

## Probability II. Solution 7.

1.(a) The ranges are independent, but the joint pdf cannot be split into  $g(x)h(y)$  for any functions  $g$  and  $h$ . Therefore  $X$  and  $Y$  are not independent.

(b) The ranges are independent and  $f_{X,Y}(x,y) = C(1+x)(1+y) = g(x)h(y)$ , where  $g(x) = C(1+x)$  for  $0 < x < 1$  and  $g(x) = 0$  elsewhere and  $h(y) = (1+y)$  for  $0 < y < 1$  and  $h(y) = 0$  elsewhere. Hence  $X$  and  $Y$  are independent.

Also  $f_X(x) = KC(1+x)$  for  $0 < x < 1$  and  $f_Y(y) = \frac{1}{K}(1+y)$ . Hence, using the result that the p.d.f. must integrate to one, we obtain  $1/K = 2/3$  and  $CK = 2/3$  so that  $K = 3/2$  and  $C = 4/9$  and  $f_X(x) = \frac{2}{3}(1+x)$  for  $0 < x < 1$  and  $f_Y(y) = \frac{2}{3}(1+y)$  for  $0 < y < 1$ .

(c) The ranges for which  $f_{X,Y}(x,y) > 0$  are dependent. Hence  $X$  and  $Y$  are not independent.

(d) The ranges are independent and  $f_{X,Y}(x,y) = g(x)h(y)$  where  $g(x) = xe^{-x}$  for  $0 < x < \infty$  and  $g(x) = 0$  elsewhere and  $h(y) = \frac{C}{y^5}$  for  $1 < y < \infty$  and  $h(y) = 0$  elsewhere. Hence  $X$  and  $Y$  are independent.

Also  $f_X(x) = Kxe^{-x}$  for  $0 < x < \infty$  and  $f_Y(y) = \frac{C}{Ky^5}$  for  $1 < y < \infty$ . Then  $X \sim \text{Gamma}(1, 2)$  so that  $K = 1$ . Also  $1 = \int_1^\infty \frac{C}{y^5} dy = \left[ \frac{-C}{4y^4} \right]_{y=1}^{y=\infty} = \frac{C}{4}$ . Hence  $C = 4$ . Therefore  $f_X(x) = xe^{-x}$  for  $0 < x < \infty$  and  $f_Y(y) = \frac{4}{y^5}$  for  $1 < y < \infty$ .

2. Note that a chi-squared with parameter  $n$  is just a  $\text{Gamma}(1/2, n/2)$ . So  $M_X(t) = (1 - 2t)^{-n/2}$  and  $M_Y(t) = (1 - 2t)^{-m/2}$ . Hence  $M_Z(t) = M_X(t)M_Y(t) = (1 - 2t)^{-n/2}(1 - 2t)^{-m/2} = (1 - 2t)^{-(n+m)/2}$ . Therefore  $Z \sim \chi_{n+m}^2$ . You could also state this result as  $Z \sim \text{Gamma}(1/2, (n+m)/2)$ .

3.  $M_X(t) = (1 - \frac{t}{10})^{-1}$  and  $M_Y(t) = (1 - \frac{t}{10})^{-1}$ .

(a)  $M_U(t) = M_X(t)M_Y(t) = (1 - \frac{t}{10})^{-2}$ . Hence  $U \sim \text{Gamma}(10, 2)$ .

(b)

$$\begin{aligned} M_V(t) &= E[e^{t(X-Y)}] = E[e^{tX}]E[e^{-tY}] = M_X(t)M_Y(-t) \\ &= \left(1 - \frac{t}{10}\right)^{-1} \left(1 + \frac{t}{10}\right)^{-1} = \left(1 - \frac{t^2}{100}\right)^{-1} \end{aligned}$$

The double exponential distribution with parameter  $\theta$  had m.g.f.  $\left(1 - \frac{t^2}{\theta^2}\right)^{-1}$ . Hence  $V$  has double exponential distribution with  $\theta = 10$ . Therefore the p.d.f. is  $f_V(v) = 5e^{-10|v|}$  for  $-\infty < v < \infty$ .

4.  $E[U] = a(E[X] + E[Y]) = a(0 + 0) = 0$ .  $Var(U) = a^2(Var(X) + Var(Y)) = a^2(1 + 1) = 2a^2$ .  $E[V] = b(E[X] + cE[Y]) = b(0 + c0) = 0$ .  $Var(V) = b^2(Var(X) + c^2Var(Y)) = b^2(1 + c^2)$ .  $Cov(U, V) = abVar(X) + abcVar(Y) = ab(1 + c)$ .

Hence we need  $c = -1$  and  $a = b = \frac{1}{\sqrt{2}}$ .

Note that  $M_X(t) = M_Y(t) = e^{t^2/2}$ .

$$M_U(t) = E[e^{ta(X+Y)}] = E[e^{atX}e^{atY}] = M_X(at)M_Y(at) = e^{a^2t^2/2}e^{a^2t^2/2} = e^{t^2/2}$$

This is the m.g.f. for an  $N(0, 1)$  distribution. So by the uniqueness of the m.g.f.  $U \sim N(0, 1)$ .

$$M_V(t) = E[e^{tb(X-Y)}] = E[e^{btX}e^{-btY}] = M_X(bt)M_Y(-bt) = e^{b^2t^2/2}e^{b^2t^2/2} = e^{t^2/2}$$

This is the m.g.f. for an  $N(0, 1)$  distribution. So by the uniqueness of the m.g.f.  $V \sim N(0, 1)$ .