

# Math 5080

## Homework 2

Magon Bowling

Due: Saturday, February 6th at 11:59 PM

**6.17** Suppose that  $X_1$  and  $X_2$  are  $\text{Gamma}(2, 1/2)$  random variables.

(a) Find the pdf of  $Y = \sqrt{X_1 + X_2}$

To find the distribution of  $Y$ , we use the Moment Generating Function. Recall the MGF of a Gamma distribution is  $M(t) = (1 - \theta t)^{-k}$  for  $t < \frac{1}{\theta}$ . If we let  $Z = X_1 + X_2$ , we get  $Z \sim \text{Gamma}(2, 1)$  for  $Z \in (0, \infty)$  as shown.

$$\begin{aligned} M_Z(t) &= M_{X_1}(t)M_{X_2}(t) \\ &= \left(\frac{2}{2-t}\right)^{\frac{1}{2}} \left(\frac{2}{2-t}\right)^{\frac{1}{2}} \\ &= \left(\frac{2}{2-t}\right) \end{aligned}$$

Now that we know the distribution, we can do a one-dimensional transformation to obtain the pdf of  $Y$ . This is done with  $\phi(x) = \sqrt{x}$ , which is one-to-one and is increasing on  $(0, \infty)$ , so its inverse is  $\phi^{-1}(x) = x^2$ . Now,

$$f_{\phi(x)}(t) = f_X(\phi^{-1}(t)) \left| (\phi^{-1})'(t) \right| \Rightarrow f_{\sqrt{X_1+X_2}}(t) = f_{X_1+X_2}(t^2) |2t| = \frac{e^{-\frac{t^2}{2}}}{2} (2t).$$

Therefore,

$$f_Y(y) = y \exp(-y^2/2) \text{ for } y > 0 \text{ and } 0 \text{ otherwise.}$$

(b) Find the pdf of  $W = X_1/X_2$

I will let  $V = X_1 + X_2$  and use a transformation of a joint pdf to solve for  $f_W(w)$ .

1. Rewrite relations in little variables:  $w = \frac{x_1}{x_2}$  and  $v = x_1 + x_2$
2. Solve for  $x_1, x_2$  in terms of  $v, w$ :

$$w = \frac{x_1}{x_2} \Rightarrow x_1 = wx_2 \Rightarrow x_1 = \frac{vw}{1+w}$$

$$v = x_1 + x_2 \Rightarrow x_2 = v - x_1 \Rightarrow x_2 = v - wx_2 \Rightarrow x_2 = \frac{v}{1+w}$$

3. Compute the Jacobian:

$$J = \det \begin{pmatrix} \frac{\partial x_1}{\partial v} & \frac{\partial x_1}{\partial w} \\ \frac{\partial x_2}{\partial v} & \frac{\partial x_2}{\partial w} \end{pmatrix} = \det \begin{bmatrix} \frac{w}{1+w} & \frac{v(1+w)-vw}{(1+w)^2} \\ \frac{1}{1+w} & \frac{-v}{(1+w)^2} \end{bmatrix} = \frac{-vw}{(1+w)^3} - \frac{v(1+w)-vw}{(1+w)^3} = \frac{v}{(1+w)^2}$$

4. Then the formula for  $f_{(V,W)}$  in terms of  $f_{(X_1,X_2)}$  is  $f_{(V,W)}(v,w) = f_{(X_1,X_2)}(x_1,x_2)|J|$ . Recall the following pdf's:

$$f_{X_1}(x_1) = \frac{e^{-\frac{x_1}{2}}}{\sqrt{2\pi x_1}} \text{ and } f_{X_2}(x_2) = \frac{e^{-\frac{x_2}{2}}}{\sqrt{2\pi x_2}} \Rightarrow f_{X_1,X_2}(x_1,x_2) = \frac{e^{-\frac{(x_1+x_2)}{2}}}{2\pi\sqrt{x_1x_2}}$$

Now through substitution we have the following:

$$\begin{aligned} f_{(V,W)}(v,w) &= \frac{\exp \left[ -\left( \frac{vw}{1+w} + \frac{v}{1+w} \right) / 2 \right]}{2\pi \sqrt{\frac{vw}{1+w} \cdot \frac{v}{1+w}}} \left| \frac{v}{(1+w)^2} \right| \\ &= \frac{\exp \left[ -\left( \frac{v(1+w)}{1+w} \right) / 2 \right]}{2\pi \sqrt{\frac{v^2}{(1+w)^2} \cdot \sqrt{w}}} \left| \frac{v}{(1+w)^2} \right| \\ &= \frac{e^{-\frac{v}{2}}}{2\pi \frac{v}{1+w} \sqrt{w}} \left| \frac{v}{(1+w)^2} \right| \\ &= \frac{e^{-\frac{v}{2}}}{2\pi(1+w)\sqrt{w}} \mathbf{1}\{v > 0, w > 0\} \end{aligned}$$

5. The final step in solving for  $f_W(w)$  is to integrate out  $v$  from the joint pdf as follows:

$$\begin{aligned} f_W(w) &= \int_0^\infty \frac{e^{-\frac{v}{2}}}{2\pi(1+w)\sqrt{w}} dv \\ &= \frac{-1}{\pi(1+w)\sqrt{w}} \int_0^\infty \frac{-1}{2} e^{-\frac{v}{2}} dv \\ &= \frac{-1}{\pi(1+w)\sqrt{w}} \left( e^{-\frac{v}{2}} \Big|_0^\infty \right) \\ &= \frac{-1}{\pi(1+w)\sqrt{w}} (0 - 1) \\ &= \frac{1}{\pi(1+w)\sqrt{w}} \mathbf{1}\{w > 0\} \end{aligned}$$

**6.26** Let  $X_1$  and  $X_2$  be independent negative binomial random variables  $X_1 \sim NB(r_1, p)$  and  $X_2 \sim NB(r_2, p)$ .

(a) Find the MGF of  $Y = X_1 + X_2$

The MGF for  $X_1 \sim NB(r_1, p)$  is:

$$M_{X_1}(t) = \begin{cases} \left( \frac{pe^t}{1-(1-p)e^t} \right)^{r_1} & \text{for } t < -\log(1-p) \\ \infty & \text{otherwise.} \end{cases}$$

The MGF for  $X_2 \sim NB(r_2, p)$  is:

$$M_{X_2}(t) = \begin{cases} \left( \frac{pe^t}{1-(1-p)e^t} \right)^{r_2} & \text{for } t < -\log(1-p) \\ \infty & \text{otherwise.} \end{cases}$$

Therefore, the MGF for  $Y = X_1 + X_2$  is obtained by

$$\begin{aligned} M_Y(t) &= M_{X_1}(t)M_{X_2}(t) \\ &= \left( \frac{pe^t}{1-(1-p)e^t} \right)^{r_1} \left( \frac{pe^t}{1-(1-p)e^t} \right)^{r_2} \\ &= \left( \frac{pe^t}{1-(1-p)e^t} \right)^{r_1+r_2} \end{aligned}$$

The MGF of  $Y = X_1 + X_2$  is:

$$M_Y(t) = \begin{cases} \left( \frac{pe^t}{1-(1-p)e^t} \right)^{r_1+r_2} & \text{for } t < -\log(1-p) \\ \infty & \text{otherwise.} \end{cases}$$

(b) What is the distribution of  $Y$ ?

Thus we see that  $Y \sim NB(r_1 + r_2, p)$ .