

Introduction to the Tidyverse

How to be a tidy data scientist

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Tidyverse

- *Tidy* pour "bien rangé" et *verse* pour "univers"
- A collection of R 📦 developed by H. Wickham and others at Rstudio



Tidyverse

- "Un modèle d'organisation des données qui vise à faciliter le travail souvent long et fastidieux de nettoyage et de préparation préalable à la mise en oeuvre de méthodes d'analyse" (Julien Barnier).
- Les principes d'un tidy data sont :
 - chaque variable est une colonne
 - chaque observation est une ligne
 - chaque type d'observation est dans une table différente

country	year	cases	population
Afghanistan	1999	745	15000000
Afghanistan	2000	666	2005360
Brazil	1999	3737	17206362
Brazil	2000	8488	17404898
China	1999	21258	127215272
China	2000	21766	128048583

variables

country	year	cases	population
Afghanistan	1999	745	15000000
Afghanistan	2000	666	2005360
Brazil	1999	3737	17206362
Brazil	2000	8488	17404898
China	1999	21258	127215272
China	2000	21766	128048583

observations

country	year	cases	population
Afghanistan	1999	745	15000000
Afghanistan	2000	666	2005360
Brazil	1999	3737	17206362
Brazil	2000	8488	17404898
China	1999	21258	127215272
China	2000	21766	128048583

values

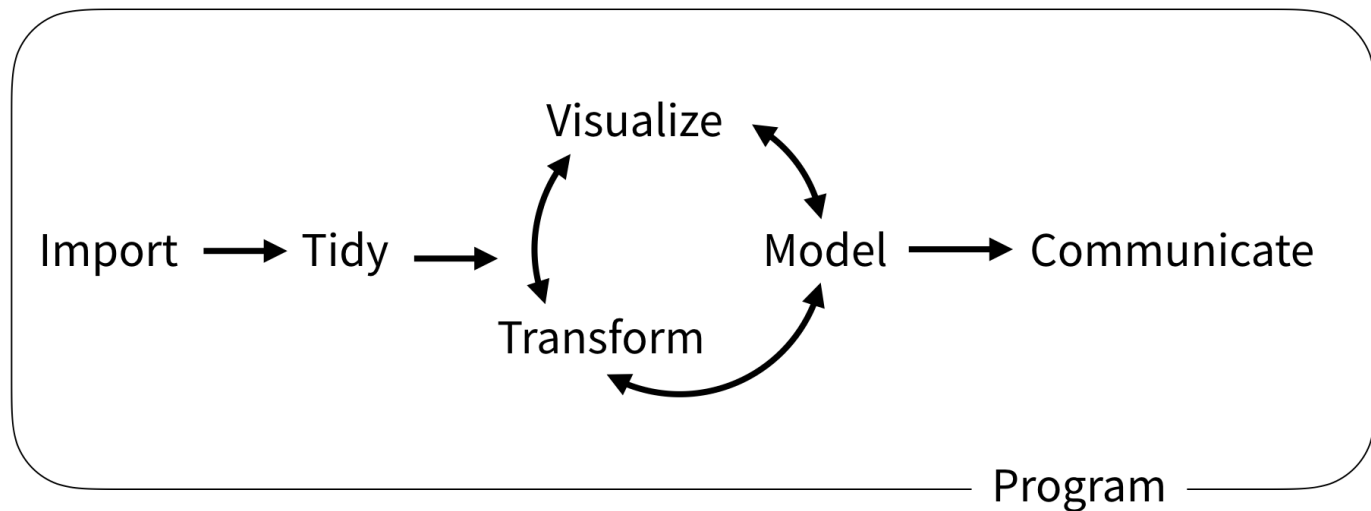
Tidyverse is a collection of R

- ggplot2 - visualisation
- dplyr - manipulation et synthèse des données
- tidyr - manipulation des données
- purrr - programmation avancée
- readr - importation de données
- tibble - tableaux de données data.frame améliorés
- forcats - variables qualitatives
- stringr - chaînes de caractères

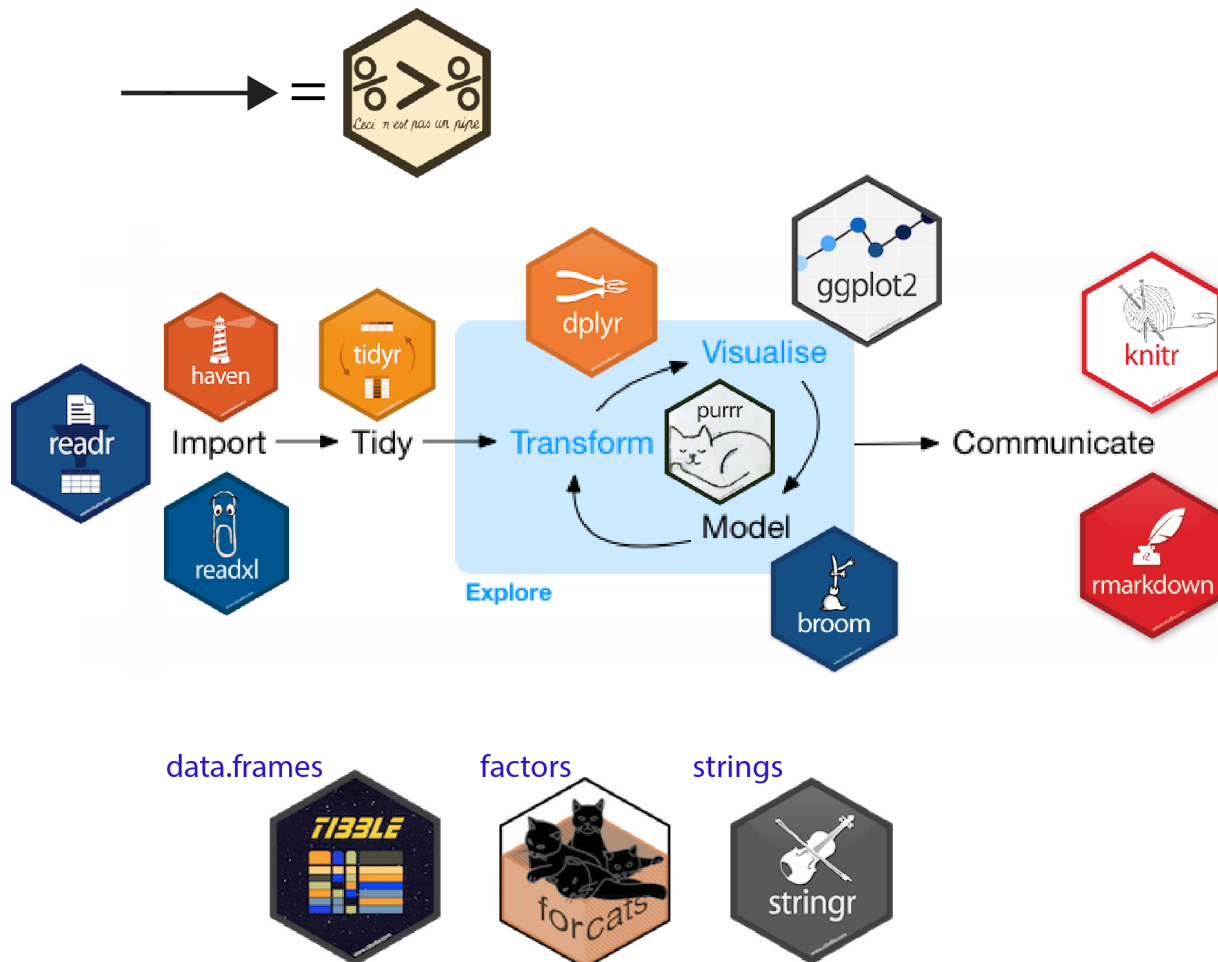
Tidyverse is a collection of R

- `ggplot2` - visualisation
- `dplyr` - manipulation et synthèse des données
- `tidyr` - manipulation des données
- `purrr` - programmation avancée
- `readr` - importation de données
- `tibble` - tableaux de données `data.frame` améliorés
- `forcats` - variables qualitatives
- `stringr` - chaînes de caractères

Workflow in data science



Workflow in data science, with Tidyverse



Load tidyverse



```
#install.packages("tidyverse")  
library(tidyverse)
```

— Attaching packages

```
## ✓ ggplot2 3.1.0.9000    ✓ purrr 0.2.5  
## ✓ tibble 2.0.1          ✓ dplyr 0.7.8  
## ✓ tidyr 0.8.2           ✓ stringr 1.3.1  
## ✓ readr 1.3.1           ✓ forcats 0.3.0
```

```
## Warning: package 'tibble' was built under R version 3.5.2
```

— Conflicts

```
## ✗ dplyr::filter() masks stats::filter()  
## ✗ dplyr::lag()     masks stats::lag()
```


Import, Tidy, Transform

Case study with Shakespeare's word usage

PeerJ Preprints

NOT PEER-REVIEWED

Declutter your R workflow with tidy tools

Zev Ross *
ZevRoss Spatial Analysis
and
Hadley Wickham
RStudio
and
David Robinson
Stack Overflow

Abstract

The *R* language has withstood the test of time. Forty years after it was initially developed (in the form of the S language) *R* is being used by millions of programmers on workflows the inventors of the language could never have imagined. Although base R packages perform well in most settings, workflows can be made more efficient by developing packages with more consistent arguments, inputs and outputs and emphasizing constantly improving code over historical code consistency. The universe of *R* packages known as the tidyverse, including `dplyr`, `tidyr` and others, aim to improve workflows and make data analysis as smooth as possible by applying a set of core programming principles in package development.

Keywords: tidy tools, tidyverse, dplyr, tidyr, tidytext, ggplot2, readr, workflow, pipe, piping, R, base R

Import data

readr::read_csv function:

- keeps input types as is (no conversion to factor)
- creates tibbles instead of `data.frame`
 - pas de noms de lignes (rownames)
 - autorisent noms de colonnes avec caractères spéciaux ou nombres
 - s'affichent plus intelligemment que les `data.frames`
 - pas de partial matching sur noms de colonnes
 - avertissement si on essaie d'accéder à une colonne inexistante
- is damn fast 🚀

Import data

readr::read_csv function:

```
shakespeare <- read_csv('https://gist.githubusercontent.com/zross/ab72ba3250a3ee  
shakespeare
```

```
## # A tibble: 164,656 x 5  
##       X1 word      word_count corpus      corpus_date  
##   <dbl> <chr>      <dbl> <chr>      <dbl>  
## 1     1 1 hive          1 loverscomplaint 1609  
## 2     2 2 plaintful      1 loverscomplaint 1609  
## 3     3 3 Are           1 loverscomplaint 1609  
## 4     4 4 Than           1 loverscomplaint 1609  
## 5     5 5 attended       1 loverscomplaint 1609  
## 6     6 6 That           7 loverscomplaint 1609  
## 7     7 7 moisture        1 loverscomplaint 1609  
## 8     8 8 praised          1 loverscomplaint 1609  
## 9     9 9 particular      1 loverscomplaint 1609  
## 10    10 10 tend           1 loverscomplaint 1609  
## # ... with 164,646 more rows
```

Group by variable to perform operation

dplyr::group_by function:

```
shakespeare_corpus <- group_by(shakespeare, corpus)
shakespeare_corpus
```

```
## # A tibble: 164,656 x 5
## # Groups:   corpus [42]
##       X1 word      word_count corpus      corpus_date
##   <dbl> <chr>      <dbl> <chr>      <dbl>
## 1     1 1 hive          1 loverscomplaint 1609
## 2     2 2 plaintful      1 loverscomplaint 1609
## 3     3 3 Are          1 loverscomplaint 1609
## 4     4 4 Than          1 loverscomplaint 1609
## 5     5 5 attended      1 loverscomplaint 1609
## 6     6 6 That          7 loverscomplaint 1609
## 7     7 7 moisture       1 loverscomplaint 1609
## 8     8 8 praised         1 loverscomplaint 1609
## 9     9 9 particular     1 loverscomplaint 1609
## 10    10 10 tend          1 loverscomplaint 1609
## # ... with 164,646 more rows
```

Summarise data by group

dplyr::summarise function:

```
mean_shakespeare_corpus <- summarise(shakespeare_corpus, avg=mean(word_count))
mean_shakespeare_corpus
```

```
## # A tibble: 42 x 2
##   corpus          avg
##   <chr>         <dbl>
## 1 1kinghenryiv    6.12
## 2 1kinghenryvi    5.24
## 3 2kinghenryiv    6.13
## 4 2kinghenryvi    5.80
## 5 3kinghenryvi    6.45
## 6 allswellthatendswell 6.23
## 7 antonyandcleopatra 5.96
## 8 asyoulikeit     6.28
## 9 comedyoferrors  5.70
## 10 coriolanus      6.35
## # ... with 32 more rows
```

Cleaner code with "pipe" operator %>%

```
shakespeare %>%  
  group_by(corpus) %>% # group by corpus  
  summarise(avg = mean(word_count)) # compute mean
```

```
## # A tibble: 42 x 2  
##   corpus      avg  
##   <chr>    <dbl>  
## 1 1kinghenryiv 6.12  
## 2 1kinghenryvi 5.24  
## 3 2kinghenryiv 6.13  
## 4 2kinghenryvi 5.80  
## 5 3kinghenryvi 6.45  
## 6 allswellthatendswell 6.23  
## 7 antonyandcleopatra 5.96  
## 8 asyoulikeit 6.28  
## 9 comedyoferrors 5.70  
## 10 coriolanus 6.35  
## # ... with 32 more rows
```

Compare to base R

```
with(shakespeare, tapply(word_count, corpus, mean))
```

##	1kinghenryiv	1kinghenryvi	2kinghenryiv
##	6.115590	5.240261	6.132682
##	2kinghenryvi	3kinghenryvi	allswellthatendswell
##	5.800769	6.449460	6.234996
##	antonyandcleopatra	asyoulikeit	comedyoferrors
##	5.960279	6.280468	5.700697
##	coriolanus	cymbeline	hamlet
##	6.347518	5.996103	6.101166
##	juliuscaesar	kinghenryv	kinghenryviii
##	6.263612	5.465125	6.200425
##	kingjohn	kinglear	kingrichardii
##	5.462972	5.853052	5.750000
##	kingrichardiii	loverscomplaint	loveslabourslost
##	6.761723	2.164017	5.474268
##	macbeth	measureforemeasure	merchantofvenice
##	4.810592	6.155045	6.104977
##	merrywivesofwindsor	midsummersnightsdream	muchadoaboutnothing
##	6.578976	5.093365	6.755714
##	othello	periclesprinceof tyre	rapeoflucrece
##	6.553688	5.286628	3.843687
##	romeoandjuliet	sonnets	tamingoftheshrew
##	6.191962	4.842263	6.044336
##	tempest	timonofathens	titusandronicus
##	4.838559	5.228195	5.663220
##	troilusandcressida	twelfthnight	twogentlemenofverona
##	5.805422	6.121392	5.923447

Deselect columns

dplyr::select function:

```
shakespeare %>%  
  select(-X1, -corpus_date) # deselect columns
```

```
## # A tibble: 164,656 x 3  
##   word      word_count corpus  
##   <chr>          <dbl> <chr>  
## 1 hive              1 loverscomplaint  
## 2 plaintful          1 loverscomplaint  
## 3 Are                1 loverscomplaint  
## 4 Than              1 loverscomplaint  
## 5 attended           1 loverscomplaint  
## 6 That              7 loverscomplaint  
## 7 moisture           1 loverscomplaint  
## 8 praised             1 loverscomplaint  
## 9 particular         1 loverscomplaint  
## 10 tend              1 loverscomplaint  
## # ... with 164,646 more rows
```

Select columns

dplyr::select function:

```
shakespeare %>%  
  select(word, word_count, corpus) # select columns
```

```
## # A tibble: 164,656 x 3  
##   word      word_count corpus  
##   <chr>      <dbl> <chr>  
## 1 hive              1 loverscomplaint  
## 2 plaintful          1 loverscomplaint  
## 3 Are                1 loverscomplaint  
## 4 Than              1 loverscomplaint  
## 5 attended          1 loverscomplaint  
## 6 That              7 loverscomplaint  
## 7 moisture          1 loverscomplaint  
## 8 praised            1 loverscomplaint  
## 9 particular        1 loverscomplaint  
## 10 tend             1 loverscomplaint  
## # ... with 164,646 more rows
```

Syntax with pipe

- Verb(Subject,Complement) replaced by Subject %>% Verb(Complement)
- No need to name unimportant intermediate variables
- Clear syntax (readability)



Create new column

dplyr::mutate function:

```
shakespeare %>%  
  mutate(word = str_to_lower(word)) # convert words to lowercase
```

```
## # A tibble: 164,656 x 5  
##       X1 word      word_count corpus      corpus_date  
##   <dbl> <chr>         <dbl> <chr>         <dbl>  
## 1     1 1 hive             1 loverscomplaint 1609  
## 2     2 2 plaintful         1 loverscomplaint 1609  
## 3     3 3 are              1 loverscomplaint 1609  
## 4     4 4 than             1 loverscomplaint 1609  
## 5     5 5 attended         1 loverscomplaint 1609  
## 6     6 6 that            7 loverscomplaint 1609  
## 7     7 7 moisture         1 loverscomplaint 1609  
## 8     8 8 praised           1 loverscomplaint 1609  
## 9     9 9 particular       1 loverscomplaint 1609  
## 10    10 10 tend            1 loverscomplaint 1609  
## # ... with 164,646 more rows
```

Count number of words in each corpus

```
shakespeare %>%  
  mutate(word = str_to_lower(word)) %>% # convert words to lowercase  
  group_by(word, corpus) %>% # group by word within corpus  
  summarize(n = sum(word_count)) # count
```

```
## # A tibble: 147,219 x 3  
## # Groups:   word [?]  
##   word corpus      n  
##   <chr> <chr>    <dbl>  
## 1 ' 1kinghenryiv    33  
## 2 ' 1kinghenryvi    14  
## 3 ' 2kinghenryiv    38  
## 4 ' 2kinghenryvi    22  
## 5 ' 3kinghenryvi    26  
## 6 ' allswellthatendswell    23  
## 7 ' antonyandcleopatra    23  
## 8 ' asyoulikeit    33  
## 9 ' comedyoferrors    22  
## 10 ' coriolanus    50  
## # ... with 147,209 more rows
```

Order stuff

dplyr::arrange function:

```
shakespeare %>%  
  mutate(word = str_to_lower(word)) %>% # convert words to lowercase  
  group_by(word, corpus) %>% # group by word within corpus  
  summarize(n = sum(word_count)) %>% # count  
  arrange(desc(n)) # decreasing order (wo desc for increasing)
```

```
## # A tibble: 147,219 x 3  
## # Groups:   word [26,928]  
##   word corpus      n  
##   <chr> <chr>    <dbl>  
## 1 the   hamlet    1143  
## 2 the   coriolanus  1127  
## 3 the   kinghenryv  1085  
## 4 and   kinghenryv  1004  
## 5 the   2kinghenryiv  997  
## 6 the   kingrichardiii 991  
## 7 the   cymbeline    971  
## 8 and   hamlet      969  
## 9 and   2kinghenryvi  949  
## 10 the  2kinghenryvi  949  
## # ... with 147,209 more rows
```

Nb of times a word occurs across all corpus

```
shakespeare %>%  
  mutate(word = str_to_lower(word)) %>% # convert words to lowercase  
  group_by(word) %>% # group by word  
  summarize(n = sum(word_count),  
            corpus = n_distinct(corpus)) %>% # count  
  arrange(desc(n)) # decreasing order
```

```
## # A tibble: 26,928 x 3  
##   word      n corpus  
##   <chr> <dbl> <int>  
## 1 the    29801     42  
## 2 and    27529     42  
## 3 i      21029     42  
## 4 to     20957     42  
## 5 of     18514     42  
## 6 a      15370     42  
## 7 you    14010     42  
## 8 my     12936     42  
## 9 in     11722     42  
## 10 that  11519     42  
## # ... with 26,918 more rows
```

Name this processed dataset

```
shakespeare_processed <- shakespeare %>%  
  mutate(word = str_to_lower(word)) %>% # convert words to lowercase  
  group_by(word) %>% # group by word  
  summarize(n = sum(word_count),  
            corpus = n_distinct(corpus)) %>% # count  
  arrange(desc(n)) # decreasing order  
shakespeare_processed
```

```
## # A tibble: 26,928 x 3  
##   word      n corpus  
##   <chr> <dbl> <int>  
## 1 the    29801     42  
## 2 and    27529     42  
## 3 i      21029     42  
## 4 to     20957     42  
## 5 of     18514     42  
## 6 a      15370     42  
## 7 you    14010     42  
## 8 my     12936     42  
## 9 in     11722     42  
## 10 that  11519     42  
## # ... with 26,918 more rows
```


Get rid of common words

dplyr::anti_join function:

```
words <- shakespeare_processed %>%  
  group_by(word) %>% # group by word within corpus  
  anti_join(tidytext::stop_words) %>% # returns all rows from  
                                         # shakespeare that do not  
                                         # match the stop words  
  arrange(desc(n))  
words
```

```
## # A tibble: 26,322 x 3  
## # Groups:   word [26,322]  
##   word      n corpus  
##   <chr> <dbl> <int>  
## 1 thou   5771     42  
## 2 thy    4269     42  
## 3 thee   3383     42  
## 4 lord   3357     40  
## 5 king   3322     40  
## 6 sir    3025     37  
## 7 enter  2406     39  
## 8 love   2249     42  
## 9 hath   2032     42  
## 10 'tis   1435     42  
## # ... with 26,312 more rows
```

Find words > 4 char not in all corpus

dplyr::filter function to select rows:

```
words <- words %>%  
  filter(corpus < 42, nchar(word) > 4) %>%  
  arrange(desc(n))  
words
```

```
## # A tibble: 23,890 x 3  
## # Groups:   word [23,890]  
##   word      n corpus  
##   <chr> <dbl> <int>  
## 1 enter   2406     39  
## 2 henry   1311     13  
## 3 speak  1194     40  
## 4 exeunt  1061     37  
## 5 queen   1005     35  
## 6 death    933     41  
## 7 night    933     41  
## 8 father   868     40  
## 9 scene    825     38  
## 10 master   803     39  
## # ... with 23,880 more rows
```



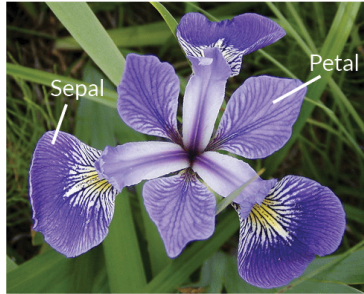
Check out other functions

- **spread()** and **gather()** from package tidyr to reshape tibbles
- **inner_join()**, **left_join()**, **right_join()**, **full_join()**, **semi_join()** and **anti_join()** from package dplyr to join two tibbles together
- **year()**, **month()**, etc... from package lubridate to manipulate dates
- Feel free to explore other Tidyverse packages, in particular forcats and stringr



Visualize

Case study on Fisher's iris dataset



Iris Versicolor



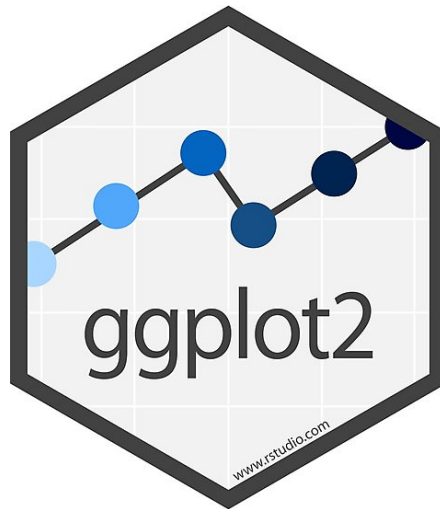
Iris Setosa



Iris Virginica

Visualization with ggplot2

- The package ggplot2 implements a **g**rammar of **g**raphics
- Operates on data.frames or tibbles, not vectors like base R
- Explicitly differentiates between the data and its representation



The ggplot2 grammar

Grammar element

Data

Geometrics

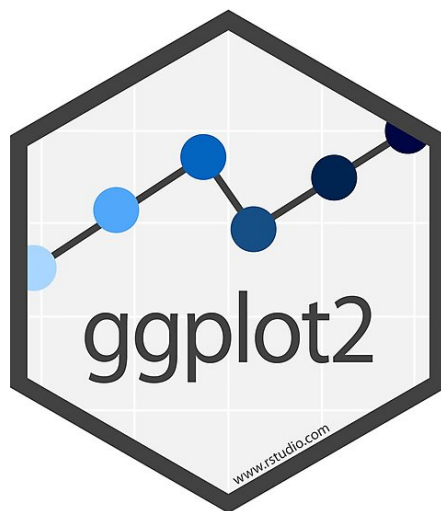
Aesthetics

What it is

The data frame being plotted

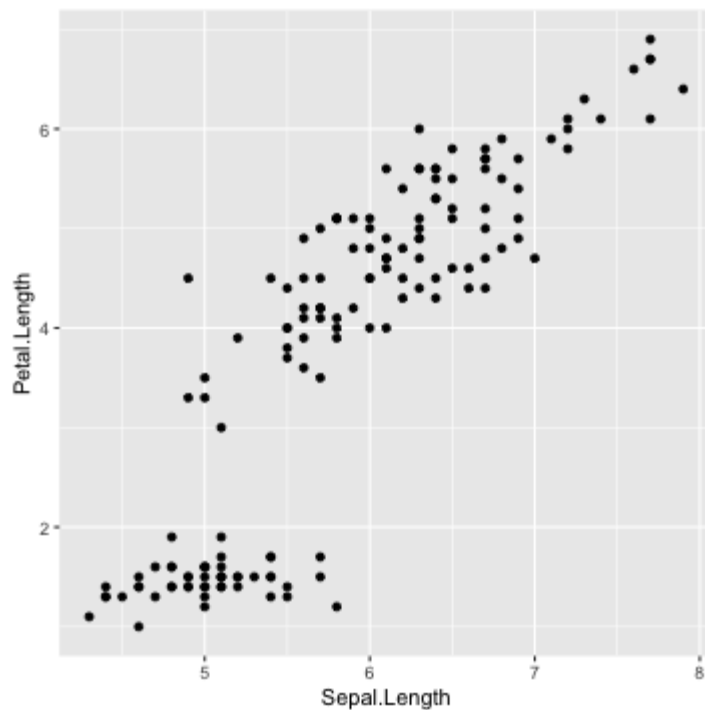
The geometric shape that will represent the data (e.g., point, boxplot, histogram)

The aesthetics of the geometric object (e.g., color, size, shape)



Scatterplots

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length, y = Petal.Length) +  
  geom_point()
```



Scatterplots

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length, y = Petal.Length) +  
  geom_point()
```

- Pass in the data frame as your first argument

Scatterplots

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length, y = Petal.Length) +  
  geom_point()
```

- Pass in the data frame as your first argument
- Aesthetics maps the data onto plot characteristics, here x and y axes

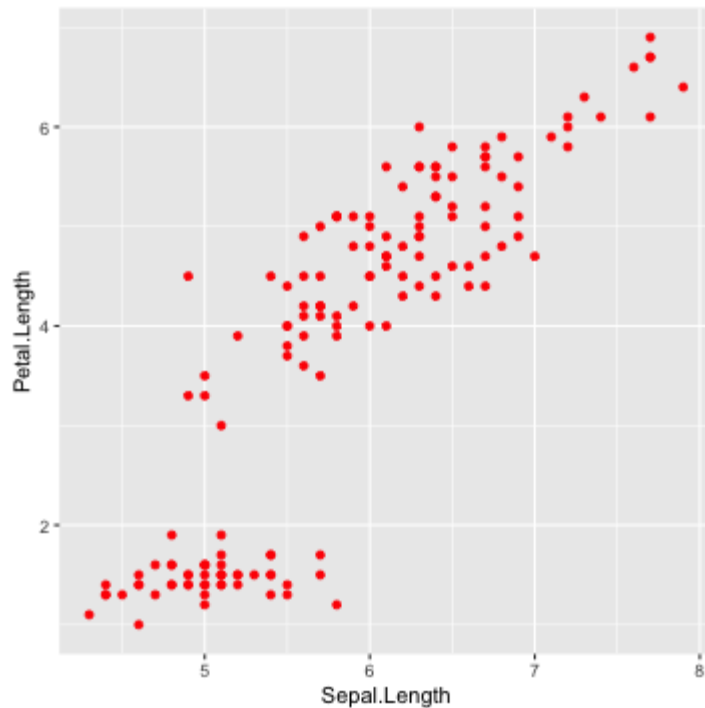
Scatterplots

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length, y = Petal.Length) +  
  geom_point()
```

- Pass in the data frame as your first argument
- Aesthetics maps the data onto plot characteristics, here x and y axes
- Display the data geometrically as points

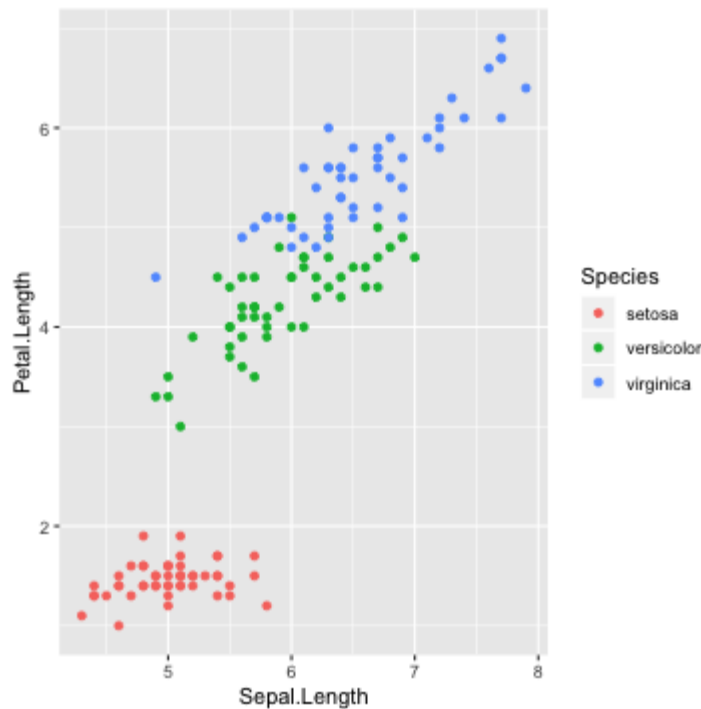
Scatterplots, with colors

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length, y = Petal.Length) +  
  geom_point(color = "red")
```



Scatterplots, with species-specific colors

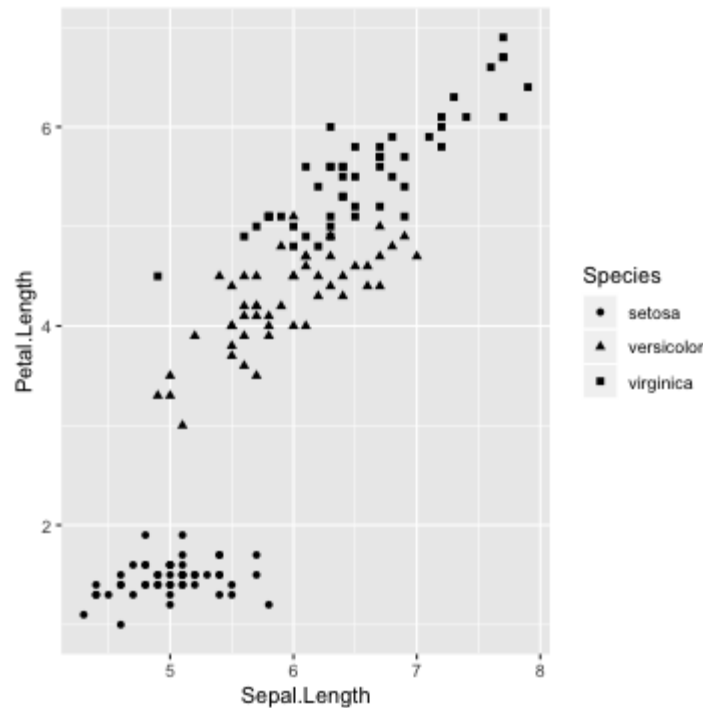
```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length, y = Petal.Length, color = Species) +  
  geom_point()
```



- Placing color inside aesthetic maps it to the data

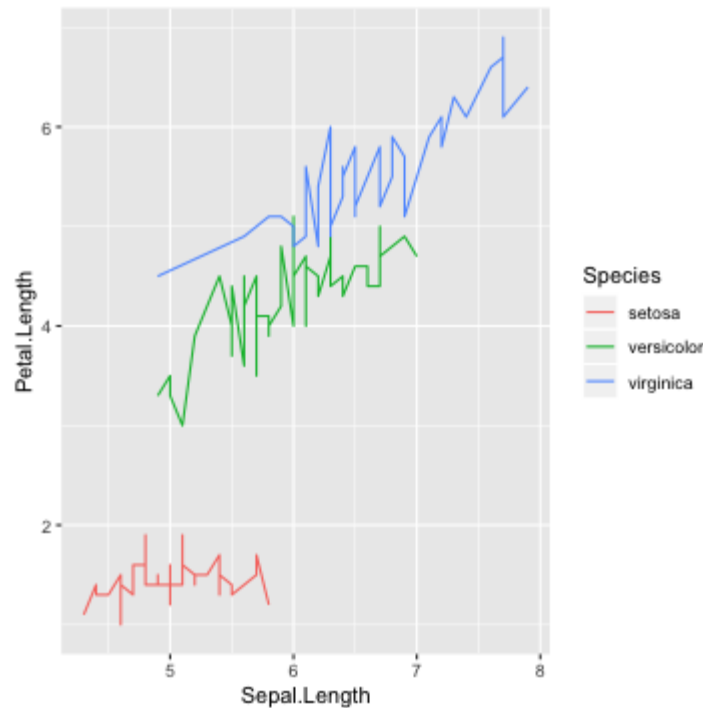
Scatterplots, with species-specific shapes

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length, y = Petal.Length, shape = Species) +  
  geom_point()
```



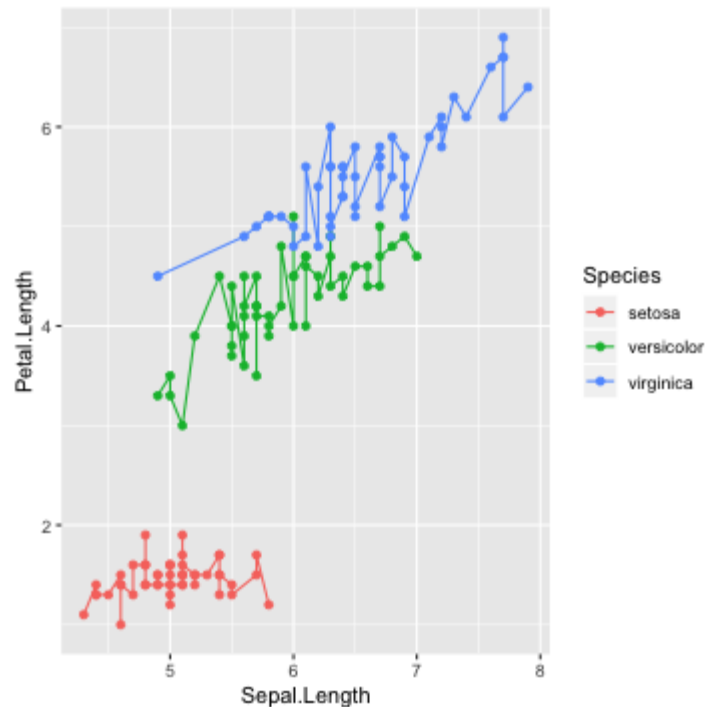
Scatterplots, lines instead of points

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length, y = Petal.Length, color = Species) +  
  geom_line()
```



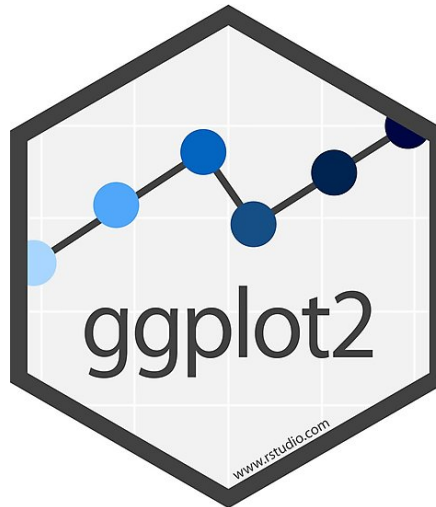
Scatterplots, add points

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length, y = Petal.Length, color = Species) +  
  geom_line() +  
  geom_point()
```



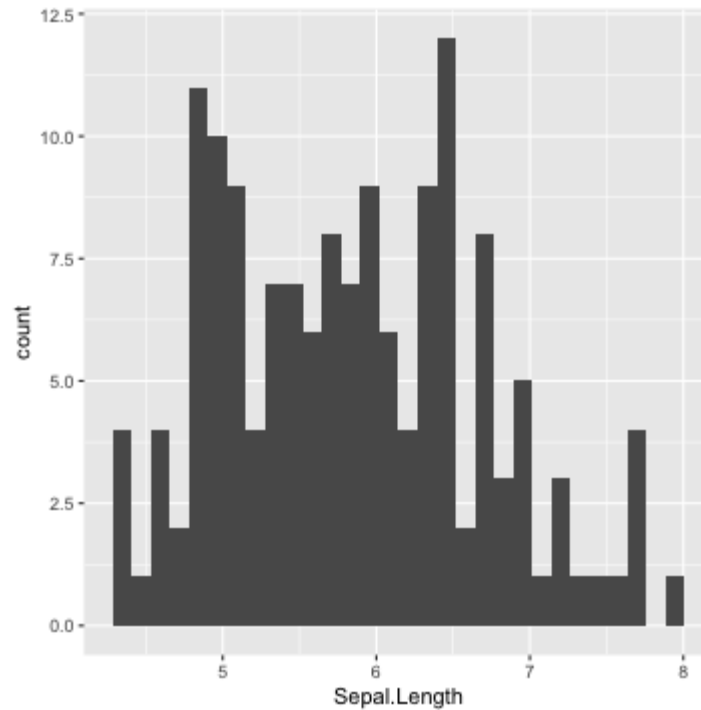
aes ou pas aes?

- Si on établit un lien entre les valeurs d'une variable et un attribut graphique, on définit un mappage, et on le déclare dans `aes()`.
- Sinon, on modifie l'attribut de la même manière pour tous les points, et on le définit en dehors de la fonction `aes()`.



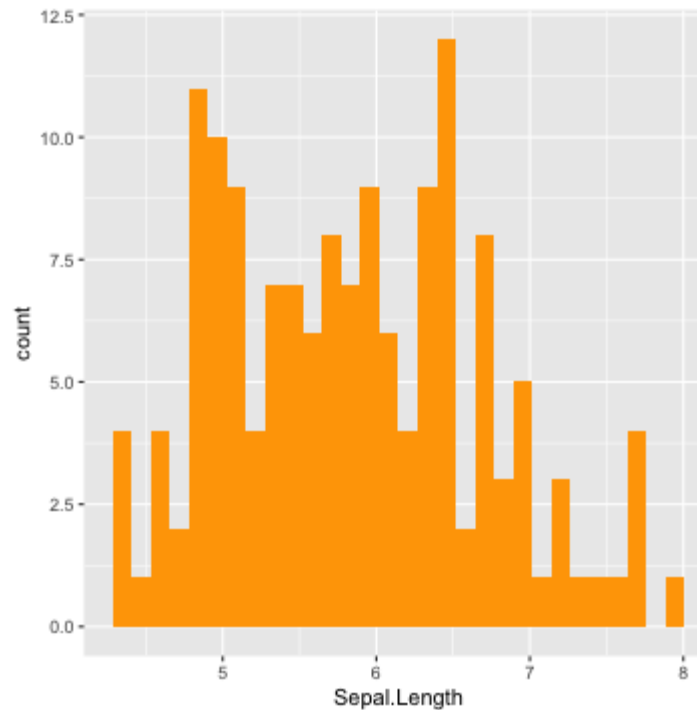
Histograms

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length) +  
  geom_histogram()
```



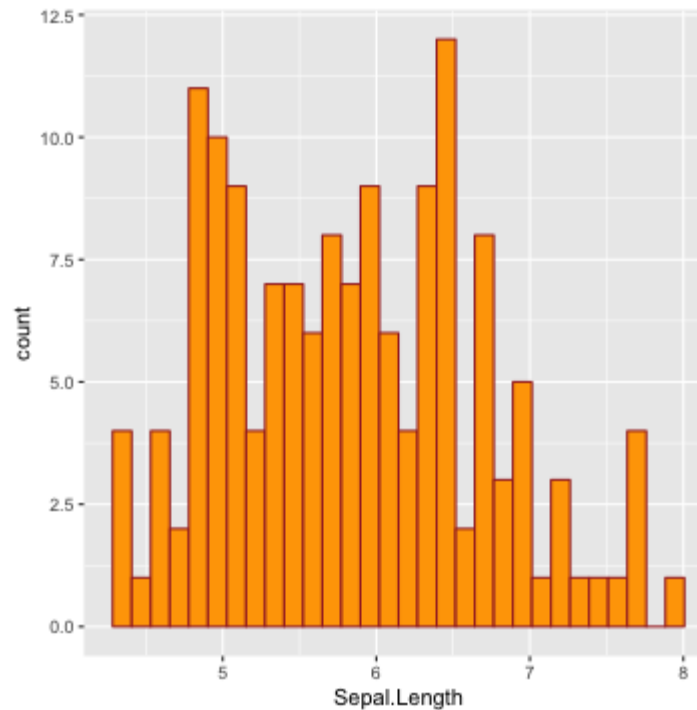
Histograms, with colors

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length) +  
  geom_histogram(fill = "orange")
```



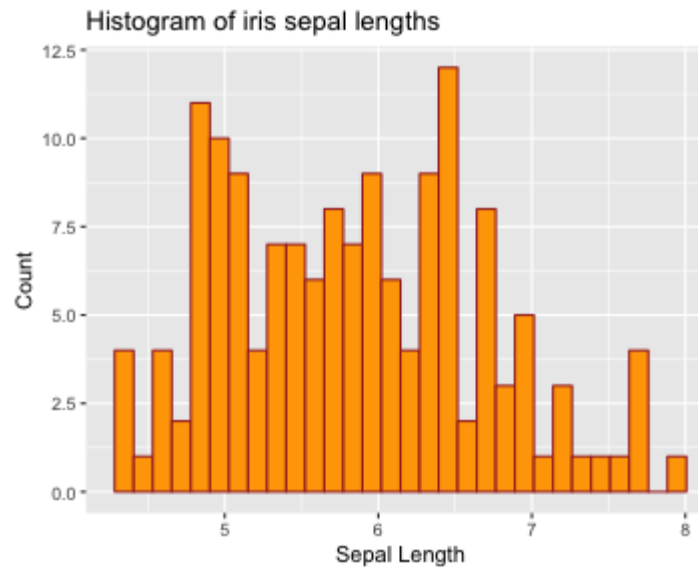
Histograms, with colors

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length) +  
  geom_histogram(fill = "orange", color = "brown")
```



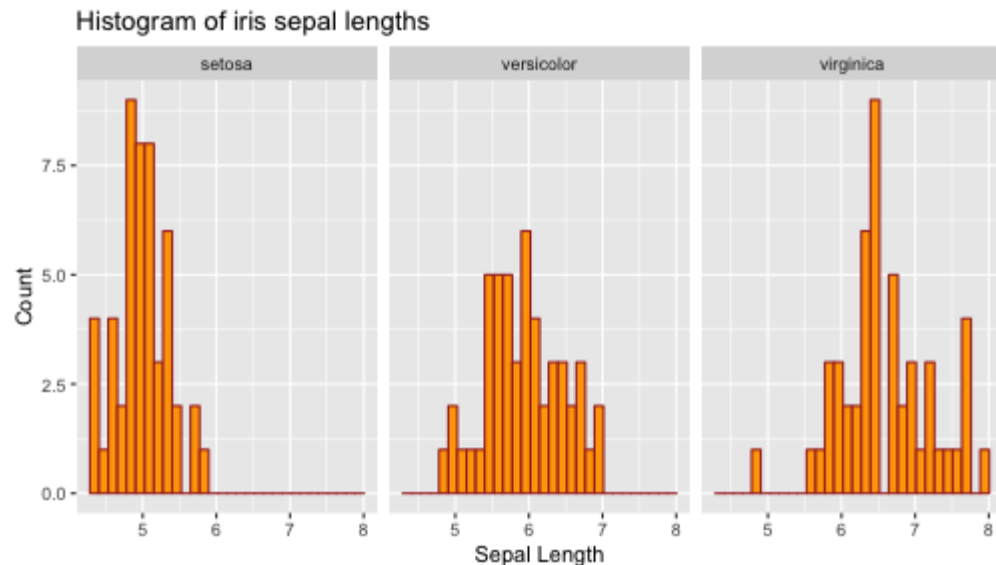
Histograms, with labels and title

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length) +  
  geom_histogram(fill = "orange", color = "brown") +  
  xlab("Sepal Length") +  
  ylab("Count") +  
  ggtitle("Histogram of iris sepal lengths")
```



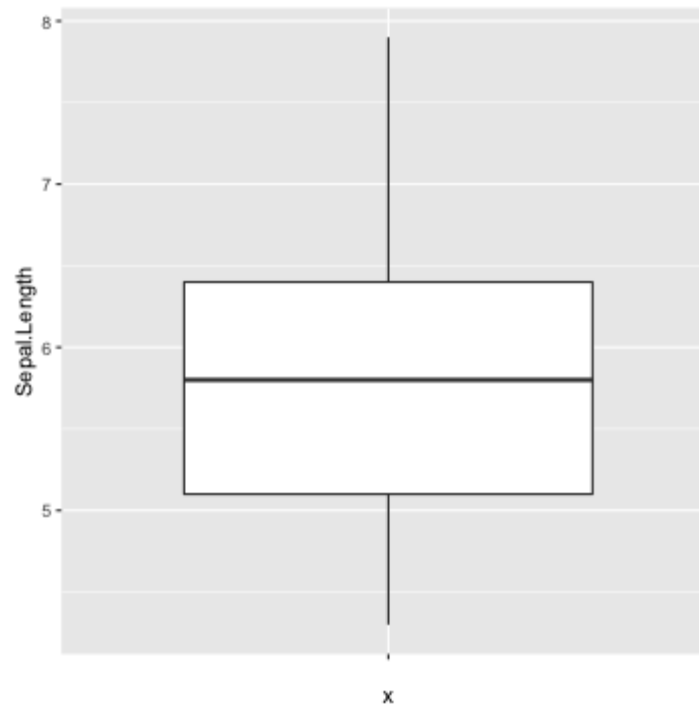
Histograms, by species

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length) +  
  geom_histogram(fill = "orange", color = "brown") +  
  xlab("Sepal Length") +  
  ylab("Count") +  
  ggtitle("Histogram of iris sepal lengths") +  
  facet_wrap(aes(Species))
```



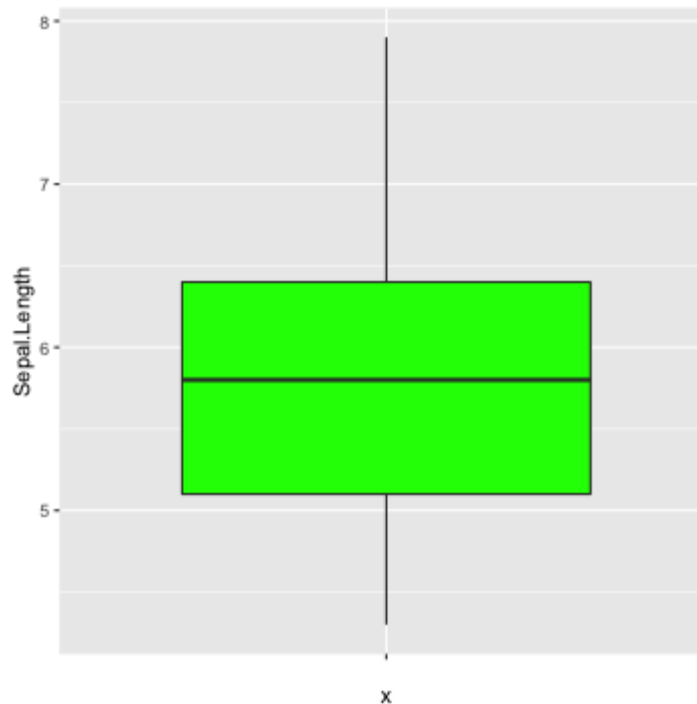
Boxplots

```
iris %>%  
  ggplot() +  
  aes(x = "", y = Sepal.Length) +  
  geom_boxplot()
```



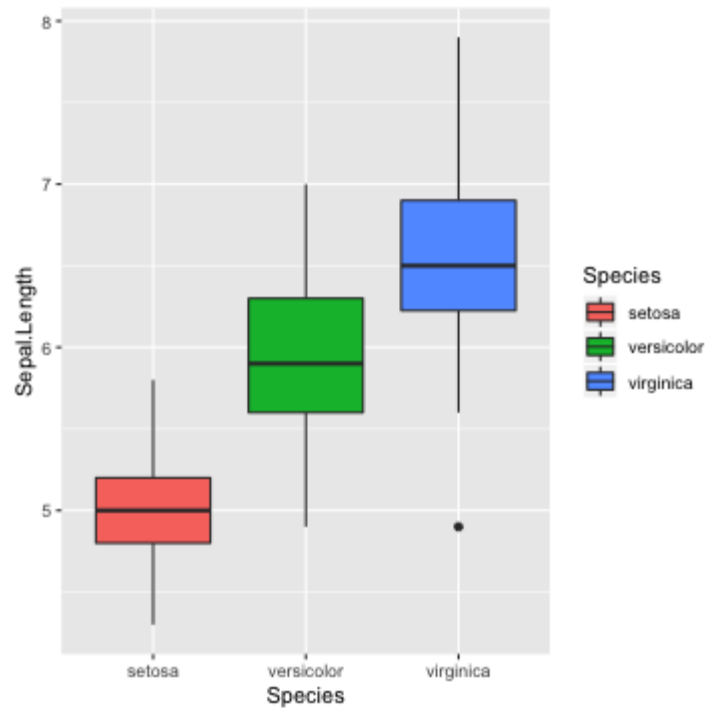
Boxplots with colors

```
iris %>%  
  ggplot() +  
  aes(x = "", y = Sepal.Length) +  
  geom_boxplot(fill = "green")
```



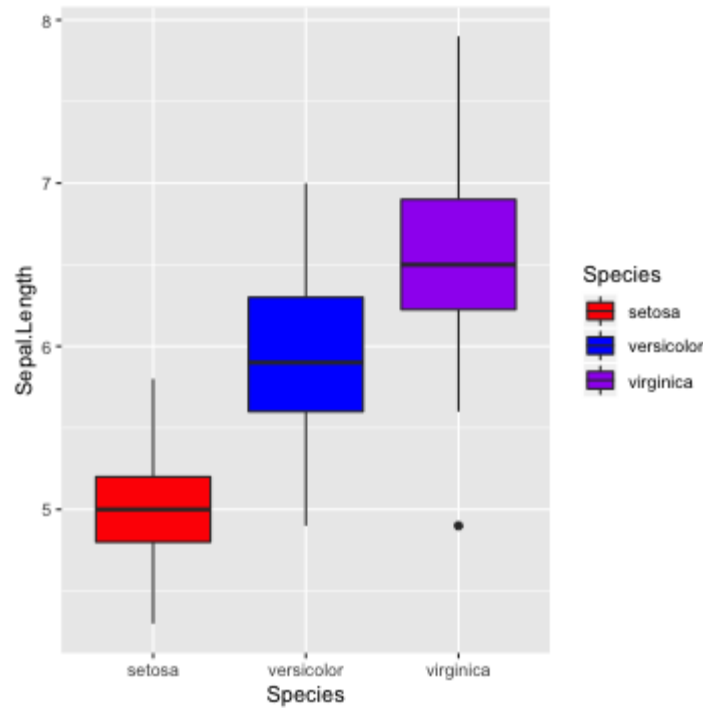
Boxplots with colors by species

```
iris %>%  
  ggplot() +  
  aes(x = Species, y = Sepal.Length, fill = Species) +  
  geom_boxplot()
```



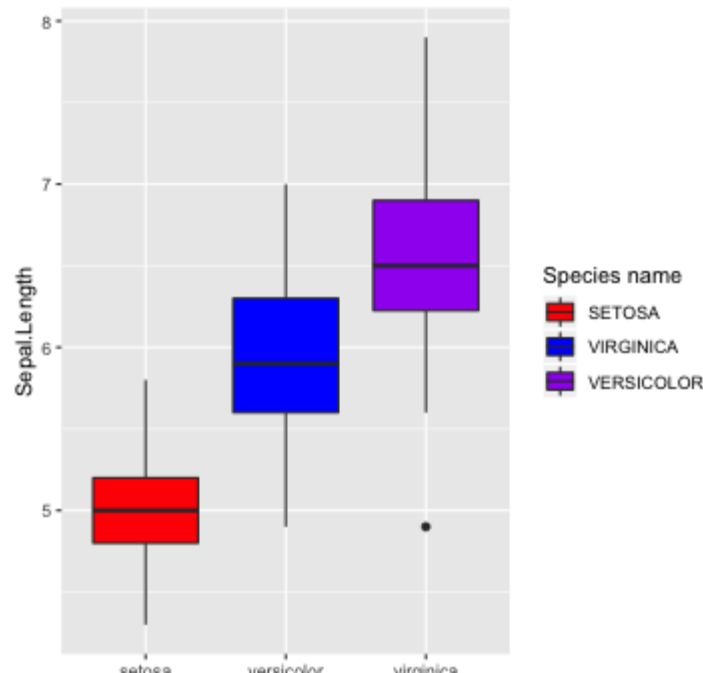
Boxplots, user-specified colors by species

```
iris %>%  
  ggplot() +  
  aes(x = Species, y = Sepal.Length, fill = Species) +  
  geom_boxplot() +  
  scale_fill_manual(  
    values=c("red", "blue", "purple")  
  )  
)
```



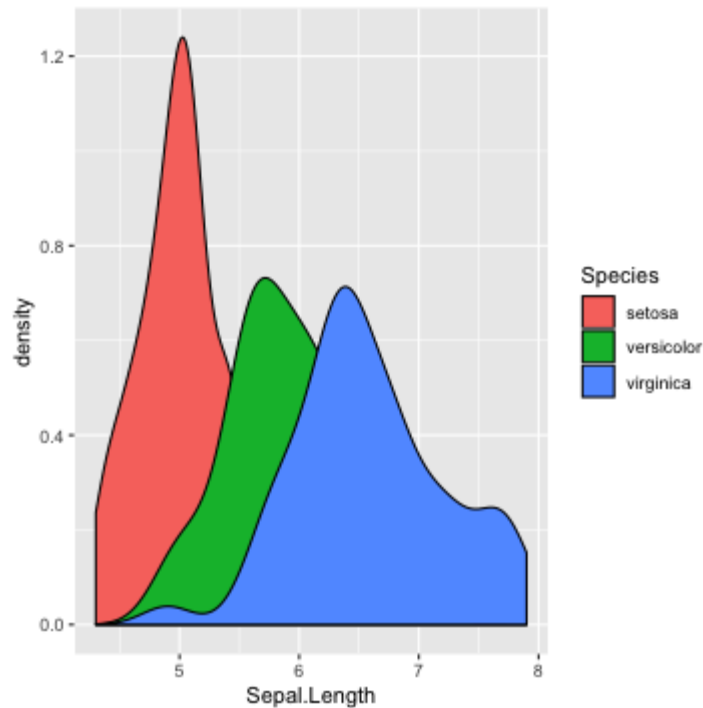
Boxplots, change legend settings

```
iris %>%  
  ggplot() +  
  aes(x = Species, y = Sepal.Length, fill = Species) +  
  geom_boxplot() +  
  scale_fill_manual(  
    values=c("red", "blue", "purple"),  
    name = "Species name",  
    labels=c("SETOSA", "VIRGINICA", "VERSICOLOR"))
```



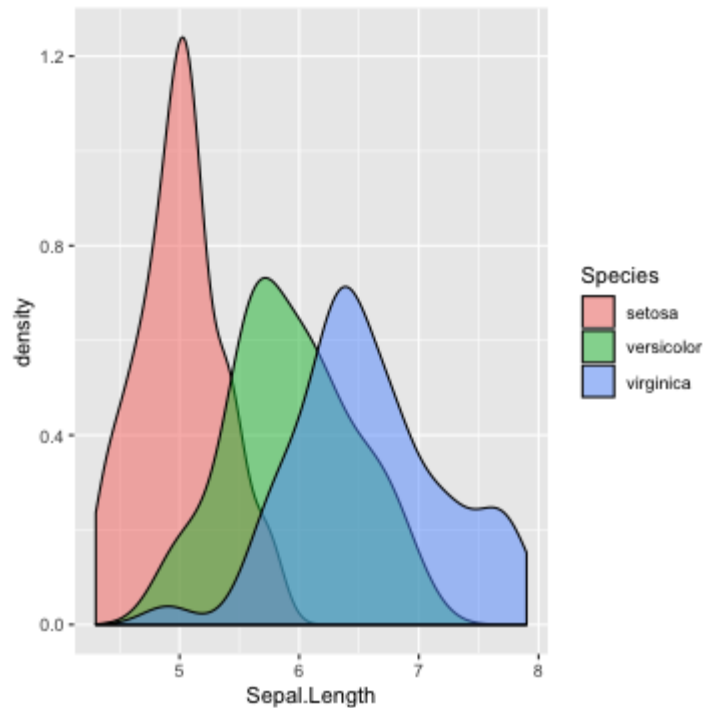
Density plots

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length, fill = Species) +  
  geom_density()
```



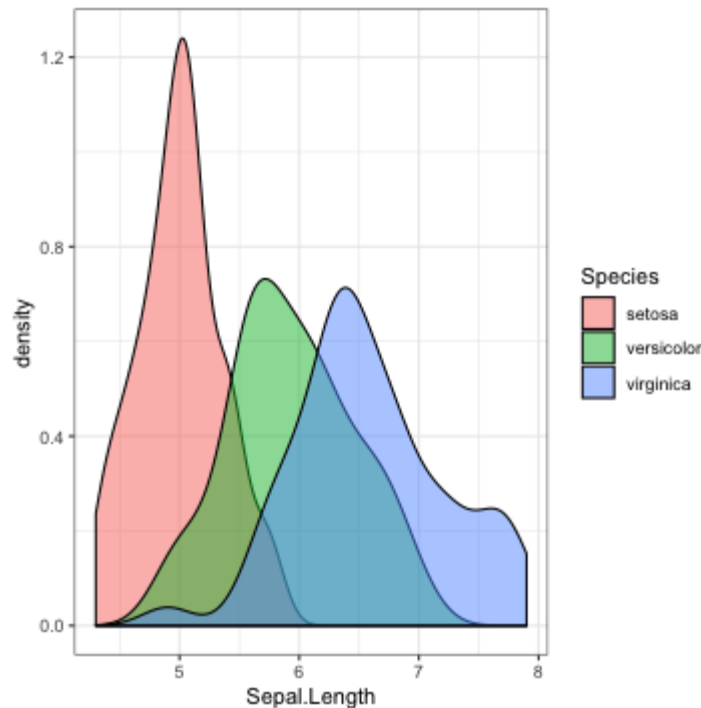
Density plots, control transparency

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length, fill = Species) +  
  geom_density(alpha = 0.5)
```



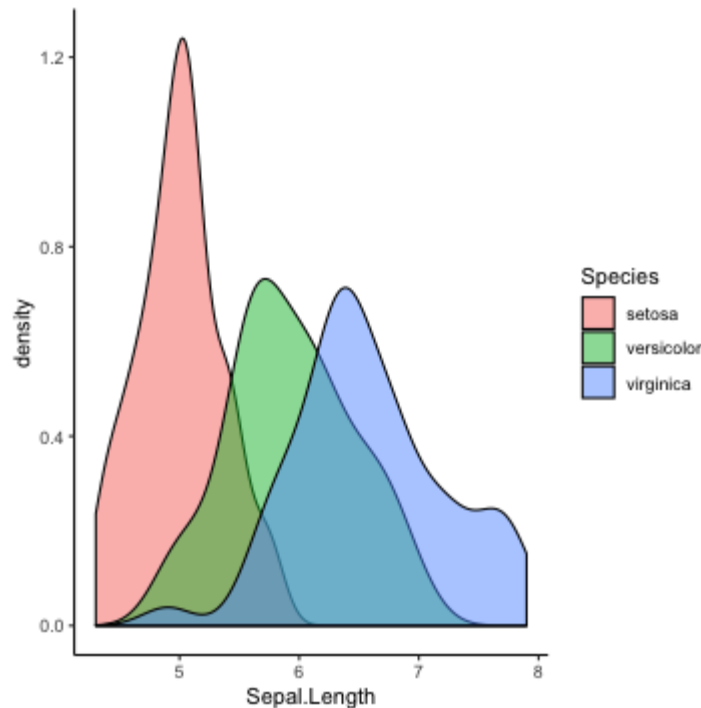
Change default background theme 1/3

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length, fill = Species) +  
  geom_density(alpha = 0.5) +  
  theme_bw()
```



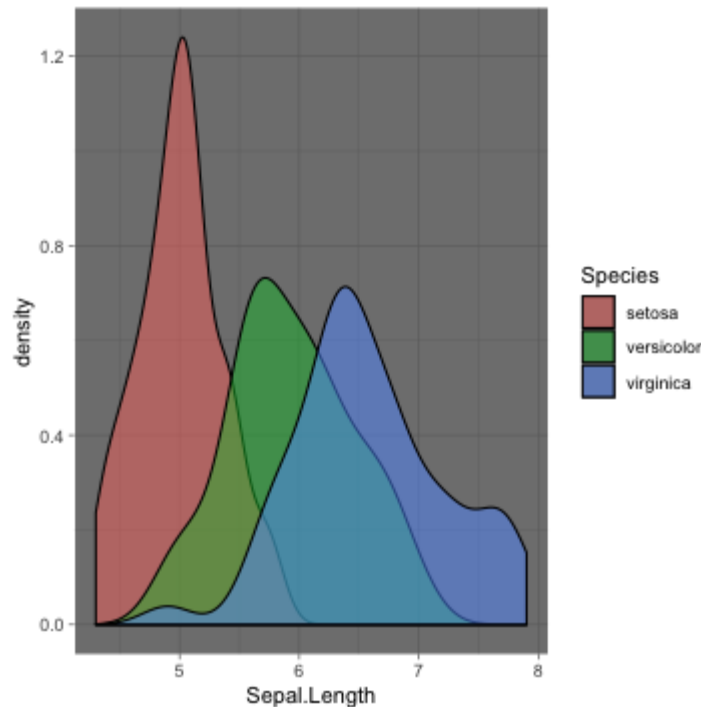
Change default background theme 2/3

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length, fill = Species) +  
  geom_density(alpha = 0.5) +  
  theme_classic()
```



Change default background theme 3/3

```
iris %>%  
  ggplot() +  
  aes(x = Sepal.Length, fill = Species) +  
  geom_density(alpha = 0.5) +  
  theme_dark()
```





To dive even deeper in the tidyverse

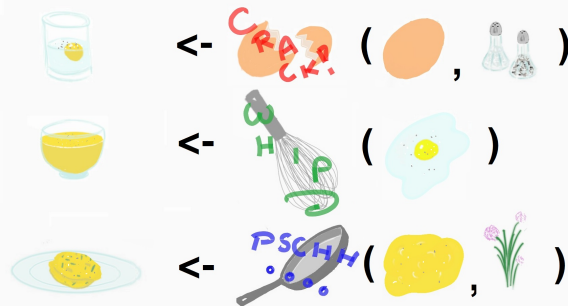
- **Learn the tidyverse**: books, workshops and online courses
- My selection of books:
 - **R for Data Science et Advanced R**
 - **Introduction à R et au tidyverse**
 - **Fundamentals of Data visualization**
 - **Data Visualization: A practical introduction**
- **Tidy Tuesdays videos** by D. Robinson chief data scientist at DataCamp

Blog Lise Vaudor

Base R

```
white_and_yolk <- crack(egg, add_seasoning)  
omelette_batter <- beat(white_and_yolk)  
omelette_with_chives <- cook(omelette_batter, add_chives)
```

Successive command lines



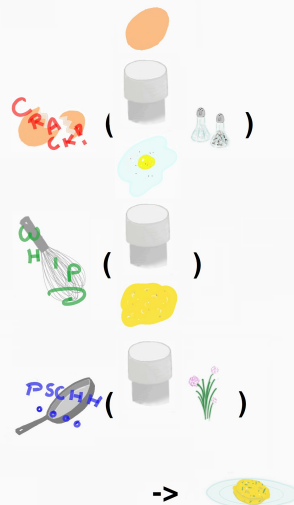
@LVaudor

Blog Lise Vaudor

Piping

```
egg %>%  
  crack(add_seasoning) %>%  
  beat() %>%  
  cook(add_chives) -> omelette_with_chives
```

Piped command line



The RStudio Cheat Sheets

Data Transformation with dplyr : : CHEAT SHEET



dplyr functions work with pipes and expect **tidy data**. In tidy data:



Each **variable** is in its own **column**



Each **observation**, or **case**, is in its own **row**



pipes
 $x \%>\% f(y)$
becomes $f(x, y)$

Summarise Cases

These apply **summary functions** to columns to create a new table of summary statistics. Summary functions take vectors as input and return one value (see back).

summary function



summarise(data, ...) Compute table of summaries.
`summarise(mtcars, avg = mean(mpg))`



count(x, ..., wt = NULL, sort = FALSE) Count number of rows in each group defined by the variables in ... Also **tally**!.
`count(iris, Species)`

VARIATIONS

summarise_all() - Apply funs to every column.
summarise_at() - Apply funs to specific columns.
summarise_if() - Apply funs to all cols of one type.

Group Cases

Use **group_by()** to create a "grouped" copy of a table. dplyr functions will manipulate each "group" separately and then combine the results.



`mtcars %>%
group_by(cyl) %>%
summarise(avg = mean(mpg))`

group_by(data, ..., add = FALSE) Returns copy of table grouped by ...
`g_iris <- group_by(iris, Species)`

ungroup(x, ...) Returns ungrouped copy of table.
`ungroup(g_iris)`

Manipulate Cases

EXTRACT CASES

Row functions return a subset of rows as a new table.



filter(data, ...) Extract rows that meet logical criteria. `filter(iris, Sepal.Length > 7)`



distinct(data, ..., keep_all = FALSE) Remove rows with duplicate values.
`distinct(iris, Species)`



sample_frac(tbl, size = 1, replace = FALSE, weight = NULL, env = parent.frame()) Randomly select fraction of rows.
`sample_frac(iris, 0.5, replace = TRUE)`



sample_n(tbl, size, replace = FALSE, weight = NULL, env = parent.frame()) Randomly select size rows. `sample_n(iris, 10, replace = TRUE)`



slice(data, ...) Select rows by position.

`slice(iris, 10:15)`

top_n(x, n, wt) Select and order top n entries (by group if grouped data). `top_n(iris, 5, Sepal.Width)`

Logical and boolean operators to use with filter()

<	<=	is.na()	%in%		xor()
>	>=	is.na()	!	&	

See ?base::logic and ?Comparison for help.

ARRANGE CASES



arrange(data, ...) Order rows by values of a column or columns (low to high), use with **desc()** to order from high to low.
`arrange(mtcars, mpg)`
`arrange(mtcars, desc(mpg))`

ADD CASES



add_row(data, ..., before = NULL, after = NULL) Add one or more rows to a table.
`add_row(faithful, eruptions = 1, waiting = 1)`

Manipulate Variables

EXTRACT VARIABLES

Column functions return a set of columns as a new vector or table.



pull(data, var = 1) Extract column values as a vector. Choose by name or index.
`pull(iris, Sepal.Length)`



select(data, ...) Extract columns as a table. Also **select_if()**.
`select(iris, Sepal.Length, Species)`

Use these helpers with **select()**, e.g. `select(iris, starts_with("Sepal"))`

contains (match)	num_range (prefix, range)	!, e.g. <code>mpg:cyl</code>
ends_with (match)	one_of (...)	~, e.g. <code>Species</code>
matches (match)	starts_with (match)	

MAKE NEW VARIABLES

These apply **vectorized functions** to columns. Vectorized funs take vectors as input and return vectors of the same length as output (see back).

vectorized function



mutate(data, ...) Compute new column(s).
`mutate(mtcars, gpm = 1/mpg)`



transmute(data, ...) Compute new column(s), drop others.
`transmute(mtcars, gpm = 1/mpg)`



mutate_all(tbl, funs, ...) Apply funs to every column. Use with **funs()**. Also **mutate_if()**.
`mutate_all(faithful, funs(log(), log2(), log()))`
`mutate_if(iris, is.numeric, funs(log(), log2(), log()))`



mutate_at(tbl, cols, funs, ...) Apply funs to specific columns. Use with **funs()**, **vars()** and the helper functions for **select()**.
`mutate_at(iris, vars(Species), funs(log(), log2(), log()))`



add_column(data, ..., before = NULL, after = NULL) Add new column(s). Also **add_count()**, **add_tally()**. `add_column(mtcars, new = 1:32)`



rename(data, ...) Rename columns.
`rename(iris, Length = Sepal.Length)`



Thanks!

I created these slides with **xaringan** and **RMarkdown** using the **rutgers css** that I slightly modified.

Credit: I used material from **Cécile Sauder**, **Stephanie J. Spielman** and **Julien Barnier**.



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