

A3

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```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.0
## v ggplot2 3.3.2      v purrr  0.3.4
## v tibble  3.0.3      v dplyr  1.0.2
## v tidyr   1.1.2      v stringr 1.4.0
## v readr   1.3.1      v forcats 0.5.0

## -- Conflicts ----- tidyverse_conflict_
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()
```

```
library(xts)
```

```
## Loading required package: zoo

##
## Attaching package: 'zoo'

## The following objects are masked from 'package:base':
##
##   as.Date, as.Date.numeric

##
## Attaching package: 'xts'

## The following objects are masked from 'package:dplyr':
##
##   first, last
```

```
library(plm)
```

```
##
## Attaching package: 'plm'

## The following objects are masked from 'package:dplyr':
##
##   between, lag, lead
```

Exercise 1 Links to the datasets

```
population <- read.csv("https://www.dropbox.com/s/s38cde88670y5mw/population.csv?dl=1")
crime_long <- read.csv("https://www.dropbox.com/s/t3vushurhm3s5my/crime_long.csv?dl=1")
officers <- read.csv("https://www.dropbox.com/s/8q2fpdb7phy86m8/officers.csv?dl=1")
```

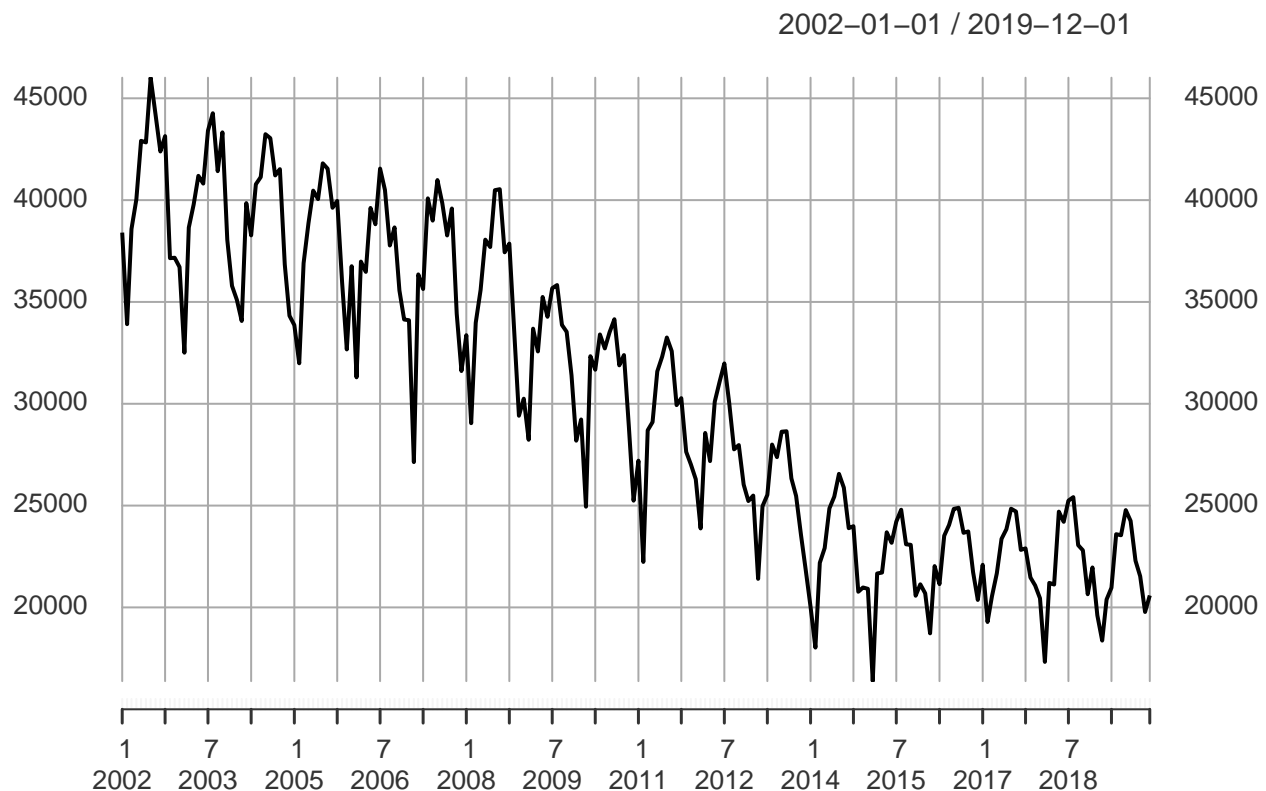
Exercise 2 Data Manipulation

Calculate total crime per month and plot the time series of crime

```
# total crime per month
crime_monthly <- crime_long %>%
  group_by(crime_month) %>%
  summarize(crime_by_month = sum(crimes))

## `summarise()` ungrouping output (override with `.groups` argument)

# plot time series of crime
crime_monthly_xts <- xts(crime_monthly$crime_by_month, as.Date(crime_monthly$crime_month, format='%Y-%m-%d'))
plot(crime_monthly_xts, type = 'l', main = '')
```



Merge the two datasets by districts-units and period

```
crime_population <- merge(crime_long, population, by.x = c("district", "crime_month"), by.y = c("district", "month"))
```

Construct a panel data of unit over time with the following variables

– Total crimes per resident – Violent crimes per resident – Property crimes per resident – Median income – Share of black, Hispanic, and white residents

```
panel_data <- crime_population %>%
  mutate(
    violent_crimes = case_when(
      crime_type == "violent" ~ crimes,
      TRUE ~ 0L
    ),
```

```

    property_crimes = case_when(
      crime_type == "property" ~ crimes,
      TRUE ~ 0L
    )
  ) %>%
  group_by(district, crime_month) %>%
  summarize(
    total_crimes = sum(crimes),
    violent_crimes = sum(violent_crimes),
    property_crimes = sum(property_crimes),
    median_income = p50_inc,
    share_of_black = tot_black/tot_pop,
    share_of_hisp = tot_hisp/tot_pop,
    share_of_white = tot_white/tot_pop
  ) %>%
  distinct()

```

`summarise()` regrouping output by 'district', 'crime_month' (override with `.groups` argument)

Exercise 3 Panel Data: Introduction

```

df <- merge(officers, panel_data, by.x = c("month", "unit"), by.y=c("crime_month", "district"), all.x=TRUE, all.y=TRUE)
panel_df <- pdata.frame(df, index=c("NUID", "month"))

```

```

# pooled model
pooled <- plm(
  formula = arrest ~ tenure + total_crimes + median_income + share_of_black + share_of_hisp + share_of_white,
  data = panel_df,
  model = "pooling"
)
# estimator
# beta
pooled$coefficients[2]

```

```

##      tenure
## -4.16066e-06

```

```

# gamma
pooled$coefficients[3:7]

```

```

##    total_crimes  median_income share_of_black  share_of_hisp share_of_white
##    2.229548e-07   1.617511e-08  -8.101603e-03  -5.362641e-03  -1.207314e-02

```

```

# we can also use ols
lm_pooled <- lm(
  formula = arrest ~ tenure + total_crimes + median_income + share_of_black + share_of_hisp + share_of_white,
  data = df
)
# estimators
# beta
lm_pooled$coefficients[2]

```

```

##      tenure
## -4.16066e-06

```

```
# gamma
lm_pooled$coefficients[3:7]

##      total_crimes  median_income share_of_black  share_of_hisp share_of_white
##      2.229548e-07   1.617511e-08  -8.101603e-03  -5.362641e-03  -1.207314e-02
```

Exercise 4 Panel Data: More controls

```
# fixed effect model
fe1 <- plm(
  formula = arrest ~ tenure + total_crimes + median_income + share_of_black + share_of_hisp + share_of_white,
  effect = "time",
  data = panel_df,
  model = "within"
)
# estimators
# beta
fe1$coefficients[1]
```

```
##      tenure
## -3.809782e-06
```

```
# gamma
fe1$coefficients[2:6]

##      total_crimes  median_income share_of_black  share_of_hisp share_of_white
## -6.320360e-06  -4.910055e-07  -9.200883e-02  -1.398480e-01  -1.012976e-01
```

```
# psi
fe1$coefficients[7:30]

## factor(unit)2 factor(unit)3 factor(unit)4 factor(unit)5 factor(unit)6
## -2.437791e-02 -2.001211e-02 -3.632805e-05 -1.580178e-02 -1.385754e-02
## factor(unit)7 factor(unit)8 factor(unit)9 factor(unit)10 factor(unit)11
## -2.108532e-02  2.342640e-02  3.109578e-03  7.681503e-03  -1.133599e-02
## factor(unit)12 factor(unit)13 factor(unit)14 factor(unit)15 factor(unit)16
##  4.412439e-04  -3.063905e-03  2.440896e-02  -2.027399e-02  7.288093e-03
## factor(unit)17 factor(unit)18 factor(unit)19 factor(unit)20 factor(unit)21
##  1.080354e-03  4.016511e-03  8.935392e-03  -1.544846e-02  -3.671653e-02
## factor(unit)22 factor(unit)23 factor(unit)24 factor(unit)25
## -6.345306e-04  -1.097174e-02  -1.539638e-02  2.202629e-02
```

```
# kappa
fixef(fe1)

## 2007-03-01 2007-04-01 2007-05-01 2007-06-01 2007-07-01 2007-08-01 2007-09-01
##      0.64160      0.63174      0.64581      0.62099      0.63370      0.62689      0.63900
## 2007-10-01 2007-11-01 2007-12-01 2008-01-01 2008-02-01 2008-03-01 2008-04-01
##      0.64170      0.63944      0.62765      0.62037      0.62824      0.62913      0.64405
## 2008-05-01 2008-06-01 2008-07-01 2008-08-01 2008-09-01 2008-10-01 2008-11-01
##      0.65507      0.63722      0.63615      0.63721      0.64554      0.63282      0.63463
## 2008-12-01 2009-01-01 2009-02-01 2009-03-01 2009-04-01 2009-05-01 2009-06-01
##      0.64711      0.63017      0.63039      0.62728      0.63126      0.63463      0.64038
## 2009-07-01 2009-08-01 2009-09-01 2009-10-01 2009-11-01 2009-12-01 2010-01-01
##      0.64164      0.63151      0.63138      0.63459      0.63096      0.62629      0.63799
## 2010-02-01 2010-03-01 2010-04-01 2010-05-01 2010-06-01 2010-07-01 2010-08-01
##      0.63979      0.62729      0.64325      0.63276      0.63030      0.62264      0.64605
```

```
## 2010-09-01 2010-10-01 2010-11-01 2010-12-01 2011-01-01 2011-02-01 2011-03-01
##      0.63839      0.63362      0.62700      0.63141      0.62797      0.63078      0.63162
## 2011-04-01 2011-05-01 2011-06-01 2011-07-01 2011-08-01 2011-09-01 2011-10-01
##      0.63498      0.63564      0.63629      0.64018      0.63607      0.62483      0.64667
## 2011-11-01 2011-12-01 2012-01-01 2012-02-01 2012-03-01 2012-04-01 2012-05-01
##      0.62920      0.63646      0.62572      0.63915      0.63799      0.64075      0.64053
## 2012-06-01 2012-07-01 2012-08-01 2012-09-01 2012-10-01 2012-11-01 2012-12-01
##      0.63411      0.63685      0.63105      0.63849      0.62160      0.63964      0.61930
## 2013-01-01 2013-02-01 2013-03-01 2013-04-01 2013-05-01 2013-06-01 2013-07-01
##      0.63997      0.62809      0.62032      0.63968      0.65028      0.64323      0.65370
## 2013-08-01 2013-09-01 2013-10-01 2013-11-01 2013-12-01 2014-01-01 2014-02-01
##      0.63344      0.63921      0.62747      0.63036      0.63949      0.62121      0.63867
## 2014-03-01 2014-04-01 2014-05-01 2014-06-01 2014-07-01 2014-08-01 2014-09-01
##      0.63173      0.61755      0.63027      0.63441      0.64175      0.63506      0.63723
## 2014-10-01 2014-11-01 2014-12-01 2015-01-01 2015-02-01 2015-03-01 2015-04-01
##      0.63444      0.62763      0.62915      0.63125      0.62686      0.64033      0.63265
## 2015-05-01 2015-06-01 2015-07-01 2015-08-01 2015-09-01 2015-10-01 2015-11-01
##      0.62945      0.62906      0.62754      0.63083      0.63516      0.62784      0.63849
## 2015-12-01 2016-01-01 2016-02-01 2016-03-01 2016-04-01 2016-05-01 2016-06-01
##      0.62880      0.63083      0.62907      0.62164      0.64599      0.64603      0.63103
## 2016-07-01 2016-08-01 2016-09-01 2016-10-01 2016-11-01 2016-12-01 2017-01-01
##      0.62535      0.61696      0.63463      0.63929      0.62219      0.62327      0.63626
## 2017-02-01 2017-03-01 2017-04-01 2017-05-01 2017-06-01 2017-07-01 2017-08-01
##      0.63033      0.64223      0.63043      0.64458      0.62398      0.63005      0.63351
## 2017-09-01 2017-10-01 2017-11-01 2017-12-01 2007-01-01 2007-02-01
##      0.62768      0.62649      0.61933      0.62725      0.63617      0.63892
```

Exercise 5 Panel Data: Individual fixed effects

Implement a within, between, and first difference estimator for the parameter β . Then, compare the estimated values.

```
# within
fe2 <- plm(
  formula = arrest ~ tenure + total_crimes + median_income + share_of_black + share_of_hisp + share_o
  effect = "twoway",
  data = panel_df,
  model = "within"
)
```

```
# between
fe3 <- plm(
  formula = arrest ~ tenure + total_crimes + median_income + share_of_black + share_of_hisp + share_o
  effect = "individual",
  data = panel_df,
  model = "between"
)
```

```
# fd
fe4 <- plm(
  formula = arrest ~ tenure + total_crimes + median_income + share_of_black + share_of_hisp + share_o
  effect = "individual",
  data = panel_df,
  model = "fd"
)
```

```
# compare beta
est_betas <- c(fe2$coefficients[1], fe3$coefficients[2], fe4$coefficients[2])
names(est_betas) <- c("within", "between", "fd")
est_betas
```

```
##           within           between           fd
## -2.767569e-04 -1.754975e-05  5.430658e-03
```

within and between estimated beta estimator are both negative, first difference estimated beta is positive.

Use a GMM approach to estimate all parameters (including fixed effects) in one step.

one-step GMM:

$$\hat{\beta}_{2SLS} = [X'Z(Z'Z)^{-1}Z'X]^{-1}X'Z(Z'Z)^{-1}Z'y$$

if $X = Z$, it is just estimator of OLS.