

ml project

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Predicting Heart Disease

The **heart** dataset (source: UCI Machine Learning Repository) contains 920 observations and 76 variables. However, we're only using the following 14 of the 76 variables in our study:

- **age**: patient age (in years)
- **sex**: gender of patient; 0 = male, 1 = female
- **cp**: chest pain type; 1 = typical angina, 2 = atypical angina, 3 = non-anginal pain, 4 = asymptomatic
- **trestbps**: resting blood pressure (in mmHg)
- **chol**: serum cholesterol (in mg/dl)
- **fbs**: fasting blood sugar > 120 mg/dl; 0 = false, 1 = true
- **restecg**: resting electrocardiographic results; 0 = normal, 1 = having ST-T wave abnormality, 2 = showing probable or definite left ventricular hypertrophy by Ester's criteria
- **thalach**: maximum heart rate achieved
- **exang**: exercise induced angina; 0 = no, 1 = yes
- **oldpeak**: ST depression induced by exercise relative to rest
- **slope**: the slope of the peak exercise ST segment; 1 = upsloping, 2 = flat, 3 = downsloping
- **ca**: number of major vessels (0-3) colored by fluoroscopy
- **thal**: thalassemia; 3 = normal, 6 = fixed defect, 7 = reversible defect
- **num**: diagnosis of heart disease; 0 = no heart disease, 1 = have heart disease

Load, Merge, and Recode Data

```
library(dplyr)
```

```
##
```

```
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
##      filter, lag
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      intersect, setdiff, setequal, union
```

```
cleveland <- read.csv("/Users/jiayizhao/Desktop/2338 Machine Learning/heart_disease_final_data/processed.cleveland.csv")
```

```
hungarian <- read.csv("/Users/jiayizhao/Desktop/2338 Machine Learning/heart_disease_final_data/processed.hungarian.csv")
```

```
switzerland <- read.csv("/Users/jiayizhao/Desktop/2338 Machine Learning/heart_disease_final_data/processed.switzerland.csv")
```

```
va <- read.csv("/Users/jiayizhao/Desktop/2338 Machine Learning/heart_disease_final_data/processed.va.csv")
```

```

nrow(cleveland)

## [1] 303
nrow(hungarian)

## [1] 294
nrow(switzerland)

## [1] 123
nrow(va)

## [1] 200
heart <- rbind(cleveland, hungarian, switzerland, va)

# check for missing values
heart[heart == "?"] <- NA
sum(is.na(heart))

## [1] 1759
apply(heart, function(x) sum(is.na(x)))

##      age      sex      cp trestbps      chol      fbs  restecg  thalach
##      0       0       0       59       30      90       2       55
##    exang oldpeak    slope      ca      thal      num
##      55      62      309      611      486       0
nrow(heart)

## [1] 920
ncol(heart)

## [1] 14

```

Univariable Analysis

```

library(dplyr)
heart$age <- as.numeric(heart$age)
heart$trestbps <- as.numeric(heart$trestbps)
heart$chol <- as.numeric(heart$chol)
heart$thalach <- as.numeric(heart$thalach)
heart$oldpeak <- as.numeric(heart$oldpeak)
heart <- na.omit(heart)
#Univariate Analysis on Continuous Data
summary(heart)

##      age      sex      cp      trestbps
##  Min.   :29.00  Min.   :0.0000  Min.   :1.000  Min.   : 94.0
## 1st Qu.:48.00  1st Qu.:0.0000  1st Qu.:3.000  1st Qu.:120.0
## Median :56.00  Median :1.0000  Median :3.000  Median :130.0
## Mean   :54.52  Mean   :0.6789  Mean   :3.164  Mean   :131.7
## 3rd Qu.:61.00  3rd Qu.:1.0000  3rd Qu.:4.000  3rd Qu.:140.0
## Max.   :77.00  Max.   :1.0000  Max.   :4.000  Max.   :200.0

```

```
##      chol      fbs      restecg      thalach
## Min.   :100.0   Length:299   Length:299   Min.    : 71.0
## 1st Qu.:211.0   Class :character   Class :character   1st Qu.:132.5
## Median :242.0   Mode  :character   Mode  :character   Median :152.0
## Mean   :246.8
## 3rd Qu.:275.5
## Max.   :564.0
##      exang      oldpeak      slope      ca
## Length:299   Min.    :0.000   Length:299   Length:299
## Class :character   1st Qu.:0.000   Class :character   Class :character
## Mode  :character   Median :0.800   Mode  :character   Mode  :character
##                      Mean    :1.059
##                      3rd Qu.:1.600
##                      Max.    :6.200
##      thal      num
## Length:299   Min.    :0.0000
## Class :character   1st Qu.:0.0000
## Mode  :character   Median :0.0000
##                      Mean    :0.9465
##                      3rd Qu.:2.0000
##                      Max.    :4.0000

sd_values <- sapply(heart[, c("age", "trestbps", "chol", "thalach", "oldpeak")], sd)
print(sd_values)
```

```
##      age trestbps      chol thalach oldpeak
## 9.030264 17.747751 52.532582 23.121062 1.162769
```

```
#Univariate Analysis on Categorical Data
sex_counts <- table(heart$sex)
sex_percentages <- prop.table(sex_counts) * 100
sex_summary <- data.frame(sex = names(sex_counts),
                          Count = as.numeric(sex_counts),
                          Percentage = sex_percentages)

print(sex_summary)
```

```
##      sex Count Percentage.Var1 Percentage.Freq
## 1      0     96              0         32.10702
## 2      1    203              1         67.89298
```

```
cp_counts <- table(heart$cp)
cp_percentages <- prop.table(cp_counts) * 100
cp_summary <- data.frame(cp = names(cp_counts),
                        Count = as.numeric(cp_counts),
                        Percentage = cp_percentages)

print(cp_summary)
```

```
##      cp Count Percentage.Var1 Percentage.Freq
## 1      1     23              1         7.692308
## 2      2     49              2        16.387960
## 3      3     83              3        27.759197
## 4      4    144              4        48.160535
```

```
fbs_counts <- table(heart$fbs)
fbs_percentages <- prop.table(fbs_counts) * 100
fbs_summary <- data.frame(fbs = names(fbs_counts),
                        Count = as.numeric(fbs_counts),
```

```

                                Percentage = fbs_percentages)
print(fbs_summary)

##    fbs Count Percentage.Var1 Percentage.Freq
## 1    0   256                0      85.61873
## 2    1    43                1      14.38127

restecg_counts <- table(heart$restecg)
restecg_percentages <- prop.table(restecg_counts) * 100
restecg_summary <- data.frame(restecg = names(restecg_counts),
                              Count = as.numeric(restecg_counts),
                              Percentage = restecg_percentages)
print(restecg_summary)

##    restecg Count Percentage.Var1 Percentage.Freq
## 1         0   149                0      49.832776
## 2         1     4                1      1.337793
## 3         2   146                2      48.829431

exang_counts <- table(heart$exang)
exang_percentages <- prop.table(exang_counts) * 100
exang_summary <- data.frame(exang = names(exang_counts),
                            Count = as.numeric(exang_counts),
                            Percentage = exang_percentages)
print(exang_summary)

##    exang Count Percentage.Var1 Percentage.Freq
## 1     0   200                0      66.88963
## 2     1    99                1      33.11037

slope_counts <- table(heart$slope)
slope_percentages <- prop.table(slope_counts) * 100
slope_summary <- data.frame(slope = names(slope_counts),
                            Count = as.numeric(slope_counts),
                            Percentage = slope_percentages)
print(slope_summary)

##    slope Count Percentage.Var1 Percentage.Freq
## 1     1   139                1      46.488294
## 2     2   139                2      46.488294
## 3     3    21                3       7.023411

ca_counts <- table(heart$ca)
ca_percentages <- prop.table(ca_counts) * 100
ca_summary <- data.frame(ca = names(ca_counts),
                        Count = as.numeric(ca_counts),
                        Percentage = ca_percentages)
print(ca_summary)

##    ca Count Percentage.Var1 Percentage.Freq
## 1    0   176                0      58.862876
## 2    1    65                1      21.739130
## 3    2    38                2      12.709030
## 4    3    20                3       6.688963

thal_counts <- table(heart$thal)
thal_percentages <- prop.table(thal_counts) * 100

```

```
thal_summary <- data.frame(thal = names(thal_counts),
                           Count = as.numeric(thal_counts),
                           Percentage = thal_percentages)

print(thal_summary)
```

```
##   thal Count Percentage.Var1 Percentage.Freq
## 1    3   164             3      54.849498
## 2    6    18             6       6.020067
## 3    7   117             7      39.130435
```

```
num_counts <- table(heart$num)
num_percentages <- prop.table(num_counts) * 100
num_summary <- data.frame(num = names(num_counts),
                           Count = as.numeric(num_counts),
                           Percentage = num_percentages)

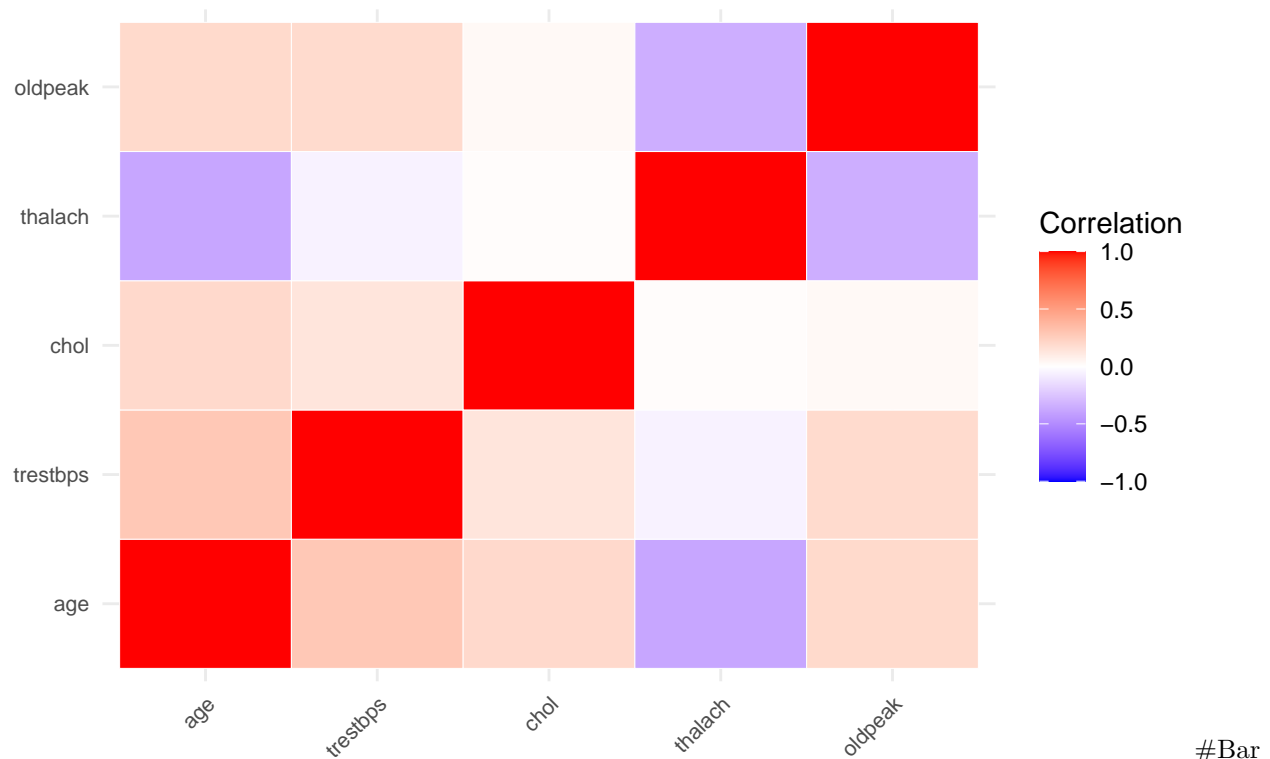
print(num_summary)
```

```
##   num Count Percentage.Var1 Percentage.Freq
## 1    0   160             0      53.511706
## 2    1    56             1      18.729097
## 3    2    35             2      11.705686
## 4    3    35             3      11.705686
## 5    4    13             4       4.347826
```

Correlation Heatmap

```
library(reshape2)
cor_matrix <- cor(heart[, c("age", "trestbps", "chol", "thalach", "oldpeak")])
melted_cor_matrix <- melt(cor_matrix)
```

```
library(ggplot2)
ggplot(melted_cor_matrix, aes(Var1, Var2, fill = value)) +
  geom_tile(color = "white") +
  scale_fill_gradient2(low = "blue", high = "red", mid = "white", midpoint = 0, limit = c(-1,1), space = "Lab") +
  theme_minimal() +
  theme(
    axis.text.x = element_text(angle = 45, vjust = 1, size = 8, hjust = 1),
    axis.text.y = element_text(size = 8)
  ) +
  labs(x = "", y = "", title = "")
```



Chart

Deal with NA's

```
#convert integers to numeric, standardization and replace na with median
heart$age <- as.numeric(heart$age)
heart$age <- scale(heart$age)
heart$age[is.na(heart$age)] <- median(heart$age, na.rm = TRUE)

heart$trestbps <- as.numeric(heart$trestbps)
heart$trestbps <- scale(heart$trestbps)
heart$trestbps[is.na(heart$trestbps)] <- median(heart$trestbps, na.rm = TRUE)

heart$chol <- as.numeric(heart$chol)
heart$chol <- scale(heart$chol)
heart$chol[is.na(heart$chol)] <- median(heart$chol, na.rm = TRUE)

heart$thalach <- as.numeric(heart$thalach)
heart$thalach <- scale(heart$thalach)
heart$thalach[is.na(heart$thalach)] <- median(heart$thalach, na.rm = TRUE)

heart$oldpeak <- as.numeric(heart$oldpeak)
heart$oldpeak <- scale(heart$oldpeak)
heart$oldpeak[is.na(heart$oldpeak)] <- median(heart$oldpeak, na.rm = TRUE)

#replace NA values in the character variables into mode
heart$fbs <- ifelse(is.na(heart$fbs), names(which.max(table(heart$fbs))), heart$fbs)

heart$restecg <- ifelse(is.na(heart$restecg), names(which.max(table(heart$restecg))), heart$restecg)
```

```

heart$exang <- ifelse(is.na(heart$exang), names(which.max(table(heart$exang))), heart$exang)

heart$slope <- ifelse(is.na(heart$slope), names(which.max(table(heart$slope))), heart$slope)

#remove ca & thal, as more than half of their observations are missing values.
heart$ca <- NULL
heart$thal <- NULL

#check for missing values
sum(is.na(heart))

## [1] 0

supply(heart, function(x) sum(is.na(x)))

##      age      sex      cp trestbps      chol      fbs  restecg  thalach
##       0       0       0       0       0       0       0       0
##  exang oldpeak  slope      num
##       0       0       0       0

#convert character variables into factors
heart$sex <- factor(heart$sex,
                    levels = c(0, 1),
                    labels = c("male", "female"))

heart$cp <- factor(heart$cp,
                  levels = c(1, 2, 3, 4),
                  labels = c("typical angina", "atypical angina", "non-anginal pain", "asymptomatic"))

heart$fbs <- factor(heart$fbs,
                  levels = c(0, 1),
                  labels = c("false", "true"))

heart$restecg <- factor(heart$restecg,
                      levels = c(0, 1, 2),
                      labels = c("normal", "abonormal", "left ventricular hypertrophy"))

heart$exang <- factor(heart$exang,
                    levels = c(0, 1),
                    labels = c("no", "yes"))

heart$slope <- factor(heart$slope,
                    levels = c(1, 2, 3),
                    labels = c("upsloping", "flat", "downsloping"))

heart$num <- as.integer(heart$num > 0)
heart$num <- factor(heart$num,
                  levels = c(0, 1),
                  labels = c("no", "yes"))

```

Initial Data Preparation

```
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v forcats 1.0.0      v stringr 1.5.0
## v lubridate 1.9.3    v tibble 3.2.1
## v purrr 1.0.2       v tidyr 1.3.1
## v readr 2.1.4
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors

library(caret)

## Loading required package: lattice
##
## Attaching package: 'caret'
##
## The following object is masked from 'package:purrr':
##
## lift

set.seed(123)

tr_ind <- sample(1:nrow(heart), 0.8 * nrow(heart))
heart_train <- heart[tr_ind, ]
heart_test <- heart[-tr_ind, ]
```

Random Forest & Gradient Boosting

```
# Random Forest
library(randomForest)

## 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
##
## The following object is masked from 'package:dplyr':
##
## combine

library(e1071)
library(caret)

set.seed(123)
rf.hd <- randomForest(num ~ ., data = heart_train, importance=TRUE)
rf.hd

##
## Call:
## randomForest(formula = num ~ ., data = heart_train, importance = TRUE)
##
## Type of random forest: classification
```



```

##                               Number of trees: 500
## No. of variables tried at each split: 3
##
##           OOB estimate of  error rate: 24.69%
## Confusion matrix:
##      no yes class.error
## no  106  25   0.1908397
## yes  34  74   0.3148148

# Split the test dataset into predictors (X_test) and outcome (y_test)
X_test <- heart_test[, -which(names(heart_test) == "num")] # Exclude the outcome variable
y_test <- heart_test$num
# Make predictions
predictions <- predict(rf.hd, X_test)
# Evaluate the model
confusion_matrix <- confusionMatrix(predictions,
                                     y_test,
                                     mode = "everything")
print(confusion_matrix)

## Confusion Matrix and Statistics
##
##           Reference
## Prediction no yes
##      no  27   8
##      yes  2  23
##
##           Accuracy : 0.8333
##           95% CI : (0.7148, 0.9171)
##      No Information Rate : 0.5167
##      P-Value [Acc > NIR] : 3.014e-07
##
##           Kappa : 0.6685
##
##  Mcnemar's Test P-Value : 0.1138
##
##           Sensitivity : 0.9310
##           Specificity : 0.7419
##           Pos Pred Value : 0.7714
##           Neg Pred Value : 0.9200
##           Precision : 0.7714
##           Recall : 0.9310
##           F1 : 0.8438
##           Prevalence : 0.4833
##           Detection Rate : 0.4500
##           Detection Prevalence : 0.5833
##           Balanced Accuracy : 0.8365
##
##           'Positive' Class : no
##

# Accuracy = 0.788
# F1 = 0.7153

# Training Error = 0.1997283

```

```

predict.train <- predict(rf.hd, new_data = heart_train,type= "response")
mean(predict.train != heart_train$num)

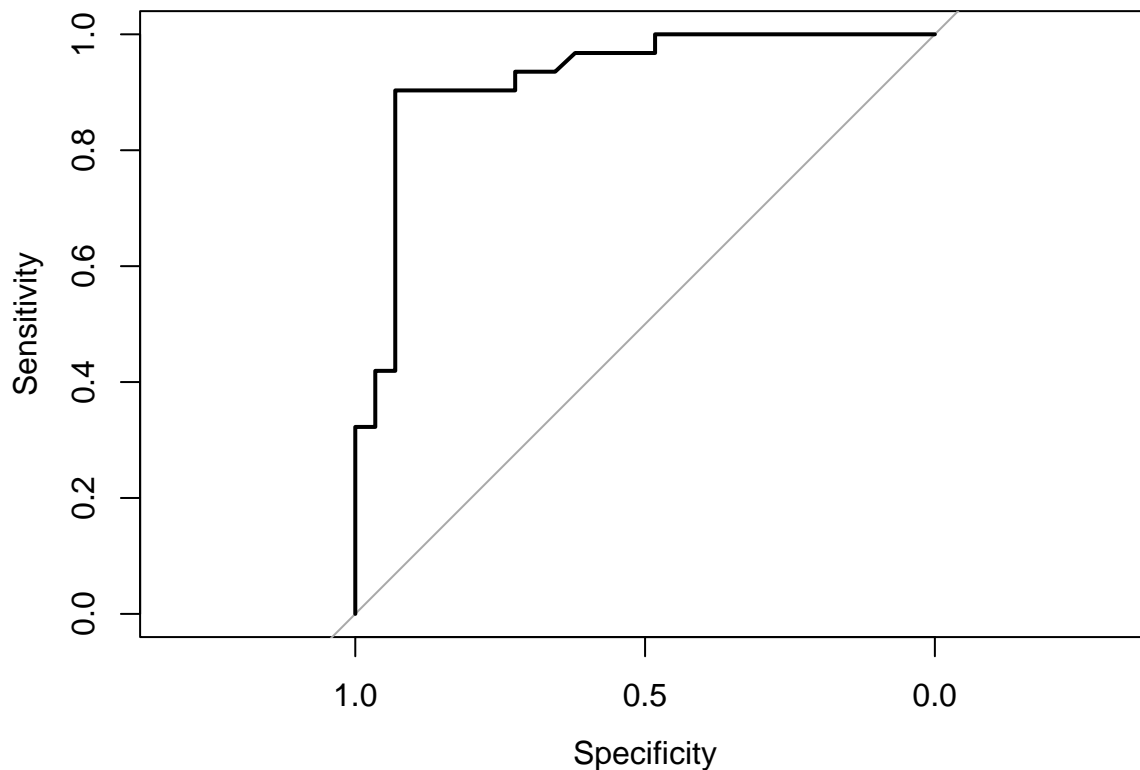
## [1] 0.2468619
# Testing Error = 0.2119565
predict.test <- predict(rf.hd,newdata = heart_test)
mean(predict.test != heart_test$num)

## [1] 0.1666667
# AUCROC
library(pROC)

## Type 'citation("pROC")' for a citation.
##
## Attaching package: 'pROC'
## The following objects are masked from 'package:stats':
##
##      cov, smooth, var
predictions.probab <- predict(rf.hd, X_test, type = "prob")
roc.rf <- roc(response = y_test, predictor = predictions.probab[,2], plot = TRUE)

## Setting levels: control = no, case = yes
## Setting direction: controls < cases

```



```

print(auc(roc.rf))

```

```

## Area under the curve: 0.926

```

```

# auc = 0.8754

#K-fold cross-validation
library(caret)
ctrl <- trainControl(method = "cv", number = 5, search = "grid")
tune_grid <- expand.grid(mtry = c(2, 4, 6, 8))

# Train the Random Forest model
rf_tuned <- train(num ~ .,
                  data = heart_train,      # Training dataset
                  method = "rf",          # Method for Random Forest
                  metric = "Accuracy",     # Metric to optimize
                  trControl = ctrl,        # Training control setup
                  tuneGrid = tune_grid)    # Grid of hyperparameters

rf_tuned

## Random Forest
##
## 239 samples
## 11 predictor
## 2 classes: 'no', 'yes'
##
## No pre-processing
## Resampling: Cross-Validated (5 fold)
## Summary of sample sizes: 192, 191, 191, 191, 191
## Resampling results across tuning parameters:
##
##  mtry  Accuracy  Kappa
##  2     0.7533688 0.4993379
##  4     0.7324468 0.4580624
##  6     0.7449468 0.4847485
##  8     0.7491135 0.4927044
##
## Accuracy was used to select the optimal model using the largest value.
## The final value used for the model was mtry = 2.
#The final value used for the model was mtry = 2.

# Make predictions on the test set using the tuned model
predictions_tuned <- predict(rf_tuned, X_test)

# Evaluate the tuned model
confusion_matrix_tuned <- confusionMatrix(predictions_tuned, y_test,,
                                           mode = "everything")
print(confusion_matrix_tuned)

## Confusion Matrix and Statistics
##
##           Reference
## Prediction no yes
##      no 27  6
##      yes  2 25
##
##           Accuracy : 0.8667
##           95% CI : (0.7541, 0.9406)

```

```
##      No Information Rate : 0.5167
##      P-Value [Acc > NIR] : 1.105e-08
##
##              Kappa : 0.7342
##
##  Mcnemar's Test P-Value : 0.2888
##
##      Sensitivity : 0.9310
##      Specificity : 0.8065
##      Pos Pred Value : 0.8182
##      Neg Pred Value : 0.9259
##      Precision : 0.8182
##      Recall : 0.9310
##      F1 : 0.8710
##      Prevalence : 0.4833
##      Detection Rate : 0.4500
##      Detection Prevalence : 0.5500
##      Balanced Accuracy : 0.8687
##
##      'Positive' Class : no
##
```

```
#Accuracy = 0.8207
```

```
#F1 = 0.7660
```

```
# Training Error = 0.0611413
```

```
predict.train <- predict(rf_tuned, newdata = heart_train)
mean(predict.train != heart_train$num)
```

```
## [1] 0.008368201
```

```
# Testing Error = 0.1793478
```

```
predict.test <- predict(rf_tuned, newdata = heart_test)
mean(predict.test != heart_test$num)
```

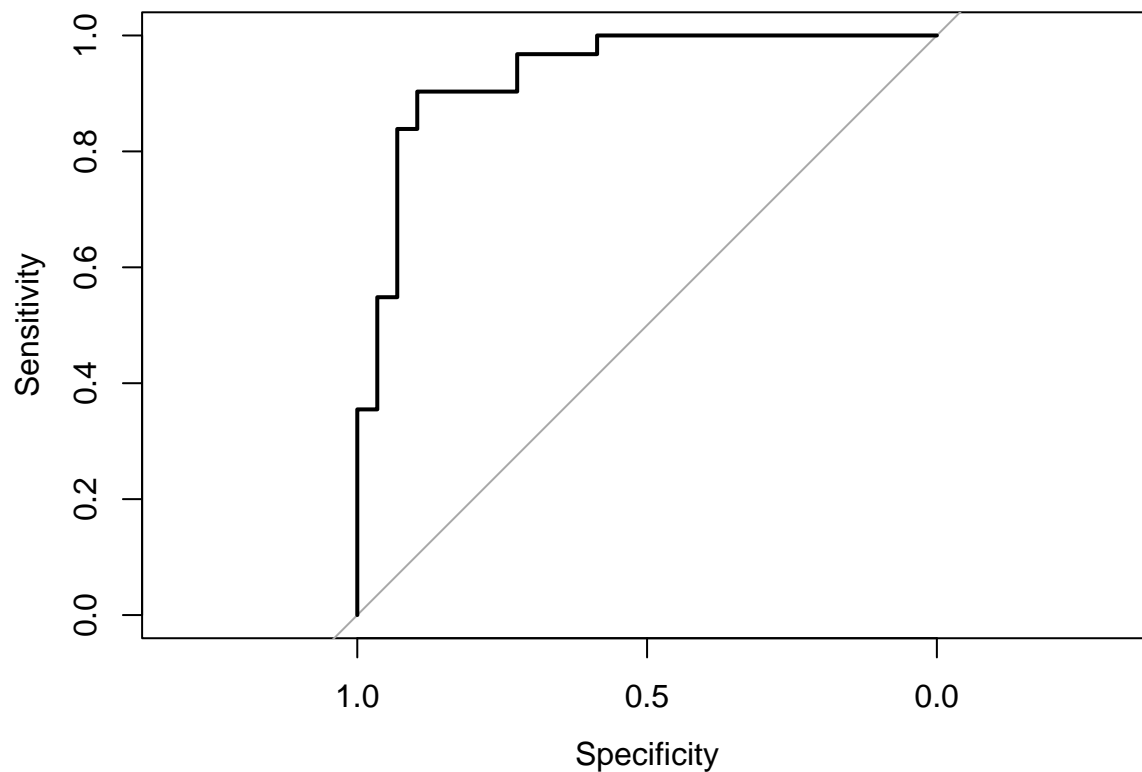
```
## [1] 0.1333333
```

```
# AUCROC
```

```
predictions.prob <- predict(rf_tuned, X_test, type = "prob")
roc.tuned.rf <- roc(response = y_test, predictor = predictions.prob[,2], plot = TRUE)
```

```
## Setting levels: control = no, case = yes
```

```
## Setting direction: controls < cases
```



```
print(auc(roc.tuned.rf))
```

```
## Area under the curve: 0.9355
```

```
# auc = 0.8963
```

```
## Loaded gbm 2.1.9
```

```
## This version of gbm is no longer under development. Consider transitioning to gbm3, https://github.com/gbm/gbm3
```

```
## [1] yes no
```

```
## Levels: no yes
```

```
## [1] yes no
```

```
## Levels: no yes
```

```
## [1] 0.07112971
```

```
## [1] 0.1333333
```

```
## Confusion Matrix and Statistics
```

```
##
```

```
##           Reference
```

```
## Prediction no yes
```

```
##           no  26   5
```

```
##           yes   3  26
```

```
##
```

```
##           Accuracy : 0.8667
```

```
##           95% CI : (0.7541, 0.9406)
```

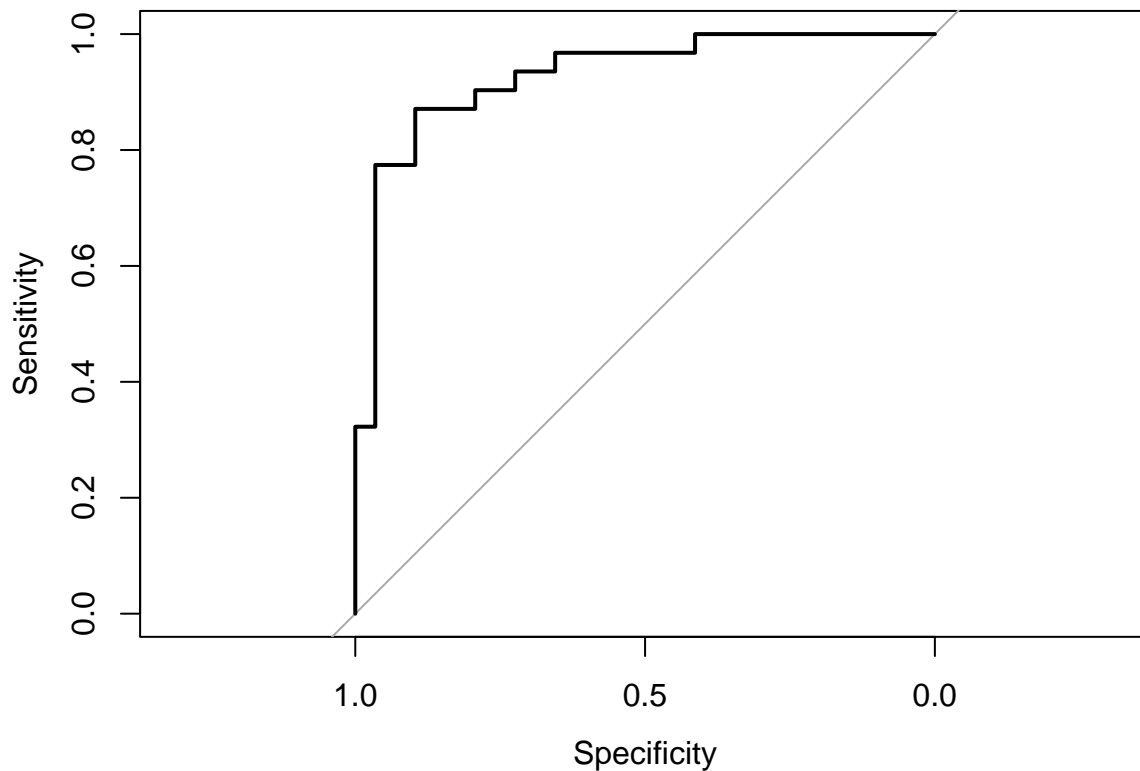
```
##           No Information Rate : 0.5167
```

```
##           P-Value [Acc > NIR] : 1.105e-08
```

```
##
```

```
##           Kappa : 0.7336
```

```
##
## McNemar's Test P-Value : 0.7237
##
##      Sensitivity : 0.8966
##      Specificity : 0.8387
##      Pos Pred Value : 0.8387
##      Neg Pred Value : 0.8966
##      Precision : 0.8387
##      Recall : 0.8966
##      F1 : 0.8667
##      Prevalence : 0.4833
##      Detection Rate : 0.4333
##      Detection Prevalence : 0.5167
##      Balanced Accuracy : 0.8676
##
##      'Positive' Class : no
##
## Using 60 trees...
## Setting levels: control = no, case = yes
## Setting direction: controls < cases
```



```
## Area under the curve: 0.9288
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1         1.3723         nan         0.0100    0.0015
##      2         1.3667         nan         0.0100    0.0021
##      3         1.3612         nan         0.0100    0.0019
##      4         1.3569         nan         0.0100    0.0020
##      5         1.3532         nan         0.0100    0.0014
```

##	6	1.3486	nan	0.0100	0.0020
##	7	1.3451	nan	0.0100	0.0011
##	8	1.3413	nan	0.0100	0.0020
##	9	1.3367	nan	0.0100	0.0016
##	10	1.3328	nan	0.0100	0.0019
##	20	1.2938	nan	0.0100	0.0017
##	40	1.2315	nan	0.0100	0.0008
##	60	1.1824	nan	0.0100	0.0011
##	80	1.1410	nan	0.0100	0.0002
##	100	1.1044	nan	0.0100	0.0007
##	120	1.0736	nan	0.0100	0.0004
##	140	1.0482	nan	0.0100	0.0004
##	150	1.0375	nan	0.0100	0.0002
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3731	nan	0.0100	0.0017
##	2	1.3685	nan	0.0100	0.0014
##	3	1.3633	nan	0.0100	0.0022
##	4	1.3594	nan	0.0100	0.0015
##	5	1.3553	nan	0.0100	0.0012
##	6	1.3514	nan	0.0100	0.0012
##	7	1.3475	nan	0.0100	0.0021
##	8	1.3440	nan	0.0100	0.0015
##	9	1.3406	nan	0.0100	0.0011
##	10	1.3366	nan	0.0100	0.0021
##	20	1.3013	nan	0.0100	0.0020
##	40	1.2381	nan	0.0100	0.0014
##	60	1.1852	nan	0.0100	0.0008
##	80	1.1433	nan	0.0100	0.0006
##	100	1.1080	nan	0.0100	0.0006
##	120	1.0795	nan	0.0100	0.0003
##	140	1.0529	nan	0.0100	0.0002
##	150	1.0415	nan	0.0100	0.0005
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3731	nan	0.0100	0.0018
##	2	1.3681	nan	0.0100	0.0015
##	3	1.3637	nan	0.0100	0.0022
##	4	1.3587	nan	0.0100	0.0021
##	5	1.3548	nan	0.0100	0.0010
##	6	1.3511	nan	0.0100	0.0016
##	7	1.3463	nan	0.0100	0.0021
##	8	1.3430	nan	0.0100	0.0009
##	9	1.3375	nan	0.0100	0.0018
##	10	1.3323	nan	0.0100	0.0017
##	20	1.2941	nan	0.0100	0.0013
##	40	1.2351	nan	0.0100	0.0008
##	60	1.1842	nan	0.0100	0.0005
##	80	1.1409	nan	0.0100	0.0006
##	100	1.1030	nan	0.0100	0.0007
##	120	1.0734	nan	0.0100	0.0004
##	140	1.0504	nan	0.0100	0.0003
##	150	1.0398	nan	0.0100	0.0003
##					

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3701	nan	0.0100	0.0023
##	2	1.3619	nan	0.0100	0.0037
##	3	1.3550	nan	0.0100	0.0022
##	4	1.3478	nan	0.0100	0.0032
##	5	1.3410	nan	0.0100	0.0029
##	6	1.3370	nan	0.0100	0.0012
##	7	1.3319	nan	0.0100	0.0019
##	8	1.3251	nan	0.0100	0.0028
##	9	1.3180	nan	0.0100	0.0026
##	10	1.3139	nan	0.0100	0.0006
##	20	1.2607	nan	0.0100	0.0006
##	40	1.1646	nan	0.0100	0.0013
##	60	1.0902	nan	0.0100	0.0011
##	80	1.0329	nan	0.0100	0.0009
##	100	0.9869	nan	0.0100	0.0002
##	120	0.9455	nan	0.0100	-0.0003
##	140	0.9124	nan	0.0100	0.0004
##	150	0.8967	nan	0.0100	0.0005
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3711	nan	0.0100	0.0020
##	2	1.3637	nan	0.0100	0.0032
##	3	1.3570	nan	0.0100	0.0018
##	4	1.3500	nan	0.0100	0.0033
##	5	1.3426	nan	0.0100	0.0031
##	6	1.3355	nan	0.0100	0.0031
##	7	1.3281	nan	0.0100	0.0028
##	8	1.3226	nan	0.0100	0.0025
##	9	1.3154	nan	0.0100	0.0035
##	10	1.3085	nan	0.0100	0.0023
##	20	1.2537	nan	0.0100	0.0016
##	40	1.1627	nan	0.0100	0.0010
##	60	1.0911	nan	0.0100	0.0013
##	80	1.0360	nan	0.0100	0.0009
##	100	0.9898	nan	0.0100	0.0006
##	120	0.9553	nan	0.0100	0.0003
##	140	0.9242	nan	0.0100	-0.0002
##	150	0.9086	nan	0.0100	-0.0001
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3714	nan	0.0100	0.0022
##	2	1.3654	nan	0.0100	0.0025
##	3	1.3593	nan	0.0100	0.0023
##	4	1.3527	nan	0.0100	0.0027
##	5	1.3457	nan	0.0100	0.0033
##	6	1.3391	nan	0.0100	0.0030
##	7	1.3319	nan	0.0100	0.0033
##	8	1.3267	nan	0.0100	0.0025
##	9	1.3220	nan	0.0100	0.0012
##	10	1.3175	nan	0.0100	0.0017
##	20	1.2651	nan	0.0100	0.0025
##	40	1.1738	nan	0.0100	0.0018
##	60	1.1064	nan	0.0100	0.0006

##	80	1.0545	nan	0.0100	0.0006
##	100	1.0094	nan	0.0100	0.0007
##	120	0.9700	nan	0.0100	-0.0001
##	140	0.9382	nan	0.0100	-0.0001
##	150	0.9248	nan	0.0100	-0.0000
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3697	nan	0.0100	0.0030
##	2	1.3621	nan	0.0100	0.0027
##	3	1.3543	nan	0.0100	0.0027
##	4	1.3462	nan	0.0100	0.0019
##	5	1.3396	nan	0.0100	0.0022
##	6	1.3314	nan	0.0100	0.0030
##	7	1.3229	nan	0.0100	0.0031
##	8	1.3149	nan	0.0100	0.0027
##	9	1.3080	nan	0.0100	0.0025
##	10	1.3000	nan	0.0100	0.0030
##	20	1.2367	nan	0.0100	0.0012
##	40	1.1275	nan	0.0100	0.0009
##	60	1.0432	nan	0.0100	0.0003
##	80	0.9803	nan	0.0100	0.0003
##	100	0.9211	nan	0.0100	0.0009
##	120	0.8741	nan	0.0100	-0.0002
##	140	0.8355	nan	0.0100	0.0001
##	150	0.8166	nan	0.0100	0.0005
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3699	nan	0.0100	0.0025
##	2	1.3615	nan	0.0100	0.0036
##	3	1.3536	nan	0.0100	0.0028
##	4	1.3474	nan	0.0100	0.0016
##	5	1.3390	nan	0.0100	0.0036
##	6	1.3320	nan	0.0100	0.0027
##	7	1.3239	nan	0.0100	0.0029
##	8	1.3160	nan	0.0100	0.0030
##	9	1.3100	nan	0.0100	0.0022
##	10	1.3022	nan	0.0100	0.0033
##	20	1.2438	nan	0.0100	0.0026
##	40	1.1468	nan	0.0100	0.0018
##	60	1.0689	nan	0.0100	0.0014
##	80	1.0045	nan	0.0100	0.0007
##	100	0.9497	nan	0.0100	0.0001
##	120	0.9061	nan	0.0100	0.0003
##	140	0.8686	nan	0.0100	-0.0004
##	150	0.8521	nan	0.0100	0.0000
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3701	nan	0.0100	0.0034
##	2	1.3628	nan	0.0100	0.0026
##	3	1.3558	nan	0.0100	0.0019
##	4	1.3497	nan	0.0100	0.0028
##	5	1.3447	nan	0.0100	0.0016
##	6	1.3383	nan	0.0100	0.0025
##	7	1.3312	nan	0.0100	0.0034

##	8	1.3259	nan	0.0100	0.0013
##	9	1.3201	nan	0.0100	0.0022
##	10	1.3134	nan	0.0100	0.0031
##	20	1.2580	nan	0.0100	0.0015
##	40	1.1698	nan	0.0100	0.0021
##	60	1.0943	nan	0.0100	0.0014
##	80	1.0388	nan	0.0100	0.0011
##	100	0.9924	nan	0.0100	0.0007
##	120	0.9534	nan	0.0100	0.0001
##	140	0.9227	nan	0.0100	-0.0001
##	150	0.9072	nan	0.0100	0.0001

##

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3709	nan	0.0100	0.0027
##	2	1.3649	nan	0.0100	0.0026
##	3	1.3599	nan	0.0100	0.0017
##	4	1.3562	nan	0.0100	0.0014
##	5	1.3518	nan	0.0100	0.0017
##	6	1.3475	nan	0.0100	0.0015
##	7	1.3416	nan	0.0100	0.0025
##	8	1.3378	nan	0.0100	0.0014
##	9	1.3333	nan	0.0100	0.0024
##	10	1.3283	nan	0.0100	0.0025
##	20	1.2856	nan	0.0100	0.0019
##	40	1.2103	nan	0.0100	0.0015
##	60	1.1560	nan	0.0100	0.0005
##	80	1.1095	nan	0.0100	0.0006
##	100	1.0745	nan	0.0100	0.0006
##	120	1.0428	nan	0.0100	0.0006
##	140	1.0154	nan	0.0100	0.0001
##	150	1.0046	nan	0.0100	0.0003

##

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3725	nan	0.0100	0.0013
##	2	1.3669	nan	0.0100	0.0027
##	3	1.3623	nan	0.0100	0.0026
##	4	1.3579	nan	0.0100	0.0015
##	5	1.3528	nan	0.0100	0.0026
##	6	1.3475	nan	0.0100	0.0025
##	7	1.3435	nan	0.0100	0.0017
##	8	1.3387	nan	0.0100	0.0024
##	9	1.3345	nan	0.0100	0.0017
##	10	1.3294	nan	0.0100	0.0026
##	20	1.2861	nan	0.0100	0.0020
##	40	1.2138	nan	0.0100	0.0012
##	60	1.1570	nan	0.0100	0.0005
##	80	1.1123	nan	0.0100	0.0006
##	100	1.0740	nan	0.0100	0.0007
##	120	1.0428	nan	0.0100	0.0002
##	140	1.0155	nan	0.0100	0.0003
##	150	1.0039	nan	0.0100	0.0005

##

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3722	nan	0.0100	0.0017

##	2	1.3685	nan	0.0100	0.0017
##	3	1.3632	nan	0.0100	0.0027
##	4	1.3572	nan	0.0100	0.0024
##	5	1.3527	nan	0.0100	0.0020
##	6	1.3491	nan	0.0100	0.0009
##	7	1.3437	nan	0.0100	0.0025
##	8	1.3386	nan	0.0100	0.0025
##	9	1.3341	nan	0.0100	0.0016
##	10	1.3288	nan	0.0100	0.0024
##	20	1.2842	nan	0.0100	0.0020
##	40	1.2118	nan	0.0100	0.0016
##	60	1.1566	nan	0.0100	0.0007
##	80	1.1128	nan	0.0100	0.0006
##	100	1.0756	nan	0.0100	0.0006
##	120	1.0438	nan	0.0100	0.0004
##	140	1.0180	nan	0.0100	0.0003
##	150	1.0058	nan	0.0100	0.0006
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3691	nan	0.0100	0.0035
##	2	1.3606	nan	0.0100	0.0039
##	3	1.3527	nan	0.0100	0.0035
##	4	1.3468	nan	0.0100	0.0021
##	5	1.3393	nan	0.0100	0.0028
##	6	1.3313	nan	0.0100	0.0032
##	7	1.3235	nan	0.0100	0.0030
##	8	1.3161	nan	0.0100	0.0032
##	9	1.3084	nan	0.0100	0.0039
##	10	1.3027	nan	0.0100	0.0029
##	20	1.2375	nan	0.0100	0.0021
##	40	1.1392	nan	0.0100	0.0018
##	60	1.0599	nan	0.0100	0.0012
##	80	0.9967	nan	0.0100	0.0008
##	100	0.9457	nan	0.0100	0.0007
##	120	0.8977	nan	0.0100	0.0011
##	140	0.8598	nan	0.0100	0.0002
##	150	0.8436	nan	0.0100	0.0004
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3700	nan	0.0100	0.0024
##	2	1.3614	nan	0.0100	0.0040
##	3	1.3532	nan	0.0100	0.0038
##	4	1.3462	nan	0.0100	0.0031
##	5	1.3395	nan	0.0100	0.0026
##	6	1.3325	nan	0.0100	0.0030
##	7	1.3247	nan	0.0100	0.0033
##	8	1.3178	nan	0.0100	0.0034
##	9	1.3101	nan	0.0100	0.0027
##	10	1.3034	nan	0.0100	0.0032
##	20	1.2403	nan	0.0100	0.0028
##	40	1.1370	nan	0.0100	0.0017
##	60	1.0596	nan	0.0100	0.0013
##	80	1.0021	nan	0.0100	0.0005
##	100	0.9521	nan	0.0100	0.0004

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##      120      0.9115      nan      0.0100      0.0004
##      140      0.8760      nan      0.0100      0.0003
##      150      0.8602      nan      0.0100     -0.0002
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.3684      nan      0.0100      0.0037
##      2      1.3610      nan      0.0100      0.0033
##      3      1.3543      nan      0.0100      0.0026
##      4      1.3462      nan      0.0100      0.0036
##      5      1.3391      nan      0.0100      0.0031
##      6      1.3313      nan      0.0100      0.0037
##      7      1.3239      nan      0.0100      0.0038
##      8      1.3164      nan      0.0100      0.0036
##      9      1.3089      nan      0.0100      0.0036
##     10      1.3027      nan      0.0100      0.0030
##     20      1.2417      nan      0.0100      0.0027
##     40      1.1400      nan      0.0100      0.0015
##     60      1.0673      nan      0.0100      0.0013
##     80      1.0041      nan      0.0100      0.0010
##    100      0.9557      nan      0.0100      0.0005
##    120      0.9120      nan      0.0100      0.0008
##    140      0.8780      nan      0.0100      0.0000
##    150      0.8607      nan      0.0100      0.0000
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.3675      nan      0.0100      0.0036
##      2      1.3595      nan      0.0100      0.0030
##      3      1.3507      nan      0.0100      0.0035
##      4      1.3418      nan      0.0100      0.0035
##      5      1.3339      nan      0.0100      0.0027
##      6      1.3249      nan      0.0100      0.0038
##      7      1.3174      nan      0.0100      0.0038
##      8      1.3090      nan      0.0100      0.0034
##      9      1.2998      nan      0.0100      0.0032
##     10      1.2926      nan      0.0100      0.0028
##     20      1.2155      nan      0.0100      0.0024
##     40      1.0921      nan      0.0100      0.0015
##     60      1.0034      nan      0.0100      0.0016
##     80      0.9339      nan      0.0100      0.0005
##    100      0.8725      nan      0.0100      0.0002
##    120      0.8212      nan      0.0100      0.0006
##    140      0.7746      nan      0.0100     -0.0002
##    150      0.7529      nan      0.0100      0.0002
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.3675      nan      0.0100      0.0038
##      2      1.3590      nan      0.0100      0.0033
##      3      1.3509      nan      0.0100      0.0034
##      4      1.3429      nan      0.0100      0.0028
##      5      1.3340      nan      0.0100      0.0038
##      6      1.3246      nan      0.0100      0.0034
##      7      1.3161      nan      0.0100      0.0040
##      8      1.3088      nan      0.0100      0.0023
##      9      1.3002      nan      0.0100      0.0039

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##	10	1.2915	nan	0.0100	0.0040
##	20	1.2228	nan	0.0100	0.0017
##	40	1.1107	nan	0.0100	0.0012
##	60	1.0239	nan	0.0100	0.0007
##	80	0.9550	nan	0.0100	0.0004
##	100	0.8951	nan	0.0100	0.0012
##	120	0.8466	nan	0.0100	0.0003
##	140	0.8100	nan	0.0100	-0.0001
##	150	0.7932	nan	0.0100	0.0002
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3695	nan	0.0100	0.0033
##	2	1.3612	nan	0.0100	0.0028
##	3	1.3528	nan	0.0100	0.0039
##	4	1.3444	nan	0.0100	0.0034
##	5	1.3391	nan	0.0100	0.0019
##	6	1.3322	nan	0.0100	0.0031
##	7	1.3260	nan	0.0100	0.0029
##	8	1.3188	nan	0.0100	0.0027
##	9	1.3110	nan	0.0100	0.0026
##	10	1.3045	nan	0.0100	0.0029
##	20	1.2382	nan	0.0100	0.0029
##	40	1.1427	nan	0.0100	0.0019
##	60	1.0622	nan	0.0100	0.0008
##	80	0.9989	nan	0.0100	0.0004
##	100	0.9505	nan	0.0100	0.0002
##	120	0.9089	nan	0.0100	0.0005
##	140	0.8708	nan	0.0100	-0.0002
##	150	0.8548	nan	0.0100	-0.0002
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3743	nan	0.0100	0.0015
##	2	1.3710	nan	0.0100	0.0013
##	3	1.3672	nan	0.0100	0.0017
##	4	1.3633	nan	0.0100	0.0022
##	5	1.3593	nan	0.0100	0.0011
##	6	1.3550	nan	0.0100	0.0021
##	7	1.3519	nan	0.0100	0.0010
##	8	1.3491	nan	0.0100	0.0010
##	9	1.3450	nan	0.0100	0.0021
##	10	1.3402	nan	0.0100	0.0020
##	20	1.3046	nan	0.0100	0.0011
##	40	1.2401	nan	0.0100	0.0013
##	60	1.1920	nan	0.0100	0.0011
##	80	1.1484	nan	0.0100	0.0008
##	100	1.1143	nan	0.0100	0.0008
##	120	1.0829	nan	0.0100	0.0004
##	140	1.0572	nan	0.0100	0.0005
##	150	1.0441	nan	0.0100	0.0004
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3719	nan	0.0100	0.0022
##	2	1.3674	nan	0.0100	0.0020
##	3	1.3628	nan	0.0100	0.0021

##	4	1.3590	nan	0.0100	0.0013
##	5	1.3559	nan	0.0100	0.0013
##	6	1.3517	nan	0.0100	0.0021
##	7	1.3483	nan	0.0100	0.0016
##	8	1.3452	nan	0.0100	0.0015
##	9	1.3414	nan	0.0100	0.0018
##	10	1.3373	nan	0.0100	0.0013
##	20	1.3046	nan	0.0100	0.0017
##	40	1.2454	nan	0.0100	0.0014
##	60	1.1955	nan	0.0100	0.0003
##	80	1.1513	nan	0.0100	0.0003
##	100	1.1153	nan	0.0100	0.0001
##	120	1.0855	nan	0.0100	0.0006
##	140	1.0584	nan	0.0100	0.0003
##	150	1.0471	nan	0.0100	-0.0002

##

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3728	nan	0.0100	0.0022
##	2	1.3691	nan	0.0100	0.0008
##	3	1.3654	nan	0.0100	0.0015
##	4	1.3610	nan	0.0100	0.0021
##	5	1.3578	nan	0.0100	0.0010
##	6	1.3536	nan	0.0100	0.0021
##	7	1.3486	nan	0.0100	0.0021
##	8	1.3445	nan	0.0100	0.0020
##	9	1.3409	nan	0.0100	0.0006
##	10	1.3372	nan	0.0100	0.0019
##	20	1.3023	nan	0.0100	0.0012
##	40	1.2411	nan	0.0100	0.0010
##	60	1.1936	nan	0.0100	0.0005
##	80	1.1541	nan	0.0100	0.0007
##	100	1.1190	nan	0.0100	0.0007
##	120	1.0878	nan	0.0100	0.0005
##	140	1.0641	nan	0.0100	-0.0005
##	150	1.0514	nan	0.0100	0.0004

##

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3710	nan	0.0100	0.0031
##	2	1.3648	nan	0.0100	0.0032
##	3	1.3577	nan	0.0100	0.0022
##	4	1.3517	nan	0.0100	0.0021
##	5	1.3441	nan	0.0100	0.0030
##	6	1.3368	nan	0.0100	0.0030
##	7	1.3311	nan	0.0100	0.0016
##	8	1.3251	nan	0.0100	0.0026
##	9	1.3193	nan	0.0100	0.0019
##	10	1.3135	nan	0.0100	0.0024
##	20	1.2591	nan	0.0100	0.0023
##	40	1.1678	nan	0.0100	0.0017
##	60	1.0957	nan	0.0100	0.0008
##	80	1.0369	nan	0.0100	0.0010
##	100	0.9855	nan	0.0100	0.0005
##	120	0.9437	nan	0.0100	0.0005
##	140	0.9042	nan	0.0100	0.0001

```

##      150      0.8872      nan      0.0100      0.0002
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.3729      nan      0.0100      0.0013
##      2      1.3665      nan      0.0100      0.0025
##      3      1.3585      nan      0.0100      0.0027
##      4      1.3521      nan      0.0100      0.0023
##      5      1.3469      nan      0.0100      0.0014
##      6      1.3398      nan      0.0100      0.0031
##      7      1.3329      nan      0.0100      0.0024
##      8      1.3262      nan      0.0100      0.0023
##      9      1.3198      nan      0.0100      0.0025
##     10      1.3127      nan      0.0100      0.0027
##     20      1.2568      nan      0.0100      0.0019
##     40      1.1694      nan      0.0100      0.0009
##     60      1.0949      nan      0.0100      0.0012
##     80      1.0352      nan      0.0100      0.0007
##    100      0.9856      nan      0.0100      0.0001
##    120      0.9435      nan      0.0100      0.0005
##    140      0.9104      nan      0.0100      0.0000
##    150      0.8935      nan      0.0100      0.0003
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.3716      nan      0.0100      0.0016
##      2      1.3665      nan      0.0100      0.0017
##      3      1.3611      nan      0.0100      0.0014
##      4      1.3542      nan      0.0100      0.0034
##      5      1.3477      nan      0.0100      0.0021
##      6      1.3423      nan      0.0100      0.0018
##      7      1.3363      nan      0.0100      0.0026
##      8      1.3288      nan      0.0100      0.0027
##      9      1.3235      nan      0.0100      0.0016
##     10      1.3174      nan      0.0100      0.0019
##     20      1.2606      nan      0.0100      0.0009
##     40      1.1697      nan      0.0100      0.0016
##     60      1.0984      nan      0.0100      0.0005
##     80      1.0391      nan      0.0100      0.0010
##    100      0.9910      nan      0.0100      0.0002
##    120      0.9528      nan      0.0100      0.0007
##    140      0.9180      nan      0.0100      0.0001
##    150      0.9012      nan      0.0100      0.0002
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.3699      nan      0.0100      0.0027
##      2      1.3622      nan      0.0100      0.0032
##      3      1.3534      nan      0.0100      0.0038
##      4      1.3450      nan      0.0100      0.0030
##      5      1.3370      nan      0.0100      0.0030
##      6      1.3291      nan      0.0100      0.0031
##      7      1.3208      nan      0.0100      0.0032
##      8      1.3145      nan      0.0100      0.0022
##      9      1.3073      nan      0.0100      0.0022
##     10      1.2988      nan      0.0100      0.0032
##     20      1.2362      nan      0.0100      0.0005

```

##	40	1.1268	nan	0.0100	0.0010
##	60	1.0346	nan	0.0100	0.0013
##	80	0.9616	nan	0.0100	0.0003
##	100	0.9015	nan	0.0100	0.0002
##	120	0.8512	nan	0.0100	0.0008
##	140	0.8063	nan	0.0100	-0.0006
##	150	0.7857	nan	0.0100	-0.0002
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3697	nan	0.0100	0.0031
##	2	1.3637	nan	0.0100	0.0013
##	3	1.3564	nan	0.0100	0.0027
##	4	1.3487	nan	0.0100	0.0027
##	5	1.3425	nan	0.0100	0.0020
##	6	1.3345	nan	0.0100	0.0035
##	7	1.3274	nan	0.0100	0.0026
##	8	1.3206	nan	0.0100	0.0024
##	9	1.3138	nan	0.0100	0.0024
##	10	1.3081	nan	0.0100	0.0020
##	20	1.2429	nan	0.0100	0.0020
##	40	1.1362	nan	0.0100	0.0014
##	60	1.0523	nan	0.0100	0.0008
##	80	0.9864	nan	0.0100	0.0005
##	100	0.9332	nan	0.0100	0.0000
##	120	0.8845	nan	0.0100	0.0011
##	140	0.8453	nan	0.0100	0.0003
##	150	0.8268	nan	0.0100	0.0002
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3705	nan	0.0100	0.0033
##	2	1.3649	nan	0.0100	0.0022
##	3	1.3583	nan	0.0100	0.0022
##	4	1.3512	nan	0.0100	0.0028
##	5	1.3466	nan	0.0100	0.0015
##	6	1.3397	nan	0.0100	0.0028
##	7	1.3339	nan	0.0100	0.0022
##	8	1.3270	nan	0.0100	0.0022
##	9	1.3212	nan	0.0100	0.0020
##	10	1.3153	nan	0.0100	0.0022
##	20	1.2604	nan	0.0100	0.0020
##	40	1.1626	nan	0.0100	0.0017
##	60	1.0927	nan	0.0100	0.0016
##	80	1.0310	nan	0.0100	0.0012
##	100	0.9815	nan	0.0100	0.0001
##	120	0.9415	nan	0.0100	-0.0005
##	140	0.9043	nan	0.0100	0.0004
##	150	0.8872	nan	0.0100	0.0002
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3731	nan	0.0100	0.0016
##	2	1.3685	nan	0.0100	0.0021
##	3	1.3635	nan	0.0100	0.0019
##	4	1.3597	nan	0.0100	0.0014
##	5	1.3555	nan	0.0100	0.0019

##	6	1.3520	nan	0.0100	0.0015
##	7	1.3485	nan	0.0100	0.0013
##	8	1.3437	nan	0.0100	0.0017
##	9	1.3394	nan	0.0100	0.0020
##	10	1.3352	nan	0.0100	0.0019
##	20	1.2998	nan	0.0100	0.0009
##	40	1.2390	nan	0.0100	0.0010
##	60	1.1896	nan	0.0100	0.0008
##	80	1.1505	nan	0.0100	-0.0004
##	100	1.1147	nan	0.0100	0.0002
##	120	1.0826	nan	0.0100	0.0003
##	140	1.0560	nan	0.0100	0.0001
##	150	1.0443	nan	0.0100	0.0004
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3727	nan	0.0100	0.0017
##	2	1.3686	nan	0.0100	0.0017
##	3	1.3638	nan	0.0100	0.0021
##	4	1.3599	nan	0.0100	0.0014
##	5	1.3553	nan	0.0100	0.0020
##	6	1.3510	nan	0.0100	0.0019
##	7	1.3474	nan	0.0100	0.0020
##	8	1.3436	nan	0.0100	0.0019
##	9	1.3395	nan	0.0100	0.0014
##	10	1.3352	nan	0.0100	0.0012
##	20	1.2980	nan	0.0100	0.0011
##	40	1.2358	nan	0.0100	0.0013
##	60	1.1863	nan	0.0100	0.0009
##	80	1.1431	nan	0.0100	0.0001
##	100	1.1072	nan	0.0100	0.0006
##	120	1.0784	nan	0.0100	0.0004
##	140	1.0527	nan	0.0100	0.0002
##	150	1.0411	nan	0.0100	0.0002
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3710	nan	0.0100	0.0020
##	2	1.3675	nan	0.0100	0.0016
##	3	1.3633	nan	0.0100	0.0011
##	4	1.3589	nan	0.0100	0.0018
##	5	1.3551	nan	0.0100	0.0016
##	6	1.3515	nan	0.0100	0.0016
##	7	1.3481	nan	0.0100	0.0009
##	8	1.3442	nan	0.0100	0.0019
##	9	1.3413	nan	0.0100	0.0006
##	10	1.3378	nan	0.0100	0.0015
##	20	1.3033	nan	0.0100	0.0011
##	40	1.2426	nan	0.0100	0.0013
##	60	1.1923	nan	0.0100	0.0011
##	80	1.1476	nan	0.0100	0.0008
##	100	1.1131	nan	0.0100	0.0006
##	120	1.0814	nan	0.0100	0.0004
##	140	1.0538	nan	0.0100	0.0001
##	150	1.0420	nan	0.0100	0.0004
##					

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3683	nan	0.0100	0.0032
##	2	1.3605	nan	0.0100	0.0030
##	3	1.3546	nan	0.0100	0.0019
##	4	1.3471	nan	0.0100	0.0028
##	5	1.3400	nan	0.0100	0.0033
##	6	1.3345	nan	0.0100	0.0020
##	7	1.3281	nan	0.0100	0.0019
##	8	1.3210	nan	0.0100	0.0024
##	9	1.3137	nan	0.0100	0.0027
##	10	1.3081	nan	0.0100	0.0022
##	20	1.2511	nan	0.0100	0.0024
##	40	1.1579	nan	0.0100	0.0001
##	60	1.0836	nan	0.0100	0.0014
##	80	1.0215	nan	0.0100	0.0006
##	100	0.9698	nan	0.0100	0.0004
##	120	0.9259	nan	0.0100	-0.0000
##	140	0.8886	nan	0.0100	0.0004
##	150	0.8728	nan	0.0100	0.0000
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3708	nan	0.0100	0.0014
##	2	1.3633	nan	0.0100	0.0038
##	3	1.3569	nan	0.0100	0.0022
##	4	1.3499	nan	0.0100	0.0026
##	5	1.3434	nan	0.0100	0.0026
##	6	1.3355	nan	0.0100	0.0027
##	7	1.3293	nan	0.0100	0.0016
##	8	1.3218	nan	0.0100	0.0034
##	9	1.3160	nan	0.0100	0.0020
##	10	1.3091	nan	0.0100	0.0029
##	20	1.2501	nan	0.0100	0.0012
##	40	1.1552	nan	0.0100	0.0017
##	60	1.0824	nan	0.0100	0.0003
##	80	1.0227	nan	0.0100	0.0012
##	100	0.9751	nan	0.0100	0.0008
##	120	0.9341	nan	0.0100	0.0000
##	140	0.8994	nan	0.0100	0.0004
##	150	0.8823	nan	0.0100	-0.0001
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3702	nan	0.0100	0.0030
##	2	1.3651	nan	0.0100	0.0015
##	3	1.3588	nan	0.0100	0.0028
##	4	1.3536	nan	0.0100	0.0018
##	5	1.3468	nan	0.0100	0.0032
##	6	1.3403	nan	0.0100	0.0019
##	7	1.3325	nan	0.0100	0.0035
##	8	1.3256	nan	0.0100	0.0035
##	9	1.3194	nan	0.0100	0.0022
##	10	1.3139	nan	0.0100	0.0021
##	20	1.2565	nan	0.0100	0.0021
##	40	1.1614	nan	0.0100	0.0011
##	60	1.0893	nan	0.0100	0.0010

##	80	1.0298	nan	0.0100	0.0008
##	100	0.9795	nan	0.0100	0.0009
##	120	0.9383	nan	0.0100	0.0007
##	140	0.9038	nan	0.0100	-0.0003
##	150	0.8862	nan	0.0100	0.0005
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3682	nan	0.0100	0.0028
##	2	1.3605	nan	0.0100	0.0025
##	3	1.3517	nan	0.0100	0.0028
##	4	1.3439	nan	0.0100	0.0033
##	5	1.3363	nan	0.0100	0.0032
##	6	1.3286	nan	0.0100	0.0026
##	7	1.3208	nan	0.0100	0.0030
##	8	1.3130	nan	0.0100	0.0035
##	9	1.3054	nan	0.0100	0.0028
##	10	1.2984	nan	0.0100	0.0023
##	20	1.2290	nan	0.0100	0.0014
##	40	1.1183	nan	0.0100	0.0012
##	60	1.0264	nan	0.0100	0.0012
##	80	0.9541	nan	0.0100	0.0005
##	100	0.8938	nan	0.0100	0.0006
##	120	0.8432	nan	0.0100	-0.0001
##	140	0.7953	nan	0.0100	-0.0001
##	150	0.7768	nan	0.0100	0.0006
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3683	nan	0.0100	0.0033
##	2	1.3608	nan	0.0100	0.0029
##	3	1.3541	nan	0.0100	0.0028
##	4	1.3466	nan	0.0100	0.0029
##	5	1.3381	nan	0.0100	0.0038
##	6	1.3309	nan	0.0100	0.0030
##	7	1.3248	nan	0.0100	0.0022
##	8	1.3187	nan	0.0100	0.0021
##	9	1.3117	nan	0.0100	0.0035
##	10	1.3043	nan	0.0100	0.0034
##	20	1.2357	nan	0.0100	0.0020
##	40	1.1307	nan	0.0100	0.0016
##	60	1.0508	nan	0.0100	0.0001
##	80	0.9823	nan	0.0100	0.0010
##	100	0.9245	nan	0.0100	0.0005
##	120	0.8771	nan	0.0100	-0.0001
##	140	0.8325	nan	0.0100	0.0004
##	150	0.8131	nan	0.0100	-0.0000
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3689	nan	0.0100	0.0022
##	2	1.3627	nan	0.0100	0.0021
##	3	1.3554	nan	0.0100	0.0031
##	4	1.3488	nan	0.0100	0.0019
##	5	1.3419	nan	0.0100	0.0025
##	6	1.3363	nan	0.0100	0.0017
##	7	1.3303	nan	0.0100	0.0022

##	8	1.3230	nan	0.0100	0.0031
##	9	1.3158	nan	0.0100	0.0029
##	10	1.3090	nan	0.0100	0.0032
##	20	1.2522	nan	0.0100	0.0023
##	40	1.1528	nan	0.0100	0.0013
##	60	1.0806	nan	0.0100	0.0008
##	80	1.0247	nan	0.0100	0.0011
##	100	0.9745	nan	0.0100	0.0002
##	120	0.9311	nan	0.0100	0.0000
##	140	0.8956	nan	0.0100	-0.0003
##	150	0.8797	nan	0.0100	-0.0001

##

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3728	nan	0.0100	0.0025
##	2	1.3678	nan	0.0100	0.0024
##	3	1.3633	nan	0.0100	0.0014
##	4	1.3579	nan	0.0100	0.0023
##	5	1.3537	nan	0.0100	0.0023
##	6	1.3490	nan	0.0100	0.0023
##	7	1.3445	nan	0.0100	0.0023
##	8	1.3412	nan	0.0100	0.0013
##	9	1.3362	nan	0.0100	0.0022
##	10	1.3310	nan	0.0100	0.0021
##	20	1.2893	nan	0.0100	0.0017
##	40	1.2261	nan	0.0100	0.0009
##	60	1.1728	nan	0.0100	0.0011
##	80	1.1312	nan	0.0100	0.0005
##	100	1.0971	nan	0.0100	0.0002
##	120	1.0661	nan	0.0100	0.0004
##	140	1.0389	nan	0.0100	-0.0002
##	150	1.0265	nan	0.0100	0.0003

##

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3730	nan	0.0100	0.0015
##	2	1.3686	nan	0.0100	0.0018
##	3	1.3634	nan	0.0100	0.0025
##	4	1.3594	nan	0.0100	0.0014
##	5	1.3574	nan	0.0100	0.0005
##	6	1.3550	nan	0.0100	0.0007
##	7	1.3499	nan	0.0100	0.0024
##	8	1.3451	nan	0.0100	0.0023
##	9	1.3406	nan	0.0100	0.0023
##	10	1.3363	nan	0.0100	0.0021
##	20	1.2954	nan	0.0100	0.0019
##	40	1.2300	nan	0.0100	0.0014
##	60	1.1766	nan	0.0100	0.0011
##	80	1.1328	nan	0.0100	0.0007
##	100	1.0956	nan	0.0100	0.0003
##	120	1.0640	nan	0.0100	0.0006
##	140	1.0377	nan	0.0100	0.0004
##	150	1.0262	nan	0.0100	0.0002

##

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3713	nan	0.0100	0.0024

##	2	1.3662	nan	0.0100	0.0023
##	3	1.3613	nan	0.0100	0.0016
##	4	1.3580	nan	0.0100	0.0012
##	5	1.3541	nan	0.0100	0.0014
##	6	1.3493	nan	0.0100	0.0024
##	7	1.3441	nan	0.0100	0.0023
##	8	1.3383	nan	0.0100	0.0021
##	9	1.3337	nan	0.0100	0.0022
##	10	1.3293	nan	0.0100	0.0022
##	20	1.2901	nan	0.0100	0.0013
##	40	1.2262	nan	0.0100	0.0009
##	60	1.1721	nan	0.0100	0.0009
##	80	1.1258	nan	0.0100	0.0006
##	100	1.0917	nan	0.0100	0.0005
##	120	1.0605	nan	0.0100	0.0004
##	140	1.0351	nan	0.0100	0.0003
##	150	1.0231	nan	0.0100	0.0004
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3702	nan	0.0100	0.0031
##	2	1.3632	nan	0.0100	0.0022
##	3	1.3558	nan	0.0100	0.0032
##	4	1.3485	nan	0.0100	0.0027
##	5	1.3425	nan	0.0100	0.0027
##	6	1.3355	nan	0.0100	0.0031
##	7	1.3289	nan	0.0100	0.0031
##	8	1.3226	nan	0.0100	0.0029
##	9	1.3156	nan	0.0100	0.0031
##	10	1.3089	nan	0.0100	0.0030
##	20	1.2525	nan	0.0100	0.0021
##	40	1.1559	nan	0.0100	0.0013
##	60	1.0812	nan	0.0100	0.0012
##	80	1.0191	nan	0.0100	0.0003
##	100	0.9653	nan	0.0100	0.0001
##	120	0.9189	nan	0.0100	0.0008
##	140	0.8805	nan	0.0100	-0.0003
##	150	0.8620	nan	0.0100	0.0005
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1.3709	nan	0.0100	0.0027
##	2	1.3652	nan	0.0100	0.0021
##	3	1.3564	nan	0.0100	0.0032
##	4	1.3488	nan	0.0100	0.0029
##	5	1.3429	nan	0.0100	0.0017
##	6	1.3378	nan	0.0100	0.0017
##	7	1.3307	nan	0.0100	0.0030
##	8	1.3236	nan	0.0100	0.0028
##	9	1.3167	nan	0.0100	0.0026
##	10	1.3094	nan	0.0100	0.0024
##	20	1.2521	nan	0.0100	0.0024
##	40	1.1563	nan	0.0100	0.0014
##	60	1.0824	nan	0.0100	0.0008
##	80	1.0235	nan	0.0100	0.0004
##	100	0.9717	nan	0.0100	0.0002

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##      120      0.9302      nan      0.0100      0.0006
##      140      0.8955      nan      0.0100      0.0002
##      150      0.8787      nan      0.0100      0.0002
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.3700      nan      0.0100      0.0031
##      2      1.3623      nan      0.0100      0.0027
##      3      1.3557      nan      0.0100      0.0030
##      4      1.3474      nan      0.0100      0.0037
##      5      1.3419      nan      0.0100      0.0023
##      6      1.3347      nan      0.0100      0.0037
##      7      1.3282      nan      0.0100      0.0029
##      8      1.3221      nan      0.0100      0.0022
##      9      1.3151      nan      0.0100      0.0033
##     10      1.3090      nan      0.0100      0.0027
##     20      1.2501      nan      0.0100      0.0016
##     40      1.1544      nan      0.0100      0.0015
##     60      1.0799      nan      0.0100      0.0011
##     80      1.0230      nan      0.0100      0.0008
##    100      0.9753      nan      0.0100      0.0004
##    120      0.9359      nan      0.0100     -0.0001
##    140      0.9028      nan      0.0100      0.0001
##    150      0.8883      nan      0.0100     -0.0000
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.3678      nan      0.0100      0.0038
##      2      1.3603      nan      0.0100      0.0025
##      3      1.3510      nan      0.0100      0.0032
##      4      1.3432      nan      0.0100      0.0033
##      5      1.3345      nan      0.0100      0.0035
##      6      1.3284      nan      0.0100      0.0021
##      7      1.3213      nan      0.0100      0.0024
##      8      1.3141      nan      0.0100      0.0023
##      9      1.3066      nan      0.0100      0.0029
##     10      1.2998      nan      0.0100      0.0020
##     20      1.2301      nan      0.0100      0.0026
##     40      1.1166      nan      0.0100      0.0013
##     60      1.0232      nan      0.0100      0.0015
##     80      0.9488      nan      0.0100      0.0005
##    100      0.8926      nan      0.0100      0.0005
##    120      0.8404      nan      0.0100      0.0004
##    140      0.7948      nan      0.0100      0.0003
##    150      0.7748      nan      0.0100      0.0004
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.3695      nan      0.0100      0.0030
##      2      1.3609      nan      0.0100      0.0032
##      3      1.3540      nan      0.0100      0.0025
##      4      1.3454      nan      0.0100      0.0042
##      5      1.3384      nan      0.0100      0.0024
##      6      1.3302      nan      0.0100      0.0029
##      7      1.3233      nan      0.0100      0.0029
##      8      1.3153      nan      0.0100      0.0032
##      9      1.3084      nan      0.0100      0.0022

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##      10      1.3017      nan      0.0100      0.0023
##      20      1.2367      nan      0.0100      0.0024
##      40      1.1306      nan      0.0100      0.0019
##      60      1.0479      nan      0.0100      0.0005
##      80      0.9795      nan      0.0100      0.0012
##     100      0.9248      nan      0.0100      0.0011
##     120      0.8766      nan      0.0100      0.0002
##     140      0.8349      nan      0.0100     -0.0004
##     150      0.8177      nan      0.0100     -0.0002
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.3709      nan      0.0100      0.0029
##      2      1.3629      nan      0.0100      0.0037
##      3      1.3555      nan      0.0100      0.0030
##      4      1.3490      nan      0.0100      0.0021
##      5      1.3427      nan      0.0100      0.0024
##      6      1.3359      nan      0.0100      0.0025
##      7      1.3295      nan      0.0100      0.0020
##      8      1.3228      nan      0.0100      0.0029
##      9      1.3157      nan      0.0100      0.0029
##     10      1.3111      nan      0.0100      0.0014
##     20      1.2527      nan      0.0100      0.0025
##     40      1.1560      nan      0.0100      0.0018
##     60      1.0800      nan      0.0100      0.0004
##     80      1.0213      nan      0.0100      0.0007
##    100      0.9693      nan      0.0100      0.0011
##    120      0.9267      nan      0.0100     -0.0002
##    140      0.8909      nan      0.0100      0.0003
##    150      0.8758      nan      0.0100      0.0001
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1.3684      nan      0.0100      0.0041
##      2      1.3596      nan      0.0100      0.0039
##      3      1.3524      nan      0.0100      0.0027
##      4      1.3442      nan      0.0100      0.0039
##      5      1.3376      nan      0.0100      0.0028
##      6      1.3299      nan      0.0100      0.0033
##      7      1.3227      nan      0.0100      0.0028
##      8      1.3149      nan      0.0100      0.0028
##      9      1.3075      nan      0.0100      0.0030
##     10      1.3017      nan      0.0100      0.0017
##     20      1.2397      nan      0.0100      0.0016
##     40      1.1378      nan      0.0100      0.0014
##     60      1.0538      nan      0.0100      0.0011
##     80      0.9888      nan      0.0100      0.0007
##    100      0.9338      nan      0.0100      0.0010
##    120      0.8894      nan      0.0100      0.0002
##    140      0.8487      nan      0.0100     -0.0002
##    150      0.8322      nan      0.0100     -0.0002

## Stochastic Gradient Boosting
##
## 239 samples
## 11 predictor

```

```

## 2 classes: 'no', 'yes'
##
## No pre-processing
## Resampling: Cross-Validated (5 fold)
## Summary of sample sizes: 192, 191, 190, 191, 192
## Resampling results across tuning parameters:
##
## interaction.depth n.minobsinnode Accuracy Kappa
## 1 5 0.7570560 0.5018753
## 1 10 0.7442086 0.4762721
## 1 15 0.7403857 0.4689432
## 3 5 0.7531517 0.4978995
## 3 10 0.7489850 0.4889287
## 3 15 0.7489886 0.4895727
## 5 5 0.7493324 0.4912688
## 5 10 0.7660027 0.5259506
## 5 15 0.7448184 0.4804674
##
## Tuning parameter 'n.trees' was held constant at a value of 150
## Tuning
## parameter 'shrinkage' was held constant at a value of 0.01
## Accuracy was used to select the optimal model using the largest value.
## The final values used for the model were n.trees = 150, interaction.depth =
## 5, shrinkage = 0.01 and n.minobsinnode = 10.
# Training Error =0.5027174
predict_train <- predict(boost_tuned, n.trees = 1000, type = "prob", newdata = heart_train)
predicted_train_classes <- ifelse(predict_train[, 2] > 0.5, 1, 0)
mean(predicted_train_classes != heart_test$num)

## Warning in predicted_train_classes != heart_test$num: longer object length is
## not a multiple of shorter object length

## [1] 0.5062762
# Testing Error = 0.1793478
predict_test <- predict(boost_tuned, n.trees = 1000, type = "prob", newdata = heart_test)
predicted_test_classes <- ifelse(predict_test[, 2] > 0.5, 1, 0)
mean(predicted_test_classes != heart_test$num)

## [1] 0.15
# Confusion Matrix
predictions_test <- factor(predicted_test_classes, levels = c(0, 1), labels = c("no", "yes"))
conf_matrix_test <- confusionMatrix(predictions_test, y_test, mode = "everything")
print(conf_matrix_test)

## Confusion Matrix and Statistics
##
## Reference
## Prediction no yes
## no 27 7
## yes 2 24
##
## Accuracy : 0.85
## 95% CI : (0.7343, 0.929)
## No Information Rate : 0.5167

```



```
##      P-Value [Acc > NIR] : 6.136e-08
##
##              Kappa : 0.7013
##
## McNemar's Test P-Value : 0.1824
##
##      Sensitivity : 0.9310
##      Specificity : 0.7742
##      Pos Pred Value : 0.7941
##      Neg Pred Value : 0.9231
##      Precision : 0.7941
##      Recall : 0.9310
##      F1 : 0.8571
##      Prevalence : 0.4833
##      Detection Rate : 0.4500
##      Detection Prevalence : 0.5667
##      Balanced Accuracy : 0.8526
##
##      'Positive' Class : no
##
```

```
# accuracy = 0.8207
```

```
# F1 = 0.7519
```

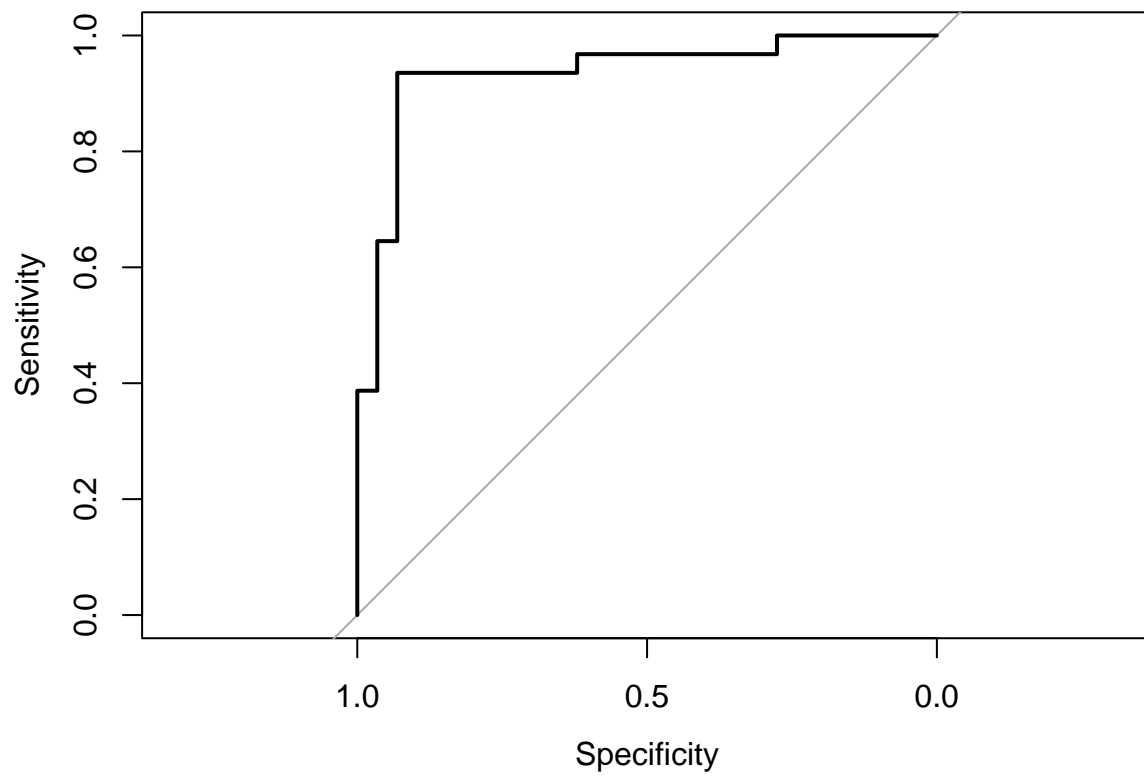
```
# AUCROC
```

```
predictions.prob <- predict(boost_tuned, X_test, type = "prob")
```

```
roc.tuned.gbm <- roc(response = y_test, predictor = predictions.prob[,2], plot = TRUE)
```

```
## Setting levels: control = no, case = yes
```

```
## Setting direction: controls < cases
```



```
print(auc(roc.tuned.gbm))
```

```
## Area under the curve: 0.9355
```

```
# auc = 0.904
```