

Lab11.R

rstudio-user

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```
#pseudorandom number generator
set.seed(11)

# Attach Packages
library(tidyverse)      # data manipulation and visualization

## -- Attaching packages ----- tidyverse 1.3.1 --
## v ggplot2 3.3.3      v purrr  0.3.4
## v tibble  3.1.1      v dplyr  1.0.5
## v tidyr   1.1.3      v stringr 1.4.0
## v readr   1.4.0      v forcats 0.5.1

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()

library(kernlab)      # SVM methodology

##
## Attaching package: 'kernlab'

## The following object is masked from 'package:purrr':
##
##   cross

## The following object is masked from 'package:ggplot2':
##
##   alpha

library(e1071)      # SVM methodology
library(ISLR)       # contains example data set "Khan"
library(RColorBrewer) # customized coloring of plots
library(caret)

## Loading required package: lattice

##
## Attaching package: 'caret'

## The following object is masked from 'package:purrr':
##
##   lift

#Read file
fl <- read.csv("heart.csv")
head(fl)
```

```
##   age sex cp trestbps chol fbs restecg thalach exang oldpeak slope ca thal
## 1  63  1  3    145  233   1         0    150    0    2.3    0  0    1
## 2  37  1  2    130  250   0         1    187    0    3.5    0  0    2
## 3  41  0  1    130  204   0         0    172    0    1.4    2  0    2
## 4  56  1  1    120  236   0         1    178    0    0.8    2  0    2
## 5  57  0  0    120  354   0         1    163    1    0.6    2  0    2
## 6  57  1  0    140  192   0         1    148    0    0.4    1  0    1
##   target
## 1      1
## 2      1
## 3      1
## 4      1
## 5      1
## 6      1
```

```
#Data structure
str(fl)
```

```
## 'data.frame':   303 obs. of  14 variables:
## $ age      : int  63 37 41 56 57 57 56 44 52 57 ...
## $ sex      : int  1 1 0 1 0 1 0 1 1 1 ...
## $ cp       : int  3 2 1 1 0 0 1 1 2 2 ...
## $ trestbps : int  145 130 130 120 120 140 140 120 172 150 ...
## $ chol     : int  233 250 204 236 354 192 294 263 199 168 ...
## $ fbs      : int  1 0 0 0 0 0 0 0 1 0 ...
## $ restecg  : int  0 1 0 1 1 1 0 1 1 1 ...
## $ thalach  : int  150 187 172 178 163 148 153 173 162 174 ...
## $ exang    : int  0 0 0 0 1 0 0 0 0 0 ...
## $ oldpeak  : num  2.3 3.5 1.4 0.8 0.6 0.4 1.3 0 0.5 1.6 ...
## $ slope    : int  0 0 2 2 2 1 1 2 2 2 ...
## $ ca       : int  0 0 0 0 0 0 0 0 0 0 ...
## $ thal     : int  1 2 2 2 2 1 2 3 3 2 ...
## $ target   : int  1 1 1 1 1 1 1 1 1 1 ...
```

```
#Summary
summary(fl)
```

```
##      age          sex          cp          trestbps
##  Min.   :29.00   Min.   :0.0000   Min.   :0.000   Min.   : 94.0
## 1st Qu.:47.50   1st Qu.:0.0000   1st Qu.:0.000   1st Qu.:120.0
## Median :55.00   Median :1.0000   Median :1.000   Median :130.0
## Mean   :54.37   Mean   :0.6832   Mean   :0.967   Mean   :131.6
## 3rd Qu.:61.00   3rd Qu.:1.0000   3rd Qu.:2.000   3rd Qu.:140.0
## Max.   :77.00   Max.   :1.0000   Max.   :3.000   Max.   :200.0
##      chol          fbs          restecg          thalach
##  Min.   :126.0   Min.   :0.0000   Min.   :0.0000   Min.   : 71.0
## 1st Qu.:211.0   1st Qu.:0.0000   1st Qu.:0.0000   1st Qu.:133.5
## Median :240.0   Median :0.0000   Median :1.0000   Median :153.0
## Mean   :246.3   Mean   :0.1485   Mean   :0.5281   Mean   :149.6
## 3rd Qu.:274.5   3rd Qu.:0.0000   3rd Qu.:1.0000   3rd Qu.:166.0
## Max.   :564.0   Max.   :1.0000   Max.   :2.0000   Max.   :202.0
##      exang          oldpeak          slope          ca
##  Min.   :0.0000   Min.   :0.00    Min.   :0.000   Min.   :0.0000
## 1st Qu.:0.0000   1st Qu.:0.00    1st Qu.:1.000   1st Qu.:0.0000
## Median :0.0000   Median :0.80    Median :1.000   Median :0.0000
## Mean   :0.3267   Mean   :1.04    Mean   :1.399   Mean   :0.7294
```

```
## 3rd Qu.:1.0000 3rd Qu.:1.60 3rd Qu.:2.000 3rd Qu.:1.0000
## Max. :1.0000 Max. :6.20 Max. :2.000 Max. :4.0000
## thal target
## Min. :0.000 Min. :0.0000
## 1st Qu.:2.000 1st Qu.:0.0000
## Median :2.000 Median :1.0000
## Mean :2.314 Mean :0.5446
## 3rd Qu.:3.000 3rd Qu.:1.0000
## Max. :3.000 Max. :1.0000
```

```
#Empty values
colSums(fl==" ")
```

```
## age sex cp trestbps chol fbs restecg thalach
## 0 0 0 0 0 0 0 0
## exang oldpeak slope ca thal target
## 0 0 0 0 0 0
```

```
#So there are no empty values
```

```
#Null values
colSums(is.na(fl))
```

```
## age sex cp trestbps chol fbs restecg thalach
## 0 0 0 0 0 0 0 0
## exang oldpeak slope ca thal target
## 0 0 0 0 0 0
```

```
#So there are no null values
```

```
#MODEL BUILDING
```

```
#Split dataset into train and test
```

```
index <- sample(1:nrow(fl), 0.75*nrow(fl))
train <- fl[index,]
test <- fl[-index,]
```

```
#Convert target into factor
```

```
train$target <- as.factor(train$target)
test$target <- as.factor(test$target)
```

```
training <- trainControl(method="repeatedcv", number=10, repeats=3)
```

```
grid <- expand.grid(C=c(0.01, 0.05, 0.1, 0.25, 0.5, 0.75, 1, 1.25, 1.5, 1.75, 2, 5))
```

```
svmgrid <- train(target~., data=train, method="svmLinear", trControl=training, preProcess=c("center", "scale"))
```

```
svmgrid
```

```
## Support Vector Machines with Linear Kernel
```

```
##
```

```
## 227 samples
```

```
## 13 predictor
```

```
## 2 classes: '0', '1'
```

```
##
```

```
## Pre-processing: centered (13), scaled (13)
```

```
## Resampling: Cross-Validated (10 fold, repeated 3 times)
```

```
## Summary of sample sizes: 205, 203, 205, 203, 205, 204, ...
```

```
## Resampling results across tuning parameters:
```

```
##
##      C      Accuracy  Kappa
##    0.01  0.8247969  0.6364641
##    0.05  0.8263834  0.6419405
##    0.10  0.8249945  0.6398748
##    0.25  0.8233476  0.6367264
##    0.50  0.8278272  0.6466485
##    0.75  0.8293423  0.6497273
##    1.00  0.8293423  0.6497273
##    1.25  0.8293423  0.6497273
##    1.50  0.8263779  0.6437773
##    1.75  0.8263779  0.6437773
##    2.00  0.8263779  0.6437773
##    5.00  0.8263779  0.6437773
##
## Accuracy was used to select the optimal model using the largest value.
## The final value used for the model was C = 0.75.
#Training with gamma=0.2 and cost=0.25
m1 <- svm(target~., data=train, kernel="linear", gamma=0.2, cost=0.25)
m1

##
## Call:
## svm(formula = target ~ ., data = train, kernel = "linear", gamma = 0.2,
##      cost = 0.25)
##
##
## Parameters:
##   SVM-Type:  C-classification
##   SVM-Kernel: linear
##      cost:   0.25
##
## Number of Support Vectors:  95
#Training with gamma=0.5 and cost=0.05
m2 <- svm(target~., data=train, kernel="linear", gamma=0.5, cost=0.05)
m2

##
## Call:
## svm(formula = target ~ ., data = train, kernel = "linear", gamma = 0.5,
##      cost = 0.05)
##
##
## Parameters:
##   SVM-Type:  C-classification
##   SVM-Kernel: linear
##      cost:   0.05
##
## Number of Support Vectors:  111
#Test with gamma=0.2 and cost=0.25
pred1 <- predict(m1, newdata=test)
confusionMatrix(pred1, test$target)

## Confusion Matrix and Statistics
```

```
##
##           Reference
## Prediction  0  1
##           0 27  2
##           1  9 38
##
##           Accuracy : 0.8553
##           95% CI : (0.7558, 0.9255)
##           No Information Rate : 0.5263
##           P-Value [Acc > NIR] : 1.432e-09
##
##           Kappa : 0.7069
##
## Mcnemar's Test P-Value : 0.07044
##
##           Sensitivity : 0.7500
##           Specificity : 0.9500
##           Pos Pred Value : 0.9310
##           Neg Pred Value : 0.8085
##           Prevalence : 0.4737
##           Detection Rate : 0.3553
##           Detection Prevalence : 0.3816
##           Balanced Accuracy : 0.8500
##
##           'Positive' Class : 0
##
```

```
#Test with gamma=0.5 and cost=0.05
pred2 <- predict(m2, newdata=test)
confusionMatrix(pred2, test$target)
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction  0  1
##           0 27  2
##           1  9 38
##
##           Accuracy : 0.8553
##           95% CI : (0.7558, 0.9255)
##           No Information Rate : 0.5263
##           P-Value [Acc > NIR] : 1.432e-09
##
##           Kappa : 0.7069
##
## Mcnemar's Test P-Value : 0.07044
##
##           Sensitivity : 0.7500
##           Specificity : 0.9500
##           Pos Pred Value : 0.9310
##           Neg Pred Value : 0.8085
##           Prevalence : 0.4737
##           Detection Rate : 0.3553
##           Detection Prevalence : 0.3816
##           Balanced Accuracy : 0.8500
```

```
##  
##      'Positive' Class : 0  
##
```

```
#We trained two different models with two sets of gamma and cost values.  
#Upon evaluating the model, we see that model 2 yeilds better accuracy of 84%,  
#than model 1, which is 82%.
```