Question 1

```
In [17]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

```
In [18]: col_names = ['sepal-length','sepal-width','petal-length','petal-width','Class']
    df = pd.read_csv('iris.csv',names = col_names)
    df
```

Out[18]:

	sepal-length	sepal-width	petal-length	petal-width	Class
0	5.1	3.5	1.4	0.2	Iris-setosa
1	4.9	3.0	1.4	0.2	Iris-setosa
2	4.7	3.2	1.3	0.2	Iris-setosa
3	4.6	3.1	1.5	0.2	Iris-setosa
4	5.0	3.6	1.4	0.2	Iris-setosa
145	6.7	3.0	5.2	2.3	Iris-virginica
146	6.3	2.5	5.0	1.9	Iris-virginica
147	6.5	3.0	5.2	2.0	Iris-virginica
148	6.2	3.4	5.4	2.3	Iris-virginica
149	5.9	3.0	5.1	1.8	Iris-virginica

150 rows × 5 columns

```
# calculate the Euclidean distance between two vectors
         def euclidean_distance(row1, row2):
             distance = 0.0
             for i in range(len(row1)-1):
                 distance += (row1[i] - row2[i])**2
             return sqrt(distance)
         # Test distance function
         dataset = [[5.1,3.5,1.4,0.2],
         [4.9,3.0,1.4,0.2],
         [4.7,3.2,1.3,0.2],
         [4.6,3.1,1.5,0.2],
         [5.0,3.6,1.4,0.2],
         [5.4,3.9,1.7,0.4],
         [4.6,3.4,1.4,0.3],
         [5.0,3.4,1.5,0.2],
         [4.4, 2.9, 1.4, 0.2],
         [4.9,3.1,1.5,0.1]]
         row0 = dataset[0]
         for row in dataset:
             distance = euclidean distance(row0, row)
             print(distance)
         0.0
         0.5385164807134502
         0.509901951359278
         0.648074069840786
         0.1414213562373093
         0.5830951894845303
         0.5099019513592785
         0.17320508075688762
         0.9219544457292882
         0.4582575694955836
In [20]:
         # Locate the most similar neighbors
         def get_neighbors(train, test_row, num_neighbors):
             distances = list()
             for train_row in train:
                 dist = euclidean distance(test row, train row)
                 distances.append((train_row, dist))
             distances.sort(key=lambda tup: tup[1])
             neighbors = list()
             for i in range(num neighbors):
                 neighbors.append(distances[i][0])
             return neighbors
         neighbors = get_neighbors(dataset, dataset[0], 3)
         for neighbor in neighbors:
             print(neighbor)
         [5.1, 3.5, 1.4, 0.2]
         [5.0, 3.6, 1.4, 0.2]
         [5.0, 3.4, 1.5, 0.2]
```

In [19]: **from** math **import** sqrt

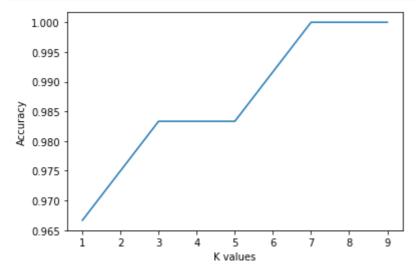
```
In [21]: # Make a classification prediction with neighbors
def predict_classification(train, test_row, num_neighbors):
    neighbors = get_neighbors(train, test_row, num_neighbors)
    output_values = [row[-1] for row in neighbors]
    prediction = max(set(output_values), key=output_values.count)
    return prediction

prediction = predict_classification(dataset, dataset[0], 3)
print('Expected %d, Got %d.' % (dataset[0][-1], prediction))
```

Expected 0, Got 0.

```
Question 2
 In [22]: from sklearn.model selection import train test split
          X = df[['sepal-length','sepal-width','petal-length','petal-width']]
          y = df[['Class']]
          X_train, X_test, y_train, y_test = train_test_split(X,y, test_size=0.4)
In [111]: from sklearn import metrics
          from sklearn.neighbors import KNeighborsClassifier
          k_range = range(1,10,2)
          scores=[]
          for k in k range:
              knn = KNeighborsClassifier(n neighbors = k)
              knn.fit(X_train, y_train)
              y_pred = knn.predict(X_test)
              scores.append(metrics.accuracy_score(y_test, y_pred))
          print(scores)
          [0.966666666666667, 0.983333333333333, 0.98333333333333, 1.0, 1.0]
          C:\Users\Nupur goel\anaconda3\lib\site-packages\sklearn\neighbors\_classification.py:1
          79: DataConversionWarning: A column-vector y was passed when a 1d array was expected.
          Please change the shape of y to (n_samples,), for example using ravel().
            return self._fit(X, y)
          C:\Users\Nupur goel\anaconda3\lib\site-packages\sklearn\neighbors\ classification.py:1
          79: DataConversionWarning: A column-vector y was passed when a 1d array was expected.
          Please change the shape of y to (n_samples,), for example using ravel().
            return self._fit(X, y)
          C:\Users\Nupur goel\anaconda3\lib\site-packages\sklearn\neighbors\_classification.py:1
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          Please change the shape of y to (n samples,), for example using ravel().
            return self._fit(X, y)
          C:\Users\Nupur goel\anaconda3\lib\site-packages\sklearn\neighbors\_classification.py:1
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          C:\Users\Nupur goel\anaconda3\lib\site-packages\sklearn\neighbors\_classification.py:1
          79: DataConversionWarning: A column-vector y was passed when a 1d array was expected.
          Please change the shape of y to (n_samples,), for example using ravel().
            return self._fit(X, y)
```

```
In [112]: plt.plot(k_range, scores)
    plt.xlabel('K values')
    plt.ylabel('Accuracy')
    plt.show()
```



In []: #This graph shows that we are getting accuracy of 0.983 when k is between 3 to 5 but whe #gives the best accuracy so knee point must be floor then k must be 7

Question 3

```
In [23]: from sklearn.model_selection import train_test_split
         from sklearn import metrics
         from sklearn.metrics import precision recall curve
         from sklearn.metrics import average_precision_score
         from sklearn.naive_bayes import GaussianNB
         df = pd.read_csv('wine.csv')
         X = df.drop('Proline', axis=1)
         y = df['Proline']
         Accuracy = []
         prec_mat = []
         split_size = [0.5, 0.4, 0.3, 0.2, 0.1]
         for i in range(0,5):
             X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = split_size[i])
             gnb = GaussianNB()
             gnb.fit(X_train, y_train)
             y_pred = gnb.predict(X_test)
             Accuracy.append(metrics.accuracy_score(y_test, y_pred)*100)
             y true = y test.values
             y_true = y_true.reshape(-1,1)
             prec_mat.insert(i,metrics.classification_report(y_true, y_pred, digits = 3))
```

_warn_prf(average, modifier, msg_start, len(result))
C:\Users\Nupur goel\anaconda3\lib\site-packages\sklearn\metrics_classification.py:
1245: UndefinedMetricWarning: Precision and F-score are ill-defined and being set t
o 0.0 in labels with no predicted samples. Use `zero_division` parameter to control
this behavior.

_warn_prf(average, modifier, msg_start, len(result))

C:\Users\Nupur goel\anaconda3\lib\site-packages\sklearn\metrics_classification.py:
1245: UndefinedMetricWarning: Recall and F-score are ill-defined and being set to
0.0 in labels with no true samples. Use `zero_division` parameter to control this b
ehavior.

_warn_prf(average, modifier, msg_start, len(result))

C:\Users\Nupur goel\anaconda3\lib\site-packages\sklearn\metrics_classification.py: 1245: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

warn prf(average, modifier, msg start, len(result))

C:\Users\Nupur goel\anaconda3\lib\site-packages\sklearn\metrics_classification.py:
1245: UndefinedMetricWarning: Recall and F-score are ill-defined and being set to
0.0 in labels with no true samples. Use `zero_division` parameter to control this behavior.

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In [25]:
          plt.show(prec_mat[i])
```

0.000

1

Question 4

In [29]: **for** i **in** range(0,5):

print(prec_mat[i])

0.000

0.000

450

```
In [28]: import pandas as pd
    df = pd.read_csv('train.csv', delimiter =',')
    df
```

Out[28]:	Phraseld		Sentenceld	Phrase	Sentiment
	0 1		1	A series of escapades demonstrating the adage	1
	1 2		1	A series of escapades demonstrating the adage	2
	2	3	1	A series	2
	3	4	1	Α	2
	4	5	1	series	2
	156055	156056	8544	Hearst 's	2
	156056	156057	8544	forced avuncular chortles	1
	156057	156058	8544	avuncular chortles	3
	156058	156059	8544	avuncular	2
	156059	156060	8544	chortles	2

156060 rows × 4 columns

```
In [29]: df.head()
Out[29]:
             Phraseld Sentenceld
                                                               Phrase Sentiment
          0
                   1
                             1 A series of escapades demonstrating the adage ...
          1
                   2
                             1 A series of escapades demonstrating the adage ...
                                                                             2
          2
                   3
                                                                             2
                                                              A series
                                                                             2
          3
                   4
                             1
                                                                   Α
          4
                   5
                             1
                                                                             2
                                                                series
In [31]: from sklearn.feature_extraction.text import CountVectorizer
         from nltk.tokenize import RegexpTokenizer
         #tokenizer to remove unwanted elements from out data like symbols and numbers
         token = RegexpTokenizer(r'[a-zA-Z0-9]+')
         cv = CountVectorizer(lowercase=True, stop_words='english', ngram_range = (1,1), tokenizer
         text counts= cv.fit transform(df['Phrase'])
 In [ ]: | Accuracy = []
         prec_mat = []
         split_size = [0.5, 0.4, 0.3, 0.2, 0.1]
         for i in range(0,5):
             X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = split_size[i])
             gnb = GaussianNB()
             gnb.fit(X_train, y_train)
             y_pred = gnb.predict(X_test)
             Accuracy.append(metrics.accuracy_score(y_test, y_pred)*100)
             y_true = y_test.values
             y_true = y_true.reshape(-1,1)
             prec_mat.insert(i,metrics.classification_report(y_true, y_pred, digits = 3))
In [32]: from sklearn.model_selection import train_test_split
         from sklearn.naive bayes import MultinomialNB
         #Import scikit-learn metrics module for accuracy calculation
         from sklearn import metrics
         # Model Generation Using Multinomial Naive Bayes
         split_size = [0.5, 0.4, 0.3, 0.2, 0.1]
         for i in range(0,5):
             X_train, X_test, y_train, y_test = train_test_split(
             text_counts, df['Sentiment'], test_size = split_size[i], random_state=0)
              clf = MultinomialNB().fit(X_train, y_train)
              predicted= clf.predict(X_test)
             print("MultinomialNB Accuracy:",metrics.accuracy_score(y_test, predicted))
         MultinomialNB Accuracy: 0.5956427015250545
         MultinomialNB Accuracy: 0.6035979751377675
```

MultinomialNB Accuracy: 0.6095305224486308 MultinomialNB Accuracy: 0.6106946046392413 MultinomialNB Accuracy: 0.6144431628860695