**Note:** Problems marked with a star ( $\star$ ) 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 (Probability review)
  - (i) Let  $(X, Y) \in \mathbb{R}^2$  be a point chosen uniformly at random from the unit disk

$$\mathbb{D} := \left\{ (x, y) \in \mathbb{R}^2 : x^2 + y^2 \le 1 \right\}.$$

Compute the marginal pdf  $f_X$  of X and the expected value  $\mathbf{E}[\sqrt{X^2 + Y^2}]$ .

- (ii) Let X be a real-valued random variable with finite second moment:  $\mathbf{E}[X^2] < \infty$ . Consider the function  $f(c) = \mathbf{E}[(X c)^2]$ . Find the value of c that minimizes f and compute the corresponding minimum value of f.
- (iii) Two real-valued random variables U and V are called uncorrelated if

$$Cov(U, V) := E[(U - E[U])(V - E[V])] = 0.$$

Let  $U = \sin X$  and  $V = \cos X$ , where  $X \sim \text{Unif}(-\pi, \pi)$ . Are *U* and *V* uncorrelated? Are they independent? Show all your work and justify all your answers!

2 In class, we saw how to generate a Bernoulli(*p*) random variable from a random variable *U* uniformly distributed on the unit interval [0,1]: if  $0 \le U < p$ , output 1; else, if  $p \le U < 1$ , output 0.

(i) Modify the above procedure for the case when, instead of  $U \sim \text{Uniform}(0, 1)$ , we can generate a random variable *V* with cdf given by

$$F_V(a) = \begin{cases} 0, & a < 0\\ a^2, & 0 \le a < 1 \\ 1, & a \ge 1 \end{cases}$$

*Hint*: Can you find a function  $f : \mathbb{R} \to \mathbb{R}$ , such that  $f(V) \sim \text{Uniform}(0, 1)$ ?

- (ii) Now suppose that we wish to generate a random variable *X* taking values in the finite set  $\{1, ..., n\}$  with probabilities  $p_i := \mathbf{P}[X = i]$ . Construct a procedure that takes  $U \sim \text{Uniform}(0, 1)$  as the input and generates a sample of *X* as the output.
- (iii) Finally, modify your procedure from part (ii) to work with input V from part (i).

**3** Consider a Markov chain  $X = (X_t)_{t \in \mathbb{Z}_+}$  with the ternary state space  $X = \{1, 2, 3\}$  and with the transition probability matrix

$$M = \begin{pmatrix} \frac{2}{3} & \frac{1}{6} & \frac{1}{6} \\ \frac{1}{6} & \frac{2}{3} & \frac{1}{6} \\ \frac{1}{6} & \frac{1}{6} & \frac{2}{3} \end{pmatrix}$$

This constitutes a declarative description of *X*. Construct an imperative description of *X* in the form  $X_{t+1} = f(X_t, U_t)$ , where the  $U_t$ 's are i.i.d. Uniform(0, 1) random variables.

4 Prove that any discrete-time Markov chain has the following property:

$$\mathbf{P}[X_{t_{n+1}} = y_1, X_{t_{n+2}} = y_2, \dots, X_{t_{n+m}} = y_m | X_{t_1} = x_1, X_{t_2} = x_2, \dots, X_{t_n} = x_n]$$
  
=  $\mathbf{P}[X_{t_{n+1}} = y_1, X_{t_{n+2}} = y_2, \dots, X_{t_{n+m}} = y_m | X_{t_n} = x_n]$ 

for all  $t_1 < t_2 < \ldots < t_n < t_{n+1} < \ldots < t_{n+m}$  and for all states  $x_0, \ldots, x_t, y_1, \ldots, y_m$ . This is why we say that, in a Markov chain, the future is conditionally independent of the past given the present.

*Hint*: Pass to a suitable imperative description of the form  $X_{t+1} = f(X_t, U_t)$  and iterate.

**5** ( $\star$ ) The moment-generating function (MGF) of a real-valued random variable X is defined as  $M(\lambda) := \mathbf{E}[e^{\lambda X}].$ 

- (i) Find the MGF of a Rademacher random variable, i.e.,  $\mathbf{P}[X = -1] = \mathbf{P}[X = +1] = \frac{1}{2}$ .
- (ii) Find the MGF of a Gaussian random variable with mean  $\mu$  and variance  $\sigma^2$ .
- (iii) Use the result of part (ii) to prove the following: If  $X \sim N(\mu, \sigma^2)$ , then, for any t > 0,

$$\mathbf{P}[X \ge \mu + t] \le e^{-t^2/2\sigma^2}.$$

*Hint:* Note that, for any  $\lambda > 0$ ,  $\mathbf{P}[X \ge \mu + t] = \mathbf{P}[e^{\lambda X} \ge e^{\lambda(\mu+t)}]$ .

(iv) Prove the following inequality for any two real-valued random variables X and Y:

$$|\operatorname{Cov}(X, Y)|^2 \le \operatorname{Var}[X] \cdot \operatorname{Var}[Y].$$

*Hint:* Assume  $\mathbf{E}[X] = \mathbf{E}[Y] = 0$  and observe that  $\mathbf{E}[(X + \alpha Y)^2] \ge 0$  for any  $\alpha \in \mathbb{R}$ .