

# REGRESSION DISCONTINUITY DESIGN

# Regression Discontinuity Design

## Introduction

- Based on a special type of natural experiments
- Occurs when the probability of participation in treatment discontinuously changes with the continuous variable  $z$  (forcing variable)
- $Z$  itself may be associated with the potential outcomes, but this association is assumed to be smooth.
- Thus, any discontinuity of the outcome as a function of this covariate at the cutoff value is interpreted as evidence of a causal effect of the treatment.
- Summary in Imbens and Lemieux (2008)

# Regression Discontinuity Design

## Sharp design

- Probability of participation in treatment  $d$  changes from 0 to 1 at threshold  $z^*$

$$\lim_{z \rightarrow z^{*-}} P(d = 1 | z) = P(z^{*-}) = 0$$

$$\lim_{z \rightarrow z^{*+}} P(d = 1 | z) = P(z^{*+}) = 1$$

- treatment status is exogenously determined by the level of threshold – thus not affected by individual decision
- We estimate the effect of the treatment around the threshold – non-treated on one side are correct “counterfactuals” of treated on the other

# RD-Sharp design

## Graphical illustration

*G.W. Imbens, T. Lemieux / Journal of Econometrics 142 (2008) 615–635*

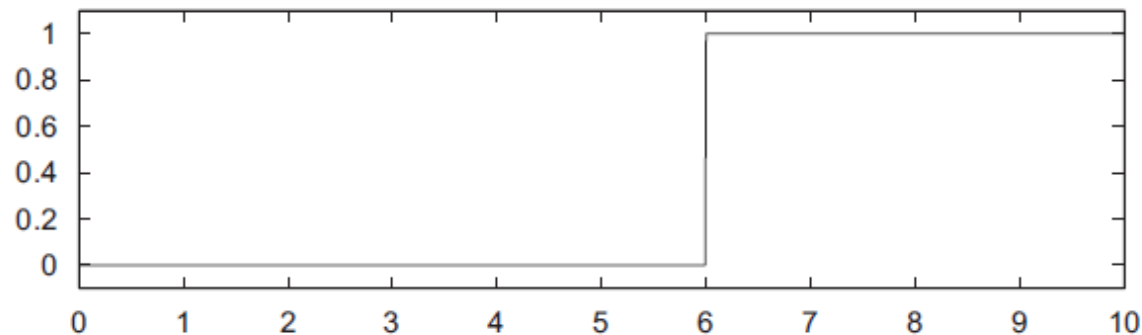


Fig. 1. Assignment probabilities (SRD).

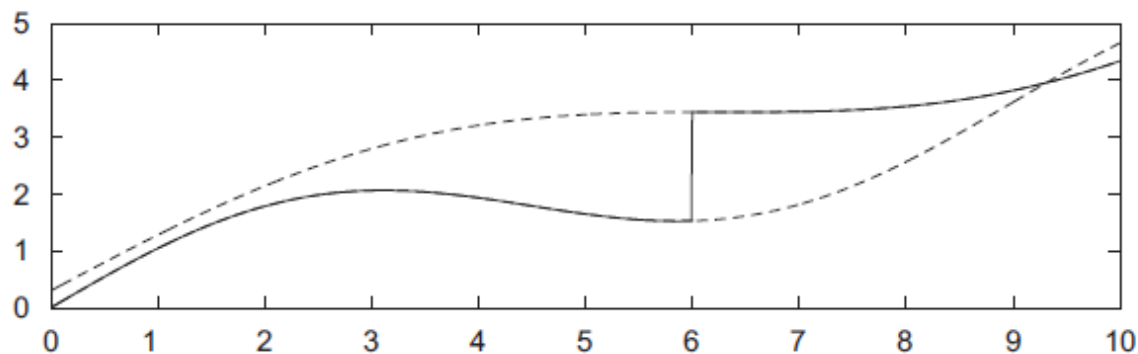


Fig. 2. Potential and observed outcome regression functions.

# RD-Sharp design

## Graphical illustration

### Eligibility based on age

- Card and Shore-Sheppard (2004):
  - ▣ expansion of the Medicaid system to cover low income children of certain age range
  - ▣ eligibility rule was based on date of birth

# Regression Discontinuity Design

## Fuzzy design

- $z$  does not fully determine participation in treatment
- Other unobserved factors also affect participation
- The discontinuity is not clear-cut and probability can change in smaller steps
- Incentives to participate in a program are not strong enough to move all from non-participation to participation over the particular threshold  $z^*$

# RD-Fuzzy design

## Graphical illustration

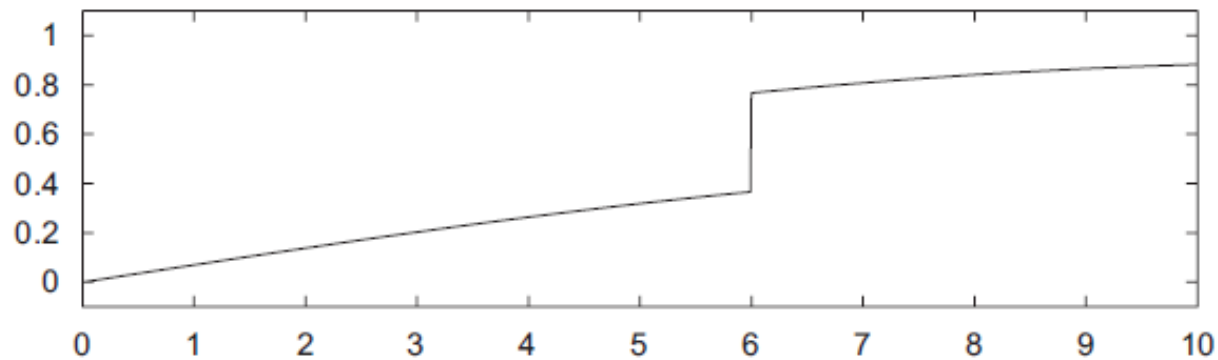


Fig. 3. Assignment probabilities (FRD).

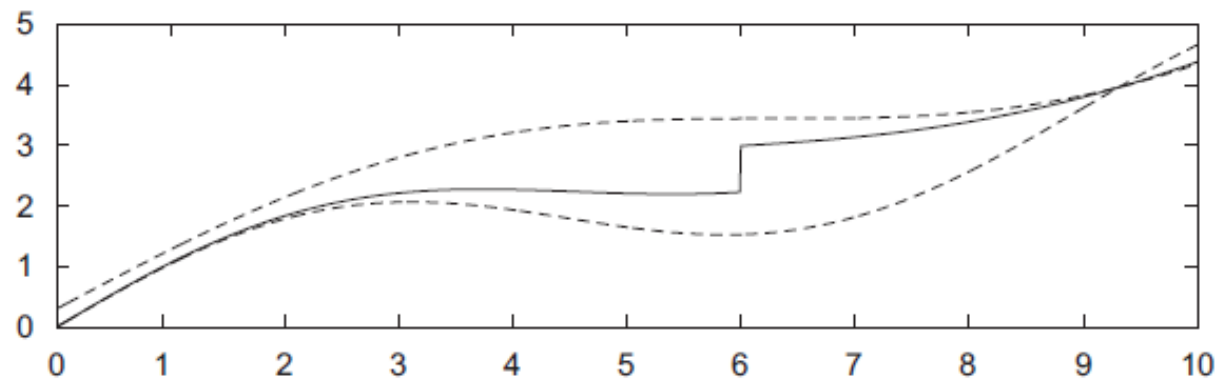


Fig. 4. Potential and observed outcome regression (FRD).

# RD-Fuzzy design

## Examples

- Effect of remedial teaching program (encouraged for students with score less than “c” (Matsudaira, 2007)
  - ▣ The effect is estimated only on compliers – those who are affected by the threshold “c” and decide to enroll
- Van der Klaauw (2002): effect of financial aid on college admission acceptance
  - ▣ SAT and other scores determine eligibility for financial aid – different groups
  - ▣ What is the effect of financial aid?
    - Causal effect: aid offer attracts more students to the college
    - Other effects: students with higher financial aid have usually better outside option from other schools
    - Other effects than discontinuity might matter too



# IV vs. RD

- **RD** is equivalent to regressing outcome  $Y$  on treatment  $d$  using  $z > c$  (forcing variable) as the instrument, applied on the subsample of  $z$  from  $[c-h, c+h]$
- **RD** is less demanding about exogeneity:
  - IV:  $z$  is exogenous (excluded)
  - RD: (i)  $Y$  is a continuous function of  $z$  at  $c$ ; (ii) cut off level is exogenous

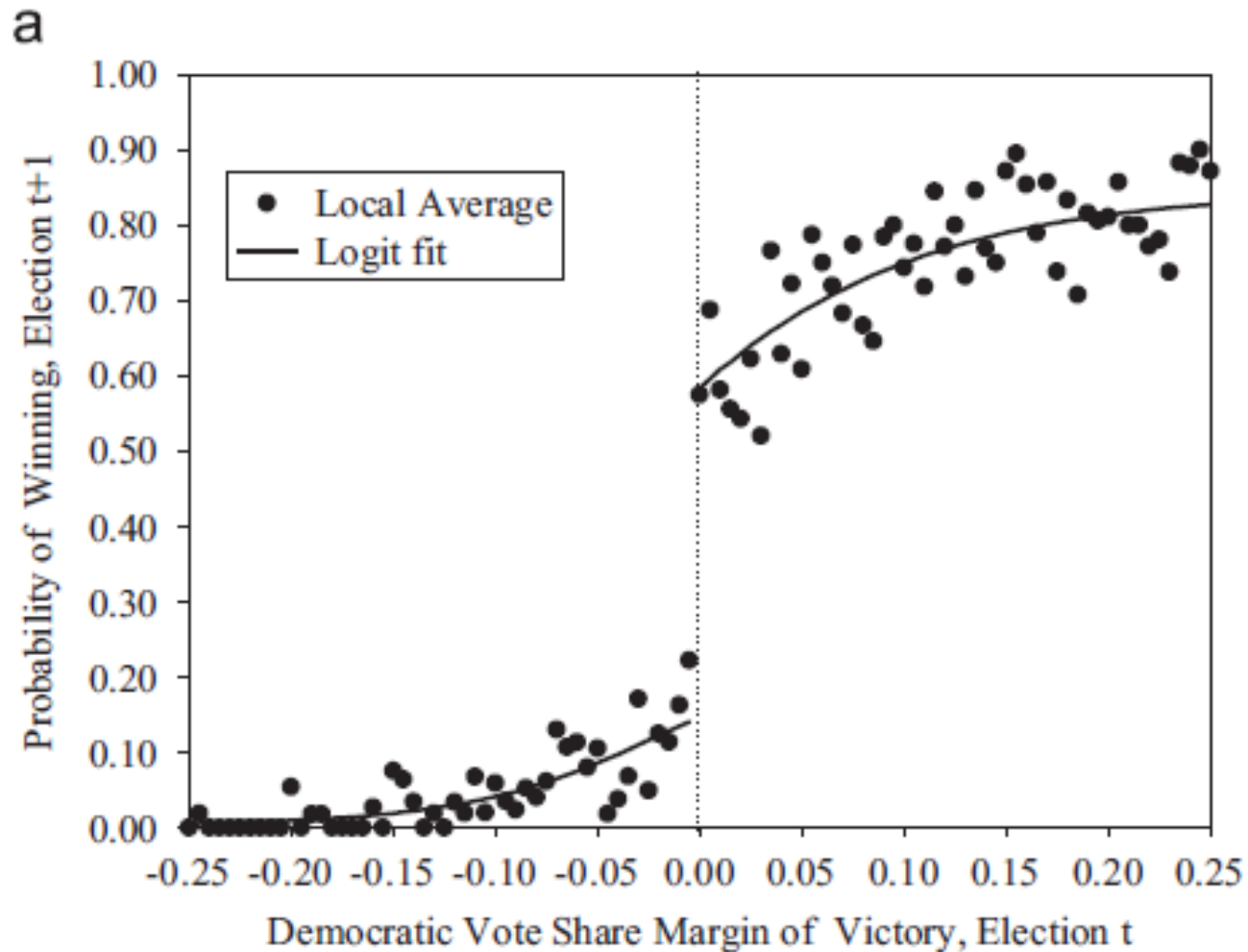
# Implementation of RD (sharp)

## Step 1: graphic analysis

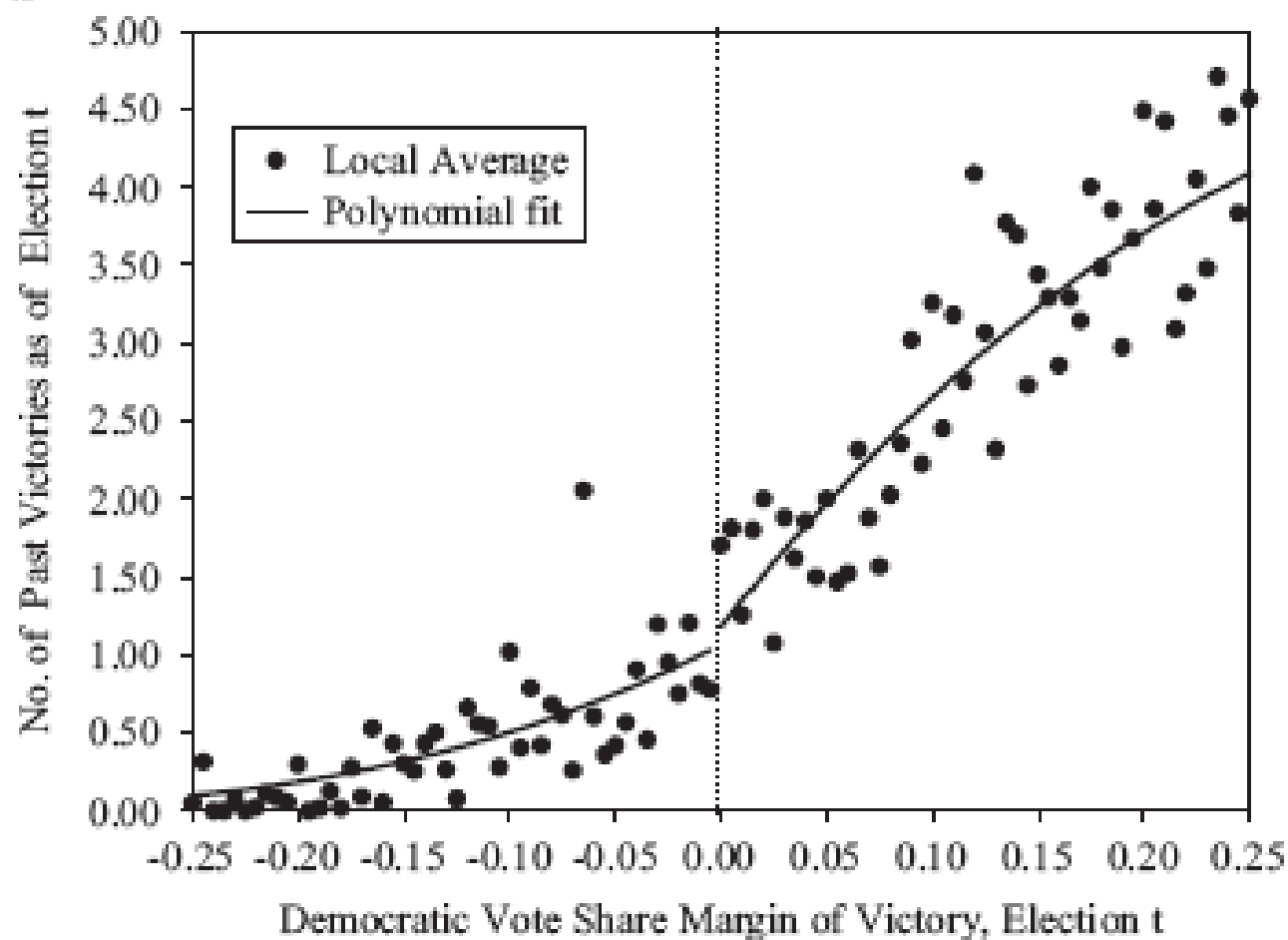
- Plot  $d$  (treatment) vs.  $z$  (forcing variable): Is there a jump at  $c$  (cutoff)?
- Plot  $Y$  (outcome) vs.  $z$ : Is there a jump at  $c$ ?
- Plot other covariates vs.  $z$ : Is there **NO** jump at  $c$ ?
- Plot the density of  $z$ : Look out for clustering of people just above  $c$  – indication of manipulation

# Example: Lee (2007)

Looking at incumbent advantage in elections for Congress



**b**



# Implementation of RD (sharp)

## Step 2: local estimation

### Simplest:

- Choose a window width  $h$
- Calculate  $E(Y \mid c-h \leq z < c)$  and  $E(Y \mid c \leq z < c+h)$ 
  - ▣ This is like fitting constant
- Problem 1: need lot of data in the neighborhood of  $c$
- Problem 2: if  $Y$  is linear in  $z$ , the bias is linear in  $h$

### Local linear regression: (non-parametric methods)

- We fit linear regression functions to the observation within distance  $h$  of discontinuity point
- Calculate value at cut-off  $\rightarrow$  use to estimate effect

# RD-summary of issues

- Identifies only **local effect** restricted to the discontinuity point
- Once the design is fuzzy
  - ▣ The discontinuity applies only to **compliers**
  - ▣ Unobserved factors can drive decision
  - ▣ If individuals can **manipulate** to which side of threshold belong based on their expectations, problems are even more severe
  - ▣ Similar problems with **IV** and **LATE**