

Lee 18

Nov. 09

$X_1 \sim N(0,1)$ ind of $X_2 \sim N(0,1)$

$$R = \frac{X_1}{X_2} = \int_{\text{supp } X_1} |x| f(x r) f(x) dx$$

$$= \int_{\mathbb{R}} |x| \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2 r^2}{2}} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx$$

$$= \frac{1}{2\pi} \int_{\mathbb{R}} |x| e^{-\frac{1}{2} x^2 (r^2 + 1)} dx = \frac{1}{2\pi} \left(\int_{-\infty}^0 (-x) e^{-\frac{1}{2} x^2 (r^2 + 1)} dx + \int_0^{\infty} x e^{-\frac{1}{2} x^2 (r^2 + 1)} dx \right)$$

$$= \frac{1}{\pi} \int_0^{\infty} x e^{-\frac{1}{2} x^2 (r^2 + 1)} dx = \frac{1}{\pi} \int_0^{\infty} x e^u - \frac{1}{x(r^2 + 1)} du$$

with $u = -\frac{1}{2} x^2 (r^2 + 1)$

$$\frac{du}{dx} = -x(r^2 + 1)$$

$$dx = -\frac{1}{x(r^2 + 1)} du$$

$$= \frac{1}{\pi(r^2 + 1)} \left[-e^u \right]_0^{\infty}$$

$$= \frac{1}{\pi(r^2 + 1)} = \text{Cauchy}(0,1)$$

must derm \uparrow

Final

$$X_1, X_2, \dots, X_n \sim f(\mu, \sigma)$$

μ, σ are unknown and we want to estimate them (inference)

\bar{X} is the average r.v. (Estimator)

\bar{x} is a realization (Estimate)

$$E[\bar{X}] = \mu$$

S^2 is the sample variance $S^2 = \frac{1}{n-1} \sum (x_i - \bar{x})^2$
is a realization

$$E[S^2] = \sigma^2$$

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\mu, \sigma^2)$$

$$\bar{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right) \quad S^2 = \frac{1}{n-1} \left((x_1 - \bar{x})^2 + \dots + (x_n - \bar{x})^2 \right)$$

$$S^2 \sim \chi^2$$

$$\vec{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

$$\sum_{i=1}^n x_i^2 = \vec{x}^T \vec{x} \sim \chi_n^2$$

$$\sum \frac{z_i^2}{\sigma^2} = \sum \left(\frac{x_i - \mu}{\sigma} \right)^2 = \frac{\sum (x_i - \mu)^2}{\sigma^2} \sim \chi^2_n$$

$$\sum (x_i - \mu)^2 = \sum (x_i - \bar{x})^2 + n(\bar{x} - \mu)^2$$

$$\sum (x_i - \mu)^2 = \sum (x_i - \bar{x})^2 + n(\bar{x} - \mu)^2$$

$$= \sum (x_i - \bar{x})^2 + 2(\sum x_i - \bar{x}) (\bar{x} - \mu) + n(\bar{x} - \mu)^2$$

$$= \sum (x_i - \bar{x})^2 + 2(\sum x_i \bar{x} - \mu \sum x_i - \sum \bar{x}^2 + \sum \bar{x} \mu) + n(\bar{x} - \mu)^2$$

$$\Rightarrow \frac{\sum (x_i - \bar{x})^2}{\sigma^2} + \frac{n(\bar{x} - \mu)^2}{\sigma^2} \sim \chi^2_n$$

$$\left(\frac{x - \mu}{\frac{\sigma}{\sqrt{n}}} \right)^2 \sim N^2(1) \sim \chi^2_1$$

$$\text{Conj. } \frac{(n-1)s^2}{\sigma^2} \sim \chi^2_{n-1}$$

Cochran's Theorem

$$z_1, \dots, z_n \stackrel{\text{i.i.d.}}{\sim} N(0, 1)$$

let Q_1, \dots, Q_k be scalar r.v.
in the quadratic forms.

$$Q_j = \vec{z}^T B_j \vec{z} \quad \text{when } B_1, \dots, B_k \text{ are positive semi-definite.}$$

a) $n \geq \sum \text{Rank}(B_j)$

b) Q_j 's are independent

c) $Q_j \sim \chi^2_{\text{Rank}(B_j)}$

$$\vec{z}^T \vec{z} = Q_1 + \dots + Q_k$$

$$\downarrow = \vec{z}^T B_1 \vec{z} + \vec{z}^T B_2 \vec{z} + \dots + \vec{z}^T B_k \vec{z}$$

$$\sim \chi^2_n \quad \vec{z}^T \vec{z} = \vec{z}^T (B_1 + \dots + B_k) \vec{z}$$

$$\Rightarrow B_n = B_1 + \dots + B_k$$

$$\bar{z} z_i = ((z_i - \bar{z}) + (\bar{z}))^2$$

$$= (z_i - \bar{z})^2 + 2(z_i - \bar{z})\bar{z} + \bar{z}^2$$

$$2(z_i \bar{z} - \bar{z}^2)$$

$$\sum \bar{z} z_i = \underbrace{\sum (z_i - \bar{z})^2}_{Q_1} + \underbrace{n \bar{z}^2}_{Q_2}$$

$$Q_2 = n \bar{z}^2 = \vec{z}^T B_2 \vec{z} = \vec{z}^T \left(\frac{1}{n} J_n \right) \vec{z}$$

$$= \vec{z}^T \begin{bmatrix} \frac{1}{n} & \dots & \frac{1}{n} \\ \vdots & & \vdots \\ \frac{1}{n} & \dots & \frac{1}{n} \end{bmatrix} \vec{z} = \bar{z}^2$$

$$J_n = \begin{bmatrix} 1 & \dots & 1 \\ \vdots & & \vdots \\ 1 & \dots & 1 \end{bmatrix}_{n \times n}$$

Theorem: Matrix A is symmetric independent $\rightarrow \text{tr}(A) = \text{rank}(A)$

Matrix A is positive semi def.
if $\forall \vec{v}, \dots, \vec{v}^T A \vec{v} \geq 0$

~~Handwritten text~~

$$n \bar{z}^2 \sim \chi_1^2$$

ind. of

$$\sum (x_i - \bar{x})^2 \sim \chi_{n-1}^2$$

$$\frac{n(\bar{x} - \mu)^2}{\sigma^2} = \frac{1}{\sigma} (\bar{x} - \mu)^T \left(\frac{1}{n} J_n \right) \frac{1}{\sigma} (\bar{x} - \mu) \sim \chi_1^2$$

$$\frac{\sum (x_i - \bar{x})^2}{\sigma^2} = \frac{1}{\sigma} (\bar{x} - \mu)^T \left(I_n - \frac{1}{n} J_n \right) \frac{1}{\sigma} (\bar{x} - \mu) \sim \chi_{n-1}^2$$

$$= \frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2$$

$$\text{with } S^2 \sim \frac{\sigma^2}{n-1} \chi_{n-1}^2 = \text{Gamma} \left(\frac{n-1}{2}, \frac{n-1}{2\sigma^2} \right)$$

S^2, \bar{x} are independent

$$\frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}} \sim N(0,1) \quad \text{"z test"}$$

T-test

$$\frac{\bar{X} - \mu}{\frac{s}{\sqrt{n}}} = \frac{\bar{X} - \mu}{\frac{1}{\sqrt{n}} \sqrt{s^2}} = \frac{\bar{X} - \mu}{\frac{1}{\sqrt{n}} \sqrt{\frac{\sigma^2}{n-1} s^2}}$$

$$= \frac{\frac{\bar{X} - \mu}{\frac{s}{\sqrt{n}}}}{\sqrt{\frac{\frac{n-1}{\sigma^2} s^2}{n-1}}} = \frac{z}{\sqrt{\frac{\chi^2}{n-1}}} = T\text{-test}$$