

Fasttext for Text Classification

Fasttext is a library created by the Facebook Research Team for **efficient learning of word representations** and **sentence classification**. FastText differs in the sense that word vectors a.k.a word2vec treats every single word as the smallest unit whose vector representation is to be found but FastText assumes a word to be formed by a n-grams of character, for example, sunny is composed of [sun, sunn,sunny],[sunny,unny,nnny] etc, where n could range from 1 to the length of the word. This document is a stepwise walkthrough to study Fasttext using a cooking dataset of stack exchange ref: <https://fasttext.cc> .

1. Install Fasttext through pip in bash. If not pip, you can clone the library from git using this link: \$ wget <https://github.com/facebookresearch/fastText.git> followed by \$ cd fastText and \$ make

```
Terminal 1
bash-3.2$ pip install fasttext
Collecting fasttext
  Using cached fasttext-0.8.3.tar.gz
Requirement already satisfied: numpy>=1 in ./anaconda3/lib/python3.6/site-packages (from fasttext)
Requirement already satisfied: future in ./anaconda3/lib/python3.6/site-packages (from fasttext)
Building wheels for collected packages: fasttext
  Running setup.py bdist_wheel for fasttext ... done
  Stored in directory: /Users/macbook/Library/Caches/pip/wheels/55/0a/95/e23f773666d3487ee7456b220f7e8d37e99b74833b20dd06a0
Successfully built fasttext
Installing collected packages: fasttext
Successfully installed fasttext-0.8.3
bash-3.2$ wget
bash: wget: command not found
bash-3.2$ git clone https://github.com/facebookresearch/fastText.git
Cloning into 'fastText'...
remote: Counting objects: 1979, done.
remote: Compressing objects: 100% (71/71), done.
remote: Total 1979 (delta 46), reused 91 (delta 33), pack-reused 1862
Receiving objects: 100% (1979/1979), 4.37 MiB | 164.00 KiB/s, done.
Resolving deltas: 100% (1230/1230), done.
bash-3.2$ cd fastText
bash-3.2$ make
c++ -pthread -std=c++0x -march=native -O3 -funroll-loops -c src/args.cc
c++ -pthread -std=c++0x -march=native -O3 -funroll-loops -c src/dictionary.cc
c++ -pthread -std=c++0x -march=native -O3 -funroll-loops -c src/productquantizer.cc
c++ -pthread -std=c++0x -march=native -O3 -funroll-loops -c src/matrix.cc
c++ -pthread -std=c++0x -march=native -O3 -funroll-loops -c src/qmatrix.cc
c++ -pthread -std=c++0x -march=native -O3 -funroll-loops -c src/vector.cc
c++ -pthread -std=c++0x -march=native -O3 -funroll-loops -c src/model.cc
c++ -pthread -std=c++0x -march=native -O3 -funroll-loops -c src/Utils.cc
c++ -pthread -std=c++0x -march=native -O3 -funroll-loops -c src/fasttext.cc
c++ -pthread -std=c++0x -march=native -O3 -funroll-loops args.o dictionary.o productquantizer.o matrix.o qmatrix.o vector.o model.o Utils.o fasttext.o src/main.cc -o fasttext
bash-3.2$ ./fasttext
usage: fasttext <command> <args>

The commands supported by fasttext are:

supervised          train a supervised classifier
```

2. To check if Fasttext has been installed properly and works, type the commands as: \$./fasttext

```
Terminal 1
bash-3.2$ ./fasttext
usage: fasttext <command> <args>

The commands supported by fasttext are:

supervised          train a supervised classifier
quantize            quantize a model to reduce the memory usage
test               evaluate a supervised classifier
predict            predict most likely labels
predict-prob       predict most likely labels with probabilities
skipgram           train a skipgram model
cbow               train a cbow model
print-word-vectors print word vectors given a trained model
print-sentence-vectors print sentence vectors given a trained model
print-ngrams       print ngrams given a trained model and word
nn                 query for nearest neighbors
analogies           query for analogies
dump               dump arguments,dictionary,input/output vectors
```

3. Fasttext works on a specific type of data that needs to be labelled. Hence we make use of such data from [the cooking section of Stackexchange](https://s3-us-west-1.amazonaws.com/fasttext-vectors/cooking.stackexchange.tar.gz), and their associated tags using

```
$ curl -O https://s3-us-west-1.amazonaws.com/fasttext-vectors/cooking.stackexchange.tar.gz
&& tar xvzf cooking.stackexchange.tar.gz
```

```
bash-3.2$ curl -O https://s3-us-west-1.amazonaws.com/fasttext-vectors/cooking.stackexchange.
tar.gz && tar xvzf cooking.stackexchange.tar.gz
% Total    % Received % Xferd  Average Speed   Time    Time     Time  Current
           Dload  Upload   Total       Spent    Left     Speed
100 446k 100 446k    0     0  65372      0  0:00:07  0:00:07 --:--:-- 84396
x cooking.stackexchange.id
x cooking.stackexchange.txt
x readme.txt
bash-3.2$ head cooking.stackexchange.txt
_label_sauce _label_cheese How much does potato starch affect a cheese sauce recipe?
_label_food-safety _label_acidity Dangerous pathogens capable of growing in acidic environments
_label_cast-iron _label_stove How do I cover up the white spots on my cast iron stove?
_label_restaurant Michelin Three Star Restaurant; but if the chef is not there
_label_knife-skills _label_dicing Without knife skills, how can I quickly and accurately dice vegetables?
_label_storage-method _label_equipment _label_bread What's the purpose of a bread box?
_label_baking _label_food-safety _label_substitutions _label_peanuts how to separte peanut oil from roasted
peanuts at home?
_label_chocolate American equivalent for British chocolate terms
```

Thus, the above data has been labelled with the text.

4. To classify this data, we need to split into train and validate. We count the number of examples using 'wc' command and split examples into train and validate. Thus 12000 examples are in train and 3000 in validate.

```
bash-3.2$ wc cooking.stackexchange.txt
15404 169582 1401900 cooking.stackexchange.txt
bash-3.2$ head -n 12404 cooking.stackexchange.txt > cooking.train
bash-3.2$ tail -n 3000 cooking.stackexchange.txt > cooking.valid
```

5. Train this classifier using \$./fasttext supervised -input cooking.train -output model_cooking

```
bash-3.2$ ./fasttext supervised -input cooking.train -output model_cooking
Read 0M words
Number of words: 14543
Number of labels: 735
Progress: 100.0% words/sec/thread: 66078 lr: 0.000000 loss: 9.966440 ETA: 0h 0m
```

6. Let us run this over validation data to test if this model works well.

```
bash-3.2$ ./fasttext test model_cooking.bin cooking.valid
N      3000
P@1    0.152
R@1    0.0656
Number of examples: 3000
```

One major point to be noted here is that this data was not pre-processed and thus we get precision and recall as above. We can improve the performance of the model by applying some pre-processing techniques like converting into lowercase etc and again repeat the same process.

```
bash-3.2$ cat cooking.stackexchange.txt | sed -e "s/([[:punct:]]|'/'/()/)/ \1 /g" | tr "[:upper:]" "[:lower:]" > cooking.p
reprocessed.txt
bash-3.2$ head -n 12404 cooking.preprocessed.txt > cooking.train
bash-3.2$ tail -n 3000 cooking.preprocessed.txt > cooking.valid
bash-3.2$ ./fasttext supervised -input cooking.train -output model_cooking
Read 0M words
Number of words: 8952
Number of labels: 735
Progress: 100.0% words/sec/thread: 71676 lr: 0.000000 loss: 9.961704 ETA: 0h 0m
bash-3.2$ ./fasttext test model_cooking.bin cooking.valid
N      3000
P@1    0.178
R@1    0.0771
Number of examples: 3000
```

Precision has been improved by 4% approx. Let us see if further this can be improved by adding epochs. By default, fasttext sees each training example only five times during training, which is small, given that our training set only have 12000 training examples. These number of examples can be increased using an epoch (set to 25) option as follows:

```

bash-3.2$ ./fasttext supervised -input cooking.train -output model_cooking -epoch 25
Read 0M words
Number of words: 8952
Number of labels: 735
Progress: 100.0% words/sec/thread: 72038 lr: 0.000000 loss: 7.195977 ETA: 0h 0m
bash-3.2$ ./fasttext test model_cooking.bin cooking.valid
N      3000
P@1    0.517
R@1    0.223
Number of examples: 3000

```

This gives us surprising results as precision goes up by almost 60%.

7. Another way, learning rate can also be used to improve the performance. This corresponds to how much the model changes after processing each example. A learning rate of 0 would mean that the model does not change at all, and thus, does not learn anything. Good values of the learning rate are in the range 0.1 - 1.0.

```

bash-3.2$ ./fasttext supervised -input cooking.train -output model_cooking -lr 1.0 -epoch 25
Read 0M words
Number of words: 8952
Number of labels: 735
Progress: 100.0% words/sec/thread: 72322 lr: 0.000000 loss: 4.320680 ETA: 0h 0m
bash-3.2$ ./fasttext test model_cooking.bin cooking.valid
N      3000
P@1    0.585
R@1    0.253
Number of examples: 3000

```

The precision goes on increasing with change in parameters.

The other available parameters that can be tuned are –

- **-lrUpdateRate** : change the rate of updates for the learning rate [100]
- **-dim** : size of word vectors [100]
- **-ws** : size of the context window [5]
- **-epoch** : number of epochs [5]
- **-neg** : number of negatives sampled [5]
- **-loss** : loss function {ns, hs, softmax} [ns]
- **-thread** : number of threads [12]
- **-pretrainedVectors** : pretrained word vectors for supervised learning []
- **-saveOutput** : whether output params should be saved [0]

This way we can improve the performance of this model by adding parameters. Fasttext outputs results quickly but it has a limitation of data to be in a very specific way i.e. it should be in the labelled format as

```
__ label __ <X> <Text>
```