W271 Live Session 11: Analysis of Panel Data 1

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Main topics covered in Week 12 (Async Unit 11)

- Introduction to panel data
- Using OLS regression model on panel data
- Exploratory panel data analysis
- Unobserved effect models
- Pooled OLS models
- First-Difference models
- Distributed Lag models

Readings

W2016: Jeffrey Wooldridge. *Introductory Econometrics: A Modern Approach.* 6th edition. Cengage Learning

- Ch. 13 (skip 13.4)
- [package plm](https://cran.r-project.org/web/packages/plm/plm.pdf)
- [plm vignettes](https://cran.r-project.org/web/packages/plm/vignettes/plm.pdf)

Agenda for this week's live session:

- 1. Questions about ARIMA and SARIMA models
- 2. Overview of the next unit: Panel data methods
- 3. EDA of panel and grouped data

Individual-level regression, pooled crossing regression, and first-difference regression

Some start-up codes:

```
#sessionInfo()

# Insert the function to *tidy up* the code when they are printed out
library(knitr)
opts_chunk$set(tidy.opts=list(width.cutoff=60),tidy=TRUE)

# Set Numeric Value Display
# See reference from https://stat.ethz.ch/R-manual/R-devel/library/base/html/options.html
options(digits=2) # Set the printed number of digits to 2. Note: It is a suggestion only. Default is 7.
#options("scipen" = 10)

# Set memory limit
memory.limit(500000000)
```

Warning: 'memory.limit()' is Windows-specific

```
## [1] Inf
# Clean up the workspace before we begin
rm(list = ls())
# Set working directory
wd <- "~/Documents/Projects/MIDS/Summer 2017/live_sessions/week11/"
setwd(wd)
# Load libraries
library(car)
library(dplyr)
##
## Attaching package: 'dplyr'
## The following object is masked from 'package:car':
##
##
       recode
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
       intersect, setdiff, setequal, union
##
library(Hmisc)
## Loading required package: lattice
## Loading required package: survival
## Loading required package: Formula
## Loading required package: ggplot2
##
## Attaching package: 'Hmisc'
## The following objects are masked from 'package:dplyr':
##
##
       combine, src, summarize
## The following objects are masked from 'package:base':
##
       format.pval, round.POSIXt, trunc.POSIXt, units
##
library(ggplot2)
library(lattice)
library(plm)
## Attaching package: 'plm'
## The following objects are masked from 'package:dplyr':
##
##
       between, lag, lead
```

1. Exploratory data analysis of panel data

- Estmated Time: Breakout session 25 minutes
- Estmated Time: Class discussion 20 minutes

In this example, we use five waves of data from "National Youth Survey". Each year, participants in the age of 11, 12, 13, 14, 15 filled out a nine-item instrument designed to assess their tolerance of deviant behiavor such as cheat on tests, sell hard drugs, etc. Response to each item is provided in a 4-point scale (1 = very wrong, 2 = wrong, 3 = a little bit wrong, and 4 = not wrong at all). At each occasion (i.e. wave), the outcome, TOL, is computed as the respondent's average across the nine responses. Two potential explanatory variables in this dataset include male, which is equal to 1 if the respondent is male, and exposure, which is a respondent's estimated proportion of their close friends who were involved in each of the nine activities, measuring on a 5-point scale (0 = none, 5 = all).

Task 1: - Import the data as data.frame. - Examine the basic structure of each of the datasets. - Print the person-level dataset. Discuss it. - Also, answer "how many male and female in the dataset?" - Construct the correlation matrix of the variables tol11 - tol15. Discuss what you observe in this matrix. Do the correlation values in the matrix make sense? Why? Why not? - Print the person-period level dataset. Discuss it. - How is the person-period level dataset different from that of the person-level dataset? - Does the person-level dataset contain exactly the same information as that in the person-period-level dataset as far as tolerance, male, and exposure variables are concern? Why? Why not? - Conduct a throughout EDA using ther person-period level dataset

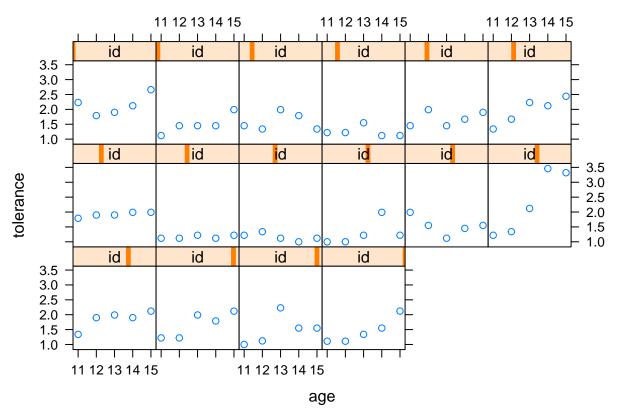
```
# YOUR CODE HERE
df <- read.table("tol_person.txt", sep = ",", header = T)

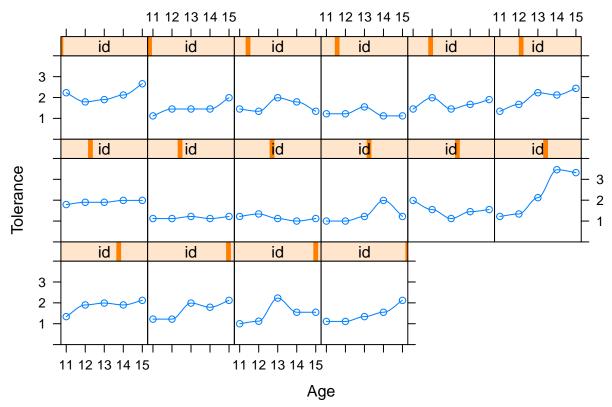
df2 <- read.table("tol_person_period.txt", sep = ",", header = T)</pre>
```

2. Growth-curve Analysis (something not covered in the book and the async)

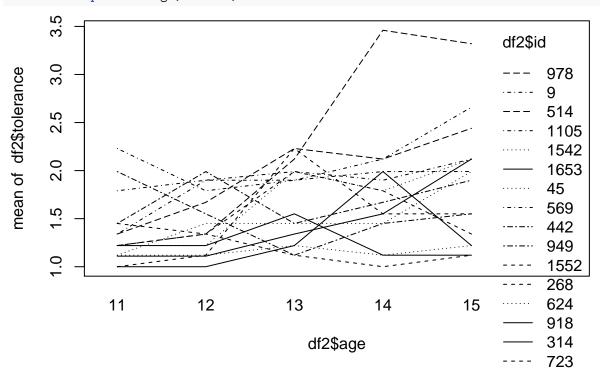
• Estmated Time: Instructor's teaching 10 minutes

```
xyplot(tolerance ~ age | id, data = df2, as.table = T)
```





plot of the raw data
interaction.plot(df2\$age, df2\$id, df2\$tolerance)



3. Build regression models to answer "Does being exposed to friends having deviant behavior increase one's tolerance of deviant behavior?"

- Estmated Time: Breakout session 20 minutes
- Estmated Time: Class discussion 15 minutes

Task 1: Estimate individual-level regression models. Remember that potential explanatory variables include *time*, *male*, and *exposure*. Can you use all of these variables in the regression models? Why? Why not? Does this regression help us answer the question? Why? Why not?

Task 2: Answer the question using a pooled-OLS appoach. Interpret the model results.

Task 3: Answer the question using a first-difference approach. Interpret the model results. Conduct EDA on both your dependent variable and explanatory variables first.

YOUR CODE HERE