Assignment 10

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Problem 1

a)

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Submitted Fisher_Score function.

Code for part 1a: hw10_1a

Top 20 dims with their fisher_scores(sorted according to Fisher score)

48 0.319180730643819 25 0.214007577983443 21 0.190969964698631 70 0.189213760572332 65 0.169320028160465 40 0.167344782823831 29 0.165045125172703 19 0.140196441194676 57 0.125453639029847 20 0.121208166902893 24 0.0995248083025423 30 0.0950242089262554 12 0.0858179354076155 47 0.0846436482183233 61 0.0606773370724610 10 0.0579159433174404 34 0.0526784350453484 27 0.0462159428948702 39 0.0461136373408911

Or sorted according to the dimension:

0.0421851727680418

- 10 0.0579159433174404
- 12 0.0858179354076155
- 19 0.140196441194676
- 20 0.121208166902893
- 21 0.190969964698631
- 24 0.0995248083025423
- 25 0.214007577983443
- 27 0.0462159428948702
- 29 0.165045125172703

30 0.0950242089262554 34 0.0526784350453484 39 0.0461136373408911 40 0.167344782823831 41 0.0421851727680418 47 0.0846436482183233 48 0.319180730643819 57 0.125453639029847 61 0.0606773370724610

65

70

b) Submitted AUROC_score function. Code for part 1b: hw10_1b

0.169320028160465

0.189213760572332

AUROC (sorted according to AUROC score):

- 25 0.733986318407961
- 48 0.713308457711443
- 40 0.688743781094527
- 29 0.683690920398010
- 21 0.683302238805970
- 67 0.672963308457711
- 70 0.670747823383085
- 11 0.669465174129353
- 47 0.666083644278607
- 65
- 0.662002487562189 12 0.645872201492537
- 24 0.643229166666667
- 39 0.641169154228856
- 6 0.638254042288557
- 19 0.631452114427861
- 57 0.626982276119403
- 20 0.620841106965174
- 34 0.617381840796020
- 5 0.616759950248756
- 14 0.609025186567164

Or sorted according to the dimension:

- 5 0.0461136373408911
- 6 0.0846436482183233

- 11 0.140196441194676
- 12 0.0995248083025423
- 14 0.0421851727680418
- 19 0.0606773370724610
- 20 0.0526784350453484
- 21 0.169320028160465
- 24 0.0950242089262554
- 25 0.319180730643819
- 29 0.189213760572332
- 34 0.0462159428948702
- 39 0.0858179354076155
- 40 0.190969964698631
- 47 0.125453639029847
- 48 0.214007577983443
- 57 0.0579159433174404
- 65 0.121208166902893
- 67 0.167344782823831
- 70 0.165045125172703

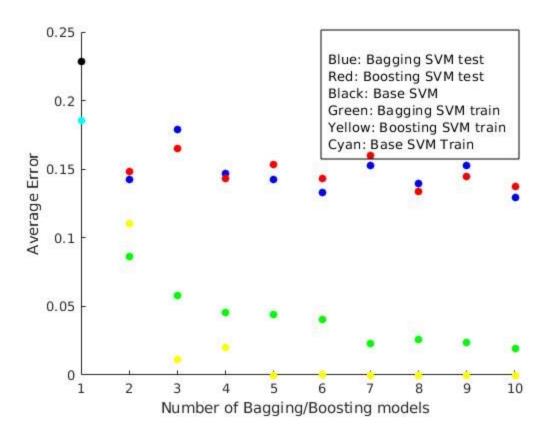
More than 50% of the dimensions are same. Both are measuring how separable the data is, so I expect them to have similar results. If means are near and SD is high, TPR and TNR should also be low and hence AUROC should be low as well as Fisher scores should be low. However, the algorithm used in both are different, so the results could be little different. AUROC doesn't consider mean or variance explicitly. It just moves a threshold and finds TPR and TNR.

Problem 2

For different curves the errors are coming different each time for all the following parts.

a) Code submitted

Test Error_5 =0.185714285714286, Test Error_base =0.20000000000000, training_Error_5 =.114285714285714, training_Error_base =0.300000000000000



Plotted are the errors for bagged and boosted for SVM

Bagging and boosting gives better result than Base.

The SVM is overfitted with boosting as the training error is 0 and test is high.

The error decrease with more number of models bagging and boosting generally.

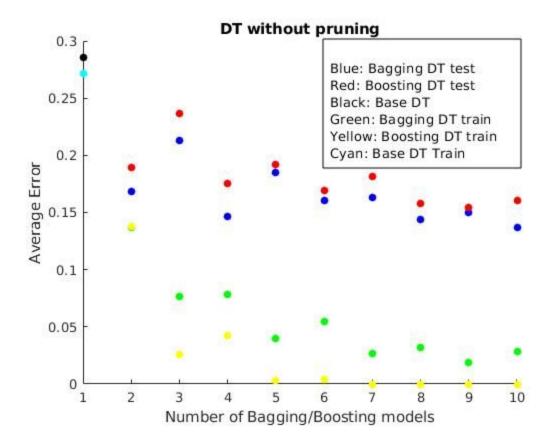
After T=6, training is decreasing but testing reached plateau bagging. Same is applicable for boosting for T=2.

b)Code submitted

Decision Trees
Plotted errors

Bagging and boosting gives better result than Base.

The error decrease with more number of models bagging and boosting generally. SVM results are better than Decision Tree. Again, The DT like SVM is overfitted with boosting as the training error is 0 and test is high. Bagging train and test both are decreasing although there is big gap between them. Bagging + regularization should help.



c)Code submitted
Decision Trees of only one split.
Plotted errors

Bagging and boosting gives better result than Base.

The error is little decrease with more number of models bagging and boosting generally. Very interesting trend it shows that test error is less than training. Here Base model has very high training error. It maybe due to less vrience in test as compared to training (the test data is easier compared to train). The algorithm **does not** overfit for bagging. Boosting training error is decreasing but increases after some steps(7 in this run) showing overfit. The gap between training and testing is lower as compared to SVM and Full Decision Trees.

