Question 1

Training Error P=3: 0.01349862258953168 Testing Error P=3: 0.051451187335092345

Training Error P=4: 0.013223140495867768 Testing Error P=4: 0.03825857519788918

Training Error P=5: 0.01046831955922865 Testing Error P=5: 0.07387862796833773

Question 2:

Training Error P=3: 0.011294765840220386 Testing Error P=3: 0.05804749340369393

Training Error P=4: 0.011019283746556474 Testing Error P=4: 0.03562005277044855

Training Error P=5: 0.01046831955922865 Testing Error P=5: 0.07387862796833773

Question 3:

Top 2 Substrings: [('TAGQE', 3573), ('DTAGQ', 5065)]

```
In [1]: import pandas as pd
        import numpy as np
In [2]: # Read in data
        training data = pd.read csv('./data/pa4train.txt', sep=" ", header=None)
        testing_data = pd.read_csv('./data/pa4test.txt', sep=" ", header=None)
        training data.head()
        print(training data.shape)
        (3630, 2)
In [3]: # Function for calculating prediction error
        def calc_error(Y_pred, Y_label):
            # Calculate Error Rate for Predicted Labels
            error = [0 for x,y in zip(Y_pred,Y_label) if x != y]
            error rate = len(error)/len(Y pred)
            return error_rate
        # Function for computing the number of common substrings of length P
        def common substrings(S,T,p, sub dict):
            common substrings = {}
            T \text{ string} = T[0]
            T label = T[1]
            for i in range((len(S) - p + 1)):
                v = S[i:(i+p)]
                v = str(v)
                if v in T string:
                    occurences T = T string.count(v)
                    occurences S = S.count(v)
                     if v in sub dict.keys():
                         sub dict[v] += (T label)*(occurences T*occurences S)
                     else:
                         sub dict[v] = (T label)*(occurences T*occurences S)
                     if T string not in common substrings.keys():
                         common substrings[v] = occurences T*occurences S
            return sum(common substrings.values())
        # Function for computing the number of common substrings of length P
        def modified substrings(S,T,p):
            common substrings = {}
            for i in range((len(S) - p + 1)):
                v = S[i:(i+p)]
                v = str(v)
                if v in T:
                     if T not in common substrings.keys():
                         common substrings[v] = 1
            return sum(common substrings.values())
```

```
In [4]: class kernelPerceptron:
             # Constructor
             def __init__(self):
                 self.w = []
                 self.substrings = {}
             # Returns predicted label for sample data point
             # W is a list where each element is a tuple (Word, label)
             def kernel(self, sample point, p):
                 num\_common = \{-1: 0, 1: 0\}
                 for pair in self.w:
                     label = pair[1]
                     common_sub = common_substrings(sample_point[0], pair, p, sel
        f.substrings)
                     num common[label] += common sub
                 label = 1
                 if num common[-1] > \text{num common}[1]:
                     label = -1
                 return label
             # Modified version of string kernel
             def modifiedKernel(self, sample point, p):
                 num\_common = \{-1: 0, 1: 0\}
                 for pair in self.w:
                     string = pair[0]
                     label = pair[1]
                     common sub = modified substrings(sample point[0], string, p)
                     num common[label] += common sub
                 label = 1
                 if num common[-1] > \text{num common}[1]:
                     label = -1
                 return label
             # train perceptron and modify w parameter
             def train(self, kernel type, training data, p):
                 self.w = []
                 label = training data.shape[1] - 1
                 for index, row in training data.iterrows():
                     true label = row[label]
                     if kernel type == 'modified':
                         kernel label = self.modifiedKernel(row ,p)
                         kernel label = self.kernel(row ,p)
                     if kernel label*true label <= 0:</pre>
                         # misclassified string, modify w
                         self.w.append((row[0], true label))
             # Make predictions for each data point in testing data
             def predict(self, kernel_type, test_data, p):
                 # Predict label for each data point
                 predictions = []
                 for index, row in test data.iterrows():
                     if kernel type == 'modified':
                         predictions.append(self.modifiedKernel(row, p))
                     else:
```

```
predictions.append(self.kernel(row, p))
return predictions
```

Question 1

```
In [5]: # Tests
         clf = kernelPerceptron()
         clf.train('normal', training data, 2)
         y pred_train = clf.predict('normal',training_data, 2)
         y_true_train = training_data[(training_data.shape[1] - 1)]
         y_pred_test = clf.predict('normal',testing_data, 2)
         y_true_test = testing_data[(testing_data.shape[1] - 1)]
         print("Training Error P=2: ",calc_error(y_pred_train, y_true_train))
         print("Testing Error P=2: ",calc_error(y pred_test, y true_test))
         Training Error P=3: 0.07024793388429752
         Testing Error P=3: 0.0870712401055409
In [14]: # Tests
         clf p 3 = kernelPerceptron()
         clf p 3.train('normal', training data, 3)
         y pred train = clf p 3.predict('normal', training data, 3)
         y true train = training data[(training data.shape[1] - 1)]
         y_pred_test = clf_p_3.predict('normal',testing_data, 3)
         y_true_test = testing_data[(testing_data.shape[1] - 1)]
         print("Training Error P=3: ",calc error(y pred train, y true train))
         print("Testing Error P=3: ",calc_error(y_pred_test, y_true_test))
         Training Error P=3: 0.01349862258953168
         Testing Error P=3: 0.051451187335092345
 In [7]: | clf p 4 = kernelPerceptron()
         clf_p_4.train('normal',training_data, 4)
         y pred train = clf p 4.predict('normal', training data, 4)
         y_true_train = training_data[(training_data.shape[1] - 1)]
         y pred test = clf p 4.predict('normal', testing data, 4)
         y_true_test = testing_data[(testing_data.shape[1] - 1)]
         print("Training Error P=4: ",calc_error(y_pred_train, y_true_train))
         print("Testing Error P=4: ",calc error(y pred test, y true test))
         Training Error P=4: 0.013223140495867768
         Testing Error P=4: 0.03825857519788918
 In [8]: | clf_p_5 = kernelPerceptron()
         clf_p_5.train('normal',training_data, 5)
         y_pred_train = clf_p_5.predict('normal',training_data, 5)
         y true train = training data[(training data.shape[1] - 1)]
         y_pred_test = clf_p_5.predict('normal',testing_data, 5)
         y true test = testing data[(testing data.shape[1] - 1)]
         print("Training Error P=5: ",calc_error(y_pred_train, y_true_train))
         print("Testing Error P=5: ",calc_error(y_pred_test, y_true_test))
         Training Error P=5: 0.01046831955922865
         Testing Error P=5: 0.07387862796833773
```

Question 2

```
In [9]: # Tests
         clf = kernelPerceptron()
         clf.train('modified',training data, 3)
         y_pred_train = clf.predict('modified',training_data, 3)
         y_true_train = training_data[(training_data.shape[1] - 1)]
         y pred test = clf.predict('modified',testing data, 3)
         y true test = testing data[(testing data.shape[1] - 1)]
         print("Training Error P=3: ",calc error(y pred train, y true train))
         print("Testing Error P=3: ",calc_error(y_pred_test, y_true_test))
         Training Error P=3: 0.011294765840220386
         Testing Error P=3: 0.05804749340369393
In [10]: # Tests
         clf = kernelPerceptron()
         clf.train('modified',training_data, 4)
         y pred train = clf.predict('modified',training data, 4)
         y true train = training data[(training data.shape[1] - 1)]
         y_pred_test = clf.predict('modified',testing_data, 4)
         y true test = testing data[(testing data.shape[1] - 1)]
         print("Training Error P=4: ",calc error(y pred train, y true train))
         print("Testing Error P=4: ",calc_error(y_pred_test, y_true_test))
         Training Error P=4: 0.011019283746556474
         Testing Error P=4: 0.03562005277044855
In [11]: # Tests
         clf = kernelPerceptron()
         clf.train('modified',training data, 5)
         y pred train = clf.predict('modified',training data, 5)
         y true train = training data[(training data.shape[1] - 1)]
         y pred test = clf.predict('modified',testing data, 5)
         y true test = testing data[(testing data.shape[1] - 1)]
         print("Training Error P=5: ",calc error(y pred train, y true train))
         print("Testing Error P=5: ",calc_error(y_pred_test, y_true_test))
         Training Error P=5: 0.01046831955922865
         Testing Error P=5: 0.07387862796833773
```

Question 3

```
In [13]: sorted_sub = sorted(clf_p_5.substrings.items(), key=lambda x: x[1])
    print("Top 2 Substrings: ",sorted_sub[-2:])

Top 2 Substrings: [('TAGQE', 3573), ('DTAGQ', 5065)]
```