

$$V(\hat{\beta}) = \sigma^2 (X'X)^{-1} \Rightarrow$$

$\begin{matrix} k \times 1 & 1 \times k \\ k \times k \end{matrix}$

$$V(\hat{\beta}) = E \left[(\hat{\beta} - \beta) (\hat{\beta} - \beta)' \right]$$

$\begin{matrix} k \times 1 & 1 \times k \end{matrix}$

$$= \begin{pmatrix} V(\hat{\beta}_1) & \text{cov}(\hat{\beta}_1, \hat{\beta}_2) & \dots & \text{cov}(\hat{\beta}_1, \hat{\beta}_k) \\ \text{cov}(\hat{\beta}_2, \hat{\beta}_1) & V(\hat{\beta}_2) & \dots & \text{cov}(\hat{\beta}_2, \hat{\beta}_k) \\ \vdots & \vdots & \ddots & \vdots \\ \text{cov}(\hat{\beta}_k, \hat{\beta}_1) & \dots & \dots & V(\hat{\beta}_k) \end{pmatrix}$$

$\begin{matrix} k \times 1 & 1 \times k \\ k \times k \end{matrix}$

$$V(\hat{\beta}_1) = \left(\sigma^2 (X'X)^{-1} \right)_{(1,1)}$$

$$V(\hat{\beta}_2) = \left(\sigma^2 (X'X)^{-1} \right)_{(2,2)}$$

$$V(\hat{\beta}_k) = \left(\sigma^2 (X'X)^{-1} \right)_{(k,k)}$$

$$s^2 = \frac{1}{n-k} \sum_{t=1}^n \hat{\varepsilon}_t^2, \quad \hat{\varepsilon}_t = y_t - x_t \hat{\beta} \\ = y_t - \hat{y}_t$$

ESTIMATOR OF σ^2

$$\hat{\beta} = (X'X)^{-1} X'Y$$

$$\hat{Y} = X\hat{\beta}$$

$$\hat{\varepsilon} = Y - \hat{Y} \Rightarrow$$

$$\Rightarrow \hat{\varepsilon} = Y - X\hat{\beta}$$

$$S^2 = \frac{1}{n-k} \sum_{t=1}^n \varepsilon_t^2 = \frac{1}{n-k} \begin{matrix} 1 & 1 \\ \varepsilon & \varepsilon \\ 1 \times n & n \times 1 \end{matrix}$$

(n-k)

$$\varepsilon \sim \underline{N}(\underline{0}, \sigma^2 I_n)$$

$$\hat{\underline{\beta}} = \underline{\beta} + (X'X)^{-1}X'\underline{\varepsilon} \Rightarrow$$

$$\begin{aligned} \hat{\underline{\beta}} &\sim N \\ E(\hat{\underline{\beta}}) &= \underline{\beta} \\ V(\hat{\underline{\beta}}) &= \sigma^2 (X'X)^{-1} \end{aligned}$$

$\left(\begin{array}{l} \Delta \text{ΙΟΤ, ΣΥΝΑΡΤΗΣΗ} \\ \text{ΓΡΑΜΜΙΚΗ} \\ \varepsilon \sim N \end{array} \right)$
 ΤΟΥ

$\Rightarrow$

$$\hat{\underline{\beta}} \sim N(\underline{\beta}, \sigma^2 (X'X)^{-1})$$

$$E \Sigma T^0 \quad k=4$$

$$b = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{pmatrix} \quad \left| \begin{array}{l} k=4 \text{ I} \\ b_2 = 0 \\ b_3 + b_4 = 3 \\ v=2 \end{array} \right.$$

$$R b = q$$

$v_{k,k} \quad k=1, \quad v_{k,k}$

Υ ΓΡΑΜΜΙΚΟΥΣ ΠΕΡΙΟΡΕ.

$$R = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{pmatrix} = \begin{pmatrix} 0 \\ 3 \end{pmatrix}$$

$$b_1 - b_2 = 1$$

$$b_3 - 2b_1 = 5$$

$$b_3 + 3b_4 = 0$$

$$\Rightarrow \begin{pmatrix} 1 & -1 & 0 & 0 \\ -2 & 0 & 1 & 0 \\ 0 & 0 & 1 & 3 \end{pmatrix} \begin{pmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{pmatrix} = \begin{pmatrix} 1 \\ 5 \\ 0 \end{pmatrix}$$

$$b_1 = b_2 \Rightarrow (b_1 - b_2 = 0)$$

$$3b_1 = 7 - 2b_2 \Rightarrow 3b_1 + 2b_2 = 7 \mid \Rightarrow$$

$$\Rightarrow \begin{pmatrix} 1 & -1 & 0 & 0 \\ 3 & 2 & 2 & 0 \end{pmatrix} \begin{pmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{pmatrix} = \begin{pmatrix} 0 \\ 7 \end{pmatrix}$$

$V < K$ $\leftarrow$ ΑΡΙΘΜΟΣ ΕΧΟΜΕΝΩΝ ΜΕΤΑΒ.
↑
ΑΡΙΘΜΟΣ
ΓΡΑΜ. ΠΕΡΙΩΡ.

$$\hat{\beta}^1 \sim N(\beta, \sigma^2 (X'X)^{-1})$$

$$E(R\hat{\beta}^1) = R E(\hat{\beta}^1) = R\beta$$

$$V(R\hat{\beta}^1) = R V(\hat{\beta}^1) R' = \sigma^2 R (X'X)^{-1} R' \quad \Bigg| \Rightarrow$$

$$R\hat{\beta}^1 \sim N$$

$$R\hat{\beta} \sim N[R\beta, \sigma^2 R(X'X)^{-1}R']$$

$$X \sim N(\mu, A) \Rightarrow (X - \mu)' A^{-1} (X - \mu) \sim \chi^2_{r(A)}$$

$r(A) = \text{rank of } A$

$$\Rightarrow (R\hat{\beta} - R\beta)' [\sigma^2 R(X'X)^{-1}R']^{-1} (R\hat{\beta} - R\beta) \sim \chi^2_{r(R(X'X)^{-1}R')} = \chi^2_v$$

$$H_0: R\beta = q$$

$\sqrt{X'X} \quad \sqrt{X'X}$

$$(R\beta - q)' \left[\sigma^2 R (X'X)^{-1} R' \right]^{-1} (R\beta - q)$$

χ^2

$$\beta = (X'X)^{-1} X'y$$

$$\sum_{t=1}^n \varepsilon_t^2 = \varepsilon' \varepsilon$$

$$\hat{\varepsilon} = Y - X\hat{\beta} = Y - X(X'X)^{-1}X'Y =$$

$$= \underbrace{[I_n - X(X'X)^{-1}X']}_M Y = MY$$

ΞΕΡΟΥΜΕ ΟΤΙ Μ ΣΥΜΜΕΤΡΙΚΗ + ΤΑΥΤΟΔΥΝΑΜΗ
 $k \times k$ $\Gamma(M) = \text{tr}(M) = n - k$

$$M' \overset{\text{ΕΠΙΣΗΣ}}{X} = M X = (I_n - X(X'X)^{-1}X')X =$$

$$= X - X \underbrace{(X'X)^{-1}X'}_{I_K} X = X - X = 0$$

M ΟΡΘΟΓΩΝΙΑ ΤΗΣ X .

$$\varepsilon^1 = M Y \Rightarrow \varepsilon^1 = M (X\beta + \varepsilon) \Rightarrow$$

$$\Rightarrow \varepsilon^1 = \underbrace{M X \beta}_{=0} + M \varepsilon \Rightarrow \boxed{\varepsilon^1 = M \varepsilon}$$

$$\boldsymbol{\varepsilon} \sim N(0, \sigma^2 \mathbf{I}_n)$$

$$\boldsymbol{\varepsilon}^1 = \mathbf{M} \boldsymbol{\varepsilon} \Rightarrow \boldsymbol{\varepsilon}^1 \sim N(0, \sigma^2 \mathbf{M})$$

$$E(\boldsymbol{\varepsilon}^1) = E(\mathbf{M} \boldsymbol{\varepsilon}) = \mathbf{M} E(\boldsymbol{\varepsilon}) = \mathbf{M} \cdot 0 = 0$$

$$V(\boldsymbol{\varepsilon}^1) = V(\mathbf{M} \boldsymbol{\varepsilon}) = \mathbf{M} V(\boldsymbol{\varepsilon}) \mathbf{M}' =$$

$$\sim \mathbf{M} \sigma^2 \mathbf{I}_n \mathbf{M}' = \sigma^2 \mathbf{M} \mathbf{M} = \sigma^2 \mathbf{M}$$

$$\begin{aligned} \boldsymbol{\varepsilon}^{1'} \boldsymbol{\varepsilon}^1 &= (\mathbf{M} \boldsymbol{\varepsilon})' (\mathbf{M} \boldsymbol{\varepsilon}) = \boldsymbol{\varepsilon}' \mathbf{M}' \mathbf{M} \boldsymbol{\varepsilon} = \\ &= \boldsymbol{\varepsilon}' \mathbf{M} \boldsymbol{\varepsilon} = \boldsymbol{\varepsilon}' \mathbf{M} \boldsymbol{\varepsilon} \end{aligned}$$

$$\varepsilon \sim N(0, \sigma^2 I_n) \Rightarrow$$

$$\frac{1}{\sigma} \varepsilon \sim N(0, I_n) \Rightarrow$$

$$\left(\frac{1}{\sigma} \varepsilon\right)' M \frac{1}{\sigma} \varepsilon \sim \chi^2_{r(M)} \Rightarrow$$

$$\frac{1}{\sigma^2} \varepsilon' M \varepsilon \sim \chi^2_{n-k} \quad \left| \begin{array}{l} \frac{1}{\sigma^2} \varepsilon' \varepsilon = \frac{1}{\sigma^2} \varepsilon' M \varepsilon \end{array} \right| \Rightarrow \frac{1}{\sigma^2} \varepsilon' \varepsilon \sim \chi^2_{n-k}$$

$$\frac{(R\delta^1 - g)' [R(x'x)^{-1}R']^{-1} (R\delta^1 - g)}{\sigma^2} \sim \chi^2_{n-k}$$

$$\frac{(R\delta^1 - g)' [R(x'x)^{-1}R']^{-1} (R\delta^1 - g)}{\sigma^2} \sim \chi^2_{n-k}$$

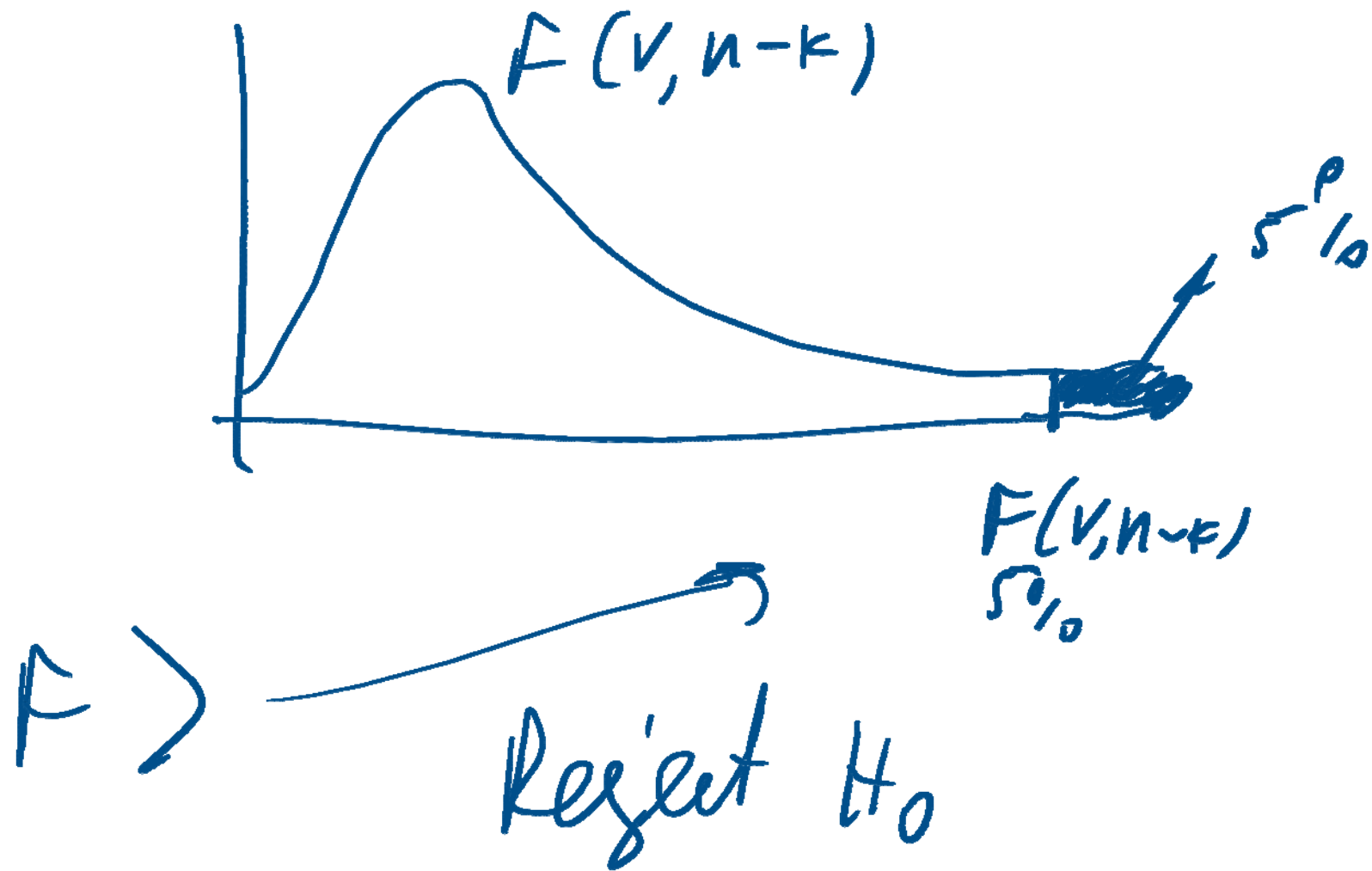
$$\frac{\frac{1}{\sigma^2} \frac{\delta^1' \delta^1}{n-k}}{F(v, n-k)}$$

$$\Rightarrow F = \frac{(R\hat{\beta} - q)' [R(X'X)^{-1}R']^{-1} (R\hat{\beta} - q)}{v} \sim F(v, n-k)$$

$\frac{\sum \sum}{n-k} = s^2$

Reject H_0 if

$$F > F_{5\%}^{(v, n-k)} \quad (\Delta E \equiv 14 \text{ OVPK})$$



$$b = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{pmatrix} \quad \text{E2T2 } K=4$$

$$H_0: b_3 = 0 \quad q = (0)$$

$$L = (0, 0, 1, 0)$$

$$Lb^1 = (0, 0, 1, 0) \begin{pmatrix} \vec{b_1} \\ \vec{b_2} \\ \vec{b_3} \\ \vec{b_4} \end{pmatrix} = \vec{b_3}$$

$$R\vec{b} - q = \vec{b}_3 - 0 = \vec{b}_3$$

$$R (X'X)^{-1} R' =$$

$$(0, 0, 1, 0) (X'X)^{-1} \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix} =$$

$$= ((X'X)^{-1})_{(3,3)}$$

$$\frac{\hat{\beta}_3 (X'X)^{-1}_{(3,3)} \hat{\beta}_3}{s^2} \sim F(1, n-k)$$

$$\text{TRPA } V(\hat{\beta}_3) = s^2 (X'X)^{-1}_{(3,3)}$$

$$\Rightarrow \left(\frac{\hat{\beta}_3}{V(\hat{\beta}_3)} \right)^2 \sim F(1, n-k) \Rightarrow$$

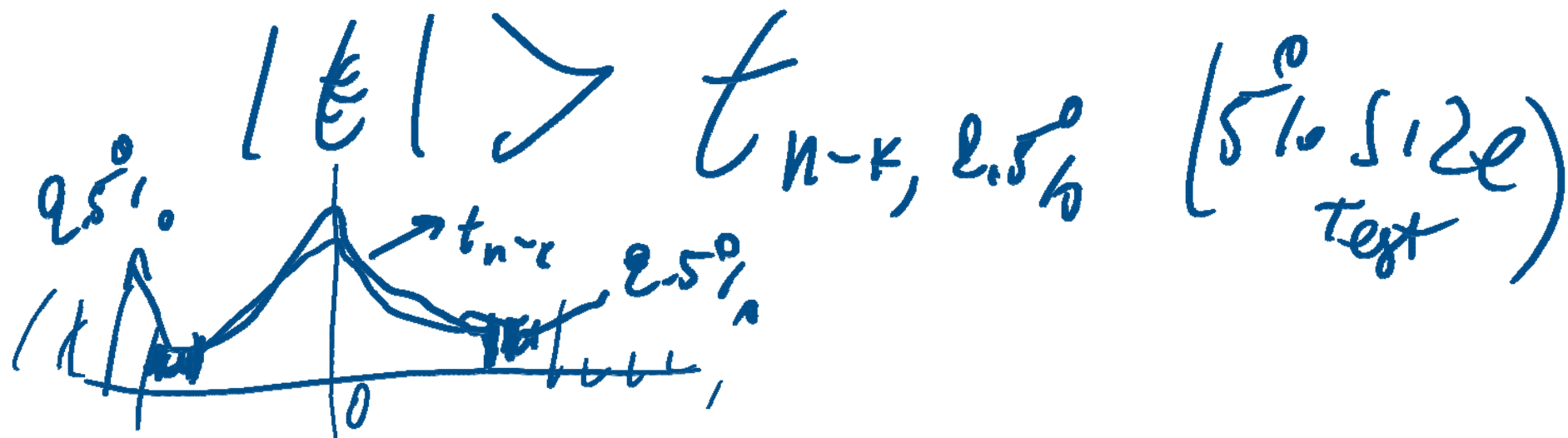
$$\frac{\hat{\beta}_3}{\sqrt{V(\hat{\beta}_3)}} \sim t_{n-k}$$

$$t = \frac{\hat{\beta}_3}{\text{S.E.}(\hat{\beta}_3)}$$

$$H_0: \beta_3 = 0 \quad \sim t_{n-k}$$

$$H_1: \beta_3 \neq 0$$

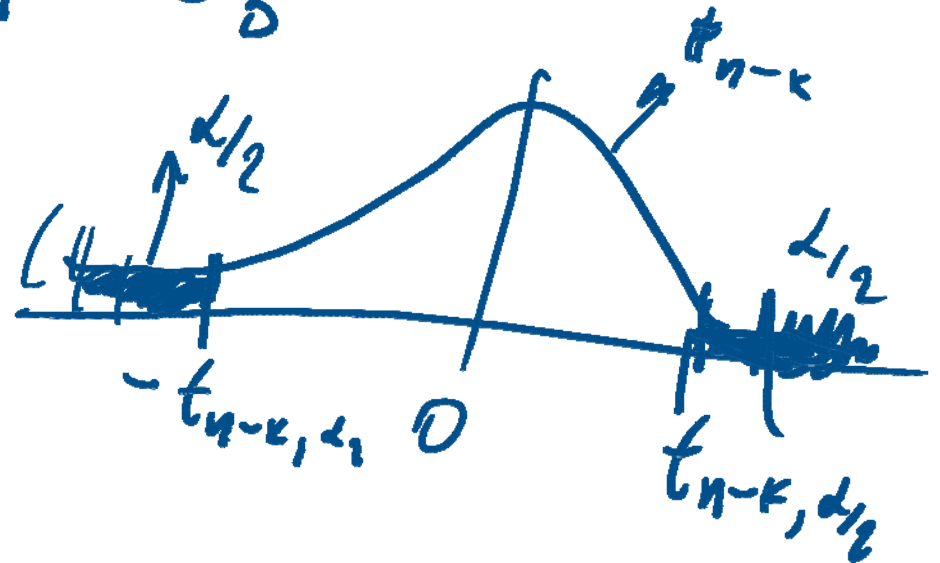
Reject H_0 iff



size test α

$$H_0: \theta_3 = 0 \text{ v.s. } H_1: \theta_3 \neq 0$$

$$t = \frac{\hat{\theta}_3 - 0}{\text{s.e.}(\hat{\theta}_3)}$$



Reject H_0 iff $|t| > t_{n-k, \alpha/2}$
 Reject H_0 iff $P.V.(t) < \alpha$

$$\begin{aligned} P.V.(t) &= P(t_{n-k} < -|t|) + P(t_{n-k} > |t|) \\ &= 2 P(t_{n-k} > |t|) \end{aligned}$$

$$H_0: \beta_2 = 0 \text{ v.s. } H_1: \beta_2 \neq 0$$

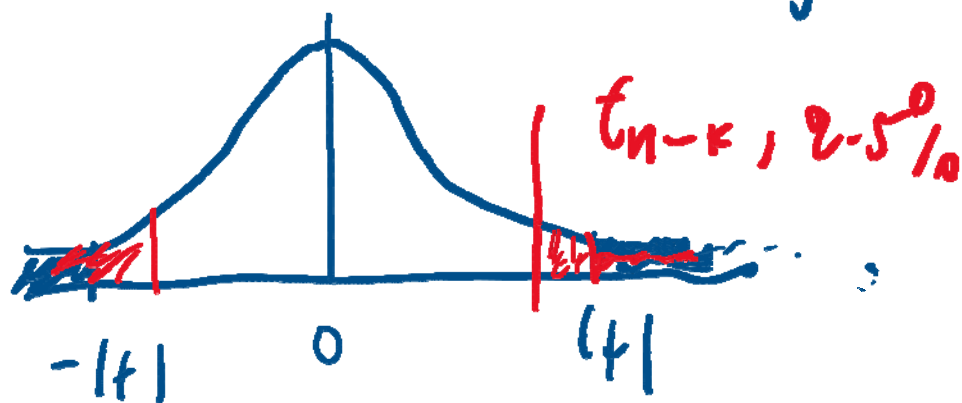
Reject H_0

$$t = \frac{\hat{\beta}_2}{\text{s.e.}(\hat{\beta}_2)} \sim_{H_0} t_{n-k}$$

iff $P.v.(t) < 5\%$

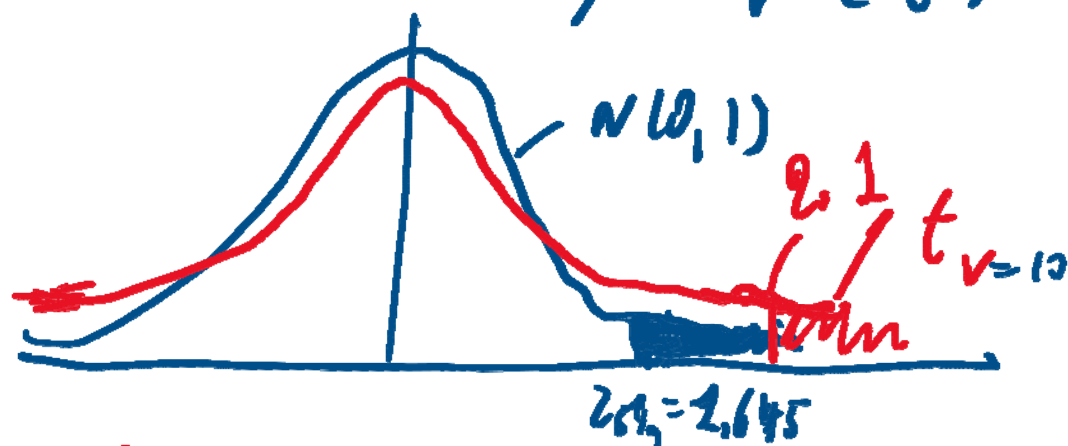
$$P.v.(t) = P(t_{n-k} < -|t|) +$$

$$+ P(t_{n-k} > |t|) = 2P(t_{n-k} > |t|)$$



t_v Gaussian as $\mu \rightarrow 0$
 $t \sim t_v$

$$E(t) = 0, \quad V(t) = \frac{v}{v-2} \quad \underline{\underline{v > 2}}$$



For $v > 50$ (METANO) $\Rightarrow$

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

Dependent Variable: R1-RF

Method: Least Squares

Date: 10/24/19 Time: 10:23

Sample: 1960M01 2003M12

Included observations: 528

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	-0.012192	0.082434	-0.147905	0.8825
RMRF	0.896798	0.019628	45.69048	0.0000
HML	0.272386	0.030257	9.002418	0.0000
SMB	1.252069	0.025830	48.47303	0.0000
UMD	0.004131	0.019786	0.208788	0.8347
R-squared	0.922954	Mean dependent var		0.791042
Adjusted R-squared	0.922365	S.D. dependent var		6.384388
S.E. of regression	1.778889	Akaike info criterion		3.999280
Sum squared resid	1655.006	Schwarz criterion		4.039707
		Hannan-Quinn		
Log likelihood	-1050.810	criter.		4.015107
F-statistic	1566.287	Durbin-Watson stat		2.088172
Prob(F-statistic)	0.000000			

Do NOT reject

0x1

$$t_{n-k} = z_{\alpha/2}$$

$$t_{528-5} = t_{523} \sim N(0,1)$$

$$\alpha = 5\%$$

$$z_{\alpha/2} = z_{2.5\%} = 1.96$$

$$H_0: \beta_i = 0 \text{ v.s. } H_1: \beta_i \neq 0$$

$$i = 1, 2, 3, 4, 5$$

$$\text{Reject if } |t| > z_{\alpha/2}$$

$$= \left| \frac{\hat{\beta}_i}{\text{S.E.}(\hat{\beta}_i)} \right| > 1.96$$

$$H_0: \beta_i = 0 \quad \text{v.s.} \quad H_1: \beta_i \neq 0 \quad i=1, 2, \dots, k$$

$$t = \frac{\hat{\beta}_i}{\text{s.e.}(\hat{\beta}_i)}$$

Reject H_0 iff $|t| > t_{n-k, \alpha/2}$ or $\equiv$

Reject H_0 iff $P_{H_0}(t) < \alpha$

If H_0 is NOT Rejected $\Rightarrow$ i Μεταβλητή
 Στατιστική α Μη-Συμμετρική Αλλαγή ο β_i
 βέλτερος είναι 0.

If H_0 is Rejected $\Rightarrow$

H i μεταβλητή $\Sigma_{i=1}^n$ Γεωμετρική
βυρναζιού Διάνοι $\theta_i \neq 0$.
