

# Assignment 4

Foodies with hoodies

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```
# load packages
if(!require(pacman)){install.packages("pacman")}

p_load(devtools,tidyverse,dplyr,ggplot2,latex2exp,
       sampleSelection, quantreg, plm, nlme, knitr,car, ivreg,stargazer)
```

## 1 Question 1

Judge Sentences	Jones		Smith	
	Prison	Other	Prison	Other
Cases	70%	30%	40%	60%
Future arrests	40%	60%	20%	50%

### 1.1 (i)

We can treat the following problem as follows:  $Y_i$  is the outcome of whether an individual is arrested later. The instrument variable  $Z_i$  is which judge they are assigned to in the first case and  $D_i$  is the treatment whether the individual is sentenced to prison or not in the first case. Then the Wald estimator can be calculated by:

$$\delta_{Wald} = \frac{E[Y_i|Z_i = 1] - E[Y_i|Z_i = 0]}{Pr(D_i = 1|Z_i = 1) - Pr(D_i = 1|Z_i = 0)} \quad (1)$$

### 1.2 (ii)

### 1.3 (iii)

## 2 Question 2

### 2.1 (i)

From what is given, we have  $MDE = 0.1$ , the power  $p = 0.7$ , the proportion of students in control group is  $p = 0.5$ . The variance of the binomial variable is  $\sigma^2 = p(1 - p) = 0.25$ . To get the number of students the teacher should include in the experiment, we use the following formula:

$$\begin{aligned}
 n &= \left( \frac{t_{1-\alpha/2} - t_{1-q}}{MDE} \right)^2 \frac{\sigma^2}{p(1-p)} \\
 &= \left( \frac{1.960 + 0.524}{0.1} \right)^2 \frac{0.25}{0.5(1-0.5)} \\
 &\approx 617
 \end{aligned} \tag{2}$$

Thus, the teacher should include at least 617 students in the experiment.

## 2.2 (ii)

This will change the proportion of students in treatment to  $p = 0.5 \times 20\% = 0.1$ , using the formula in Equation (2), the number of students required to participate in the experiment is:

$$\begin{aligned}
 n &= \left( \frac{1.960 + 0.524}{0.1} \right)^2 \frac{0.25}{0.1(1-0.1)} \\
 &\approx 1713
 \end{aligned} \tag{3}$$

Thus, the number of students required to participate in the experiment increases by  $6856 - 2468 = 4388$  students.

## 3 Question 3

### 3.1 (i)

```

# Load data
dfData = read.csv("AngristEvans80.csv")
attach(dfData)

# Fraction of girls among the first born child
count_girl1 = table(dfData$SEXK)
fraction_girl1 = count_girl1[[2]] / (count_girl1[[1]] + count_girl1[[2]])

# Fraction of girls among the second born child

```

```
count_girl2 = table(dfData$SEX2ND)
fraction_girl2= count_girl2[[2]]/(count_girl2[[1]]+count_girl2[[2]])

cat("Fraction of girls among the first born child is:
↪  ",fraction_girl1,"\n", "Fraction of girls among the second born
↪  child is: ",fraction_girl2)
```

Fraction of girls among the first born child is: 0.4876463  
 Fraction of girls among the second born child is: 0.4884266

```
#Regress gender of second child on gender of first child
lm_second_first = lm(SEX2ND~SEXK, data = dfData)
summary(lm_second_first)
```

Call:

```
lm(formula = SEX2ND ~ SEXK, data = dfData)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.4908	-0.4862	-0.4862	0.5092	0.5138

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.4861744	0.0008672	560.626	<2e-16 ***
SEXK	0.0046185	0.0012418	3.719	2e-04 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.4999 on 648470 degrees of freedom

Multiple R-squared: 2.133e-05, Adjusted R-squared: 1.979e-05

F-statistic: 13.83 on 1 and 648470 DF, p-value: 2e-04

### 3.2 (ii)

```
# First stage regression
lm_first_stage = lm(CHILD3 ~ SAMESEX, data= dfData)
summary(lm_first_stage)
```

```
Call:
lm(formula = CHILD3 ~ SAMESEX, data = dfData)

Residuals:
    Min       1Q   Median       3Q      Max
-0.4093 -0.4093 -0.3552  0.5907  0.6448

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 0.3552366   0.0008544  415.79  <2e-16 ***
SAMESEX      0.0540534   0.0012051   44.85  <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.4852 on 648470 degrees of freedom
Multiple R-squared:  0.003093, Adjusted R-squared:  0.003091
F-statistic: 2012 on 1 and 648470 DF, p-value: < 2.2e-16
```

Is the instrumental variable sufficiently strong? => yes

```
# Regress number of children on whether the first two children have the
↳ same gender
lm_total = lm(KIDCOUNT ~ SAMESEX, data= dfData)
summary(lm_total)
```

```
Call:
lm(formula = KIDCOUNT ~ SAMESEX, data = dfData)

Residuals:
    Min       1Q   Median       3Q      Max
-0.5752 -0.5752 -0.5040  0.4248  9.4960

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 2.504033   0.001458 1716.93  <2e-16 ***
SAMESEX      0.071200   0.002057  34.61  <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.8283 on 648470 degrees of freedom
Multiple R-squared:  0.001844, Adjusted R-squared:  0.001842
F-statistic: 1198 on 1 and 648470 DF, p-value: < 2.2e-16
```

### 3.3 (iii)

In this study, the treatment group includes those who have a third child and the control group includes those who have two children or less. The variables that affect decision for mothers to be assigned into treatment or control group is  $Z = SAMESEX$ , indicating whether the first two child are of the same sex or not.

The always takers are those who have a third child regardless of whether the first two children is of the same sex or not.

```
df_always = dfData[dfData$CHILD3 == 1 & dfData$SAMESEX == 0,]  
cat("The share of always takers is: ", nrow(df_always)/nrow(dfData))
```

The share of always takers is: 0.1766969

The compliers are those who only have a third child if the first two kids are of the same sex.

```
df_compliers1 = dfData[dfData$CHILD3 == 1 & dfData$SAMESEX == 1,]  
df_compliers0 = dfData[dfData$CHILD3 == 0 & dfData$SAMESEX == 0,]  
cat("The share of compliers is: ",  
    ↪ (nrow(df_compliers1)+nrow(df_compliers0))/nrow(dfData))
```

The share of compliers is: 0.5264159

The never takers are those who will never have the third child regardless of whether the first two children are of the same sex or not.

```
df_never = dfData[dfData$CHILD3 == 0 & dfData$SAMESEX == 1,]  
cat("The share of never takers is: ", nrow(df_never)/nrow(dfData))
```

The share of never takers is: 0.2968871

Lastly, the defiers are those who will have a third child if the first two kids are of different sexes and will not have a third child if the first two kids are of the same sex. We cannot observe this as they are divided among the always taker and never taker's group.

### 3.4 (iv)

```
iv_reg_hour <- ivreg(HOURSM ~ CHILD3 | SAMESEX, data = dfData)
iv_reg_income <- ivreg(INCOME1M ~ CHILD3 | SAMESEX, data = dfData)

stargazer(iv_reg_hour, iv_reg_income,
           type="text", report="vc*stp",
           keep.stat=c("n","adj.rsq"),
           title = "...") #remember to change title
```

...

Dependent variable:		
	HOURSM	INCOME1M
	(1)	(2)
CHILD3	-3.585*** (0.864) t = -4.150 p = 0.00004	-786.830*** (244.939) t = -3.212 p = 0.002
Constant	20.304*** (0.331) t = 61.311 p = 0.000	3,825.464*** (93.899) t = 40.740 p = 0.000
Observations	648,472	648,472
Adjusted R2	0.007	0.008
Note: *p<0.1; **p<0.05; ***p<0.01		

### 3.5 (v)

```
# Subgroup 1: Always taker
hour1=mean(df_always$HOURSM)
income1=mean(df_always$INCOME1M)
```

```
cat("The mean working hour of always takers is: ", hour1, ", the mean
  ↪ income of always takers is: ",income1)
```

The mean working hour of always takers is: 17.04711 , the mean income of always takers is: 30.00000

```
# Subgroup 2: never takers
hour2=mean(df_never$HOURLSM)
income2=mean(df_never$INCOME1M)
cat("The mean working hour of never takers is: ", hour2, ", the mean
  ↪ income of never takers is: ",income2)
```

The mean working hour of never takers is: 20.20379 , the mean income of never takers is: 30.00000

```
# Subgroup 3: complier 1
hour3=mean(df_compliers1$HOURLSM)
income3=mean(df_compliers1$INCOME1M)
cat("The mean working hour of complier in treatment group is: ", hour3,
  ↪ ", the mean income of this group is: ",income3)
```

The mean working hour of complier in treatment group is: 16.8629 , the mean income of this group is: 30.00000

```
# Subgroup 4: complier 0
hour4=mean(df_compliers0$HOURLSM)
income4=mean(df_compliers0$INCOME1M)
cat("The mean working hour of complier in control group is: ", hour4, ",
  ↪ the mean income of this group is: ",income4)
```

The mean working hour of complier in control group is: 20.12279 , the mean income of this group is: 30.00000

To-dos: USE these means to say something about the preference of having a third child

### 3.6 (vi)

### 3.7 (vii)

First, we stratify the sample by gender of the first child:



```
df_first_girl = dfData[dfData$SEXK == 1,]
df_first_boy = dfData[dfData$SEXK == 0,]
```

(But they ask to use the first stage result?) I try to to it manually below:

```
# First stage regression
lm_1st_girl = lm(CHILD3 ~ SAMESEX, data=df_first_girl)
lm_1st_boy = lm(CHILD3 ~ SAMESEX, data=df_first_boy)
stargazer(lm_1st_girl, lm_1st_boy,
           type="text", report="vc*stp",
           keep.stat=c("n","adj.rsq"),
           title = "...") #remember to change title
```

```
...
=====
Dependent variable:
-----
CHILD3
(1)          (2)
-----
SAMESEX      0.063***      0.046***
              (0.002)      (0.002)
              t = 36.381    t = 27.351
              p = 0.000      p = 0.000

Constant     0.355***      0.356***
              (0.001)      (0.001)
              t = 293.159    t = 294.887
              p = 0.000      p = 0.000

-----
Observations  316,225      332,247
Adjusted R2   0.004        0.002
=====
Note:          *p<0.1; **p<0.05; ***p<0.01
```

If the first child is a girl and first two children are of the same sex, one is more likely to have a third child.

Then, we perform instrumental variable regressions:

```

iv_reg_girl_hour <- ivreg(HOURSM ~ CHILD3 | SAMESEX, data =
  ↪ df_first_girl)
iv_reg_girl_income <- ivreg(INCOME1M ~ CHILD3 | SAMESEX, data =
  ↪ df_first_girl)
iv_reg_boy_hour <- ivreg(HOURSM ~ CHILD3 | SAMESEX, data = df_first_boy)
iv_reg_boy_income <- ivreg(INCOME1M ~ CHILD3 | SAMESEX, data =
  ↪ df_first_boy)
stargazer(iv_reg_girl_hour,
  ↪ iv_reg_girl_income, iv_reg_boy_hour, iv_reg_boy_income,
  type="text", report="vc*stp",
  keep.stat=c("n", "adj.rsq"),
  title = "...") #remember to change title

```

...

Dependent variable:				
	HOURSM	INCOME1M	HOURSM	INCOME1M
	(1)	(2)	(3)	(4)
CHILD3	-1.104 (1.064) t = -1.037 p = 0.300	-343.922 (303.245) t = -1.134 p = 0.257	-6.695*** (1.425) t = -4.698 p = 0.00001	-1,320.535*** (400.868) t = -3.294 p = 0.001
Constant	19.409*** (0.412) t = 47.115 p = 0.000	3,676.891*** (117.357) t = 31.331 p = 0.000	21.424*** (0.541) t = 39.567 p = 0.000	4,006.674*** (152.305) t = 26.307 p = 0.000
Observations	316,225	316,225	332,247	332,247
Adjusted R2	0.004	0.005	-0.001	0.007
Note: *p<0.1; **p<0.05; ***p<0.01				

Here we see that if the first child is a girl, having a third child does not significantly influence the hour and income.