

Bill's Drills Book

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2020-04-12

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Chapter 1

Drills: Part of Every Healthy Intellectual Diet

The goal of this book is to organize my R drills into reasonable chunks, the better to understand my strengths and weaknesses, and to plan new forays into data science.

Notes on **bookdown**:

- the `_bookdown.yml` file contains a snippet that is important to inserting the word “Chapter” before the chapter number in each of the Rmd files.
- packages are indicated in bold, like **dplyr**
- inline code and filenames are indicated in typewriter face using backticks, like `_bookdown.yml`
- `_output.yml` is modified from that used by Xie in his `bookdown-demo` (Xie, 2020); it evokes `style.css`, `toc.css`, `preamble.tex`, which are also borrowed from Xie.

Chapter 2

Data Exploration

Data exploration is one of the most important aspects of data science and forms the cornerstone of my drills. Nonetheless, I have lots of room for improvement.

I like Hadely Wickham's writing and find his approach exceptionally clear. Therefore, I'll use the *tidyverse*.

```
library(tidyverse)
```

2.1 Counting things. The naming of parts.

```
starwars %>%  
  filter(!is.na(species)) %>%  
  count(species = fct_lump(species, 5), sort = TRUE) %>%  
  mutate(species = fct_reorder(species, n)) %>%  
  ggplot(aes(species, n)) +  
  geom_col() + coord_flip()
```

I like stacked bars for their economy, but it's easy to over do it. Supperimposing gender onto the columns seems easy...

```
starwars %>%  
  filter(!is.na(species)) %>%  
  count(species = fct_lump(species, 5), gender = fct_lump(gender, 2), sort = TRUE) %>%  
  mutate(species = fct_reorder(species, n)) %>%  
  ggplot(aes(species, n, fill = gender)) +  
  geom_col() + coord_flip()
```

```
## Warning: Factor `gender` contains implicit NA, consider using  
## `forcats::fct_explicit_na`
```

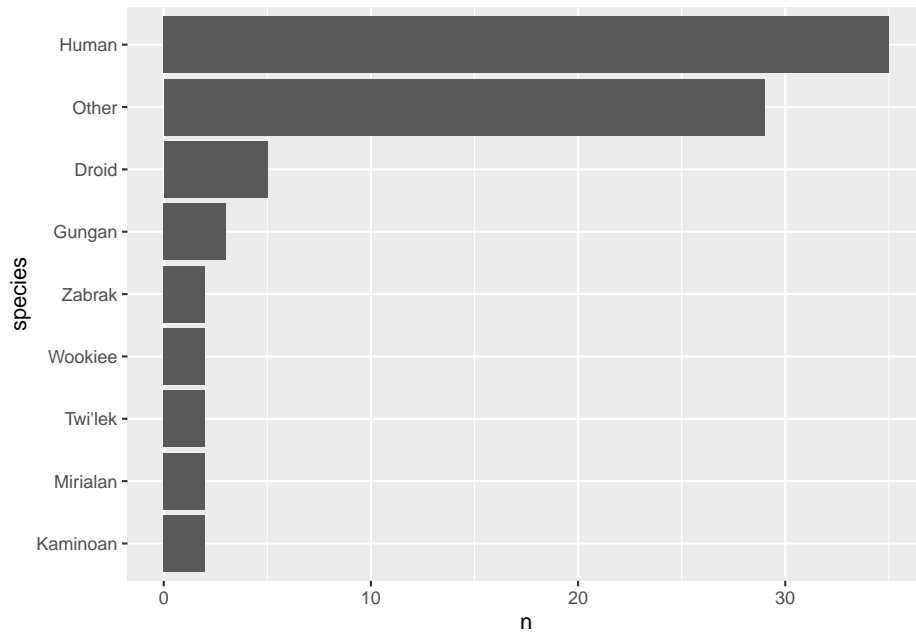
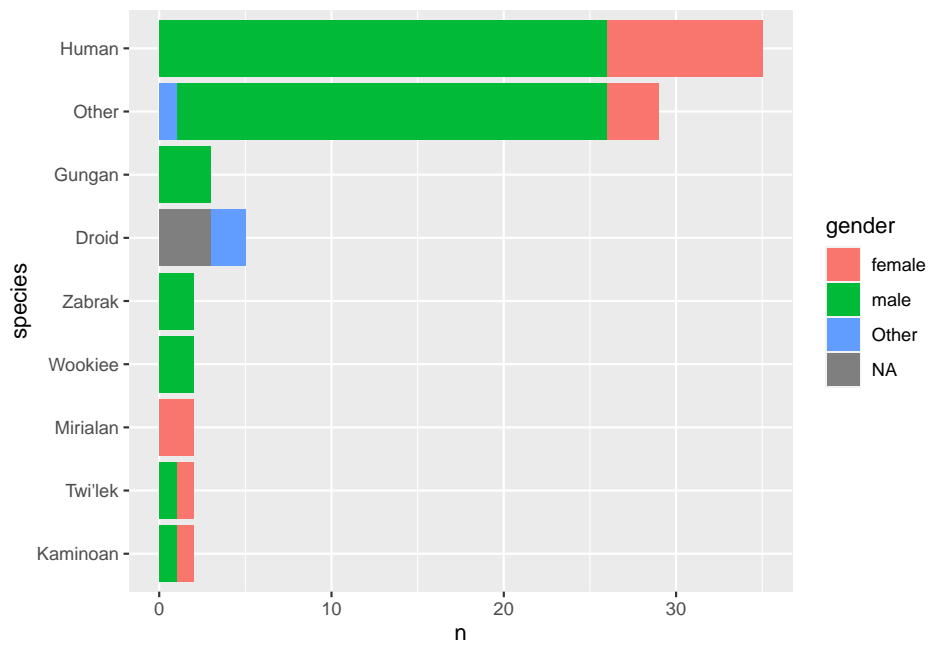


Figure 2.1: Starwars Figure 1



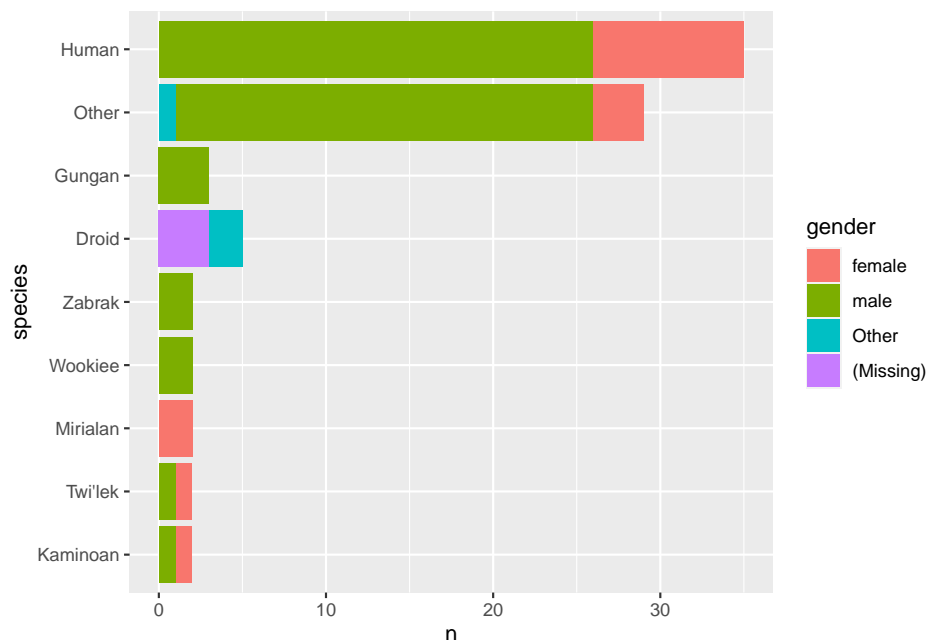
But note that I've got a problem: the Droids, which outnumber the Gungans, are now reordered to *after* the Gungans. This happens because the n that we're

counting comprises subcategories of species *and* gender. Only three Gungan males exist (and no females), but that is enough to tie the Droid NA category. The Droid NA category come after the Gungan category, presumably because *male* comes before *NA*, or because *NA* comes last (more likely).

Exploring this, I see that I'm getting warning messages about the implicit NA's in gender. Note that the following renders a slightly different plot. I *still* have not fixed the order of the species.

```
starwars %>%
  filter(!is.na(species)) %>%
  count(species = fct_lump(species, 5), gender = fct_lump(gender, 2), sort = TRUE) %>%
  mutate(gender = fct_explicit_na(gender),
         species = fct_reorder(species, n)) %>%
  ggplot(aes(species, n, fill = gender)) +
  geom_col() + coord_flip()
```

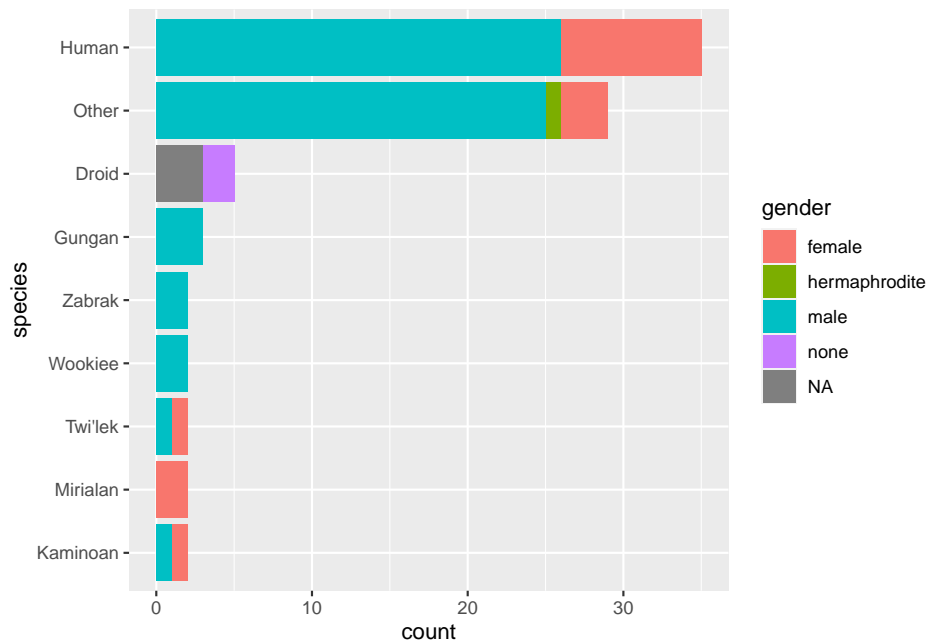
```
## Warning: Factor `gender` contains implicit NA, consider using
## `forcats::fct_explicit_na`
```



The trick here is to use `group_by()` and `ungroup()` wisely.

```
starwars %>% filter(!is.na(species)) %>%
  mutate(species = fct_lump(species, 5)) %>%
  group_by(species) %>%
  mutate(typeCount = n()) %>%
  ungroup() %>%
```

```
mutate(species = fct_reorder(species, typeCount)) %>%
  ggplot()+
  geom_bar(aes(species, fill = gender))+
  coord_flip()
```



As opposed to using `count()`, which progressively narrows the information available to be used, by using `group_by()/mutate()/ungroup()` with `geom_bar()` we have all of the variables still available for plotting.

2.2 Summarize is another very useful function:

```
starwars %>%
  filter(!is.na(species)) %>%
  group_by(species) %>%
  summarize(n=n(), mean = mean(height, na.rm = TRUE)) %>%
  arrange(desc(n))
```

```
## # A tibble: 37 x 3
##   species      n mean
##   <chr>    <int> <dbl>
## 1 Human      35 177.
## 2 Droid       5 140
## 3 Gungan      3 209.
```

```
## 4 Kaminoan      2 221
## 5 Mirialan      2 168
## 6 Twi'lek       2 179
## 7 Wookiee       2 231
## 8 Zabrak        2 173
## 9 Aleena        1  79
## 10 Besalisk     1 198
## # ... with 27 more rows
```

2.3 Referencing other parts of the document

This is a good place to practice referencing figures. Say that I want to refer the reader back to my first starwars figure. See Figure 2.1.

I can reference other pages in a similar fashion. See Chapter 9. Note that this works by referencing a `{#label}` placed in the chapter title.

See Chapter 1

See Chapter 2

Note that the `{#label}` uses a single run-together word. It does not tolerate spaces and this cannot be overcome by ‘quoting’ it.

2.4 Referencing citations:

In order to insert citations, one needs a `.bib` file in the project. I’ve included one in this project as *book.bib*. The yml header in Chapter 1 needs to have a *bibliography* : and *biblio — style* : line added.

To insert a citation, use the **citr** Addin from RStudio. **bookdown**, for instance, is cited thusly (Xie, 2020). Note that I need to figure out an adequate workflow of references. The convenience of Endnote in MS Word will not be available. Nonetheless, if I populate the *book.bib* and *packages.bib* files carefully, with `.txt` files generated in Endnote, I should be OK.

For instance, a recent dump of my Endnote library is in *bookFromEndnote.txt*. This can be opened in RStudio, and I can copy-and-paste references from the `.txt` file to my *book.bib*. For instance, if I have a breast paper that I want to cite here (Stevens and Parekh, 2016), I’d copy-and-paste the reference from *bookFromEndnot.txt* to *book.bib*.

Of note, Yihui Xie includes a nifty bit of code to automatically generate a *bib* database for R packages:

```
knitr::write_bib(c(.packages(), 'bookdown', 'knitr', 'rmarkdown', 'tidyverse', 'Comple
```

References appear automatically at the end of a chapter.

Chapter 3

Sampling

3.1 Think about throwing a bunch of dice.

```
sample(1:6, size=100, replace=TRUE)
```

```
## [1] 1 5 5 2 4 1 3 4 1 2 3 3 6 6 1 1 5 3 3 2 3 4 2 2 3 3 3 4 6 5 4 6 5 1 3 6 5
## [38] 2 2 1 5 5 5 1 1 2 3 2 3 2 3 1 2 3 1 1 5 2 6 4 2 4 4 4 6 6 6 2 2 5 1 2 3 1
## [75] 5 5 3 3 2 4 3 6 1 6 1 6 5 5 6 5 3 4 4 5 3 1 3 4 5 2
```

```
sample(1:6, size=100, replace=TRUE) %>% table()
```

```
## .
## 1 2 3 4 5 6
## 18 20 20 7 14 21
```

```
sample(1:6, size=100, replace=TRUE) %>% table() %>% prop.table()
```

```
## .
## 1 2 3 4 5 6
## 0.18 0.20 0.20 0.16 0.12 0.14
```

3.2 A keen way to divide up a dataset into testing and training components.

```
x <- 1:50
y <- 51:100
```

```
df <- data.frame(x,y)
df
```

```
##      x  y
## 1    1 51
## 2    2 52
## 3    3 53
## 4    4 54
## 5    5 55
## 6    6 56
## 7    7 57
## 8    8 58
## 9    9 59
## 10   10 60
## 11   11 61
## 12   12 62
## 13   13 63
## 14   14 64
## 15   15 65
## 16   16 66
## 17   17 67
## 18   18 68
## 19   19 69
## 20   20 70
## 21   21 71
## 22   22 72
## 23   23 73
## 24   24 74
## 25   25 75
## 26   26 76
## 27   27 77
## 28   28 78
## 29   29 79
## 30   30 80
## 31   31 81
## 32   32 82
## 33   33 83
## 34   34 84
## 35   35 85
## 36   36 86
## 37   37 87
## 38   38 88
## 39   39 89
## 40   40 90
## 41   41 91
```

3.2. A KEEN WAY TO DIVIDE UP A DATASET INTO TESTING AND TRAINING COMPONENTS.15

```
## 42 42 92
## 43 43 93
## 44 44 94
## 45 45 95
## 46 46 96
## 47 47 97
## 48 48 98
## 49 49 99
## 50 50 100
```

```
set.seed(0)
train_indexes = sample(1:nrow(df), .7 * nrow(df))

train_set <- df[train_indexes,]
test_set <- df[-train_indexes,]
```


Chapter 4

Factor Practice

```
cups <- c("small", "medium", "large")
manyCups <- sample(cups, size = 100, replace = TRUE)
sizesCups <- factor(manyCups, levels = c("small", "medium", "large"))
sizesCups
```

```
## [1] medium small large medium small small large medium medium large
## [11] large medium medium medium medium small medium medium medium medium
## [21] small large large medium large large medium large large small
## [31] small small small large medium large small small medium small
## [41] small small small large medium small small large large large
## [51] medium medium medium large medium medium large large large small
## [61] medium medium small large large medium large medium small medium
## [71] small large large small medium small large medium large large
## [81] small small medium medium medium small small small medium small
## [91] large medium large large medium large large small small medium
## Levels: small medium large
```

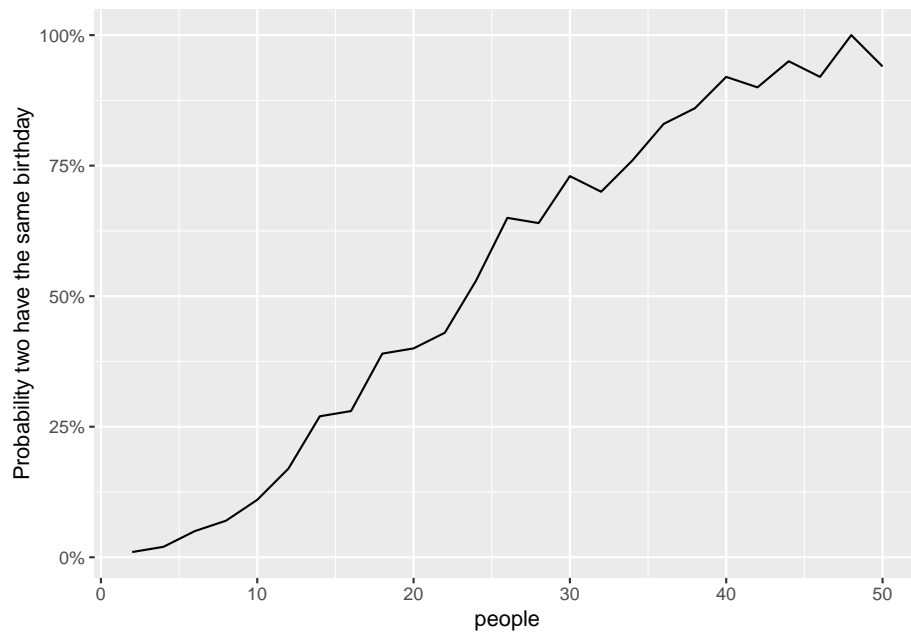

Chapter 5

Crossing Trial

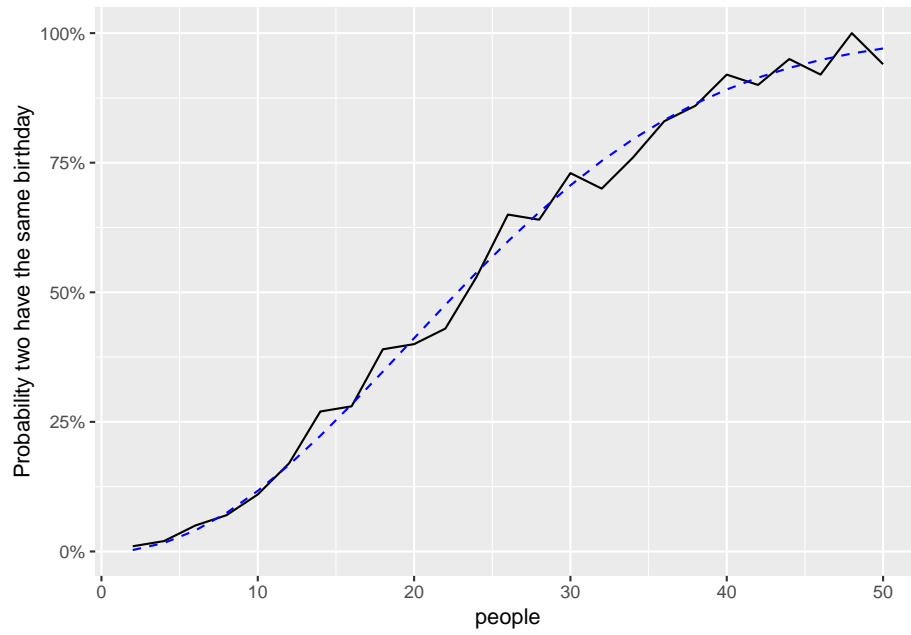
From David Robinson birthday paradox Rblogger at <https://www.r-bloggers.com/the-birthday-paradox-puzzle-tidy-simulation-in-r/>

```
summarized <- crossing(people = seq(2, 50, 2),
                      trial = 1:100) %>%
  mutate(birthday = map(people, ~ sample(365, .x, replace = TRUE)),
         multiple = map_lgl(birthday, ~ any(duplicated(.x)))) %>%
  group_by(people) %>%
  summarize(chance = mean(multiple))

ggplot(summarized, aes(people, chance)) +
  geom_line() +
  scale_y_continuous(labels = scales::percent_format()) +
  labs(y = "Probability two have the same birthday")
```



```
# Checking the work with pbirthday function
summarized %>%
  mutate(exact = map_dbl(people, pbirthday)) %>%
  ggplot(aes(people, chance)) +
  geom_line() +
  geom_line(aes(y = exact), lty = 2, color = "blue") +
  scale_y_continuous(labels = scales::percent_format()) +
  labs(y = "Probability two have the same birthday")
```



Chapter 6

By Any Other Name

This deceptively simple-seeming idea gets complex quickly. The following YouTube was a nice description of the process: <https://www.youtube.com/watch?v=Okc0IL5uTnA>

```
my.data <- data.frame(colOne=1:3, column2=4:6, column_3=7:9)
rownames(my.data) <- c("ant", "bee", "cat")
names(my.data)
```

```
## [1] "colOne" "column2" "column_3"
```

```
colnames(my.data)
```

```
## [1] "colOne" "column2" "column_3"
```

```
#make some changes
```

```
names(my.data) <- c("col_1", "col_2", "col_3")
my.data
```

```
##      col_1 col_2 col_3
## ant      1     4     7
## bee      2     5     8
## cat      3     6     9
```

```
names(my.data)[3] <- "col.3"
my.data
```

```
##      col_1 col_2 col.3
## ant      1     4     7
## bee      2     5     8
## cat      3     6     9
```

```
names(my.data)[names(my.data)=="col_2"]
```

```
## [1] "col_2"
my.data["col_2"]

##      col_2
## ant      4
## bee      5
## cat      6
my.data$col_2

## [1] 4 5 6
my.data[,2]

## [1] 4 5 6
names(my.data)[names(my.data)=="col_2"] <- "col.2"
my.data

##      col_1 col.2 col.3
## ant      1     4     7
## bee      2     5     8
## cat      3     6     9
names(my.data) <- gsub("_", ".", names(my.data))
my.data

##      col.1 col.2 col.3
## ant      1     4     7
## bee      2     5     8
## cat      3     6     9
rownames(my.data)

## [1] "ant" "bee" "cat"
my.data$species <- rownames(my.data)
my.data

##      col.1 col.2 col.3 species
## ant      1     4     7     ant
## bee      2     5     8     bee
## cat      3     6     9     cat
rownames(my.data) <- NULL
my.data

##      col.1 col.2 col.3 species
## 1         1     4     7     ant
## 2         2     5     8     bee
```



```
## 3      3      6      9      cat
colnames(my.data) <- c("good", "better", "best", "species")
my.data
```

```
##   good better best species
## 1     1      4     7      ant
## 2     2      5     8      bee
## 3     3      6     9      cat
```

```
keep <- 2:ncol(my.data)
my.data[,keep]
```

```
##   better best species
## 1      4     7      ant
## 2      5     8      bee
## 3      6     9      cat
```


Chapter 7

Correlation Plots

```
iris
```

##	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
## 1	5.1	3.5	1.4	0.2	setosa
## 2	4.9	3.0	1.4	0.2	setosa
## 3	4.7	3.2	1.3	0.2	setosa
## 4	4.6	3.1	1.5	0.2	setosa
## 5	5.0	3.6	1.4	0.2	setosa
## 6	5.4	3.9	1.7	0.4	setosa
## 7	4.6	3.4	1.4	0.3	setosa
## 8	5.0	3.4	1.5	0.2	setosa
## 9	4.4	2.9	1.4	0.2	setosa
## 10	4.9	3.1	1.5	0.1	setosa
## 11	5.4	3.7	1.5	0.2	setosa
## 12	4.8	3.4	1.6	0.2	setosa
## 13	4.8	3.0	1.4	0.1	setosa
## 14	4.3	3.0	1.1	0.1	setosa
## 15	5.8	4.0	1.2	0.2	setosa
## 16	5.7	4.4	1.5	0.4	setosa
## 17	5.4	3.9	1.3	0.4	setosa
## 18	5.1	3.5	1.4	0.3	setosa
## 19	5.7	3.8	1.7	0.3	setosa
## 20	5.1	3.8	1.5	0.3	setosa
## 21	5.4	3.4	1.7	0.2	setosa
## 22	5.1	3.7	1.5	0.4	setosa
## 23	4.6	3.6	1.0	0.2	setosa
## 24	5.1	3.3	1.7	0.5	setosa
## 25	4.8	3.4	1.9	0.2	setosa
## 26	5.0	3.0	1.6	0.2	setosa

## 27	5.0	3.4	1.6	0.4	setosa
## 28	5.2	3.5	1.5	0.2	setosa
## 29	5.2	3.4	1.4	0.2	setosa
## 30	4.7	3.2	1.6	0.2	setosa
## 31	4.8	3.1	1.6	0.2	setosa
## 32	5.4	3.4	1.5	0.4	setosa
## 33	5.2	4.1	1.5	0.1	setosa
## 34	5.5	4.2	1.4	0.2	setosa
## 35	4.9	3.1	1.5	0.2	setosa
## 36	5.0	3.2	1.2	0.2	setosa
## 37	5.5	3.5	1.3	0.2	setosa
## 38	4.9	3.6	1.4	0.1	setosa
## 39	4.4	3.0	1.3	0.2	setosa
## 40	5.1	3.4	1.5	0.2	setosa
## 41	5.0	3.5	1.3	0.3	setosa
## 42	4.5	2.3	1.3	0.3	setosa
## 43	4.4	3.2	1.3	0.2	setosa
## 44	5.0	3.5	1.6	0.6	setosa
## 45	5.1	3.8	1.9	0.4	setosa
## 46	4.8	3.0	1.4	0.3	setosa
## 47	5.1	3.8	1.6	0.2	setosa
## 48	4.6	3.2	1.4	0.2	setosa
## 49	5.3	3.7	1.5	0.2	setosa
## 50	5.0	3.3	1.4	0.2	setosa
## 51	7.0	3.2	4.7	1.4	versicolor
## 52	6.4	3.2	4.5	1.5	versicolor
## 53	6.9	3.1	4.9	1.5	versicolor
## 54	5.5	2.3	4.0	1.3	versicolor
## 55	6.5	2.8	4.6	1.5	versicolor
## 56	5.7	2.8	4.5	1.3	versicolor
## 57	6.3	3.3	4.7	1.6	versicolor
## 58	4.9	2.4	3.3	1.0	versicolor
## 59	6.6	2.9	4.6	1.3	versicolor
## 60	5.2	2.7	3.9	1.4	versicolor
## 61	5.0	2.0	3.5	1.0	versicolor
## 62	5.9	3.0	4.2	1.5	versicolor
## 63	6.0	2.2	4.0	1.0	versicolor
## 64	6.1	2.9	4.7	1.4	versicolor
## 65	5.6	2.9	3.6	1.3	versicolor
## 66	6.7	3.1	4.4	1.4	versicolor
## 67	5.6	3.0	4.5	1.5	versicolor
## 68	5.8	2.7	4.1	1.0	versicolor
## 69	6.2	2.2	4.5	1.5	versicolor
## 70	5.6	2.5	3.9	1.1	versicolor
## 71	5.9	3.2	4.8	1.8	versicolor
## 72	6.1	2.8	4.0	1.3	versicolor

## 73	6.3	2.5	4.9	1.5 versicolor
## 74	6.1	2.8	4.7	1.2 versicolor
## 75	6.4	2.9	4.3	1.3 versicolor
## 76	6.6	3.0	4.4	1.4 versicolor
## 77	6.8	2.8	4.8	1.4 versicolor
## 78	6.7	3.0	5.0	1.7 versicolor
## 79	6.0	2.9	4.5	1.5 versicolor
## 80	5.7	2.6	3.5	1.0 versicolor
## 81	5.5	2.4	3.8	1.1 versicolor
## 82	5.5	2.4	3.7	1.0 versicolor
## 83	5.8	2.7	3.9	1.2 versicolor
## 84	6.0	2.7	5.1	1.6 versicolor
## 85	5.4	3.0	4.5	1.5 versicolor
## 86	6.0	3.4	4.5	1.6 versicolor
## 87	6.7	3.1	4.7	1.5 versicolor
## 88	6.3	2.3	4.4	1.3 versicolor
## 89	5.6	3.0	4.1	1.3 versicolor
## 90	5.5	2.5	4.0	1.3 versicolor
## 91	5.5	2.6	4.4	1.2 versicolor
## 92	6.1	3.0	4.6	1.4 versicolor
## 93	5.8	2.6	4.0	1.2 versicolor
## 94	5.0	2.3	3.3	1.0 versicolor
## 95	5.6	2.7	4.2	1.3 versicolor
## 96	5.7	3.0	4.2	1.2 versicolor
## 97	5.7	2.9	4.2	1.3 versicolor
## 98	6.2	2.9	4.3	1.3 versicolor
## 99	5.1	2.5	3.0	1.1 versicolor
## 100	5.7	2.8	4.1	1.3 versicolor
## 101	6.3	3.3	6.0	2.5 virginica
## 102	5.8	2.7	5.1	1.9 virginica
## 103	7.1	3.0	5.9	2.1 virginica
## 104	6.3	2.9	5.6	1.8 virginica
## 105	6.5	3.0	5.8	2.2 virginica
## 106	7.6	3.0	6.6	2.1 virginica
## 107	4.9	2.5	4.5	1.7 virginica
## 108	7.3	2.9	6.3	1.8 virginica
## 109	6.7	2.5	5.8	1.8 virginica
## 110	7.2	3.6	6.1	2.5 virginica
## 111	6.5	3.2	5.1	2.0 virginica
## 112	6.4	2.7	5.3	1.9 virginica
## 113	6.8	3.0	5.5	2.1 virginica
## 114	5.7	2.5	5.0	2.0 virginica
## 115	5.8	2.8	5.1	2.4 virginica
## 116	6.4	3.2	5.3	2.3 virginica
## 117	6.5	3.0	5.5	1.8 virginica
## 118	7.7	3.8	6.7	2.2 virginica

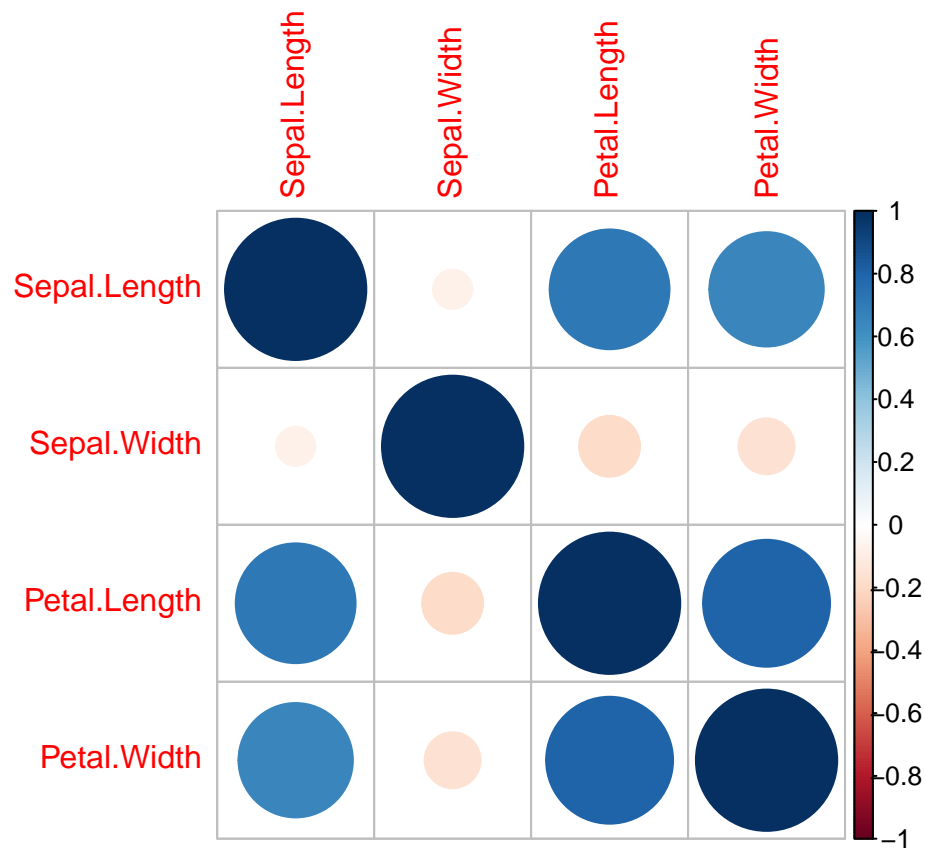
```
## 119      7.7      2.6      6.9      2.3 virginica
## 120      6.0      2.2      5.0      1.5 virginica
## 121      6.9      3.2      5.7      2.3 virginica
## 122      5.6      2.8      4.9      2.0 virginica
## 123      7.7      2.8      6.7      2.0 virginica
## 124      6.3      2.7      4.9      1.8 virginica
## 125      6.7      3.3      5.7      2.1 virginica
## 126      7.2      3.2      6.0      1.8 virginica
## 127      6.2      2.8      4.8      1.8 virginica
## 128      6.1      3.0      4.9      1.8 virginica
## 129      6.4      2.8      5.6      2.1 virginica
## 130      7.2      3.0      5.8      1.6 virginica
## 131      7.4      2.8      6.1      1.9 virginica
## 132      7.9      3.8      6.4      2.0 virginica
## 133      6.4      2.8      5.6      2.2 virginica
## 134      6.3      2.8      5.1      1.5 virginica
## 135      6.1      2.6      5.6      1.4 virginica
## 136      7.7      3.0      6.1      2.3 virginica
## 137      6.3      3.4      5.6      2.4 virginica
## 138      6.4      3.1      5.5      1.8 virginica
## 139      6.0      3.0      4.8      1.8 virginica
## 140      6.9      3.1      5.4      2.1 virginica
## 141      6.7      3.1      5.6      2.4 virginica
## 142      6.9      3.1      5.1      2.3 virginica
## 143      5.8      2.7      5.1      1.9 virginica
## 144      6.8      3.2      5.9      2.3 virginica
## 145      6.7      3.3      5.7      2.5 virginica
## 146      6.7      3.0      5.2      2.3 virginica
## 147      6.3      2.5      5.0      1.9 virginica
## 148      6.5      3.0      5.2      2.0 virginica
## 149      6.2      3.4      5.4      2.3 virginica
## 150      5.9      3.0      5.1      1.8 virginica
```

```
iris %>% select(-Species) %>% cor()
```

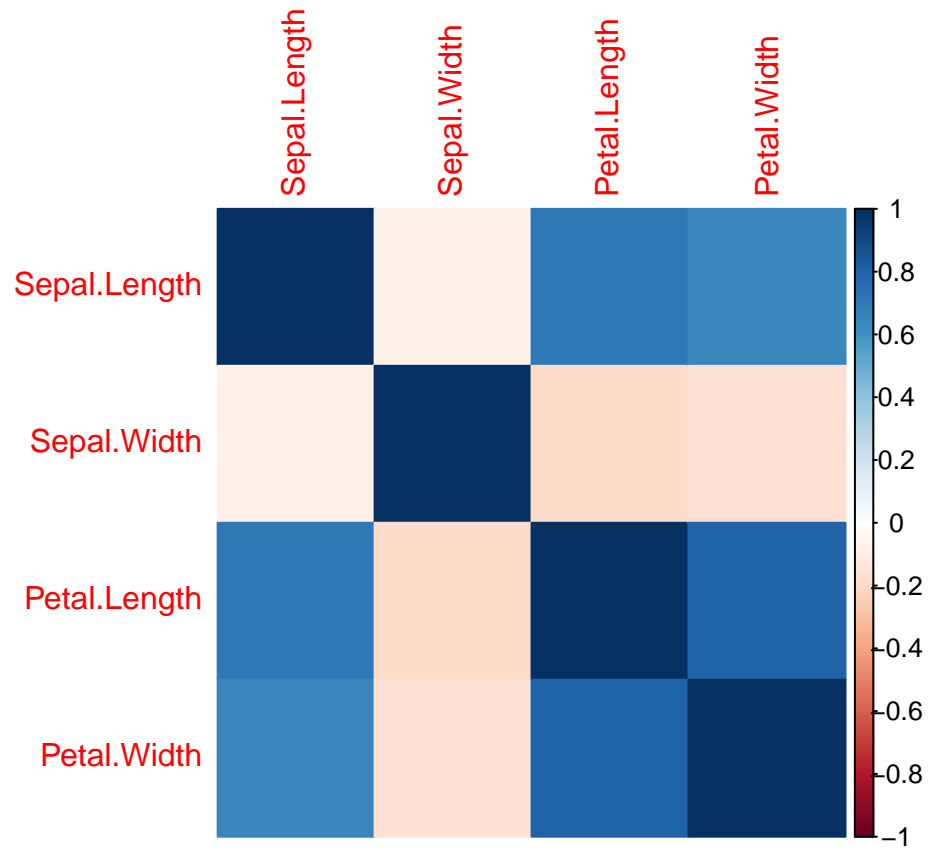
```
##           Sepal.Length Sepal.Width Petal.Length Petal.Width
## Sepal.Length      1.0000000 -0.1175698   0.8717538   0.8179411
## Sepal.Width      -0.1175698   1.0000000  -0.4284401  -0.3661259
## Petal.Length      0.8717538  -0.4284401   1.0000000   0.9628654
## Petal.Width      0.8179411  -0.3661259   0.9628654   1.0000000
```

```
M <- iris %>% select(-Species) %>% cor(method = "kendall")
```

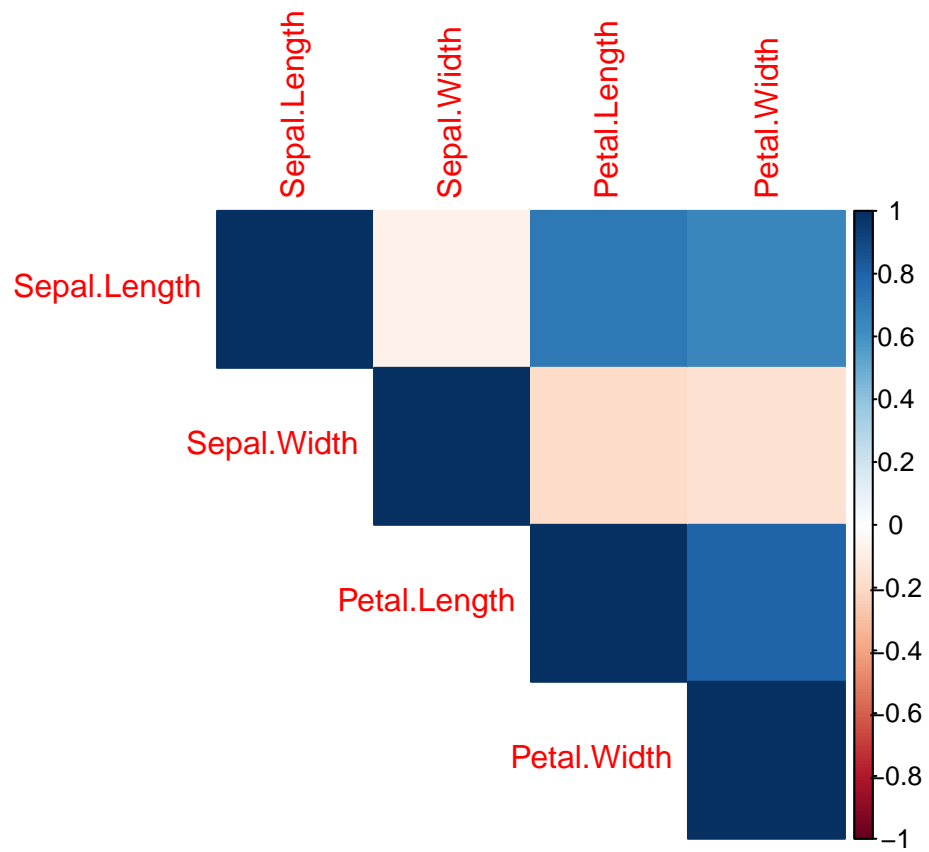
```
corrplot::corrplot(M)
```



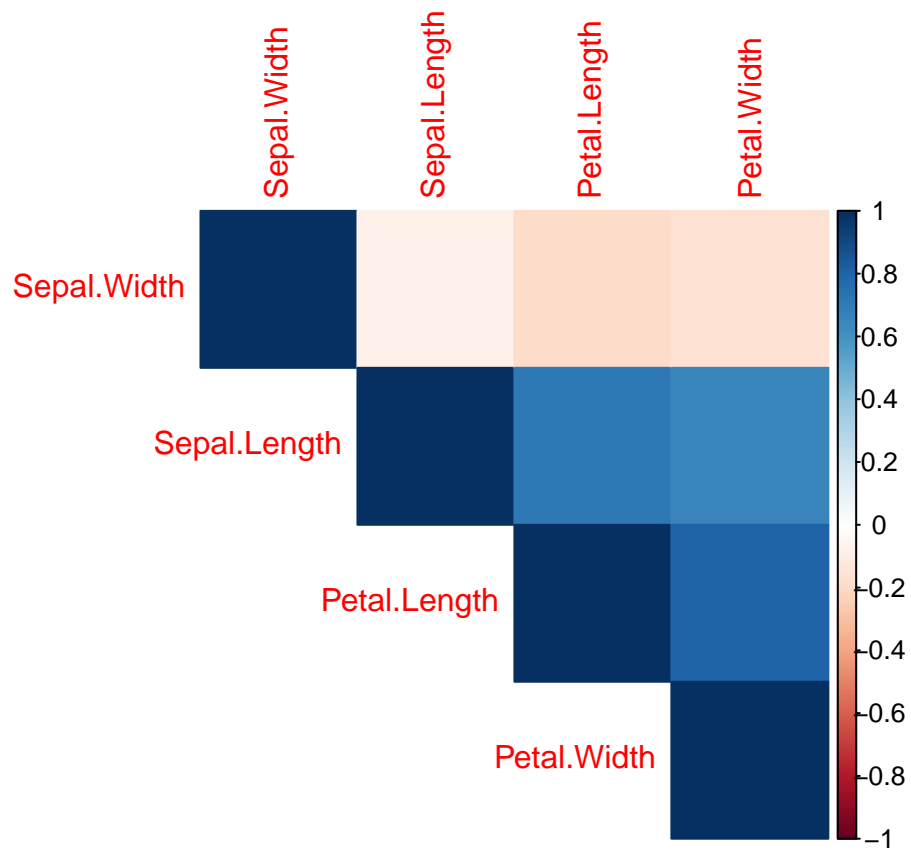
```
corrplot::corrplot(M, method = "color")
```



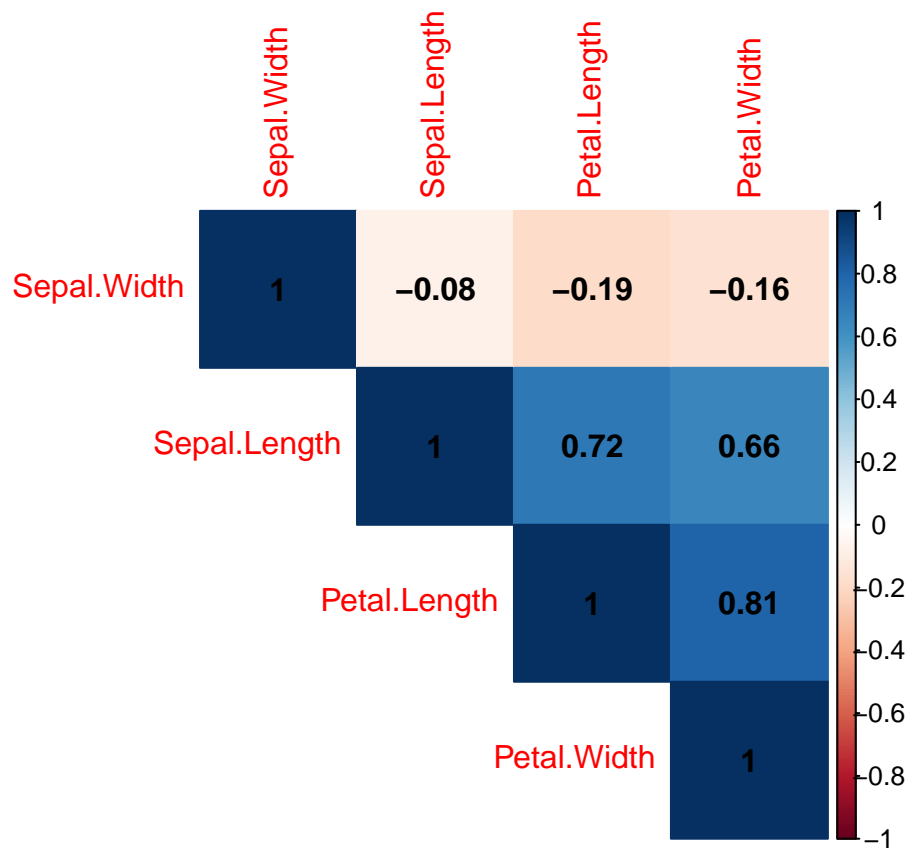
```
corrplot::corrplot(M, method = "color", type = "upper")
```

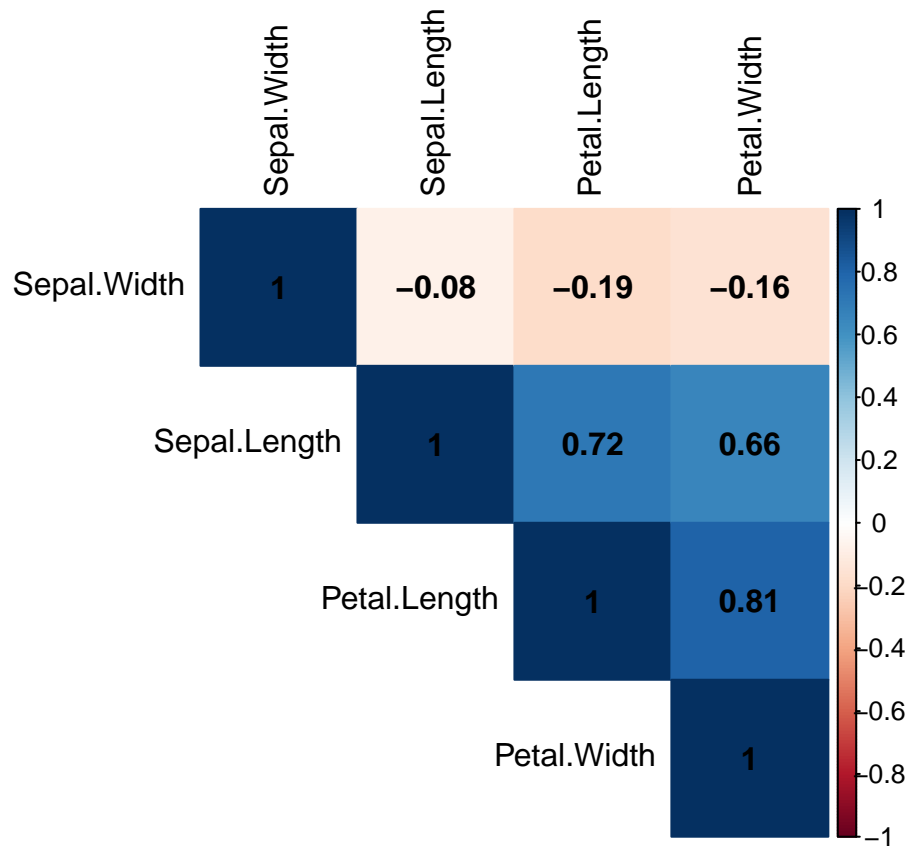
```
corrplot::corrplot(M, method = "color", type = "upper", order = "hclust")
```



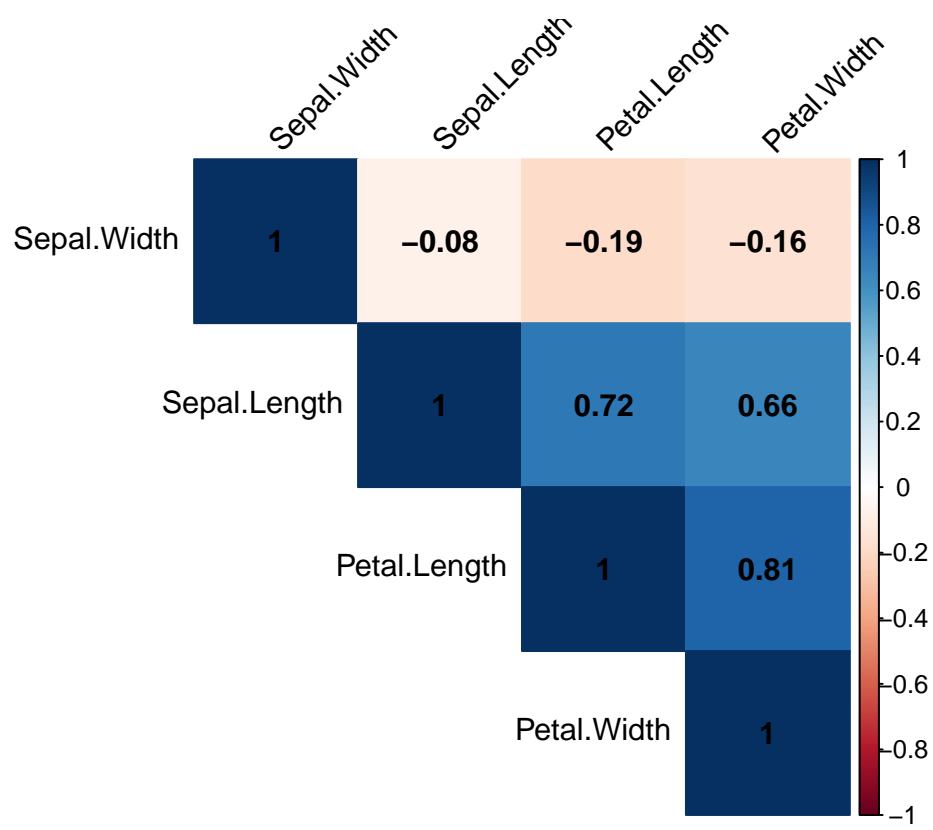
```
corrplot::corrplot(M, method = "color", type = "upper", order = "hclust", addCoef.col = TRUE)
```



```
corrplot::corrplot(M, method = "color", type = "upper", order = "hclust", addCoef.col = "black",
```



```
corrplot::corrplot(M, method = "color", type = "upper", order = "hclust", addCoef.col = TRUE)
```



Chapter 8

if__else() and case__when(): Comparison

8.1 case__when()

case_when() from https://www.rdocumentation.org/packages/dplyr/versions/0.7.8/topics/case_when

```
x <- 1:50  
y <- 51:100
```

```
df <- data.frame(x,y)  
df
```

```
##      x  y  
## 1    1 51  
## 2    2 52  
## 3    3 53  
## 4    4 54  
## 5    5 55  
## 6    6 56  
## 7    7 57  
## 8    8 58  
## 9    9 59  
## 10  10 60  
## 11  11 61  
## 12  12 62  
## 13  13 63  
## 14  14 64  
## 15  15 65
```

```
## 16 16 66
## 17 17 67
## 18 18 68
## 19 19 69
## 20 20 70
## 21 21 71
## 22 22 72
## 23 23 73
## 24 24 74
## 25 25 75
## 26 26 76
## 27 27 77
## 28 28 78
## 29 29 79
## 30 30 80
## 31 31 81
## 32 32 82
## 33 33 83
## 34 34 84
## 35 35 85
## 36 36 86
## 37 37 87
## 38 38 88
## 39 39 89
## 40 40 90
## 41 41 91
## 42 42 92
## 43 43 93
## 44 44 94
## 45 45 95
## 46 46 96
## 47 47 97
## 48 48 98
## 49 49 99
## 50 50 100
```

```
case_when(
  x %% 35 == 0 ~ "fizz buzz",
  x %% 5 == 0 ~ "fizz",
  x %% 7 == 0 ~ "buzz",
  TRUE ~ as.character(x)
)
```

```
## [1] "1"      "2"      "3"      "4"      "fizz"   "6"
## [7] "buzz"   "8"      "9"      "fizz"   "11"     "12"
## [13] "13"     "buzz"   "fizz"   "16"     "17"     "18"
```



```
## [19] "19"      "fizz"      "buzz"      "22"      "23"      "24"
## [25] "fizz"     "26"      "27"      "buzz"     "29"      "fizz"
## [31] "31"      "32"      "33"      "34"      "fizz buzz" "36"
## [37] "37"      "38"      "39"      "fizz"     "41"      "buzz"
## [43] "43"      "44"      "fizz"     "46"      "47"      "48"
## [49] "buzz"     "fizz"
```

8.2 Compare this with if_else()

```
if_else(x %% 2 == 0, "even", "odd")
```

```
## [1] "odd" "even" "odd" "even" "odd" "even" "odd" "even" "odd" "even"
## [11] "odd" "even" "odd" "even" "odd" "even" "odd" "even" "odd" "even"
## [21] "odd" "even" "odd" "even" "odd" "even" "odd" "even" "odd" "even"
## [31] "odd" "even" "odd" "even" "odd" "even" "odd" "even" "odd" "even"
## [41] "odd" "even" "odd" "even" "odd" "even" "odd" "even" "odd" "even"
```


Chapter 9

Subsetting

From <https://www.r-bloggers.com/5-ways-to-subset-a-data-frame-in-r/>

Note: since this is down for maintenance, I will turn off evaluation on these chunks:

```
education <- read.csv("https://vincentarelbundock.github.io/Rdatasets/csv/robustbase/education.csv")
colnames(education) <- c("X", "State", "Region", "Urban.Population", "Per.Capita.Income", "Minor.Population", "Education.Expenditures")
glimpse(education)
```

```
## Rows: 50
## Columns: 7
## $ X               <int> 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 1...
## $ State           <chr> "ME", "NH", "VT", "MA", "RI", "CT", "NY", "N...
## $ Region          <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, ...
## $ Urban.Population <int> 508, 564, 322, 846, 871, 774, 856, 889, 715, ...
## $ Per.Capita.Income <int> 3944, 4578, 4011, 5233, 4780, 5889, 5663, 57...
## $ Minor.Population <int> 325, 323, 328, 305, 303, 307, 301, 310, 300, ...
## $ Education.Expenditures <int> 235, 231, 270, 261, 300, 317, 387, 285, 300, ...
```

9.1 Subsetting using brackets

```
education[c(10:21), c(2, 6:7)]
```

```
##      State Minor.Population Education.Expenditures
## 10    OH                324                    221
## 11    IN                329                    264
## 12    IL                320                    308
```

```
## 13    MI                337                379
## 14    WI                328                342
## 15    MN                330                378
## 16    IA                318                232
## 17    MO                309                231
## 18    ND                333                246
## 19    SD                330                230
## 20    NB                318                268
## 21    KS                304                337
```

9.2 Subset using brackets by omitting the rows and columns we don't want

```
education[-c(1:9,22:50),-c(1,3:5)]
```

```
##      State Minor.Population Education.Expenditures
## 10     OH                324                221
## 11     IN                329                264
## 12     IL                320                308
## 13     MI                337                379
## 14     WI                328                342
## 15     MN                330                378
## 16     IA                318                232
## 17     MO                309                231
## 18     ND                333                246
## 19     SD                330                230
## 20     NB                318                268
## 21     KS                304                337
```

9.3 Subset using brackets in combination with the which() function and the %in% operator

```
education[which(education$Region == 2),names(education) %in% c("State", "Minor.Population")]
```

```
##      State Minor.Population Education.Expenditures
## 10     OH                324                221
## 11     IN                329                264
## 12     IL                320                308
## 13     MI                337                379
## 14     WI                328                342
## 15     MN                330                378
```

```
## 16    IA                318                232
## 17    MO                309                231
## 18    ND                333                246
## 19    SD                330                230
## 20    NB                318                268
## 21    KS                304                337
```

9.4 Subset using the subset() function

```
subset(education, Region == 2, select = c("State", "Minor.Population", "Education.Expenditures"))
```

```
##      State Minor.Population Education.Expenditures
## 10    OH                324                221
## 11    IN                329                264
## 12    IL                320                308
## 13    MI                337                379
## 14    WI                328                342
## 15    MN                330                378
## 16    IA                318                232
## 17    MO                309                231
## 18    ND                333                246
## 19    SD                330                230
## 20    NB                318                268
## 21    KS                304                337
```

9.5 Subset using dplyr's filter() and select()

```
select(filter(education, Region == 2), c(State, Minor.Population:Education.Expenditures))
```

```
##      State Minor.Population Education.Expenditures
## 1     OH                324                221
## 2     IN                329                264
## 3     IL                320                308
## 4     MI                337                379
## 5     WI                328                342
## 6     MN                330                378
## 7     IA                318                232
## 8     MO                309                231
## 9     ND                333                246
## 10    SD                330                230
## 11    NB                318                268
## 12    KS                304                337
```


Bibliography

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