

Introductory Econometrics

Remained Discussions

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Model Specification and Model Comparison

All models are wrong, but some are useful.

George E. P. Box

- Model misspecification mainly refers to
 - Selecting variables incorrectly.
 - Specifying model functional form incorrectly.
- Model misspecification may induce an increase in estimator variance and is closely related to bias and inconsistency issues.
- To compare and select specific model(s), we can use various information criteria. AIC, BIC, and DIC,...

Serial Correlation and Time Series Data

- For linear regression model with observations indexed by t

$$Y_t = \beta_0 + \beta_1 X_{t1} + \beta_2 X_{t2} + \dots + \beta_k X_{tk} + u_t,$$

if for $t \neq s$, $\text{Cov}(u_t, u_s) = \text{E}(u_t u_s) \neq 0$, then obviously

$$\begin{aligned}\text{Cov}(\mathbf{u}) &= \text{E}(\mathbf{u}\mathbf{u}') = \begin{pmatrix} \sigma^2 & \cdots & \text{E}(u_1 u_n) \\ \vdots & \ddots & \vdots \\ \text{E}(u_n u_1) & \cdots & \sigma^2 \end{pmatrix} \\ &= \sigma^2 \mathbf{\Omega} \neq \sigma^2 \mathbf{I}\end{aligned}$$

- $\text{E}(u_t u_s) \neq 0$ is referred to as the **serial correlation**, which violates the classical assumptions again.

Testing Serial Correlation

- Durbin-Watson Test

$$\text{D.W.} = \frac{\sum_{t=2}^n (e_t - e_{t-1})^2}{\sum_{t=1}^n e_t^2}.$$

- LM Test. Consider auxiliary regression

$$e_t = \beta_0 + \beta_1 X_{t1} + \cdots + \beta_k X_{tk} + \rho_1 e_{t-1} + \cdots + \rho_p e_{t-p} + \varepsilon_t$$

then under the null $H_0 : \rho_1 = \rho_2 = \cdots = \rho_p = 0$

$$\text{LM}_n = (n - p)R^2 \xrightarrow{d} \chi^2(p).$$

R^2 refers the R^2 of the auxiliary regression.

Remedies for Serial Correlation

- GLS if we know or we can estimate correlation coefficient properly.
- For the scenario with serial correlation of order 1,

$$u_t = \rho u_{t-1} + \varepsilon_t, \quad \varepsilon_t \sim (0, \sigma_\varepsilon^2)$$

then

$$\text{Cov}(\mathbf{u}\mathbf{u}') = \frac{\sigma_\varepsilon^2}{1 - \rho^2} \begin{pmatrix} 1 & \rho & \cdots & \rho^{n-1} \\ \rho & 1 & \cdots & \rho^{n-2} \\ \cdots & \cdots & \cdots & \cdots \\ \rho^{n-1} & \rho^{n-2} & \cdots & 1 \end{pmatrix}$$

Remedies for Serial Correlation

- Difference method. Suppose

$$u_t = \rho_1 u_{t-1} + \rho_2 u_{t-2} + \cdots + \rho_p u_{t-p} + \varepsilon_t,$$

then we can transform the original model by taking difference

$$\begin{aligned} & Y_t - \rho_1 Y_{t-1} - \cdots - \rho_p Y_{t-p} \\ &= \beta_0 (1 - \rho_1 - \cdots - \rho_p) + \beta_1 (X_{t1} - \rho_1 X_{t-1,1} - \cdots - \rho_p X_{t-p,1}) \\ &+ \cdots + \beta_k (X_{tk} - \rho_1 X_{t-1,k} - \cdots - \rho_p X_{t-p,k}) + \varepsilon_t. \end{aligned}$$

- Since $\varepsilon_t \stackrel{i.i.d.}{\sim} (0, \sigma_\varepsilon^2)$, the transformed model satisfy the no serial-correlation assumption.