# Homework4

## Q1:

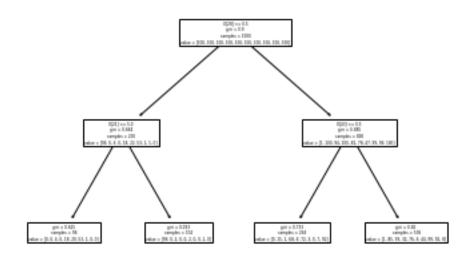
## tree for depth = 2

```
'train err 25 2': 0.712,
```

'test\_err\_25\_2': 0.7545909849749582,

'train\_err\_100\_2': 0.657,

'test\_err\_100\_2': 0.6878130217028381



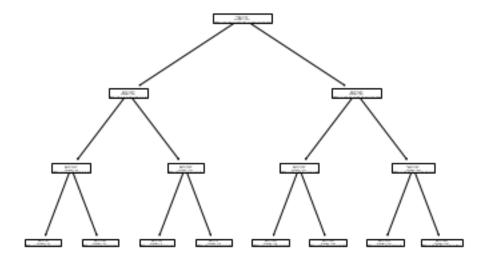
# tree for depth = 3

'train err 25 3': 0.536,

'test\_err\_25\_3': 0.6104618809126321,

'train\_err\_100\_3': 0.485,

'test\_err\_100\_3': 0.5375626043405677



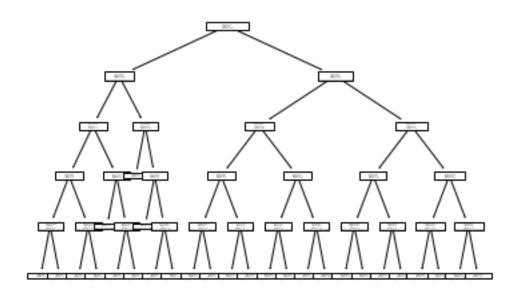
# tree for depth = 5

'train\_err\_25\_5': 0.1999999999999996,

'test\_err\_25\_5': 0.38842515303283254,

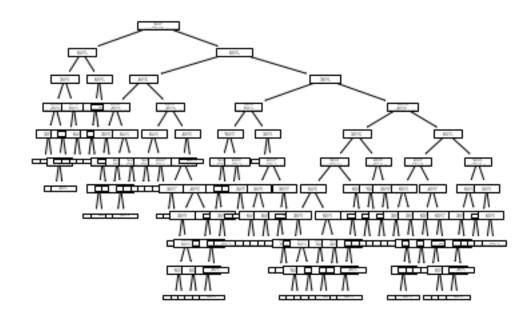
'train\_err\_100\_5': 0.1969999999999995,

'test\_err\_100\_5': 0.320534223706177



## tree for depth = 10

```
'train_err_25_10': 0.0,
```



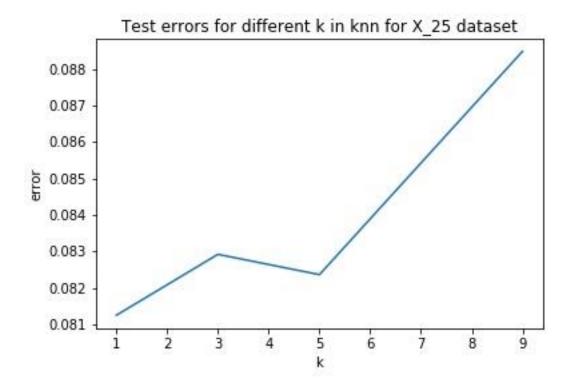
**Q3**: For X\_25 data set k-nearest neighbor algorithm with neighbors K = 1 performs better. However, as the number of instances increases in X\_100 dataset K = 3 performs better. In both the cases k = 9 does not perform well and error shoots up. Also, K = 1 performs

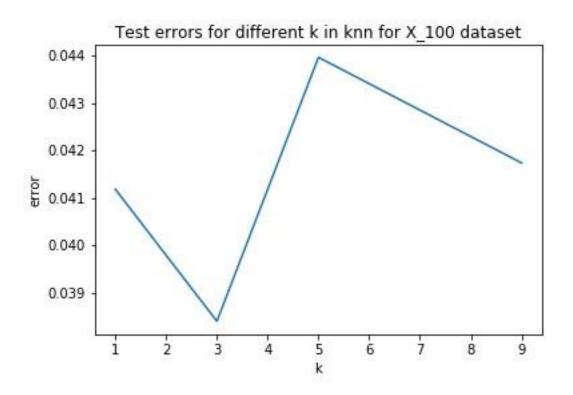
Please see the graph for test error with the models trained on  $X_25$  and  $X_100$  datasets respectively for k = [1,3,5,9]

<sup>&#</sup>x27;test\_err\_25\_10': 0.29771841958820255,

<sup>&#</sup>x27;train\_err\_100\_10': 0.00700000000000000,

<sup>&#</sup>x27;test\_err\_100\_10': 0.2309404563160824





**Q4:** Predicting pixels whose value is all zero gives 0 error and hence is the easiest to predict. Also, as we can see in the graph below, finding pixel 57 and 54 are with the minimal error which does not have all zero values.

