

Fiscal Policy and Inequality

21. Regression Discontinuity Designs

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Regression Discontinuity Design (RDD)

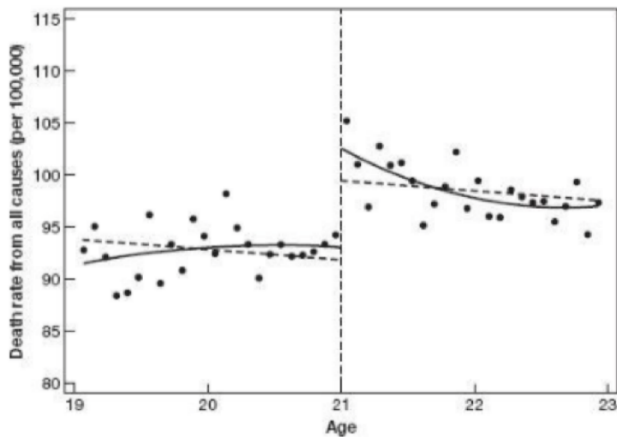
- ▶ Often rules are used to assign individuals to “treatments” which can be exploited for estimating causal effects.
 - ▶ in particular: threshold rules that are based on some ex-ante variable
- ▶ Example “running variables”:
 - ▶ Score in entry exams
 - ▶ Income for subsidy eligibility
 - ▶ Project quality score for public R&D subsidies
 - ▶ Age limit for alcohol consumption
 - ▶ Votes in an election
- ▶ The idea in RDD is to *exploit randomness around this threshold*.

Example: Effect of Minimum Legal Drinking Age on Death Rates

Carpenter and Dobkin (2009)

- ▶ outcome variable y_i : death rate
- ▶ treatment D_i : legal drinking status
- ▶ running variable x_i : age
- ▶ cutoff: at age 21, minors can suddenly drink legally

RDD Viz: Drinking Age and Death Rate



Running Variable

- ▶ The ex-ante variable is called the **running (forcing, assignment) variable**.
 - ▶ Selected threshold of the running variable divides individuals into “treated” and “not treated”
- ▶ Examples:
 - ▶ Score in entry exams
 - ▶ Income for subsidy eligibility
 - ▶ Project quality score for public R&D subsidies
 - ▶ Age limit for alcohol consumption
 - ▶ Votes in an election

RDD Conditions

- ▶ Two conditions for credible and precise RDD estimates:
 1. Variation in the treatment status near the threshold is as good as random
 - 1.1 How likely is this? Were there a way to anticipate the threshold and to manipulate the running variable?
 2. Sufficient number of observations around the threshold

RDD Estimation

- ▶ OLS regression:

$$Y_i = \alpha + \rho \mathbb{I}[x_i > c] + f(x_i)' \beta + \epsilon_i$$

- ▶ $f(x_i)$ includes polynomials in the forcing variable
 - ▶ generally linear or quadratic
 - ▶ also interact with being above or below the cutoff

Local Linear Regression

$$Y_i = \alpha + \rho \mathbb{I}[x_i > c] + \epsilon_i$$

- ▶ Limit sample to a given window (bandwidth) of width h around the cutoff point
- ▶ How to choose the bandwidth?
 - ▶ Trade-off: the closer you get the better it is for identification, but the less data you have...
- ▶ Use existent statistical algorithms for selecting the “optimal bandwidth” (e.g.: Imbens-Kalyanaraman 2011, Calonico, Cattaneo and Titiunik 2014).
 - ▶ Explore the robustness of results to different bandwidths

Testing the validity of RDD

- ▶ RD Design can be invalid if individuals can precisely manipulate the assignment variable x_i in order to get (or to avoid) treatment.
- ▶ Testing for validity:
 1. Density of the running variable should be continuous (McCrary test)
 2. Predetermined characteristics should have the same distribution just above and just below the cut off

Sharp vs. Fuzzy RD

- ▶ **Sharp RD:** treatment status (D_i) is a **deterministic** and **discontinuous** function of the running (assignment, forcing) variable (x_i):
 - ▶ $D_i = 1$ if $x_i > c$,
 - ▶ $D_i = 0$ if $x_i \leq c$
- ▶ **Fuzzy RD:** being above threshold increases **probability** of receiving treatment, rather than deterministically changing treatment.
 - ▶ In this case, use RD as first stage in two-stage least squares estimation:

$$S_i = \alpha + \gamma \mathbb{I}[x_i > c] + \eta_i$$

$$Y_i = \alpha + \rho S_i + \epsilon_i$$

RDD in Python

- ▶ For sharp RD: estimate OLS (**linearmodels.PanelOLS**)
 - ▶ subsample around cutoff, and/or with polynomials in the running variable.
- ▶ For fuzzy RD: estimate 2SLS (**linearmodels.IV2SLS**)
 - ▶ can subsample around cutoff
 - ▶ instrument is a dummy variable for being above cutoff
 - ▶ endogenous variable is whether treatment is actually assigned.
 - ▶ include polynomials in running variable as covariates.

Electoral RD: Causal Effect of Incumbency

Lee, David (2008), 'Randomized experiments from non-random selection in U.S. House elections', *Journal of Econometrics*.

- ▶ Does a Democratic candidate for a seat in the U.S. House of Representatives have an advantage if his party won the seat last time?
- ▶ Exploits the fact that (previous) election winner is determined by majority rule in vote share:
 - ▶ forcing variable is margin of victory (difference between Democratic and Republican votes shares).
- ▶ Because D_i is a deterministic function of x_i , there should be no confounding factors.
 - ▶ in particular, no electoral fraud around the cutoff.

Viz: Electoral RDD

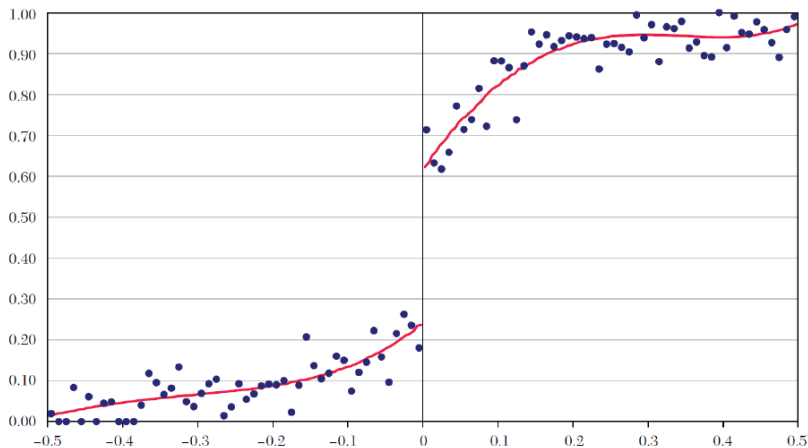


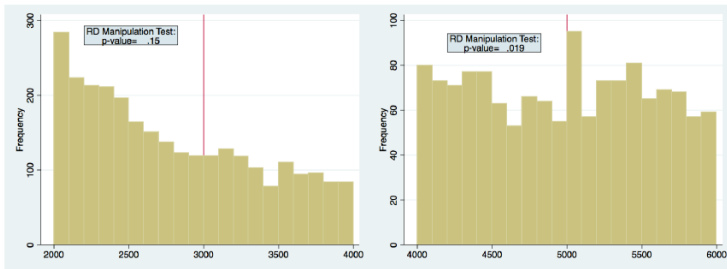
Figure 10. Winning the Next Election, Bandwidth of 0.01 (100 bins)

Electoral RDD: Notes

- ▶ Incumbency raises party re-election probabilities by 40 p.p.
- ▶ Checks:
 - ▶ Any discontinuities in predetermined covariates at the cut-off?
 - ▶ Is there bunching near the cutoff?
 - ▶ Are results robust to different bandwidths or functional form assumptions?

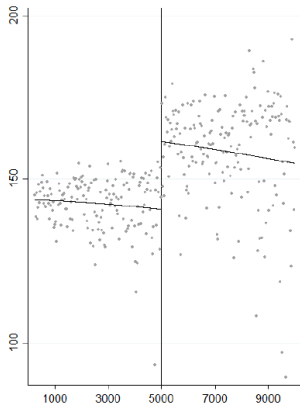
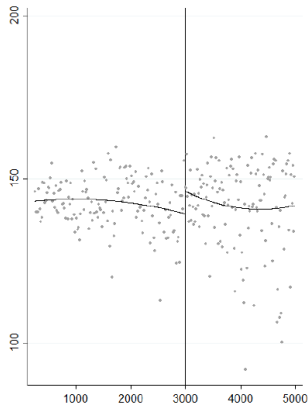
Manipulation Test: Density Around Cutoff

Bagues and Campa (2017): Histograms of Population Around Population Thresholds



Manipulation Test: Effect on Past Covariates

Bagues and Campa (2017): Federal Transfers Per Capita



RDD: Recap

- ▶ Useful method to analyze the impact of treatment when the assignment varies discontinuously due to some rules!
 - ▶ (test score, electoral results, income threshold, etc.)
- ▶ Graphical analysis is key, and can be very convincing
- ▶ Only feasible when the number of observations around the threshold is sufficiently large
- ▶ Condition for consistency: no precise manipulation at the threshold
 - ▶ may not hold if agents anticipate the threshold
- ▶ Make sure there are no other treatments at the same threshold
- ▶ RDD estimates treatment only for individuals around the threshold – effect could be different for others.