

# **Second Assignment**

## **Machine Learning Models**

### **Titanic**

- 1. Cleaning the data**
- 2. Making the best tree by Hyperparameter Selection**
- 3. Making the best random forest by Hyperparameter Selection**
- 4. Conclusion of the questions of the first Assignment**
- 5. Conclusion**

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# 1. Cleaning the data

Survived
TRUE
FALSE
FALSE
TRUE

We replace the variable survived filled with (0, 1), with True False or “SURVIVED” and “DIED”

```
#We change the variable embarked :
titanic.train$Embarked= factor(titanic.train$Embarked,
                                levels=c("C", "S", "Q"),
                                labels=c(1,2,3))
titanic.train$Embarked = as.numeric(titanic.train$Embarked)

#We do the same for the Sex:
titanic.train$Sex = factor(titanic.train$Sex,
                           levels = c("male", "female"),
                           labels= c(1,2))
```

We change the **non-numeric** variables in numeric

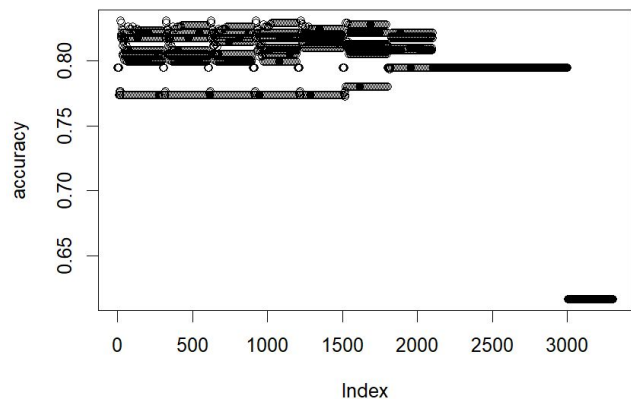
```
titanic.train[, "Ticket"] = NULL
titanic.train[, "Cabin"] = NULL
```

Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
TRUE	1	2	38.0	1	0	71.2833	1
FALSE	3	1	35.0	0	0	8.0500	2
FALSE	1	1	54.0	0	0	51.8625	2
TRUE	2	2	14.0	1	0	30.0708	1
TRUE	3	2	4.0	1	1	16.7000	2
FALSE	3	1	39.0	1	5	31.2750	2

- The change of the variable Sex is not necessary.
- The variable **Ticket** doesn't seem to be relevant
- We decide not to use the variable **Cabin**, as in the First Assignment
- We don't replace, by the mean, the values for **Fare = 0**.
- We **standardize** to see if the accuracy is the same, but **we don't use it**.

## 2. Making the best tree by Hyperparameter Selection

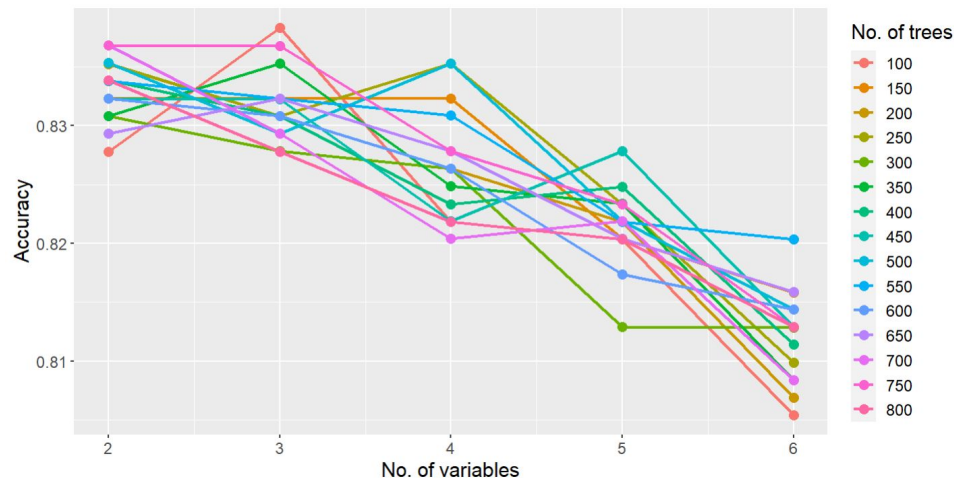
```
d_minsplit=seq(from=10, to= 100,by=10)
d_maxdepth = seq(from=1, to = 30, by = 1)
d_cp = 2^(-11:-1)
paramet= expand.grid(d_minsplit, d_maxdepth, d_cp)
```



- Find the **accuracy**, of each parameter thanks to the **k-fold cross validation**.
- Find the values of the **best hyper parameters**.
- Create the tree and plot it

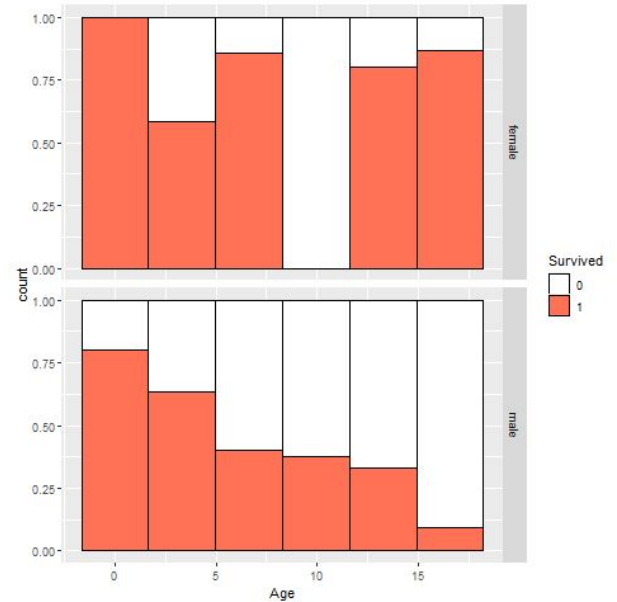
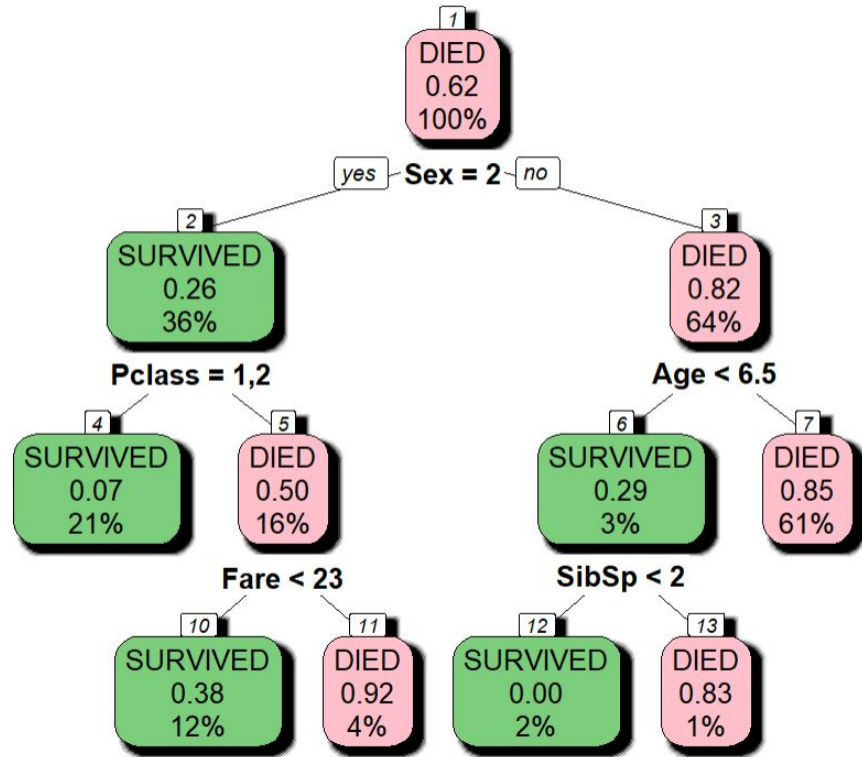
## 3. Making the best random forest by Hyperparameter Selection

```
d_mtry=seq(from=2, to=6, by=1)
d_ntree=seq(from=100, to=800, by=50)
parameters = expand.grid(mtry=d_mtry, ntree=d_ntree)
```



- Create the random forest and find the one with the best accuracy
- Find the hyper parameters for the best random forest
- Create the forest and the **model**.

## 4. Conclusion of the questions of the first Assignment



-Relationship between **survival**, **gender** and **minors**

-Relationship between **survival** and the **Family size**.

-Relationship between **money** and **Survival**.

-The **stereotype** of a **Survival** and the **stereotype** of a **non Survival**.

## 5. Conclusion

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### PROS OF MACHINE LEARNING MODELS

- Helpful to **predict** values
- Shows relations between variables that are difficult to find
- Create a **hierarchy** between the variables by the order of influence

### CONS OF MACHINE LEARNING MODELS

- Sometimes difficult to visualise the meaning of some relation between variables.
- Some calculus are very **time consuming**, specialty for random forest.

### OUR CONCLUSION

-Mix **Exploratory Data Analysis** and **Machine Learning Models** is the best way to find clear and relevant information.