Machine Learning

Final Project Report

Team members: Shilp Patel, Divy Patel, Smit Patel, Jay Patel

Machine Learning Final Project

In this project, we had two primary tasks; one was to do classification with the given 5 train, test, label datasets and second one was to estimate the missing values in two additional datasets. After a little bit of research and execution, we have implemented all of the projects with the use of Java. To achieve the target for part-one, we used the KNN algorithm in order to predict the missing values in the data set. We utilized two dimensional arrays for our data sets to make our algorithm work smoothly and more essential. Our main goal to use the KNN algorithm was, it can help us very efficiently in handling the endless data and also the primary concern was in regarding the algorithm that it works very smoothly with multi-dimensions.

In the first part, the most important task was the classification part. In that, we had to check for the missing values in all of the datasets provided. Surprisingly, most of them had missing values, it was unpredicted. For this case, our KNN algorithm helped a lot in finding the nearest k-neighbors inside the missing values. We had many similar types of value, so we used Euclidean distance for our dataset. The KNN formula was used in order to get the results. After using that formula, we found the corresponding value for every nearest K neighbor. By calculating the mean value of the attribute between the closest K neighbors, except such missing values. After replacing the

present missed value with the equivalent mean value, the approach was used in this classification. To predict the labels for the test data, the same approach as above was used again. For instance, we used the KNN algorithm and the Euclidean distance to predict the values. The next thing we applied was that the decision set was calculated by dividing the distance by 1 and a request instance and obtaining for each possible mark the sum of all weights. The best prediction rating shall be the mark with the highest weight score.

In the second challenge, the Gene expression data are similar and reconsidered from the previous parts. The mean we found from the attribute value of the nearest neighbors was calculated using KNN. To get the absolute distances, we used Euclidean Distance Formula. The appropriate calculated mean was used to replace all the missing values in order to get the best results.

Code	Comments
	This part of the code is used to find missing value estimates using KNN algorithm. Training Dataset are gathered from the file. This part is used to find estimated missing values. It outputs the complete data into a new file. This part is used to find estimated missing values.
	This section concerns the classification of the test data. After performing the code above we now finally have 2d arrays named train_Data and test_Data & for the label we will use vectorized data named lab_data. Now we will find nearest k- neighbors for every case using classify().

```
    String [] micros = {"MissingData1.txt", "MissingData2.txt"};
    int [] microSamples = {14, 50};
    int [] microFeatures = {242, 758};


    fixed = false;
    for(int i = 0; i < micros.length; i++) {

       getData(micros[i], microSamples[i], microFeatures[i]);
       currentData = microData;
       fixData();
       writeData(micros[i]);
       System.out.println();
    }
    fixed = true;

}
```

Question 2: Missing Value Estimation
The following part is about Gene Expression Data. Datasets are gathered from the file.This part is used to find estimated missing values.It outputs the complete data into a new file.

```
System.out.println("Retrieving data from "+filename);
try {
    FileReader fileReader = new FileReader("./input/"+filename);
       BufferedReader bufferedReader = new BufferedReader(fileReader);

       String line;


       if(filename.contains("Label")) {

          labelData = new String[samples];

          for(int i = 0; i < labelData.length; i++) {
             line = bufferedReader.readLine().trim();
             labelData[i] = line;
          }
          System.out.println("Success!\n");

       }

       else if(filename.contains("Missing")) {
          System.out.println(filename+" Samples: "+samples+" Features: "+features);
          microData = new String[features][samples];

          String[] lineSplit;
          for(int i = 0; i < microData.length; i++) {
             line = bufferedReader.readLine().trim();
             lineSplit = line.split("\\s+|\\s*,\\s*");
              for (int j = 0; j < microData[i].length; j++) {
                   microData[i][j] = lineSplit[j];
              }
          }
```

Reads the training data collection and inserts a 2d string sequence called train_Data & test_Data. This codes shown are are used to for Labels and for Gene Data

```
          double d = Math.sqrt(microData.length) / 2;
          k = (int)d;
          bufferedReader.close();
          System.out.println("Success!\n");
       }


       else if(filename.contains("Train")) {

          System.out.println(filename+" Samples: "+samples+" Features: "+features);
          trainingData = new String[samples][features];

          String[] lineSplit;
          for(int i = 0; i < trainingData.length; i++) {
             line = bufferedReader.readLine().trim();
             lineSplit = line.split("\\s+|\\s*,\\s*");
              for (int j = 0; j < trainingData[i].length; j++) {
                   trainingData[i][j] = lineSplit[j];
              }
          }
```

The following formula is used to identify the specific K-value. Following code is used for training data.

```


          double d = Math.sqrt(trainingData.length) / 2;
          k = (int)d;
          bufferedReader.close();
          System.out.println("Success!\n");
       }

       else if(filename.contains("Test")) {


          testingData = new String[samples][features];

          String[] lineSplit;
          for(int i = 0; i < testingData.length; i++) {
             line = bufferedReader.readLine().trim();
             lineSplit = line.split("\\s+|\\s*,\\s*");
              for (int j = 0; j < testingData[i].length; j++) {
                   testingData[i][j] = lineSplit[j];
              }
          }

          double d = Math.sqrt(testingData.length) / 2;
          k = (int)d;
          bufferedReader.close();
          System.out.println("Success!\n");
       }
    }
    catch(IOException io) {
       System.out.println("File: "+filename+" not found.");
       System.exit(1);
    }
}
```

The following formula is used to identify the specific K-value. For testing data. The following formula is used to identify the specific K-value

```
177  public static void fixData() {
178
179     System.out.println("Examining data for missing values..");
180     for(int i = 0; i < currentData.length; i++) {
181        int missing = 0;
182        System.out.print("Line "+(i+1)+"... ");
183        for(int j = 0; j < currentData[i].length; j++) {
184           if(currentData[i][j].contains("+99")) {
185              missing++;
186              String[] query = currentData[i];
187
188
189              Distance[] kNearest = myKnn(i, query);
190
191
192              double sum = 0;
193              for(int n = 0; n < k; n++) {
194                 Double val = Double.parseDouble(currentData[kNearest[n].index][j]);
195                 sum += val;
196              }
197
198              String mean = Double.toString(sum / k);
199
200              currentData[i][j] = mean;
201           }
202        }
203        System.out.println("Missing values fixed: "+missing);
204
205     }
206     System.out.println("All Missing Values fixed!!");
207  }
208
209  public static Distance[] myKnn(int index, String[] query) {
210
211     Distance[] distances;
212
213     if(fixed == false) {
214
```

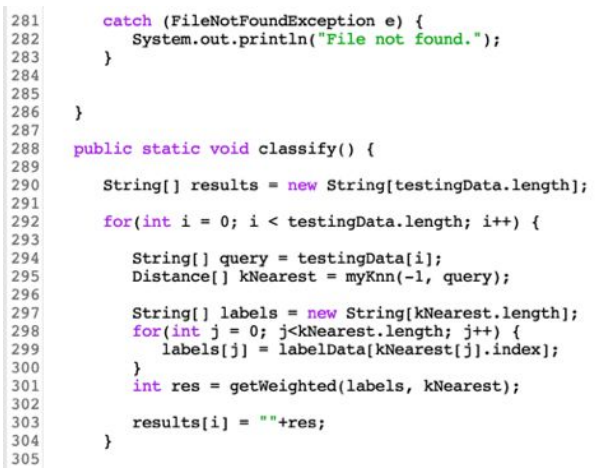
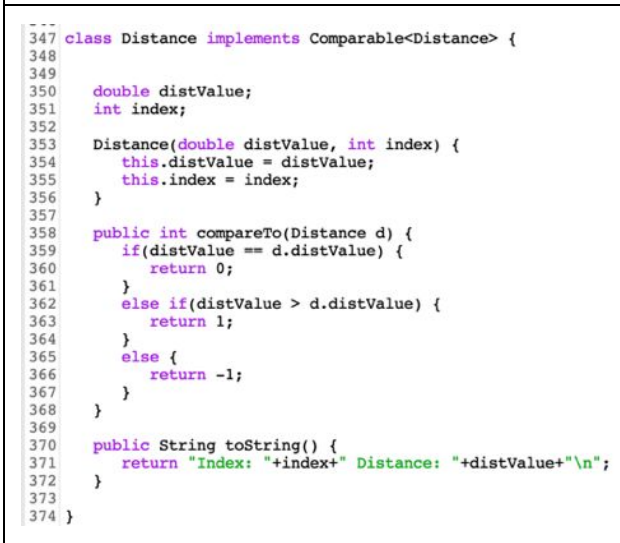
Queries for missing attributes, gets the nearest K-neighbors for data, and substitutes the missing value with the mean value for that attribute. Outputs the k nearest value.

```
216        distances = new Distance[currentData.length-1];
217        int j = 0;
218
219        for(int i = 0; i < currentData.length; i++) {
220              String[] instance = currentData[i];
221              Distance d = new Distance(getDistance(instance, query), i);
222              if(i != index) {
223                 distances[j] = d;
224                 j++;
225              }
226        }
227     }
228     else {
229        distances = new Distance[trainingData.length];
230        for(int i = 0; i < trainingData.length ; i++) {
231              String[] instance = trainingData[i];
232              Distance d = new Distance(getDistance(instance, query), i);
233              distances[i] = d;
234        }
235     }
236     Arrays.sort(distances);
237     Distance[] kNearest = Arrays.copyOfRange(distances, 0, k);
238     return kNearest;
239  }
```

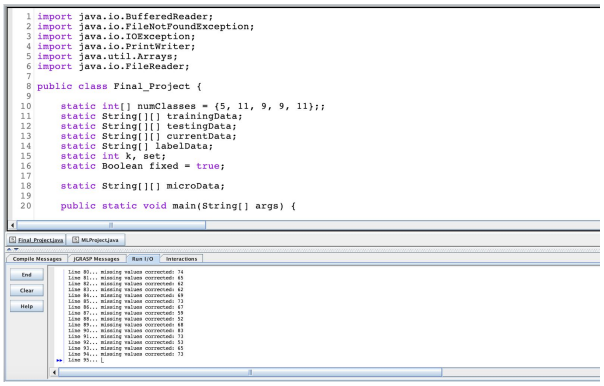
Finds the closest k-neighbor to an object with the missing attribute and returns a collection of 'Distance' objects made up of the distance value and index of each of the closest k instances.

```
241  public static double getDistance(String[] s1, String[] s2) {
242
243
244     double distance = 0;
245     for(int i = 0; i < s2.length ; i++) {
246        if(s1[i].contains("+99") || s2[i].contains("+99")) {
247           distance += 0;
248        }
249        else {
250           double d1 = Double.parseDouble(s1[i]);
251           double d2 = Double.parseDouble(s2[i]);
252           double d = Math.pow((d2 - d1),2);
253           distance += d;
254
255        }
256
257     }
258     return Math.sqrt(distance);
259  }
260
261  public static void writeData(String filename) {
262
263     String file;
264     if(filename.contains("Missing")) {
265        file = "./output/PatelNew"+filename;
266     }
267     else {
268        file = "./input/New"+filename;
269     }
270     try {
271        PrintWriter pw = new PrintWriter(file);
272        for(int k=0; k<currentData.length; k++) {
273              for(int l=0; l<currentData[k].length; l++) {
274                 pw.print(currentData[k][l]+"\t");
275              }
276              pw.print("\n");
277             }
278             pw.close();
279             System.out.println("Fixed data saved to: "+file);
280     }
```

Calculates the difference between two vectors for Euclidean while overlooking any other incomplete data values. Writes the complete data into a new file.

	This result is compared to every possible case in the testing data
	This part creates a new file with all the results.
	This part gives out the label values regarding the weights(high) of the nearest k-neighbor.
	The distance entity used in myKnn system includes the case index and the distance from the query-instance. This is made equivalent by the distance attribute for sorting.

Screenshots taken throughout the process of the code and the output of code is shown below:



```
import java.io.BufferedReader;
import java.io.FileNotFoundException;
import java.io.IOException;
import java.io.PrintWriter;
import java.util.Arrays;
import java.io.FileReader;

public class Final_Project {

    static int[] numClasses = {5, 11, 9, 9, 11};;
    static String[][] trainingData;
    static String[][] testingData;
    static String[][] currentData;
    static String[] labelData;
    static int k, set;
    static Boolean fixed = true;

    static String[][] microData;

    public static void main(String[] args) {
```