Class 5: Data Visualization

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Base R graphics vs ggplot2

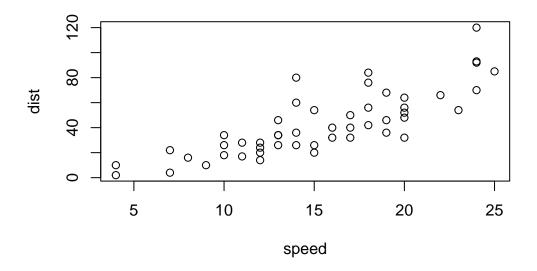
There Are many graphics systems available in R, including so-called "base" R graphics and the very popular **ggplot2** package.

To compare these let's play with the inbuilt cars dataset.

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:

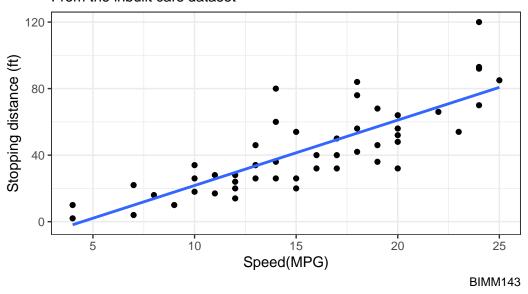
- data (i.e. what I want to plot)
- aes the aesthetic mapping of the data to the plot I want.
- the \mathbf{geoms} i.e. How I was nt to plot the data

```
ggplot(data=cars) +
aes(x=speed, y=dist) +
geom_point() +
geom_smooth(se=FALSE, method="lm") + labs(title="Stopping distance for old cars", subtitle
x="Speed(MPG)",
y="Stopping distance (ft)") +
theme_bw()
```

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

Stopping distance for old cars

From the inbuilt cars dataset



GGplot is much more verbose than base R plots for standard plots but it has a consistent layer system that I can use to make just about any plot.

A more complicated plot

Let's plot some gene expression data. The code below reads the results on 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
```

Q. How many genes are in this dataset

```
nrow(genes)
```

[1] 5196

Q. How can we summarize that last column - the "State" column?

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

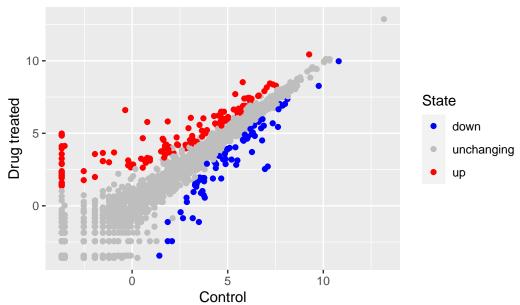
```
down unchanging up
    72    4997    127

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

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

```
p + labs(title="Gene Expression changes upon drug treatment",
    x="Control", y="Drug treated") + scale_colour_manual( values=c("blue", "gray", "red") )
```

Gene Expression changes upon drug treatment



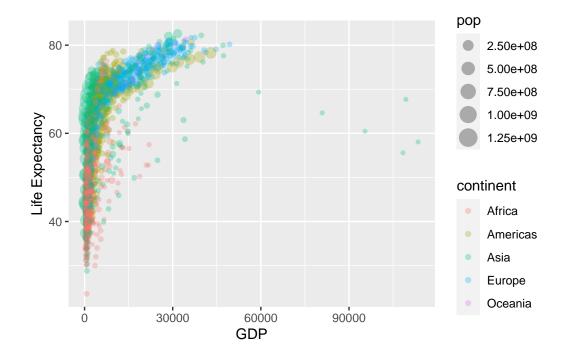
##Going further

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)
head(gapminder)</pre>
```

```
countrycontinentyearlifeExppopgdpPercap1 AfghanistanAsia195228.8018425333779.44532 AfghanistanAsia195730.3329240934820.85303 AfghanistanAsia196231.99710267083853.10074 AfghanistanAsia196734.02011537966836.19715 AfghanistanAsia197236.08813079460739.98116 AfghanistanAsia197738.43814880372786.1134
```

```
ggplot(data=gapminder) +
  aes(x=gdpPercap, y=lifeExp, color=continent, size=pop) +
  geom_point(alpha=0.3) +
  labs(x="GDP", y="Life Expectancy")
```



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

```
ggplot(data=gapminder) +
  aes(x=gdpPercap, y=lifeExp, color=continent, size=pop) +
  geom_point(alpha=0.3) +
  labs(x="GDP", y="Life Expectancy") +
  facet_wrap(~continent)
```

