Exercise 4

Applications of Data Analysis

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1 Data preprocessing and creating distance matrix

Data was given in three csv-files. We read the input data in x, output data in y and coordinates in z. Also the input data is standardized and stored into stdX.

```
basepath = os.path.dirname(__file__)
inputpath = os.path.abspath(os.path.join(basepath, "Data4/INPUT.csv"))
outputpath = os.path.abspath(os.path.join(basepath, "Data4/OUTPUT.csv"))
coordinatespath = os.path.abspath(os.path.join(basepath,
    "Data4/COORDINATES.csv"))

x = np.genfromtxt(inputpath, delimiter=',')
y = np.genfromtxt(outputpath, delimiter=',')
z = np.genfromtxt(coordinatespath, delimiter=',')

xArr = np.asarray(x)
stdX = (xArr - xArr.mean()) / xArr.std()
```

Distances between every pair of points is stored into a distance matrix. M_{ij} is the euclidean distance between *i*th and *j*th point in data. If i=j, the M_{ij} is set to -1.

2 FINDING OUT VALUES IN DEADZONE

The Method *calculateDeadZone* finds the 10 nearest points (where the distance is larger or equal to 0.0) for every data point and sets the distance to those points as -1. Thus our implementation of leave-on-out cross-validation ignores those points (see below).

```
def calculateDeadZone(matrix):
    for i in range(len(matrix)):
        xAxis = matrix[i]
        for _ in range(10):
            minIndex = xAxis.index(min(filter(lambda x:x>=0.0, xAxis)))
            xAxis[minIndex] = -1.0

return matrix
```

3 CALCULATING 5-NEAREST-NEIGHBOR AND PREDICTING LABEL

The distance between two features is calculated only if the M_{ij} is larger or equals with 0. This limitation is done because we need to leave the test instance out of calculation and also we need to ignore features that are in the deadzone. The distance of a feature with itself and the distances to the features in the deadzone are set to -1. The method returns k-nearest neighbors. The value of k is given as an argument and in this case it is 5.

The predicted value for the test instance is the mean value of its neighbors' classes.

```
def chooseMajorityLabel(neighbors, k):
    predictedOutcome = []
    sumOfMod = 0.0
    for i in range(len(neighbors)):
        sumOfMod = sumOfMod + neighbors[i][1]
    predictedOutcome.append(sumOfMod/k)
    return predictedOutcome
```

4 Leave-one-out Cross-validation

Leave-one-out cross-validation – method gets the number of nearest neighbors to consider and a distance matrix as its arguments. The distance matrix consists of distances between every pair of points.

The method takes every instance from the training set one by one to use as a test set and calculates the instance's neighbors. The predicted label for the test instance is calculated based on its nearest neighbors and the label is added to a list of predictions made throughout the execution of the method.

After that the c-index is calculated based on pseudo-code we got in last exercise (Third Exercise: Prediction of metal ion content from multi-parameter data).

```
def LooCV(k, distanceMatrix):
    yPredictions = []
    for i in range(len(stdX)):
        neighbors = inferNeighbors(stdX, stdX[i], y, k, distanceMatrix[i])
        yPredictions.append(chooseMajorityLabel(neighbors,k))

cIndex = calculateCIndex(yPredictions, y)
    printCIndexes(cIndex)

return cIndex
```

5 SUMMING ALL THE METHODS

Firstly, the distance matrix is created. This matrix is used and modified during calculation. C-index is calculated for each of the dead zone radius "cases" using leave-one-out implementation

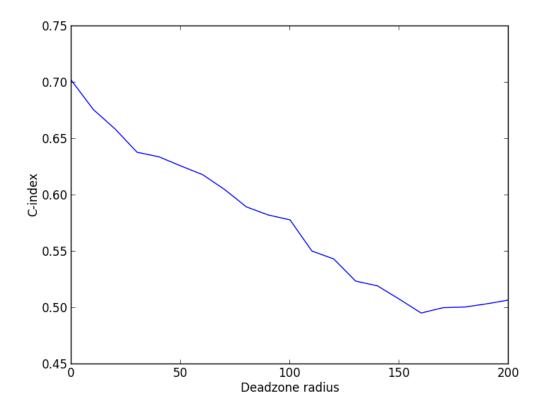
of cross validation (see above) together with the pattern recognition algorithmn k(5)-nearest-neighbors. After each dead zone radius case (0,10,20...,200) the distance matrix is modified by disabling the next 10 nearest points. In the first round it disables 1-10 nearest features, in second round 11-20 nearest features and so on.

```
def main():
    startTime = int(round(time.time() * 1000))
    distanceMatrix = calculateDistanceMatrix()
    cIndexes = []
    deadZoneValues = []
    for i in range(21):
        print 'Leave-one-out CV with deadzone radius ' + str(i * 10) + ':'
        cIndexes.append(LooCV(5, distanceMatrix))
        deadZoneValues.append(i * 10)
        distanceMatrix = calculateDeadZone(distanceMatrix)

    endTime = int(round(time.time() * 1000))
    print 'Running time: ' + str(endTime - startTime) + 'ms'

    plotCIndexVsDeadZone(cIndexes, deadZoneValues)
```

All the used deadzone radius and c-indexes are plotted:



And same data as plotted above:

Leave-one-out CV with deadzone radius 0: C-Index: 0.701855867534 Leave-one-out CV with deadzone radius 10: C-Index: 0.676058930829 Leave-one-out CV with deadzone radius 20: C-Index: 0.658731830851 Leave-one-out CV with deadzone radius 30: C-Index: 0.638168826969 Leave-one-out CV with deadzone radius 40: C-Index: 0.634122722163 Leave-one-out CV with deadzone radius 50: C-Index: 0.626040661936 Leave-one-out CV with deadzone radius 60: C-Index: 0.618301230945 Leave-one-out CV with deadzone radius 70: C-Index: 0.605187175657 Leave-one-out CV with deadzone radius 80: C-Index: 0.589694664503 Leave-one-out CV with deadzone radius 90: C-Index: 0.582504700215 Leave-one-out CV with deadzone radius 100: C-Index: 0.578135914964 Leave-one-out CV with deadzone radius 110: C-Index: 0.550505789349 Leave-one-out CV with deadzone radius 120: C-Index: 0.543447067108 Leave-one-out CV with deadzone radius 130: C-Index: 0.523801007659 Leave-one-out CV with deadzone radius 140: C-Index: 0.519612461485 Leave-one-out CV with deadzone radius 150: C-Index: 0.507752730185 Leave-one-out CV with deadzone radius 160: C-Index: 0.495467424674 Leave-one-out CV with deadzone radius 170: C-Index: 0.500218736745 Leave-one-out CV with deadzone radius 180: C-Index: 0.500782202599 Leave-one-out CV with deadzone radius 190: C-Index: 0.503624030384 Leave-one-out CV with deadzone radius 200: C-Index: 0.507006575402

```
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Task steps:
1. Implement a Leave-One-Out cross validation with deadzone radius R = 0, 10, 20,
..., 200. So you will do 21 <u>analyses</u> in total here.
    Use 5-nearest neighbor as the prediction method. Remember normalization.
2. Calculate the C-index value for each of the deadzone radius cases.
3. Plot the C-index vs. Deadzone radius in a graph to visualize, how the prediction
performance changes with the deadzone radius.
    Set Y-axis to be the C-index and X-axis to be Deadzone radius
4. Return your implementation and the graph in a written report.
. . .
import os
import operator
import scipy.spatial.distance as ssd
import numpy as np
import matplotlib.pyplot as pp
import time
if __name__ == '__main__':
    pass
basepath = os.path.dirname( file )
inputpath = os.path.abspath(os.path.join(basepath, "Data4/INPUT.csv"))
outputpath = os.path.abspath(os.path.join(basepath, "Data4/OUTPUT.csv"))
coordinatespath = os.path.abspath(os.path.join(basepath, "Data4/COORDINATES.csv"))
x = np.genfromtxt(inputpath, delimiter=',')
y = np.genfromtxt(outputpath, delimiter=',')
z = np.genfromtxt(coordinatespath, delimiter=',')
xArr = np.asarray(x)
stdX = (xArr - xArr.mean()) / xArr.std()
def calculateDistanceMatrix():
    distanceMatrix = []
    for i in range(len(z)):
        xAxis = []
        for j in range(len(z)):
            if i == j:
                xAxis.append(-1.0)
```

```
else:
                 xAxis.append(ssd.euclidean(z[i], z[j]))
        distanceMatrix.append(xAxis)
    return distanceMatrix
def calculateCIndex(predictions, labels):
    n = 0
    h sum = 0
    for i in range(len(labels)):
        t = labels[i]
        p = predictions[i]
        for j in range(i+1,len(labels)):
            nt = labels[j]
            np = predictions[j]
            if t != nt:
                n = n + 1
                 if (p < np \text{ and } t < nt) \text{ or } (p > np \text{ and } t > nt):
                     h_sum = h_sum + 1
                 elif (p < np \text{ and } t > nt) \text{ or } (p > np \text{ and } t < nt):
                     h_sum = h_sum + 0
                 elif (p == np):
                     h_sum = h_sum + 0.5
    if n == 0:
        return 0
    else:
        return h_sum/n
def LooCV(k, distanceMatrix):
    yPredictions = []
    for i in range(len(stdX)):
        neighbors = inferNeighbors(stdX, stdX[i], y, k, distanceMatrix[i])
        yPredictions.append(chooseMajorityLabel(neighbors,k))
    cIndex = calculateCIndex(yPredictions, y)
    printCIndexes(cIndex)
    return cIndex
def chooseMajorityLabel(neighbors, k):
    predictedOutcome = []
    sumOfMod = 0.0
    for i in range(len(neighbors)):
        sumOfMod = sumOfMod + neighbors[i][1]
    predictedOutcome.append(sumOfMod/k)
    return predictedOutcome
def inferNeighbors(trainSet, testInstance, labels, k, distRow):
    distances = []
    for x in range(len(trainSet)):
        if distRow[x] >= 0.0:
            distances.append((ssd.euclidean(trainSet[x], testInstance), labels[x]))
```

```
distances.sort(key=operator.itemgetter(0))
    return distances[0:k]
def printCIndexes(cIndex):
    print 'C-Index: {a}'.format(a=cIndex)
    print
def calculateDeadZone(matrix):
    for i in range(len(matrix)):
        xAxis = matrix[i]
        for _ in range(10):
            minIndex = xAxis.index(min(filter(lambda x:x>=0.0, xAxis)))
            xAxis[minIndex] = -1.0
    return matrix
def plotCIndexVsDeadZone(cIndexes, deadZoneValues):
    pp.ylabel('C-index')
    pp.xlabel('<u>Deadzone</u> radius')
    pp.plot(deadZoneValues, cIndexes)
    pp.show()
def main():
    startTime = int(round(time.time() * 1000))
    distanceMatrix = calculateDistanceMatrix()
    cIndexes = []
    deadZoneValues = []
    for i in range(21):
        print 'Leave-one-out CV with deadzone radius ' + str(i * 10) + ':'
        cIndexes.append(LooCV(5, distanceMatrix))
        deadZoneValues.append(i * 10)
        distanceMatrix = calculateDeadZone(distanceMatrix)
    endTime = int(round(time.time() * 1000))
    print 'Running time: ' + str(endTime - startTime) + 'ms'
    plotCIndexVsDeadZone(cIndexes, deadZoneValues)
main()
```