# POLS/CS&SS 503: Assignment 3

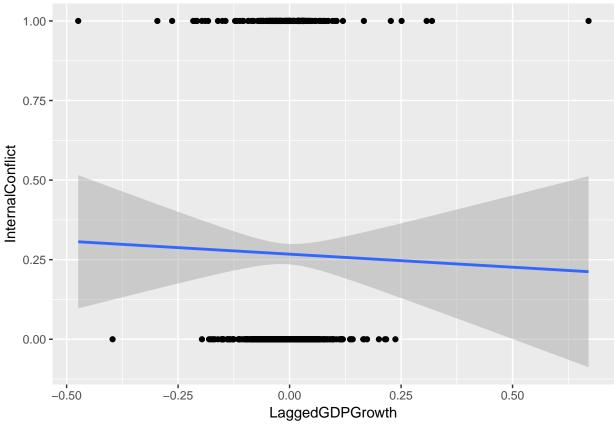
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# Setup

```
library("tidyverse")
## Loading tidyverse: ggplot2
## Loading tidyverse: tibble
## Loading tidyverse: tidyr
## Loading tidyverse: readr
## Loading tidyverse: purrr
## Loading tidyverse: dplyr
## Conflicts with tidy packages ------
## filter(): dplyr, stats
## lag():
            dplyr, stats
library("haven")
library("AER")
## Loading required package: car
## Attaching package: 'car'
## The following object is masked from 'package:dplyr':
##
##
      recode
## The following object is masked from 'package:purrr':
##
##
      some
## Loading required package: lmtest
## Loading required package: zoo
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##
      as.Date, as.Date.numeric
## Loading required package: sandwich
## Loading required package: survival
library("Formula")
```

## 0.1 Problem 1

```
Bailey (2016) Ex 9.1
RainIV <- read_csv( "Ch09.Ex1.CivilWarRainInstrumentData/RainIV.csv")</pre>
## Parsed with column specification:
## cols(
##
     .default = col_integer(),
     country_name = col_character(),
##
##
     country_code = col_character(),
##
     GPCP = col_double(),
##
     RainfallGrowth = col_double(),
##
     LaggedRainfallGrowth = col_double(),
##
     pop = col_double(),
##
    lpopl1 = col_double(),
##
    Mountains = col_double(),
##
    lmtnest = col_double(),
##
    EthnicFrac = col_double(),
##
    ReligiousFrac = col_double(),
    GDPGrowth = col_double(),
    LaggedGDPGrowth = col_double(),
##
    InitialGDP = col_double()
## )
## See spec(...) for full column specifications.
ggplot(RainIV, aes(y = InternalConflict, x = LaggedGDPGrowth)) +
  geom_point() +
  geom_smooth(method = "lm")
```



mod1\_f <- InternalConflict ~ LaggedGDPGrowth
mod1 <- lm(mod1\_f, data = RainIV)
summary(mod1)</pre>

```
##
## Call:
## lm(formula = mod1_f, data = RainIV)
##
## Residuals:
##
      Min
               1Q Median
                               3Q
                                      Max
##
  -0.2999 -0.2689 -0.2660 0.7228 0.7876
##
## Coefficients:
                   Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                   0.26738
                              0.01631 16.389
                                                <2e-16 ***
## LaggedGDPGrowth -0.08206
                              0.22485 -0.365
                                                 0.715
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.4434 on 741 degrees of freedom
## Multiple R-squared: 0.0001797, Adjusted R-squared: -0.00117
## F-statistic: 0.1332 on 1 and 741 DF, p-value: 0.7152
```

**b.** Add controls for initial GDP (InitialGDP), democracy (') mountains, and ethnic and religious fractionalization to the model.

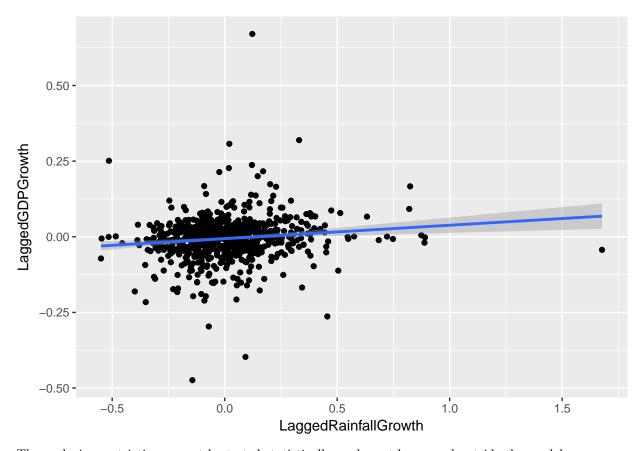
```
mod2 <- lm(mod2_f, data = RainIV)</pre>
summary(mod2)
##
## Call:
## lm(formula = mod2_f, data = RainIV)
##
## Residuals:
##
      Min
               1Q Median
                               3Q
                                      Max
## -0.5654 -0.2811 -0.2221 0.4570 0.9459
##
## Coefficients:
                    Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                   0.0703555 0.0731012
                                          0.962 0.33614
## LaggedGDPGrowth -0.1087977 0.2200999
                                        -0.494 0.62123
## InitialGDP
                  -0.0569091 0.0182258 -3.122 0.00186 **
## Democracy
                   0.0012242 0.0028894
                                          0.424 0.67193
                   0.0038654 0.0009527
                                          4.057 5.49e-05 ***
## Mountains
## EthnicFrac
                   0.3247931 0.0918181
                                          3.537 0.00043 ***
## ReligiousFrac
                   0.0105162 0.0958907
                                          0.110 0.91270
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4334 on 736 degrees of freedom
## Multiple R-squared: 0.05106,
                                   Adjusted R-squared: 0.04332
## F-statistic: 6.6 on 6 and 736 DF, p-value: 8.276e-07
```

No. The coefficient on  $\beta$  would only be a causal estimate under an assumption of no omitted confounders. This seems highly unlikely.

- ${\bf c.}$  The two conditions needed for a good instrument are
  - inclusion restriction: Rainfall must be correlated with economic growth
  - exclusion restriction: Rainfall must not be correled with war except through its effect on economic growth.

The inclusion restriction can be tested (in the first stage):

```
ggplot(RainIV, aes(x = LaggedRainfallGrowth, y = LaggedGDPGrowth)) +
geom_point() +
geom_smooth(method = "lm")
```



The exclusion restriction cannot be tested statistically, and must be argued outside the model.

d. Instrumenting GDP growth with rainfall randomly assigns some part of GDP growth to countries.

```
mod3_f <- InternalConflict ~ LaggedGDPGrowth + InitialGDP + Democracy + Mountains + EthnicFrac + Religi
mod3 <- ivreg(mod3_f, data = RainIV)</pre>
summary(mod3)
##
## Call:
## ivreg(formula = mod3_f, data = RainIV)
##
## Residuals:
##
                1Q Median
                                 3Q
                                        Max
  -1.1693 -0.3106 -0.1897
                            0.4203
##
##
## Coefficients:
##
                    Estimate Std. Error t value Pr(>|t|)
                    0.062506
                                0.077268
                                           0.809 0.418802
## (Intercept)
## LaggedGDPGrowth -2.063153
                                1.845106
                                         -1.118 0.263857
## InitialGDP
                   -0.058080
                                0.019209
                                          -3.024 0.002584 **
                    0.002361
                                0.003221
## Democracy
                                           0.733 0.463785
## Mountains
                    0.004069
                                0.001020
                                           3.988 7.34e-05 ***
## EthnicFrac
                    0.328851
                                0.096686
                                           3.401 0.000707 ***
## ReligiousFrac
                    0.004724
                                0.101042
                                           0.047 0.962721
```

## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

## ---

```
##
## Residual standard error: 0.456 on 736 degrees of freedom
## Multiple R-Squared: -0.05059,
                                    Adjusted R-squared: -0.05916
## Wald test: 6.133 on 6 and 736 DF, p-value: 2.748e-06
```

The coefficient on LaggedGDPGrowth is 20 times larger than in the OLS regression. It is still not stastistically significant.

```
f. Redo the 2SLS with country fixed effects.
mod4_f <- InternalConflict ~ LaggedGDPGrowth + InitialGDP + Democracy + Mountains + EthnicFrac + Religi
mod4 <- ivreg(mod4_f, data = RainIV)</pre>
summary(mod4)
##
## Call:
## ivreg(formula = mod4_f, data = RainIV)
## Residuals:
       Min
                 1Q
                      Median
                                    3Q
                                           Max
## -1.60872 -0.18282 -0.01501 0.13649
## Coefficients:
##
                   Estimate Std. Error t value Pr(>|t|)
                   -0.752079 0.473516 -1.588 0.11267
## (Intercept)
## LaggedGDPGrowth -2.853380
                              1.535631 -1.858 0.06357
## InitialGDP
                  -0.011809
                              0.072129 -0.164 0.86999
                             0.003276
                                        0.325 0.74518
## Democracy
                   0.001065
## Mountains
                   0.098156
                             0.017018
                                        5.768 1.21e-08 ***
                              0.376584
                                         1.096 0.27326
## EthnicFrac
                   0.412905
## ReligiousFrac
                   0.749920
                              0.871221
                                         0.861 0.38966
## country_codeBDI -6.579658
                              1.236583
                                        -5.321 1.39e-07 ***
## country_codeBEN 0.166612
                              0.116083
                                         1.435 0.15165
## country_codeBFA
                   0.234677
                              0.134340
                                         1.747 0.08110
## country_codeBWA
                   0.251482
                              0.166556
                                         1.510 0.13152
## country_codeCAF -0.580915
                              0.217323
                                        -2.673 0.00769 **
## country_codeCIV -0.114346
                              0.165163
                                        -0.692
                                                0.48896
## country_codeCMR -1.768851
                              0.200940
                                        -8.803
                                                < 2e-16 ***
## country_codeCOG 0.289731
                              0.144290
                                         2.008 0.04503 *
## country_codeDJI 0.101831
                              0.358274
                                         0.284 0.77632
## country_codeETH -6.086221
                              1.112293 -5.472 6.21e-08 ***
## country_codeGAB
                   0.162233
                              0.283837
                                         0.572 0.56780
## country_codeGHA 0.073439
                              0.223433
                                         0.329 0.74249
## country_codeGIN 0.064057
                              0.263453
                                         0.243 0.80796
                                         1.162 0.24573
## country_codeGMB
                   0.334810
                              0.288195
## country_codeGNB
                   0.123575
                              0.122652
                                         1.008
                                                0.31403
## country_codeKEN -2.632596
                              0.305915 -8.606 < 2e-16 ***
## country_codeLBR 0.104590
                              0.173242
                                         0.604 0.54622
## country_codeLSO -7.567227
                              1.466494
                                        -5.160 3.22e-07 ***
## country_codeMDG -2.976911
                              0.558717
                                        -5.328 1.34e-07 ***
## country_codeMLI
                  0.368259
                              0.321301
                                         1.146 0.25213
## country_codeMOZ
                                         2.915 0.00367 **
                   0.423353
                              0.145244
## country_codeMRT
                              0.418388
                                         1.511 0.13122
                   0.632220
## country_codeMWI -0.812187
                              0.124855
                                        -6.505 1.48e-10 ***
## country_codeNAM -0.548080
                              0.272637 -2.010 0.04478 *
## country_codeNER 0.266671
                              0.216693
                                         1.231 0.21887
```

```
## country_codeNGA -0.310737
                              0.118896 -2.614 0.00915 **
## country_codeRWA -6.395050
                              1.222877
                                       -5.230 2.25e-07 ***
## country codeSDN 0.416879
                              0.158764
                                         2.626 0.00883 **
## country_codeSEN
                                         2.404
                                                0.01646 *
                  0.726010
                              0.301963
## country_codeSLE 0.239721
                              0.107686
                                         2.226
                                                0.02632 *
## country_codeSOM 0.447733
                              0.598064
                                         0.749 0.45433
## country codeSWZ -1.232084
                              0.248451
                                        -4.959 8.90e-07 ***
## country_codeTCD -0.034781
                              0.111964
                                        -0.311 0.75616
## country_codeTGO 0.201914
                              0.117503
                                         1.718 0.08617 .
## country_codeTZA -2.216959
                              0.288023
                                       -7.697 4.75e-14 ***
## country_codeUGA -0.095200
                              0.162157 -0.587 0.55734
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3709 on 700 degrees of freedom
## Multiple R-Squared: 0.3391, Adjusted R-squared: 0.2995
## Wald test: 13.55 on 42 and 700 DF, p-value: < 2.2e-16
```

The coefficient on LaggedGDPGrowth has increased by 40% (-2 to -2.8) and now has a p-value of 0.06. These regressions use only variation within each country and thus remove the country-level unobserved variables that may be correlated with economic growth.

f. Regression LaggedGDPGrowth on rainfall and controls:

```
mod_f1 <- lm(LaggedGDPGrowth ~ LaggedRainfallGrowth + InitialGDP + Democracy + Mountains + EthnicFrac +
```

Save the residuals. I use broom::augment for convenience, but still need to add InternalConflict since it wasn't in the original data.

```
data_f <- broom::augment(mod_f1) %>%
  left_join(select(RainIV, country_code, InternalConflict), by = "country_code")
mod_f2 <- lm(InternalConflict ~ .resid + LaggedGDPGrowth + InitialGDP + Democracy + Mountains + EthnicFccef(mod_f2)["LaggedGDPGrowth"]
## LaggedGDPGrowth</pre>
```

The coefficient is the same as that in the 2SLS.

It controls for endogeneity by controlling for the part of LaggedGDPGrowth not explained by rainfall growth (i.e. the non-random part of GDP growth).

#### 0.2 Problem 2

-6.536277e-14

Bailey (2016) Ex 9.2

## 0.3 Problem 3

Bailey (2016) Ex 9.4

#### 0.4 Problem 4

Bailey (2016) Ex 9.5

## References

Bailey, Michael A. 2016. Real Stats: Using Econometrics for Political Science and Public Policy. Oxford University Press.