



## Advanced Graphics Using R

Osama Mahmoud

**Website:** <http://osmahmoud.com>

**E-mail:** [o.mahmoud@bristol.ac.uk](mailto:o.mahmoud@bristol.ac.uk)

In this session, I aim to cover:

- What is data visualisation?
- Why do we need visualisation?
- How to use `ggplot2` in R.
- What are the building-blocks of `ggplot2` graphs
- How to use facets to produce multiple panels in same plot
- How to use themes to control the appearance of a plot.

# Contents

- 1 Introduction to data visualisation
- 2 Overview of ggplot2
- 3 Plot building-blocks
- 4 Facets
- 5 Themes

# Graphics reveal data

*“The greatest value of a picture is when it forces us to notice what we never expected to see.”*

JOHN W. TUKEY (1915 - 2000)

# Why do we need visualisations?

I		II		III	
x	y	x	y	x	y
10.0	8.04	10.0	9.14	10.0	7.46
8.0	6.95	8.0	8.14	8.0	6.77
13.0	7.58	13.0	8.74	13.0	12.74
9.0	8.81	9.0	8.77	9.0	7.11
11.0	8.33	11.0	9.26	11.0	7.81
14.0	9.96	14.0	8.10	14.0	8.84
6.0	7.24	6.0	6.13	6.0	6.08
4.0	4.26	4.0	3.10	4.0	5.39
12.0	10.84	12.0	9.13	12.0	8.15
7.0	4.82	7.0	7.26	7.0	6.42
5.0	5.68	5.0	4.74	5.0	5.73

$n = 11$

mean  $x = 9.0$

mean  $y = 7.5$

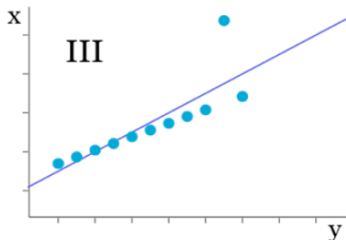
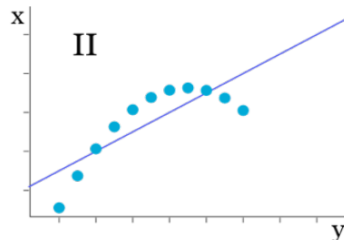
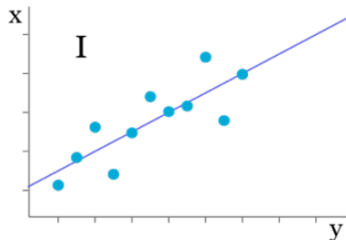
variance  $x = 11.0$

variance  $y = 4.12$

correlation  $x$  &  $y = 0.816$

regression line:  $y = 3 + 0.5x$

# Why do we need visualisations?



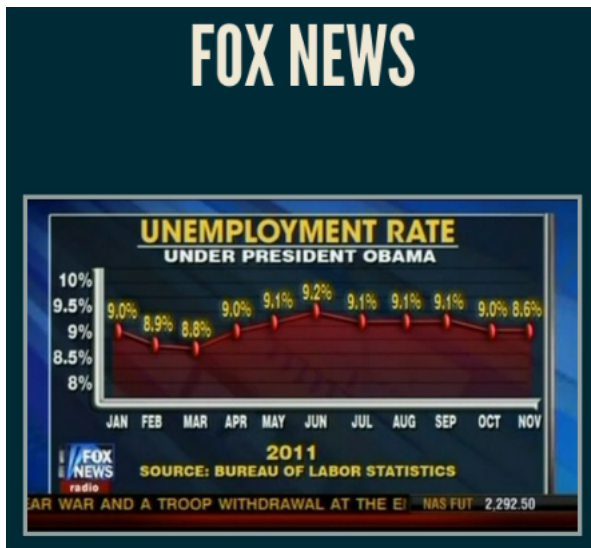
# THE UGLY

# EXCEL

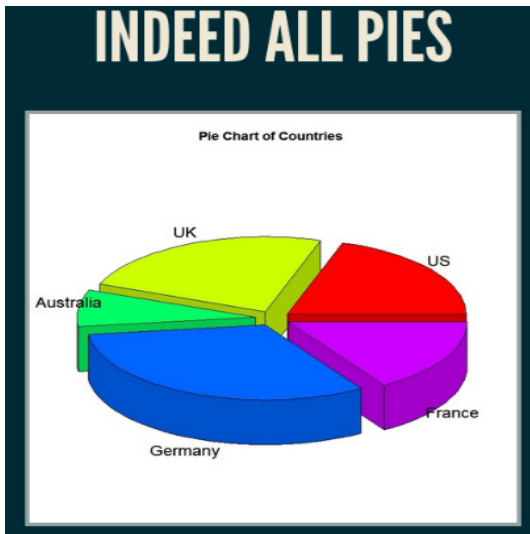




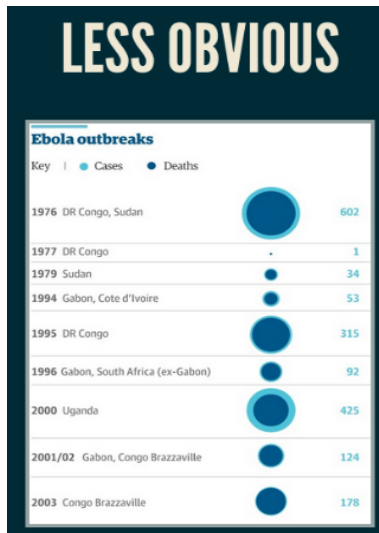
## The ugly



# The ugly



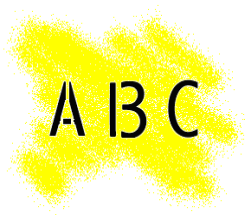
## The ugly



# FEW ISSUES

## Few issues

- Human perception can be ...



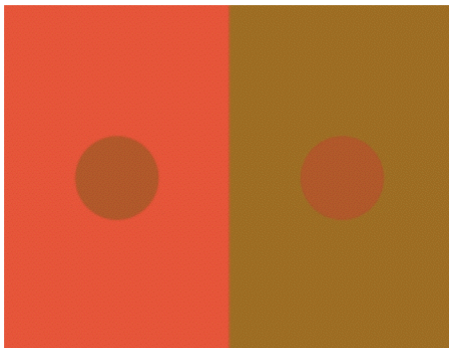
## Few issues

- Human perception can be deceiving



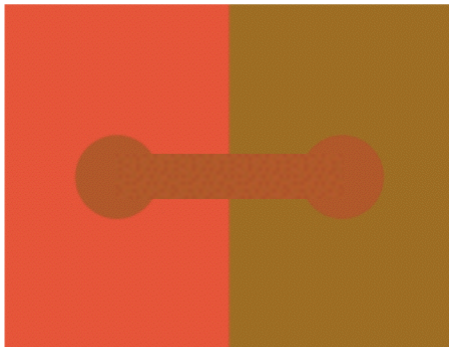
## Few issues

- Another example:



## Few issues

- Another example:





## Few issues

- Selective attention



Selective attention test

# Graphics in R

# Graphics in R: Background

Installing packages in R is straightforward. For example:

```
> install.packages("ggplot2")
```

Then, you can simply load it to your R session whenever needed:

```
> library("ggplot2")
```

# Types of R graphics

# Types of R graphics

- Base graphics.
- Grid graphics.
- Lattice graphics.
- ggplot2 graphics.

# Overview of ggplot2

# Installing Course R package: BristolVis

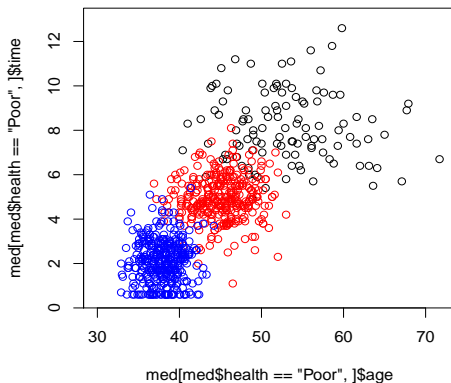
```
> install.packages("drat")
```

```
> drat::addRepo("statcourses")
```

```
> install.packages("BristolVis", type="source")
```

# Basic plots using base graphics

```
> plot(med[med$health=="Poor",]$age, med[med$health=="Poor",]$time,  
xlim=c(30,72), ylim=c(0.5,13))  
> points(med[med$health=="Fair",]$age, med[med$health=="Fair",]$time,  
col=2)  
> points(med[med$health=="Good",]$age, med[med$health=="Good",]$time,  
col=4)
```





## Basic plots using base graphics

- We had to manually set the scales using the `xlim` and `ylim` parameters.
- We had not created a legend. We would need to use the `legend` function to create one.
- The default axis labels were terrible!

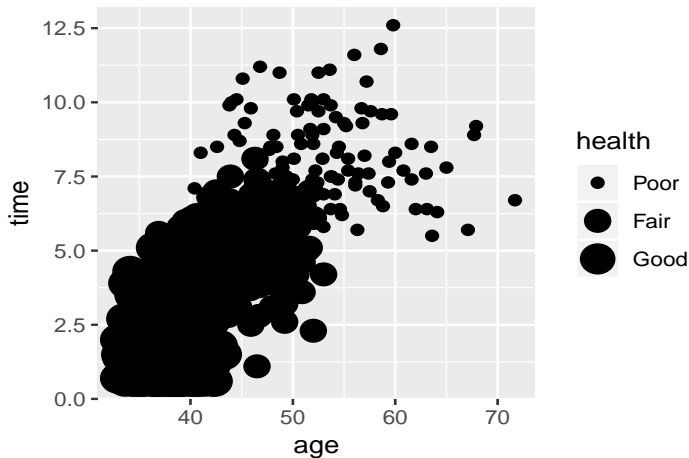
# Equivalent graphics using ggplot2

```
> library(ggplot2)
> g = ggplot(data=med, aes(x=age, y=time))
> g + geom_point(aes(colour=health))
```



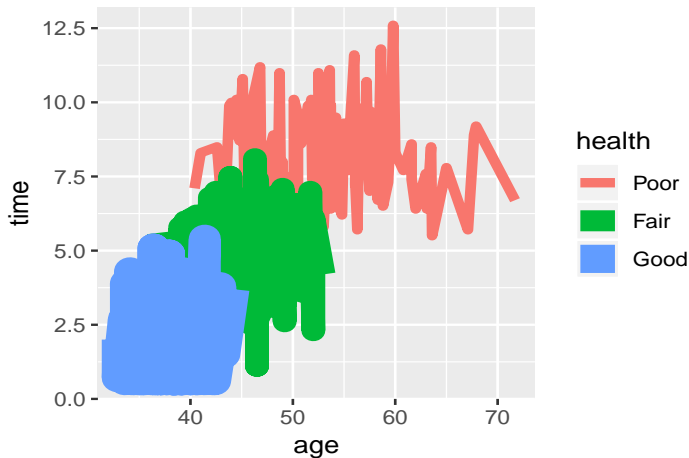
# Factor of point sizes

```
g + geom_point(aes(size=health))
```



# Example of a line chart

```
g + geom_line(aes(colour=health, size = health))
```



# Geometric objects

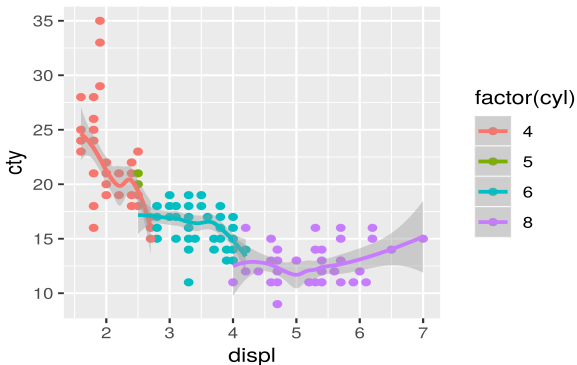
Points, bars and lines are examples of geom's. Some useful standard geoms and their equivalent base graphic counter part:

Plot Name	Geom	Base graphic
Bar chart	bar	<code>barplot</code>
Box-and-whisker	boxplot	<code>boxplot</code>
Histogram	histogram	<code>hist</code>
Line plot	line	<code>plot</code> and <code>lines</code>
Scatter plot	point	<code>plot</code> and <code>points</code>

## More complicated functions

The idea of graphical layers, enables constructing more complicated functions, e.g.:

```
p = ggplot(mpg, aes(x = displ, y = cty)) +  
  geom_point(aes(colour=factor(cyl))) +  
  stat_smooth(aes(colour=factor(cyl)))
```



# Understanding ggplot2 philosophy

Each `ggplot` command adds iteratively layers. A single layer may comprise of four elements:

- an aesthetic and data mapping;
- a geometric object (`geom`);
- a statistical transformation (`stat`);
- a position adjustment, i.e. how should overlapped objects be handled.

# Understanding ggplot2 philosophy

For example, the command:

```
g + geom_point(aes(colour=health))
```

actually calls (in the background) the command:

```
g + layer(data = med, #inherited  
mapping = aes(color=health), #x and y are inherited.  
stat = "identity",  
geom = "point",  
position = "identity",  
params = list(na.rm=FALSE))
```



# Plot building-blocks

## Initial plot object

An initial ggplot object, can be setup using the `ggplot()` function which has two arguments:

- `data` (*takes a data frame*)
- an aesthetic `mapping` (*creates default aesthetic attributes*)

```
g = ggplot(data=mpg, mapping=aes(x=displ, y=cty,  
colour=factor(cyl)))
```

Or equivalently,

```
g = ggplot(mpg, aes(displ, cty, colour=factor(cyl)))
```

doesn't actually produce anything to be displayed, it just sets the initial plot object. We need to add layers for that to happen.

# The `geom_` functions

- The `geom_` functions perform the actual rendering in a plot, e.g. a line geom will create a line plot and a point geom creates a scatter plot.
- Each geom has a list of aesthetics that it accepts such as `x` , `y` , `colour` and `size`.
- However, some geoms have unique elements. For example, the `geom_errorbar` requires arguments `ymin` and `ymax`.
- The full list of aesthetics can be displayed by:  

```
> ggplot2:::.all_aesthetics
```

# The geom\_ functions

- This table gives names and descriptions of some commonly used geoms:

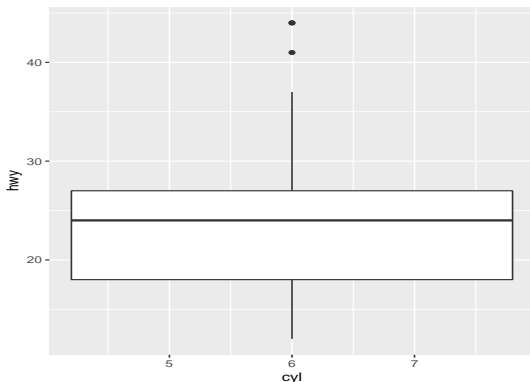
Name	Description
<code>abline</code>	Line, specified by slope and intercept
<code>boxplot</code>	Box and whiskers plot
<code>density</code>	Kernel density plot
<code>histogram</code>	Histograms
<code>jitter</code>	Individual points are jittered to avoid overlap
<code>step</code>	Connect observations by stairs

## Combining geoms

I will show how to combine more than one `geom` function to produce a bit more complex plots. If we consider the `mpg` data set, a base `ggplot` object:

`g = ggplot(mpg, aes(x=factor(cyl), y=hwy))` will do nothing.

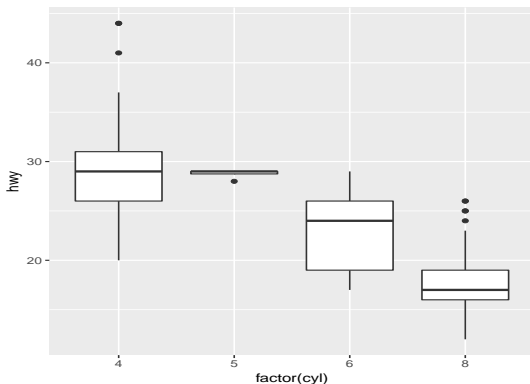
Now we'll create a boxplot: `(g1 = g + geom_boxplot())`



## Combining geoms

Previous figure was a boxplot of all the `mpg` data, a more useful plot would be to have individual boxplots conditional on number of cylinders:

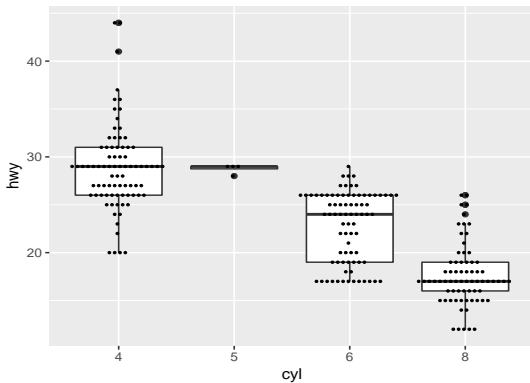
```
(g2 = g + geom_boxplot(aes(x=factor(cyl), group=cyl)))
```



## Combining geoms

We are not restricted to a single geom. When data sets are reasonably small, it is useful to display the data on top of the boxplots:

```
(g3 = g2 + geom_dotplot(aes(x=factor(cyl), group=cyl),  
  binaxis="y", stackdir="center", binwidth=0.25,  
  stackratio=2))
```



# Standard plots

There are a few standard geom 's that are particular useful:

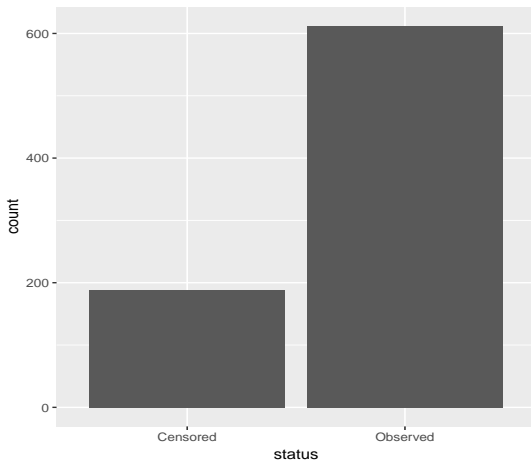
- `geom_line` : a line plot.
- `geom_boxplot` : produces a boxplot.
- `geom_point` : a scatter plot.
- `geom_dotplot`: a dot plot.
- `geom_bar` : produces a standard barplot that counts the x values.
- `geom_text` : adds labels to specified points (as `geom_point` but draw labels rather than points).
- `geom_raster`: Similar to `levelplot` (heatmap).



## Standard plots

To generate a bar plot, e.g. for the status of effect observations in the `med` data set, the following code can be used:

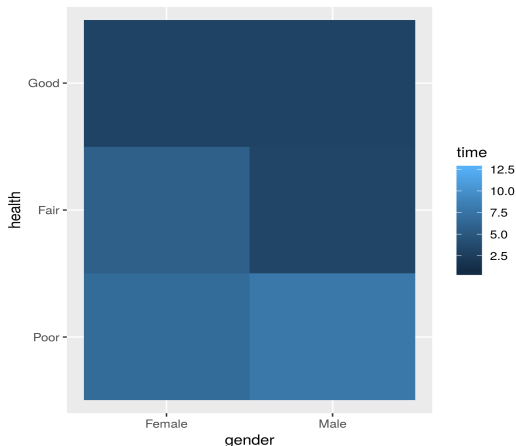
```
ggplot(med, aes(x=status)) + geom_bar()
```



# Standard plots

An example of a `geom_raster`:

```
ggplot(med, aes(gender, health)) +  
  geom_raster(aes(fill=time))
```



# Facets

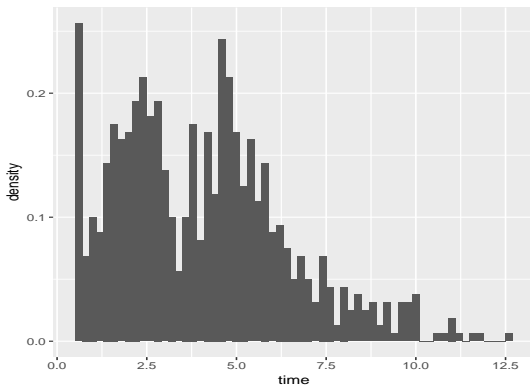
# Facets

- Faceting is a mechanism for automatically laying out multiple plots on a page.
- The data is split into subsets, each subset onto a different panel.
- `ggplot2` has two types of faceting:
  - `facet_grid`: produces a 2d panel of plots where variables define rows and columns.
  - `facet_wrap`: produces a 1d ribbon of panels which can be wrapped into 2d.

# Facet grid

Suppose we are interested in time to relief from the `med` data set. A first plot we could generate is a basic histogram:

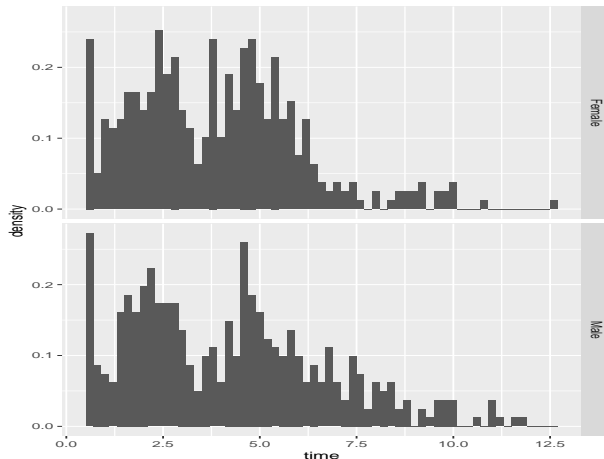
```
g = ggplot(med, aes(x=time)) +  
  geom_histogram(aes(y=..density..), binwidth=0.2)
```



# Facet grid

We will now use faceting to explore the data further:

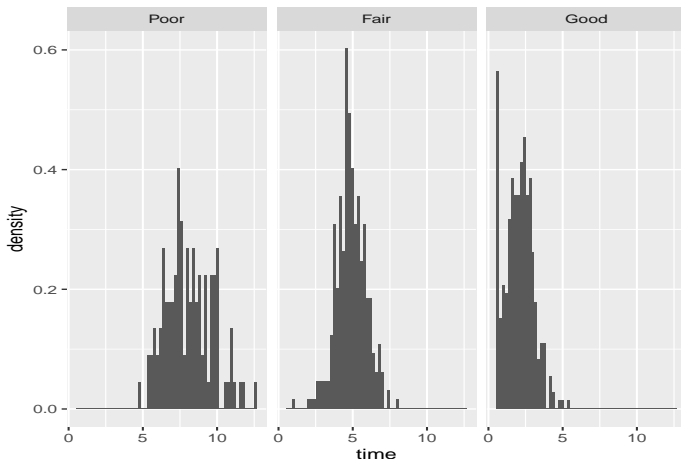
```
g + facet_grid(gender ~ .)
```



## Facet grid

We could also split data by health condition in columns:

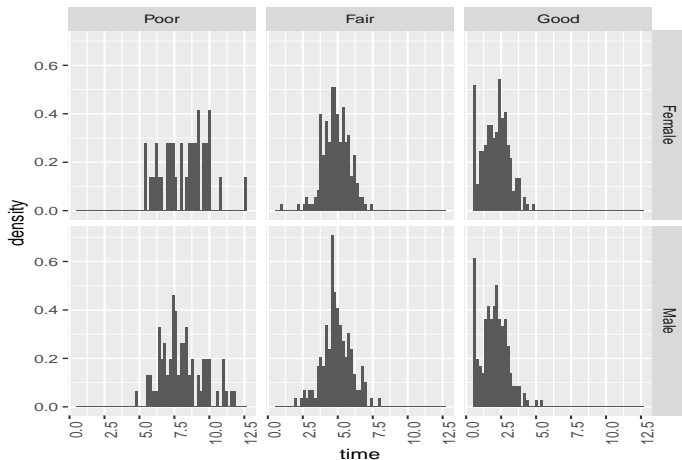
```
g + facet_grid(. ~ health) + scale_x_continuous(breaks =  
c(0, 5, 10))
```



# Facet grid

Both rows and columns can be used to split data by:

```
g + facet_grid(gender ~ health) + theme(axis.text.x =  
  element_text(angle = 90, hjust = 1))
```

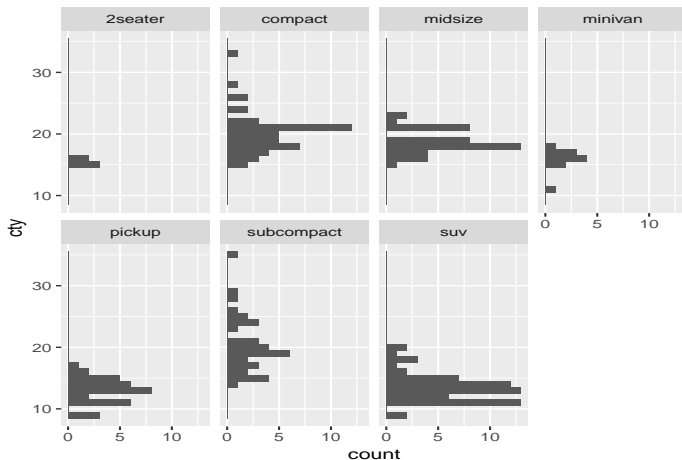




# Facet wrap

The `facet_wrap` function creates a 1d ribbon of plots:

```
ggplot(mpg, aes(x=cty)) + geom_histogram(binwidth=1) +  
facet_wrap(~class, ncol=4) + coord_flip()
```



# Themes

# What are themes?

Themes are used in `ggplot2` to ease making consistent changes to all your plots. For example:

- change the font size;
- change background colour;
- control angels of axis label.
- etc.

## Choice of themes

A few useful standard themes are available in the [ggthemes](#) package which can be loaded using:

```
require(ggthemes)
```

Once loaded, a range of standard themes would be available to control the plot appearance:

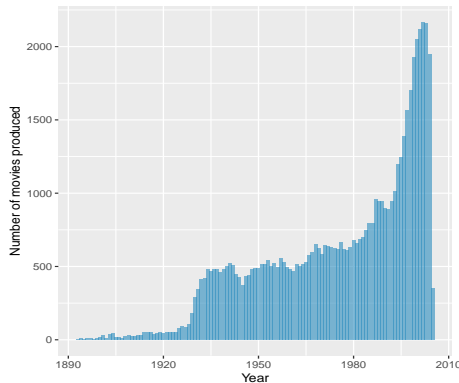
- `theme_bw()`
- `theme_economist()`
- `theme_stata()` - useful for stata users
- `theme_pander()`
- `theme_hc(style = "darkunica")`

# Theme examples

I will use the `movies` data from the `ggplot2movies` to show few examples of standard themes:

```
data(movies, package="ggplot2movies")
```

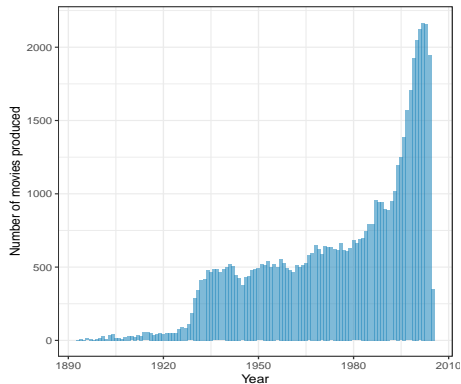
```
(h=  
ggplot(movies, aes(year))+  
geom_histogram(binwidth=1,  
fill="#2b8cbe",alpha=0.6)+  
xlab("Year")+ylab("Number  
of movies produced"))
```



# Theme examples

Using the `theme_bw()`, the appearance can be changed to:

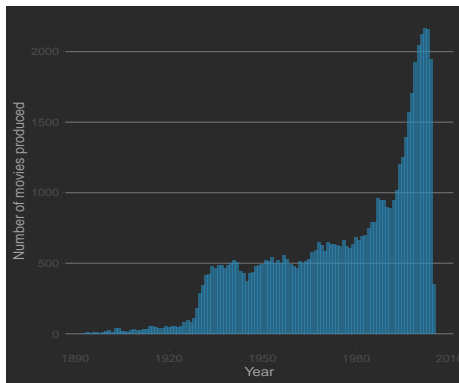
```
h + theme_bw()
```



# Theme examples

Using the `theme_hc(style = "darkunica")`, the appearance can be changed to:

```
h + theme_hc(style =  
"darkunica")
```



# Define our own theme

We can setup our own theme:

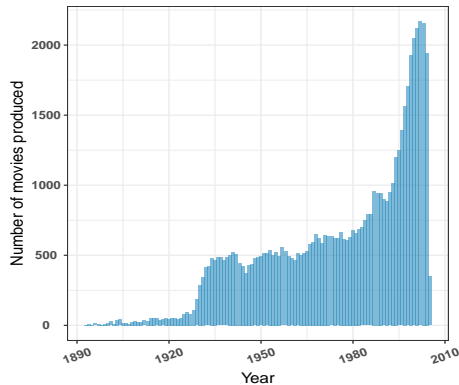
```
My_theme <-  
theme(axis.text=element_text(size=10, face="bold"),  
axis.text.x=element_text(angle=20, vjust=0.5),  
axis.text.y=element_text(vjust=0.5, hjust = 0.5),  
plot.margin = unit(c(0.5,0.5,0.5,0.5), "cm"),  
axis.title=element_text(size = rel(1.2)),  
axis.title.x=element_text(vjust=-0.5),  
axis.title.y=element_text(vjust=1.5))
```



# Define our own theme

Then, we can use it using:

```
h + theme_bw() + My_theme
```



## Useful links

### R Courses

Details and materials of a set of author's courses:

<http://osmahmoud.com/R-courses/>

### Github

Repository of course packages:

<https://statcourses.github.io/>

### Shiny web-page

The RVis tool for learners of data visualisation using R:

<http://bristol-medical-stat.bristol.ac.uk:3838/RVis/>

# Thank You