

Econometrics . 2022-5-27

$$H_0 : \theta = \theta_0 \quad (=2)$$

$$H_1 : \theta = \theta_1 \quad (=5) \quad | \text{ assume } \theta_1 > \theta_0$$

Test statistic : $\hat{\theta}$

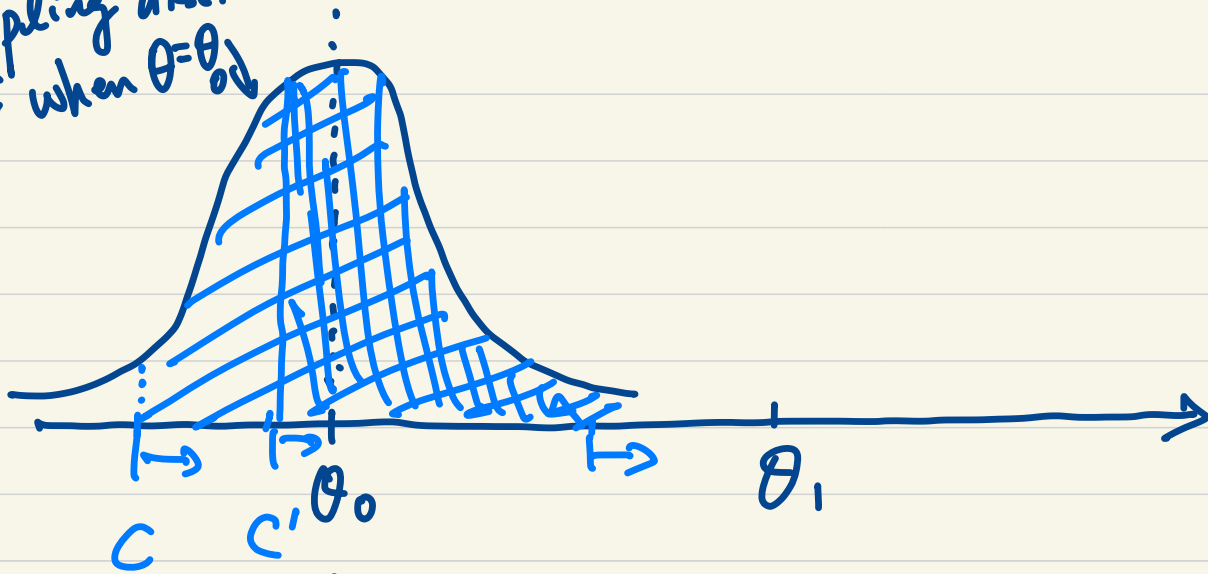
Rejection region : $\hat{\theta} > \underline{c}$

What if $\hat{\theta} < c$

\Rightarrow Acceptance : $\hat{\theta} \leq c$
region

Sampling dist. of $\hat{\theta}$.

Sampling dist.
of $\hat{\theta}$ when $\theta = \theta_0$

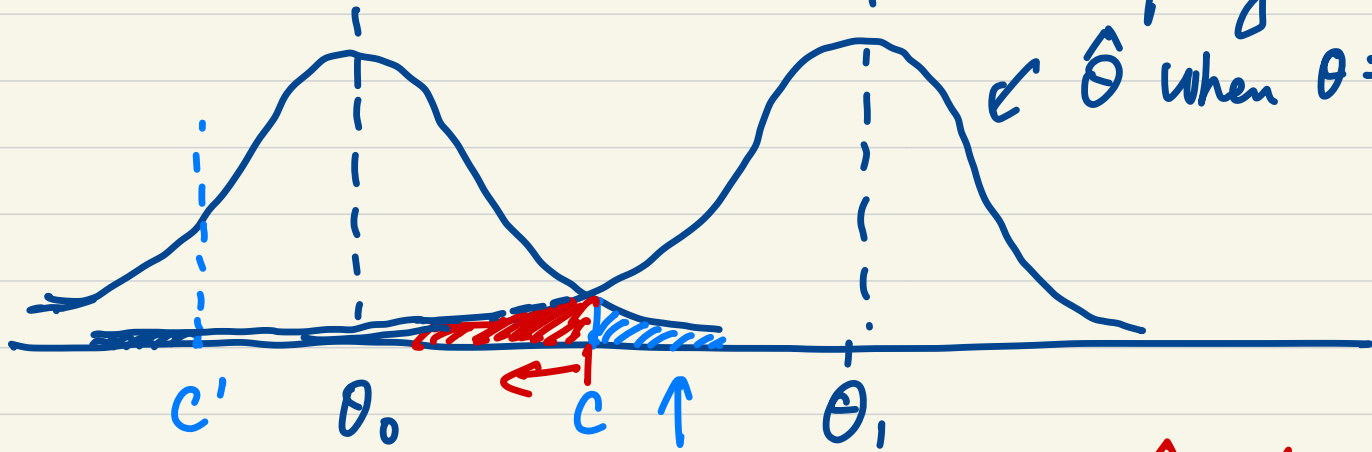


$$\alpha = \Pr(\hat{\theta} > c | \theta = \theta_0)$$

$$\alpha' = \Pr(\hat{\theta} > c' | \theta = \theta_0)$$

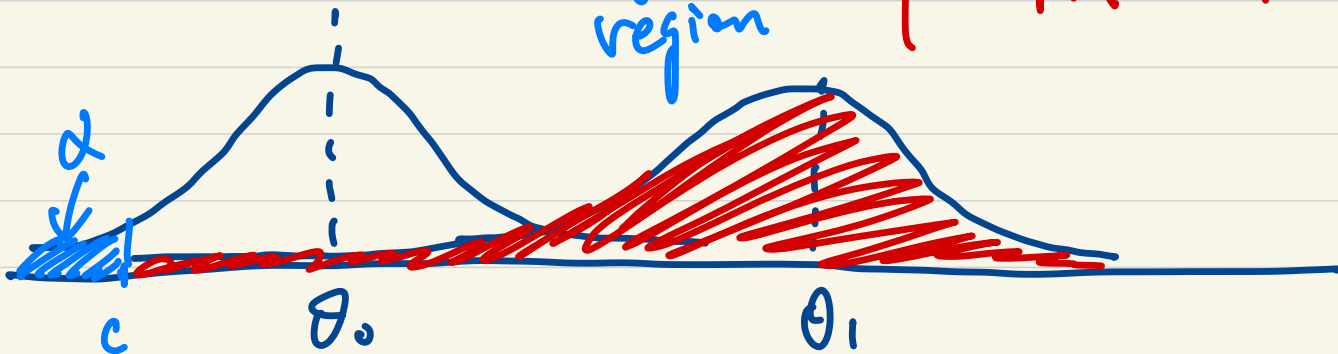
$$\alpha = 0.05, 0.01, 0.1$$

sampling dist. of
 $\hat{\theta}$ when $\theta = \theta_1$.



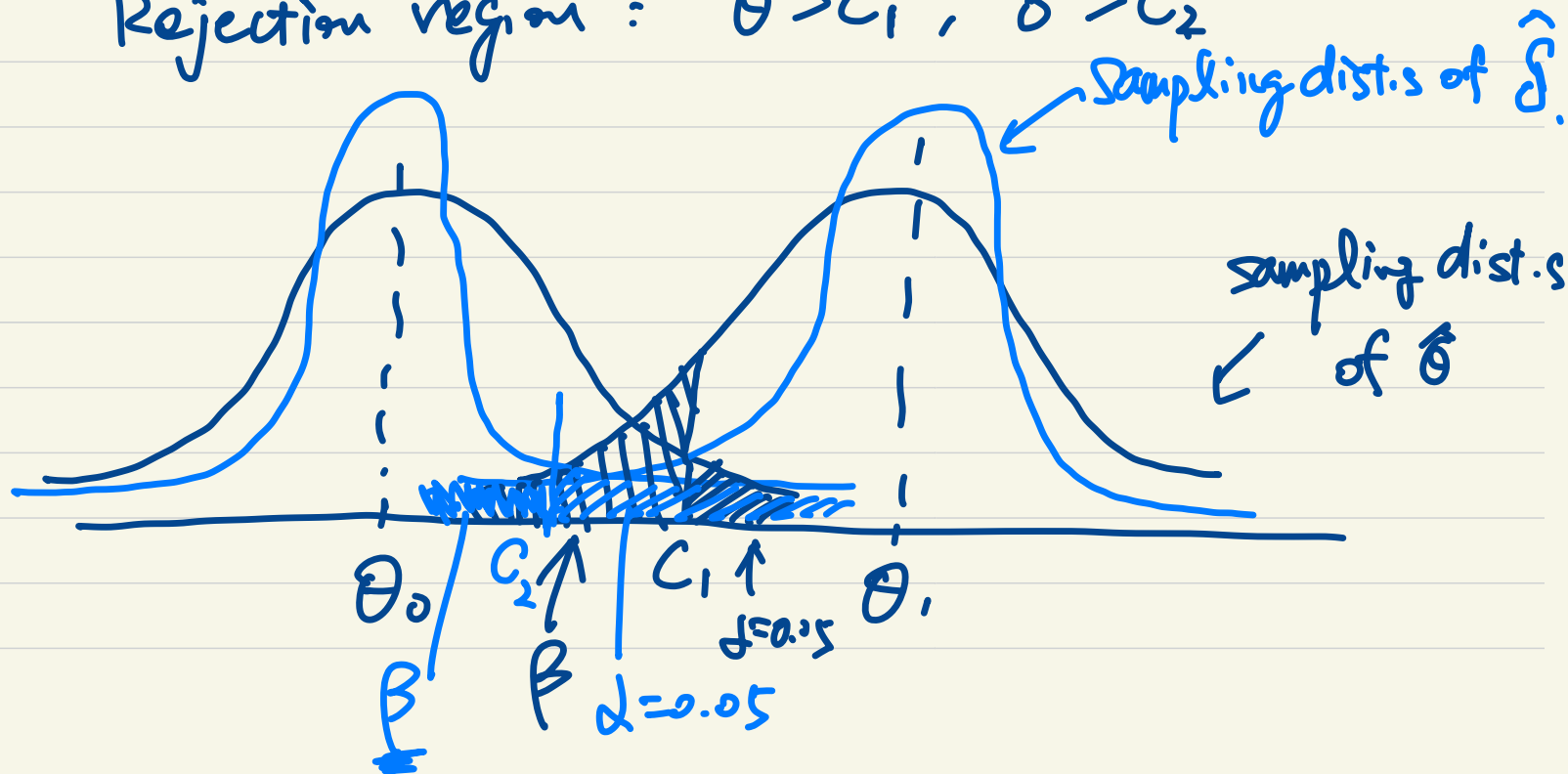
rejection
region

$$\beta = \Pr(\hat{\theta} \leq c | \theta = \theta_1)$$



Test statistics $\hat{\theta}$ vs. $\hat{\delta}$

Rejection region: $\hat{\theta} > C_1, \hat{\delta} > C_2$



$H_0 : \theta = \theta_0$, $H_1 : \theta = \theta_1 \leftarrow$ Simple hypothesis

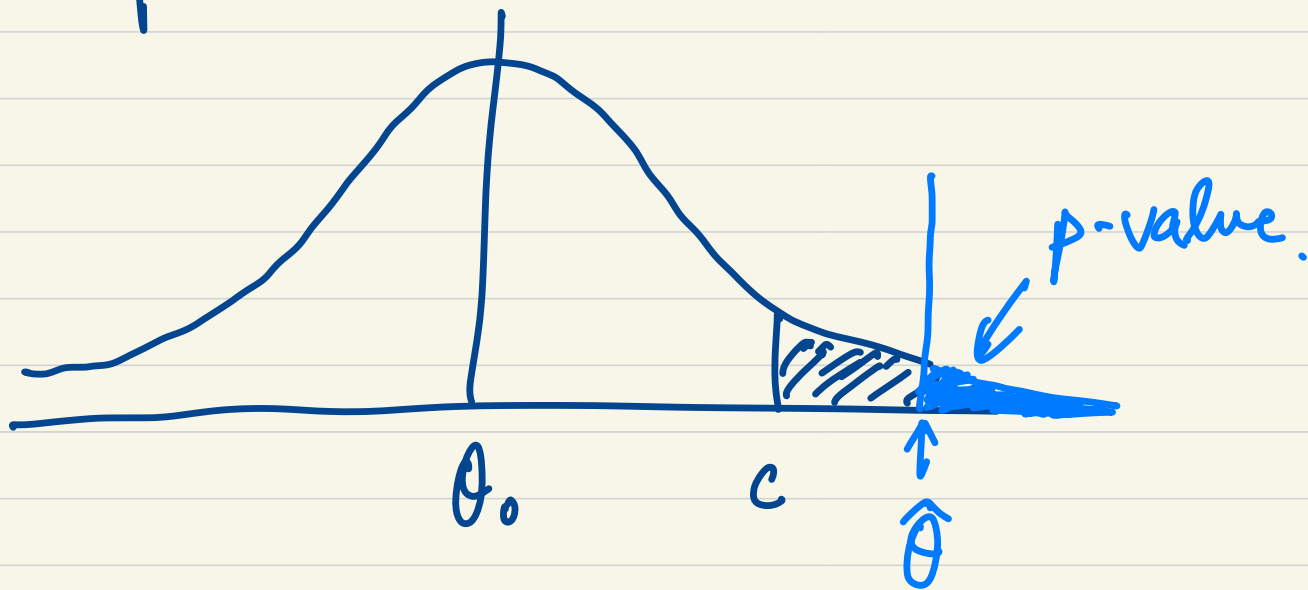
$H_0 : \theta = \theta_0$, $H_1 : \theta \neq \theta_0 \leftarrow$ Composite hypothesis.

Uniformly most powerful test. (UMP test).

\rightarrow Neyman-Pearson Lemma.

\rightarrow Likelihood ratio test is UMP.

The p-value.



p-value \downarrow

sample size \uparrow

Three widely used tests (for large samples)

- The likelihood ratio test (LR)
 - The Wald test (W)
 - The Lagrangian multiplier test (LM).
- } → Chapter 14.

Linear hypotheses :

$$H_0: R\beta = q, \quad \begin{matrix} R \\ (J \times k) \end{matrix} \left\{ \underbrace{\begin{bmatrix} R \end{bmatrix}}_k \right\}^J \left\{ \begin{bmatrix} \beta \end{bmatrix} \right\}_k = \left\{ \begin{bmatrix} q \end{bmatrix} \right\}_J.$$

$$H_1: R\beta \neq q$$

E.g. $\beta_2 = 0$: $R = [0 \mid 0 \cdots 0]$ $\begin{matrix} \uparrow \\ \text{for } \beta_2 \end{matrix}$ $\begin{bmatrix} \beta_1 \\ \beta_2 \\ \vdots \\ \beta_k \end{bmatrix} = 0$ $\rightarrow [0 \mid 0 \cdots 0][\beta] = 0$

$\beta_3 + \beta_4 = 1$ $\rightarrow [0 \mid 0 \mid 1 \mid 1 \mid 0 \cdots 0][\beta] = 1$

regression under H_0 : restricted regression \rightarrow LM
original regression : unrestricted regression \rightarrow LR
 \rightarrow W

The Wald test.

If $H_0 : R\beta = q$ is true, using the unrestricted estimator b , $Rb - q$ must be small.

Let $v = Rb - q$, Under H_0 .

$$v = Rb - q = Rb - R\beta = R(b - \beta).$$

$$E[b|X] = \beta, \quad \text{Var}[b|X] = \sigma^2(X'X)^{-1}.$$

Under A.6, $b - \beta | X \sim N(0, \sigma^2(X'X)^{-1}).$

$$\Rightarrow R(b - \beta) | X \sim N(0, \sigma^2 R(X'X)^{-1} R').$$

$$\Rightarrow W = \underbrace{[R(b - \beta)]'}_{X'} \cdot \underbrace{[\sigma^2 R(X'X)^{-1} R']^{-1}}_{\Sigma^{-1}} \cdot \underbrace{[R(b - \beta)]}_X.$$

$$\sim \chi^2_J$$

\Rightarrow Wald test: reject H_0
when W is large.

• The F-test.

Let σ be unknown. $S^2 = \frac{e'e}{n-k}$ is an unbiased estimator of σ^2 .

Let $r = \frac{e'e}{\sigma^2}$, then it can be shown that

$$r | X \sim \chi^2_{(n-k)}.$$

w and r are independent conditional on X .

$$\Rightarrow \text{Let } F = \frac{w/J}{r/(n-k)} \sim F(J, n-k).$$

\Rightarrow reject H_0 if F is large.