EE6550 Machine Learning HW2 README

Linear Support Vector Machine for Binary Classification trained with Sequential Minimal Optimization

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User Manual

Dev Environment

- Developed under Anaconda 4.3.0 (x86_64).
- Require Numpy for matrix operations.
- Tested on Python 2.7.12 and Python 3.6.0.

File Structure

- /dataset: (Important) The program reads datasets from this folder and performs training and testing on the specified training and test data files. (See the *Dataset Format* section below)
- svm.py: Support vector machine (SVM) model.
- utils.py: Some utilities used by this program, such as loading dataset, normalize labels, etc.
- main.py: (Important) The main program. User should train a SVM by running this program.
- <K>-fold-result-[HH:MM:SS].txt: (Important) Logs, containing cross validation errors with corresponding hyper-parameter (c in our case), optimal hyper-parameter, optimal hypothesis (weight vector w and bias term b is our case) and performance. This will be generated when running the program without specified hyper-parameter. Note that the file name <K> is the number of "K"-fold cross-validation and HH:MM:SS is the time you run the program.

Dataset Format

- Currently, the program only supports reading .csv file.
- The class label of each item should locate at the first column. The class labels should only be binary, e.g. {+1, -1}, {1, 0} or {'+', '-'}, etc.
- If you want to train your SVM with your own dataset, please be sure that you've followed the required format described above, and have placed your own training and test data files in the /dataset folder.

Getting Started

Train your SVM by running python main.py in terminal. Be sure that your terminal is under the same directory as main.py.

Note that we've set default values for required input arguments. Run python main.py --help to view input arguments information shown below.

Here we show some running examples for different test scenarios:

- For inputting a training data file, which contains a labeled training sample. This labeled training sample is used to train the SVM-learning algorithm which will return a hypothesis after K-fold cross-validation, place the training data file in the /dataset folder.
- For inputting a testing data file, which contains a labeled testing sample. This labeled testing sample is used to evaluate the performance of the returned hypothesis from the SVM-learning algorithm based on the labeled training sample, place the testing data file in the /dataset folder.
- For inputting a positive integer K to perform K-fold cross-validation to determine the optimal value C of the free parameter C to minimize the cross-validation error, specify the argument -- k when running main.py (default value is 5).
- Summing up the above, you can also run, for example, python main.py
 --train_filename="xxx.csv" --test_filename="xxx.csv" --K=5. Note that training data and test data files must be specified.
- For checking the obtained optimal value C, optimal hypothesis and performance, see the generated log file <K>-fold-result-[HH:MM:SS].txt.

Next, we describe reported results in the *Report* section.

Report

Note that training on a single hyper-parameter c takes about 1~3 minutes. The larger c is, the more time it is needed to train a SVM.

Experiment Set

Tested on 20 different hyper-parameters c

```
[0.1, 0.15, 0.19, 0.24, 0.29, 0.34, 0.38, 0.43, 0.48, 0.53, 0.57, 0.62, 0.67, 0.72, 0.76, 0.81, 0.86, 0.91, 0.95, 1.0]
```

5-Fold Cross-validation

- ~2 hours to perform the experiment.
- The cross-validation errors over 20 different c is shown as follows:

C Cross-validation Error

- 0.1 0.270808929632
- 0.15 0.269552669553
- 0.19 0.279976232917
- 0.24 0.26301672184
- 0.29 0.274789915966
- 0.34 0.268253968254
- 0.38 0.260402342755
- 0.43 0.262999745353
- 0.48 0.277353365589
- 0.53 0.278652066887
- 0.57 0.264306934895
- 0.62 0.257813428402
- 0.67 0.268236991766
- 0.72 0.277327900857
- 0.76 0.263050674815
- 0.81 0.269544181309
- 0.86 0.266929802224
- 0.91 0.268236991766
- 0.95 0.269552669553
- 1.0 0.272158560394
- Optimal C: 0.62
- Weight vector: [0.41376851, -1.1783888, 0.76335893, -0.2726952, -0.37456999,
 - -0.15391376, 0.02827039, 0.00133731, 0.0055075, -0.00362205, -0.02664064, -0.03642613,
 - -0.02312384, 0.52445173, 0.92577699, 0.57587795, 0.20615108, -0.24830772, -0.0214078]
- Bias term: -0.553474057815
- Performance:
 - o training error: 0.247395833333
 - test error: 0.240208877285

10-Fold Cross-validation

- ~5 hours to perform the experiment.
- The cross-validation errors over 20 different c is shown as follows:

C Cross-validation Error

0.1 0.25517771702

- 0.15 0.270830485304
- 0.19 0.265447710185

C Cross-validation Error

- 0.24 0.256459330144
- 0.29 0.266866028708
- 0.34 0.262918660287
- 0.38 0.268147641832
- 0.43 0.270745044429
- 0.48 0.279887218045
- 0.53 0.257740943267
- 0.57 0.260269993165
- 0.62 0.261637047163
- 0.67 0.27462406015
- 0.72 0.259039644566
- 0.76 0.264200273411
- 0.81 0.269480519481
- 0.86 0.256425153794
- 0.91 0.261637047163
- 0.95 0.266814764183
- 1.0 0.265516062884
- Optimal C: 0.1
- Weight vector: [0.2, -0.70799666, 0.73837806, -0.24449533, -0.38217573, -0.13843935, -0.00088726, 0.02371915, 0.00757344, -0.01023123, -0.01775169, -0.00509908, 0.00981942, 0.17055394, 0.21350525, 0.11907159, 0.03387551, -0.06414395, -0.08501416]
- Bias term: -0.721050592969
- Performance:
 - training error: 0.255208333333
 - test error: 0.22454308094