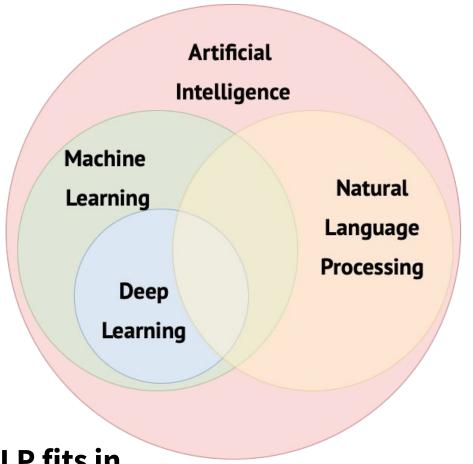
Introduction to Natural Language Processing & Computational Linguistics

Lecture 1: 21-479-0105 Computational Linguistics

Dr. Jeena Kleenankandy

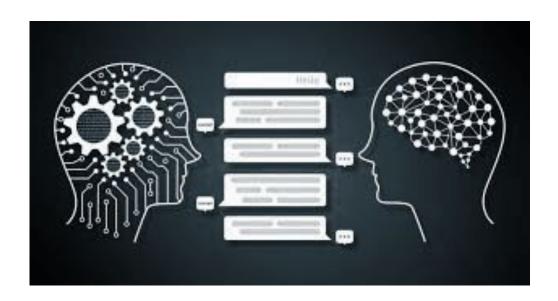
Agenda

- Introduction to NLP
- Key NLP tasks
- Application of NLP in various domains
 - Healthcare
 - E-commerce
 - Social media
 - Human Resource Management
 - Legal Domain
 - Finance/Banking
- Why NLP is challenging



Where NLP fits in..

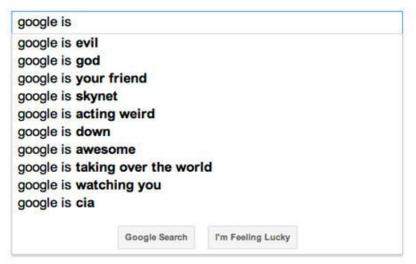
What is Natural Language Processing?

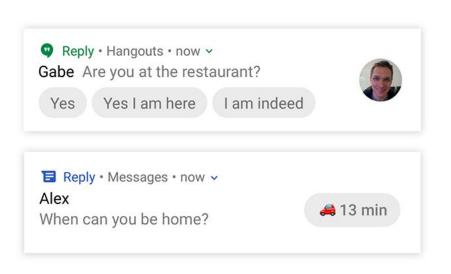


- Designing/building computational models to understand and generate human languages to get some useful task done

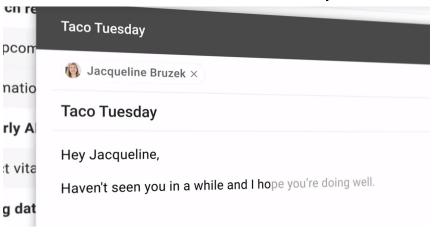
Google Search
Google Autocomplete
Personalised Ads
Video suggestions
Chat bots







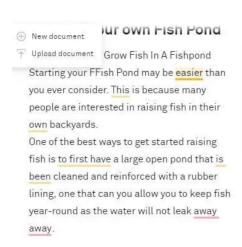
Gmail's Smart Compose



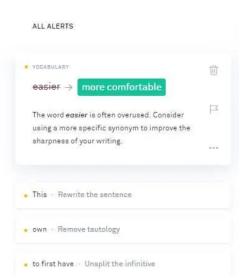
Spam filter

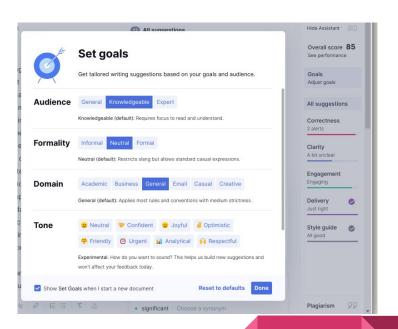


Google's Smart reply



Grammarly - writing assistant





Not just Grammar correction!





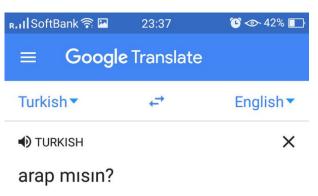


Personal assistants using voice commands

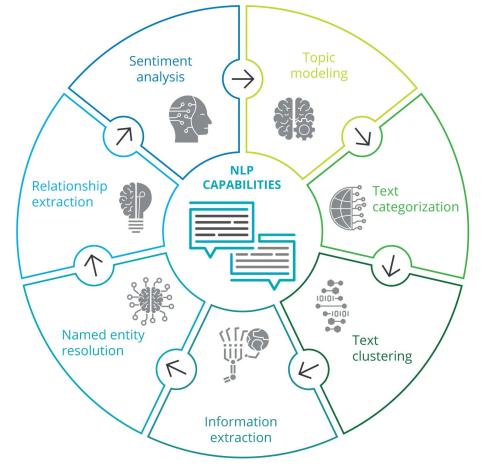
Speech Recognition

Language Translation

Converts text from one language to another







Key NLP tasks

Image courtesy:

https://www2.deloitte.com/us/en/insights/focus/cognitive-technologies/natural-language-processing-examples-in-government-data.html

NLP Applications

NLP in Healthcare

Named Entity Recognition (NER) and Relation Extraction (RE) in Electronic Health Records (EHR):

Disease Risk Prediction

- facilitate earlier detection of diseases and potentially improving patient outcomes.

Personalized Healthcare

- suggesting personalized treatment plans
- Chatbot therapist helping people with anxiety and other disorders.

Disease Evolution Prediction

Drug Reaction Detection

NLP in ecommerce & digital marketing

Sentiment Analysis, Machine Translation and Language understanding in:

- Understanding User Intent
- Semantics for Search Engine Experience
- Autocorrect and Autocomplete
- Virtual Assistants & Chat-based Product Recommendations
- Target Advertising
- Customer Review Analysis
- Translation for Global Reach

NLP and Social media content

Identifying fake news and hate speech in posts involves

- Stance detection
- Automatic summarization
- Fact checking
- Sentiment Analysis

Stock market prediction

- Tracking news, reports, comments about possible mergers between companies
- Sentiment Analysis on Twitter & other social media platforms

Motivations, Methods and Metrics of Misinformation Detection: An NLP Perspective, Qi Su, Mingyu Wan Xiaoqian Liu, Chu-Ren Huang

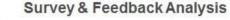
NLP and Human Resource Management

Recruitment



- Classifying & ranking
- Identifying personal traits
- · Identifying gaps in record
- Identifying fraud
- Deep information extraction
- Removing human biasness







- Identifying friction areas
- Human emotions
- 360 feedback analysis
- Survey analysis





- Improving Competencies
- Boosting SMART Goals
- Approval rating
- Comments analysis
- 360 feedback analysis





- · Identifying potential
- · Identifying training needs
- Matching fitments
- Designing succession
- · Conflict resolution



Social Media Analysis



- · Identifying potential/ talent
- Identifying competence and interest areas
- Behavior trends







NLP in Legal domain

Legal Judgment Prediction

Similar Case Matching

Legal Question Answering

Legal Document Summarization

Reviewing Legal contracts

Source: How Does NLP Benefit Legal System: A Summary of Legal Artificial Intelligence Haoxi Zhong, Chaojun Xiao, , Cunchao Tu,, Tianyang Zhang, Zhiyuan Liu1, Maosong Sun

NLP in Finance

Credit Scoring

- assess creditworthiness of borrowers digital footprints across social media, browsing history, geo-location
- Fraud detection
- Claim Approval in Insurance

Customer Service

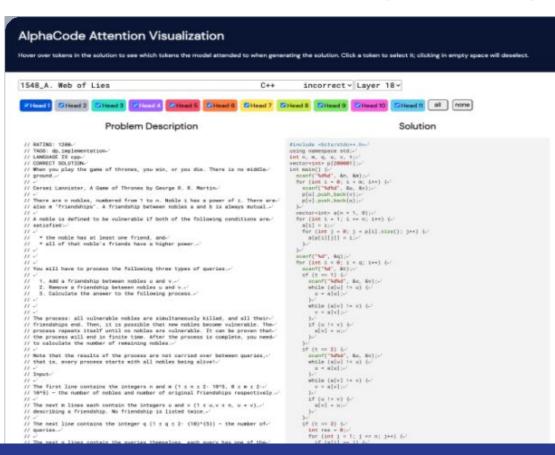
- Chatbots / virtual assistance
- savings \$7.3 billion globally in operational cost in banking

Market Data Collection

Stock market prediction

Analyse Banking contracts

NLP in Competitive programming



AlphaCode

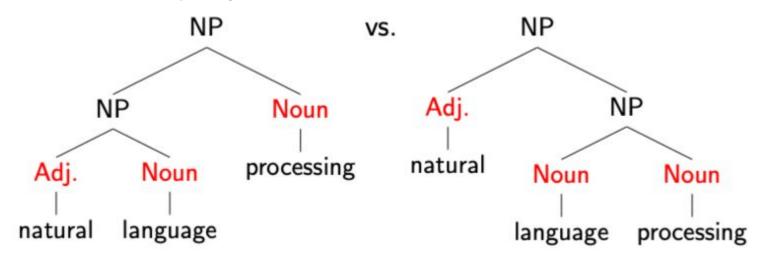
Why is NLP challenging?

- Language is Ambiguous
 - Lexical, Syntactic, Semantic, Pragmatic etc.
- Language has Irregularities like
 - o Sarcasm, idioms, metaphors and lot more...
- Language keeps evolving with variations across domains
- Linguistic communication is compressed
- Expressivity, unmodelled variables, unknown representations....

Even "words" are challenging

- Segmenting text into words
- Morphological variation
- Words with multiple meanings: bank, mean
- Domain-specific meanings: latex
- Multiword expressions: make a decision, take out, make up

Syntactic ambiguity



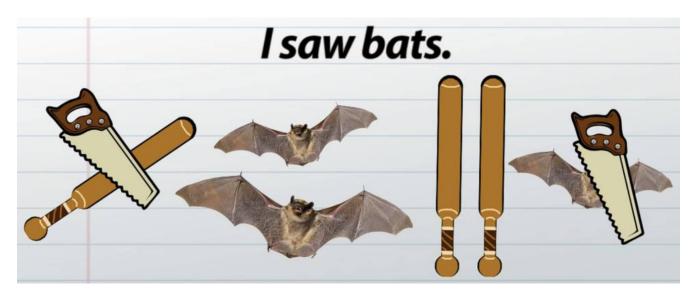


"I saw a girl with a telescope"



PP-attachment problem : An example of **Syntactic Ambiguity**

Semantic Ambiguity



Four different interpretation for a simple sentence!!

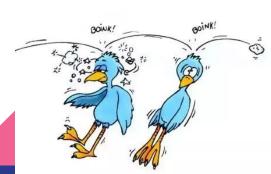
Irregularities.....

- → He has a heart of gold
- → Meet me at the bank.
- → Kill two birds with one stone.
- → It's a piece of cake
- → It's raining cats and dogs
- → She is a walking encyclopedia.
- → He broke my heart.
- → A great movie for a sunday nap!

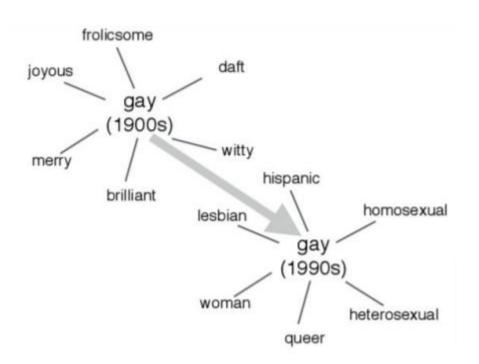








Language keeps evolving



Meaning changes

Language keeps evolving....

ikr smh he asked fir yo last name

New words are introduced

so he can add u on fb lololol

Language keeps evolving....

```
ikr smh he asked fir yo last name
```

so he can add u on fb lololol

Linguistic communication is compressed

We all know what we all know, but machine don't!!
 (Missing Text Phenomenon)

```
plastic cup
plastic factory
coffee cup
coffee machine
computer store
neighborhood store
etc.
```

```
is a [cup made of plastic]
is a [factory that produces plastic]
is a [cup to hold coffee]
is a [machine to make coffee]
is a [store that sells computers]
is a [store in the neighborhood]
etc.
```

The Missing Text Phenomenon, Again: the case of Compound Nominals

Expressivity

- Not only can one form have different meanings (ambiguity) but the same meaning can be expressed with different forms:
 - She gave the book to Tom vs. She gave Tom the book
 - Some kids popped by vs. A few children visited
 - Is that window still open? vs. Please close the window

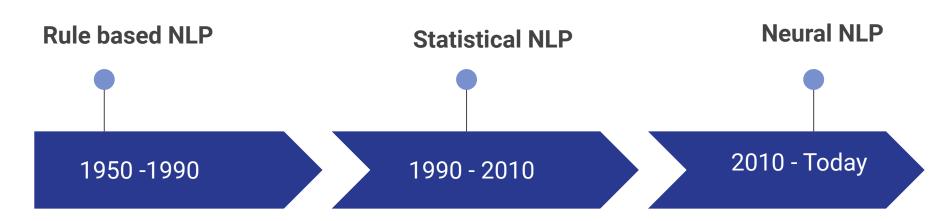
Unmodeled Variables

World knowledge

I dropped the glass on the floor and it broke

I dropped the hammer on the glass and it broke

Evolution of NLP over the years



- Uses a well-defined set of rules
- Not scalable
- Need domain expertise

- Uses statistics to learn rules
- Relies on hand-crafted features
- Need domain expertise

- Uses artificial neural networks
- Extracts features from raw data
- No need of domain expertise

Rule - Based NLP System

Think of some useful rules to filter spam emails ??

Rule - Based NLP System

Think of some useful rules to filter spam emails ??

R1: misspelling in company names

R2: overuse of words like "lottery", "dollars"

R3: Unnecessary urgency

R4: Inappropriate languages

R5:...

Relies on probability and statistics to learn rules or pattern from data
 Eg:

"...won 5 million dollars...bank account......click"



What is the *probability of finding these words* in a spam vs genuine email?

Check whether P(spam|email) > P (genuine|email) ?

Auto completion can be done the same way!!!

I love chocolate _____

What of these words is most probable to occur next in the above sentence?

(Pizza , dog , van, Cake , drink , pencil)

P(Cake|chocolate) > P(drink |chocolate) > P(pizza |chocolate)

Auto completion can be done the same way!!!

I love chocolate _____

What of these words is most probable to occur next in the above sentence?

(Pizza , dog , van, Cake , drink , pencil)

P(Cake|chocolate) > P(drink |chocolate) > P(pizza |chocolate)

You shall know a word by the company it keeps (Firth, J. R. 1957:11).

A bottle of tezgüino is on the table.

Everyone likes tezgüino.

Tezgüino makes you drunk.

We make tezgüino out of corn.

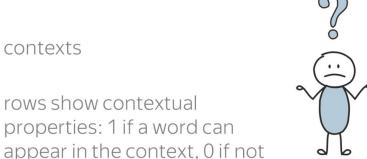
Can you understand what tezgüino means?



You shall know a word by the company it keeps (Firth, J. R. 1957:11).

- (1) A bottle of _____ is on the table.
- (2) Everyone likes _____.
- (3) ____ makes you drunk.
- (4) We make out of corn.

rows show contextual properties: 1 if a word can



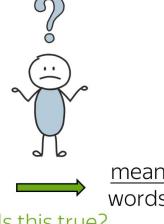
What other words fit

into these contexts?

You shall know a word by the company it keeps (Firth, J. R. 1957:11).

- (1) A bottle of _____ is on the table.
- (2) Everyone likes _____.
- (3) _____ makes you drunk.
- (4) We make _____ out of corn.

(1)	(2)	(3)	(4)	
1	1	1	1	
0	0	0	0	
1	0	0	1	
0	1	0	1	
1	1	1	0	
	1 0 1	1 1 0 0 1 0	1 1 1 0 0 0 1 0 0	1 0 0 1



meanings of the words are similar

Is this true?

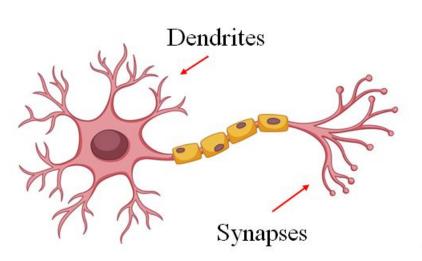
rows are

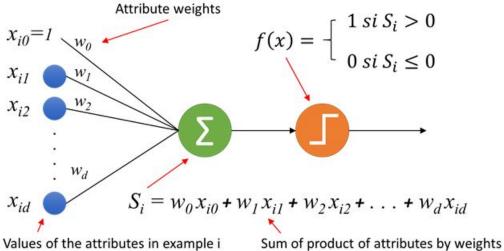
similar

Neural NLP System

Artificial Neural Networks

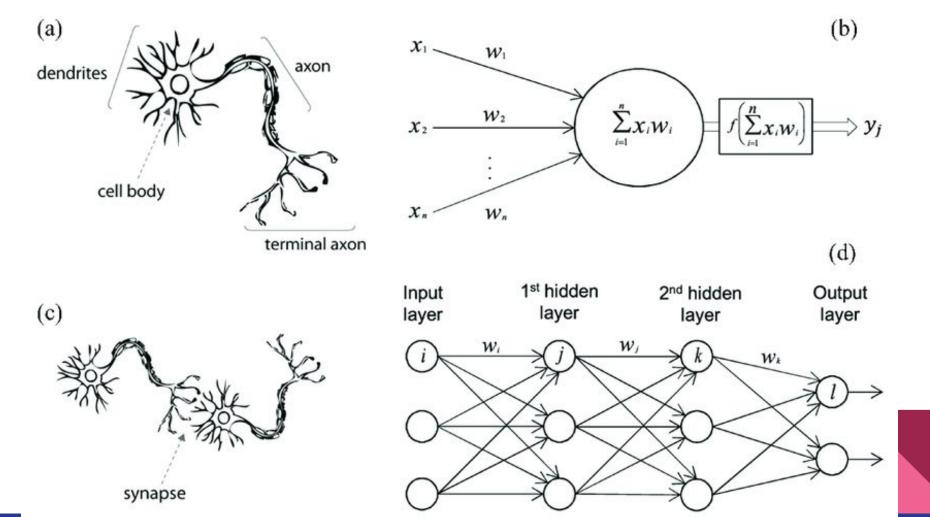
Inspired by neurons in human brain





NEURON

PERCEPTRON



Classic NLU problems (still very relevant)

Sentiment Analysis	Is the movie review positive, negative, or neutral?	"The movie is funny, smart, visually inventive, and most of all, alive." = .93056 (Very Positive)
Paraphrase Identification	Is the sentence B a paraphrase of sentence A?	A) "Yesterday, Taiwan reported 35 new infections, bringing the total number of cases to 418." B) "The island reported another 35 probable cases yesterday, taking its total to 418." = A Paraphrase
Similarity scoring	How similar are sentences A and B?	A) "Elephants are walking down a trail." B) "A herd of elephants are walking along a trail." = 4.6 (Very Similar)
Duplicate question	Are the two questions similar?	A) "How can I increase the speed of my internet connection while using a VPN?" B) "How can Internet speed be increased by hacking through DNS?" = Not Similar
Language Inference	Does sentence A entail or contradict sentence B?	A) "Tourist Information offices can be very helpful." B) "Tourist Information offices are never of any help." = Contradiction
Question Answering	Does sentence B contain the answer to the question in sentence A?	A) "What is essential for the mating of the elements that create radio waves?" B) "Antennas are required by any radio receiver or transmitter to couple its electrical connection to the electromagnetic field." = Answerable
Recognizing Textual Entailment	Does sentence A entail sentence B?	A) "In 2003, Yunus brought the microcredit revolution to the streets of Bangladesh to support more than 50,000 beggars, whom the Grameen Bank respectfully calls Struggling Members." B) "Yunus supported more than 50,000 Struggling Members." = Entailed
Coreference Resolution	Sentence B replaces sentence A's ambiguous pronoun with one of the nouns - is this the correct noun?	A) "Lily spoke to Donna, breaking her concentration." B) "Lily spoke to Donna, breaking Lily's concentration." = Incorrect Referent

Research/Project Ideas in Core NLP

Reading Comprehension **Visual Question Answering** Dialogue System **Event Extraction Emotion Recognition** Semantic Parsing Relational Reasoning Abuse Detection Stance Detection

Hate Speech Detection Fake News Detection Language Identification **Code Generation Bias Detection** Intent Detection **Authorship Verification** Clickbait Detection

About the Course:

Course Objectives:

CO1: Understand the fundamentals of written language processing

CO2: Applying theses fundamentals in real world problems like POS tagging, Corpus development, WordNet, Dialogue processing, document retrieval, Machine translation etc etc

CO3: Creating resources for less resource languages

CO4: Case study of various typical Language processing tools.

Thank You!!