## E1 222 Stochastic Models and Applications Problem Sheet 3.3

(You need not submit solutions)

1. Consider a communication system. Let Y denote the bit sent by transmiter. (Y is a binary random variable). The receiver makes a measurement, X, and based on its value decides what is sent. The decision at the receiver can be represented by a function  $h: \Re \to \{0, 1\}$ . For any specific h, let  $R_0(h)$  represent the set of all  $x \in \Re$  for which h(x) = 0 and let  $R_1(h)$  represent the set of  $x \in \Re$  for which h(x) = 1. An error occurs if a wrong decision is made. Argue that the event of error is:  $[h(X) = 0, Y = 1] \cup [h(X) = 1, Y = 0]$ . Show that probability of error for a decision rule h is

$$\int_{R_0(h)} p_1 f_{X|Y}(x|1) dx + \int_{R_1(h)} p_0 f_{X|Y}(x|0) dx$$

where  $p_i = P[Y = i]$ . Now consider a h given by

$$h(x) = 1$$
 if  $f_{Y|X}(1|x) \ge f_{Y|X}(0|x)$ 

(Otherwise h(x) = 0). Show that this h would achieve minimum probability of error.

- 2. Let X, Y have a joint distribution that is uniform over the quadrilateral with vertices at (-1,0), (1,0), (0,-1) and (0,1). Find P[X>Y]. Are X, Y independent? (Hint: Can you decide on independence here without calculating the marginal densities?)
- 3. Let X, Y be *iid* uniform over (0, 1). Let  $Z = \max(X, Y)$  and  $W = \min(X, Y)$ . Find the density of Z W.
- 4. Let X, Y be iid exponential random variables with mean 1. Let Z = X + Y and W = X Y. Find the conditional density  $f_{W|Z}$
- 5. Let X, Y be independent Gaussian random variables with mean zero and variance unity. Define random variables D and  $\theta$  by

$$D = X^2 + Y^2; \quad \theta = \tan^{-1}(Y/X)$$

(where, by convention, we assume  $\theta$  takes values in  $[0, 2\pi]$ ; for this we first calculate  $\tan^{-1}(|Y|/|X|)$  in the range  $[0, \pi/2]$  and then put

that angle in the appropriate quadrant based on signs of Y and X). Find the joint density of D and  $\theta$  and their marginal densities. Are D and  $\theta$  independent? (Hint: Note that the  $(D,\theta)$  to (X,Y) mapping is invertible. Hence you can use the formula).

- 6. Consider the following algorithm for generating random numbers X and Y:
  - 1. Generate  $U_1$  and  $U_2$  uniform over [0, 1].

2. Set 
$$X = \sqrt{-2\log(U_1)}\cos(2\pi U_2)$$
 and  $Y = \sqrt{-2\log(U_1)}\sin(2\pi U_2)$ .

What would be the joint distribution of X and Y? (Hint: Recall that when U is uniform over [0, 1], we know that  $-a \log(U)$  is exponential with parameter 1/a and  $2\pi U$  is uniform over  $[0, 2\pi]$ ).

- 7. Consider the following algorithm for generating random variables  $V_1$  and  $V_2$ :
  - 1. Generate  $X_1$  and  $X_2$  uniform over [-1, 1].
  - 2. If  $X_1^2 + X_2^2 > 1$  then go to step 1; else set  $V_1 = X_1$ ,  $V_2 = X_2$  and exit.

What would be the joint distribution of  $V_1$  and  $V_2$ ?

- 8. Suppose we have access to a random number generator that can generate random numbers uniformly distributed over (0, 1). Using the results of the previous problems, suggest a method for generating samples of X when X has Gaussian density with mean zero and variance unity.
- 9. Let  $p_i, q_i, i = 1, \dots, N$ , be positive numbers such that  $\sum_{i=1}^{N} p_i = \sum_{i=1}^{N} q_i = 1$  and  $p_i \leq Cq_i$ ,  $\forall i$  for some positive constant C. Consider the following algorithm to simulate a random variable, X:
  - 1. Generate a random number Y such that  $P[Y = j] = q_j$ ,  $j = 1, \dots, N$ . (That is, the mass function of Y is  $f_Y(j) = q_j$ ).
  - 2. Generate U uniform over [0, 1].
  - 3. Suppose the value generated for Y in step-1 is j. If  $U < (p_j/Cq_j)$ , then set X = Y and exit; else go to step-1.

On any iteration of the above algorithm, if condition in step-3 becomes true, we say the generated Y is accepted. Find the value of  $P[Y \text{ is accepted} \mid Y = j]$ . Show that  $P[Y \text{ is accepted}, Y = j] = p_j/C$ . Now calculate P[Y is accepted]. Use these to calculate the mass function of X.

10. Suppose X is a discrete rv taking values  $\{x_1, x_2, \dots, x_m\}$  with probabilities  $p_1, \dots p_m$ . The usual method of simulating such a rv is as follows. We divide the [0, 1] interval into bins of length  $p_1, p_2$  etc. Then we generate a rv, uniform over [0, 1] and depending on the bin it falls in, we decide on the value for X. That is, if  $U \leq p_1$  we assign  $X = x_1$ ; if  $p_1 < U \leq p_1 + p_2$  then we assgn  $X = x_2$  and so on. Suppose X is a discrete random variable taking values  $1, 2, \dots, 10$ . Its mass function is:  $f_X(1) = 0.08, f_X(2) = 0.13, f_X(3) = 0.07, f_X(4) = 0.15, f_X(5) = 0.1, f_X(6) = 0.06, f_X(7) = 0.11, f_X(8) = 0.1, f_X(9) = 0.1, f_X(10) = 0.1$ . Can you use the result of previous problem to suggest an efficient method for simulating X.