# 1 Solution

### 1. Read File

- (a) bag-of-words: use regex to retrieve all the data and store as term frequency vector
- (b) bigram: just like bag of words but put two word as a symbol instead.
- (c) store the file in the container as pattern:[dictionary(key,frequency),sentiment]

# 2. Shuffle

- (a) use a seed to reproduce the result
- (b) cut the train and validate data into two part
- (c) shuffle the train data

### 3. Train

- (a) update the weight if the prediction is wrong
- (b) only update the word occur in a single document(sparse vector)
- (c) average the weight while updating weight to enhance the run-time and reduce memory usage

#### 4. Evaluate

- (a) use the fit function to predict the classification, return 1 as positive, and -1 as negative
- (b) sum the total of correct classifications to compute the accuracy
- (c) the top10 function sort all the weight in order to find the most positive words
- (d) use the progress function to plot the learning progress

# 2 Result

### 1. Accuracy

- (a) The starting precision will be 0.5 because the weight will not update before shuffling. However, all the result will be all positive or negative.
- (b) After shuffling the document, the weight will update and the precision will comes to around 0.5.
- (c) Multiple passes over the training instances will improve the accuracy of prediction. Instead of using the last time weight, I average the weight of each train and iterate for 20 time. The precision will stop at around 0.8375 and stop training. See Figure 1.

(d) With another feature: Bigram, compare to bag-of-words, bigram put two words as one symbol. The initial accuracy ranked up to 0.67 which is much more higher than bad-of-words. However, I don't think it really help improve the accuracy because the training stop at accuracy of 0.8025, but it only took 5 time iteration to meet the result with a very high initial accuracy.

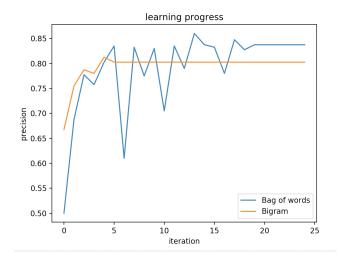


Figure 1: Learning Progress

- 2. The top 10 most positive word
  - (a) bag-of-words:
    - $i. \ seen: 247, great: 230, both: 188, best: 180, quite: 165, see: 159, life: 159, back, 156, ever: 156, most, 150, and 150, back, 156, ever: 156, most, 150, and 1$
  - (b) bigram:
    - i. black cauldron:31,he is:29, ,which is:28,the best:28,one of:27,and it:26,to the:25,the black:25,to see :24,a little:23
  - (c) Better feature:
    - i. As for laptop and restanrant: some symbol may be very important like battery assumption, distance(distant) that will be considered as an important factor. We can train a seperate weight with this value to make then become more influential. Just to said that we can pre-make a weight rather then set them to 0.
  - (d) run time
    - i. bag of words:6.36
    - ii. bigram:11.9