## CS 165A Discussion 3

--Break Through MP1

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#### Announcement

• HW2 was out, start early!

## Agenda

Gaussian Naïve Bayes Classifier with Bag-of-Words

Gaussian Naïve Bayes Classifier with TF-IDF

Multinomial Naïve Naïve Bayes Classifier with Bag-of-Words

#### Our task: Movies Review Classification

Compare P (positive | review) and P (negative | review)

## Apply the Bayes Rule

$$P(C_i | review) = \frac{P(review | C_i) P(C_i)}{P(review)}$$

## Apply the Bayes Rule

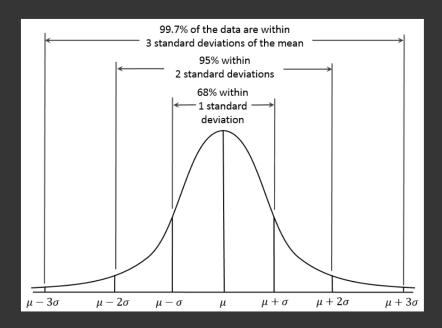
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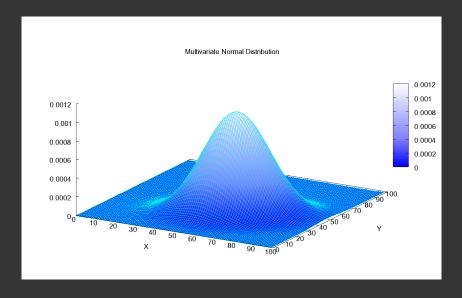
#### How do we know $P(review | C_i)$ ?

- 1. A Gaussian Perspective
  - Assume the probability P(review  $\mid C_i$ ) follows a **Multivariate** Gaussian Distribution.
  - Use Feature Vector represent each review  $\boldsymbol{X}(x_1, x_2 ... x_n) \sim N(\mu, \Sigma)$

μ: Mean

Σ: Covariance Matrix





#### The Multivariate Gaussian Distribution

$$p(x|\mu, \Sigma) = \frac{1}{(2\pi)^{n/2} |\Sigma|^{1/2}} \exp\left\{-\frac{1}{2} (x - \mu)^T \Sigma^{-1} (x - \mu)\right\}$$

How do we know  $\mu$ ,  $\Sigma$  ?

Maximum Likelihood Estimate a.k.a. learn it from the training data.

Find the best parameter  $\mu$ ,  $\Sigma$  that fit the data.

# How to calculate this? Try it out yourself in HW 2:P

$$\hat{\mu} = \frac{1}{N} \sum_{n=1}^{N} x_n$$

$$\hat{\Sigma} = \frac{1}{N} \sum_{n=1}^{N} (x_n - \hat{\mu}) (x_n - \hat{\mu})^T.$$

#### Feature Vector x

Bag-of-Words

Eg. In positive class

Review 1:" another movie about trip "

Review 2: "a very very great movie "

	а	great	another	movie	trip	about	very
review 1	0	0	1	1	1	1	0
review 2	1	1	0	1	0	0	2

$$x1 = (0,0,1,1,1,1,0)$$
  
 $X2 = (1,1,0,1,0,0,2)$ 

#### TF-IDF Feature

- Recall in the BoW feature, each review is represented by a vector.
- Vector contains the frequency of each word in the vocabulary.
- Replace this frequency counts with TF-IDF score for each word.
- TF = (Number of times  $word_i$  appears in the review ) / (Total number of words in this review)
- IDF = log(Total number of reviews / Number of reviews with  $word_i$  in it)

## Multinomial Naïve Bayes Classifier see lecture slide6 - 36

•  $P(review | Positive) = \prod P(word_i | Positive)$ 

$$P(word_i \mid Positive) = \frac{\# ofword_i \text{ in Positive}}{total \# of words \text{ in entire Positive data}}$$

- Multiply by zero --- Apply Laplace Smoothing
- Underflow --- Take log term

## Summary

• Use training data, build your vocabulary for two classes, separately.

- Estimate the parameter of your Gaussian distribution
  - Use BoW as feature
  - Use TF-IDF as feature
- Or use Multinomial Naïve Bayes model

Classify the testing data according to Naïve Bayes Rule.

## Thanks!

Happy Chinese New Year! (Feb 5)