

Introduction to Machine Learning

Graphical visualisation with ggplot2

- **Geom functions are used to display the data in a multitude of ways**
 - `geom_point()`: scatterplot
 - `geom_smooth()`: smoothed conditional means
 - `geom_histogram()`: histogram
 - `geom_density()`: smoothed variant of the histogram
 - `geom_bar()` / `geom_col()`: bars with bases on the x-axis
 - `geom_line()`: lines
 - `geom_boxplot()`: boxes-and-whiskers
- **Geoms differ in the aesthetics they require:**
 - `geom_point()` requires both x and y values
 - `geom_histogram()` only requires x values
- On the following slides we will give examples for each of the new ones

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Graphical visualisation with ggplot2

- **geom_histogram()**: histogram
- It is used to plot the distribution of a single, continuous variable

```
>ggplot(data = plantData, aes(x = Sepal.Length)) + geom_histogram()
```

```
# important parameter: binwidth: change the width of bins
```

```
>ggplot(data = plantData, aes(x = Sepal.Length)) + geom_histogram(binwidth = 2)
```

```
# aes(fill = ...): color the bars according to their groups
```

```
> bplot= ggplot(data = plantData, aes(x = Sepal.Length)) + geom_histogram(binwidth = 2, aes(fill = Species))
```

```
# scale_fill_manual(): change the color manually,
```

```
>bplot + scale_fill_manual(values=c("#999999", "#E69F00", "#56B4E9"))
```

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Graphical visualisation with ggplot2

- **geom_density()**: smoothed variant of the histogram
- It is used to plot the distribution of a single, continuous variable

```
>ggplot(data = plantData, aes(x = Sepal.Length) ) + geom_density()
```

aes(color = ...): used to create multiple density curves

```
>ggplot(data = plantData, aes(x = Sepal.Length )) +geom_density(aes(color=Species))
```

#**aes(fill = ...)**: is not usefull due to overlapping regions

```
>ggplot(data = plantData, aes(x = Sepal.Length )) + geom_density(aes(fill=Species))
```

scale_color_manual(): change the color manully of lines and points,

```
>ggplot(data = plantData, aes(x = Sepal.Length )) +geom_density(aes(color=Species))  
+ scale_color_manual(values=c("#999999", "#E69F00", "#56B4E9"))
```

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Graphical visualisation with ggplot2

- **geom_bar()**: bars with bases on the x-axis
 - It is used to plot the distribution of a single, discrete/categorical variable

```
>ggplot(data = plantData, aes(x = Species)) + geom_bar()
```



```
# aes(fill = ...): color the bars according to their groups
```

```
# color the bars using this parameter
```

```
>ggplot(data = plantData, aes(x = Species) ) + geom_bar(aes(fill=Species))
```



```
# create a pie chart from a bar chart with a single bar using coord_polar()
```

```
>ggplot(data = plantData, aes(x = "", fill= Species)) + geom_bar()+ coord_polar(theta = "y")
```

```
# x is assigned nothing so that we get a single bar divided into the different species
```

```
# theta indicates the variable to map the angle to (x or y)
```
- If you want the heights of bars to represent values in the data, use **geom_col()** instead.

```
>ggplot(data = plantData, aes(x = Species, y = Petal.Length)) + geom_col()
```

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Exercises with R: Graphical visualisation with ggplot2

Exercise:

- create random subsets with 75 sample size from plantData without replacement
- draw the graphics bar plot, pie chart and histogram again.

```
randomNumbers= sample(1:nrow(plantData), 75, replace=F)
```

```
newData= plantData[randomNumbers, ]
```

```
ggplot(data = newData, aes(x = Species)) + geom_bar(aes(fill=Species))
```

```
ggplot(data = newData, aes(x = "", fill= Species)) + geom_bar()+ coord_polar(theta = "y")
```

```
ggplot(data = newData, aes(x = Sepal.Length)) + geom_histogram(binwidth = 1, aes(fill=Species) )
```

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Exercises with R: Graphical visualisation with ggplot2

Exercise: Write a **function** that should create random subsets with different sample size from plantData with and without replacement. Further, draw the graphic (scatterplot, histogram or pie chart) selected by the user.

```
randomSubsets=function(dataSet, sampleSize, replacement, graphic){  
  randomNumbers= sample(1:nrow(dataSet), sampleSize, replace=replacement)  
  newData= dataSet[randomNumbers, ]  
  
  if(graphic=="scatterplot"){  
    ggplot(data = newData, aes(x = Sepal.Width,y = Sepal.Length)) + geom_point(aes(colour = Species))  
  }  
  else if (graphic=="histogram"){  
    ggplot(data = newData, aes(x = Sepal.Length)) + geom_histogram(binwidth = 1, aes(fill=Species) )  
  }  
  else if (graphic=="piechart"){  
    ggplot(data = newData, aes(x = "", fill= Species)) + geom_bar()+ coord_polar(theta = "y")  
  }  
}  
randomSubsets(plantData, 75,FALSE, "piechart")  
randomSubsets(plantData, 100, TRUE, "histogram")
```

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Exercises with R: Graphical visualisation with ggplot2

- **geom_line()**: lines
- It is used to plot the distribution of two variables representing points
- We don't have an index variable in the dataset so we simulate one using the function **seq_along(vector)**, which creates the integer sequence 1, 2, ..., vector.length

```
>seq_along(c(4,7,12)) #[1] 1 2 3
```

```
>ggplot(data = plantData, aes(x = seq_along(Sepal.Length),y = Sepal.Length)) +  
  geom_line() # A single line is created connecting all points
```

```
# aes(group = ...): group the points
```

```
>ggplot(data = plantData, aes(x = seq_along(Sepal.Length),y = Sepal.Length)) +  
  geom_line(aes(group = Species))
```

```
#aes(color = ...): colors each line differently
```

```
#aes(linetype = ...): changes the type of the line for each group
```

```
>ggplot(data = plantData, aes(x = seq_along(Sepal.Length), y=Sepal.Length)) +  
  geom_line(aes(group=Species, color=Species, linetype=Species))
```

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Exercises with R: Graphical visualisation with ggplot2

- **geom_boxplot()**: boxes-and-whiskers
- It is used to plot the distribution of a variable grouped by another one and is useful for showing outliers and differences between the groups

```
>ggplot(data = plantData, aes(x = Species, y = Sepal.Length)) + geom_boxplot()
```

```
# aes(fill = ...): and aes(color = ...) : color the boxes based on their group
```

```
>ggplot(data = plantData, aes(x = Species, y = Sepal.Length)) + geom_boxplot(aes(fill = Species))
```

```
# outlier.colour, shape, and size: highlight the outlier
```

```
>boxPlot= ggplot(data = plantData, aes(x = Species, y=Sepal.Length))
```

```
>prm1=geom_boxplot(aes(fill=Species), outlier.colour="red", outlier.shape=8, outlier.size=4)
```

```
>boxPlot+prm1
```

```
#rotate boxplot
```

```
>boxPlot+prm1 + coord_flip()
```

```
#Change the legend position
```

```
boxPlot+prm1 + theme(legend.position="top") # position-> top, bottom, none
```


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Exercises with R: Graphical visualisation with ggplot2

- **Exercise:** boxplots for the combination of two variables
 - Create a vector which contains the variables “**small**” and “**medium**” for plant sizes
 - Using **sample** function and **plant size vector**, create a **new vector with 150 random variables**
 - Insert this vector into the **plantData** data frame as a new column **size**
 - Using **interaction** function, combine the variables size and species in plantData and save it in a new column
 - Create a boxplot with respect to the combined variables and Sepal.Length of the plants

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Exercises with R: Graphical visualisation with ggplot2

- **Exercise:** boxplots for the combination of two variables
 - Create a vector which contains the size of plants as “small” and “medium”
`>size=c("small", "medium")`
 - Using **sample** function and **size of plants**, create a **new vector with 150** elements
`>rndSample=sample(size, 150, replace = T)`
 - Insert this vector into the **plantData** data frame as a new column **size**
`>plantData$size=rndSample`
 - Using **interaction** function, combine the variables size and Species in plantData and save it in a new column
`>plantData$comb= interaction(plantData$Species, plantData$size)`
 - Create a boxplot with respect to the combined variables and Sepal.Length of the plants
`>ggplot(data = plantData, aes(x = comb, y=Sepal.Length)) + geom_boxplot(aes(fill=Species))`

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Exercises with R: Graphical visualisation with ggplot2

- Additional functions for changing labels and axes
 - **lims()**: This function allows us to restrict the range of the axes to certain values
*>ggplot(data = plantData, aes(x = seq_along(Sepal.Length), y = Sepal.Length, shape = Species)) +
geom_point() + lims(x = c(50,120), y = c(5,7))*
 - **labs()**: Give labels to the axes and also allows to add title, subtitle and caption
*>ggplot(data = plantData, aes(x = seq_along(Sepal.Length), y = Sepal.Length, shape = Species)) +
geom_point() + labs(x = "X", y = "Y", title = "Title", subtitle = "Sub", caption = "Caption")*
 - Remove legends added by for example **aes(fill = ...)** with **guides()**
*>ggplot(data = plantData, aes(x = Species, y = Sepal.Length)) +
geom_boxplot(aes(fill = Species)) + guides(fill = "none")*

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Exercises with R: Graphical visualisation with ggplot2

- **Faceting** allows to split the plot into multiple small plots

#Facet_grid(): split the plot into multiple small plots

for the row-splitting -> put the variable before ~,

```
>ggplot(data = plantData, aes(x = comb, y=Sepal.Length)) + geom_boxplot(aes(fill=Species))  
+facet_grid(Species~.)
```

for the column-splitting -> put the variable after ~

```
>ggplot(data=plantData, aes(x=size, y=Sepal.Length)) + geom_boxplot(aes(fill=Species))  
+facet_grid(~Species)
```

- **Save the Plots**

```
>p = ggplot(data = plantData, aes(x = Species, y = Sepal.Length)) + geom_boxplot() #assign it to a variable  
>ggsave("myplot.jpg", plot = p) # ggsave(): save a plot to a file
```

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Exercises with R: Graphical visualisation with ggplot2

Exercise: Write a **function** that should create random subsets with different sample size from plantData with replacement. Further,

- It should normalize the **Petal.Width** and **Petal.Length** values with min-max normalization

$$x_i^{norm} = \frac{x_i - \min(x)}{\max(x) - \min(x)}$$

- It should draw multiple graphics on a page
 - You will need “**gridExtra**” package and **grid.arrange** function
 - **Graphic 1:** a scatter plot & use **red**, **black** and **orange** colors for the points according to species & **shape** the points of each species (the function is **shape** used with **aes**)
 - **Graphic 2:** a histogram for Petal.Length & use **red**, **black** and **orange** colors for bars according to species
 - **Graphic 3:** a density plot for Petal.Length with **red**, **black** and **orange** colors
 - **Graphic 4:** a pie chart for plant sizes with **red**, and **orange** colors
 - **Graphic 5 -6:** box plots with **red**, **black** and **orange** colors according to
 - plant sizes and Petal.Width as well as Petal.Length
 - split them in columns according to Species and highlight the outliers
- Save the graphics in a png file, its name is given by the user