Support Vector Machines INF552

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Work Breakup

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Non Linear SVM and documentation

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Linear SVM and Standard Library Function

INTRODUCTION

Objectives

Implement Support vector machines for 2D data in linsep.txt and nonlinsep.txt.

The result should report the support vectors and the equation of the separation line for linear separation SVM. It should also report the support vectors and the kernel used for non-linearly separable data.

Functionality Linear Separable

- 1. Input
- 2. Solve for alpha using quadratic programmer
- 3. Get indices using alpha
- 4. Calculate weights
- 5. Calculate the bias value (intercept).
- 6. Get the equation of the line with y = mx+c where, m = -weights[0][0]/weights[0][1] and c = -intercept/weights[0][1]

Non Linear Separable

- 1. Input
- 2. Solve for alpha using quadratic programmer
- 3. Get indices from alpha
- 4. Calculate support vectors using indices

Design Data Structure

- 1. Input data is stored in Numpy arrays in both linear and non linear seperable data
- 2. **Alpha** is stored in Numpy array and so is the **indices** and weights.
- 3. Support vectors are stored in List in python

Procedures

- 1. Input
- 2. Calculate Alpha using the quadratic programming solver.
 - The **kernel** used for non-linear separable data is polynomial kernel

$$K(x,z) = (Z^T x + c)^n$$

- 3. Calculate indices using the value in alpha that are non zero.
- 4. Remove all zero values from alpha
- 5. Get the support vectors using the indices on X.

Comparison with Scikit Learn

- Machine Learning Library in Python

sklearn.svm.SVC and sklearn.svm.LinearSVC

LinearSVC is particularly meant for Linear kernels when dealing with linearly separable data, whereas SVC accepts a parameter called kernel where you can define the type of kernel transformation that you would like to apply for the non-linear separable data. Our algorithm is developed currently only to deal with hard classification problems. SVC and LinearSVC both consist of an argument 'C' that corresponds to the error term that needs to be included to allow certain misclassification of points. The different kinds of kernels available are 'linear', 'poly', 'rbf', 'sigmoid',

'precomputed'. If none is given, 'rbf' will be used. Also our implemented algorithm only works for binary classification whereas LinearSVC

also includes support for multi-class classification through one-vs-the-rest. SVC on the other hand works on one-against-one approach for multi- class classification. If there are n classes, then a combination of their classifiers are trained and then the 'density_function_shape' parameter is used in order to provide a consistent interface to all the classifiers.

Optimizations and Challenges

Optimization

Avoiding dot operator:

Calculating b we do not use any for loop and thus code is faster than using loop for every value

• Removing non-zero and getting there incides using flatnonzero of numpy is faster than iterating through a python list

Challenges

Understanding which kernel to choose. Figuring out computation using quadratic functions and using constraints while doing so. Understanding how to convert the data points using kernel function and computing the weights in the new dimension.

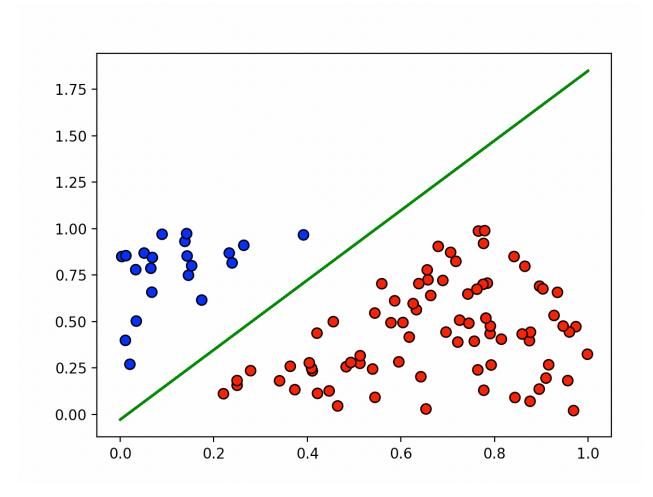
Usage Instruction

Use Scikit-learn, Python3 the code also requires cvxopt for solving quadratic equations

Results

```
(venv) aviral:HW-6 aviral$ python3 nonlinsep.py
     pcost
                  dcost
                               gap
                                      pres
                                             dres
 0: -4.0666e+01 -1.0206e+02
                               5e+02
                                      2e+01
                                              3e+00
                                             1e+00
 1: -1.5924e+02 -2.2789e+02
                               3e+02
                                      1e+01
 2: -2.9280e+02 -3.6244e+02
                               3e+02
                                      1e+01
                                             1e+00
                               4e+02
 3: -5.7710e+02 -6.0303e+02
                                      9e+00
                                             1e+00
 4: -1.2873e+03 -1.2409e+03
                               5e+02
                                      9e+00
                                             1e+00
 5: -1.2647e+03 -1.0924e+03
                               7e+02
                                             9e-01
                                      8e+00
 6: -6.9076e+02 -4.0802e+02
                               1e+03
                                      5e+00
                                             6e-01
 7: -1.8688e+02 -2.9779e+01
                               4e+02
                                      1e+00
                                             2e-01
 8: -3.4731e+00 -5.2038e-02
                               1e+01
                                      3e-02
                                             4e-03
 9: -3.5053e-02 -3.8447e-02
                               1e-01
                                      3e-04
                                             3e-05
10: -2.1413e-02 -2.7448e-02
                               6e-03
                                      3e-18
                                             2e-13
11: -2.6166e-02 -2.6328e-02
                               2e-04
                                      6e-18
                                             2e-13
12: -2.6293e-02 -2.6295e-02
                               2e-06
                                      5e-18
                                             2e-13
13: -2.6295e-02 -2.6295e-02
                               2e-08
                                      2e-18
                                             1e-13
Optimal solution found.
[[0.0125012]
 [0.0149904]
 [0.00664213]
 [0.00466215]
 [0.00740837]
 [0.00638511]]
The Suppport Vectors are:
[[ -8.47422847
                  5.15621613]
 [-10.260969
                  2.07391791]
    1.3393313
                -10.290988221
                  4.3759541 1
    9.67917724
 [ -6.80002274
                 -7.02384335]
    9.90143538
                 -0.31483149]]
(venv) aviral: HW-6 aviral$
```

Non Linear Separable data



Linear Separable Data Graph

```
(venv) aviral:HW-6 aviral$ python3 linsep.py
                 dcost
     pcost
                             gap
                                    pres
                                           dres
 0: -2.0636e+01 -4.3905e+01
                                    2e+01
                             3e+02
                                           2e+00
 1: -2.2372e+01 -3.7202e+01
                             9e+01
                                    5e+00
                                           5e-01
 2: -2.3112e+01 -3.8857e+01
                             5e+01
                                    2e+00
                                           2e-01
 3: -2.8318e+01 -3.3963e+01
                             1e+01
                                    4e-01
                                           4e-02
 4: -3.2264e+01 -3.3927e+01
                             2e+00
                                    1e-02
                                           1e-03
 5: -3.3568e+01 -3.3764e+01
                             2e-01
                                    1e-03
                                           1e-04
 6: -3.3737e+01 -3.3739e+01
                             2e-03
                                           1e-06
                                    1e-05
 7: -3.3739e+01 -3.3739e+01
                             2e-05
                                    1e-07
                                           1e-08
 8: -3.3739e+01 -3.3739e+01 2e-07
                                    1e-09 1e-10
Optimal solution found.
Weight Values : [[ 7.25005369 -3.86189521]]
Intercept Value: -0.10698559836302746
The equation of the line is :
[7.250053688191241] \times + [-3.861895209680986] y + -0.10698559836302746
```

Linear Separable Result and equation of line

```
((venv) aviral:HW-6 aviral$ python linsepsklearn.py
Weight Values: [[ 3.59965788 -2.03198838]]
Intercept: [0.21848298]
[[0.23307747 0.86884518]
 [0.23918196 0.81585285]
 [0.14301642 0.85313079]
 [0.14586533 0.74931925]
 [0.26419794 0.91067489]
 [0.06756031 0.65812372]
 [0.17422964 0.6157447 ]
 [0.01107579 0.39873158]
 [0.15267995 0.8006936 ]
 [0.03436631 0.50247843]
 [0.3917889
             0.96675591]
 [0.02066458 0.27003158]
 [0.55919837 0.70372314]
 [0.2498981
             0.15693917]
 [0.65628367 0.77812372]
 [0.27872572 0.23552777]
 [0.24979414 0.18230306]
 [0.70631503 0.87261758]
 [0.22068726 0.11139981]
 [0.36354491 0.25915653]
 [0.42066002 0.43762265]
 [0.76570056 0.98727513]
 [0.45552411 0.49956489]
             0.90468041]]
 [0.6798148
(venv) aviral:HW-6 aviral$
```

Linear Separable Result and equation of line (Sci-Kit LEARN)

```
(venv) aviral:HW-6 aviral$ python nonlinsepsklearn.py
Support Vectors: [[ -8.47422847
                                    5.15621613]
  -2.7471552
                 -8.47244873]
  -6.90647562
                  7.14833849]
  -7.47227246
                 -5.09869845]
   -9.53754332
                 -0.51895777]
   -9.46760885
                 2.36139525]
  -8.92844214
                 -1.80373857]
    6.3666283
                 -6.49712918]
    0.20162846
                 -8.81260121]
    6.99249663
                 -6.41143087]
   -6.80002274
                 -7.02384335]
    9.90143538
                 -0.31483149]
   -4.98349411
                 8.31816584]
    4.27289989
                 8.67079427]
  10.24592717
                 7.95373492]
 [-15.64719728
                  3.32039056]
 [-11.64621294
                 -0.87217731]
  12.74780931
                 0.19913032]
 [-10.02833317
                 11.09354511]
               -14.15854536]
    3.28969027
    2.91722251 -12.27214032]
  -1.08933763 -14.10562483]
 [-10.260969
                  2.07391791]
                 12.68562818]
    1.66404809
    1.3393313
                -10.29098822]
                  6.07221393]
  16.42108453
  11.75880948
                 -9.85890377]
    9.67917724
                 4.3759541 ]]
(venv) aviral:HW-6 aviral$
```

Non-linear Separable Result and equation of line (Sci-Kit LEARN)

Applications

SVMs are helpful in text and hypertext categorization as their application can significantly reduce the need for labeled training instances.

- In Quantitative finance, for example predictive tasks, where x consists of features derived from a historical stock indicator time series and y is a sell or buy signal.
- Text based SVM, filter e-mail to be forwarded to either the inbox or the spam folder
- Used in Bioinformatics and research (Eg: Protein classification, Cancer classification)
- Another outstanding application of SVMs is the detection of human faces in graylevelimages. The problem is to determine in an image the location of human faces and, if there are any, return an encoding of their position.
- Hand-written characters can be recognized using SVM.
- Many problems of image processing and image classification are accomplished by SVM.