Panel Data 2: Implementation in R

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Panel

Preliminary:

- I use the following package
 - lfe package.

Panel Data Regression

- I use the dataset Fatalities in AER package.
 - See https://www.rdocumentation.org/packages/AER/versions/1.2-6/topics/Fatalities for details.

```
library(AER)
data(Fatalities)
str(Fatalities)
## 'data.frame':
                 336 obs. of 10 variables:
   $ state : Factor w/ 48 levels "al", "az", "ar", ...: 1 1 1 1 1 1 1 2 2 2 ...
##
##
   $ year
             : Factor w/ 7 levels "1982", "1983", ...: 1 2 3 4 5 6 7 1 2 3 ...
   $ spirits : num 1.37 1.36 1.32 1.28 1.23 ...
##
##
   $ unemp
                    14.4 13.7 11.1 8.9 9.8 ...
             : num
   $ income : num
##
                    10544 10733 11109 11333 11662 ...
##
   $ emppop : num 50.7 52.1 54.2 55.3 56.5 ...
##
   $ beertax : num 1.54 1.79 1.71 1.65 1.61 ...
   $ baptist : num 30.4 30.3 30.3 30.3 30.3 ...
   $ mormon : num 0.328 0.343 0.359 0.376 0.393 ...
##
##
   $ drinkage: num 19 19 19 19.7 21 ...
```

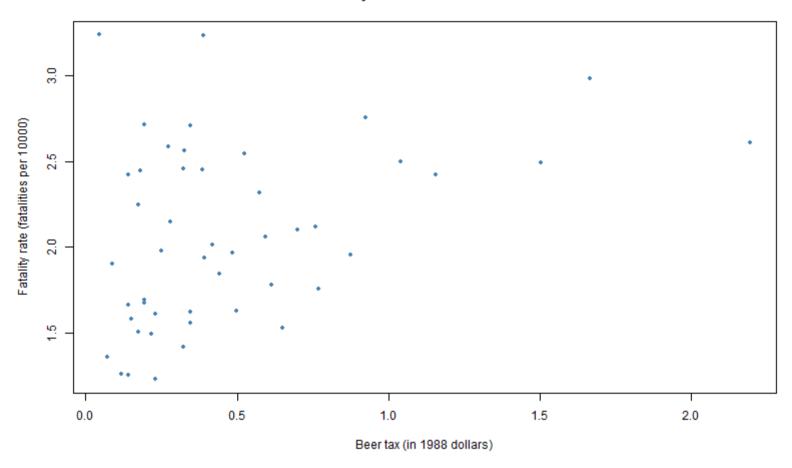
```
## 'data.frame': 336 obs. of 24 variables:
   $ drv
##
                 : num 25 23 24 23.6 23.5 ...
   $ youngdrivers: num 0.212 0.211 0.211 0.211 0.213 ...
##
   $ miles
                 : num 7234 7836 8263 8727 8953 ...
   $ breath
                 : Factor w/ 2 levels "no", "yes": 1 1 1 1 1 1 1 1 1 1 ...
##
##
   $ iail
                 : Factor w/ 2 levels "no", "yes": 1 1 1 1 1 1 1 2 2 2 ...
##
   $ service
                 : Factor w/ 2 levels "no", "yes": 1 1 1 1 1 1 1 2 2 2 ...
##
   $ fatal
                 : int 839 930 932 882 1081 1110 1023 724 675 869 ...
##
   $ nfatal
                 : int
                       146 154 165 146 172 181 139 131 112 149 ...
##
   $ sfatal
                 : int 99 98 94 98 119 114 89 76 60 81 ...
   $ fatal1517
                 : int 53 71 49 66 82 94 66 40 40 51 ...
##
##
   $ nfatal1517
                 : int 9 8 7 9 10 11 8 7 7 8 ...
##
   $ fatal1820
                       99 108 103 100 120 127 105 81 83 118 ...
                 : int
   $ nfatal1820
                 : int
                       34 26 25 23 23 31 24 16 19 34 ...
##
   $ fatal2124
##
                 : int
                       120 124 118 114 119 138 123 96 80 123 ...
   $ nfatal2124
##
                 : int 32 35 34 45 29 30 25 36 17 33 ...
##
   $ afatal
                 : num
                        309 342 305 277 361 ...
##
   $ pop
                       3942002 3960008 3988992 4021008 4049994 ...
                 : num
##
   $ pop1517
                        209000 202000 197000 195000 204000 ...
                 : num
##
   $ pop1820
                 : num
                        221553 219125 216724 214349 212000 ...
##
   $ pop2124
                        290000 290000 288000 284000 263000 ...
                 : num
##
   $ milestot
                        28516 31032 32961 35091 36259 ...
                 : num
                        9.7 9.6 7.5 7.2 7 ...
##
   $ unempus
                 : num
##
   $ emppopus
                        57.8 57.9 59.5 60.1 60.7 ...
                 : num
##
   $ gsp
                        -0.0221 0.0466 0.0628 0.0275 0.0321 ...
                 : num
```

• As a preliminary analysis, let's plot the relationship between fatality rate and beer tax in 1998.

```
Fatalities %>%
  mutate(fatal_rate = fatal / pop * 10000) %>%
  filter(year == "1988") -> data

plot(x = data$beertax,
     y = data$fatal_rate,
     xlab = "Beer tax (in 1988 dollars)",
     ylab = "Fatality rate (fatalities per 10000)",
     main = "Traffic Fatality Rates and Beer Taxes in 1988",
     pch = 20,
     col = "steelblue")
```

Traffic Fatality Rates and Beer Taxes in 1988



• Positive correlation between alcohol tax and traffic accident. Possibly due to omitted variable bias.

- Run fixed effect regression using felm command in lfe package.
 - https://www.rdocumentation.org/packages/lfe/versions/2.8-3/topics/felm

```
library("lfe")

Fatalities %>%
    mutate(fatal_rate = fatal / pop * 10000) -> data

# OLS

result_ols <- felm( fatal_rate ~ beertax | 0 | 0 | 0, data = data )
summary(result_ols, robust = TRUE)</pre>
```

```
##
## Call:
     felm(formula = fatal rate ~ beertax | 0 | 0 | 0, data = data)
##
##
## Residuals:
       Min
                 10 Median
##
                                  30
                                          Max
## -1.09060 -0.37768 -0.09436 0.28548 2.27643
##
## Coefficients:
##
              Estimate Robust s.e t value Pr(>|t|)
## (Intercept) 1.85331 0.04713 39.324 < 2e-16 ***
## beertax
           0.36461 0.05285 6.899 2.64e-11 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5437 on 334 degrees of freedom
## Multiple R-squared(full model): 0.09336 Adjusted R-squared: 0.09065
## Multiple R-squared(proj model): 0.09336 Adjusted R-squared: 0.09065
## F-statistic(full model, *iid*):34.39 on 1 and 334 DF, p-value: 1.082e-08
## F-statistic(proj model): 47.59 on 1 and 334 DF, p-value: 2.643e-11
```

```
# State FE
result_stateFE <- felm( fatal_rate ~ beertax | state | 0 | state, data = data )
summary(result stateFE, robust = TRUE)
##
## Call:
##
     felm(formula = fatal_rate ~ beertax | state | 0 | state, data = data)
##
## Residuals:
                 10 Median
##
       Min
                                  30
                                          Max
## -0.58696 -0.08284 -0.00127 0.07955 0.89780
##
## Coefficients:
##
          Estimate Cluster s.e. t value Pr(>|t|)
## beertax -0.6559
                    0.2919 -2.247 0.0294 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1899 on 287 degrees of freedom
## Multiple R-squared(full model): 0.905 Adjusted R-squared: 0.8891
## Multiple R-squared(proj model): 0.04074 Adjusted R-squared: -0.1197
## F-statistic(full model, *iid*):56.97 on 48 and 287 DF, p-value: < 2.2e-16
## F-statistic(proj model): 5.05 on 1 and 47 DF, p-value: 0.02936
```

```
# State and Year FF
result_bothFE <- felm( fatal_rate ~ beertax | state + year | 0 | state, data = data )
summary(result bothFE, robust = TRUE)
##
## Call:
##
     ##
## Residuals:
               1Q Median
##
      Min
                                      Max
                               30
## -0.59556 -0.08096 0.00143 0.08234 0.83883
##
## Coefficients:
##
         Estimate Cluster s.e. t value Pr(>|t|)
## beertax -0.6400
                  0.3539 - 1.809 0.0769 .
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1879 on 281 degrees of freedom
## Multiple R-squared(full model): 0.9089 Adjusted R-squared: 0.8914
## Multiple R-squared(proj model): 0.03606 Adjusted R-squared: -0.1492
## F-statistic(full model, *iid*):51.93 on 54 and 281 DF, p-value: < 2.2e-16
## F-statistic(proj model): 3.271 on 1 and 47 DF, p-value: 0.07692
```

Report results using texreg. Note that

- Setting "robust" option TRUE reports Heteroskedasticity-robust SE for the first column.
- Automatically report Cluster-Robust SE for the second and the third columns.

```
library(texreg)
screenreg(l = list(result_ols, result_stateFE, result_bothFE),
         digits = 3,
          # caption = 'title',
         # custom.model.names = c("(I)", "(II)", "(III)", "(IV)", "(V)"),
          custom.coef.names = NULL, # add a class, if you want to change the names of variables.
          include.ci = F,
          include.rsguared = FALSE, include.adjrs = TRUE, include.nobs = TRUE,
          include.pvalues = FALSE, include.df = FALSE, include.rmse = FALSE,
          robust = T, # robust standard error
          custom.header = list("fatal_rate" = 1:3),
          custom.gof.rows = list("State FE"=c("No", "Yes", "Yes"), "Year FE"=c("No", "No", "Yes")),
          stars = numeric(0)
```

```
##
                            fatal_rate
##
##
                    Model 1 Model 2 Model 3
##
## (Intercept)
                     1.853
##
                   (0.047)
## beertax
                     0.365 -0.656 -0.640
##
                     (0.053) (0.292) (0.354)
## State FE
                     No Yes Yes
## Year FE
                   No No
                                Yes
## Num. obs. 336
                            336 336
## Adj. R^2 (full model) 0.091 0.889 0.891
## Adj. R^2 (proj model) 0.091 -0.120 -0.149
## Num. groups: state
                             48
                                     48
## Num. groups: year
```

What if we do not use the cluster-robust standard error?

```
# State FF w.o. CRS
result wo CRS <- felm( fatal rate ~ beertax | state | 0 | 0, data = data )
# State FF w. CRS
result_w_CRS <- felm( fatal_rate ~ beertax | state | 0 | state, data = data )
# Report heteroskedasticity robust standard error and cluster-robust standard errors
screenreg(l = list(result_wo_CRS, result_w_CRS),
         digits = 3,
         # caption = 'title'.
         # custom.model.names = c("(I)", "(II)", "(III)", "(IV)", "(V)"),
         custom.coef.names = NULL, # add a class, if you want to change the names of variables.
          stars = numeric(0),
         include.ci = F,
          include.rsquared = FALSE, include.adjrs = TRUE, include.nobs = TRUE,
         include.pvalues = FALSE, include.df = FALSE, include.rmse = FALSE,
          robust = T, # robust standard error
          custom.header = list("fatal_rate" = 1:2),
          custom.note = 'SE of `Model 1` is "Heteroskedasticity-Robust", while one of `Model 2` is '
```

```
##
                           fatal_rate
##
##
##
                       Model 1
                                Model 2
                        -0.656 -0.656
## beertax
##
                        (0.203) (0.292)
## Num. obs.
                       336
                                 336
## Adj. R^2 (full model) 0.889 0.889
## Adj. R^2 (proj model) -0.120 -0.120
## Num. groups: state
                        48
## SE of `Model 1` is "Heteroskedasticity-Robust", while one of `Model 2` is "Cluster-Robust."
```

Panel + IV

Panel Data with Instrumental Variables

- Revisit the demand for Cigaretts
- Consider the following model

$$\log(Q_{it}) = eta_0 + eta_1 \log(P_{it}) + eta_2 \log(income_{it}) + u_i + e_{it}$$

where

- $\circ \ Q_{it}$ is the number of packs per capita in state i in year t,
- $\circ \; P_{it}$ is the after-tax average real price per pack of cigarettes, and
- $\circ income_{it}$ is the real income per capita. This is demand shifter.

- As an IV for the price, we use the followings:
 - \circ $SalesTax_{it}$: the proportion of taxes on cigarettes arising from the general sales tax.
 - Relevant as it is included in the after-tax price
 - Exogenous(indepndent) since the sales tax does not influence demand directly, but indirectly through the price.
 - $\circ \ CigTax_{it}$: the cigarett-specific taxes

```
# load the data set and get an overview
library(AER)
data("CigarettesSW")
CigarettesSW %>%
  mutate( rincome = (income / population) / cpi) %>%
  mutate( rprice = price / cpi ) %>%
  mutate( salestax = (taxs - tax) / cpi ) %>%
  mutate( cigtax = tax/cpi ) -> Cigdata
```

• Run IV regression with panel data.

```
# OLS
result_1 <- felm( log(packs) ~ log(rprice) + log(rincome) | 0 | 0 | state, data = Cigdata )
# State FF
result_2 <- felm( log(packs) ~ log(rprice) + log(rincome) | state | 0 | state, data = Cigdata )
# TV without FF
result_3 <- felm( log(packs) ~ log(rincome) | 0 | (log(rprice) ~ salestax + cigtax) |
                     state, data = Cigdata )
# TV with FF
result 4 <- felm( log(packs) ~ log(rincome) | state | (log(rprice) ~ salestax + cigtax) |
                     state, data = Cigdata )
screenreg(l = list(result_1, result_2, result_3, result_4), digits = 3,
          custom.coef.names = NULL, # add a class, if you want to change the names of variables.
          include.ci = F.
          include.rsguared = FALSE, include.adjrs = TRUE, include.nobs = TRUE,
          include.pvalues = FALSE, include.df = FALSE, include.rmse = FALSE,
          robust = T, # robust standard error
          custom.header = list("log(packs)" = 1:4),
         stars = numeric(0))
```

```
##
                                 log(packs)
##
##
                      Model 1 Model 2 Model 3 Model 4
##
                      10.067
                                       9.736
## (Intercept)
##
                      (0.464)
                              (0.555)
## log(rprice)
                      -1.334 -1.210
##
                      (0.174) (0.143)
                      0.318
                             0.121
## log(rincome)
                                       0.257 0.204
##
                      (0.212) (0.218) (0.204) (0.238)
## `log(rprice)(fit)`
                                      -1.229
                                              -1.268
##
                                      (0.183) (0.162)
## Num. obs.
             96 96
                                      96
                                              96
## Adj. R^2 (full model) 0.542
                            0.929 0.539 0.929
## Adj. R^2 (proj model) 0.542 0.793
                                       0.539
                                             0.792
## Num. groups: state
```

felm command

Report heteroskedasticity robust standard error

```
# Run felm command without specifying cluster.
result_1 <- felm( log(packs) ~ log(rprice) + log(rincome) | 0 | 0 | state, data = Cigdata )
screenreg(l = list(result_1),
         digits = 3.
          # caption = 'title',
          custom.model.names = c(" Model 1 "),
          custom.coef.names = NULL, # add a class, if you want to change the names of variables.
          include.ci = T,
          include.rsguared = FALSE, include.adjrs = TRUE, include.nobs = TRUE,
          include.pvalues = FALSE, include.df = FALSE, include.rmse = FALSE,
          robust = F, # robust standard error
          custom.header = list("log(packs)" = 1),
          stars = numeric(0),
# stargazer::stargazer(result 1, type = "text",
                      se = list(result 1$rse ) )
```

```
##
##
                           log(packs)
##
##
                            Model 1
## (Intercept)
                          10.067
##
                         (0.464)
## log(rprice)
                         -1.334
##
                         (0.174)
## log(rincome)
                         0.318
##
                          (0.212)
## Num. obs.
                          96
## Adj. R^2 (full model) 0.542
## Adj. R^2 (proj model) 0.542
```

How to conduct F test after felm

49,22642

```
# Run felm command without specifying cluster.
result_1 <- felm( packs ~ rprice + rincome | 0 | 0 | 0, data = Cigdata )
# The following tests H0: _b[rincome] = 0 & _b[rprice] = 0
ftest1 = waldtest(result_1, ~ rincome | rprice )
ftest1
            p chi2 df1
                                                                        df2
##
                                               p.F F
## 4.180596e-22 9.845284e+01 2.000000e+00 2.621701e-15 4.922642e+01 9.300000e+01
## attr(,"formula")
## ~rincome | rprice
## <environment: 0x0000000313c7f18>
# ftest[5] corresponds to F-value
fval1 = ftest1[5]
fval1
##
```

```
# The following tests HO: _b[rincome] - 1 = 0 & _b[rprice] = 0
ftest2 = waldtest(result_1, ~ rincome - 1 | rprice )
ftest2
##
                     chi2 df1
                                              p.F F
                                                                      df2
## 2.048665e-24 1.090897e+02 2.000000e+00 2.121544e-16 5.454485e+01 9.300000e+01
## attr(,"formula")
## ~rincome - 1 | rprice
## <environment: 0x000000030a984e0>
# ftest[5] corresponds to F-value
fval2 = ftest1[5]
fval2
## F
## 49.22642
```