

# APPLIED STATISTICAL ANALYSIS I

## Regression diagnostics

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# Today's Agenda

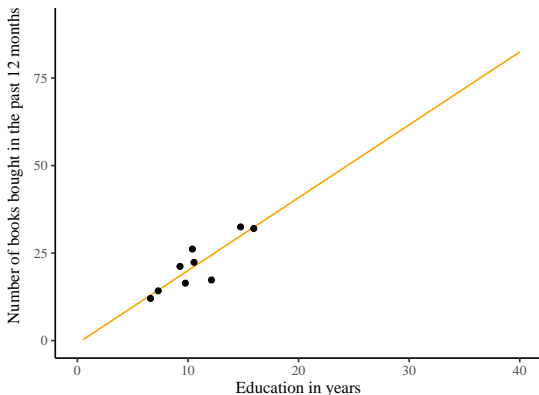
- (1) Validating quadratic effects (from last week)
- (2) Lecture recap
- (3) Tutorial exercises: What is the relationship between education and Euroscepticism?

# Discrepancy, Leverage and Influence

*What are influential cases/outliers?*

## Discrepancy, Leverage and Influence

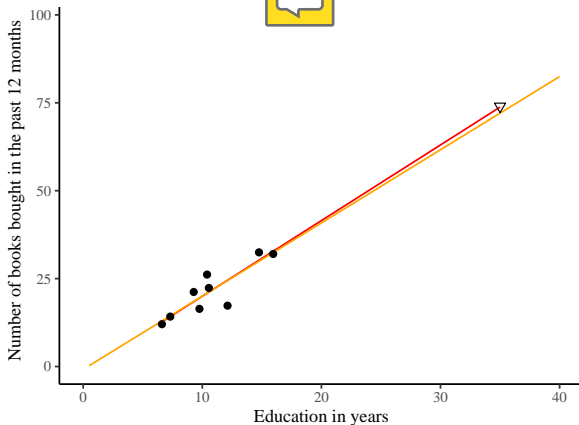
Not all outliers are concerning, because leverage  $\neq$  influence, and discrepancy  $\neq$  influence.  $\rightarrow$  Influence = leverage  $\times$  discrepancy



\*These are fictional data.

## Leverage

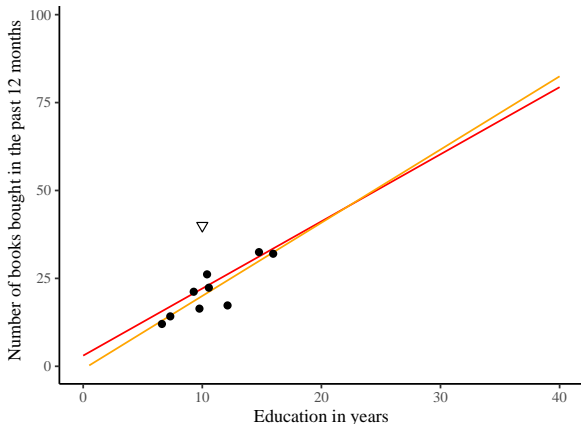
Observation is unusual in its value on X, has high leverage, but low discrepancy. → Low influence



→ Hat values ( $h_i$ ), distance of each observation from the data center

## Discrepancy

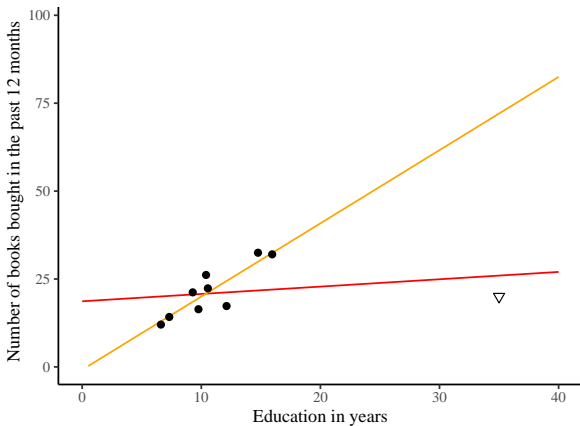
Observation is unusual in its value on Y, given its value on X, has high discrepancy, but low leverage. → Low influence



→ Standardized ( $\hat{\epsilon}_i'$ ) and studentized residuals ( $\hat{\epsilon}_i^*$ ), because  $\epsilon_i$  is scale-dependent and high leverage leads to low  $\epsilon_i$

# Influence

Observation has high leverage and discrepancy, an unusual value on X and Y. → High influence



# Influence

Validate through

1. Cook's Distance, difference in predicted values when observation  $i$  is included and not included
2. Difference in betas (DFBeta), difference in coefficients when observation  $i$  is included and not included
3. Leverage versus residual plot

Remedies

1. Check for coding errors
2. Think carefully about omitted variables







# OLS assumptions

*What are the assumptions of linear regression?*

## Assumptions of linear regression

Assumptions about the error ( $\epsilon_i$ ),  $Y_i = \alpha + \beta X_i + \epsilon_i$

$$\epsilon_i \sim N(0, \sigma^2)$$

- \*  $\epsilon_i$  is normally distributed  $\rightarrow$  needed for inference
- \*  $E(\epsilon_i) = 0$ , no bias  $\rightarrow$  violated if error is not random, but correlated with omitted variable
- \*  $\epsilon_i$  has constant variance  $\sigma^2$  (Homoscedasticity  $\leftrightarrow$  Heteroscedasticity) 
- \* No autocorrelation,  correlation occurs when the stochastic terms for any two or more cases are systematically related to each other". 
- \* X values are ured without error

(Kellstedt and Whitten 2018, 190–194)

# Assumptions of linear regression

Assumptions about the model specification,  $Y_i = \alpha + \beta X_i + \epsilon_i$

- \* No causal variables left out and no noncausal variables included
- \* Parametric linearity

(Kellstedt and Whitten 2018, 190–194)

# Assumptions of linear regression

Minimal mathematical requirements,  $Y_i = \alpha + \beta X_i + \epsilon_i$

- \* X must vary
- \* Number of observations must be larger than the number of predictors
- \* In multiple regression: No perfect multicollinearity

(Kellstedt and Whitten 2018, 190–194)

$\epsilon_i$  is normally distributed

Validate through

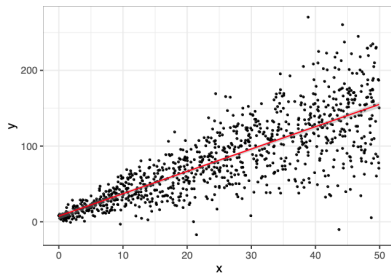
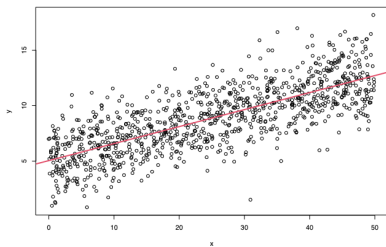
1. Histogram for  $\epsilon_i$
2. QQ (Quantile-quantile) plot

→ If violated, standard errors are unreliable

Remedies

1. Gather more data

$\epsilon_i$  has constant variance  $\sigma^2$



$\epsilon_i$  has constant variance  $\sigma^2$

Validate through

1. Residual versus fitted plot

→ If violated, standard errors are unreliable

Remedies

1. Log-transform Y

2. Roust standard er



## Parametric linearity

Validate through

1. Scatter plot
2. Residual plot

→ If violated, slope coefficients are unreliable

Remedies

1. Transform  $X$



## No perfect multicollinearity

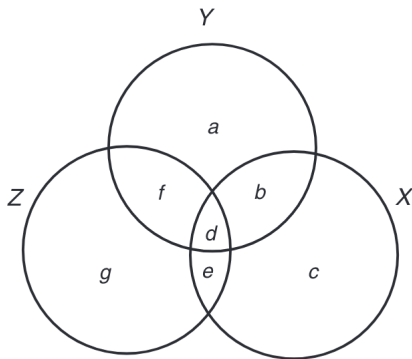


Figure 9.1. Venn diagram in which X, Y, and Z are correlated.

(Kellstedt and Whitten 2018, 212).

## No perfect multicollinearity

Validate through

1. Correlation matrix
2. Variance Inflation Factor (VIF), indicates how much variation in  $X$  is explained by other independent variables



→ Mathematical requirement, slope cannot be estimated

Remedies

1. Gather more data
2. Combine variables in index

# References I



Kellstedt, Paul M., and Guy D. Whitten. 2018. *The fundamentals of political science research*. Cambridge: Cambridge University Press.