Text Extraction

FinTech

Lesson 12.3

Class Objectives

By the end of the class, you will be able to:

- Understand spaCy capabilities and where to find documentation.
- Be able to use POS tagged text to extract specific words.
- Use dependency parsed text to extract descriptors.
- Extract specific types of entities from text.
- Correlate text features to real-world series like stock prices.
- Create a dashboard from NLP sentiment features.

spaCy

spaCy

spaCy

- Core functions depend on language models learned from tagged text
- Fast and flexible
- Designed specifically for production use

NLTK

- Core functions depend on language models learned from programmed rules
- Accurate
- Intended for educational and prototyping purposes

spaCy

We will be using spaCy for:

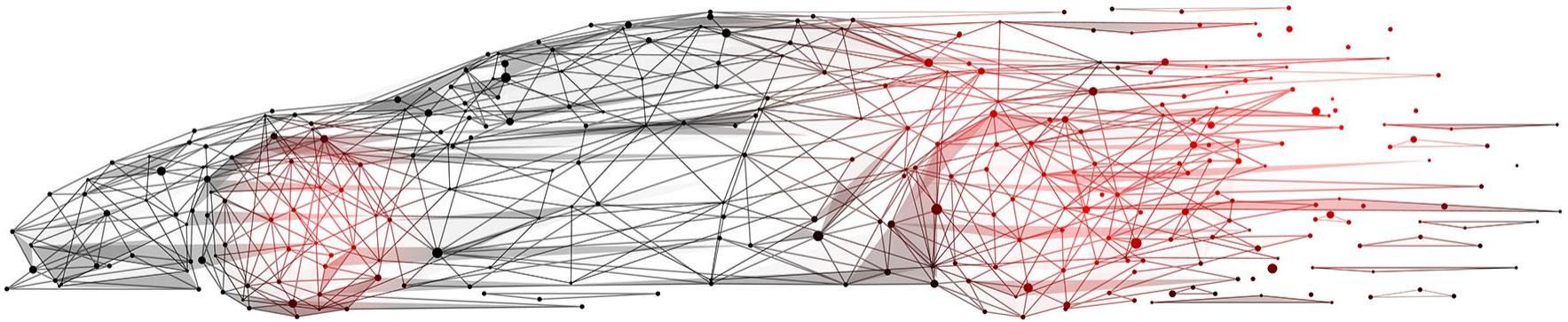
- Part of speech tagging
- Named entity recognition
- Dependency parsing

These tasks are more suitable for **model-based solutions** because they are complex and depend highly on context.

spaCy also provides tools for tasks like **tokenization** and **lemmatization**, which we've already learned with NLTK, and creating word vectors.

spaCy

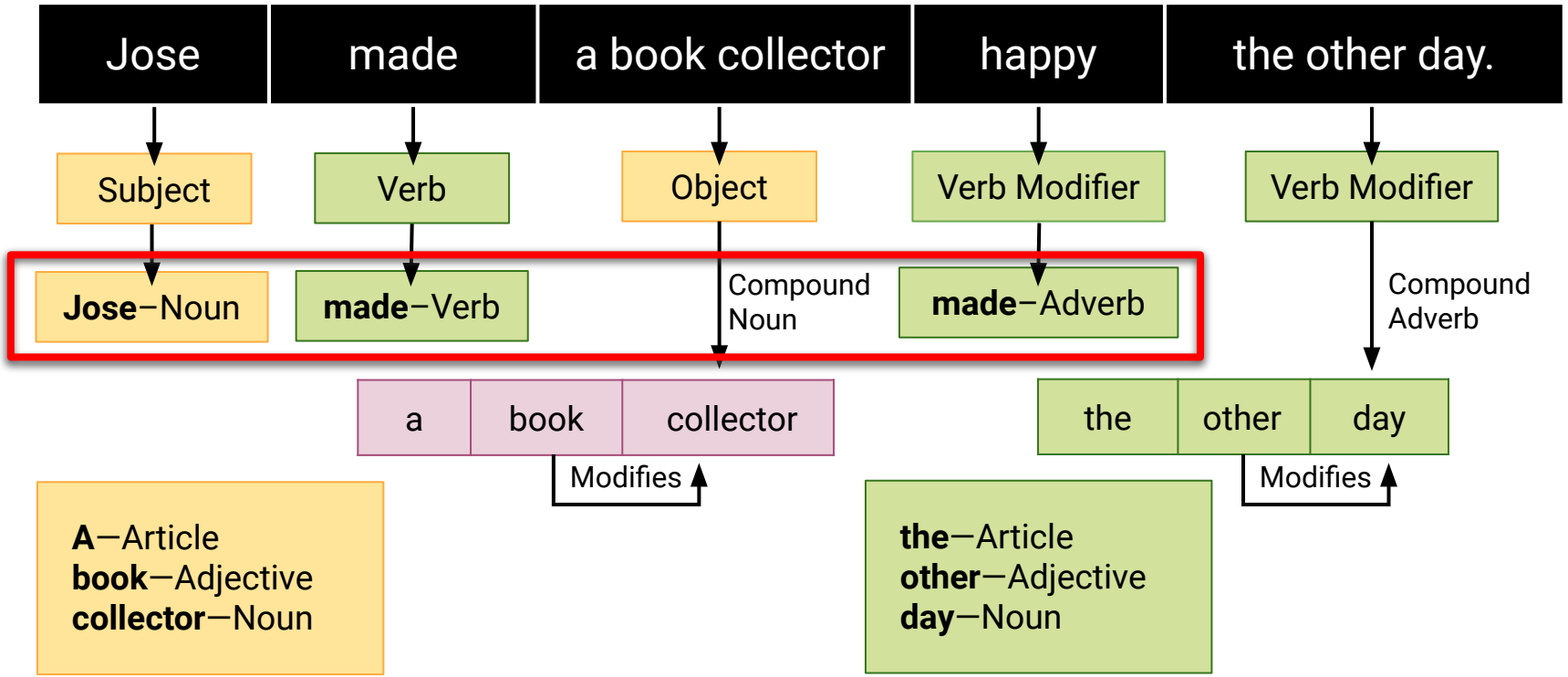
In comparison to NLTK, spaCy's language models trades off accuracy for speed, so if the corpus is large then you may prefer a simpler, rule-based solution.



POS Tagging and Dependency Parsing

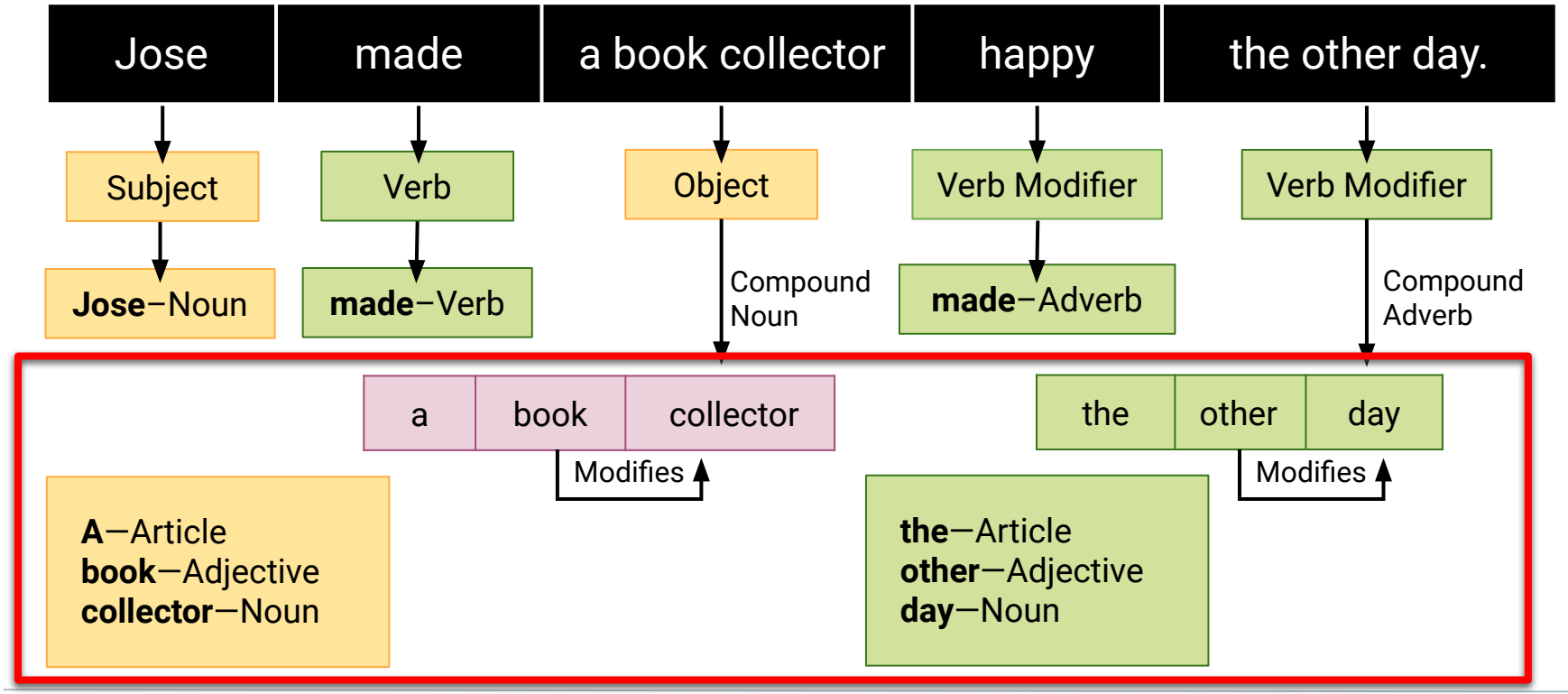
Part-of-Speech (POS) Tagging

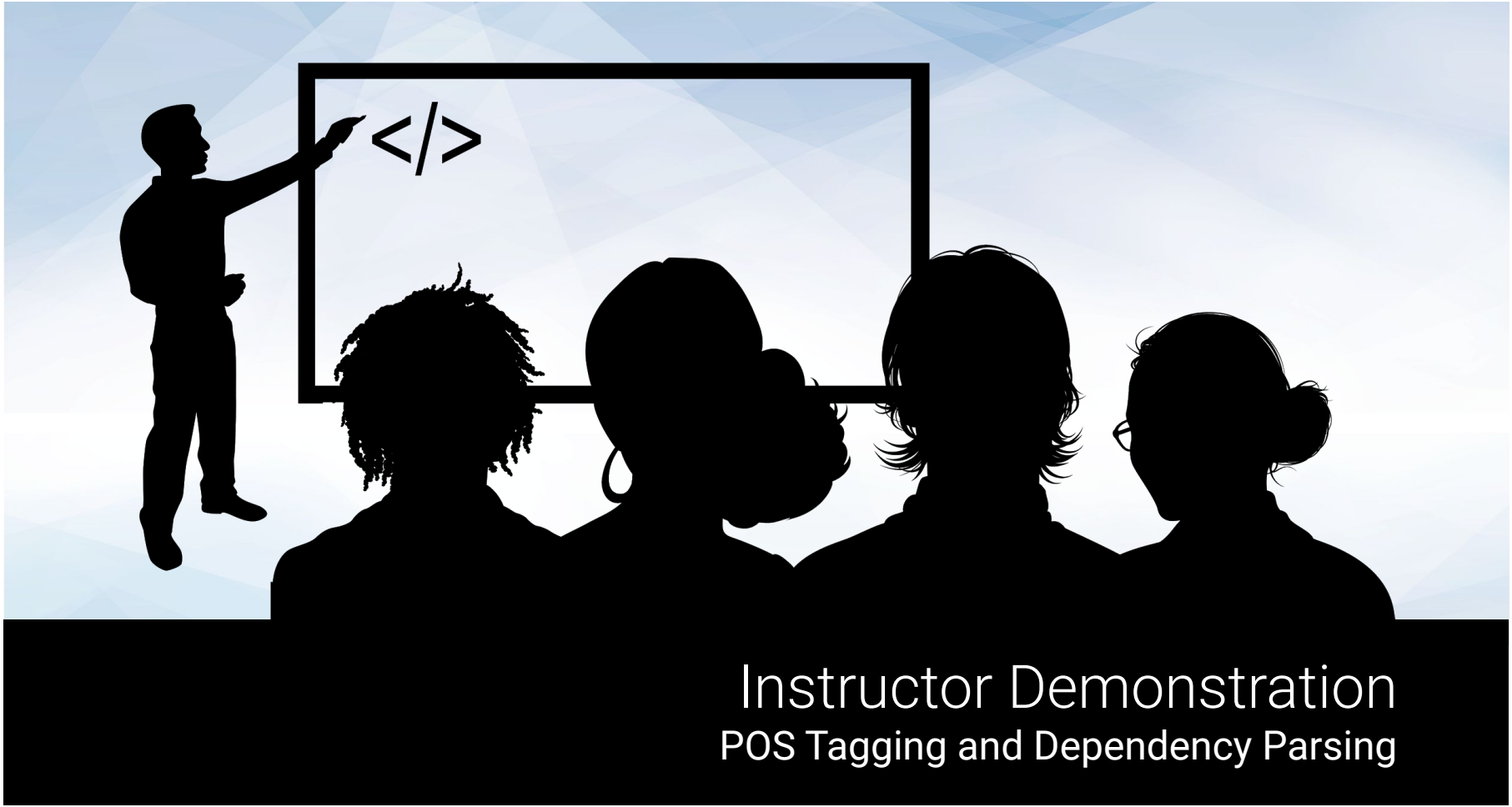
POS tagging categorizes each word in a sentence by its grammatical role.



Part-of-Speech (POS) Tagging

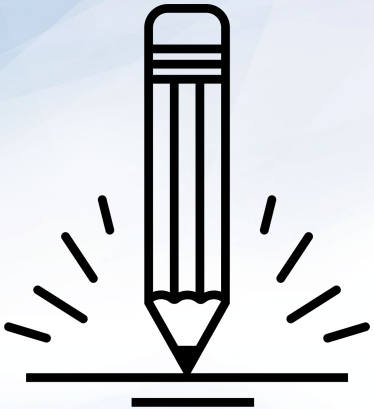
POS tagging categorizes each word in a sentence by its grammatical role.





Instructor Demonstration

POS Tagging and Dependency Parsing



Activity: Describing America

In this activity you will use the inaugural address corpus from NLTK and spacy's parsing and tagging modules to analyze the text that presidents have used to describe America.

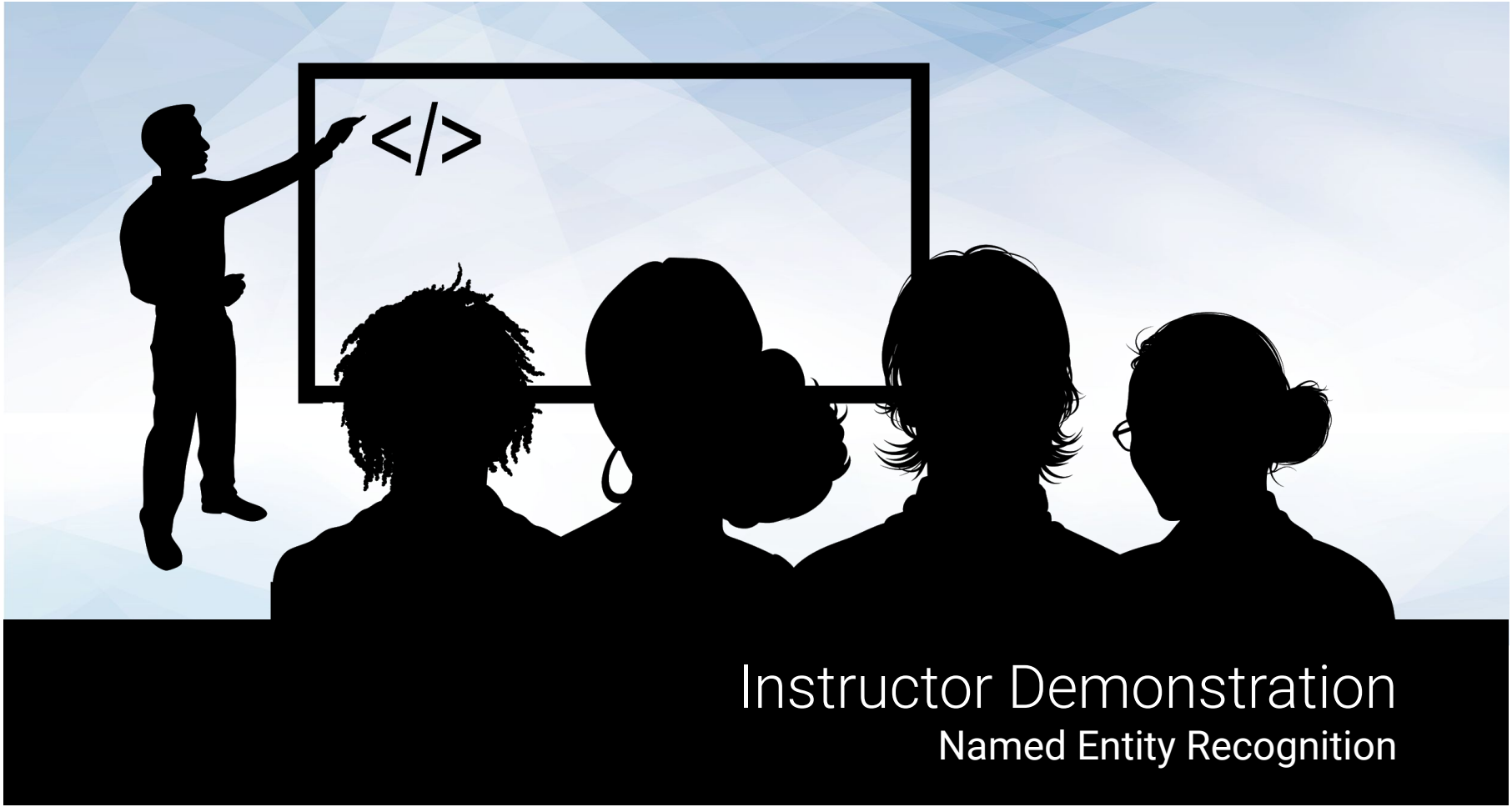
Suggested Time:
15 minutes

Time’s Up! Let’s Review.

Named Entity Recognition

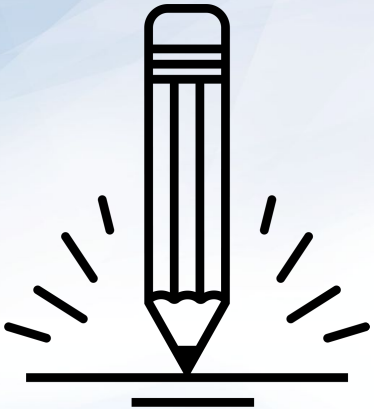
Extracting named entities, which include proper nouns and other specific types of nouns such as currencies, from a text.

US GPE unveils world's most powerful supercomputer, beats China GPE. The US GPE has unveiled the world's most powerful supercomputer called 'Summit', beating the previous record-holder China GPE 's Sunway TaighuLight ORG. With a peak performance of 200,000 CARDINAL trillion calculations per Second ORDINAL, it is over twice as fast as Sunway TaighuLight ORG, which is capable of 93,000 CARDINAL Trillion calculations per second. Summit has 4,608 CARDINAL servers, which reportedly take up the size of two CARDINAL tennis courts.



Instructor Demonstration

Named Entity Recognition



Activity: NER Clouds

In this activity you will extract named entities of your choosing from the Reuters corpus and build a word cloud from these entities.

Suggested Time:
15 minutes

Time’s Up! Let’s Review.


Break

Text as Feature

Tools and Techniques

Tools and techniques used to create numerical features (structured data) from text (unstructured data):

Tools

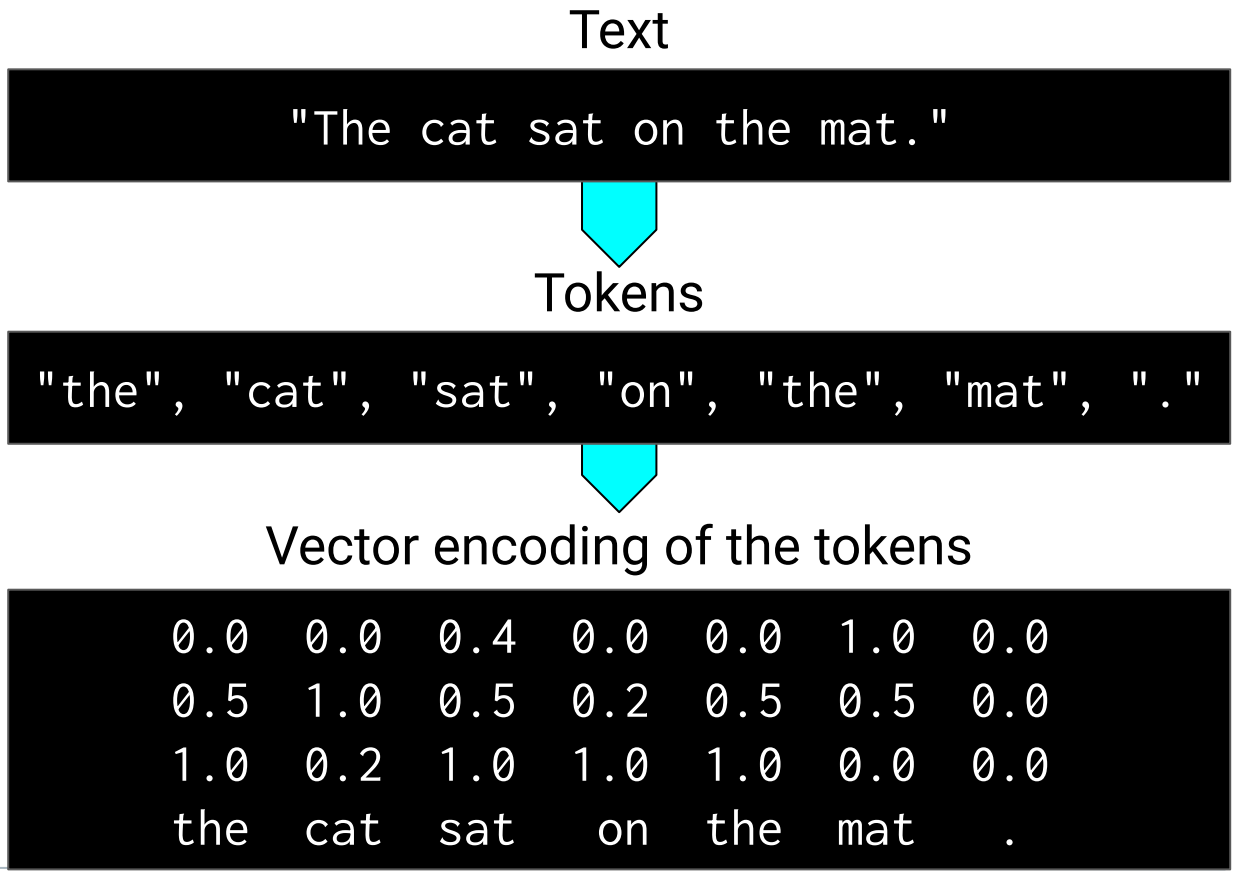
- NLTK
- wordcloud
- spaCy

Techniques

- preprocessing
- tokenizing text
- lemmatizing text
- aggregating word counts
- creating n-grams
- normalizing to tf-idf weights
- sentiment analysis
- parsing and pos-tagging text
- named entity recognition

Text as Feature

In order to use this data for classification or prediction, we need to make them features—numerical representations of unstructured text.



What are some examples of features that can be created from text documents?

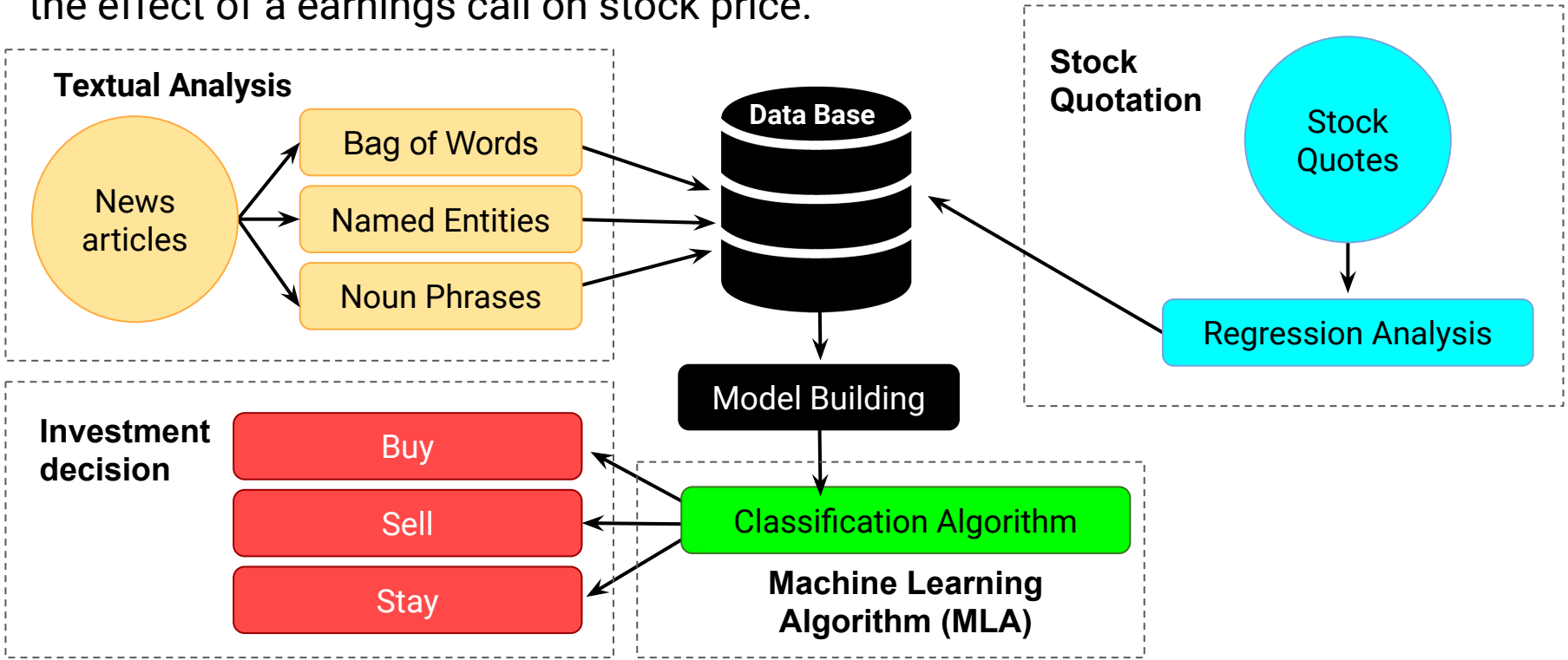
Text as Feature

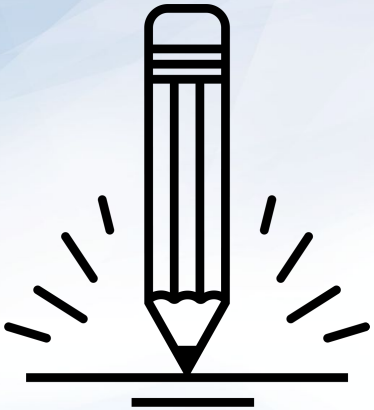
Examples:

- count of a keyword
- count of named entities
- sentiment scores
- percent of words that are adjectives
- total tf-idf score

Text as Feature

These features can then be used to classify a document to a category or predict the effect of a earnings call on stock price.





Activity: Correlating Returns

In this activity you will create a sentiment index from newsapi headlines and correlated it to S&P 500 daily returns, looking for text topic that generates the highest correlation.

Suggested Time:
15 minutes

Time’s Up! Let’s Review.

Machine Learning Review

Questions?