#### CS5560 Knowledge Discovery and Management

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#### References

https://www.analyticsvidhya.com/blog/2015/09/naive-bayes-explained/https://nlp.stanford.edu/IR-book/html/htmledition/text-classification-and-naive-bayes-1.htmlhttp://www.nltk.org/book/ch06.html

I. Consider the problem of classifying the origination point of passenger travel itineraries. Suppose we have the following training set of travel itineraries:

Itinerary	Document	Class
1	"smith: new york - chicago - san francisco - new york"	JFK
2	"chen: san francisco - london - paris - san francisco"	SFO
3	"chen: san francisco - tokyo - singapore- san francisco"	SFO
4	"o'brien: chicago - buenos aires - new york - chicago"	ORD

- a) Assume that we use a Bernoulli (i.e., binary) Naive Bayes model. Compute the following feature probabilities:
  - P(Xfrancisco=true | Class=SFO)
  - P(Xlondon=true | Class=SFO)
  - P(Xfrancisco=true | Class=JFK)
- b) Assume that we use a multinomial NB model instead. Compute the following probabilities:
  - P(X=francisco | Class=SFO)
  - P(X=london | Class=SFO)
  - P(X=francisco | Class=JFK)
- c) Consider a standard Naive Bayes classifier trained on the training set and applied to a similar test set. How accurate is this classifier for:
  - (i) the Bernoulli model, and
  - (ii) the multinomial model?
- d) Construct a non-standard feature representation that is 100% accurate for either model.

II. This problem concerns smoothing Naïve Bayes classifiers. Consider the following formula for Laplace (add-1) smoothing for Naïve Bayes

$$\hat{P}(w_i | c) = \frac{count(w_i, c) + 1}{\sum_{w \in V} (count(w, c) + 1)}$$

$$= \frac{count(w_i, c) + 1}{\left(\sum_{w \in V} count(w, c)\right) + |V|}$$

- a) Suppose we build a Naive Bayes classifier (multinomial or Bernoulli) with no smoothing of the respective P(word | class) probabilities. If a word was unseen in a class, it will thus have a probability of 0. Describe in words the decision procedure of this classifier (emphasizing the effect of the lack of smoothing, and how its decisions will differ from a
- b) Suppose we take a smoothed multinomial classifier and double the amount of smoothing (e.g., for a variant of "add 1 smoothing", add 2 to each count, and add to the denominator 2k, where k is the number of samples). What qualitative effect will this have on decisions
- III. An IR system returns 3 relevant documents, and 2 irrelevant documents. There are a total of 8 relevant documents in the collection.
- a) What is the precision of the system on this search, and what is its recall?
- b) Instead of using recall/precision for evaluating IR systems, we could use accuracy of classification. Consider a classifier that classifies documents as being either relevant or non-relevant. The accuracy of a classifier that makes c correct decisions and i incorrect
  - Why do the recall and precision measures reflect the utility (i.e., quality or usefulness) of an IR system better than accuracy does? (ii)
  - Suppose that we have a collection of 10 documents, and two different boolean retrieval systems A and B. Give an example of two result sets, Aq and Bq, assumed to have been returned by the system in response to a query q, constructed such that Aq has clearly higher utility and a better score for precision than Bq, but such that Aq and Bq have the same scores on accuracy.

## Ber noulli Novie Bayes Model-

- · P ( Xtrancisco = true / class = SFO) = 1.0
  - · P (XLONDON= true | Class=SFO) = 0.5
  - · P(Xfroncisco=frue / class=JFK) 1.0
- (b) Multinomial Naive Bayes modeli-
  - · P (x=Francisco=true / class-SFO)= 4/14
  - · P (x=10ndon/class=sFO) = 1/14
  - ·P(x = Francisco / class = TFK) = 1/8.

# (i) Bernoulli's Modelin

'Not very accurate model because it ignores the frequency information which is important in this domain.

### (ii) Multinomial Model:

"More accurate", because it used frequency information flowever it ignores position information, sodavit distinguish between a city name occurring at the beginning | end of the Hernary from one occurring in the middle.

(d) construct a non-standard feature representation that is 100% accurate for either model.

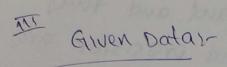
-> use a feature the term that occurs in the last position of each document.

 $\frac{II}{p'(\omega;|c)} = \frac{count(\omega;c)+1}{(Sev(count(\omega,c))+1)}$   $= \frac{count(\omega;c)+1}{(Sev(count(\omega,c))+1)}$ 

without smoothing!

(a) It will never choose a category unless all words in a document never seen for that category for the training set (unless there is no category for which all words were seen similarly to the smoothed camp all words were seen similarly to the smoothing!

e) Doubling the smoothing valuer-It will be more likely to woose categories for which some many of the words in the document users unseen.



Relewant Documents = 3

irrelevant Documents = 2

Relevant Documents = 8

1 Setriened	Mat 16 trished	5
relevant 3 (TP)	5 (FN)	8
(Welevant) 2 (FP)	3 (TN) 1 8	13.

Relevant, noot Not relevant = Relevant - Retrieved relevant

8-3=5 (FN).

An information retrieved system always returns no results will have high accuracy for most queries, since the corpus usually contains only a few relevant documents. Documents that are truly relevant are the only ones that will be

mistakenly classified as non-relevant and thus accuracy is close to 1. Recall and precision are two different measures that can gointly capture the trade off petmeen returing more relevand results are returning temen irrelevant results.

(ii) Assuming a collection of 10 documents.

Two required systems A &B. Lets assume, document 1 is the only relevant document.

Aa -> {1,2,3}

Ba -) {3}.

Both Ag & Ba made 2 mistaker, so they have same

accuracy = 801.

The Precision of Aq=113 " Boy = 0.

since Ba didn't geturn any relevant documenty

as there is no we of it.