Machine Learning - Solution/Exercises

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EXERCISE - RPART KYPHOSIS

Consider the Kyphosis data frame

- 1. Which variables are in the kyphosis dataset
- 2. Build a tree to classify Kyphosis from Age, Number and Start.

CONSIDER THE TREE BUILD ABOVE.

- 3. Which variables are used to explain Kyphosis presence?
- 4. How many observations contain the terminal nodes.

Consider the Kyphosis data frame.

- 5. Build a tree using the first 60 observations of kyphosis.
- 6. Predict the kyphosis presence for the other 21 observations.
- 7. Which is the misclassification rate (prediction error)

THE DATASET KYPHOSIS

THE DATASET CONTAINS (1):

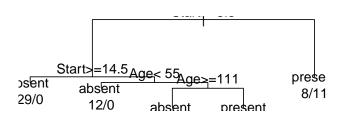
- ► Kyphosis: a factor with levels absent present indicating if a kyphosis (a type of deformation) was present after the operation.
- ► Age: in months.
- Number: the number of vertebrae involved.
- ▶ Start: the number of the first (topmost) vertebra operated on.

Build the tree (2)

```
library('rpart')
TREE <- rpart(Kyphosis ~ Age + Number + Start,
              data=kyphosis,method="class")
TREE
## n= 81
##
  node), split, n, loss, yval, (yprob)
         * denotes terminal node
##
##
##
    1) root 81 17 absent (0.79012346 0.20987654)
##
      2) Start>=8.5 62 6 absent (0.90322581 0.09677419)
##
        4) Start>=14.5 29 0 absent (1.00000000 0.00000000) *
        5) Start< 14.5 33 6 absent (0.81818182 0.18181818)
##
         10) Age< 55 12 0 absent (1.00000000 0.00000000) *
##
         11) Age>=55 21 6 absent (0.71428571 0.28571429)
##
           22) Age>=111 14 2 absent (0.85714286 0.14285714) *
##
##
           23) Age< 111 7 3 present (0.42857143 0.57142857) *
```

PLOT THE RESULT

```
plot(TREE)
text(TREE,use.n=T)
```



Answers

- 3. Which variables are used to explain Kyphosis presence?
- ▶ The variables are Start and Age
- 4. How many observations contain the terminal nodes.
- *denotes terminal nodes. The nodes have 29, 12, 14, 7 and 19 observations

Consider the Kyphosis data frame.

5. Build a tree using the first 60 observations of kyphosis.

6. Predict the kyphosis presence for the other 21 observations.

```
PR <- predict(TREE,kyphosis[61:81,],type='class')
```

7. Which is the misclassification rate (prediction error)

```
test=kyphosis$Kyphosis[61:81]
table(PR,test)
```

```
## test
## PR absent present
## absent 14 2
## present 3 2
```

```
rate <- 100*length(which(PR!=test))/length(PR)</pre>
```

EXERCISE RPART - IRIS

CONSIDER THE IRIS DATA FRAME

- 1. Build a tree to classify Species from the other variables.
- 2. Plot the trees, add nodes information.

Consider the tree build before

- 3. Prune the the using median complexity parameter (cp) associated to the tree.
- 4. Plot in the same window, the pruned and the original tree.
- 5. In which terminal nodes is clasified each oobservations of iris?
- 6. Which Specie has a flower of Petal.Length greater than 2.45 and Petal.Width less than 1.75.

SOLUTION - RPART - IRIS (I)

1. Build a tree to classify Species from the other variables.

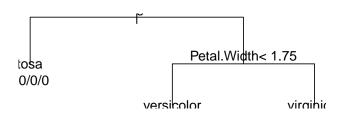
TREE2=rpart(Species ~ ., data=iris,method="class")

```
TREE2
## n = 150
##
## node), split, n, loss, yval, (yprob)
       * denotes terminal node
##
##
  1) root 150 100 setosa (0.33333333 0.33333333 0.33333333)
    ##
##
    3) Petal.Length>=2.45 100 50 versicolor (0.00000000 0.5000
##
      6) Petal.Width< 1.75 54 5 versicolor (0.00000000 0.9074
##
      7) Petal.Width>=1.75 46
                            1 virginica (0.00000000 0.02173
```

SOLUTION - RPART - IRIS (II)

2. Plot the trees, add nodes information.

```
plot(TREE2)
text(TREE2,use.n=T)
```



SOLUTION - RPART - IRIS (III)

3. Prune the the using median complexity parameter (cp) associated to the tree.

```
TP <- prune(TREE2,cp=median(TREE2$cptable[,'CP']))</pre>
```

4. Plot in the same window, the pruned and the original tree.

```
par(mfrow=c(1,2))
plot(TREE2);text(TREE2,use.n=T)
plot(TP);text(TP,use.n=T)
```



Solution - rpart - iris (IV)

5. In which terminal nodes is clasified each oobservations of iris?

TREE2\$where

| ## | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 12 | 13 | 14 | 15 | |
|----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|---|
| ## | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | |
| ## | 19 | 20 | 21 | 22 | 23 | 24 | 25 | 26 | 27 | 28 | 29 | 30 | 31 | 32 | 33 | |
| ## | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | |
| ## | 37 | 38 | 39 | 40 | 41 | 42 | 43 | 44 | 45 | 46 | 47 | 48 | 49 | 50 | 51 | |
| ## | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 2 | 4 | |
| ## | 55 | 56 | 57 | 58 | 59 | 60 | 61 | 62 | 63 | 64 | 65 | 66 | 67 | 68 | 69 | |
| ## | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | |
| ## | 73 | 74 | 75 | 76 | 77 | 78 | 79 | 80 | 81 | 82 | 83 | 84 | 85 | 86 | 87 | |
| ## | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | |
| ## | 91 | 92 | 93 | 94 | 95 | 96 | 97 | 98 | 99 | 100 | 101 | 102 | 103 | 104 | 105 | 1 |
| ## | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 4 | 5 | 5 | 5 | 5 | 5 | |
| ## | 109 | 110 | 111 | 112 | 113 | 114 | 115 | 116 | 117 | 118 | 119 | 120 | 121 | 122 | 123 | 1 |
| ## | 5 | 5 | 5 | 5 | 5 | 5 | 5 | 5 | 5 | 5 | 5 | 4 | 5 | 5 | 5 | |

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