README

- Review of the 1st chapter of the DataCamp course
- Add your name and (pledged) in the #+AUTHOR: meta headline
- When you've completed the file, submit it in Canvas
- You'll get solutions after the deadline has passed

DONE Identify yourself

- 1. Add your name and (pledged) at the top
- 2. Run the #+PROPERTY and #+STARTUP lines with C-c C-c
- 3. You should open the R session in the same directory as this file
- 4. To check, run getwd() in the R console window

DONE Answer conceptual questions

• What is the requirement for this course and where can you find this information? - "Introduction to the Tidyverse" (Bottom of dashboard)

This is an introduction to the programming language R, focused on a powerful set of tools known as the Tidyverse. You'll learn the intertwined processes of data manipulation and visualization using the tools dplyr and ggplot2. You'll learn to manipulate data by filtering, sorting, and summarizing a real dataset of historical country data in order to answer exploratory questions. You'll then learn to turn this processed data into informative line plots, bar plots, histograms, and more with the ggplot2 package. You'll get a taste of the value of exploratory data analysis and the power of Tidyverse tools. This is a suitable introduction for those who have no previous experience in R and are interested in performing data analysis.

• What is the "Tidyverse"? It's a bundle of R packages including ggplot2 (which predates the "Tidyverse" by several years), dplyr for data frame manipulation, and many more. Its functions rely on data being "tidy", which corresponds to Codds 3rd normal form for relational or tabular data.



(Image source: hbctraining.github.io)

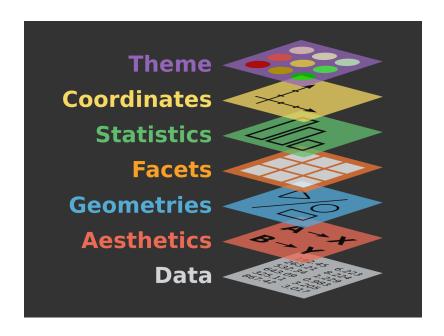
• Base R is the foundation that every data scientist should know.

"If the user knows base-R (not difficult), she can handle any situation with just a few simple operations. The old adage applies: "Give a man a fish, and he can eat for a day. Teach him how to fish, and he can eat for a lifetime." From: "TidyverseSceptic" (Matloff, 2022)

• What is the "Grammar of Graphics"?

The "Grammar of Graphics" (gg) is a plotting framework by Leland Wilkinson (1999) implemented in R's ggplot2 plotting package. Its core ideas are:

- 1. Graphics are distinct layers of 'grammatical' elements
- 2. Plots are given meaning through 'aesthetic' mappings
- What are the layers of the 'Grammar of Graphics'?



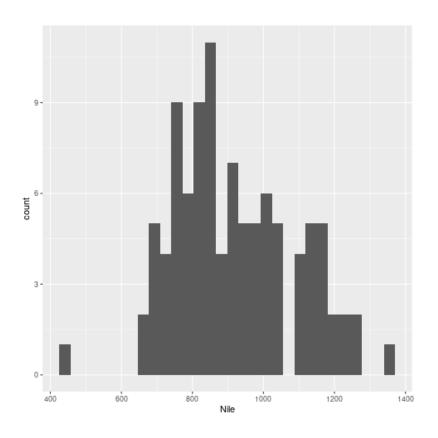
(Image source: r.qcbs.ca)

• What are some examples for these elements in ggplot2?

```
Data
                                    {variables of interest}
                    x-axis
y-axis
                                colour
fill
                                              size
                                                         alpha
shape
                                                                    line width
line type
  Aesthetics
                                             labels
 Geometries
                     point
                                  line
                                           histogram
                                                          bar
                                                                     boxplot
     Themes non-data ink
   Statistics
                    binning
                              smoothing descriptive inferential
Coordinates
                                                          limits
                   cartesian
                                 fixed
                                             polar
       Facets
                   columns
                                 rows
```

• Can you save a ggplot2 plot as an R object?

```
library(ggplot2)
g <- ggplot(
  data = data.frame(Nile),
  aes(Nile)) +
  geom_histogram()
g</pre>
```



• Show the structure and attributes of g:

```
attributes(g)
str(g)
```

```
$names
```

```
[1] "data" "layers" "scales" "guides" "mapping" "theme" [7] "coordinates" "facet" "plot_env" "layout" "labels"
```

\$class

[1] "gg" "ggplot"

List of 11

- \$ data :'data.frame': 100 obs. of 1 variable:
- ..\$ Nile: Time-Series [1:100] from 1871 to 1970: 1120 1160 963 1210 1160 1160 8
- \$ layers :List of 1
- ..\$:Classes 'LayerInstance', 'Layer', 'ggproto', 'gg' <ggproto object: Class La

```
aes_params: list
  compute_aesthetics: function
  compute_geom_1: function
  compute_geom_2: function
  compute_position: function
  compute_statistic: function
  computed_geom_params: list
  computed_mapping: uneval
  computed_stat_params: list
  constructor: call
  data: waiver
  draw_geom: function
  finish_statistics: function
  geom: <ggproto object: Class GeomBar, GeomRect, Geom, gg>
aesthetics: function
default aes: uneval
draw_group: function
draw_key: function
draw_layer: function
draw_panel: function
extra_params: just na.rm orientation
handle_na: function
non_missing_aes: xmin xmax ymin ymax
optional_aes:
parameters: function
rename_size: TRUE
required_aes: x y
setup_data: function
setup_params: function
use_defaults: function
super: <ggproto object: Class GeomRect, Geom, gg>
  geom_params: list
  inherit.aes: TRUE
  layer_data: function
  map_statistic: function
  mapping: NULL
  position: <ggproto object: Class PositionStack, Position, gg>
compute_layer: function
compute_panel: function
fill: FALSE
```

```
required_aes:
reverse: FALSE
setup_data: function
setup_params: function
type: NULL
vjust: 1
super: <ggproto object: Class Position, gg>
  print: function
  setup_layer: function
  show.legend: NA
  stat: <ggproto object: Class StatBin, Stat, gg>
 aesthetics: function
compute_group: function
compute_layer: function
compute_panel: function
default_aes: uneval
dropped_aes: weight
extra_params: na.rm orientation
finish_layer: function
non_missing_aes:
optional_aes:
parameters: function
required_aes: x | y
retransform: TRUE
setup_data: function
setup_params: function
super: <ggproto object: Class Stat, gg>
  stat_params: list
  super: <ggproto object: Class Layer, gg>
$ scales
             :Classes 'ScalesList', 'ggproto', 'gg' <ggproto object: Class Scales
  add: function
  add_defaults: function
  add_missing: function
  backtransform_df: function
  clone: function
  find: function
  get_scales: function
  has_scale: function
  input: function
```

map_df: function

```
n: function
  non_position_scales: function
  scales: list
  train_df: function
  transform_df: function
   super: <ggproto object: Class ScalesList, gg>
$ guides
             :Classes 'Guides', 'ggproto', 'gg' <ggproto object: Class Guides, gg
  add: function
  assemble: function
  build: function
  draw: function
  get_custom: function
  get_guide: function
  get_params: function
  get_position: function
  guides: NULL
  merge: function
  missing: <ggproto object: Class GuideNone, Guide, gg>
add_title: function
arrange_layout: function
assemble_drawing: function
available_aes: any
build_decor: function
build_labels: function
build_ticks: function
build_title: function
draw: function
draw_early_exit: function
elements: list
extract_decor: function
extract_key: function
extract_params: function
get_layer_key: function
hashables: list
measure_grobs: function
merge: function
override_elements: function
params: list
process_layers: function
```

setup_elements: function

```
setup_params: function
train: function
transform: function
super: <ggproto object: Class GuideNone, Guide, gg>
  package_box: function
  print: function
  process_layers: function
  setup: function
  subset_guides: function
  train: function
  update_params: function
   super: <ggproto object: Class Guides, gg>
$ mapping
            :List of 1
 ..$ x: language ~Nile
 ....- attr(*, ".Environment")=<environment: R_GlobalEnv>
 ... attr(*, "class")= chr "uneval"
$ theme
             : list()
$ coordinates:Classes 'CoordCartesian', 'Coord', 'ggproto', 'gg' <ggproto object
  aspect: function
  backtransform_range: function
  clip: on
  default: TRUE
  distance: function
  expand: TRUE
  is_free: function
  is_linear: function
  labels: function
  limits: list
  modify_scales: function
  range: function
  render_axis_h: function
  render_axis_v: function
  render_bg: function
  render_fg: function
  setup_data: function
  setup_layout: function
   setup_panel_guides: function
   setup_panel_params: function
   setup_params: function
  train_panel_guides: function
```

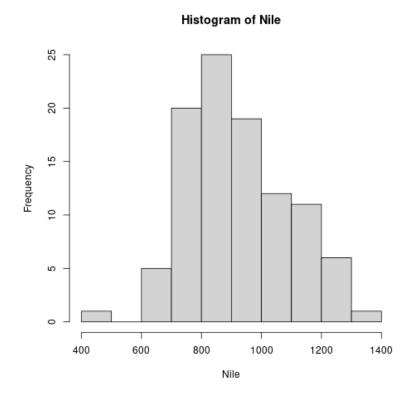
```
transform: function
  super: <ggproto object: Class CoordCartesian, Coord, gg>
             :Classes 'FacetNull', 'Facet', 'ggproto', 'gg' <ggproto object: Clas
   compute_layout: function
  draw_back: function
  draw_front: function
  draw_labels: function
  draw_panels: function
  finish_data: function
  init_scales: function
  map_data: function
  params: list
  setup_data: function
  setup_params: function
  shrink: TRUE
  train_scales: function
  vars: function
  super: <ggproto object: Class FacetNull, Facet, gg>
$ plot_env :<environment: R_GlobalEnv>
             :Classes 'Layout', 'ggproto', 'gg' <ggproto object: Class Layout, gg
$ layout
  coord: NULL
  coord_params: list
  facet: NULL
  facet_params: list
  finish_data: function
  get_scales: function
  layout: NULL
  map_position: function
  panel_params: NULL
  panel_scales_x: NULL
  panel_scales_y: NULL
  render: function
  render_labels: function
  reset_scales: function
  resolve_label: function
  setup: function
  setup_panel_guides: function
   setup_panel_params: function
  train_position: function
```

super: <ggproto object: Class Layout, gg>

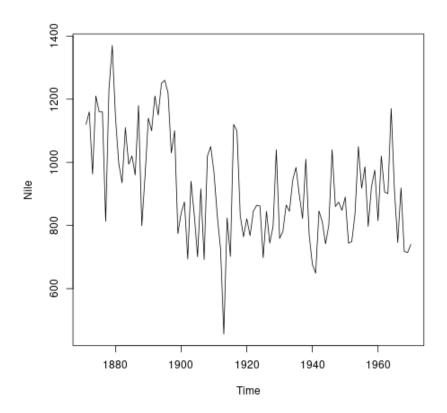
```
$ labels :List of 3
..$ x : chr "Nile"
..$ y : chr "count"
.. ..- attr(*, "fallback")= logi TRUE
..$ weight: chr "weight"
.. ..- attr(*, "fallback")= logi TRUE
- attr(*, "class")= chr [1:2] "gg" "ggplot"
```

• Can you save a base R plot as an R object?

```
h <- hist(Nile)
h</pre>
```



```
p <- plot(Nile)
p</pre>
```



attributes(h) # saving histogram to access its attributes and
 # structure
str(h)

```
$names
```

[1] "breaks" "counts" "density" "mids" "xname" "equidist"

\$class

[1] "histogram"

List of 6

\$ breaks : int [1:11] 400 500 600 700 800 900 1000 1100 1200 1300 ...

\$ counts : int [1:10] 1 0 5 20 25 19 12 11 6 1

\$ density : num [1:10] 0.0001 0 0.0005 0.002 0.0025 0.0019 0.0012 0.0011 0.0006 0.000

\$ mids : num [1:10] 450 550 650 750 850 950 1050 1150 1250 1350

```
$ xname : chr "Nile"
$ equidist: logi TRUE
- attr(*, "class")= chr "histogram"
```

• Can you combine ggplot2 and base R graphics in one plot array? Answer: no. Base R graphics and ggplot2 graphics are completely different and cannot be mixed. In base R, plots are created by opening graphics devices, in ggplot2, plots are layered R objects.

DONE Create simple scatterplots

We're going to work with MASS::mammals using ggplot2 and base R.

1. Load the relevant packages.

```
library(MASS)
library(ggplot2)
```

2. Show the data structure of mammals.

```
str(mammals)
```

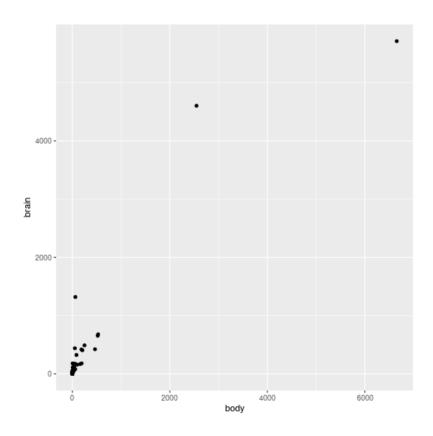
```
'data.frame': 62 obs. of 2 variables:

$ body : num 3.38 0.48 1.35 465 36.33 ...

$ brain: num 44.5 15.5 8.1 423 119.5 ...
```

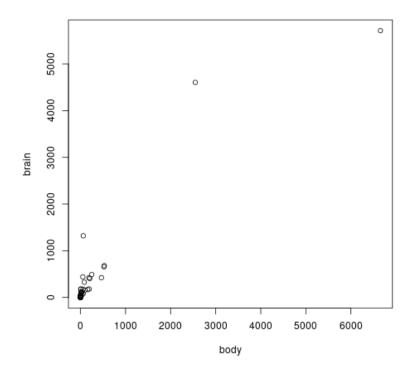
3. Create a scatterplot of brain vs body of the mammals data set in ggplot2.

```
ggplot(mammals, aes(x=body,y=brain)) +
  geom_point()
```



4. Create a scatterplot of brain vs body of the mammals data set in base ${\tt R}.$

plot(mammals)



DONE Transform plots

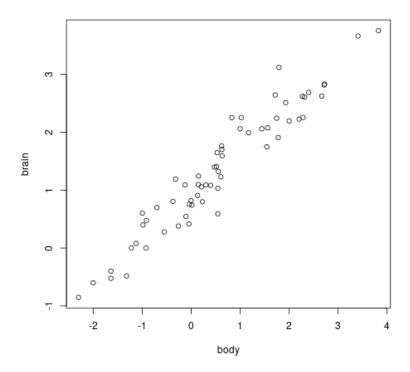
1. What's the problem with these plots and what could you do about it?

The problem: the data points are too bunched up because mammals have a wide spectrum of body and brain weights (there are very small and very large ones).

The solution: transform the x- and y-axis logarithmically.

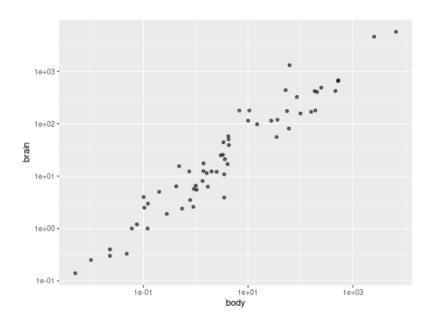
2. Implement the solution with plot.

plot(log10(mammals))



3. Implement the solution with ggplot - save the plot as gg for later, and print it.

```
gg <- ggplot(
  data=mammals,
  aes(x=body,y=brain)) +
  geom_point(alpha=0.6) +
  coord_fixed() +
  scale_x_log10() +
  scale_y_log10()
gg</pre>
```



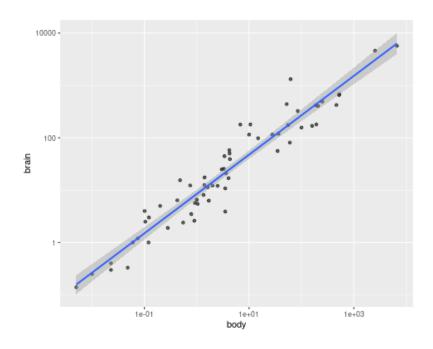
4. What does the geom_point argument alpha do?

Answer: it reduces the transparency of the points by 40%.

$\ensuremath{\mathsf{DONE}}$ Create trendlines with ggplot2 and base R

1. Create a linear trendline for the ggplot2 plot gg. Inside the smoothing geometry, use method="lm" to fix the model.

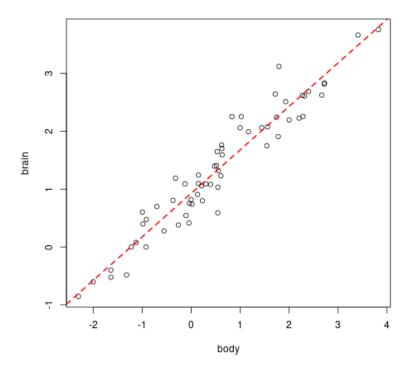
```
gg +
  geom_smooth(method="lm")
```



2. Create a linear model in base R using lm and data-log10(mammals). Save it as line and print it.

3. Create a trendline plot in base R using the linear model. The line should be red, dashed and double wide.

```
plot(log10(mammals))
abline(line, col="red", lty=2, lwd=2)
```

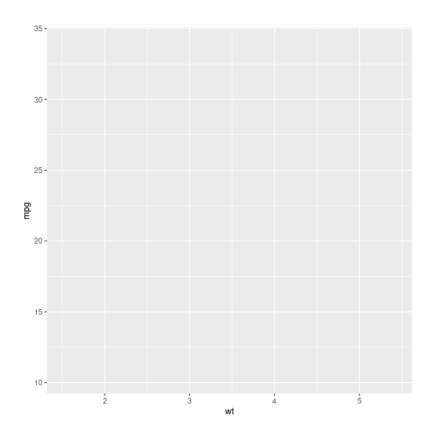


DONE Map 'aesthetics' to variables

Recall that the mtcars data frame lists the characteristics mileage (mpg), weight (wt) and number of cylinders (cyl) as numeric variables.

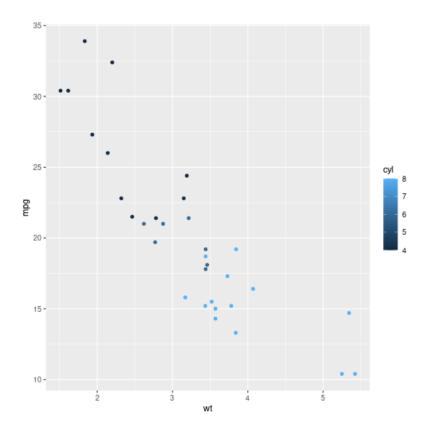
1. Create a ggplot of mileage vs. weight using ggplot2, save it as gg and print it.

```
gg <- ggplot(data=mtcars, aes(wt,mpg))
gg</pre>
```



2. Create a scatterplot where the color 'aesthetic' is mapped to the number of cylinders by adding a 'geometry' to gg.

```
ggplot(data=mtcars,
  aes(wt,mpg)) +
  geom_point(
   aes(color=cyl))
```



3. What's the difference between mapping 'aesthetics' inside the 'geometry' or inside the ggplot function?

Answer: the aes function knows about the dataset from data. You can also pipe the data set into the function using |> or %>%. Without aes, you need to specify geom_point(mtcars\$cyl). You can subset data for a specific geometry by putting aes into the function.