

## Lec 2/8

Wednesday, February 8, 2017 15:00

Thm 11.8:  $X_1 \sim \text{Bin}(n_1, \theta_1)$   $X_2 \sim \text{Bin}(n_2, \theta_2)$ ,  $X_1, X_2$  independent,  $n_1, n_2$  large,  $\hat{\theta}_1 = \frac{X_1}{n_1}$   $\hat{\theta}_2 = \frac{X_2}{n_2}$ .

$$\hat{\theta}_1 - \hat{\theta}_2 \pm z_{\frac{\alpha}{2}} \sqrt{\frac{\hat{\theta}_1(1-\hat{\theta}_1)}{n_1} + \frac{\hat{\theta}_2(1-\hat{\theta}_2)}{n_2}} \quad \text{is a } (1-\alpha) \times 100\% \text{ CI for } \theta_1 - \theta_2$$

Ex: Wicking. Day 1: observe 10 wicking boxes in 100 ( $X_1$ )  
Day 2: observe 6 wicking boxes in 50 ( $X_2$ )

Find a 95% CI for  $\theta_2 - \theta_1$ :

$$0.12 - 0.1 \pm 1.96 \sqrt{\frac{0.1(0.9)}{100} + \frac{0.12(0.88)}{50}}$$

$$= (-0.0876, 0.1276)$$

Note: in example from last time,  $\hat{\theta}$  in getting number  $n$  required is a new  $\hat{\theta}$ .

### §11.6 estimation of variances

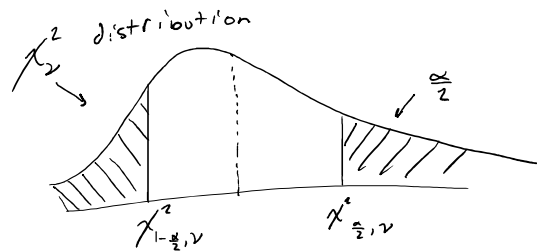
Recall:  $S^2$  sample variance;

$\frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2$  (chi-squared)  
depends on  $\sigma^2$ ,  $S^2$  doesn't depend on any parameter

Use pivot idea:

$$P\left(\chi_{1-\frac{\alpha}{2}, n-1}^2 < \frac{(n-1)S^2}{\sigma^2} < \chi_{\frac{\alpha}{2}, n-1}^2\right) = 1-\alpha$$

$$= P\left(\frac{(n-1)S^2}{\chi_{\frac{\alpha}{2}, n-1}^2} < \sigma^2 < \frac{(n-1)S^2}{\chi_{1-\frac{\alpha}{2}, n-1}^2}\right) = 1-\alpha$$



so a  $(1-\alpha) \times 100\%$  CI for  $\sigma^2$  is  $\left( \frac{(n-1)S^2}{\chi_{\frac{\alpha}{2}, n-1}^2}, \frac{(n-1)S^2}{\chi_{1-\frac{\alpha}{2}, n-1}^2} \right)$

to get a CI for  $\sigma$ , just take  $\sqrt{\quad}$  of endpoints

Ex: (usually assumption is that samples are iid from normal dist).

find a 95% CI for  $\sigma^2$  if a RS of 20 boxes gives  $S = 0.25$ .

from table,  $\chi^2_{\frac{0.05}{2}, 19} = 32.852$   $\chi^2_{1-\frac{\alpha}{2}, 19} = 8.907$

$$\Rightarrow 95\% \text{ CI} : \left( \frac{19(0.25)^2}{32.852}, \frac{19(0.25)^2}{8.907} \right) = (0.0361, 0.1333)$$

### §11.7 Estimation of Ratio of two variances.

Recall F distribution:  $U \sim \chi^2_{\nu_1}$   $V \sim \chi^2_{\nu_2}$

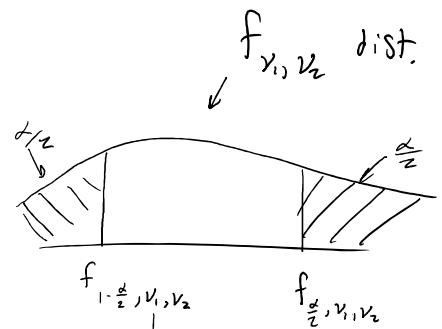
$F = \frac{U/\nu_1}{V/\nu_2}$  has an F distribution w/ d.f.s  $\nu_1$  and  $\nu_2$ .

Thm 8.15:  $F = \frac{S_1^2/\sigma_1^2}{S_2^2/\sigma_2^2} = \frac{S_1^2 \sigma_2^2}{S_2^2 \sigma_1^2} \sim f_{\nu_1-1, \nu_2-1}$

$$P\left(f_{1-\frac{\alpha}{2}, \nu_1-1, \nu_2-1} < \frac{\sigma_1^2 S_2^2}{\sigma_2^2 S_1^2} < f_{\frac{\alpha}{2}, \nu_1-1, \nu_2-1}\right) = 1-\alpha$$

$$\Rightarrow P\left(\frac{S_1^2}{S_2^2} \frac{1}{f_{\frac{\alpha}{2}, \nu_1-1, \nu_2-1}} < \frac{\sigma_1^2}{\sigma_2^2} < \frac{S_1^2}{S_2^2} f_{\frac{\alpha}{2}, \nu_1-1, \nu_2-1}\right)$$

using this.  $\bullet = \frac{1}{\bullet}$



So a  $(1-\alpha) \times 100\%$  CI for  $\frac{\sigma_1^2}{\sigma_2^2}$  is  $\left( \frac{S_1^2}{S_2^2} \frac{1}{f_{\frac{\alpha}{2}, \nu_1-1, \nu_2-1}}, \frac{S_1^2}{S_2^2} f_{\frac{\alpha}{2}, \nu_1-1, \nu_2-1} \right)$ .

if you want a CI for  $\frac{\sigma_1}{\sigma_2}$ , take  $\sqrt{\quad}$  of endpoints.

Ex: Variation in weights of dog food in two days.

Day 1:  $n_1 = 100$   $S_1 = 0.25$

Day 2:  $n_2 = 100$

find a 90% CI for  $\frac{\sigma_1}{\sigma_2}$ .

$$\text{Day 1: } n_1 = 100 \quad S_1 = 0.25$$

$$\text{Day 2: } n_2 = 50 \quad S_2 = 0.18$$

find a 90% for  $\frac{\sigma_1^2}{\sigma_2^2}$ .

$$\text{Note } f_{\frac{0.05}{2}, 19, 49} = 1.58.$$

$$\text{So } CI = \left( \frac{0.25^2}{0.18^2} \frac{1}{1.58}, \frac{0.25^2}{0.18^2} 1.58 \right) = (1.22, 2.894).$$