### **Plots**

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# Slide with R Code and Output

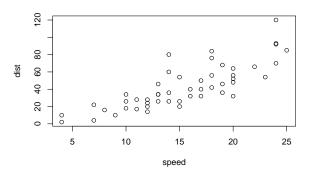
#### summary(cars)

```
##
       speed
                     dist
   Min. : 4.0 Min. : 2.00
##
   1st Qu.:12.0 1st Qu.: 26.00
##
##
   Median: 15.0 Median: 36.00
##
   Mean :15.4
                Mean : 42.98
##
   3rd Qu.:19.0
                 3rd Qu.: 56.00
   Max. :25.0
                Max. :120.00
##
```

#### Slide with Plot

A very simple plot using example data included in R and base R plot functions

plot(cars)



### ggplot2

Nowadays the package most frequently used for *normal* plotting is ggplot2

- it is based on a grammar of graphics
- ▶ in plain words, we assemble plots by adding different *pieces*
- these pieces are quite modular
- this allows making many different types of plots from not so many pieces
- the basic pieces are layers
- the name reflects the fact that order in which we add them is significant
- later additions are plotted on top of the first ones

### what is a ggplot object?

- a ggplot object is independent of the output format
- ▶ it is not purely visual, it is to a significant extent semantic
- ▶ the size and the theme can be decided at the time of generating output
- a ggplot can be modified by adding and sometimes replacing elements
- a ggplot object can be large (in memory) because it contains a copy of the data plotted
- this means that the ggplot object is self-sufficient (you can copy it to a new 'empty' R session and still generate the output)

# What are the words of the grammar

- data
- aesthetics
- statistics
- scales
- geometries

#### data

- data tells R where to look for the variables used in the plot, data is supplied as a data.frame object
- aesthetics (aes) tells how to map variables to features in the plot
- stablishes a connection between a variable an aesthetic dimension
- for example: x coordinate, colour, etc.
- statistics or some operation on the data before plotting \_ scales the mapping between data values and aesthetic values
- e.g. is value 1 by plotted as red and value 2 with say blue, or using some other colours

#### geometries

- they are easier to describe with examples
- geom\_line, geom\_point, geom\_text
- each geom you add to a ggplot adds a new layer to the plot (on top of the layers added earlier)

#### annotations

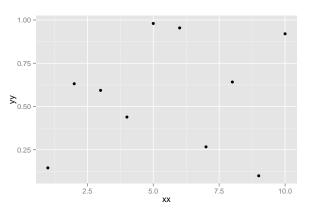
- they add layers that are not directly 'connected' to the data
- e.g. a label not directly related to a single observation
- annotations can also use different geoms and aesthetics, buy they are not inherited, they are selected locally to each annotation

### A very simple plot

```
library(ggplot2)
# runif generates random numbers
# we create a data frame to play with
my.df <- data.frame(xx = 1:10, yy = runif(10))
# we create a plot and save it as 'my.plot'
my.plot <- ggplot(data = my.df, aes(x = xx, y = yy)) +
geom_point()</pre>
```

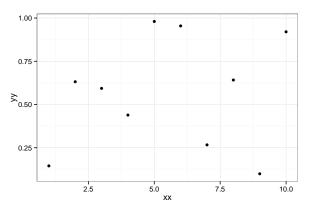
# Printing the plot

### print(my.plot)



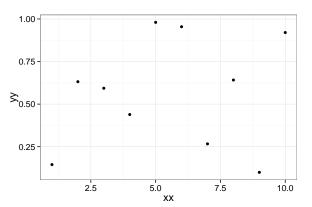
# Printing the plot using a different theme

print(my.plot + theme\_bw())



# Printing the plot using a different base font size

print(my.plot + theme\_bw(15))



# What is in my.plot

```
str(my.plot)
```

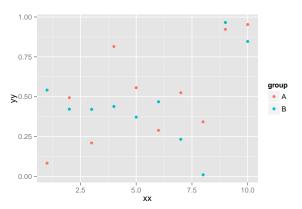
```
## List of 9
   $ data :'data.frame': 10 obs. of 2 variables:
##
     ..$ xx: int [1:10] 1 2 3 4 5 6 7 8 9 10
##
   ..$ yy: num [1:10] 0.145 0.632 0.593 0.439 0.98 ...
##
   $ layers :List of 1
##
##
   ...$ :Classes 'proto', 'environment' <environment: 0x00
##
   $ scales :Reference class 'Scales' [package "ggplo
     ..$ scales: list()
##
##
     .. and 21 methods, of which 9 are possibly relevant:
##
     .. add, clone, find, get_scales, has_scale, initializ
##
     .. non position scales
##
    $ mapping :List of 2
##
     ..$ x: symbol xx
##
     ..$ y: symbol yy
   $ theme : list()
##
                                   4 D > 4 B > 4 E > 4 E > E 990
##
    $ coordinates I ist of 1
```

### A not so simple plot

```
# runif generates random numbers
# we create a data frame to play with
my.2nd.df <-
  data.frame(xx = rep(1:10, 2),
             yy = runif(20),
             group = factor(rep(c("A", "B"), c(10, 10)))
# we create a plot and save it as 'my.plot'
my.2nd.plot <- ggplot(data = my.2nd.df,
                      aes(x = xx, y = yy, colour = group))
  geom_point()
```

# *Printing* the plot

#### print(my.2nd.plot)

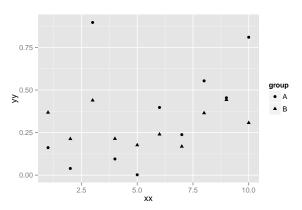


### A not so simple plot

```
# runif generates random numbers
# we create a data frame to play with
my.2nd.df <-
 data.frame(xx = rep(1:10, 2),
             yy = runif(20),
             group = factor(rep(c("A", "B"), c(10, 10)))
# we create a plot and save it as 'my.plot'
my.3rd.plot <- ggplot(data = my.2nd.df,
                      aes(x = xx, y = yy, shape = group))
 geom_point()
```

# *Printing* the plot

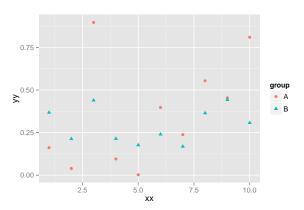
#### print(my.3rd.plot)



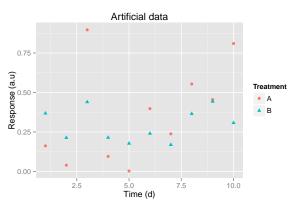
# A plot with two aesthetics for group

# *Printing* the plot

#### print(my.4th.plot)

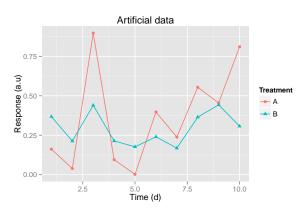


### Replacing the default labels



# Adding another layer on the fly

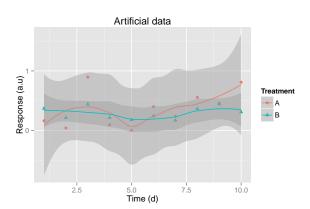
print(my.5th.plot + geom\_line())



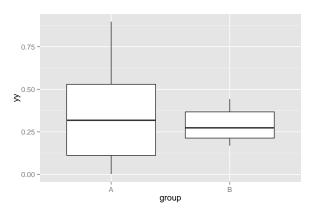
# Adding another layer on the fly

```
print(my.5th.plot + geom_smooth())
```

 $\mbox{\tt \#\#}$  geom\_smooth: method="auto" and size of largest group is

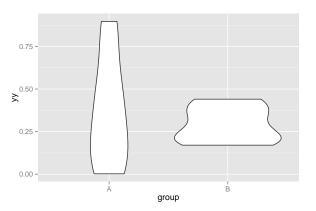


### Another way of plotting the same data

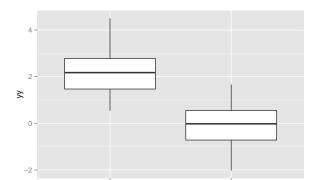


# Another way of plotting the same data

print(my.6th.plot + geom\_violin())



# A larger set of normally distributed data



# The plot with a few additional layers

