#### Rotman

# INTRO TO R

R Workshop - 1



#### Plan for the 4 Sessions

Overview (Today)

- Data Manipulation (tidyverse)
- Data Visualization (ggplot2)
- Modeling Workflow (tidymodels)

Time Series & Finance Applications

## Plan for Today

- Intro to Overview
  - What is R and what can R do?
  - Setup R
  - Motivation examples
- Overview of R programing and Data Science
  - Basics of R programming
  - Data science with R
- Learning Resources and Road Map

## Goal for Today – Answer These Questions

- What's R?
- What will I use R for?
- How to Run R code?
- What are the R coding basics?
  - Data and programming structure
  - Data science workflow
- How to learn R (on my own)?
  - with get-started help from this mini-course

#### What's R?



- R = a language + an eco-system
  - A free and open-source programming language
  - An eco-system of many high-quality user-contributed libraries/packages

• In the past R is mostly known for its statistical analysis toolkits

- Nowadays R is capable of (and very good at) many other tasks
  - Tools that cover the whole data analysis workflow
  - Tools for web technology...

#### What can R do – Statistics & related

#### Statistics & Econometrics

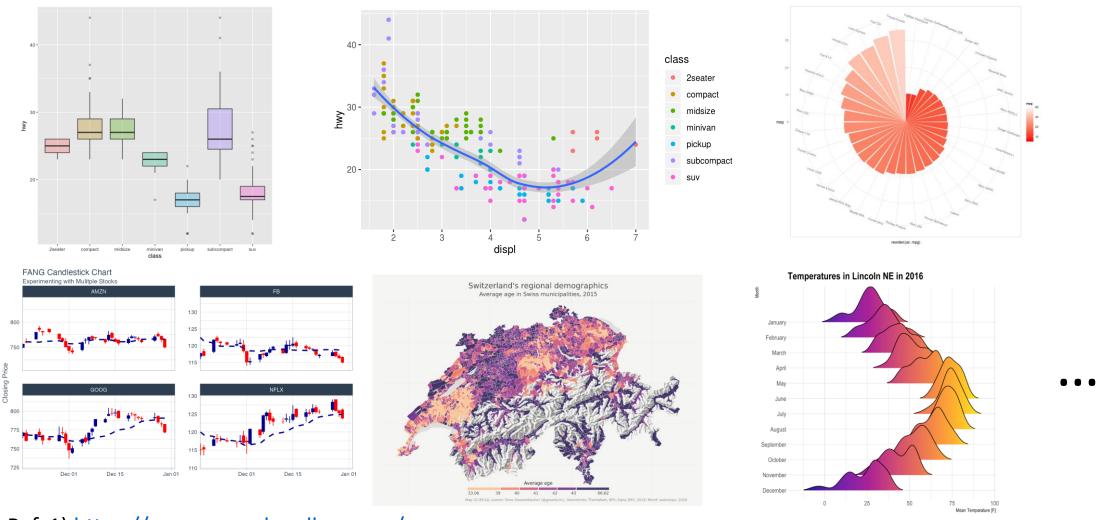
- Regressions
- Time series analysis
- Bayesian inference
- Survival analysis
- ...
- Numerical Mathematics
  - Optimization
  - Solver
  - Differential equations
  - •

#### • Finance

- Portfolio management
- Risk management
- Option pricing
- ..

•

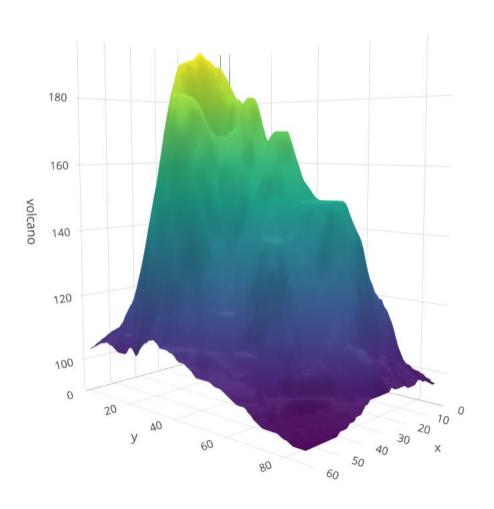
## What can R do – Graphics (static ones)

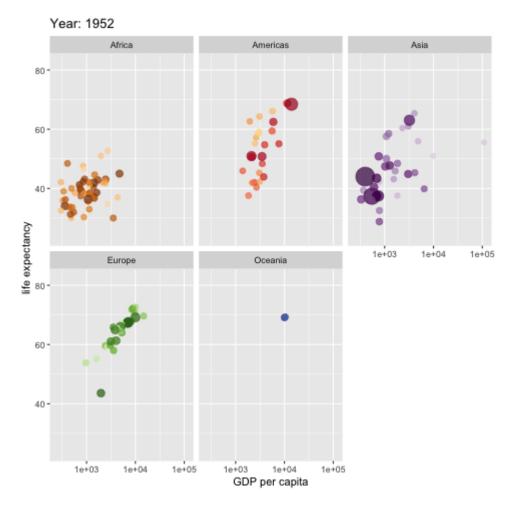


Ref. 1) <a href="https://www.r-graph-gallery.com/">https://www.r-graph-gallery.com/</a>

2) <a href="https://timogrossenbacher.ch/2016/12/beautiful-thematic-maps-with-ggplot2-only/">https://timogrossenbacher.ch/2016/12/beautiful-thematic-maps-with-ggplot2-only/</a>

## What can R do – Graphics (dynamic ones)





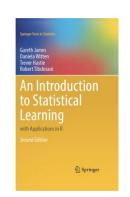
https://plot.ly/r/3d-surface-plots/;

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

#### What can R do – ML & NLP

Deep learning (neural networks)

- Machine learning
  - Statistical learning (clustering, decision tree, etc.)
    - An Introduction to Statistical Learning (with Applications in R)







- Interface to **Keras** and **Tensorflow** (via **reticulate**, an R to Python interface)
- <u>Torch for R</u> (natively from R; similar as PyTorch in Python)





- Natural language processing (e.g., <u>tidytext</u>, <u>topicmodels</u>)
- 1. See more R Machine Learning Packages on R Task View ML & Statistical Learning
- 2. See more R Natural Language Processing Packages on R Task View NLP

## What can R do – Web & Reporting

- Web technology
  - Web scraping (e.g., <u>rvest</u>)
  - API wrapper (e.g., Twitter: <u>rtweet</u>; bigquery: <u>bigrquery</u>; Quandl: <u>Quandl</u>)
  - Shiny web app (<a href="https://shiny.rstudio.com/">https://shiny.rstudio.com/</a>)

- Reporting
  - R Markdown (write reports, slides, blogs, books, etc. See a gallery <u>here</u>.)

• ... (see R Task View for more)

#### R vs Excel and BI Tools vs Python





- Excel & Business Intelligence (BI) Tools (e.g., Tableau, Power BI, etc.)
  - 2-D tables as basic data structure
  - Good UI (User Interface) and minimum programming
  - Limited modeling tools
  - Not easy to reproduce an analysis (because it's hard to store UI clicks)
  - Not flexible enough for complicated analysis problems, i.e., problems with
    - Many data cleaning steps/pipelines
    - Many different models to try



- Python
  - Python is more general purpose, R is more specialized in statistical analysis
  - R is much easier to learn (in my opinion)



## Why learn R (What can R do for You)?

- Beyond Excel Data Analysis
  - I wish Excel could...
- Automate boring repeating tasks
  - e.g., daily data collection from different sources, weekly dashboard update
- Prototype ideas
  - e.g., a novel trading strategy, a new credit risk model
- Really, find anything that interests you and use R...

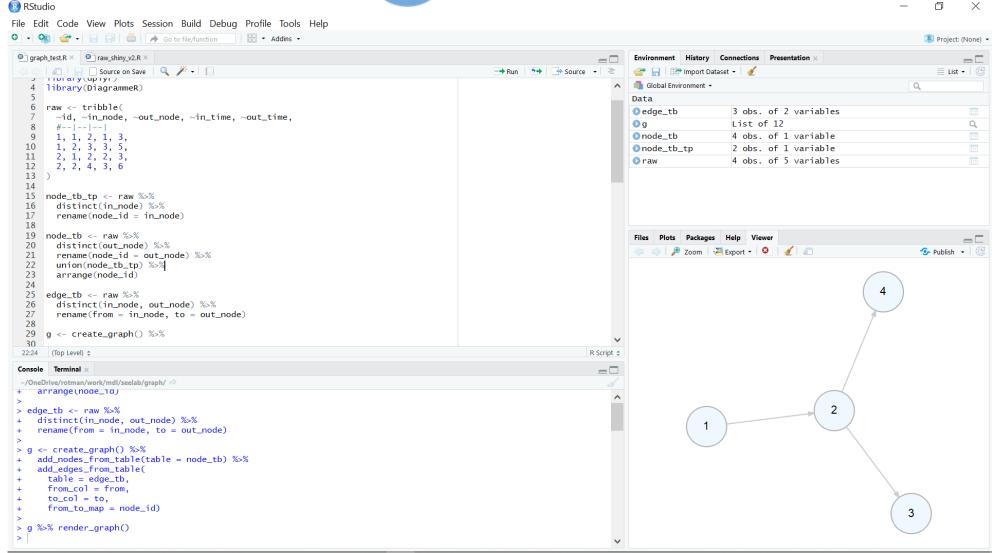
## Plan for Today

- Intro to Overview
  - What is R and what can R do?
  - Setup R
  - Motivation examples
- Overview of R programing and Data Science
  - Basics of R programming
  - Data science with R
- Learning Resources and Road Map

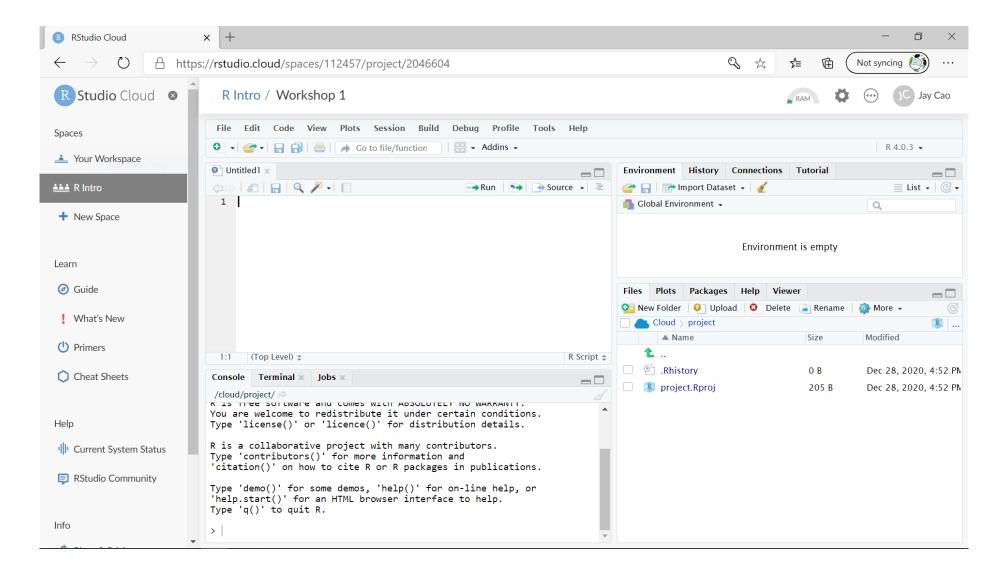
#### Setup R

- R & RStudio on your computer (most of you should choose this one)
  - Install R (<a href="https://www.r-project.org/">https://www.r-project.org/</a>)
  - Install RStudio (<a href="https://rstudio.com/products/rstudio/download/">https://rstudio.com/products/rstudio/download/</a>)
- R & RStudio in the Cloud (run R without installation)
  - RStudio Cloud (https://rstudio.cloud/)
  - UofT JupyterHub RStudio (<a href="https://jupyter.utoronto.ca/hub/login">https://jupyter.utoronto.ca/hub/login</a>)
- R & Notebook in the Cloud (run R without installation)
  - UofT JupyterHub Notebook (<a href="https://jupyter.utoronto.ca/hub/login">https://jupyter.utoronto.ca/hub/login</a>)
  - Google Colab (<a href="https://colab.to/r">https://colab.to/r</a>)

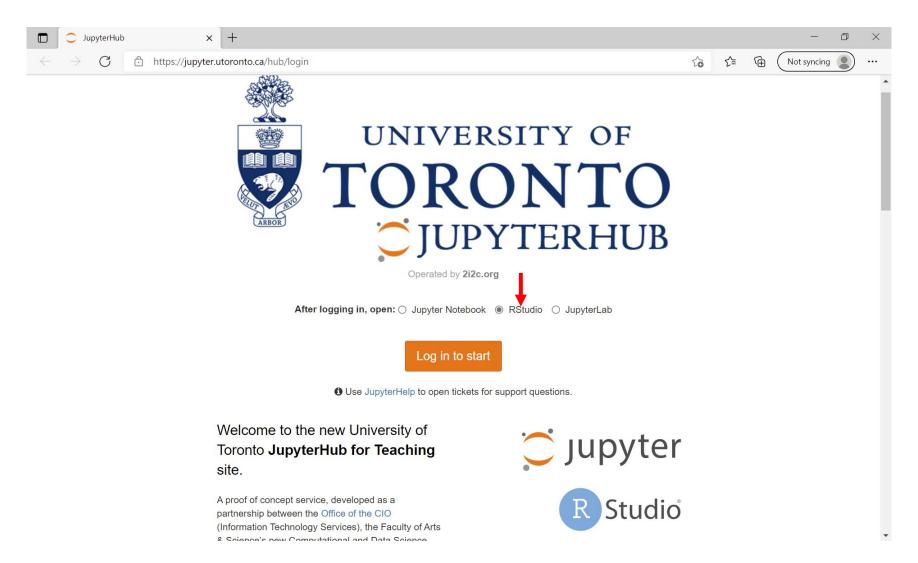
# What's RStudio? R Studio



#### RStudio Cloud

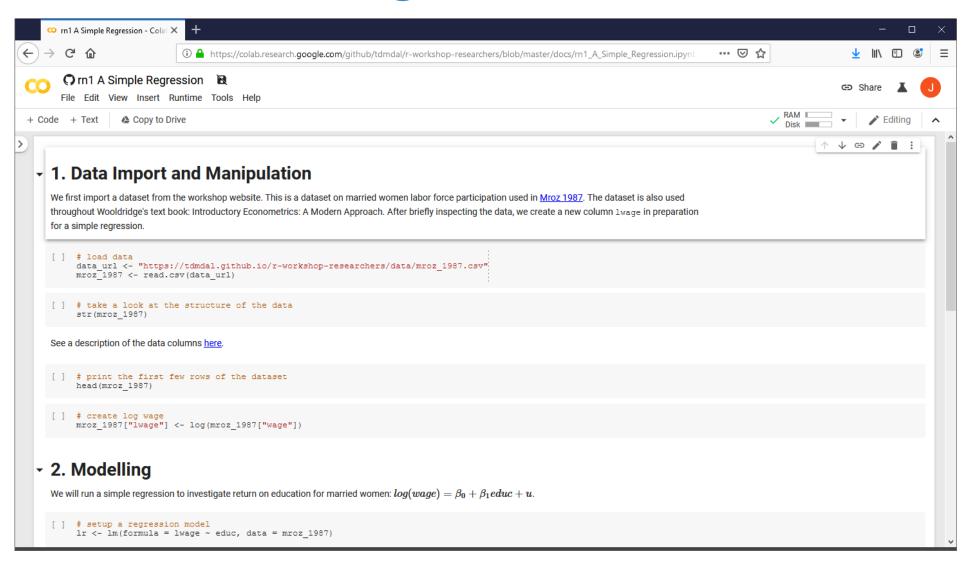


## RStudio at UofT Jupyterhub

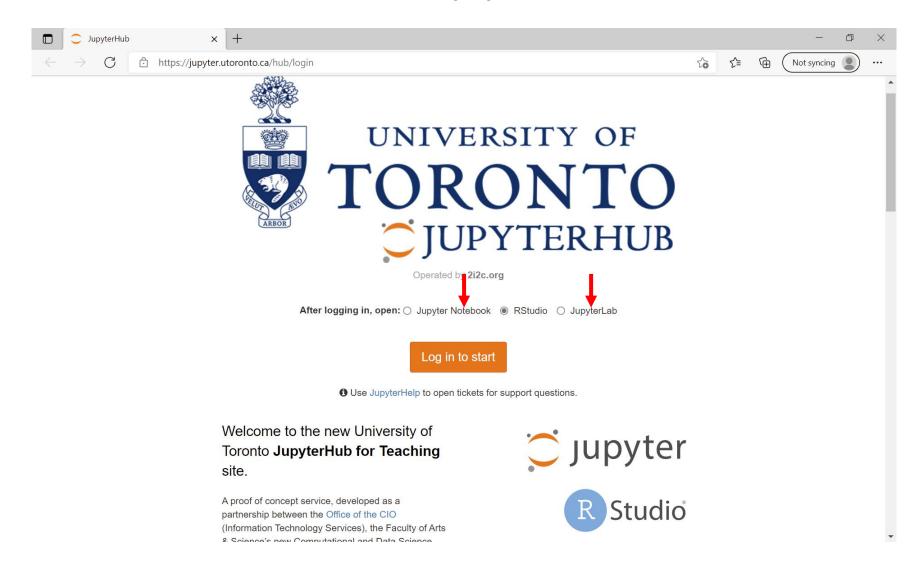


# R Notebook in Google Colab





## R Notebook at UofT Jupyterhub



## Plan for Today

- Intro to Intro
  - What is R and what can R do?
  - Setup R
  - Motivation examples
- Overview of R programing and Data Science
  - Basics of R programming
  - Data science with R
- Learning Resources and Road Map

#### A Few Examples

Analyze portfolio performance

 Look for trends in R community through Twitter

Recognize handwritten digits - an example of deep learning



PerformanceAnalytics
Package









#### A Few Examples: What to Look For

- Focus on analysis workflow (by reading the code comments)
  - Import and manipulate data
  - Model data
  - Report and visualize results
- Don't focus on R syntax

- Do notice everything is done in a sequential way
  - no conditional branching or looping

## Plan for Today

- Intro to Intro
- Overview of R programing and Data Science
  - Basics of R programming
    - Expression & Assignment
    - Data Structure
    - Programming Structure (control flow & function)
    - Turn ideas into code
  - Data science with R
- Learning Road Map and Resources

## Expression and Assignment

```
# expression
2 + sqrt(4) + log(exp(2)) + 2^2
# assignment
x < -3
y \leftarrow (pi == 3.14)
```

#### R Data Structure - Overview

	Homogeneous	Heterogeneous
1-d	<b>Atomic vector</b>	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
                                                          World!
vec_character <- c("Hello,", "World!")</pre>
                                                 Hello,
vec_integer <- c(1L, 2L, 3L)</pre>
                                                             3
vec double < c(1.1, 2.2, 3.3)
                                             1.1 2.2
                                                            3.3
vec_logical <- c(TRUE, TRUE, FALSE)</pre>
                                                  TRUE
                                            TRUE
                                                           FALSE
```

#### List

#### Data Frame

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

X	У	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
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)
)</pre>

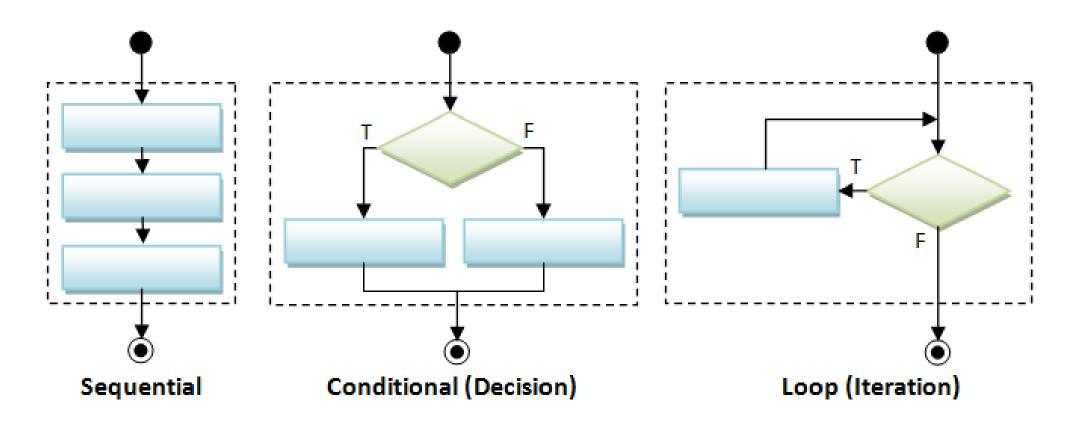
x y z
1 "a" 1.1
2 "b" 2.2
3 "c" 3.3
```

#### A Cousin to Data Frame - Tibble

```
# load tibble library (part of tidyverse lib)
library(tibble)
# create a tibble
tb1 <- tibble(</pre>
  x = 1:3,
  y = letters[1:3],
  z = c(1.1, 2.2, 3.3)
```

https://r4ds.had.co.nz/tibbles.html#tibbles-vs.data.frame

#### Programming Structure: Control Flows



**Today** 

**Learn on your own (See Appendix)** 

## Sequential

• Example: Sum of Squares

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

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

## Sequential

• Example: Sum of Squares

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

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

## Sequential

• Example: Sum of Squares

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

```
# sum of squares
t <- 1:3
y \leftarrow sum(t^2)
print(y)
                    1 4 9
              t^2
         sum(t^2) 14
```

#### 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) {</pre>
  t <- 1:n
  sum(t^2)
# calling the ss() function
print(ss(2))
print(ss(3))
```

#### 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) {</pre>
  t <- 1:n
  sum(t^2)
# calling the ss() function
print(ss(2))
print(ss(3))
```

#### 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) {</pre>
  t <- 1:n
  sum(t^2) # return(sum(t^2))
# calling the ss() function
print(ss(2))
print(ss(3))
```

#### Turn Ideas into Code

- Solve problems using code: three main ingredients
  - Data Structure + Programming Structure + Algorithm (sorting, searching, optimization, etc.)
  - Examples
    - Sort a list of integers
    - Generate and solve Sudoku puzzles
    - Implement and backtest a trading strategy
- For us (data analysis tasks), in most cases, we solve problems by
  - Using algorithm implemented by other peoples (i.e., functions from R packages)
  - Combine algorithms (and data & programming structures) to achieve our goal (still not easy; need practices to write good code.)

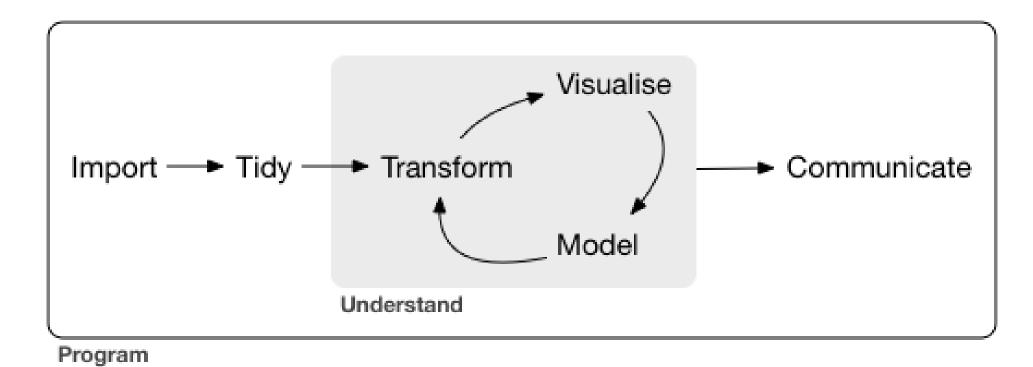
# Plan for Today

Intro to Intro

- Overview of R programing and Data Science
  - Basics of R programming
  - Data science with R
    - A Typical data analysis workflow
    - Choice of R packages
    - An example: regression analysis
- Learning Road Map and Resources

# Data Science/Analysis Workflow

Use this workflow to organize your thoughts and code



https://r4ds.had.co.nz/introduction.html

# An Example: Housing Price & Clean Air

- Manipulate data
  - Load data
  - Create new columns
  - Filter columns and rows
- Build models
  - Multiple linear regressions

Obs: 506

1.	price	median	housing	price,	\$
----	-------	--------	---------	--------	----

- 2. crime crimes committed per capita
- 3. **nox** nitrous oxide, parts per 100 mill.
- 4. rooms avg number of rooms per house
- 5. dist weighted dist. to 5 employ centers
- 6. radial accessibiliy index to radial hghwys
- 7. proptax property tax per \$1000
- 8. stratio average student-teacher ratio
- 9. lowstat % of people 'lower status'

- Report and graph
  - Build a publication-ready table for regression results

### Many Ways to Achieve the Same Goal

- The "pure" R way if possible
  - Mostly use functions/packages in <u>R standard library</u> (those shipped with R)
    - for structuring and manipulating data
    - for modeling if possible (e.g. regressions)
  - An example of a simple linear regression
- The "modern" way
  - Use specialized packages to manipulate data and assist modeling tasks
    - Data are stored in improved data structures (in most cases still compatible with base R data structure)
  - What we will focus on

### R Packages: Many choices, which one to use

- Often, a task can be achieved using functions in different libraries
  - R is open and extensible!

- Example: load a csv file to a data frame (or its variations)
  - Use read.csv() function from the utils library in Base R
  - Use <u>read csv()</u> function from the <u>readr</u> library
  - Use <a href="fread">fread()</a> function from the <a href="data.table">data.table</a> library
  - Use <a href="mailto:vroom">vroom</a> library

### R Packages: 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 date
  - packages maintained by companies (e.g., RStudio Co.) or academic teams

Choose one that suits your task

# Great Choice for Data Science Work,

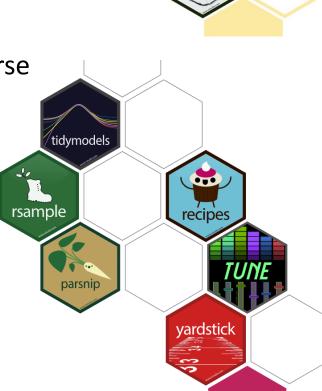
#### Tidyverse

- "an opinionated <u>collection of R packages</u> designed for data science"
- "All packages share an underlying design philosophy, grammar, and data structures."
- Handle data manipulation, visualization, and much more

 an eco-system: many package developers started to follow tidyverse principles too

#### <u>Tidymodels</u>

- "a collection of packages for modeling and machine learning using <u>tidyverse</u> principles"
- Manage modeling process but does not do modeling itself



readr

#### Our Choice: the Regression Example

- Manipulate data (<u>tidyverse</u> eco-system)
  - Load data (read csv() from the readr)
  - Create new columns (<u>mutate()</u> from <u>dplyr</u>)
  - Filter columns and rows (<u>select()</u> and <u>filter()</u> from <u>dplyr</u>)
- Build models
  - Multiple regression (<u>lm()</u> from stats library in R base)
- Report and graph
  - Build a publication-ready table (<u>huxreg()</u> from <u>huxtable</u> library)

# Using R packages/libraries

• Install an R library (only need to install a library once)

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

Load an R library (before you use a library)

```
library(library_name)
```

- <u>CRAN</u> (The Comprehensive R Archive Network)
  - CRAN Task Views

#### Load a CSV file

• read csv() from the readr

```
read_csv(file)
```

```
e.g. hprice <- read_csv("hprice.csv")</pre>
```

- More about <u>read csv()</u>
- More about <a href="readr">readr</a>

#### Load Data – Other file formats and sources

- readxl for Excel sheets
- <a href="haven">haven</a> for SPSS, Stata and SAS data
- jsonlite for JSON
- xml2 for XML
- <a href="httr">httr</a> for web APIs
- <u>rvest</u> for web scraping
- DBI for connecting to DataBase engine

• ...

#### Load Data - Financial Dataset

- tq get() from tidyquant library
  - collect financial and economic data from many online sources
    - Yahoo Finance, FRED, Quandl, Tiingo, Alpha Vantage, Bloomberg
- simfinapi library
  - download financial statements balance sheet, cash flow and income statement – and adjusted daily price of stocks through the simfin project
- a few others (try to look for them yourselves...)

#### A Quick Detour - Load Data from Web API

- Modern Web API: An example using Yahoo Finance
  - Yahoo Finance API Specification
  - How does a call look like
    - Apple on Yahoo Finance from the web, <a href="https://finance.yahoo.com/quote/AAPL/history?p=AAPL">https://finance.yahoo.com/quote/AAPL/history?p=AAPL</a>
    - An API call for AAPL recommendation trend data, <a href="https://query2.finance.yahoo.com/v11/finance/quoteSummary/AAPL?formatted=true&modules=recommendationTrend&corsDomain=finance.yahoo.com">https://query2.finance.yahoo.com/v11/finance/quoteSummary/AAPL?formatted=true&modules=recommendationTrend&corsDomain=finance.yahoo.com</a>
  - R packages to get data from Yahoo finance (API call wrappers), quantmod; tidyquant
- An R package to deal with raw Web API call, <a href="https://https://html.ncbi.nlm.ncbi.n
  - You can build your own Yahoo Finance R package using it

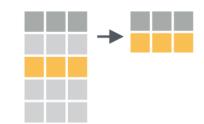
#### A Quick Detour – Get Data From Bloomberg

- Bloomberg is working hard on its Web APIs
  - https://www.bloomberg.com/professional/blog/modernizing-capitalmarkets-tech-with-web-apis/
- Meanwhile, you can get data using its traditional <u>server API</u>
  - To connect, need authentication (i.e. hostname, userid, etc.)
- R packages
  - Rblpai (a wrapper of Bloomberg Server API)
  - Tidyquant's tq get(x, "rblpapi") (a wrapper of Rblpai package)

#### Data Manipulation: dplyr basics

• Filter observations (rows): <a href="filter">filter()</a>

```
filter(my_dataframe, condition1, ...)
e.g., hprice_reg <- filter(hprice, price > 20000)
```



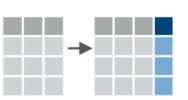
• Select variables (columns): select()

```
select(my_dataframe, var1, ...)
e.g., hprice_reg <- select(hprice_reg, lprice, rooms)</pre>
```



• Create new variables: <a href="mutate()">mutate()</a>

```
mutate(my_dataframe, new_var1 = expression1, ...)
e.g., hprice_reg <- mutate(hprice_reg, lprice = log(price))</pre>
```



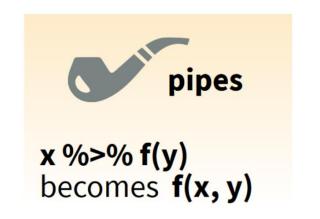
Ref. dplyr data wrangling <u>cheat sheet</u>

# Data Manipulation: Data Pipe (%>%)

```
hprice_reg <- filter(hprice, price > 20000)
hprice_reg <- mutate(hprice_reg, lprice = log(price))
hprice_reg <- select(hprice_reg, lprice, rooms)</pre>
```



```
hprice_reg <- hprice %>%
  filter(price > 20000) %>%
  mutate(lprice = log(price)) %>%
  select(lprice, rooms)
```



Ref. dplyr data wrangling cheat sheet

#### Regression

• Multiple regressions: <u>lm()</u> from stats library in base R

my\_model <- lm(y 
$$\sim$$
 x1 + x2, data) 
$$Y_i = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \epsilon_i$$

my\_model <- lm(y ~ x1 + x2 + I(x1 \* x2), data) 
$$Y_i = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 X_1 X_2 + \epsilon_i$$

Regression result summary: summary()

Ref. <a href="https://faculty.chicagobooth.edu/richard.hahn/teaching/FormulaNotation.pdf">https://faculty.chicagobooth.edu/richard.hahn/teaching/FormulaNotation.pdf</a>

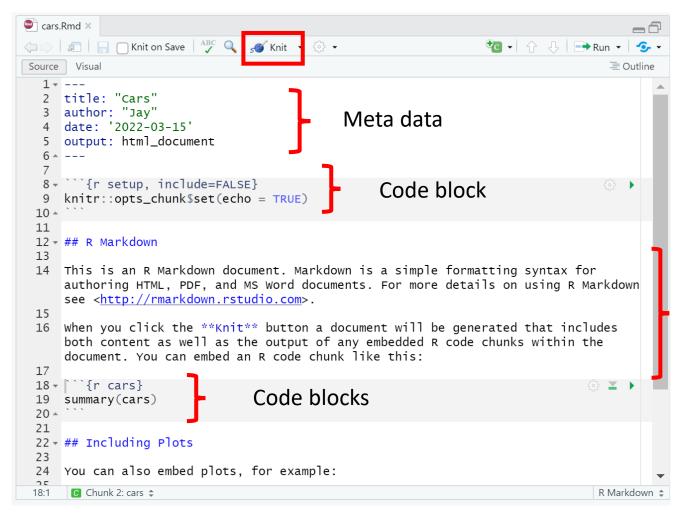
#### Report

- Summary table
  - <u>Summary for Im()</u>: summary(my\_model)
- publication-ready table: <a href="https://example.com/huxtable">huxreg()</a> from <a href="https://example.com/huxtable">huxtable</a> library

```
huxtable(my model1, my model2, ...)
```

Ref. <a href="https://hughjonesd.github.io/huxtable/huxreg.html">https://hughjonesd.github.io/huxtable/huxreg.html</a>

# Report - Publish to The Web (1) / R Markdown

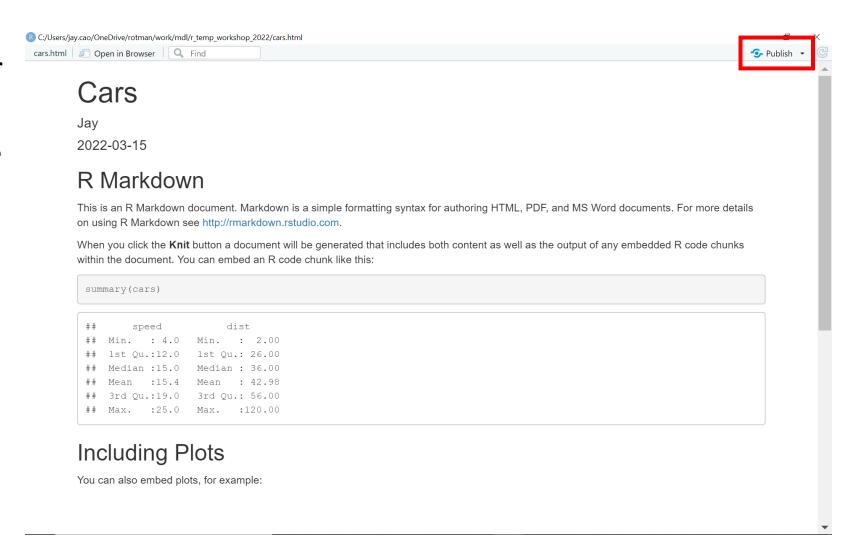


R Markdown text

Ref. 1) <a href="https://rmarkdown.rstudio.com/lesson-1.html">https://rmarkdown.rstudio.com/lesson-1.html</a>; R Markdown <a href="https://cheat.Sheet">Cheat.Sheet</a>

# Report - Publish to The Web (2) / R Markdown

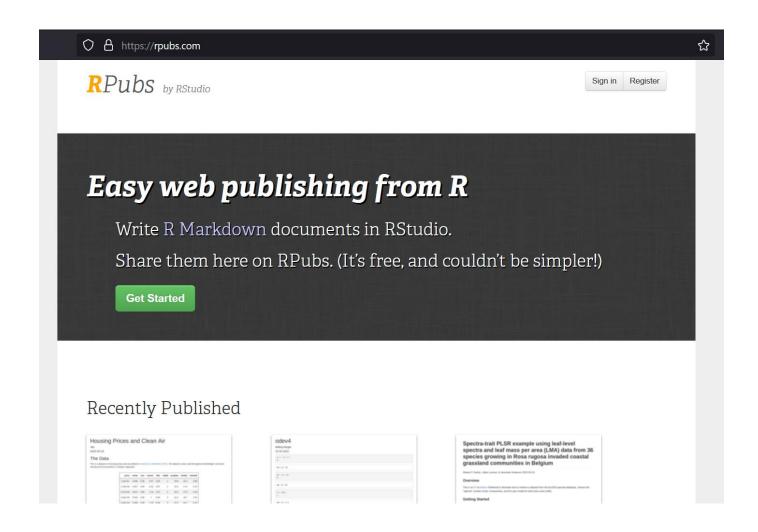
- Knit button for rendering
  - knitr package



### Report - Publish to The Web (3) / RPubs

 Publish button for publishing

- RPubs
  - Publish and share your RMarkdown output
- Demo



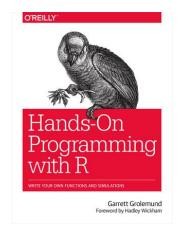
# Plan for Today

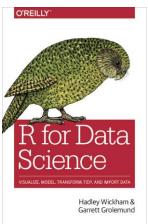
Intro to Intro

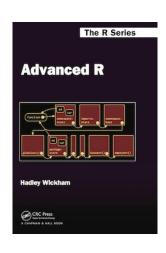
- Overview of R programing and Data Science
  - Basics of R programming
  - Data science with R
- Learning Resources and Road Map

# Learning Road Map (Three Free Books)

- Step 1. Basic R programming skills (Never programmed before? Start here.)
  - Data and programming structure; how to turn an idea into code;
  - Book: <u>Hands-On Programming with R</u>
- Step 2. R Data Science skills
  - Data wrangling, modeling, and visualization/reporting; Best practice;
  - Book: R for Data Science
- Step 3. Take your R Skill to the next level
  - Book: Advanced R







# Learning Approach

- Learn the underlying principles
  - e.g., why organize data in a certain way
- Learn best practices
  - follow a consistent analysis workflow

#### Free Learning Resource

- RStudio Education
  - Choose Your Learning Paths
- RStudio Video Resources Site
- More free R books? Check <u>bookdown.org</u> often
- Coursera: Search R and learn
  - free for <u>UofT students</u> (mostly always free if you just audit the courses)
- Twitter (a few seeds: #rstat, @hadleywickham, @WeAreRLadies)

# Appendix

- Programming Structure Continued
  - Conditional
  - Iteration

### 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 (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")
              y>10?
     "great..."
                      "less..."
```

### 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)
```