DS 501: STATISTICAL & MATHEMATICAL METHODS FOR DATA SCIENCE FALL 2019 INDIVIDUAL ASSIGNMENT 6

DUE: Wednesday, December 04 (no late assignment is accepted)

NOTE: Hard copy is due on Thursday 12:00 noon. No late submissions of soft or hard copy of the assignment will be accepted due to semester deadlines.

Reading

Tom Mitchel's Machine Learning, chapter 4

DEVELOP AN OCR SYSTEM USING A LINEAR PERCEPTRON

DATASET

Dataset is the same as for assignment 4. This data is a subset of OCR data taken from http://cmp.felk.cvut.cz/cmp/software/stprtool/index.html

View an image

You have 4 files: train2_5.txt, train2_5Labels.txt, test2_5.txt, test2_5Labels.txt.

Each row of train and test contains the image of a digit (either 2 or 5). Each image is a 16x16 image, stored in one row as a 256 dimensional vector. You can view any images as follows:

```
X = \text{read.table}(\text{`train2}\_5.\text{txt'})

r = X[1,] #first digit image, i.e., image in row 1

r = \text{data.matrix}(r)

\dim(r) = c(16,16)

image(r) #view image
```

Similarly you can view any image you like for any row of train and test matrices

ALGORITHM FOR STOCHASTIC GRADIENT DESCENT

Dimensions of the data matrix are mxn (m rows and n column)

```
    Add a column of ones to data matrix X (this will cater for bias unit).
    Initialize all weights randomly #(n+1 weights because of the column of ones)
```

```
3. Repeat each epoch as:

#i is a row index and j is a column #index

a. for i = 1 to m

i. compute \hat{y}_i = \sum_{j=0}^{\hat{n}} x_{ij} w_j #prediction for one row

ii. update the weights as:

#i is a row index and j is a column #index

# for each row

#prediction for one row

# each weight has to be updated. In # R it can be done without a loop.

# Just do matrix/vector math # j=0..n (note n+1 weights)

\Delta w_j \leftarrow \propto \Delta w_j + \eta(y_i - \hat{y}_i)(x_{ij}) \text{ #alpha} = \text{momentum},

#eata = learning rate

w_j \leftarrow w_j + \Delta w_j
```

MODEL BUILDING & EVALUATION

Training part

Write a function for training:

trainGradientDescent <- function(X,Y,learningRate,momentum)

The above function should return the weights when given the training data X, target values y and the learning rate and momentum. Don't forget to add a column of ones to X so that you can cater for the bias unit. Alternatively, you can keep a separate variable for bias.

Test part

Write a test function for getting predictions

testGradientDescent <- function (testX,regressionCoefficients)

This function should also not have any loops or iterations

Evaluation part

Read about the confusion matrix as for assignment 4 (YOU HAVE TO READ THIS AS IT IS PART OF THE COURSE):

https://en.wikipedia.org/wiki/Confusion_matrix

Write an evaluation function that makes a confusion matrix

Now compute the following:

BalancedAccuracyRate (BAC) = (TP/totalPositiveLabels+TN/totalNegativeLabels)/2

Main script: Bring it all together

Write a main script that:

- a. Reads the training data and builds a perceptron model. Next it gets predictions from the model using the training set as well as the test set.
- b. Find a way of mapping the OCR labels to the predictions. So for example if your prediction is 10, then how will you map it to a label?
- c. For the training data as well as the test data, make the confusion matrix for different values of lambda, as given in the report.

Simulations

You have to repeat for different values of learning rate and momentum and the number of iterations (parameters are given in the report).

TO SUBMIT

- 1. Source code and soft copy of report on slate
- 2. Hard copy of the report

NOTE

Plagiarism will not be tolerated and can result in an F grade in the course.

Plagiarism from the internet or for cheating amongst yourselves will NOT be tolerated