# Causes of Cross-country Income Gaps

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# **Objectives**

- ► We start looking into the question: Why are some countries rich and others poor?
- We think about methods that could be used to answer such questions.

# Why Are Some Countries Rich and Others Poor?

Fact: Rich countries are 25 times richer than poor countries.

What do poor countries lack?

Some candidates...

#### Methods

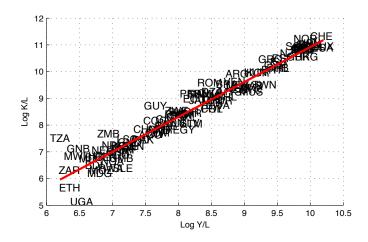
What methods could be used to answer questions such as:

How important is capital for cross-country income differences?

- Regression analysis (we will look at this one next)
- ▶ Others?

# Regression Analysis

# GDP and Capital Stock: 1990 data



# Regression Analysis

- ► The figure above suggests that output and capital are related by a linear (straight line) relationship.
- We could postulate the (statistical) model:

$$\log(Y_i/L_i) = \alpha + \beta \log(K_i/L_i) + \varepsilon_i \tag{1}$$

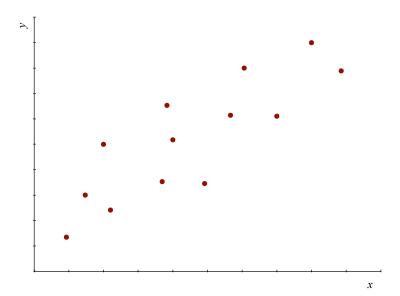
- i indexes the country
- ► The model "explains" the  $\alpha + \beta \log(K_i/L_i)$  part of the variation in  $\log(Y_i/L_i)$ .
- ε<sub>i</sub> is the unexplained residual everything we have not modeled.

# Ordinary Least Squares (OLS)

- ▶ OLS is a method for fitting a line through the data.
- ▶ OLS finds the coefficients  $(\alpha, \beta)$  that minimize the sum of squared residuals.
- Formally, OLS solves:

$$\min_{\alpha,\beta} \sum_{i} (\varepsilon_{i})^{2} = \min_{i} \sum_{i} (\log(Y_{i}/L_{i}) - \alpha - \beta \log(K_{i}/L_{i}))^{2}$$
 (2)

# **OLS Illustration**



# Multiple regression

- ► Typically one would add "covariates" to a regression.
- ▶ The idea is to "hold constant" other things.
- ▶ The model

$$\log(Y_i/L_i) = \alpha + \beta \log(K_i/L_i) + \sum_k \gamma_k X_{ik} + \varepsilon_i$$
 (3)

- $ightharpoonup X_{ik}$  is the value of regressor k for country i
- ► More compact:  $\log(Y/L) = X\beta + \varepsilon$ where Y/L and  $\varepsilon$  are  $N \times 1$  vectors and X is an  $N \times K$  matrix
- Examples:

# Example

 ${\it Table~2}$  Education as Determinant of Growth of Income per Capita, 1960-2000

	Dependent variable: average annual growth rate in GDP per capita, 1960–2000			
	(1)	(2)	(3) <sup>a</sup>	(4)
GDP per capita 1960	-0.379 (4.24)	-0.302 (5.54)	-0.277 (4.43)	-0.351 (6.01)
Years of schooling 1960	0.369 (3.23)	0.026 (0.34)	0.052 (0.64)	0.004 (0.05)
Test score (mean)		1.980 (9.12)	1.548 (4.96)	1.265 (4.06)
Openness				0.508 (1.39)
Protection against expropriation				0.388 (2.29)
Constant	2.785 (7.41)	-4.737 $(5.54)$	-3.701 $(3.32)$	-4.695 $(5.09)$
$\frac{N}{R^2}$ (adj.)	50 0.252	50 0.728	50 0.741	47 0.784

Notes: t-statistics in parentheses.

Source: Hanushek and Woessman (2008)

# Reading a Regression Table

A made-up example:

$$Y = \alpha + \beta X + \varepsilon \tag{4}$$

$$= 0.123 + 2.34X + \varepsilon \tag{5}$$

Point estimate  $\hat{\beta} = 2.34$ : the estimated "effect" of the regressor on the dependent variable

Standard error of  $\hat{\beta}$  in parentheses (0.45)

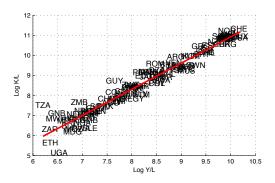
what does that mean?

R<sup>2</sup>: measure of "fit"

- ▶  $R^2 = 1 [residual sum of squares] / [total sum of squares]$
- ► fraction of (squared) variation in *Y* that is "explained" by the regression

# Application to capital and output

The OLS estimate of  $\beta$  is about 0.5.



Just eyeballing the figure shows: variation in capital "explains" almost the entire variation in Y/L.

Are we done?

Regression Analysis: Interpretation Issues

# Interpreting Regression Results

#### Suppose we regress

$$\ln(Y/L) = \alpha + \beta \ln(K/L) + \varepsilon \tag{6}$$

and find  $\beta = 0.5$ 

What does  $\beta = 0.5$  mean in words?

What do we learn about the question:

By how much would Y/L rise, if we gave a country 10% more K/L?

# Interpreting Regression Results

### Key point

The OLS regression has nothing to say about this question.

Is there an easy way to prove this?

# Regressions Do Not Answer Causality Questions

Proof: I can run the regression in reverse:

$$\log(K_i/L_i) = \hat{\alpha} + \hat{\beta}\log(Y_i/L_i) + \hat{\varepsilon}_i \tag{7}$$

I will get something close to  $\hat{\beta} = 1/\beta = 2$ .

Either regression is equally valid.

This means the regression says nothing about whether K causes Y or the other way around (or neither).

## Interpretation issues

#### **Fact**

OLS does nothing more than describe the data.

OLS answers the question:

If two observations differ by a given x, by how much do their y's differ on average?

This has nothing to do with causality.

We learn nothing about the question:

If Greece increased its K/L by 10%, by how much would Y/L increase?

# Interpretation Issues

#### Fact

No statistical method can answer cause-effect questions.

# How Can We Answer Cause/Effect Questions?

#### Possible methods:

- controlled experiments almost never possible in economics
- natural experiments (see below) these are rare
- case studies subject to interpretation issues
- 4. quantitative models

# Natural Experiments

This is as close as we can get to experimental evidence in social sciences.

The idea:

By a fluke of nature, something varies "at random" across countries

Examples?

# Summary

Statistical methods can describe data (useful).

• e.g.: capital and output are highly correlated across countries

They cannot answer cause-effect questions

e.g.: by how much would output rise, if we gave a country more capital?

How can we answer cause-effect questions?

- natural experiments (rare)
- quantitative models

# Reading

A good reference for econometrics (practical issues and interpretation) is Kennedy (2008).

The blog entry Against Multiple Regression and the interview it points to highlight the limitations of regression analysis.

#### References I

Hanushek, E. A. and Woessman, L. (2008). The role of cognitive skills in economic development. *Journal of Economic Literature*, **46** (3), 607–668.

Kennedy, P. (2008). A Guide to Econometrics. 6th edition. Wiley-Blackwell, 6th edn.