Introduction to Natural Language Processing

Learning Objectives

- Discuss the major tasks involved with natural language processing.
- Discuss, on a low level, the components of natural language processing.
- Identify why natural language processing is difficult.
- Demonstrate text classification.
- Demonstrate common text preprocessing techniques

What is Natural Language Processing?

- Using computers to process (analyze, understand, generate) natural human languages.
- Making sense of human knowledge stored as unstructured text.
- Building probabilistic models using data about a language.

What are some of the higher level task areas?





Information Extraction

Subject: curriculum meeting

Date: January 15, 2012

Event: Curriculum mtg

Date: Jan-16-2012

Start: 10:00am

End: 11:30am

To: Dan Jura Where: Gates 159

Hi Dan, we've now scheduled the curriculum meeting.

It will be in Gates 159 tomorrow from 10:00-11:30.



-Chris

Create new Calendar entry



Information Extraction & Sentiment Analysis



Attributes: zoom affordability size and weight flash ease of use

has a manageral and hardyne. He women you had live weight from it ments to first this size. The distance to Prevent, in praces. I are privey that course with the Server castle to reporter wroning to 4 Mill year answer. Milly for my wife, it is a great until files tracing to water they care May Flavor service, Ter a high quality, book Earlings syman which is not to and other Edition formy the of it If They married at the hours. material residence for more as the deep our spray with a price while it because allertup in yet at veloping a 18th Both discover you have be atomic products. Grill marrie promote for many or con-A moral amon accomp displicative tests where it is a distally to Lon P you put word because the could have place about theory not brig age are incompanied if you on a vect to have photographic. The only integraphing to the 2 water same with the appropriate regions become as a soft. The to the present X Plus oursers deple made troops, it honor group, horse group, and tolker comgreat principle. I from your on same manifestion that it was different to of maris, the proprietary forces. Provey on Towarding a process and in such curries with the first and real wheel I will remain the Child Fee (Series) Responsible or seek an expension on air I was a subspicio recesso in this bicarregion II will are code increasing half. I would so for this year to provide and supplies of sometic flat by the indian on Arquain, commission and years after place I stuff had

Size and weight



nice and compact to carry!



since the camera is small and light, I around those heavy, bulky professio



 the camera feels flimsy, is plastic and very light in weight you have to be very delicate in the handling of this camera



Machine Translation

Fully automatic

Enter Source Text:

这不过是一个时间的问题.

Translation from Stanford's Phrasal:

This is only a matter of time.

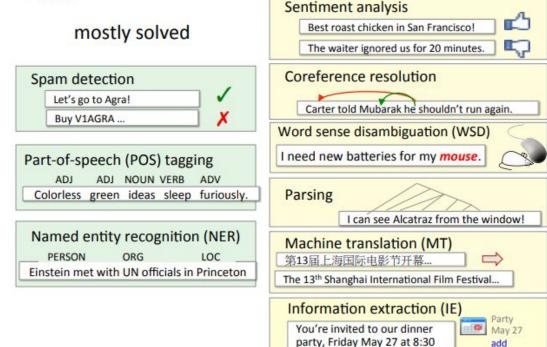
Helping human translators

عائية تحولت لي ب# +ها ذا الموضوع	تعرض الرئيس اللبنائي اميل لحود ل# حملة عنيقة في مجلس النواب الذي العقد امس في جلسة تشريعية ع الي " محاكمة " ل# رئيس الجمهورية على موقف +ه من المحكمة الدولية و " الملاحظات " التي ادا . حول هذا
Translate	inslation:
ebanese	president
	suffered
	exposed
	president emile
	before
	presented
Donel	offer

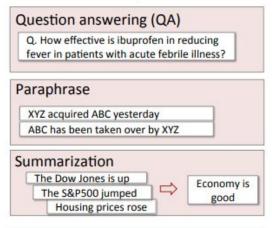


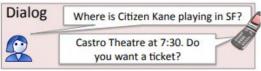
Language Technology

making good progress



still really hard





What are some of the Lower level Components?

What are some of the Lower level Components?

- **Tokenization:** Breaking text into tokens (words, sentences, n-grams)
- Stop-word removal: a/an/the
- Stemming and lemmatization: root word
- TF-IDF: word importance
- Part-of-speech tagging: noun/verb/adjective
- Named entity recognition: person/organization/location
- Spelling correction: "New Yrok City"
- Word sense disambiguation: "buy a mouse"
- Segmentation: "New York City subway"
- Language detection: "translate this page"
- Machine learning: specialized models that work well with text

Why is NLP hard?

- Ambiguity:
 - Hospitals Are Sued by 7 Foot Doctors
 - Juvenile Court to Try Shooting Defendant
 - Local High School Dropouts Cut in Half
- Non-standard English: text messages/ tweets
- Idioms: "throw in the towel"
- Newly coined words: "retweet"
- Tricky entity names: "Where is A Bug's Life playing?"
- World knowledge: "Mary and Sue are sisters", "Mary and Sue are mothers"

NLP Terms

Corpus: A collection of documents (or words)

Corpora: Plural of corpus

Bag-of-words: All possible words in the corpus

Text Vectorization: Converting all text in a corpus into numerical values

Countvectorizer: Converts each document into a set of words and their

counts

Text Classification

- Predicting a category or topic from a text sample
 - Sentiment Analysis e.g. Positive or negative sentiment?
 - Category classification e.g. Sports or Business Story?
 - Rating
- Words are used as the features
- Numeric value is given to each word which could be the number of times they appear in a document
- Text is vectorized and referred to as bag-of-words

Dataset: Yelp Reviews

Countvectorizer

Doc 1: The quick brown fox jumped over the lazy dog

Doc 2: The lazy dog could not outrun the fox

INDEX

	0	1	2	3	4	5	6	7	8	9
1	the	quick	brown	fox	jumped	over	lazy	dog	could	not
Doc 1	2	1	1	1	1	1	1	1	0	0
Doc 2	2	0	0	1	0	0	1	1	1	1

VOCABULARIES



Sparse Matrix

A matrix which contains very few non-zero elements

Sparse Matrix

[1.1	0	0	0	0	0	0.5	1
0	1.9	0	0	0	0	0.5	
0	0	2.6	0	0	0	0.5	
0	0	7.8	0.6	0	0	0	
0	0	0	1.5	2.7	0	0	ı
1.6	0	0	0	0.4	0	0	ı
0	0	0	0	0	0.9	1.7	

ComputerHope.com

	the	quick	brown	fox	jumped	over	lazy	dog	could	not
Doc 1	0	0	0	1	1	0	1	0	0	0
Doc 2	1	0	0	0	0	1	0	0	1	0
Doc 3	0	1	0	0	1	0	0	0	0	0

- Vectorizing text produces a sparse matrix
- A sparse matrix can be converted to the full form by calling .toarray() on the object





N-Grams

Features which consist of N consecutive words

Text	My cat is awesome
1-gram	'My', 'cat', 'is', 'awesome'
2-gram	'My cat', 'cat is', 'is awesome'
3-gram	'My cat is', 'cat is awesome',

Ngram_range: the upper and lower boundary of ngrams

EX: How many features do we get from the above examples with ngram_range=(1,3)



Stop Words

- Stop words are some of the most common words in a language
- They are used so that a sentence makes sense grammatically, such as prepositions and determiners, e.g., "to," "the," "and."
- they are so commonly used that they are generally worthless for predicting the class of a document
- They contribute noise to our model

Example:

Original Sentence: "The dog jumped over the fence"

After stop-word removal: "dog jumped over fence"



TextBlob

 provides a simple API for diving into common natural language processing (NLP) tasks such as part-of-speech tagging, noun phrase extraction, sentiment analysis, classification, translation, and more.



Stemming and Lemmatization

Stemming

- Reducing a word to its base form.
- Removes common ending such as 'ly', 'ing', 's', 'es', 'ed'
- It helps in reducing the number of features

Lemmatization

- A more refined process that uses specific language and grammar rules to derive the root of a word
- It can be better than stemming e.g 'best' to 'good', 'better' to 'good'

Stemming and Lemmatization

Lemmatization	Stemming
shouted → shout	badly → bad
best → good	computing → comput
better → good	computed → comput
good → good	wipes → wip
wiping → wipe	wiped → wip
hidden → hide	wiping → wip



Term Frequency-Inverse Document Frequency (TF-IDF)

- TF-IDF computes the relative frequency with which a word appears in a document compared to frequency across all documents.
- It analyses the uniqueness of words between documents to find distinguishing characters.



Sentiment Analysis with TextBlob

- Understanding how positive or negative a review is.
- There are many ways in practice to compute a sentiment value. For example:
 - Have a list of "positive" words and a list of "negative" words and count how many occur in a document.
 - Train a classifier given many examples of "positive" documents and "negative" documents.
 - Use Generic models



Any Questions