

# **Naive Bayes Text Classification**

# CS 585 Homework-1



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### You can run the code like:

python NaiveBayes.py data/aclImdb 1.0

# Classification and Evaluation (40 Points)

```
self.P_positive = P(+) = 0.5

self.P_negative = P(-) = 0.5

size of self.count_positive = 252192

size of self.count_negative = 252192

self.vocab_len = 252192

self.total_positive_words = 2958696

self.total_negative_words = 2885722
```

ALPHA	Accuracy	self.deno_pos	self.deno.neg
0.1	0.81688	2983915.2	2910941.2
0.5	0.82668	3084792.0	3011818.0
1.0	0.8304	3210888.0	3137914.0
5.0	0.83632	4219656.0	4146682.0
10.0	0.83712	5480616.0	5407642.0

## Probability Prediction (20 Points)

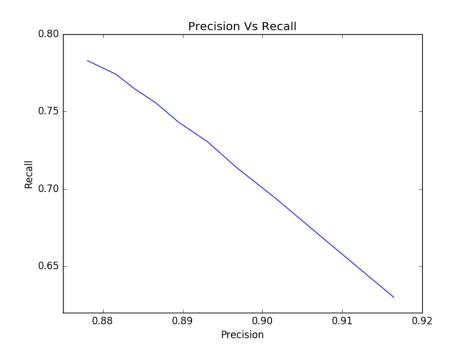
The probability estimated for the first 10 reviews in the test data.

ALPHA = 1.0

Index of test.X[i]	test.Y[i]	predicted_prob_positive	predicted_prob_positive
0	-1.0	0.9365132665902932	0.06348673340965141
1	-1.0	1.4294347203216804e-07	0.999998570565469
2	-1.0	4.388356852437946e-17	1.0
3	-1.0	4.078027965195846e-13	0.99999999995453
4	-1.0	4.775228165409764e-13	0.99999999995453
5	-1.0	8.66332518621793e-08	0.999999133667681
6	-1.0	3.3444380516436173e-08	0.999999665556066
7	1.0	0.020877247067161847	0.9791227529328611
8	-1.0	0.9794966557702657	0.020503344229679848
9	-1.0	1.4350093016703943e-05	0.9999856499070662

## Precision and Recall (20 Points)

Graph precision vs. recall for the positive and negative classes by varying the threshold.



Precision  $\alpha$  (1/Recall)

i.e. there is **inverse relationship** between Precision and Recall.

### Features (20 points)

The 20 most positive and 20 most negative words in the vocabulary sorted by their weight according to model.

```
negative_word_weight = log(P(-|w|) - log(P(+|w|))
positive_word_weight = log(P(+|w|) - log(P(-|w|))
```

**Top +ve words** = [('edie', 4.395851321341416), ('gundam', 4.320816135398502), ('antwone', 4.10414509858991), ('yokai', 3.8482117244527085), ('/>8/10', 3.8482117244527085), ('gunga', 3.827158315254877), ('/>7/10', 3.827158315254877), ('/>10/10', 3.805652110033913), ('din', 3.7836732033151375), ('gypo', 3.7836732033151375), ('othello', 3.7382108292383798), ('7/10.', 3.6145968732712035), ('tsui', 3.560529652000927), ('paulie', 3.546543410026187), ('blandings', 3.532358775034231), ('goldsworthy', 3.4735182750112976), ('kells', 3.442746616344545), ('gino', 3.442746616344545), ('/>9/10', 3.442746616344545), ('harilal', 3.410997918029963)]

**Top -ve words** = [('/>4/10', 4.0660405542897315), ('seagal', 4.057229924607578), ('2/10', 3.914809584565809), ('boll', 3.9045530843986196), ('uwe', 3.894190297363073), ('\*1/2', 3.8516306829442772), ('unwatchable.', 3.829651776225502), ('thunderbirds', 3.76065890473855), ('/>3/10', 3.736561353159491), ('gamera', 3.736561353159491), ('4/10', 3.673647527748921), ('wayans', 3.633907199099406), ('awful!', 3.5783373479445952), ('slater', 3.488725189254909), ('/>avoid', 3.488725189254909), ('segal', 3.4569764909403276), ('drivel.', 3.4569764909403276), ('tashan', 3.4569764909403276), ('kareena', 3.424186668117338)]

#### **Resources:**

- [1] <a href="http://docs.scipy.org/doc/scipy-0.15.1/reference/generated/scipy.sparse.csr">http://docs.scipy.org/doc/scipy-0.15.1/reference/generated/scipy.sparse.csr</a> matrix.html <a href="mailto:scipy.sparse.csr">scipy.sparse.csr</a> matrix
- [2] <a href="https://en.wikipedia.org/wiki/List">https://en.wikipedia.org/wiki/List</a> of logarithmic identities
- [3] <a href="https://stats.stackexchange.com/questions/105602/example-of-how-the-log-sum-%20exp-trick-works-in-naive-bayes">https://stats.stackexchange.com/questions/105602/example-of-how-the-log-sum-%20exp-trick-works-in-naive-bayes</a>