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RDD Replication

Github repo and summary

- 1. https://github.com/EvelynCheng-Github/RDD
- 2. What is his research question? His research question is that the effect of harsher punishments and sanctions on driving under the influence.

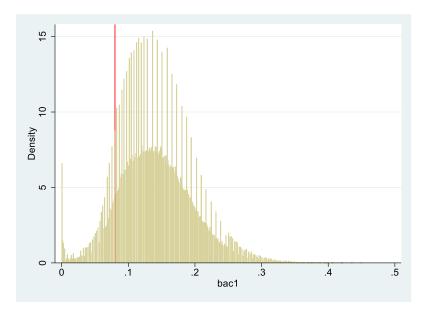
What data does he use? He utilizes the administrative records of 512,964 drunk driving parking spots in Washington State. What's more, blood alcohol content thresholds are important value to control drinking and driving.

What is his research design, or "identification strategy"? This article provides quasi-experimental evidence on the impact of severity of punishment on future crimes. In order to provide evidence for these alternative mechanisms, this article examines the degree of change in sanctions and punishments in terms of thresholds, multiple time windows for recidivism, and alcohol-related alternative crimes.

What are his conclusions? Conclusion is that the additional sanctions experienced by drunk drivers at BAC thresholds are effective in reducing repeat drunk driving.

Reproducing somewhat Hansen's results

- 3. Create a dummy
- 4. Any evidence for manipulation



BAC histogram I draw presented that there are no obvious changes around 0.08. I didn't see manipulations in these data. I find same results with Hansen and there are no evidence for sorting on the running variable.

5. Recreate Table 2 Panel A

. rdrobust white bac1, c(0.08) h(0.03 0.13) kernel(uniform)

Sharp RD estimates using local polynomial regression.

Cutoff c = .08 | Left of c Right of c Number of obs =214558 -----+------BW type = Manual Number of obs | 23010 = Uniform 191548 Kernel Eff. Number of obs | 16399 169805 VCE method = NN Order est. (p) 1 1 Order bias (q) 2 2 BW est. (h) | 0.030 0.130 BW bias (b) | 0.030 0.130 rho (h/b) | 1.000 1.000

Outcome: white. Running variable: bac1.

Method | Coef. Std. Err. z P>|z| [95% Conf. Interval]

-----+------

Robust | - - -0.1135 0.910 -.014432 .012852

Sharp RD estimates using local polynomial regression.

[.] est store model1

[.] rdrobust aged bac1, c(0.08) h(0.03 0.13) kernel(uniform)

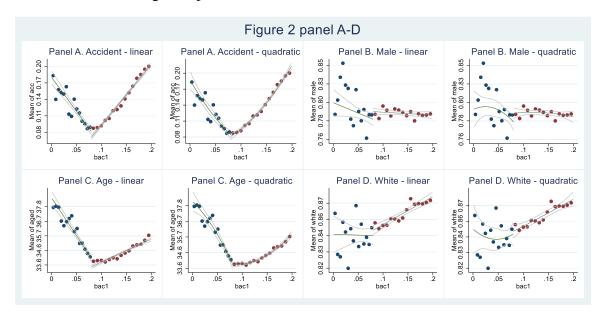
```
Cutoff c = .08 | Left of c Right of c
                                Number of obs = 214558
-----+-----+------
                              BW type
                                           Manual
  Number of obs | 23010
                       191548
                                 Kernel
                                         = Uniform
Eff. Number of obs | 16399
                       169805
                                  VCE method =
                                                  NN
 Order est. (p) | 1
                     1
 Order bias (q) |
                2
                      2
   BW est. (h) | 0.030
                     0.130
   BW bias (b) | 0.030
                     0.130
   rho (h/b) | 1.000
                    1.000
Outcome: aged. Running variable: bac1.
     Method | Coef. Std. Err. z P>|z| [95% Conf. Interval]
-----+-----+
  Robust | - - -0.5166 0.605 -.562484
                                         .327809
 ______
. est store model2
. rdrobust acc bac1, c(0.08) h(0.03 0.13) kernel(uniform)
Sharp RD estimates using local polynomial regression.
 Cutoff c = .08 | Left of c Right of c
                             Number of obs =
                                              214558
 -----+-----+------
                              BW type
                                        = Manual
  Number of obs | 23010
                                 Kernel
                                         = Uniform
                      191548
                       169805
                                  VCE method =
Eff. Number of obs | 16399
                                                  NN
 Order est. (p)
               1
                     1
 Order bias (q)
                2
                      2
```

BW est. (h) | 0.030 0.130 BW bias (b) | 0.030 0.130 rho (h/b) | 1.000 1.000 Outcome: acc. Running variable: bac1. Method | Coef. Std. Err. z P>|z| [95% Conf. Interval] ------Robust | - - -1.4959 0.135 -.019453 .002612 . est store model3 . esttab model1 model2 model3 (1) (2) (3) white aged acc -----RD_Estimate 0.00180 -0.610*** -0.0130** (0.36) (-3.72) (-3.19)N 214558 214558 214558 t statistics in parentheses

* p<0.05, ** p<0.01, *** p<0.001

When we consider white male as dependent variables, p-value is 0.719 which means we can't reject null hypothsis. The covariates aren balanced at the cutoff. But if we take age and accident into account, p-value is near 0 which presents these two covariates aren't balanced at the cutoff. Hansen's result is that age and acc are exogenous. And balance test on table 2 shows that the results of age and acc isn't statistically significant. There are no cutoff about all variables which is different than my results.

6. Recreate Figure 2 panel A-D



In Hansen's paper, demographic factors such as age, race, and gender are stable across the DUI punishment thresholds.

7. Replicate Table 3

. reg recidivism bac1 male white acc aged if bac1 > 0.03 & bac1 < 0.13

-----+-----+

. rdrobust recidivism bac1, c(0.08) $h(0.03\ 0.13)$ kernel(uniform) covs(male white acc aged)

Covariate-adjusted sharp RD estimates using local polynomial regression.

Outcome: recidivism. Running variable: bac1.

Method | Coef. Std. Err. z P>|z| [95% Conf. Interval]

[.] est store model4

Robust | - - -2.9039 0.004 -.029956 -.005814 Covariate-adjusted estimates. Additional covariates included: 4 . est store model5 . rdrobust recidivism bac1, c(0.08) h(0.03 0.13) kernel(uniform) p(2) covs(male white acc aged) Covariate-adjusted sharp RD estimates using local polynomial regression. Cutoff c = .08 | Left of c Right of c Number of obs = 214558-----+------BW type = Manual Number of obs | 23010 191548 Kernel = Uniform Eff. Number of obs | 16399 169805 VCE method = NN Order est. (p) | 2 2 Order bias (q) | 3 3 BW est. (h) | 0.030 0.130 BW bias (b) | 0.030 0.130 rho (h/b) | 1.000 1.000 Outcome: recidivism. Running variable: bac1. Method | Coef. Std. Err. z P>|z| [95% Conf. Interval] -----+------Robust | - - -2.6092 0.009 -.035159 -.004996 ._____ Covariate-adjusted estimates. Additional covariates included: 4

. est store model6

```
. esttab model4 model5 model6
         (1) (2) (3)
      recidivism recidivism recidivism
bac1 -0.0755
       (-1.60)
male
          0.0332***
        (13.07)
          0.0161***
white
        (5.54)
   0.00477
acc
        (1.40)
aged -0.000838***
        (-9.41)
RD_Estimate -0.0202*** -0.0179**
               (-4.54) (-2.90)
_cons 0.103***
        (15.87)
N
         89967
                  214558
                            214558
```

t statistics in parentheses

^{*} p<0.05, ** p<0.01, *** p<0.001

```
. reg recidivism bac1 male white acc aged if bac1 > 0.055 & bac1 < 0.105
  Source |
         SS
               df
                   MS
                       Number of obs = 46.957
F(5, 46951) = 34.24
  Residual | 4405.02672 | 46,951 | .093821787 | R-squared | = 0.0036
-----
recidivism | Coef. Std. Err. t P>|t| [95% Conf. Interval]
-----+-----
  bac1 | -.4758923 .1091708 -4.36 0.000 -.6898687 -.2619159
  male | .0357643 .0034757 10.29 0.000 .0289519 .0425767
  white | .0174948 .0039742 4.40 0.000 .0097052 .0252843
   aged | -.0007526 .0001216 -6.19 0.000 -.0009909 -.0005142
  _cons | .1284739 .0112569 11.41 0.000 .1064103 .1505375
 -----
. est store model7
. rdrobust recidivism bac1, c(0.08) h(0.055 0.105) kernel(uniform) covs(male white acc
aged)
Covariate-adjusted sharp RD estimates using local polynomial regression.
 Cutoff c = .08 \mid \text{Left of } c \mid \text{Right of } c
                         Number of obs = 214558
-----+------
                         BW type = Manual
 Number of obs | 23010 191548
                            Kernel = Uniform
Eff. Number of obs | 19435 148004
                            VCE method =
                                          NN
 Order est. (p) 1 1
```

Outcome: recidivism. Running variable: bac1.

$$Method \mid \ Coef. \ Std. \ Err. \ z \ P \!\!>\!\! |z| \ [95\% \ Conf. \ Interval]$$

-----+-----

Covariate-adjusted estimates. Additional covariates included: 4

. est store model8

. rdrobust recidivism bac1, c(0.08) $h(0.055\ 0.105)$ kernel(uniform) p(2) covs(male white acc aged)

Covariate-adjusted sharp RD estimates using local polynomial regression.

Cutoff
$$c = .08$$
 | Left of c Right of c Number of obs = 214558

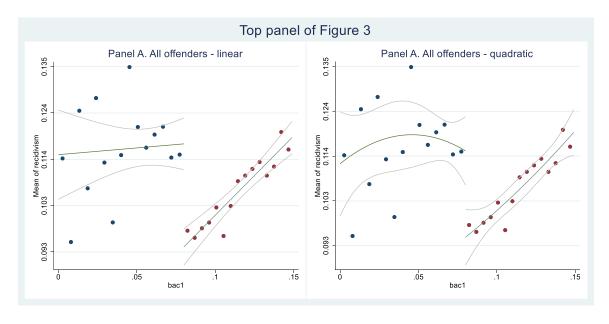
Order est.
$$(p)$$
 | 2 2

Order bias
$$(q)$$
 | 3

```
Outcome: recidivism. Running variable: bac1.
    Method | Coef. Std. Err. z P>|z| [95% Conf. Interval]
Robust | - - -2.7437 0.006 -.032208 -.005367
______
Covariate-adjusted estimates. Additional covariates included: 4
. est store model9
. esttab model7 model8 model9
           (2)
        (1)
                     (3)
     recidivism recidivism recidivism
bac1 -0.476***
      (-4.36)
        0.0358***
male
      (10.29)
        0.0175***
white
       (4.40)
       0.00434
acc
       (0.88)
       -0.000753***
aged
      (-6.19)
RD_Estimate -0.0205*** -0.0201***
```

* p<0.05, ** p<0.01, *** p<0.001

8. Recreate the top panel of Figure 3



9. The hypothesis I tested is that raw data hasn't be manipulated. And I find that there are no evidence for manipulations. Then, we check for covariate balance which I discover different results with author about age and acc. But the white variable is smooth around 0.08 that author and I get the same result. What's more, I tested regression discontinuity of having BAC above the threshold.