



Python and Web Data Extraction: Introduction

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- Overview
- Text Representation
- The Natural Language Toolkit (NLTK)
- Tutorial 3: Computing TF and TF-IDF

Natural Language Processing (NLP)

• Natural language:

 Language that is used for everyday communication by humans

• Natural Language Processing (NLP):

Any kind of computer manipulation of natural language.

Tools

- Text representation
 - Tokenization
 - Stop words removal
 - Stemming
 - Simple summarization
 - Frequency
 - TF-IDF

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Text Representation: A Sample Text

The raw text format is not convenient for any statistical analysis

Google is a global technology leader focused on improving the ways people connect with information. We aspire to build products and provide services that improve the lives of billions of people globally.

Tokenization

- Tokenization: splitting text into words and sentences
- The "bag of words" representation
 - Each document is a "bag"
 - The "bag" contains word tokens
 - Word order is ignored

Stopwords Removal

- Stopwords:
 - Typically function words: a, an, and, as, for, in, of, the, to
 - Are usually discarded from a text representation
 - Google global technology leader focused improving ways people connect information

Stemming

- A common root may have multiple variants
 - Accounting, accountant, accountants
 - Manage, management, managing, manager
- Stemming is the process of reducing words to their word "stem"
 - Accounting, accountant, accountants => account
 - Manage, management, managing, manager => manag
- May not always be used

Term frequency

• Term frequency (tf)

How often a word occurs in the document

- Vector Space Model
 - Each document in the corpus is represented by a vector in the word space

$$d_i = \{tf_{i1}, \dots, tf_{ij}, tf_{iM}\}$$

- tf_{ij} represents the term frequency of word j in doc i
- *M* is the number of unique words in the corpus

tf-idf Model

• The tf-idf model further considers the distinctive power of words (i.e., IDF)

 $d_i = \{tf_{i1} * idf_1, \dots, tf_{ij} * idf_j, tf_{iM} * idf_M\}$

- tf_{ij} represents the term frequency of word j in doc i. The log scale $log(1 + tf_{ij})$ is often used in practice
- idf_j represents the inverse document frequency of word j. The log scale is $log\left(\frac{N}{df_j}\right)$ is often used in practice

tf-idf versus tf

| Document 1 | | Doc | Document 2 | |
|------------|------------|---------|------------|--|
| Term | Term Count | Term | Term Count | |
| this | 1 | this | 1 | |
| is | 1 | is | 1 | |
| a | 2 | another | 2 | |
| sample | 1 | example | 3 | |

 $ext{tf}(\mathsf{example}, d_2) = 3 \quad ext{ idf}(\mathsf{example}, D) = \log rac{2}{1} pprox 0.3010$

 $\mathrm{tfidf}(\mathsf{example}, d_2) = \mathrm{tf}(\mathsf{example}, d_2) \times \mathrm{idf}(\mathsf{example}, D) = 3 \times 0.3010 \approx 0.9030$

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Installing NLTK package

- The Natural Language Toolkit (NLTK) provides:
 A set of tools for the common NLP processes
- Use pip in your command line interface to install pip install nltk

NLTK Modules

| Task | NLTK modules | |
|----------------------------|--------------------------------|--|
| Accessing corpora | nltk.corpus | |
| String processing | nltk.tokenize, nltk.stem | |
| Collocation discovery | nltk.collocations | |
| Part-of-speech tagging | nltk.tag | |
| Classification | nltk.classify, nltk.cluster | |
| Chunking | nltk.chunk | |
| Parsing | nltk.parse | |
| Semantic interpretation | nltk.sem, nltk.inference | |
| Evaluation metrics | nltk.metrics | |
| Probability and estimation | nltk.probability | |
| Applications | nltk.app, nltk.chat | |
| Linguistic fieldwork | nltk.toolbox | |

Functionality

standardized interfaces to corpora and lexicons tokenizers, sentence tokenizers, stemmers t-test, chi-squared, point-wise mutual information n-gram, backoff, Brill, HMM, TnT decision tree, maximum entropy, naive Bayes, EM, k-means regular expression, n-gram, named-entity chart, feature-based, unification, probabilistic, dependency lambda calculus, first-order logic, model checking precision, recall, agreement coefficients frequency distributions, smoothed probability distributions graphical concordancer, parsers, WordNet browser, chatbots

manipulate data in SIL Toolbox format

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Tutorial 3: Computing TF and TF-IDF

• Download the 5tfidf.py and put it in the same folder with previous files

- Run the script.
- You will find two new files: tf.csv and tfidf.csv

Other Resources

• <u>Natural Language Processing with Python</u> (for Python 2)