

# In-Class Lab 11

ECON 4223

October 17, 2023

The purpose of this in-class lab is to use R to practice with instrumental variables estimation. The lab should be completed in your group. To get credit, upload your .R script to the appropriate place on Canvas.

## For starters

You may need to install the packages `AER`, `flextable` and `modelsummary`. (`AER` may have already been installed when you previously installed `car` and `zoo`.)

Open up a new R script (named `ICL11_XYZ.R`, where `XYZ` are your initials) and add the usual “preamble” to the top:

```
# Add names of group members HERE
library(tidyverse)
library(wooldridge)
library(broom)
library(AER)
library(magrittr)
library(modelsummary)
```

## Load the data

We’re going to use data on fertility of Botswanian women.

```
df <- as_tibble(fertil2)
```

## Summary statistics

Let’s look at summary statistics of our data by using the `modelsummary` package. We can export this to a word document format if we’d like:

```
df %>% datasummaryskim(histogram=F,output="myfile.docx")
```

```
## [1] "myfile.docx"
```

1. What do you think is going on when you see varying numbers of observations across the different variables?

## Determinants of fertility

Suppose we want to see if education causes lower fertility (as can be seen when comparing more- and less-educated countries):

$$children = \beta_0 + \beta_1 educ + \beta_2 age + \beta_3 age^2 + u$$

where *children* is the number of children born to the woman, *educ* is years of education, and *age* is age (in years).

2. Interpret the estimates of the regression:

```
est.ols <- lm(children ~ educ + age + I(age^2), data=df)
```

(Note: include `I(age^2)` puts the quadratic term in automatically without us having to use `mutate()` to create a new variable called `age.sq`.)

We can also use `modelsummary` to examine the output. It puts the standard errors of each variable in parentheses under the estimated coefficient.

```
modelsummary(est.ols)
```

## Instrumenting for endogenous education

We know that education is endogenous (i.e. people choose the level of education that maximizes their utility). A possible instrument for education is *firsthalf*, which is a dummy equal to 1 if the woman was born in the first half of the calendar year, and 0 otherwise.

Let's create this variable:

```
df %<>% mutate(firsthalf = mnthborn<7)
```

We will assume that *firsthalf* is uncorrelated with *u*.

3. Check that *firsthalf* is correlated with *educ* by running a regression. (I will suppress the code, since it should be old hat) Call the output `est.iv1`.

## IV estimation

Now let's do the IV regression:

```
est.iv <- ivreg(children ~ educ + age + I(age^2) | firsthalf + age + I(age^2), data=df)
```

The variables on the right hand side of the `|` are the instruments (including the *x*'s that we assume to be exogenous, like *age*). The endogenous *x* is the first one after the `~`.

Now we can compare the output for each of the models:

```
modelsummary(list(est.ols,est.iv1,est.iv))
```

We can also save the output of `modelsummary()` to an image, a text file or something else:

	(1)	(2)	(3)
(Intercept)	−4.138 (0.241)	6.363 (0.087)	−3.388 (0.548)
educ	−0.091 (0.006)		−0.171 (0.053)
age	0.332 (0.017)		0.324 (0.018)
	−0.003 (0.000)		−0.003 (0.000)
firsthalfTRUE		−0.938 (0.118)	
Num.Obs.	4361	4361	4361
R2	0.569	0.014	0.550
R2 Adj.	0.568	0.014	0.550
AIC	15 681.2	24 249.6	15 864.3
BIC	15 713.1	24 268.7	15 896.2
Log.Lik.	−7835.592	−12 121.779	
F	1915.196	62.620	
RMSE	1.46	3.90	1.49

```
modelsummary(list(est.ols,est.iv1,est.iv), output="results.jpg")
```

## save\_kable will have the best result with magick installed.

```
modelsummary(list(est.ols,est.iv1,est.iv), output="results.docx")
```

4. Comment on the IV estimates. Do they make sense? Discuss why the IV standard error is so much larger than the OLS standard error.