



Trade Flows and Trade Policy Analysis

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Selected econometric methodologies and STATA applications

Content

- a) Classical regression model
 - Linear prediction
 - Ordinary least squares estimator (OLS)
 - OLS in STATA
 - Interpretation of coefficients

- b) Introduction to panel data analysis: fixed effects
 - Definition and advantages
 - Fixed effects versus random effects models
 - Fixed effects (FE) estimator
 - Alternative: Brute force OLS (dummy variable estimator)
 - FE and dummy variable estimator estimation in STATA

- c) Binary dependent variable models
 - i. Cross-section
 - ii. Panel data
 - iii. Binary dependent variable models in STATA

a) Classical regression model

- Linear prediction
- Ordinary least squares estimator (OLS)
- OLS in STATA
- Interpretation of coefficients

Linear prediction

1. Starting from an economic model and/or an economic intuition, the purpose of regression is to test a theory and/or to estimate a relationship
2. Regression analysis studies the conditional prediction of a dependent (or endogenous) variable y given a vector of regressors (or predictors or covariates) $\mathbf{x}$, $E[y|\mathbf{x}]$
3. The classical regression model is:
 - A stochastic model: $y = E[y|\mathbf{x}] + \varepsilon$, where ε is an error (or disturbance) term
 - A parametric model: $E[y|\mathbf{x}] = g(\mathbf{x}, \boldsymbol{\beta})$, where $g(\cdot)$ is a specified function and $\boldsymbol{\beta}$ a vector of parameters to be estimated
 - A linear model in parameters: $g(\cdot)$ is a linear function, so: $E[y|\mathbf{x}] = \mathbf{x}'\boldsymbol{\beta}$

Ordinary least squares (OLS) estimator

- With a sample of N observations ($i = 1, \dots, N$) on y and $\mathbf{x}$, the linear regression model is:

$$y_i = \mathbf{x}_i' \boldsymbol{\beta} + \varepsilon_i$$

where $\mathbf{x}_i$ is a $K \times 1$ regression vector and $\boldsymbol{\beta}$ is a $K \times 1$ parameter vector (the first element of $\mathbf{x}_i$ is a 1 for all i)

- In matrix notation, this is written as $\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$
- OLS estimator of $\boldsymbol{\beta}$ minimizes the sum of squared errors:

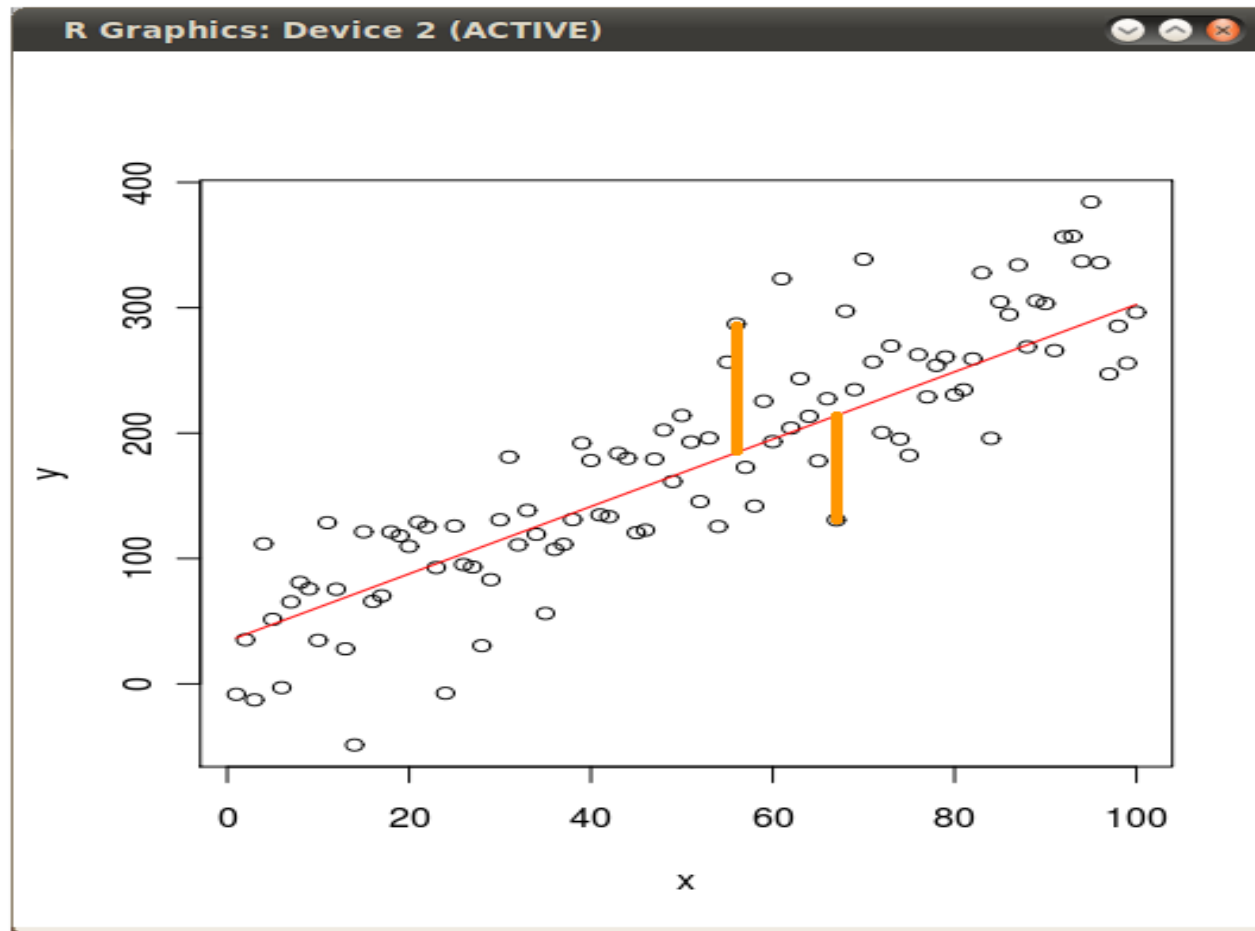
$$\sum_{i=1}^N \varepsilon_i^2 = \boldsymbol{\varepsilon}' \boldsymbol{\varepsilon} = (\mathbf{y} - \mathbf{X}\boldsymbol{\beta})'(\mathbf{y} - \mathbf{X}\boldsymbol{\beta})$$

which (provided that $\mathbf{X}$ is of full column rank K) yields:

$$\hat{\boldsymbol{\beta}}_{OLS} = (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{y} = \left(\sum_i \mathbf{x}_i \mathbf{x}_i' \right)^{-1} \left(\sum_i \mathbf{x}_i y_i \right)$$

- This is the best linear predictor of y given $\mathbf{x}$ if a squared loss error function $L(e) = e^2$ is used (where $e \equiv y - \hat{y}$ is the prediction error)

Ordinary least squares (OLS) estimator



OLS in STATA

- Stata's regress command runs a simple OLS regression
 - *Regress depvar indepvar1 indepvar2, options*
- Always use the option **robust** to ensure that the covariance estimator can handle **heteroskedasticity** of unknown form
- Usually apply the **cluster** option and specify an appropriate level of clustering to account for correlation within groups
- Rule of thumb: apply cluster to the most aggregated level of variables in the model

预览已结束，完整报告链接和二维码如下：

https://www.yunbaogao.cn/report/index/report?reportId=5_6557

