

Statistical Natural Language Processing (SS-2018)

Submission Deadline: 08.06.2018, 23:59

June 1, 2018

Feature Selection (6 points)

1) (1.5 points) Miscellaneous Problems

- Part a (0.75 points):
 - (i) Consider a document containing 100 words wherein the word 'Eutopia' appears 3 times. What is the term frequency for the term 'Eutopia'?
 - (ii) Now, assume we have 100 million documents and the word 'Eutopia' appears in ten thousand of these. What is the inverse document frequency (idf) of this term?
Note: Take log to the base 10. Click on this link to learn about tf-idf
 - (iii) Finally, calculate the tf-idf weight from the values obtained from Part a and Part b.
- Part b (0.25 points):

Choose only one best option and explain your choice.

When training a language model, if we use an overly narrow corpus, the probabilities

 - a. Doesn't reflect the task
 - b. Reflect all possible wordings
 - c. Reflect intuition
 - d. Dont generalize
- Part c (0.5 points)

You are an English Literature teacher and you ask your class to write a play in the style of Shakespeare. You want to score their plays using a trigram language model you computed from a corpus of all Shakespeare plays but you find that the data is too sparse and most of your students sentences receive a score of zero.

How would you use a back-off model to alleviate this problem? Your short answer should be between 50-100 words.

2) (2 points) Imagine we have a predefined set of class labels 'Coffee' and 'Tea'. The following table containing the entry of the counts is given for this purpose:

	black	beans	leaves	rest
Class = Coffee	500	1000	100	400
Class = ~Coffee	500	50	1200	9450
Class = Tea	750	110	1300	400
Class = ~Tea	1000	1500	200	7350

- (1 point) What are good features for predicting class 'Coffee'? Explain your findings.
 - (1 point) What are good features for predicting class 'Tea'? Explain your findings.
Hint: Do this task by checking χ^2 value for all possible features and perform the feature selection.
Ex: $\chi^2(\text{black, Coffee})$.
- #### 3) (2.5 points) You learnt about the feature selection for unsupervised learning (Slide 23 onwards, Chapter 6) in the lecture. Observe the two plots below and comment about the following.
- (1 point) In Plot 1, are features x and y redundant/useful/irrelevant for defining the clusters? Explain your choice.

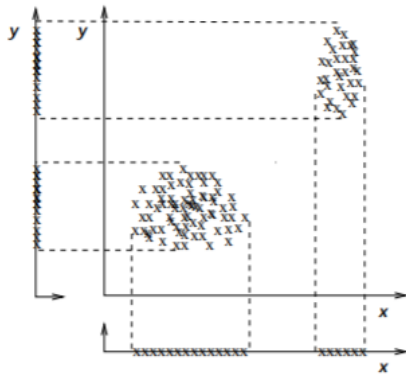


Figure 1: Plot 1

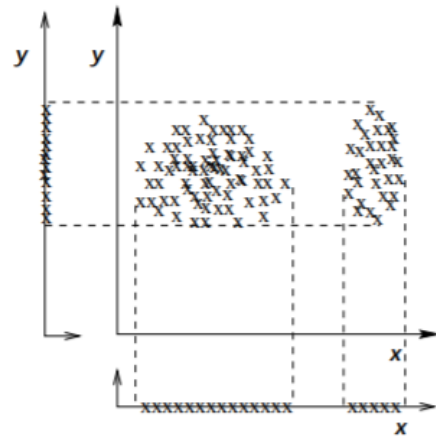


Figure 2: Plot 2

- (1 point) In Plot 2, are features x and y redundant/useful/irrelevant for defining the clusters? Explain your choice.
- (0.5 point) Draw an example plot which portrays feature x and y as the remaining choice from 'redundant/useful/irrelevant' for defining the clusters which was not chosen as an answer in above two parts for plot 1 and plot 2. Explain your plot.

Mutual Information (4 points)

4) (4 points)

- Use the documents provided in "Materials/train" to construct the vocabulary. You need this vocabulary for the next exercise as well. Remember to do the text preprocessing:
 - stopwords removal with the stopwords.txt given in the Materials Folder
 - lowercasing (Can use NLTK 3.3: Reference Link)
 - lemmatization + stemming Reference Link (Can use NLTK 3.3: Reference Link)
 - tokenization (Can use NLTK 3.3: Reference Link)
- (1 point) Find the mutual information between each term and each class (topic). Compute $pmi(t)$ in the case we want each term to discriminate well for a single category.
- (1 point) Use the $pmi(t)$ s to do the feature selection such that it results in 10 features and report them. How much has your problem's dimension decreased?
- (1 point) Do the feature selection this time by MI¹ and select the 10 terms with greatest MI. How do these features differ from the previous part of the question? Report these features and their differences with previous part.
- (1 point) Use the features obtained from each case separately to classify each test file by Naïve Bayes Classifier.
 - Compute the likelihoods for each word (after feature selection) in each class (topic)
 - Assume uniform prior probability for classes
 - Classify by posterior probability

¹https://en.wikipedia.org/wiki/Mutual_information in which each term is a random variable

Submission Instructions

- You must form groups of 2 to 3 people
- Submit only 1 archive file in the ZIP format with name containing the MN of all the team members, e.g.:

`Exercise_07_MatriculationNumber1_MatriculationNumber2.zip`

- Provide in the archive:
 - i. your code, accompanied with sufficient comments
 - ii. a PDF report with answers, solutions, plots and brief instructions on executing your code
 - iii. a README file with the group member names, matriculation numbers and emails
 - iv. Data necessary to reproduce your results
- The subject of your submission mail **must** contain the string [SNLP] (including the braces) and explicitly denoting that it is an exercise submission, e.g:

`[SNLP] Exercise# Submission MatriculationNumber1 MatriculationNumber2`

- Depending on your tutorial group, please send your assignment to the **corresponding tutor**:
 - * Mo. 16-18: Lukas Lange *s9lslang@stud.uni-saarland.de*
 - * Th. 14-16: Harshita Jhavar *snlp18.thursday@gmail.com*
 - * Fr. 10-12: Marimuthu Kalimuthu *mkalimuthu@lsv.uni-saarland.de*
- If two teams submit same solutions, both teams will be given 0 points and no presentation chance will be given to the two team members. If you do this again, you will be disqualified from the exam.