

LING572 HW3 (Naïve Bayes)

Due: 11pm on Jan 30, 2020

The example files are under `dropbox/19-20/572/hw3/examples/`.

Q1 (5 points): Run the Mallet NB learner (i.e., the trainer's name is NaiveBayes) with **train.vectors.txt** as the training data and **test.vectors.txt** as the test data. In your note file, write down the training accuracy and the test accuracy.

Q2 (35 points): Write a script, **build_NB1.sh**, that implements the Multi-variate Bernoulli NB model. It builds a NB model from the training data, classifies the training and test data, and calculates the accuracy.

- The learner should treat all features as binary; that is, the feature is considered present iff its value is nonzero.
- The format is: `build_NB1.sh training_data test_data class_prior_delta cond_prob_delta model_file sys_output > acc_file`
- `training_data` and `test_data` are the vector files in the text format (cf. **train.vectors.txt**).
- `class_prior_delta` is the δ used in add- δ smoothing when calculating the class prior $P(c)$; `cond_prob_delta` is the δ used in add- δ smoothing when calculating the conditional probability $P(f | c)$.
- `model_file` stores the values of $P(c)$ and $P(f | c)$ (cf. **modell1**).
Comment lines start with “%”. The line for $P(c)$ has the format “classname $P(c)$ logprob”, where logprob is 10-based log of $P(c)$.
The line for $P(f | c)$ has the format “featname classname $P(f|c)$ logprob”, where logprob is 10-based log of $P(f | c)$.
- `sys_output` is the classification result on the training and test data (cf. **sys1**). Each line has the following format:
`instanceName true_class_label c1 p1 c2 p2 ...`,
where $p_i = P(c_i | x) = \frac{P(c_i, x)}{P(x)}$. The (c_i, p_i) pairs should be sorted according to the value of p_i in descending order.
- `acc_file` shows the confusion matrix and the accuracy for the training and the test data (cf. **acc1**).
- As always, **modell1**, **sys1**, and **acc1** are NOT gold standard. These files were created with a much smaller training dataset.

Run `build_NB1.sh` with **train.vectors.txt** as the training data, **test.vectors.txt** as the test data, and `class_prior_delta` set to 0:

- Fill out Table 1 with different values of `cond_prob_delta`.

- Store the `model_file`, `sys_output` and `acc_file` for the second row (when `cond_prob_delta` is 0.5) under `q2/`.

Table 1: Results of your **Bernoulli** NB model

<code>cond_prob_delta</code>	Training accuracy	Test accuracy
0.1		
0.5		
1.0		

Q3 (35 points): Write a script, `build_NB2.sh`, that implements the multinomial NB model. Other than the modeling (e.g., the features in the multinomial NB model are real-valued), everything else (e.g., the input/output files) is the same as in Q2.

- Fill out Table 2.
- Store the `model_file`, `sys_output` and `acc_file` for the second row (when `cond_prob_delta` is 0.5) under `q3/`.

Table 2: Results of your **multinomial** NB model

<code>cond_prob_delta</code>	Training accuracy	Test accuracy
0.1		
0.5		
1.0		

Submission: Submit the following to Canvas:

- Your note file `readme.(txt | pdf)` that includes Table 1 and 2, and any notes that you want the TA to read.
- `hw3.tar.gz` that includes all the files specified in `dropbox/19-20/572/hw3/submit-file-list`, plus any source code (and binary code) used by the shell scripts.
- Make sure that you run `check_hw3.sh` before submitting your `hw.tar.gz`.