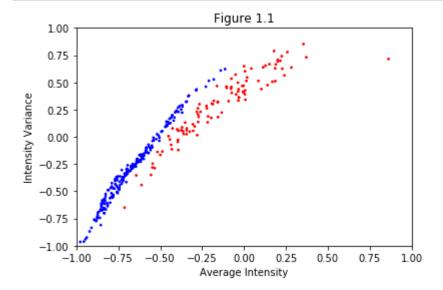
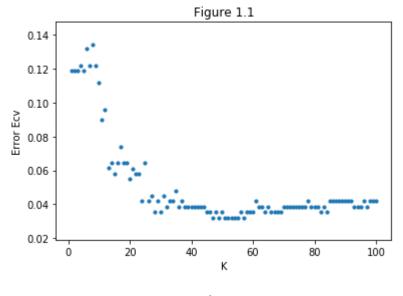
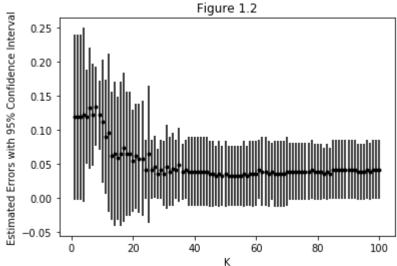
```
In [1]: #From the console, run the following
         #pip install numpy
         #pip install scipy
         #pip install scikit-learn
         #pip install matplotlib
         # Import required packages here (after they are installed)
         import numpy as np
         from sklearn.neighbors import KNeighborsClassifier
         import matplotlib.pyplot as mp
         from pylab import show
         from sklearn.model_selection import cross_val_score, cross_val_predict
         from sklearn.ensemble import AdaBoostClassifier
         from statistics import mean, stdev, median, mode
In [14]: def minimum(x,y):
             min = np.argmin(y)
             return x[min]
In [2]: # Load data. csv file should be in the same folder as the notebook for this to
         work, otherwise
         # give data path.
         data = np.loadtxt("data.csv")
In [3]: #shuffle the data and select training and test data
         np.random.seed(100)
         np.random.shuffle(data)
         features = []
         digits = []
         for row in data:
             #import the data and select only the 1's and 5's
             if(row[0] == 1 or row[0] == 5):
                 features.append(row[1:])
                 digits.append(str(row[0]))
         #Select the proportion of data to use for training.
         #Notice that we have set aside 80% of the data for testing
         numTrain = int(len(features)*.2)
         trainFeatures = features[:numTrain]
         testFeatures = features[numTrain:]
         trainDigits = digits[:numTrain]
         testDigits = digits[numTrain:]
```

```
In [4]: #Convert the 256D data (trainFeatures) to 2D data
        #We need X and Y for plotting and simpleTrain for building the model.
        #They contain the same points in a different arrangement
        X = []
        Y = []
        simpleTrain = []
        #Colors will be passed to the graphing library to color the points.
        #1's are blue: "b" and 5's are red: "r"
        colors = []
        for index in range(len(trainFeatures)):
            #produce the 2D dataset for graphing/training and scale the data so it is
         in the [-1,1] square
            xNew = 2*np.average(trainFeatures[index])+.75
            yNew = 3*np.var(trainFeatures[index])-1.5
            X.append(xNew)
            Y.append(yNew)
            simpleTrain.append([xNew,yNew])
            #trainDigits will still be the value we try to classify. Here it is the st
        ring "1.0" or "5.0"
            if(trainDigits[index]=="1.0"):
                 colors.append("b")
            else:
                 colors.append("r")
        #plot the data points
        ### https://matplotlib.org/api/_as_gen/matplotlib.pyplot.scatter.html
        mp.scatter(X,Y,s=3,c=colors)
        #specify the axes
        mp.xlim(-1,1)
        mp.xlabel("Average Intensity")
        mp.ylim(-1,1)
        mp.ylabel("Intensity Variance")
        mp.title("Figure 1.1")
        #display the current graph
        show()
```



```
In [19]: # USING 2D dimensional data
         X = []
         y = []
         z = []
         p = []
         m = []
         std =[]
         for i in range(1,101):
             #print(i)
             model = AdaBoostClassifier(n_estimators = i)
             #model2.predict(testFeatures)
             cvs = cross_val_score(model, simpleTrain, trainDigits, cv = 10, scoring='a
         ccuracy')
             err = 1-cvs
             evsm = err.mean()
             temp = stdev(err)
             temp2 = evsm + temp
             m.append(evsm)
             std.append(2*temp)
             p.append(temp2)
             x.append(i)
             y.append(evsm)
             z.append([x,evsm])
         # print(len(x))
         # print(len(y))
         # print(count)
         mp.scatter(x,y, s=10)
         mp.xlabel("K")
         mp.ylabel("Error Ecv")
         mp.title("Figure 1.1")
         mp.show()
         mp.errorbar(x, m, yerr=std, fmt='.k');
         mp.xlabel("K")
         mp.ylabel("Estimated Errors with 95% Confidence Interval")
         mp.title("Figure 1.2")
         show()
```





```
In [23]: b = minimum(x,m)
b
```

Out[23]: 47

Considering 95 % confident Interval

```
In [22]: b = minimum(x,p)
b
Out[22]: 28
```

Considering error mean and 95 % confidence interval, The optimal number of estimators is 28