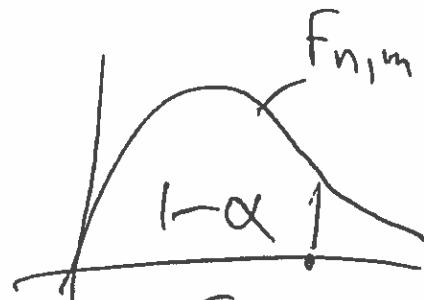


Homework 5 solutions

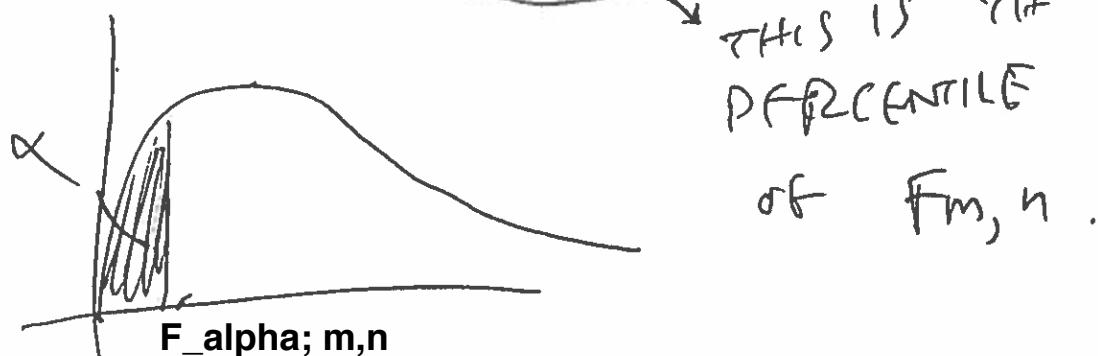
$$(a). X \sim F_{m,n} \rightarrow \frac{1}{X} \sim f_{n,m}$$

$$P(F_{n,m} < F_{1-\alpha;n,m}) = 1-\alpha$$



$$P(F_{m,n} > \frac{1}{F_{1-\alpha;n,m}}) = 1-\alpha$$

$$P(F_{m,n} < F_{1-\alpha;n,m}) = \alpha$$



Let $X \sim F_{1,n}$, then $P(X < F_{1-\alpha;1,n}) = 1-\alpha$

But $X = t_n^2$, so $P(t_n^2 < F_{1-\alpha;1,n}) = 1-\alpha$

Or $P[-\sqrt{F_{1-\alpha;1,n}} < t_n < +\sqrt{F_{1-\alpha;1,n}}] = 1-\alpha$

Therefore, $\sqrt{F_{1-\alpha;1,n}} = t_{1-\alpha/2;n}$

Finally, $t_{1-\alpha/2;n}^2 = F_{1-\alpha;1,n}$

$$(B). \bar{X} - \bar{Y} \sim N\left(\mu_1 - \mu_2, \sqrt{\frac{\sigma_1^2}{13} + \frac{\sigma_2^2}{16}}\right)$$

$$\text{OR } \bar{X} - \bar{Y} \sim N\left(\mu_1 - \mu_2, \sigma_1 \sqrt{\frac{1}{13} + \frac{3}{16}}\right)$$

Ans

$$\frac{(13-1)S_x^2}{\sigma_1^2} + \frac{(16-1)S_y^2}{\sigma_2^2} \sim \chi_{27}^2$$

$$\text{OR } \frac{12S_x^2}{\sigma_1^2} + \frac{15S_y^2}{5\sigma_1^2} \sim \chi_{27}^2$$

$$\text{or } \frac{12S_x^2}{\sigma_1^2} + \frac{3S_y^2}{\sigma_1^2} \sim \chi_{27}^2$$

Then

$$\frac{\bar{X} - \bar{Y} - (\mu_1 - \mu_2)}{\frac{\sigma_1 \sqrt{\frac{1}{13} + \frac{3}{16}}}{\sqrt{\frac{12S_x^2 + 3S_y^2}{\sigma_1^2}} / 27}} \sim t_{27}$$

(c). $\sum X_i \sim \Gamma(n\alpha, \beta) \rightarrow M_{\sum X_i}(t) = (1-\beta t)^{-n\alpha}$

$$M_{\bar{X}}(t) = M_{\frac{\sum X_i}{n}}(t) = M_{\sum X_i}\left(\frac{t}{n}\right)$$
$$= \left(1 - \frac{\beta t}{n}\right)^{-n\alpha}$$

LET $Q = \frac{2n}{\beta} \bar{X}$

Then

$$M_Q(t) = (1-2t)^{-n\alpha}$$
$$- \frac{2n\alpha}{2}$$

OR $M_Q(t) = (1-2t)^{-2n\alpha}$

$$\therefore Q = \frac{2n\bar{X}}{\beta} \sim \tilde{X}_{2n\alpha}$$

(Assume α is integer).

(f). $X \sim f_{n,m}$

OR $\frac{\hat{X}_n/n}{\hat{X}_m/m} =$

$$EX = \left(E \frac{\hat{X}_n}{n} \right) E \left(\frac{\hat{X}_m}{m} \right)^{-1}$$

BFAUSE
 \hat{X}_n, \hat{X}_m
ARE ID.

BUT $\hat{X}_m \sim F(\frac{n}{2}, 2)$

AND $E \frac{\hat{X}_n}{n} = 1$

So, $EX = E(F(\frac{n}{2}, 2))^{-1}$

USE PROPERTY OF $F(\alpha, \beta)$:

$$(E X)^k = \frac{F(\alpha+k)}{F(\alpha)}$$

Hence $k = -1$.

$$(e). \quad X_n \sim \mathcal{N}(0, 1) \quad z_j v \text{ are independent}$$

$$X = \frac{\bar{Z}}{\sqrt{U/n}}$$

$$\mathbb{E} X = \sqrt{n} \mathbb{E} Z e^{\bar{U}/2} = 0 \quad \text{because } \mathbb{E} Z = 0$$

$$M_R(X) = n [\mathbb{E} Z^2] [\mathbb{E} \bar{U}] = \infty$$

but $U \sim \chi_n^2$ or $U \sim \Gamma\left(\frac{n}{2}, 2\right)$

$$= n [\sigma^2 + \mu^2] \frac{\Gamma\left(\frac{n}{2} - 1\right) \bar{z}^{\frac{n}{2}}}{\Gamma\left(\frac{n}{2}\right)}$$

$$= n (1 + 0) \frac{\Gamma\left(\frac{n}{2} - 1\right)}{\left(\frac{n}{2} - 1\right) \Gamma\left(\frac{n}{2} - 1\right)} \cdot \bar{z}^{\frac{n}{2}}$$

$$= \frac{n}{n-2}.$$

(d)

$$Y' \Sigma^{-1} Y = (y_1, y_2) \begin{pmatrix} \sigma_1^2 & \sigma_{12} \\ \sigma_{12} & \sigma_2^2 \end{pmatrix}^{-1} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$$

$$= (y_1, y_2) \begin{pmatrix} \frac{\sigma_2^2}{\sigma_1^2 \sigma_2^2 - \sigma_{12}^2} & -\frac{\sigma_{12}}{\sigma_1^2 \sigma_2^2 - \sigma_{12}^2} \\ -\frac{\sigma_{12}}{\sigma_1^2 \sigma_2^2 - \sigma_{12}^2} & \frac{\sigma_1^2}{\sigma_1^2 \sigma_2^2 - \sigma_{12}^2} \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$$

$$= \begin{pmatrix} \frac{y_1 \sigma_2^2 - y_2 \sigma_{12}}{\sigma_1^2 \sigma_2^2 - \sigma_{12}^2}, & -\frac{y_1 \sigma_{12} + y_2 \sigma_1^2}{\sigma_1^2 \sigma_2^2 - \sigma_{12}^2} \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$$

$$= \frac{y_1^2 \sigma_2^2 - y_1 y_2 \sigma_{12} - y_1 y_2 \sigma_{12} + y_2^2 \sigma_1^2}{\sigma_1^2 \sigma_2^2 - \sigma_{12}^2}$$

$$= \frac{y_1^2 \sigma_2^2 - 2 y_1 y_2 \rho \sigma_1 \sigma_2 + y_2^2 \sigma_1^2}{\sigma_1^2 \sigma_2^2 - \rho^2 \sigma_1^2 \sigma_2^2} = \frac{y_1^2 \sigma_2^2 - 2 y_1 y_2 \rho \sigma_1 \sigma_2 + y_2^2 \sigma_1^2 - \sigma_1^2 (1-\rho^2) y_1^2}{\sigma_1^2 \sigma_2^2 (1-\rho^2)}$$

$$\text{Now: } Y' \Sigma^{-1} Y - \frac{y_1^2}{\sigma_1^2} = \frac{y_1^2 \sigma_2^2 - 2 y_1 y_2 \rho \sigma_1 \sigma_2 + y_2^2 \sigma_1^2 - \sigma_1^2 (1-\rho^2) y_1^2}{\sigma_1^2 \sigma_2^2 (1-\rho^2)}$$

$$= \frac{y_1^2 \sigma_2^2 - 2 y_1 y_2 \rho \sigma_1 \sigma_2 + y_2^2 \sigma_1^2 - y_1^2 \sigma_2^2 + \rho^2 y_1^2 \sigma_2^2}{\sigma_1^2 \sigma_2^2 (1-\rho^2)}$$

$$= \frac{1}{1-\rho^2} \left[-\frac{\rho^2 y_1^2}{\sigma_1^2} - \frac{2 y_1 y_2 \rho}{\sigma_1 \sigma_2} + \frac{y_2^2}{\sigma_2^2} \right] = \frac{1}{1-\rho^2} \left(\frac{\rho y_1}{\sigma_1} - \frac{y_2}{\sigma_2} \right)^2$$

$$= \frac{Q^2}{1-\rho^2}, \text{ where } Q \sim N(0, \sqrt{1-\rho^2}) \quad \therefore \sim \chi^2_1$$

(Q). from CLASS NOTES

$$(\bar{y} - \mu) \stackrel{D}{\sim} (\bar{y} - \mu) \sim \mathcal{X}_n^2 \quad (1)$$

$(x_1, \bar{y}_1), \dots, (x_n, \bar{y}_n)$ is a random sample from $N_n(\mu, \Sigma)$

therefore

$$t_1 \bar{x} + t_2 \bar{y}$$

$$M_{\bar{x}, \bar{y}}(t_1, t_2) = E e^{t_1 \bar{x} + t_2 \bar{y}}$$

$$= E e$$

$$= \left[E e^{\frac{t_1}{n} x_1 + \frac{t_2}{n} y_1} \right] \times \dots \times \left[E e^{\frac{t_1}{n} x_n + \frac{t_2}{n} y_n} \right]$$

$$= \left\{ M_{x_i, y_i} \left(\frac{t_1}{n}, \frac{t_2}{n} \right) \right\}^n$$

$$= \left\{ e^{\frac{t_1}{n} \mu + \frac{1}{2} \frac{t_1^2}{n} + \frac{t_2}{n} \mu + \frac{1}{2} \frac{t_2^2}{n}} \right\}^n = e^{\frac{t_1^2}{n} + \frac{1}{2} \frac{t_1^2}{n} + \frac{t_2^2}{n}}$$

$$\therefore \left(\begin{pmatrix} \bar{x} \\ \bar{y} \end{pmatrix} \right) \sim N_2 \left(\begin{pmatrix} \mu \\ \mu \end{pmatrix}, \frac{\Sigma}{n} \right) \text{ using } (1)$$

$$\text{we get } (\bar{x} - \mu_1, \bar{y} - \mu_2) \left(\frac{\Sigma}{n} \right)^{-1} \left(\begin{pmatrix} \bar{x} - \mu_1 \\ \bar{y} - \mu_2 \end{pmatrix} \right) \sim \mathcal{X}_2^2.$$

$$(h) z \sim N(0, 1) \quad u \sim X_n^2$$

z and u are independent.

$$f(z, u) = f(z) \cdot f(u) = \frac{1}{\sqrt{n}} e^{-\frac{1}{2}z^2} \frac{u^{\frac{n}{2}-1}}{\Gamma(\frac{n}{2})} \frac{e^{-\frac{u}{2}}}{2^{\frac{n}{2}}}$$

$$\left. \begin{array}{l} x = \frac{z}{\sqrt{u/n}} \\ w = u \end{array} \right\} \rightarrow z = x\sqrt{\frac{w}{n}}$$

$$\rightarrow u = w$$

$$J = \begin{pmatrix} \frac{\partial x}{\partial z} & \frac{\partial x}{\partial w} \\ \frac{\partial w}{\partial z} & \frac{\partial w}{\partial u} \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{u/n}} & -\frac{1}{2} \sqrt{\frac{z}{u}} \\ 0 & 1 \end{pmatrix} = \frac{1}{\sqrt{u/n}}$$

Therefore,

$$f(x, w) = \frac{1}{\sqrt{n}} e^{-\frac{1}{2}x^2} \frac{w^{\frac{n}{2}-1}}{\Gamma(\frac{n}{2})} \frac{e^{-\frac{w}{2}}}{2^{\frac{n}{2}}} \left(\frac{w}{n}\right)^{1/2}$$

Now INTEGRATE THE joint
PDF OF X, ω w.r.t. ω
to find $f(x)$:

$$\begin{aligned}
 f(x) &= \int_0^\infty f(x, \omega) d\omega \\
 &= \frac{1}{\sqrt{2n} \Gamma\left(\frac{n}{2}\right) 2^{n/2} \sqrt{n}} \int_0^\infty w^{\frac{n+1}{2}-1} e^{-\frac{w}{(C\frac{x^2}{n} + \frac{1}{2})}} dw \\
 &\quad \text{KERNEL FUNCTION} \\
 &= \frac{\left(\frac{2}{x^2+n+1}\right)^{\frac{n+1}{2}}}{\sqrt{2n} \Gamma\left(\frac{n}{2}\right) 2^{n/2} \sqrt{n}} \Gamma\left(\frac{n+1}{2}\right) \left(1 + \frac{x^2}{n}\right)^{-\frac{n+1}{2}}
 \end{aligned}$$

$$(i). \quad z \sim N(\delta, 1)$$

$$v \sim \chi_n^2$$

$$T \text{ (t-tn)} \quad t = \frac{\bar{z}}{\sqrt{\frac{\bar{v}}{n}}} \sim t_n \quad (N\delta = \sigma)$$

$$\mathbb{E} t = \sqrt{n} (\mathbb{E} \bar{z}) \left(\mathbb{E} \bar{v}^{-1/2} \right)$$

$$= \sqrt{n} \cdot \sigma \cdot \frac{\Gamma\left(\frac{n}{2} - \frac{1}{2}\right)^{-1/2}}{\Gamma\left(\frac{n}{2}\right)}$$

$$\text{Var}(t) \subseteq \mathbb{E} t^2 - \tilde{(\mathbb{E} t)}^2$$

(*). $U \sim \mathcal{X}_{n_1} \text{ (NCF} = \theta\text{)}$

$V \sim \mathcal{X}_{n_2}$

$\frac{U/n_1}{V/n_2} \sim F_{n_1, n_2} \text{ (NCF} = \theta\text{)}$

$$E \frac{U/n_1}{V/n_2} = \frac{n_2}{n_1} (E U) (E V')$$

NOTE: $M_U(t) = (-2t)^{\frac{-n_1}{2}} e^{\theta \frac{t}{t-2t}}$