



```

        name: <unnamed>
        log: C:/Users/yfkas/Documents/GitHub/ARE213_Fall2023/PSet 4/Stata/pset4_logfil
> e.smcl
    log type: smcl
    opened on: 4 Dec 2023, 11:50:56

1 .
2 .
3 . // install programs
4 . // do "$do_loc/Code/01_programs.do"
5 .
6 . // analyze
7 . do "$do_loc/02_q1a.do"

8 . /*
> Title:          02_q1a.do
> Purpose:       Question 1.a, PSet 4
>
> */
9 .
10.
11.
12. use "$dta_loc/pset4_data.dta", clear

13. foreach v of varlist *vote* election* {
    2.      char `v'[_de_col_width_] 14
    3. }

14.
15.
16. /* -----
> 1a Define running var, X*/
17. gen votes_tot = votes_for + votes_against

18. gen x = votes_for/votes_tot

19. assert win == 1 if x > 0.5

20. gen w = eligible_voters

21. gen margin = x-0.5

22.
23.
24. save "$dta_loc/pset4_clean.dta", replace
    file C:/Users/yfkas/Dropbox (Personal)/ARE213/Pset4/pset4_clean.dta saved

25.
    end of do-file

26. do "$do_loc/02_q1b.do"

27. /*
> Title:          02_q1b.do
> Purpose:       Question 1.b, PSet 4
>
> */
28.
29.

```

```

30. /* -----
> lb Check RDD assumption, E[W|X]*/
31.
32.
33. use "$dta_loc/pset4_clean.dta", clear

34.
35.
36. // hist x
37. sort votes_for votes_against eligible_voters

38.
39. /*
> // Testing scatter and binscatter
> scatter w x
> xtile pctx = x, nq(100)
> preserve
>         collapse (mean) w, by(pctx)
>         scatter w pctx
> restore
> */
40.
41. /*
> rdrobust w x, ///
>         p(2) ///
>         c(0.5) ///
>         h(0.5) ///
>         masspoints(adjust) ///
>         bwselect(mserd) ///
>         kernel(tri)
>
> local h_l = e(h_l)
> local h_r = e(h_r)
> */
42.
43. // Local linear regression
44. reg w i.win##c.margin

```

Source	SS	df	MS	Number of obs	=	76,740
Model	44643727.3	3	14881242.4	F(3, 76736)	=	136.32
Residual	8.3770e+09	76,736	109167.039	Prob > F	=	0.0000
				R-squared	=	0.0053
				Adj R-squared	=	0.0053
Total	8.4217e+09	76,739	109744.532	Root MSE	=	330.4

w	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
1.win	-69.0819	4.187571	-16.50	0.000	-77.28952	-60.87428
margin	106.8179	13.95248	7.66	0.000	79.47107	134.1646
win#c.margin						
1	103.1523	17.93967	5.75	0.000	67.99067	138.314
_cons	151.7831	2.846744	53.32	0.000	146.2035	157.3627

```

45. rdrobust w x, p(1) c(0.5) h(0.5) kernel(uniform)

```

**Sharp RD estimates using local polynomial regression.**

Cutoff c = .5	Left of c	Right of c	Number of obs	=	76740
Number of obs	41656	35084	BW type	=	Manual
Eff. Number of obs	41656	35084	Kernel	=	Uniform
Order est. (p)	1	1	VCE method	=	NN
Order bias (q)	2	2			
BW est. (h)	0.500	0.500			
BW bias (b)	0.500	0.500			
rho (h/b)	1.000	1.000			

Outcome: w. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>-77.577</b>	<b>3.5019</b>	<b>-22.1532</b>	<b>0.000</b>	<b>-84.4406</b>	<b>-70.7136</b>
Robust	-	-	<b>3.5594</b>	<b>0.000</b>	<b>6.68396</b>	<b>23.0655</b>

```

46. // close
47.
48. // Visual test
49. rdplot w x, ///
    >      p(2) ///
    >      c(0.5) ///
    >      masspoints(adjust) ///
    >      /// bwselect(mserd) ///
    >      kernel(tri) ///
    >      binselect(espr) ///
    >      graph_options(legend(position(6)) ///
    >                      xtitle("Running variable") ///
    >                      ytitle("Eligible voters")) ///
    >      ci(95) ///
    >      shade

```

Mass points detected in the running variable.

RD Plot with evenly spaced number of bins using polynomial regression.

Cutoff c = .5	Left of c	Right of c	Number of obs	=	76740
			Kernel	=	Triangular
Number of obs	<b>41656</b>	<b>35084</b>			
Eff. Number of obs	<b>41656</b>	<b>35084</b>			
Order poly. fit (p)	<b>2</b>	<b>2</b>			
BW poly. fit (h)	<b>0.500</b>	<b>0.500</b>			
Number of bins scale	<b>1.000</b>	<b>1.000</b>			

Outcome: w. Running variable: x.

	Left of c	Right of c
Bins selected	<b>11</b>	<b>18</b>
Average bin length	<b>0.045</b>	<b>0.028</b>
Median bin length	<b>0.045</b>	<b>0.028</b>
IMSE-optimal bins	<b>11</b>	<b>18</b>
Mimicking Var. bins	<b>786</b>	<b>600</b>
Rel. to IMSE-optimal:		
Implied scale	<b>1.000</b>	<b>1.000</b>
WIMSE var. weight	<b>0.500</b>	<b>0.500</b>
WIMSE bias weight	<b>0.500</b>	<b>0.500</b>

```

50. graph export "$do_loc/graphs/qlb_rdplot.png", ///
    >      width(1200) height(900) ///
    >      replace
file C:/Users/yfkas/Documents/GitHub/ARE213_Fall12023/PSet
4/Stata/graphs/qlb_rdplot.png saved as PNG format

```

51.

```
52. // Statistical test
53. rdperm x w, c(0.5) perm(500) // rejects equality
```

RD Distribution Test using permutations.

Cutoff c =	.5	Left of c	Right of c	Number of obs =	76740
				Fixed q =	462
Number of obs		41656	35084	Number of perms =	500
Eff. number of obs		462	462	Test statistic =	CvM
Eff. neighbourhood		-.0090909	0		
Running variable :	X				
Covariates :	W				
		P.value			
Result		0			

```
54. cdfplot w if w < 500, by(win) // visually apparent that CDFs diverge early on
(0 observations deleted)
```

```
55. graph export "$do_loc/graphs/q1b_cdfs.png", ///
>         width(1200) height(900) 7///
>         replace
file C:/Users/yfkas/Documents/GitHub/ARE213_Fall2023/PSet
4/Stata/graphs/q1b_cdfs.png saved as PNG format
```

```

56.
57. /*
  > Q: is checking whether  $E[W_i | X_i]$  is continuous at the cutoff
  > be a useful placebo check for the RDD assumptions
  >
  > Testing for the continuity of this expectation can be done visually using a
  > scatter plot. If a ruler can be placed across the entire graph, then
  > the continuity of  $E[W_i | X_i]$  is satisfied in a local linear model. The ruler
  > analogy can be extended to a polynomial of degree  $p$ , hence the result is
  > somewhat arbitrary. The bandwidth introduces more arbitrariness. Canay and Kamat
  > (2017) suggested comparing CDFs of  $W$  at either side of the cutoff. The
  > result of rdperm shows that the null of equality of distributions can be
  > rejected. Visually inspecting the CDFs confirms this.
  >
  > */
58.
59.
60.
end of do-file

```

```
61. do "$do loc/02 q1c.do"
```

```
62. /*
> Title:          02_q1c.do
> Purpose:       Question 1.c, PSet 4
>
> */
```

```
63.
64. /* -----
> 1c Check RDD assumption, density of Xi is continuous around c.
> */
```

```
65.  
66.  
67. use "$dta loc/pset4 clean.dta", clear
```

```

68.
69. hist x // histogram seems ok
    (bin=48, start=0, width=.02083333)

70.
71. // try rddensity
72. rddensity x, c(0.45) plot kernel(triangular) all
    Computing data-driven bandwidth selectors.

```

Point estimates and standard errors have been adjusted for repeated observations.  
(Use option *nomasspoints* to suppress this adjustment.)

RD Manipulation test using local polynomial density estimation.

c =	0.450	Left of c	Right of c	Number of obs =	76740
				Model =	unrestricted
Number of obs		35726	41014	BW method =	comb
Eff. Number of obs		8397	7756	Kernel =	triangular
Order est. (p)		2	2	VCE method =	jackknife
Order bias (q)		3	3		
BW est. (h)		0.059	0.064		

Running variable: x.

Method	T	P> T
Conventional	-0.7890	0.4301
Robust	-0.9183	0.3584

P-values of binomial tests. (H0: prob = .5)

Window Length / 2	<c	>=c	P> T
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000

```

73. graph export "$do_loc/graphs/qlc_45.png", ///
    > width(1200) height(900) ///
    > replace
    file C:/Users/yfkas/Documents/GitHub/ARE213_Fall2023/PSet 4/Stata/graphs/qlc_45.png
    saved as PNG format

```

```

74.
75. rddensity x, c(0.5) plot kernel(triangular) all
    Computing data-driven bandwidth selectors.

```

Point estimates and standard errors have been adjusted for repeated observations.  
(Use option *nomasspoints* to suppress this adjustment.)

RD Manipulation test using local polynomial density estimation.

c =	0.500	Left of c	Right of c	Number of obs =	76740
				Model =	unrestricted
Number of obs		41656	35084	BW method =	comb
Eff. Number of obs		6592	4859	Kernel =	triangular
Order est. (p)		2	2	VCE method =	jackknife
Order bias (q)		3	3		
BW est. (h)		0.055	0.041		

Running variable: x.

Method	T	P> T
Conventional	<b>-13.3528</b>	<b>0.0000</b>
Robust	<b>-10.9449</b>	<b>0.0000</b>

P-values of binomial tests. (H0: prob = .5)

[illegible]

```
76. graph export "$do_loc/graphs/q1c_50.png", ///
>         width(1200) height(900) 7///
>         replace
file C:/Users/yfkas/Documents/GitHub/ARE213_Fall12023/PSet 4/Stata/graphs/q1c_50.png
      saved as PNG format
```

```
77.
78. rddensity x, c(0.55) plot kernel(triangular) all
    Computing data-driven bandwidth selectors.
```

Point estimates and standard errors have been adjusted for repeated observations.  
(Use option *nomasspoints* to suppress this adjustment.)

RD Manipulation test using local polynomial density estimation.

c =	<b>0.550</b>	Left of c	Right of c	Number of obs =	<b>76740</b>
				Model =	<b>unrestricted</b>
Eff. Number of obs		<b>47372</b>	<b>29368</b>	BW method =	<b>comb</b>
Order est. (p)		<b>2</b>	<b>2</b>	Kernel =	<b>triangular</b>
Order bias (q)		<b>3</b>	<b>3</b>	VCE method =	<b>jackknife</b>
BW est. (h)		<b>0.062</b>	<b>0.065</b>		

Running variable: x.

Method	T	P> T
Conventional	<b>-0.4940</b>	<b>0.6213</b>
Robust	<b>2.3461</b>	<b>0.0190</b>

P-values of binomial tests. (H0: prob = .5)

[illegible]

```

79. graph export "$do_loc/graphs/qlc_55.png", ///
>       width(1200) height(900) 7///
>       replace
file C:/Users/yfkas/Documents/GitHub/ARE213_Fall2023/PSet 4/Stata/graphs/qlc_55.png
    saved as PNG format

80.
81. /*
> This algorithm compares densities below and above the cutoff using
> bandwidths calculated as follows:
>       1) minimize the MSE of each density estimator to the L and R separately
>       2) minimize the MSE of the difference of the two density estimators
>       3) repeat (2) for the sum
>       3) Take the median of the three bandwidths.
> I have also set it to use a triangular kernel to emphasize observations
> closer to the cutoff. One notices that shifting the cutoff slightly around
> c=0.5 shows that indeed there appears to be manipulation at c = 0.5 but not
> at those other points.
> */
82.
83.
    end of do-file

84. do "$do_loc/02_qld.do"

85. /*
> Title:           02_qld.do
> Purpose:         Question 1.d, PSet 4
>
> */
86.
87.
88. /* -----
> 1d Drop inconclusive races and repeat qlb and c
> */
89.
90. use "$dta_loc/pset4_clean.dta", clear

91.
92. drop if x == 0.5
    (956 observations deleted)

93.
94.
95.
96. // repeat 1b -----
97. // Visual test
98. rdplot w x, ///
>       p(2) ///
>       c(0.5) ///
>       masspoints(adjust) ///
>       /// bwselect(mserd) ///
>       kernel(tri) ///
>       binselect(espr) ///
>       graph_options(legend(position(6))) ///
>       ci(95) ///
>       shade

```

**Mass points detected in the running variable.**

**RD Plot with evenly spaced number of bins using polynomial regression.**

Cutoff c = .5	Left of c	Right of c	Number of obs =	75784
			Kernel	= Triangular
Number of obs	41656	34128		
Eff. Number of obs	41656	34128		
Order poly. fit (p)	2	2		
BW poly. fit (h)	0.500	0.500		
Number of bins scale	1.000	1.000		

Outcome: w. Running variable: x.

	Left of c	Right of c
Bins selected	11	20
Average bin length	0.045	0.025
Median bin length	0.045	0.025
IMSE-optimal bins	11	20
Mimicking Var. bins	778	609
Rel. to IMSE-optimal:		
Implied scale	1.000	1.000
WIMSE var. weight	0.500	0.500
WIMSE bias weight	0.500	0.500

99.

100 // Statistical test

101 rdperm x w, c(0.5) perm(500) // fails to reject equality

RD Distribution Test using permutations.

Cutoff c =	.5	Left of c	Right of c	Number of obs =	75784
				Fixed q =	383
Number of obs		41656	34128	Number of perms =	500
Eff. number of obs		383	383	Test statistic =	CvM
Eff. neighbourhood		-.00819671	.00887573		
Running variable :	x				
Covariates :	w				
		P.value			
Result		.186			

102 cdfplot w if w < 500, by(win)  
(0 observations deleted)

103 graph export "\$do\_loc/graphs/qld\_cdfs.png", ///  
> width(1200) height(900) ///  
> replace

file C:/Users/yfkas/Documents/GitHub/ARE213\_Fall2023/PSet  
4/Stata/graphs/qld\_cdfs.png saved as PNG format

104 // although still visually apparent that CDFs diverge early on

105

106

107

108 // repeat 1c -----

109 hist x // histogram seems ok

(bin=48, start=0, width=.02083333)

110

111 // try rddensity

112 rddensity x, c(0.45) plot kernel(triangular) all

Computing data-driven bandwidth selectors.

Point estimates and standard errors have been adjusted for repeated observations.  
(Use option *nomasspoints* to suppress this adjustment.)

RD Manipulation test using local polynomial density estimation.

c =	0.450	Left of c	Right of c	Number of obs =	75784
				Model =	unrestricted
Number of obs		35726	40058	BW method =	comb
Eff. Number of obs		8399	8200	Kernel =	triangular
Order est. (p)		2	2	VCE method =	jackknife
Order bias (q)		3	3		
BW est. (h)		0.059	0.076		



Running variable: **x**.

Method	T	P> T
Conventional	<b>4.5365</b>	<b>0.0000</b>
Robust	<b>0.3655</b>	<b>0.7148</b>

**P-values of binomial tests.** (H0: prob = .5)

Window Length / 2	<c	>=c	P> T
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000
0.000	89	0	0.0000

```
113 graph export "$do_loc/graphs/qld_45.png", ///
> width(1200) height(900) 7///
> replace
file C:/Users/yfkas/Documents/GitHub/ARE213_Fall2023/PSet 4/Stata/graphs/qld_45.png
saved as PNG format
```

```
114
115 rddensity x, c(0.5) plot kernel(triangular) all
Computing data-driven bandwidth selectors.
```

Point estimates and standard errors have been adjusted for repeated observations.  
(Use option *nomasspoints* to suppress this adjustment.)

**RD Manipulation test using local polynomial density estimation.**

c =	<b>0.500</b>	Left of c	Right of c	Number of obs =	<b>75784</b>
Number of obs		<b>41656</b>	<b>34128</b>	Model =	<b>unrestricted</b>
Eff. Number of obs		<b>6591</b>	<b>5156</b>	BW method =	<b>comb</b>
Order est. (p)		<b>2</b>	<b>2</b>	Kernel =	<b>triangular</b>
Order bias (q)		<b>3</b>	<b>3</b>	VCE method =	<b>jackknife</b>
BW est. (h)		<b>0.055</b>	<b>0.053</b>		

Running variable: **x**.

Method	T	P> T
Conventional	<b>-3.8780</b>	<b>0.0001</b>
Robust	<b>-3.4202</b>	<b>0.0006</b>

**P-values of binomial tests.** (H0: prob = .5)

Window Length / 2	<c	>=c	P> T
0.002	11	9	0.8238
0.003	62	45	0.1215
0.005	140	120	0.2386
0.007	269	237	0.1681
0.008	404	345	0.0340
0.010	575	484	0.0057
0.012	808	649	0.0000
0.013	1008	795	0.0000
0.015	1236	976	0.0000
0.017	1544	1200	0.0000

```

116 graph export "$do_loc/graphs/qld_50.png", ///
>       width(1200) height(900) 7///
>       replace
file C:/Users/yfkas/Documents/GitHub/ARE213_Fall2023/PSet 4/Stata/graphs/qld_50.png
    saved as PNG format

```

```

117
118 rddensity x, c(0.55) plot kernel(triangular) all
    Computing data-driven bandwidth selectors.

```

Point estimates and standard errors have been adjusted for repeated observations.  
(Use option *nomasspoints* to suppress this adjustment.)

RD Manipulation test using local polynomial density estimation.

c =	0.550	Left of c	Right of c	Number of obs =	75784
				Model =	unrestricted
Number of obs		46416	29368	BW method =	comb
Eff. Number of obs		5638	6763	Kernel =	triangular
Order est. (p)		2	2	VCE method =	jackknife
Order bias (q)		3	3		
BW est. (h)		0.062	0.065		

Running variable: x.

Method	T	P> T
Conventional	-2.0580	0.0396
Robust	5.0569	0.0000

P-values of binomial tests. (H0: prob = .5)

Window Length / 2	<c	>=c	P> T
0.000	0	66	0.0000
0.000	0	66	0.0000
0.000	0	66	0.0000
0.000	0	66	0.0000
0.000	0	66	0.0000
0.000	0	66	0.0000
0.000	0	66	0.0000
0.000	0	66	0.0000
0.000	0	66	0.0000
0.000	0	66	0.0000

```

119 graph export "$do_loc/graphs/qld_55.png", ///
>       width(1200) height(900) 7///
>       replace
file C:/Users/yfkas/Documents/GitHub/ARE213_Fall2023/PSet 4/Stata/graphs/qld_55.png
    saved as PNG format

```

```

120
121
122 /*
> Ans: Explain.
> Dropping observations right at the cutoff leads us to fail to reject the
> null hypothesis that the densities are not continuous at the cutoff.
>
> What occurs is that there was no bunching over 0.5. Rather, many races end up
> tied. Dropping such inconclusive races shows that even though there are
> relatively fewer observations right before the cutoff, there also appear to be
> a relatively lower observations from conclusive races right above the cutoff.
> So even though there is a visual dip in the estimated density plot, there is
> no discontinuity.
>
> */

```

```

123
124 save "$dta_loc/pset4_trim.dta", replace
    file C:/Users/yfkas/Dropbox (Personal)/ARE213/Pset4/pset4_trim.dta saved
125
126
127     end of do-file

128 do "$do_loc/02_q1e.do"

129 /*
    > Title:          02_q1e.do
    > Purpose:        Question 1.e, PSet 4
    >
    > */
130
131
132
133 /* -----
    > 1d Redefine x and
    > */
134 use "$dta_loc/pset4_trim.dta", clear

135 gen xp = votes_for - votes_against, after(x)

136 sort xp

137
138 // cdfplot xp
139
140 xtile xp_100 = xp, nq(100)

141 // scatter xp_100 xp
142 // drop if inlist(xp_100, 1,2,99,100)
143
144 // Test for continuity of mean firm size around the cutoff.
145 // Visual test
146 rdplot w xp if !inlist(xp_100, 1,2,99,100) , ///
    >     p(2) ///
    >     c(0) ///
    >     masspoints(adjust) ///
    >     bwselect(mserd) ///
    >     h(`h_l' `h_r') ///
    >     kernel(tri) ///
    >     binselect(espr) ///
    >     graph_options(legend(position(6))) ///
    >     ci(95) ///
    >     shade

```

**Mass points detected in the running variable.**

**RD Plot with evenly spaced number of bins using polynomial regression.**

Cutoff c = 0	Left of c	Right of c	Number of obs =	72756
			Kernel	= Triangular
Number of obs	40138	32618		
Eff. Number of obs	40138	32618		
Order poly. fit (p)	2	2		
BW poly. fit (h)	159.000	148.000		
Number of bins scale	1.000	1.000		

Outcome: w. Running variable: xp.

	Left of c	Right of c
Bins selected	30	16
Average bin length	5.300	9.250
Median bin length	5.300	9.250
IMSE-optimal bins	30	16
Mimicking Var. bins	346	137
Rel. to IMSE-optimal:		
Implied scale	1.000	1.000
WIMSE var. weight	0.500	0.500
WIMSE bias weight	0.500	0.500

```

147 graph export "$do_loc/graphs/q1e.png", ///
>       width(1200) height(900) ///
>       replace
file C:/Users/yfkas/Documents/GitHub/ARE213_Fall2023/PSet 4/Stata/graphs/q1e.png
    saved as PNG format

148
149
150 /* 1e and 1f Ans:
>
> 1e: Redefining x as the difference shows that firms tend to be smaller if the race
> is close. That is, larger firms have larger majorities either for or against.
>
> 1f: From the exercise in q1d, it appears that there is indeed bunching
> that causes close races. Dropping inconclusive races results in balance tests
> that fail to reject the identifying assumption of the continuity of potential
> outcomes which we test using balance of w and x at the cutoff.
>
> */
151
152 save "$dta_loc/pset4_trim2.dta", replace
file C:/Users/yfkas/Dropbox (Personal)/ARE213/Pset4/pset4_trim2.dta saved

153
154
155
156
end of do-file

157 do "$do_loc/02_q2a.do"

158 /*
> Title:          02_q2a.do
> Purpose:        Question 2.a, PSet 4
>
> */
159
160
161
162 use "$dta_loc/pset4_trim2.dta", clear

163

```

```

164 gen y = logwage
165
166 scatter y x
167 graph export "$do_loc/graphs/q2a_yx.png", ///
>         width(1200) height(900) ///
>         replace
file C:/Users/yfkas/Documents/GitHub/ARE213_Fall2023/PSet 4/Stata/graphs/q2a_yx.png
saved as PNG format
168
169 // statistically inspect ATE at c
170 reg y i.win##c.margin // Local linear regression

```

Source	SS	df	MS	Number of obs	=	75,784
Model	1098.54679	3	366.182262	F(3, 75780)	=	435.10
Residual	63776.1907	75,780	.841596605	Prob > F	=	0.0000
				R-squared	=	0.0169
				Adj R-squared	=	0.0169
Total	64874.7375	75,783	.856059242	Root MSE	=	.91739

  

y	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
1.win	.1551971	.0118305	13.12	0.000	.1320094	.1783848
margin	.1714447	.0397388	4.31	0.000	.0935568	.2493326
win#c.margin						
1	.0565245	.0505913	1.12	0.264	-.0426343	.1556833
_cons	11.34204	.0082005	1383.09	0.000	11.32597	11.35812

```

171 rdrobust y x, p(1) c(0.5) h(0.5) kernel(uniform)

```

**Sharp RD estimates using local polynomial regression.**

Cutoff c = .5	Left of c	Right of c	Number of obs	=	75784
Number of obs	41656	34128	BW type	=	Manual
Eff. Number of obs	41656	34128	Kernel	=	Uniform
Order est. (p)	1	1	VCE method	=	NN
Order bias (q)	2	2			
BW est. (h)	0.500	0.500			
BW bias (b)	0.500	0.500			
rho (h/b)	1.000	1.000			

**Outcome: y. Running variable: x.**

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	.1552	.01161	13.3654	0.000	.132438	.177956
Robust	-	-	8.7974	0.000	.118795	.1869

```

172
173 // visually inspect ATE at c
174 rdplot y x, ///
>         p(1) ///
>         c(0.5) ///
>         masspoints(adjust) ///
>         /// bwselect(mserd) ///
>         kernel(tri) ///
>         binselect(espr) ///
>         graph_options(legend(position(6))) ///
>         ci(95) ///
>         shade

```

**Mass points detected in the running variable.**

**RD Plot with evenly spaced number of bins using polynomial regression.**

Cutoff c = .5	Left of c	Right of c	Number of obs =	75784
Number of obs	41656	34128	Kernel	= Triangular
Eff. Number of obs	41656	34128		
Order poly. fit (p)	1	1		
BW poly. fit (h)	0.500	0.500		
Number of bins scale	1.000	1.000		

Outcome: y. Running variable: x.

	Left of c	Right of c
Bins selected	6	6
Average bin length	0.083	0.083
Median bin length	0.083	0.083
IMSE-optimal bins	6	6
Mimicking Var. bins	604	599
Rel. to IMSE-optimal:		
Implied scale	1.000	1.000
WIMSE var. weight	0.500	0.500
WIMSE bias weight	0.500	0.500

```
175 graph export "$do_loc/graphs/q2a_h50.png", ///
> width(1200) height(900) ///
> replace
file C:/Users/yfkas/Documents/GitHub/ARE213_Fall2023/PSet 4/Stata/graphs/q2a_h50.png
saved as PNG format
```

176

177 // setting h by minimizing MSE tightens the bandwidth and exaggerates the ATE

178 rdrobust y x, p(1) c(0.5) kernel(uniform)

Mass points detected in the running variable.

Sharp RD estimates using local polynomial regression.

Cutoff c = .5	Left of c	Right of c	Number of obs =	75784
Number of obs	41656	34128	BW type	= mserd
Eff. Number of obs	14499	11060	Kernel	= Uniform
Order est. (p)	1	1	VCE method	= NN
Order bias (q)	2	2		
BW est. (h)	0.110	0.110		
BW bias (b)	0.206	0.206		
rho (h/b)	0.533	0.533		
Unique obs	7352	5274		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	.17225	.02428	7.0948	0.000	.124668 .219839
Robust	-	-	6.0272	0.000	.117361 .230472

Estimates adjusted for mass points in the running variable.

```
179 local h_1 `e(h_1)'
```

```

180 local h_r `e(h_r)'
181 rdplot y x, ///
>      p(1) ///
>      c(0.5) ///
>      masspoints(adjust) ///
>      /// bwselect(mserd) ///
>      h(`h_l' `h_r') ///
>      kernel(tri) ///
>      binselect(espr) ///
>      graph_options(legend(position(6))) ///
>      ci(95) ///
>      shade

```

**Mass points detected in the running variable.**

**RD Plot with evenly spaced number of bins using polynomial regression.**

Cutoff c = .5	Left of c	Right of c	Number of obs =	75784
			Kernel =	Triangular
Number of obs	41656	34128		
Eff. Number of obs	14499	11060		
Order poly. fit (p)	1	1		
BW poly. fit (h)	0.110	0.110		
Number of bins scale	1.000	1.000		

**Outcome: y. Running variable: x.**

	Left of c	Right of c
Bins selected	6	6
Average bin length	0.083	0.083
Median bin length	0.083	0.083
IMSE-optimal bins	6	6
Mimicking Var. bins	604	599
Rel. to IMSE-optimal:		
Implied scale	1.000	1.000
WIMSE var. weight	0.500	0.500
WIMSE bias weight	0.500	0.500

```

182 graph export "$do_loc/graphs/q2a_hopt.png", ///
>      width(1200) height(900) ///
>      replace
file C:/Users/yfkas/Documents/GitHub/ARE213_Fall2023/PSet
4/Stata/graphs/q2a_hopt.png saved as PNG format

```

```

183
184 /*
> // plotting xp against y is very different and sensitive to outliers
> rdplot y xp if !inlist(xp_100, 1,2,99,100), ///
>      p(2) ///
>      c(0.5) ///
>      masspoints(adjust) ///
>      /// bwselect(mserd) ///
>      kernel(tri) ///
>      binselect(espr) ///
>      graph_options(legend(position(6))) ///
>      ci(95) ///
>      shade
>
> rdplot y xp, ///
>      p(2) ///
>      c(0.5) ///
>      masspoints(adjust) ///
>      /// bwselect(mserd) ///
>      kernel(tri) ///
>      binselect(espr) ///
>      graph_options(legend(position(6))) ///
>      ci(95) ///
>

```

```

>          shade
> */
185
186
187 /* 2a Discuss:
>
> Upon visual inspection, the outcome appears to follow a local linear trend
> on both sides of the cutoff. I therefore implement a local linear regression
> using the full range of the running variable as a bandwidth. I also use a
> uniform kernel but a triangular one does not change the results by much. My
> approach is thus equivalent to a specification where the treatment and the
> margin (running variable less cutoff) are fully interacted. A local linear
> specification is not always the most informative because the conditional
> expectation of Y given X can have some curvature. In our case, it apparently
> does not.
>
> The effect is a statistically significant positive number.
>
> 2c: see discussion above.
>
> */
188
189
190     end of do-file

191 do "$do_loc/02_q2b.do"

192 /*
> Title:          02_q2b.do
> Purpose:        Question 2.b, PSet 4
>
> */
193
194 /*
> What's the estimand? The estimand is  $\hat{\tau}$ , the same one as in slide 15.
> Larger or smaller firms? As we showed in q 1
>
> */
195
196 use "$dta_loc/pset4_trim2.dta", clear

197
198 gen y = logwage

199
200 rdrobust y x, p(1) c(0.5) kernel(uniform)

```

**Mass points detected in the running variable.**

**Sharp RD estimates using local polynomial regression.**

Cutoff c = .5	Left of c	Right of c	Number of obs =	75784
			BW type =	mserd
Number of obs	41656	34128	Kernel =	Uniform
Eff. Number of obs	14499	11060	VCE method =	NN
Order est. (p)	1	1		
Order bias (q)	2	2		
BW est. (h)	0.110	0.110		
BW bias (b)	0.206	0.206		
rho (h/b)	0.533	0.533		
Unique obs	7352	5274		

**Outcome: y. Running variable: x.**

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	.17225	.02428	7.0948	0.000	.124668 .219839
Robust	-	-	6.0272	0.000	.117361 .230472

Estimates adjusted for mass points in the running variable.



```

201 local h_l `e(h_l)'
202 local h_r `e(h_r)'
203 rdplot y x, ///
>     p(1) ///
>     c(0.5) ///
>     masspoints(adjust) ///
>     /// bwselect(mserd) ///
>     h(`h_l' `h_r') ///
>     kernel(tri) ///
>     binselect(espr) ///
>     graph_options(legend(position(6))) ///
>     ci(95) ///
>     shade

```

**Mass points detected in the running variable.**

**RD Plot with evenly spaced number of bins using polynomial regression.**

Cutoff c = .5	Left of c	Right of c	Number of obs =	<b>75784</b>
			Kernel =	<b>Triangular</b>
Number of obs	<b>41656</b>	<b>34128</b>		
Eff. Number of obs	<b>14499</b>	<b>11060</b>		
Order poly. fit (p)	<b>1</b>	<b>1</b>		
BW poly. fit (h)	<b>0.110</b>	<b>0.110</b>		
Number of bins scale	<b>1.000</b>	<b>1.000</b>		

**Outcome: y. Running variable: x.**

	Left of c	Right of c
Bins selected	<b>6</b>	<b>6</b>
Average bin length	<b>0.083</b>	<b>0.083</b>
Median bin length	<b>0.083</b>	<b>0.083</b>
IMSE-optimal bins	<b>6</b>	<b>6</b>
Mimicking Var. bins	<b>604</b>	<b>599</b>
Rel. to IMSE-optimal:		
Implied scale	<b>1.000</b>	<b>1.000</b>
WIMSE var. weight	<b>0.500</b>	<b>0.500</b>
WIMSE bias weight	<b>0.500</b>	<b>0.500</b>

```

204
205 // around the cutoff, firms are not larger
206 gen localtobw = inrange(x, 0.5-`h_l', 0.5+`h_r')
207 ttest w, by(localtobw)

```

Two-sample t test with equal variances

Group	Obs	Mean	Std. err.	Std. dev.	[95% conf. interval]	
0	<b>50,225</b>	<b>131.7441</b>	<b>1.663188</b>	<b>372.736</b>	<b>128.4842</b>	<b>135.004</b>
1	<b>25,559</b>	<b>134.3151</b>	<b>1.482614</b>	<b>237.0283</b>	<b>131.4091</b>	<b>137.2211</b>
Combined	<b>75,784</b>	<b>132.6112</b>	<b>1.210376</b>	<b>333.203</b>	<b>130.2389</b>	<b>134.9835</b>
diff		<b>-2.571003</b>	<b>2.560152</b>		<b>-7.588889</b>	<b>2.446883</b>

```

diff = mean(0) - mean(1)
H0: diff = 0
Ha: diff < 0
Pr(T < t) = 0.1576

```

t = **-1.0042**  
Degrees of freedom = **75782**

```

Ha: diff != 0
Pr(|T| > |t|) = 0.3153
Ha: diff > 0
Pr(T > t) = 0.8424

```

```

208
    end of do-file

209 do "$do_loc/02_q2d.do"

210 /*
    > Title:          02_q2d.do
    > Purpose:        Question 2.d, PSet 4
    >
    > */
211
212
213
214 use "$dta_loc/pset4_trim2.dta", clear

215
216 gen y = logwage

217
218 local h_max = 50

219 local step = 50

220 forval i = 1(`h_max' / `step') `h_max' {
    2.      dis "rdrobust with h(`i'/100)"
    3.      qui rdrobust y x, p(1) c(0.5) h(`i'/100) kernel(uniform)
    4.
221      // get stats
222      local tau_i = `e(tau_cl)'
    5.      local z_bc = e(tau_bc) / e(se_tau_rb)
    6.      local bc_lb_i = e(tau_bc) - invnormal(0.975)*e(se_tau_rb)
    7.      local bc_ub_i = e(tau_bc) + invnormal(0.975)*e(se_tau_rb)
    8. }
    rdrobust with h(1/100)
    rdrobust with h(2/100)
    rdrobust with h(3/100)
    rdrobust with h(4/100)
    rdrobust with h(5/100)
    rdrobust with h(6/100)
    rdrobust with h(7/100)
    rdrobust with h(8/100)
    rdrobust with h(9/100)
    rdrobust with h(10/100)
    rdrobust with h(11/100)
    rdrobust with h(12/100)
    rdrobust with h(13/100)
    rdrobust with h(14/100)
    rdrobust with h(15/100)
    rdrobust with h(16/100)
    rdrobust with h(17/100)
    rdrobust with h(18/100)
    rdrobust with h(19/100)
    rdrobust with h(20/100)
    rdrobust with h(21/100)
    rdrobust with h(22/100)
    rdrobust with h(23/100)
    rdrobust with h(24/100)
    rdrobust with h(25/100)
    rdrobust with h(26/100)
    rdrobust with h(27/100)
    rdrobust with h(28/100)
    rdrobust with h(29/100)
    rdrobust with h(30/100)
    rdrobust with h(31/100)
    rdrobust with h(32/100)
    rdrobust with h(33/100)
    rdrobust with h(34/100)
    rdrobust with h(35/100)
    rdrobust with h(36/100)
    rdrobust with h(37/100)
    rdrobust with h(38/100)
    rdrobust with h(39/100)

```

```
rdrobust with h(40/100)
rdrobust with h(41/100)
rdrobust with h(42/100)
rdrobust with h(43/100)
rdrobust with h(44/100)
rdrobust with h(45/100)
rdrobust with h(46/100)
rdrobust with h(47/100)
rdrobust with h(48/100)
rdrobust with h(49/100)
rdrobust with h(50/100)
```

224

226 clear

Number of observations (**N**) was 0, now 50.

(50 missing values generated)

```
(50 missing values generated)
```

(50 missing values generated)

(50 missing values generated)

```
233 forval i = 1/'=_N' {
2.     replace h      = `i'/'`h_max' in `i'
3.     replace tau    = `tau_`i' in `i'
4.     replace rb_lb  = `bc_lb_`i' in `i'
5.     replace rb_ub  = `bc_ub_`i' in `i'
6. }
```

(1 real change made)

(1 real change made)

(1 real change made)

(1 real change made)

(1 real change made)

(1 real change made)

(1 real change made)

```
(1 real change made)
```

(1 real change made)

(1 real change made)

(1 real change made)

```
(1 real change made)
(1 real change made)
```

```
(1 real change made)
(1 real change made)
```

```
(1 real change made)
(1 real change made)
```

(1 real change made)  
(1 real change made)

```
(1 Real Change Made)
(1 real change made)
```

```
(1 real change made)
(1 real change made)
```

(1 real change made)  
(1 real change made)

```
(1 real change made)
(1 real change made)
```

```
(1 real change made)
(1 real change made)
```

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```
(1 real change made)
```

(1 real change made)

```
(1 real change made)
(1 real change made)
```

```
(1 real change made)
(1 real change made)
```

(1 real change made)

[illegible]

[illegible]

[illegible]

```

248 /*
> Title:          02_q2e.do
> Purpose:       Question 2.e, PSet 4
>
> */
249 /*
> 2.E Alter the cutoff.
>
> */
250
251 use "$dta_loc/pset4_trim2.dta", clear

252
253 gen y = logwage

254
255 local c_range 30(10)70

256
257 forval i = 1/41 {
2.      local c = 29+`i' // c= 30(1)70
3.
258      gen margin_`c' = x - `c'/100
          gen win_`c'   = x > `c'/100
4.
5. }

259
260 // Local linear regression
261 /*
> forval i = 1/41 {
>      local c = 29+`i' // c= 30(1)70
>      if `c' < 50 reg y i.win_`c'##c.margin_`c' if x < 0.5
>      else if `c' == 50 reg y i.win_`c'##c.margin_`c'
>      else if `c' > 50 reg y i.win_`c'##c.margin_`c' if x > 0.5
> }
> */
262
263 local rdopts ""

264 forval i = 1/41 {
2.
265      // get c
266      local c = 29+`i' // c= 30(1)70
3.      local c_reg = `c'/100
4.
267      if `c' < 50 rdrobust y x if x < 0.5, c(`c_reg') p(1) h(50/100) kernel(
> uniform)
5.      else if `c' == 50 rdrobust y x, c(`c_reg') p(1) h(50/
> 100) kernel(uniform)
6.      else if `c' > 50 rdrobust y x if x > 0.5, c(`c_reg') p(1) h(50/100) kern
> el(uniform)
7.
268      // get stats
269      local tau_`c' = `e(tau_cl)'
8.      dis "tau_`c' = `tau_`c'""
9.      local z_bc = e(tau_bc) / e(se_tau_rb)
10.     local bc_lb_`c' = e(tau_bc) - invnormal(0.975)*e(se_tau_rb)
11.     local bc_ub_`c' = e(tau_bc) + invnormal(0.975)*e(se_tau_rb)
12. }

```

**Sharp RD estimates using local polynomial regression.**

Cutoff c = .3	Left of c	Right of c	Number of obs =	41656
			BW type =	Manual
Number of obs	15195	26461	Kernel =	Uniform
Eff. Number of obs	15195	26461	VCE method =	NN
Order est. (p)	1	1		
Order bias (q)	2	2		
BW est. (h)	0.500	0.500		
BW bias (b)	0.500	0.500		
rho (h/b)	1.000	1.000		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	.01502	.01845	0.8140	0.416	-.021149	.051192
Robust	-	-	-0.1407	0.888	-.056689	.049096

tau\_30 = .015021428558839

Sharp RD estimates using local polynomial regression.

Cutoff c = .31	Left of c	Right of c	Number of obs =	41656
			BW type =	Manual
Number of obs	16406	25250	Kernel =	Uniform
Eff. Number of obs	16406	25250	VCE method =	NN
Order est. (p)	1	1		
Order bias (q)	2	2		
BW est. (h)	0.500	0.500		
BW bias (b)	0.500	0.500		
rho (h/b)	1.000	1.000		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	.00878	.01824	0.4810	0.631	-.026983	.044535
Robust	-	-	-0.5599	0.576	-.067179	.037325

tau\_31 = .0087761259873247

Sharp RD estimates using local polynomial regression.

Cutoff c = .32	Left of c	Right of c	Number of obs =	41656
			BW type =	Manual
Number of obs	17674	23982	Kernel =	Uniform
Eff. Number of obs	17674	23982	VCE method =	NN
Order est. (p)	1	1		
Order bias (q)	2	2		
BW est. (h)	0.500	0.500		
BW bias (b)	0.500	0.500		
rho (h/b)	1.000	1.000		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	.01224	.0181	0.6764	0.499	-.023234	.047722
Robust	-	-	-0.1910	0.849	-.056797	.046711

tau\_32 = .0122437935695849

Sharp RD estimates using local polynomial regression.

Cutoff c = .33	Left of c	Right of c	Number of obs =	41656
			BW type =	Manual
Number of obs	18859	22797	Kernel =	Uniform
Eff. Number of obs	18859	22797	VCE method =	NN
Order est. (p)	1	1		
Order bias (q)	2	2		
BW est. (h)	0.500	0.500		
BW bias (b)	0.500	0.500		
rho (h/b)	1.000	1.000		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	.02172	.01802	1.2052	0.228	-.013602	.057044
Robust	-	-	0.6917	0.489	-.0333	.069627

tau\_33 = .0217211098977259



**Sharp RD estimates using local polynomial regression.**

Cutoff c = .34	Left of c	Right of c	Number of obs =	<b>41656</b>
			BW type =	<b>Manual</b>
Number of obs	<b>20061</b>	<b>21595</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>20061</b>	<b>21595</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

**Outcome: y. Running variable: x.**

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>.01245</b>	<b>.018</b>	<b>0.6914</b>	<b>0.489</b>	<b>-.022836</b>	<b>.047727</b>
Robust	<b>-</b>	<b>-</b>	<b>0.1006</b>	<b>0.920</b>	<b>-.048578</b>	<b>.053833</b>

tau\_34 = .0124454009049089

**Sharp RD estimates using local polynomial regression.**

Cutoff c = .35	Left of c	Right of c	Number of obs =	<b>41656</b>
			BW type =	<b>Manual</b>
Number of obs	<b>21507</b>	<b>20149</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>21507</b>	<b>20149</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

**Outcome: y. Running variable: x.**

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>.01003</b>	<b>.01816</b>	<b>0.5525</b>	<b>0.581</b>	<b>-.025558</b>	<b>.045622</b>
Robust	<b>-</b>	<b>-</b>	<b>0.1085</b>	<b>0.914</b>	<b>-.049084</b>	<b>.054836</b>

tau\_35 = .0100319166975602

**Sharp RD estimates using local polynomial regression.**

Cutoff c = .36	Left of c	Right of c	Number of obs =	<b>41656</b>
			BW type =	<b>Manual</b>
Number of obs	<b>22802</b>	<b>18854</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>22802</b>	<b>18854</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

**Outcome: y. Running variable: x.**

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>.01462</b>	<b>.01828</b>	<b>0.8000</b>	<b>0.424</b>	<b>-.0212</b>	<b>.050444</b>
Robust	<b>-</b>	<b>-</b>	<b>0.6753</b>	<b>0.499</b>	<b>-.03406</b>	<b>.069868</b>

tau\_36 = .0146219233892815

**Sharp RD estimates using local polynomial regression.**

Cutoff c = .37	Left of c	Right of c	Number of obs =	<b>41656</b>
			BW type =	<b>Manual</b>
Number of obs	<b>24211</b>	<b>17445</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>24211</b>	<b>17445</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>-.00777</b>	<b>.01851</b>	<b>-0.4198</b>	<b>0.675</b>	<b>-.04406</b>	<b>.028515</b>
Robust	<b>-</b>	<b>-</b>	<b>-0.9555</b>	<b>0.339</b>	<b>-.078265</b>	<b>.026964</b>

tau\_37 = -.0077726241658844

Sharp RD estimates using local polynomial regression.

Cutoff c = .38	Left of c	Right of c	Number of obs =	<b>41656</b>
			BW type =	<b>Manual</b>
Number of obs	<b>25775</b>	<b>15881</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>25775</b>	<b>15881</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>-.01689</b>	<b>.01906</b>	<b>-0.8863</b>	<b>0.375</b>	<b>-.054254</b>	<b>.020466</b>
Robust	<b>-</b>	<b>-</b>	<b>-1.6830</b>	<b>0.092</b>	<b>-.102506</b>	<b>.007793</b>

tau\_38 = -.0168941355655257

Sharp RD estimates using local polynomial regression.

Cutoff c = .39	Left of c	Right of c	Number of obs =	<b>41656</b>
			BW type =	<b>Manual</b>
Number of obs	<b>27154</b>	<b>14502</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>27154</b>	<b>14502</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>.00011</b>	<b>.01954</b>	<b>0.0058</b>	<b>0.995</b>	<b>-.038175</b>	<b>.038403</b>
Robust	<b>-</b>	<b>-</b>	<b>-0.3402</b>	<b>0.734</b>	<b>-.066784</b>	<b>.047028</b>

tau\_39 = .0001140979514993

Sharp RD estimates using local polynomial regression.

Cutoff c = .4	Left of c	Right of c	Number of obs =	<b>41656</b>
			BW type =	<b>Manual</b>
Number of obs	<b>28253</b>	<b>13403</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>28253</b>	<b>13403</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	<b>-.01523</b>	<b>.0195</b>	<b>-0.7812</b>	<b>0.435</b>	<b>-.053441 .022982</b>
Robust	<b>-</b>	<b>-</b>	<b>-1.4866</b>	<b>0.137</b>	<b>-.095815 .013159</b>

tau\_40 = -.0152294819179382

Sharp RD estimates using local polynomial regression.

Cutoff c = .41	Left of c	Right of c	Number of obs =	<b>41656</b>
			BW type =	<b>Manual</b>
Number of obs	<b>29990</b>	<b>11666</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>29990</b>	<b>11666</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	<b>-.01446</b>	<b>.02069</b>	<b>-0.6986</b>	<b>0.485</b>	<b>-.055012 .026099</b>
Robust	<b>-</b>	<b>-</b>	<b>-1.6120</b>	<b>0.107</b>	<b>-.108443 .010563</b>

tau\_41 = -.0144561424662797

Sharp RD estimates using local polynomial regression.

Cutoff c = .42	Left of c	Right of c	Number of obs =	<b>41656</b>
			BW type =	<b>Manual</b>
Number of obs	<b>31455</b>	<b>10201</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>31455</b>	<b>10201</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	<b>-.00157</b>	<b>.0219</b>	<b>-0.0715</b>	<b>0.943</b>	<b>-.04449 .041358</b>
Robust	<b>-</b>	<b>-</b>	<b>-0.8279</b>	<b>0.408</b>	<b>-.089807 .036469</b>

tau\_42 = -.0015662699972339

Sharp RD estimates using local polynomial regression.

Cutoff c = .43	Left of c	Right of c	Number of obs =	<b>41656</b>
			BW type =	<b>Manual</b>
Number of obs	<b>32904</b>	<b>8752</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>32904</b>	<b>8752</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>.00041</b>	<b>.0236</b>	<b>0.0176</b>	<b>0.986</b>	<b>-.045843</b>	<b>.046671</b>
Robust	<b>-</b>	<b>-</b>	<b>-0.6693</b>	<b>0.503</b>	<b>-.094178</b>	<b>.04623</b>

tau\_43 = .0004143517504502

Sharp RD estimates using local polynomial regression.

Cutoff c = .44	Left of c	Right of c	Number of obs =	<b>41656</b>
			BW type =	<b>Manual</b>
Number of obs	<b>34314</b>	<b>7342</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>34314</b>	<b>7342</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>.00958</b>	<b>.0252</b>	<b>0.3801</b>	<b>0.704</b>	<b>-.03982</b>	<b>.058982</b>
Robust	<b>-</b>	<b>-</b>	<b>0.1102</b>	<b>0.912</b>	<b>-.0696</b>	<b>.077894</b>

tau\_44 = .0095809283334063

Sharp RD estimates using local polynomial regression.

Cutoff c = .45	Left of c	Right of c	Number of obs =	<b>41656</b>
			BW type =	<b>Manual</b>
Number of obs	<b>35726</b>	<b>5930</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>35726</b>	<b>5930</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>.02032</b>	<b>.0279</b>	<b>0.7283</b>	<b>0.466</b>	<b>-.034359</b>	<b>.074992</b>
Robust	<b>-</b>	<b>-</b>	<b>1.2153</b>	<b>0.224</b>	<b>-.031722</b>	<b>.135257</b>

tau\_45 = .0203162987549561

Sharp RD estimates using local polynomial regression.

Cutoff c = .46	Left of c	Right of c	Number of obs =	<b>41656</b>
			BW type =	<b>Manual</b>
Number of obs	<b>36962</b>	<b>4694</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>36962</b>	<b>4694</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>-.03429</b>	<b>.03026</b>	<b>-1.1333</b>	<b>0.257</b>	<b>-.093603</b>	<b>.025016</b>
Robust	<b>-</b>	<b>-</b>	<b>-1.3311</b>	<b>0.183</b>	<b>-.141402</b>	<b>.02702</b>

tau\_46 = -.0342936662187014

Sharp RD estimates using local polynomial regression.

Cutoff c = .47	Left of c	Right of c	Number of obs =	<b>41656</b>
			BW type =	<b>Manual</b>
Number of obs	<b>38420</b>	<b>3236</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>38420</b>	<b>3236</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>.01087</b>	<b>.0372</b>	<b>0.2923</b>	<b>0.770</b>	<b>-.062032</b>	<b>.083775</b>
Robust	<b>-</b>	<b>-</b>	<b>1.3783</b>	<b>0.168</b>	<b>-.031512</b>	<b>.180843</b>

tau\_47 = .0108714762746409

Sharp RD estimates using local polynomial regression.

Cutoff c = .48	Left of c	Right of c	Number of obs =	<b>41656</b>
			BW type =	<b>Manual</b>
Number of obs	<b>39677</b>	<b>1979</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>39677</b>	<b>1979</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>-.08226</b>	<b>.04421</b>	<b>-1.8606</b>	<b>0.063</b>	<b>-.168909</b>	<b>.004393</b>
Robust	<b>-</b>	<b>-</b>	<b>-1.4809</b>	<b>0.139</b>	<b>-.239282</b>	<b>.033311</b>

tau\_48 = -.0822580530512731

Sharp RD estimates using local polynomial regression.

Cutoff c = .49	Left of c	Right of c	Number of obs =	<b>41656</b>
			BW type =	<b>Manual</b>
Number of obs	<b>41088</b>	<b>568</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>41088</b>	<b>568</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>.03427</b>	<b>.08067</b>	<b>0.4248</b>	<b>0.671</b>	<b>-.123841</b>	<b>.192383</b>
Robust	<b>-</b>	<b>-</b>	<b>0.8097</b>	<b>0.418</b>	<b>-.141107</b>	<b>.33976</b>

tau\_49 = .0342712501831883

Sharp RD estimates using local polynomial regression.

Cutoff c = .5	Left of c	Right of c	Number of obs =	<b>75784</b>
			BW type =	<b>Manual</b>
Number of obs	<b>41656</b>	<b>34128</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>41656</b>	<b>34128</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>.1552</b>	<b>.01161</b>	<b>13.3654</b>	<b>0.000</b>	<b>.132438</b>	<b>.177956</b>
Robust	<b>-</b>	<b>-</b>	<b>8.7974</b>	<b>0.000</b>	<b>.118795</b>	<b>.1869</b>

tau\_50 = .155197106007666

Sharp RD estimates using local polynomial regression.

Cutoff c = .51	Left of c	Right of c	Number of obs =	<b>34128</b>
			BW type =	<b>Manual</b>
Number of obs	<b>482</b>	<b>33646</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>482</b>	<b>33646</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>-.01941</b>	<b>.06182</b>	<b>-0.3141</b>	<b>0.753</b>	<b>-.140574</b>	<b>.101744</b>
Robust	<b>-</b>	<b>-</b>	<b>-0.5866</b>	<b>0.557</b>	<b>-.215795</b>	<b>.116378</b>

tau\_51 = -.0194146579748917

Sharp RD estimates using local polynomial regression.

Cutoff c = .52	Left of c	Right of c	Number of obs =	<b>34128</b>
			BW type =	<b>Manual</b>
Number of obs	<b>1666</b>	<b>32462</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>1666</b>	<b>32462</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	<b>-.04578</b>	<b>.03491</b>	<b>-1.3116</b>	<b>0.190</b>	<b>-.114198 .022631</b>
Robust	<b>-</b>	<b>-</b>	<b>-0.8126</b>	<b>0.416</b>	<b>-.131196 .054293</b>

tau\_52 = -.0457833334740378

Sharp RD estimates using local polynomial regression.

Cutoff c = .53	Left of c	Right of c	Number of obs =	<b>34128</b>
			BW type =	<b>Manual</b>
Number of obs	<b>2632</b>	<b>31496</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>2632</b>	<b>31496</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	<b>-.01349</b>	<b>.03108</b>	<b>-0.4340</b>	<b>0.664</b>	<b>-.074413 .047435</b>
Robust	<b>-</b>	<b>-</b>	<b>-0.3319</b>	<b>0.740</b>	<b>-.103083 .073228</b>

tau\_53 = -.0134892261037152

Sharp RD estimates using local polynomial regression.

Cutoff c = .54	Left of c	Right of c	Number of obs =	<b>34128</b>
			BW type =	<b>Manual</b>
Number of obs	<b>3769</b>	<b>30359</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>3769</b>	<b>30359</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	<b>.01915</b>	<b>.02587</b>	<b>0.7401</b>	<b>0.459</b>	<b>-.031555 .069846</b>
Robust	<b>-</b>	<b>-</b>	<b>1.0729</b>	<b>0.283</b>	<b>-.033464 .114405</b>

tau\_54 = .0191453971836175

Sharp RD estimates using local polynomial regression.

Cutoff c = .55	Left of c	Right of c	Number of obs =	<b>34128</b>
			BW type =	<b>Manual</b>
Number of obs	<b>4760</b>	<b>29368</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>4760</b>	<b>29368</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>.01545</b>	<b>.02415</b>	<b>0.6395</b>	<b>0.522</b>	<b>-.031895</b>	<b>.062789</b>
Robust	<b>-</b>	<b>-</b>	<b>0.5433</b>	<b>0.587</b>	<b>-.051888</b>	<b>.091689</b>

tau\_55 = .0154470996587825

Sharp RD estimates using local polynomial regression.

Cutoff c = .56000000000000001	Left of c	Right of c	Number of obs =	<b>34128</b>
			BW type =	<b>Manual</b>
Number of obs	<b>5868</b>	<b>28260</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>5868</b>	<b>28260</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>.00461</b>	<b>.02197</b>	<b>0.2097</b>	<b>0.834</b>	<b>-.038451</b>	<b>.047664</b>
Robust	<b>-</b>	<b>-</b>	<b>-0.5253</b>	<b>0.599</b>	<b>-.082074</b>	<b>.047381</b>

tau\_56 = .0046067195915391

Sharp RD estimates using local polynomial regression.

Cutoff c = .57	Left of c	Right of c	Number of obs =	<b>34128</b>
			BW type =	<b>Manual</b>
Number of obs	<b>6972</b>	<b>27156</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>6972</b>	<b>27156</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>-.00931</b>	<b>.02072</b>	<b>-0.4492</b>	<b>0.653</b>	<b>-.049921</b>	<b>.031305</b>
Robust	<b>-</b>	<b>-</b>	<b>-1.4678</b>	<b>0.142</b>	<b>-.106717</b>	<b>.015322</b>

tau\_57 = -.0093076136005159

Sharp RD estimates using local polynomial regression.



Cutoff c = .58	Left of c	Right of c	Number of obs =	<b>34128</b>
			BW type =	<b>Manual</b>
Number of obs	<b>8095</b>	<b>26033</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>8095</b>	<b>26033</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	<b>.02369</b>	<b>.01955</b>	<b>1.2121</b>	<b>0.225</b>	<b>-.014619 .062006</b>
Robust	<b>-</b>	<b>-</b>	<b>1.0032</b>	<b>0.316</b>	<b>-.027958 .086587</b>

tau\_58 = .0236935265799367

Sharp RD estimates using local polynomial regression.

Cutoff c = .59	Left of c	Right of c	Number of obs =	<b>34128</b>
			BW type =	<b>Manual</b>
Number of obs	<b>9067</b>	<b>25061</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>9067</b>	<b>25061</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	<b>.03116</b>	<b>.01912</b>	<b>1.6299</b>	<b>0.103</b>	<b>-.006312 .068641</b>
Robust	<b>-</b>	<b>-</b>	<b>1.4291</b>	<b>0.153</b>	<b>-.015403 .098337</b>

tau\_59 = .0311646223006825

Sharp RD estimates using local polynomial regression.

Cutoff c = .6	Left of c	Right of c	Number of obs =	<b>34128</b>
			BW type =	<b>Manual</b>
Number of obs	<b>9868</b>	<b>24260</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>9868</b>	<b>24260</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	<b>.03126</b>	<b>.01894</b>	<b>1.6509</b>	<b>0.099</b>	<b>-.005854 .068377</b>
Robust	<b>-</b>	<b>-</b>	<b>1.1919</b>	<b>0.233</b>	<b>-.022494 .092316</b>

tau\_60 = .0312615341529892

Sharp RD estimates using local polynomial regression.

Cutoff c = .61	Left of c	Right of c	Number of obs =	34128
			BW type =	Manual
Number of obs	11060	23068	Kernel =	Uniform
Eff. Number of obs	11060	23068	VCE method =	NN
Order est. (p)	1	1		
Order bias (q)	2	2		
BW est. (h)	0.500	0.500		
BW bias (b)	0.500	0.500		
rho (h/b)	1.000	1.000		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	.02702	.01788	1.5115	0.131	-.008017 .062065
Robust	-	-	1.0029	0.316	-.025451 .078789

tau\_61 = .0270239747992704

Sharp RD estimates using local polynomial regression.

Cutoff c = .62	Left of c	Right of c	Number of obs =	34128
			BW type =	Manual
Number of obs	12020	22108	Kernel =	Uniform
Eff. Number of obs	12020	22108	VCE method =	NN
Order est. (p)	1	1		
Order bias (q)	2	2		
BW est. (h)	0.500	0.500		
BW bias (b)	0.500	0.500		
rho (h/b)	1.000	1.000		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	.01711	.01757	0.9742	0.330	-.017316 .05154
Robust	-	-	0.2231	0.823	-.045703 .057442

tau\_62 = .0171116405654494

Sharp RD estimates using local polynomial regression.

Cutoff c = .63	Left of c	Right of c	Number of obs =	34128
			BW type =	Manual
Number of obs	12981	21147	Kernel =	Uniform
Eff. Number of obs	12981	21147	VCE method =	NN
Order est. (p)	1	1		
Order bias (q)	2	2		
BW est. (h)	0.500	0.500		
BW bias (b)	0.500	0.500		
rho (h/b)	1.000	1.000		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	.01478	.01718	0.8607	0.389	-.018882 .048451
Robust	-	-	0.2014	0.840	-.04477 .055024

tau\_63 = .0147847834244317

Sharp RD estimates using local polynomial regression.

Cutoff c = .64	Left of c	Right of c	Number of obs =	<b>34128</b>
			BW type =	<b>Manual</b>
Number of obs	<b>13819</b>	<b>20309</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>13819</b>	<b>20309</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>.01531</b>	<b>.01712</b>	<b>0.8946</b>	<b>0.371</b>	<b>-.018238</b>	<b>.048865</b>
Robust	<b>-</b>	<b>-</b>	<b>0.3826</b>	<b>0.702</b>	<b>-.04043</b>	<b>.060045</b>

tau\_64 = .0153138155797485

Sharp RD estimates using local polynomial regression.

Cutoff c = .65	Left of c	Right of c	Number of obs =	<b>34128</b>
			BW type =	<b>Manual</b>
Number of obs	<b>14679</b>	<b>19449</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>14679</b>	<b>19449</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>.02027</b>	<b>.01713</b>	<b>1.1833</b>	<b>0.237</b>	<b>-.013306</b>	<b>.053847</b>
Robust	<b>-</b>	<b>-</b>	<b>0.8962</b>	<b>0.370</b>	<b>-.027532</b>	<b>.073919</b>

tau\_65 = .0202708768525568

Sharp RD estimates using local polynomial regression.

Cutoff c = .66	Left of c	Right of c	Number of obs =	<b>34128</b>
			BW type =	<b>Manual</b>
Number of obs	<b>15489</b>	<b>18639</b>	Kernel =	<b>Uniform</b>
Eff. Number of obs	<b>15489</b>	<b>18639</b>	VCE method =	<b>NN</b>
Order est. (p)	<b>1</b>	<b>1</b>		
Order bias (q)	<b>2</b>	<b>2</b>		
BW est. (h)	<b>0.500</b>	<b>0.500</b>		
BW bias (b)	<b>0.500</b>	<b>0.500</b>		
rho (h/b)	<b>1.000</b>	<b>1.000</b>		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Conventional	<b>.00671</b>	<b>.01717</b>	<b>0.3907</b>	<b>0.696</b>	<b>-.026948</b>	<b>.040366</b>
Robust	<b>-</b>	<b>-</b>	<b>-0.1697</b>	<b>0.865</b>	<b>-.055517</b>	<b>.04667</b>

tau\_66 = .0067088055904883

Sharp RD estimates using local polynomial regression.

Cutoff c = .67	Left of c	Right of c	Number of obs =	34128
			BW type =	Manual
Number of obs	16212	17916	Kernel =	Uniform
Eff. Number of obs	16212	17916	VCE method =	NN
Order est. (p)	1	1		
Order bias (q)	2	2		
BW est. (h)	0.500	0.500		
BW bias (b)	0.500	0.500		
rho (h/b)	1.000	1.000		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	.00166	.01732	0.0957	0.924	-.032289 .035605
Robust	-	-	-0.4357	0.663	-.063507 .040408

tau\_67 = .001657936994885

Sharp RD estimates using local polynomial regression.

Cutoff c = .68	Left of c	Right of c	Number of obs =	34128
			BW type =	Manual
Number of obs	16905	17223	Kernel =	Uniform
Eff. Number of obs	16905	17223	VCE method =	NN
Order est. (p)	1	1		
Order bias (q)	2	2		
BW est. (h)	0.500	0.500		
BW bias (b)	0.500	0.500		
rho (h/b)	1.000	1.000		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	.01096	.01746	0.6278	0.530	-.023262 .045187
Robust	-	-	0.5266	0.598	-.038432 .066672

tau\_68 = .0109626271500858

Sharp RD estimates using local polynomial regression.

Cutoff c = .6899999999999999	Left of c	Right of c	Number of obs =	34128
			BW type =	Manual
Number of obs	17638	16490	Kernel =	Uniform
Eff. Number of obs	17638	16490	VCE method =	NN
Order est. (p)	1	1		
Order bias (q)	2	2		
BW est. (h)	0.500	0.500		
BW bias (b)	0.500	0.500		
rho (h/b)	1.000	1.000		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	.00599	.01755	0.3411	0.733	-.028415 .040391
Robust	-	-	0.2943	0.769	-.044674 .060463

tau\_69 = .0059880332223443

Sharp RD estimates using local polynomial regression.

Cutoff c = .7	Left of c	Right of c	Number of obs =	34128
			BW type =	Manual
Number of obs	18315	15813	Kernel =	Uniform
Eff. Number of obs	18315	15813	VCE method =	NN
Order est. (p)	1	1		
Order bias (q)	2	2		
BW est. (h)	0.500	0.500		
BW bias (b)	0.500	0.500		
rho (h/b)	1.000	1.000		

Outcome: y. Running variable: x.

Method	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Conventional	.00137	.01773	0.0773	0.938	-.033371 .036112
Robust	-	-	0.0731	0.942	-.051141 .055102

tau\_70 = .0013703275602168

270

271

272 clear

273 set obs 41

Number of observations (\_N) was 0, now 41.

274 gen c = .

(41 missing values generated)

275 gen tau = .

(41 missing values generated)

276 gen rb\_lb = .

(41 missing values generated)

277 gen rb\_ub = .

(41 missing values generated)

278

```
279 forval i = 1/'= N' {
    2.     local c = 29+'i' // c= 30(1)70
    3.     replace c = `c' in `i'
    4.     replace tau = `tau_`c'' in `i'
    5.     replace rb_lb = `bc_lb_`c'' in `i'
    6.     replace rb_ub = `bc_ub_`c'' in `i'
    7. }
```

(1 real change made)

(1 real change made)

(1 real change made)

(1 real change made)

(1 real change made)

(1 real change made)

(1 real change made)

(1 real change made)

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(1 real change made)

(1 real change made)

(1 real change made)

(1 real change made)

(1 real change made)

[illegible]

[illegible]

```

280
281 label var rb_ub "Bias-corrected upper bound (95% CI)"
282 label var tau    "Conventional local-polynomial RD estimate"
283 label var rb_lb "Bias-corrected lower bound (95% CI)"
284 label var c "Cutoff"

285 twoway (line rb_ub c, lpattern(dash) lcolor(grey)) ///
>         (line tau c, lcolor(black)) ///
>         (line rb_lb c, lpattern(dash) lcolor(grey)), ///
>         legend(position(6)) ytitle("Estimate") ///
>         yline(0, lcolor(red) lpattern(solid))
(note: named style grey not found in class color, default attributes used)
(note: named style grey not found in class color, default attributes used)
(note: named style grey not found in class color, default attributes used)
(note: named style grey not found in class color, default attributes used)

286 graph export "$do_loc/graphs/q2e.png", ///
>         width(1200) height(900) ///
>         replace
file C:/Users/yfkas/Documents/GitHub/ARE213_Fall2023/PSet 4/Stata/graphs/q2e.png
    saved as PNG format

287
288
289
290
    end of do-file

291
292
293
294 log close
    name: <unnamed>
    log: C:/Users/yfkas/Documents/GitHub/ARE213_Fall2023/PSet 4/Stata/pset4_logfil
> e.smcl
    log type: smcl
    closed on: 4 Dec 2023, 11:53:26

```

---