# Sample Questions for Module 2

- Module 2 is about sampling distributions.
- Main aim is to be familiar with the t,  $\chi^2$ , and F-distributions, and be able to derive CI based on a pivotal quantity.
- Important results: sampling distribution of the sample mean and the sample variance (can use these without proof in the exam)

#### Sample Question 4

Consider two independent samples of size  $n_1$  and  $n_2$ :

$$Y_{i1} \sim i.i.d.N(\mu_1,\sigma_1^2), \qquad i=1,2,...,n_1$$
  $Y_{j2} \sim i.i.d.N(\mu_2,\sigma_2^2), \qquad j=1,2,...,n_2$  and writing  $\sigma_2^2 = k\sigma_1^2$ , where  $k>0$  is a constant. Let  $\overline{Y_1}$  and  $Y_2$  be their samples means and  $S_1^2$  and  $S_2^2$  be their sample variances.

- a) Show that  $Z = \frac{(\bar{Y}_1 \bar{Y}_2) (\mu_1 \mu_2)}{\sigma_1 \sqrt{\frac{1}{n_1} + \frac{k}{n_2}}}$  has a standard normal distribution.
- b) Show that  $W = \frac{(n_1 1)S_1^2 + (n_2 1)S_2^2/k}{\sigma_1^2}$  has a  $\chi^2$ -distribution.
- c) Show that  $T = \frac{(\bar{Y}_1 \bar{Y}_2) (\mu_1 \mu_2)}{S_p \sqrt{\frac{1}{n_1} + \frac{k}{n_2}}}$  has a t-distribution, where  $s_p^2 = \frac{\sigma_1^2}{n_1 + n_2 2} W$ .
- d) Using the results of part c), construct a  $100(1 \alpha)\%$  confidence interval for  $\mu_1 \mu_2$ .

(a) 
$$\overline{Y}_1 \sim \mathcal{N}(\mu_1, \frac{\sigma_1^2}{n_1})$$
 independent of  $\overline{Y}_2 \sim \mathcal{N}(\mu_2, \frac{\sigma_2^2}{n_2})$ .  
By the property of linear combination of independent normal random variables,  $\overline{Y}_1 - \overline{Y}_2 \sim \mathcal{N}(\mu_1 - \mu_2, \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2})$ .

$$Z = \frac{\left(\overline{Y}_{1} - \overline{Y}_{2}\right) - E\left(\overline{Y}_{1} - \overline{Y}_{2}\right)}{\sqrt{var\left(\overline{Y}_{1} - \overline{Y}_{2}\right)}} \sim N(0,1)$$

$$= \frac{\left(\overline{Y}_{1} - \overline{Y}_{2}\right) - \left(\mu_{1} - \mu_{2}\right)}{\sqrt{\frac{\sigma_{1}^{2}}{n_{1}} + \frac{\sigma_{2}^{2}}{n_{2}}}}$$

$$= \frac{\left(\overline{Y}_{1} - \overline{Y}_{2}\right) - \left(\mu_{1} - \mu_{2}\right)}{\sigma_{1}\sqrt{\frac{1}{n_{1}} + \frac{k}{n_{2}}}} \qquad (\text{Write } \sigma_{2}^{2} = k\sigma_{1}^{2})$$

$$\therefore Z \sim N(0,1).$$

(b) We know 
$$U_1 = \frac{(n_1-1)S_1^2}{\sigma_1^2} \sim \chi_{n_1-1}^2$$
 independent of  $U_2 = \frac{(n_2-1)S_2^2}{\sigma_2^2} \sim \chi_{n_2-1}^2$ 

$$W = U_1 + U_2 \sim \chi_{n_1+n_2-2}^2$$

$$= \frac{(n_1-1)S_1^2}{\sigma_1^2} + \frac{(n_2-1)S_2^2}{\sigma_2^2}$$

$$= \frac{(n_1-1)S_1^2}{\sigma_1^2} + \frac{(n_2-1)S_2^2}{k\sigma_1^2}$$

$$= \frac{(n_1-1)S_1^2}{\sigma_1^2} + \frac{(n_2-1)S_2^2}{k\sigma_1^2} \sim \chi_{n_1+n_2-2}^2$$

(c) From parts a and c, we have  $Z \sim N(0,1)$  independent of  $W \sim \chi^2_{n,tn_2-2}$ .

By the definition of the t-distribution, we have

$$T = \frac{Z}{\int \frac{\omega}{n_1 + n_2 - 2}} \sim t_{n_1 + n_2 - 2}$$

$$= \frac{\overline{Y_{1} - \overline{Y_{2}} - (\mu_{1} - \mu_{2})}}{\overline{\sigma_{1}} \cdot \overline{\frac{1}{n_{1}} + \frac{k}{n_{2}}}}$$

$$= \frac{\overline{W}_{1} - \overline{W}_{2} - (\mu_{1} - \mu_{2})}{\overline{\omega_{1} + n_{2} - 2}}$$

$$= \frac{Y_{1} - Y_{2} - (\mu_{1} - \mu_{2})}{\sqrt{\sigma_{1}^{2} \left(\frac{1}{n_{1}} + \frac{k}{n_{2}}\right) \left(\frac{\omega}{n_{1} + n_{2} - 2}\right)}}$$

$$= \frac{S\rho^{2}}{\left(\frac{1}{n_{1}} + \frac{k}{n_{2}}\right) \left(\frac{\omega}{n_{1} + n_{2} - 2}\right)}$$

$$= \frac{Y_1 - Y_2 - (\mu_1 - \mu_2)}{\int S\rho^2(\frac{1}{n_1} + \frac{k}{n_2})}$$

$$= \frac{\overline{Y_1} - \overline{Y_2} - (\mu_1 - \mu_2)}{S\rho\sqrt{\frac{1}{n_1} + \frac{k}{n_2}}}$$

d) 
$$1-d = P(-t_{n_1+n_2-2}, \leq T < t_{n_1+n_2-2}, \leq)$$

$$= P(-t_{n_1+n_2-2}, \leq \frac{(\bar{Y}_1 - \bar{Y}_2) - (\mu_2 - \mu_2)}{Sp\sqrt{\frac{1}{n_1} + \frac{k}{n_2}}} < t_{n_1+n_2-2}, \leq)$$

$$= P((\bar{Y}_1 - \bar{Y}_2) - t_{n_1+n_2-2}, \leq Sp\sqrt{\frac{1}{n_1} + \frac{k}{n_2}} < \mu_1 - \mu_2 <)$$

$$(\bar{Y}_1 - \bar{Y}_2) + t_{n_1+n_2-2}, \leq Sp\sqrt{\frac{1}{n_1} + \frac{k}{n_2}})$$

$$(\bar{Y}_1 - \bar{Y}_2) \pm t_{\Lambda_1 + \Omega_2 - 2} \leq S_P \sqrt{\frac{1}{n_1} + \frac{k}{n_2}}$$

#### Sample Question 5

Suppose  $Y_1, Y_2, ..., Y_n$  is a sample of size n from a Beta $(\theta, 1)$ . Consider the quantity  $X = -2\theta \sum_{i=1}^{n} \log Y_i$ .

- a) Find the distribution of *X* and show that *X* is a pivotal quantity.
- b) Use the above pivotal quantity to construct a symmetric  $100(1-\alpha)\%$  confidence interval for  $\theta$ .
- Suppose we have observed the following values for Y: 0.72, 0.48, 0.80, 0.45, 0.67 Using the results of part b), give a 95% confidence interval for  $\theta$ .

$$Y_{1},Y_{2},...,Y_{n} \sim iid \text{ Beta } (\theta,1)$$

$$f_{Y_{1}}(y) = \theta y^{\theta-1} \quad \text{for } 0 < y < 1$$

$$X = -2\theta \sum_{i=1}^{\infty} \log Y_{i}$$

$$(a) \quad \text{let } X_{i} = -2\theta \log Y_{i}, \text{ then } Y_{i} = e^{-\frac{X_{i}}{2\theta}} \quad \text{(inverse transformation)}$$

$$\frac{dY_{i}}{dX_{i}} = -\frac{1}{2\theta} e^{-\frac{X_{i}}{2\theta}}$$

$$f_{X_{i}}(x) = \theta \left(e^{-\frac{X_{i}}{2\theta}}\right)^{\frac{1}{2}(X_{i})} \left(-\frac{1}{2\theta}e^{-\frac{X_{i}}{2\theta}}\right) \quad \text{(transformation formula)}$$

$$= \frac{1}{2}e^{-\frac{X_{i}}{2}} \quad \text{for } x > 0$$

$$\therefore \quad X_{i} \sim iid \quad \text{Exp}(\frac{1}{2})$$

$$\therefore \quad X = \sum_{i=1}^{\infty} X_{i} \sim \text{Gamma}(n, \frac{1}{2}) \quad \text{or } X_{2n}^{2}.$$

Since X is a function of and the sample Y. Yz, ..., Yn, O is the only unknown parameter in X, and the distribution of X does not depend on O, so X is a pivotal quantity of O.

(b) 
$$1-\alpha = \mathcal{P}(\chi_{2n,1-\frac{\alpha}{2}}^2 < \chi < \chi_{2n,\frac{\alpha}{2}}^2)$$

$$= \mathcal{P}(\chi_{2n,1-\frac{\alpha}{2}}^2 < -20 \frac{\hat{\Sigma}}{|\Sigma|} \log Y; < \chi_{2n,\frac{\alpha}{2}}^2)$$

$$= \mathcal{P}(\frac{\chi_{2n,1-\frac{\alpha}{2}}^2}{-2 \frac{\hat{\Sigma}}{|\Sigma|} \log Y;} < 0 < \frac{\chi_{2n,\frac{\alpha}{2}}^2}{-2 \frac{\hat{\Sigma}}{|\Sigma|} \log Y;})$$

$$\therefore CI = (\frac{\chi_{2n,1-\frac{\alpha}{2}}^2}{-2 \frac{\hat{\Sigma}}{|\Sigma|} \log Y;}, \frac{\chi_{2n,\frac{\alpha}{2}}^2}{-2 \frac{\hat{\Sigma}}{|\Sigma|} \log Y;})$$

c) 
$$-2 \sum_{i=1}^{n} \log i$$
; =  $-2$  (log 0.72 + log 0.48 + log 0.80 + log 0.45 + log 0.67)  
=  $4.9692$   
 $n = 5$ ,  
 $\chi^{2}_{10,0.025} = 20.4832$ , qchisq(0.975,10)  
 $\chi^{2}_{10,0.975} = 3.2470$ , qchisq(0.025,10)  
 $CI = \left(\frac{3.2470}{4.9692}, \frac{20.4832}{4.9692}\right)$   
 $\approx (0.653, 4.122)$