

Project Data Mining

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2024-03-18

Load libraries

```
library(caret)
```

```
## Loading required package: ggplot2
```

```
## Warning: package 'ggplot2' was built under R version 4.3.1
```

```
## Loading required package: lattice
```

```
library(tidyverse) # for data manipulation and visualization
```

```
## Warning: package 'readr' was built under R version 4.3.1
```

```
## Warning: package 'dplyr' was built under R version 4.3.1
```

```
## Warning: package 'stringr' was built under R version 4.3.1
```

```
## Warning: package 'lubridate' was built under R version 4.3.1
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
```

```
## v dplyr      1.1.4      v readr      2.1.5
```

```
## v forcats   1.0.0      v stringr   1.5.1
```

```
## v lubridate 1.9.3      v tibble    3.2.1
```

```
## v purrr     1.0.2      v tidyr     1.3.0
```

```
## -- Conflicts ----- tidyverse_conflicts() --
```

```
## x dplyr::filter() masks stats::filter()
```

```
## x dplyr::lag()     masks stats::lag()
```

```
## x purrr::lift()    masks caret::lift()
```

```
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
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:dplyr':
##
##     combine
##
## The following object is masked from 'package:ggplot2':
##
##     margin
```

```
library(e1071) # for SVM
```

```
## Warning: package 'e1071' was built under R version 4.3.1
```

```
library(nnet) # for neural network
install.packages("kernlab")
```

```
##
## The downloaded binary packages are in
## /var/folders/zw/9p94jsvn2mb68h0wkh7y2zx80000gn/T//RtmpLok40y/downloaded_packages
```

```
library(kernlab)
```

```
##
## Attaching package: 'kernlab'
##
## The following object is masked from 'package:purrr':
##
##     cross
##
## The following object is masked from 'package:ggplot2':
##
##     alpha
```

Step 1. Import and parse breastcancer data set

```
library(readr)
data <- read_csv("/Users/danaabdirakhym/Downloads/breast_cancer.csv")

## Rows: 569 Columns: 32
## -- Column specification -----
## Delimiter: ","
## chr  (1): diagnosis
## dbl (31): id, radius_mean, texture_mean, perimeter_mean, area_mean, smoothne...
##
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
```

```
##Coverted Malignant to 1, Benign to 0
data$diagnosis <- as.factor(ifelse(data$diagnosis == "M", 1, 0))
breast_cancer <- na.omit(data)

breast_cancer <- subset(breast_cancer, select = -id)
colnames(breast_cancer)
```

```
## [1] "diagnosis"          "radius_mean"
## [3] "texture_mean"       "perimeter_mean"
## [5] "area_mean"          "smoothness_mean"
## [7] "compactness_mean"   "concavity_mean"
## [9] "concave points_mean" "symmetry_mean"
## [11] "fractal_dimension_mean" "radius_se"
## [13] "texture_se"         "perimeter_se"
## [15] "area_se"            "smoothness_se"
## [17] "compactness_se"     "concavity_se"
## [19] "concave points_se"  "symmetry_se"
## [21] "fractal_dimension_se" "radius_worst"
## [23] "texture_worst"      "perimeter_worst"
## [25] "area_worst"         "smoothness_worst"
## [27] "compactness_worst"  "concavity_worst"
## [29] "concave points_worst" "symmetry_worst"
## [31] "fractal_dimension_worst"
```

###Step 2. Create at least four different classifiers. Determine the needed features and create training and test data.

```
library(caret)
set.seed(123) # for reproducibility

### Splitting to train and test data 80 to 20
index <- createDataPartition(breast_cancer$diagnosis, p=0.8, list=FALSE)
train_data <- breast_cancer[index, ]
test_data <- breast_cancer[-index, ]

head(train_data)
```

```
## # A tibble: 6 x 31
##   diagnosis radius_mean texture_mean perimeter_mean area_mean smoothness_mean
##   <fct>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>
## 1 1          18.0        10.4        123.        1001        0.118
## 2 1          20.6        17.8        133.        1326        0.0847
## 3 1          19.7        21.2        130         1203        0.110
## 4 1          11.4        20.4         77.6        386.        0.142
## 5 1          20.3        14.3        135.        1297        0.100
## 6 1          12.4        15.7         82.6        477.        0.128
## # i 25 more variables: compactness_mean <dbl>, concavity_mean <dbl>,
## #   'concave points_mean' <dbl>, symmetry_mean <dbl>,
## #   fractal_dimension_mean <dbl>, radius_se <dbl>, texture_se <dbl>,
## #   perimeter_se <dbl>, area_se <dbl>, smoothness_se <dbl>,
## #   compactness_se <dbl>, concavity_se <dbl>, 'concave points_se' <dbl>,
```

```
## # symmetry_se <dbl>, fractal_dimension_se <dbl>, radius_worst <dbl>,
## # texture_worst <dbl>, perimeter_worst <dbl>, area_worst <dbl>, ...
```

```
head(test_data)
```

```
## # A tibble: 6 x 31
##   diagnosis radius_mean texture_mean perimeter_mean area_mean smoothness_mean
##   <fct>         <dbl>         <dbl>         <dbl>         <dbl>         <dbl>
## 1 1           19.2           24.8           132.           1123           0.0974
## 2 1           15.8           24.0           104.            783.           0.0840
## 3 1           13.7           22.6           93.6            578.           0.113
## 4 1           14.7           20.1           94.7            684.           0.0987
## 5 0            9.50           12.4           60.3            274.           0.102
## 6 1           17.1           16.4           116             913.           0.119
## # i 25 more variables: compactness_mean <dbl>, concavity_mean <dbl>,
## #   'concave points_mean' <dbl>, symmetry_mean <dbl>,
## #   fractal_dimension_mean <dbl>, radius_se <dbl>, texture_se <dbl>,
## #   perimeter_se <dbl>, area_se <dbl>, smoothness_se <dbl>,
## #   compactness_se <dbl>, concavity_se <dbl>, 'concave points_se' <dbl>,
## #   symmetry_se <dbl>, fractal_dimension_se <dbl>, radius_worst <dbl>,
## #   texture_worst <dbl>, perimeter_worst <dbl>, area_worst <dbl>, ...
```

```
# Check the levels before conversion
levels(breast_cancer$diagnosis)
```

```
## [1] "0" "1"
```

```
# Convert factor levels "0" and "1" to actual numeric values 0 and 1
# This conversion ensures that 'diagnosis' is now a numeric variable instead of a factor.
breast_cancer$diagnosis <- as.numeric(as.character(breast_cancer$diagnosis))
train_data$diagnosis <- as.numeric(as.character(train_data$diagnosis))
test_data$diagnosis <- as.numeric(as.character(test_data$diagnosis))
```

```
###Creating the model using recursive partitioning
```

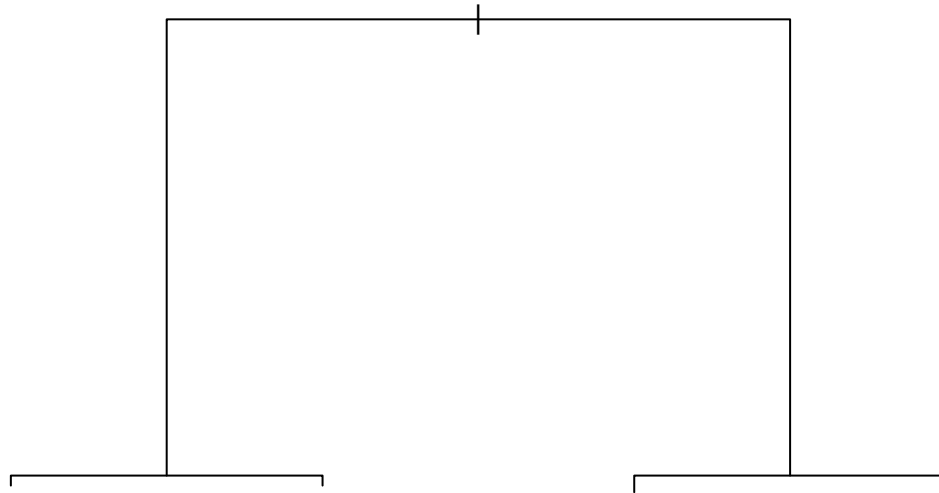
```
if (!requireNamespace("rpart", quietly = TRUE)) {
  install.packages("rpart")
}
library(rpart)
```

```
# Fit a decision tree model 1
```

```
modell1_rpart <- rpart(diagnosis ~ ., data = train_data, method = "class")
modell1_pred <- predict(modell1_rpart, type="class", newdata= test_data)
modell1_prob <-predict(modell1_rpart, type="prob", newdata= test_data)

plot(modell1_rpart, main= "decision tree model 1")
```

decision tree model 1



```
##create model using conditional inferences
if (!requireNamespace("party", quietly = TRUE)) {
  install.packages("party")
}
library(party)
```

```
## Warning: package 'party' was built under R version 4.3.1
```

```
## Loading required package: grid
```

```
## Loading required package: mvtnorm
```

```
## Warning: package 'mvtnorm' was built under R version 4.3.1
```

```
## Loading required package: modeltools
```

```
## Loading required package: stats4
```

```
##
```

```
## Attaching package: 'modeltools'
```

```
## The following object is masked from 'package:kernlab':
```

```
##
```

```
## prior
```

```

## Loading required package: strucchange

## Loading required package: zoo

##
## Attaching package: 'zoo'

## The following objects are masked from 'package:base':
##
##   as.Date, as.Date.numeric

## Loading required package: sandwich

##
## Attaching package: 'strucchange'

## The following object is masked from 'package:stringr':
##
##   boundary

##
## Attaching package: 'party'

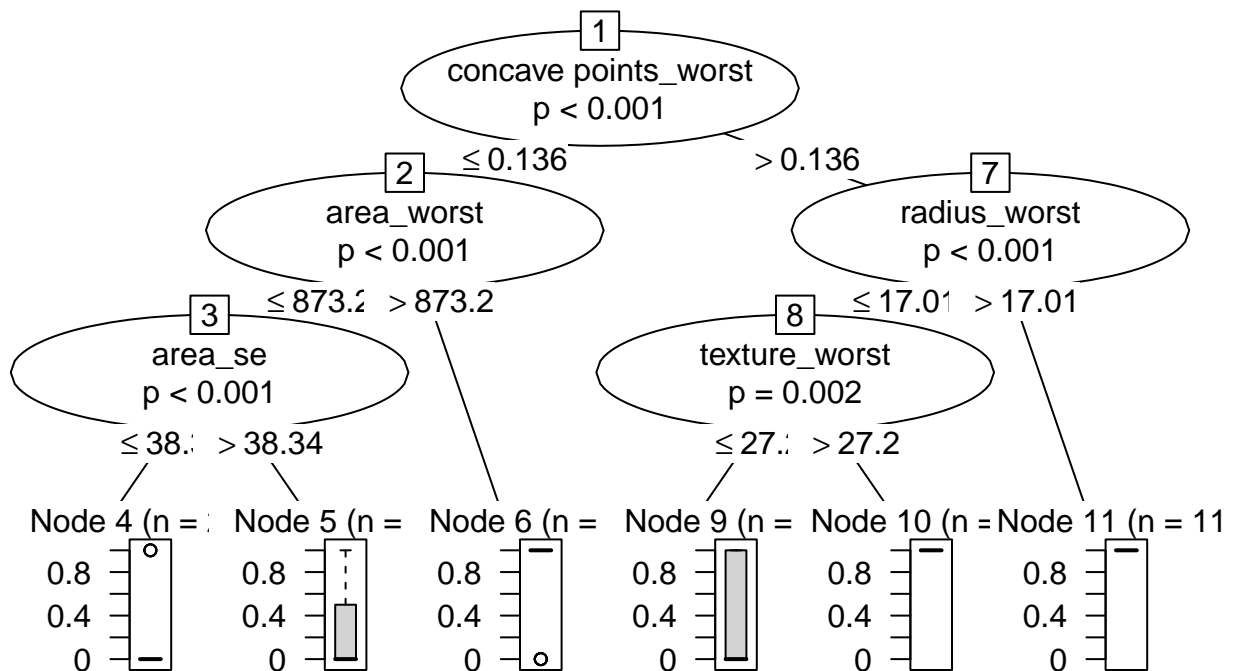
## The following object is masked from 'package:dplyr':
##
##   where

model2 <- ctree(diagnosis ~ ., data = train_data)
# Predict class labels
model2_pred <- predict(model2, newdata = test_data, type = "response")
# Predict probabilities
model2_prob <- predict(model2, newdata = test_data, type = "prob")

plot(model2, main= "decision tree model using conditional inference tree")

```

decision tree model using conditional inference tree



```
###create random forest using conditional inference tree

if (!requireNamespace("ipred", quietly = TRUE)) {
  install.packages("ipred")
}
library(ipred)
# Fit a Bagging model
model3_bagging <- bagging(diagnosis ~ ., data = train_data, nbagg = 25) # 'nbagg' denotes the number of
model3_bagging_pred <- predict(model3_bagging, newdata = test_data, type = "class")
model3_bagging_pred <- predict(model3_bagging, newdata = test_data, type = "prob")
```

##Step 3 Provide your performance measure: confusion matrix, accuracy, recall, etc.

```
if (!requireNamespace("caret", quietly = TRUE)) {
  install.packages("caret")
}
library(caret)

if (!requireNamespace("e1071", quietly = TRUE)) { # Required for SVM
  install.packages("e1071")
}
if (!requireNamespace("class", quietly = TRUE)) { # Might be required for KNN
  install.packages("class")
}
# Convert 'diagnosis' to a factor with two levels
```

```
train_data$diagnosis <- factor(train_data$diagnosis, levels = c('0', '1'))
test_data$diagnosis <- factor(test_data$diagnosis, levels = c('0', '1'))
```

```
# Support vector machines
```

```
model_svm <- train(diagnosis ~ ., data = train_data, method = "svmRadial")
model_svm_pred <- predict(model_svm, newdata = test_data)
model_svm_prob <- predict(model_svm, newdata = test_data, type = "prob")
```

```
## Warning in method$prob(modelFit = modelFit, newdata = newdata, submodels =
## param): kernlab class probability calculations failed; returning NAs
```

```
#Knn model
```

```
model_knn <- train(diagnosis ~ ., data = train_data, method = "knn")
model_knn_pred <- predict(model_knn, newdata = test_data)
model_knn_prob <- predict(model_knn, newdata = test_data, type = "prob")
```

```
#Logistic regression model
```

```
model_log <- train(diagnosis ~ ., data = train_data, method = "glm", family = "binomial")
```

```
## Warning: glm.fit: algorithm did not converge
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
## Warning: glm.fit: algorithm did not converge
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
## Warning: glm.fit: algorithm did not converge
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
## Warning: glm.fit: algorithm did not converge
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
## Warning: glm.fit: algorithm did not converge
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
## Warning: glm.fit: algorithm did not converge
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
## Warning: glm.fit: algorithm did not converge
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
## Warning: glm.fit: algorithm did not converge
```


Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

Warning: glm.fit: algorithm did not converge

Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

Warning: glm.fit: algorithm did not converge

Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

Warning: glm.fit: algorithm did not converge

Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

Warning: glm.fit: algorithm did not converge

Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

Warning: glm.fit: algorithm did not converge

Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

Warning: glm.fit: algorithm did not converge

Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

Warning: glm.fit: algorithm did not converge

Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

Warning: glm.fit: algorithm did not converge

Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

Warning: glm.fit: algorithm did not converge

Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

Warning: glm.fit: algorithm did not converge

Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

Warning: glm.fit: algorithm did not converge

Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

Warning: glm.fit: algorithm did not converge

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: algorithm did not converge
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: algorithm did not converge
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: algorithm did not converge
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: algorithm did not converge
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: algorithm did not converge
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: algorithm did not converge
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: algorithm did not converge
```

```
model_log_pred <- predict(model_log, newdata = test_data)
model_log_prob <- predict(model_log, newdata = test_data, type = "prob")
```

```
# Calculate performance metrics for the SVM model
```

```
conf_matrix_svm <- confusionMatrix(factor(model_svm_pred, levels = levels(test_data$diagnosis)), test_data$diagnosis)
precision_svm <- conf_matrix_svm$byClass['Pos Pred Value'] # This is the precision
recall_svm <- conf_matrix_svm$byClass['Sensitivity'] # This is the recall
accuracy_svm <- conf_matrix_svm$overall['Accuracy']
```

```
# Calculate performance metrics for the KNN model
```

```
conf_matrix_knn <- confusionMatrix(factor(model_knn_pred, levels = levels(test_data$diagnosis)), test_data$diagnosis)
precision_knn <- conf_matrix_knn$byClass['Pos Pred Value']
recall_knn <- conf_matrix_knn$byClass['Sensitivity']
accuracy_knn <- conf_matrix_knn$overall['Accuracy']
```

```
# Calculate performance metrics for the Logistic Regression model
```

```
conf_matrix_log <- confusionMatrix(factor(model_log_pred, levels = levels(test_data$diagnosis)), test_data$diagnosis)
precision_log <- conf_matrix_log$byClass['Pos Pred Value']
recall_log <- conf_matrix_log$byClass['Sensitivity']
accuracy_log <- conf_matrix_log$overall['Accuracy']
```

```
# Print out the metrics
```

```
cat("SVM Metrics: Accuracy =", accuracy_svm, ", Precision =", precision_svm, ", Recall =", recall_svm, "\n")
```

```
## SVM Metrics: Accuracy = 0.9823009 , Precision = 1 , Recall = 0.971831
```

```
cat("KNN Metrics: Accuracy =", accuracy_knn, ", Precision =", precision_knn, ", Recall =", recall_knn, "
```

```
## KNN Metrics: Accuracy = 0.9115044 , Precision = 0.942029 , Recall = 0.915493
```

```
cat("Logistic Regression Metrics: Accuracy =", accuracy_log, ", Precision =", precision_log, ", Recall =", recall_log, "
```

```
## Logistic Regression Metrics: Accuracy = 0.9380531 , Precision = 1 , Recall = 0.9014085
```

```
###Step 4 Combine the classifiers in an ensemble
```

```
classifier1_svm <- as.numeric(model_svm_pred) - 1 # Convert factors to numeric
classifier2_knn <- as.numeric(model_knn_pred) - 1
classifier3_log <- as.numeric(model_log_pred) - 1
classifier4_dtree <- as.numeric(model1_pred) - 1

# Assuming your factor levels are correctly ordered, like c('0', '1')
# If they are not, replace '0' and '1' with the actual factor levels from your predictions
classifier1_svm <- ifelse(model_svm_pred == '1', 1, 0) # Adjust based on your actual levels
classifier2_knn <- ifelse(model_knn_pred == '1', 1, 0)
classifier3_log <- ifelse(model_log_pred == '1', 1, 0)
classifier4_dtree <- ifelse(model1_pred == '1', 1, 0)

combine.df <- data.frame(classifier1_svm, classifier2_knn, classifier3_log, classifier4_dtree)
combine.df
```

```
##      classifier1_svm classifier2_knn classifier3_log classifier4_dtree
## 1                1                1                1                1
## 2                1                1                1                1
## 3                1                0                1                1
## 4                1                1                1                1
## 5                0                0                0                0
## 6                1                1                1                1
## 7                1                1                1                0
## 8                1                0                1                1
## 9                1                1                1                1
## 10               0                0                0                0
## 11               0                0                0                0
## 12               0                0                0                0
## 13               0                0                0                0
## 14               1                0                1                1
## 15               1                1                1                1
## 16               1                1                1                1
## 17               0                0                0                0
## 18               1                1                1                1
## 19               1                0                1                1
## 20               1                1                1                1
## 21               1                1                1                1
## 22               1                1                1                1
## 23               1                1                1                1
## 24               1                1                1                1
```

## 25	0	1	1	0
## 26	0	0	0	0
## 27	0	0	0	0
## 28	0	0	0	0
## 29	1	0	1	1
## 30	0	0	0	1
## 31	0	0	0	0
## 32	1	1	1	0
## 33	1	1	1	1
## 34	0	0	0	0
## 35	1	1	1	1
## 36	1	1	1	0
## 37	0	0	0	0
## 38	1	1	1	1
## 39	1	1	1	0
## 40	0	0	0	0
## 41	0	0	1	0
## 42	0	0	0	0
## 43	1	1	1	1
## 44	1	1	1	1
## 45	0	0	0	0
## 46	0	0	0	1
## 47	1	1	1	1
## 48	1	1	1	1
## 49	1	1	1	1
## 50	0	0	0	0
## 51	0	0	0	0
## 52	0	0	0	0
## 53	1	1	1	1
## 54	0	0	0	0
## 55	0	0	1	0
## 56	1	1	1	1
## 57	1	1	1	1
## 58	0	0	0	0
## 59	0	0	1	0
## 60	0	0	0	0
## 61	0	0	0	0
## 62	0	0	0	0
## 63	0	0	0	0
## 64	0	0	0	0
## 65	1	1	1	1
## 66	0	0	0	0
## 67	1	1	1	1
## 68	0	0	1	0
## 69	0	0	0	0
## 70	0	0	0	0
## 71	1	1	1	1
## 72	0	0	0	0
## 73	0	0	0	0
## 74	1	1	1	1
## 75	0	0	0	0
## 76	0	1	0	0
## 77	0	0	0	0
## 78	1	1	1	1

## 79	0	0	0	0
## 80	0	0	0	0
## 81	0	0	0	0
## 82	0	1	0	0
## 83	1	1	1	1
## 84	0	0	0	0
## 85	1	0	1	1
## 86	0	0	0	0
## 87	0	0	0	0
## 88	0	0	0	1
## 89	1	1	1	1
## 90	0	0	0	0
## 91	1	1	1	1
## 92	0	0	0	0
## 93	0	0	0	0
## 94	0	1	0	1
## 95	0	0	0	0
## 96	0	0	0	1
## 97	0	0	0	0
## 98	0	0	0	1
## 99	0	0	0	0
## 100	0	1	0	0
## 101	1	1	1	1
## 102	0	0	0	0
## 103	0	0	0	0
## 104	0	0	0	0
## 105	0	0	0	0
## 106	0	1	0	0
## 107	0	0	0	0
## 108	0	0	0	1
## 109	0	0	0	0
## 110	0	0	0	0
## 111	0	0	0	0
## 112	0	0	0	0
## 113	1	1	1	1

```

combine.df$vote <- rowSums(combine.df)
combine.df$class <- ifelse(combine.df$vote >= 2,1,0)
# Majority is 'malignant' if 2 or more classifiers predict 1
combine.df

```

##	classifier1_svm	classifier2_knn	classifier3_log	classifier4_dtree	vote
## 1	1	1	1	1	4
## 2	1	1	1	1	4
## 3	1	0	1	1	3
## 4	1	1	1	1	4
## 5	0	0	0	0	0
## 6	1	1	1	1	4
## 7	1	1	1	0	3
## 8	1	0	1	1	3
## 9	1	1	1	1	4
## 10	0	0	0	0	0
## 11	0	0	0	0	0
## 12	0	0	0	0	0

## 13	0	0	0	0	0
## 14	1	0	1	1	3
## 15	1	1	1	1	4
## 16	1	1	1	1	4
## 17	0	0	0	0	0
## 18	1	1	1	1	4
## 19	1	0	1	1	3
## 20	1	1	1	1	4
## 21	1	1	1	1	4
## 22	1	1	1	1	4
## 23	1	1	1	1	4
## 24	1	1	1	1	4
## 25	0	1	1	0	2
## 26	0	0	0	0	0
## 27	0	0	0	0	0
## 28	0	0	0	0	0
## 29	1	0	1	1	3
## 30	0	0	0	1	1
## 31	0	0	0	0	0
## 32	1	1	1	0	3
## 33	1	1	1	1	4
## 34	0	0	0	0	0
## 35	1	1	1	1	4
## 36	1	1	1	0	3
## 37	0	0	0	0	0
## 38	1	1	1	1	4
## 39	1	1	1	0	3
## 40	0	0	0	0	0
## 41	0	0	1	0	1
## 42	0	0	0	0	0
## 43	1	1	1	1	4
## 44	1	1	1	1	4
## 45	0	0	0	0	0
## 46	0	0	0	1	1
## 47	1	1	1	1	4
## 48	1	1	1	1	4
## 49	1	1	1	1	4
## 50	0	0	0	0	0
## 51	0	0	0	0	0
## 52	0	0	0	0	0
## 53	1	1	1	1	4
## 54	0	0	0	0	0
## 55	0	0	1	0	1
## 56	1	1	1	1	4
## 57	1	1	1	1	4
## 58	0	0	0	0	0
## 59	0	0	1	0	1
## 60	0	0	0	0	0
## 61	0	0	0	0	0
## 62	0	0	0	0	0
## 63	0	0	0	0	0
## 64	0	0	0	0	0
## 65	1	1	1	1	4
## 66	0	0	0	0	0

## 67	1	1	1	1	4
## 68	0	0	1	0	1
## 69	0	0	0	0	0
## 70	0	0	0	0	0
## 71	1	1	1	1	4
## 72	0	0	0	0	0
## 73	0	0	0	0	0
## 74	1	1	1	1	4
## 75	0	0	0	0	0
## 76	0	1	0	0	1
## 77	0	0	0	0	0
## 78	1	1	1	1	4
## 79	0	0	0	0	0
## 80	0	0	0	0	0
## 81	0	0	0	0	0
## 82	0	1	0	0	1
## 83	1	1	1	1	4
## 84	0	0	0	0	0
## 85	1	0	1	1	3
## 86	0	0	0	0	0
## 87	0	0	0	0	0
## 88	0	0	0	1	1
## 89	1	1	1	1	4
## 90	0	0	0	0	0
## 91	1	1	1	1	4
## 92	0	0	0	0	0
## 93	0	0	0	0	0
## 94	0	1	0	1	2
## 95	0	0	0	0	0
## 96	0	0	0	1	1
## 97	0	0	0	0	0
## 98	0	0	0	1	1
## 99	0	0	0	0	0
## 100	0	1	0	0	1
## 101	1	1	1	1	4
## 102	0	0	0	0	0
## 103	0	0	0	0	0
## 104	0	0	0	0	0
## 105	0	0	0	0	0
## 106	0	1	0	0	1
## 107	0	0	0	0	0
## 108	0	0	0	1	1
## 109	0	0	0	0	0
## 110	0	0	0	0	0
## 111	0	0	0	0	0
## 112	0	0	0	0	0
## 113	1	1	1	1	4
##	class				
## 1	1				
## 2	1				
## 3	1				
## 4	1				
## 5	0				
## 6	1				

## 7	1
## 8	1
## 9	1
## 10	0
## 11	0
## 12	0
## 13	0
## 14	1
## 15	1
## 16	1
## 17	0
## 18	1
## 19	1
## 20	1
## 21	1
## 22	1
## 23	1
## 24	1
## 25	1
## 26	0
## 27	0
## 28	0
## 29	1
## 30	0
## 31	0
## 32	1
## 33	1
## 34	0
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## 36	1
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## 42	0
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## 45	0
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## 52	0
## 53	1
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## 55	0
## 56	1
## 57	1
## 58	0
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## 60	0

## 61	0
## 62	0
## 63	0
## 64	0
## 65	1
## 66	0
## 67	1
## 68	0
## 69	0
## 70	0
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## 87	0
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## 89	1
## 90	0
## 91	1
## 92	0
## 93	0
## 94	1
## 95	0
## 96	0
## 97	0
## 98	0
## 99	0
## 100	0
## 101	1
## 102	0
## 103	0
## 104	0
## 105	0
## 106	0
## 107	0
## 108	0
## 109	0
## 110	0
## 111	0
## 112	0
## 113	1