

Lecture 18

11/9/17

$$X_1 \sim N(0,1)$$

and of

$$X_2 \sim N(0,1)$$

$$R = \frac{X_1}{X_2} = \int_{\sup[x]} |x| f_{X_1}(xr) f_{X_2}(x) dx$$

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

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

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

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

$$u = -\frac{1}{2}x^2(r^2+1) \quad x=0 \Rightarrow u=0$$

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

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

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

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

Problem 2 End  
Here

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

$\bar{X}$  is the average  $n, \sqrt{\bar{X}} = \frac{X_1 + \dots + X_n}{n}$  realisation  
 "Estimator" "estimate"

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

$S^2$  is a sample variance  
 random variable  
 "Estimator"

$$s^2 = \frac{1}{n-1} \sum (x_i - \bar{x})^2$$

is realization  
 "estimate"

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

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

$$\bar{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right)$$

What is  $S^2 \sim$ ?

$$S^2 = \frac{1}{n-1} \left( (X_1 - \bar{X})^2 + \dots + (X_n - \bar{X})^2 \right)$$

$$\vec{Z} = \begin{bmatrix} z_1 \\ \vdots \\ z_n \end{bmatrix}, \quad \sum_{i=1}^n z_i^2 = \vec{Z}^T \vec{Z} \sim \chi_n^2$$

$$\sum z_i^2 = \sum \left( \frac{X_i - \mu}{\sigma} \right)^2 = \frac{\sum (X_i - \mu)^2}{\sigma^2} \sim \chi_n^2$$



$$\begin{aligned}
\sum (x_i - \mu)^2 &= \sum ((x_i - \bar{x}) + (\bar{x} - \mu))^2 \\
&= \sum (x_i - \bar{x})^2 + 2(x_i - \bar{x})(\bar{x} - \mu) + (\bar{x} - \mu)^2 \\
&= \sum (x_i - \bar{x})^2 + 2 \left( \sum x_i \bar{x} - \mu \sum x_i - \sum \bar{x}^2 + \sum \bar{x} \mu \right) \\
&\quad \quad \quad n\bar{x}^2 - \mu n\bar{x} - n\bar{x}^2 + n\bar{x}\mu + n(\bar{x} - \mu)^2 \\
&= \frac{\sum (x_i - \bar{x})^2}{\sigma^2} + \frac{n(\bar{x} - \mu)^2}{\sigma^2} \sim \chi_n^2 \\
&\quad \quad \quad \parallel \quad \quad \quad \parallel \\
&\quad \quad \quad \frac{(n-1)s^2}{\sigma^2} \sim \chi_{n-1}^2 \quad \left( \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}} \right)^2 \sim \chi_1^2
\end{aligned}$$

### Cochran's Theorem (1934)

$$Z_1, \dots, Z_n \stackrel{i.i.d.}{\sim} N(0, 1)$$

let  $Q_1, \dots, Q_k$  be scalar r.v.s with the quadratic form

$$Q_j = \vec{Z}^T B_j \vec{Z} \quad \text{where } B_1, \dots, B_k \text{ are}$$

positive semi-definite matrices such that

$$\begin{aligned}
(a) \quad n &= \sum \text{rank}(B_j) \quad \chi_n^2 & \vec{Z}^T \vec{Z} &= Q_1 + \dots + Q_k \\
&\quad \quad \quad \uparrow & &= \vec{Z}^T B_1 \vec{Z} + \dots + \vec{Z}^T B_k \vec{Z} \\
(b) \quad Q_j \text{'s are independent} & & &= \vec{Z}^T (B_1 + \dots + B_k) \vec{Z} \\
&\quad \quad \quad \uparrow & &\Rightarrow I_n = B_1 + \dots + B_k \\
(c) \quad Q_j &\sim \chi_{\text{rank}(B_j)}^2
\end{aligned}$$

Defn: Positive semi-definite Matrix  $A$  is positive semi-definite if  $\forall \vec{v}$   
 $\vec{v}^T A \vec{v} \geq 0$



$$S(\mu - \bar{y})$$

$$(x^2 + x - 5)(x^2 + x - 5)$$

$$(u-x)$$

2021.11.11

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$$\sqrt{7} \neq 0$$

$$C \leq A$$

[illegible]

7-24

$$5.25 +$$

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Q.E.D.

$$\frac{(n-1)S^2}{\sigma^2} \sim \chi^2_{n-1} \Rightarrow$$

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

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

$$\frac{\bar{X} - \mu}{\frac{s}{\sqrt{n}}} = \frac{\bar{X} - \mu}{\frac{1}{\sqrt{n}} \sqrt{s^2}} = \frac{\bar{X} - \mu}{\frac{1}{\sqrt{n}} \sqrt{\frac{\sigma^2}{n-1} \chi^2_{n-1}}} \text{ "Student 1908"}$$

$$= \frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \sim N(0,1)$$

$$\sqrt{\frac{\frac{n-1}{\sigma^2} S^2}{n-1}} \sim \chi^2_{n-1}$$

$$= \frac{Z}{\sqrt{\frac{\chi^2_{n-1}}{n-1}}} = T_{n-1}$$

Both independent



$$\sum (z_i - \bar{z})^2 = \vec{Z}^T B_1 \vec{Z}$$

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

$$= \sum z_i^2 - 2n\bar{z}^2 + n\bar{z}^2$$

$$= \sum z_i^2 - n\bar{z}^2$$

$$= \vec{Z}^T I_n \vec{Z} - \vec{Z}^T \frac{1}{n} J_n \vec{Z} = \vec{Z}^T (I_n - \frac{1}{n} J_n) \vec{Z} = B_2$$

$$\begin{bmatrix} 1 & & & 0 \\ & 1 & & \\ & & \ddots & \\ 0 & & & 1 \end{bmatrix} - \begin{bmatrix} \frac{1}{n} & \dots & \frac{1}{n} \\ \vdots & \ddots & \vdots \\ \frac{1}{n} & \dots & \frac{1}{n} \end{bmatrix} = \begin{bmatrix} 1-\frac{1}{n} & -\frac{1}{n} & \dots & -\frac{1}{n} \\ \vdots & \ddots & \ddots & \vdots \\ -\frac{1}{n} & \dots & 1-\frac{1}{n} \end{bmatrix}$$

Proven:  $n\bar{z}^2 \sim \chi_1^2$  iid of  $\sum (z_i - \bar{z})^2 \sim \chi_{n-1}^2$

$$\frac{n(\bar{X} - \mu)^2}{\sigma^2} = \frac{1}{\sigma^2} (\bar{X} - \mu) \underbrace{\left( \frac{1}{n} J_n \right)}_{\vec{Z}^T} \frac{1}{\sigma^2} (\bar{X} - \mu) \sim \chi_1^2$$

$$\frac{\sum (X_i - \bar{X})^2}{\sigma^2} = \frac{1}{\sigma^2} (\bar{X} - \mu)^T (I_n - \frac{1}{n} J_n) \frac{1}{\sigma^2} (\bar{X} - \mu) \sim \chi_{n-1}^2$$

$$\Rightarrow \frac{n(\bar{X} - \mu)^2}{\sigma^2} \text{ ind of } \frac{(n-1)S^2}{\sigma^2} \Rightarrow \bar{X}, S^2 \text{ are independent}$$

"Fisher 1935"

Gerny, 1936 Proved this is unique to the normal