

COL774 - MACHINE LEARNING

ASSIGNMENT 3 REPORT

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2018MCS2143

1. Decision Trees

Continuous Attributes	Binary Attributes	Categorical Attributes
X1	X2	X3
X5		X4
X12		X6
X13		X7
X14		X8
X15		X9
X16		X10
X17		X11
X18		
X19		
X20		
X21		
X22		
X22		

Table 1: Different types of attributes in the data set

For continuous attributes, I converted them to binary based on whether the value is greater than the median threshold or not.

For binary attr, I did boolean(two-way) split.

For categorical, I did the multi-way split.

a. Accuracies against number of nodes in the tree as tree grows

BFS growth

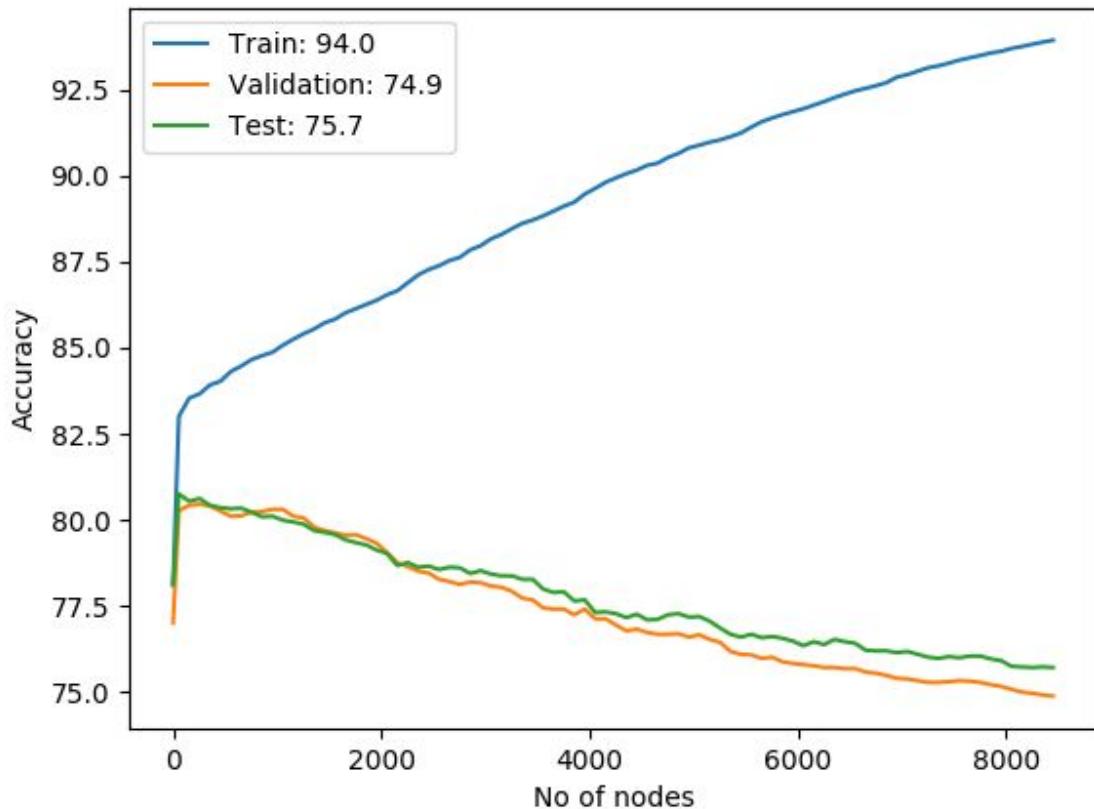
Tree height 22

Number of nodes 8555

Accuracy (training set) 93.97222222222223

Accuracy (validation set) 74.83333333333333

Accuracy (testing set) 75.68333333333334



Observations:

Decision Tree with a single node predicts the majority class giving the accuracy of ~78%. As number of nodes increases, Training accuracy increases while Testing and Validation accuracies decrease i.e. overfitting happens.

b. Post pruning based on validation set

BFS growth

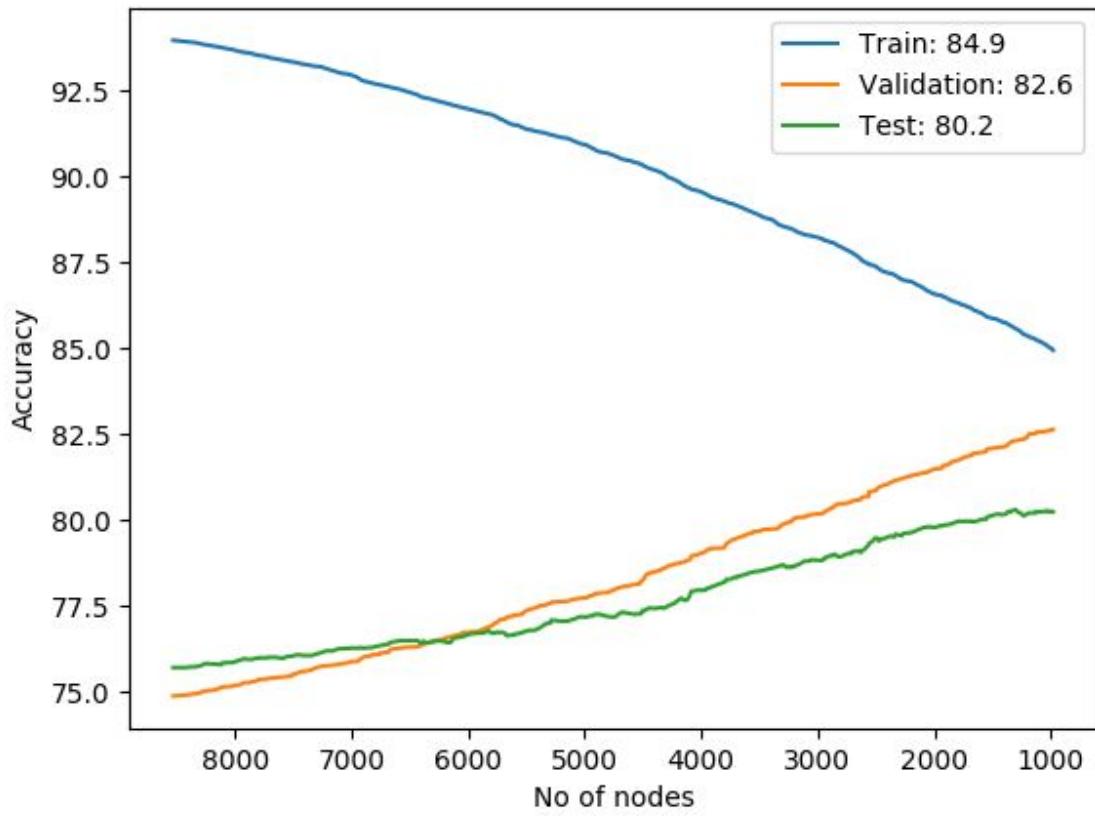
Tree height 15

Number of nodes 981

Accuracy (training set) 84.93888888888888

Accuracy (validation set) 82.63333333333334

Accuracy (testing set) 80.23333333333333



Observations:

Pruning decreases the height of tree to 15 and number of nodes from 8555 to 981 improving validation accuracy from 74% to 82% and testing accuracy from 75% to 80%. And hence helps in generalizing well while reducing overfitting.

c. Using medians dynamically (without pruning)

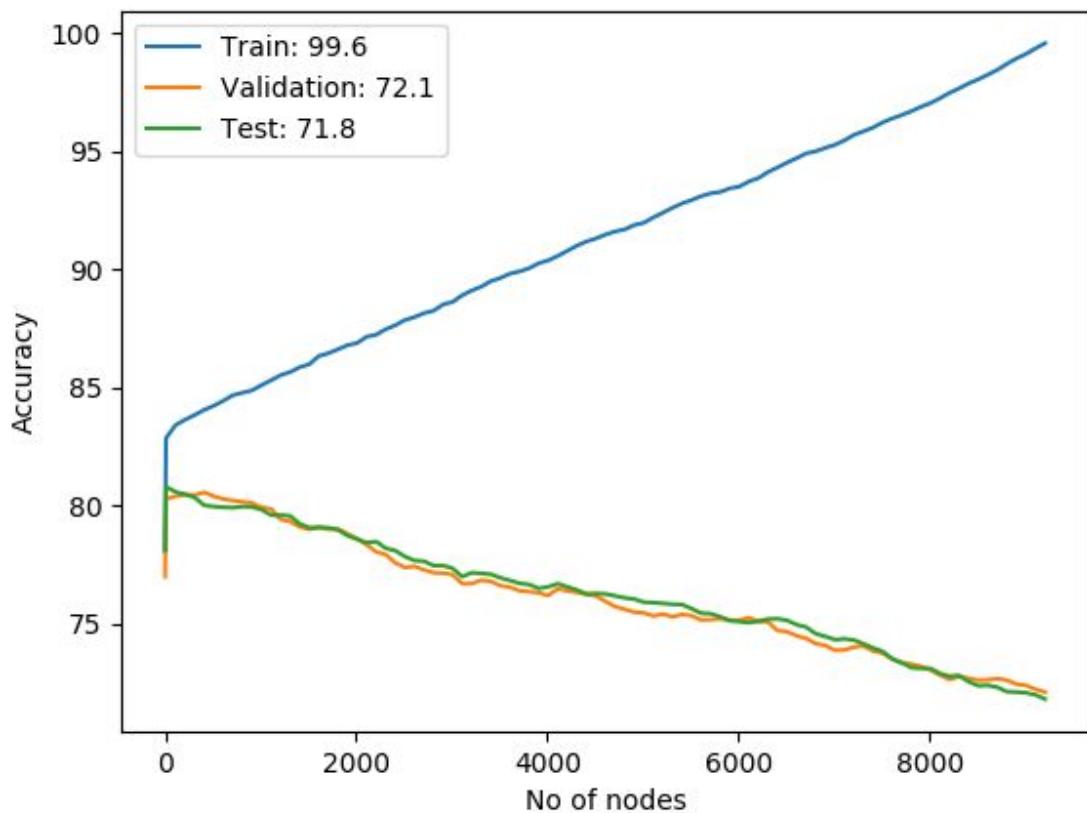
Tree height 19

Number of nodes 9309

Accuracy (training set) 99.79444444444444

Accuracy (validation set) 72.01666666666667

Accuracy (testing set) 71.75



Numerical Attributes split multiple times in a branch:

1 [120000.0, 50000.0, 70000.0, 80000.0, 95000.0, 110000.0]

5 [36.0, 41.5, 48.0, 50.0, 51.0, 54.0]

12 [107948.0, 34711.5, 59307.0, 46385.0]

13 [44413.0, 49256.0, 59156.5]

14 [19633.0, 28010.0, 25103.0, 27376.0]
15 [40385.0, 48097.0, 48635.0]
16 [39369.0, 45794.0, 41296.0]
17 [15500.5, 18929.0, 17132.0]
18 [3101.5, 2000.0, 1530.0]
19 [1500.0, 1602.0, 1803.0]
20 [1500.0, 1287.0, 1058.0]
21 [2882.5, 3360.0, 5000.0]
22 [1162.0, 2000.0, 1458.0]
23: [2453.0, 4031.0]

Observations:

The training accuracy boosts to 99.6 while testing and validation set accuracy is ~72% which shows how bad it overfits the data. Also as same attributes are split multiple times based on median, number of nodes of tree is increased.

d. Using Sklearn library

Scikit-learn implementation :

- (i) min_sample_leaf : A split at any depth will only be considered if it leaves at least min_sample_leaf samples in both left and right branches. node.
- (ii) min_sample_split : Min samples required to split an internal
- (iii) max_depth : Max height of the tree.

With default parameters

(max_depth=None,min_samples_split=2,min_samples_leaf=1)

Accuracy on training set 99.96111111111111

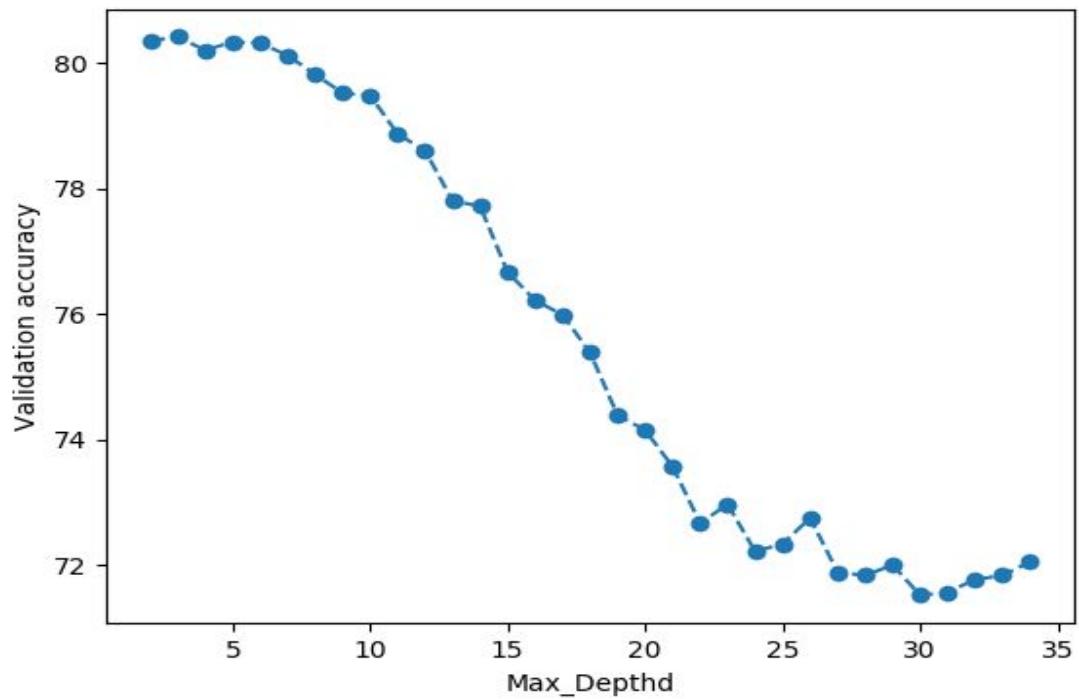
Accuracy on validation set 72.39999999999999

Accuracy on testing set 72.85000000000001

Height of tree 53

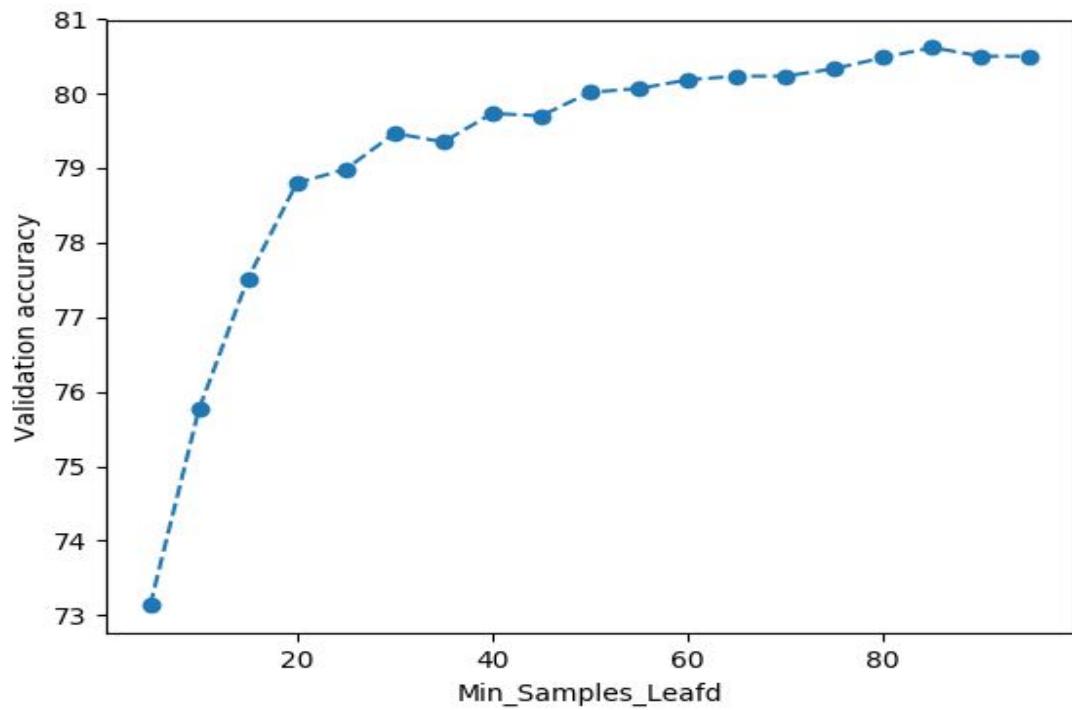
No of nodes 5127

max_dept h	2	5	7	10	15	20	50
Training	82.86	83.45	83.95	85.73	91.10	95.53	100
Testing	80.85	80.86	80.6	79.83	75.73	72.8	70.88
Validation	80.35	80.41	80.3	79.61	75.11	72.91	71.41



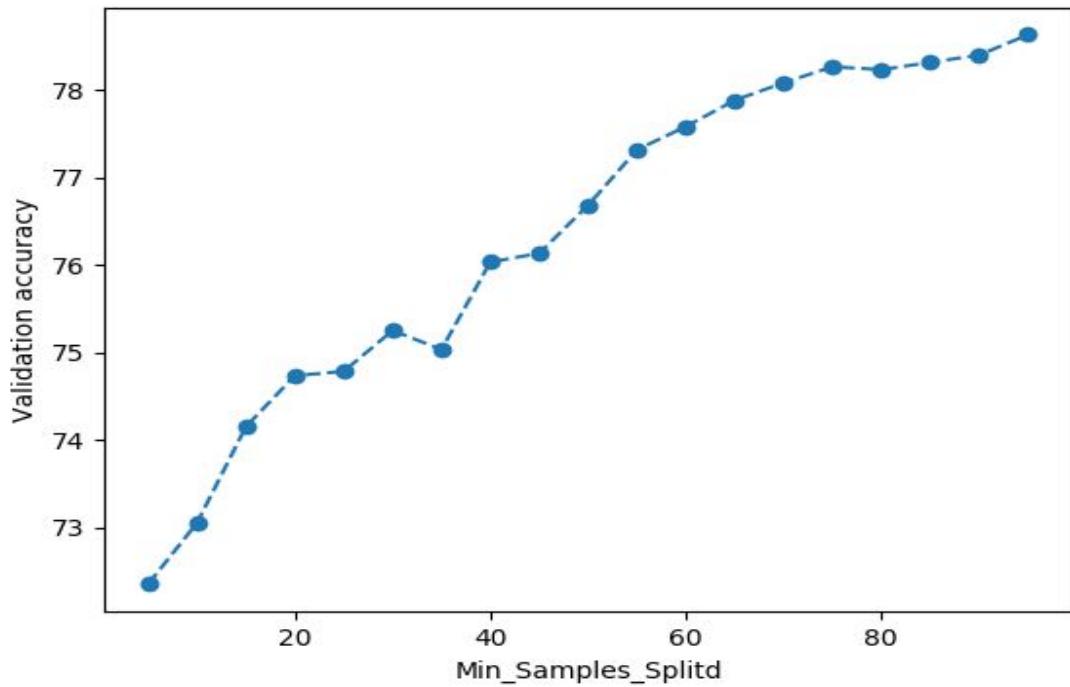
Validation accuracy drops with increase in max_depth.

min_sample_leaf	5	15	20	30	50	150	500
Training	92.03	85.98	85.13	84.16	83.70	83.09	82.86
Testing	71.91	77.36	77.4	79.98	80.03	80.76	80.85
Validation	72.86	77.9	78.6	79.68	79.88	80.26	80.35



Validation accuracy increases with increase in min_samples_leaf.

min_sample_split	2	5	10	20	50	100	500
Training	100	98.23	94.95	91.20	87.01	85.26	83.43
Testing	70.71	70.31	72.38	73.33	75.56	77.6	80.36
Validation	71.26	70.45	72.85	73.16	75.95	78.35	80.06



Validation accuracy increases with increase in min_samples_split.

After running grid parameter search, parameters with best validation accuracy are:

Parameters : {'min_samples_split': 95, 'max_depth': 5, 'min_samples_leaf': 70}

Train set accuracy: 83.16666666666667

Validation set accuracy: 80.48333333333333

Test set accuracy: 80.78333333333333

Execution time 3050.06785297

Observations:

Training accuracy decreases than part c and is almost same as of part b. However, testing and validation accuracy is increased than part c and is almost same as that in part b. Therefore, the results it produces are close to the results produced by post pruning.

e. Using one hot encoding

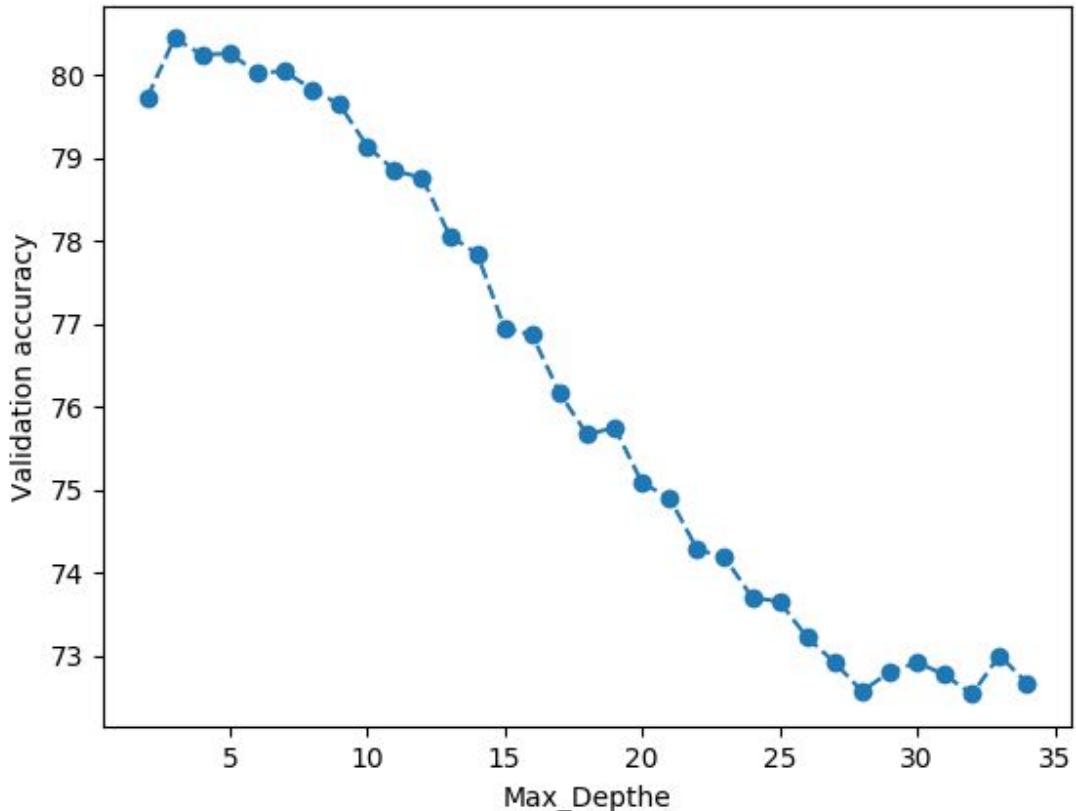
Accuracy on training set 99.96111111111111

Accuracy on validation set 72.38333333333333

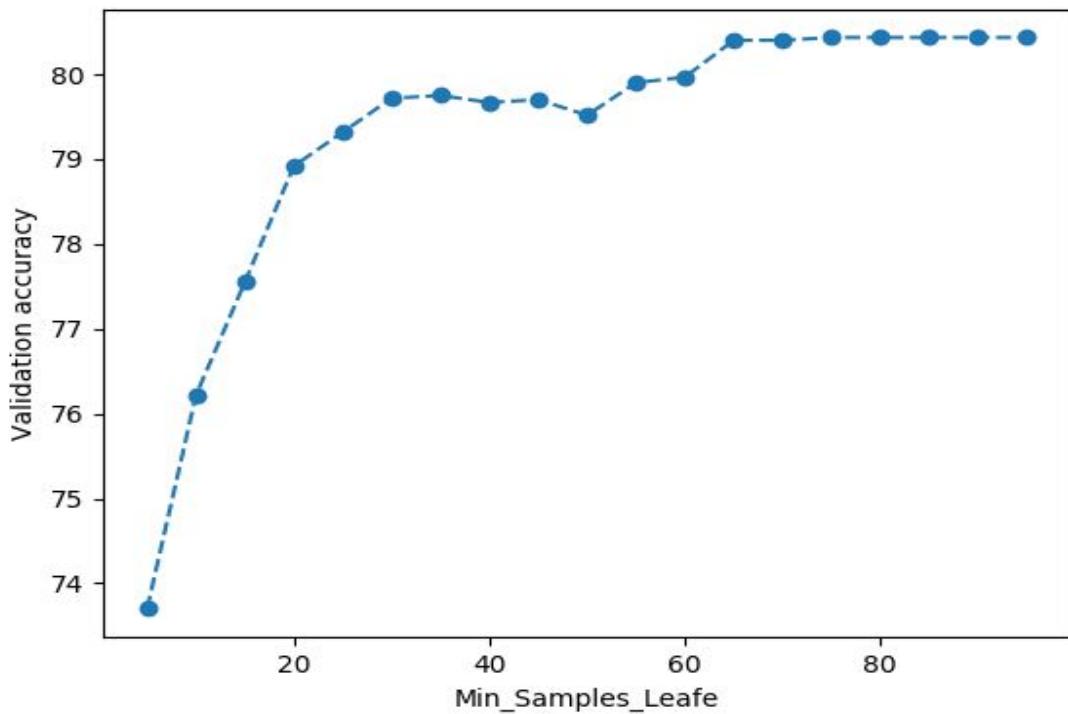
Accuracy on testing set 72.45

Height of tree 40

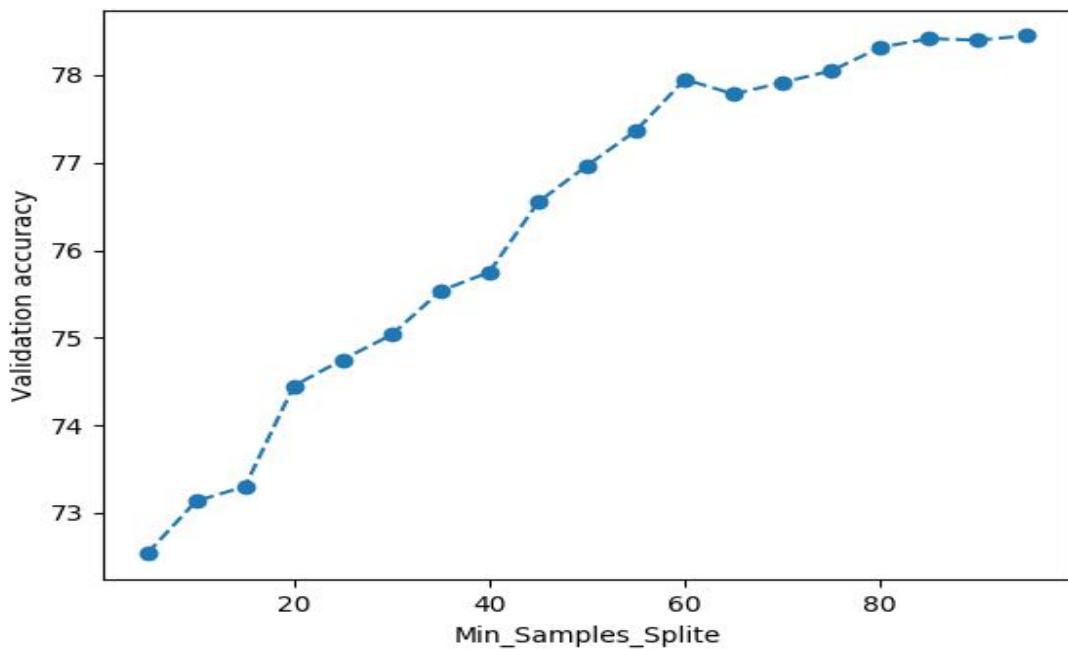
No of nodes 5197



Validation accuracy decreases with increase in max_depth.



Validation accuracy increases with increase in `min_samples_leaf`



Validation accuracy increases with increase in `min_samples_leaf`.

Using grid parameter search:

Parameters : {'min_samples_split': 95, 'max_depth': 7, 'min_samples_leaf': 35}

Train set accuracy: 83.43888888888888

Validation set accuracy: 80.48333333333333

Test set accuracy: 80.61666666666667

Execution time 2115.11174202

With previous settings i.e. max_depth = 5, min_samples_split = 95, min_samples_leaf = 70

Accuracy on training set 83.0111111111112

Accuracy on validation set 80.15

Accuracy on testing set 80.93333333333334

Height of tree 5

No of nodes 41

Observations:

Training accuracy decreases than part c and is almost same as of part b & d.

However, testing and validation accuracy is increased than part c and is almost same as that in part b & d. Therefore, one hot encoding is not showing improvement over the accuracies.

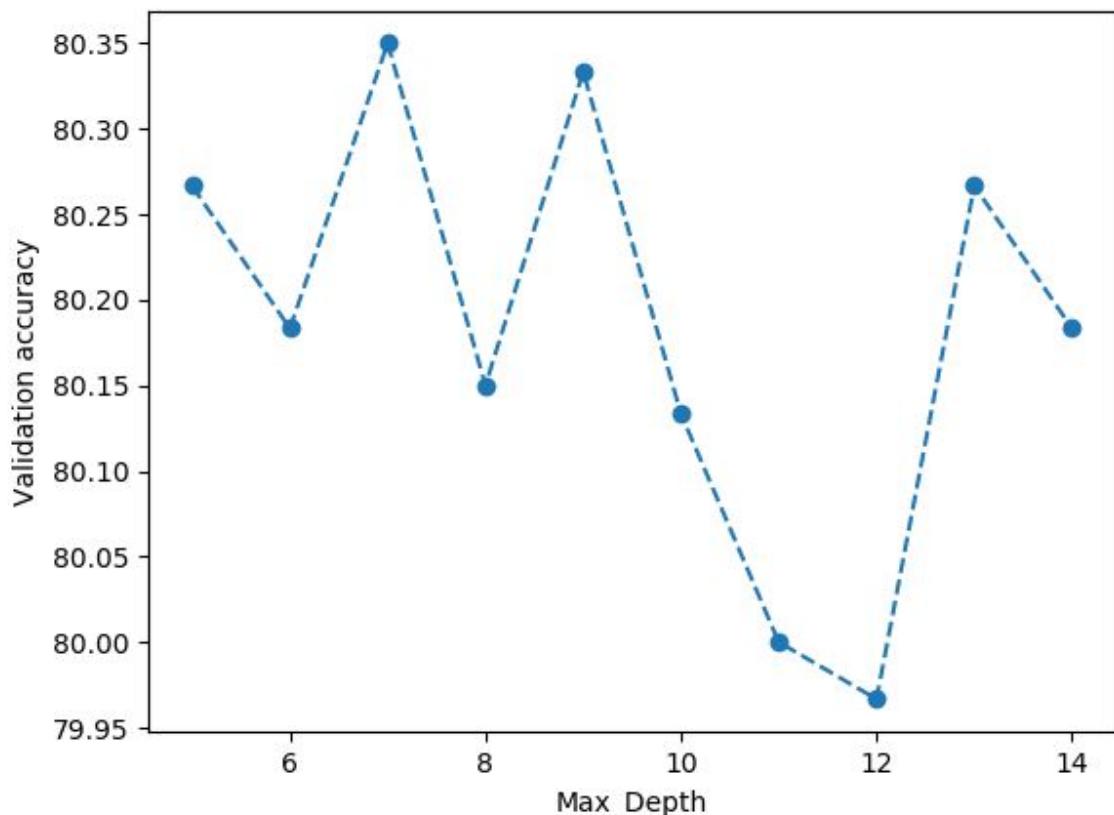
f. Random Forest using sklearn

Using default parameters: 'max_features': auto, 'n_estimators': 10, 'bootstrap': True, 'max_depth': None

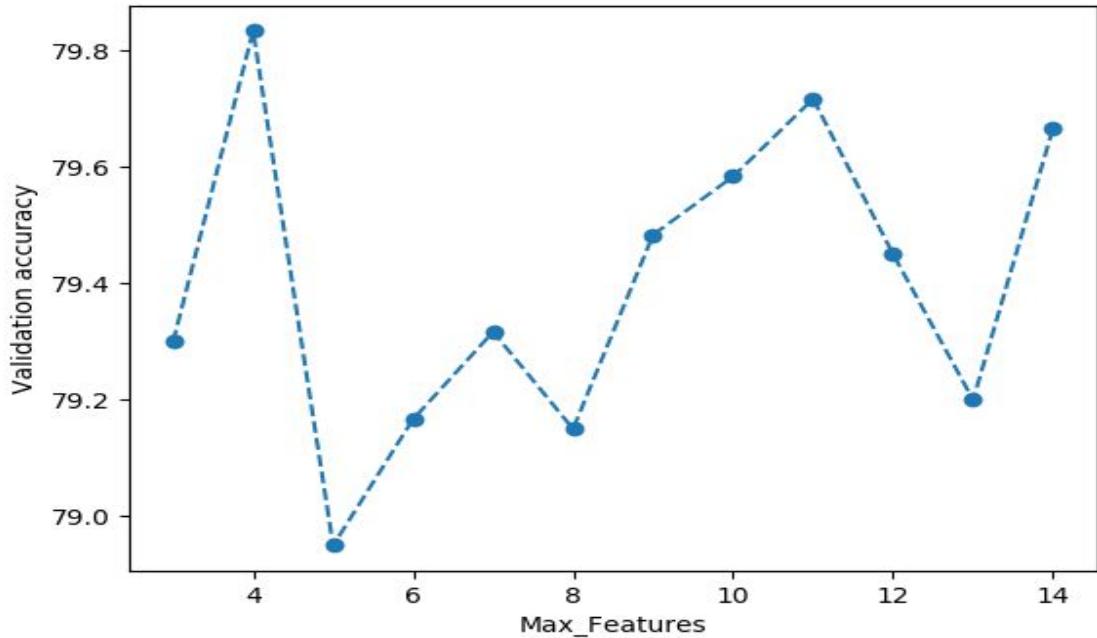
Accuracy on training set 98.26111111111111

Accuracy on validation set 79.83333333333333

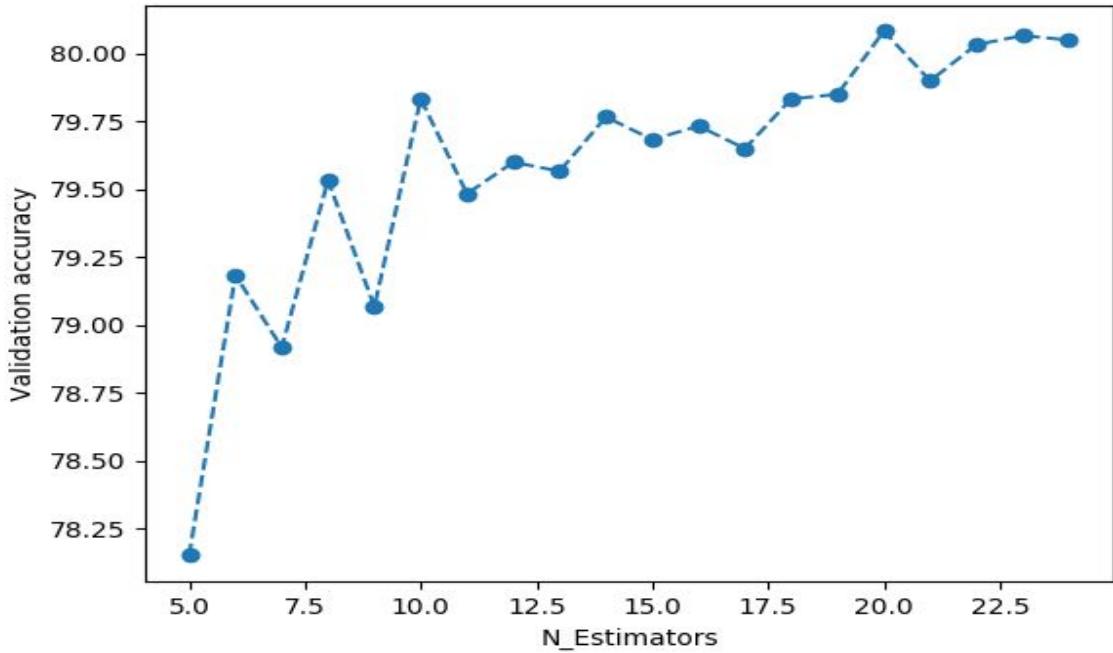
Accuracy on testing set 79.63333333333334



Validation accuracy decreases at some values and increases at some i.e. no fixed pattern is there.



On max_features also, validation accuracy increases at some values and decreases at some i.e. no fixed pattern is followed.



Validation accuracy increases with n_estimators i.e. number of trees in forest.

bootstrap	True	False
Training accuracy	98.18	100
Testing accuracy	79.25	79.43
Validation accuracy	79.43	78.83

Best parameters using grid search:

Parameters : {'max_features': 7, 'n_estimators': 21, 'bootstrap': False, 'max_depth': 11}

Train set accuracy: 87.59444444444443

Validation set accuracy: 80.78333333333333

Test set accuracy: 80.85

Execution time 8202.64859104

Observations :

Training accuracy decreases than part c and is almost same as of part b, d & e. However, testing and validation accuracy is increased than part c and is almost same as that in part b, d & e. Therefore, random forest generalizes quite well as done by post-pruning.

2. Neural Networks

- a. The link for the one-hot encoding of train and test data is as follows:

For one_hot_train.csv

<https://drive.google.com/open?id=13zdqEg1qHc4VyA-DjyS67KoHkSmpQo3A>

For one_hot_test.csv

<https://drive.google.com/open?id=1-CL8AayM93nS4hTmB2nOu27vQffktJfV>

- b. Neural Network implemented

- c. Single hidden layer.

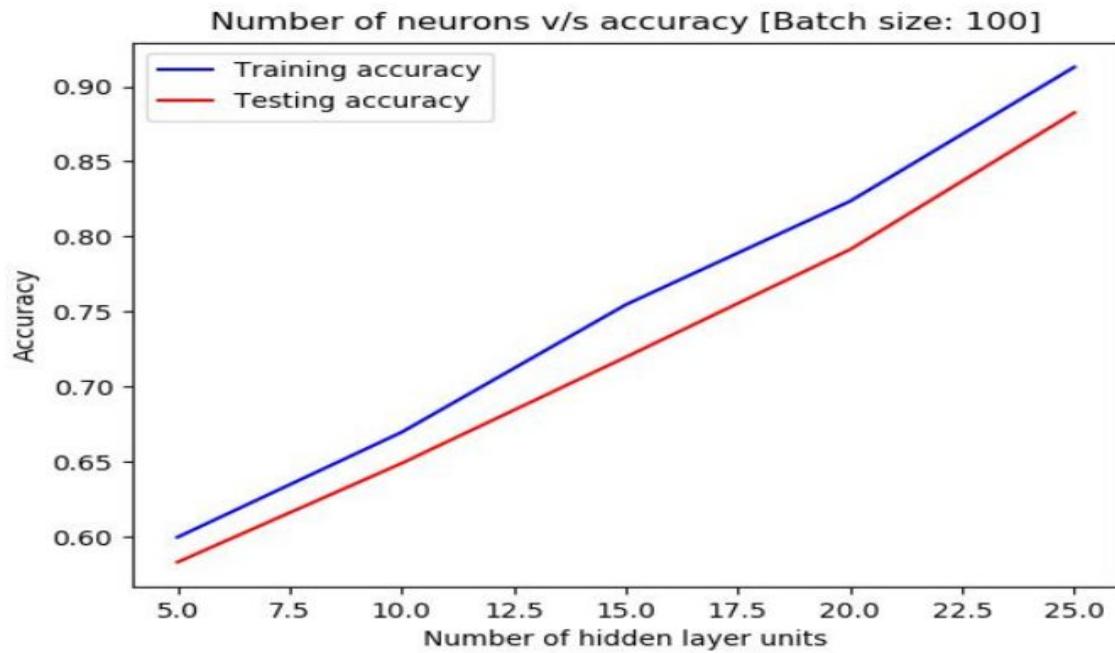
The neural network was tested with a single hidden layer and by varying number of units in that layer.

Number of neurons: [5, 10, 15, 20, 25]

Stopping criteria :

- Max epochs: 2500
- $\text{abs}(\text{Loss}_{t+1} - \text{Loss}_t) \leq 10^{-9}$

Neurons	Train Accuracy	Test Accuracy	Training time(sec)
5	0.59956017	0.582915	4697.125
10	0.6695721	0.648863	4948.525
15	0.7545781	0.71961	5032.4703
20	0.82351059	0.791329	5156.3362
25	0.91291483	0.88253	5557.0028



By increasing the number of units in the hidden layer accuracy has gone up. This may be because of the fact that with more neurons we get more parameters and our model learns better. But if we increase it by large number, the model may overfit.

For 5 neurons in single hidden layer

For 10 neurons in hidden layer

For 15 neurons in hidden layer

For 20 neurons in hidden layer

For layers [85,20,10]										
Actual	Predicted									
	0	1	2	3	4	5	6	7	8	9
	0	465011	36198	0	0	0	0	0	0	0
	1	96180	326318	0	0	0	0	0	0	0
	2	1556	46066	0	0	0	0	0	0	0
	3	818	20303	0	0	0	0	0	0	0
	4	2481	1404	0	0	0	0	0	0	0
	5	1879	117	0	0	0	0	0	0	0
	6	4	1420	0	0	0	0	0	0	0
	7	0	213	0	0	0	0	0	0	0
	8	8	4	0	0	0	0	0	0	0
	9	2	1	0	0	0	0	0	0	0

For 25 neurons in hidden layer

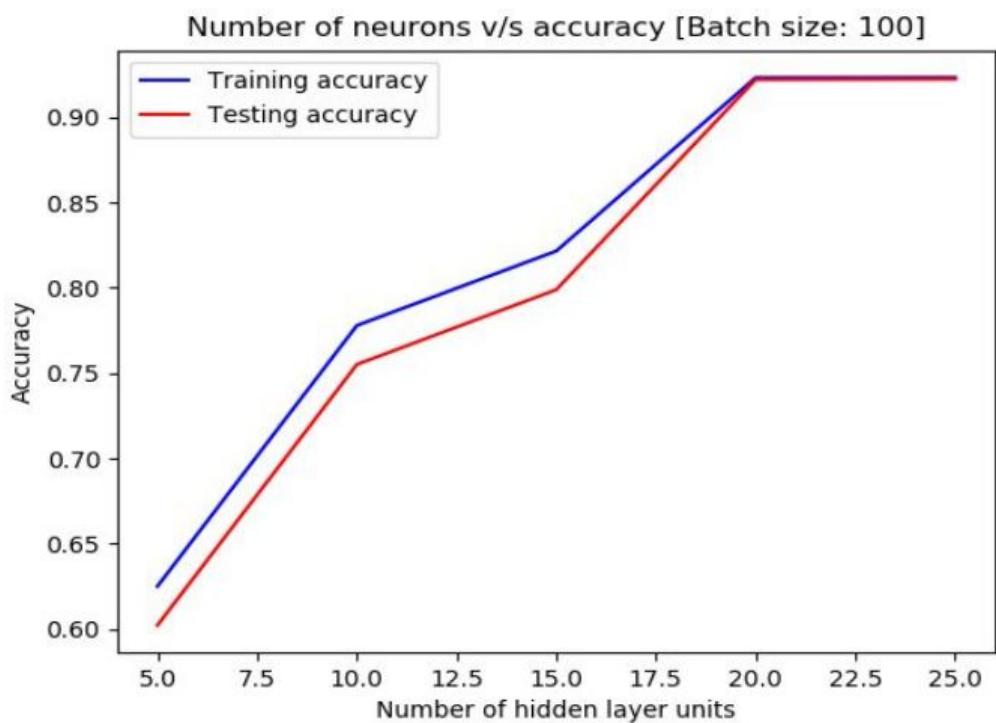
d. 2 hidden layers and same neurons in both of them

Stopping criteria : 2000 epochs

Error threshold: 10^{-9}

The accuracy improved with addition of one more hidden layer. However, there wasn't much improvement after 20 neurons units.

Neurons	Train Accuracy	Test Accuracy	Training time(sec)
5	0.62514994	0.6023	4997.125
10	0.777968812	0.755127	5991.868798
15	0.8215920	0.79895	6031.3261
20	0.92323070	0.9221	6186.97881
25	0.92331075	0.922361	8264.47905



For 5 5 neurons in two hidden layers

For 2 layers [85, 5, 5, 10] False False

For 10 neurons in two hidden layers

For 2 layers [85, 10, 10, 10] FalseFalse

For 15 neurons in two hidden layers

For 2 layers [85, 15, 15, 10] False False

For 20 20 neurons in two hidden layers

For 2 layers [85, 20, 20, 10] False False

For 25 25 neurons in two hidden layers

		For layers [85,25,25,10]FalseFalse									
		Predicted									
		0	1	2	3	4	5	6	7	8	9
0	500929	280	0	0	0	0	0	0	0	0	0
1	1066	421432	0	0	0	0	0	0	0	0	0
2	39	47622	0	0	0	0	0	0	0	0	0
3	0	21121	0	0	0	0	0	0	0	0	0
4	3762	123	0	0	0	0	0	0	0	0	0
5	1995	1	0	0	0	0	0	0	0	0	0
6	0	1424	0	0	0	0	0	0	0	0	0
7	0	230	0	0	0	0	0	0	0	0	0
8	10	2	0	0	0	0	0	0	0	0	0
9	0	3	0	0	0	0	0	0	0	0	0

e. Adaptive learning rate

There wasn't any improvement in the accuracy when adaptive learning with tol = 10 ** -4 was used. Some accuracies remained same as earlier while some became even worse.

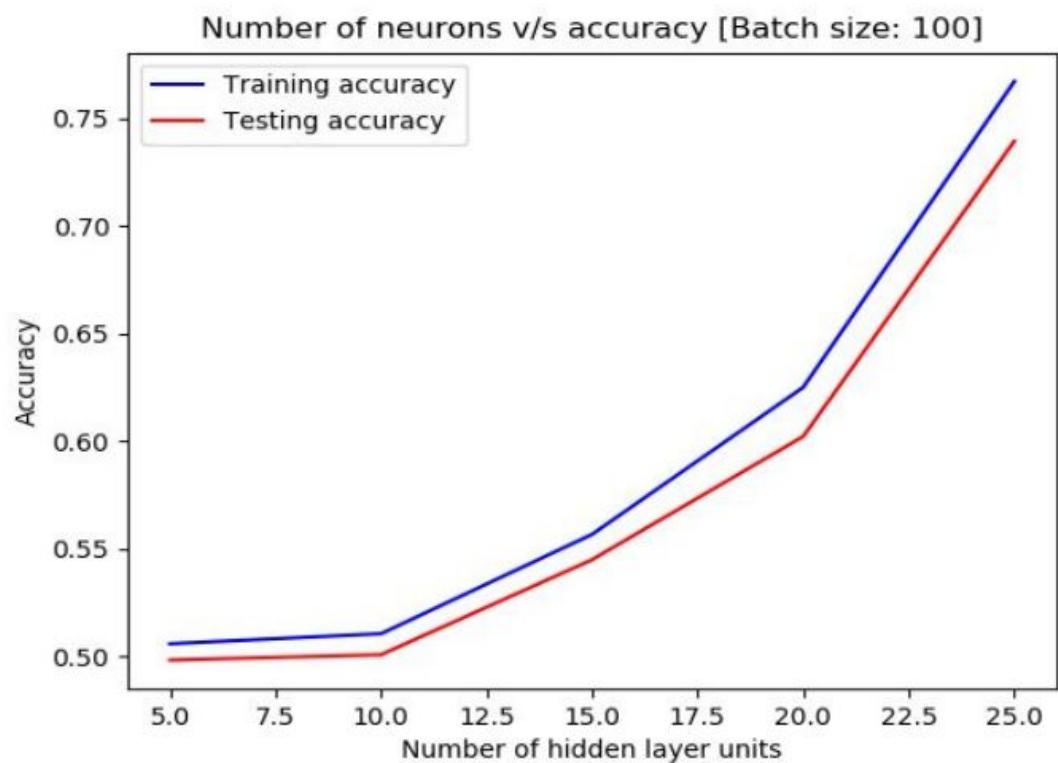
Stopping condition:

Error threshold: 10 ** -9

Epochs: 1000

i. Single hidden layers

Neurons	Train Accuracy	Test Accuracy	Training time(sec)
5	0.50603758	0.49842	1440.8093
10	0.51067572	0.500934	1421.091
15	0.556857	0.54508	2308.1100
20	0.62514994	0.6023	3638.7678
25	0.76721201	0.73948	4826.8413



For 5 neurons in single layer

For layers [85,5,10]

For 10 neurons in single layer

For layers [85,10,10]

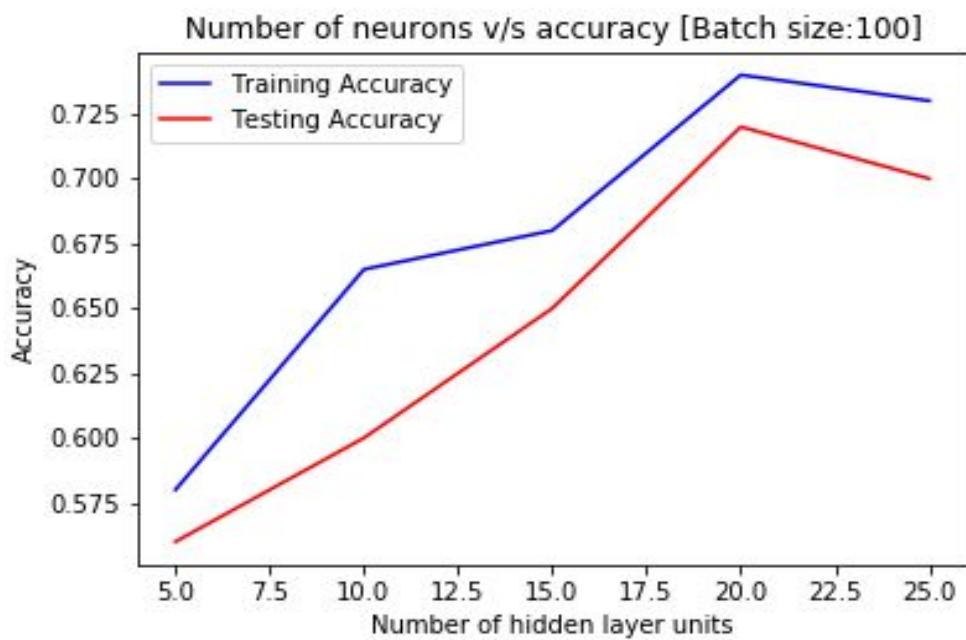
For 15 neurons in single layer

For 20 neurons in single layer

For 25 neurons in single layer

		Predicted									
		0	1	2	3	4	5	6	7	8	9
Actual	0	445353	55856	0	0	0	0	0	0	0	0
	1	112721	301684	0	0	0	0	0	0	0	0
	2	1491	46131	0	0	0	0	0	0	0	0
	3	2467	18654	0	0	0	0	0	0	0	0
	4	2076	1717	0	0	0	0	0	0	0	0
	5	1795	121	0	0	0	0	0	0	0	0
	6	13	1417	0	0	0	0	0	0	0	0
	7	7	225	0	0	0	0	0	0	0	0
	8	7	5	0	0	0	0	0	0	0	0
	9	3	0	0	0	0	0	0	0	0	0

ii. Two hidden layers



For 5 5 neurons in two hidden layers

For layers [85,5,5,10]

For 10 neurons in two hidden layer

For layers [85,10,10,10]

For 15 15 neurons in two hidden layers

For layers [85,15,15,10]

For 20 20 neurons in two hidden layers

For layers [85,20,20,10]

For 25 25 neurons in two hidden layers

		Predicted									
		0	1	2	3	4	5	6	7	8	9
Actual	0	406218	35789	0	0	0	0	0	0	0	0
	1	122178	290173	0	0	0	0	0	0	0	0
	2	3414	34651	0	0	0	0	0	0	0	0
	3	543	9676	0	0	0	0	0	0	0	0
	4	2442	1443	0	0	0	0	0	0	0	0
	5	1811	124	0	0	0	0	0	0	0	0
	6	16	1241	0	0	0	0	0	0	0	0
	7	20	210	0	0	0	0	0	0	0	0
	8	18	2	0	0	0	0	0	0	0	0
	9	3	0	0	0	0	0	0	0	0	0

e. Relu activation in hidden layers

Relu improved the accuracy to some extent but the key point is it was quite faster than sigmoid activation. The training time was comparatively low.

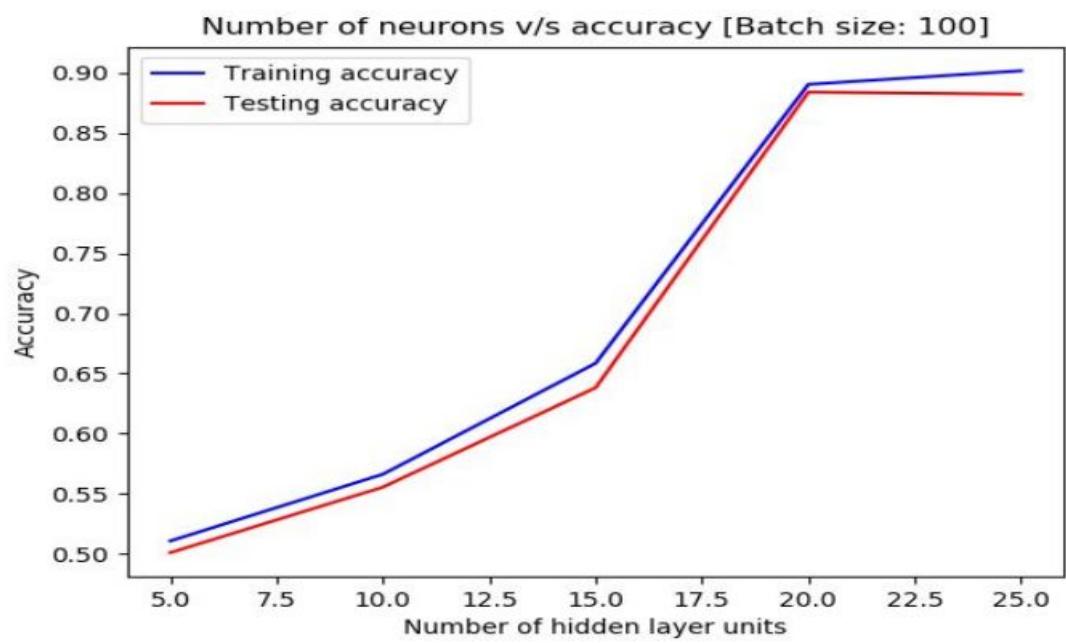
Stopping condition:

Error threshold: 10 ** -12

Epochs : 1000

(i) For 1 hidden layer

Neurons	Train Accuracy	Test Accuracy	Training time(sec)
5	0.51067572	0.500934	690.53810
10	0.566213514	0.55549	705.665904
15	0.658536585	0.638176	2531.8523
20	0.89056377	0.884128	3896.766
25	0.901879248	0.882424	3231.8668



For 5 neurons in single layer

For layers [85,5,10] relu

For 10 neurons in single layer

For layers [85,10,10] relu

For 15 neurons in single layer

For layers [85,15,10] relu

For 20 neurons in single layer

For layers [85,20,10] relu

For 25 neurons in single layer

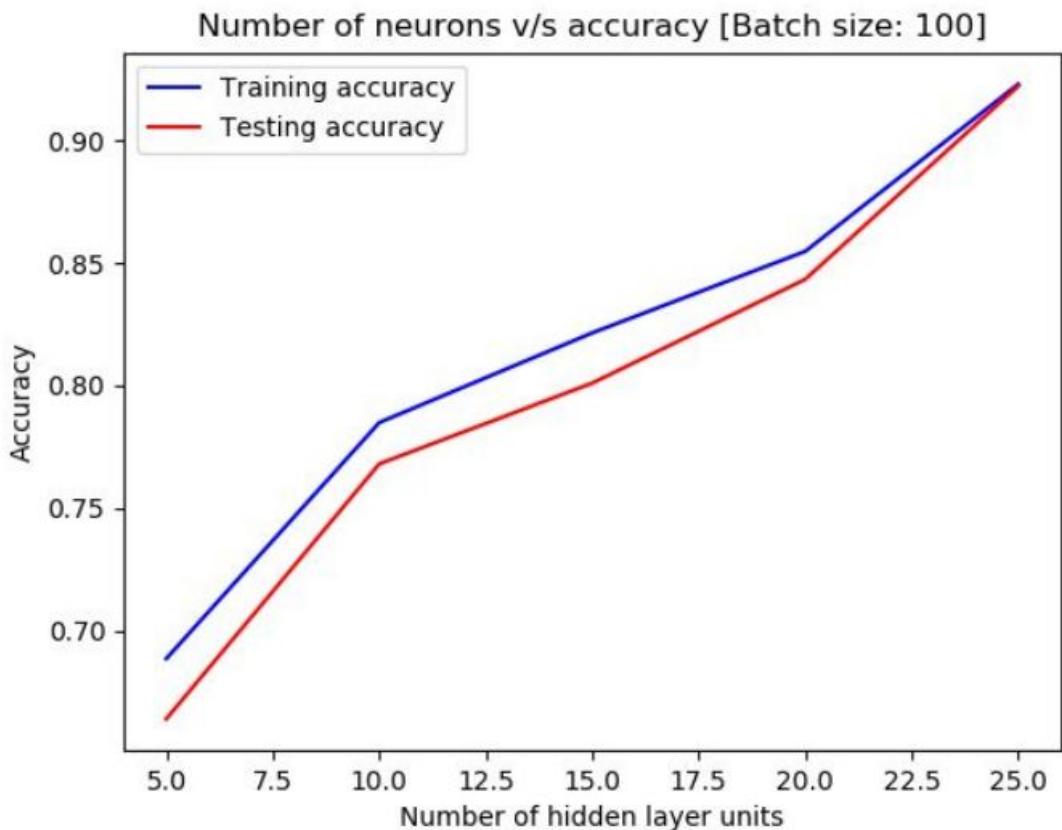
		For layers [85,25,10] relu									
		Predicted									
		0	1	2	3	4	5	6	7	8	9
0	-	488973	12236	0	0	0	0	0	0	0	0
1	-	29047	393451	0	0	0	0	0	0	0	0
2	-	367	47255	0	0	0	0	0	0	0	0
3	-	55	21066	0	0	0	0	0	0	0	0
4	-	3508	377	0	0	0	0	0	0	0	0
5	-	1944	52	0	0	0	0	0	0	0	0
6	-	5	1419	0	0	0	0	0	0	0	0
7	-	0	230	0	0	0	0	0	0	0	0
8	-	10	2	0	0	0	0	0	0	0	0
9	-	3	0	0	0	0	0	0	0	0	0

(ii) 2 hidden layers

Error threshold : 10^{-12}

Epochs: 1000

Neurons	Train Accuracy	Test Accuracy	Training time(sec)
5	0.68844622	0.663819	4697.125
10	0.784886045	0.768007	4717.97415
15	0.82151532	0.801006	5395.63291
20	0.854818072	0.84341	5641.15869
25	0.9231305	0.922361	6464.49075



For 5 5 neurons in two layers

For 10 neurons in two hidden layers

For layers [85,10,10,10] relu

For 15 15 neurons in two hidden layers

For layers [85,15,15,10] relu

For 20 20 neurons in two hidden layer

For 25 neurons in two hidden layer

