

3.3 Hausman's Specification Error (特定化誤差) Test

Regression model:

$$y = X\beta + u, \quad y : n \times 1, \quad X : n \times k, \quad \beta : k \times 1, \quad u : n \times 1.$$

Suppose that X is stochastic.

If $E(u|X) = 0$, OLSE $\hat{\beta}$ is unbiased because of $\hat{\beta} = (X'X)^{-1}X'y = \beta + (X'X)^{-1}X'u$ and $E((X'X)^{-1}X'u) = 0$.

However, If $E(u|X) \neq 0$, OLSE $\hat{\beta}$ is biased and inconsistent.

Therefore, we need to check if X is correlated with u or not.

⇒ **Hausman's Specification Error Test**

The null and alternative hypotheses are:

- H_0 : X and u are independent, i.e., $\text{Cov}(X, u) = 0$,
- H_1 : X and u are not independent.

Suppose that we have two estimators $\hat{\beta}_0$ and $\hat{\beta}_1$, which have the following properties:

- $\hat{\beta}_0$ is consistent and efficient under H_0 , but is not consistent under H_1 ,
- $\hat{\beta}_1$ is consistent under both H_0 and H_1 , but is not efficient under H_0 .

Under the conditions above, we have the following test statistic:

$$(\hat{\beta}_1 - \hat{\beta}_0)' \left(V(\hat{\beta}_1) - V(\hat{\beta}_0) \right)^{-1} (\hat{\beta}_1 - \hat{\beta}_0) \longrightarrow \chi^2(k).$$

Example: $\hat{\beta}_0$ is OLS, while $\hat{\beta}_1$ is IV such as 2SLS.

Hausman, J.A. (1978) “Specification Tests in Econometrics,” *Econometrica*, Vol.46, No.6, pp.1251–1271.

3.4 Choice of Fixed Effect Model or Random Effect Model

3.4.1 The Case where X is Correlated with u — Review

The standard regression model is given by:

$$y = X\beta + u, \quad u \sim N(0, \sigma^2 I_n)$$

OLS is:

$$\hat{\beta} = (X'X)^{-1}X'y = \beta + (X'X)^{-1}X'u.$$

If X is not correlated with u , i.e., $E(X'u) = 0$, we have the result: $E(\hat{\beta}) = \beta$.

However, if X is correlated with u , i.e., $E(X'u) \neq 0$, we have the result: $E(\hat{\beta}) \neq \beta$.

$\implies \hat{\beta}$ is biased.

Assume that in the limit we have the followings:

$$\begin{aligned} \left(\frac{1}{n}X'X\right)^{-1} &\longrightarrow M_{xx}^{-1}, \\ \frac{1}{n}X'u &\longrightarrow M_{xu} \neq 0 \text{ when } X \text{ is correlated with } u. \end{aligned}$$

Therefore, even in the limit,

$$\text{plim } \hat{\beta} = \beta + M_{xx}^{-1}M_{xu} \neq \beta,$$

which implies that $\hat{\beta}$ is not a consistent estimator of β .

Thus, in the case where X is correlated with u , OLSE $\hat{\beta}$ is neither unbiased nor consistent.

3.4.2 Fixed Effect Model or Random Effect Model

Usually, in the random effect model, we can consider that v_i is correlated with X_{it} .

[Reason:]

v_i includes the unobserved variables in the i th individual, i.e., ability, intelligence, and so on.

X_{it} represents the observed variables in the i th individual, i.e., income, assets, and so on.

The unobserved variables v_i are related to the observed variables X_{it} .

Therefore, we consider that v_i is correlated with X_{it} .

Thus, in the case of the random effect model, usually we cannot use OLS or GLS.

In order to use the random effect model, we need to test whether v_i is uncorrelated with X_{it} .

Apply Hausman's test.

- H_0 : X_{it} and e_{it} are independent (\rightarrow Use the random effect model),
- H_1 : X_{it} and e_{it} are not independent (\rightarrow Use the fixed effect model),

where $e_{it} = v_i + u_{it}$.

Note that:

- We can use the random effect model under H_0 , but not under H_1 .
- We can use the fixed effect model under both H_0 and H_1 .
- The random effect model is more efficient than the fixed effect model under H_0 .

Therefore, under H_0 we should use the random effect model, rather than the fixed effect model.

3.5 Applications

Example of Panel Data in Section 3: Production Function of Prefectures from 2001 to 2010.

pref: 都道府県（通し番号 1 ~ 47）

year: 年度（2001 ~ 2010 年）

y : 県内総生産（支出側、実質：固定基準年方式），出所：県民経済計算（平成 13 年度 - 平成 24 年度）(93SNA , 平成 17 年基準計数)

k : 都道府県別民間資本ストック（平成 12 歳年価格，年度末，国民経済計算ベース 平成 23 年 3 月時点）一期前（2000 ~ 2009 年）

l : 県内就業者数，出所：県民経済計算（平成 13 年度 - 平成 24 年度）(93SNA , 平成 17 年基準計数)

. tset pref year

panel variable: pref (strongly balanced)
time variable: year, 2001 to 2010
delta: 1 unit

```
. gen ly=log(y)  
. gen lk=log(k)  
. gen ll=log(l)  
. reg ly lk ll
```

| Source | SS | df | MS | Number of obs | = | 470 |
|----------|------------|-----|------------|---------------|---|----------|
| Model | 316.479302 | 2 | 158.239651 | F(2, 467) | = | 19374.95 |
| Residual | 3.81409572 | 467 | .008167229 | Prob > F | = | 0.0000 |
| Total | 320.293398 | 469 | .682928354 | R-squared | = | 0.9881 |

| ly | | Coef. | Std. Err. | t | P> t | [95% Conf. Interval] |
|-------|--|----------|-----------|-------|-------|----------------------|
| lk | | .0941587 | .0081273 | 11.59 | 0.000 | .0781881 .1101294 |
| ll | | .9976399 | .0102641 | 97.20 | 0.000 | .9774703 1.017809 |
| _cons | | .5970719 | .0773137 | 7.72 | 0.000 | .4451461 .7489978 |

```
. xtreg ly lk ll,fe
```

Fixed-effects (within) regression
 Group variable: pref
 Number of obs = 470
 Number of groups = 47

R-sq: within = 0.1721
 between = 0.9456
 overall = 0.9439
 Obs per group: min = 10
 avg = 10.0
 max = 10

corr(u_i, Xb) = 0.8803
 F(2, 421) = 43.77
 Prob > F = 0.0000

| ly | Coef. | Std. Err. | t | P> t | [95% Conf. Interval] |
|---------|-----------|-----------------------------------|------|-------|----------------------|
| lk | .2329208 | .0252321 | 9.23 | 0.000 | .1833242 .2825175 |
| ll | .3268537 | .0810662 | 4.03 | 0.000 | .1675088 .4861987 |
| _cons | 7.691145 | 1.376677 | 5.59 | 0.000 | 4.985128 10.39716 |
| sigma_u | .41045507 | | | | |
| sigma_e | .03561437 | | | | |
| rho | .99252757 | (fraction of variance due to u_i) | | | |

F test that all u_i=0: F(46, 421) = 56.22 Prob > F = 0.0000

. est store fixed

. xtreg ly lk ll,re

Random-effects GLS regression
 Group variable: pref
 Number of obs = 470
 Number of groups = 47

| R-sq: | within = 0.1058 | Obs per group: | min = 10 | | |
|--------------|------------------|-----------------------------------|------------|-------|----------------------|
| | between = 0.9805 | | avg = 10.0 | | |
| | overall = 0.9787 | | max = 10 | | |
| | | Wald chi2(2) | = 3875.75 | | |
| corr(u_i, X) | = 0 (assumed) | Prob > chi2 | = 0.0000 | | |
| <hr/> | | | | | |
| ly | Coef. | Std. Err. | z | P> z | [95% Conf. Interval] |
| lk | .2457767 | .0153094 | 16.05 | 0.000 | .2157708 .2757827 |
| ll | .8105099 | .0220256 | 36.80 | 0.000 | .7673406 .8536793 |
| _cons | .8332015 | .2411141 | 3.46 | 0.001 | .3606265 1.305776 |
| <hr/> | | | | | |
| sigma_u | .081609 | | | | |
| sigma_e | .03561437 | | | | |
| rho | .8400205 | (fraction of variance due to u_i) | | | |
| <hr/> | | | | | |

. hausman fixed

| | ----- Coefficients ----- | | | |
|----|--------------------------|------------|---------------------|-----------------------------|
| | (b) fixed | (B) . . | (b-B) Difference | sqrt(diag(V_b-V_B)) S.E. |
| | | | Difference | S.E. |
| lk | .2329208 | .2457767 | -.0128559 | .020057 |
| ll | .3268537 | .8105099 | -.4836562 | .0780167 |

b = consistent under H_0 and H_a ; obtained from xtreg
B = inconsistent under H_a , efficient under H_0 ; obtained from xtreg

Test: H_0 : difference in coefficients not systematic

chi2(2) = $(b-B)'[(V_b-V_B)^{-1}](b-B)$
= 144.66
Prob>chi2 = 0.0000

4 Generalized Method of Moments (GMM, 一般化積率法)

4.1 Method of Moments (MM, 積率法)

As $n \rightarrow \infty$, we have the result: $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i \rightarrow E(X) = \mu$.

⇒ Law of Large Number (大数の法則)

X_1, X_2, \dots, X_n are n realizations of X .

[Review] Chebyshev's inequality (チェビシェフの不等式) is given by:

$$P(|X - \mu| > \epsilon) \leq \frac{\sigma^2}{\epsilon^2} \quad \text{or} \quad P(|X - \mu| \leq \epsilon) \geq 1 - \frac{\sigma^2}{\epsilon^2},$$

where $\mu = E(X)$, $\sigma^2 = V(X)$ and any $\epsilon > 0$.

Note that $P(|X - \mu| > \epsilon) + P(|X - \mu| \leq \epsilon) = 1$.

Replace X , $E(X)$ and $V(X)$ by \bar{X} , $E(\bar{X}) = \mu$ and $V(\bar{X}) = \frac{\sigma^2}{n}$.

As $n \rightarrow \infty$,

$$P(|\bar{X} - \mu| \leq \epsilon) \geq 1 - \frac{\sigma^2}{n\epsilon^2} \rightarrow 1.$$

That is, $\bar{X} \rightarrow \mu$ as $n \rightarrow \infty$.

[End of Review]

\bar{X} is an approximation of $E(X) = \mu$.

Therefore, $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$ is taken as an estimator of μ .
 $\implies \bar{X}$ is MM estimator of $E(X) = \mu$.