EE6550 Machine Learning HW5 Neural Network

A simple neural network trained with SGD for regression.

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User Manual

Dev Environment

- Developed under Anaconda 4.3.0 (x86_64).
- Require Numpy for matrix operations.
- Tested on 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)
- /logs: (Important) Output reports are stored in this folder.
- /hypothesis: (Important) Output hypotheses are stored in this folder. (There is timestamp at the end of the file name, which corresponds to the report)
- core.py: Components to build our model.
- utils.py: Some utilities used by this program, such as loading dataset, normalize labels, etc.
- regression nn.py: Our model implementation.
- main.py: (Important) The main program. User should train the model by running this program.

Dataset Format

- Currently, the program only supports reading .csv file.
- The true label (real value) of each item should locate at the first column.
- (Important) If you want to train your SVR 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 model 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.

```
usage: main.py [-h] [--header_filename HEADER_FILENAME]
[--train_filename TRAIN_FILENAME]
[--test_filename TEST_FILENAME] [--depth DEPTH]
```

```
[--learning rate LEARNING RATE] [--reg lambda REG LAMBDA]
Nerual network (NN) for regression.
optional arguments:
                        show this help message and exit
 --header filename HEADER FILENAME
                       Training dataset with header csv. This is used to
                        output hypothesis. (Default:
                        "energy efficiency cooling load training header.csv")
  --train filename TRAIN FILENAME
                        Training dataset csv. (Default:
                        "energy_efficiency_cooling_load_training.csv")
  --test filename TEST FILENAME
                        Training dataset csv. (Default:
                        "energy efficiency cooling load testing.csv")
  --depth DEPTH
                       Number of weight matrices (Defualt: 2. (depth-1)
                        hidden layers)
  --learning_rate LEARNING_RATE
                        Step size of updating weights. (Default: 0.01)
  --reg lambda REG LAMBDA
                        Strength of L2 regularization for weights. (Default:
  --K K
                        Denotes for "K"-fold cross-validation for determine
                        the optimal hyper-parameters for NN. (Default: None)
  --max iteration MAX ITERATION
                        Max iteration for NN training algorithm to avoid not
                        converging. (Defualt: 30000)
```

For Grading Session

Here we show some guides for different test scenarios:

- Place the training data file(e.g. xxx_training.csv), testing data file(e.g. xxx_testing.csv) and training header file(e.g. xxx_training_header.csv) in the /dataset folder before running main.py with specified --train_filename, --test_filename and --header_filename.
- To specify model depth, for example, run --depth=2 will generate 2-layer neural net with hidden size 20 (20 is our default hidden size)
- To specify learning rate, run --learning rate=0.01. (Default: 0.0.1)
- To specify regularization term, run --reg lambda=0.1. (Default: 0.0)
- To specify max iterations, run --max iteration=10000. (Default: 30000)
- To perform K-fold cross-validation, for example, specify --K=5. (Default: None)
- All the required output information, such as class label mapping, cross-validation history, optimal hyper-parameters, etc., are stored at the <code>logs/</code> folder. Note that the log file name indicates what kernel type and number of K-fold you choosed, and when you run the program. This naming convension aims to help graders to choose which report to check after running the program.
- Note that the csv file name of our hypothesis in the hypothesis/ folder is concatenated with the timestamp. This aims to let the graders know which hypothesis file to choose to test after running the program. (You may remove the timestamp for grading)

• Here is one of a example for running the following commands in the test scienarios:

```
(For 5-fold CV)
$ python main.py --header_filename="xxx_training_header.csv"
--train_filename="xxx_training.csv" --test_filename="xxx_testing.csv" --K=5
```

Report

All required output information are stored at logs/.

Snippet (See more in logs/):

```
[*] Cross validation history:
- Parameter: {'depth': 4, 'reg lambda': 0.0, 'max iteration': 30000} | Cross
validation error: [ 4.08458964]
- Parameter: {'depth': 2, 'reg lambda': 0.0, 'max iteration': 30000} | Cross
validation error: [ 5.14624725]
- Parameter: {'depth': 3, 'reg lambda': 0.0, 'max iteration': 30000} | Cross
validation error: [ 5.33495332]
  Parameter: {'depth': 3, 'reg_lambda': 0.001, 'max_iteration': 30000} | Cross
validation error: [ 5.40552295]
- Parameter: {'depth': 2, 'reg lambda': 0.001, 'max iteration': 30000} | Cross
validation error: [ 5.64238829]
- Parameter: {'depth': 5, 'reg lambda': 0.001, 'max iteration': 30000} | Cross
validation error: [ 5.97097392]
- Parameter: {'depth': 4, 'reg_lambda': 0.001, 'max_iteration': 30000} | Cross
validation error: [ 6.02975134]
- Parameter: {'depth': 6, 'reg_lambda': 0.0, 'max_iteration': 30000} | Cross
validation error: [ 6.19784738]
- Parameter: {'depth': 2, 'reg lambda': 0.01, 'max iteration': 30000} | Cross
validation error: [ 6.27289822]
- Parameter: {'depth': 2, 'reg lambda': 0.005, 'max iteration': 30000} | Cross
validation error: [ 6.37716251]
- Parameter: {'depth': 5, 'reg lambda': 0.0, 'max iteration': 30000} | Cross
validation error: [ 6.58103282]
- Parameter: {'depth': 3, 'reg_lambda': 0.005, 'max_iteration': 30000} | Cross
- Parameter: {'depth': 3, 'reg lambda': 0.01, 'max iteration': 30000} | Cross
validation error: [ 7.7029069]
- Parameter: {'depth': 4, 'reg lambda': 0.005, 'max iteration': 30000} | Cross
- Parameter: {'depth': 4, 'reg lambda': 0.01, 'max iteration': 30000} | Cross
validation error: [ 8.09004805]
- Parameter: {'depth': 6, 'reg_lambda': 0.001, 'max_iteration': 30000} | Cross
validation error: [ 8.62358754]
- Parameter: {'depth': 5, 'reg lambda': 0.01, 'max iteration': 30000} | Cross
- Parameter: {'depth': 5, 'reg lambda': 0.005, 'max iteration': 30000} | Cross
- Parameter: {'depth': 6, 'reg lambda': 0.005, 'max iteration': 30000} | Cross
validation error: [ 10.59537494]
 - Parameter: {'depth': 6, 'reg lambda': 0.01, 'max iteration': 30000} | Cross
```

Our hypothesis is located at hypothesis/SGD_hypothesis_header-[14:11:18].csv. The saving format is as the same as what professor described at *ilms*.

Note that the corresponding hypothesis csv file is indicated by the HH:MM:SS timestamp, which is the time that the program finished training.

Discussion

- I've run dozens of hidden layer size and finally turned out that 20 is good enough.
- If regularization term is specified, test error might increase, but the overfitting problem might be improved (training error is not much larger than validation error)