

# Introduction to R and econometrics - Part II

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# Causal effects

## Some methods to consistently estimate causal effects

We will discuss several methods that could be used to overcome endogeneity problems in order consistently estimate regression parameters that describe causal effects (like the slope of a demand function).

1. Conduct a randomized experiment.
2. Add control variables
3. Use instrumental variable estimation.

## Conduct a Randomized Experiment

- The ideal method to estimate a causal effect is to run a randomized experiment. We have already illustrated this.
- Randomised experiments are often called the *Scientific Gold Standard* to establish causal effects. They are for example required by regulators when a pharmaceutical company wants to establish that a new drug has positive effects on patients.
- However, it is not always possible, or too costly, to run a randomized experiment. We thus learn the other approaches

## Hypothesis tests: Null hypothesis

- A hypothesis test consists of a **null hypothesis**  $H_0$  and a corresponding **alternative hypothesis**  $H_1$  about some features of a data generating process. Examples for hypotheses for a linear regression model:
  - $H_0: \beta_1 = 0, H_1 : \beta_1 \neq 0$
  - $H_0$ : The explanatory variable  $x_k$  is exogenous,  $H_1 : x_k$  is endogenous
  - $H_0$ : The disturbance  $\varepsilon$  is not auto-correlated,  $H_1: \varepsilon$  is auto-correlated

## Example: t-test for a regression coefficient

- Consider a linear regression model  $y = \beta_0 + \beta_1 x_1 + \dots + \beta_K x_K + \varepsilon$  that satisfies a multiple regression equivalent to assumptions (A1)-(A4) and the null hypothesis:

$$H_0 : \beta_k = 0$$

- Every hypothesis test is based on a **test statistic** that can be computed from the data. In our example, it has *t-value*:

$$t_k = \frac{\hat{\beta}_k}{\hat{sd}(\hat{\beta}_k)}$$

- We can also view a test statistic as a random variable. Here  $t_k$  is a transformation of the random variable  $\varepsilon$  and the explanatory variables.
- Key of every hypothesis test is that one knows the distribution

## P-values and significance levels

- The p-value measures the probability to find the realized or more extreme test statistic if  $H_0$  is true (see plot above).
- One often considers critical levels of the p-value like 5% or 1%, which are called significance levels.
- We say we can reject the  $H_0$  at significance level  $\alpha$  if the p-value is smaller than  $\alpha$ ,
  - e.g. if we have p-value=0.043 we can reject  $H_0$  at a significance level of 5%.
- Significance levels are often marked with one or several stars \*\* in regression outputs.