

EC475 Problem set 1

Data handling and mechanics of OLS

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- Class webpage:
[http://personal.lse.ac.uk/goujard/Under Teaching/EC475](http://personal.lse.ac.uk/goujard/Under%20Teaching/EC475).
 - 1 Introduction to Gauss (in progress) ;
 - 2 Examples of codes (in Gauss and Stata).
- Useful references for STATA programming (nothing is required):
 - 1 Baum, An Introduction to Stata Programming ;
 - 2 Cameron and Trivedi, Microeconometrics using STATA.
- Useful references for GAUSS programming:
 - 1 <http://www.aae.wisc.edu/aae637/gausscode.htm> ;
 - 2 Gauss user's guide (Aptech) ;

According to the lecture notes, we set up (for $s \in \{1, \dots, S\}$):

$$y_s = \beta_1 \cdot x_{s1} + \dots + \beta_k \cdot x_{sk} + \epsilon_s$$

Or in matrix notations: $y_s = \mathbf{x}'_s \cdot \boldsymbol{\beta} + \epsilon_s$ where $\mathbf{x}_s, \boldsymbol{\beta}$ are $k \times 1$ vectors.

So that stacking the S equations: $\mathbf{y} = \mathbf{X} \cdot \boldsymbol{\beta} + \boldsymbol{\epsilon}$ where

$$\mathbf{y} = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_S \end{pmatrix}, \quad \boldsymbol{\epsilon} = \begin{pmatrix} \epsilon_1 \\ \epsilon_2 \\ \vdots \\ \epsilon_S \end{pmatrix}, \quad \mathbf{X} = \begin{pmatrix} \mathbf{x}'_1 \\ \mathbf{x}'_2 \\ \vdots \\ \mathbf{x}'_S \end{pmatrix}$$

The OLS estimator minimizes the RSS (wrt \mathbf{b}):

$$RSS(\mathbf{b}) = \sum_{s=1}^S (y_s - \mathbf{x}'_s \cdot \mathbf{b})^2 = (\mathbf{y} - \mathbf{X} \cdot \mathbf{b})'(\mathbf{y} - \mathbf{X} \cdot \mathbf{b})$$

Under A_1 (full column rank), we have $\hat{\beta}_{OLS} = (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{y}$.

We can then define: $\hat{\epsilon}_s := y_s - \mathbf{x}'_s \cdot \hat{\beta}_{OLS}$ and $\hat{\epsilon}$.

And we can compute the indicators of this PS:

- The $RSS_{OLS} = RSS = \hat{\epsilon}'\hat{\epsilon}$
- If a constant is included, the $R^2 = 1 - \frac{\hat{\epsilon}'\hat{\epsilon}}{(\mathbf{y}-\bar{\mathbf{y}})'(\mathbf{y}-\bar{\mathbf{y}})} := 1 - \frac{RSS}{TSS^*}$
(rk: $\bar{\mathbf{y}} := \bar{y}\mathbf{1}$).
- The estimated regression variance: $s^2_{OLS} = s^2 = RSS \times 1/(S - k)$
(rk: k includes the constant).

Lqdata.dat is a panel of 10000 observations corresponding to individuals (uniqid) over time (yearcur).

The variables from lqdata.dta are:

- uniqid
- yearcur
- choice12
- age
- race
- dispy
- constant

Stata command: **reg y x** reports the **centered** $R^2 = R_c^2$ which corresponds to what we want to compute. This is "stored" in **e(r2)**.

But: **reg y x, noconstant** reports the **un-centered** $R^2 = R_u^2$ instead of the **centered** R_c^2 . This is "stored" in **e(r2)**.

This does not corresponds to our definition of the R^2 .

$R_u^2 = ESS/TSS$ is based on the decomposition:

$$TSS = ESS + RSS \text{ or } \mathbf{y}' \cdot \mathbf{y} = \hat{\mathbf{y}}' \hat{\mathbf{y}} + \hat{\boldsymbol{\varepsilon}}' \hat{\boldsymbol{\varepsilon}}$$

$R_c^2 = 1 - RSS^*/TSS^*$ is based on the decomposition:

$$TSS^* = ESS^* + RSS^* = ESS^* + RSS = (\hat{\mathbf{y}} - \bar{\mathbf{y}})'(\hat{\mathbf{y}} - \bar{\mathbf{y}}) + \hat{\boldsymbol{\varepsilon}}' \hat{\boldsymbol{\varepsilon}}$$

(If the regression includes a constant: $\bar{\hat{\boldsymbol{\varepsilon}}} = 0$ (handout1, p4.)).

The reason to use the centered R_c^2 instead of R_u^2 is that it will be invariant to a re-scaling of \mathbf{y} by adding a constant α to each observation while it is not the case for R_u^2 .

(If we have a constant in x , $R_u^2(\alpha) = \frac{\|P_x(\mathbf{y} + \alpha \cdot \mathbf{z})\|^2}{\|\mathbf{y} + \alpha \cdot \mathbf{z}\|^2} = \frac{\|P_x \cdot \mathbf{y} + \alpha \cdot \mathbf{z}\|^2}{\|\mathbf{y} + \alpha \cdot \mathbf{z}\|^2} \xrightarrow{\alpha \rightarrow \infty} 1$).

Gauss