

# Report AML

Prob1)

a) When we compare the performance of inbuilt and our implementation, the inbuilt gives better accuracy in less time. Our implementation takes roughly 3 to 4 times more time and accuracy is nearly 10% less on an average.

For  $n_{\text{estimators}} = 80$  and  $m = 8$

## OUR IMPLEMENTATION

```
The accuracy is :  
0.908695652174  
Time in seconds :  
1.101469
```

```
The accuracy is :  
0.868115942029  
Time in seconds :  
1.398869
```

```
The accuracy is :  
0.901449275362  
Time in seconds :  
1.255522
```

## Inbuilt

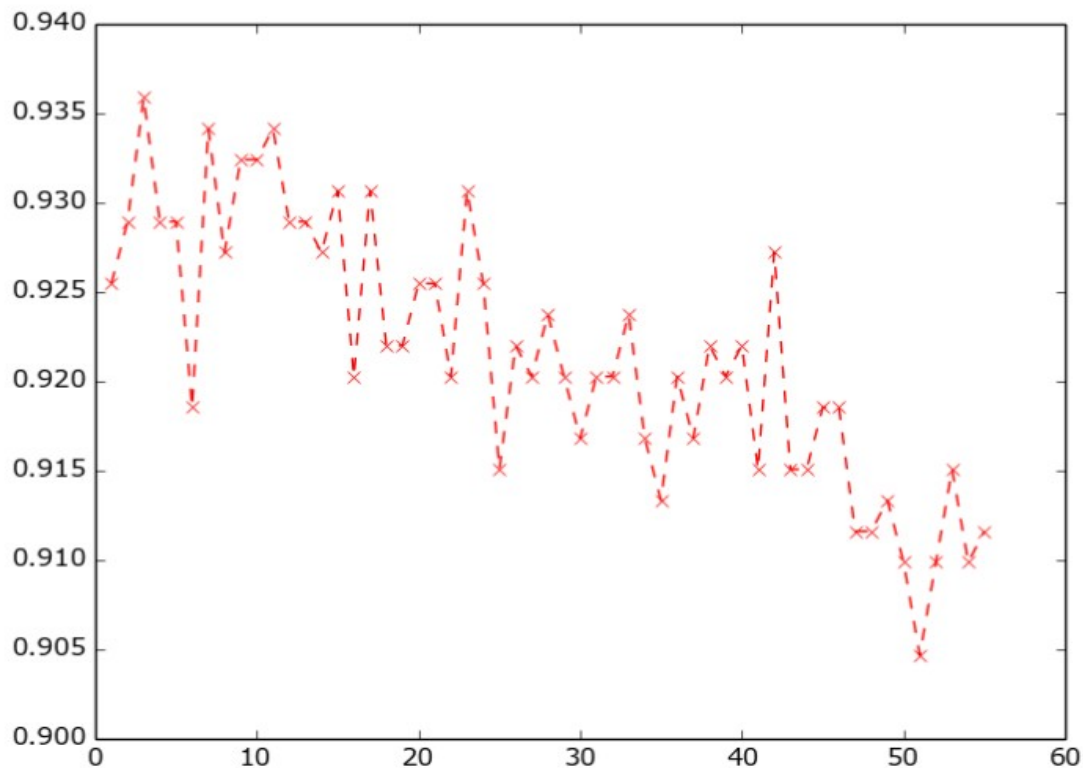
```
The accuracy is :  
0.954380883418  
Time in seconds :  
0.389075
```

```
The accuracy is :  
0.955104996379  
Time in seconds :  
0.420919
```

```
The accuracy is :  
0.958001448226  
Time in seconds :  
0.369429
```

b) Comparing the sensitivity vs the m values

We notice that sensitivity decreases as value of m increases



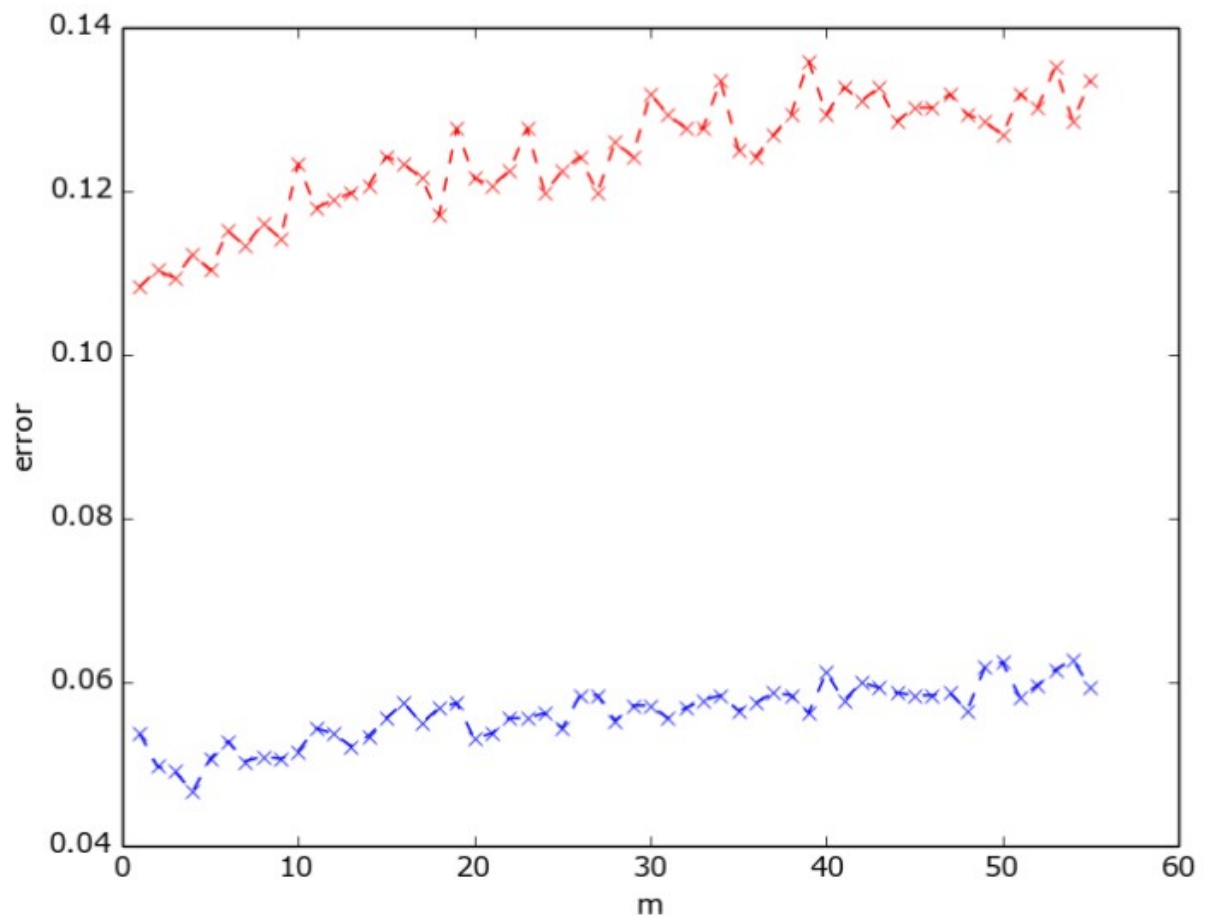
Sensitivity, also called Recall or True Positive Rate, is the proportion of actual positives which are correctly identified as positives by the classifier.

$$\text{Sensitivity} = \frac{TP}{(TP + FN)}$$

### c) OOB and test error vs m values

Each tree is constructed using a different bootstrap sample from the original data. About one-third of the cases are left out of the bootstrap sample and not used in the construction of the kth tree.

We put each case left out in the construction of the kth tree down the kth tree to get a classification. In this way, a test set classification is obtained for each case in about one-third of the trees. At the end of the run, take  $j$  to be the class that got most of the votes every time case  $n$  was oob. The proportion of times that  $j$  is not equal to the true class of  $n$  averaged over all cases is the oob error estimate.



OOB ERROR is BLUE  
TEST ERROR is RED

We notice that test error and oob error increases as value of  $m$  increases.  
We find optimal split value of  $m$  between 0-10

3 prob)

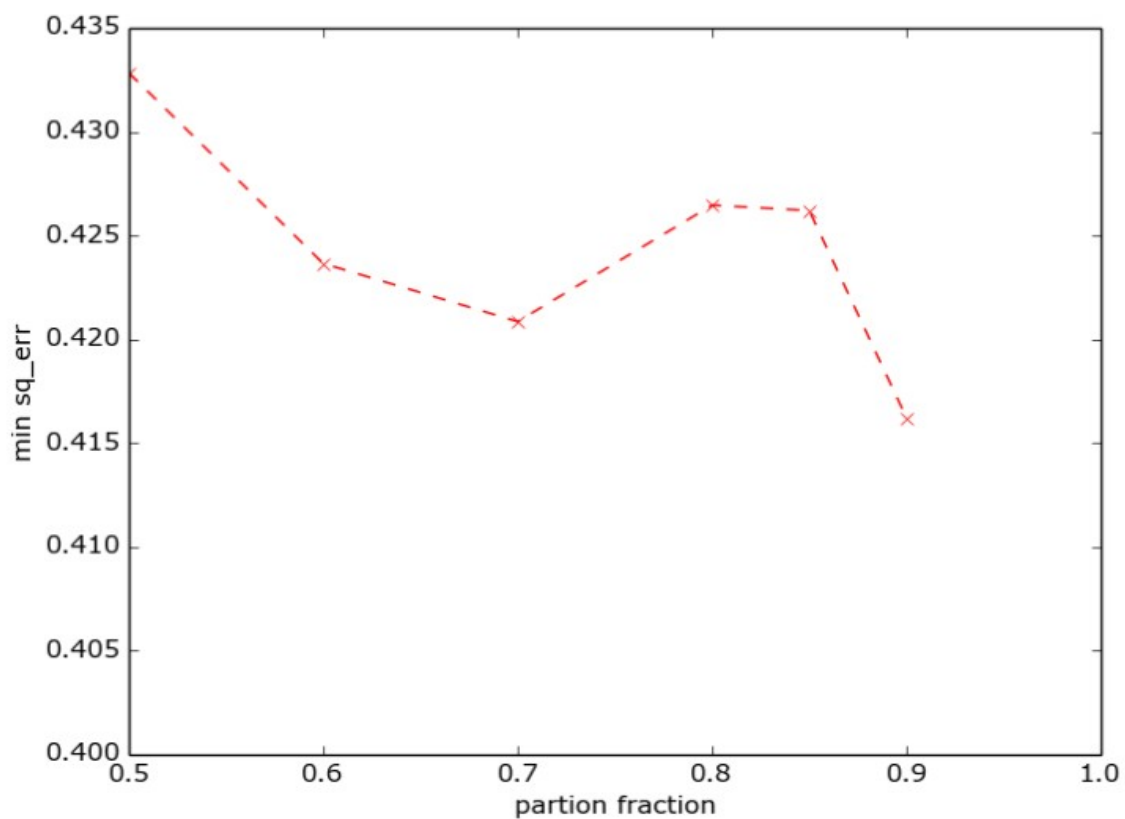
e) We find that  $\lambda$  around 50-60 is the best value we can get for  $\lambda$ .  
We find that the fifth attribute is the most significant attribute.

When we remove two to three least significant attributes, we find that square mean error doesn't change much.

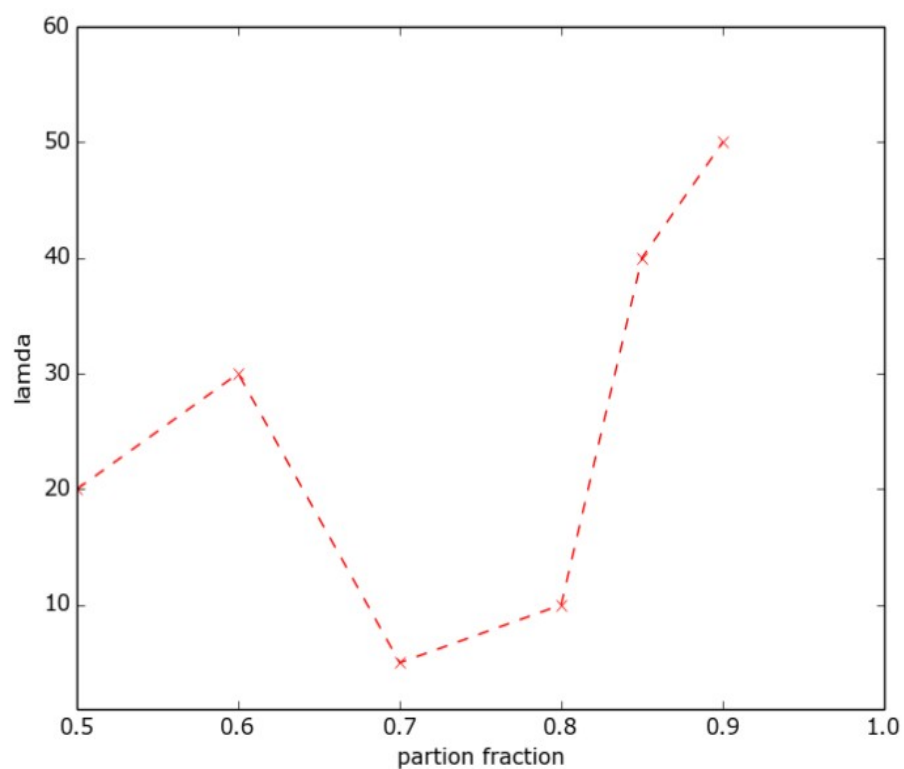
f) Yes the effect of  $\lambda$  changes for different partitions of data. This can be seen from the graphs plotted which have been submitted.

g) minsq\_error vs partition fraction graph

We observe that as partition size increases error decreases.



Lamda vs partition fraction graph

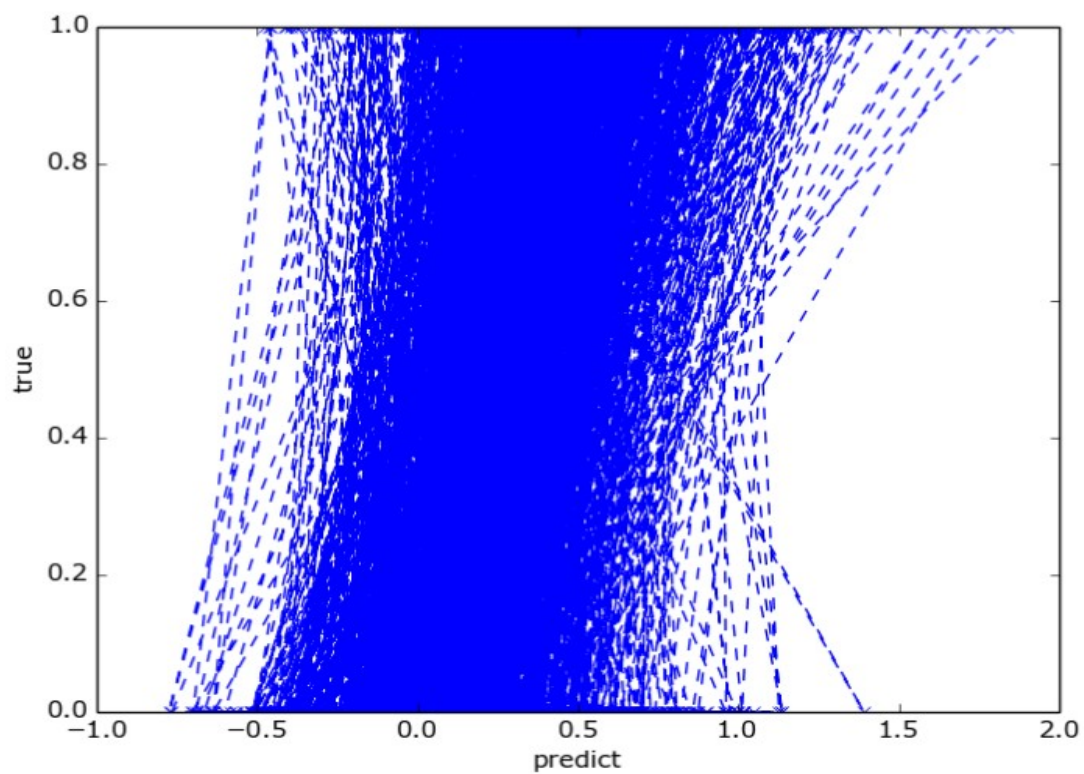


h) Train data

True vs predicted

We can see that the data is around the 45 degree axis

We can say the model is good



Test data: true vs predicted

We can see that the data is around the 45 degree axis

We can say the model is good even for test set

