













Inspire...Educate...Transform.

Text mining: Full day lab 20160910 - Batch 17 - CSE7306c

Group Activity -Author Identification

- 1. Make groups with 3 benches into 1
- 2. This is a group activity
- 3. 10 minutes will be given for this task.
- 4. Task: Author identification
 - a. Identifying the author of a given text.
 - b. Classification problem
 - c. Documents with known authors are used for training
 - d. The aim is to automatically determine the corresponding author of an anonymous text using the training data
- 5. You are given total 300 paragraphs of 3 different authors (100 paragraphs for each other)
- 6. Write the steps that you follow to solve this classification problem in detail







Steps for the GroupActivity: Author Identification

- 1. Read the text data
- 2. build a corpus using the text mining (tm) package
- 3. Data processing
 - Text cleansing
 - Key word extraction: candidate keyword list, TF-IDF
 - Term extraction, dimensionality reduction, feature selection, etc
- 4. clean up the corpus using tm_map()
 - a. removeNumbers
 - b. removePunctuation
 - c. stripWhitespace
 - d. content_transformer(tolower)
 - e. removeWords using stopwords("english")
 - f. stemDocument
 - g. Covert to PlainTextDocument
- 5. Vectorization: Create DocumentTermMatrix
- 6. Convert DTM into dataframe
- 7. Do the Visualizations to understand the important attributes of each author
- 8. Decide up on the removal of sparse terms from a document-term matrix decide the sparse percentage
- 9. Add the class label to the dataframe
- 10. Split the dataset into train and test
- 11. Build the Model (you can experiment with different models also)
- 12. Predict for the test dataset
- 13. Compute the evaluation metrics







Text mining

Text categorization (a.k.a. **text classification**) is the task of assigning predefined categories to free-**text** documents.

- Pre-given categories and labelled document examples
- Classify a new document
- A standard classification modelling (Supervised learning)





Steps involved in text classification

- Labeled text documents
- Text data pre-processing
 - Text cleansing
 - Key word extraction: candidate keyword list, TF-IDF
 - Term extraction, dimensionality reduction, feature selection, etc.
- Define Train and test sets
- Create classification model on train data
 - Vector space models:
 - SVM, KNN, Decision trees, Neural nets, etc.
 - Probabilistic models
 - Naïve Bayes classifier.
- Classification model evaluation
- Classification of unknown text documents.





Preprocessing steps

- library(tm)
- Create the corpus : Corpus(DataframeSource(data.frame(text)))
- Preprocessing using tm_map
- removeNumbers removes numbers,
- removePunctuation removes punctuation symbols,
- stripWhitespace removes extra spaces,
- removeWords stopwords(language='english') removes stopwords for the language specified -
- content_transformer(tolower) transforms all upper case letters to lower case,





DocumentTermMatrix

- Once the data is cleansed, create the document term matrix.
 - Use as is DTM matrix
 - Use TFIDF matrix
 - Use Binary form
- Convert the document term matrix as a data.frame to find features and perform modeling





Keyword extraction (ngram)

- Manually inspect a sample of documents in each category and list down the keywords that are representative of the class. Example of some 1-grams are:
- "auto"
 - Airbags, acceleration, gear, speeding, automotives, etc.
- "med"
 - Diagnosis, treatment, illness, disease, death, symptoms, etc.
- "comp"
 - Graphics, visuals, image, processor, resolution, compiler, etc.

Define scores based on the number of keywords found for each class.





Adding class attribute

Add the class attribute to the document term dataframe

```
class <- c(rep("auto",1000), rep("med",1000), rep("comp",1000))
FullData.p.DtM.C = cbind(dt_matrix,class)
```

At this stage, build models to check for accuracy of classification.

To improve further, reduce features, build dictionary of terms and score the document for classification.





Feature reduction: remove sparseness

1. Remove sparse terms

dt_matrix <- removeSparseTerms(FullData.p.DtM, 0.99)</pre>

terms with frequency at least >N(1-0.99) will be retained. N is total number of docs.





Feature reduction: Random Forest

- Use random forest on the data and find terms that are important.
- From the random forest model summary, you obtain the importance of terms.
- varImp(rfmodel) or
- importance(rfmodel,type=2)
- Use reduced number of terms only in the model and ignore the rest.





Model building

- Split the data into train and test data sets
- Build the model
- Obtain train and test accuracy





Several aspects to consider

- There is no single way of approaching a text analysis.
- Text preprocessing
 - Dates/Numbers/Abbreviations/Alphanumerics/stop words/stemming/spell-check,
 etc. might be important based on the domain and quality of data. Need careful analysis and use of regular expressions to remove or retain.
- Keywords extractions
 - Manual inspection, frequency of words/wordcloud, POS, Named-entities, wordnet, etc.
 - Ngrams
 - Synonyms dictionary lookup, Abbreviations-Full forms dictionary, etc.
 - Sentiment, confounding characteristics, opinions, etc.
- Feature engineering
 - Use DTM matrix/TF-IDF/Binary data
 - Defining scores based on keywords
 - Defining semantic based rules, etc.





Several aspects to consider

- Feature reduction
 - Sparsity, C5.0 importance, RF importance, PCA, SVD, etc.
- Modeling and evaluation
 - Train-Test split, Stratified split in case of class imbalance, Merging categories, etc.
 - Semantic rules
 - LSA, text summarization Ranking
 - Clustering stability check, etc.
 - Classifier confusion matrix, ROC, etc.





References http://finzi.psych.upenn.edu/library/LSAfun/html/genericSummary.html

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- https://eight2late.wordpress.com/2015/07/22/a-gentle-introduction-to-cluster-analysis-using-r/
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Text mining infrastructure in R

- http://www3.ntu.edu.sg/sce/pakdd2006/tutorial/pakdd06-Tutorial%20Text%20Clustering.pdf
 Mahout implementation of Text classification
- https://mahout.apache.org/users/classification/twenty-newsgroups.html
 Wordnet dictionary in R
- https://cran.r-project.org/web/packages/wordnet/vignettes/wordnet.pdf







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