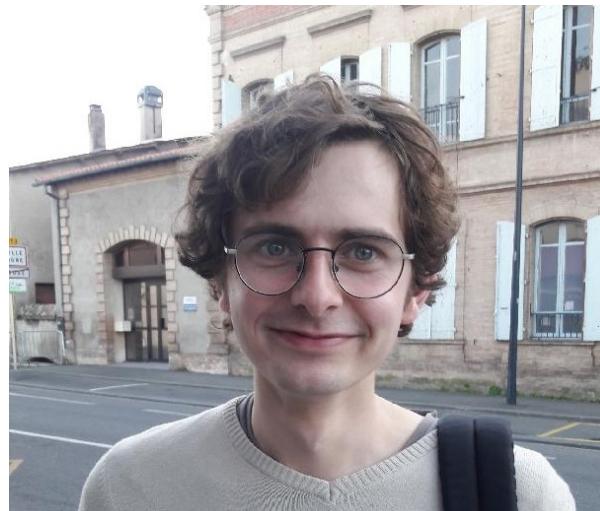


Data visualization on R

Rémi Mahmoud

Who is talking ?



INRAe

What you have to do:

Open Rstudio

What you have to do:

Open Rstudio

Stop me if anything is unclear

What we will talk about today

What we will talk about today

- Visualize data in R

What we will talk about today

- Visualize data in R
- Customize your plots

What we will talk about today

- Visualize data in R
- Customize your plots
- Avoid caveats in data visualization

What we will talk about today

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What we will NOT talk about today

What we will talk about today

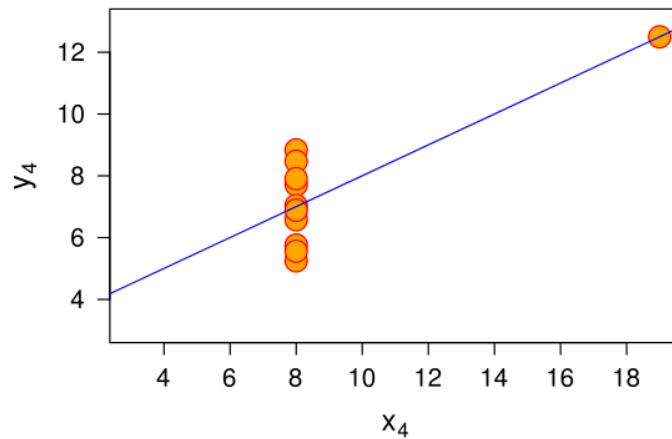
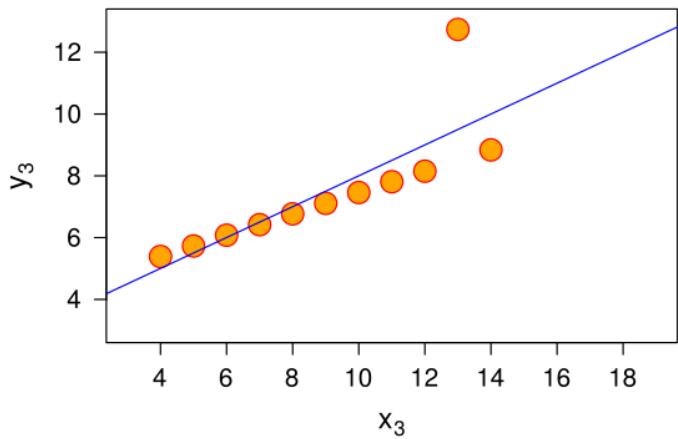
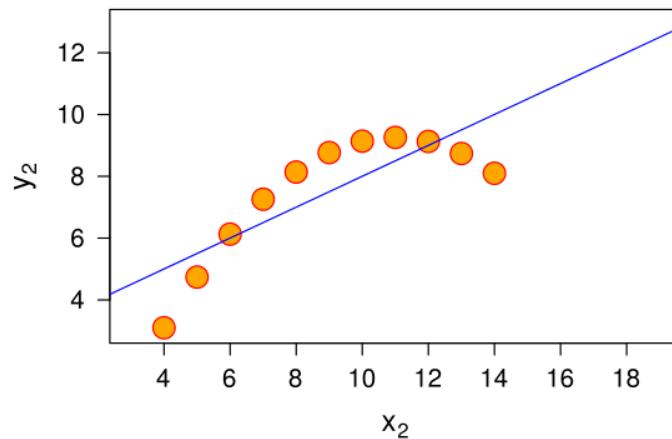
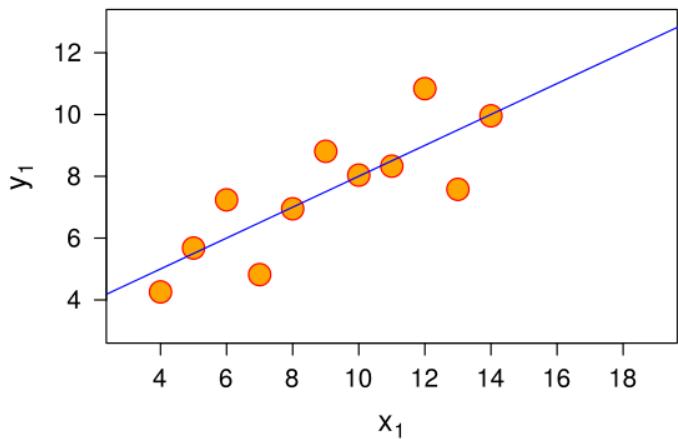
- Visualize data in R
- Customize your plots
- Avoid caveats in data visualization

What we will NOT talk about today

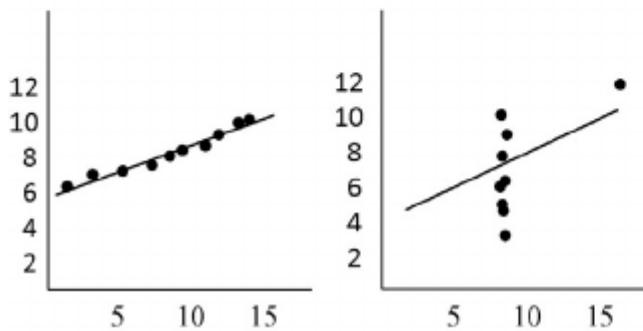
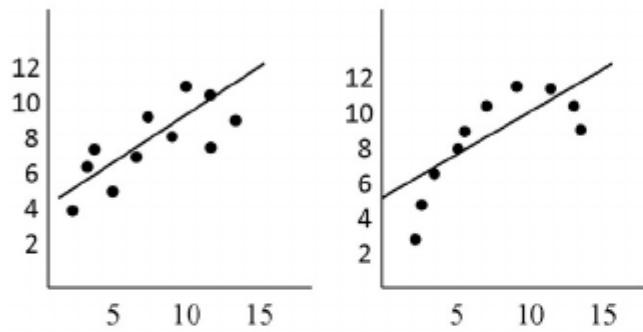
- Manipulate data etc. (next week :-)

Data-visualisation: WHY ?

What have these sets of points in common ?



Anscombe's Quartet



Property

<u>Property</u>	<u>Value</u>
Mean of X (average)	9 in all 4 XY plots
Sample variance of X	11 in all four XY plots
Mean of Y	7.50 in all 4 XY plots
Sample variance of Y	4.122 or 4.127 in all 4 XY plots
Correlation (r)	0.816 in all 4 XY plots
Linear regression	$y = 3.00 + (0.500 x)$ in all 4 XY plots

Data sets for the 4 XY plots

I II III IV

x	y	x	y	x	y	x	y
10.0	8.04	10.0	9.14	10.0	7.46	8.0	6.58
8.0	6.95	8.0	8.14	8.0	6.77	8.0	5.76
13.0	7.58	13.0	8.74	13.0	12.74	8.0	5.76
9.0	8.81	9.0	8.77	9.0	7.11	8.0	8.84
11.0	8.33	11.0	9.26	11.0	8.81	8.0	8.47
14.0	9.96	14.0	8.10	14.0	8.84	8.0	7.04
6.0	7.24	6.0	6.13	6.0	6.08	8.0	5.25
4.0	4.26	4.0	3.10	4.0	5.39	19.0	12.50
12.0	10.84	12.0	7.26	12.0	8.15	8.0	5.56
7.0	4.82	7.0	7.26	7.0	6.42	8.0	7.91
5.0	5.68	5.0	4.74	5.0	5.73	8.0	6.89

DATA-VISUALIZATION IS AN ESSENTIAL STEP

**DATA-VISUALIZATION IS AN ESSENTIAL STEP
OUR BRAINS ARE NOT MADE TO SEE PATTERNS IN TABLES**

Framework

All manipulations will be done in the `tidyverse` framework.

Framework

All manipulations will be done in the `tidyverse` framework.

Hence, you should, if not already done, run the following command in R **NOW**

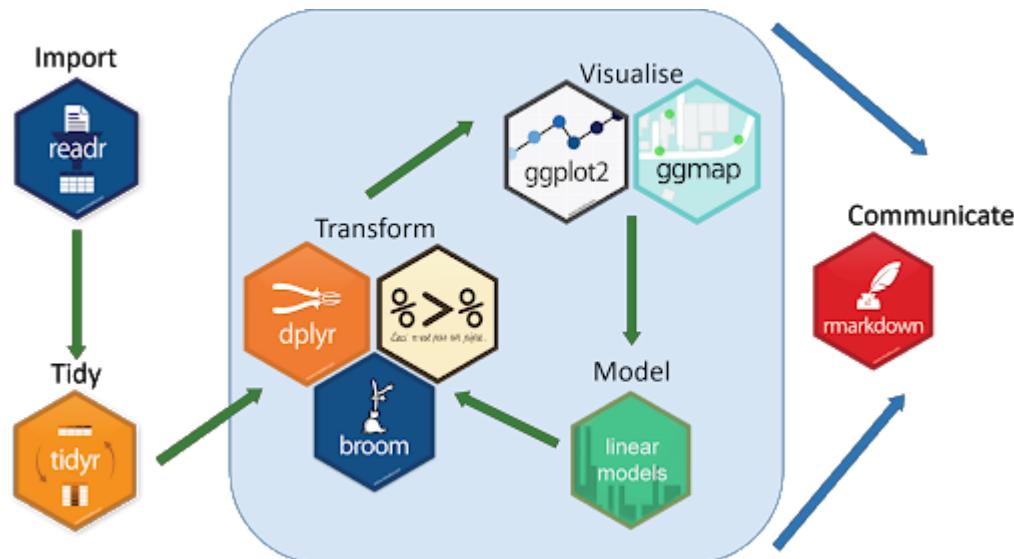
```
install.packages("tidyverse")
```

Tidyverse = Tidy universe

Tidyverse is a set of packages with different purposes, that share the same syntax and that are designed to work in a complementary way

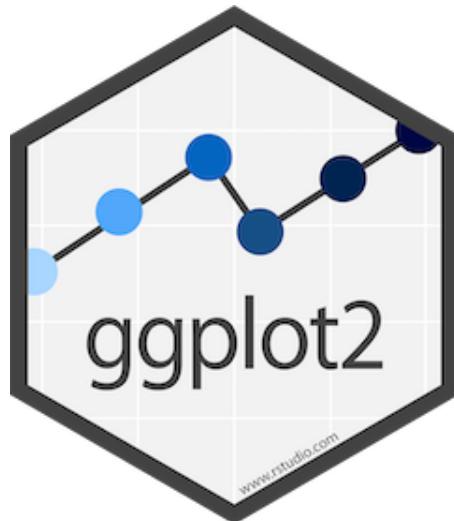
Tidyverse = Tidy universe

Tidyverse is a set of packages with different purposes, that share the same syntax and that are designed to work in a complementary way

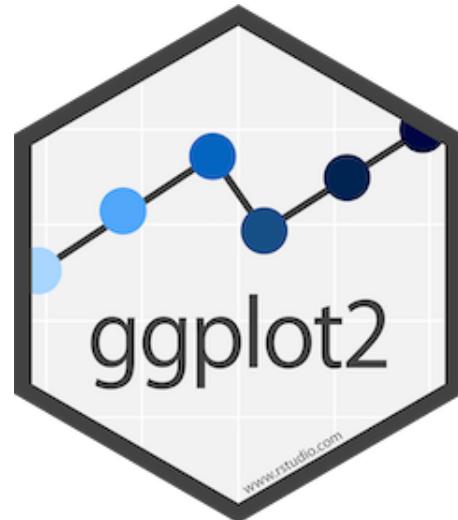


Today, we'll focus on the package `ggplot2`.

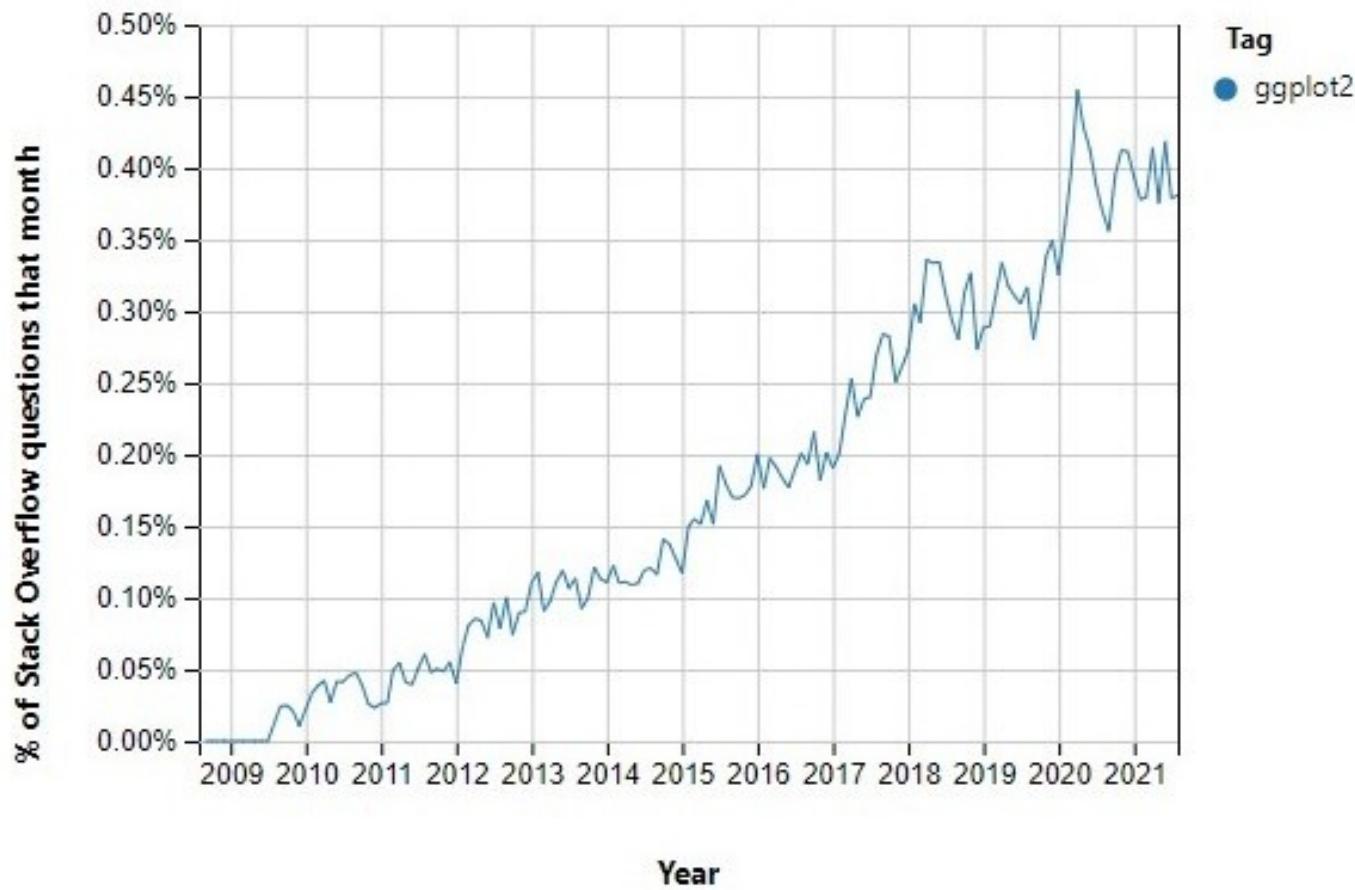
Today, we'll focus on the package `ggplot2`.



Today, we'll focus on the package `ggplot2`.



ggplot2, WHY ?



Base R plots vs ggplot2: the clash

Base graphics:

- directly available from R
- easy to use for beginners
- is less verbose for simple / canned graphics
- has methods (plot works for many different objects)

ggplot2:

- R graphing package by Hadley Wickham based on the Grammar of Graphics (Wilkinson, 2005)
- is less verbose for complex / custom graphics: plots can be iteratively built up and edited later
- carefully chosen defaults = publication-quality graphics in seconds

Harder ? Not really

GGPLOT2: BASICS



imgflip.com

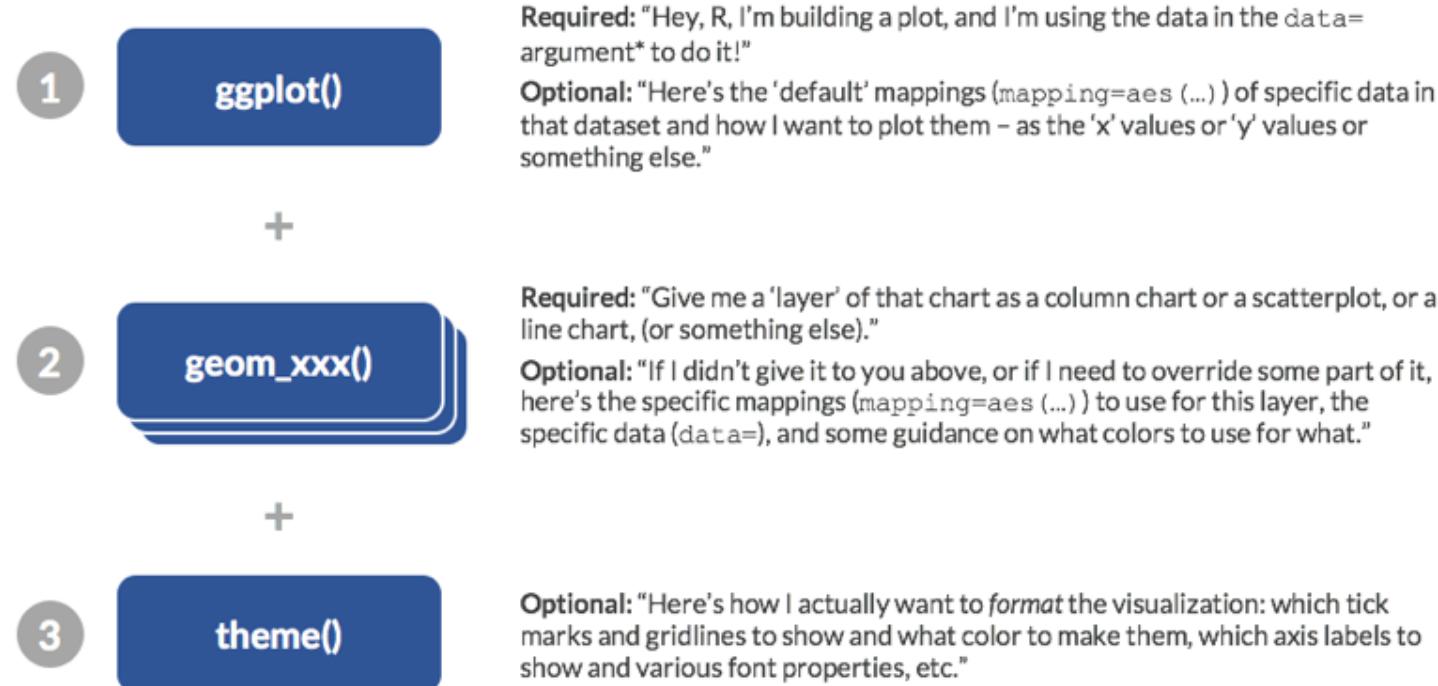
ggplot2 building blocks

- **Data:** a `data.frame` and nothing else!
- **Aesthetic mapping:** describe how data are mapped to things we can see on the plot through the function `aes()`.
- **Geometric object:** perform the actual rendering of the plot and control the type of plot to create (`points`, `line`, `histogram`, `boxplot`...).
- **Statistical transformations:** transform the data for instance by summarising it in some manner (`sum`, `density`, `smooth`...)
- **Position adjustments:** apply minor changes to the position of elements (`jitter`, `fill`, `stack`, `dodge`, `identity`)

In a nutshell

The Basics of ggplot2

There are three main components that get *added* together to build up a plot.



* Not always the case, and not strictly required, but ggplot2 works best with data in a long format (or, at least, a tidy format). `gather()` from the `tidyverse` package is your friend here.

Syntax

```
library(ggplot2)
ggplot(data = <DATA>,
       aes(x = <X AXIS VARIABLE>,
           y = <Y AXIS VARIABLE>, ... ), ...) +
  geom_<TYPE>(aes(size = <SIZE VARIABLE>, ...),
               data = <DATA>,
               stat = <FUNCTION>,
               position = <POSITION>,
               color = <"COLOR">, ...) +
  scale_<AESTHETIC>_<TYPE>(name = <NAME>,
                             breaks = <WHERE>,
                             labels = <LABELS>, ... ) +
  theme(...) +
  facet_<TYPE>(<FORMULA>)
```

What happens when we create a simple ggplot() object ?

```
library(ggplot2)  
ggplot()
```

Almost nothing

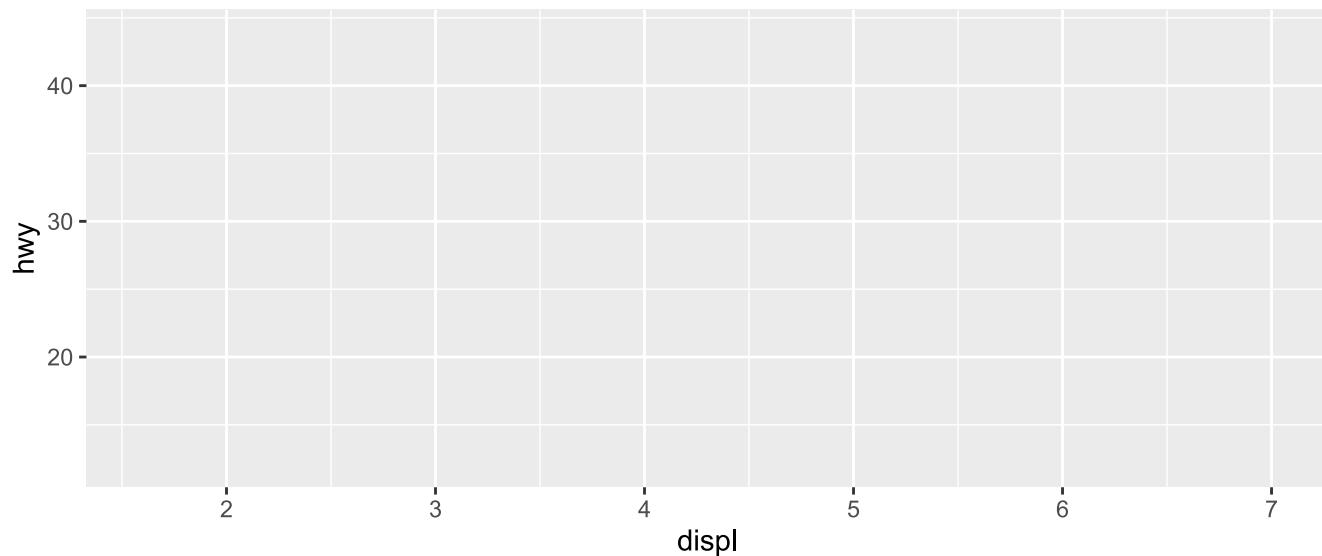
Specify data

```
data("mpg")
head(mpg)
```

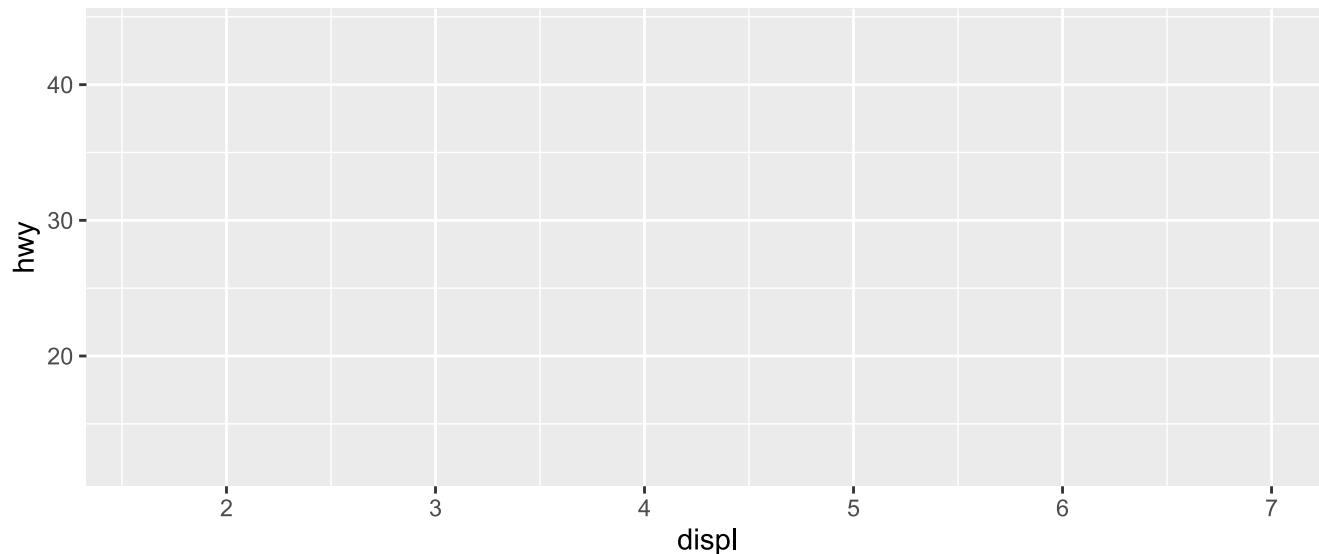
```
## # A tibble: 6 x 11
##   manufacturer model displ year cyl trans     drv     cty     hwy
##   <chr>        <chr>  <dbl> <int> <int> <chr>     <chr> <int> <int>
## 1 audi         a4      1.8  1999     4 auto(l5)   f       18     29
## 2 audi         a4      1.8  1999     4 manual(m5) f       21     29
## 3 audi         a4      2.0  2008     4 manual(m6) f       20     31
## 4 audi         a4      2.0  2008     4 auto(av)   f       21     30
## 5 audi         a4      2.8  1999     6 auto(l5)   f       16     26
## 6 audi         a4      2.8  1999     6 manual(m5) f       18     26
```

```
# See ?mpg for the description of the dataset
```

```
p <- ggplot(data = mpg, aes(x =displ,    y= hwy  ))  
p
```



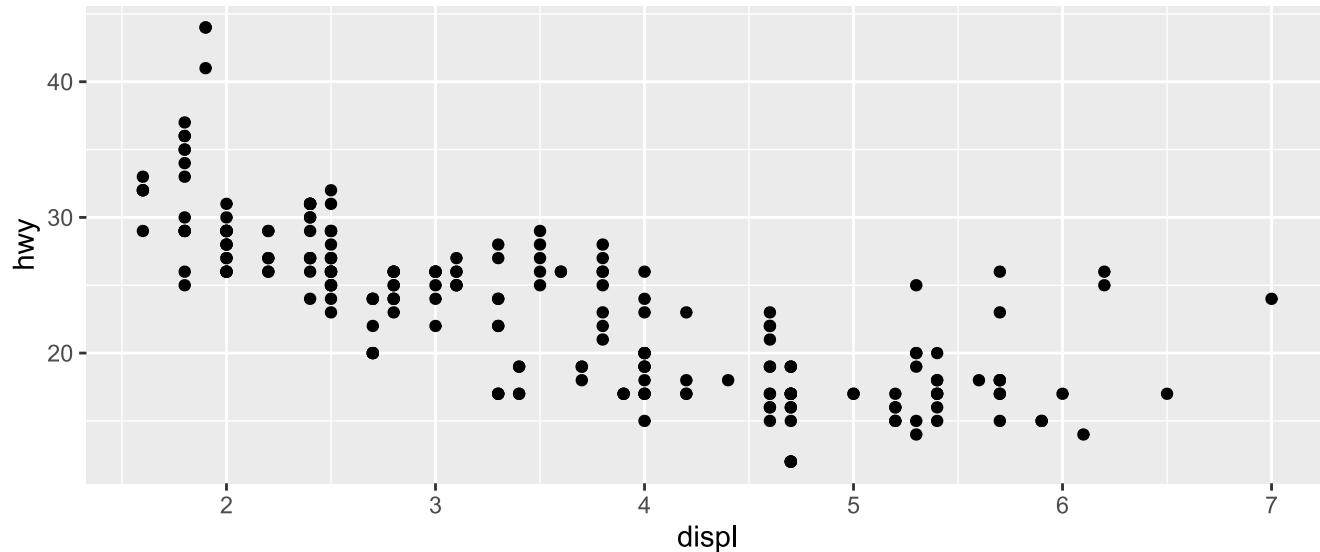
```
p <- ggplot(data = mpg, aes(x =displ,    y= hwy ))  
p
```



Almost nothing? Look at the axes :-)

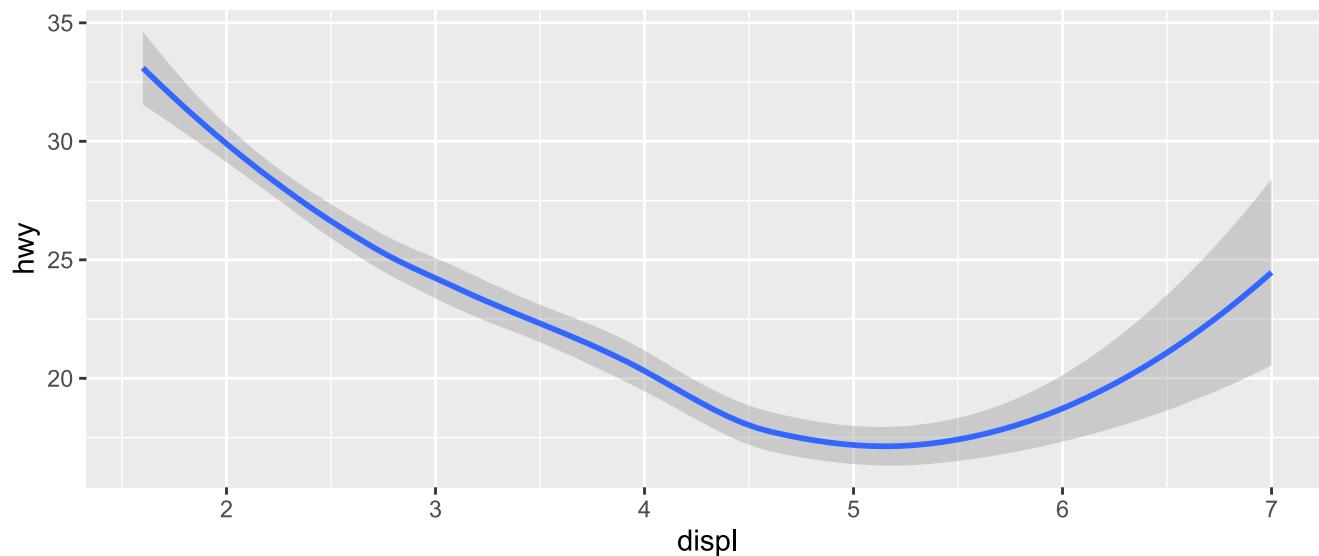
Let's specify what sort of plot we want, using a `geom_*` functions

```
p <- ggplot(data = mpg, aes(x =displ,    y= hwy ))  
p + geom_point()
```



Another geom_*

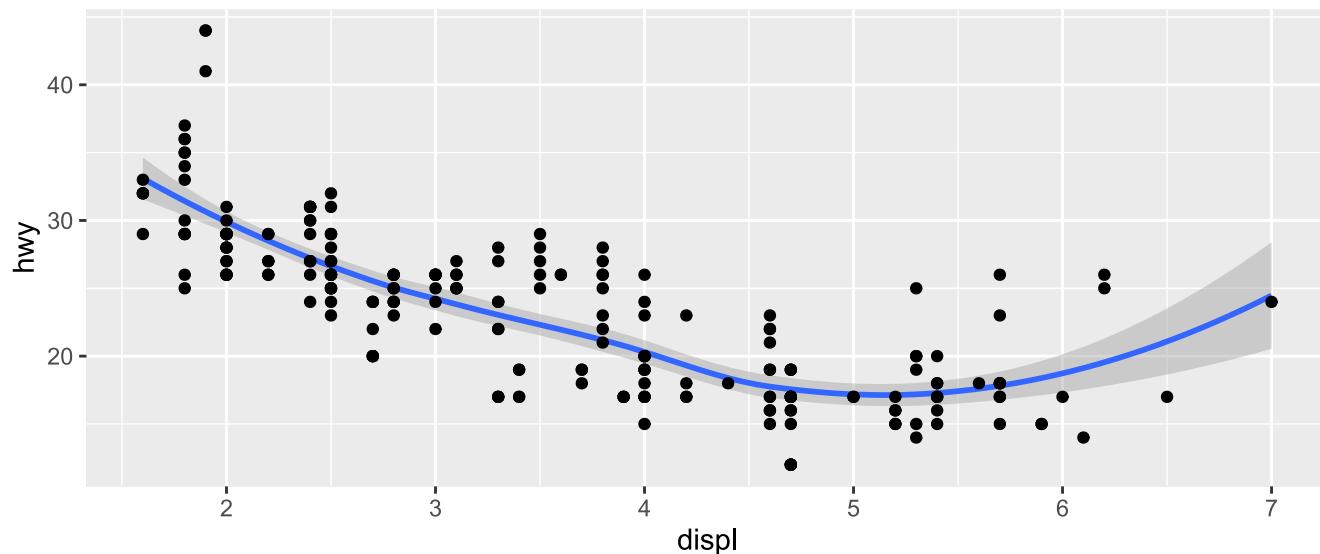
```
p <- ggplot(data = mpg, aes(x =displ,   y= hwy ))  
p + geom_smooth()
```



Combine two geom?

Simply **add** as many geom_functions as you want

```
p <- ggplot(data = mpg, aes(x =displ, y= hwy ))  
p +  
  geom_smooth() +  
  geom_point()
```



Combine three geom?

- Suppose we want to also look a linear thrend
- Need for help

```
?geom_smooth
```

Arguments

mapping Set of aesthetic mappings created by `aes()` or `aes_()`. If specified and `inherit.aes = TRUE` (the default), it is combined with the default mapping at the top level of the plot. You must supply `mapping` if there is no plot mapping.

data The data to be displayed in this layer. There are three options:

If `NULL`, the default, the data is inherited from the plot data as specified in the call to `ggplot()`.

A `data.frame`, or other object, will override the plot data. All objects will be fortified to produce a data frame. See `fortify()` for which variables will be created.

A `function` will be called with a single argument, the plot data. The return value must be a `data.frame`, and will be used as the layer data. A `function` can be created from a `formula` (e.g. `~ head(.x, 10)`).

position Position adjustment, either as a string, or the result of a call to a position adjustment function.

... Other arguments passed on to `layer()`. These are often aesthetics, used to set an aesthetic to a fixed value, like `colour = "red"` or `size = 3`. They may also be parameters to the paired geom/stat.

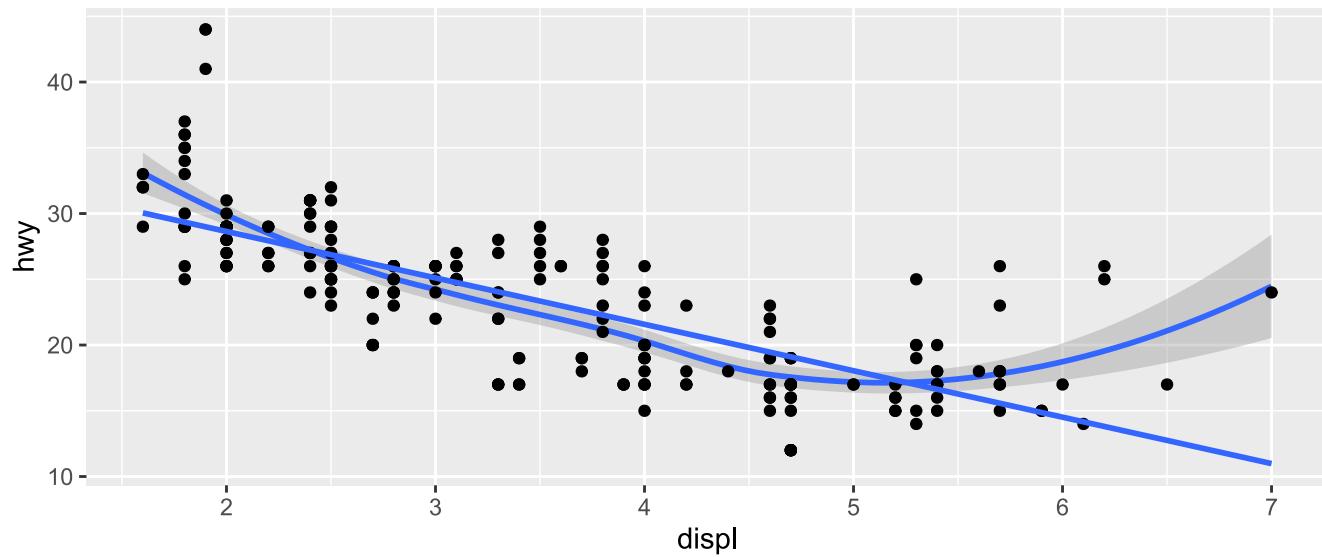
method Smoothing method (function) to use, accepts either `NULL` or a character vector, e.g. `"lm"`, `"glm"`, `"gam"`, `"loess"` or a function, e.g. `MASS::rlm` or `mgcv::gam`, `stats::lm`, or `stats::loess`. `"auto"` is also accepted for backwards compatibility. It is equivalent to `NULL`.

For `method = NULL` the smoothing method is chosen based on the size of the largest group (across all panels). `stats::loess()` is used for less than 1,000 observations; otherwise `mgcv::gam()` is used with `formula = y ~ s(x, bs = "cs")` with `method = "REML"`. Somewhat anecdotally, `loess` gives a better appearance, but is $O(N^2)$ in memory, so does not work for larger datasets.

If you have fewer than 1,000 observations but want to use the same `gam()` model that `method = NULL` would use, then set `method = "gam"`, `formula = y ~ s(x, bs = "cs")`.

formula Formula to use in smoothing function, eg. `y ~ x`, `y ~ poly(x, 2)`, `y ~ log(x)`. `NULL` by default, in which case `method = NULL` implies `formula = y ~ x` when there are fewer than

```
p <- ggplot(data = mpg, aes(x = displ, y= hwy ))  
p +  
  geom_smooth() +  
  geom_point() +  
  geom_smooth(method = "lm", se = FALSE)
```



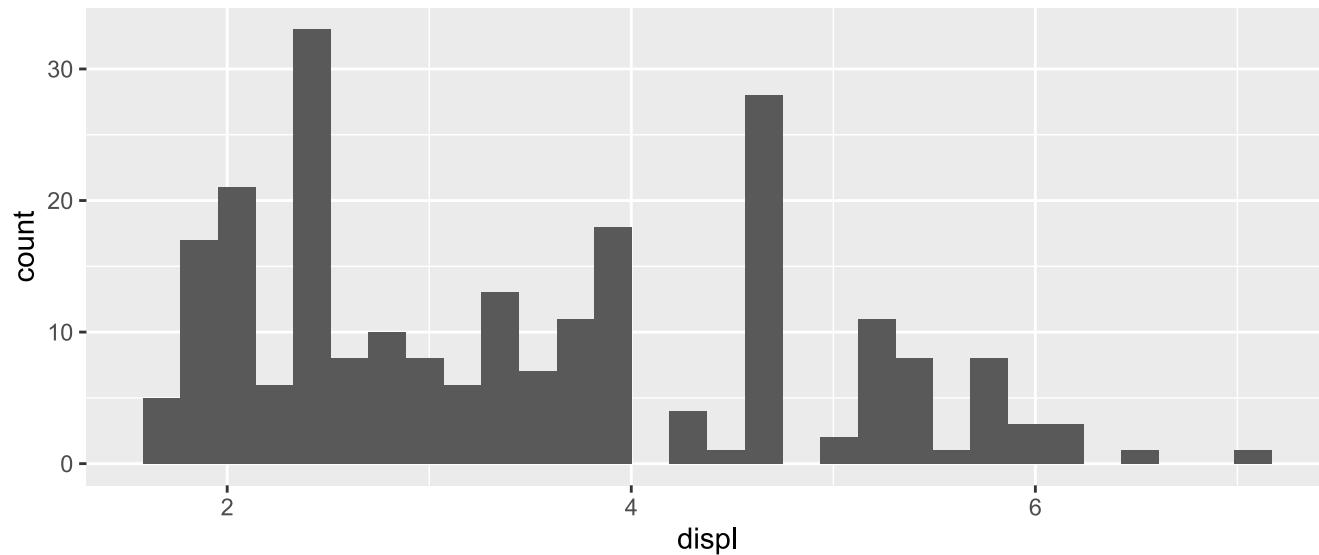
There are *many* geom_ functions:

```
ls("package:ggplot2")[grep(ls("package:ggplot2"), pattern = "geom")]
```

```
## [1] "geom_abline"  
## [4] "geom_bin2d"  
## [7] "geom_col"  
## [10] "geom_count"  
## [13] "geom_density"  
## [16] "geom_density2d"  
## [19] "geom_errorbar"  
## [22] "geom_function"  
## [25] "geom_hline"  
## [28] "geom_line"  
## [31] "geom_path"  
## [34] "geom_polygon"  
## [37] "geom_quantile"  
## [40] "geom_ribbon"  
## [43] "geom_sf"  
## [46] "geom_smooth"  
## [49] "geom_text"  
## [52] "geom_vline"  
  
"geom_area"  
"geom_blank"  
"geom_contour"  
"geom_crossbar"  
"geom_density_2d"  
"geom_density2d_filled"  
"geom_errorbarh"  
"geom_hex"  
"geom_jitter"  
"geom_linerange"  
"geom_point"  
"geom_qq"  
"geom_raster"  
"geom_rug"  
"geom_sf_label"  
"geom_spoke"  
"geom_tile"  
"guide_geom"  
  
"geom_bar"  
"geom_boxplot"  
"geom_contour"  
"geom_curve"  
"geom_dens"  
"geom_dotplot"  
"geom_freq"  
"geom_hist"  
"geom_label"  
"geom_map"  
"geom_point"  
"geom_qq_li"  
"geom_rect"  
"geom_segm"  
"geom_sf_te"  
"geom_step"  
"geom_styl"  
"geom_violin"  
"update_geor
```

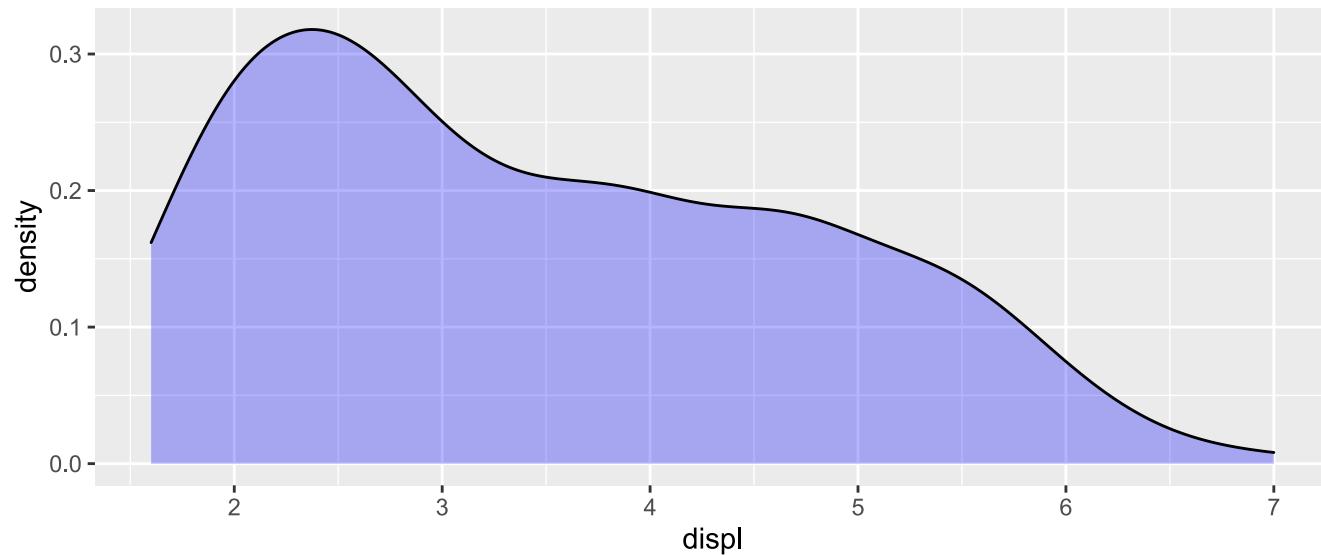
Quick look around of some classical geom_* functions: geom_histogram

```
ggplot(data = mpg, aes(x = displ)) +  
  geom_histogram()
```



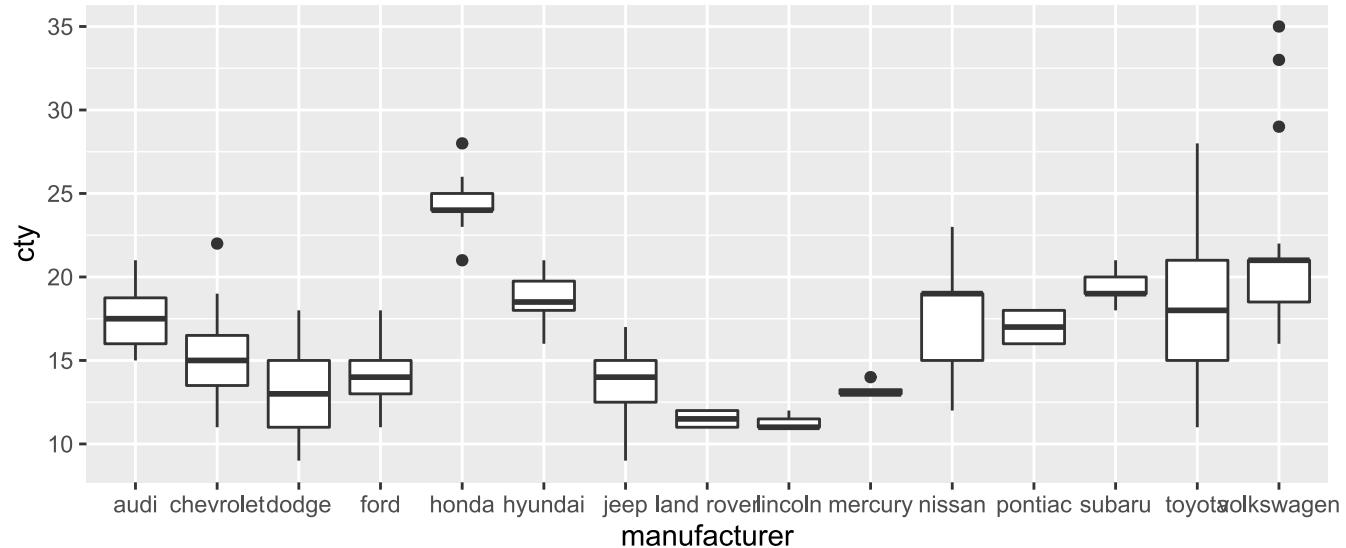
Quick look around of some classical geom_* functions: geom_density

```
ggplot(data = mpg, aes(x = displ)) +  
  geom_density(fill = "blue", alpha = .3) # alpha: transparency
```



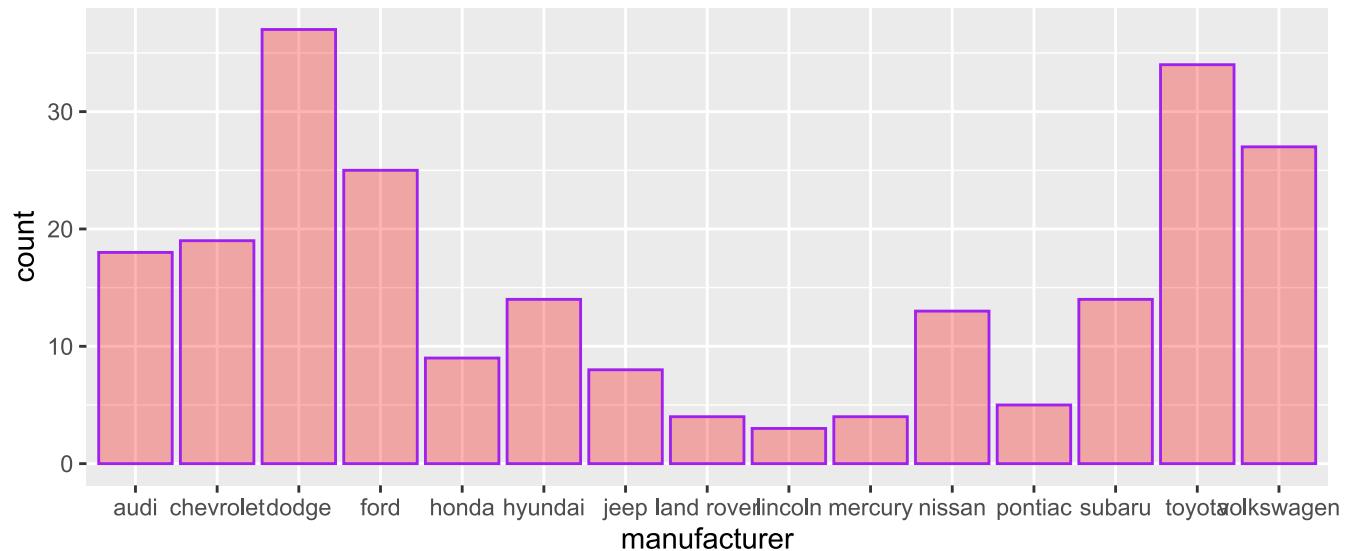
Quick look around of some classical geom_* functions: geom_boxplot

```
ggplot(data = mpg, aes(x = manufacturer, y = cty)) +  
  geom_boxplot()
```



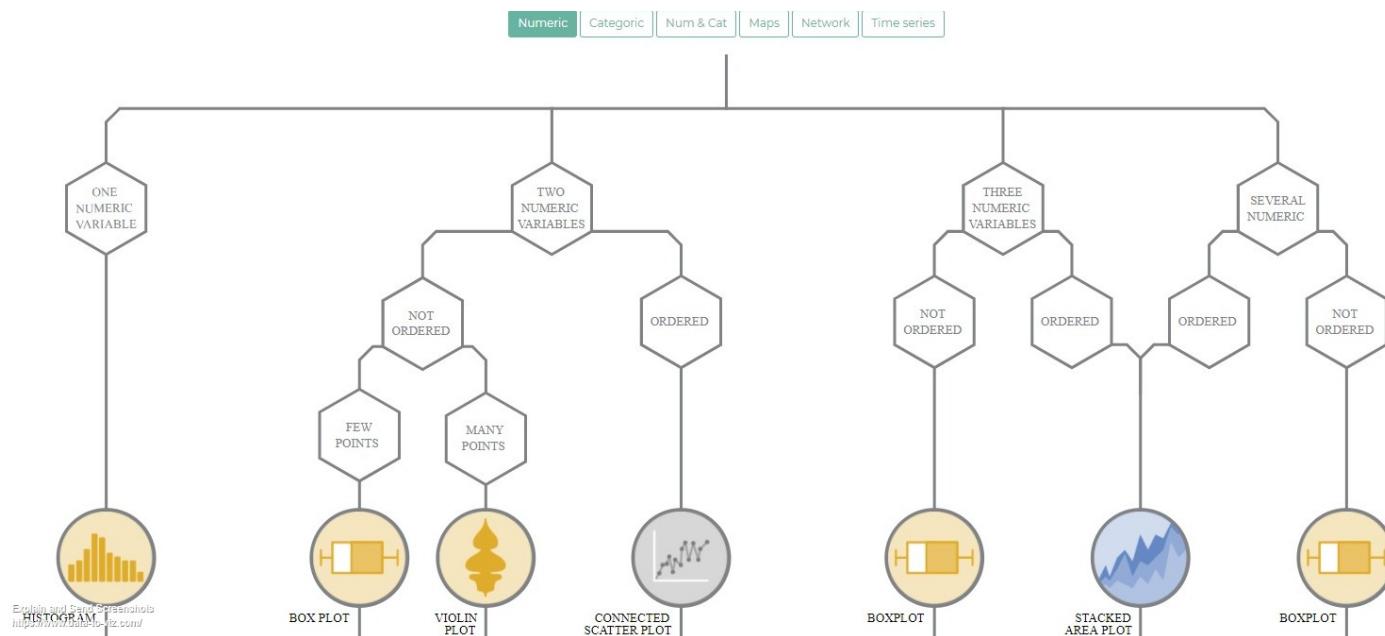
Quick look around of some classical geom_* functions: geom_bar

```
ggplot(data = mpg, aes(x = manufacturer)) +  
  geom_bar(alpha = .3, col = "purple", fill = "red") # col = borders
```



Take a look at this site to choose a plot

<https://www.data-to-viz.com/>



Is everything clear ?



Exercises

~ 20 minutes

Exercises

~ 20 minutes

Open "ggplot2_basics" with Rstudio

GGPLOT2: Aesthetics / Facets / Labs



**GGPLOT(DATA,
AESO)
+ GEOM_POINTO**



**GGPLOT(DATA,
AESO) +
GEOM_POINTO + LABSO
+ FACET_WRAPO**



imgflip.com

- One may want to :

- One may want to :

add color/transparency to the points

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add color/transparency to the points

do one plot for each manufacturer (or modality of a given factor)

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do one plot for each manufacturer (or modality of a given factor)

specify a title/subtitle to the plot

Aesthetics, basics

Aesthetic mapping vs. parameter setting

Aesthetic mapping:

- Data value determines visual characteristic
- use `aes()`

```
ggplot(diamonds, aes(x = carat, y = price, color = clarity)) + geom_p
```

Setting:

- Constant value determines visual characteristic
- Use parameter in `geom_<TYPE>`

```
ggplot(diamonds, aes(x = carat, y = price)) + geom_point(color = "red")
```

- Let's take another dataset

```
data(diamonds)
data_work <- diamonds[sample(1:nrow(diamonds), 5000),] # take 10000 rows
head(data_work)

## # A tibble: 6 x 10
##   carat    cut      color clarity depth table price     x     y
##   <dbl> <ord>    <ord> <ord>   <dbl> <dbl> <int> <dbl> <dbl> <dbl>
## 1 0.3   Very Good I       SI1     63.3   59    506  4.3   4.23  2.7
## 2 1     Ideal      E       SI2     61.8   57    4743  6.39  6.43  3.90
## 3 0.73  Fair       I       VS1     55.9   66    2330  6.11  6.01  3.39
## 4 0.38  Ideal      E       VS1     60.7   56    933   4.7   4.72  2.80
## 5 0.58  Ideal      F       VVS2    61.8   56    2263  5.37  5.41  3.31
## 6 1     Ideal      F       VVS2    61.9   57    9354  6.47  6.42  3.99

# See ?diamonds for the description of the dataset
```

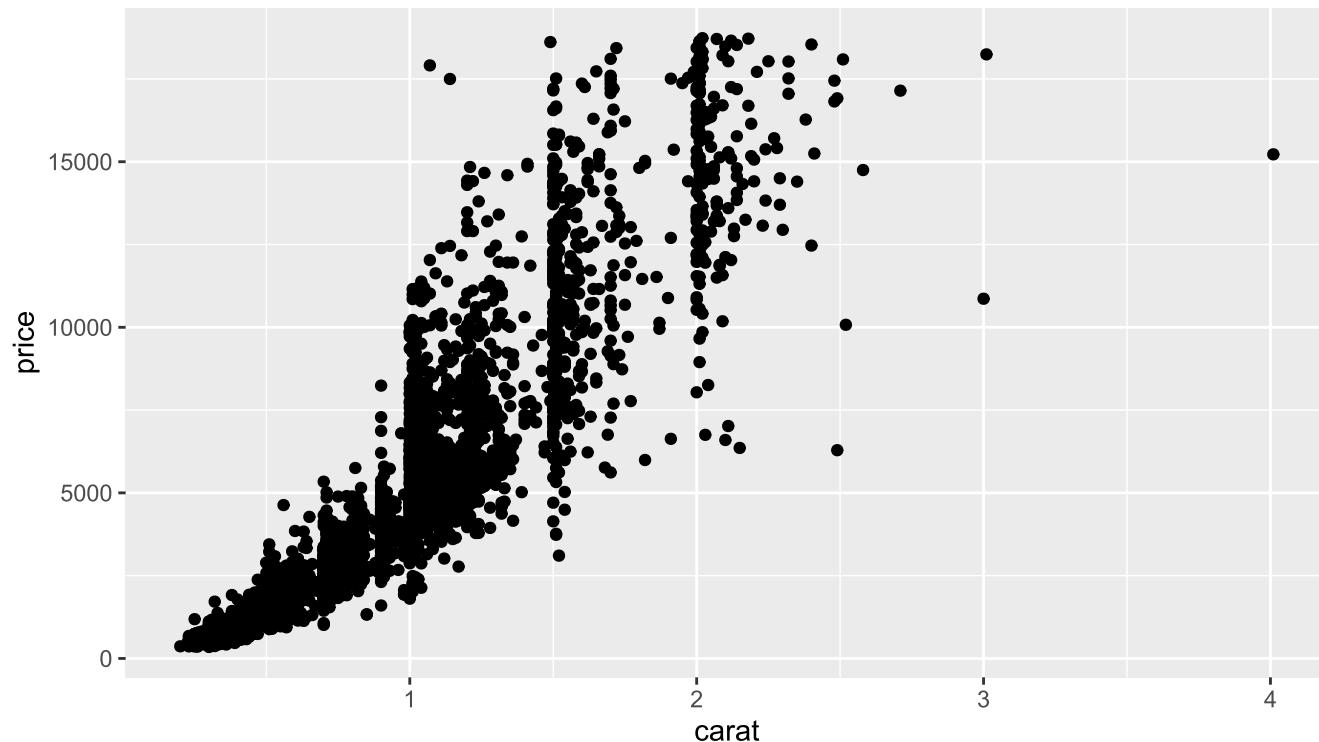
```
str(data_work)
```

```
## # tibble [5,000 x 10] (S3: tbl_df/tbl/data.frame)
## $ carat   : num [1:5000] 0.3 1 0.73 0.38 0.58 1 0.35 1 0.43 0.61 .
## $ cut      : Ord.factor w/ 5 levels "Fair"<"Good"<...: 3 5 1 5 5 5 3
## $ color    : Ord.factor w/ 7 levels "D"<"E"<"F"<"G"<...: 6 2 6 2 3 3
## $ clarity  : Ord.factor w/ 8 levels "I1"<"SI2"<"SI1"<...: 3 2 5 5 6 1
## $ depth    : num [1:5000] 63.3 61.8 55.9 60.7 61.8 61.9 60.7 62.4 61.5 59.8 ...
## $ table    : num [1:5000] 59 57 66 56 56 57 61 58 58 57 ...
## $ price    : int [1:5000] 506 4743 2330 933 2263 9354 706 4574 1113 1000 ...
## $ x        : num [1:5000] 4.3 6.39 6.11 4.7 5.37 6.47 4.54 6.37 4.81 5.0 ...
## $ y        : num [1:5000] 4.23 6.43 6.01 4.72 5.41 6.42 4.58 6.42 4.0 4.2 ...
## $ z        : num [1:5000] 2.7 3.96 3.39 2.86 3.33 3.99 2.77 3.99 2.1 2.3 ...
```

- Price vs carat ?

- Price vs carat ?

```
p_diamonds <- ggplot(data_work, aes(x = carat, y = price)) +  
  geom_point()  
p_diamonds
```

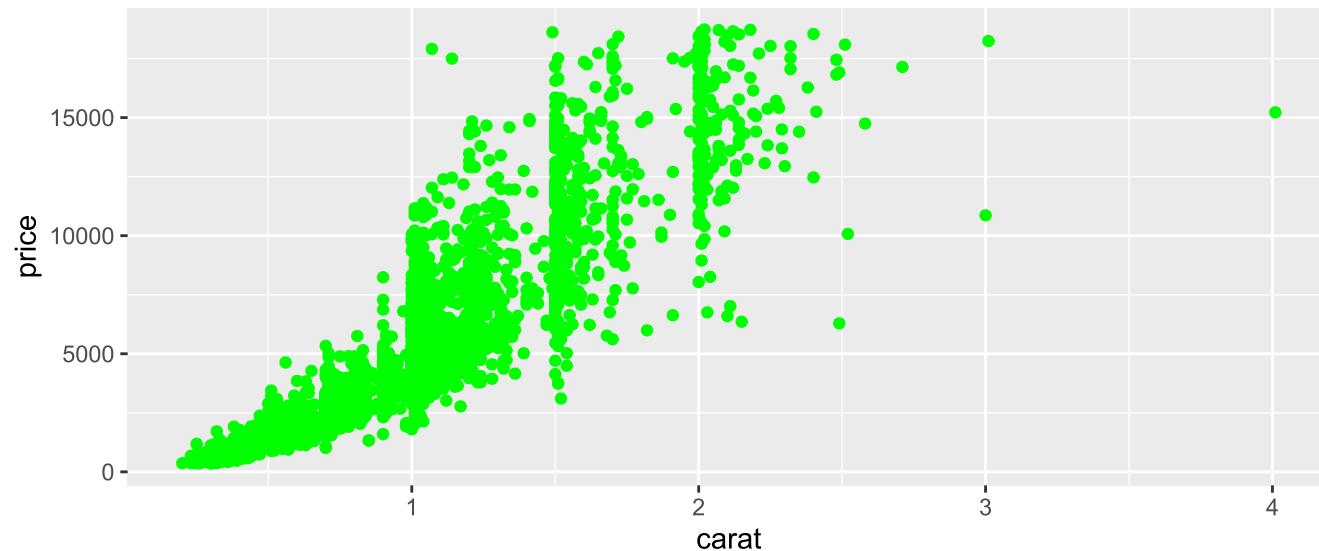


- We want the points to be green

- We want the points to be green
- This does not depend on the data → We specify the color argument *outside aes*

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- This does not depend on the data → We specify the color argument *outside aes*

```
p_diamonds <- ggplot(data_work, aes(x = carat, y = price)) +  
  geom_point(col = "green")  
p_diamonds
```

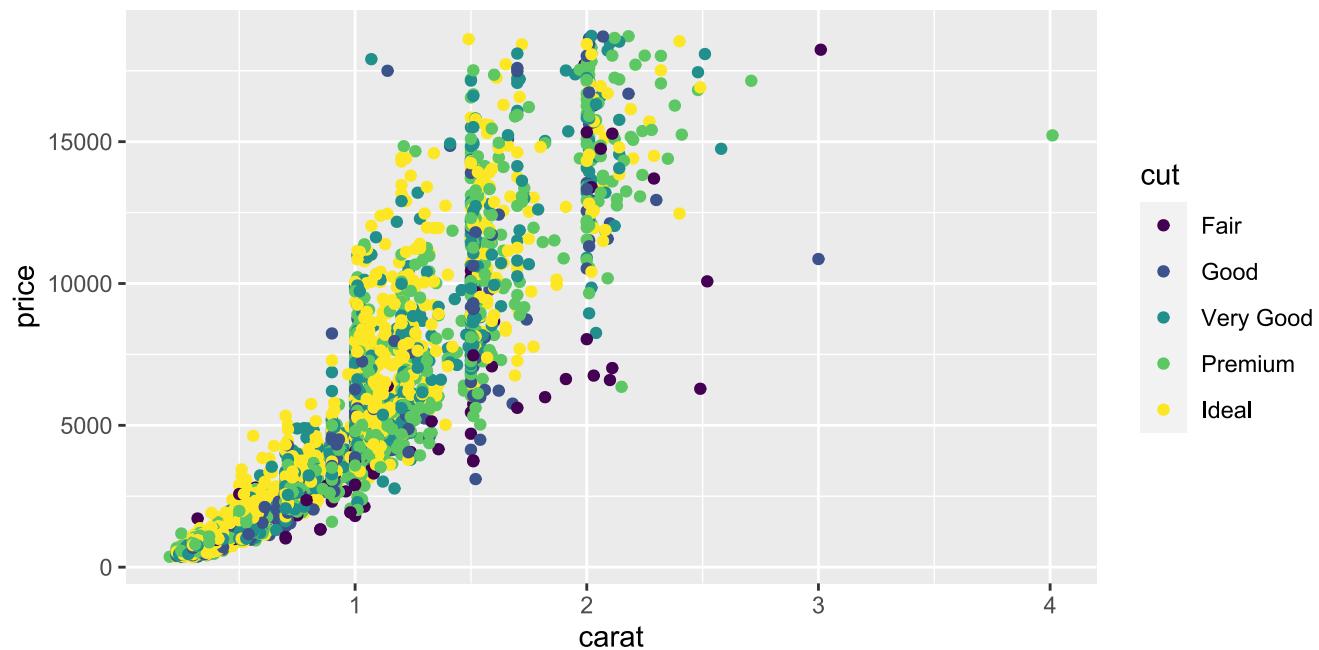


- One color for each cut of the diamond

- One color for each cut of the diamond
- This *depends* on the data → We specify the color argument *inside aes*

- One color for each cut of the diamond
- This *depends* on the data → We specify the color argument *inside* aes

```
p_diamonds <- ggplot(data_work, aes(x = carat, y = price)) +  
  geom_point(aes(col = cut)) #Says to ggplot2, give one color to each  
p_diamonds
```

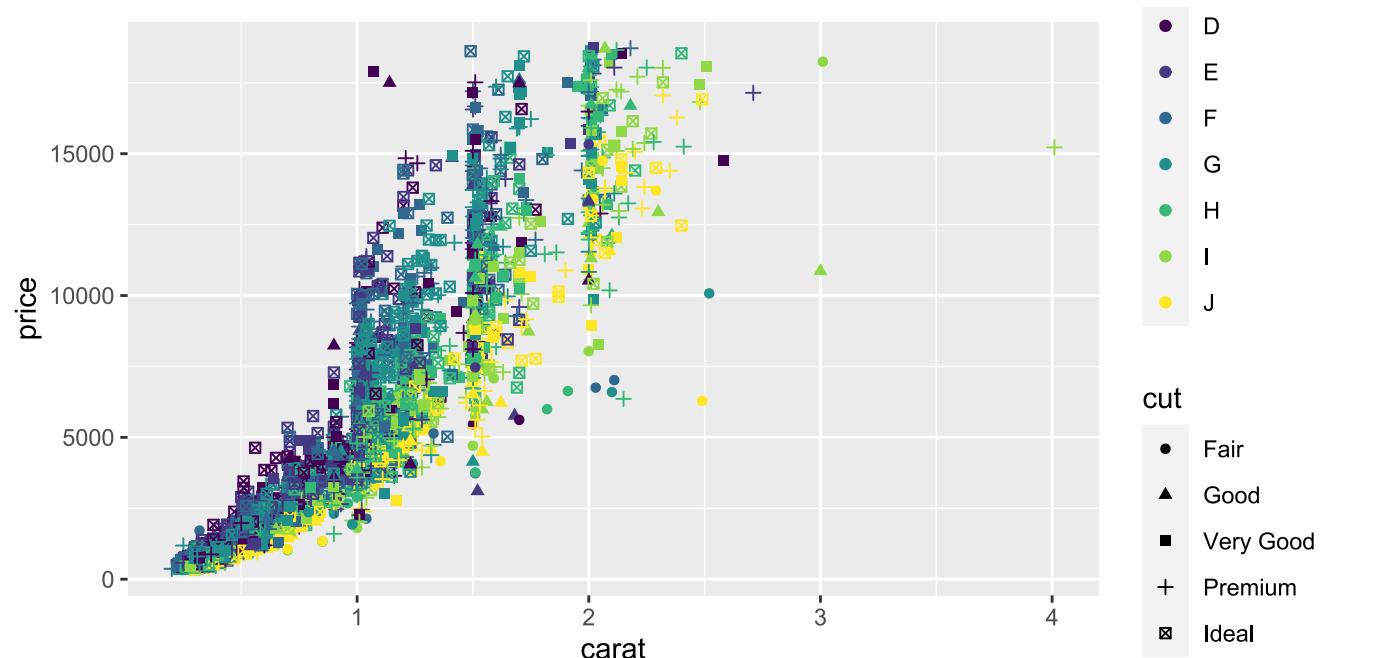


- One color for each for each color of the diamond, one shape for each cut of the diamond

- One color for each for each color of the diamond, one shape for each cut of the diamond
- This *depends* on the data → We specify the color and shape arguments *inside aes*

- One color for each for each color of the diamond, one shape for each cut of the diamond
- This *depends* on the data → We specify the color and shape arguments *inside aes*

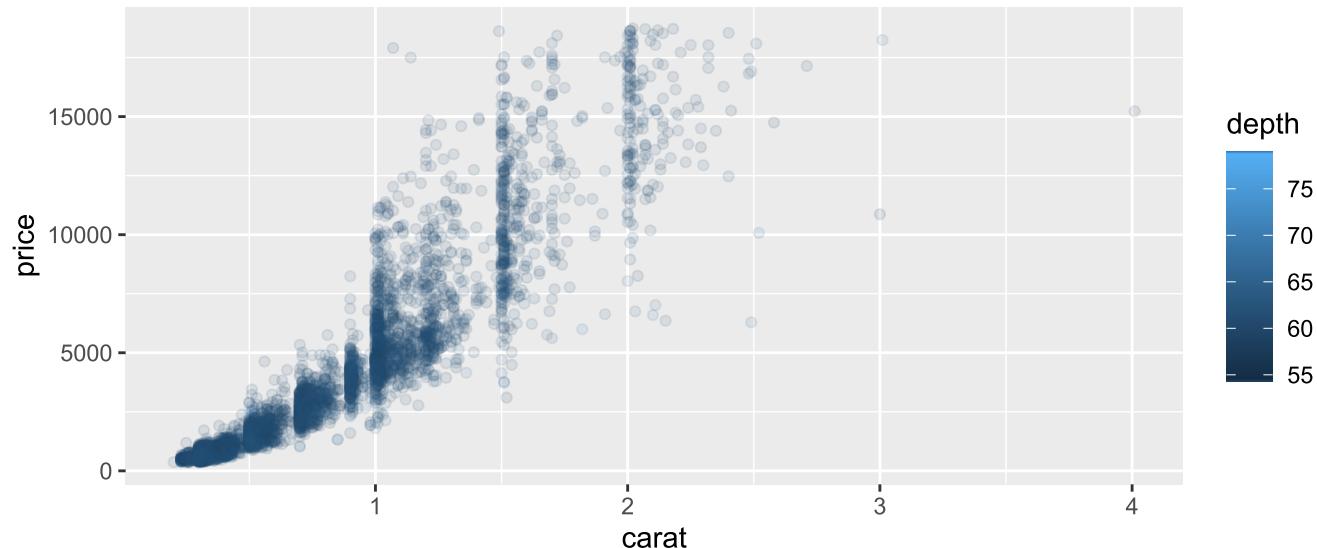
```
p_diamonds <- ggplot(data_work, aes(x = carat, y = price)) +
  geom_point(aes(col = color, shape = cut)) #Says to ggplot2, give one point per row
p_diamonds
```



- If the variable mapped to color is continuous, then the scale of the colour becomes automatically continuous

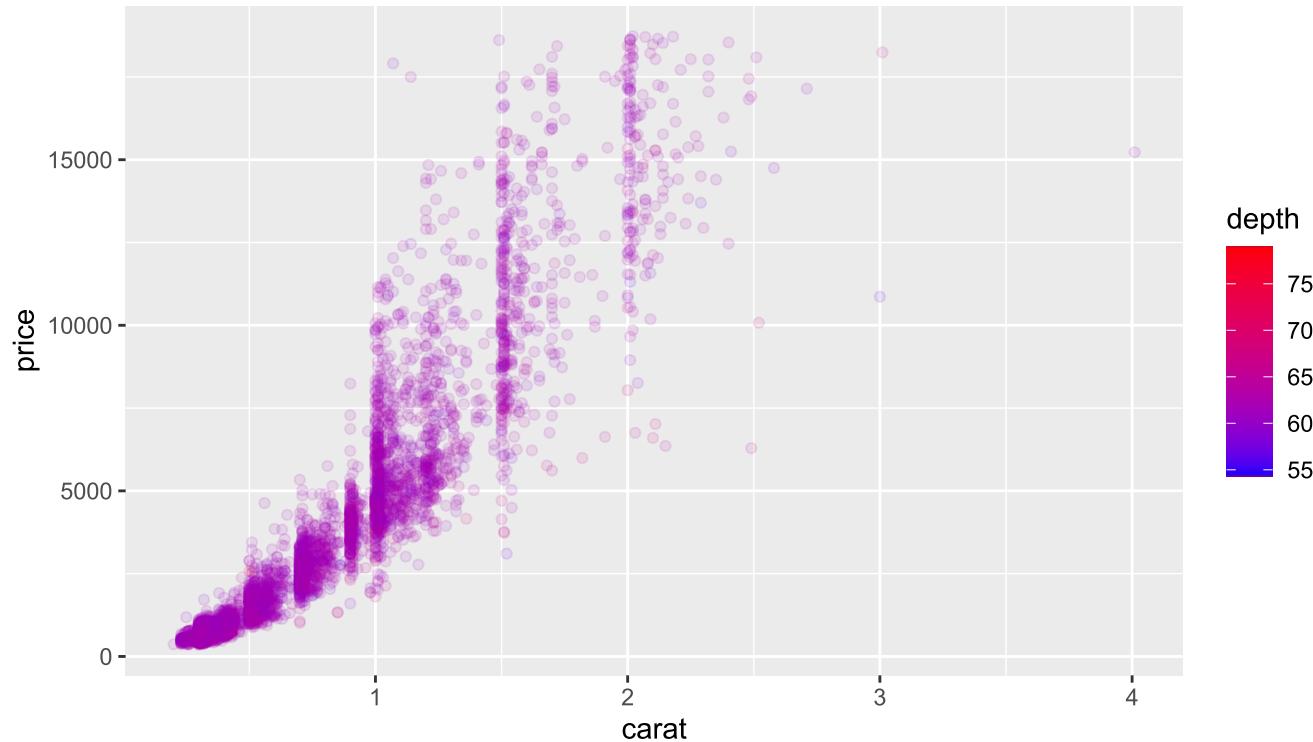
- If the variable mapped to color is continuous, then the scale of the colour becomes automatically continuous

```
p_diamonds <- ggplot(data_work, aes(x = carat, y = price)) +  
  geom_point(aes(col = depth),  
             alpha = .1) #Says to ggplot2, give one color based on th  
#Put transparency on 0.1, regardless of the points' values  
p_diamonds
```



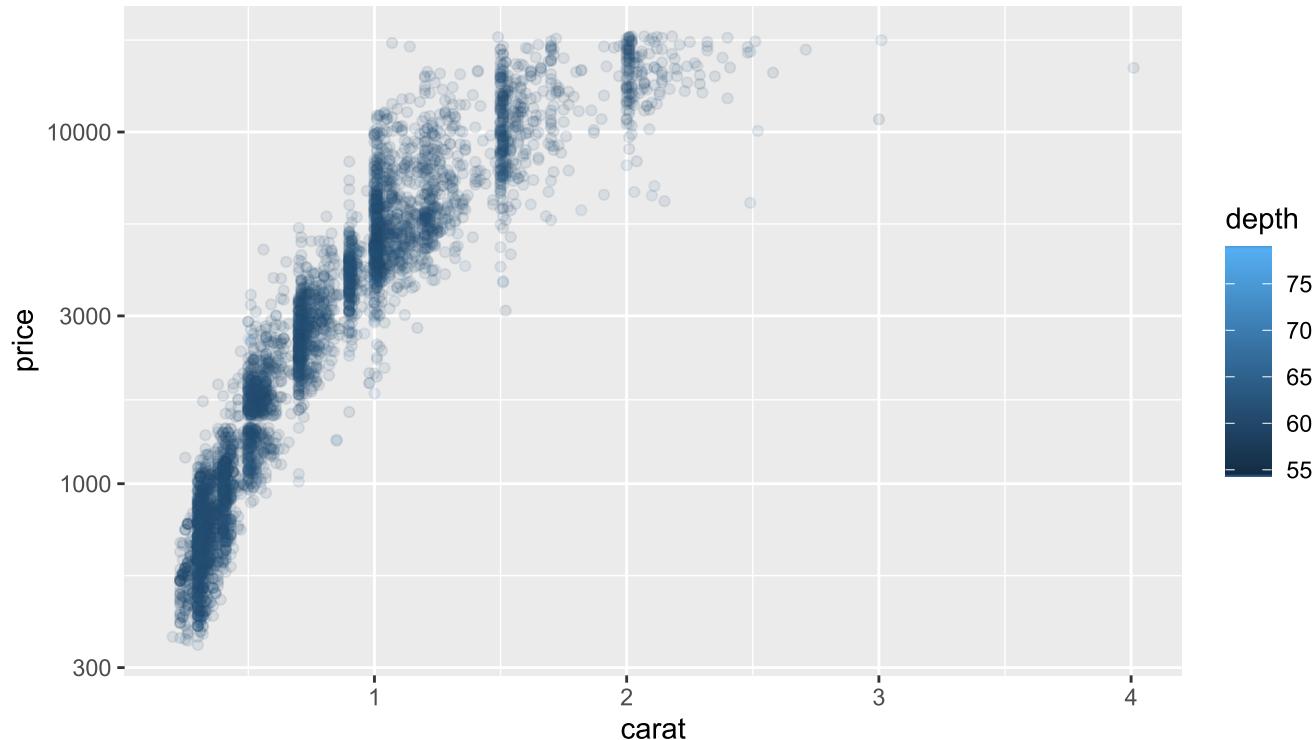
Working with scales (color boundaries)

```
p_diamonds + scale_color_continuous(low = "blue", high = "red")
```



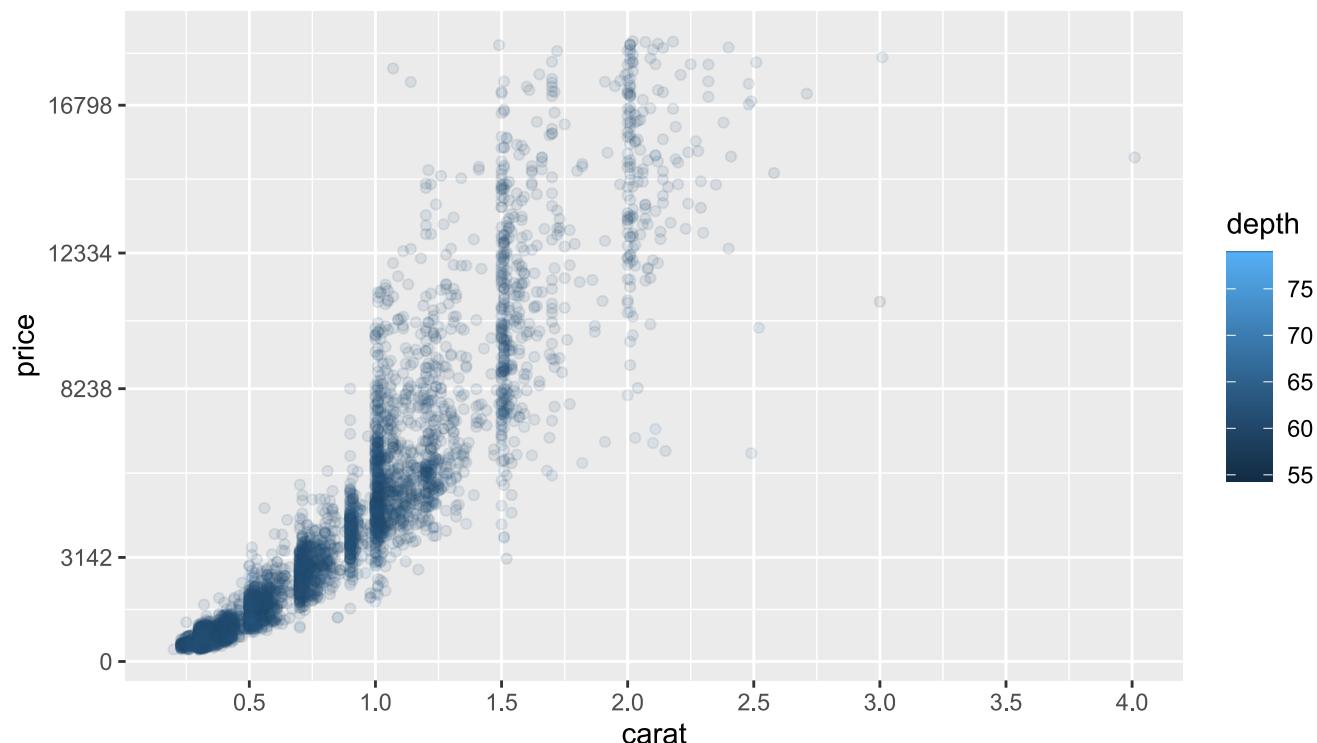
Working with scales: put a log scale

```
p_diamonds + scale_y_log10()
```



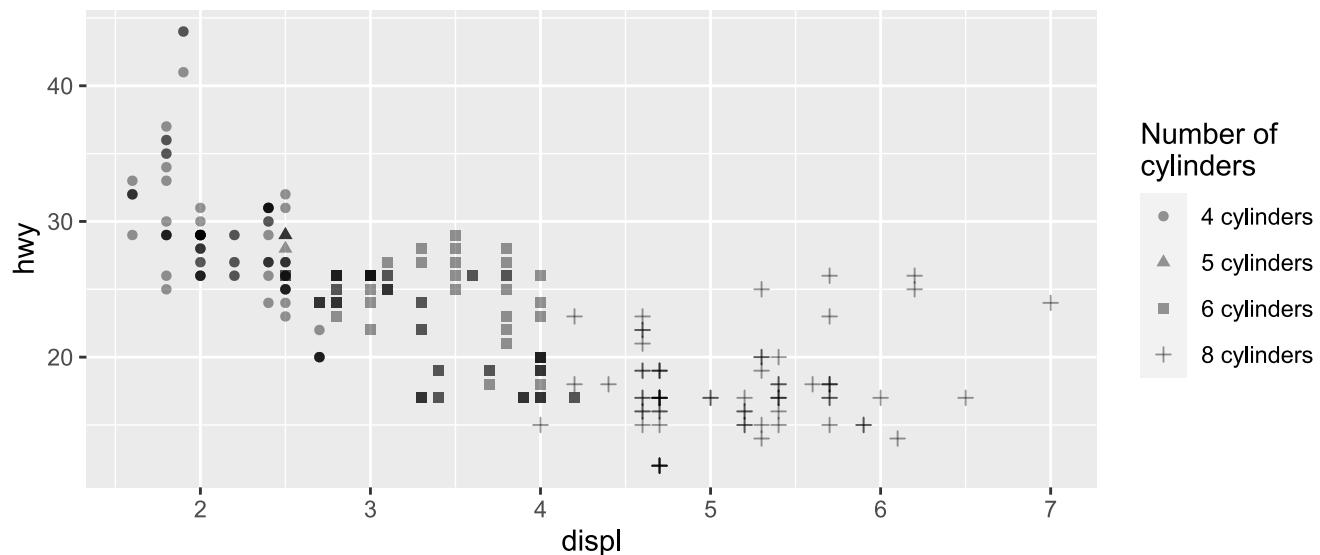
Working with scales: set the breaks of an axis

```
p_diamonds +  
  scale_x_continuous(n.breaks = 8) +  
  scale_y_continuous(breaks = c(0,3142, 8238,12334, 16798 ))
```



Working with scales: modify the shape's legend

```
p <- ggplot(data = mpg, aes(x = displ, y= hwy ))+  
  geom_point(aes(shape = factor(cyl)),  
             alpha = .3) +  
  scale_shape_discrete(labels = paste(c(4,5,6,8), "cylinders") ,  
                       name = "Number of cylinders") # Modify label  
p
```



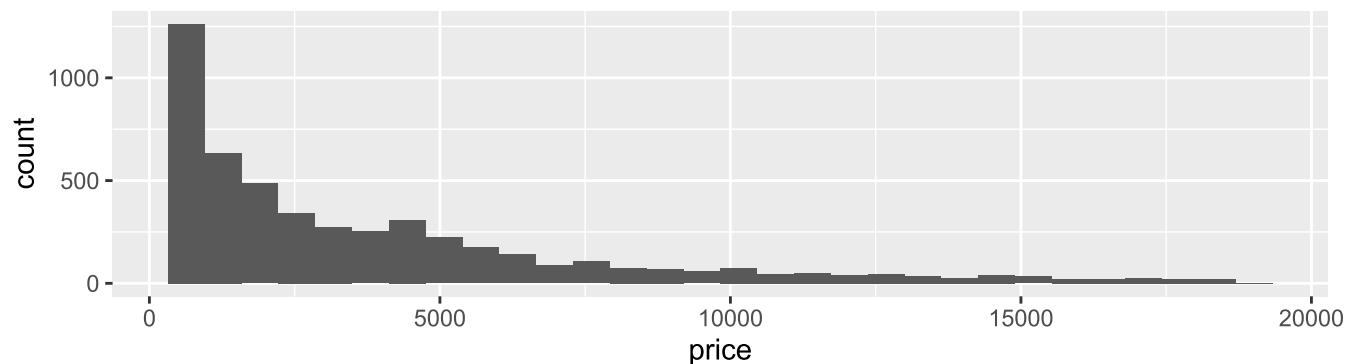
Facet: draw multiple plots easily

- Distribution of prices for the whole dataset:

Facet: draw multiple plots easily

- Distribution of prices for the whole dataset:

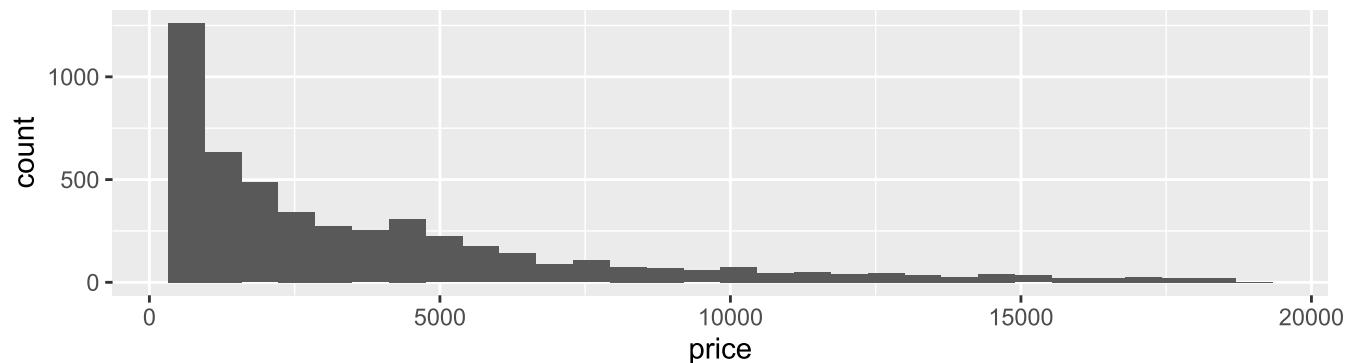
```
histogram_prices <- ggplot(data_work, aes(x = price)) + # No need for  
  geom_histogram()  
histogram_prices
```



Facet: draw multiple plots easily

- Distribution of prices for the whole dataset:

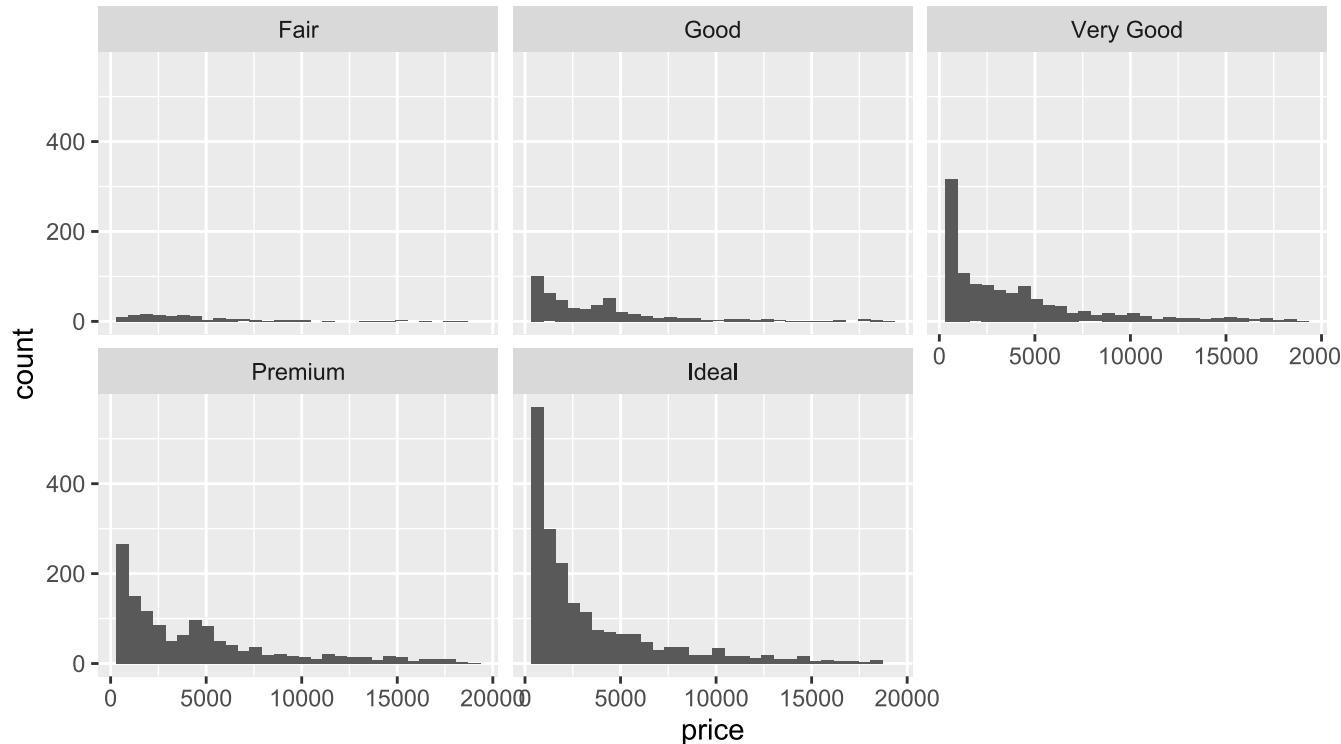
```
histogram_prices <- ggplot(data_work, aes(x = price)) + # No need for  
  geom_histogram()  
histogram_prices
```



- What if we want the distributions of prices *for each modality of the cut variable*

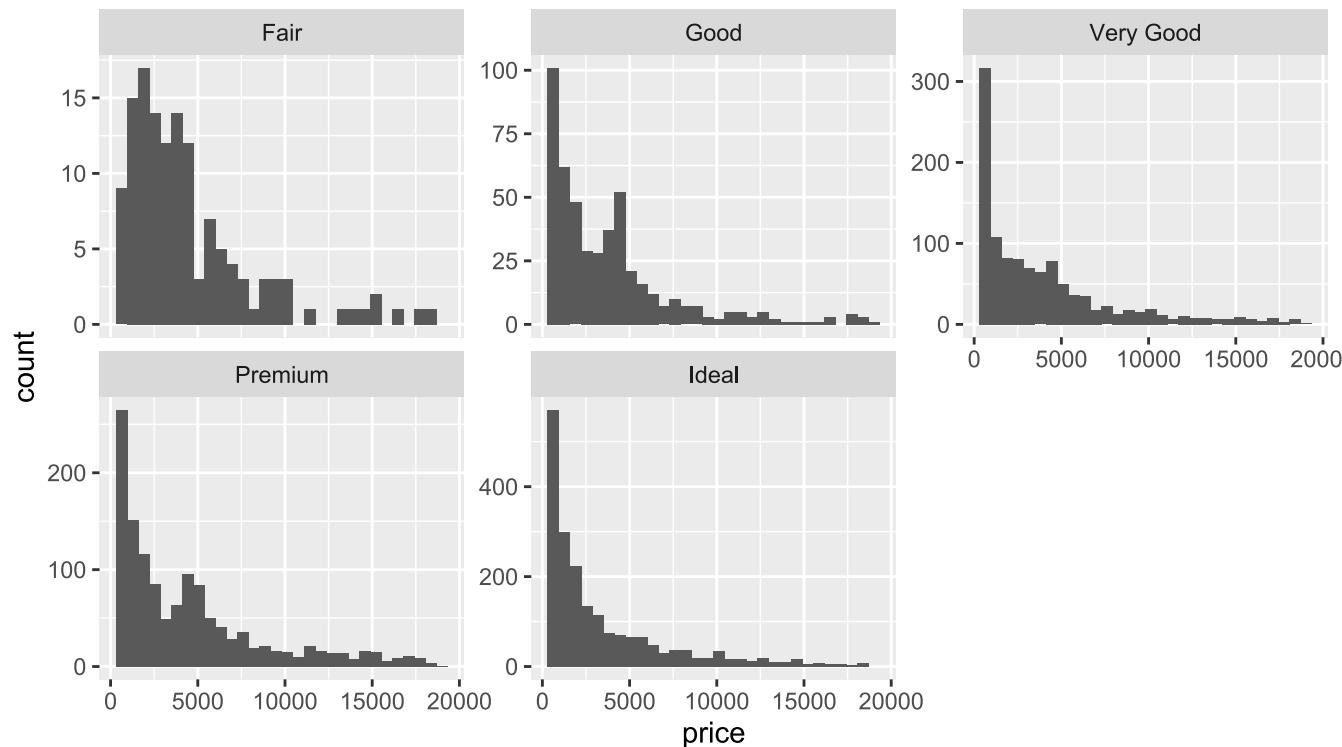
We simply need to use the `facet_wrap` function:

```
histogram_prices +  
  facet_wrap(cut~. )
```



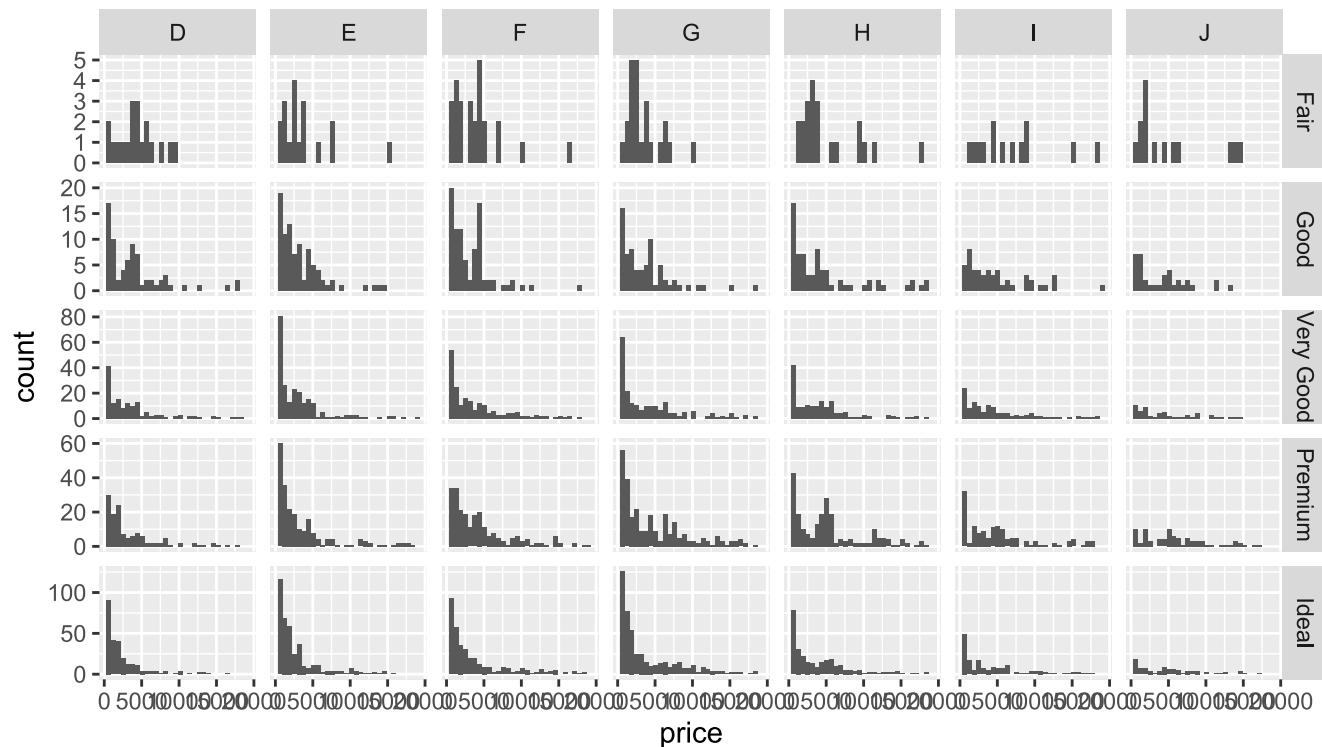
We can use the `scales` argument, to let scales vary freely across plots

```
histogram_prices +  
  facet_wrap(cut~., scales = "free_y" ) # Arg "free_x" for free x ax
```



- We can also make a grid of plots, using `facet_grid` function

```
histogram_prices +  
  facet_grid(cut~color,scales = "free_y")
```



Titles and subtitles: say what you show, and show what you say

- Many ways exist to specify titles in ggplot

Titles and subtitles: say what you show, and show what you say

- Many ways exist to specify titles in ggplot
- The easiest one may be to use the `labs` function

Titles and subtitles: say what you show, and show what you say

- Many ways exist to specify titles in ggplot
- The easiest one may be to use the `labs` function
- Let's take another dataset

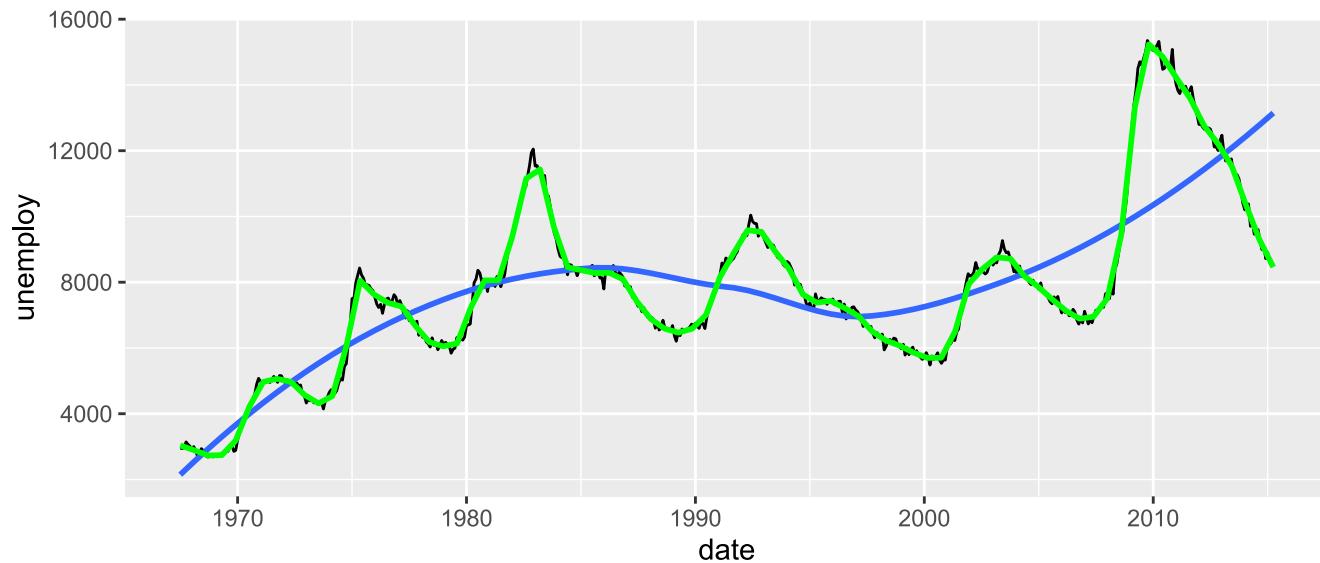
```
data(economics)
head(economics)
```

```
## # A tibble: 6 x 6
##   date          pce      pop psavert uempmed unemploy
##   <date>     <dbl>    <dbl>    <dbl>    <dbl>    <dbl>
## 1 1967-07-01  507.  198712    12.6     4.5    2944
## 2 1967-08-01  510.  198911    12.6     4.7    2945
## 3 1967-09-01  516.  199113    11.9     4.6    2958
## 4 1967-10-01  512.  199311    12.9     4.9    3143
## 5 1967-11-01  517.  199498    12.8     4.7    3066
## 6 1967-12-01  525.  199657    11.8     4.8    3018
```

```
str(economics)
```

- Here is a plot, without any title (btw. note the span argument to make the fit closer to the points)

```
plot_unemployed <- ggplot(data = economics, aes(x = date, y = unemployed)) +  
  geom_line() +  
  geom_smooth(se = FALSE) +  
  geom_smooth(span = 0.05, col = "green", se = FALSE )  
  
plot_unemployed
```



- Adding a title is straightforward

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```
plot_unemployed +  
  labs(title = "Number of unemployed people (in thousands) in the USA")
```



- Adding subtitle/caption is also straightforward

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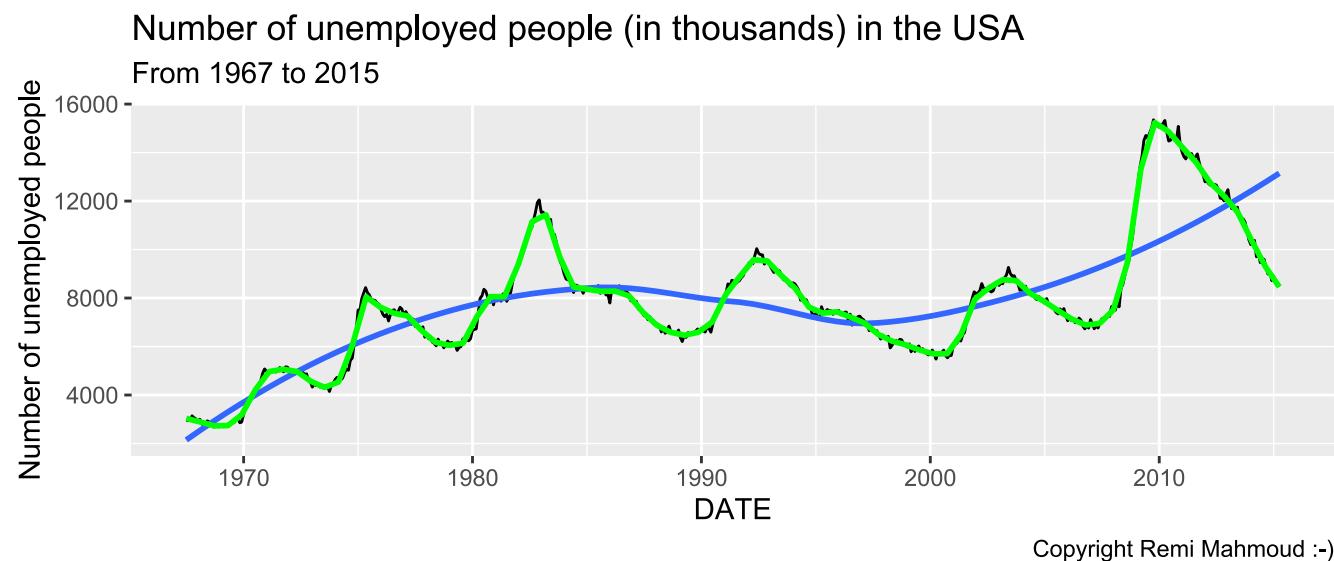
```
plot_unemployed +  
  labs(title = "Number of unemployed people (in thousands) in the USA",  
        subtitle = "From 1967 to 2015",  
        caption = "Copyright Remi Mahmoud :-)")
```



- Adding modifying axes titles is also straightforward

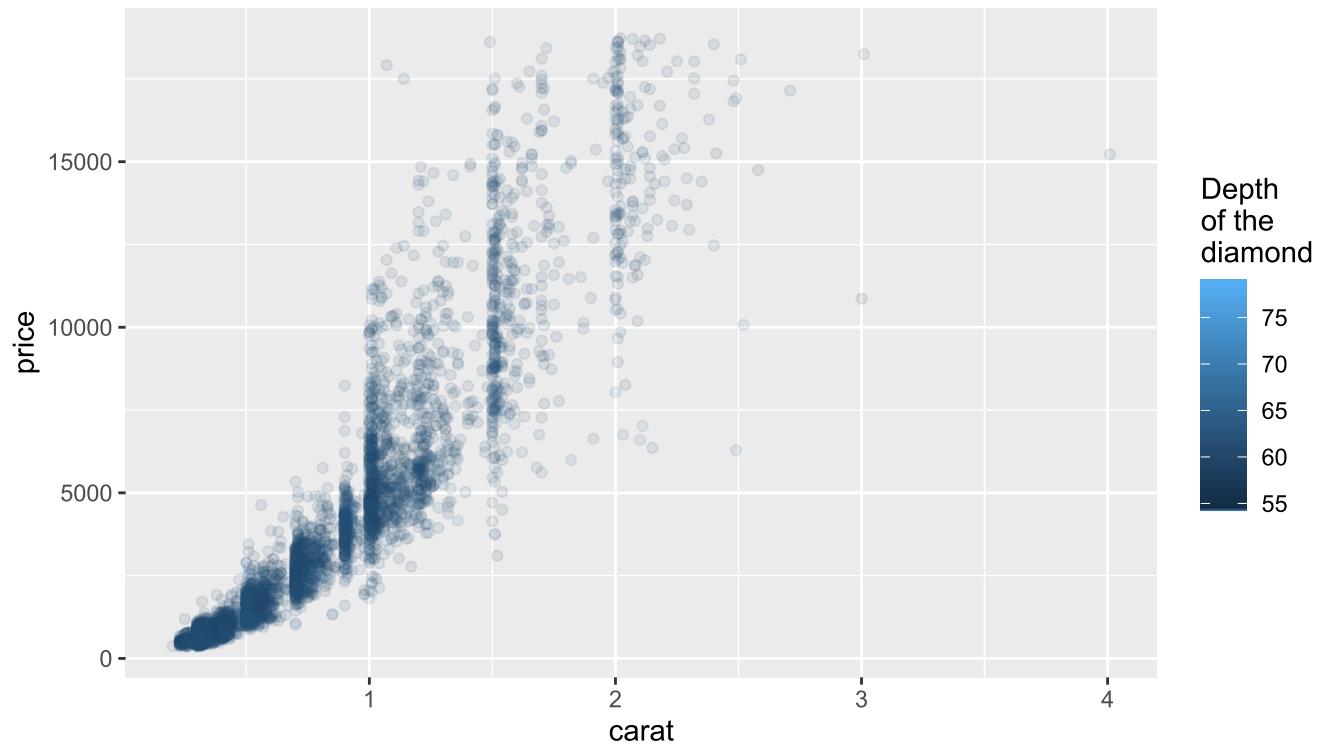
- Adding modifying axes titles is also straightforward

```
plot_unemployed +  
  labs(title = "Number of unemployed people (in thousands) in the USA/  
           From 1967 to 2015",  
        subtitle = "",  
        caption = "Copyright Remi Mahmoud :-)",  
        x = "DATE", y = "Number of unemployed people")
```



- Title of a legend can also be changed using the `labs` function

```
p_diamonds+ labs(colour = "Depth\nof the\ndiamond")
```



Is everything clear ?





Exercises

~ 20 minutes

Exercises

~ 20 minutes

Open "ggplot2_basics" with Rstudio

GGPLOT2: Customizing your plot



**GGPLOT(DATA,
AESO)
+ GEOM_POINTO**

**GGPLOT(DATA,
AESO) +
GEOM_POINTO + LABSO
+ FACET_WRAPO**

**GGPLOT(DATA,
AESO) + GEOM_POINTO
+ LABSO +
FACET_WRAPO + THEME_BWO**

- One may want to :

- One may want to :

specify color / shapes of points / boxplots / bars

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specify color / shapes of points / boxplots / bars

change the sizes/font/orientation of labels / titles

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specify color / shapes of points / boxplots / bars

change the sizes/font/orientation of labels / titles

change the background

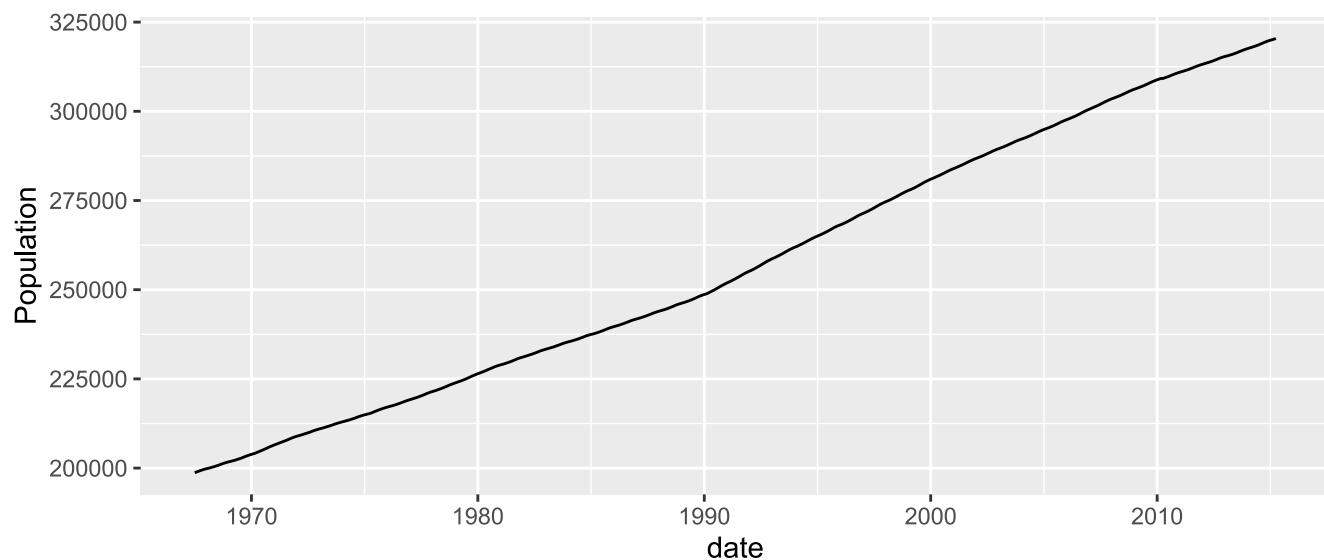
Themes: global appearance of the plot

- You may have noticed the (horrible) grey background of the default plots

Themes: global appearance of the plot

- You may have noticed the (horrible) grey background of the default plots

```
plot_population <- ggplot(data = economics, aes(x = date, y = pop)) +  
  geom_line() +  
  labs(y = "Population")  
  
plot_population
```



Non-data elements, like background property, legend appearance, axis labels are handled by *themes* in ggplot.

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We will see how to specify our own themes, but there are already some predefined themes (`theme_bw()`, `theme_minimal()`).

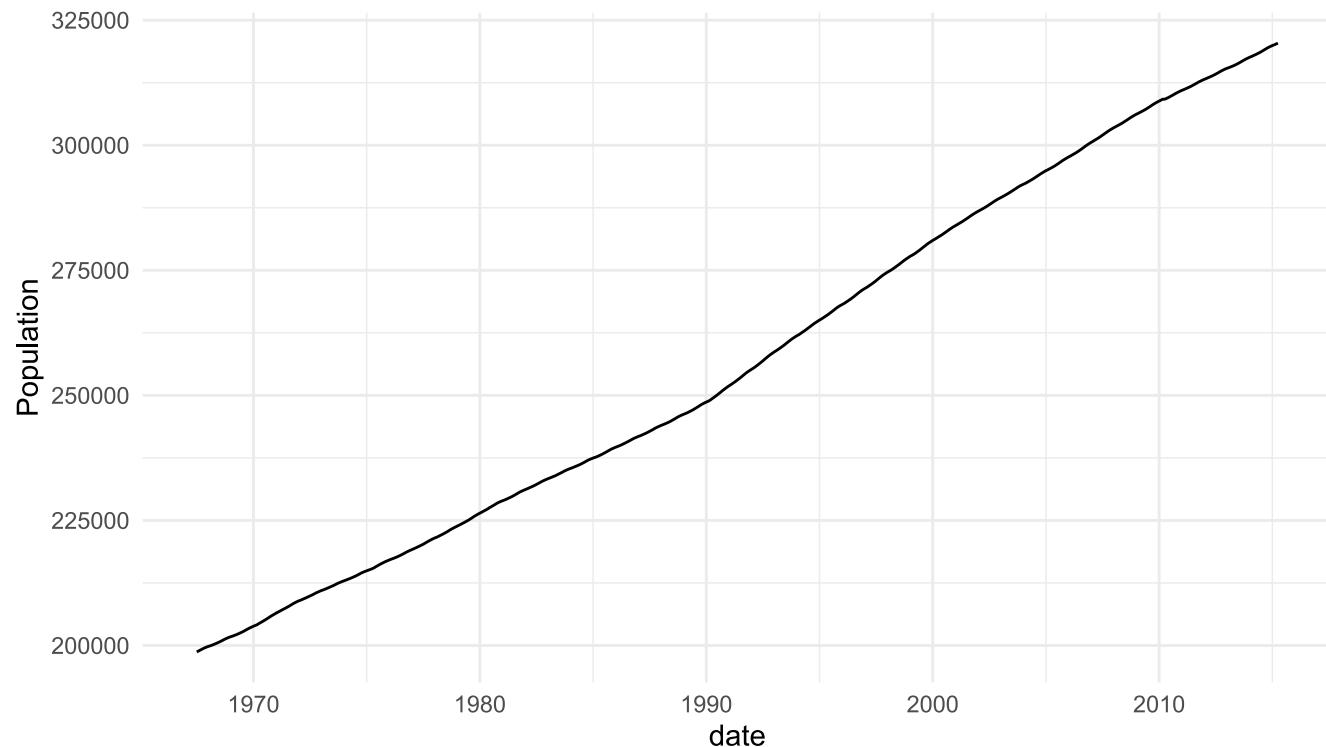
Non-data elements, like background property, legend appearance, axis labels are handled by *themes* in ggplot.

We will see how to specify our own themes, but there are already some predefined themes (`theme_bw()`, `theme_minimal()`).

To modify the theme of a plot, simply add the desired theme to the ggplot object

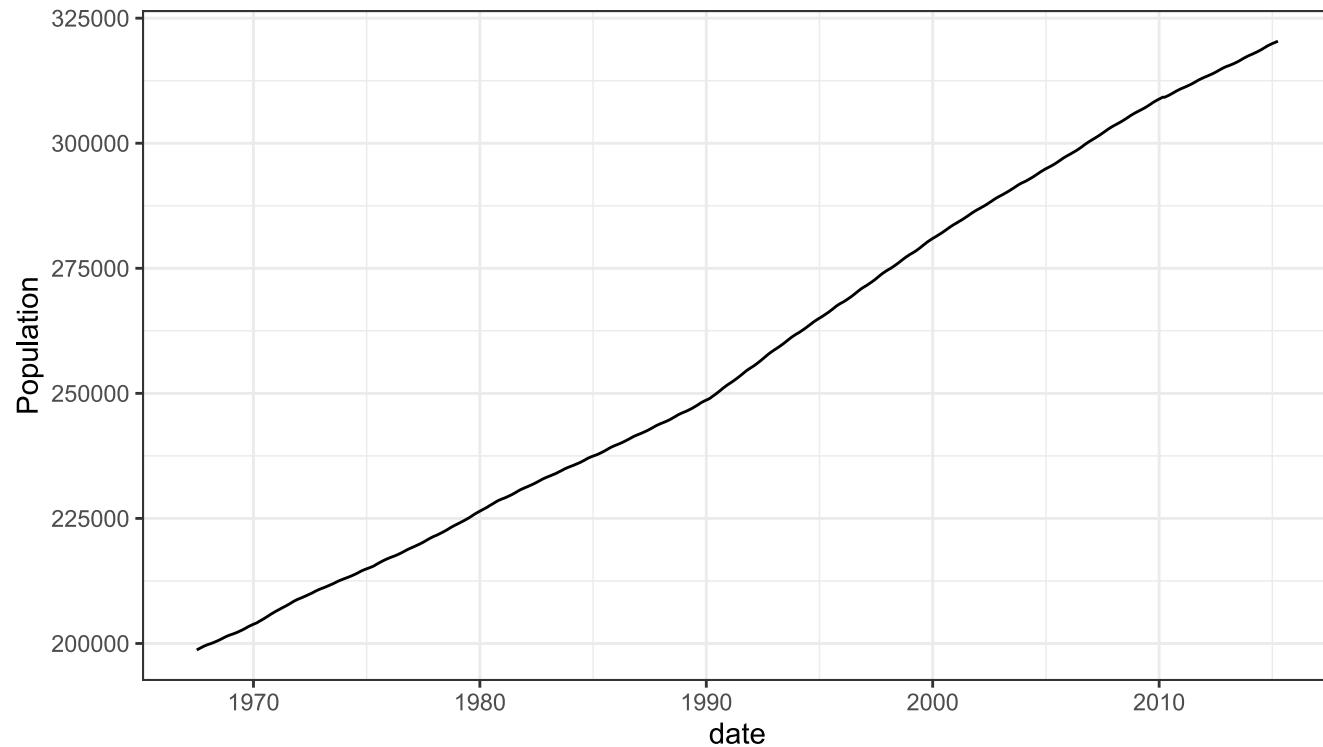
Predefined themes: theme_minimal()

```
plot_population + theme_minimal()
```



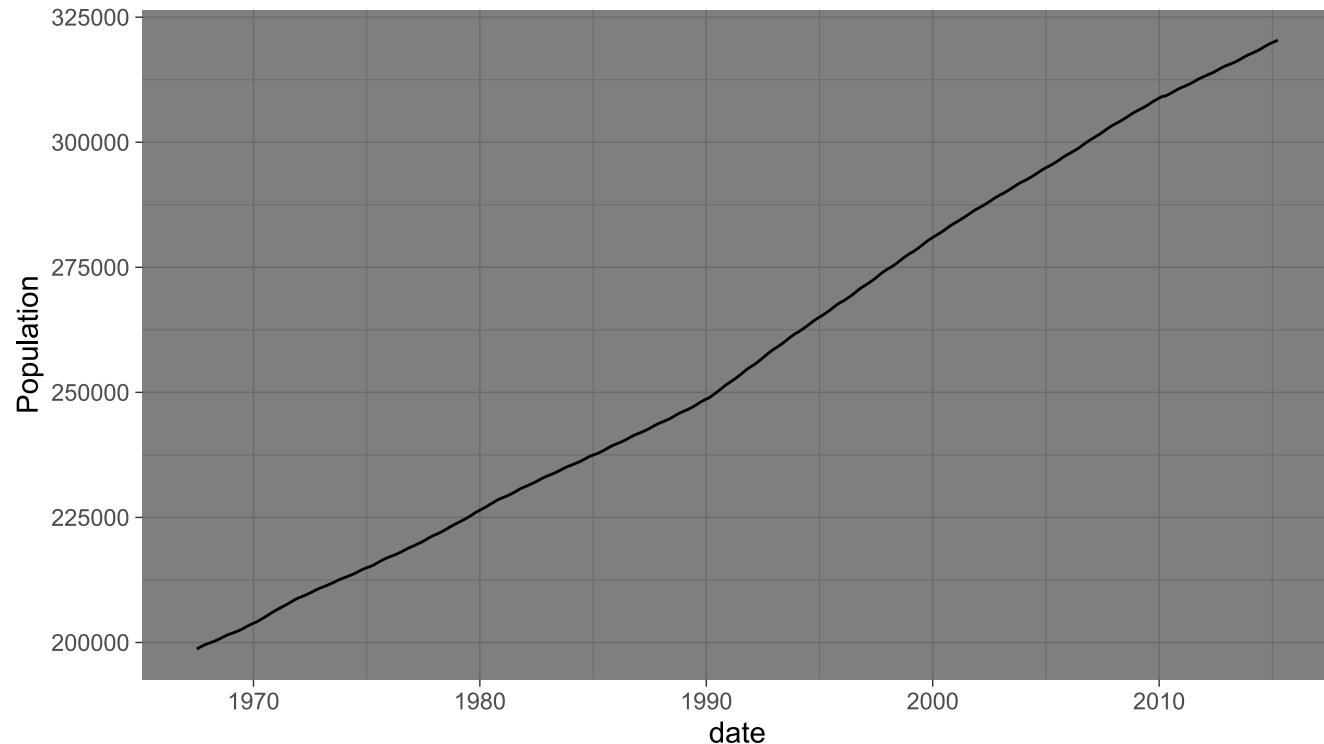
Predefined themes: theme_bw()

```
plot_population + theme_bw()
```



Predefined themes: theme_dark()

```
plot_population + theme_dark()
```



- Look at the (long) list of things you can modify in a theme

```
(attributes_theme <- names(theme_bw()))
```

```
## [1] "line"                      "rect"
## [3] "text"                       "title"
## [5] "aspect.ratio"               "axis.title"
## [7] "axis.title.x"                "axis.title.x.top"
## [9] "axis.title.x.bottom"         "axis.title.y"
## [11] "axis.title.y.left"           "axis.title.y.right"
## [13] "axis.text"                   "axis.text.x"
## [15] "axis.text.x.top"              "axis.text.x.bottom"
## [17] "axis.text.y"                 "axis.text.y.left"
## [19] "axis.text.y.right"            "axis.ticks"
## [21] "axis.ticks.x"                "axis.ticks.x.top"
## [23] "axis.ticks.x.bottom"          "axis.ticks.y"
## [25] "axis.ticks.y.left"            "axis.ticks.y.right"
## [27] "axis.ticks.length"            "axis.ticks.length.x"
## [29] "axis.ticks.length.x.top"       "axis.ticks.length.x.bottom"
## [31] "axis.ticks.length.y"          "axis.ticks.length.y.left"
## [33] "axis.ticks.length.y.right"     "axis.line"
## [35] "axis.line.x"                  "axis.line.x.top"
## [37] "axis.line.x.bottom"            "axis.line.y"
## [39] "axis.line.y.left"              "axis.line.y.right"
## [41] "legend.background"             "legend.margin"
## [43] "legend.spacing"                "legend.spacing.x"
## [45] "legend.spacing.y"              "legend.key"
## [47] "legend.key.size"               "legend.key.height"
```

- Some of the attributes of a `theme` regard text's elements (titles, axis labels etc.)

```
attributes_theme[grep(attributes_theme, pattern = "text|title")]
```

```
## [1] "text"                  "title"                 "axis.title"
## [4] "axis.title.x"           "axis.title.x.top"    "axis.title.x.bot·
## [7] "axis.title.y"           "axis.title.y.left"   "axis.title.y.rigl
## [10] "axis.text"              "axis.text.x"        "axis.text.x.top"
## [13] "axis.text.x.bottom"     "axis.text.y"        "axis.text.y.left"
## [16] "axis.text.y.right"      "legend.text"        "legend.text.alig·
## [19] "legend.title"          "legend.title.align" "plot.title"
## [22] "plot.title.position"   "plot.subtitle"     "strip.text"
## [25] "strip.text.x"          "strip.text.y"       "strip.text.y.lef·
```

- Some of the attributes of a theme regard text's elements (titles, axis labels etc.)

```
attributes_theme[grepl(attributes_theme, pattern = "text|title")]
```

```
## [1] "text"                  "title"                 "axis.title"
## [4] "axis.title.x"           "axis.title.x.top"    "axis.title.x.bot·
## [7] "axis.title.y"           "axis.title.y.left"   "axis.title.y.rigl·
## [10] "axis.text"              "axis.text.x"         "axis.text.x.top"
## [13] "axis.text.x.bottom"     "axis.text.y"         "axis.text.y.left"
## [16] "axis.text.y.right"      "legend.text"        "legend.text.align"
## [19] "legend.title"          "legend.title.align" "plot.title"
## [22] "plot.title.position"   "plot.subtitle"      "strip.text"
## [25] "strip.text.x"          "strip.text.y"        "strip.text.y.lef·
```

- Texts elements are modified using the `element_text()` function

- Some of the attributes of a `theme` regard text's elements (titles, axis labels etc.)

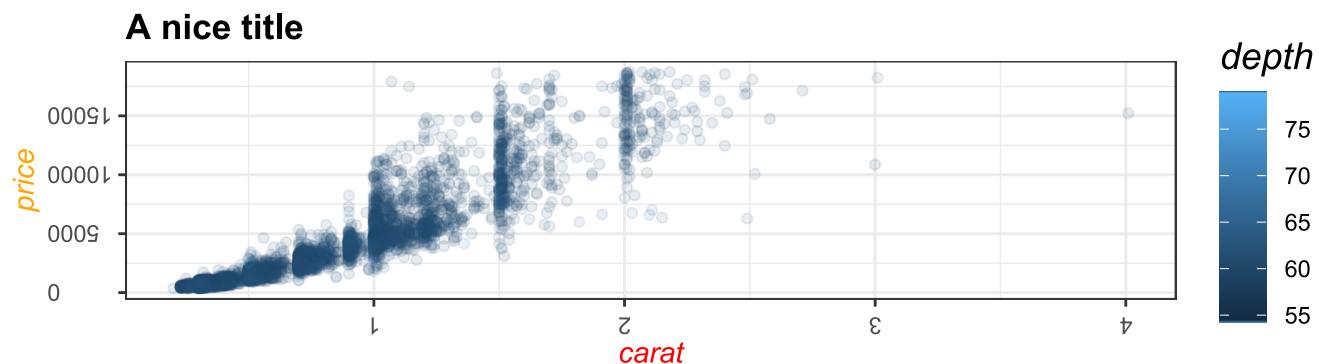
```
attributes_theme[grepl(attributes_theme, pattern = "text|title")]
```

```
## [1] "text"                  "title"                 "axis.title"
## [4] "axis.title.x"           "axis.title.x.top"    "axis.title.x.bot·
## [7] "axis.title.y"           "axis.title.y.left"   "axis.title.y.rigl
## [10] "axis.text"               "axis.text.x"         "axis.text.x.top"
## [13] "axis.text.x.bottom"     "axis.text.y"         "axis.text.y.left"
## [16] "axis.text.y.right"      "legend.text"        "legend.text.align"
## [19] "legend.title"           "legend.title.align" "plot.title"
## [22] "plot.title.position"    "plot.subtitle"      "strip.text"
## [25] "strip.text.x"           "strip.text.y"        "strip.text.y.lef·
```

- Texts elements are modified using the `element_text()` function
- `element_text()` can set the font, the color, the angle of the text setted (see `?element_text()` for a complete list)

Modify texts elements

```
p_diamonds + labs(title = "A nice title") +  
  theme_bw() +  
  theme(axis.title.x = element_text(color = "red"), # Modify title of x axis  
        axis.title.y = element_text(color = "orange"), # Modify title of y axis  
        axis.text = element_text(angle = 180), # Modify the angle of axis labels  
        legend.title = element_text(size = 14 ),# Modify the size of legend title  
        plot.title = element_text(face = "bold"), # Modify the face of plot title  
        title = element_text(face = "italic")) # Modify the face of main title
```



Modify background of elements

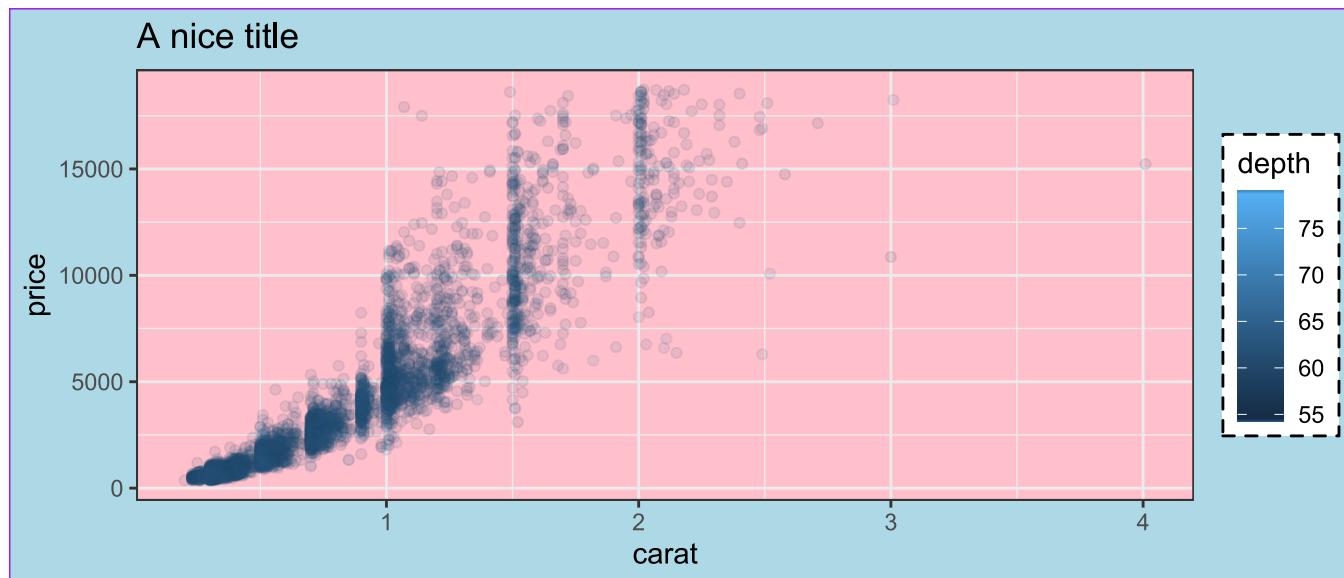
- Background elements are modified using the `element_rect()` function

```
attributes_theme[grepl(attributes_theme, pattern = "background")]
```

```
## [1] "legend.background"      "legend.box.background" "panel.backgroi
## [4] "plot.background"        "strip.background"       "strip.backgroi
## [7] "strip.background.y"
```

```
p_diamonds + labs(title = "A nice title") +  
  theme_bw() +  
  theme(legend.background = element_rect(color = "black", linetype =  
    plot.background = element_rect(fill = "lightblue", colour =  
    panel.background = element_rect(fill = "pink")) #guess what
```

```
p_diamonds + labs(title = "A nice title") +  
  theme_bw() +  
  theme(legend.background = element_rect(color = "black", linetype =  
    plot.background = element_rect(fill = "lightblue", colour =  
    panel.background = element_rect(fill = "pink")) #guess what
```



Save your plot using `ggsave`

- Saving your plot at a fixed size is a good practice

Save your plot using ggsave

- Saving your plot at a fixed size is a good practice

```
#Many parameters exist to save your plots  
ggsave(filename = "my_path/title_of_the_plot.jpg",  
        plot = my_plot, units = "cm",  
        height = 25, width = 20)
```

Is everything clear ?



Exercises

~ 20 minutes

Exercises

~ 20 minutes

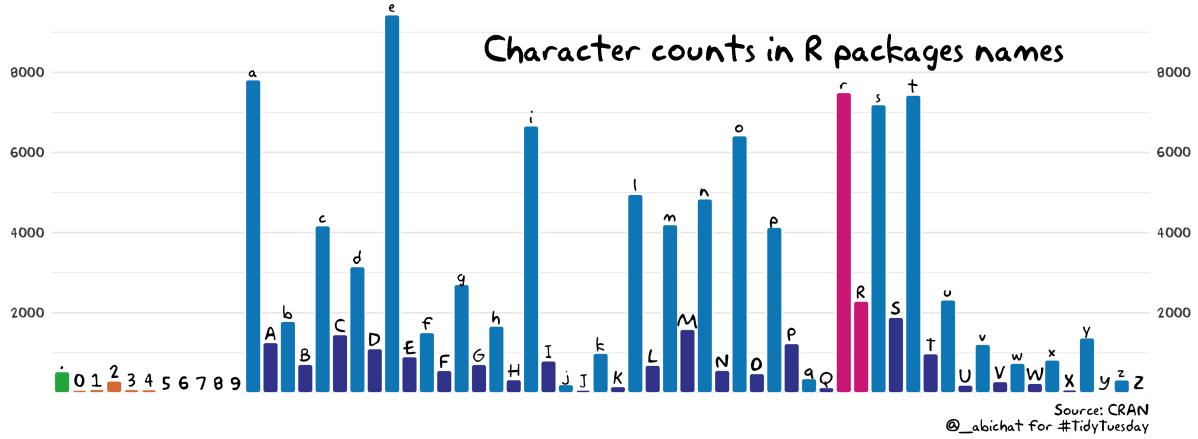
Open "ggplot2_customization" with Rstudio

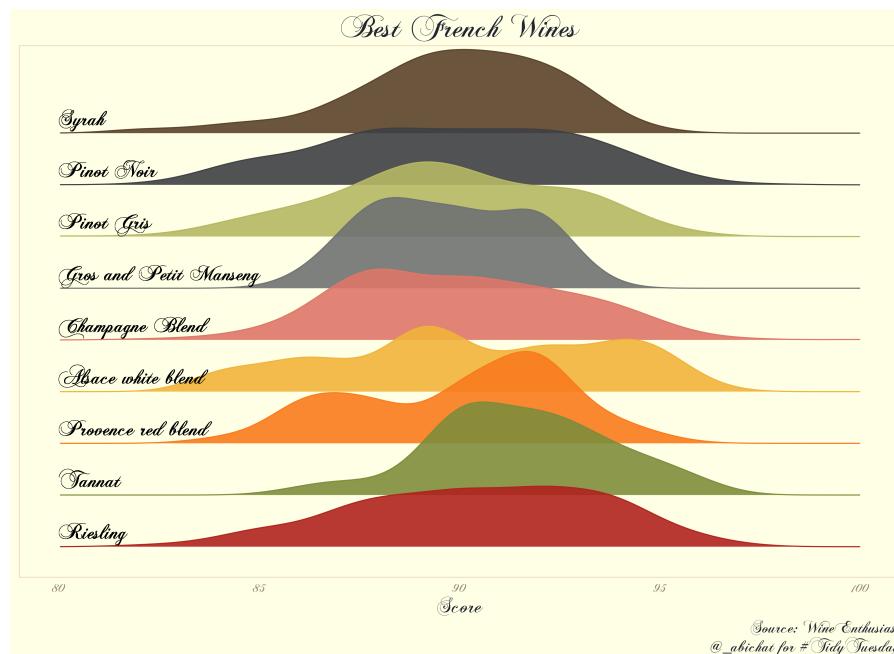
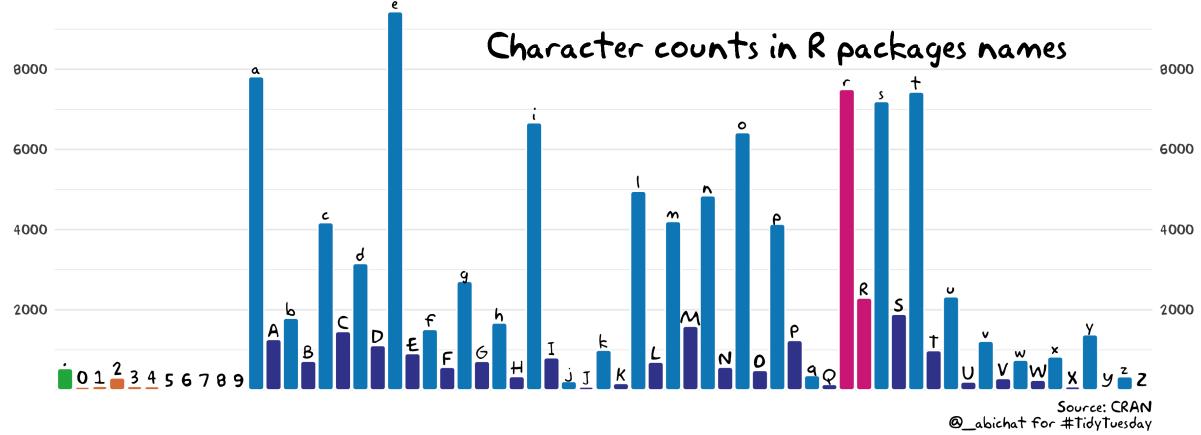
Conclusion

- This was just an introduction

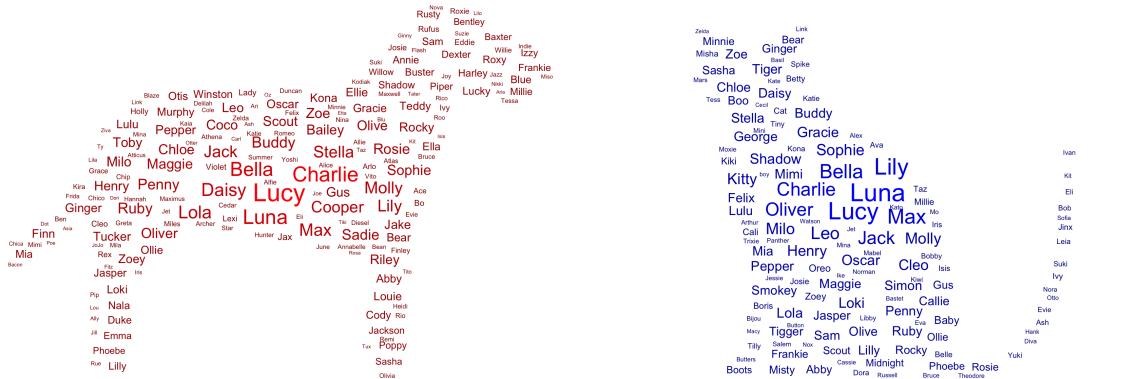
- This was just an introduction
- ggplot2 offer *many* possibilities

- This was just an introduction
- ggplot2 offer *many* possibilities
- Here are some wonderful plots by @abichat to give you some tastes



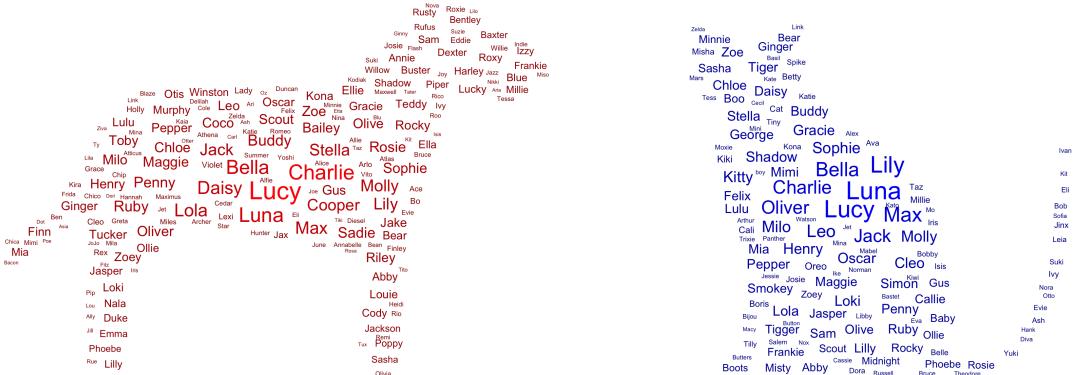


Most used names for dogs and cats in Seattle



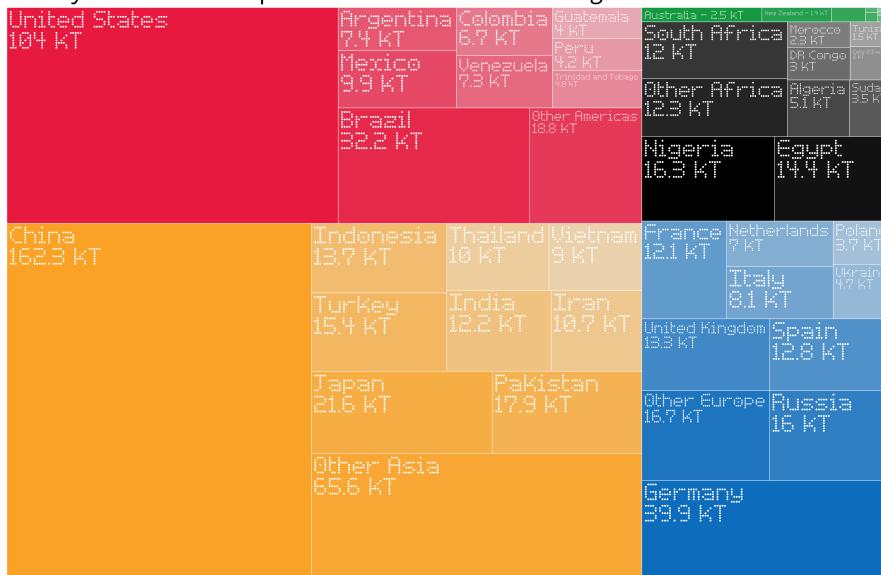
Source: Seattle's open data portal
@_abichat for #TidyTuesday

Most used names for dogs and cats in Seattle



Source: Seattle's open data portal
 @_abichat for #TidyTuesday

Daily amount of plastic waste entering the ocean



Source: Our World in Data
@ abichat for #TidyTuesday

ggplot2 resources online

- wonderful and complete article about `ggplot`: my holy bible for ggplot2
- a bunch of `tutorials` on ggplot2, really clear
- the `ggplot2 book`, recall and explanations

References

- a lesson from N. Vialaneix
- the `ggplot2` book
- wonderful plots of A. Bichat: go take a look !

THANKS

THANKS

Any remark, questions ?

THANKS

Any remark, questions ?

remi.mahmoud@inrae.fr

THANKS

Any remark, questions ?

remi.mahmoud@inrae.fr

Lesson contents available at:

https://github.com/RemiMahmoud/ggplot2_lesson