## Quiz 2

Consider the following linear model

$$Y_t = X_t^T \beta_0 + \varepsilon_t \tag{1}$$

Assume that  $\{Y_t, X_t^T\}^T$  is jointly stationary and ergodic,  $E(\varepsilon_t | X_t) = 0$ ,  $E(\varepsilon_t) = \sigma^2$ , and  $\{X_t \varepsilon_t\}$  is an martingale difference sequence. Define  $Q = E(X_t X'_t)$  which is finite and nonsingular,  $V = E(X_t X'_t \varepsilon_t^2)$  is finite and positive definite.

- 1. Show step by step that the OLS estimator,  $\hat{\beta}$ , is consistent.
- 2. Show step by step the asymptotic normality of  $\hat{\beta}$ .
- 3. How to test  $R\beta_0 = r$ .

## ANSWER:

1.

$$\hat{\beta} = (X'X)^{-1}X'Y$$
$$= \hat{Q}^{-1}n^{-1}\sum_{t=1}^{n}X_{t}Y_{t}$$

where  $\hat{Q}^{-1} = n^{-1} \sum_{t=1}^{n} X_t X_t'$ . Substituting  $Y_t = X_t' \beta^0 + \varepsilon_t$ , we have

$$\hat{\beta} - \beta^0 = \hat{Q}^{-1} n^{-1} \sum_{t=1}^n X_t \varepsilon_t$$

Since  $X_t$  is ergodic stationary,  $\{X_tX'_t\}$  is also ergodic stationary.

$$\hat{Q} \xrightarrow{p} E\left(X_t X'\right) = Q$$

 $\hat{Q}^{-1} \xrightarrow{p} Q^{-1}$ 

Since  $\{Y_t, X'_t\}$  is ergodic stationary, so  $\varepsilon_t = Y_t - X'_t \beta^0$  is ergodic stationary, so is  $X_t \varepsilon_t$ .

$$n^{-1} \sum_{t=1}^{n} X_{t} \varepsilon_{t} \xrightarrow{p} E\left(X_{t} \varepsilon_{t}\right) = E\left[E\left(X_{t} \varepsilon_{t} \mid X_{t}\right)\right] = E\left[X_{t} E\left(\varepsilon_{t} \mid X_{t}\right)\right] = E\left(X_{t} \cdot 0\right) = 0$$

Thus,

$$\hat{\beta} - \beta^0 = \hat{Q}^{-1} n^{-1} \sum_{t=1}^n X_t \varepsilon_t = 0$$

This completes the proof.

2. Since  $\sqrt{n}\left(\hat{\beta}-\beta^0\right)=\hat{Q}^{-1}n^{-1}\sum_{t=1}^nX_t\varepsilon_t$ , and  $n^{-1}\sum_{t=1}^nX_t\varepsilon_t=\sqrt{n}\bar{Z}_n$ ,  $\bar{Z}_n=n^{-1}\sum_{t=1}^nZ_t$ ,  $Z_t=X_t\varepsilon_t$ . Because  $\{Y_t,X_t\}$  is ergodic stationary, so  $X_t\varepsilon_t$  is also stationary ergodic.

$$V = E\left(X_t X'_t \varepsilon_t^2\right) = Var\left(X_t \varepsilon_t\right)$$

$$n^{-1} \sum_{t=1}^{n} X_{t} \varepsilon_{t} \stackrel{d}{\to} N(0, V)$$

$$\hat{Q}^{-1} \stackrel{p}{\to} Q^{-1}$$

$$\sqrt{n} \left( \hat{\beta} - \beta^{0} \right) = \hat{Q}^{-1} n^{-1} \sum_{t=1}^{n} X_{t} \varepsilon_{t} \stackrel{d}{\to} N\left( 0, Q^{-1} V Q^{-1} \right)$$

$$\sqrt{n} \left( \hat{\beta} - \beta^{0} \right) \sim N\left( 0, Q^{-1} V Q^{-1} \right)$$

$$\sqrt{n} R\left( \hat{\beta} - \beta^{0} \right) \sim N\left( 0, RQ^{-1} V Q^{-1} R' \right)$$

$$\sqrt{n} \left( R \hat{\beta} - \beta^{0} \right) \left( RQ^{-1} V Q^{-1} R' \right)^{-1} \sqrt{n} \left( R \hat{\beta} - \beta^{0} \right) \sim \chi_{J}^{2}$$

3.