# Regression Diagnostics and Troubleshooting II

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# Topics for Today

- 1. Heteroskedasticity and correlated errors
- 2. Bootstrapping
- 3. Outliers
- What are they?
- What problems do they cause? Biased  $\hat{\beta}$ ? Biased or large se?

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Heteroskedasticity and Correlated errors

Classical Linear Regression Model

$$y = \boldsymbol{X}\boldsymbol{\beta} + \varepsilon$$

where

$$E(\varepsilon_i) = 0$$
  
Var $(\varepsilon_i) = \sigma^2 \operatorname{Cov}(\varepsilon_i, \varepsilon_j) = 0$  for all  $i \neq j$ .

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## Examples of Heteroskedasticity and Correlated errors

- 1. Clustering: variables in same geographic area or people in group are Correlated
- Autocorrelation: observations close in time are correlated
  3.

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## General Methods to Deal with It

- 1. Weighted Least Squared
- 2. Adjusted Variance-Covariance Matrices ("Robust" standard errors)

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3. Time-series or spatial weighting methods

### Weighted Least Squares

Minimize the 3we

$$\mathop{rg\,min}\limits_{b}\sum_{i}w_{i}(y_{i}-oldsymbol{x}_{i}^{\prime}oldsymbol{eta})^{2}$$

- If right weighting: unbiased and efficient
- Different estimates than OLS,  $\hat{\beta}_{WLS} \neq \hat{\beta}_{OLS}$
- Most often used:
  - different populations
  - known measurement errors
- If don't know weights: use adjusted standard standard errors

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## "Robust" Standard errors

Run OLS, but use different variance covariance matrix

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- 1. OLS  $\hat{\beta}$
- 2. Different  $Var(\hat{\beta})$
- Different methods for different correlations
- Heteroskedasticity

## Bootstrapping

- Flexible way of generating standard errors by resampling data
  - 1. Parametric: resample from a model
  - 2. Non-parametric: resample data itself
- Why? Can calculate standard errors for unusual functions or complicated DGPs

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See examples in Assignment 03

#### Outliers

Three concepts:

- 1. Leverage: unusual points in X
- 2. Outliers: large errors  $\varepsilon_i$
- 3. Influential points:
  - points with a large effect on  $\hat{\beta}$
  - influential = leverage \* outlier

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#### What to do about unusual data?

- Can have large effects on  $\hat{\beta}$  and se
- Whether these are "wrong" depends on the DGP of those points
- Do we drop them? Generally not
  - 1. Why is it unusual? Is it bad data?
  - 2. Reformulate the model
  - 3. Learn from what the model is not capturing
  - 4. Robust and resistant methods least absolute deviations