

# **CS498 AML HW1**

Huiyun Wu  
hwu63

## **Part1 A**

0.736601307189543

## **Part1 B**

0.74640522875817

## **Part1 D**

0.777777777777778

```
C:\Users\Sharo\Desktop\aml_hwl1 - RStudio
File Edit Code View Plots Session Build Debug Profile Tools Help
Go to File/Function Addins

hwl1_1A.R hwl1_1B.R Untitled1*
Source on Save Run Source

9 trscore<-array(dim=10)
10 tescore<-array(dim=10)
11
12 for (wi in 1:10){
13   #test-train split
14   wtd<-createDataPartition(y=labels, p=.8, list=FALSE)
15   nbx<-features
16   ntrbx<-nbx[wtd, ]
17   ntrby<-labels[wtd]
18   trposflag<-ntrby==0
19   ptrs<-ntrbx[trposflag, ]
20   ntrgs<-ntrbx[!trposflag, ]
21
22   ntebx<-nbx[-wtd, ]
23   nteby<-labels[-wtd]
24
25   ptrmean<-sapply(ptrs, mean, na.rm=TRUE)
26   ntrmean<-sapply(ntrgs, mean, na.rm=TRUE)
27   ptrsd<-sapply(ptrs, sd, na.rm=TRUE)
28   ntrsd<-sapply(ntrgs, sd, na.rm=TRUE)
29
30   ptroffsets<-t(t(ntrbx)-ptrmean)
31   ptrscales<-t(t(ptroffsets)/ptrsd)
32   ptrlogs<--(1/2)*rowSums(apply(ptrscales,c(1, 2), function(x)x^2), na.rm=TRUE)-sum(log(ptrsd))
33
34   ntroffsets<-t(t(ntrbx)-ntrmean)
35   ntrscales<-t(t(ntroffsets)/ntrsd)
36   ntrlogs<--(1/2)*rowSums(apply(ntrscales,c(1, 2), function(x)x^2), na.rm=TRUE)-sum(log(ntrsd))
37
38   #probability calculations
39   ptr_rate<-sum(ntrby)/length(ntrby)
40   ntr_rate<-sum(!ntrby)/length(ntrby)
41
42   ptr_prior<-ptrlogs + log(ptr_rate)
43   ntr_prior<-ntrlogs + log(ntr_rate)
44
45   lvwtr<-ptr_prior>ntr_prior
46
47   gotrighttr<-lvwtr==ntrby
48   trscore[wi]<-sum(gotrighttr)/(sum(gotrighttr)+sum(!gotrighttr))
49
50   pteoffsets<-t(t(nteby)-ptrmean)
51   ptescales<-t(t(pteoffsets)/ptrsd)
52   ptelogs<--(1/2)*rowSums(apply(ptescales,c(1, 2), function(x)x^2), na.rm=TRUE)-sum(log(ptrsd))
53
54   nteoffsets<-t(t(nteby)-ntrmean)
55   ntescales<-t(t(nteoffsets)/ntrsd)
56   ntelogs<--(1/2)*rowSums(apply(ntescales,c(1, 2), function(x)x^2), na.rm=TRUE)-sum(log(ntrsd))
57
58   #evaluations
59   lvwte<-ptelogs>ntelogs
60   gotright<-lvwte==nteby
61   tescore[wi]<-sum(gotright)/(sum(gotright)+sum(!gotright))
62 }
63 print(c("Training score: ", mean(trscore)))
64 print(c("Testing score: ", mean(tescore)))
65
31:36 (Top Level) R Script
Console
```

Accuracy	depth = 4	depth = 8	depth = 16
#trees = 10	0.8481	0.933	0.933
#trees = 30	0.8744	0.9394	0.9649

Untouched (I was not able to write these test result to files so I just showed the confusionMatrix results...)

<p>Accuracy : 0.9624  95% CI : (0.9585, 0.966)  No Information Rate : 0.1135  P-Value [Acc &gt; NIR] : &lt; 2.2e-16</p> <p>Kappa : 0.9582  McNemar's Test P-Value : NA</p> <p>Accuracy : 0.9367  95% CI : (0.9317, 0.9414)  No Information Rate : 0.1135  P-Value [Acc &gt; NIR] : &lt; 2.2e-16</p> <p>Kappa : 0.9296  McNemar's Test P-Value : NA</p> <p>Accuracy : 0.9394  95% CI : (0.9345, 0.944)  No Information Rate : 0.1135  P-Value [Acc &gt; NIR] : &lt; 2.2e-16</p> <p>Kappa : 0.9326  McNemar's Test P-Value : NA</p>	<p>Accuracy : 0.8653  95% CI : (0.8585, 0.8719)  No Information Rate : 0.1135  P-Value [Acc &gt; NIR] : &lt; 2.2e-16</p> <p>Kappa : 0.8502  McNemar's Test P-Value : &lt; 2.2e-16</p> <p>Accuracy : 0.8744  95% CI : (0.8677, 0.8808)  No Information Rate : 0.1135  P-Value [Acc &gt; NIR] : &lt; 2.2e-16</p> <p>Kappa : 0.8603  McNemar's Test P-Value : NA</p> <p>Accuracy : 0.9649  95% CI : (0.9611, 0.9684)  No Information Rate : 0.1135  P-Value [Acc &gt; NIR] : &lt; 2.2e-16</p> <p>Kappa : 0.961  McNemar's Test P-Value : NA</p>
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Accuracy	depth = 4	depth = 8	depth = 16
#trees = 10	0.8011	0.9059	0.9385
#trees = 30	0.9385	0.9151	0.9484

Bounded

<p>Accuracy : 0.8011  95% CI : (0.7931, 0.8089)  No Information Rate : 0.1135  P-Value [Acc &gt; NIR] : &lt; 2.2e-16</p> <p>Kappa : 0.7787  McNemar's Test P-Value : NA</p> <p>Accuracy : 0.9059  95% CI : (0.9, 0.9116)  No Information Rate : 0.1135  P-Value [Acc &gt; NIR] : &lt; 2.2e-16</p> <p>Kappa : 0.8954  McNemar's Test P-Value : NA</p> <p>Accuracy : 0.9151  95% CI : (0.9095, 0.9205)  No Information Rate : 0.1135  P-Value [Acc &gt; NIR] : &lt; 2.2e-16</p> <p>Kappa : 0.9056  McNemar's Test P-Value : NA</p>	<p>Accuracy : 0.9385  95% CI : (0.9336, 0.9431)  No Information Rate : 0.1135  P-Value [Acc &gt; NIR] : &lt; 2.2e-16</p> <p>Kappa : 0.9316  McNemar's Test P-Value : NA</p> <p>Accuracy : 0.9385  95% CI : (0.9336, 0.9431)  No Information Rate : 0.1135  P-Value [Acc &gt; NIR] : &lt; 2.2e-16</p> <p>Kappa : 0.9316  McNemar's Test P-Value : NA</p> <p>Accuracy : 0.9484  95% CI : (0.9439, 0.9527)  No Information Rate : 0.1135  P-Value [Acc &gt; NIR] : &lt; 2.2e-16</p> <p>Kappa : 0.9426  McNemar's Test P-Value : &lt; 2.2e-16</p>
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Accuracy	Gaussian	Bernoulli
Untouched images	0.5352	0.1135
Stretched bounding box	0.8289	0.8145

Gaussian is better in both untouched and stretched bounding box images. Because Bernoulli only recognize 0 or 1 but Gaussian is continuous.

class: 0	class: 1	class: 2	class: 3	class: 4	class: 5	class: 6	class: 7	class: 8	class: 9
0.8796	0.9524	0.2054	0.3069	0.1334	0.03924	0.9290	0.2218	0.6591	0.9524
0.9707	0.9791	0.9982	0.9859	0.9979	0.99704	0.9437	0.9970	0.8124	0.8019
0.7655	0.8539	0.9298	0.7094	0.8733	0.56452	0.6362	0.8941	0.2749	0.3505
0.9867	0.9938	0.9161	0.9268	0.9136	0.91377	0.9921	0.9179	0.9567	0.9934
0.0980	0.1135	0.1032	0.1010	0.0982	0.08920	0.0958	0.1028	0.0974	0.1009
0.0862	0.1081	0.0212	0.0310	0.0131	0.00350	0.0890	0.0228	0.0642	0.0961
0.1126	0.1266	0.0228	0.0437	0.0150	0.00620	0.1399	0.0255	0.2335	0.2742
0.9252	0.9658	0.6018	0.6464	0.5656	0.51814	0.9364	0.6094	0.7358	0.8772

  

class: 0	class: 1	class: 2	class: 3	class: 4	class: 5	class: 6	class: 7	class: 8	class: 9
0.000	1.0000	0.0000	0.000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
1.000	0.0000	1.0000	1.000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
NaN	0.1135	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
0.902	NaN	0.8968	0.899	0.9018	0.9108	0.9042	0.8972	0.9026	0.8991
0.098	0.1135	0.1032	0.101	0.0982	0.0892	0.0958	0.1028	0.0974	0.1009
0.000	0.1135	0.0000	0.000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
0.000	1.0000	0.0000	0.000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
0.500	0.5000	0.5000	0.500	0.5000	0.5000	0.5000	0.5000	0.5000	0.5000

  

class: 0	class: 1	class: 2	class: 3	class: 4	class: 5	class: 6	class: 7	class: 8	class: 9
0.9541	0.7956	0.8091	0.8446	0.8289	0.7881	0.8800	0.8074	0.7659	0.8196
0.9911	0.9877	0.9816	0.9889	0.9799	0.9847	0.9894	0.9777	0.9629	0.9660
0.9212	0.8923	0.8350	0.8951	0.8181	0.8349	0.8978	0.8058	0.6901	0.7299
0.9950	0.9742	0.9781	0.9826	0.9813	0.9794	0.9873	0.9779	0.9744	0.9795
0.0980	0.1135	0.1032	0.1010	0.0982	0.0892	0.0958	0.1028	0.0974	0.1009
0.0935	0.0903	0.0835	0.0853	0.0814	0.0703	0.0843	0.0830	0.0746	0.0827
0.1015	0.1012	0.1000	0.0953	0.0995	0.0842	0.0939	0.1030	0.1081	0.1133
0.9726	0.8916	0.8954	0.9167	0.9044	0.8864	0.9347	0.8926	0.8644	0.8928

  

class: 0	class: 1	class: 2	class: 3	class: 4	class: 5	class: 6	class: 7	class: 8	class: 9
0.9571	0.7463	0.7771	0.8347	0.8096	0.7814	0.8820	0.8054	0.7567	0.8057
0.9899	0.9873	0.9809	0.9878	0.9792	0.9852	0.9885	0.9737	0.9595	0.9621
0.9116	0.8823	0.8243	0.8846	0.8087	0.8377	0.8904	0.7782	0.6682	0.7045
0.9953	0.9681	0.9745	0.9815	0.9793	0.9787	0.9875	0.9776	0.9734	0.9778
0.0980	0.1135	0.1032	0.1010	0.0982	0.0892	0.0958	0.1028	0.0974	0.1009
0.0938	0.0847	0.0802	0.0843	0.0795	0.0697	0.0845	0.0828	0.0737	0.0813
0.1029	0.0960	0.0973	0.0953	0.0983	0.0832	0.0949	0.1064	0.1103	0.1154
0.9735	0.8668	0.8790	0.9112	0.8944	0.8833	0.9353	0.8896	0.8581	0.8839



Library:

```
1 library(readr)
2 source('Reader.R')
3 library(naivebayes)
4 library(caret)
5
6 source('Reader.R')
7
8 library(caret)
9 library(h2o)
10
11 library(klaR)
12 library(caret)
```

Evaluations:

```
34 #result
35 predictions<-as.data.frame(h2o.predict(rfut_4_10,h2o_df_te_ut))
36 confusionMatrix(predictions[,1], te_labels)
37 predictions<-as.data.frame(h2o.predict(rfut_4_30,h2o_df_te_ut))
38 confusionMatrix(predictions[,1], te_labels)
39
40 predictions<-as.data.frame(h2o.predict(rfut_8_10,h2o_df_te_ut))
41 confusionMatrix(predictions[,1], te_labels)
42 predictions<-as.data.frame(h2o.predict(rfut_8_30,h2o_df_te_ut))
43 confusionMatrix(predictions[,1], te_labels)
44
45 predictions<-as.data.frame(h2o.predict(rfut_16_10,h2o_df_te_ut))
46 confusionMatrix(predictions[,1], te_labels)
47 predictions<-as.data.frame(h2o.predict(rfut_16_30,h2o_df_te_ut))
48 confusionMatrix(predictions[,1], te_labels)
49
50 predictions<-as.data.frame(h2o.predict(rfBounded_4_10,h2o_df_te_bounded))
51 confusionMatrix(predictions[,1], te_labels)
52 predictions<-as.data.frame(h2o.predict(rfBounded_8_10,h2o_df_te_bounded))
53 confusionMatrix(predictions[,1], te_labels)
54 predictions<-as.data.frame(h2o.predict(rfBounded_16_10,h2o_df_te_bounded))
55 confusionMatrix(predictions[,1], te_labels)
56
57 predictions<-as.data.frame(h2o.predict(rfBounded_4_30,h2o_df_te_bounded))
58 confusionMatrix(predictions[,1], te_labels)
59 predictions<-as.data.frame(h2o.predict(rfBounded_8_30,h2o_df_te_bounded))
60 confusionMatrix(predictions[,1], te_labels)
61 predictions<-as.data.frame(h2o.predict(rfBounded_16_30,h2o_df_te_bounded))
62 confusionMatrix(predictions[,1], te_labels)
```