

Today we will implement a spam filter using a Multinomial Naive Bayes classifier on a <u>bag of</u> words model.

This exercise will loosely follow the exercise from Andrew Ng's online Machine Learning course: http://openclassroom.stanford.edu/MainFolder/DocumentPage.php?course=MachineLearning&doc=exercises/ex6/ex6.html

However, we will not use the feature data and first use different probability formulas.

It also worth taking a look at the wiki page:

https://en.wikipedia.org/wiki/Naive_Bayes_spam_filtering

1. Download the prepared pre-processed emails from (<u>do not</u> download the features data from the exercise page):

http://openclassroom.stanford.edu/MainFolder/courses/MachineLearning/exercises/ex6materials/ex6
DataEmails.zip

- 2. To create our classifier, it is recommended to create the following helper functions:
 - a. Read all words from a file into a list
 - b. Count number of instances of words in a list

Optional: Try using Python's <u>defaultdict</u> class which provides a convenient dictionary with default value, thus simplifying the counting instances code.

- Compute the probability of word appearing in a spam or a ham (not a spam) message by counting the number of times it appears in all appropriate messages divided by the number of words in all those messages.
- 4. To combine the probability for all words in a message we inspect, we need to multiply them. Since we are multiplying many small numbers we are likely to run into <u>floating pointing underflow</u>. Hence we will calculate the log probabilities instead, and add them up!
- 5. Now build the classifier which will simply compare the probability of message being ham or spam, and check the accuracy of prediction for the test data.
- 6. Alternatively to the method in 3 you can calculate the probabilities of each word being part of a ham or spam message, using other methods such as the one suggested in the exercise link, or the wiki page. Try to experiment with a few methods, as well as dealing with issues such as words which didn't appear in train data, or rare words.

Good Luck!