CS 498 AML Homework 1 Report

Part 1 (all performed with 80% and 20% train test split, iteration of 10):

1.A accuracy (Naive Bayes, treat 0 as it is):

The average accuracy over 10 times: 0.7418301

2.B accuracy (Naive Bayes, treat 0 in certain columns as NA):

The average accuracy over 10 times: 0.7098039

3.D accuracy (SVM, treat 0 as it is):

The average accuracy over 10 times: 0.7300654

Part 1 Code:

1.A (1.B has similar code):

```
trainIndex <- createDataPartition(dataset$V9, p = 0.8, list = FALSE)</pre>
train.label1 <- as.matrix(train[which(train$V9 == 1),])[, 1:num.features]</pre>
train.label0.prior <- nrow(train.label0) / numTrain</pre>
train.label0.expect <- colMeans(train.label0)</pre>
train.label1.expect <- colMeans(train.label1)</pre>
train.label0.std <- colSds(train.label0)</pre>
train.label1.std <- colSds(train.label1)</pre>
test$predict <- NA
  x <- as.numeric(test[idx, 1:num.features])</pre>
  pred.zero <-sum(dnorm(x, mean = train.label0.expect, sd = train.label0.std, log = TRUE)) + log(train.label0.prior)</pre>
  pred.one <- sum(dnorm(x, mean = train.label1.expect, sd = train.label1.std, log = TRUE)) + log(train.label1.prior)</pre>
  if(pred.one >= pred.zero){
    test[idx, ]$predict <- 1</pre>
    test[idx, ]$predict <- 0</pre>
acc list[each] <- sum(test$V9 == test$predict) / nrow(test)</pre>
```

1.D (Train/test split is similar to 1.A and 1.B, omitted for the space):

```
svm model <- svmlight(train.feature, train.label, pathsvm = path)</pre>
prediction <- predict(svm_model, test.feature)</pre>
acc list[each] <- sum(test.label == prediction$class) / num test
```

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Part 2 (Table & Submission Screenshot):

#	Method	Testing Accuracy (%)
1	Gaussian + untouched	55.75
2	Gaussian + stretched	81.70
3	Bernoulli + untouched	83.385
4	Bernoulli + stretched	75.57
5	10 trees + 4 depth + untouched	75.12
6	10 trees + 4 depth + stretched	75.675
7	10 trees + 16 depth + untouched	96.12
8	10 trees + 16 depth + stretched	96.595
9	30 trees + 4 depth + untouched	76.995
10	30 trees + 4 depth + stretched	78.255
11	30 trees + 16 depth + untouched	97.175
12	30 trees + 16 depth + stretched	97.425

cwu72_12.csv 2 minutes ago by Caiting Wu add submission details	0.97425	∀
cwu72_11.csv 3 minutes ago by Caiting Wu add submission details	0.97175	⋖
cwu72_10.csv 4 minutes ago by Caiting Wu add submission details	0.78255	⋖
cwu72_9.csv 4 minutes ago by Caiting Wu add submission details	0.76995	♂
cwu72_8.csv 6 minutes ago by Caiting Wu add submission details	0.96595	♂
cwu72_7.csv 6 minutes ago by Calting Wu add submission details	0.96120	
cwu72_6.csv 7 minutes ago by Caiting Wu add submission details	0.75675	
cwu72_5.csv 9 minutes ago by Caiting Wu add submission details	0.75120	∀
cwu72_4.csv 25 minutes ago by Caiting Wu add submission details	0.75570	∀
cwu72_3.csv 27 minutes ago by Caiting Wu add submission details	0.83385	∀
cwu72_2.csv 28 minutes ago by Caiting Wu add submission details	0.81700	∀
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Part 2 (40 Pictures):

Gauss & untouch Gauss + stretch Bern + untouch Bern + stretch

0	12 13 14 15 15 15 15 15 16 17 18 18 18 18 18 18 18 18 18 18 18 18 18	0	O
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2	13	2	12 23 24 24 24 25 26 26 27 28 28 28 28 28 28 28 28 28 28 28 28 28
3	10 10 10 10 10 10 10 10 10 10 10 10 10 1	3	14 15 15 15 15 15 15 15 15 15 15 15 15 15
ų	12 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2	4	
5	12 13 13 13 13 13 13 13	5	12 14 16 10 10 10 10 10 10 10 10 10 10 10 10 10
6	12 13 14 15 16 17 16 18 18 18 18 18 18 18 18 18 18 18 18 18	6	10 10 10 10 10 10 10 10 10 10 10 10 10 1
7	12 23 24 25 25 25 25 25 25 25 25 25 25 25 25 25	7	
8	12 12 12 12 12 12 12 12 12 12 12 12 12 1	8	
9	10 10 10 10 10 10 10 10 10 10 10 10 10 1	9	9

Part 2 (Codes):

```
# The script for CS 298 AML MP1 Part 2
import pandas as pd
import numpy as np
from skimage.transform import resize
from sklearn import preprocessing
import matplotlib.pyplot as plt
from sklearn.naive_bayes import GaussianNB
from sklearn.naive_bayes import BernoulliNB
from sklearn.ensemble import RandomForestClassifier
```

```
# Begin Naive Bayes Training
g_nb_1 = GaussianNB()
g_nb_2 = GaussianNB()
b_nb_1 = BernoulliNB()
b_nb_2 = BernoulliNB()

# For untouched
g_nb_1.fit(train_untouched, train_labels)
b_nb_1.fit(train_untouched, train_labels)
# For streched
g_nb_2.fit(train_scaled, train_labels)
b_nb_2.fit(train_scaled, train_labels)
# Validating
print('Gaussian + untouched validation acc:' , g_nb_1.score(val_untouched, val_labels))
print('Gaussian + stretched validation acc:' , b_nb_1.score(val_untouched, val_labels))
print('Bernoulli + untouched validation acc:' , b_nb_1.score(val_untouched, val_labels))
print('Bernoulli + stretched validation acc:' , b_nb_1.score(val_scaled, val_labels))
```

```
rfc 1 = RandomForestClassifier(n estimators=10, max depth=4)
rfc 2 = RandomForestClassifier(n_estimators=10, max_depth=4)
rfc 3 = RandomForestClassifier(n estimators=10, max depth=16)
rfc 4 = RandomForestClassifier(n estimators=10, max depth=16)
rfc 5 = RandomForestClassifier(n estimators=30, max depth=4)
rfc 6 = RandomForestClassifier(n estimators=30, max depth=4)
rfc_7 = RandomForestClassifier(n_estimators=30, max_depth=16)
rfc 8 = RandomForestClassifier(n estimators=30, max depth=16)
rfc 1.fit(total_untouched, total labels)
rfc_2.fit(total_scaled, total_labels)
rfc_3.fit(total_untouched, total_labels)
rfc_4.fit(total_scaled, total_labels)
rfc_5.fit(total_untouched, total_labels)
rfc_6.fit(total_scaled, total_labels)
rfc_7.fit(total_untouched, total_labels)
rfc 8.fit(total scaled, total labels)
# @NOTE: Same as before, still lazy
prediction = rfc_1.predict(test_untouched)
pred df = pd.DataFrame(data={'Label': prediction})
pred_df.to_csv('cwu72_5.csv', index=True)
pred_df = pd.read_csv('cwu72_5.csv')
pred df.columns = ['ImageId', 'Label']
pred_df.to_csv('cwu72_5.csv', index=False)
```