

INTRODUCTION TO R PROGRAMMING

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Outline

- ① Getting started
- ② Data structures
 - Vectors and matrices
 - Data frame
- ③ Data import-export
- ④ Data manipulation using dplyr/tidyverse

What is R? I

- R is a programming language
- R is an environment for statistical computing and graphics
- R is free and open source (since 1995)
- R has LOTS of packages (+1300 on CRAN, +1600 on Bioconductor, ??? on Github)
- R is used in industry: Google, IBM, HP, Microsoft, Oracle etc. See <https://www.r-consortium.org/members>
- R plays well with other programming languages

What is RStudio? I

- RStudio is an *integrated development environment* (IDE) for R
- There are many IDEs (and graphical user interfaces (GUIs)) for R such as Rcmdr, Emacs+ESS, Revolution-R, Tinn-R, Eclipse, JGR, ...
- RStudio is also a company :(. It is the company providing Rstudio IDE
- RStudio is available in two versions: open source and commercial
- RStudio has good community and commercial support

R resources I

- <http://cran.r-project.org/web/views/>
- Official (and difficult): <http://cran.r-project.org/manuals.html>
- Contributed documentation (customized):
<http://cran.r-project.org/other-docs.html>
- Books: <http://www.r-project.org/doc/bib/R-books.html>
- Help: Stackoverflow and (lots of) mailing lists
<https://www.r-project.org/mail.html>

Working with R I

Standard arithmetic operators: +, -, *, /, and ^ :

```
(2 + 3*5 - 3^2 )/5
```

```
## [1] 1.6
```

```
2^4
```

```
## [1] 16
```

Use **Ctrl-Enter** to run selected lines or current line

Mathematical functions: R has `log()`, `exp()`, `sqrt()`, `log()`, `abs()`, `min()`, `max()`, `sin()`, `cos()`, `tan()`, `sign()`, ...

```
# log() # by default base = e
```

```
log(10, base=2) + sin(pi/4)
```

Working with R II

```
## [1] 4.03
```

Getting help: Google it! But also:

```
help("solve") # help(solve) works as well
```

```
? "solve"      # ?solve works as well
```

```
? "for"       # ?for does not work
```

```
help.search("binomial distribution") # help(solve) works as well
```

Finding, installing and removing packages:

- Packages : <http://cran.r-project.org/web/packages/>
- Task views: <http://cran.r-project.org/web/views/>
- To install/remove a package:

```
install.packages("dplyr", dep = TRUE)
```

```
install.packages(c("dplyr", "tidyr"), dep = TRUE)
```

```
remove.packages("fclust")
```

```
remove.packages(c("fclust", "pmatch"))
```


Working with R IV

Getting and setting working directory:

```
getwd()  
setwd("c:\\my work\\R")  
setwd("c:/my work/R")
```

In RStudio we can change:

- Working directory: Session → Set Working Directory → Choose directory
- Default working directory: Tools → Global Options → General → Default working directory

Calling an R script: If R commands are stored in a file, say `do.R` in `c:/my work/R`, the command to use is

```
source("c:/my work/R/do.R") # from anywhere  
source("do.R") # If I am already in "c:/my work/R"
```

Assignments I

Assignment operators:

- <- and = are both OK. But the former is somehow more popular!

```
x <- 5
```

```
y = 6
```

- An object's name can not begin with a number. Names are case sensitive.
- The following names are used by R; they should not be used as object name: **Inf, NA, NaN, NULL, TRUE, FALSE, break, else, for, function, if, in, next, repeat, return, while.**
- The following ones can be used but it is not recommended **c, q, t, C, D, I, diff, length, mean, pi, range, var.**

Saving and loading objects I

- Basic save and load process:

```
# history() # default is max.show=25
# history(max.show=Inf) # all history
# savehistory(file="myRsession.R") # default is ".Rhistory"
### save some variables
x=5; y=10
save(x, y, file = "xy.RData")
### save all
#save(list=ls(all=TRUE), file = "myRsession.RData")
save.image(file = "myRsession.RData") # default is ".RData"
load("myRsession.RData") # for loading back
```

- One problem with `save()` is it saves the objects and their names together. When `load()` loads a file saved by `save()` this may overwrite objects in memory already. Use `saveRDS()` and `readRDS()` for avoiding this danger.

Saving and loading objects II

- To see the current objects the command is

`objects()`

`ls()`

- To remove objects, say `x`, `y`, `z`, `foo` and `bar`, from the workspace we use the `rm()` command

`rm(x, y)`

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Data structures I

- 2 factors determine the type of the data structures in R
- dimension (1d, 2d or 3d+) and content (homogenous or heterogenous)

	Homogeneous	Heterogeneous
	-----	-----
1d	Vector	List
2d	Matrix	Data frame
nd	Array	

Vectors I

- Generation of vectors: Using combine function, `c()`.

```
x <- c(1.8, 3.14, 4)
```

```
2 * log(x) + 3
```

```
## [1] 4.18 5.29 5.77
```

```
x + c(10,10,0)
```

```
## [1] 11.8 13.1 4.0
```

- Generating vectors with a given pattern:

```
seq(from = 0, to = 1, by = 0.2)
```

```
## [1] 0.0 0.2 0.4 0.6 0.8 1.0
```

```
seq(0, 1, 0.2)
```

```
## [1] 0.0 0.2 0.4 0.6 0.8 1.0
```

Vectors II

```
# seq(from = 0, to = 1, length.out = 6)
```

```
# seq(0, 1, 6) ## !!!
```

```
s2 = c("a", "b", "c")
```

```
(s3 <- rep(s2, times=2))
```

```
## [1] "a" "b" "c" "a" "b" "c"
```

```
(s4 <- rep(s2, each=2))
```

```
## [1] "a" "a" "b" "b" "c" "c"
```

```
# (s5 <- 3:10)
```

```
# (s6=seq_along(s5))
```

- Generating random vectors:

```
set.seed(127) # To make below reproducible
```

```
v1 = rnorm(n=5, mean=3, sd=1) # default m=0, sd=1
```

```
v1
```

```
## [1] 2.43 2.19 2.51 3.00 3.82
```


Vectors III

```
v2 = runif(5,min=1,max=4)#default min=0,max=1
```

```
v2
```

```
## [1] 3.52 1.96 3.32 1.60 2.35
```

```
## random numbers from a given set
```

```
sample(1:10,size=10,replace=FALSE)
```

```
## [1] 4 7 2 10 1 3 8 5 9 6
```

```
sample(1:10,size=10,replace=TRUE)
```

```
## [1] 4 4 5 5 4 9 10 5 3 3
```

- Basic vector operations:

```
(v1 = 1:4)
```

```
## [1] 1 2 3 4
```

```
(v2 = 4:1)
```

```
## [1] 4 3 2 1
```

Vectors IV

```
(v3 <- 1:2)

## [1] 1 2

v1+v2

## [1] 5 5 5 5

v1+v3 ##Attention !!!

## [1] 2 4 4 6

v1*v2

## [1] 4 6 6 4

sum(v1*v2)

## [1] 20

v1 %*% v2# dot product: sum_i^n (v1_i*v2_i)

##      [,1]
## [1,]    20
```

Matrices I

- Matrices have two dimensions, rows and columns. A 2×3 matrix containing the elements 1:6, by column, is generated via

```
A <- matrix(1:6, nrow = 2) # tray also: matrix(1:6, ncol = 3)
```

A

```
##      [,1] [,2] [,3]
## [1,]    1    3    5
## [2,]    2    4    6
```

```
A2 <- matrix(1:6, nrow = 2, byrow=T)
```

A2

```
##      [,1] [,2] [,3]
## [1,]    1    2    3
## [2,]    4    5    6
```

Matrices II

- Solving a system of linear equations: Let's say, we want to solve the following linear system

$$5x + 7y = 1$$

$$4x + 3y + 2z = 2$$

$$6x - 2y - z = 3$$

```
A = matrix(c(5,7,0,4,3,2,6,-2,-1),nrow=3, byrow=T)
```

```
A
```

```
##      [,1] [,2] [,3]
## [1,]    5    7    0
## [2,]    4    3    2
## [3,]    6   -2   -1
```

Matrices III

```
b=1:3
x=solve(A,b)# solution of A*x=b
x

## [1]  0.487 -0.205  0.333

Ainv = solve(A) # A^(-1) = solve(A)
Ainv %*% b # A^(-1)*b should be equal to x

##      [,1]
## [1,]  0.487
## [2,] -0.205
## [3,]  0.333
```

Indexing - subsetting I

Vectors:

- Extract elements by their index.
- Exclude elements with negative index.

```
x
```

```
## [1]  0.487 -0.205  0.333
```

```
x[c(1, 4)]
```

```
## [1] 0.487    NA
```

```
x[-c(1, 4)]
```

```
## [1] -0.205  0.333
```

Matrices are vectors with an additional dimension attribute enabling row/column-type indexing

Indexing - subsetting II

- `A[i, j]` extracts element a_{ij} of matrix A .
- `A[i,]` extracts i th row.
- `A[, j]` extracts j th column.
- Results of these operations are *vectors*, i.e., dimension attribute is dropped (by default).
- `A[i, j, drop = FALSE]` avoids dropping and returns a matrix.

```
A[1:2, c(1, 3)]
```

```
##      [,1] [,2]  
## [1,]    5    0  
## [2,]    4    2
```

```
A[-(1:2), c(1, 3)]
```

```
## [1]  6 -1
```

Data frame I

- A data frame is a 2-dimensional list with the following properties:
 - ① The components are vectors of the same length but they can have different data types, i.e. numeric, character, or logical.
 - ② Each column has a title by which the whole vector may be addressed.
 - ③ Numeric vectors, logicals and factors are included as is, and character vectors are coerced to be **factors**.
- To create a data frame, DF, and entering some values manually for variables:

```
d <- data.frame(v1= 1:4, v2 = c("B", "C", "A", "D"),  
               v3=c(33, 43, 61, 91))  
d
```


Data frame II

```
##   v1 v2 v3
## 1  1  1  B 33
## 2  2  2  C 43
## 3  3  3  A 61
## 4  4  4  D 91
```

```
d[order(d$v2, d$v3), ]
```

```
##   v1 v2 v3
## 3  3  3  A 61
## 1  1  1  B 33
## 2  2  2  C 43
## 4  4  4  D 91
```

- In order delete a column permanently:

```
d$log_v3 = log(d$v3) # creates new var
d$v3 <- NULL # deletes v3
d
```

Data frame III

```
##   v1 v2 log_v3
## 1  1  B   3.50
## 2  2  C   3.76
## 3  3  A   4.11
## 4  4  D   4.51
```

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Reading-writing Excel and CSV files I

```
#For CSV no need for extra package
write.csv(d, "ver1.csv") # sep = "," and dec = "."
write.csv2(d, "ver2.csv") # sep = ";" and dec = ","
v1 = read.csv("ver1.csv") # sep = "," and dec = "."
v2 = read.csv2("ver2.csv") # sep = ";" and dec = ","
#For Excel we can use openxlsx package
library(openxlsx)
write.xlsx(d,"ver3.xlsx") #write excel
# first get get an excel file
f_url = "https://github.com/obakis/econ_data/raw/master/illere_gore_ihracat.xlsx"
download.file(url = f_url, destfile = "il_ihracat.xlsx", mode="wb")
# then, load it using read.xlsx package
dat = read.xlsx("il_ihracat.xlsx",
                cols = 1:16, rows=5:1458, colNames = TRUE)
head(dat)
```

Reading-writing non-excel files I

- `read.table()` for any type of delimited ASCII file (both numeric and character values). We can specify optional arguments for decimals, missing values, headers, separator etc.
- `foreign` package can be used to read and write data in 'Minitab', 'SAS', 'SPSS', 'Stata' etc. format.

#Ex.

```
read.dta      #Read Stata binary files
read.spss     #Read an SPSS data file
read.xport    #Read a SAS XPORT Format Library
read.dbf      #Read a DBF File
read.octave   #Read Octave Text Data Files
```

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Data manipulation I

```
library(dplyr)
library(tidyr)
library(openxlsx)
f_url = "https://github.com/obakis/econ_data/raw/master/illere_gore_ihracat.xlsx"
download.file(url = f_url, destfile = "il_ihracat.xlsx", mode="wb")
dat = read.xlsx("il_ihracat.xlsx",
               cols = 1:16, rows=5:1458, colNames = TRUE)

head(dat)
dat = dat[,-c(3,4)]
names(dat)[1:2] = c("year", "province")
names(dat)

dat = as_data_frame(dat)
str(dat)
# dat %>%
#   mutate_each(funs(extract_numeric), year:december) -> dat1
```

Data manipulation II

```
dat %>%
  transmute_all(extract_numeric) -> dat1
print(dat1[1350:1405,], n=10)

dat2 = fill(dat1, year, .direction = "down")
dat2 = dat2 %>%
  filter(! province %in% c(0,NA))
#print(dat2[,1:4], n=35, width=Inf)
dat_x1 = gather(data=dat2, key=month, value=export, -province, -year)
head(dat_x1)
dat_x1 %>%
  mutate(month = factor(month, levels = month.name)) %>%
  arrange(year,month, province) -> dat_x
dat_x
#write.xlsx(dat_x, "../data/tur_x.xlsx")
saveRDS(dat_x, "../data/tur_x.rds")
```


Data manipulation III

```
f_url = "https://github.com/obakis/econ_data/raw/master/illere_gore_ithalat.xlsx"
download.file(url = f_url, destfile = "il_ithalat.xlsx", mode="wb")
dat = read.xlsx("il_ithalat.xlsx",
               cols = 1:16, rows=5:1471, colNames = TRUE)
dat = dat[,-c(3,4)]
names(dat)[1:2] = c("year", "province")
names(dat)

dat = as_data_frame(dat)
dat %>%
  transmute_all(extract_numeric) -> dat1

dat2 = fill(dat1, year, .direction = "down")
dat2 = dat2 %>%
  filter(! province %in% c(0,NA,99)) # imports
dat2
```

Data manipulation IV

```
dat_i1 = gather(data=dat2, key=month, value=import, -c(province, year))
dat_i1 %>%
  mutate(month = factor(month, levels = month.name)) %>%
  arrange(year, month, province) -> dat_i
dat_i
#write.xlsx(dat_i, "../data/tur_i.xlsx")
saveRDS(dat_i, "../data/tur_i.rds")
```

Data manipulation V

```
##See http://dplyr.tidyverse.org/reference/join.html for more on joining
tur_xi = full_join(tur_i, tur_x, by=c("year","province","month"))
tur_xi %>%
  arrange(year,month, province) -> tur_xi
tur_xi
#write.xlsx(tur_xi, "../data/tur_xi.xlsx")
saveRDS(tur_xi, "../data/tur_xi.rds")
```

Data manipulation with dplyr I

```
library(dplyr)
f_url = "https://github.com/obakis/econ_data/raw/master/tur_x.rds"
download.file(url = f_url, destfile = "tur_x.rds", mode="wb")
dat_x = readRDS("tur_x.rds")

dat_x %>%
  filter (province == 34) %>% print(n=4)

## # A tibble: 204 x 4
##   year province month      export
##   <dbl>   <dbl> <fct>      <dbl>
## 1  2002       34 January  1530915.
## 2  2002       34 February 1362440.
## 3  2002       34 March    1644882.
## 4  2002       34 April    1619166.
## # ... with 200 more rows
```

Data manipulation with dplyr II

```
#group_by(): How to group data (to facilitate "split-apply-combine")  
#top_n(): to select the top or bottom entries in each group, ordered by wt  
# top 3 provinces by export share by year and month
```

```
dat_x %>%  
  filter(!is.na(export) & year == 2017) %>%  
  group_by(year, month) %>%  
  mutate(sh_x = 100*export/sum(export)) %>%  
  top_n(n=3, wt = sh_x) %>%  
  arrange(month, -sh_x)
```

```
## # A tibble: 36 x 5  
## # Groups:   year, month [12]  
##   year province month      export  sh_x  
##   <dbl>      <dbl> <fct>      <dbl> <dbl>  
## 1  2017         34 January  5571150. 49.5  
## 2  2017         16 January   745933.  6.63  
## 3  2017         35 January   690017.  6.13
```

Data manipulation with dplyr III

```
## 4 2017      34 February 6182411. 51.1
## 5 2017      16 February 881377.  7.29
## 6 2017      41 February 712545.  5.89
## 7 2017      34 March    7453151. 51.5
## 8 2017      16 March    952632.  6.58
## 9 2017      41 March    880979.  6.09
## 10 2017     34 April    6828343. 53.1
## # ... with 26 more rows
```

#yearly exports by province

```
dat_x %>%
  group_by(province, year) %>%
  summarise(
    export = sum(export, na.rm=TRUE)
  ) -> dat_x2
head(dat_x2)
```

Data manipulation with dplyr IV

```
## # A tibble: 6 x 3
## # Groups:   province [1]
##   province year   export
##   <dbl> <dbl>   <dbl>
## 1      1    2002  461040.
## 2      1    2003  565281.
## 3      1    2004  816249.
## 4      1    2005  883833.
## 5      1    2006  958987.
## 6      1    2007 1166028.
```

```
## provinces with highest exports after 2014, default is increasing
```

```
dat_x2 %>%
```

```
  filter (province !=34) %>%
```

```
  group_by(year) %>%
```

```
  filter(min_rank(desc(export))==1 & year > 2014)
```

Data manipulation with dplyr V

```
## # A tibble: 4 x 3
## # Groups:   year [4]
##   province year    export
##   <dbl> <dbl>    <dbl>
## 1      16  2015  8634502.
## 2      16  2016  9765910.
## 3      16  2017 10535563.
## 4      35  2018   839148.

## provinces with lowest exports before 208
dat_x2 %>%
  group_by(year) %>%
  filter(min_rank(export)==1 & year < 2008)
```


Data manipulation with dplyr VI

```
## # A tibble: 6 x 3
## # Groups:   year [6]
##   province year export
##   <dbl> <dbl> <dbl>
## 1      29  2002  23.5
## 2      29  2004 119.
## 3      29  2006   9.27
## 4      49  2005  41.1
## 5      62  2007  67.4
## 6      69  2003 140.
```

#The `ntile()` function is used to divide the data into N bins.

```
dat_x2 %>%
  group_by(year) %>%
  mutate(pct = ntile(export,10)) %>%
  filter(year==2017) %>%
  count(pct)
```

Data manipulation with dplyr VII

```
## # A tibble: 10 x 3
## # Groups:   year [1]
##   year    pct     n
##   <dbl> <int> <int>
## 1  2017     1     9
## 2  2017     2     8
## 3  2017     3     8
## 4  2017     4     8
## 5  2017     5     8
## 6  2017     6     8
## 7  2017     7     8
## 8  2017     8     8
## 9  2017     9     8
## 10 2017    10     8
```

Data manipulation with dplyr VIII

```
dat_x2 %>%  
  group_by(year) %>%  
  mutate(million_x=ifelse(export>=1000000, 1, 0)) %>%  
  count(million_x)
```

```
## # A tibble: 34 x 3  
## # Groups:   year [17]  
##   year million_x     n  
##   <dbl>      <dbl> <int>  
## 1  2002         0     75  
## 2  2002         1      5  
## 3  2003         0     74  
## 4  2003         1      5  
## 5  2004         0     71  
## 6  2004         1      8  
## 7  2005         0     73  
## 8  2005         1      8
```

Data manipulation with dplyr IX

```
## 9 2006      0    73
## 10 2006     1     8
## # ... with 24 more rows
```

```
dat_x2 %>%
```

```
  group_by(year) %>%
```

```
  mutate(million_x = ifelse(export >= 1000000, 1, 0)) %>%
```

```
  count(year, wt = million_x)
```

```
## # A tibble: 17 x 2
```

```
## # Groups:   year [17]
```

```
##   year      n
```

```
##   <dbl> <dbl>
```

```
## 1 2002     5
```

```
## 2 2003     5
```

```
## 3 2004     8
```

```
## 4 2005     8
```

Data manipulation with dplyr X

##	5	2006	8
##	6	2007	10
##	7	2008	13
##	8	2009	11
##	9	2010	14
##	10	2011	15
##	11	2012	16
##	12	2013	18
##	13	2014	16
##	14	2015	15
##	15	2016	15
##	16	2017	17
##	17	2018	1

Data manipulation with dplyr XI

```
dat_x2 %>%  
  group_by(year) %>%  
  mutate(million_x=ifelse(export>=1000000, 1, 0)) %>%  
  count(year, wt=millions_x, sort = TRUE)
```

```
## # A tibble: 17 x 2  
## # Groups:   year [17]  
##   year      n  
##   <dbl> <dbl>  
## 1  2013     18  
## 2  2017     17  
## 3  2012     16  
## 4  2014     16  
## 5  2011     15  
## 6  2015     15  
## 7  2016     15  
## 8  2010     14
```

Data manipulation with dplyr XII

```
## 9 2008 13
## 10 2009 11
## 11 2007 10
## 12 2004 8
## 13 2005 8
## 14 2006 8
## 15 2002 5
## 16 2003 5
## 17 2018 1
```

```
dat_x2 %>%
```

```
  group_by(year) %>%
```

```
  mutate(million_x=ifelse(export>=1000000, 1, 0)) %>%
```

```
  tally(million_x)
```

Data manipulation with dplyr XIII

```
## # A tibble: 17 x 2
```

```
##   year      n
```

```
##   <dbl> <dbl>
```

```
## 1  2002     5
```

```
## 2  2003     5
```

```
## 3  2004     8
```

```
## 4  2005     8
```

```
## 5  2006     8
```

```
## 6  2007    10
```

```
## 7  2008    13
```

```
## 8  2009    11
```

```
## 9  2010    14
```

```
## 10 2011    15
```

```
## 11 2012    16
```

```
## 12 2013    18
```

```
## 13 2014    16
```

```
## 14 2015    15
```


Data manipulation with dplyr XIV

```
## 15  2016    15
## 16  2017    17
## 17  2018     1
```

```
dat_x2 %>%
  group_by(province) %>%
    summarise(avg_x = mean(export),
              min_x = min(export),
              max_x = max(export),
              nobs = n()) %>%
  print(n=8)
```

Data manipulation with dplyr XV

```
## # A tibble: 81 x 5
##   province    avg_x    min_x    max_x  nobs
##   <dbl>    <dbl>    <dbl>    <dbl> <int>
## 1         1 1258992. 150322. 1916196.    17
## 2         2  113379.   8097.   542820.    17
## 3         3  211896.  24786.  362111.    17
## 4         4   43001.   2776.   87645.    17
## 5         5   43974.   1312.   98218.    17
## 6         6 4846643. 529935. 8102722.    17
## 7         7  706496. 115651. 1240524.    17
## 8         8   47918.   3480.   91620.    17
## # ... with 73 more rows
```

Data manipulation with dplyr XVI

```
## export growth rates by province
dat_x2 %>%
  group_by(province) %>%
  arrange(province, year) %>%
  mutate(lag_x = lag(export, n = 1)) %>%
  mutate(gr_x = 100*(export - lag_x)/lag_x) %>%
  filter(year == 2017) %>%
  arrange(-year, province)
```

```
## # A tibble: 81 x 5
## # Groups:   province [81]
##   province year  export    lag_x    gr_x
##   <dbl> <dbl>    <dbl>    <dbl>    <dbl>
## 1         1  2017 1822782. 1607018.   13.4
## 2         2  2017  131623.  338329.  -61.1
## 3         3  2017  318463.  296187.    7.52
## 4         4  2017   43092.   52134.  -17.3
```

Data manipulation with dplyr XVII

```
## 5          5 2017  92477.  76165.  21.4
## 6          6 2017 6740237. 6463475.   4.28
## 7          7 2017 1240524.  997112.  24.4
## 8          8 2017  52065.  55859.  -6.79
## 9          9 2017  702007.  617679.  13.7
## 10         10 2017  537182.  519260.   3.45
## # ... with 71 more rows
```

```
# dat_x2 %>%
#   group_by(province) %>%
#   arrange(province, year) %>%
#   mutate(lag_x = lag(export, n = 1)) %>%
#   mutate(gr_x = 100*(export - lag_x)/lag_x) %>%
#   filter(!is.na(gr_x)) %>%
#   group_by(year) %>%
#   top_n(n=1, wt=gr_x) %>%
#   arrange(-year)
```

Data manipulation with dplyr XVIII

```
## best performing province: cumulative change of x_sh
dat_x2 %>%
  filter (!is.na(export)) %>%
  group_by(year) %>%
  mutate(sh_x = 100*export/sum(export)) %>%
  arrange(province, year) %>%
  group_by(province) %>%
  mutate(c_diff_x = sh_x - first(sh_x)) %>%
  arrange(-c_diff_x) %>%
  filter(year==2017)
```

Data manipulation with dplyr XIX

```
## # A tibble: 81 x 5
## # Groups:   province [81]
##   province year  export sh_x c_diff_x
##   <dbl> <dbl>   <dbl> <dbl>   <dbl>
## 1      27  2017 6607631. 4.21    2.49
## 2      54  2017 5249859. 3.34    2.16
## 3      41  2017 8095543. 5.16    1.64
## 4      42  2017 1548194. 0.986   0.625
## 5      31  2017 2333958. 1.49    0.517
## 6      47  2017  910275. 0.580   0.515
## 7      45  2017 1989714. 1.27    0.402
## 8       7  2017 1240524. 0.790   0.330
## 9      46  2017  955209. 0.608   0.302
## 10     73  2017  471547. 0.300   0.242
## # ... with 71 more rows
```

Data manipulation with dplyr XX

```
dat_x2 %>%  
  filter (!is.na(export)) %>%  
  group_by(year) %>%  
  summarise(p10=quantile(export, probs=0.1),  
            p50=quantile(export, probs=0.5),  
            p90=quantile(export, probs=0.90)  
            )-> x_perc  
x_perc %>% filter(year > 2013)
```

```
## # A tibble: 5 x 4  
##   year    p10    p50    p90  
##   <dbl> <dbl> <dbl> <dbl>  
## 1  2014  6062. 214064. 2115434.  
## 2  2015  7361. 186820. 1839282.  
## 3  2016  9219. 167067. 1874348.  
## 4  2017 12737. 162682. 2333958.  
## 5  2018   966. 15323. 188769.
```

Data manipulation with dplyr XXI

```
x_perc %>%
```

```
  mutate_each(funs("log"=log), p10,p50) -> x_perc2
```

```
## `mutate_each()` is deprecated.
```

```
## Use `mutate_all()`, `mutate_at()` or `mutate_if()` instead.
```

```
## To map `funs` over a selection of variables, use `mutate_at()`
```

```
x_perc2
```

```
## # A tibble: 17 x 6
```

	year	p10	p50	p90	p10_log	p50_log
	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
## 1	2002	576.	21338.	431330.	6.36	9.97
## 2	2003	1190.	26689.	620829.	7.08	10.2
## 3	2004	2015.	34725.	892258.	7.61	10.5
## 4	2005	2095.	46125.	965882.	7.65	10.7
## 5	2006	1759.	59381.	958987.	7.47	11.0
## 6	2007	3893.	87802.	1280695.	8.27	11.4

Data manipulation with dplyr XXII

##	7	2008	5337.	103431.	1762181.	8.58	11.5
##	8	2009	6060.	93267.	1417802.	8.71	11.4
##	9	2010	6414.	120037.	1698405.	8.77	11.7
##	10	2011	5910.	143675.	2050555.	8.68	11.9
##	11	2012	6226.	150370.	2039566.	8.74	11.9
##	12	2013	9829.	190564.	2097846.	9.19	12.2
##	13	2014	6062.	214064.	2115434.	8.71	12.3
##	14	2015	7361.	186820.	1839282.	8.90	12.1
##	15	2016	9219.	167067.	1874348.	9.13	12.0
##	16	2017	12737.	162682.	2333958.	9.45	12.0
##	17	2018	966.	15323.	188769.	6.87	9.64

Describing data I

```
f_url = "https://github.com/obakis/econ_data/raw/master/hls2011.rds"
download.file(url = f_url, destfile = "hls2011.rds", mode="wb")
hls = readRDS("hls2011.rds")
```

```
myvars = c("hwage", "educ", "female", "exper", "emp_sect")
head(hls[,myvars])
```

```
##   hwage educ female exper emp_sect
## 1  8.75   2      0    33      pub
## 2  2.92   2      1     2      priv
## 3  2.53   5      1    22      priv
## 4 58.33  15      0    21      priv
## 5  3.89   8      0    16      priv
## 6  1.46   8      0    49      priv
```

```
str(hls[,myvars])
```

Describing data II

```
## 'data.frame': 762 obs. of 5 variables:
## $ hwage : num 8.75 2.92 2.53 58.33 3.89 ...
## $ educ : int 2 2 5 15 8 8 5 15 5 15 ...
## $ female : int 0 1 1 0 0 0 0 1 0 1 ...
## $ exper : int 33 2 22 21 16 49 22 6 17 2 ...
## $ emp_sect: Factor w/ 3 levels "other","priv",...: 3 2 2 2 2 2 2 2 2 3 ...
```

```
summary(hls[,myvars])
```

##	hwage	educ	female	exper
## Min. :	1.2	Min. : 0.00	Min. : 0.00	Min. : 0.0
## 1st Qu.:	2.9	1st Qu.: 5.00	1st Qu.: 0.00	1st Qu.: 10.0
## Median :	3.9	Median : 8.00	Median : 0.00	Median : 18.0
## Mean :	6.2	Mean : 9.26	Mean : 0.22	Mean : 18.9
## 3rd Qu.:	8.2	3rd Qu.: 15.00	3rd Qu.: 0.00	3rd Qu.: 27.0
## Max. :	58.3	Max. : 15.00	Max. : 1.00	Max. : 72.0
## emp_sect				

Describing data III

```
## other: 9
## priv :557
## pub  :196
##
##
##
```

```
summary(hls[, "exper"])
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##       0.0    10.0    18.0    18.9    27.0    72.0
```

```
seq(0,8,by=3)
```

```
## [1] 0 3 6
```

```
cut(0:8, breaks=seq(0,8,by=3))
```

Describing data IV

```
## [1] <NA> (0,3] (0,3] (0,3] (3,6] (3,6] (3,6] <NA> <NA>  
## Levels: (0,3] (3,6]
```

```
cut(0:8, breaks=seq(0,8,by=3), include.lowest = TRUE)
```

```
## [1] [0,3] [0,3] [0,3] [0,3] (3,6] (3,6] (3,6] <NA> <NA>  
## Levels: [0,3] (3,6]
```

```
cut(0:8, breaks=seq(0,9,by=3), include.lowest = TRUE)
```

```
## [1] [0,3] [0,3] [0,3] [0,3] (3,6] (3,6] (3,6] (6,9] (6,9]  
## Levels: [0,3] (3,6] (6,9]
```

```
cut(0:8, breaks=seq(0,9,by=3), include.lowest = TRUE, right=FALSE)
```

```
## [1] [0,3) [0,3) [0,3) [3,6) [3,6) [3,6) [6,9] [6,9] [6,9]  
## Levels: [0,3) [3,6) [6,9]
```

Describing data V

```
hls$exper_gr = cut(hls$exper, breaks = seq(0,60,by=10), right=FALSE,  
                  include.lowest = TRUE)
```

```
tab = table(hls$exper_gr)  
tab
```

```
##  
## [0,10) [10,20) [20,30) [30,40) [40,50) [50,60]  
##      184      234      200      108      33      1
```

```
round( 100*prop.table(tab), 1)
```

```
##  
## [0,10) [10,20) [20,30) [30,40) [40,50) [50,60]  
##      24.2      30.8      26.3      14.2      4.3      0.1
```

Describing data VI

```
## 2 categorical variables
#tab2 = with(hls, table(gender, e_sect)) # or better
tab2 = xtabs( ~ female+emp_sect, data=hls)
tab2
```

```
##      emp_sect
## female other priv pub
##      0      3  456 135
##      1      6  101  61
```

```
prop.table(tab2, margin=1) # row sum = 100
```

```
##      emp_sect
## female  other  priv   pub
##      0 0.00505 0.76768 0.22727
##      1 0.03571 0.60119 0.36310
```

```
prop.table(tab2, margin=2) # col sum = 100
```

Describing data VII

```
##      emp_sect
## female other  priv   pub
##      0 0.333 0.819 0.689
##      1 0.667 0.181 0.311
```

```
tab3 = xtabs(wts ~ female+emp_sect, data=hls)
prop.table(tab3, margin=2)
```

```
##      emp_sect
## female other  priv   pub
##      0 0.333 0.807 0.683
##      1 0.667 0.193 0.317
```

```
## one categorical and one numerical variable, mean
aggregate(hwage ~ exper_gr, FUN = mean, data=hls)
```


Describing data VIII

```
##   exper_gr hwage
## 1  [0,10)  5.34
## 2  [10,20) 6.81
## 3  [20,30) 6.02
## 4  [30,40) 6.27
## 5  [40,50) 7.09
## 6  [50,60] 2.16
```

```
## one categorical and one numerical variable, weighted mean
```

```
sapply(
  split(hls, hls$exper_gr),
  FUN=function(dat) weighted.mean(dat$hwage, dat$wts)
)
```

```
##  [0,10) [10,20) [20,30) [30,40) [40,50) [50,60]
##    5.26    6.89    6.04    6.50    7.08    2.16
```

Describing data IX

```
## or
library(dplyr)
hls %>%
  group_by(exper_gr) %>%
  summarise(vmean = weighted.mean(hwage, wts))
```

```
## # A tibble: 7 x 2
##   exper_gr vmean
##   <fct>    <dbl>
## 1 [0,10)    5.26
## 2 [10,20)   6.89
## 3 [20,30)   6.04
## 4 [30,40)   6.50
## 5 [40,50)   7.08
## 6 [50,60]   2.16
## 7 <NA>      3.84
```

Describing data X

```
plot(log(hwage) ~ factor(female), hls)
```

```
plot(log(hwage) ~ educ, hls)
```

