau}}{2} & \frac{1-e^{-2\alpha\tau}}{2} \\ \frac{1-e^{-2\alpha\tau}}{2} & \frac{1+e^{-2\alpha\tau}}{2} \end{pmatrix}.$$

3 In the lecture on  $1/f$  noise, we have considered the probability distribution of the relaxation time  $T_0 = ce^{\Delta E/kT}$ , where  $c$  is a positive constant,  $k$  is the Boltzmann constant,  $T$  is the ambient temperature in Kelvin, and  $\Delta E$  is a random energy gap distributed uniformly on the interval  $[\Delta_0, \Delta_1]$ .

- Prove that the pdf of  $T_0$  is given by

$$g(t_0) = \begin{cases} \frac{kT}{(\Delta_1 - \Delta_0)t_0}, & ce^{\Delta_0/kT} \leq t_0 \leq ce^{\Delta_1/kT} \\ 0, & \text{otherwise} \end{cases}.$$

- Compute the mean and the variance of  $T_0$ .

4 (★) Let  $N = (N_t)_{t \geq 0}$  be a Poisson process with rate  $\lambda$ , and  $T = (T_k)_{k \in \mathbb{Z}_+}$  be the arrival times of  $N$  (with  $T_0 = 0$ ). Let  $M$  be a given  $n \times n$  Markov matrix. Consider a continuous-time stochastic signal  $X = (X_t)_{t \geq 0}$  with finite state space  $\mathcal{X} = \{0, \dots, n-1\}$  that evolves as follows: it starts from  $X_0 = 0$  and stays the same until the next arrival, at which point it changes randomly to a different state with probabilities prescribed by  $M$ . That is,  $X_t = 0$  for  $t < T_1$ ; then at  $t = T_1$ ,  $X_t = y$  with probability  $M(X_0, y) = M(0, y)$ , for each  $y \in \mathcal{X}$ . Then the state  $X_t$  stays the same until  $t = T_2$ , at which point it changes randomly to a new state  $y'$  with probability  $M(X_{T_1}, y')$ , etc.

- (a) Prove that  $X$  is a Markov process.
- (b) Let  $p_t$  denote the probability distribution of  $X_t$ , i.e.,  $p_t(x) = \mathbf{P}[X_t = x]$  for each  $x \in \mathcal{X}$ . Prove the following explicit formula for  $p_t$ :

$$p_t = p_0 e^{\lambda t(M - I_n)},$$

where  $I_n$  is the  $n \times n$  identity matrix,  $p_0$  is the initial state distribution (in this case,  $p_0(x) = 1$  if  $x = 0$  and 0 otherwise), and the matrix exponential  $e^A$  for a square matrix  $A$  is defined as

$$e^A \triangleq \sum_{k=0}^{\infty} \frac{A^k}{k!}.$$

*Hint:* Use the fact that the number of state transitions between times 0 and  $t$  is equal to  $N_t$ , the number of arrivals by time  $t$ , then apply the law of total probability.

- (c) Consider the binary case  $\mathcal{X} = \{0, 1\}$  with  $M(0, 0) = M(1, 1) = \frac{1}{2}$ . Compute the matrix  $e^{\lambda t(M - I_2)}$  explicitly. What can you say about the long-term behavior of  $p_t$  – i.e., will it converge to a limiting distribution, and, if the answer is “yes,” how fast is the convergence?
- (d) How does this relate to the continuous-time Markov process we have used to model conductance fluctuations in our analysis of  $1/f$  noise?