# Research Methods for Political Science PO3110 (TCD)

HT: Tutorial 6 - Week 7

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10-11 March 2020

## Today's topics

- Categorical predictors (Recoding) (Stats HT05);
- Interaction terms (Stats HT05);
- Bootstraping (Stats HT05);
- Non-parametric tests (Stats HT06).

#### Download the data

Data: https://tinyurl.com/anes16sav Codebook: https://www.electionstudies.org/wp-content/uploads/2016/02/anes\_pilot\_2016\_CodebookUserGuide.pdf

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- Recode missing values (into new variables)
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- Bootstrapping: Random sampling with replacement from an observed dataset in order to estimate properties of the sampling distribution
- Remember to tick "Correct": This corrects for the fact that our bootstrap samples might be, on average, slightly different from our sample.
- Keep "simple" for the sample as you have not extracted more complex samples (like stratified samples etc.).
- We estimate 2000 regressions and their betas. SPSS shows the mean.

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- Perform a Shapiro-Wilk Normality Test on all scale variables: (Analyze ... Descriptives ... Explore ... Plots)
- What does the Sig. column under Shapiro-Wilk tell us? What is the null hypothesis?
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