

## Problem 1

(a)

$$X_1, \dots, X_n \sim P(X_i \leq x | \alpha, \beta) = \begin{cases} 0 & x < 0 \\ (x/\beta)^\alpha & 0 \leq x \leq \beta \\ 1 & x > \beta \end{cases}$$

$\alpha, \beta$  are positive

$$\hat{\beta}_{MLE} = X_{(n)} \text{ (from problem 7.10)}$$

$$\alpha = \alpha_0$$

$$\text{Pivot} = X_{(n)}/\beta$$

$$.05 = P(X_{(n)}/\beta \leq c) = P(X_1, \dots, X_n \leq c\beta) = \left(\frac{c\beta}{\beta}\right)^{\alpha_0 n} = c^{\alpha_0 n}$$

$$.05^{1/(\alpha_0 n)} = c$$

$$.95 = P(X_{(n)}/\beta > c) = P(X_{(n)}/c > \beta)$$

$$= P(X_{(n)}/.05^{1/(\alpha_0 n)} > \beta)$$

$$\{\beta : \beta < X_{(n)}/(.05^{1/(\alpha_0 n)})\}$$

(b)

$$\hat{\beta}_{MLE} = X_{(n)} = 25$$

$$\hat{\alpha}_{MLE} = 12.59 \quad n = 14$$

$$\{\beta : \beta < 25/ (.05^{1/(12.59*14)})\}$$

$$25/ (.05^{1/(12.59*14)}) = 25.42853 = 25.43$$

$$\beta < 25.43$$

Since  $\beta$  is positive, the lower bound cannot be below 0

Thus the interval for  $\beta$  is  $(0, 25.43)$

## Problem 2

(a)

$$X_1, \dots, X_n \sim N(0, \sigma_X^2) \quad Y_1, \dots, Y_m \sim N(0, \sigma_Y^2) \quad X \perp Y$$

$$\lambda = \sigma_Y^2 / \sigma_X^2$$

$$H_0 : \lambda = \lambda_0 \text{ vs } H_1 : \lambda \neq \lambda_0$$

$$\lambda(x, y) = \frac{\sup_{\lambda=\lambda_0} L(\sigma_X^2, \sigma_Y^2 | x, y)}{\sup_{\lambda \in \Theta} L(\sigma_X^2, \sigma_Y^2 | x, y)}$$

Unrestricted MLEs:

$$\hat{\sigma}_{XMLE}^2 = \sum_{i=1}^n X_i^2 / n \quad \hat{\sigma}_{YMLE}^2 = \sum_{i=1}^m Y_i^2 / m$$

$$L(\sigma_X^2, \sigma_Y^2 | x, y) = (2\pi)^{-(n+m)/2} (\sigma_X^2)^{-n/2} (\sigma_Y^2)^{-m/2} \exp\left(-\sum x_i^2 / (2\sigma_X^2)\right) \exp\left(-\sum y_i^2 / (2\sigma_Y^2)\right)$$

$$\text{Under } H_0 : \quad \lambda = \lambda_0 = \sigma_Y^2 / \sigma_X^2$$

$$\sigma_Y^2 = \lambda_0 \sigma_X^2$$

$$L(\sigma_X^2, \lambda_0 \sigma_X^2 | x, y) = (2\pi \sigma_X^2)^{-n/2} (2\pi \lambda_0 \sigma_X^2)^{-m/2} \exp\left(-\sum x_i^2 / (2\sigma_X^2)\right) \exp\left(-\sum y_i^2 / (2\lambda_0 \sigma_X^2)\right)$$

$$= (2\pi)^{-(n+m)/2} (\sigma_X^2)^{-(n+m)/2} \lambda_0^{-m/2} \exp\left(-\left[\lambda_0 \sum x_i^2 + \sum y_i^2\right] / (2\lambda_0 \sigma_X^2)\right)$$

$$\ell = -((n+m)/2) \log(2\pi) - ((n+m)/2) \log(\sigma_X^2) - (m/2) \log(\lambda_0) - \left[\lambda_0 \sum x_i^2 + \sum y_i^2\right] / (2\lambda_0 \sigma_X^2)$$

$$\frac{\partial \ell}{\partial \sigma_X^2} = -\frac{(n+m)/2}{\sigma_X^2} + \frac{\lambda_0 \sum x_i^2 + \sum y_i^2}{2\lambda_0 (\sigma_X^2)^2} = 0$$

$$\frac{(n+m)/2}{\sigma_X^2} = \frac{\lambda_0 \sum x_i^2 + \sum y_i^2}{2\lambda_0 (\sigma_X^2)^2}$$

$$\hat{\sigma}_0^2 = \frac{\lambda_0 \sum x_i^2 + \sum y_i^2}{\lambda_0 (n+m)}$$

$$\frac{\partial \ell^2}{\partial (\sigma_X^2)^2} = \frac{(n+m)/2}{(\sigma_X^2)^2} - \frac{\lambda_0 \sum x_i^2 + \sum y_i^2}{\lambda_0 (\sigma_X^2)^3}$$

Plugging in  $\hat{\sigma}_0^2$ :

$$= \left(\frac{\lambda_0 (n+m)}{\lambda_0 \sum x_i^2 + \sum y_i^2}\right)^2 \left[(n+m)/2 - \frac{\lambda_0 \sum x_i^2 + \sum y_i^2}{\lambda_0} \frac{\lambda_0 (n+m)}{\lambda_0 \sum x_i^2 + \sum y_i^2}\right]$$

$$= \left(\frac{\lambda_0 (n+m)}{\lambda_0 \sum x_i^2 + \sum y_i^2}\right)^2 (-(n+m)/2)$$

$$= -\frac{(\sigma_0^2)^2 (n+m)}{2} < 0$$

Thus  $\hat{\sigma}_0^2$  is the MLE

$$\begin{aligned}
\lambda(x, y) &= \frac{(\hat{\sigma}_0^2)^{-(n+m)/2} \lambda_0^{-m/2} \exp\left(-[\lambda_0 \sum x_i^2 + \sum y_i^2] / (2\lambda_0 \hat{\sigma}_0^2)\right)}{(\hat{\sigma}_X^2)^{-n/2} (\hat{\sigma}_Y^2)^{-m/2} \exp\left(-\sum x_i^2 / (2\hat{\sigma}_X^2) - \sum y_i^2 / (2\hat{\sigma}_Y^2)\right)} \\
&= \frac{(\hat{\sigma}_X^2)^{n/2} (\hat{\sigma}_Y^2)^{m/2} \exp\left(-[\lambda_0 \sum x_i^2 + \sum y_i^2] / (2\lambda_0 \hat{\sigma}_0^2) + \sum x_i^2 / (2\hat{\sigma}_X^2) + \sum y_i^2 / (2\hat{\sigma}_Y^2)\right)}{(\hat{\sigma}_0^2)^{(n+m)/2} \lambda_0^{m/2}} \\
&= \frac{(\hat{\sigma}_X^2)^{n/2} (\hat{\sigma}_Y^2)^{m/2} \exp(-(n+m)/2 + (n/2) + (m/2))}{(\hat{\sigma}_0^2)^{(n+m)/2} \lambda_0^{m/2}} \\
\lambda(x, y) &= \frac{(\hat{\sigma}_X^2)^{n/2} (\hat{\sigma}_Y^2)^{m/2}}{(\hat{\sigma}_0^2)^{(n+m)/2} \lambda_0^{m/2}}
\end{aligned}$$

$R = \{(x, y) : \lambda(x, y) < c\}$  Where  $c$  is chosen to give the test size  $\alpha$

(b)

$$\begin{aligned}
&\text{Under } H_0 : \lambda_0 = \lambda = \sigma_Y^2 / \sigma_X^2 \\
&\text{Let } A = \sum Y_i^2 / (\lambda_0 \sigma_X^2) = \sum Y_i^2 / \sigma_Y^2 \sim \chi_m^2 \\
&\text{Let } B = \sum X_i^2 / \sigma_X^2 \sim \chi_n^2 \\
&A \perp B \\
&\text{Let } F = \frac{A}{B} \frac{n}{m} = \frac{(\sum Y_i^2 / \sigma_Y^2) / m}{(\sum X_i^2 / \sigma_X^2) / n} = \frac{\sum Y_i^2 / (\lambda_0 m)}{\sum X_i^2 / n} \sim F_{m,n} \\
\lambda(x, y) &= \left( \frac{\hat{\sigma}_X^2}{\hat{\sigma}_0^2} \right)^{n/2} \left( \frac{\hat{\sigma}_Y^2}{\hat{\sigma}_0^2 \lambda_0} \right)^{m/2} \\
\frac{\hat{\sigma}_X^2}{\hat{\sigma}_0^2} &= \frac{n+m}{n} \frac{\sum X_i^2 \lambda_0}{\lambda_0 \sum X_i^2 + \sum Y_i^2} \\
&= \frac{n+m}{n} \frac{\sum X_i^2 \lambda_0}{\lambda_0 \sum X_i^2 + \sum Y_i^2} \\
&= \frac{n+m}{n} \frac{1}{1 + (\sum Y_i^2 / \sigma_Y^2) / (\sum X_i^2 / \sigma_X^2)} \\
&= \frac{n+m}{n + nA/B} \\
&= 1 / \left( \frac{n + nA/B}{n+m} \right) \\
&= 1 / \left( \frac{n}{n+m} + \frac{1}{m+n} \frac{mnA}{mB} \right) \\
&= 1 / \left( \frac{n}{n+m} + \frac{m}{m+n} F \right)
\end{aligned}$$

$$\begin{aligned}
\frac{\hat{\sigma}_Y^2}{\hat{\sigma}_0^2 \lambda_0} &= \frac{n+m}{m} \frac{\sum Y_i^2}{\lambda_0 \sum X_i^2 + \sum Y_i^2} \\
&= \frac{n+m}{m} \frac{1}{1 + (\sigma_Y^2 \sum X_i^2) / (\sum Y_i^2 \sigma_X^2)} \\
&= 1 / \left( \frac{m}{n+m} \left[ 1 + \frac{\sigma_Y^2 \sum X_i^2}{\sigma_X^2 \sum Y_i^2} \right] \right) \\
&= 1 / \left( \frac{m}{n+m} (1 + B/A) \right) \\
&= 1 / \left( \frac{m}{n+m} + \frac{1}{n+m} \frac{mB}{A} \frac{n}{n} \right) \\
&= 1 / \left( \frac{m}{n+m} + \frac{n}{n+m} \frac{mB}{nA} \right) \\
&= 1 / \left( \frac{m}{n+m} + \frac{n}{n+m} F^{-1} \right) \\
\lambda(x, y) &= \left( \frac{1}{\frac{n}{n+m} + \frac{m}{m+n} F} \right)^{n/2} \left( \frac{1}{\frac{m}{n+m} + \frac{n}{n+m} F^{-1}} \right)^{m/2} \\
R &= \left\{ (x, y) : \left( \frac{1}{\frac{n}{n+m} + \frac{m}{m+n} F} \right)^{n/2} \left( \frac{1}{\frac{m}{n+m} + \frac{n}{n+m} F^{-1}} \right)^{m/2} < c_\alpha \right\}
\end{aligned}$$

Where  $c_\alpha$  satisfies:

$$P \left( \left( \frac{1}{\frac{n}{n+m} + \frac{m}{m+n} F} \right)^{n/2} \left( \frac{1}{\frac{m}{n+m} + \frac{n}{n+m} F^{-1}} \right)^{m/2} < c_\alpha \right) = \alpha$$

(c)

$$\begin{aligned}
\lambda(x, y) &= \left( \frac{n}{n+m} + \frac{m}{n+m} \frac{\sum Y_i^2}{\sum X_i^2 \lambda} \frac{n}{m} \right)^{-n/2} \left( \frac{m}{n+m} + \frac{n}{n+m} \frac{\sum X_i^2 \lambda}{\sum Y_i^2} \frac{m}{n} \right)^{-m/2} \\
&= \left( \frac{n}{n+m} + \frac{n}{n+m} \frac{\sum Y_i^2}{\sum X_i^2 \lambda} \right)^{-n/2} \left( \frac{m}{n+m} + \frac{m}{n+m} \frac{\sum X_i^2 \lambda}{\sum Y_i^2} \right)^{-m/2} \\
&\quad \text{Let } D = \frac{n}{n+m} \text{ and } E = D \frac{\sum Y_i^2}{\sum X_i^2} = \frac{n}{n+m} \frac{\sum Y_i^2}{\sum X_i^2} \\
\lambda(x, y) &= \left( D + \frac{E}{\lambda} \right)^{-n/2} \left( (1-D) + (1-D) \frac{D \sum Y_i^2}{D \sum X_i^2 \lambda} \right)^{-m/2}
\end{aligned}$$

$$\lambda(x, y) = \left(D + \frac{E}{\lambda}\right)^{-n/2} \left((1-D) + (1-D)D\frac{\lambda}{E}\right)^{-m/2}$$

The acceptance region is:

$$A = \left\{ \lambda(x, y) : \left(D + \frac{E}{\lambda}\right)^{-n/2} \left((1-D) + (1-D)D\frac{\lambda}{E}\right)^{-m/2} \geq c_a \right\}$$

Inverting the acceptance region we have the  $1 - \alpha$  CI for  $\lambda$  :

$$C(\lambda) = \left\{ \lambda : \left(D + \frac{E}{\lambda}\right)^{-n/2} \left((1-D) + (1-D)D\frac{\lambda}{E}\right)^{-m/2} \geq c_a \right\}$$

Multiplying both sides by  $\left(\frac{E}{1-D}\right)^{-m/2}$  we have:

$$\begin{aligned} \left(D + \frac{E}{\lambda}\right)^{-n/2} \left([ (1-D) + (1-D)D\frac{\lambda}{E} ] \left(\frac{E}{1-D}\right)\right)^{-m/2} &\geq c_a \left(\frac{E}{1-D}\right)^{-m/2} \\ \left(D + \frac{E}{\lambda}\right)^{-n/2} (E + D\lambda)^{-m/2} &\geq c_a \left(\frac{E}{1-D}\right)^{-m/2} \end{aligned}$$

taking the derivative of the log with respect to  $\lambda$  of the left side:

$$\begin{aligned} &\frac{\partial}{\partial \lambda} (-n/2) \log(D + E/\lambda) - (m/2) \log(E + D\lambda) \\ &= \frac{nE - mD\lambda}{2\lambda(E + D\lambda)} \Rightarrow (1/2) \frac{n \frac{n}{n+m} \frac{\sum Y_i}{\sum X_i} - m \frac{n}{n+m} \lambda}{\lambda \left( \frac{n}{n+m} \frac{\sum Y_i}{\sum X_i} + \frac{n}{n+m} \lambda \right)} \end{aligned}$$

$$\text{Let } S = \frac{\sum Y_i}{\sum X_i}$$

$$= (1/2) \frac{nS - m\lambda}{\lambda(S + \lambda)} \Rightarrow (1/2) \frac{nS/\lambda - m}{S + \lambda}$$

For  $\lambda \geq 0$  the derivative changes sign from positive to negative,

thus  $C(\lambda)$  increases and decreases and is therefore an interval

The graph of  $C(\lambda)$  is a parabola

### Problem 3

(a)

$$X_1, \dots, X_n \sim f(x|\theta) = 1, \quad \theta - 1/2 < x < \theta + 1/2$$

$$Y = X - \theta$$

$$f_Y(y) = 1 \quad -1/2 < y < 1/2 \quad Y \perp \theta$$

$$\begin{aligned}
Y &\sim U(-1/2, 1/2) \\
1 - \alpha &= P(\alpha_1 - 1/2 \leq X - \theta \leq 1/2 - \alpha_2) \\
1 - \alpha &= P(X_1 - (1/2 - \alpha_2) \leq \theta \leq X_1 - (\alpha_1 - 1/2)) \\
1 - \alpha \text{ CI} &= (\alpha_2 - 1/2, 1/2 - \alpha_1)
\end{aligned}$$

(b)

$$\begin{aligned}
X_1, \dots, X_n &\sim f(x|\theta) = 2x/\theta^2, \quad 0 < x < \theta, \quad \theta > 0 \\
Y = X/\theta \quad \frac{dy}{dx} &= 1/\theta \\
f_y(y) = f_x(\theta y) &= \frac{1}{|1/\theta|} = \frac{2y\theta}{\theta^2} \theta \\
f_Y(y) = 2y \quad 0 \leq y \leq 1 \quad Y &\perp \theta \\
P(a \leq X/\theta \leq b) &= \int_a^b 2y \, dy = b^2 - a^2 \\
b^2 - a^2 = 1 - \alpha &\Rightarrow b^2 - a^2 = \sqrt{1 - \alpha/2}^2 - \sqrt{\alpha/2}^2 \\
b = \sqrt{1 - \alpha/2} \quad a &= \sqrt{\alpha/2}
\end{aligned}$$

## Problem 4

(a)

$$\begin{aligned}
X_1, \dots, X_n &\sim f(x|\theta) = (a/\theta)(x/\theta)^{a-1} \quad 0 < x < \theta \\
CDF &\sim U(0, 1) \\
1 - \alpha &= P(\alpha_1 < F_{X_{(n)}}(x) < 1 - \alpha_2) \\
F_{X_{(n)}}(x) &= P(X_{(n)} \leq x) = [P(X_1 \leq x)]^n \\
P(X_1 \leq x) &= \int_0^x (a/\theta)(y/\theta)^{a-1} \, dy = (x/\theta)^a \\
[P(X_1 \leq x)]^n &= (x/\theta)^{an} \\
1 - \alpha &= P(\alpha_1 < (X_{(n)}/\theta)^{an} < 1 - \alpha_2) \\
&= P(\alpha_1^{1/(an)} < X_{(n)}/\theta < (1 - \alpha_2)^{1/(an)}) \\
&= P(X_{(n)}/(1 - \alpha_2)^{1/(an)} < \theta < X_{(n)}/\alpha_1^{1/(an)}) \\
1 - \alpha \text{ CI} &= \left( \frac{X_{(n)}}{(1 - \alpha_2)^{1/(an)}}, \frac{X_{(n)}}{\alpha_1^{1/(an)}} \right)
\end{aligned}$$

(b)

$$\begin{aligned}
Y &= \left( \frac{X_{(n)}}{\theta} \right)^{na} \\
F_Y(y) &= P((X_{(n)}/\theta)^{na} \leq y) = [P(X_{(n)} \leq \theta y^{1/na})]^n \\
&= \left( \frac{\theta y^{1/na}}{\theta} \right)^{na} = y \Rightarrow Y \sim U(0, 1) \\
F_Y(y) &\sim U(0, 1) \perp \theta \\
1 - \alpha &= P(\alpha_1 \leq (X_{(n)}/\theta)^{na} < 1 - \alpha_1)
\end{aligned}$$

(c)

The intervals are the same

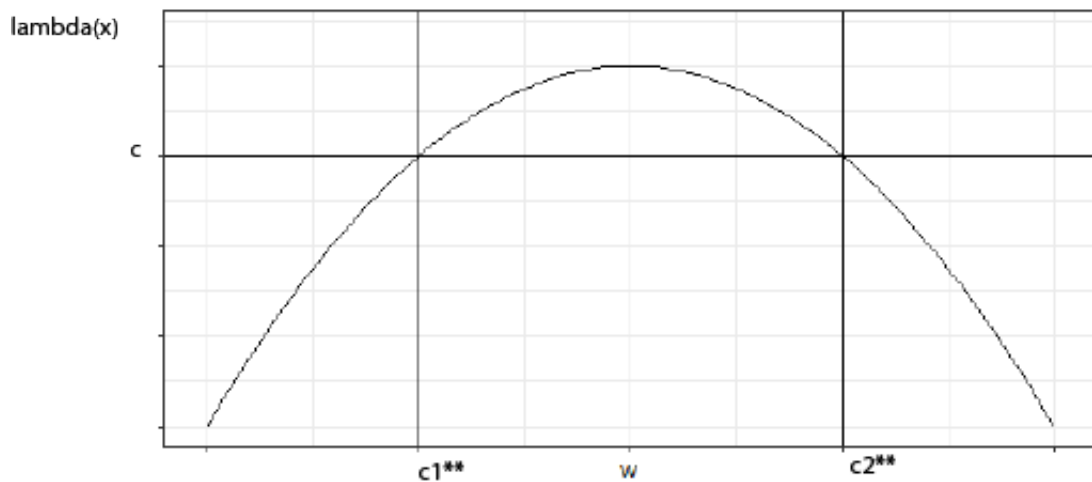
## Problem 5

(a)

$$\begin{aligned}
X_1, \dots, X_m &\sim f(x|\mu_1) = \frac{1}{\mu_1} \exp(-x/\mu_1) \\
Y_1, \dots, Y_n &\sim f(y|\mu_2) = \frac{1}{\mu_2} \exp(-y/\mu_2) \\
H_0 : \mu_1 &= \mu_2 = \mu_0 \text{ vs } H_1 : \mu_1 \neq \mu_2 \\
\lambda(x) &= \frac{\sup_{\theta \in \Theta_0} L(\theta)}{\sup_{\theta \in \Theta} L(\theta)} \\
&= \frac{L(\hat{\mu}_0)}{L(\hat{\mu}_1, \hat{\mu}_2)} = \frac{L(\hat{\mu}_0)}{L(\bar{X}, \bar{Y})} \\
L(\mu_1, \mu_2) &= \prod \frac{1}{\mu_1} e^{-x_i/\mu_1} \prod \frac{1}{\mu_2} e^{-y_i/\mu_2} \\
L(\mu_0) &= (1/\mu_0)^{m+n} e^{-\sum x_i/\mu_0} e^{-\sum y_i/\mu_0} \\
&= (1/\mu_0)^{m+n} e^{-(\sum x_i + \sum y_i)/\mu_0} \\
\hat{\mu}_0 &= \frac{\sum x_i + \sum y_i}{m+n}
\end{aligned}$$

$$\begin{aligned}
\frac{L(\hat{\mu}_0)}{L(\bar{X}, \bar{Y})} &= \frac{\left(\frac{m+n}{\sum x_i + \sum y_i}\right)^{m+n} e^{-(m+n)}}{(1/\bar{x})^m e^{-m} (1/\bar{y})^n e^{-n}} \\
&= \left(\frac{m+n}{\sum x_i + \sum y_i}\right)^{m+n} \bar{x}^m \bar{y}^n \\
&= \frac{(m+n)^{m+n}}{n^n m^m} \frac{(\sum x_i)^m (\sum y_i)^n}{(\sum x_i + \sum y_i)^{m+n}} \\
&= \frac{(m+n)^{m+n}}{n^n m^m} \left(\frac{\sum x_i}{\sum x_i + \sum y_i}\right)^m \left(\frac{\sum y_i}{\sum x_i + \sum y_i}\right)^n \\
&= \frac{(m+n)^{m+n}}{n^n m^m} \left(\frac{\sum x_i + \sum y_i}{\sum x_i}\right)^{-m} \left(\frac{\sum x_i + \sum y_i}{\sum y_i}\right)^{-n} \\
&= \frac{(m+n)^{m+n}}{n^n m^m} \left(1 + \frac{\sum y_i}{\sum x_i}\right)^{-m} \left(1 + \frac{\sum x_i}{\sum y_i}\right)^{-n} \\
&= \frac{(m+n)^{m+n}}{n^n m^m} \left(1 + \frac{n\bar{y}}{m\bar{x}}\right)^{-m} \left(1 + \frac{m\bar{x}}{n\bar{y}}\right)^{-n} \\
&\quad \text{Let } r = \bar{y}/\bar{x} \quad w = m/(m+nr) \\
&= \frac{(m+n)^{m+n}}{n^n m^m} \left(1 + \frac{nr}{m}\right)^{-m} \left(1 + \frac{m}{nr}\right)^{-n} \\
&= \frac{(m+n)^{m+n}}{n^n m^m} \left(\frac{m}{m+nr}\right)^m \left(\frac{nr}{m+nr}\right)^n \\
&= \frac{(m+n)^{m+n}}{n^n m^m} w^m (1-w)^n
\end{aligned}$$

(b)





$$w = \frac{m}{m + n \left( \frac{m \sum y_i}{n \sum y_i} \right)} = \frac{\sum x_i}{\sum x_i + \sum y_i} \Rightarrow 0 \leq w \leq 1$$

$\lambda(x, y)$  is a quadratic function of  $w$

$w$  is a monotone decreasing function of  $r$

$$R = \{\lambda(x, y) \leq c\} \Leftrightarrow \{w < c_1^*\} \text{ or } \{w > c_2^*\}$$

$$\Leftrightarrow \{r < c_1^{**}\} \text{ or } \{r > c_2^{**}\}$$

$$X_i \sim \text{Exp}(\mu_1) = \text{Gamma}(1, \mu_1)$$

$$\sum X_i \sim \text{Gamma}(m, \mu_1) \Rightarrow 2 \sum X_i / \mu_1 \sim \text{Gamma}(m, 2) = \chi_{2m}^2$$

$$2 \sum Y_i / \mu_2 \sim \chi_{2n}^2$$

$$\frac{2 \sum X_i / \mu_1 / 2m}{2 \sum Y_i / \mu_2 / 2n} = \frac{n \sum X_i \mu_2}{m \sum Y_i \mu_1} = \frac{\bar{X} \mu_2}{\bar{Y} \mu_1} \sim F_{2m, 2n}$$

$$\alpha_1 = P\left(\frac{\bar{Y}}{\bar{X}} < c_1^* | H_0\right) = P\left(\frac{\mu_1 \bar{Y}}{\mu_2 \bar{X}} < \frac{c_1^* \mu_1}{\mu_2} | H_0\right)$$

$$\frac{\mu_1 \bar{Y}}{\mu_2 \bar{X}} \sim F_{2n, 2m} \quad \frac{c_1^* \mu_1}{\mu_2} = c_1^* \text{ (since } \mu_1 = \mu_2 \text{ under } H_0)$$

$$c_1^* = F_{2n, 2m, \alpha_1}$$

$$\alpha_2 = P\left(\frac{\bar{Y}}{\bar{X}} > c_2^* | H_0\right) = P\left(\frac{\mu_1 \bar{Y}}{\mu_2 \bar{X}} > \frac{c_2^* \mu_1}{\mu_2} | H_0\right)$$

$$c_2^* = F_{2n, 2m, 1-\alpha_2}$$

(c)

$$\psi = \mu_2 / \mu_1$$

$$\frac{\psi \bar{X}}{\bar{Y}} = \frac{\mu_2 \bar{X}}{\mu_1 \bar{Y}} \sim F_{2m, 2n} \perp \mu_1, \mu_2 \text{ (} \perp \psi \text{)}$$

Thus  $\frac{\psi \bar{X}}{\bar{Y}}$  is a pivotal quantity

Exact 95% CI for  $\psi$  :

$$1 - \alpha = P(F_{2m, 2n, \alpha_1} < \frac{\psi \bar{X}}{\bar{Y}} < F_{2m, 2n, 1-\alpha_2})$$

$$95\% \text{ CI} = P\left(\frac{\bar{Y}}{\bar{X}} F_{2m, 2n, \alpha_1} < \psi < \frac{\bar{Y}}{\bar{X}} F_{2m, 2n, 1-\alpha_2}\right)$$