## Naive Bayes Classifier

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## Read in data

```
library(caret)
mydata <- read.table(params$file, sep=",")</pre>
head(mydata)
     V1 V2 V3 V4 V5 V6
                               V7 V8 V9
##
## 1 6 148 72 35 0 33.6 0.627 50 1
## 2 1 85 66 29 0 26.6 0.351 31 0
## 3 8 183 64 0 0 23.3 0.672 32 1
## 4 1 89 66 23 94 28.1 0.167 21 0
## 5 0 137 40 35 168 43.1 2.288 33 1
## 6 5 116 74 0 0 25.6 0.201 30 0
X <- mydata[, 1:8]</pre>
y <- mydata[, 9]</pre>
partition <- createDataPartition(y, p=0.8, list=FALSE)</pre>
trainX <- X[partition,]</pre>
trainy <- y[partition]</pre>
testX <- X[-partition,]</pre>
testy <- y[-partition]</pre>
```

Function to train a Naiva Bayes classifier, assume each attribute has a gaussian distribution.

Function to make predictions on test data.

```
naiveBayesPred <- function(NBmodel, testX) {
  testX.centered.pos <- t(t(testX) - NBmodel$meanPos)
  testX.scaled.pos <- t(t(testX.centered.pos) / NBmodel$sdPos)
  logLik.pos <- -(1/2)*rowSums(testX.scaled.pos^2) - sum(log(NBmodel$sdPos)) + log(NBmodel$pPos)</pre>
```

```
testX.centered.neg <- t(t(testX) - NBmodel$meanNeg)
testX.scaled.neg <- t(t(testX.centered.neg) / NBmodel$sdNeg)
logLik.neg <- -(1/2)*rowSums(testX.scaled.neg^2) - sum(log(NBmodel$sdNeg)) + log(1-NBmodel$pPos)
pred <- as.numeric(logLik.pos > logLik.neg)
return(pred)
}
```

## Evaluate model on test data.

```
pred <- naiveBayesPred(NBmodel, testX)
error <- sum(pred != testy) / length(testy)
print(paste("Test error rate is", error))</pre>
```

## [1] "Test error rate is 0.209150326797386"