# Social Science Inquiry II

Week 4: Joint relationships, part I

Molly Offer-Westort

Department of Political Science, University of Chicago

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# Loading packages for this class

> library(ggplot2)

### Homework

- ► Solution sets will be posted at the same time as problem sets.
- ▶ Do as much as you can on the problem set before checking the solutions.
- Check your work, and then fill out a form on how you did, what you understood and didn't.
- You get marked both on completion of the problem set, AND filling out the form.
- ► (If you find errors in the solution set, post them on the class StackOverflow and you will get extra credit)
- ► For homework assignments, always submit *both* your .R file showing your work, and and a compiled .pdf file on Canvas.

# Homework grading

### check(+/-)

- ► Check: You fully completed the assignment, and submitted all components. (A)
- Check plus: You went above and beyond, your solutions were clear and detailed. (A+)
- Check minus: You made an attempt, but it wasn't complete. Maybe you didn't submit all components, or didn't fully answer some of the questions. (B or C)
- ▶ Unmarked: You did not submit enough of an assignment for credit.

Angrist, Joshua D., and Alan B. Krueger. (1991) "Does compulsory school attendance affect schooling and earnings?"

### An aside on Nobels

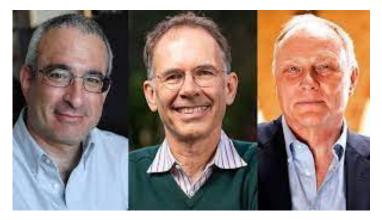


Figure: Joshua Angrist, Guido Imbens, David Card

### An aside on Nobels



Figure: Alan Krueger

Krueger and Card were economists at Princeton when they started collaborating (Card started his career at the Booth School). Josh Angrist started collaborating on the returns to schooling project as a PhD student in the department.

## Reading papers

What to get out of reading a research paper:

- ▶ What is the main question of the paper?
- ► What method do the authors use to address the question? For empirical papers:
  - ▶ Data (Where does it come from/how is it generated? What is the sample population? What is being measured?)
  - Research design/strategy
  - Statistical tools
- ▶ What is the answer that the authors get to the main question?

How would you answer these questions with the Angrist and Keueger (1991) paper?

Establishing evidence: relationship between birth quarter and education

# Loading the data

```
> dat <- read.csv('../data/angrist-krueger.csv', as.is = TRUE)
> head(dat)
```

	log_weekly_wage	education	year_of_birth	quarter_of_birth	place_of_birth
1	5.790019	12	30	1	45
2	5.952494	11	30	1	45
3	5.315949	12	30	1	45
4	5.595926	12	30	1	45
5	6.068915	12	30	1	37
6	5.793871	11	30	1	45

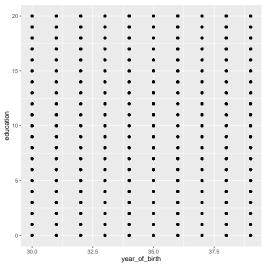
## Examining the data

```
> str(dat)
'data.frame':
                 329509 obs. of 5 variables:
$ log_weekly_wage : num
                      5.79 5.95 5.32 5.6 6.07 ...
$ education : int 12 11 12 12 12 11 11 12 11 7 ...
$ year_of_birth : int 30 30 30 30 30 30 30 30 30 30 ...
$ place_of_birth : int 45 45 45 45 37 45 36 51 45 45 ...
> summary(dat)
log_weekly_wage
                 education
                             vear_of_birth
                                          quarter_of_birth
Min.
      :-2.342
               Min. : 0.00
                             Min.
                                   :30.0
                                          Min.
                                                :1,000
                            1st Qu.:32.0
1st Qu.: 5.637 1st Qu.:12.00
                                          1st Qu.:2.000
Median: 5.952 Median: 12.00
                             Median:35.0
                                          Median :3.000
Mean : 5.900 Mean : 12.77 Mean : 34.6 Mean : 2.506
3rd Qu.: 6.257
               3rd Qu.:15.00
                             3rd Qu.:37.0
                                          3rd Qu.:3.000
Max.
      :10.532
               Max. :20.00
                             Max. :39.0
                                          Max.
                                                :4.000
place_of_birth
Min.
      : 1.00
1st Qu.:19.00
Median :34.00
Mean
      :30.69
3rd Qu.:42.00
Max.
      :56.00
```

### Data

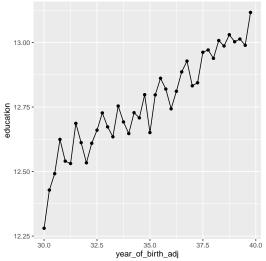
- ▶ Where does it come from/how is it generated?
- ▶ What is the sample population?
- ▶ What is being measured?

> ggplot(dat, aes(x = year\_of\_birth, y = education)) +
+ geom\_point()



```
> dat_agg <- aggregate(x = dat[, c('log_weekly_wage', 'education')],</pre>
                      by = list(`year_of_birth` = dat$year_of_birth,
                                 `quarter_of_birth` = dat$quarter_of_birth),
                      FUN = mean)
> dat_agg$year_of_birth_adj <- dat_agg$year_of_birth +</pre>
    0.25 * (dat_agg$quarter_of_birth-1)
> head(dat_agg)
  year_of_birth quarter_of_birth log_weekly_wage education year_of_birth_adj
1
             30
                                          5.889133 12.28041
                                                                             30
             31
                                          5.902136 12.54043
                                                                             31
3
             32
                                         5.899809 12.53393
                                                                             32
             33
                                         5.891946 12.67319
                                                                             33
             34
                                         5.895157 12.64726
                                                                             34
             35
                                         5.879843 12.65091
                                                                             35
```

- > ggplot(dat\_agg, aes(x = year\_of\_birth\_adj, y = education)) +
  + geom\_point() + # points
- + geom\_line() # lines



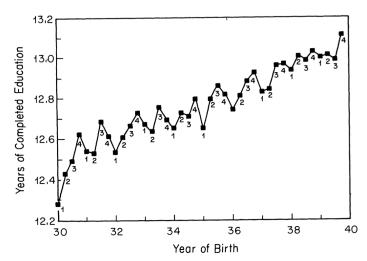
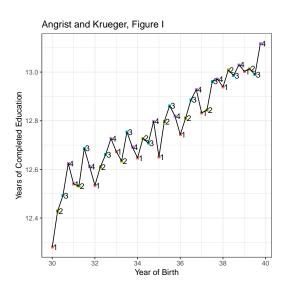


FIGURE I
Years of Education and Season of Birth
1980 Census
Note. Quarter of birth is listed below each observation.

```
> ggplot(dat_agg, aes(x = year_of_birth_adj,
                     y = education,
+
                     label = quarter of birth)) +
   geom_point(pch = 15,
               aes(color = as.factor(quarter_of_birth) )) + # points with color
+
   geom line() + # lines
   geom\_text(hjust = 0, nudge\_x = 0.05) + # text labels on points
    theme_bw() + # plot style
    theme(legend.position = '') + # remove legend from colored text labels
   vlab('Years of Completed Education') + # y-axis label
   xlab('Year of Birth') + # x-axis label
   ggtitle('Angrist and Krueger, Figure I') + # title
    scale_x_continuous(breaks = seq(30, 40, 2)) + # x-axis ticks
    scale_y_continuous(breaks = seq(12.2, 13.2, .2)) # y-axis ticks
>
```



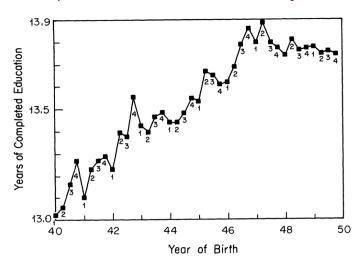


FIGURE II
Years of Education and Season of Birth
1980 Census
Note. Quarter of birth is listed below each observation.

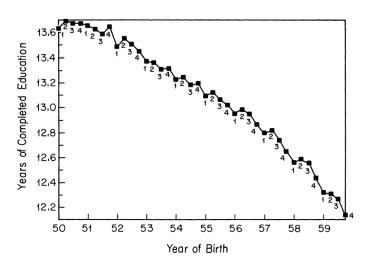
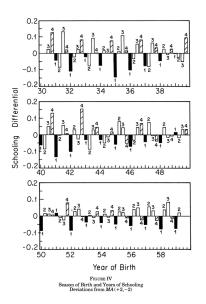
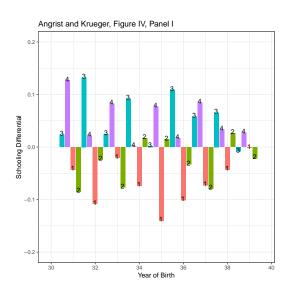


FIGURE III
Years of Education and Season of Birth
1980 Census
Note. Quarter of birth is listed below each observation.



```
> # function for moving average
> ma \leftarrow function(x, n = 5){
+ ma_x <- as.numeric(filter(x, rep(1 / n, n), sides = 2))
+ ma_x^2 < (ma_x - x/5)*5/4
+ return(ma x2)
+ }
> # get dat_agg in right order
> dat_agg <- dat_agg[order(dat_agg$year_of_birth_adj),]</pre>
> # calculate moving average
> dat_agg$moving_average <- ma(dat_agg$education)</pre>
> # update adjusted birth year in main dataset
> dat$year_of_birth_adj <- dat$year_of_birth + 0.25 * (dat$quarter_of_birth-1)</pre>
> # and match aggregated moving average to main data
> dat$moving_average <- dat_agg$moving_average[match(dat$year_of_birth_adj,</pre>
                                                     dat_agg$year_of_birth_adj)]
> # calculate deviation from moving average
> dat$deviation <- dat$education-dat$moving_average</pre>
> # get aggregate deviation
> dat_agg$deviation <- aggregate(x = dat$deviation,
                                 by = list(dat$year_of_birth_adj), mean)$x
```



What is the case that the difference in education across quarters is due to compulsory schooling?

### Inference

- ▶ Over what population do these effects apply?
  - ► Time frame
  - ► Geography
  - Policy

# Policy implications

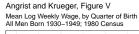
What should we do with this evidence?

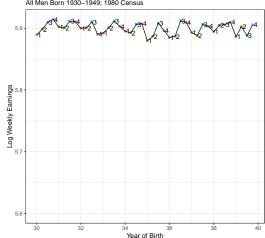
- ▶ Should we change compulsory school attendance laws in the US?
- ▶ If you were hired as a consultant for another country, would you recommend to change compulsory school attendance laws? Under what conditions?

Estimating causal effects: returns to education

> ggplot(dat\_agg, aes(x = year\_of\_birth\_adj, y = log\_weekly\_wage,
+ label = quarter\_of\_birth)) +

```
geom point(pch = 15.
               aes(color = as.factor(quarter_of_birth) )) + # points with color
+
   geom_line() + # lines
   geom_text(hjust = 0, nudge_x = 0.05) + # text labels on points
   theme_bw() + # plot style
   theme(legend.position = '') + # remove legend from colored text labels
   scale_x_continuous(breaks = seq(30, 40, 2)) + # x-axis ticks
+
   scale_v_continuous(breaks = seg(5.6, 6.1, .1)) + # y-axis ticks
+
   coord\_cartesian(ylim = c(5.6, NA)) +
+
   vlab('Log Weekly Earnings') + # y-axis label
   xlab('Year of Birth') + # x-axis label
   ggtitle('Angrist and Krueger, Figure V',
           subtitle = 'Mean Log Weekly Wage, by Quarter of Birth\nAll Men Born 1930
```





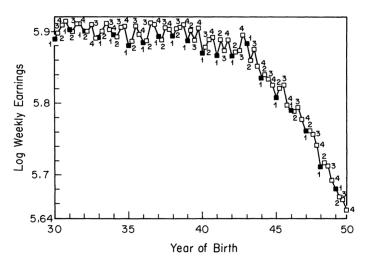


FIGURE V Mean Log Weekly Wage, by Quarter of Birth All Men Born 1930–1949; 1980 Census

### Wald estimator

#### Computes returns to education as ratio:

- ▶ numerator: the difference in earning by quarter of birth
- ▶ denominator: the difference in education by quarter of birth
- ► comparison: men born in first quarter vs. men born in last three quarters

The Wald estimator is a simple example of **Instrumental Variables** analysis, where you *instrument* for changes in X with changes in some instrument, Z.

### Wald estimator

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### Wald estimator

The Wald estimator is a simple example of **Instrumental Variables** analysis, where you *instrument* for changes in X with changes in some instrument, Z.

$$\beta_{IV} = \frac{\delta y/\delta z}{\delta x/\delta z}$$

With the Wald estimator, Z is binary.

$$\hat{\beta}_{\textit{Wald}} = \frac{\bar{y}_1 - \bar{y}_0}{\bar{x}_1 - \bar{x}_0}$$

TABLE III
PANEL A: WALD ESTIMATES FOR 1970 CENSUS—MEN BORN 1920–1929\*

	(1) Born in 1st quarter of year	(2) Born in 2nd, 3rd, or 4th quarter of year	(3) Difference (std. error) (1) - (2)
ln (wkly. wage)	5.1484	5.1574	-0.00898
			(0.00301)
Education	11.3996	11.5252	-0.1256
			(0.0155)
Wald est. of return to education			0.0715
			(0.0219)
OLS return to education <sup>b</sup>			0.0801
			(0.0004)

Panel B: Wald Estimates for 1980 Census—Men Born 1930-1939 (1) (3) Born in Born in 2nd, Difference 1st quarter 3rd, or 4th (std. error) of year quarter of year (1) - (2)ln (wkly. wage) 5.8916 5 9027 -0.01110(0.00274)Education -0.108812.6881 12,7969 (0.0132)0.1020 Wald est, of return to education (0.0239)OLS return to education 0.0709 (0.0003)

a. The sample size is 247,199 in Panel A, and 327,509 in Panel B. Each sample consists of males born in the United States who had positive earnings in the year preceding the survey. The 1990 Census sample is drawn from the 5 percent sample, and the 1970 Census sample is from the State, County, and Neighborhoods 1 percent samples.

b. The OLS return to education was estimated from a bivariate regression of log weekly earnings on years of education.

What makes this paper so compelling? What is its contribution to research methods in the social sciences?

### References I

Angrist, J. D. and Keueger, A. B. (1991). Does compulsory school attendance affect schooling and earnings? The Quarterly Journal of Economics, 106(4):979–1014.