

Report: PRML Assignment 3

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Dataset reference:

<http://www.aueb.gr/users/ion/data/enron-spam/preprocessed/enron1.tar.gz>

Algorithm:

We have implemented a Naive Bayes classifier. That corresponding steps are given below.

1) Data loading and Tokenization

We read all emails one by one and split it into words. Then remove special characters out of words and made a dictionary. Dictionary contains words with its frequencies.

2) Training the model

We made 2 csv files one for spam and another for ham. Both are dictionaries, that contains words with its frequencies. so first we tokenize the mail and adds those words to corresponding dictionaries.

3) Testing/Prediction

Suppose we have email $E = \{e_1, e_2, e_3, \dots\}$ where e_i is a word. Now from bayes theorem.

$$P(spam | e_1 \cap e_2 \cap e_3 \dots) = \frac{P(spam)P(e_1 \cap e_2 \cap e_3 \dots | spam)}{P(e_1 \cap e_2 \cap e_3 \dots)}$$

And in naive bayes we assumes all words are independent hence...

$$P(spam | e_1 \cap e_2 \cap e_3 \dots) = \frac{P(spam) P(e_1 | spam) P(e_2 | spam) P(e_3 | spam) \dots}{P(e_1 \cap e_2 \cap e_3 \dots)}$$

So to classify upcoming email spam or ham we need to find

$$P(spam | e_1 \cap e_2 \cap e_3 \dots) \text{ and } P(ham | e_1 \cap e_2 \cap e_3 \dots).$$

Whichever probability is maximum, upcoming email will go into that class.

$$P(e_1) = \frac{\text{total number of } e_1 \text{ in dataset}}{\text{total number of words in dataset}}$$

$$P(e_1 | spam) = \frac{\text{total number of } e_1 \text{ in spam emails}}{\text{total number of words in spam emails}}.$$

We encountered the case where some features are not available in the dictionary so that feature become zero. That issue solved by additive smoothing that formulae given below.

$$P(e_k | spam) = \frac{n_k + 1}{n + |dictionary|}$$

Where, n_k = number of e_k in spam messages.

n = total words in spam messages.

And last we computed $P(spam | e_1 \cap e_2 \cap e_3 \dots)$ and

$P(ham | e_1 \cap e_2 \cap e_3 \dots)$ to classify email as spam or ham.