

# Econometrics 2 PS4: Regression Discontinuity

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## Problem 1: (Sharp) Regression Discontinuity

“Take any dataset with covariate  $X$  and outcome  $Y$  that are related in some way. For instance, you can use the data on birth weight and smoking from here: <http://www.stata.com/texts/eacsap/>, or any other relevant dataset. Alternatively, feel free to simulate your own data. In any case, please provide an explanation of your dataset. Construct a placebo treatment by applying a rule such that  $D_i = 1$  when  $X_i \geq x_0$  for some  $x_0$ . That is, modify the outcome variable  $Y_i$  for those units with  $X_i \geq x_0$  by adding a constant treatment effect, for example, add one standard deviation of the outcome plus some noise (with mean zero). Include an explanation of what you ended up doing”

```
# Change to test how RMarkdown works with VSCode
# Change 2--- synced?

# Data Simulation
library(truncnorm)

## Warning: package 'truncnorm' was built under R version 4.3.3

n = 10000

data = data.table(
  gpa = rtruncnorm(n, a = 0, b = 4, mean = 3, sd = 1),
  fam_income = runif(n, min = 20000, max = 150000)
)
gpa_cutoff = 2

# if gpa is below cutoff, school provides tutoring that increases score
data[, treat := ifelse(gpa < gpa_cutoff, 1, 0)]
data[, outcome_score := gpa*300 + fam_income/1000 + treat*200 + rnorm(n, mean = 0, sd = 50)]

datasummary_skim(data, out = "markdown", title = "Summary Statistics", histogram = F)
```

Table 1: Summary Statistics

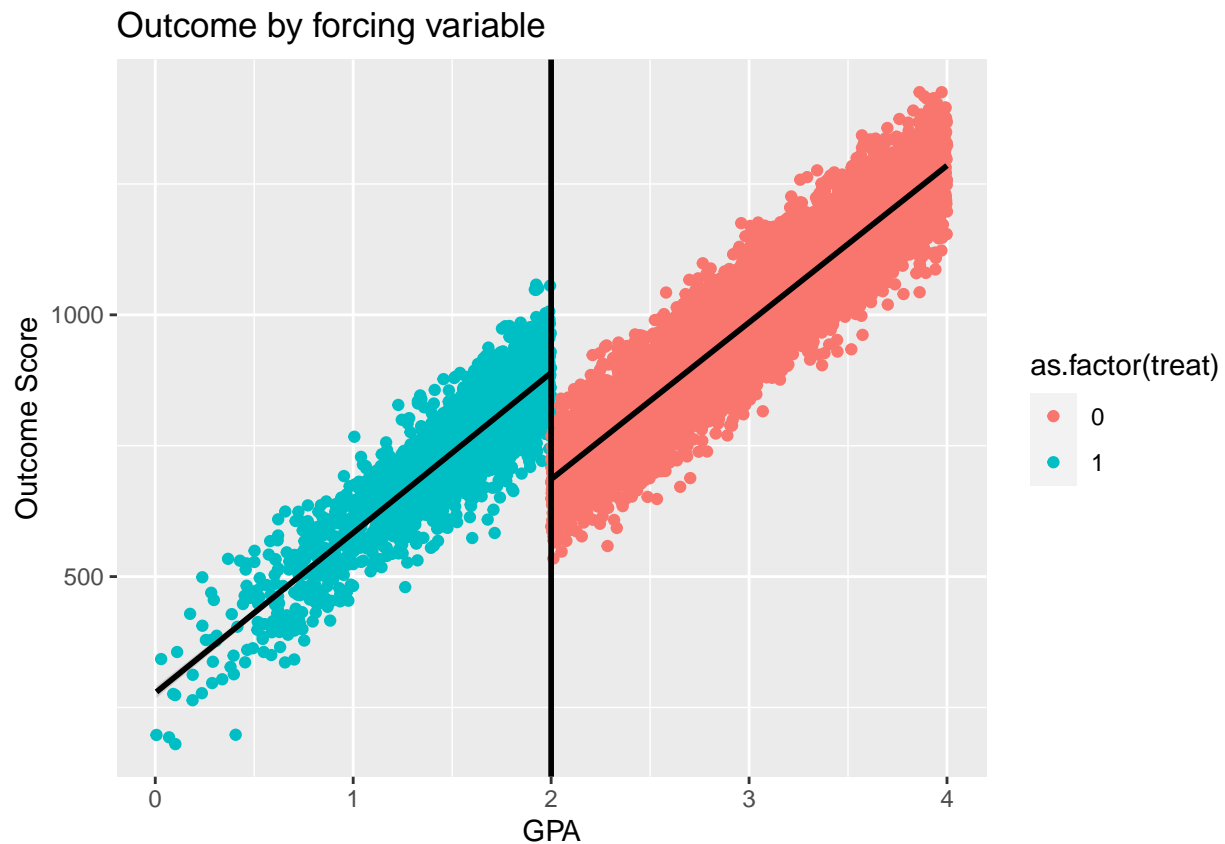
	Unique	Missing	Pct.	Mean	SD	Min	Median	Max
gpa	10000	0		2.7	0.8	0.0	2.8	4.0
fam_income	10000	0		85245.7	37431.9	20018.7	85403.9	149967.3
treat	2	0		0.2	0.4	0.0	0.0	1.0

	Unique	Missing	Pct.	Mean	SD	Min	Median	Max
outcome_score	10000		0	940.1	194.7	180.1	931.7	1425.4

### 1.1) Plot the outcome by forcing variable (the standard graph showing the discontinuity)

```
# Generate input data for output plot
```

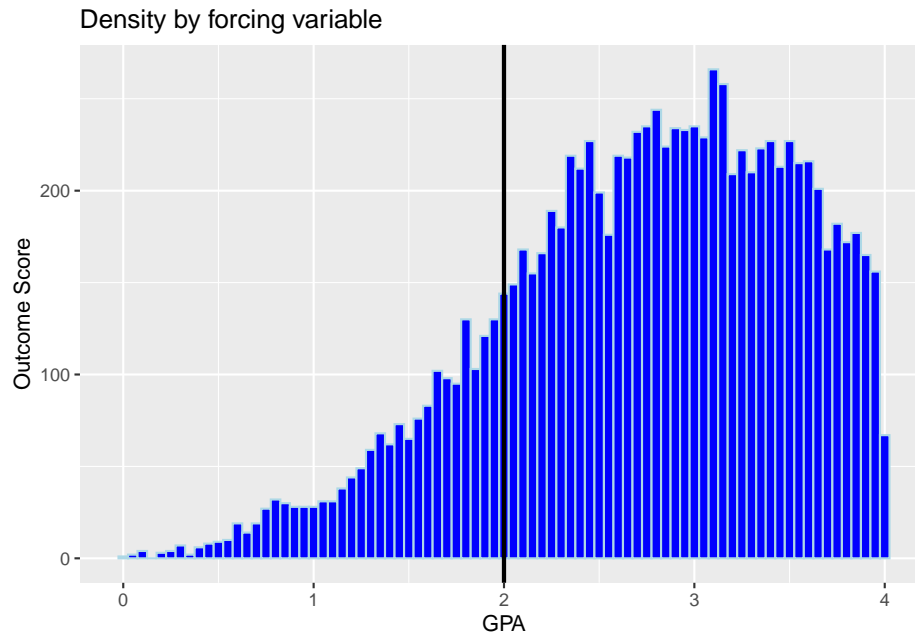
```
plot1 <- ggplot(data, aes(x = gpa, y = outcome_score, color = as.factor(treat), group = as.factor(treat))) +
  geom_point() +
  #geom_smooth(aes(fill = as.factor(treat))) +
  geom_smooth(method = "lm", color = "black", formula = y~x) +
  labs(x = "GPA", y = "Outcome Score") + ggtitle("Outcome by forcing variable") +
  geom_vline(xintercept = gpa_cutoff, linewidth = 1)
plot1
```



### 1.2) Plot the density of the forcing variable

You can also embed plots, for example:

```
plot2 <- ggplot(data, aes(gpa)) +
  geom_histogram(fill = "blue", color = "lightblue", binwidth = 0.05) +
  labs(x = "GPA", y = "Outcome Score") + ggtitle("Density by forcing variable") +
  geom_vline(xintercept = gpa_cutoff, linewidth = 1)
plot2
```



### 1.3) Estimate the effect using a local linear regression

```
reg3 = lm(outcome_score ~ treat + I(gpa-gpa_cutoff) + I(treat*(gpa-gpa_cutoff)) , data = data)
stargazer(reg3, type = "latex", title = "Local Linear Regression", table.placement = "H",
  header=FALSE, no.space = T, omit.stat = "f")
```

Table 2: Local Linear Regression

	<i>Dependent variable:</i>
	outcome_score
treat	203.537*** (2.780)
I(gpa - gpa_cutoff)	299.659*** (1.281)
I(treat *(gpa - gpa_cutoff))	6.113 (3.812)
Constant	685.806*** (1.459)
Observations	10,000
R <sup>2</sup>	0.896
Adjusted R <sup>2</sup>	0.896
Residual Std. Error	62.791 (df = 9996)
<i>Note:</i>	*p<0.1; **p<0.05; ***p<0.01

#### 1.4) Estimate the effect using a local polynomial (of order 2 and 3) regression

```
reg41 = lm(outcome_score ~ treat + I(gpa-gpa_cutoff) + I(treat*(gpa-gpa_cutoff))+
           I((gpa-gpa_cutoff)^2) + I(treat*(gpa-gpa_cutoff)^2), data = data)
#stargazer(reg41, type = "latex", title = "Estimate Effect Using Local Polynomial of Order 2", table.pl
```

```
reg42 = lm(outcome_score ~ treat + I(gpa-gpa_cutoff) + I(treat*(gpa-gpa_cutoff))+
           I((gpa-gpa_cutoff)^2) + I(treat*(gpa-gpa_cutoff)^2)+
           I((gpa-gpa_cutoff)^3) + I(treat*(gpa-gpa_cutoff)^3), data = data)
stargazer(reg3, reg41, reg42, type = "latex", title = "Estimate Effect Using Local Polynomial of Order 3",
           table.placement = "H", header=FALSE, no.space = T, omit.stat = "f")
```

Table 3: Estimate Effect Using Local Polynomial of Order 2 and 3

	<i>Dependent variable:</i>		
	outcome_score		
	(1)	(2)	(3)
treat	203.537*** (2.780)	203.802*** (3.941)	206.974*** (5.085)
I(gpa - gpa_cutoff)	299.659*** (1.281)	298.048*** (5.077)	290.306*** (12.789)
I(treat *(gpa - gpa_cutoff))	6.113 (3.812)	11.720 (12.205)	58.001** (26.691)
I((gpa - gpa_cutoff)^2)		0.805 (2.457)	10.347 (14.674)
I(treat *(gpa - gpa_cutoff)^2)		2.012 (7.800)	55.470 (37.519)
I((gpa - gpa_cutoff)^3)			-3.181 (4.823)
I(treat *(gpa - gpa_cutoff)^3)			29.224** (14.753)
Constant	685.806*** (1.459)	686.375*** (2.268)	687.755*** (3.086)
Observations	10,000	10,000	10,000
R <sup>2</sup>	0.896	0.896	0.896
Adjusted R <sup>2</sup>	0.896	0.896	0.896
Residual Std. Error	62.791 (df = 9996)	62.797 (df = 9994)	62.791 (df = 9992)

*Note:*

\*p&lt;0.1; \*\*p&lt;0.05; \*\*\*p&lt;0.01