Crash Course on Data Analytics

Assoc Prof Peter Julian Cayton, PhD

2024-09-23

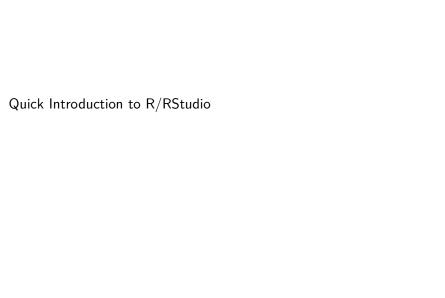
Flow of Presentation

- 1. Quick Introduction to R/RStudio
- 2. A Typical Data Analytics Workflow
- 3. A Crash Course Demonstration
- 4. Case Study: Collaborative Data Science and COVID-19
- 5. Further Sources and References

Materials Available Online

They may be found in the website below:

 $https://github.com/pacayton/Crash_course_On_Data_Analytics$



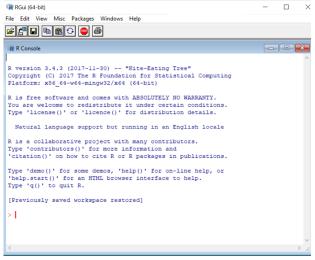
Quick Introduction to R/RStudio

R Programming Language

- ▶ R is a freely available language and environment for statistical computing and graphics which provides a wide variety of statistical and graphical techniques: linear and nonlinear modelling, statistical tests, time series analysis, classification, clustering, etc. (source: The Comprehensive R Archive Network)
- Website: https://cran.r-project.org/

Quick Introduction to R/RStudio

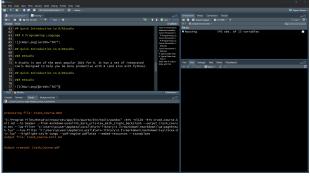
R Programming Language

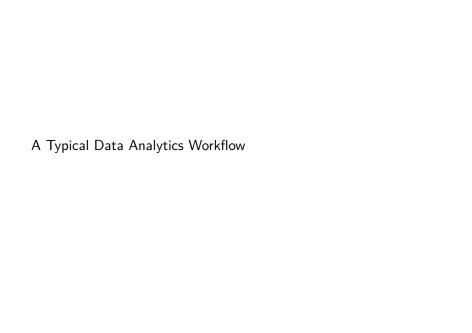


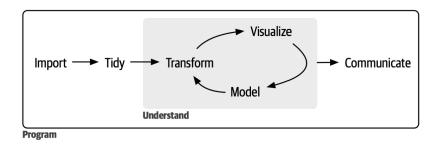
Quick Introduction to R/RStudio

RStudio

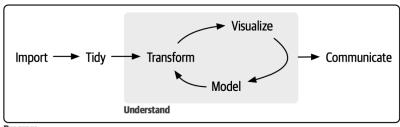
- RStudio is one of the most popular IDEs for R. It has a set of integrated tools designed to help you be more productive with R (and also with Python).
- Website: https://posit.co/downloads/







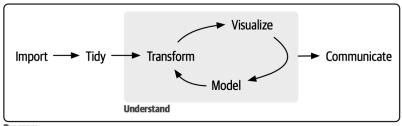
Source: Wickham, H and Grolemund, G (2017). R for Data Science. O'Reilly. https://r4ds.had.co.nz/introduction.html



Program

Import

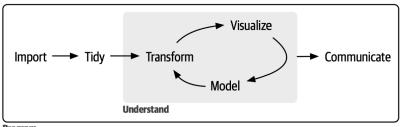
 Extracting the data from an internal database, a file, an online website, or thru a web application programming interface (API), to be loaded in R/RStudio



Program

Tidy

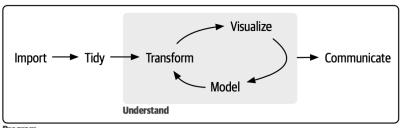
- Arranging the data into a neat data structure, with variable as columns and data points as rows.
- Included in this step would be data cleaning, data augmentation, missing data imputation, and others



Program

Understand

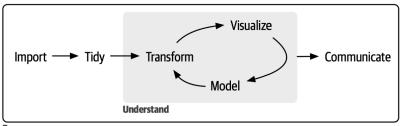
Generally, the steps to extract insights from data after tidying up.



Program

Transform

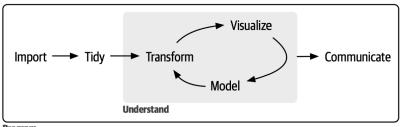
- Processing the data in preparation for further steps. Examples are:
- 1. Narrowing the data, e.g., by region or by age,
- Computing new variables, e.g., length of days until recovery, or delays in reporting cases
- 3. Aggregating data, e.g., counting cases or solving rates/means



Program

Visualize

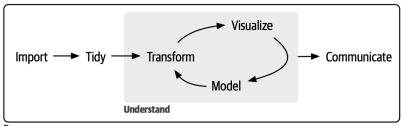
▶ Plot data into graphs so that patterns and features may be explored and insights be extracted from what is seen.



Program

Model

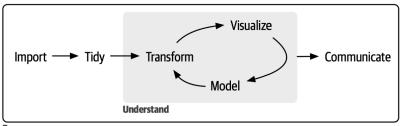
- ► When necessary, models help in summarizing the complex relationships and the patterns found from visualizations
- Designing models for prediction or forecasting



Program

Communicate

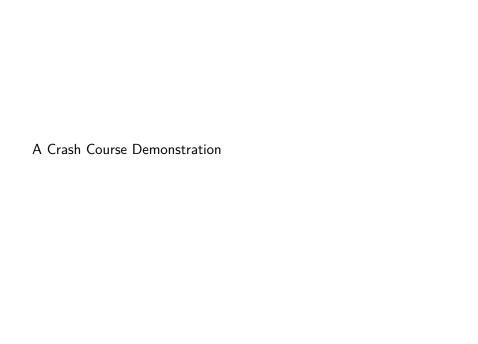
Writing reports, creating dashboards, making presentations, compilations, etc.



Program

Program

- ► All these processes to be encapsulated in a data science project plan
- Possible to be encapsulated in one software, but it's not impossible to use more than one depending on team members' capabilities to process and analyze data.



Data for Demonstration:

M Yasser H. Housing Prices Dataset.

https://www.kaggle.com/datasets/yasserh/housing-prices-dataset

Program

- upload the necessary packages
- pre-installation of packages should be done first

```
## Preamble Code
# install.packages("tidyverse", "knitr", "kableExtra",
#
                    "moments", "stargazer")
## Packages to Use
library(tidyverse)
library(knitr)
library(kableExtra)
library(moments)
library(stargazer)
```

Program

below is a set of codes for me to compute descriptive statistics later.

```
### generate basic stats
stat vec <- function(x) {</pre>
  v <-as.numeric(x)
  vec \leftarrow c(mean(v, na.rm = TRUE),
            sd(v, na.rm = TRUE),
            skewness(v, na.rm = TRUE),
            kurtosis(v, na.rm = TRUE))
  return(vec)
```

Import

- get the dataset from an online repository (Github)
- ▶ location of the file: https://github.com/pacayton/Crash_cour se_On_Data_Analytics/raw/main/Housing.csv

housing <- read_csv("https://github.com/pacayton/Crash_cour</pre>

Import

```
housing[,1:3]
   # A tibble: 545 x 3
##
         price area bedrooms
##
         <dbl> <dbl>
                          <dbl>
                7420
##
      13300000
    2 12250000
##
                 8960
    3 12250000
                9960
                              3
##
##
    4 12215000
                 7500
##
    5 11410000
                 7420
##
      10850000
                 7500
                              3
##
      10150000
                 8580
##
      10150000
                16200
                              5
       9870000
                 8100
##
    9
                              4
       9800000
                 5750
                              3
##
   10
   # i 535 more rows
```

Import

```
summary(housing[,1:3])
## price area bedrooms
## Min. : 1750000 Min. : 1650 Min. :1.000
```

```
1st Qu.: 3430000
                                   1st Qu.:2.000
##
                     1st Qu.: 3600
##
   Median: 4340000
                     Median: 4600
                                   Median :3.000
##
   Mean : 4766729
                     Mean : 5151
                                   Mean :2.965
##
   3rd Qu.: 5740000
                     3rd Qu.: 6360
                                   3rd Qu.:3.000
##
   Max. :13300000
                     Max. :16200
                                   Max. :6.000
```

colnames(housing)

- [1] "price" "area" "bedrooms" "bathrooms"
- [5] "stories" "mainroad" "guestroom" "basement"
- [9] "hotwaterheating" "airconditioning" "parking" "prefarea"
- [13] "furnishingstatus"

Tidy

- ▶ fix the data for easier processing later
- ex: price_in_thousands & with_parking

```
housing <- housing %>%
  mutate(
    price_in_thousands = price / 1000,
    with_parking = ifelse(parking > 0, TRUE, FALSE)
)
```

Tidy

```
head(housing[,c("price", "parking",
                 "price_in_thousands", "with_parking")])
## # A tibble: 6 x 4
##
        price parking price_in_thousands with_parking
##
        <dbl>
                <dbl>
                                    <dbl> <lgl>
## 1 13300000
                                    13300 TRUE
                     3
## 2 12250000
                                    12250 TRUE
## 3 12250000
                                    12250 TRUE
## 4 12215000
                     3
                                    12215 TRUE
## 5 11410000
                                    11410 TRUE
## 6 10850000
                                    10850 TRUE
```

Understand / Transform

► Computing basic statistics

Understand / Transform

▶ Presenting basic statistics on a table

kable(t(stats_table), format="pipe")

| comp_stats |
|------------|
| 4766.73 |
| 1870.44 |
| 1.21 |
| 4.93 |
| |

Understand / Transform

 Presenting average house price by whether they are next to a main road (yes/no)

| mainroad | Average House Price (in thousands) |
|----------|------------------------------------|
| no | 3398.905 |
| yes | 4991.777 |

Understand / Transform

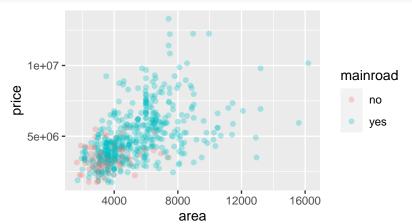
Presenting average house price by furnishing status (un/semi/furnished)

| furnishingstatus | Average House Price (in thousands) |
|------------------|------------------------------------|
| furnished | 5495.696 |
| semi-furnished | 4907.524 |
| unfurnished | 4013.831 |

Understand / Transform

Presenting average house price by parking availability (TRUE/FALSE)

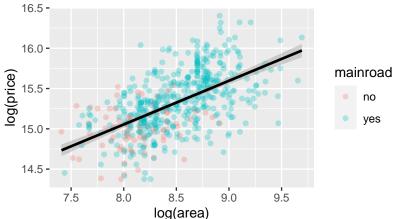
| with_parking | Average House Price (in thousands) |
|--------------|------------------------------------|
| FALSE | 4136.017 |
| TRUE | 5533.327 |



```
ggplot(housing) +
  geom_point(aes(x = log(area), y = log(price),
                    color = mainroad), alpha = 0.3)
    16.5 -
    16.0 -
log(price)
                                                      mainroad
   15.5 -
                                                           no
                                                           yes
    14.5 -
          7.5
                   8.0
                           8.5
                                    9.0
                                             9.5
                         log(area)
```

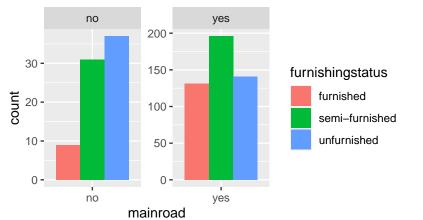
```
ggplot(housing) +
  geom_point(aes(x = log(area), y = log(price),
                   color = with_parking), alpha = 0.3)
    16.5 -
    16.0 -
log(price)
                                                  with parking
   15.5 -
                                                       FALSE
                                                       TRUE
    14.5 -
          7.5
                  8.0
                         8.5
                                 9.0
                                         9.5
                       log(area)
```

```
ggplot(housing, aes(x = log(area), y = log(price))) +
  geom_point(aes(color = mainroad), alpha = 0.3) +
  geom_smooth(method = "lm", color = "black")
```



```
ggplot(housing, aes(x = mainroad)) +
  geom_bar(aes(fill = furnishingstatus), stat = "count",
             position = "dodge")
   200 -
   150 -
                                             furnishingstatus
                                                 furnished
 count
   100 -
                                                 semi-furnished
                                                 unfurnished
    50 -
                no
                               ves
                    mainroad
```

A Crash Course Demonstration Understand / Visualize



Understand / Model

we assume the following:

$$Price_{in thousands} = \beta_0 + \beta_1 \times area + \beta_2 \times bedrooms + \epsilon$$

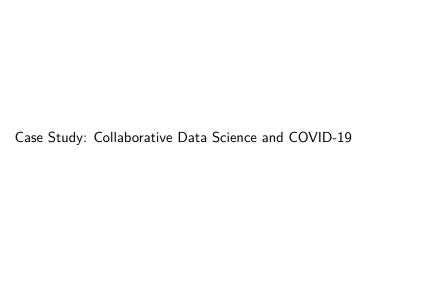
Understand / Model

▶ the result is:

 $PredictedPrice_{in\ thousands} = 391.126 + 0.424 \times area + 739.566 \times bedrooms$

Communicate

► The current slide presentation is produced within the RStudio interface, helping with both writing reports and running statistical procedures in 1 code file!



Case Study: Collaborative Data Science and COVID-19

You can access the case study handout here:

https://github.com/pacayton/Crash_course_On_Data_Analytics/blob/main/Collaborative%20Data%20Science%20and%20COVID-19.pdf



Further Sources and References

Offered Trainings

- Philippine Statistical Research and Training Institute: https://psrti.gov.ph/ (Website currently under maintenance) or https://www.facebook.com/PSRTI.Official
- ▶ UP Statistical Center Research Foundation: For more details on trainings offered, follow https://www.facebook.com/UPDStat

Online Reference on Data Science using R/RStudio

R for Data Science (2e): https://r4ds.hadley.nz/

