

## Endsem: Probability and Statistics (100 marks)

Instruction:

- Please state reasons wherever applicable.
- Use precise mathematical arguments, no speeches.

### Each question: 10 marks

1. Stochastic simulation: Suppose you want to generate samples from a discrete random variable  $X$  having pmf  $\{p_j, j \geq 0\}$ . Now assume that you have access to samples from another discrete random variable  $Y$  with pmf  $\{q_j, j \geq 0\}$  with the property that  $\frac{p_j}{q_j} \leq c$  for some constant  $c$  and for all  $j$  such that  $p_j > 0$ . The rejection method generates samples of  $X$  as follows.
  - (a) Simulate/Generate the value of  $Y$  with mass function  $\{q_j, j \geq 0\}$ .
  - (b) Generate random number  $U$  which is uniform in the interval  $[0, 1]$ .
  - (c) If  $U < \frac{p_Y}{cq_Y}$ , set  $X = Y$  and stop. Otherwise return to Step (a).

Prove that samples of  $X$  generated using the above algorithm indeed have pmf  $\{p_j, j \geq 0\}$ .

2. MGF: Derive the expression for the MGF of a Gaussian  $\mathcal{N}(\mu, \sigma^2)$  random variable and use the MGF to identify the first and the second moment. Furthermore, using MGF, show that sum of  $n$  independent Gaussian  $\mathcal{N}(\mu, \sigma^2)$  random variables is also a Gaussian random variable. What are the resulting mean and variance parameters ?
3. MLE: Consider a Gaussian random variable  $X$  with unknown mean  $\mu$  and unknown standard deviation  $\sigma$ . Suppose you observe  $k$  iid samples from this random variable which is denoted by  $\mathcal{D} = \{x_1, x_2, \dots, x_k\}$ . Find the maximum likelihood estimate for  $\mu$  and  $\sigma$ . Is the MLE for the standard deviation biased ? justify why.
4. Functions of random vectors: Let  $X = [X_1, X_2]$  where  $X_1$  and  $X_2$  are independent exponential random variables with parameter 1. Find the probability density function of  $U = [U_1, U_2]$  where  $U_1 = X_1 + X_2$  and  $U_2 = \frac{X_1}{X_1 + X_2}$ .

## Each question: 15 marks

- Bayesian Inference problem: Suppose  $D = \{x_1, \dots, x_n\}$  is a data set consisting of independent samples of a Bernoulli random variable with unknown parameter  $\theta$ , i.e.,  $f(x_i|\theta) = \theta^{x_i}(1-\theta)^{1-x_i}$  for  $x_i \in \{0, 1\}$ . Now assume a uniform  $U[0, 1]$  prior on the unknown parameter  $\theta$ . Obtain an expression for the posterior distribution on  $\theta$ . Using this obtain  $\theta_{MAP}$  and the conditional expectation estimator  $\theta_{CE}$ . (Hint: you may use the fact that  $\int_0^1 \theta^m (1-\theta)^r d\theta = \frac{m!r!}{(m+r+1)!}$ )
- (8 mks) Let  $X_1, X_2, \dots$  be a sequence of random variables with density  $f_{X_n}(x) = \frac{n}{2} e^{-n|x|}$ . Show that  $X_n$  converges to 0 in probability and in distribution. (Do not use the result that convergence in probability implies convergence in distribution)
  - (7 mks) Let  $X_n$  be  $Poisson(n\lambda)$  random variable for  $n = 1, 2, 3, \dots$ . Consider the sequence of random variables  $Y_n = \frac{X_n}{n}$  for  $n = 1, 2, 3, \dots$ . Show that  $Y_n$  converges in mean square sense to  $\lambda$ .
- (8mks): Let  $\mathcal{D} = \{x_1, \dots, x_n\}$  denote i.i.d samples from a uniform random variable  $U[0, a]$  where  $a$  is unknown. Find an  $MLE$  estimate for the unknown parameter  $a$ .
  - (7mks): Let  $\mathcal{D} = \{x_1, \dots, x_n\}$  denote i.i.d samples from a Poisson random variable with unknown parameter  $\gamma$ . Find an  $MLE$  estimate for the unknown parameter  $\gamma$ .
- (7mks): Consider a discrete time Markov chain with the following transition probabilities:  $p_{ij} = 0$  when  $i = j$ . What is the probability of head and tail in the  $n$ th step and when the initial distribution is  $\mu = [\mu_1, \mu_2]$ .
  - (8mks): Find the limiting distribution and the stationary distribution  $\pi$  for Markov Chain with the following transition probability matrix
$$P = \begin{bmatrix} 0.5 & 0.5 & 0 & 0 \\ 0.5 & 0.5 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$
Now obtain the values of  $F_{ii}$  (probability of ever returning to state  $i$ , having started in state  $i$ ) for each of the 4 states and based on the values identify if each state is transient or recurrent.