

# Introduction to Natural Language Processing

Unit 1: Introduction to NLP

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July 18, 2025

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# Introduction

# What is Language?

Definition by Britannica:

Language is a **system of conventional spoken, manual (signed), or written symbols** by means of which human beings, as members of a social group and participants in its culture, express themselves.

The function of language include - communication, the expression of identity, play, imaginative expression, and emotional release

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- We think (partly) with language
- We tell stories in language
- We build **Scientific Theories** with language
- We make friends/build relationships

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# What is Language? - another perspective

- **Language:** A system of symbols and rules used for communication.
- **Types of Language:**
  - Natural Language (e.g., English, Nepali)
  - Programming Language (e.g., Python, C++)

What make Natural Language different over programming language?

- **Characteristics of Natural Language:**
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  - Evolving and diverse

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It is a sub-field of artificial intelligence that is concerned with the human computer interaction using natural language. It is **interdisciplinary** field Computer and Linguistics.

We can 1) **Analyze** and 2) **Produce** the natural language using NLP techniques

## Goals:

- Understand, interpret, and generate human language
- Automate language-related tasks

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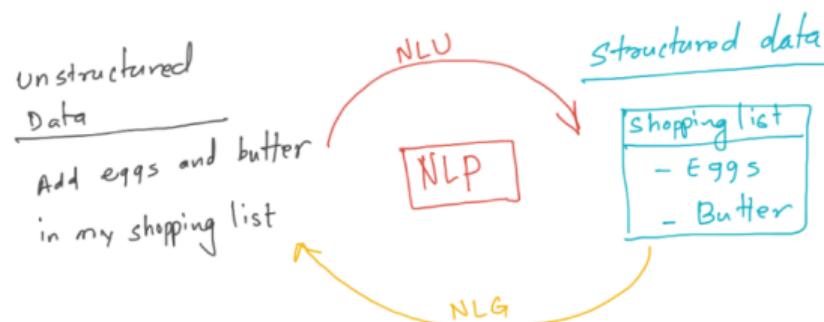
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# What is NLP? (cont.)

The NLP can be divided into two categories: Text and Speech Processing.

## Text Processing

- Machine translation
- Spam email detection
- Document classification
- Text summary generation
- Sentiment Analysis

## Speech Processing

- Text-to-speech Generation (Speech Synthesis)
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Access Knowledge

- search engine, recommend system

Communicate

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- Social Media: Facebook, Instagram, WeChat, Twitter
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# Brief History of NLP

Era	Key Developments
1950s-60s	Rule-based systems (ELIZA, symbolic AI)
1980s-90s	Statistical NLP (HMMs, n-grams)
2000s	Machine Learning (SVMs, CRFs)
2010s-Now	Deep Learning and Transformers (BERT, GPT)

- Transition from rules → statistics → deep learning
- Rise of large pre-trained language models

# Level of Linguistics

# Five level of linguistics

- Phonology
- Morphology
- Syntax
- Semantics
- Pragmatics (context)
- Extra-linguistic - other material along with language

Analysis in context	Extra-linguistic context		<i>Found him in the street inside a bag. I think he is happy with his new life</i> <small><a href="http://tiny.cc/meyarw">http://tiny.cc/meyarw/Found-him-in-the-street-inside-a-bag-I-think-he-is-happy-with-his-new-life</a></small>
	Linguistic context	<ul style="list-style-type: none"> <li>— You know what? <b>John</b> gave <b>Peter</b> a <b>Christmas present</b> yesterday</li> <li>— Wow, was <b>he</b> surprised? What was <b>it</b> like?</li> <li>— <b>Surprisingly good.</b> <b>He</b> spent quite a bit on <b>it</b>.</li> </ul>	
	Semantic level	<b>The landlord<sub>SPEAKER</sub></b> has not yet <b>REPLIED<sub>Communication_response</sub></b> in writing <sub>MEDIUM</sub> to the tenant <sub>ADRESSEE</sub> objecting the proposed alterations <sub>MESSAGE<sub>DNI<sub>TRIGGER</sub></sub></sub>	
	Syntactic level		<i>John saw a dog yesterday which was a Yorkshire Terrier</i>
	Morphological level	<i>brav+itude, bio+terror-isme/-iste, skype+(e)r</i> <i>mang-er-i-ons = MANGER+cond+1pl</i>	
	Phonological level	<small>International Phonetic Alphabet</small> [ai pʰi: ei]	Graphemic level <i>enough, cough, draught, although, brought, through, thorough, hiccough</i>

# Phonology

It is a study of the **sounds** of a language

Every language has its own **inventory of sounds** and logical rules for combining those sounds to create words.

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# Morphology

Study of the **internal structure of the words** of a language

There are many words to which a speaker can add a **suffix**, **prefix**, or **infix** to create a new word

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Study of **meaning** in language

Linguists attempt to identify not only how speakers of a language distinguish the meanings of words in their language, but also **how the logical rules speakers apply to determine the meaning of phrases, sentences, and entire paragraphs**

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Study of the **social use** of language

All speakers of a language use different registers, or different **conversational styles**, depending on the places

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# Challenges of NLP

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1. Productivity
  2. Ambiguous
  3. Variability
  4. Diversity
  5. Sparsity

# Productivity

**Definition:** “property of the language-system which enables native speakers to **construct and understand** an **indefinitely large number of utterances**, including utterances that they have never previously encountered.” (Lyons, 1977)

New words, senses, structure are introduced in languages all the time

Examples: social distance were added to the Oxford Dictionary in 2021

Why Sanskrit is not in common use?

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# Ambiguous

Most linguistic observations (speech, text) are open to several **interpretations**

How We (Humans) disambiguate?  
i.e. find the correct interpretation

- using all kind of signals
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Ambiguity can appear at all levels

phonology, graphemics, morphology, syntax, semantics

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# Semantic Ambiguity

## Polysemy:

eg. set, arm, head - Head of New-Zealand is woman **Name Entity**:

eg. Michael Jordan - Michael Jordan is a professor at Berkeley **Object/Color**:

eg. cherry - Your cherry coat

# Variability

Language varies at all levels

- Phonetics (accent)
- Morphological, Lexical (spelling)
- Syntactic
- Semantic

# Variation Determiners

## Who is talking?

- To Whom?
- Where? Work, Home, Restaurant
- When? 19th century, 2008, 2022...
- About what? Specialised domain, the Weather,...

Essentially, the Variability of a language depends on

- Social Context
- Geography
- Sociology
- Date
- Topic

# Diversity

About **7000 languages** spoken in the world. About **60% are found in the written form**.  
Diversity are in

- Phonologic Diversity
- Graphemic Diversity (latin, arabic, devanagari, greek)

## Syntactic Diversity

A key characteristics of the syntax of a given language is the word order

- Word order differs across languages
- Word order degree of freedom also differs across languages
- We characterize word orders with: Subject (S), Verb (V), Object (O) order

# Terminologies

# Tokenization

Splits longer strings of text into **smaller pieces, or tokens**

Larger chunks of text can be tokenized into sentences

Sentences can be tokenized into words, etc.

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# Normalization (Pre-processing)

Normalization generally refers to a series of related tasks

- converting all text to the same case (upper or lower)
- removing punctuation
- expanding contractions
- converting numbers to their word equivalents, and so on.

Normalization puts all words on equal footing, and allows processing to proceed uniformly.

# Stemming

Stemming is the process of **eliminating affixes**

- suffixed, prefixes, infixes, circumfixes

## Stemmer

Tool to obtain a word stem.

- Running – run
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Lemmatization is related to stemming, differing in that lemmatization is able to **capture canonical forms** based on a **word's lemma**.

For example, stemming the word "better" would fail to return its citation form (another word for lemma); however, lemmatization would result in the following:

better → good

Implementation of a stemmer would be the less difficult

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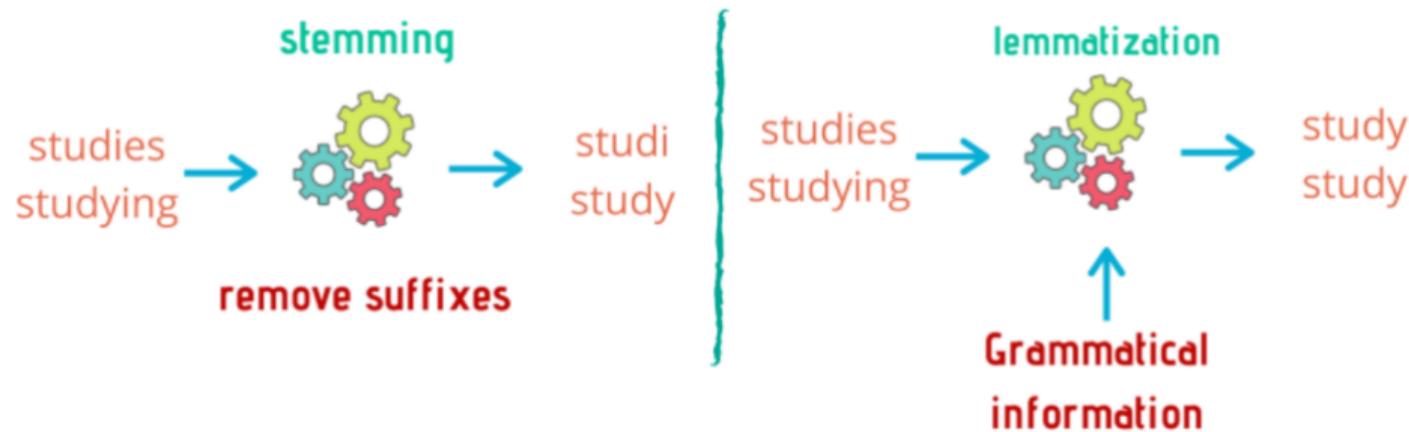
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# Stemming vs. Lemmatization



# Corpus / Corpora

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Sources: Single, Multiple, Multilingual

Corpora are generally solely used for statistical linguistic analysis and hypothesis testing.

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A, an, the, in , on etc.

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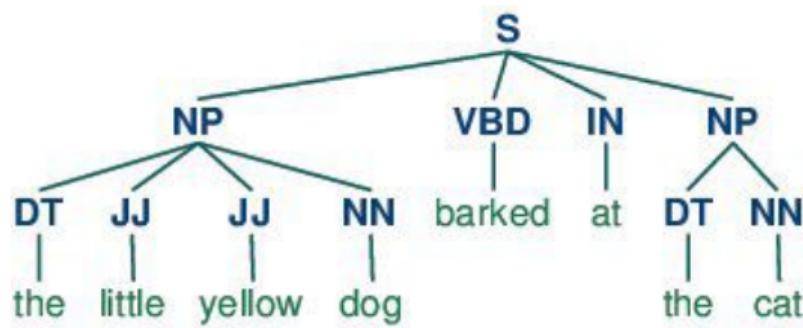
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# Bag of Words

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iThe bag of words model omits grammar and word order, but is interested in the number of occurrences of words within the text.

"Well, well, well," said John. "There, there," said James. "There, there."

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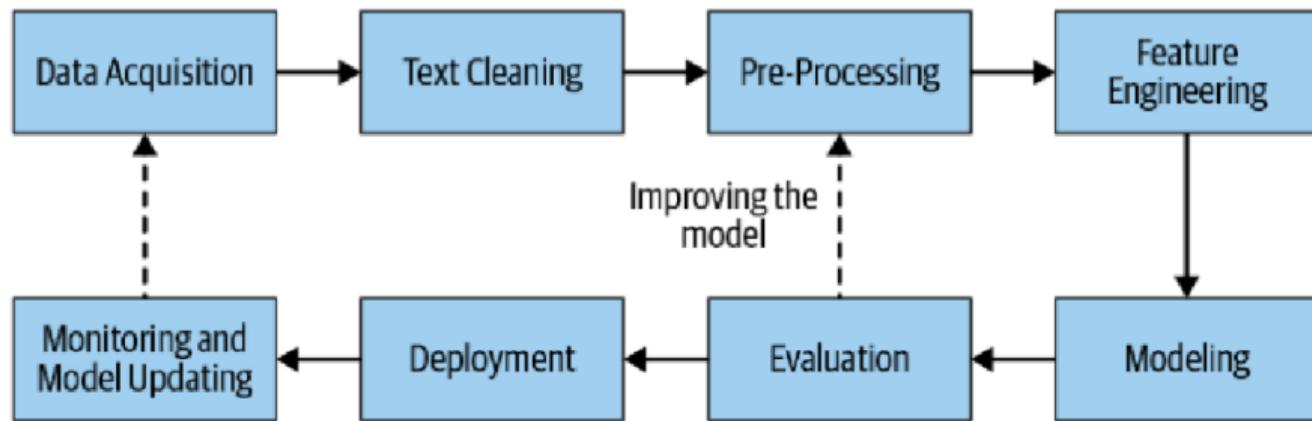
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# NLP Pipeline

# Standard NLP Pipeline



# Performing NLP research

Assume we have a Research, Engineering, Product Problem

- Define a **NLP System** to solve it. Split into modules, each one performing a task
- Define **Evaluation Metric(s)** for your system and sub-modules
- **Collect Data** to build/train your models
- Build **Baseline Models** (i.e. most simple model you can think of that have a non trivial performance metric)
- Build **Better Models** using **symbolic/statistical/DL** methods

# NLP problems

# Sentiment Analysis

**Definition:** Identify the emotional tone (e.g., positive, negative, neutral) in a piece of text.

**Example:**

- Input: "The movie was absolutely fantastic!"
- Output: Positive

$$f_{\text{sentiment}} : \mathcal{X} \rightarrow \mathcal{Y}, \quad \mathcal{Y} = \{\text{positive, negative, neutral}\}$$

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# Document Summarization

**Definition:** Automatically generate a short and coherent summary from a long document.

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- Input: "The government passed the new tax reform bill."
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- Input: "I love you."
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# Automatic Speech Recognition (ASR)

## Definition:

Convert spoken audio signals into written text. **Example:**

- Input: (audio) WAV file with "Hello, how are you?"
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# Questions?

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# Discussion