

# class05: 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 bvery popular **ggplot2** package.

To compare these let’s play with teh inbuilt “cars” dataset.

```
head(cars)
```

	speed	dist
1	4	2
2	4	10
3	7	4
4	7	22
5	8	16
6	9	10

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 everytime I render my report...

The main function in this package is called “ggplot()”

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

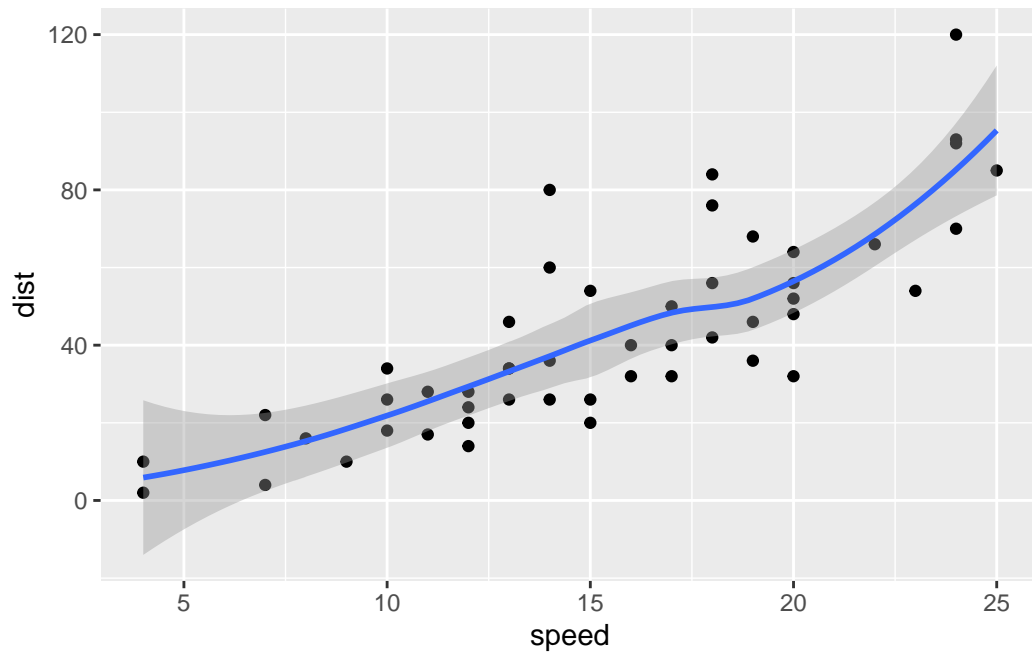


To make a figure with ggplot I need 3 things:

- **data** (what I want to plot)
- **aesthetic** mapping of data to plot i want
- the **geoms** (how I want to plot the data)

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

``geom_smooth()`` using method = 'loess' and formula = 'y ~ x'



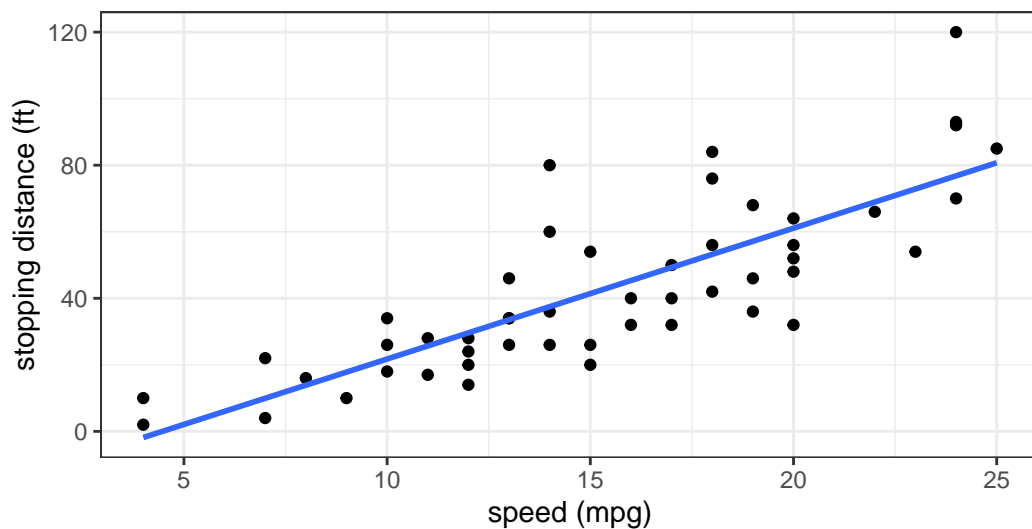
ggplot is much more verbose than base R plots but it has a consistent layer system that I can use to make just about any plot

```
ggplot(data=cars) +
  aes(x=speed, y=dist) +
  geom_point() +
  geom_smooth(se=F, method="lm") +
  labs(title="Stopping distance for old cars",
        subtitle= "From inbuilt cars dataset",
        caption="BIMM 143",
        x= "speed (mpg)", y="stopping distance (ft)") +
  theme_bw()
```

`geom\_smooth()` using formula = 'y ~ x'

## Stopping distance for old cars

From inbuilt cars dataset



BIMM 143

## A more complicated plot

Lets plot some gene expression data

The code below reads the results of a differential expression analysis where a new anti-viral drug is being tested. So

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

	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 the last column - the “State” column?

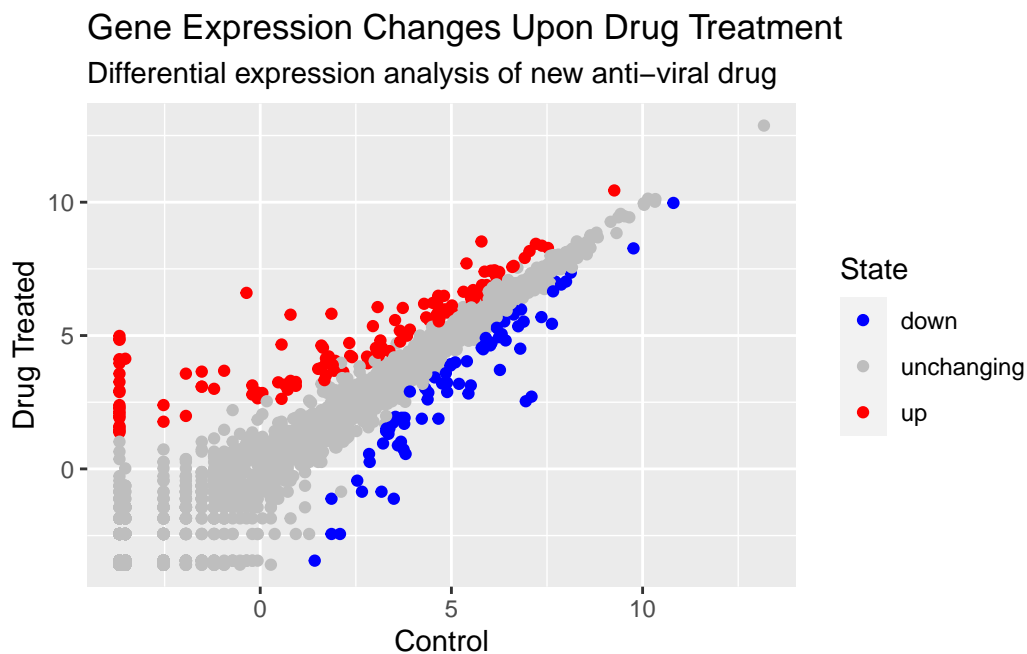
```
table(genes$State)
```

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

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

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

```
p + labs(title="Gene Expression Changes Upon Drug Treatment",
  subtitle = "Differential expression analysis of new anti-viral drug",
  x="Control", y="Drug Treated") +
  scale_color_manual(values=c("blue", "gray", "red"))
```



## Going further

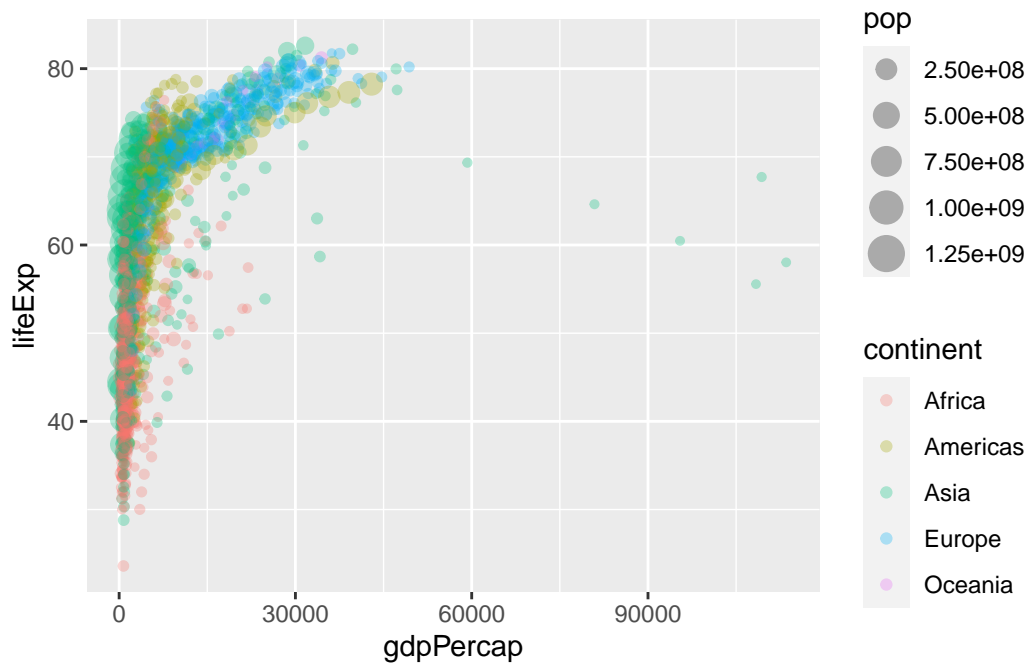
Here I read a slightly larger dataset

```
#File location online
url <- "https://raw.githubusercontent.com/jennybc/gapminder/master/inst/extdata/gapminder."

gapminder <- read.delim(url)
head(gapminder)
```

	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, col=continent, size=pop) +
  geom_point(alpha=0.3)
```



A very useful layer to add sometimes is for “faceting”.

```
ggplot(gapminder) +  
  aes(x=gdpPercap, y=lifeExp, col=continent, size=pop) +  
  geom_point(alpha=0.3) +  
  facet_wrap(~continent)
```

