



Exercise 11

Information Retrieval



15..19. Text Classification

Exercise 15..19.1

- Are the following statements true or false? Give reasons for your answer.
 - a) Classification is about assigning a document to one (or more) out of several predefined categories.
 - b) The accuracy of a classifier is always one minus its error rate.
 - c) When using supervised learning for classification, we first train a classifier on unlabeled documents.
 - d) Overfitting means that a classifier is too general.
 - e) With n -times k -fold cross-validation, we partition a set of n labeled documents into k (nearly) equally-sized subsets and for each subset S_i we train a classifier on the documents in S_i and evaluate its performance by applying it to the documents in S_i .
 - f) Given a set of labeled documents, the sequential covering algorithm determines a set of rules for rule-based text classification.

Exercise 15..19.1

- Are the following statements true or false? Give reasons for your answer.
 - g) With probabilistic classification, a document d is assigned to category c_i if the probability $P(c_i | d)$ is the maximum of the probabilities $P(c_k | d)$ for all categories c_k .
 - h) Feature selection can reduce the training time of a classifier, but it cannot improve its quality.
 - i) The Rocchio approach for vector-based classification assumes that the document vectors in each category are close to each other, but distant from the document vectors in the other categories.
 - j) The k NN classifier assigns a doc. d to the k categories whose centroid vectors are closest to \vec{d} .
 - k) The idea of support vector machines (SVMs) is to separate the vector space using an optimal hyperplane and to assign a document d to one of two classes depending on whether \vec{d} lies on the one or on the other side of the hyperplane.
 - l) The SVM approach can only be applied to linearly separable datasets.

ok

Multinomial Model vs. Bernoulli Model

Exercise 15..19.2

- Consider the following Bernoulli and multinomial estimates for the word “the”
 - multinomial model: $\hat{P}(t = \text{"the"} \mid c) \approx 0.05$
 - Bernoulli model: $\hat{P}(t = \text{"the"} \mid c) \approx 1.00$
- Explain the difference

??

Naive Bayes Classification

Exercise 15..19.3

- Based on the data in the table below,
 - estimate a **multinomial Naive Bayes classifier** and apply it to the test document
 - estimate a **Bernoulli Naive Bayes classifier** and apply it to the test document
- You don't need to estimate parameters that you don't need for classifying the test document

| | docId | Tokens in document | In c = China? |
|--------------|-------|-----------------------|---------------|
| Training set | 1 | Taipei Taiwan | Yes |
| | 2 | Macao Taiwan Shanghai | Yes |
| | 3 | Japan Sapporo | No |
| | 4 | Sapporo Osaka Taiwan | No |
| Test set | 5 | Taiwan Taiwan Sapporo | ? |

两套大公式，还得练

Exercise 15..19.4

三种属性选择，目前没答案 NEXT

- Assume we have a collection of 100k documents
- Consider the following frequencies $N_{e_t e_c}$ for the class **coffee** for three terms

| Term t | N_{00} | N_{10} | N_{01} | N_{11} | Frequency $N(t, c)$ | Mutual Inf. $I(U_t; C_c)$ | Chi-Square $\chi^2(t, c)$ |
|----------|----------|----------|----------|----------|------------------------|------------------------------|------------------------------|
| brazil | 98,012 | 102 | 1835 | 51 | | 0,00155 | 818,9 |
| roasted | 99,824 | 143 | 23 | 10 | | 0,00065 | 1964,3 |
| producer | 98,729 | 119 | 1118 | 34 | | | |

$$N_{e_t e_c}$$

\uparrow \uparrow
 $0 \rightarrow t \notin d$ $0 \rightarrow d \notin c$
 $1 \rightarrow t \in d$ $1 \rightarrow d \in c$

a) Fill in the empty cells and select two of these three terms based on

frequency

mutual information

χ^2

b) What are the values of $I(U_t; C_c)$ and $\chi^2(t, c)$ if term and class are completely

dependent

independent