

2010/11/1 Exercise

1. Wald test : standard linear regression model with heteroskedasticity

Generate random samples from the linear specification:

$$y_t = \beta_1 + (\beta_2 + \Delta) * x_{t2} + \beta_3 * x_{t3} + \varepsilon_t$$

$$x_{2t} \sim N(0,1), \quad x_{3t} \sim N(0,1)$$

$$\varepsilon_t \sim N(0,1) \quad (\text{homoskedasticity}) \quad \text{or} \quad .$$

$$\varepsilon_t \sim N(0, x_{t2}^2) \quad (\text{heteroskedasticity})$$

where  $\beta_1 = 2, \quad \beta_2 = 3, \quad \beta_3 = 5, \quad \Delta = 0.5$

Regressing y on x, we obtain OLS coefficients  $\hat{\beta} = (\hat{\beta}_1, \hat{\beta}_2, \hat{\beta}_3)'$ . Please use

Eicker-White covariance matrix estimator to construct a Wald test based on the

following designs, and check whether  $\hat{\beta}_2$  is "sufficiently close" to 3.

$$H_0 : \beta_2 = 3$$

$$H_1 : \beta_2 \neq 3$$

$$W_T \xrightarrow{D} \chi^2(q)$$

- (1) When heteroskedasticity is present in your data, please use Eicker-White covariance matrix estimator to construct your test. Given  $\Delta = (0, 0.1, 0.2, 0.5, 0.8, 0.9, 1)$ , plot power curve of each sample size.
- (2) When heteroskedasticity is present in your data, compare two results of Wald Test by using different estimators **under the null**. (1) Use classical OLS variance estimator  $\hat{\sigma}_T^2 = \sum_{t=1}^T \hat{\varepsilon}_t^2 / (T - k)$  to estimate  $V_0$ , then construct Wald statistics . (2) Use Eicker-White (sandwich-type) estimator to estimate  $\widehat{D}_T$ , then construct Wald statistics

◆ When a form of heteroskedasticity is specified in the following form,

$$\varepsilon_t \sim N(0, \text{abs}(u_t))$$

$$u_t \sim N(0, \sigma_\alpha)$$

- (3) (Optional) Given  $\Delta = 0.1$ , change  $\sigma_\alpha = (1, 5, 2, 3, 5)$ . Plot power curves.
- (4) (Optional) Given  $\Delta = 0.5$ , change  $\sigma_\alpha = (1, 5, 2, 3, 5)$ . Plot power curves.
- (5) (Optional) Given  $\Delta = 0$ , change  $\sigma_\alpha = (1, 5, 2, 3, 5)$ . Plot power curves.

**Note 1:** For question (3)~(5), You may add one more power curve under homoscedasticity into your plot, and compare together.

**Note 2:** For each case, consider the sample sizes  $T = (100, 200, 500, 800, 1000)$ . For each sample size, please simulate the test at least 1000 times and evaluate the proportion of projection. Please explain **in detail what you see and why**.

**Hint (It's just a hint, not complete programs):**

```
# test whether b2 is equal to zero
b1_no <- 2
b2_no <- 0
b3_no <- 25
delta <- 0.5

R <- matrix(c(0, 1, 0), ncol = 3) # test x2
dimR <- dim(R)
r <- matrix(c(b2_no), ncol = 1)
beta_no <- t(matrix(c(b1_no, b2_no + delta, b3_no), ncol = 3))

..... (neglect details).....

##### (Please compare results below)
library(aod) # aod library supports wald.test()
ols <- lm(y~x2+x3) # regress y on x2 and x3
## using R build-in function ---- vcov
w1 = wald.test(b = coef(ols), Sigma = vcov(ols), L = R, H0 = r)
## using heteroskedasticity-consistent covariance matrix estimator
w2 = wald.test(b = coef(ols), Sigma = Dt/(sample size), L = R, H0 = r)

## using R build-in function ---- vcov , and calculate wald statistics
RDR_inv_ols <- solve( R %*% vcov(ols) %*% t(R) )
w3 = t(R %*% b_hat - r) %*% RDR_inv_ols %*% (R %*% b_hat - r)
## note : w1 = w3
# you may compare w1, w2, and w3

## calculate power
pw <- 1- pchisq(W, dimR[1])
##### (End of Comparison)
```