# PS2

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I am trying out using markstat package for better reproducability, All code and asnwers will be in the same document, I will include the .stmd file and the boot file I use for running it (you will need to change the directories)

Also, all files can be found on my github at:

https://github.com/MattarKlein/Econometrics-B.git

## **Import Data**

```
. clear all
. use "star_problem_set.dta"
```

#### Question 1

(a) Run Regression Basic Regression

```
. quietly gen sc = .
. quietly gen ra = .
. foreach v in k 1 2 3{
  2. quietly gen tss`v' = treadss`v' + tmathss`v'
  3. quietly replace sc = cltype`v' == 1
  4. quietly replace ra = cltype`v' == 3
  5. quietly regress tss`v' sc ra, robust
  6. quietly eststo a_reg_`v'
  7 }
```

The regression model is

$$Y_i = \beta_0 + \beta_1 s c_i + \beta_2 r a_i + u_i$$

Where  $Y_i$  is the outcome variable done for 4 different grades (Kindergarden though 3rd),  $sc_i$  is a binary variable for being randomized into a small class, and  $ra_i$  is a binary variable for being randomized into a class with an assistant.

Therefore, the constant term is the average total score of the control group (i.e. no assistant assigned and reduction in class size), and the coefficients of  $sc_i$  and  $ra_i$  are the change in total test scores for those groups respectively.

As can be seen in the output table, being randomized into a small class has an effect across all classes and is largest in 1st grade.

- . label variable tssk "K"
- . label variable tss1 "1st"
- . label variable tss2 "2nd"
- . label variable tss3 "3rd"
- . label variable sc "Small Class"
- . label variable ra "Regular + Ass"

. esttab a\_reg\_k a\_reg\_1 a\_reg\_2 a\_reg\_3, label se ar2

	4.3	4-2	(-)	
	(1)	(2)	(3)	(4)
	K	1st	2nd	3rd
Small Class	13.90***	29.78***	19.39***	15.59***
	(2.454)	(2.831)	(2.712)	(2.396)
Regular + Ass	0.314	11.96***	3.479	-0.291
	(2.271)	(2.652)	(2.545)	(2.273)
Constant	918.0***	1039.4***	1157.8***	1228.5***
	(1.633)	(1.785)	(1.815)	(1.680)
Observations	5786	6379	6049	5967
Adjusted R-squared	0.007	0.017	0.009	0.010

Standard errors in parentheses \* p<0.05, \*\* p<0.01, \*\*\* p<0.001

## (b) Including controls

```
. quietly gen totexp = .
. quietly gen ses = .
. foreach v in k 1 2 3{
    2. quietly replace tss`v´ = treadss`v´ + tmathss`v´
    3. quietly replace sc = cltype`v´ == 1
    4. quietly replace ra = cltype`v´ == 3
    5. quietly replace totexp = totexp`v´
    6. quietly replace ses = ses`v´
    7. quietly regress tss`v´ sc ra totexp ssex srace ses, robust
    8. quietly eststo b_reg_`v´
    9. }
```

The regression model is the same as before with added demographic controls for teacher experience and the gender race and socio-economic status of the student.

$$Y_i = \beta_0 + \beta_1 sc_i + \beta_2 ra_i + \beta_3 totexp_i + \beta_4 ssex_i + \beta_5 srace_i + \beta_6 ses_i + u_i$$

Comparing the size of the coefficients of small class to the coefficients of different demographics shows that the effects aren't very big. Take socio-economic status for instance this is measured by whether the sutdent is entitled to a free lunch or not, the coefficients on this are systematically larger and more than double the effect of small class.

The standard errors on  $sc_i$  went down for all coefficients, this is b/c by adding more controls the model is getting more exact. Also the adjusted R-Squared has gone up by quite a lot.

```
. label variable totexp "Exp Teacher"
. label variable ssex "Female"
. label variable srace "Black"
. label variable ses "Lower SE status"
.
. esttab b_reg_k b_reg_1 b_reg_2 b_reg_3, label se ar2
```

	(1)	(2)	(3)	(4)
	K	1st	2nd	3rd
Small Class	14.17***	24.80***	16.78***	11.99***
	(2.354)	(2.625)	(2.544)	(2.277)
Regular + Ass	0.776	8.250***	4.811*	-0.524
	(2.156)	(2.459)	(2.385)	(2.146)
Exp Teacher	1.117***	0.235*	0.124	0.0784
	(0.169)	(0.112)	(0.113)	(0.105)
Female	13.98***	11.91***	10.47***	7.654***
	(1.850)	(2.090)	(2.020)	(1.783)
Black	-10.36***	-27.97***	-31.55***	-20.06***
	(2.063)	(2.789)	(3.145)	(2.616)
Lower SE status	34.95***	56.98***	46.00***	42.02***
	(2.013)	(2.431)	(2.511)	(2.151)
Constant	847.5***	974.5***	1115.0***	1181.2***
	(6.044)	(7.459)	(8.124)	(6.940)

Observations	5748	6227	5707	5733
Adjusted R-squared	0.101	0.184	0.171	0.144

Standard errors in parentheses \* p<0.05, \*\* p<0.01, \*\*\* p<0.001

# (c) Test for randomization

```
. foreach v in k 1 2 3{
   2. quietly replace sc = cltype`v´ == 1
   3. quietly replace ra = cltype`v´ == 3
   4. quietly replace ses = ses`v´
   5. quietly regress totexp`v´ sc ra, robust
   6. quietly eststo c_reg_`v´_exp
   7. quietly regress ssex sc ra, robust
   8. quietly eststo c_reg_`v´_ssex
   9. quietly regress srace sc ra, robust
   10. quietly eststo c_reg_`v´_srace
   11. quietly regress ses sc ra, robust
   12. quietly eststo c_reg_`v´_ses
   13. }
```

The following tables check the distribution of observables among different test groups. It seems that while some of the randomizations (like gender) did actually randomize across demographics there were some demographics which weren't split randomly (specifically race and teacher experience).

```
. label variable totexpk "Exp K". label variable totexp1 "Exp 1st". label variable totexp2 "Exp 2nd"
```

. label variable totexp3 "Exp 3rd"

. esttab c\_reg\_k\_exp c\_reg\_1\_exp c\_reg\_2\_exp c\_reg\_3\_exp, label se

	(1)	(2)	(3)	(4)
	Exp K	Exp 1st	Exp 2nd	Exp 3rd
Small Class	-0.149 (0.181)	1.899***	0.311 (0.268)	-0.172 (0.265)
Regular + Ass	0.670*** (0.175)	2.353*** (0.257)	1.102*** (0.250)	1.335*** (0.254)
Constant	9.068*** (0.122)	10.30*** (0.171)	12.66*** (0.182)	13.49*** (0.189)
Observations	6304	6810	6738	6751

Standard errors in parentheses \* p<0.05, \*\* p<0.01, \*\*\* p<0.001

. esttab c\_reg\_k\_ssex c\_reg\_1\_ssex c\_reg\_2\_ssex c\_reg\_3\_ssex, label se

	(1) Female	(2) Female	(3) Female	(4) Female
	Temale		. emare	remare
Small Class	0.0217	0.0176	0.0250*	0.0232
	(0.0128)	(0.0128)	(0.0126)	(0.0123)
Regular + Ass	0.0187	0.00230	0.0108	0.00287
	(0.0121)	(0.0119)	(0.0116)	(0.0116)
Constant	1.464***	1.468***	1.465***	1.466***
	(0.00578)	(0.00582)	(0.00593)	(0.00602)
Observations	11578	11578	11578	11578

Standard errors in parentheses

- \* p<0.05, \*\* p<0.01, \*\*\* p<0.001
- . esttab c\_reg\_k\_srace c\_reg\_1\_srace c\_reg\_2\_srace c\_reg\_3\_srace, label se

	(1)	(2)	(3)	(4)
	Black	Black	Black	Black
Small Class	-0.0803***	-0.0895***	-0.0665***	-0.0902***
	(0.0139)	(0.0139)	(0.0135)	(0.0130)
Regular + Ass	-0.0633***	-0.116***	-0.0386**	-0.0577***
	(0.0128)	(0.0128)	(0.0128)	(0.0123)
Constant	1.415***	1.427***	1.409***	1.419***
	(0.00658)	(0.00656)	(0.00671)	(0.00702)
Observations	11453	11453	11453	11453

Standard errors in parentheses

- \* p<0.05, \*\* p<0.01, \*\*\* p<0.001
- . esttab c\_reg\_k\_ses c\_reg\_1\_ses c\_reg\_2\_ses c\_reg\_3\_ses, label se

	(1)	(2)	(3)	(4)
	Lower SE s <sub>~</sub> s	Lower SE s <sub>z</sub> s	Lower SE s <sub>z</sub> s	Lower SE s <sub>z</sub> s
Small Class	0.00622	0.0418**	0.0451**	0.0334*
	(0.0157)	(0.0152)	(0.0155)	(0.0157)
Regular + Ass	-0.0256	0.00609	0.0130	-0.000146
	(0.0151)	(0.0145)	(0.0148)	(0.0151)
Constant	1.523***	1.470***	1.468***	1.484***
	(0.0107)	(0.00992)	(0.0106)	(0.0113)
Observations	6302	6651	6496	6520

Standard errors in parentheses \* p<0.05, \*\* p<0.01, \*\*\* p<0.001

#### (d) Treatment Heterogeneity

```
. quietly gen sc_totexp = .
. quietly gen sc_ssex = .
. quietly gen sc_srace = .
. quietly gen sc_ses = .
. foreach v in k 1 2 3{
  2. quietly replace tss`v´ = treadss`v´ + tmathss`v´
 3. quietly replace sc = cltype`v´ == 1
4. quietly replace ra = cltype`v´ == 3
  5. quietly replace totexp = totexp`v´
  6. quietly replace ses = ses`v´
 7. quietly replace sc_totexp = sc*totexp
 8. quietly replace sc_ssex = sc*ssex
 9. quietly replace sc_srace = sc*srace
 10. quietly replace sc_ses = sc*ses
11. quietly regress tss`v' sc ra totexp sc_totexp ssex sc_ssex srace sc_srace ses sc_ses, robus
> t
 12. quietly eststo d_reg_`v´
```

In order to account for heterogeniety across students with different co-variates I have added interaction terms for the 4 observed demographics with small class. It does not seem like there is much heterogeniety in the sample. However, as can be seen in the previous question, I am not very convinced about the randomization anymore.

```
. label variable sc_totexp "SC X Exp"
. label variable sc_ssex "SC X Female"
. label variable sc_srace "SC X Black"
. label variable sc_ses "SC X Lower SE status"
.
. esttab d_reg_k d_reg_1 d_reg_2 d_reg_3, label se ar2
```

	(1)	(2)	(3)	(4)
	K	1st	2nd	3rd
Small Class	39.53**	-5.587	5.243	-3.557
	(12.93)	(15.69)	(17.58)	(14.00)
Regular + Ass	0.423	8.341***	4.825*	-0.233
	(2.154)	(2.468)	(2.390)	(2.145)
Exp Teacher	1.607***	0.0948	0.0852	-0.181
	(0.198)	(0.132)	(0.134)	(0.123)
SC X Exp	-1.667***	0.462	0.111	0.791***
	(0.366)	(0.250)	(0.249)	(0.231)
Female	16.82***	11.37***	10.30***	8.242***
	(2.159)	(2.427)	(2.377)	(2.131)
SC X Female	-9.567*	1.785	0.572	-1.774
	(4.143)	(4.745)	(4.496)	(3.870)
Black	-10.89***	-30.63***	-33.50***	-21.49***
	(2.360)	(3.448)	(3.767)	(3.364)
SC X Black	1.066	8.155	6.771	4.937
	(4.610)	(5.910)	(6.901)	(5.258)
Lower SE status	34.81***	54.56***	46.00***	42.25***

	(2.373)	(2.886)	(2.978)	(2.678)
SC X Lower SE status	1.614	7.439	0.0482	0.589
	(4.419)	(5.383)	(5.553)	(4.544)
Constant	839.7***	984.1***	1118.4***	1185.5***
	(7.094)	(9.084)	(9.669)	(8.662)
Observations	5748	6227	5707	5733
Adjusted R-squared	0.105	0.184	0.171	0.146

Standard errors in parentheses \* p<0.05, \*\* p<0.01, \*\*\* p<0.001

(e) As I stated before I am not convinced about the randomization, it seems that there was some confounding factor which affected the randomization (or maybe just chance). My guess is that after the randomization there was some movement between treatments. This kind of movement will only be made worse in a real world environment if only some classes actually reduce size. Additionally if all classes reduce size then I am not sure how good the estimates actually help.

However, the experiment might be usefull for policy if researches try to go back and understand how such changes were done, and how they affected different demographic groups after the fact.

# Question 2

$$Y_{0i} \perp D_i \rightarrow E[Y_{0i}|D_i = 0] = E[Y_{0i}|D_i = 1]$$

Adding this to the definition of ATT (1), we get:

$$ATT = E[Y_{1i}|D_i = 1] - E[Y_{0i}|D_i = 1] \stackrel{(1)}{=} E[Y_{1i}|D_i = 1] - E[Y_{0i}|D_i = 0]$$

## Question 3

(a)