

1a. 0.7538961

1b. 0.7053127

1d. 0.7650897

```

```{r}
#install.packages("dplyr")
library(dplyr)
sample1_train = sample_n(data, as.integer(nrow(data)*0.8))
sample1_test = anti_join(data, sample1_train)

P_c1 = length(sample1_train$X1[sample1_train$X1 == 1])/nrow(sample1_train)
P_c1
P_c0 = 1 - P_c1
P_c0
col_1_mean = sapply(sample1_train[sample1_train$X1 == 1,], mean)
col_0_mean = sapply(sample1_train[sample1_train$X1 == 0,], mean)
col_1_sd = sapply(sample1_train[sample1_train$X1 == 1,], sd)
col_0_sd = sapply(sample1_train[sample1_train$X1 == 0,], sd)
sample1_test$X6_P1 = dnorm(sample1_test$X6, col_1_mean[1], col_1_sd[1], log = TRUE)
sample1_test$X6_P0 = dnorm(sample1_test$X6, col_0_mean[1], col_0_sd[1], log = TRUE)
sample1_test$X148_P1 = dnorm(sample1_test$X148, col_1_mean[2], col_1_sd[2], log = TRUE)
sample1_test$X148_P0 = dnorm(sample1_test$X148, col_0_mean[2], col_0_sd[2], log = TRUE)
sample1_test$X72_P1 = dnorm(sample1_test$X72, col_1_mean[3], col_1_sd[3], log = TRUE)
sample1_test$X72_P0 = dnorm(sample1_test$X72, col_0_mean[3], col_0_sd[3], log = TRUE)
sample1_test$X35_P1 = dnorm(sample1_test$X35, col_1_mean[4], col_1_sd[4], log = TRUE)
sample1_test$X35_P0 = dnorm(sample1_test$X35, col_0_mean[4], col_0_sd[4], log = TRUE)
sample1_test$X0_P1 = dnorm(sample1_test$X0, col_1_mean[5], col_1_sd[5], log = TRUE)
sample1_test$X0_P0 = dnorm(sample1_test$X0, col_0_mean[5], col_0_sd[5], log = TRUE)
sample1_test$X336_P1 = dnorm(sample1_test$X33.6, col_1_mean[6], col_1_sd[6], log = TRUE)
sample1_test$X336_P0 = dnorm(sample1_test$X33.6, col_0_mean[6], col_0_sd[6], log = TRUE)
sample1_test$X627_P1 = dnorm(sample1_test$X0.627, col_1_mean[7], col_1_sd[7], log = TRUE)
sample1_test$X627_P0 = dnorm(sample1_test$X0.627, col_0_mean[7], col_0_sd[7], log = TRUE)
sample1_test$X50_P1 = dnorm(sample1_test$X50, col_1_mean[8], col_1_sd[8], log = TRUE)
sample1_test$X50_P0 = dnorm(sample1_test$X50, col_0_mean[8], col_0_sd[8], log = TRUE)
sample1_test$sum_p1 = sample1_test$X6_P1+sample1_test$X148_P1+sample1_test$X72_P1+sample1_test$X35_P1+sample1_test$X0_P1+sample1_test$X336_P1+sample1_test$X627_P1+sample1_test$X50_P1 + log(P_c1)
sample1_test$sum_p0 = sample1_test$X6_P0+sample1_test$X148_P0+sample1_test$X72_P0+sample1_test$X35_P0+sample1_test$X0_P0+sample1_test$X336_P0+sample1_test$X627_P0+sample1_test$X50_P0 + log(P_c0)
sample1_test$sum_p1
sample1_test$sum_p0
sample1_test$result = (sample1_test$sum_p1 > sample1_test$sum_p0)
sample1_test$result
accuracy = nrow(sample1_test[sample1_test$result == sample1_test$X1])/nrow(sample1_test)


sample2_train = sample_n(data, as.integer(nrow(data)*0.8))
sample2_test = anti_join(data, sample1_train)
P_c1 = length(sample1_train$X1[sample1_train$X1 == 1])/nrow(sample1_train)
P_c0 = 1 - P_c1

col_1_mean = sapply(na.omit(sample2_train[sample2_train$X1 == 1,]), mean)
col_0_mean = sapply(na.omit(sample2_train[sample2_train$X1 == 0,]), mean)
col_1_sd = sapply(na.omit(sample2_train[sample2_train$X1 == 1,]), sd)
col_0_sd = sapply(na.omit(sample2_train[sample2_train$X1 == 0,]), sd)

install.packages('klaR')
library('klaR')
library(e1071)
data = read.csv("pima-indians-diabetes.csv")
sample3_train = sample_n(data, as.integer(nrow(data)*0.8))
sample3_test = anti_join(data, sample3_train)
#model = svmLight(x = sample3_train[,1:8], grouping = sample3_train$X1)
1 - 144/nrow(sample3_train)
```

```

| x  | Method                          | Probability |
|----|---------------------------------|-------------|
| 1  | Gaussian + untouched            | 0.52354     |
| 2  | Gaussian + stretched            | 0.83115     |
| 3  | Bernoulli + untouched           | 0.83375     |
| 4  | Bernoulli + stretched           | 0.79005     |
| 5  | 10 trees + 4 depth + untouched  | 0.10315     |
| 6  | 10 trees + 4 depth + stretched  | 0.75810     |
| 7  | 10 trees + 16 depth + untouched | 0.10035     |
| 8  | 10 trees + 16 depth + stretched | 0.96660     |
| 9  | 30 trees + 4 depth + untouched  | 0.10205     |
| 10 | 30 trees + 4 depth + stretched  | 0.79535     |
| 11 | 30 trees + 16 depth + untouched | 0.10115     |
| 12 | 30 trees + 16 depth + stretched | 0.97385     |

|   |         |   |
|---|---------|---|
| <b>xinruiy2_12.csv</b><br>2 minutes ago by <a href="#">RAYING</a><br><a href="#">add submission details</a> | 0.97385 |    |
| <b>xinruiy2_11.csv</b><br>3 minutes ago by <a href="#">RAYING</a><br><a href="#">add submission details</a> | 0.10115 |    |
| <b>xinruiy2_10.csv</b><br>4 minutes ago by <a href="#">RAYING</a><br><a href="#">add submission details</a> | 0.79535 |    |
| <b>xinruiy2_9.csv</b><br>4 minutes ago by <a href="#">RAYING</a><br><a href="#">add submission details</a>  | 0.10205 |  |
| <b>xinruiy2_8.csv</b><br>6 minutes ago by <a href="#">RAYING</a><br><a href="#">add submission details</a>  | 0.96660 |  |
| <b>xinruiy2_7.csv</b><br>7 minutes ago by <a href="#">RAYING</a><br><a href="#">add submission details</a>  | 0.10035 |  |
| <b>xinruiy2_6.csv</b><br>9 minutes ago by <a href="#">RAYING</a><br><a href="#">add submission details</a>  | 0.75810 |  |
| <b>xinruiy2_5.csv</b><br>10 minutes ago by <a href="#">RAYING</a><br><a href="#">add submission details</a> | 0.10315 |  |
| <b>xinruiy2_4.csv</b><br>12 minutes ago by <a href="#">RAYING</a><br><a href="#">add submission details</a> | 0.79005 |  |
| <b>xinruiy2_3.csv</b><br>15 minutes ago by <a href="#">RAYING</a><br><a href="#">add submission details</a> | 0.83375 |  |
| <b>xinruiy2_2.csv</b><br>17 minutes ago by <a href="#">RAYING</a><br><a href="#">add submission details</a> | 0.83115 |  |
| <b>xinruiy2_1.csv</b><br>20 minutes ago by <a href="#">RAYING</a><br><a href="#">add submission details</a> | 0.52354 |  |

Gaussian + untouched : **0 1 2 3 4 5 6 7 8 9**

Bernoulli + Untouched: **0 1 2 3 4 5 6 7 8 9**

Gaussian + Stretched: **0/23456789**

Bernoulli + Stretched: **0/23456789**

```

#install.packages("e1071")
library(e1071)
library(caret)
library(naivebayes)
model = naive_bayes(train[,3:786], train[,2])
pred = predict(model, newdata = val[,2:785])
sum(pred == val$label)/2000
pred_t = predict(model, newdata = test[,1:784])

library(quantda)
bernoulli_model = textmodel_nb(as.dfm(train[,3:786]), as.factor(train[,2]), distribution = "Bernoulli")
bernoulli_pred = predict(bernoulli_model, newdata = as.dfm(val[,2:785]))
correct = as.vector(as.integer(val$label == bernoulli_pred))
sum(correct)/2000
bernoulli_pred_t = predict(bernoulli_model, newdata = as.dfm(test[,1:784]))

library(OpenImageR)
resize = function(image){

model = naive_bayes(new_train[,2:401], train[,2])
pred_s = predict(model, newdata = new_val[,2:401])
sum(pred_s == val$label)/2000

pred_t1 = predict(model, newdata = new_test)
as.data.frame(new_test)
new_test$V401 = as.integer(pred_t1) - 1

bernoulli_model = textmodel_nb(as.dfm(new_train[,2:401]), as.factor(train[,2]), distribution = "Bernoulli")
bernoulli_pred = predict(bernoulli_model, newdata = as.dfm(new_val[,2:401]))
correct = as.vector(as.integer(val$label == bernoulli_pred))
sum(correct)/2000
bernoulli_pred_t1 = predict(bernoulli_model, newdata = new_test[,2:401])

import numpy as np
import pandas as pd
import math
from sklearn import metrics
from sklearn.naive_bayes import GaussianNB
from sklearn.naive_bayes import BernoulliNB
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
from skimage import io
from skimage.transform import resize

def get_rf_acc(train_x, train_y, test_x, depth, trees):
    rf = RandomForestClassifier(max_depth=depth, n_estimators=trees)
    rf.fit(train_x, train_y)
    # test
    test_result = rf.predict(test_x)
    #cm = confusion_matrix(test_y, test_result)
    #acc = accuracy_score(test_y, test_result)
    return(test_result)

```