Econometrics

Introduction to Linear Regression Analysis

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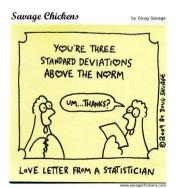
Lecture 2

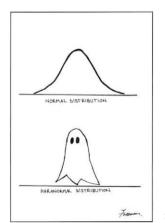
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PREVIOUS LECTURE...

Introduction, organization, review of statistical background

- random variables
- mean, variance, standard deviation
- covariance, correlation, independence
- normal distribution
- standardized random variables





WARM-UP EXERCISE

► What is the correlation between X and Y?

$$\begin{pmatrix}
X & Y \\
5 & 10 \\
3 & 6 \\
-1 & -4 \\
6 & 8 \\
2 & 5
\end{pmatrix}$$

- Correlation: $Corr(X, Y) = \frac{Cov(X, Y)}{\sigma_X \sigma_Y}$
- Covariance:

Cov(X, Y) = E[(X - E[X])(Y - E[Y])] = E[XY] - E[X]E[Y]

- Standard deviation : $\sigma_X = \sqrt{Var[X]}$
- ► Variance: $Var[X] = E\left[(X E[X])^2\right] = E[X^2] (E[X])^2$

LECTURE 2.

Introduction to simple linear regression analysis

- Sampling and estimation
- OLS principle

Readings:

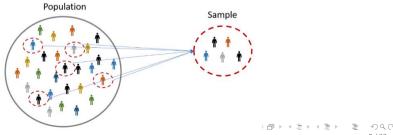
Studenmund, A. H., Using Econometrics: A Practical Guide, Chapters 1, 2.1, 16.1, 16.2Wooldridge, J. M., Introductory Econometrics: A Modern

Approach, Chapters 2.1, 2.2

SAMPLING

- Population: the entire group of items that interests us
- **Sample**: the part of the population that we actually observe
- **Statistical inference**: use of the sample to draw conclusion about the characteristics of the population from which the sample came

Examples: medical experiments, opinion polls



RANDOM SAMPLING VS SELECTION BIAS

Correct statistical inference can be performed only on a **random sample** - a sample that reflects the true distribution of the population

Biased sample: any sample that differs systematically from the population that it is intended to represent

Selection bias: occurs when the selection of the sample systematically excludes or underrepresents certain groups Example: opinion poll about tuition payments among undergraduate students vs all citizens

Self-selection bias: occurs when we examine data for a group of people who have chosen to be in that group

Example: accident records of people who buy collision insurance

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EXERCISE 1

- American Express and the French tourist office sponsored a survey that found that most visitors to France do not consider the French to be especially unfriendly.
- The sample consisted of 1,000 Americans who have visited France more than once for pleasure over the past two years.
- Is this survey unbiased?

ESTIMATION

Parameter: a true characteristic of the distribution of a variable, whose value is unknown, but can be estimated

Example: population mean E[X]

Estimator: a sample statistic that is used to estimate the value of the parameter

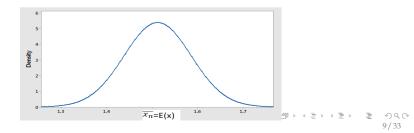
Example: sample mean \overline{X}_n

Note that the estimator is a random variable (it has a probability distribution, mean, variance,...)

Estimate: the specific value of the estimator that is obtained on a specific sample

PROPERTIES OF AN ESTIMATOR

- An estimator is **unbiased** if the mean of its distribution is equal to the value of the parameter it is estimating
- An estimator is **consistent** if it converges to the value of the true parameter as the sample size increases
- An estimator is **efficient** if the variance of its sampling distribution is the smallest possible



EXERCISE 2

- A young econometrician wants to estimate the relationship between foreign direct investments (FDI) in her country and firm profitability.
- Her reasoning is that better managerial skills introduced by foreign owners increases firms' profitability.
- She collects a random sample of 8,750 firms and finds that one sixth of the firms were entered within last few years by foreign investors. The rest of the firms are owned domestically.
- When she compares indicators of profitability, such as ROA and ROE, between the domestic and foreign-owned firms, she finds significantly better outcomes for foreign-owned firms.
- She concludes that FDI increases firms' profitability. Is this conclusion correct?

ECONOMETRIC MODELS

Econometric model is an estimable formulation of a theoretical relationship

Theory says: $Q = f(P, P_s, Y)$ $Q \dots$ quantity demanded $P \dots$ commodity's price $P_s \dots$ price of substitute good $Y \dots$ disposable income We simplify: $Q = \beta_0 + \beta_1 P + \beta_2 P_s + \beta_3 Y$

We estimate: $Q = 31.50 - 0.73P + 0.11P_s + 0.23Y$

ECONOMETRIC MODELS

- Today's econometrics deals with different, even very general models
- During this course we will cover linear regression models
- We will see how these models are estimated by
 - Ordinary Least Squares (OLS)
 - Generalized Least Squares (GLS)
 - Instrumental Variables (IV)
- We will perform estimation on different types of data

DATA USED IN ECONOMETRICS

cross-section

sample of units (eg. firms, individuals) taken at a given point in time

repeated cross-section

several independent samples of units (eg. firms, individuals) taken at different points in time

time-series

observations of variable(s) in different points in time (eg. GDP)

panel data

time series for each cross-sectional unit in the data set (eg. GDP of various countries)

DATA USED IN ECONOMETRICS - EXAMPLES

- Country's macroeconomic indicators (GDP, inflation rate, net exports, etc.) month by month
- Data about firms' employees or financial indicators as of the end of the year
- Records of bank clients who were given a loan
- · Annual social security or tax records of individual workers

STEPS OF AN ECONOMETRIC ANALYSIS

- 1. Formulation of an economic model (rigorous or intuitive)
- 2. Formulation of an econometric model based on the economic model
- 3. Collection of data
- 4. Estimation of the econometric model
- 5. Interpretation of results

EXAMPLE - ECONOMIC MODEL

- Denote:
 - p \dots price of the goodc \dots firm's average cost per one unit of outputq(p) \dots demand for firm's output

Firm profit:

Demand for good:

- $\pi = q(p) \cdot (p-c) \qquad \qquad q(p) = a b \cdot p$
- Derive:

$$q = \frac{a}{2} - \frac{b}{2} \cdot c$$

• We call *q* dependent variable and *c* explanatory variable

EXAMPLE - ECONOMETRIC MODEL

• Write the relationship in a simple linear form

$$q = \beta_0 + \beta_1 c$$

(have in mind that $\beta_0 = \frac{a}{2}$ and $\beta_1 = -\frac{b}{2}$

• There are other (unpredictable) things that influence firms' sales ⇒ add disturbance term

$$q = \beta_0 + \beta_1 c + \varepsilon$$

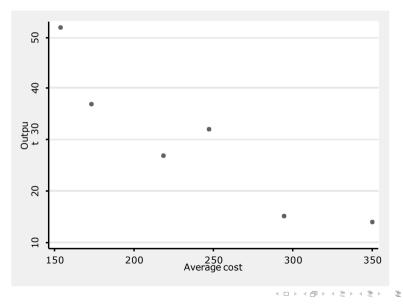
• Find the value of parameters β_1 (slope) and β_0 (intercept)

EXAMPLE - DATA

- · Ideally: investigate all firms in the economy
- Reality: investigate a sample of firms We need a random (unbiased) sample of firms
- Collect data:

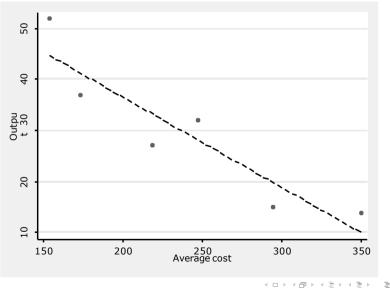
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q	15	32	52	14	37	27
С	294	247	153	350	173	218

EXAMPLE - DATA



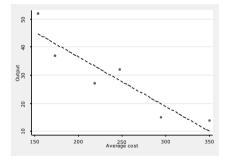
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EXAMPLE - ESTIMATION



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EXAMPLE - ESTIMATION



OLS method:

Make the fit as good as possible ↓ Make the misfit as low as possible

Minimize the (vertical) distance between data points and regression line

Minimize the sum of squared deviations

TERMINOLOGY

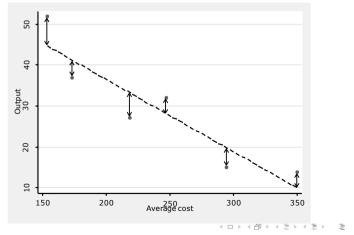
$$y_i = \beta_0 + \beta_1 x_i + \varepsilon_i \dots$$
 regression line

 y_i ... dependent/explained variable (*i*-thobservation)

 x_i . . . independent/explanatory variable (*i*-th observation) ε_i . . . random error term/disturbance (of *i*-th observation) β_0 . . . intercept parameter (β_0° . . . estimate of this parameter) β_1 . . . slope parameter (β_1° . . . estimate of this parameter)

ORDINARY LEAST SQUARES

• OLS = fitting the regression line by minimizing the sum of vertical distance between the regression line and the observed points



ORDINARY LEAST SQUARES - PRINCIPLE

Take the squared differences between observed point *y_i* and regression line β₀ + β₁*x_i*:

$$\varepsilon_i^2 = (y_i - \beta_0 - \beta_1 x_i)^2$$

• Sum them over all *n* observations:

$$\sum_{i=1}^{n} (y_i - \beta_0 - \beta_1 x_i)^2$$

• Find $\hat{\beta}_0$ and $\hat{\beta}_1$ such that they minimize this sum

$$\left[\widehat{\beta}_{0},\widehat{\beta}_{1}\right] = \operatorname*{argmin}_{\beta_{0},\beta_{1}} \sum_{i=1}^{n} (y_{i} - \beta_{0} - \beta_{1} x_{i})^{2}$$

ORDINARY LEAST SQUARES - DERIVATION

$$\left[\widehat{\beta}_{0},\widehat{\beta}_{1}\right] = \operatorname*{argmin}_{\beta_{0},\beta_{1}} \sum_{i=1}^{n} (y_{i} - \beta_{0} - \beta_{1}x_{i})^{2}$$

► FOC:

$$\frac{\partial}{\partial \beta_0} : \qquad -2\sum_{i=1}^n \left(y_i - \widehat{\beta}_0 - \widehat{\beta}_1 x_i \right) = 0$$
$$\frac{\partial}{\partial \beta_1} : \qquad -2\sum_{i=1}^n x_i \left(y_i - \widehat{\beta}_0 - \widehat{\beta}_1 x_i \right) = 0$$

► We express:

$$\widehat{\beta}_{0} = \overline{y}_{n} - \widehat{\beta}_{1}\overline{x}_{n} \qquad \qquad \widehat{\beta}_{1} = \frac{\sum_{i=1}^{n} (x_{i} - \overline{x}_{n}) (y_{i} - \overline{y}_{n})}{\sum_{i=1}^{n} (x_{i} - \overline{x}_{n})^{2}}$$

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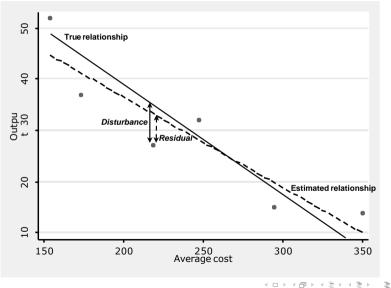
RESIDUAL

- Residual is the vertical difference between the estimated regression line and the observation points
- OLS minimizes the sum of squares of all residuals
- It is the difference between the true value y_i and the estimated value $\hat{y}_i = \hat{\beta}_0 + \hat{\beta}_1 x_i$

• Wedefine:
$$e_i = y_i - \widehat{\beta}_0 - \widehat{\beta}_1 x_i$$

- Residual *e_i* (observed) is not the same as the disturbance *ε_i* (unobserved)!!!
- Residual is an estimate of the disturbance: $e_i = \hat{\varepsilon}_i$

Residual vs. disturbance



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GETTING BACK TO THE EXAMPLE

• We have the economic model

$$q = \frac{a}{2} - \frac{b}{2} \cdot c$$

• Weestimate

$$q_i = \beta_0 + \beta_1 c_i + \varepsilon_i$$
(having in mind that $\beta_0 = \frac{a}{2}$ and $\beta_1 = -\frac{b}{2}$)

• Over data:

Firm	1	2	3	4	5	6
q	15	32	52	14	37	27
С	294	247	153	350	173	218

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When we plug in the formula:

$$\widehat{\beta}_{1} = \frac{\sum_{i=1}^{6} (c_{i} - \overline{c}) (q_{i} - \overline{q})}{\sum_{i=1}^{6} (c_{i} - \overline{c})^{2}} = -0.177$$

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GETTING BACK TO THE EXAMPLE

• When we plug in the formula:

$$\widehat{\beta}_{1} = \frac{\sum_{i=1}^{6} (c_{i} - \overline{c}) (q_{i} - \overline{q})}{\sum_{i=1}^{6} (c_{i} - \overline{c})^{2}} = -0.177$$
$$\widehat{\beta}_{0} = \overline{q} - \widehat{\beta}_{1}\overline{c} = 71.74$$

► The estimated equation is

$$\widehat{q} = 71.74$$
 -0.177c

and so

$$\hat{a} = 2\hat{\beta}_0 = 143.48$$
 and $\hat{b} = -2\hat{\beta}_1 = 0.353$

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MEANING OF REGRESSION COEFFICIENT

Consider themodel

$$q = \beta_0 + \beta_1 c$$

estimated as $\hat{\eta} = 71.74 - 0.1776$

 $q \dots$ demand for firm's output

c . . . firm's average cost per unit of output

- Meaning of β₁ is the impact of a one unit increase in *c* on the dependent variable *q*
- When average costs increase by 1 unit, quantity demanded decreases by 1.77 units

BEHIND THE ERROR TERM

- The stochastic error term must be present in a regression equation because of:
 - 1. omission of many minor influences (unavailable data)
 - 2. measurement error
 - 3. possibly incorrect functional form
 - 4. stochastic character of unpredictable human behavior
- Remember that all of these factors are included in the error term and may alter its properties
- The properties of the error term determine the properties of the estimates

SUMMARY

- We have learned that an econometric analysis consists of
 - 1. definition of the model
 - 2. estimation
 - 3. interpretation
- We have explained the principle of OLS: minimizing the sum of squared differences between the observations and the regression line
- We have derived the formulas of the estimates:

$$\widehat{\beta}_1 = \frac{\sum_{i=1}^n (x_i - \overline{x}_n) (y_i - \overline{y}_n)}{\sum_{i=1}^n (x_i - \overline{x}_n)^2} \qquad \widehat{\beta}_0 = \overline{y}_n - \widehat{\beta}_1 \overline{x}_n$$

WHAT'S NEXT

In the next lectures, we will

- derive estimation formulas for multivariate models
- specify properties of the OLS estimator
- start using Gretl for data description and estimation