

Last Lecture! → Next week review.

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Recall:  $\chi^2$ -distribution.

$X_1, X_2, \dots, X_n$  iid  $N(0, 1)$

HW 9



$$Y = X_1^2 + X_2^2 + \dots + X_n^2 \sim \chi^2(n) = \text{Gamma}\left(\frac{n}{2}, \frac{1}{2}\right)$$

"Chi-squared w/  $n$  degrees of freedom"

Useful because  $X_1, X_2, \dots, X_n \sim N(\mu, \sigma^2)$

$$Q = \frac{(n-1)}{\sigma^2} s^2 = \frac{1}{\sigma^2} \sum_{i=1}^n (X_i - \bar{X})^2 \sim \chi^2(n-1)$$

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$t$ -distribution. (Student's  $t$ -distribution).

↳ Discovered by William Gosset (1908)  
- under the pseudonym "student".

Def Let  $Z \sim N(0, 1)$ ,  $Y \sim \chi^2(n)$

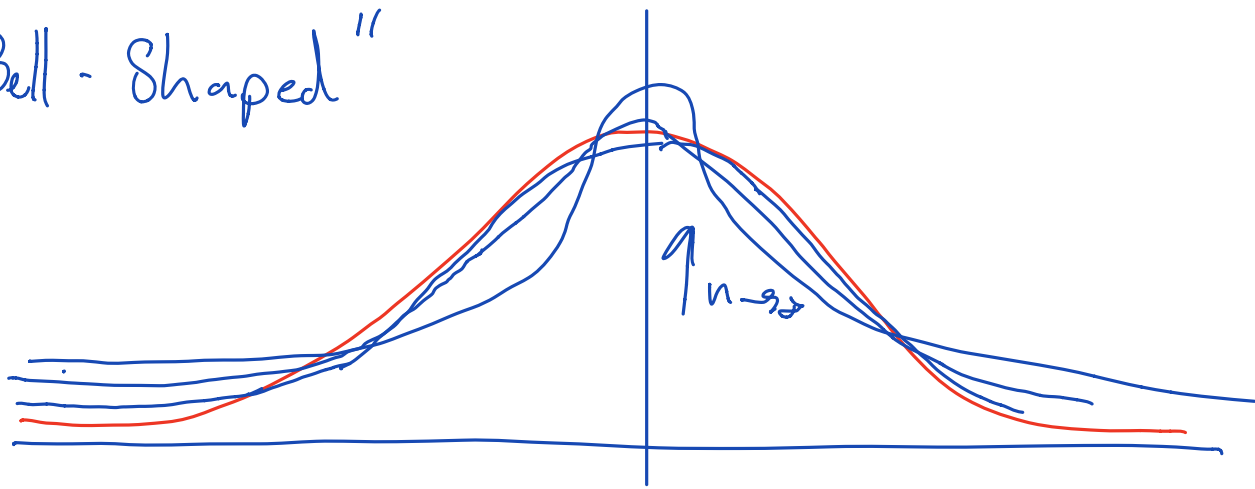
then  $T = \frac{Z}{\sqrt{Y/n}} \sim T(n)$

is a random variable satisfy a  $t$ -distribution with  $n$ -degree's of freedom.

Properties: ① PDF

$$f_T(t) = \frac{\Gamma(\frac{n+1}{2})}{\sqrt{n\pi} \Gamma(\frac{n}{2})} \left(1 + \frac{t^2}{n}\right)^{-\frac{n+1}{2}} \quad -\infty < t < \infty.$$

"Bell-Shaped"

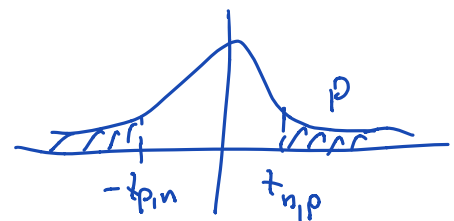


②  $E T = 0$  ,  $Var(T) = \frac{n}{n-2}$   
 $n > 1$  ,  $n > 2$

③ CLT  $T(n) \xrightarrow{d} N(0, 1) \quad n \rightarrow \infty$

④ Let  $t_{p,n}$  be such that.

$$P(T > t_{p,n}) = p$$



Symmetry  $t_{1-p,n} = -t_{p,n}$

MATLAB

$$t_{p,n} = F_{T(n)}^{-1}(1-p) = \text{tinv}(1-p, n)$$

Also - table on course webpage.

Main Application.

$$X_1, X_2, X_3, \dots, X_n \sim N(\mu, \sigma^2)$$

Find confidence interval for  $\mu$ .

$$T = \frac{\bar{X} - \mu}{S/\sqrt{n}} \sim T(n-1)$$

↑  
note are-less.

Proof

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

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

So,

$$\frac{\bar{X} - \mu}{S/\sqrt{n}} = \frac{Z}{\sqrt{Y/n-1}} \sim T(n-1).$$

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Find  $(1-\alpha)\%$  confidence interval for  $\mu$ .

$$\textcircled{1} \quad Q = \frac{\bar{X} - \mu}{S/\sqrt{n}} \sim T(n-1) \quad \div \text{Pivot.}$$

$$\textcircled{2} \quad P\left(-t_{\alpha/2, n-1} \leq Q \leq t_{\alpha/2, n-1}\right) = 1 - \alpha$$
$$\hookrightarrow t_{\alpha/2, n-1} = F_{T(n-1)}^{-1}\left(1 - \frac{\alpha}{2}\right)$$

$\textcircled{3}$  Algebra:

$$P\left(\bar{X} - \frac{t_{\alpha/2, n-1} S}{\sqrt{n}} \leq \mu \leq \bar{X} + \frac{t_{\alpha/2, n-1} S}{\sqrt{n}}\right) = 1 - \alpha$$

$$\left[\bar{X} - \frac{t_{\alpha/2, n-1} S}{\sqrt{n}}, \bar{X} + \frac{t_{\alpha/2, n-1} S}{\sqrt{n}}\right], \quad 1 - \alpha \% \text{ confidence interval.}$$

Ex: A farmer weigh 10 watermelons.

7.72, 9.58, 12.38, 2.77, 11.27

8.70, 11.10, 7.80, 10.17, 6.00

Assume weights are normally distributed.

95% confidence interval on  $\mu, \sigma^2$ ?

① Numerically calculate.

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i = 9.26, \quad S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2 = 3.96.$$

②  $n = 10$ ,  $\alpha = 1 - .95 = 0.05$ ,  $\alpha/2 = 0.025$ .

$$t_{\alpha/2, n-1} = 2.262$$

$$\chi_{\alpha/2, n-1}^2 = 19.023$$

$$\chi_{1-\alpha/2, n-1}^2 = 2.700$$

Mean:  $\mu \in \left[ \bar{X} - \frac{t_{\alpha/2, n-1} S}{\sqrt{n}}, \bar{X} + \frac{t_{\alpha/2, n-1} S}{\sqrt{n}} \right]$

$$= \left[ 9.26 - \frac{2.266 \sqrt{3.96}}{\sqrt{10}}, 9.26 + \frac{2.266 \sqrt{3.96}}{\sqrt{10}} \right]$$
$$= [7.834, 10.686] \quad 95\% \text{ confidence}$$

Variance  $\sigma^2: \left[ \frac{(n-1)S^2}{\chi_{\alpha/2, n-1}^2}, \frac{(n-1)S^2}{\chi_{1-\alpha/2, n-1}^2} \right]$

$$= \left[ \frac{9(3.96)}{19.023}, \frac{9(3.96)}{2.700} \right]$$
$$= [1.874, 13.2]$$

What about other distributions?

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↳ Can be challenging. → Here is an example

Ex  $X_1, X_2, \dots, X_n \sim \text{Exponential}(\theta)$

What is a  $(1-\alpha)\%$  confidence interval for  $\theta$ ?

Recall  $E X_i = 1/\theta$   $f_X(x) = \begin{cases} \theta e^{-\theta x} & x \geq 0 \\ 0 & \text{otherwise} \end{cases}$

What is an estimator for  $\theta$ ?

MLE

$$L(x_1, x_2, \dots, x_n; \theta) = \prod_{i=1}^n \theta e^{-\theta x_i}$$

$$\ln(L(x_1, x_2, \dots, x_n; \theta)) = n \ln \theta - \theta \sum_{i=1}^n x_i$$

$$\frac{\partial}{\partial \theta} \rightarrow \frac{n}{\theta} - \sum_{i=1}^n x_i = 0$$

$$\Rightarrow \hat{\theta} = \frac{1}{\frac{1}{n} \sum_{i=1}^n x_i} = \frac{1}{\bar{X}}$$

Confidence interval?

Pivot,  $X_1 + X_2 + \dots + X_n \sim \text{Gamma}(n, \theta)$

$\frac{X_i}{\theta} \sim \text{Exponential}(1)$

$$Q = \frac{X_1 + X_2 + \dots + X_n}{\theta} = \frac{n\bar{X}}{\theta} \sim \text{Gamma}(n, 1)$$

Defn.  $\gamma_{p,n}$   $P(Q \geq \gamma_{p,n}) = p$

$$\gamma_{p,n} = F_{\Gamma(n,1)}^{-1}(1-p) = \text{gaminv}(1-p, n)$$

↑  
MATLAB.

$$P(\gamma_{1-\alpha/2, n} \leq Q \leq \gamma_{\alpha/2, n}) = 1 - \alpha$$

Algebra

$$\gamma_{1-\alpha/2, n} \leq \frac{n\bar{X}}{\theta} \leq \gamma_{\alpha/2, n}$$

$$\theta \leq \frac{n\bar{X}}{\gamma_{1-\alpha/2, n}}, \quad \theta \geq \frac{n\bar{X}}{\gamma_{\alpha/2, n}}$$

$$P\left(\frac{n\bar{X}}{z_{\alpha/2, n}} \leq \theta \leq \frac{n\bar{X}}{z_{1-\alpha/2, n}}\right) = 1 - \alpha.$$

↑  $(1 - \alpha)\%$  confidence interval  
for  $\theta$ .