Class 5: Data visualization

Moises Gonzalez (PID: A17579866)

Base R graphics vs ggplot2

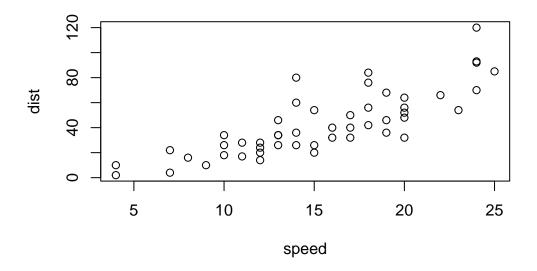
There are many graphics systems available in R, including so-called "base" R graphics and the very popular **ggplot2** package.

To compare these let's play with the inbuilt carsdataset.

head(cars)

To use "base" R I can simply call the plot() function:

```
plot(cars)
```



To use ggplot2 package I first need to install it with the function install.packages("ggplot2").

I will run this in my R console (i.e. the R brain) as I do not want to re-install it every time I render my report...

The main function in this package is called ggplot(). Can I just call it

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

To make a figure with ggplot I need always at least 3 things:

- Data (i.e. What I want to plot.)
- $\bf Aes$ (The aesthetic mapping of the data to the plot I want.)
- The Geoms (i.e. How I want to plot the data.)

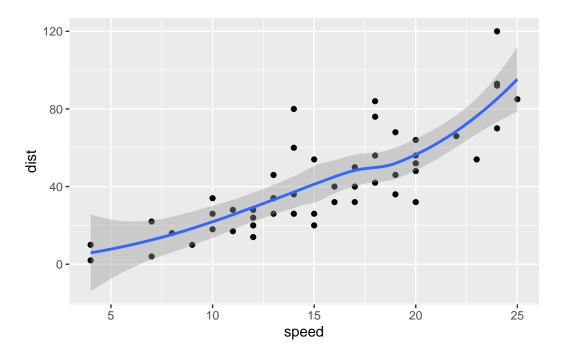
```
ggplot(data=cars) +
  aes(x=speed, y=dist) +
  geom_point()
```



Let's make a plot with a smooth line

```
ggplot(data=cars) +
  aes(x=speed, y=dist) +
  geom_point() +
  geom_smooth()
```

 $\ensuremath{\mbox{`geom_smooth()`}}\ \mbox{using method} = \ensuremath{\mbox{'loess'}}\ \mbox{and formula} = \ensuremath{\mbox{'y}}\ \sim \ensuremath{\mbox{x'}}\ \mbox{'}$

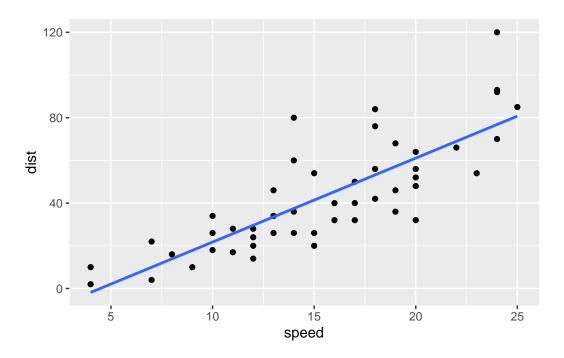


Ggplot is much more verbose that base R plot but it has a consistent layer system that I can use to make just about my plot.

Let's make a plot with a straight line fit - i.e. a linear model and no standard error shown.

```
ggplot(data=cars) +
  aes(x=speed, y=dist) +
  geom_point() +
  geom_smooth(se=FALSE, method="lm")
```

[`]geom_smooth()` using formula = 'y ~ x'

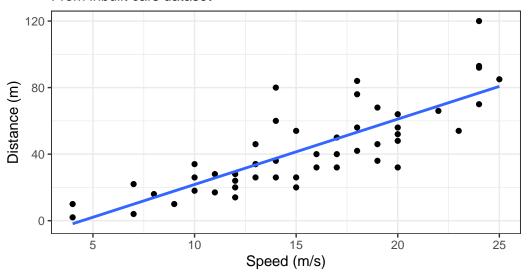


Adding various labels & annotations with the labs() function and changing the plot look to a more conservative "black & white" theme by adding the theme_bw() function:

[`]geom_smooth()` using formula = 'y ~ x'

Cars: Distance VS Speed

From inbuilt cars dataset



3 the speed of cars and the distances taken to stop. Note that the data were recorded in the 1920s.

A more complicated plot

Let's load some gene expression data. The code below reads the results of a differential expression analysis where a new anti-viral drug is being tested.

```
url <- "https://bioboot.github.io/bimm143_S20/class-material/up_down_expression.txt"
genes <- read.delim(url)
head(genes)</pre>
```

```
Gene Condition1 Condition2 State
1 A4GNT -3.6808610 -3.4401355 unchanging
2 AAAS 4.5479580 4.3864126 unchanging
3 AASDH 3.7190695 3.4787276 unchanging
4 AATF 5.0784720 5.0151916 unchanging
5 AATK 0.4711421 0.5598642 unchanging
6 AB015752.4 -3.6808610 -3.5921390 unchanging
```

Q1: How many genes are in this dataset?

```
nrow(genes)
```

[1] 5196

Q2: How can we summarize that last column - the "State" column?

There are different ways to call genes the "State" column:

- genes[,4]
- genes[,"State"]
- genes\$State

```
table(genes$State)
```

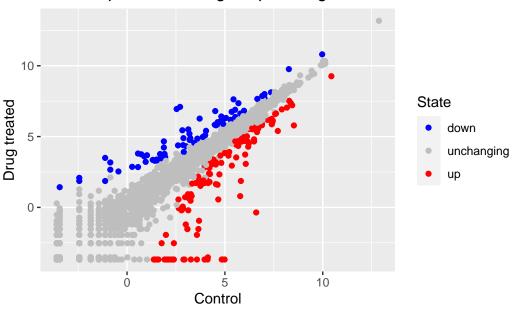
```
down unchanging up
72 4997 127
```

Note we can save our plot as an object and use it later to add more layers:

```
p <- ggplot(genes) +
  aes(x=Condition2, y=Condition1, color=State) +
  geom_point()</pre>
```

I can now just call **p** when I want to plot or add to it.

Gene Expression changes upon drug treatment

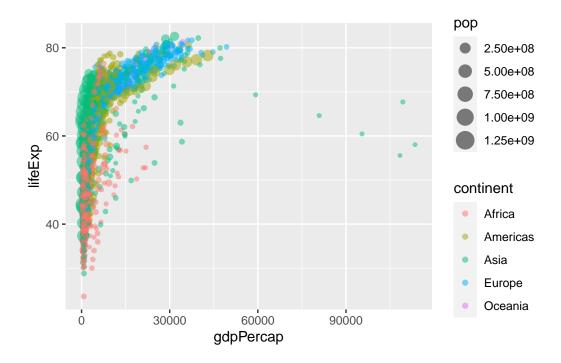


Going Further

Here I read a slightly larger dataset

```
# File location online
  url <- "https://raw.githubusercontent.com/jennybc/gapminder/master/inst/extdata/gapminder.
  gapminder <- read.delim(url)</pre>
  head(gapminder)
      country continent year lifeExp
                                         pop gdpPercap
1 Afghanistan
                   Asia 1952 28.801 8425333 779.4453
2 Afghanistan
                  Asia 1957 30.332 9240934 820.8530
3 Afghanistan
                  Asia 1962 31.997 10267083 853.1007
4 Afghanistan
                  Asia 1967 34.020 11537966 836.1971
5 Afghanistan
                  Asia 1972 36.088 13079460 739.9811
6 Afghanistan
                  Asia 1977 38.438 14880372 786.1134
```

```
ggplot(gapminder) +
  aes(x=gdpPercap, y=lifeExp, color=continent, size=pop) +
  geom_point(alpha=0.5)
```



A very useful layer to add sometimes is for "faceting"

```
ggplot(gapminder) +
  aes(x=gdpPercap, y=lifeExp, color=continent, size=pop) +
  geom_point(alpha=0.5) +
  facet_wrap(~continent)
```

