

# L4: Hypotheses Testing with Linear Regression - Summary

Michal Hakala, Pavel Fišer

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# Mean Linear Regression

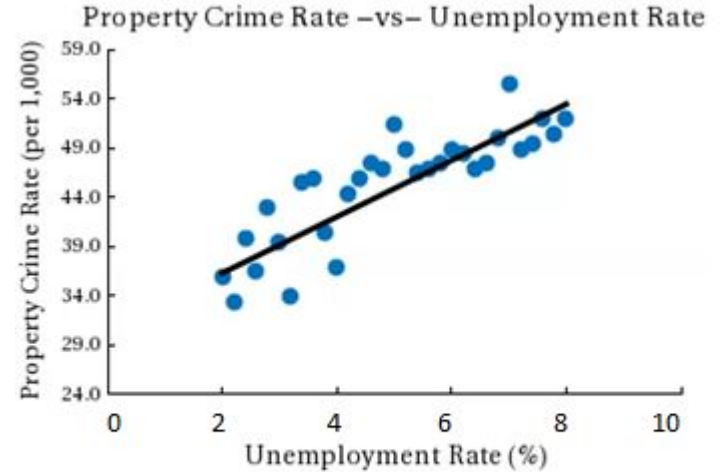
- Dependent/explained variable  $y_i$
- Independent/explanatory variable(s)  $x_i$
- Regression error  $\varepsilon_i$

## Scalar Notation

$$y_i = \beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i} + \dots + \beta_k x_{ki} + \varepsilon_i$$

## Matrix Notation

$$y_i = x_i' \beta + \varepsilon_i \quad \beta = \begin{pmatrix} \beta_0 \\ \beta_1 \\ \vdots \\ \beta_k \end{pmatrix} \quad x_i = \begin{pmatrix} 1 \\ x_{1i} \\ \vdots \\ x_{ki} \end{pmatrix}$$



## Test of Independence

$$\text{crime rate} = \beta_0 + \beta_1 \text{unemployment rate} + \varepsilon$$

Null hypothesis  $H_0$ : There is no relationship between unemployment rate and crime rate.

$$H_0: \beta_1 = 0$$

We reject the null hypothesis, if our calculated  $\beta_1$  is *"far enough from zero"*.

# Test of Independence

Null hypothesis  $H_0$ : There is no relationship between unemployment rate and crime rate.

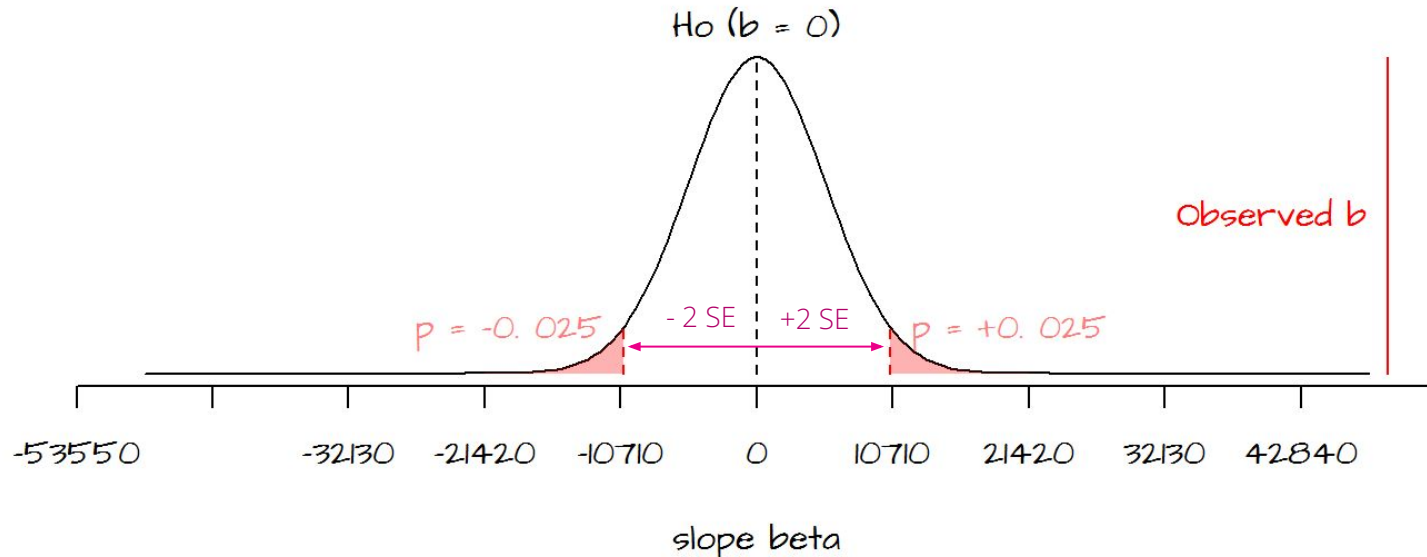
$$H_0: \beta_1 = 0$$

## Distribution of Beta Estimates

- $\widehat{\beta}_1$  – estimate,
- $\beta_1$  – true beta (not observed)
- Test = answering question how unlikely it is to obtain  $\widehat{\beta}_1$  when  $\beta_1 = 0$

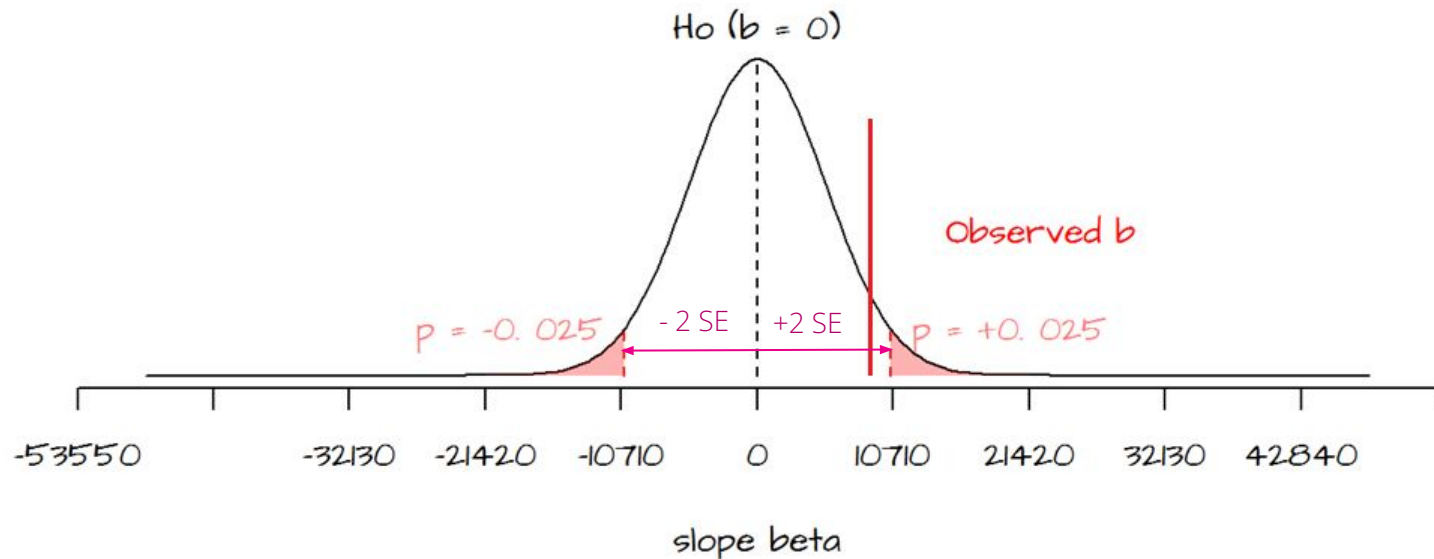
$$\widehat{\beta}_1 \sim N(\beta_0, \text{Standard Error}^2)$$

## Hypotheses Testing with OLS: Is Beta “far enough from zero”?



Here, we reject  $H_0$  with 99,99% confidence

## Hypotheses Testing with OLS: Is Beta “far enough from zero”?



And so on..

# Interpretation of Results

	coef	std err	t	P> t	[0.025	0.975]
Intercept	3526.6672	374.290	9.422	0.000	2788.515	4264.820
facebook	0.1885	0.009	21.893	0.000	0.172	0.206
newspaper	-0.0010	0.006	-0.177	0.860	-0.013	0.011
youtube	0.0458	0.001	32.809	0.000	0.043	0.049

- Coef – beta estimate
- Std err – standard error
- t – t-ratio

$$t = \frac{\hat{\beta}_j}{\text{Standard Error}(\hat{\beta}_j)} \sim N(0,1)$$

- P>|t| – p-value

“Probability of obtaining  $\hat{\beta}_j$  or worse value with respect to  $\mathbb{H}_0$ ”

## 4. No Omitted Variable

### Causal Impact

“Impact of X on Y while everything else remains the same.”

### Wage Example

Consider two models

$$wage = \alpha_0 + \alpha_1 education + e$$

$$wage = \beta_0 + \beta_1 education + \beta_2 ability + u$$

- **Problem:** education and ability is correlated  $\Rightarrow$   
 $\alpha_1$  captures impact of education and partially impact of ability
- **“Solution”:** we have to add ability to the model to control it (“keep it the same”)
- **Omitted Variable Bias:**  $\alpha_1 - \beta_1$



# Assumptions Summary

We need to check that **model assumptions** hold.

1. Linear relationship → biased betas
2. No multicollinearity → high variance of beta estimates
3. Random sample → biased betas
4. No omitted variable → biased betas
5. Homoskedasticity → invalid inference (use robust errors)
6. Normality → invalid inference (large sample desired)

Values of  
betas

Validity of  
inference