Decision tree and random forest

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```
library(mclust)
## Warning: package 'mclust' was built under R version 4.1.3
## Package 'mclust' version 5.4.10
## Type 'citation("mclust")' for citing this R package in publications.
library(DAAG)
## Warning: package 'DAAG' was built under R version 4.1.3
library(party)
## Warning: package 'party' was built under R version 4.1.3
## Loading required package: grid
## Loading required package: mvtnorm
## Warning: package 'mvtnorm' was built under R version 4.1.1
## Attaching package: 'mvtnorm'
## The following object is masked from 'package:mclust':
##
       dmvnorm
## Loading required package: modeltools
## Warning: package 'modeltools' was built under R version 4.1.1
## Loading required package: stats4
## Loading required package: strucchange
## Warning: package 'strucchange' was built under R version 4.1.3
```

```
## Loading required package: zoo
## Warning: package 'zoo' was built under R version 4.1.3
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
       as.Date, as.Date.numeric
##
## Loading required package: sandwich
## Warning: package 'sandwich' was built under R version 4.1.3
library(rpart)
## Warning: package 'rpart' was built under R version 4.1.3
library(rpart.plot)
## Warning: package 'rpart.plot' was built under R version 4.1.3
library(mlbench)
## Warning: package 'mlbench' was built under R version 4.1.3
library(pROC)
## Warning: package 'pROC' was built under R version 4.1.3
## Type 'citation("pROC")' for a citation.
## Attaching package: 'pROC'
## The following objects are masked from 'package:stats':
##
##
       cov, smooth, var
library(tree)
## Warning: package 'tree' was built under R version 4.1.3
library(caret)
## Warning: package 'caret' was built under R version 4.1.3
## Loading required package: ggplot2
## Warning: package 'ggplot2' was built under R version 4.1.3
## Loading required package: lattice
```

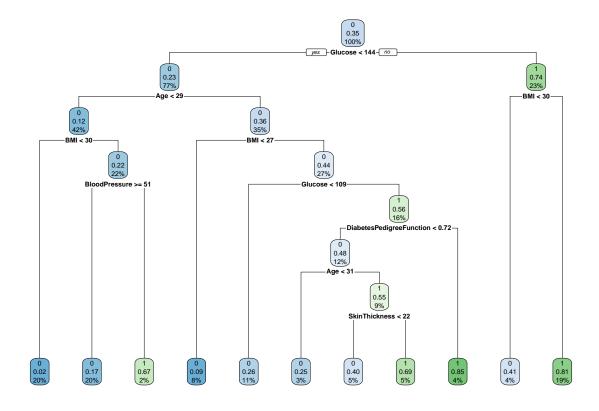
Data Cleaning

```
Diabetes<-read.csv("diabetes.csv")
set.seed(123)
trainIndex <- createDataPartition(Diabetes$Outcome, p = 0.7, list = FALSE)
bn.training <- Diabetes[trainIndex,]
bn.test <- Diabetes[-trainIndex,]
head(bn.training)</pre>
```

```
Pregnancies Glucose BloodPressure SkinThickness Insulin BMI
##
## 2
                                                           0 26.6
                1
                      85
                                    66
                                                  29
               0
                                                  35
## 5
                      137
                                    40
                                                         168 43.1
                                                           0 25.6
## 6
               5
                     116
                                    74
                                                   0
                                                  32
## 7
               3
                                    50
                                                          88 31.0
                      78
              10
                                                           0 35.3
## 8
                     115
                                     0
                                                   0
               8
                                    96
                                                           0.0
## 10
                     125
                                                   0
##
     DiabetesPedigreeFunction Age Outcome
                        0.351 31
## 5
                        2.288 33
                                        1
                        0.201 30
## 6
                                        0
## 7
                        0.248 26
                                        1
## 8
                        0.134 29
                                        0
## 10
                        0.232 54
                                        1
```

tree

```
tree <- rpart(Outcome ~., data = bn.training,method='class',minsplit=10,minbucket=10)
rpart.plot(tree)</pre>
```



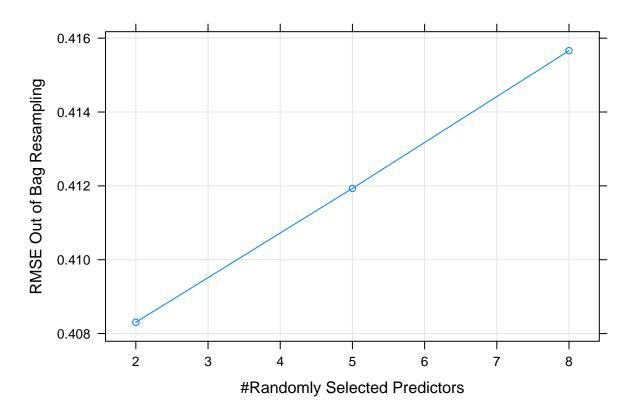
tree

```
## n= 538
##
##
  node), split, n, loss, yval, (yprob)
##
        * denotes terminal node
##
##
    1) root 538 187 0 (0.65241636 0.34758364)
      2) Glucose< 143.5 415 96 0 (0.76867470 0.23132530)
##
        4) Age< 28.5 227 28 0 (0.87665198 0.12334802)
##
##
         8) BMI< 29.95 107
                           2 0 (0.98130841 0.01869159) *
##
         9) BMI>=29.95 120 26 0 (0.78333333 0.21666667)
##
          18) BloodPressure>=51 108 18 0 (0.83333333 0.16666667) *
##
          19) BloodPressure< 51 12
                                   4 1 (0.33333333 0.66666667) *
        5) Age>=28.5 188 68 0 (0.63829787 0.36170213)
##
##
         10) BMI< 27.05 43
                          4 0 (0.90697674 0.09302326) *
##
         11) BMI>=27.05 145 64 0 (0.55862069 0.44137931)
##
          ##
          23) Glucose>=108.5 87 38 1 (0.43678161 0.56321839)
            46) DiabetesPedigreeFunction< 0.7205 67 32 0 (0.52238806 0.47761194)
##
##
              92) Age< 30.5 16 4 0 (0.75000000 0.25000000) *
##
              93) Age>=30.5 51 23 1 (0.45098039 0.54901961)
##
               186) SkinThickness< 21.5 25 10 0 (0.60000000 0.40000000) *
               ##
##
            47) DiabetesPedigreeFunction>=0.7205 20 3 1 (0.15000000 0.85000000) *
      3) Glucose>=143.5 123 32 1 (0.26016260 0.73983740)
##
```

```
## 6) BMI< 29.85 22 9 0 (0.59090909 0.40909091) *
## 7) BMI>=29.85 101 19 1 (0.18811881 0.81188119) *
```

Random Forests

```
set.seed(123)
oob_train_control <- trainControl(method="oob", classProbs = TRUE, savePredictions = TRUE)</pre>
forestfit <- train(Outcome ~., data = bn.training, method = 'rf', importance = FALSE, trControl = oob_tr</pre>
## Warning in train.default(x, y, weights = w, ...): You are trying to do
## regression and your outcome only has two possible values Are you trying to do
## classification? If so, use a 2 level factor as your outcome column.
## Warning in train.default(x, y, weights = w, ...): cannnot compute class
## probabilities for regression
## Warning in randomForest.default(x, y, mtry = param$mtry, ...): The response has
## five or fewer unique values. Are you sure you want to do regression?
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## five or fewer unique values. Are you sure you want to do regression?
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## five or fewer unique values. Are you sure you want to do regression?
## Warning in randomForest.default(x, y, mtry = param$mtry, ...): The response has
## five or fewer unique values. Are you sure you want to do regression?
plot(forestfit)
```



```
recommended.mtry <-floor(sqrt(ncol(bn.training)))</pre>
tunegrid <- expand.grid(mtry=recommended.mtry)</pre>
set.seed(123)
forestfit.m <- train(Outcome ~.,data = bn.training, method = 'rf', importance = FALSE, trControl = oob_</pre>
## Warning in train.default(x, y, weights = w, ...): You are trying to do
## regression and your outcome only has two possible values Are you trying to do
## classification? If so, use a 2 level factor as your outcome column.
## Warning in train.default(x, y, weights = w, ...): cannnot compute class
## probabilities for regression
## Warning in randomForest.default(x, y, mtry = param$mtry, ...): The response has
## five or fewer unique values. Are you sure you want to do regression?
## Warning in randomForest.default(x, y, mtry = param$mtry, ...): The response has
## five or fewer unique values. Are you sure you want to do regression?
print(forestfit.m, digits = 2)
## Random Forest
##
## 538 samples
     8 predictor
```

```
##
## No pre-processing
## Resampling results:
##
    RMSE Rsquared
##
    0.41 0.26
## Tuning parameter 'mtry' was held constant at a value of 3
set.seed(123)
forestfit.ntree <- train(Outcome ~., data = bn.training, method = 'rf', ntree = 500, importance = T, trCo.
## Warning in train.default(x, y, weights = w, ...): You are trying to do
## regression and your outcome only has two possible values Are you trying to do
## classification? If so, use a 2 level factor as your outcome column.
## Warning in train.default(x, y, weights = w, ...): cannnot compute class
## probabilities for regression
## Warning in randomForest.default(x, y, mtry = param$mtry, ...): The response has
## five or fewer unique values. Are you sure you want to do regression?
## Warning in randomForest.default(x, y, mtry = param$mtry, ...): The response has
## five or fewer unique values. Are you sure you want to do regression?
print(forestfit.ntree, digits = 2)
## Random Forest
## 538 samples
    8 predictor
##
## No pre-processing
## Resampling results:
##
##
    RMSE Rsquared
##
    0.41 0.26
##
## Tuning parameter 'mtry' was held constant at a value of 3
forestfit.ntree$finalModel
##
## Call:
   randomForest(x = x, y = y, ntree = 500, mtry = param$mtry, importance = ..2)
                  Type of random forest: regression
##
                        Number of trees: 500
## No. of variables tried at each split: 3
##
             Mean of squared residuals: 0.167706
##
                       % Var explained: 26.05
```

```
library(randomForest)

## Warning: package 'randomForest' was built under R version 4.1.3

## randomForest 4.7-1.1

## Type rfNews() to see new features/changes/bug fixes.

## ## Attaching package: 'randomForest'

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

## ## margin

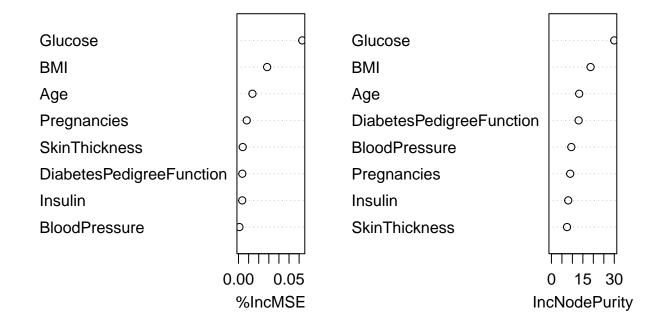
forestfit.RF <- randomForest(Outcome ~., data = bn.training, ntree = 500, importance = TRUE)

## Warning in randomForest.default(m, y, ...): The response has five or fewer

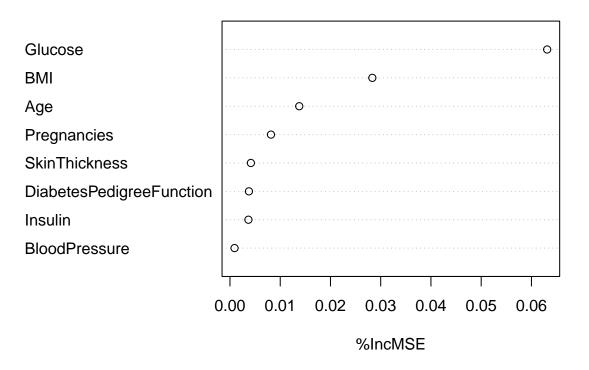
## unique values. Are you sure you want to do regression?

varImpPlot(forestfit.RF, scale = F)</pre>
```

forestfit.RF

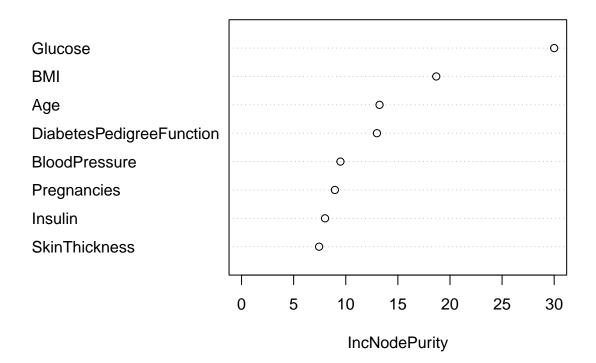


forestfit.RF

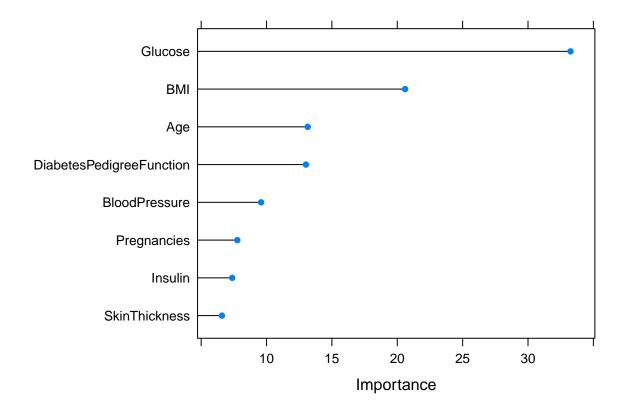


varImpPlot(forestfit.RF, type = 2, scale = F)

forestfit.RF



```
RFimp <- varImp(forestfit.m, scale = F)
plot(RFimp)</pre>
```



According to the result above, "Glucose" is the most important variable.