$ML_original_data$

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Packages & Data

```
library(readr)
library(performance)
library(gmodels)
library(caret)
library(tidyverse)
library(tree)
library(randomForest)
```

Loading and setting up the data

```
# Load data into the environment and creates partition for 50-50 validation
german_credit <- read_csv("data/german_credit.csv")

Train50 <- read_csv("data/Training50.csv")

Test50 <- read_csv("data/Test50.csv")

attach(Train50)</pre>
```

Logistic regression model with 50:50 Cross-validation

```
LogisticModel50final <- glm(Creditability ~ Account.Balance +

Payment.Status.of.Previous.Credit + Purpose +

Length.of.current.employment +

Sex...Marital.Status, family=binomial,

data = Train50)

fit50 <- fitted.values(LogisticModel50final)

Threshold50 <- rep(0,500)
```

```
##
##
##
  Cell Contents
## I
   N / Col Total |
## |-----|
## Total Observations in Table: 500
##
##
             | Threshold50
##
                     1 | Row Total |
## Train50$Creditability | 0 |
 -----|
               47 | 96 |
0.6 | 0.2 |
           0 |
##
##
            ## -----|-----|
                      327 |
           1 |
                30 |
           0.4 | 0.8 |
##
## -----|----|
     Column Total | 77 |
                     423 |
##
      1
                0.2 |
                      0.8 |
## -----|----|
##
##
```

```
for (i in 1:500) {
   if (Threshold50[i] == '1') {
     Threshold50[i] <- 'Creditable'
   }

   if (Threshold50[i] == '0') {
     Threshold50[i] <- 'Non-Creditable'
   }
}

for (i in 1:500) {
   if (Test50$Creditability[i] == '0') {
     Test50$Creditability[i] <- 'Non-Creditable'
   }

   if (Test50$Creditability[i] == '1') {
     Test50$Creditability[i] <- 'Creditable'
   }
}</pre>
```

Confusion matrix for GLM

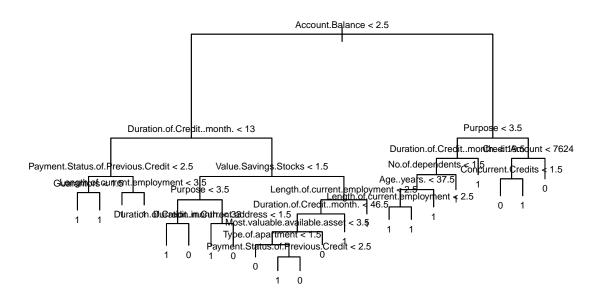
```
confusion <- confusionMatrix(data = factor(Threshold50),</pre>
                             reference = factor(Test50$Creditability))
confusion
## Confusion Matrix and Statistics
##
##
                   Reference
## Prediction
                    Creditable Non-Creditable
##
     Creditable
                           291
                                           132
##
     Non-Creditable
                            52
                                            25
##
                  Accuracy: 0.632
##
                    95% CI: (0.588, 0.6744)
##
##
       No Information Rate: 0.686
       P-Value [Acc > NIR] : 0.9956
##
##
##
                     Kappa: 0.0089
##
    Mcnemar's Test P-Value : 5.747e-09
##
##
##
               Sensitivity: 0.8484
##
               Specificity: 0.1592
##
            Pos Pred Value: 0.6879
##
            Neg Pred Value: 0.3247
##
                Prevalence: 0.6860
            Detection Rate: 0.5820
##
      Detection Prevalence: 0.8460
##
##
         Balanced Accuracy: 0.5038
##
##
          'Positive' Class : Creditable
```

Supervised Tree based method

##

summary(Train50_tree)

```
##
## Classification tree:
## tree(formula = as.factor(Creditability) ~ Account.Balance + Duration.of.Credit..month. +
       Payment.Status.of.Previous.Credit + Purpose + Credit.Amount +
       Value.Savings.Stocks + Length.of.current.employment + Instalment.per.cent +
##
##
       Sex...Marital.Status + Guarantors + Duration.in.Current.address +
##
       Most.valuable.available.asset + Age..years. + Concurrent.Credits +
##
       Type.of.apartment + No.of.Credits.at.this.Bank + Occupation +
##
       No.of.dependents + Telephone, data = Train50, method = "class")
## Variables actually used in tree construction:
## [1] "Account.Balance"
                                            "Duration.of.Credit..month."
## [3] "Payment.Status.of.Previous.Credit" "Guarantors"
## [5] "Length.of.current.employment"
                                            "Value.Savings.Stocks"
                                            "Duration.in.Current.address"
## [7] "Purpose"
## [9] "Most.valuable.available.asset"
                                            "Type.of.apartment"
## [11] "No.of.dependents"
                                            "Age..years."
## [13] "Credit.Amount"
                                            "Concurrent.Credits"
## Number of terminal nodes: 22
## Residual mean deviance: 0.7682 = 367.2 / 478
## Misclassification error rate: 0.168 = 84 / 500
# Tree visual representation
plot(Train50_tree)
text(Train50_tree, pretty=0,cex=0.6)
```



```
# Simple confusion matrix without pruning
Test50_pred <- predict(Train50_tree, Test50, type="class")
table(Test50_pred, Test50$Creditability)

##
## Test50_pred 0 1
## 0 46 45
## 1 111 298

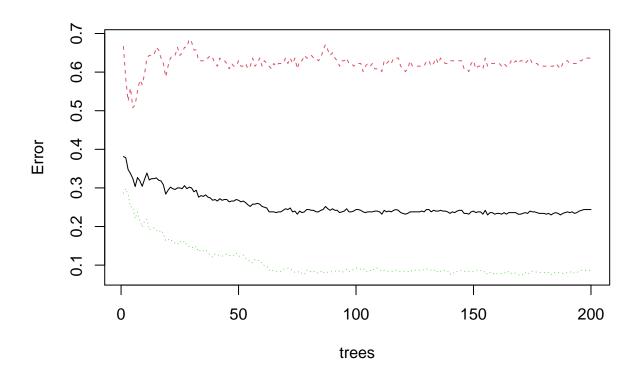
Train50_prune8 <- prune.misclass(Train50_tree, best=8)
Test50_prune8_pred <- predict(Train50_prune8, Test50, type="class")</pre>
```

Confusion matrix for supervised trees (with pruning)

```
1 119 314
##
##
##
                  Accuracy: 0.704
##
                    95% CI : (0.6619, 0.7437)
       No Information Rate: 0.686
##
       P-Value [Acc > NIR] : 0.207
##
##
##
                     Kappa: 0.1865
##
##
   Mcnemar's Test P-Value : 2.559e-13
##
##
               Sensitivity: 0.2420
##
               Specificity: 0.9155
            Pos Pred Value: 0.5672
##
##
           Neg Pred Value: 0.7252
                Prevalence: 0.3140
##
##
           Detection Rate: 0.0760
      Detection Prevalence: 0.1340
##
##
         Balanced Accuracy: 0.5787
##
          'Positive' Class : 0
##
##
```

Unsupervised Random Forest based method

```
rf50 <- randomForest(as.factor(Creditability) ~., data = Train50, ntree=200, importance=T, proximity=T)
plot(rf50, main="")</pre>
```



```
rf50
##
## Call:
   Type of random forest: classification
##
                  Number of trees: 200
##
## No. of variables tried at each split: 4
##
        OOB estimate of error rate: 24.4%
##
## Confusion matrix:
       1 class.error
## 0 52 91 0.63636364
## 1 31 326 0.08683473
Test50_rf_pred <- predict(rf50, Test50, type="class")</pre>
```

Confusion matrix for unsupervised random forest

```
## Confusion Matrix and Statistics
##
##
            Reference
## Prediction 0 1
           0 52 29
##
           1 105 314
##
##
##
                 Accuracy: 0.732
                   95% CI : (0.6909, 0.7704)
##
##
       No Information Rate : 0.686
       P-Value [Acc > NIR] : 0.01417
##
##
##
                     Kappa : 0.2839
##
##
   Mcnemar's Test P-Value : 9.232e-11
##
##
              Sensitivity: 0.3312
##
              Specificity: 0.9155
##
           Pos Pred Value : 0.6420
##
           Neg Pred Value: 0.7494
##
               Prevalence: 0.3140
##
           Detection Rate: 0.1040
##
      Detection Prevalence : 0.1620
##
         Balanced Accuracy: 0.6233
##
          'Positive' Class : 0
##
##
```