

## Solutions for Week12 Discussion Session

### 6.15

Let  $U$  have a uniform distribution on  $(0, 1)$ .

The distribution function for  $U$  is  $F_U(u) = P(U < u) = u; 0 \leq u \leq 1$ .

For a function  $G$ , we require  $G(U) = Y$  where  $Y$  has distribution function  $F_Y(y) = 1 - e^{-y^2}; y \geq 0$ .

Note that,

$$F_Y(y) = P(Y \leq y) = P(G(U) \leq y) = P(U \leq G^{-1}(y)) = F(G^{-1}(y)) = u.$$

So it must be true that  $G^{-1}(y) = 1 - e^{-y^2} = u$

Hence,  $G(u) = [-\ln(1 - u)]^{-1/2}$

Therefore, the random variable  $Y = [-\ln(1-U)]^{-1/2}$  has distribution function  $F_Y(y)$ .

### 6.20

$Y$  follows uniform  $(0,1)$ .

(a) We have to find the distribution of  $W = Y^2$ . We will use the CDF method.

$$F_W(w) = P(W \leq w) = P(Y^2 < w) = P(Y < \sqrt{w}) = \sqrt{w}.$$

And the range for  $w$  is:  $0 < y < 1 \implies 0 < y^2 < 1 \implies 0 < w < 1$ .

(b) We have to find the distribution of  $W = \sqrt{Y}$ . We will use the CDF method.

$$F_W(w) = P(W \leq w) = P(\sqrt{Y} < w) = P(Y < w^2) = w^2$$

And the range for  $w$  is:  $0 < y < 1 \implies 0 < \sqrt{y} < 1 \implies 0 < w < 1$ .

### 6.23

Given,  $Y$  has pdf  $f(y) = 2(1 - y); 0 \leq y \leq 1$

(a) We have to find the distribution of  $U_1 = 2Y - 1$ . We will use the Jacobian method to find the distribution of  $U_1$ .

$$U_1 = 2Y - 1 \implies Y = \frac{U_1 + 1}{2}$$

Take the first derivative of  $Y$  with respect to  $U_1$ ;  $dy/du_1 = 1/2$ .

Hence, the Jacobian;  $|J| = |dy/du_1| = 1/2$ .

$$f(u_1) = |J|f_{u_1}(y) = \frac{1}{2}2(1 - \frac{U_1 + 1}{2}) = \frac{1 - u_1}{2}$$

Now, for the range of  $u_1$ , we have  $0 \leq y \leq 1 \implies 0 \leq 2y \leq 2 \implies -1 \leq 2y - 1 \leq 1 \implies -1 \leq u_1 \leq 1$ .

Hence,  $f(u_1) = \frac{1 - u_1}{2}; -1 \leq u_1 \leq 1$ .

(b) We have to find the distribution of  $U_2 = 1 - 2Y$ . We will use the Jacobian method to find the distribution of  $U_2$ .

$$U_2 = 1 - 2Y \implies Y = \frac{1 - U_2}{2}$$

Take the first derivative of  $Y$  with respect to  $U_2$ ;  $dy/du_2 = -1/2$ .

Hence, the Jacobian;  $|J| = |dy/du_2| = 1/2$ .

$$f(u_2) = |J|f_{u_2}(y) = \frac{1}{2}2(1 - \frac{1 - U_2}{2}) = \frac{1 + u_2}{2}$$

Now, for the range of  $u_2$ , we have  $0 \leq y \leq 1 \implies 0 \leq 2y \leq 2 \implies 0 \geq -2y \geq -2 \implies 1 \geq 1 - 2y \geq$

$-1 \implies -1 \leq u_2 \leq 1$ .  
Hence,  $f(u_2) = \frac{1+u_2}{2}; -1 \leq u_2 \leq 1$ .

(c) Again, to find the distribution of  $U_3 = Y^2$ . We will use the Jacobian method to find the distribution of  $U_3$ .

$$U_3 = Y^2 \implies Y = \sqrt{U_3}$$

Take the first derivative of Y with respect to  $U_3$ ;  $dy/du_3 = \frac{1}{2\sqrt{u_3}}$ .

Hence, the Jacobian;  $|J| = |dy/du_3| = \frac{1}{2\sqrt{u_3}}$ .

$$f(u_3) = |J|f_{u_3}(y) = \frac{1}{2\sqrt{u_3}}2(1 - \sqrt{u_3}) = \frac{1-\sqrt{u_3}}{\sqrt{u_3}}$$

Now, for the range of  $u_3$ , we have  $0 \leq y \leq 1 \implies 0 \leq y^2 \leq 1 \implies 0 \leq u_3 \leq 1$ .

Hence,  $f(u_3) = \frac{1-\sqrt{u_3}}{\sqrt{u_3}}; 0 \leq u_3 \leq 1$ .

## 6.28

Given, Y follows Uniform(0,1). So,  $f(y) = 1; 0 < y < 1$

We have to find the distribution of  $U = -2\ln(Y)$ . We will use the Jacobian method to find the distribution of U.

$$U = -2\ln(Y) \implies -U/2 = \ln Y \implies Y = \exp(-U/2)$$

Take the first derivative of Y with respect to U;  $dy/du = \exp(-u/2)(-1/2)$ .

Hence, the Jacobian;  $|J| = |dy/du_1| = \exp(-u/2)(1/2)$ .

$$f(u) = |J|f_u(y) = \exp(-u/2)(1/2).1 = \exp(-u/2)(1/2)$$

Now, for the range of u, we have  $-\infty < \ln y < 0 \implies 0 < -2\ln y < \infty \implies 0 < u < \infty$ . [as log is a decreasing function; so as  $y \rightarrow 0$  then,  $\ln(y) \rightarrow \infty$ ]

Hence,  $f(u) = \frac{1}{2}\exp(-u/2); 0 < u < \infty$ .

It is clear that U is now an exponential distribution with mean 2.

## 6.46

Given that, Y has a Gamma distribution with  $\alpha = n/2$  for some positive integer n and  $\beta$  equal to some specified value.

Here we will use method of MGFs to find the distribution of  $W = 2Y/\beta$ .

We know the MGF of Y is:  $M_Y(t) = E(e^{tY}) = (1 - \beta t)^{-\alpha}$

Now, to find the MGF of W is:

$$\begin{aligned} M_W(t) &= E(e^{tW}) \\ &= E(e^{t(2Y/\beta)}) \\ &= E(e^{2tY/\beta}) \\ &= E(e^{t'Y}); \text{ where, } t' = 2t/\beta \\ &= (1 - \beta t')^{-\alpha} \\ &= (1 - \beta(2t/\beta))^{-\alpha} \\ &= (1 - 2t)^{-\alpha} \end{aligned}$$

As  $\alpha = n/2$ , so,  $M_W(t) = (1 - 2t)^{-\alpha} = (1 - 2t)^{-n/2}$ . Which is a MGF of Chi-square distribution. Hence,  $W = 2Y/\beta$  has a  $\chi^2$  distribution with n degrees of freedom.