EXERCISES - RANDOM FORESTS

Jan-Philipp Kolb

27 Mai, 2019

EXERCISE: RANDOM FORESTS

DOWNLOAD AND IMPORT EXAMPLE DATA

- Download the Data from here and read in the adult.csv file with header=FALSE. Store this in df. Use str() command to see the dataframe.
- 2) Get the column names from the **meta data** and add them to the data frame. Notice the df is ordered V1,V2,V3,... and so on.

GET AN OVERVIEW OF THE DATA

- 3) Use the table command to get the distribution of the class feature.
- Make a binary variable class.
- 5) Use the cor() command to see the corelation of all the numeric and integer columns including the class column.

SOLUTION: DOWNLOAD AND IMPORT (I) EXERCISE 1

12 <-"/uploads/2016/11/adult.csv"

 $link \leftarrow paste0(11,12)$

11 <- "http://www.r-exercises.com/wp-content"

```
df <- read.csv(link.header=FALSE)</pre>
str(df)
## 'data.frame': 15916 obs. of 15 variables:
##
   $ V1: int 39 50 38 53 28 37 49 52 31 42 ...
##
   $ V2 : Factor w/ 9 levels " ?", "Federal-gov", ...: 8 7 5 5
##
   $ V3: int 77516 83311 215646 234721 338409 284582 160187 2
   $ V4 : Factor w/ 16 levels " 10th", " 11th", ...: 10 10 12 2 10
##
##
   $ V5: int 13 13 9 7 13 14 5 9 14 13 ...
   $ V6 : Factor w/ 7 levels " Divorced", " Married-AF-spouse",.
##
   $ V7 : Factor w/ 15 levels " ?", " Adm-clerical", ...: 2 5 7 7
##
   $ V8 : Factor w/ 6 levels " Husband", " Not-in-family", ...: 2
##
    $ V9 : Factor w/ 5 levels " Amer-Indian-Eskimo",..: 5 5 5 3
##
    $ V10: Factor w/ 2 levels " Female". " Male": 2 2 2 2 1 1 1 2
```

SOLUTION: DOWNLOAD AND IMPORT (II)

Exercise 2

SOLUTION: GET OVERVIEW

```
Exercise 3
table(df$class)
##
    <=50K
             >50K
##
##
    12097
             3819
Exercise 4
df$class <- ifelse(df$class==" >50K", 1, 0)
Exercise 5
\operatorname{cor}(\operatorname{df}[,c(1,3,5,11,12,13,15)])
##
                                        fnlwgt education-num capital
                             age
## age
```

age 1.00000000 -0.079506361 0.02668698 0.0664 ## fnlwgt -0.07950636 1.000000000 -0.04671504 0.0006 ## education-num 0.02668698 -0.046715043 1.00000000 0.1174 ## capital-gain 0.06646649 0.000653693 0.11745307 1.0000

EXERCISE: RANDOM FORESTS SPLIT THE DATASET

- 6) Split the dataset into Train and Test sample. You may use caTools::sample.split() and use the ratio as 0.7 and set the seed to be 1000.
- 7) Check the number of rows of Train and Test
- 8) We are ready to use decision tree in our dataset. Load the package rpart and rpart.plot
- Use rpart to build the decision tree on the Train set. Include all features. Store this model in dec
- 10) Use prp() to plot the decision tree.

```
6
set.seed(1000)
library(caTools)
split=sample.split(df$class, SplitRatio=0.8)
Train=df[split==TRUE,]
Test=df[split==FALSE,]
```

EXERCISE

- use the predict() command to make predictions on the Train data. Set the method to class. Class returns classifications instead of probability scores. Store this prediction in pred_dec.
- 2) Print out the confusion matrix
- 3) What is the accuracy of the model. Use the confusion matrix.
- 4) What is the misclassification error rate? Refer to Basic_decision_tree exercise to get the formula.
- 5) Lets say we want to find the baseline model to compare our prediction improvement. We create a base model using this code

```
length(Test$class)
## [1] 3183
base=rep(1,3183)
```

- Use the table() command to create a confusion matrix between the base and Test\$class.
- 6) What is the number difference between the confusion matrix accuracy

SOLUTION

```
library(caTools)
colnames(df)=c("age", "workclass", "fnlwgt", "education", "education")
df$class=ifelse(df$class==" >50K", 1, 0)
df$class=as.factor(df$class)
set.seed(1000)
split=sample.split(df$class, SplitRatio=0.8)
Train=df [split==TRUE,]
Test=df[split==FALSE,]
library(rpart)
library(rpart.plot)
dec=rpart(class~., data=Train)
par(mar = rep(2, 4))
1)
pred dec=predict(dec,newdata = Test,type="class")
```

JAN-I