

Rotman

INTRO TO R

R Workshop

January 24, 2020 Prepared by Jay Cao / TDMDAL

Website: <https://tdmdal.github.io/r-tutorial-201920-winter/>



Rotman School of Management
UNIVERSITY OF TORONTO

What's R?



- A programming language
 - Free and open source
 - Extensible with many high-quality user-contributed libraries/packages
- Great for statistical analysis, graphics and many other things (ex?)

General purpose
languages

- C/C++
- Python



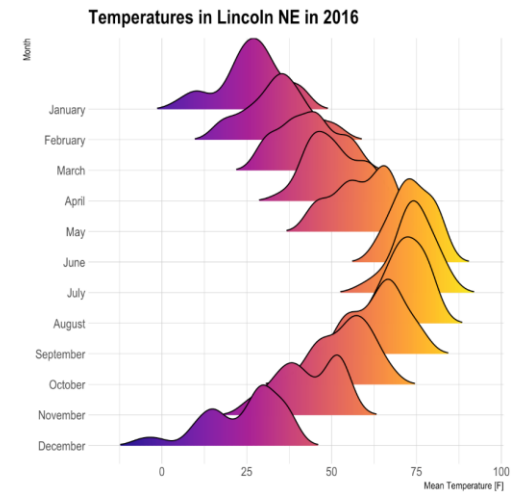
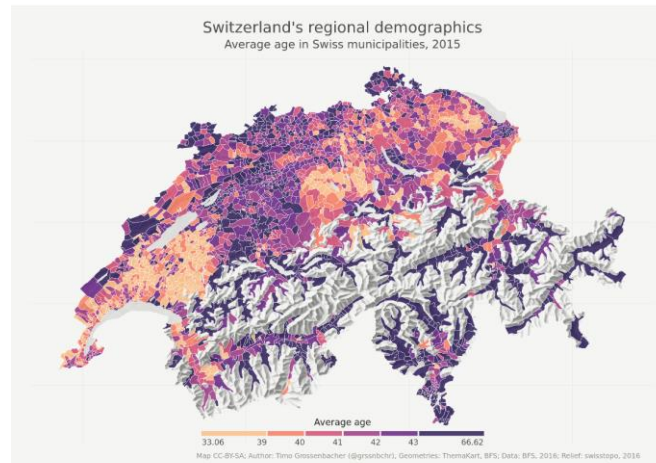
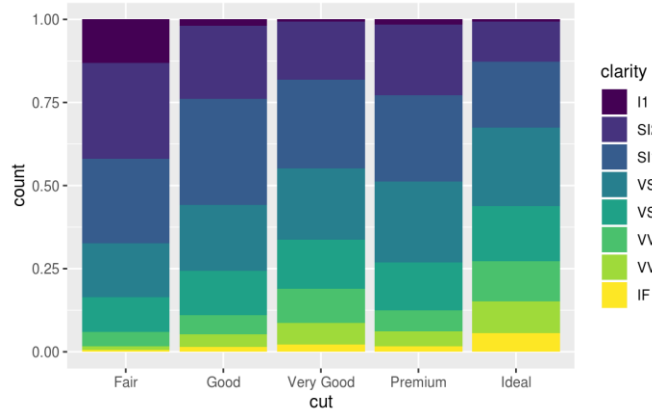
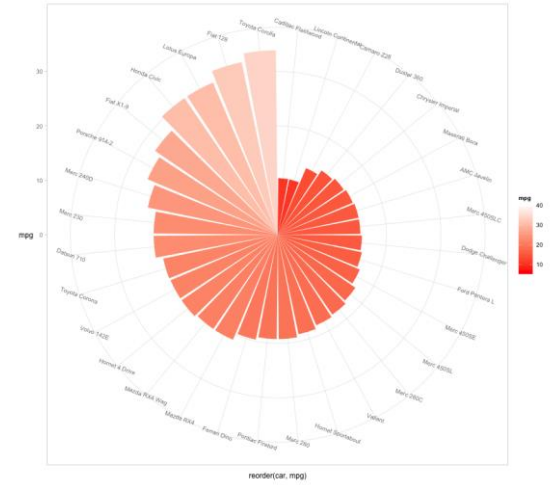
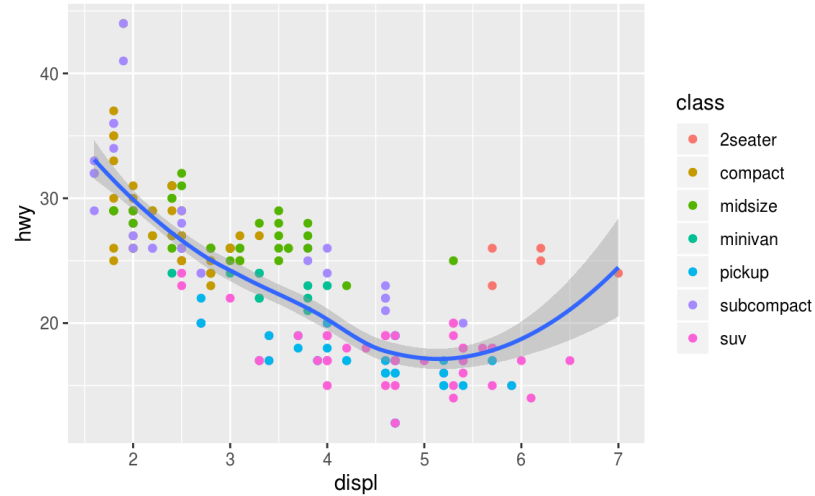
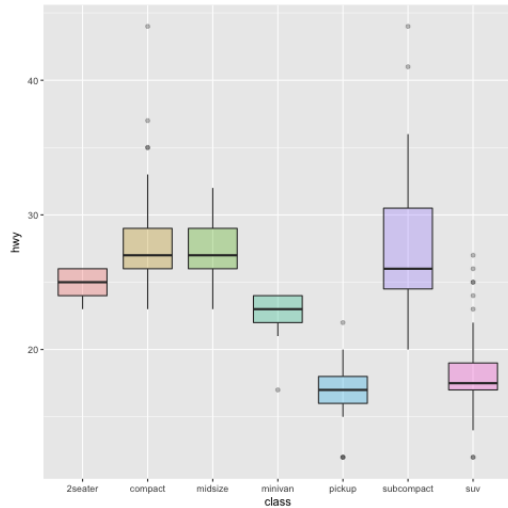
Statistical Analysis
Languages/Toolkits

- SPSS
- Stata

What can R do – Statistics & related

- Statistics & Econometrics
 - Regressions
 - Time series analysis
 - Bayesian inference
 - Survival analysis
 - ...
- Numerical Mathematics
 - Optimization
 - Solver
 - Differential equations
 - ...
- Finance
 - Portfolio management
 - Risk management
 - Option pricing
 - ...
- ...

What can R do – Graphics (static ones)

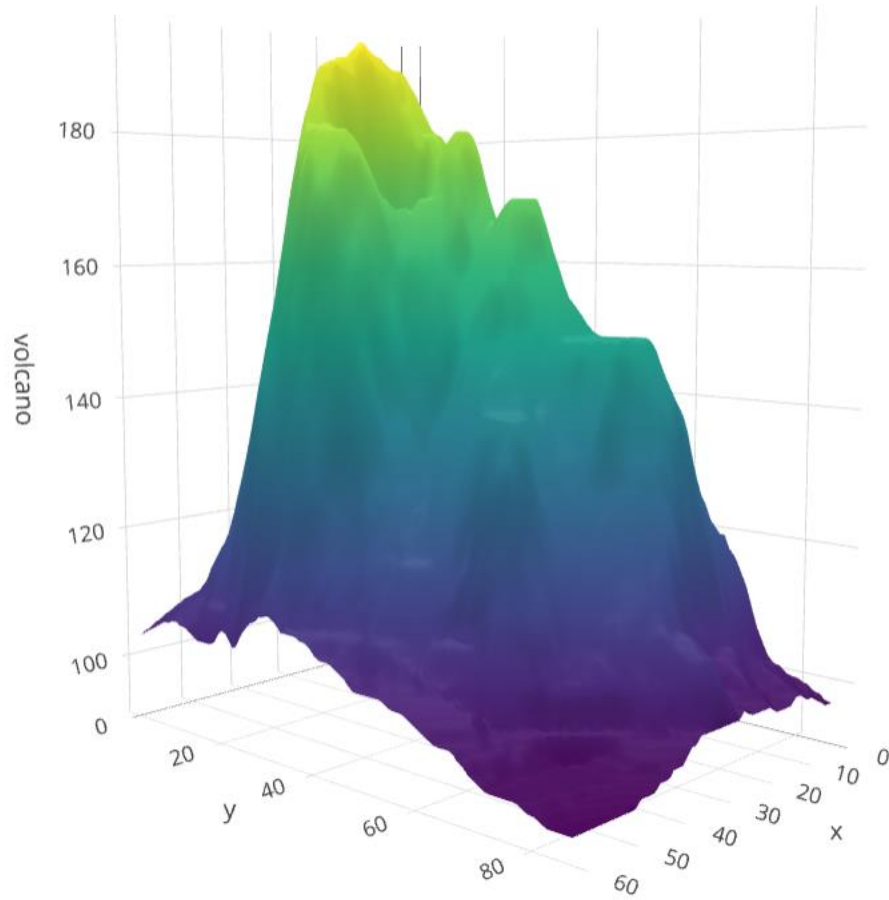


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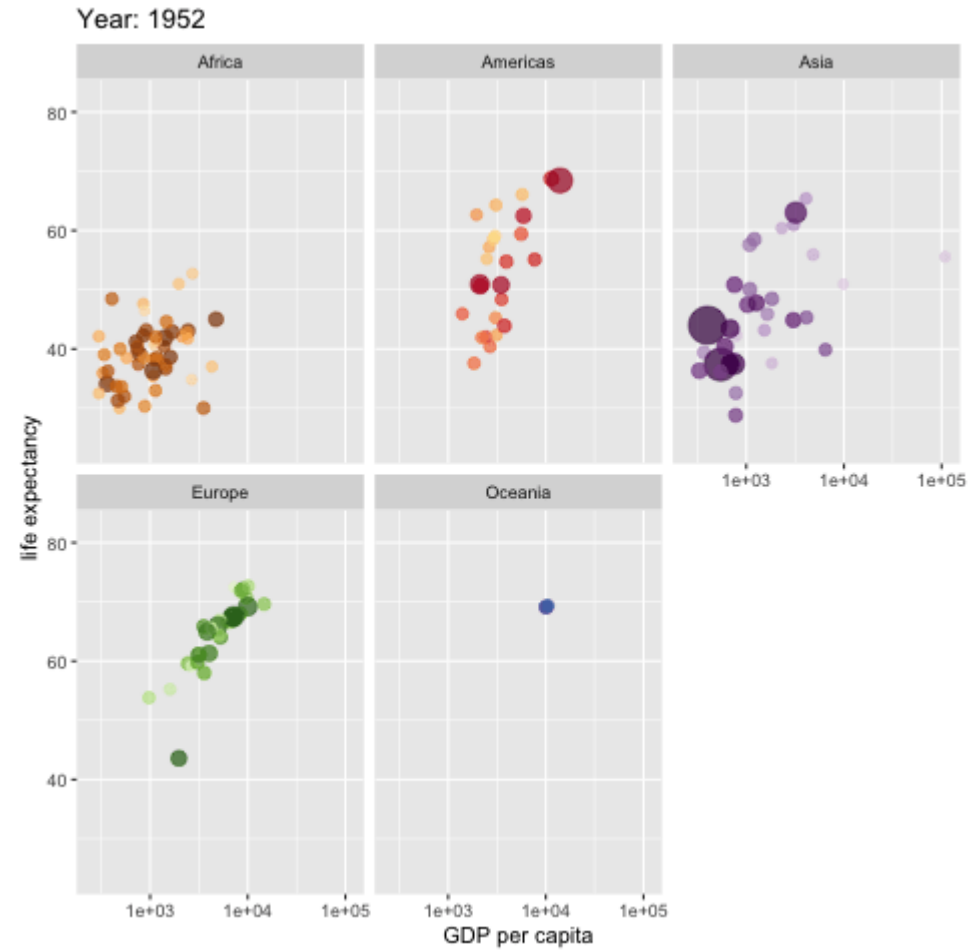
<https://www.r-graph-gallery.com/>

<https://timogrossenbacher.ch/2016/12/beautiful-thematic-maps-with-ggplot2-only/>;

What can R do – Graphics (dynamic ones)



<https://plot.ly/r/3d-surface-plots/>;



<https://github.com/thomasp85/gganimate>;

What can R do – Others

- Machine learning (ex. R interface to Keras: [keras](#))
- Natural language processing (ex. [tidytext](#), [topicmodels](#))
- Web technology
 - Web scraping (ex. [rvest](#))
 - API wrapper (ex. Twitter: [rtweet](#); bigquery: [bigrquery](#); Quandl: [Quandl](#))
 - Shiny web app (<https://shiny.rstudio.com/>)
- Reporting
 - [R Markdown](#) (write reports, slides, blogs, books, etc. See a gallery [here](#).)
- ... (see [R Task View](#) for more)

What can R do, for You?

- Beyond Excel Data Analysis
- Automate boring tasks
- Prototype ideas
- ...

Plan for the 4 Sessions

- Overview
- Data Manipulation
- Graphs
- Time Series (Finance Applications)

Plan for Today (~2 hrs)

- Motivation: three examples
 - A simple regression (housing price and pollution)
 - Twitter API
 - Deep learning “Hello World!”
- Basics of R
 - Data structure
 - Programming structure
- A typical analysis workflow: extending the regression example
 - Import and manipulate data
 - Build models
 - Report results

What's RStudio?



RStudio

File Edit Code View Plots Session Build Debug Profile Tools Help

graph_test.R raw_shiny_v2.R

```
1 library(Diagrammer)
2
3 raw <- tribble(
4   ~id, ~in_node, ~out_node, ~in_time, ~out_time,
5   #--|--|
6   1, 1, 2, 1, 3,
7   1, 2, 3, 3, 5,
8   2, 1, 2, 2, 3,
9   2, 2, 4, 3, 6
10 )
11
12 node_tb_tp <- raw %>%
13   distinct(in_node) %>%
14   rename(node_id = in_node)
15
16 node_tb <- raw %>%
17   distinct(out_node) %>%
18   rename(node_id = out_node) %>%
19   union(node_tb_tp) %>%
20   arrange(node_id)
21
22 edge_tb <- raw %>%
23   distinct(in_node, out_node) %>%
24   rename(from = in_node, to = out_node)
25
26 g <- create_graph() %>%
27   add_nodes_from_table(table = node_tb) %>%
28   add_edges_from_table(
29     table = edge_tb,
30     from_col = from,
31     to_col = to,
32     from_to_map = node_id
33   )
34
35 g %>% render_graph()
```

Environment History Connections Presentation

Global Environment

Data

edge_tb	3 obs. of 2 variables
g	List of 12
node_tb	4 obs. of 1 variable
node_tb_tp	2 obs. of 1 variable
raw	4 obs. of 5 variables

Files Plots Packages Help Viewer

Zoom Export Publish

```
graph LR
  1((1)) --> 2((2))
  2((2)) --> 4((4))
  2((2)) --> 3((3))
```

Google Colab



The screenshot shows a Google Colab notebook interface. The browser address bar displays the URL: `https://colab.research.google.com/github/tdmdal/r-workshop-researchers/blob/master/docs/rn1_A_Simple_Regression.ipynb`. The notebook title is "rn1 A Simple Regression". The menu bar includes "File", "Edit", "View", "Insert", "Runtime", "Tools", and "Help". Below the menu, there are buttons for "+ Code", "+ Text", and "Copy to Drive". On the right, there are status indicators for "RAM" and "Disk" usage, and a button labeled "Editing".

The notebook content is organized into sections:

- 1. Data Import and Manipulation**

We first import a dataset from the workshop website. This is a dataset on married women labor force participation used in [Mroz 1987](#). The dataset is also used throughout Wooldridge's text book: Introductory Econometrics: A Modern Approach. After briefly inspecting the data, we create a new column `lwage` in preparation for a simple regression.

```
[ ] # load data
data_url <- "https://tdmdal.github.io/r-workshop-researchers/data/mroz_1987.csv"
mroz_1987 <- read.csv(data_url)

[ ] # take a look at the structure of the data
str(mroz_1987)

See a description of the data columns here.

[ ] # print the first few rows of the dataset
head(mroz_1987)

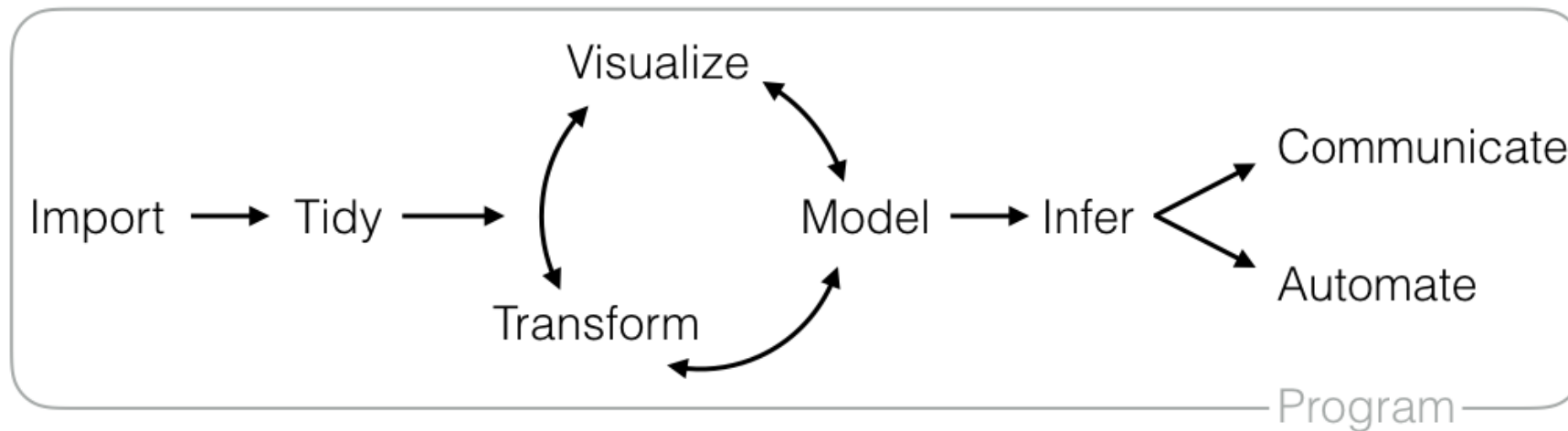
[ ] # create log wage
mroz_1987["lwage"] <- log(mroz_1987["wage"])
```
- 2. Modelling**

We will run a simple regression to investigate return on education for married women: $\log(wage) = \beta_0 + \beta_1 educ + u$.

```
[ ] # setup a regression model
lr <- lm(formula = lwage ~ educ, data = mroz_1987)
```

Data Analysis Workflow: Three Examples

- A simple regression
- Twitter API
- Deep learning [“Hello World!”](#)



R Basics

- Data structures
- Programming structures

Expression and Assignment

```
# expression
```

```
2 + sqrt(4) + log(exp(2)) + 2^2
```

```
# assignment
```

```
x <- 3
```

```
y <- (pi == 3.14)
```

R Data Structure - Overview

	Homogeneous	Heterogeneous
1-d	Atomic vector	List
2-d	Matrix	Data frame
n-d	Array	

R Data Structure - Overview

	Homogeneous	Heterogeneous
1-d	Atomic vector →	List
2-d	Matrix	↓ Data frame
n-d	Array	

Atomic Vectors

```
# create R vectors
```

```
vec_character <- c("Hello,", "World!")
```

Hello,

World!

```
vec_integer <- c(1L, 2L, 3L)
```

1

2

3

```
vec_double <- c(1.1, 2.2, 3.3)
```

1.1

2.2

3.3

```
vec_logical <- c(TRUE, TRUE, FALSE)
```

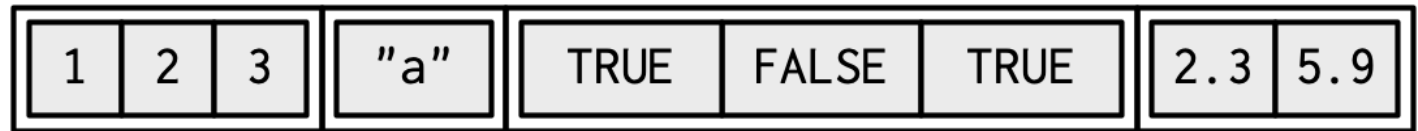
TRUE

TRUE

FALSE

List

```
# create an R list  
l1 <- list(  
  1:3,  
  "a",  
  c(TRUE, FALSE, TRUE),  
  c(2.3, 5.9)  
)
```



Data Frame

```
# create a data frame
df1 <- data.frame(
  x = 1:3,
  y = letters[1:3],
  z = c(1.1, 2.2, 3.3)
)
```

x	y	z
1	"a"	1.1
2	"b"	2.2
3	"c"	3.3

Data Frame

```
# create a data frame
df1 <- data.frame(
  x = 1:3,
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)
```

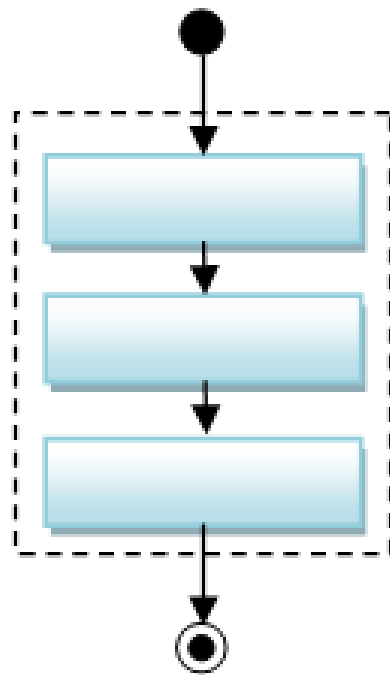
x	y	z
1	"a"	1.1
2	"b"	2.2
3	"c"	3.3

Data Frame

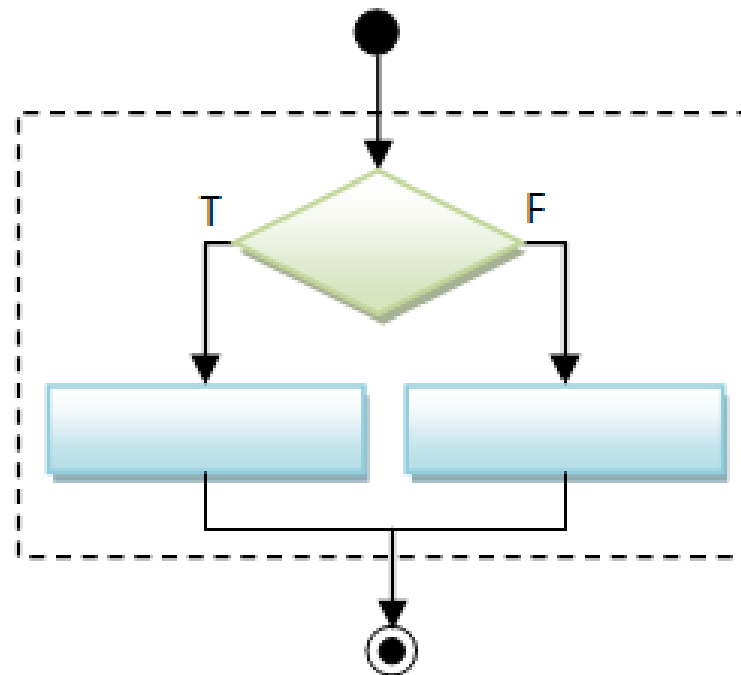
```
# create a data frame
df1 <- data.frame(
  x = 1:3,
  y = letters[1:3],
  z = c(1.1, 2.2, 3.3)
)
```

x	y	z
1	"a"	1.1
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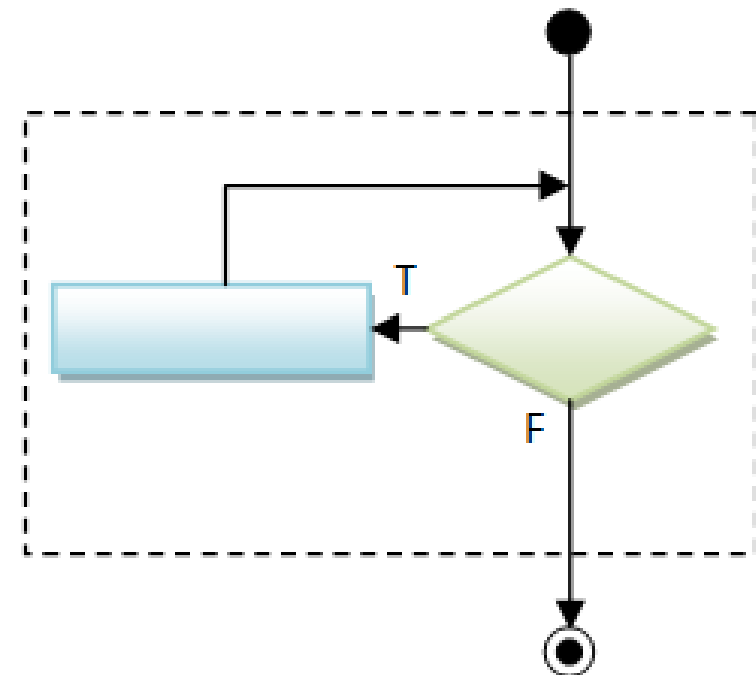
Programming Structure: Control Flows



Sequential



Conditional (Decision)



Loop (Iteration)

Sequential

- Example: Sum of Squares

$$\sum_{t=1}^3 t^2$$

```
# sum of squares  
t <- 1:3  
y <- sum(t^2)  
print(y)
```

Conditional (if...else...)

```
if (cond) {  
    # run here if cond is TRUE  
} else {  
    # run here if cond is FALSE  
}
```

```
# y greater than 10?  
if (y > 10) {  
    print("greater than 10")  
} else {  
    print("less or equal to 10")  
}
```


Conditional (if...else if...else...)

```
if (cond1) {  
    # run here if cond1 is TRUE  
} else if (cond2) {  
    # run here if cond1 is FALSE but cond2 is TRUE  
} else {  
    # run here if neither cond1 nor cond2 is TRUE  
}
```

Iteration

```
for (var in seq) {  
  do something  
}
```

```
while (cond) {  
  do something if cond is TRUE  
}
```

```
# sum of squares  
t <- 1:3  
y <- 0  
  
for (x in t) {  
  y <- y + x^2  
}  
  
print(y)
```

Programming Structure: Functions

- What's a function
 - a logical block of code
 - input -> output
- Why write functions
 - Reusability
 - Abstraction
 - Maintainability
- Example: $\sum_{t=1}^n t^2$

```
# sum of squares from 1 to n
ss <- function(n) {
  t <- 1:n
  sum(t^2)
}

# calling the ss() function
print(ss(2))
print(ss(3))
```

Extending the regression example

- Manipulate data
 - Load data
 - Create new columns
 - Filter columns and rows
- Build models
 - Multiple regression
 - IV regression
- Report and graph
 - Build a publication-ready table for regression results

Using R libraries

- Install and load an R library

```
install.packages("library_name")
```

```
library(library_name)
```

- [CRAN](#) (The Comprehensive R Archive Network)
 - [CRAN Task Views](#)

Many choices, which one to use

- Often time, many choices of functions/libraries to do one task
 - R is open and extensible!
- Example: load a csv file to a data frame
 - Use [read.csv\(\)](#) function from the `utils` library
 - Use [read_csv\(\)](#) function from the [readr](#) library
 - Use [fread\(\)](#) function from the [data.table](#) library
 - Use [vroom\(\)](#) from the [vroom](#) library

Many choices, which one to use

- Start with the one most people use
- Choose one that is well maintained
 - check document, github, etc. for last update
- Choose one that suits your task

Our Choice: extending the regression example

- Manipulate data ([tidyverse](#) eco-system)
 - Load data ([read_csv\(\)](#) from the [readr](#))
 - Create new columns ([mutate\(\)](#) from [dplyr](#))
 - Filter columns and rows ([select\(\)](#) and [filter\(\)](#) from [dplyr](#))
- Build models
 - Multiple regression ([lm\(\)](#) from stats library in R base)
- Report and graph
 - Build a publication-ready table ([stargazer\(\)](#) from [stargazer](#) library)

Load a CSV file

- [read_csv\(\)](#) from the [readr](#)

```
read_csv(file)
```

- More about [read_csv\(\)](#)
- More about [readr](#)

Load Data – Many other libraries

- [readxl](#) for Excel sheets
- [haven](#) for SPSS, Stata and SAS data
- [jsonlite](#) for JSON
- [xml2](#) for XML
- [httr](#) for web APIs
- [rvest](#) for web scraping
- [DBI](#) for connecting to DataBase engine
- ...

Data Manipulation: dplyr basics

- Filter observations: filter()
- Select variables: select()
- Reorder rows: `arrange()`
- Create new variables: mutate()
- Collapse column values to a single summary: `summarise()`
- Group by: `group_by()`

Data Manipulation: filter()

```
filter(my_dataframe, condition1, ...)
```

Data Manipulation: mutate()

```
mutate(my_dataframe, new_var1 = expression1, ...)
```

Data Manipulation: select()

```
select(my_dataframe, var1, ...)
```

Data Manipulation: Data Pipe (%>%)

```
iris_cleaned <- filter(iris, Species == "setosa")  
iris_cleaned <- select(iris_cleaned, Sepal.Length)
```

Data Manipulation: Data Pipe (%>%)

```
iris_cleaned <- filter(iris, Species == "setosa")  
iris_cleaned <- select(iris_cleaned, Sepal.Length)
```

```
iris_cleaned <- iris %>%  
  filter(., Species == "setosa") %>%  
  select(., Sepal.Length)
```


Data Manipulation: Data Pipe (%>%)

```
iris_cleaned <- filter(iris, Species == "setosa")  
iris_cleaned <- select(iris_cleaned, Sepal.Length)
```

```
iris_cleaned <- iris %>%  
  filter(Species == "setosa") %>%  
  select(Sepal.Length)
```

Data Manipulation: Others

- Join two data frames
 - [join\(\)](#) family in dplyr
- Reshape data frames
 - [pivot longer\(\)](#) and [pivot wider\(\)](#) in tidyr

Regression

- Multiple regressions: [lm\(\)](#) from stats library in base R

```
my_model <- lm(y ~ x1 + x2, data)
```

- Multiple regressions with interactive terms

```
my_model <- lm(y ~ x1 + x2 + I(x1 * x2), data)
```

- Regression result summary: `summary()`

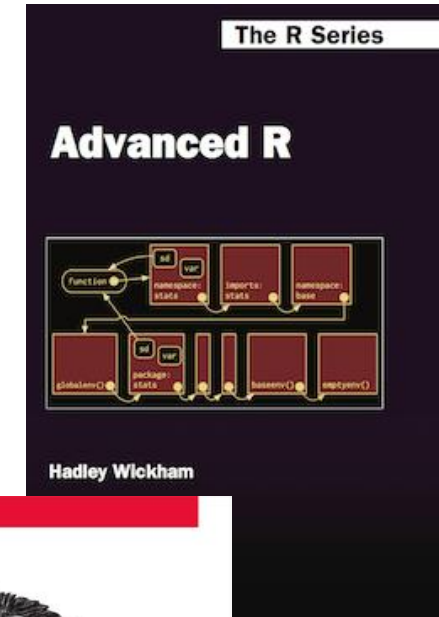
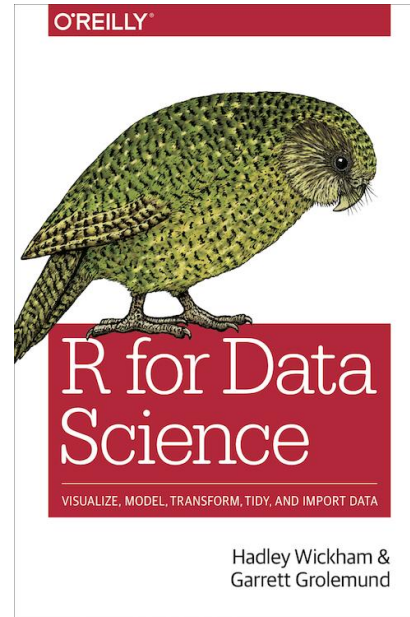
Report

- Summary table
 - [Summary for lm\(\)](#): `summary(my_model)`
- publication-ready table: [stargazer\(\)](#) from [stargazer](#) library

```
stargazer(my_model1, my_model2, ...)
```

Free Learning Resources - Books

- [R for Data Science](#)
- [Advanced R](#)
- [Hands-On Programming with R](#)
- Check bookdown.org often



Free Learning Resources – Video Courses

- [RStudio Resources Site](#)
- LinkedIn Learning (used to be lynda.com)
 - free for [UofT students](#) and [Toronto Public Library users](#)
 - Search R and learn

Free Learning Resources – Others

- [CRAN Task View](#)
- Sample notebooks / reports at <http://rpubs.com/>
- Twitter (a few seeds: [#rstat](#), [@hadleywickham](#), [@WeAreRLadies](#))