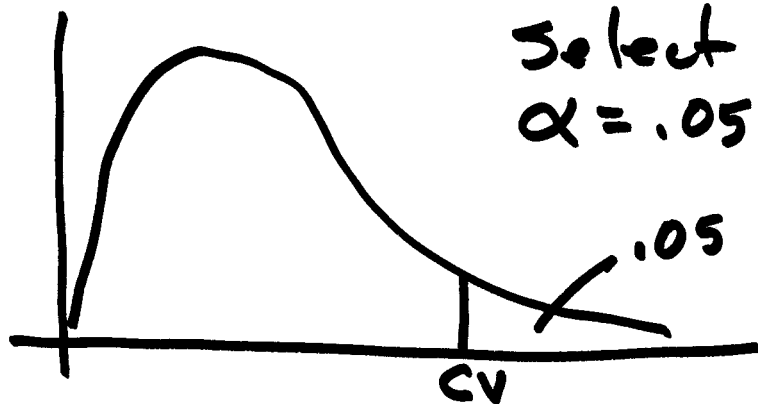


# Global Test

$$H_0: \sigma^2 = \sigma_0^2$$

$$H_1: \sigma^2 > \sigma_0^2$$

$$\chi^* \quad \frac{VTWV}{\sigma_0^2} \sim \chi_r^2 \quad \begin{array}{l} r: \text{redundancy} \\ r: \text{degrees of freedom} \end{array}$$



Select significance level  $\alpha = .05$

one  
sided

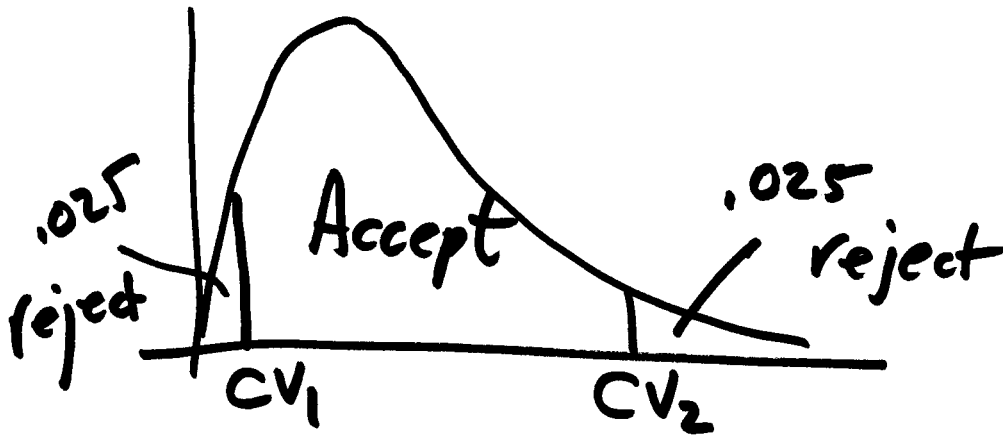
Probability of (Type I error) = .05  
 Prob (reject  $H_0$  when it's true)

two-sided test  $H_0: \sigma = \sigma_0$

vs.

$H_1: \sigma \neq \sigma_0$

$$\chi^* = \frac{\sqrt{TWU}}{\sigma_0^2} \leftarrow \text{"prior" value}$$



if  $CV_1 < \chi^* < CV_2$   
accept  $H_0$

otherwise  
reject

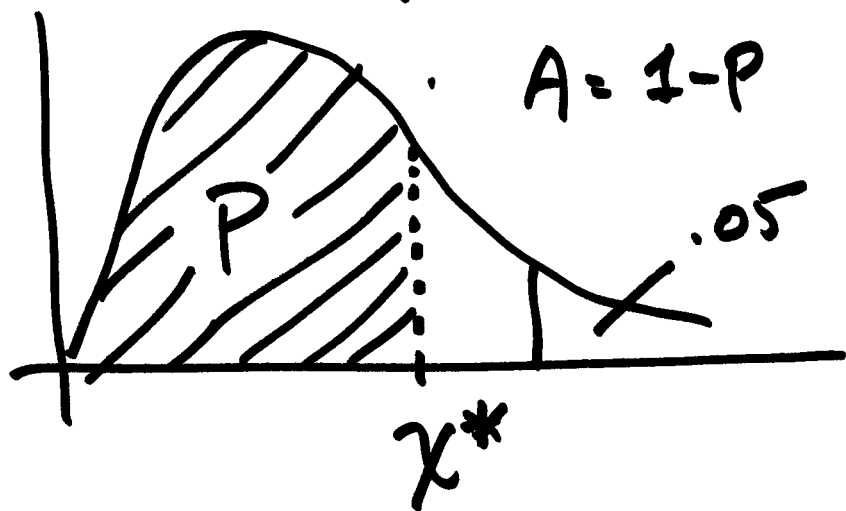
14-3

$$CV_1 = \text{icdf}('chi2', .025, r)$$

$$CV_2 = \text{icdf}('chi2', 1 - .025, r)$$

$1 - \alpha/2$

alternate way for decision: probability domain



reject if  $A < \alpha$

$$P = \text{cdf}('chi2', x^*, r)$$

if accept  $H_0$  in global test (F-test) <sup>14-4</sup>

use  $\sigma_0^2$  to scale  $Q$ 's

if reject  $H_0$  in global test

then use  $\hat{\sigma}_0^2 = \frac{V^T W V}{r}$  to scale  $Q$ 's

$$\Sigma_{xx} = \sigma_0^2 Q_{xx}$$

↓

$$\hat{\sigma}_0^2$$

$$\begin{array}{l}
 l_1 \quad \sigma_1 = 0.5 \\
 l_2 \quad \sigma_2 = 1.5 \\
 l_3 \quad \sigma_3 = 1.0
 \end{array}
 \left| \begin{array}{l}
 \sigma_0 = 0.5, \quad l_1: w = 1 \\
 w_1 = (0.5)^2 / (0.5)^2 = 1 \\
 w_2 = (0.5)^2 / (1.5)^2 = 0.111 \\
 w_3 = (0.5)^2 / 1^2 = 0.25
 \end{array} \right.$$

$$\sigma_0 = 1.5, \quad w_2 = 1$$

$$w_1 = (1.5)^2 / (0.5)^2 = 9$$

$$w_2 = (1.5)^2 / (1.5)^2 = 1$$

$$w_3 = (1.5)^2 / 1^2 = 2.25$$

$$\sigma_0 = 1.0, \quad w_3 = 1$$

$$w_1 = 4$$

$$w_2 = 0.444$$

$$w_3 = 1$$

$$H_0: \sigma = \sigma_0$$

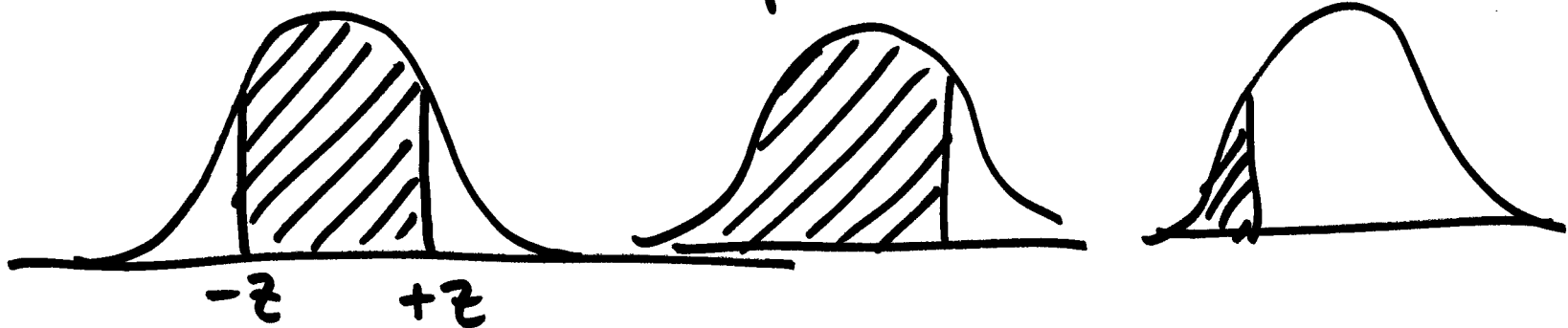
$$\chi^* = \frac{v^T w v}{\sigma_0^2}$$

$$, \quad \begin{bmatrix} \sigma_0^2 / \sigma_1^2 \\ \vdots \\ \sigma_0^2 / \sigma_n^2 \end{bmatrix} = \sigma_0^2 \cdot \Sigma^{-1} = w$$

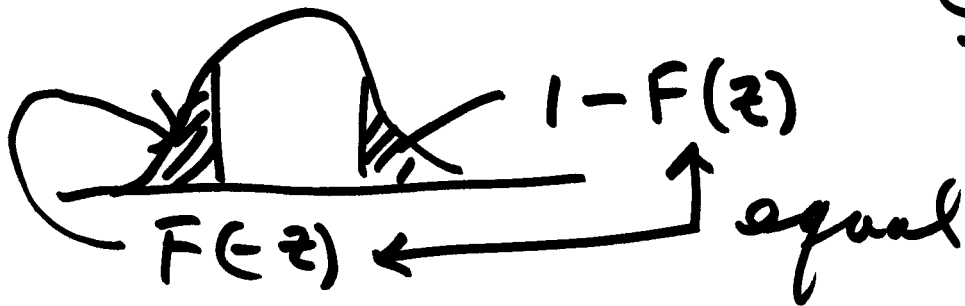
$$\chi^* = \frac{v^T \cdot \sigma_0^2 \Sigma^{-1} v}{\sigma_0^2} = v^T \Sigma^{-1} v$$

Confidence intervals

assume normality for observations



$$P(-z < RV < +z) = \overbrace{F(z) - F(-z)}^{\Phi(z) - \Phi(-z)}$$



$$F(z) - (1 - F(z))$$

$$2F(z) - 1$$

result of adjustment  $\hat{x}$

Standardize  $\hat{x}$  :  $z = \frac{\hat{x} - \mu_x}{\sigma_x}$

||  
mean 0,  $\sigma = 1$

14-8  
prior  
 $\sigma_0^2$   
passed  
test

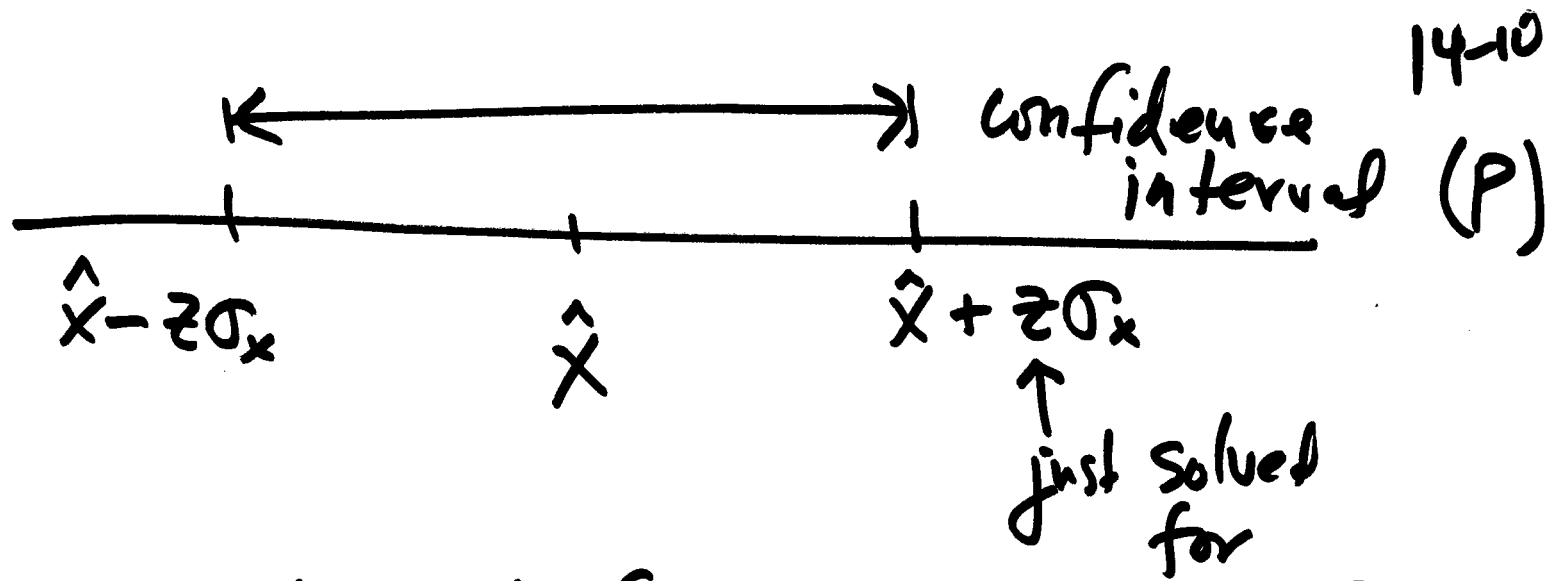
$$P\left(-z < \frac{\hat{x} - \mu_x}{\sigma_x} < +z\right) = \underline{2F(z) - 1}$$

$$P\left(-z\sigma_x < \hat{x} - \mu_x < +z\sigma_x\right)$$

$$P\left(-\hat{x} - z\sigma_x < -\mu_x < -\hat{x} + z\sigma_x\right)$$

$$P\left(\hat{x} + z\sigma_x > \mu_x > \hat{x} - z\sigma_x\right)$$





this derivation ok for use of prior  $\sigma_0^2$   
 passed global test

$$\frac{\hat{x} - \mu_x}{\hat{\sigma}_x} \sim t_r$$

$$\Sigma = \hat{\sigma}_0^2 Q_{xx}$$

$$\hat{\sigma}_0^2 = \frac{v^T w v}{r}$$

$$P\left(-t < \frac{\hat{x} - \mu_x}{\hat{\sigma}_x} < +t\right) = \frac{F(t) - F(-t)}{2F(t) - 1}$$

$$P\left(-t\hat{\sigma}_x < \hat{x} - \mu_x < t\hat{\sigma}_x\right) =$$

$$P\left(-\hat{x} - t\hat{\sigma}_x < -\mu_x < -\hat{x} + t\hat{\sigma}_x\right)$$

$$P\left(\hat{x} + t\hat{\sigma}_x > \mu_x > \hat{x} - t\hat{\sigma}_x\right)$$

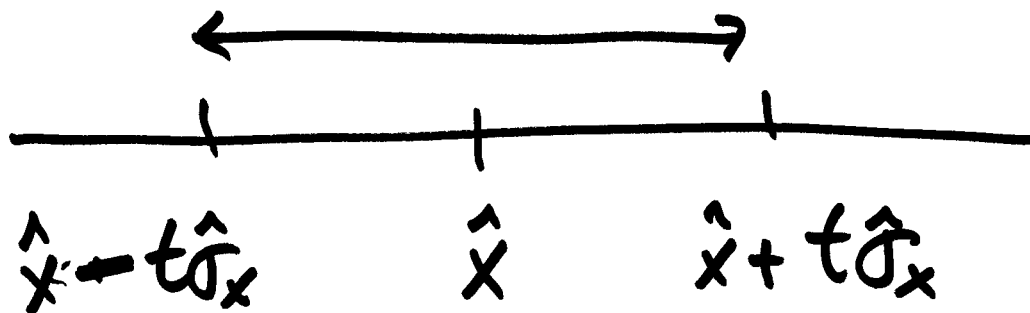
$$= 2F(t) - 1$$

(use this one  
if fail to  
pass global  
test)

14-12

select  $P$   
 solve  $P = 2F(t) - 1$ , for  $F(t)$   
 $F(t) = \frac{P+1}{2}$

$t = \text{icdf}('t', (P+1)/2, r)$   $\rightarrow$  degrees of freedom



recall

$$\hat{x} - z\sigma_x$$

$$\hat{x} + z\sigma_x$$