

# Assignment #5

*ECON 5783 — University of Arkansas*

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These assignments should be completed in groups of 1 or 2 but submitted individually. My preference is for you to use Rmarkdown files to have your code, results, and your answers to the questions intermixed. Since I am not requiring you to code in R for these assignments, you can use latex or microsoft word to write up answers alternatively. Not that in these cases, I would like you to upload your code seperately.

## Theoretical Questions

1. In the following examples, give an example why you think the following exclusion restrictions might fail
  - i. You regress future wages ( $y_i$ ) on going to post-secondary school ( $D_i$ ) with an instrument of being within a 15-minute drive of a community college ( $Z_i$ )
  - ii. You regress house prices ( $y_i$ ) on the town's school quality ( $X_i$ ) with an instrument for whether a town levied an extra school-funding tax ( $Z_i$ )
  - iii. You regress a person's happiness ( $y_i$ ) on their wealth ( $X_i$ ) with an instrument for the parent's wealth ( $Z_i$ )
2. In your own words, why is inference hard on the two-stage least squares coefficient when you have a 'weak' instrument

## Coding Exercise: Gerber, Huber, and Washington (2010, APSR)

This exercise will have you work with replication material from Gerber, Huber, and Washington.<sup>1</sup> The dataset can be loaded from `gerber_et_al_2010.csv`. This paper wants to test they hypothesis that registering with a political party (instead of as an 'Independent') changes a person's political beliefs. To do so, they run a randomized experiment in Connecticut for the 2008 election. In Connecticut, you have to be registered with a political party (Democrat or Republican) to vote

1. Full data documentation is available at <https://isps.yale.edu/research/data/d055>.

in that party's primary election. From a sample of unregistered voters, the authors randomly assigned some voters to receive information ( $Z = \text{treat}$ ) that let them know this (so that they register with a political party).

Each voter they contacted was asked questions to gauge which political party they most closely aligned with. They then recorded whether the vote registered with their ideological party ( $D = \text{pt\_id\_with\_lean}$ ). The outcome of interest is the voter's political ideology a few months after registering with a party, which they measured via a follow-up survey with a similar set of policy questions. They created a variable ( $y = \text{pt\_voteevalalignindex}$ ) which has larger values if the person more strongly agrees with *their* political party's stances.

### Overview of experiment

- $Z_i$ : Voters are randomly told before the primary (say January) that you must register with a party to vote in the primary
- $D_i$ : Voters (possibly) register to vote with a party before the primary
- $y_i$ : Voters are surveyed later in the year (say July) about their political beliefs.  $y_i$  is larger if you are more 'ideologically aligned' with your registered party

### Questions

1. First, let's see if the experimental design actually caused some voters to register with their political party. To do so, run the first-stage regression of  $Z$  on  $D$ . Interpret, using the  $F$ -statistic, whether the first-stage is "strong".
2. Second, run the reduced-form regression of  $y$  on  $Z$ . Then, form the  $\beta_{2\text{SLS}}$  estimate by hand
3. Estimate the two-stage least squares estimate using `feols`. From this result, interpret whether or not registering with a political party changes a voter's ideology. Is the estimate statistically significant?
4. Now, let's try and characterize compliers using the method highlighted in the lecture slides (2SLS regression of  $D_i * X_i$  on  $D_i$  instrument by  $Z_i$ ). For the following variables, compare these complier's means to the overall population means: the voter's age in 2008 (`age`); whether the voter voted in the 2006 midterm elections (`voted2006`); and whether the voter attends church or not (`pt_church`).
  - Given this information, how does this change how you interpret the LATE estimate?