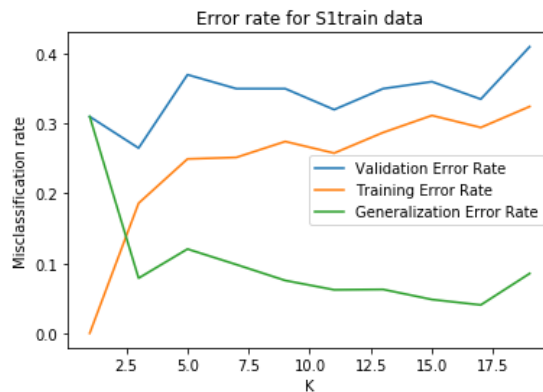


STAT 339 Spring 2020
Homework 1 Solutions
Garrett Robins

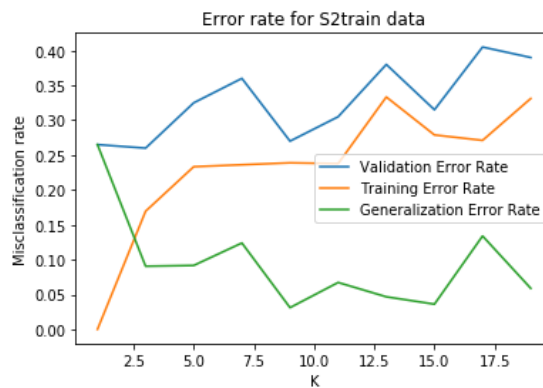
Honor Code: I affirm I have adhered to the Honor Code in this assignment - Garrett Robins

Part 1: KNN Implementation

Below is the graph I made for the KNN neighbors on the S1Train dataset. As you can see, the misclassification rate of the classifier was minimized on the validation set when $K = 3$. The misclassification rate of the training set was trivially minimized at $K = 1$ and non-trivially minimized at $K = 3$. Thus, as the graph has a general upward trend for larger values of K , I believe $K = 3$ to be the value that minimizes misclassification rate.



Below is the graph I made for the KNN neighbors on the S2Train dataset. As you can see, the misclassification rate of the classifier was minimized on the validation set when $K = 3$. The misclassification rate of the training set was trivially minimized at $K = 1$ and non-trivially minimized at $K = 3$. Thus, as the graph has a general upward trend for larger values of K , I believe $K = 3$ to be the value that minimizes misclassification rate.



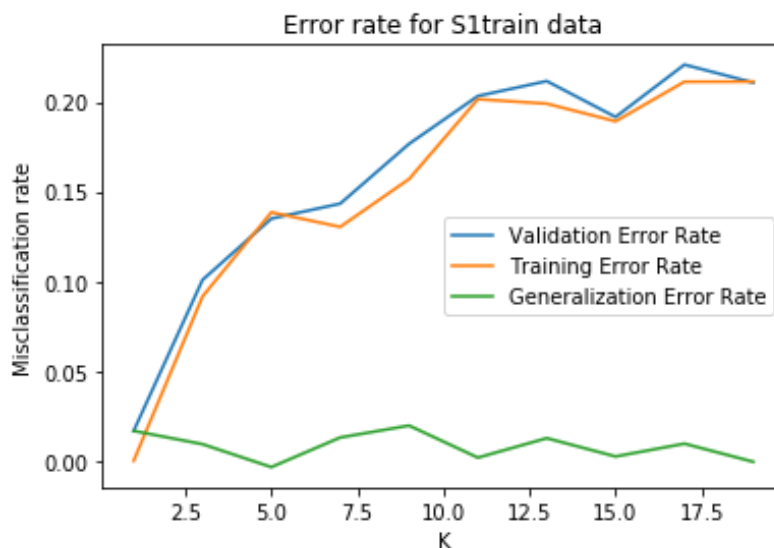
From the sealed test sets, we should be able to find generalized performance for our $K = 3$ classifiers. In fact, when we train a KNN algorithm, with $K = 3$, with the training datasets, we see that the test datasets have these errors:

- a. When the model is trained with S1train, the S1test dataset has a generalization error of **.267**
- b. When the model is trained with S2train, the S2test dataset has a generalization error of **.295**

Although both of these data distributions had minimized classification error at $K = 3$, I don't know if I believe this to be true in general. It would seem that a more complex distribution would require more neighbors as reference points to minimize misclassification rate. To that end, it remains to be seen whether a more complex or simpler distribution leads to a higher K .

Part 2: KNN on Images

Find below the graph of KNN neighbors on the image data given. Notice that this graph has minimized error for both the training and validation sets when $K = 1$. Although I am hesitant to use $K = 1$, as it has trivially 0 training error, the benefits seem to outweigh the costs and thus $K = 1$ is the optimal choice for K . (Note: the graph's title is wrong)



When running a KNN algorithm trained on the training data on the test set, if we set $K = 1$ we get a **.0137** misclassification rate. This is quite small and makes sense. In 28 dimensional data, you don't need many neighbors to confirm the type of image you're seeing.