

Module 2: Reporting, Data Wrangling and Graphing (I)

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Outline

We will review R, Rstudio, and Syntax of R together.

- LaTeX/Markdown
- Tidy data, processing (tidyverse)
- Graphing (ggplot2)

LaTeX and Markdown

LaTeX is useful for documents with mathematical formulas.

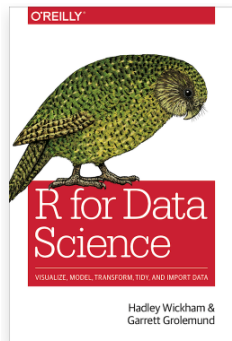
- Overleaf - an online, collaborative LaTeX editor
- LaTeX mathematical symbols
- Inline equation e.g. (`\alpha`) returns α
- Equation e.g. (`e = mc^2`) returns

$$e = mc^2$$

Markdown is appealing for formatting, e.g. headings, bold text, text with codes, ...

Resources

“R for Data Science: Import, Tidy, Transform, Visualize, and Model Data”
by Hadley Wickham.



Data import

```
df <- read.table("mtcars.txt", header = TRUE)
head(df) # Show the first 6 rows.
```

```
##   Cntry lper100k weight length
## 1   US      19.8   2178   5.92
## 2 Japan      9.9   1026   4.32
## 3   US     10.8   1188   4.27
## 4   US     12.5  1444   5.11
## 5   US     12.5  1485   5.03
## 6   US     12.5  1485   5.03
```

Other options

CSV files.

- `base::read.csv()` in the base `r`.
- `readr::read.csv()` in “`readr`” package (much faster).
- `data.table::fread()` in “`data.table`” package (much more faster).

Rdata.

- `load()` in the base `r`.

Tidy data

The goal is to clean the dataset so it is much easier to use.

Specifically,

- Each variable must have its own column.
- Each observation must have its own row.
- Each value must have its own cell.

We will focus on the functions from “tidyverse” package.

```
library(tidyverse)
```

Tidy data 1: pivoting

For a dataset having column names are not names of variables, but values of a variable, e.g.

```
table4a
```

```
## # A tibble: 3 x 3
##   country    `1999` `2000`
##   <chr>      <dbl> <dbl>
## 1 Afghanistan    745   2666
## 2 Brazil        37737  80488
## 3 China         212258 213766
```

- Need to change 1999, 2000 to a column? named as "year".
- Need to change the values of 1999, 2000 as "cases".

We can use `pivot_longer()` from the "tidyverse" package.

Pivot longer

```
table4a %>%  
  pivot_longer(c(`1999`, `2000`),  
    names_to = "year", values_to = "cases")
```

```
## # A tibble: 6 x 3  
##   country    year  cases  
##   <chr>    <chr> <dbl>  
## 1 Afghanistan 1999     745  
## 2 Afghanistan 2000    2666  
## 3 Brazil      1999   37737  
## 4 Brazil      2000   80488  
## 5 China       1999  212258  
## 6 China       2000  213766
```

Another example

```
table2 %>% head(5)
```

```
## # A tibble: 5 x 4
##   country      year type      count
##   <chr>      <dbl> <chr>    <dbl>
## 1 Afghanistan 1999 cases      745
## 2 Afghanistan 1999 population 19987071
## 3 Afghanistan 2000 cases      2666
## 4 Afghanistan 2000 population 20595360
## 5 Brazil      1999 cases     37737
```

- case and population are two variables and should be converted into columns.

We can use `pivot_wider()`.

Pivot wider

```
table2 %>%  
  pivot_wider(names_from = type, values_from = count)
```

```
## # A tibble: 6 x 4  
##   country      year  cases population  
##   <chr>      <dbl> <dbl>      <dbl>  
## 1 Afghanistan 1999     745    19987071  
## 2 Afghanistan 2000    2666   20595360  
## 3 Brazil      1999   37737   172006362  
## 4 Brazil      2000   80488   174504898  
## 5 China       1999  212258  1272915272  
## 6 China       2000  213766  1280428583
```

Transform data

Use the “pipes” from the “tidyverse” package, a powerful tool for clearly expressing a sequence of multiple operations, with the combination of the following functions:

- `select()`
- `filter()`
- `arrange()`
- `mutate()`
- `summarise()`
- `group_by()`

Dataset - Diamonds

A dataset containing the prices and other attributes of almost 54,000 diamonds.

```
head(diamonds)
```

```
## # A tibble: 6 x 10
##   carat cut      color clarity depth table price      x      y      z
##   <dbl> <ord>    <ord> <ord>    <dbl> <dbl> <int> <dbl> <dbl> <dbl>
## 1  0.23 Ideal    E     SI2     61.5   55   326   3.95   3.98   2.43
## 2  0.21 Premium E     SI1     59.8   61   326   3.89   3.84   2.31
## 3  0.23 Good    E     VS1     56.9   65   327   4.05   4.07   2.31
## 4  0.29 Premium I     VS2     62.4   58   334   4.2    4.23   2.63
## 5  0.31 Good    J     SI2     63.3   58   335   4.34   4.35   2.75
## 6  0.24 Very Good J     VVS2     62.8   57   336   3.94   3.96   2.48
```

Select

Use `select()` to get a column, e.g. “color”

```
diamonds %>%  
  select(color) %>%  
  head()
```

```
## # A tibble: 6 x 1  
##   color  
##   <ord>  
## 1 E  
## 2 E  
## 3 E  
## 4 I  
## 5 J  
## 6 J
```

```
# Equivalent to...  
head(diamonds$color)
```

```
## [1] E E E I J J  
## Levels: D < E < F < G < H < I < J
```

Select

Use `select()` to remove a column, e.g. “color”

```
diamonds %>%  
  select(-color)
```

```
## # A tibble: 53,940 x 9  
##   carat cut      clarity depth table price      x      y      z  
##   <dbl> <ord>    <ord>    <dbl> <dbl> <int> <dbl> <dbl> <dbl>  
## 1  0.23 Ideal    SI2      61.5    55   326   3.95   3.98   2.43  
## 2  0.21 Premium SI1      59.8    61   326   3.89   3.84   2.31  
## 3  0.23 Good     VS1      56.9    65   327   4.05   4.07   2.31  
## 4  0.29 Premium VS2      62.4    58   334   4.2    4.23   2.63  
## 5  0.31 Good     SI2      63.3    58   335   4.34   4.35   2.75  
## 6  0.24 Very Good VVS2     62.8    57   336   3.94   3.96   2.48  
## 7  0.24 Very Good VVS1     62.3    57   336   3.95   3.98   2.47  
## 8  0.26 Very Good SI1      61.9    55   337   4.07   4.11   2.53  
## 9  0.22 Fair     VS2      65.1    61   337   3.87   3.78   2.49  
## 10 0.23 Very Good VS1      59.4    61   338   4      4.05   2.39  
## # i 53,930 more rows
```

Need to assign the change to the original dataset, otherwise, the deletion won't affect the dataset.

```
diamonds <- diamonds %>%  
  select(-color)
```

Filter

Use `filter()` to filter by some condition, e.g. filter all price > 335

```
diamonds %>%  
  filter(price > 335)
```

```
## # A tibble: 53,935 x 10  
##   carat cut      color clarity depth table price      x      y      z  
##   <dbl> <ord>    <ord> <ord>    <dbl> <dbl> <int> <dbl> <dbl> <dbl>  
## 1  0.24 Very Good J      VVS2    62.8    57   336   3.94   3.96   2.48  
## 2  0.24 Very Good I      VVS1    62.3    57   336   3.95   3.98   2.47  
## 3  0.26 Very Good H      SI1     61.9    55   337   4.07   4.11   2.53  
## 4  0.22 Fair      E       VS2     65.1    61   337   3.87   3.78   2.49  
## 5  0.23 Very Good H      VS1     59.4    61   338    4      4.05   2.39  
## 6  0.3   Good      J      SI1     64      55   339   4.25   4.28   2.73  
## 7  0.23 Ideal      J      VS1     62.8    56   340   3.93   3.9    2.46  
## 8  0.22 Premium F      SI1     60.4    61   342   3.88   3.84   2.33  
## 9  0.31 Ideal      J      SI2     62.2    54   344   4.35   4.37   2.71  
## 10 0.2   Premium E      SI2     60.2    62   345   3.79   3.75   2.27  
## # i 53,925 more rows
```


Filters with multiple conditions

```
diamonds %>%  
  filter(price > 335 & depth < 64)
```

```
## # A tibble: 51,849 x 10  
##   carat cut      color clarity depth table price      x      y      z  
##   <dbl> <ord>    <ord> <ord>    <dbl> <dbl> <int> <dbl> <dbl> <dbl>  
## 1  0.24 Very Good J      VVS2    62.8    57    336  3.94  3.96  2.48  
## 2  0.24 Very Good I      VVS1    62.3    57    336  3.95  3.98  2.47  
## 3  0.26 Very Good H      SI1     61.9    55    337  4.07  4.11  2.53  
## 4  0.23 Very Good H      VS1     59.4    61    338   4     4.05  2.39  
## 5  0.23 Ideal      J      VS1     62.8    56    340  3.93  3.9  2.46  
## 6  0.22 Premium    F      SI1     60.4    61    342  3.88  3.84  2.33  
## 7  0.31 Ideal      J      SI2     62.2    54    344  4.35  4.37  2.71  
## 8  0.2  Premium    E      SI2     60.2    62    345  3.79  3.75  2.27  
## 9  0.32 Premium    E      I1     60.9    58    345  4.38  4.42  2.68  
## 10 0.3  Ideal      I      SI2     62     54    348  4.31  4.34  2.68  
## # i 51,839 more rows
```

Filters with multiple conditions

```
diamonds %>%  
  filter(cut == "Very Good" | cut == "Fair")
```

```
## # A tibble: 13,692 x 10  
##   carat cut      color clarity depth table price      x      y      z  
##   <dbl> <ord>    <ord> <ord>    <dbl> <dbl> <int> <dbl> <dbl> <dbl>  
## 1  0.24 Very Good J      VVS2    62.8   57   336   3.94   3.96   2.48  
## 2  0.24 Very Good I      VVS1    62.3   57   336   3.95   3.98   2.47  
## 3  0.26 Very Good H      SI1     61.9   55   337   4.07   4.11   2.53  
## 4  0.22 Fair     E      VS2     65.1   61   337   3.87   3.78   2.49  
## 5  0.23 Very Good H      VS1     59.4   61   338    4     4.05   2.39  
## 6  0.3  Very Good J      SI1     62.7   59   351   4.21   4.27   2.66  
## 7  0.23 Very Good E      VS2     63.8   55   352   3.85   3.92   2.48  
## 8  0.23 Very Good H      VS1     61     57   353   3.94   3.96   2.41  
## 9  0.31 Very Good J      SI1     59.4   62   353   4.39   4.43   2.62  
## 10 0.31 Very Good J      SI1     58.1   62   353   4.44   4.47   2.59  
## # i 13,682 more rows
```

Filter after select

This is an example of “a sequence of operations”.

```
diamonds %>%  
  select(price) %>%  
  filter(price > 335)
```

```
## # A tibble: 53,935 x 1  
##   price  
##   <int>  
## 1    336  
## 2    336  
## 3    337  
## 4    337  
## 5    338  
## 6    339  
## 7    340  
## 8    342  
## 9    344  
## 10   345  
## # i 53,925 more rows
```

Arrange

Use `arrange()` to order data.

```
diamonds %>%  
  arrange(price)
```

```
## # A tibble: 53,940 x 10  
##   carat cut      color clarity depth table price      x      y      z  
##   <dbl> <ord>    <ord> <ord>    <dbl> <dbl> <int> <dbl> <dbl> <dbl>  
## 1  0.23 Ideal    E      SI2     61.5    55   326   3.95   3.98   2.43  
## 2  0.21 Premium E      SI1     59.8    61   326   3.89   3.84   2.31  
## 3  0.23 Good    E      VS1     56.9    65   327   4.05   4.07   2.31  
## 4  0.29 Premium I      VS2     62.4    58   334   4.2    4.23   2.63  
## 5  0.31 Good    J      SI2     63.3    58   335   4.34   4.35   2.75  
## 6  0.24 Very Good J      VVS2    62.8    57   336   3.94   3.96   2.48  
## 7  0.24 Very Good I      VVS1    62.3    57   336   3.95   3.98   2.47  
## 8  0.26 Very Good H      SI1     61.9    55   337   4.07   4.11   2.53  
## 9  0.22 Fair    E      VS2     65.1    61   337   3.87   3.78   2.49  
## 10 0.23 Very Good H      VS1     59.4    61   338   4      4.05   2.39  
## # i 53,930 more rows
```

Arrange descending order

e.g. from the cheapest!

```
diamonds %>%  
  arrange(-price)
```

```
## # A tibble: 53,940 x 10  
##   carat cut      color clarity depth table price      x      y      z  
##   <dbl> <ord>    <ord> <ord>    <dbl> <dbl> <int> <dbl> <dbl> <dbl>  
## 1  2.29 Premium  I     VS2     60.8   60 18823  8.5   8.47  5.16  
## 2  2      Very Good G     SI1     63.5   56 18818  7.9   7.97  5.04  
## 3  1.51 Ideal   G     IF      61.7   55 18806  7.37  7.41  4.56  
## 4  2.07 Ideal   G     SI2     62.5   55 18804  8.2   8.13  5.11  
## 5  2      Very Good H     SI1     62.8   57 18803  7.95  8     5.01  
## 6  2.29 Premium  I     SI1     61.8   59 18797  8.52  8.45  5.24  
## 7  2.04 Premium  H     SI1     58.1   60 18795  8.37  8.28  4.84  
## 8  2      Premium  I     VS1     60.8   59 18795  8.13  8.02  4.91  
## 9  1.71 Premium  F     VS2     62.3   59 18791  7.57  7.53  4.7  
## 10 2.15 Ideal   G     SI2     62.6   54 18791  8.29  8.35  5.21  
## # i 53,930 more rows
```

Arrange by multiple conditions

```
diamonds %>%  
  arrange(price, cut)
```

```
## # A tibble: 53,940 x 10  
##   carat cut      color clarity depth table price      x      y      z  
##   <dbl> <ord>    <ord> <ord>    <dbl> <dbl> <int> <dbl> <dbl> <dbl>  
## 1  0.21 Premium E      SI1     59.8   61   326   3.89   3.84   2.31  
## 2  0.23 Ideal  E      SI2     61.5   55   326   3.95   3.98   2.43  
## 3  0.23 Good   E      VS1     56.9   65   327   4.05   4.07   2.31  
## 4  0.29 Premium I      VS2     62.4   58   334   4.2    4.23   2.63  
## 5  0.31 Good   J      SI2     63.3   58   335   4.34   4.35   2.75  
## 6  0.24 Very Good J      VVS2    62.8   57   336   3.94   3.96   2.48  
## 7  0.24 Very Good I      VVS1    62.3   57   336   3.95   3.98   2.47  
## 8  0.22 Fair   E      VS2     65.1   61   337   3.87   3.78   2.49  
## 9  0.26 Very Good H      SI1     61.9   55   337   4.07   4.11   2.53  
## 10 0.23 Very Good H      VS1     59.4   61   338   4      4.05   2.39  
## # i 53,930 more rows
```

Filter, select, arrange

```
diamonds %>%  
  filter(table < 340) %>%  
  select(carat, cut, price) %>%  
  arrange(price, cut)
```

```
## # A tibble: 53,940 x 3  
##   carat cut      price  
##   <dbl> <ord>   <int>  
## 1  0.21 Premium    326  
## 2  0.23 Ideal     326  
## 3  0.23 Good      327  
## 4  0.29 Premium    334  
## 5  0.31 Good      335  
## 6  0.24 Very Good  336  
## 7  0.24 Very Good  336  
## 8  0.22 Fair      337  
## 9  0.26 Very Good  337  
## 10 0.23 Very Good  338  
## # i 53,930 more rows
```

Mutate

Create new variables using `mutate()`.

- Create a boolean variable, 0 = not affordable, 1 = affordable.

```
diamonds %>%  
  mutate(affordable = price < 400)
```

```
## # A tibble: 53,940 x 11  
##   carat cut      color clarity depth table price      x      y      z affordable  
##   <dbl> <ord>    <ord> <ord>    <dbl> <dbl> <int> <dbl> <dbl> <dbl> <lgl>  
## 1  0.23 Ideal    E      SI2     61.5   55   326   3.95   3.98   2.43 TRUE  
## 2  0.21 Premium E      SI1     59.8   61   326   3.89   3.84   2.31 TRUE  
## 3  0.23 Good    E      VS1     56.9   65   327   4.05   4.07   2.31 TRUE  
## 4  0.29 Premium I      VS2     62.4   58   334   4.2    4.23   2.63 TRUE  
## 5  0.31 Good    J      SI2     63.3   58   335   4.34   4.35   2.75 TRUE  
## 6  0.24 Very Good J      VVS2    62.8   57   336   3.94   3.96   2.48 TRUE  
## 7  0.24 Very Good I      VVS1    62.3   57   336   3.95   3.98   2.47 TRUE  
## 8  0.26 Very Good H      SI1     61.9   55   337   4.07   4.11   2.53 TRUE  
## 9  0.22 Fair    E      VS2     65.1   61   337   3.87   3.78   2.49 TRUE  
## 10 0.23 Very Good H      VS1     59.4   61   338   4      4.05   2.39 TRUE  
## # i 53,930 more rows
```


Mutate (cont'd)

- Create a variable containing string with `case_when()`:

```
diamonds %>%  
  mutate(affordable = case_when(price<400 ~ "affordable",  
                                TRUE ~ "not affordable"))
```

```
## # A tibble: 53,940 x 11  
##   carat cut      color clarity depth table price      x      y      z affordable  
##   <dbl> <ord>    <ord> <ord>    <dbl> <dbl> <int> <dbl> <dbl> <dbl> <chr>  
## 1  0.23 Ideal      E      SI2      61.5    55   326   3.95   3.98   2.43 affordable  
## 2  0.21 Premium    E      SI1      59.8    61   326   3.89   3.84   2.31 affordable  
## 3  0.23 Good       E      VS1      56.9    65   327   4.05   4.07   2.31 affordable  
## 4  0.29 Premium    I      VS2      62.4    58   334   4.2    4.23   2.63 affordable  
## 5  0.31 Good       J      SI2      63.3    58   335   4.34   4.35   2.75 affordable  
## 6  0.24 Very Good J      VVS2     62.8    57   336   3.94   3.96   2.48 affordable  
## 7  0.24 Very Good I      VVS1     62.3    57   336   3.95   3.98   2.47 affordable  
## 8  0.26 Very Good H      SI1      61.9    55   337   4.07   4.11   2.53 affordable  
## 9  0.22 Fair       E      VS2      65.1    61   337   3.87   3.78   2.49 affordable  
## 10 0.23 Very Good H      VS1      59.4    61   338   4      4.05   2.39 affordable  
## # i 53,930 more rows
```

Group by and Summarise

Use `group_by` and `summarise` to group variables:

```
diamonds %>%  
  group_by(cut) %>%  
  summarise(n = n())
```

```
## # A tibble: 5 x 2  
##   cut      n  
##   <ord>   <int>  
## 1 Fair     1610  
## 2 Good     4906  
## 3 Very Good 12082  
## 4 Premium  13791  
## 5 Ideal   21551
```

More examples

```
diamonds %>%  
  group_by(cut) %>%  
  summarise(n = n(), price_avg = mean(price))
```

```
## # A tibble: 5 x 3  
##   cut          n price_avg  
##   <ord>      <int>      <dbl>  
## 1 Fair       1610      4359.  
## 2 Good       4906      3929.  
## 3 Very Good 12082      3982.  
## 4 Premium   13791      4584.  
## 5 Ideal     21551      3458.
```

Proportions

```
diamonds %>%  
  group_by(cut) %>%  
  summarise(n = n(), price_avg = mean(price)) %>%  
  ungroup() %>%  
  mutate(prop = n/sum(n))
```

```
## # A tibble: 5 x 4  
##   cut      n price_avg prop  
##   <ord> <int>   <dbl> <dbl>  
## 1 Fair    1610    4359. 0.0298  
## 2 Good    4906    3929. 0.0910  
## 3 Very Good 12082   3982. 0.224  
## 4 Premium 13791    4584. 0.256  
## 5 Ideal  21551    3458. 0.400
```

With percentage

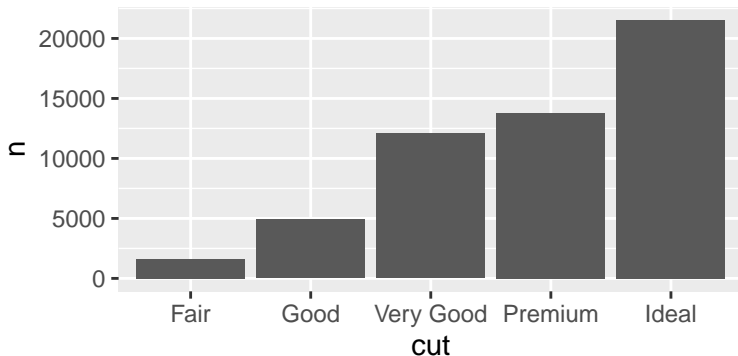
Use `scales::percent()` to add %.

```
diamonds %>%  
  group_by(cut) %>%  
  summarise(n = n(), price_avg = mean(price)) %>%  
  ungroup() %>%  
  mutate(prop = scales::percent(n/sum(n)))
```

```
## # A tibble: 5 x 4  
##   cut      n price_avg prop  
##   <ord> <int>   <dbl> <chr>  
## 1 Fair    1610    4359. 3.0%  
## 2 Good    4906    3929. 9.1%  
## 3 Very Good 12082   3982. 22.4%  
## 4 Premium 13791    4584. 25.6%  
## 5 Ideal   21551    3458. 40.0%
```

Graphing after transformation

```
diamonds %>%  
  group_by(cut) %>%  
  summarise(n = n(), price_avg = mean(price)) %>%  
  ggplot() +  
  geom_bar(aes(x = cut, y = n), stat = "identity")
```



ggplot

Here we used functions from “ggplot2” package. Same pattern as “tidyverse”, but using “+” to connect.

How to write?

- Specify the data using `ggplot(data = diamonds)`
- Specify the x-/y-axis, `ggplot(data = diamonds, mapping = aes(x = cut))`
- Specify the types of plots with `geom`, e.g. `+ geom_bar()`

```
ggplot(data = diamonds) +  
  geom_bar(mapping = aes(x = cut))
```

More plots

- `geom_histogram()`, `geom_density()`, `geom_line()`, `geom_point()`
- `geom_facet()` generates subplots
- color package
 - “RColorBrewer”
 - “ggsci”

Resources

This module is based on

- Brendan R. E. Ansell's "Introduction to R - tidyverse" [\[link\]](#)