

# Econometrics CheatSheet

## Basic concepts

### Definition of econometrics

**Econometrics** - is a social science discipline with the objective of quantify the relationships between economic agents, contrast economic theories and evaluate and implement government and business policies.

**Econometric model** - is a simplified representation of the reality to explain economic phenomena.

### Data types

1. Cross section: data taken at a given moment in time, an static "photo". Order does not matter.
2. Temporal series: observation of one/many vairable/s across time. Order does matter.
3. Panel data: consist of a temporal serie for each observation of a cross section.
4. Pooled cross sections: combines cross sections from different temporal periods.

### Phases of an econometric model

1. Specification
2. Estimation
3. Validation
4. Utilization

### Assumptions of the econometric model

Under this assumptions the estimators of the parameters will present "good properties". GAUSS MARKOV ASSUMPTIONS (EXTENDED)

- Parameters linearity.
- The sample of the poblation is random. Characteristics:
  - Independence: independence, that guarantees that all the covariances between independents are zero.

- Identical distribution: that guarantees that the  $n$  expected values and variances of the observations are the same.

- $E(u/X_1, X_2, \dots, X_k) = 0$ , guarantees that the estimations are unbiased, that have some implications:

- $E(u) = 0$  there are none systematic errors.

- $Cov(u, X_1) = Cov(u, X_2) = \dots = Cov(u, X_k) = 0$  there are no relevant variables not included in the model.

- $E(Y/X_1, X_2, \dots, X_k) = \beta_0 + \beta_1 X_1 + \beta_k X_k$  the lineal relation between  $Y$  and  $X_1, \dots, X_k$  is fulfilled, at least in average.

- Homocedasticity:  $Var(u_i/X_{1i}, X_{2i}, \dots, X_{ki}) = \sigma^2$ , the variability of the error is the same for all levels of  $x$ . Guarantees that the estimations are efficient. Implies that:  $Var(Y_i/X_{1i}, X_{2i}, \dots, X_{ki}) = \sigma^2$ , the variability of the dependent variable is the same for all levels of  $x$ .

- No autocorrelation:  $Cov(u_i, u_j) = 0 \rightarrow Cov(Y_i Y_j / X) = 0$  for every  $i$  different from  $j$ . The errors do not contain information about other errors.

- The distribution of the errors is normal (is not always necessary).

- No multicollineality: none of the independent variables is constant nor exist an exact (or aproximate) linear relation between them, they are linearly independents.

- The number of available data is greater than  $k+1$  ( $\beta$  parameters to estimate).

The homocedasticity and no autocorrelation assumptions can also be written in matrix form:  $Var(u/X) = \sigma^2 I_n$

## Interpretation of the coefficients

Model	Dependent	Independent	Interpretation $\beta_1$
Level-level	$y$	$x$	$\Delta y = \beta_1 \Delta x$
Level-log	$y$	$\log(x)$	$\Delta y = (\beta_1/100)[1\% \Delta x]$
Log-level	$\log(y)$	$x$	$\% \Delta y = (100\beta_1) \Delta x$
Log-log	$\log(y)$	$\log(x)$	$\% \Delta y = \beta_1 \% \Delta x$
Quadratic	$y$	$x + x^2$	$\Delta y = (\beta_1 + 2\beta_2 x) \Delta x$

## OLS estimation of the model

### Simple regression model

$$Y_i = \beta_0 + \beta_1 X_{1i} + u_i, i = 1, \dots, n$$

#### Definitions

$$\hat{y}_i = \hat{\beta}_0 + \hat{\beta}_1 X_i$$

$$\hat{u}_i = Y_i - \hat{Y}_i = Y_i - (\hat{\beta}_0 + \hat{\beta}_1 X_i)$$

Objective is minimize the square sum of resid:

$$\text{Min} \sum_{i=1}^n \hat{u}_i^2 = \text{Min} \sum_{i=1}^n [Y_i - (\hat{\beta}_0 + \hat{\beta}_1 X_i)]^2$$

With

$$\hat{\beta}_0 = \bar{Y} - \hat{\beta}_1 \bar{X}$$

$$\hat{\beta}_1 = \frac{Cov(Y, X)}{Var(X)}$$

### Multiple regression model

$$Y_i = \beta_0 + \beta_1 X_{1i} + \dots + \beta_k X_{ki} + u_i, i = 1, \dots, n$$

$$\hat{u}_i = Y_i - \hat{Y}_i = Y_i - (\hat{\beta}_0 + \hat{\beta}_1 X_i + \dots + \hat{\beta}_k X_{ki})$$

Objective:

$$\text{Min} \sum_{i=1}^n \hat{u}_i^2$$

Then

$$\hat{\beta}_0 = \bar{Y} - \hat{\beta}_1 \bar{X}_1 - \dots - \hat{\beta}_k \bar{X}_k$$

$$\hat{\beta}_j = \frac{Cov(Y, \text{resid}(X_j))}{Var(\text{resid}(X_j))}$$

## Properties of OLS

- Lineality in  $Y$ .
- Normality:  $Y/X \sim N(\beta_0 + \beta_1 X, \sigma^2)$

- Expected value of the estimator:  $E(\hat{\beta}_1/X_i) = \beta_1$ , then  $\hat{\beta}_1$  is an unbiased estimator of  $\beta_1$
- Variance of the estimator:  $Var(\hat{\beta}_1/X_i) = \frac{\sigma^2}{nVar(X_i)}$

**Efficiency of OLS estimators, Gauss-Markov Theorem.** In the context of the simple or multiple linear regression model, the OLS estimators of the parameters are those with the lowest variance between the lineal and unbiased estimators

## Central Limit Theorem

Under the CLT,  $\hat{\beta}_j$  is a consistent estimator of the poblational parameter  $\beta_j$ .

$$plim \hat{\beta}_i = \beta_i$$

## Regression Analysis

Study and predict the mean value of a variable regarding the base of fixed values of other variables. We usually use Ordinary Least Squares (OLS).

## Correlation Analysis

The correlation analysis not distinguish between dependent and independent variables. **Simple Correlation** Measure the grade of lineal association between two variables.

## Utilization

### Interpretation of the model

### Heterocedasticity

The residuals  $u_i$  of the poblational regression function don't have the same variance  $\sigma^2$ :

$$Var(u_i | x_i) = \sigma_i^2; i = 1, \dots, n$$

### Consequences

Under the Gauss-Markov Theorem assumptions, OLS estimators are not efficient. The estimations of the

variance of the estimators are biased. The hypothesis contrast and the confidence intervals are not reliable.

## Detection

Plots (look for structures in plots with the square residuals) and contrasts: Park test, Goldfield-Quandt, Bartlett, Breush-Pagan, CUSUMQ, Spearman, White. White's null hypothesis:

$$H_0 = HOMOCEDASTICITY$$

## Correction

- When the variance structure is known, use weighted least squares.
- When the variance structure is not known: make assumptions of the possible structure and apply weighted least squares
- Supposing that  $\sigma_i^2$  is proportional to  $x_i^2$ , divide by  $x_i$

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