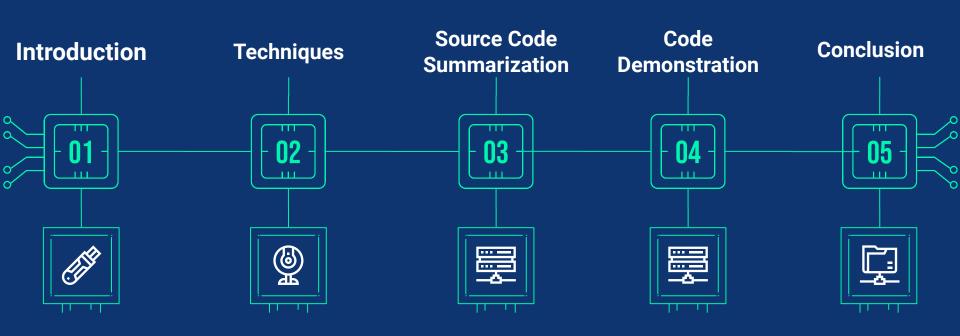


TABLE OF CONTENTS









WHAT IS COMPUTATIONAL LINGUISTICS?

- Interdisciplinary field
 - Linguistics & Computer Science
- Goal: Use computational methods to study human language
- Natural Language Processing (NLP)
- Goal: The implementation of algorithms for processing and analyzing human language

CHALLENGES WITH PROCESSING NATURAL LANGUAGE

- Ambiguity: Natural language inherently ambiguous
- Variation: Variation in language use
- Pragmatics: Context in which language is used
- Human bias

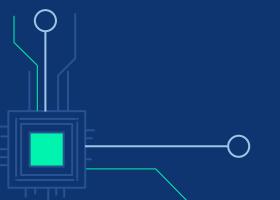
CAREERS INVOLVING COMPUTATIONAL LINGUISTICS / NLP

- NLP / Machine Learning Engineer
- Data Scientist
- Linguistics Researcher
- Software Developer



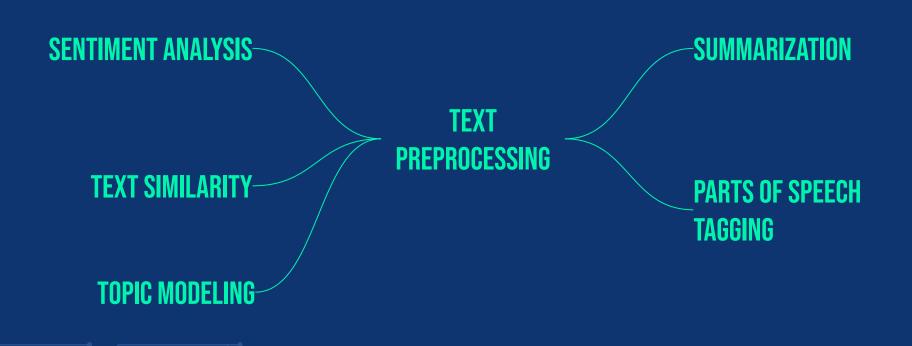


NLP TECHNIQUES





NLP TECHNIQUES



TEXT PREPROCESSING

- Remove
 - Punctuation
 - Stop words
 - Commonly used words that do not carry much meaning
- Lower case
- Tokenization
 - Splitting text into smaller units, "tokens".
- Stemming
 - Removing suffixes to obtain root of the word
- Lemmatization
 - Reducing a word to its base or dictionary form, "lemma"

SENTIMENT ANALYSIS

- Classify the emotional tone / sentiment of a piece of text
- You own a restaurant and you want to improve customer satisfaction by identifying areas where your guests are most satisfied and dissatisfied. You have a lot of customer feedback and reviews on your website, but it's difficult to read and analyze them all manually
- Specific keywords / phrases



TEXT SIMILARITY

- Quantify the similarity between two pieces of text
- Lexical similarity & semantic similarity
- Jaccard similarity, cosine similarity, kmeans clustering
- Convert words to vectors

$$ext{similarity} = \cos(heta) = rac{\mathbf{A} \cdot \mathbf{B}}{\|\mathbf{A}\| \|\mathbf{B}\|} = rac{\sum\limits_{i=1}^n A_i B_i}{\sqrt{\sum\limits_{i=1}^n A_i^2} \sqrt{\sum\limits_{i=1}^n B_i^2}},$$

"How old are you?"

"What is your age?"

"My phone is good."

Embed

[0.3, 0.2, ...]

[0.4, 0.1, ...]

[0.9, 0.6, ...]

How old are you?

What is your age?

My phone is good.

Your cellphone looks great.

How old are you?

What is your age?

My phone is good.

Your cellphone looks great.

How old are you?

What is your age?

My phone is good.

Your cellphone looks great.

How old are you?

What is your age?

Ny phone is good.

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How old are you?

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How old are you?

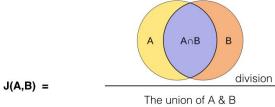
Ny phone is good.

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How old are you?

Ny phone is good.

Your cellphone looks great.

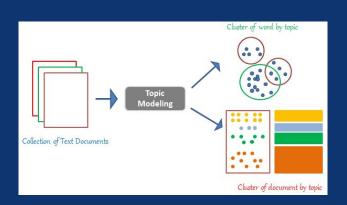


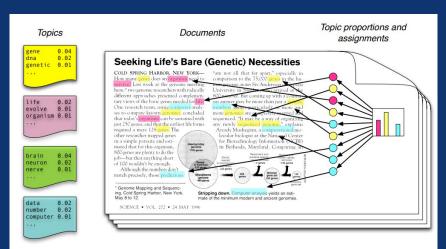
The intersect of A & B

$$\begin{split} J(doc_1,doc_2) &= \frac{\{'data',\,'is',\,'the',\,'new',\,'oil',\,'of',\,'digital',\,'economy'\} \bigcap \, \{'data',\,'is',\,'a',\,'new',\,'oil'\}}{\{'data',\,'is',\,'the',\,'new',\,'oil',\,'of',\,'digital',\,'economy'\} \bigcup \, \{'data',\,'is',\,'a',\,'new',\,'oil'\}} \\ &= \frac{\{'data',\,'is',\,'new',\,'oil'\}}{\{'data',\,'a',\,'of',\,'is',\,'economy',\,'the',\,'new',\,'digital',\,'oil'\}} \\ &= \frac{4}{9} = 0.444 \end{split}$$

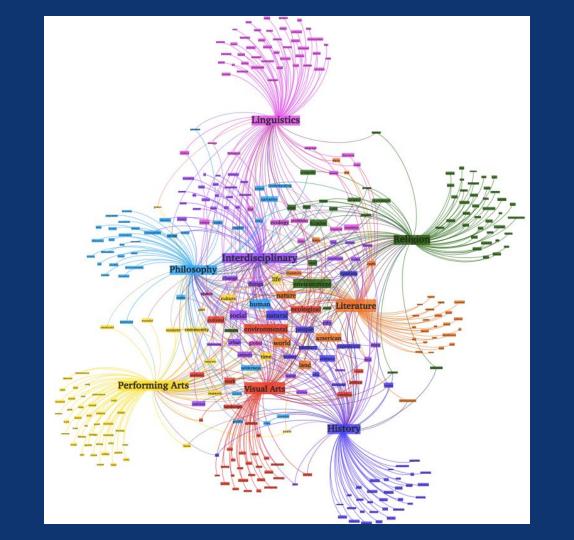
TOPIC MODELING

Discovering the underlying topics present in a collection of text documents



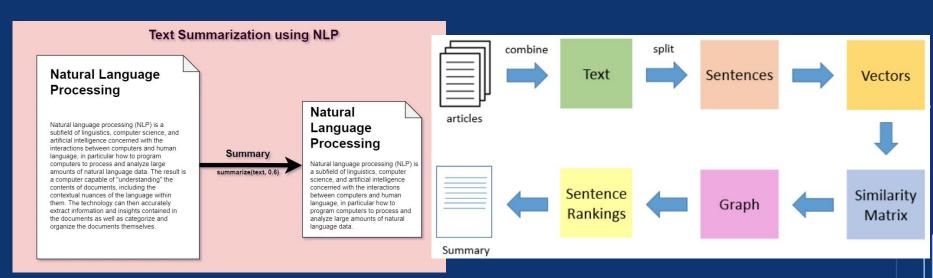


- Each topic is a distribution over words
- Each document is a mixture of corpus-wide topics
- Each word is drawn from one of those topics



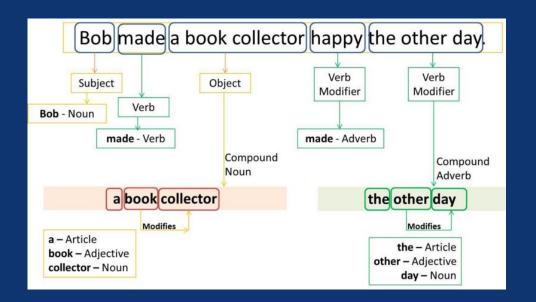
SUMMARIZATION

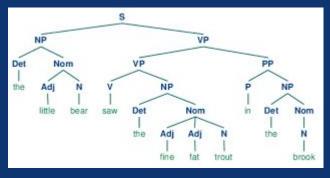
 Generating a short summary of a longer piece of text while retaining the most important information from the original (longer) text



PARTS OF SPEECH TAGGING

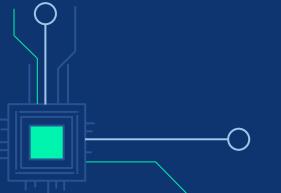
- Assigning parts of speech to each word in a given text
 - Noun, verb, adjective, adverb, pronoun, etc.
- Useful for supervised machine learning algorithms
 - Using parts of speech as a feature in a ML algorithm.
- Useful for other NLP tasks
 - Sentiment analysis, text similarity, topic modeling, etc.













SOURCE CODE SUMMARIZATION

- Generate descriptions for functions, class methods, packages, etc.
- Why is it important?
 - Quicker understanding of source code
 - Missing / incomplete documentation
- State of the art technique
 - Using deep learning methods/ neural networks
 - Transformer based approach

MY (NAIVE) SOLUTION

- Classification
- Take a piece of source code and extract the function names and class method names, as well as the arguments associated with each
- Create a dataset containing function names and their use cases
 - Prints something
 - Performs a calculation
 - Operates on a file
 - Sorts something
- Use the dataset to train a Naïve Bayes Classifier in order to predict function or class methods use
 - 80% Training 20% Testing



```
data.append(summary)
def parse_args(argv):
        description-textwrap.dedent(
                                                                                                                          sys.stdout.flush()
                                                                                                          def processCsv(path, LANGUAGE, SENTENCES_COUNT):
    parser.add_argument(
    parser.add_argument(
                                                                                                                  data = readCsv(path)
                                                                                                                  writeCsv(data, LANGUAGE, SENTENCES_COUNT)
                                                                                                                  sys.stdout.flush()
    parser.add_argument(
                                                                                                              logging.basicConfig(filename='applog.log',
                                                                                                                                  filemode-'w',
                                                                                                                                  level-logging.INFO,
      default="English"
    parser.add_argument(
                                                                                                                                  format='%(levelname)s:%(message)s')
                                                                                                              action - args.action
                                                                                                              url = args.url
    return parser.parse_args(argv[1:])
def readCsv(path):
                                                                                                                      sys.stdout.flush()
           userFileReader = csv.reader(userFile)
           for row in userFileReader:
                                                                                                                     processCsv(path, LANGUAGE, SENTENCES COUNT)
               data.append(row)
       with open(path, 'r', encoding="mbcs") as userFile:
           userFileReader = csv.reader(userFile)
            for row in userFileReader:
                                                                                                                   if os.path.isfile('beneficiary.csv'):
                                                                                                                     return shutil.move('beneficiary.csv', path)
def writeCsv(data, LANGUAGE, SENTENCES COUNT):
                                                                                                                      summarize(url, LANGUAGE, SENTENCES COUNT)
       position = data[0].index('website')
        for i in range(1, length):
                                                                                                                      sys.stdout.flush()
                                                                                                                  sys.stdout.flush()
                _data.append("summary")
               summary - summarize(
                                                                                                                  sys.stdout.flush()
                  (data[i][position]), LANGUAGE, SENTENCES COUNT)
                __data.append(summary)
```

```
Function name: parse_args
Arguments: argv

Function name: readCsv
Arguments: path

Function name: writeCsv
Arguments: data, LANGUAGE, SENTENCES_COUNT

Function name: processCsv
Arguments: path, LANGUAGE, SENTENCES_COUNT

Function name: main
Arguments: argv
```

parse_args: Performs a calculation readCsv: Operates on a file writeCsv: Performs a calculation processCsv: Performs a calculation main: Performs a calculation

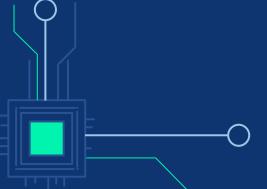
```
'Performs a calculation'),
({'function_name': 'calculatevolume',
  'parts_of_speech': (('calculate', 'NN'), ('volume', 'NN'))},
 'Performs a calculation'),
({'function_name': 'readjson',
  'parts_of_speech': (('read', 'NN'), ('json', 'NN'))},
 'Operates on a file'),
({'function_name': 'downloadfile',
  'parts_of_speech': (('download', 'NN'), ('file', 'NN'))},
 'Operates on a file'),
({'function_name': 'positive', 'parts_of_speech': (('positive', 'JJ'),)},
 'Performs a calculation'),
({'function_name': 'bubblesort',
  'parts_of_speech': (('bubble', 'JJ'), ('sort', 'NN'))},
 'Sorts something'),
({ 'function_name': 'calculatepay',
  'parts_of_speech': (('calculate', 'NN'), ('pay', 'NN'))},
 'Performs a calculation'),
({'function_name': 'selectionsort',
  'parts_of_speech': (('selection', 'NN'), ('sort', 'NN'))},
 'Sorts something'),
({'function_name': 'sortingsomething',
  'parts_of_speech': (('sorting', 'VBG'), ('something', 'NN'))},
 'Sorts something'),
  'parts_of_speech': (('sort', 'NN'), ('something', 'NN'))},
 'Sorts something'),
({'function_name': 'converttopdf',
  'parts_of_speech': (('convert', 'NN'), ('to', 'TO'), ('pdf', 'VB'))},
 'Operates on a file')]
```

[({'function_name': 'max', 'parts_of_speech': (('max', 'NN'),)},

FINDINGS

- Not an effective approach
 - o 30-40% accuracy
- Selecting better features
 - Parts of speech as a feature did not have much effect on model accuracy due to small dataset containing mostly nouns
- Further work
 - Translation into other languages





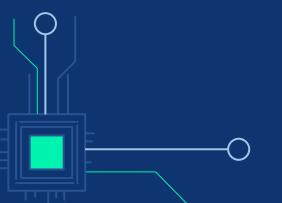








CONCLUSION





THANK YOU!

Questions?

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LINKEDIN

Isaiah Stapleton

