# Getting to plot in ${\bf R}$

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IN THE FORSCHUNGSVERBUND BERLIN E.V.

### Plotting in R

- Introduction
- Plotting with traditional graphics
- Plotting with ggplot:
- 4 Conclusion: traditional graphics vs. ggplot2

### Why plot in **R**?

- Powerful (large range of plot types)
- Fully customizable (make your own style)
- Practical (integrate your plots and your code together)

# Graphics paradigms in R

They are three dominant graphics paradigms in R:

- traditional graphics (based on graphics)
- lattice (based on grid)
- ggplot2 (based on grid)

# Graphics paradigms in ${\bf R}$

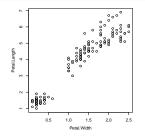
They are three dominant graphics paradigms in R:

- traditional graphics (based on graphics)
- lattice (based on grid)
- ggplot2 (based on grid)

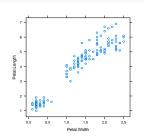
#### Note:

- graphics and grid are part of any basic installation of R
- lattice is part of the so-called list of CRAN recommended packages
- ggplot2 is part of the tidyverse universe (from RStudio)
- we will focus on traditional graphics and ggplot2 (lattice is excellent but a bit more difficult)
- some other packages are sometimes useful too (e.g. rgl, plotly)

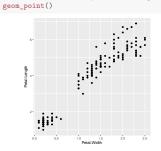
# An example (using default settings)



#### library(lattice)



# library(tidyverse) ## or ggplot2 ggplot(data = iris, aes(x = Petal.Width, y = Petal.Length)) +



# How to learn on your own?

1. Check the examples readily available in R, e.g.

```
demo(graphics)
demo(lange)
demo(persp)
demo(colors)
demo(plotmath)
demo(Hershey)

example(plot)
example(boxplot)
example(boxplot)
example(bartplot)
browseVignettes(package = "ggplot2")
```

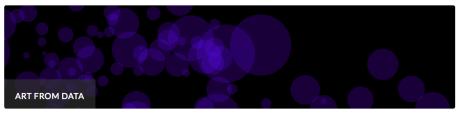
### How to learn on your own?

### 2. Scroll the web:

(e.g. http://www.r-graph-gallery.com/all-graphs/)



HOME GGPLOT2 ALL GRAPHS BLOG ABOUT PYTHON



ALL GRAPHS













If you are looking for something in particular, please use the search tool below. It works even if you are looking for informations concerning an R graph function that is used in the website.

Type an R function, graph type, graph number...

Search ...

# How to learn on your own?

#### 3. Read books:



### Plotting in R

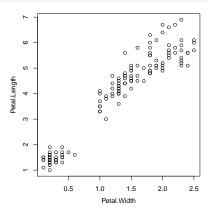
- Plotting with traditional graphics

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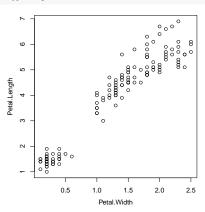
### In traditional graphics, use plot() to draw a scatter plot:

plot(Petal.Length ~ Petal.Width, data = iris)



You can choose what type of scatter plot to display with argument type:

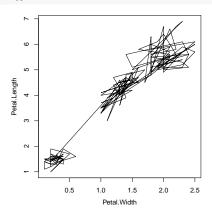
plot(Petal.Length ~ Petal.Width, data = iris, type = "p")



Note: see "?plot.default".

You can choose what type of scatter plot to display with argument type:

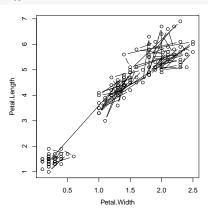
plot(Petal.Length ~ Petal.Width, data = iris, type = "1")



Note: it makes more sense when data are ordered...

You can choose what type of scatter plot to display with argument type:

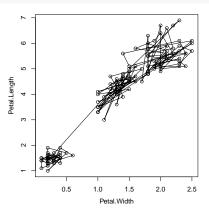
plot(Petal.Length ~ Petal.Width, data = iris, type = "b")



Note: it makes more sense when data are ordered...

You can choose what type of scatter plot to display with argument type:

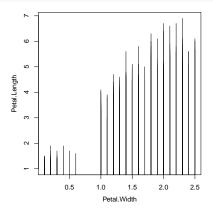
plot(Petal.Length ~ Petal.Width, data = iris, type = "o")



Note: it makes more sense when data are ordered...

You can choose what type of scatter plot to display with argument type:

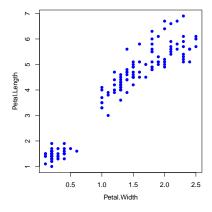
plot(Petal.Length ~ Petal.Width, data = iris, type = "h")



Note: it makes more sense when x-values are unique. . .

You can change point shapes (pch) and colour (col):

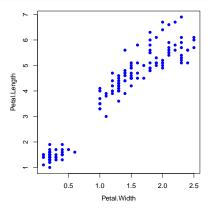
plot(Petal.Length ~ Petal.Width, data = iris, pch = 16, col = "blue")



Note: you can use colour names.

You can change point shapes (pch) and colour (col):

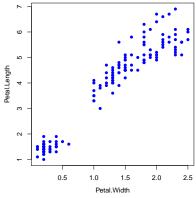
plot(Petal.Length ~ Petal.Width, data = iris, pch = 16, col = 4)



Note: you can use numbers to describe basic colours.

### You can change point shapes (pch) and colour (col):

```
plot(Petal.Length ~ Petal.Width, data = iris, pch = 16, col = "#0000FFFF")
```

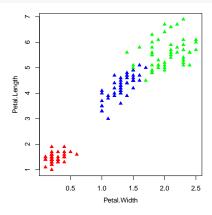


```
rgb(red = 0, green = 0, blue = 255, alpha = 255, maxColorValue = 255)
## [1] "#0000FFFF"
```

Note: you can have full control using hexadecimal!!

You can change point shapes (pch) and colour (col):

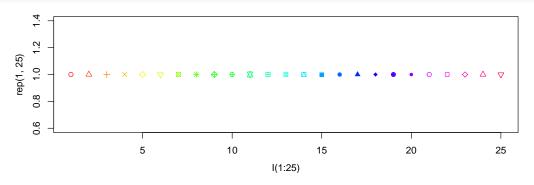
```
palette(c("red", "blue", "green"))
plot(Petal.Length ~ Petal.Width, data = iris, pch = 17, col = iris$Species)
```



Note: you can use a palette to match the levels of a factor.

You can change point shapes (pch) and colour (col):

```
plot(rep(1, 25) ~ I(1:25), data = iris, pch = 1:25, col = rainbow(25))
```



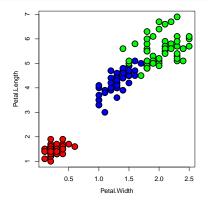
#### Note:

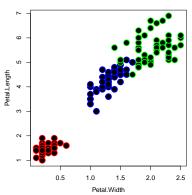
- there are 25 basic symbols (but other ways allow to use many more)
- check "?rainbow" for a list of different color palettes
- the I() allows for the creation of the vector before being interpreted by plot()

For some point shapes you can set both an outline colour (col) and background colour (bg):

```
palette(c("red", "blue", "green"))
plot(Petal.Length - Petal.Width, data = iris, cex = 2,
    pch = 21, col = "black", bg = iris$Species, lwd = 2)
```

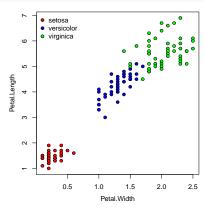
```
palette(c("red", "blue", "green"))
plot(Petal.Length - Petal.Width, data = iris, cex = 2,
    pch = 21, col = iris$Species, bg = "black", lwd = 2)
```





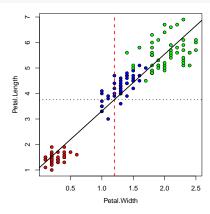
### You can add a legend to the plot:

```
palette(c("red", "blue", "green"))
plot(Petal.Length - Petal.Width, data = iris, pch = 21, bg = iris$Species)
legend(x = "topleft", legend = c("setosa", "versicolor", "virginica"), pch = 21, pt.bg = c("red", "blue", "green"), bty = "n")
```



### You can add lines to the plot:

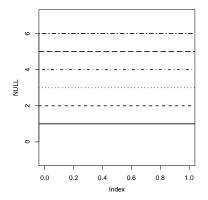
```
palette(c("red", "blue", "green"))
plot(Petal.Length ~ Petal.Width, data = iris, pch = 21, bg = iris$Species)
abline(v = mean(iris$Petal.Width), lty = 2, col = "red", lwd = 2)
abline(h = mean(iris$Petal.Length), lty = 3, col = "black", lwd = 2)
abline(a = 1.084, b = 2.23, lty = 1, lwd = 2)
```

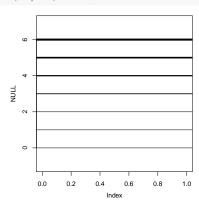


#### You can add lines to the plot:

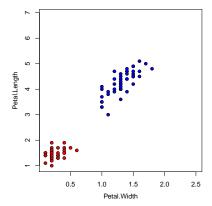
```
plot(NULL, xlim = c(0, 1), ylim = c(-1, 7))
abline(h = 0, lty = 0, lwd = 2)
abline(h = 1, lty = 1, lwd = 2)
abline(h = 2, lty = 2, lwd = 2)
abline(h = 3, lty = 3, lwd = 2)
abline(h = 4, lty = 4, lwd = 2)
abline(h = 5, lty = 5, lwd = 2)
abline(h = 6, lty = 6, lwd = 2)
```

```
plot(NULL, xlim = c(0, 1), ylim = c(-1, 7))
abline(h = 0, lty = 1, lwd = 0.2)
abline(h = 1, lty = 1, lwd = 0.5)
abline(h = 2, lty = 1, lwd = 1)
abline(h = 3, lty = 1, lwd = 2)
abline(h = 4, lty = 1, lwd = 3)
abline(h = 5, lty = 1, lwd = 4)
abline(h = 6, lty = 1, lwd = 5)
```



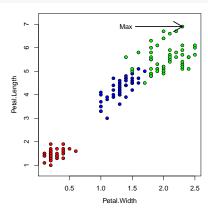


#### You can also add additional points to the plot:

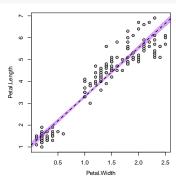


Note: You need to make sure the axis limits are the same!

#### You can add text and arrows:



### Including an interval around a prediction line requires you to build a polygon:



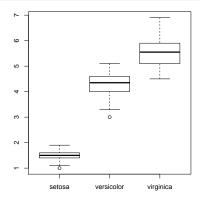
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# Traditional graphics: box plots

### In traditional graphics, use boxplot() to draw a box plot:

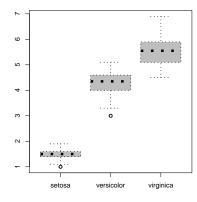
boxplot(Petal.Length ~ Species, data = iris)



# Traditional graphics: box plots

Many of the same changes made to scatter plots can be made here:

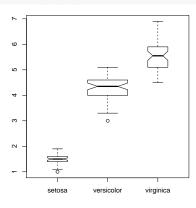
boxplot(Petal.Length ~ Species, data = iris, col = "grey", lwd = 2, lty = 3)



# Traditional graphics: box plots

### There are also some boxplot specific arguments:

boxplot(Petal.Length ~ Species, data = iris, width = c(1, 2, 1), notch = TRUE)



box plots

### Traditional graphics: box plots

### You can retrieve information by storing the output in an object:

```
my_boxcox <- boxplot(Petal.Length ~ Species, data = iris, plot = FALSE)
my_boxcox
## $stats
        [,1] [,2] [,3]
## [1,] 1.1 3.30 4.50
## [2,]
        1.4 4.00 5.10
## [3.] 1.5 4.35 5.55
## [4.] 1.6 4.60 5.90
## [5,] 1.9 5.10 6.90
##
## $n
## [1] 50 50 50
## $conf
            [,1]
                   [,2]
                              [,3]
## [1,] 1.455311 4.215933 5.371243
## [2,] 1.544689 4.484067 5.728757
##
## $out
## [1] 1 3
##
## $group
## [1] 1 2
## $names
## [1] "setosa" "versicolor" "virginica"
```

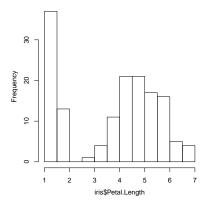
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## Traditional graphics: histograms

### In traditional graphics, use hist() to draw an histogram:

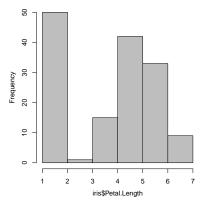
hist(iris\$Petal.Length, main = "") ## main used here to remove the automatic title



# Traditional graphics: histograms

You can change the number and location of breaks between bins:

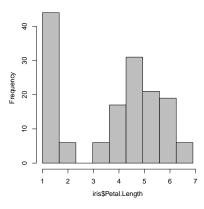
hist(iris\$Petal.Length, main = "", breaks = 5, col = "grey")



# Traditional graphics: histograms

### You can change the number and location of breaks between bins:

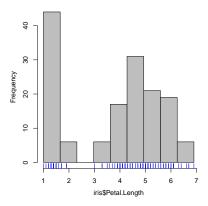
```
hist_breaks <- seq(min(iris$Petal.Length), max(iris$Petal.Length), length.out = 10)
hist(iris$Petal.Length, main = "", breaks = hist_breaks, col = "grey")</pre>
```



# Traditional graphics: histograms

### You can change the number and location of breaks between bins:

```
hist_breaks <- seq(min(iris$Petal.Length), max(iris$Petal.Length), length.out = 10)
hist(iris$Petal.Length, main = "", breaks = hist_breaks, col = "grey")
rug(x = iris$Petal.Length, col = "blue")</pre>
```



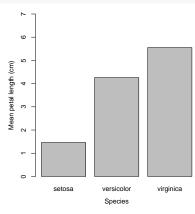
Note: it never hurts to add a rug under an histogram!

## Plotting in **R**

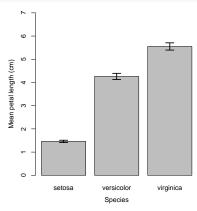
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```
spp_means <- data.frame(Species = c("setosa", "versicolor", "virginica"),</pre>
                        mean = as.numeric(by(iris$Petal.Length, iris$Species, mean)),
                        SE = as.numeric(by(iris$Petal.Length, iris$Species, function(x)sd(x)/sqrt(length(x))))
spp_means
        Species mean
         setosa 1.462 0.02455980
## 2 versicolor 4.260 0.06645545
## 3 virginica 5.552 0.07804970
```

### In traditional graphics, use barplot() to draw a bar plot:

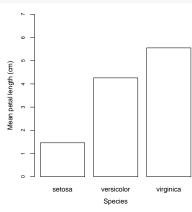


### Adding error bars can be done with the arrows function:



### As before, there are similar arguments available:

```
barplot(height = spp_means$mean, names.arg = spp_means$species, ylim = c(0, 7),
    xlab = "Species", ylab = "Mean petal length (cm)",
    col = "white", cex.axis = 0.75)
```



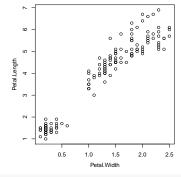
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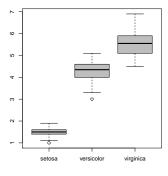
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### Combining plots

You can easily combine multiple traditional graphics using global parameters:

```
oldpar <- par(mfrow = c(1, 2))
plot(Petal.Length - Petal.Width, data = iris)
boxplot(Petal.Length - Species, data = iris, col = "grey")</pre>
```





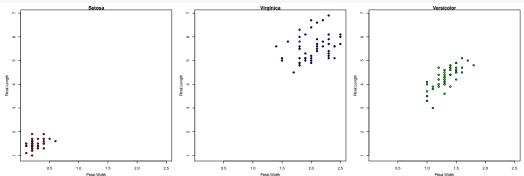
```
par(oldpar) ## restore default values
```

Note: if you want to combine graphics of different sizes it is a bit more complex (see ?layout).

### **Facetting**

#### This can be used to created 'facet' plots:

```
oldpar \leftarrow par(mfrow = c(1, 3))
split_data <- split(iris, iris$Species)</pre>
plot(Petal.Length ~ Petal.Width, data = split data$setosa, main = "Setosa", pch = 21,
     bg = "red", col = "black", xlim = range(iris$Petal.Width), ylim = range(iris$Petal.Length))
plot(Petal.Length ~ Petal.Width, data = split_data$virginica, main = "Virginica", pch = 21,
     bg = "blue", col = "black", xlim = range(iris$Petal.Width), ylim = range(iris$Petal.Length))
plot(Petal.Length ~ Petal.Width, data = split_data$versicolor, main = "Versicolor", pch = 21,
     bg = "green", col = "black", xlim = range(iris$Petal, Width), vlim = range(iris$Petal, Length))
```



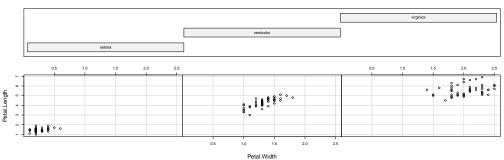
par(oldpar) ## restore default values

## Facetting

#### Note: there is also an easier function for that:

coplot(Petal.Length ~ Petal.Width | Species, data = iris, rows = 1)

Given : Species

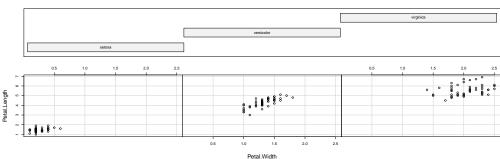


## Facetting

#### Note: there is also an easier function for that:

coplot(Petal.Length ~ Petal.Width | Species, data = iris, rows = 1)

Given : Species

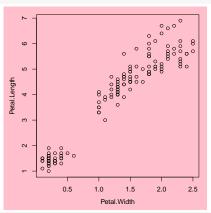


Challenge: have a look at the relationship between the happiness and the price of present using the dataset Surprise.

### Background

Global parameters can also be used to change the background colour:

```
oldpar <- par(bg = "pink")
plot(Petal.Length ~ Petal.Width, data = iris)</pre>
```



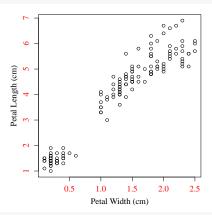
par(oldpar) ## restore default values

Note: you can also plot a background image but you need to use specific packages for that.

# Modifying typefaces

### Global parameters can also be used to change the typeface:

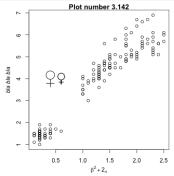
```
oldpar <- par(family = "serif")</pre>
plot(Petal.Length ~ Petal.Width, data = iris,
    xlab = "Petal Width (cm)", ylab = "Petal Length (cm)",
     cex.lab = 1.25, cex.axis = 1.25, col.axis = "red")
```



par(oldpar) ## restore default values

# Special characters

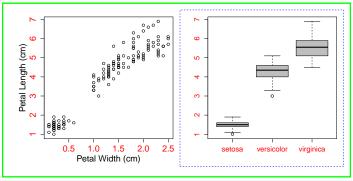
#### You can use weird characters:



Note: unicode characters are nice but source of problems. How to make it work depends on both the OS, graphical device and typefaces installed on your computer.

## Plot margins

### Global parameters can also be used to change individual plot margins (blue) and outer margins (green):

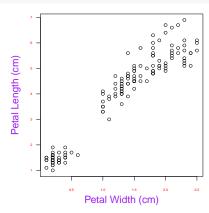


par(oldpar) ## restore default values

# Change axes

You can change the size, colour and orientation of the axis labels and text easily:

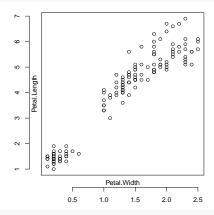
```
plot(Petal.Length ~ Petal.Width, data = iris,
    xlab = "Petal Width (cm)", ylab = "Petal Length (cm)",
    cex.lab = 1.5, cex.axis = 0.5, col.axis = "red", col.lab = "purple", las = 1)
```



## Change axes

### You can change the position of the axes:

```
oldpar <- par(mgp = c(0.4, 2, 1.5))
plot(Petal.Length ~ Petal.Width, data = iris)</pre>
```



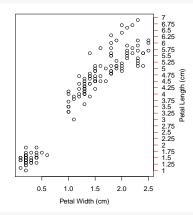
par(oldpar) ## restore default values

Note: check ?par for a huge list of the things you can easily change!

### Change axes

#### You can change the axes themselves:

```
oldpar <- par(mar = c(4, 4, 1, 4))
plot(Petal.Length ~ Petal.Width, data = iris, axes = FALSE,
    xlab = "Petal Width (cm)", ylab = "")
axis(side = 1)
axis(side = 4, at = seq(1, 7, 0.25), labels = seq(1, 7, 0.25), col.ticks = "red", las = 1)
mtext("Petal Length (cm)", side = 4, line = 3)
box()
```



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# Traditional graphics: exporting

```
?pdf ?jpeg ?tiff ?bmp ?postscript
```

```
pdf("base_plot.pdf", width = 15, height = 5)
plot(Petal.Length ~ Petal.Width, data = iris)
dev.off()
```

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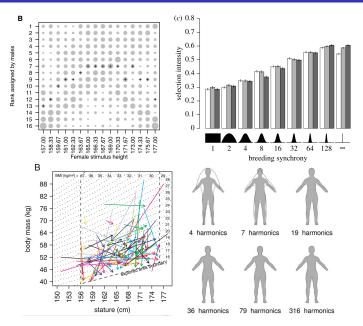
# Challenge

Run the examples of the following traditional plot functions:

- plot.ecdf()
- curve()
- pie()
- cdplot()
- image()
- contour()
- persp()
- o dotchart()
- mosaicplot()
- stars()
- matplot()
- pairs()
- sunflowerplot()

Note: there are a few other ones, but they do not seem very interesting. . .

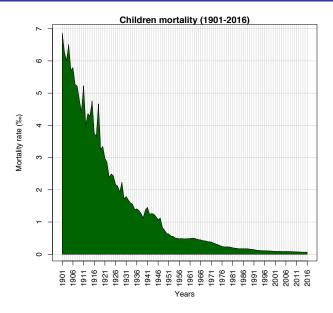
# You can create home-made R graphics too



## Challenge

Use the datasets called population\_UK and deaths\_UK to compute yearly mortality rates for individuals below 15 yrs only and make a really nice plot of the outcome!

## Possible solution



#### Possible solution

```
deaths UK %>%
  filter(Age_cat %in% levels(Age_cat)[1:3]) %>%
  group by (Year) %>%
  summarize(tot Deaths = sum(Deaths)) -> deaths processed
population_UK %>%
  group_by(Year) %>%
  summarize(tot_Pop = sum(Pop)) -> pop_processed
full join(deaths processed, pop processed) %>%
  mutate(rel Deaths = 1000*tot Deaths/tot Pop) -> data processed
par(mgp = c(3.2, 1, 0), mar = c(5, 5, 1, 1))
plot(rel_Deaths ~ Year, data = data_processed, col = NULL, axes = FALSE,
     ylab = "Mortality rate (\u2030)", xlab = "Years",
     main = paste0("Children mortality (", min(data_processed$Year), "-", max(data_processed$Year), ")"))
abline(v = seq(min(data processed$Year), max(data processed$Year), by = 1), col = "lightgrey", lwd = 0.5)
abline(h = 0:ceiling(max(data processed$rel Deaths)), col = "lightgrey", lwd = 0.5)
polygon(x = c(data_processed$Year, rev(x = data_processed$Year)),
        v = c(\text{data processed} \text{srel Deaths. } \text{rep}(x = 0, \text{times} = \text{nrow}(\text{data processed})))
        col = "darkgreen")
axis(side = 1, at = seq(min(data_processed$Year), max(data_processed$Year), by = 5), las = 2)
axis(side = 2)
box()
```

## Plotting in **R**

- Plotting with ggplot2

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## ggplot2: introduction

 ${\tt ggplot (defunct) and ggplot2 were both created by Hadley Wickham (now chief Scientist at Rstudio) during his PhD: {\tt http://had.co.nz/thesis/practical-tools-hadley-wickham.pdf}}$ 

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The idea was to create a grammar of graphics for R.

It is inspired from the seminal Leland Wilkson's book The Grammar of Graphics: "This book [...] presents a unique foundation for producing almost every quantitative graphic found in scientific journals, newspapers, statistical packages, and data visualization systems."

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#### In Wickham's words:

"The grammar of graphics is an answer to a question: what is a statistical graphic? [...] In brief, the grammar tells us that a statistical graphic is a mapping from data to aesthetic attributes (colour, shape, size) of geometric objects (points, lines, bars). The plot may also contain statistical transformations of the data and is drawn on a specific coordinate system. Facetting can be used to generate the same plot for different subsets of the dataset. It is the combination of these independent components that make up a graphic." (Wickham ggplot2 book)

### ggplot2: introduction

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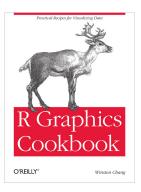
"The grammar of graphics is an answer to a question: what is a statistical graphic? [...] In brief, the grammar tells us that a statistical graphic is a mapping from data to aesthetic attributes (colour, shape, size) of geometric objects (points, lines, bars). The plot may also contain statistical transformations of the data and is drawn on a specific coordinate system. Facetting can be used to generate the same plot for different subsets of the dataset. It is the combination of these independent components that make up a graphic." (Wickham ggplot2 book)

In plain english, that means that the graphics are build by considering building blocs or layers. The originality is that the user directly handles different functions corresponding to each layer and can combine them freely.

## How to learn ggplot2 on your own?

- forum: https://community.rstudio.com
- reference:http://ggplot2.tidyverse.org/reference/
- the RStudio cheatsheet
- the books:



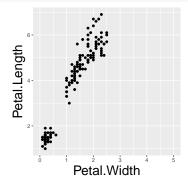


Note: Wickham's books (or their sources) are free and open source!

# ggplot2: introduction

#### Overall structure of a ggplot2 call:

```
ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length)) + ## we specify the source of data and map the variables
geom_point() + ## we use a function that specifies the type of plot (e.g. scatterplot, histogram). A geom_* function is always required!
scale_x_continuous(limits = c(0, 5)) + ## we can fiddle with the scale_* * functions to adjust the axes, colouration etc.
theme(axis.title = element_text(size = 22)) # we can use one of the theme.* functions to change things like legends, font size etc.
```



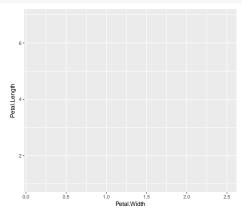
Note: to better understand, execute the code progressively adding one command at a time!

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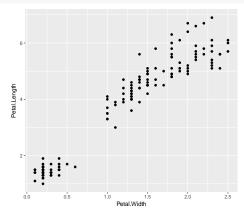
### We start by preparing the data:

```
ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length))
```



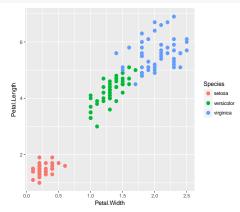
### We add points on the empty plot:

```
ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length)) + geom_point()
```



You can change point shapes shape (= pch in traditional graphics) and colour (= col in traditional graphics):

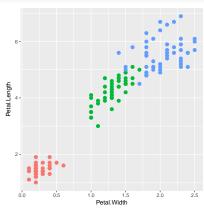
```
ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length, colour = Species)) +
    geom_point(shape = 16, size = 3)
```



Note: everytime we want to create a connection between the data and some elements of the plot, we use the same function: aes()!

#### In ggplot the legend is included by default but you can remove it:

```
ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length, colour = Species)) +
  geom_point(shape = 16, size = 3) +
 theme(legend.position = "none")
```



The aesthetic function in ggplot is a powerful tool for changing plot aesthetics.

If you specify an aesthetic argument inside aes(), it will give each point a different aesthetic based on its value. If you specify the same aesthetic argument outside aes(), it will give all data points the same aesthetic:

```
ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length, ggplot(data = iris,
                                      colour = Species)) +
                                                                                      mapping = aes(x = Petal.Width, y = Petal.Length)) +
  geom point(shape = 16, size = 3)
                                                                                geom point(colour = "blue", shape = 16, size = 3)
           Petal.Length
                                                       Species
                                                       setosa

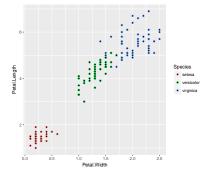
    virginica

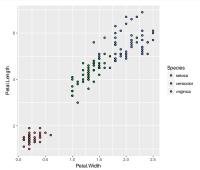
                                                                                                                  Potal Width
```

Note: we have used aes() to apply colours but it can be used to make other changes (see later).

Again, for many elements you can set both an outline colour (= col for traditional graphics) and background fill colour (= bg for traditional graphics):

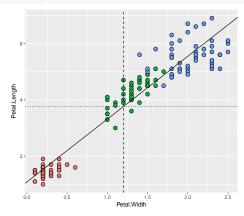
```
ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length, ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length,
                                  colour = Species)) +
                                                                                                        fill = Species)) +
 geom point(fill = "black", shape = 21)
                                                                        geom point(colour = "black", shape = 21)
```



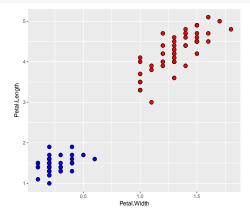


#### You can also add lines to the plot:

```
ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length, fill = Species)) +
  geom_point(shape = 21, size = 3) +
 theme(legend.position = "none") +
  geom_hline(yintercept = mean(x = iris$Petal.Length), lty = 3) +
  geom_vline(xintercept = mean(x = iris$Petal.Width), lty = 2) +
  geom abline(intercept = 1.084, slope = 2.23, lty = 1)
```



You can add extra points to the plot by using two geom\_point layers:

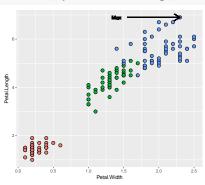


Note 1: unlike traditional plots, ggplot will automatically adjust the axes limits. Note 2: if needed you can define the aes() in the geoms!!

## Traditional graphics: scatter plots

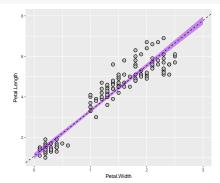
#### You can add text and arrows:

```
max.value <- iris[iris$Petal.Length == max(iris$Petal.Length), ]</pre>
ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length, fill = Species)) +
  geom_point(shape = 21, size = 3) +
  theme(legend.position = "none") +
  geom segment(aes(x = max.value$Petal.Width - 0.75,
                   xend = max.value$Petal.Width.
                   v = max.value$Petal.Length,
                   yend = max.value$Petal.Length), size = 1,
               arrow = arrow(length = unit(0.5, "cm"))) +
  geom text(mapping = aes(x = max.value$Petal.Width - 0.9, v = max.value$Petal.Length, label = "Max"))
```



In ggplot, confidence intervals can be added with specialised function geom\_ribbon():

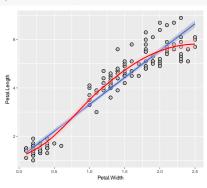
```
test_mod <- lm(Petal.Length ~ Petal.Width, data = iris)
        <- data.frame(Petal.Width = seq(0, 3, length.out = 100))
newdat
        <- predict(object = test mod, newdata = newdat, interval = "confidence")</pre>
pred
ggplot() +
 geom point(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length), fill = "grey", shape = 21, size = 3) +
 geom_ribbon(data = newdat, mapping = aes(x = Petal.Width, ymin = pred[, 2], ymax = pred[, 3]), fill = "purple", alpha = 0.5) +
 geom abline(intercept = coef(test mod)[1], slope = coef(test mod)[2], ltv = 2)
```



Note: because we are building the plot with two different datasets (i.e. iris & newdat), we specify the data and aes() separately in each line of the ggplot code.

#### In ggplot, you can also do basic statistics directly:

```
ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length)) +
  geom_point(fill = "grey", shape = 21, size = 3) +
 geom_smooth(method = "lm") +
  geom_smooth(method = "loess", fill = "pink", colour = "red")
```



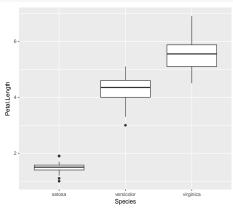
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# ggplot2: box plots

#### In ggplot, use geom\_boxplot() to draw a box plot:

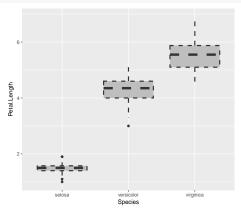
```
ggplot(data = iris, mapping = aes(x = Species, y = Petal.Length)) +
    geom_boxplot()
```



# ggplot2: box plots

#### Many of the same changes made to scatter plots can be made here:

```
ggplot(data = iris, mapping = aes(x = Species, y = Petal.Length)) +
    geom_boxplot(fill = "grey", size = 1, lty = 2)
```



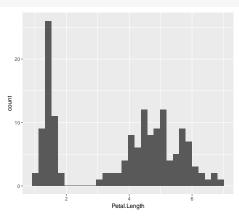
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# ggplot2: histograms

#### In ggplot, use geom\_histogram() to draw an histogram:

```
ggplot(data = iris, mapping = aes(x = Petal.Length)) +
geom_histogram()
## 'stat_bin()' using 'bins = 30'. Pick better value with
# 'binuidth'.
```



# ggplot2: histograms

30 -

10-

You can change the width of bins with binwidth or their number with bins:

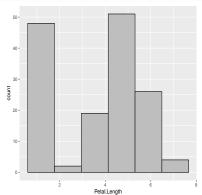
```
ggplot(data = iris, mapping = aes(x = Petal.Length)) +
geom_histogram(binwidth = 0.5, colour = "black", fill = "grey")
```



Petal.Length



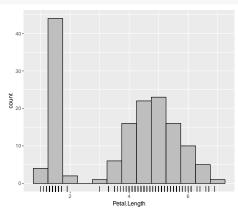




# ggplot2: histograms

#### Add a rug below the histogram with geom\_rug:

```
ggplot(data = iris, mapping = aes(x = Petal.Length)) +
geom_histogram(binwidth = 0.5, colour = "black", fill = "grey") +
geom_rug()
```



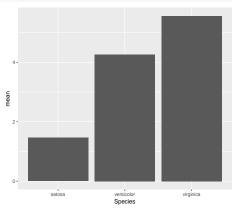
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# ggplot2: bar plots

### In ggplot, use geom\_col() to draw a bar plot:

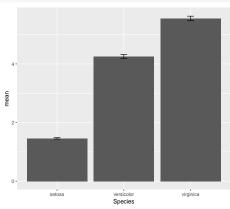
```
ggplot(data = spp_means, mapping = aes(x = Species, y = mean)) +
 geom_col()
```



# ggplot2: bar plots

## Adding error bars in ggplot is much easier than in traditional graphics:

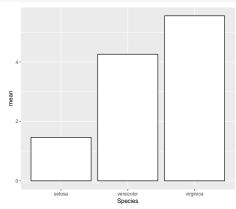
```
ggplot(data = spp_means, mapping = aes(x = Species, y = mean)) +
 geom_col() +
 geom_errorbar(aes(ymin = mean - SE, ymax = mean + SE), width = 0.1)
```



# ggplot2: bar plots

#### As before, modifications are straigthforward:

```
ggplot(data = spp_means, mapping = aes(x = Species, y = mean)) +
 geom_col(fill = "white", colour = "black")
```



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#### Introduction to aesthetics

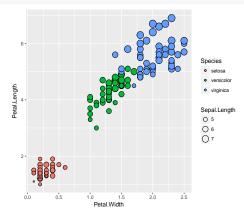
In ggplot, you can change aesthetics in individual segments of the code (aes()) or you can change information for the whole plot using themes (theme()).

aesthetics

# Using the aesthetic function more

We'll start by looking back at aes(). We can use it to change multiple aesthetics of a plot:

```
ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length, fill = Species, size = Sepal.Length)) +
    geom_point(shape = 21)
```

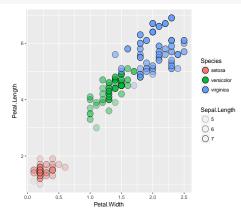


aesthetics

## Using the aesthetic function more

#### Here we change the transparency:

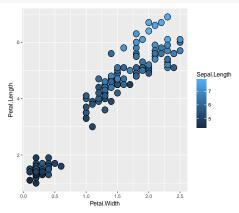
```
ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length, fill = Species, alpha = Sepal.Length)) +
 geom_point(shape = 21, size = 5)
```



# Using the aesthetic argument more

### Applying aesthetics to continuous variables will be different to categorical variables:

```
ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length, fill = Sepal.Length)) +
    geom_point(shape = 21, size = 5)
```

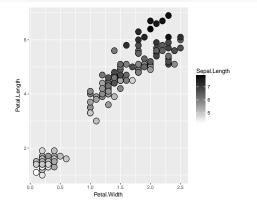


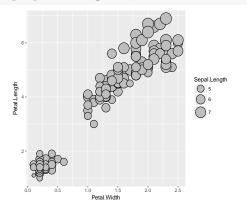
## Using the aesthetic argument more

#### You can indicate values for a third dimension in different ways:

```
fill = Sepal.Length)) +
geom point(shape = 21, size = 5) +
scale_fill_continuous(low = "white", high = "black")
```

```
ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length, ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length,
                                                                                                        size = Sepal.Length)) +
                                                                        geom point(shape = 21, fill = "grev") +
                                                                        scale_size_continuous(range = c(3, 10))
```

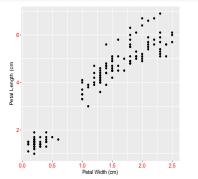




#### Change text

You can change size and colour of axis text easily:

```
ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length)) +
 geom_point() +
 xlab("Petal Width (cm)") +
 ylab("Petal Length (cm") +
 theme(axis.text = element_text(size = 12, colour = "red"), axis.title = element_text(size = 12))
```



Note: the text size uses different measurement units than traditional graphics.

## Combining plots

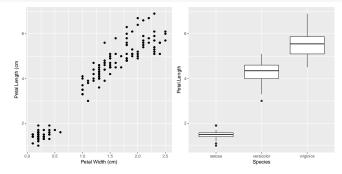
Combining plots is less straightforward in ggplot2...

#### You need to use an additional package:

```
scatter <- ggplot(iris, mapping = aes(x = Petal.Width, y = Petal.Length)) + geom_point() +
xlab("Petal Width (cm)") + ylab("Petal Length (cm")

box <- ggplot(iris, mapping = aes(x = Species, y = Petal.Length)) + geom_boxplot()

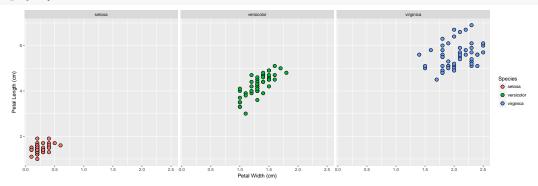
library(gridExtra)
grid.arrange(scatter, box, nrow = 1)</pre>
```



### Faceting

Although combining multiple plots is cumbersome, there is an inbuilt function to create facets:

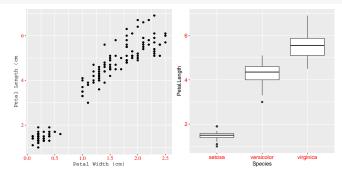
```
ggplot(data = iris, mapping = aes(x = Petal.Width, y = Petal.Length, fill = Species)) +
geom_point(shape = 21, colour = "black", size = 3) +
xlab("Petal Width (cm)") +
ylab("Petal Length (cm)") +
facet_wrap(~ Species)
```



## Modifying typefaces

Unlike traditional graphics, in ggplot2 you can easily change the typeface of individual elements in theme:

aesthetics



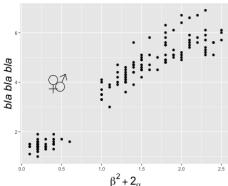
# Special characters

#### You can use weird characters:

```
foo <- round(pi, 3)
ggplot(iris, mapping = aes(x = Petal.Width, y = Petal.Length)) +
  geom_point() +
  labs(x = expression(beta^2 + 2[alpha]), y = expression(paste(italic("bla bla bla"))), title = paste("Plot number", foo)) +
  annotate("text", x = c(0.4, 0.5), y = c(4, 4), label = c("\u2640", "\u2642"), size = 12) + ## requires font with unicode installed (try it
  theme(plot.title = element_text(hjust = 0.5, size = 20),
       axis.title = element text(size = 20))
```

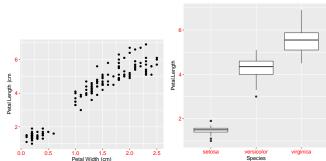
aesthetics





### Plot margins

#### Plot margins are also controlled in theme of each plot individually:

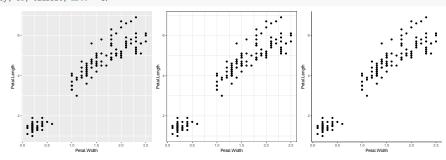


#### Preset themes

#### ggplot2 also has a number of preset themes that you can use:

```
grey <- ggplot(data = iris, aes(x = Petal.Width, y = Petal.Length)) +</pre>
  geom point() +
 theme grey()
bw <- ggplot(data = iris, aes(x = Petal.Width, y = Petal.Length)) +</pre>
  geom point() +
 theme_bw()
classic <- ggplot(data = iris, aes(x = Petal.Width, y = Petal.Length)) +
  geom_point() +
 theme_classic()
```

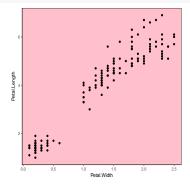
grid.arrange(grey, bw, classic, nrow = 1)



#### Preset themes

#### You can create your own theme once for all!

```
theme_pink <- theme_classic() %+replace% theme(panel.background = element_rect(fill = "pink"))
ggplot(data = iris, aes(x = Petal.Width, y = Petal.Length)) +
    geom_point() +
    theme_pink</pre>
```

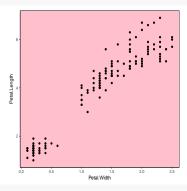


Note: check ?theme for the list of options that can be changed.

#### Preset themes

#### You can set a theme for all plots with theme\_set

```
theme_old <- theme_set(theme_pink)
ggplot(data = iris, aes(x = Petal.Width, y = Petal.Length)) +
    geom_point()</pre>
```



theme\_set(theme\_old) ## restore original theme

#### Plotting in **R**

- Introduction
- Plotting with traditional graphics
  - scatter plot
  - box plots
  - histograms
  - bar plots
  - aesthetics
  - saving your plot
  - other plots
- Plotting with ggplot2
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# ggplot2: Exporting

```
?ggsave
classic <- ggplot(iris, aes(x = Petal.Width, y = Petal.Length)) +</pre>
 geom_point() +
 theme_classic()
ggsave("ggplot.pdf", plot = classic, width = 15, height = 5)
```

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# There are other geom out there!

There are a few other plot types in ggplot2 that are worth looking at (check the cheatsheet!):

- geom\_violin()
- geom\_area()
- geom\_smooth()
- geom\_density()

# There are other geom out there!

There are a few other plot types in ggplot2 that are worth looking at (check the cheatsheet!):

- geom\_violin()
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- geom\_density()

But the true force of having developed something as modular as ggplot2 is that there are many more packages each week providing ggplot2 extensions!

A few random examples of ggplot2 extensions:

- ggExtra (for marginal distribution)
- ggthemes (for extra themes, including theme\_excel()!! :-/)
- ggmap (for maps)
- ggrepel (for adding labels to plots)
- ggalt (for creating cluster plots)
- cowplot (for creating nested figures)

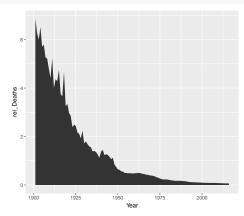
Note: see also http://www.ggplot2-exts.org/gallery/ (but the list is incomplete).

## Challenge

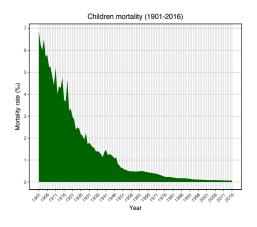
Use the datasets called population\_UK and deaths\_UK to compute yearly mortality rates for individuals below 15 yrs only and make a nice plot of the outcome using ggplot2!

# Simple solution

```
ggplot(data = data_processed, mapping = aes(y = rel_Deaths, x = Year)) +
    geom_area()
```



# More complex solution



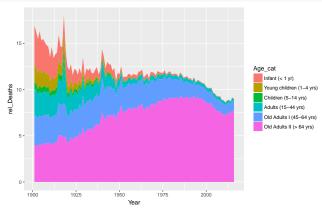
## More complex solution

### Challenge

Use the datasets called population\_UK and deaths\_UK to compute yearly mortality rates for individuals of each age class and make a quick plot of the outcome using ggplot2!

#### Possible solution

```
deaths_UK %>% group_by(Year, Age_cat) %>% summarize(tot_Deaths = sum(Deaths)) -> deaths_processed_all
population UK %>% group by(Year) %>% summarize(tot Pop = sum(Pop)) -> pop processed## as before
full_join(deaths_processed_all, pop_processed) %>% mutate(rel_Deaths = 1000*tot_Deaths/tot_Pop) -> data_processed_all
## Joining, by = "Year"
ggplot(data = data_processed_all, mapping = aes(y = rel_Deaths, x = Year, fill = Age_cat)) +
  geom_area()
```



# Plotting in R

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- Plotting with ggplot2
- 4 Conclusion: traditional graphics vs. ggplot2

## Which plotting tool should you use?

#### Traditional graphics:

- no new packages required
- arguments often differ between functions
- help files useful
- easier for doing something simple
- more difficult for doing something complex
- main restriction: always add things on top of what has already been plotted

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### ggplot2:

- requires multiple packages for best results
- uniform grammar
- help files often useless (due to modularity)
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- easier for doing something complex
- faster for large datasets
- more accessible than lattice

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 $\rightarrow$  It may be useful to know both!?