Exercise_ADH

2022-09-05

Loading data, packages

```
library(readr)
library(tibble)
library(fixest)
library(stringr)
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
library(lmtest)
## Loading required package: zoo
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##
       as.Date, as.Date.numeric
library(sandwich)
df = read_csv("adh_data.csv")
## Rows: 1444 Columns: 208
## -- Column specification --
## Delimiter: ","
## chr
        (1): city
## dbl (207): czone, statefip, yr, t2, timepwt48, reg_midatl, reg_encen, reg_wn...
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
head(df)
## # A tibble: 6 x 208
                                yr t2 timepwt48 reg_m~1 reg_e~2 reg_w~3 reg_s~4
    czone statefip city
##
##
     <dbl>
           <dbl> <chr>
                             <dbl> <dbl>
                                             <dbl> <dbl>
                                                             <dbl>
                                                                     <dbl>
                                                                             <dbl>
## 1
      100
                47 undefined 1990 0 0.00211
                                                        0
                                                                 0
                                                                         0
                                                                                 0
## 2
    100
               47 undefined 2000
                                       1 0.00207
                                                         0
                                                                 0
                                                                         0
```

```
## 3
       200
                 47 undefined 1990
                                        0 0.000732
                                                                   0
                                                                           0
                                                                                   0
## 4
       200
                 47 undefined 2000
                                           0.000815
                                                                                   0
                                        1
                                                           0
                                                                   0
                                                                           0
                                           0.000261
## 5
       301
                 47 undefined 1990
                                        0
                                                           0
                                                                           0
                                                                                   0
                                                                                   0
## 6
       301
                 47 undefined 2000
                                        1 0.000240
                                                           Ω
                                                                   0
                                                                           0
## # ... with 198 more variables: reg_escen <dbl>, reg_wscen <dbl>,
      reg_mount <dbl>, reg_pacif <dbl>, l_popcount <dbl>,
       l_no_workers_totcbp <dbl>, l_shind_manuf_cbp <dbl>, l_sh_popedu_c <dbl>,
## #
       1_sh_popfborn <dbl>, 1_sh_empl_f <dbl>, 1_sh_routine33 <dbl>,
## #
## #
       1_task_outsource <dbl>, 1_sh_empl <dbl>, 1_sh_empl_mfg <dbl>,
## #
       l_sh_empl_mfg_edu_c <dbl>, l_sh_empl_mfg_edu_nc <dbl>,
       1_sh_empl_mfg_age1634 <dbl>, l_sh_empl_mfg_age3549 <dbl>, ...
```

OLS regression

The core of the paper is looking at what happened to laborer's when theres an increase in us imports from china. Let's try and replicate part of Table 9 - namely the estimate from panel A column 2.

Their y variable is relchg_avg_hhincwage_pc_pw.

The important x variable is decadal trade between the us and china d_tradeusch_pw.

1. Run that simple regression

```
ols_1 = lm(relchg_avg_hhincwage_pc_pw ~ d_tradeusch_pw, data = df)
summary(ols_1)
```

```
##
## Call:
## lm(formula = relchg_avg_hhincwage_pc_pw ~ d_tradeusch_pw, data = df)
##
## Residuals:
##
      Min
                1Q Median
                               3Q
                                      Max
## -28.789 -8.411 -0.663
                            7.715
                                   49.684
##
## Coefficients:
                 Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                  16.0720
                              0.3889
                                       41.33
                                               <2e-16 ***
## d_tradeusch_pw -1.6466
                              0.1212 -13.59
                                               <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 11.89 on 1442 degrees of freedom
## Multiple R-squared: 0.1135, Adjusted R-squared: 0.1129
## F-statistic: 184.6 on 1 and 1442 DF, p-value: < 2.2e-16
```

2. Now add heteroskedasticity robust standard (HC1). Hint: Use the sandwich and lmtest packages

```
# ols_1 = lm(relchg_avg_hhincwage_pc_pw ~ d_tradeusch_pw, data = df)
lmtest::coeftest(ols_1, sandwich::vcovHC(ols_1, "HC1"))
```

```
##
## t test of coefficients:
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept) 16.07198 0.57211 28.0923 < 2.2e-16 ***
## d_tradeusch_pw -1.64663 0.28496 -5.7785 9.219e-09 ***
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1</pre>
```

```
ols_rob = feols(relchg_avg_hhincwage_pc_pw ~ d_tradeusch_pw, data = df, se = "HC1")
summary(ols_rob)
## OLS estimation, Dep. Var.: relchg_avg_hhincwage_pc_pw
## Observations: 1,444
## Standard-errors: Heteroskedasticity-robust
##
                 Estimate Std. Error t value
                                                Pr(>|t|)
## (Intercept)
                 16.07198
                             0.572114 28.09229 < 2.2e-16 ***
## d_tradeusch_pw -1.64663
                             0.284956 -5.77854 9.2187e-09 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## RMSE: 11.9
               Adj. R2: 0.112885
Now we will start to add extra x variables.
  3. Start by adding t2 - a dummy variable for whether observation is in the second decade. Fit again with
    HC1 robust standard errors.
ols_2 = feols(relchg_avg_hhincwage_pc_pw ~ d_tradeusch_pw + t2,
                data = df, vcov = "HC1")
summary(ols_2)
## OLS estimation, Dep. Var.: relchg_avg_hhincwage_pc_pw
## Observations: 1,444
## Standard-errors: Heteroskedasticity-robust
##
                   Estimate Std. Error t value
                                                    Pr(>|t|)
## (Intercept)
                  21.590958
                             0.398893 54.12714 < 2.2e-16 ***
## d_tradeusch_pw -0.883161
                               0.168044 -5.25555 1.6977e-07 ***
                  -13.947688
                             0.578687 -24.10229 < 2.2e-16 ***
## t2
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## RMSE: 9.82092
                   Adj. R2: 0.393606
ols_2_hc3 = feols(relchg_avg_hhincwage_pc_pw ~ d_tradeusch_pw + t2,
                  data = df, vcov = function (x) sandwich::vcovHC(x, "HC3"))
summary(ols_2_hc3)
## OLS estimation, Dep. Var.: relchg_avg_hhincwage_pc_pw
## Observations: 1,444
## Standard-errors: vcovHC(x, "HC3")
                   Estimate Std. Error
##
                                        t value
                                                    Pr(>|t|)
## (Intercept)
                  21.590958
                              0.410042 52.65551 < 2.2e-16 ***
## d_tradeusch_pw -0.883161
                               0.187696 -4.70526 2.7788e-06 ***
## t2
                 -13.947688
                             0.588642 -23.69469 < 2.2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## RMSE: 9.82092
                  Adj. R2: 0.393606
Clustering
Let us now use clustertered standard errors instead. ADH cluster by statefip.
  1. Run the basic regression with clustering
ols_clust = feols(relchg_avg_hhincwage_pc_pw ~ d_tradeusch_pw,
                  data = df, cluster = c("statefip"))
```

summary(ols_clust)

```
## OLS estimation, Dep. Var.: relchg_avg_hhincwage_pc_pw
## Observations: 1,444
## Standard-errors: Clustered (statefip)
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                16.07198
                          0.957363 16.78776 < 2.2e-16 ***
## d tradeusch pw -1.64663
                          0.370000 -4.45034 5.2454e-05 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## RMSE: 11.9
             Adj. R2: 0.112885
  2. Add the following controls to your last regression:
      • l_shind_manuf_cbp
      • l_sh_popedu_c
      • l_sh_popfborn
      • l_sh_empl_f
      • l_sh_routine33
      • l_task_outsource
ols_clust = feols(relchg_avg_hhincwage_pc_pw ~ d_tradeusch_pw + l_shind_manuf_cbp +
                  l_sh_popedu_c + l_sh_popfborn + l_sh_popfborn + l_sh_empl_f +
                  1_sh_routine33 + 1_task_outsource,
                data = df, cluster = c("statefip"))
summary(ols_clust)
## OLS estimation, Dep. Var.: relchg_avg_hhincwage_pc_pw
## Observations: 1,444
## Standard-errors: Clustered (statefip)
##
                   Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                   98.430164 11.191489 8.795091 1.7032e-11 ***
## d_tradeusch_pw
                  -1.063005
                             0.287892 -3.692369 5.7754e-04 ***
## l_sh_popedu_c
                  ## l_sh_popfborn
                   0.090310 0.748463 4.5791e-01
## l_sh_empl_f
                   0.067594
## 1_sh_routine33
                   -2.041397
                              0.287715 -7.095208 5.8625e-09 ***
## l_task_outsource 17.046616
                              2.083908 8.180119 1.3734e-10 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## RMSE: 10.4
             Adj. R2: 0.317146
  3. Add region fixed effects to your regression.
      • First find all variables in the dataset that start with reg
      • Add these to your last regression
# reg_vars <- df %>% select(starts_with("reg_")) %>% names()
reg_vars <- df %>% colnames() %>% str_subset("reg_")
fe str = str c(reg vars, collapse = " + ")
form str = str c(
   "relchg_avg_hhincwage_pc_pw ~ d_tradeusch_pw + l_shind_manuf_cbp",
   "l_sh_popedu_c + l_sh_popfborn + l_sh_popfborn + l_sh_empl_f",
   "l_sh_routine33 + l_task_outsource",
   fe str,
   sep = " + "
```

form <- as.formula(form_str)</pre>

```
fe_model = feols(form, data = df, cluster = c("statefip"))
summary(fe_model)
## OLS estimation, Dep. Var.: relchg_avg_hhincwage_pc_pw
## Observations: 1,444
## Standard-errors: Clustered (statefip)
                                              Estimate Std. Error t value
                                                                                                                 Pr(>|t|)
## (Intercept)
                                         103.331722 13.422215 7.698560 7.2020e-10 ***
## d_tradeusch_pw
                                           ## l_shind_manuf_cbp -0.139771 0.045212 -3.091447 3.3447e-03 **
## l_sh_popedu_c
                                         ## l_sh_popfborn
## l_sh_empl_f
                                            ## 1_sh_routine33
## 1_task_outsource 17.653643 2.200501 8.022556 2.3573e-10 ***
## reg_midatl
                                          -1.139319 2.335920 -0.487739 6.2800e-01
                                           -0.511749 2.735801 -0.187056 8.5242e-01
## reg_encen
                                             5.788149 2.668650 2.168943 3.5180e-02 *
## reg_wncen
## reg_satl
                                            -4.025975 2.784034 -1.446094 1.5479e-01
## reg_escen
                                           -3.523777 2.846048 -1.238130 2.2182e-01
                                             1.113610 2.460586 0.452579 6.5293e-01
## reg wscen
## reg mount
                                             1.913916 2.617786 0.731120 4.6834e-01
## reg_pacif
                                            -1.892620 2.649640 -0.714293 4.7858e-01
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## RMSE: 10.1
                              Adj. R2: 0.358537
df = df %>% mutate(region = case_when(
    reg midatl == 1 ~ 'midatl',
   reg encen == 1 ~ 'encen',
    reg_wncen == 1 ~ 'wncen',
   reg_satl == 1 ~ 'satl',
   reg_escen == 1 ~ 'escen',
   reg_wscen == 1 ~ 'wscen',
   reg_mount == 1 ~ 'mount',
   reg_pacif == 1 ~ 'pacif',
    TRUE ~ NA_character_
))
fe_model <- feols(</pre>
        \verb|relchg_avg_hhincwage_pc_pw| ~ \texttt{d}_tradeusch_pw + \texttt{l}_shind_manuf_cbp + \texttt{l}_shind_man
                l_sh_popedu_c + l_sh_popfborn + l_sh_popfborn + l_sh_empl_f +
                l_sh_routine33 + l_task_outsource + i(region),
        data = df, cluster = c("statefip")
)
## NOTE: 32 observations removed because of NA values (RHS: 32).
summary(fe_model)
## OLS estimation, Dep. Var.: relchg_avg_hhincwage_pc_pw
## Observations: 1,412
## Standard-errors: Clustered (statefip)
##
                                               Estimate Std. Error
                                                                                            t value
                                                                                                                 Pr(>|t|)
## (Intercept)
                                          102.733914 13.411295 7.660253 1.9580e-09 ***
```

```
## d tradeusch pw
                  -0.949723
                           0.288561 -3.291238 2.0571e-03 **
## 1_shind_manuf_cbp -0.139459 0.045500 -3.065072 3.8397e-03 **
## 1 sh popedu c
                 ## l_sh_popfborn
                 ## l_sh_empl_f
                 -0.219595 0.123019 -1.785055 8.1652e-02 .
## 1 sh routine33
                 ## 1 task outsource
                 17.618681 2.217757 7.944369 7.9153e-10 ***
                 -3.026390 2.013417 -1.503111 1.4047e-01
## region::escen
## region::midatl
                 -0.605299 1.579094 -0.383320 7.0346e-01
## region::mount
                 2.493887
                           2.296060 1.086159 2.8375e-01
## region::pacif
                 -1.276165 2.209543 -0.577570 5.6671e-01
                           2.035516 -1.725170 9.2028e-02
## region::satl
                 -3.511611
## region::wncen
                  6.270456
                          ## region::wscen
                  1.649646
                          1.668402 0.988758 3.2858e-01
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## RMSE: 10.1 Adj. R2: 0.358638
fe_model <- feols(</pre>
   relchg_avg_hhincwage_pc_pw ~ d_tradeusch_pw + l_shind_manuf_cbp +
      l_sh_popedu_c + l_sh_popfborn + l_sh_popfborn + l_sh_empl_f +
      1_sh_routine33 + 1_task_outsource | region,
   data = df, cluster = c("statefip")
)
## NOTE: 32 observations removed because of NA values (Fixed-effects: 32).
summary(fe_model)
## OLS estimation, Dep. Var.: relchg_avg_hhincwage_pc_pw
## Observations: 1,412
## Fixed-effects: region: 8
## Standard-errors: Clustered (statefip)
                 Estimate Std. Error t value
                                           Pr(>|t|)
                ## d_tradeusch_pw
## l_sh_popedu_c
## l_sh_popfborn
                ## l_sh_empl_f
                -0.219595
                          0.123019 -1.78505 8.1652e-02 .
## l_sh_routine33
                 -1.641218
                          0.323082 -5.07989 8.6710e-06 ***
## l_task_outsource 17.618681
                          2.217757 7.94437 7.9153e-10 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## RMSE: 10.1
              Adj. R2: 0.358638
##
             Within R2: 0.303106
Instrument Variables
 1. Instrument d_tradeusch_pw with d_tradeotch_pw_lag in your last regression
iv reg = feols(
```

```
summary(iv_reg)
## TSLS estimation, Dep. Var.: relchg_avg_hhincwage_pc_pw, Endo.: d_tradeusch_pw, Instr.: d_tradeotch_p
## Second stage: Dep. Var.: relchg_avg_hhincwage_pc_pw
## Observations: 1,444
## Standard-errors: Clustered (statefip)
##
                     Estimate Std. Error
                                       t value
                                                 Pr(>|t,|)
## (Intercept)
                    84.382112 11.863723 7.112616 5.5171e-09 ***
## 1_shind_manuf_cbp -0.024517 0.065685 -0.373258 7.1063e-01
## l_sh_popedu_c
                   ## l_sh_popfborn
                   0.091119 0.099534 0.915459 3.6462e-01
## l_sh_empl_f
## l_sh_routine33
                   -1.725561 0.296334 -5.823030 4.9822e-07 ***
## 1_task_outsource 14.615441 2.030773 7.196985 4.1110e-09 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## RMSE: 10.1
             Adj. R2: 0.239452
## F-test (1st stage), d_{tradeusch_pw}: stat = 907.0, p < 2.2e-16, on 1 and 1,436 DoF.
                        Wu-Hausman: stat = 111.1, p < 2.2e-16, on 1 and 1,435 DoF.
  2. Weight your regression by timepwt48
iv_reg_2 = feols(
   relchg_avg_hhincwage_pc_pw ~ l_shind_manuf_cbp +
       l_sh_popedu_c + l_sh_popfborn + l_sh_popfborn + l_sh_empl_f +
       l_sh_routine33 + l_task_outsource | 0 | d_tradeusch_pw ~ d_tradeotch_pw_lag,
   data = df, cluster = c("statefip"),
   weights = ~timepwt48
summary(iv_reg_2)
## TSLS estimation, Dep. Var.: relchg_avg_hhincwage_pc_pw, Endo.: d_tradeusch_pw, Instr.: d_tradeotch_p
## Second stage: Dep. Var.: relchg_avg_hhincwage_pc_pw
## Observations: 1,444
## Standard-errors: Clustered (statefip)
##
                    Estimate Std. Error t value
                                                  Pr(>|t|)
## (Intercept)
                    73.961602 11.854845 6.238934 1.1694e-07 ***
                             0.570072 -6.640973 2.8672e-08 ***
## fit_d_tradeusch_pw -3.785835
## l_shind_manuf_cbp 0.169233 0.069850 2.422798 1.9308e-02 *
                   -0.186881 0.131546 -1.420653 1.6202e-01
## l_sh_popedu_c
## l_sh_popfborn
                   -0.080177 0.086035 -0.931921 3.5614e-01
                             0.138788 -1.029976 3.0829e-01
## l_sh_empl_f
                    -0.142948
## l_sh_routine33
                   -1.217598
                              0.352090 -3.458202 1.1655e-03 **
                              1.944024 5.778867 5.8079e-07 ***
## l_task_outsource 11.234254
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## RMSE: 0.330337 Adj. R2: 0.185213
## F-test (1st stage), d_tradeusch_pw: stat = 1,117.1, p < 2.2e-16, on 1 and 1,436 DoF.
                        Wu-Hausman: stat = 154.6, p < 2.2e-16, on 1 and 1,435 DoF.
iv_reg_2$iv_first_stage
## $d_tradeusch_pw
## TSLS estimation, Dep. Var.: d_tradeusch_pw, Endo.: d_tradeusch_pw, Instr.: d_tradeotch_pw_lag
## First stage: Dep. Var.: d_tradeusch_pw
```

```
## Observations: 1,444
## Standard-errors: Clustered (statefip)
##
                  Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                  -1.832015 1.268263 -1.444508 1.5523e-01
## d_tradeotch_pw_lag 0.746107 0.072278 10.322808 1.1372e-13 ***
## l_shind_manuf_cbp 0.044772 0.008211 5.452934 1.7923e-06 ***
## l_sh_popedu_c
                  ## l_sh_popfborn
## l_sh_empl_f
                  0.005347 0.010529 0.507884 6.1391e-01
## l_sh_routine33
                 -0.003746 0.041013 -0.091337 9.2761e-01
## 1_task_outsource -0.241380 0.276901 -0.871719 3.8779e-01
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## RMSE: 0.043006 Adj. R2: 0.562323
## F-test (1st stage): stat = 1,117.1, p < 2.2e-16, on 1 and 1,436 DoF.
```