

# hw6

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## Question 1

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Let  $X_i$ ,  $i = 1, 2, 3$  be independent with  $N(i, i^2)$  distributions. For each of the following situations, use the  $X_i$ 's to construct a statistic with the indicated distributions.

- a.  $\chi^2$  with 3 degrees of freedom
- b.  $t$  distribution with 2 degrees of freedom
- c.  $F$  distribution with 1 and 2 degrees of freedom

**(a) Answer:**

$$Z_i = \frac{X_i - i}{\sqrt{i^2}} = \frac{X_i - i}{i} \text{ for } i = 1, 2, 3$$

$$Z_i \sim N(0, 1)$$

$$\sum_{i=1}^3 \left( \frac{X_i - i}{i} \right)^2 \sim \chi^2(3)$$

**(b) Answer:**

$$\sum_{i=1}^2 \left( \frac{X_i - i}{i} \right)^2 \sim \chi^2(2)$$

By independence of  $X_1$ ,  $X_2$ , and  $X_3$ ,

$$\frac{X_1 - 1}{\sqrt{\sum_{i=2}^3 \left( \frac{X_i - i}{i} \right)^2 / 2}} \sim t_2$$

**(c) Answer:**

$$(X_1 - 1)^2 \sim \chi^2(1)$$

$$\frac{\chi^2(1)/1}{\chi^2(2)/2} \sim F_{1,2}$$

By independence of  $X_1$ ,  $X_2$ , and  $X_3$ ,

$$\frac{(X_1 - 1)^2}{\sum_{i=2}^3 \left( \frac{X_i - i}{i} \right)^2 / 2} \sim F_{1,2}$$

## Question 2

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Suppose  $X$  and  $Y$  are iid normal(0,1). Prove that  $W = X^2 + 2Y^2$  does not have a chi-square distribution.

**Answer:**

By independence of  $X$  and  $Y$ ,

$$M_W(t) = M_{X^2+2Y^2}(t) = E[e^{X^2t+2Y^2t}] = E[e^{X^2t}]E[e^{2Y^2t}] = M_{X^2}(t)M_{Y^2}(2t)$$

Since  $X$  and  $Y$  are iid normal(0,1)

$$X^2 \sim \chi^2(1)$$

$$Y^2 \sim \chi^2(1)$$

$$M_{X^2}(t) = \frac{1}{(1-2t)^{1/2}}, t < 1/2$$

$$M_{Y^2}(2t) = \frac{1}{(1-4t)^{1/2}}, t < 1/4$$

$$M_W(t) = M_{X^2}(t)M_{Y^2}(2t) = \frac{1}{(1-2t)^{1/2}} \frac{1}{(1-4t)^{1/2}}, t < 1/4$$

Since it is not in the shape of the MGF of chisquare distribution, we can conclude that  $W$  does not follow a chisquare distribution.

## Question 3

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Let  $X_1$  and  $X_2$  be two independent random variables. Let  $X_1$  and  $Y = X_1 + X_2$  have Poisson distributions with mean  $\mu_1$  and  $\mu \geq \mu_1$ , respectively. Derive the distribution of  $X_2$ .

**Answer:**

$$M_Y(t) = E[e^{X_1t+X_2t}] = M_{X_1}(t)M_{X_2}(t)$$

$$M_{X_2}(t) = \frac{M_Y(t)}{M_{X_1}(t)}$$

Since  $Y \sim \text{Poisson}(\mu)$  and  $X_1 \sim \text{Poisson}(\mu_1)$ ,

$$M_Y(t) = e^{\mu(e^t-1)} \forall t$$

$$M_{X_1}(t) = e^{\mu_1(e^t-1)} \forall t$$

$$M_{X_2}(t) = \frac{M_Y(t)}{M_{X_1}(t)} = e^{\mu(e^t-1)-\mu_1(e^t-1)} = e^{(\mu-\mu_1)(e^t-1)} \forall t$$

Therefore,  $X_2 \sim \text{Poisson}(\mu - \mu_1)$

## Question 4

Suppose  $X$  is distributed as Geometric( $p$ ) with  $0 < p < 1$  and  $q = 1 - p$ . The probability mass function is  $f(x) = pq^x$  for  $x = 0, 1, 2, \dots$

- Find the moment generating function of  $X$ .
- Find the mean and variance  $\mu_x$  and  $\sigma_x^2$ .
- Suppose that  $X_1, \dots, X_n$  are iid Geo( $p$ ),  $0 < p < 1$ . Discuss in detail the limiting distribution of  $\sqrt{n}(\bar{X} - \mu_x)$ .

**(a) Answer:**

Since  $\sum r^x = \frac{1}{1-r}$  for  $|r| < 1$ ,

$$M_X(t) = E[e^{tX}] = \sum_{x=0}^{\infty} e^{tx} p(1-p)^x = p \sum_{x=0}^{\infty} e^{tx} (1-p)^x = p \sum_{x=0}^{\infty} (e^t(1-p))^x = \frac{p}{1-e^t(1-p)}$$

To find the bound of  $t$ ,

$$|e^t(1-p)| < 1 \Rightarrow e^t(1-p) < 1 \Rightarrow t + \log(1-p) < 0 \Rightarrow t < -\log(1-p)$$

**(b) Answer:**

$$\text{Let } \psi_X(t) = \log M_X(t) = \log(p) - \log(1 - e^t(1-p))$$

$$\mu_X = \psi'_X(t) \Big|_{t=0} = \frac{e^t(1-p)}{1-e^t(1-p)} \Big|_{t=0} = \frac{1-p}{p}$$

$$\sigma_X^2 = \psi''_X(t) \Big|_{t=0} = \frac{(1-p)e^t\{1-(1-p)e^t\} + (1-p)^2 e^{2t}}{\{1-e^t(1-p)\}^2} \Big|_{t=0} = \frac{(1-p)p + (1-p)^2}{p^2} = \frac{1-p}{p^2}$$

**(c) Answer:**

By central limit theorem:

$$\frac{\sqrt{n}(\bar{X} - \mu_x)}{\sigma_x} \sim N(0, 1)$$

$$E[\sqrt{n}(\bar{X} - \mu_x)] = 0$$

$$\text{Var}[\sqrt{n}(\bar{X} - \mu_x)] = \sigma_x^2$$

$$\sqrt{n}(\bar{X} - \mu_x) \sim N(0, \sigma_x^2)$$

## Question 5

Let  $Y_1, Y_2$  be iid  $N(0, 1)$ . Find the moment generating function of  $Y = Y_1 Y_2$ .

**Answer:**

$$E[e^{Y_1 Y_2 t}] = E_{Y_2} \{E[e^{Y_1 Y_2 t} | Y_2]\}$$

Let  $Y_2 = y_2$

$$E[e^{Y_1 Y_2 t} | Y_2 = y_2] = E[e^{Y_1 y_2 t} | Y_2 = y_2] = E[e^{Y_1 y_2 t}] = M_{Y_1}(y_2 t)$$

Since  $Y_1 \sim N(0, 1)$

$$M_{Y_1}(y_2 t) = e^{\frac{1}{2} y_2^2 t^2} \text{ where } -\infty < t < \infty$$

Replace  $y_2$  with  $Y_2$

$$E_{Y_2}[e^{\frac{1}{2} Y_2^2 t^2}] = M_{Y_2^2}(\frac{1}{2} t^2)$$

Since  $Y_2 \sim N(0, 1)$ , we have  $Y_2^2 \sim \chi^2(1)$

$$M_{Y_2^2}(\frac{1}{2} t^2) = \frac{1}{(1 - 2 \cdot \frac{1}{2} t^2)^{1/2}} = \frac{1}{(1 - t^2)^{1/2}}$$

The support of  $t$  can be found below:

$$\frac{t^2}{2} < \frac{1}{2} \Rightarrow -1 < t < 1$$

$$M_Y(t) = \frac{1}{(1 - t^2)^{1/2}}$$

## Question 6

Suppose  $X$  is distributed as  $n(\mu, \sigma^2)$  and suppose  $Y = e^X$ .  $Y$  is said to have a log-normal distribution.

- Find the pdf of  $Y$ .
- Find  $E(Y^k)$  for any  $k$ .

Note that although  $E(Y^k)$  exists for every  $k$ ,  $Y$  does not have a mgf.

**(a) Answer:**

$$Y = e^X \Rightarrow \log Y = X$$

$$-\infty < X < \infty \Rightarrow 0 < Y < \infty$$

$$f_X(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \text{ with support: } -\infty < x < \infty$$

$$\left| \frac{dx}{dy} \right| = \left| \frac{d \log y}{dy} \right| = \left| \frac{1}{y} \right| = \frac{1}{y}$$

$$f_Y(y) = f_X(\log y) \left| \frac{dx}{dy} \right| = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(\log(y)-\mu)^2}{2\sigma^2}} \frac{1}{y} \text{ with support: } 0 < y < \infty$$

**(b) Answer:**

$$E[Y^k] = E[e^{Xk}] = M_X(k) = e^{k\mu + \frac{k^2\sigma^2}{2}} \text{ for all } k$$