

HW 4

a Sample Mean  $\bar{x} = \frac{\sum x_i}{n} = \frac{T}{n}$

$$M_{\bar{x}} = M_{x_1} + x_2 + \dots + x_n (t)$$

$$= M_{x_1} \cdot M_{x_2} \cdot \dots \cdot M_{x_n} \left( \frac{t}{n} \right)$$

Bcs it's independent,

$$= M_{x_1} \cdot M_{x_2} \cdot \dots \cdot M_{x_n} \left( \frac{t}{n} \right)$$

$$= \left( 1 - B \frac{t}{n} \right)^{-\alpha} \cdot \left( 1 - B \frac{t}{n} \right)^{-\alpha} \cdot \dots \cdot \left( 1 - B \frac{t}{n} \right)^{-\alpha}$$

$$= \left( 1 - B \frac{t}{n} \right)^{-n\alpha}$$

Let  $\bar{x} = \frac{2tn}{B}$

$$M_{\bar{x}} \left( \frac{2tn}{B} \right) = \left( 1 - B \cdot \frac{2tn}{Bn} \right)^{-n\alpha} = (1 - 2t)^{-n\alpha}$$

This follows pdf of  $\Gamma(n\alpha, 2)$

then, the degree of freedom of this transformation is  $2n\alpha$ .

b  $x$  has uniform distribution  $(0,1)$

$$\begin{bmatrix} x \\ x^2 \end{bmatrix} = \begin{pmatrix} \mu_x \\ \mu_{x^2} \end{pmatrix}$$

$$\mu_x = \frac{a+b}{2} = \frac{0+1}{2} = \frac{1}{2} \quad E(x) = \frac{1}{2} \quad \text{Var}(x) = \frac{1}{12}$$

$$\text{Var}(x) = E(x^2) - [E(x)]^2$$

$$\frac{1}{12} = E(x^2) - \frac{1}{4}$$

$$E(x^2) = \frac{1}{12} + \frac{1}{4} = \frac{4}{12} = \frac{1}{3}$$

$$\begin{bmatrix} x \\ x^2 \end{bmatrix} = \begin{pmatrix} \frac{1}{2} \\ \frac{1}{3} \end{pmatrix}$$

$$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \rightarrow \begin{pmatrix} \sigma_1^2 & \sigma_1 \sigma_2 \\ \sigma_2 \sigma_1 & \sigma_2^2 \end{pmatrix}$$

$$\text{Cov}(x, x^2) = E[x x^2] - E(x)E(x^2) = E[x^3] - E(x)E(x^2)$$

$$E[x^3] = \int g(x) f(x) dx \text{ pdf}$$

$$= \int_0^1 x^3 \cdot 1 = \frac{x^4}{4} \Big|_0^1 = \frac{1}{4}$$

$$\text{Cov}(x, x^2) = \frac{1}{4} - \left( \frac{1}{2} \right) \left( \frac{1}{3} \right) = \frac{3-2}{12} = \frac{1}{12}$$

$$\text{Var}(x^2) = E(x^4) - (E(x^2))^2$$

$$E(x^4) = \int_0^1 x^4 = \frac{1}{5} x^5 = \frac{1}{5}$$

$$\text{Var}(x^2) = \frac{1}{5} - \left(\frac{1}{3}\right)^2 = \frac{9-5}{45} = \frac{4}{45}$$

$$\begin{pmatrix} x \\ x^2 \end{pmatrix} \rightarrow \begin{pmatrix} 1/12 & 1/12 \\ 1/12 & 4/45 \end{pmatrix}$$

$$C \quad Y = 2 \cdot x_1^{1/2} \cdot x_2^{1/2}$$

$$E(Y) = 2 E(x_1^{1/2}) E(x_2^{1/2})$$

$$E(x_1^{1/2}) = \frac{\Gamma(\alpha + \frac{1}{2}) B^{1/2}}{\Gamma(\alpha)}$$

$$\hookrightarrow \Gamma(\alpha, 1) \rightarrow E(x_1^{1/2}) = \frac{\Gamma(\alpha + \frac{1}{2})}{\Gamma(\alpha)}$$

$$\hookrightarrow \Gamma(\alpha + \frac{1}{2}, 1) \rightarrow E(x_2^{1/2}) = \frac{\Gamma(\alpha + 1)}{\Gamma(\alpha + \frac{1}{2})}$$

$$\begin{aligned} E(Y) &= 2 \cdot \frac{\Gamma(\alpha + \frac{1}{2})}{\Gamma(\alpha)} \cdot \frac{\Gamma(\alpha + 1)}{\Gamma(\alpha + \frac{1}{2})} = 2 \frac{\Gamma(\alpha + 1)}{\Gamma(\alpha)} = 2 \cdot \frac{\alpha!}{(\alpha-1)!} \\ &= 2 \cdot \frac{\alpha(\alpha-1)!}{(\alpha-1)!} = 2\alpha \end{aligned}$$

$$\text{Var}(y) = E(y^2) - E(y)^2$$

$$y^2 = 4 x_1 x_2$$

$$E(y^2) = 4 E(x_1) E(x_2)$$

$$\hookrightarrow E(x_1) = \alpha \quad E(x_2) = \alpha + \frac{1}{2}$$

$$E(y^2) = 4 \cdot (\alpha^2 + \frac{1}{2}\alpha) = 4\alpha^2 + 2\alpha = 2\alpha(2\alpha + 1)$$

$$\text{Var}(y) = 4\alpha^2 + 2\alpha - 4\alpha^2 = 2\alpha$$



d. Joint distribution:

$$\begin{aligned} M_{\bar{x}, \bar{y}}(s, t) &= E e^{\bar{x}s + \bar{y}t} = E e^{(x_1 + x_2 + \dots + x_n) \frac{s}{n} + (y_1 + y_2 + \dots + y_n) \frac{t}{n}} \\ &= E e^{\frac{x_1 s}{n} + \frac{y_1 t}{n}} \cdot E e^{\frac{x_2 s}{n} + \frac{y_2 t}{n}} \cdot \dots \cdot E e^{\frac{x_n s}{n} + \frac{y_n t}{n}} \\ &= E e^{(x_1 y_1) \left( \frac{s}{n} \right)} \dots E e^{(x_n y_n) \left( \frac{s}{n} \right)} \\ &= \left[ E e^{(x_1 y_1) \left( \frac{s}{n} \right)} \right]^n \end{aligned}$$

$$\text{From } M_{y_i}(t) = E e^{t y_i} = E e^{(\sum t_i y_i)} = e^{t' \mu + \frac{1}{2} t' \Sigma t'}$$

$$\begin{aligned} &= \left( e^{(\frac{s}{n} \frac{t}{n}) \left( \begin{matrix} \mu_1 \\ \mu_2 \end{matrix} \right) + \frac{1}{2} \left( \frac{s}{n} \frac{t}{n} \right) \Sigma \left( \frac{s}{n} \frac{t}{n} \right)} \right)^n \\ &= e^{(s \ t) \left( \begin{matrix} \mu_1 \\ \mu_2 \end{matrix} \right) + \frac{1}{2} \left( \frac{s}{n} \frac{t}{n} \right) \Sigma \left( \frac{s}{n} \frac{t}{n} \right)} \\ &= e^{(s \ t) \left( \begin{matrix} \mu_1 \\ \mu_2 \end{matrix} \right) + \frac{1}{2} (s \ t) \left( \frac{\Sigma}{n} \right) \left( \frac{s}{n} \frac{t}{n} \right)} \end{aligned}$$

$$\text{Thus, } \bar{x}, \bar{y} \sim N \left[ \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix}, \left( \frac{\Sigma}{n} \right) \right]$$

$$(x - \mu)' \Sigma^{-1} (x - \mu) \sim \chi_{n-2}^2$$

$$n (\bar{x} - \mu_1, \bar{y} - \mu_2) \Sigma^{-1} \begin{pmatrix} \bar{x} - \mu_1 \\ \bar{y} - \mu_2 \end{pmatrix}$$

$$= (\bar{x} - \mu_1, \bar{y} - \mu_2) \left( \frac{\Sigma}{n} \right)^{-1} \begin{pmatrix} \bar{x} - \mu_1 \\ \bar{y} - \mu_2 \end{pmatrix} \rightarrow \text{degree of freedom } 2$$