Regression Discontinuity

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Structure

- RDD interpretation
- RDD estimation
- Placebo tests
- Sorting
- Other stuff

Interpretation

- It's a LATE!
- A different kind of LATE!
- It can be interpreted as a weighted average over all units (Lee & Lemieux 2010)
- (W, U) are observed and unobserved factors which explain all heterogeneity.
- X = c is the cutpoint on the running variable, Y is the outcome $\lim_{\epsilon \downarrow 0} E[Y|X = c + \epsilon] \lim_{\epsilon \uparrow 0} E[Y|X = c + \epsilon]$
 - $= \sum_{w,u} \tau(w,u) p(W = w, U = u | X = c)$ = $\sum_{w,u} \tau(w,u) \frac{f(c|W=w,U=u)}{f(c)} p(W = w, U = u)$
- What does this mean?
- It's a weight of individual treatment effects weighted by the likelihood that a unit will lie near the threshold on the running variable.
- Keep this in mind as you interpret results.

Estimation

- If only someone wrote a package to do this. . .
- http://github.com/ddimmery/rdd

- The current best pracices is to use local polynomial regression.
- Typically linear
- There are also some interesting methods using randomization inference, though. (Cattaneo et al n.d.)

Replication

- I'll be replicating the recent Meyersson paper that's been making noise.
- Replication materials
- The paper shows a (local) result that when Islamic parties won elections in Turkey, this resulted in better outcomes for women.
- Running variable: vote margin (but not exclusively 2 party system as in Lee)
- Outcome that we'll look at: high school education

. . .

```
require(foreign, quietly = TRUE)
d <- read.dta("regdata0.dta")
summary(d$iwm94)</pre>
```

##	Min.	1st Qu.	Median	Mean 3rd Qu.	Max.	NA's
##	-1.0	-0.5	-0.3	-0.3 -0.1	1.0	544

Explore data

• Plot the raw data.

```
. . .
```

```
with(d, plot(iwm94, hischshr1520f, pch = 19, cex = 0.2, xlim = c(-0.5, 0.5)))
left.lm <- lm(hischshr1520f ~ iwm94, d, subset = iwm94 < 0)
right.lm <- lm(hischshr1520f ~ iwm94, d, subset = iwm94 >= 0)
left.x <- seq(-0.5, 0, 0.01)
right.x <- -left.x
lines(left.x, predict(left.lm, newd = data.frame(iwm94 = left.x)), col = "red")
lines(right.x, predict(right.lm, newd = data.frame(iwm94 = right.x)), col = "red")</pre>
```



iwm94

Estimation

- So the basic estimation would just take the difference of the intercepts from left.lm and right.lm.
- And there's an equivalency to just running a single regression as Cyrus showed in class.
- But I'm just going to use rdd

```
. . .
require(rdd, quietly = TRUE)
## Loading required package: zoo
##
## Attaching package: 'zoo'
##
## The following objects are masked from 'package:base':
##
##
       as.Date, as.Date.numeric
##
## Loading required package: car
## Loading required package: survival
## Loading required package: splines
rd.out <- RDestimate(hischshr1520f ~ iwm94, d)</pre>
rd.out
##
## Call:
## RDestimate(formula = hischshr1520f ~ iwm94, data = d)
##
## Coefficients:
##
        LATE
              Half-BW Double-BW
      0.0296
             0.0250
                            0.0228
##
```

Full Results

summary(rd.out)

```
##
## Call:
## RDestimate(formula = hischshr1520f ~ iwm94, data = d)
```

```
##
## Type:
## sharp
##
## Estimates:
##
              Bandwidth Observations Estimate Std. Error z value
## LATE
              0.24
                         1020
                                       0.0296
                                                  0.0124
                                                              2.39
              0.12
                                                              1.52
## Half-BW
                          589
                                       0.0250
                                                  0.0165
## Double-BW 0.48
                         2050
                                       0.0228
                                                  0.0101
                                                              2.26
##
              Pr(|z|)
              0.0169
## LATE
## Half-BW
              0.1286
## Double-BW 0.0240
                        *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## F-statistics:
##
              F
                     Num. DoF Denom. DoF
                                           р
## LATE
               4.99 3
                               1016
                                           3.86e-03
## Half-BW
               1.70 3
                                585
                                           3.30e-01
## Double-BW 25.77 3
                               2046
                                           4.44e-16
```

Plot it

plot(rd.out, range = c(-0.4, 0.4))
title(xlab = "Islamic Party Vote Margin", ylab = "Female High School Education Share")

Placebo tests

- Do placebo tests on other covariates and other outcomes.
- They're "placebo" because there "shouldn't" be an effect on them (except occasionally by chance)

```
. . .
```

```
# Age 19+
RDestimate(ageshr19 ~ iwm94, d)[c("est", "se")]
## $est
## LATE Half-BW Double-BW
## -0.003737 0.006946 -0.004117
```



Islamic Party Vote Margin

```
##
## $se
## [1] 0.010314 0.013783 0.008307
# Log Population
RDestimate(lpop1994 ~ iwm94, d)[c("est", "se")]
## $est
##
       LATE
             Half-BW Double-BW
##
    0.06921 -0.04339
                         0.03000
##
## $se
## [1] 0.2384 0.3276 0.1879
# Household Size
```

RDestimate(shhs ~ iwm94, d)[c("est", "se")]

```
## $est
## LATE Half-BW Double-BW
## -0.006963 0.321148 -0.091759
##
## $se
## [1] 0.3543 0.5431 0.2557
```

More Placebos

```
# Men in 2000
RDestimate(hischshr1520m ~ iwm94, d)[c("est", "se")]
## $est
##
       LATE
             Half-BW Double-BW
  0.009632 0.016188 0.007619
##
##
## $se
## [1] 0.009037 0.011807 0.007435
# Women in 1990 (pre-treatment)
RDestimate(c90hischshr1520f ~ iwm94, d)[c("est", "se")]
## $est
##
       LATE Half-BW Double-BW
## 0.0079389 0.0007974 0.0130517
##
## $se
## [1] 0.012239 0.017631 0.009308
```

```
# Men in 1990 (pre-treatment)
RDestimate(c90hischshr1520m ~ iwm94, d)[c("est", "se")]
## $est
## LATE Half-BW Double-BW
## 0.005930 0.002779 0.003861
##
## $se
## [1] 0.009770 0.013259 0.007891
```

Sorting

• As Cyrus discussed, density tests are also a good way to examine the possibility of sorting.

```
. . .
```

DCdensity(d\$iwm94, verbose = TRUE, plot = FALSE)

```
## Assuming cutpoint of zero.
## Using calculated bin size: 0.009
## Using calculated bandwidth: 0.165
## Log difference in heights is -0.095 with SE 0.147
## this gives a z-stat of -0.650
## and a p value of 0.515
```

[1] 0.5154

Density Plot

DCdensity(d\$iwm94)

[1] 0.5154

Fuzzy designs

- I don't have an example for this, but it's quite easy.
- Do it the same way as before, but with RDestimate(Y~runvar+treatment)



Overall

- Some big things for RDD:
 - Lots of plots
 - Think about locality in interpretation
 - Use your covariates for robustness/placebo tests
 - Everything should be robust to different bandwidths, etc
 - If effects start disappearing as by goes down, that's a bad sign.
 - Your bandwidth is probably to wide.
- If there's still more time, maybe I'll go through some high points of the rdd code.