

Lecture 1: Introduction to NLP

Natural Language Processing or NLP is a ~~not~~ multidisciplinary field combining math, linguistics and Computer science, the goal is to get Computer to do useful things with natural language data.

Common NLP tasks and applications:

- Translation
- Summarization
- Question answering
- Speech recognition
- classification
- assisted writing

some reasons why NLP is hard:

1. Content
2. Inflections
3. Context
4. misspelled or misused words.
5. physical gesture.

Fundamentals of NLP

→ Preprocessing:

- tokenization
- simplification technique
- tagging and
- simple rules-based approaches.

→ Basic vectorization:

→ turning text to numbers.

→ measuring similarity between docs.

→ Modelling overview:

→ types of machine learning

→ algorithms

→ vs. models

→ evaluation

→ first steps into classification:

→ classifying text using Naive Bayes.

→ evaluation with precision and recall

→ Topic modeling

→ Automatically finding topics
in docs using α latent
Dirichlet allocation.

Deep learning For NLP:

- word vectors: capturing word meaning
the concept of embedding.
- Recurrent Neural Networks:
capturing sequence information
and generating language.
- Neural Networks:
what they are, how they work
and details around training
- Sequence2sequence
Seq2Seq and attention: training
a neural network to transform
one sequence to another.
- Transformation: The dominant mainstream
architecture today. Pretraining and
transfer learning

Processing:

1. tokenization: The process of segmenting our documents into tokens.

Code:

```
// inserting libraral libraries
```

```
() ! pip install -U spacy == 3.*
```

```
() ! python -m spacy: info
```

```
() import spacy
```

```
() ! python -m spacy download en-core-web-sm
```

```
() nlp = spacy.load('en-core-web-sm')
```

```
() type(nlp)
```

```
// sample sentence testing
```

```
s = "I eat nice"
```

```
doc = nlp(s)
```

```
Print([t.text for t in doc])
```

```
// output :
```

```
["I", "eat", "nice"]
```


Basic Programming

case folding, stop word removal, stemming
lemmatization

case folding

sentence: "Mr. Cook went into the
* kitchen to cook dinner"

without cf: with cf()

{ Cook, dinner, into, kitchen, mr, the, to, went, cook } { Cook, dinner, into, kitchen, mr, the, to, went }

Code:

```
Print([t.lower_ for t in doc])
```

// to skip for first word

```
Print([t.lower_ if not t.is_sent_start else  
t for t in doc])
```

Stop word removal:

stop words \rightarrow {the, a, of, an, this, that}

Code:

```
Print (nlp.Defaults.stop_words)
```

```
Print (len(nlp.Defaults.stop_words))
```

```
Print([t for t in doc if not t.is_stop])
```

Stemming

removing word suffixes or prefixes.

Banking } Bank
Banks }

Lemmatization

Reduce a word down to its lemma or dictionary form.

Did }
Done } DO
Doing }

Code: $[(t.\text{text}, t.\text{lemma}) \text{ for } t \text{ in doc}]$

week-2 || Advance processing

Part of speech (POS) Tagging:

{noun, verb, adjective, ...}

Example:

"john watched an old movie at the cinema?"

Prop. N Verb Det Noun ADP DET Adj P Noun

// Code

→ [(t.text, t.pos_) for t in doc]

→ [(t.text, t.tag_) for t in doc]

Named Entity Recognition (NER)

{a person, a location, an organization ...}

Named entity: anything that can be referred by a proper name. They often have a proper Noun (PROPN) pos tag.

Example:

Person → PER

Location → LOC

Geopolitical Entity → GPE

Organization → ORG

Usefulness:

- Organizing / categorizing Corpus
- Question answering
- Critical in information extraction.

Challenge:

An entity can speak multiple tokens.

Hamilton

US president?

City

F1 driven?

Code:

```
doc = nlp(s) // s is a sentence or string
```

```
[t.text, t.ent_type_] for t in doc)
```

```
// entity non-zero check
```

```
[(t.text, t.ent_type_) for t in doc if  
t.ent_type_ != 0]
```

```
// iterating through ents.
```

```
[(ent.text, ent.label) for ent in doc.ents)
```

Parsing

Determining the syntactic structure of a sentence.

type:

→ Constituency Parsing

→ Dependency Parsing

Constituency parsing using CFG

NP → Noun phrase

VP → verb phrase

PP → Prepositional phrase

NN → Noun

PPP → personal pronoun

NNP → proper Noun

VB → verb (Base form)

DT → Determiner

IN → preposition

Production Rules

$S \rightarrow NP VP$

$NP \rightarrow PRP | NNP | DT NN$

$VP \rightarrow VB | VB NP | VP PP$

$PP \rightarrow IN NP$

Lexicon

DT → the | a | this | that

PRP → I | she | he

IN → in | at

NN → book | hotel | room

"She enrolled in the course at university."

She
NP

enrolled
VP

"She enrolled in the course at university."

She
NP

enrolled in the course at university
VP

.
.

She
PRP

enrolled
VBD

In the course
PP

at the university
PP

In
IN

the course
NP

at
IN

the univer
NP

the
DT

course
NN

the
DT

University
NN

NP
N
the

V

NP
N
the

PP
P
at

NN
N
University

Rule

$S \rightarrow NP VP$

$NP \rightarrow DET N \mid DET ADJ N$

$VP \rightarrow V NP$

Lexicon

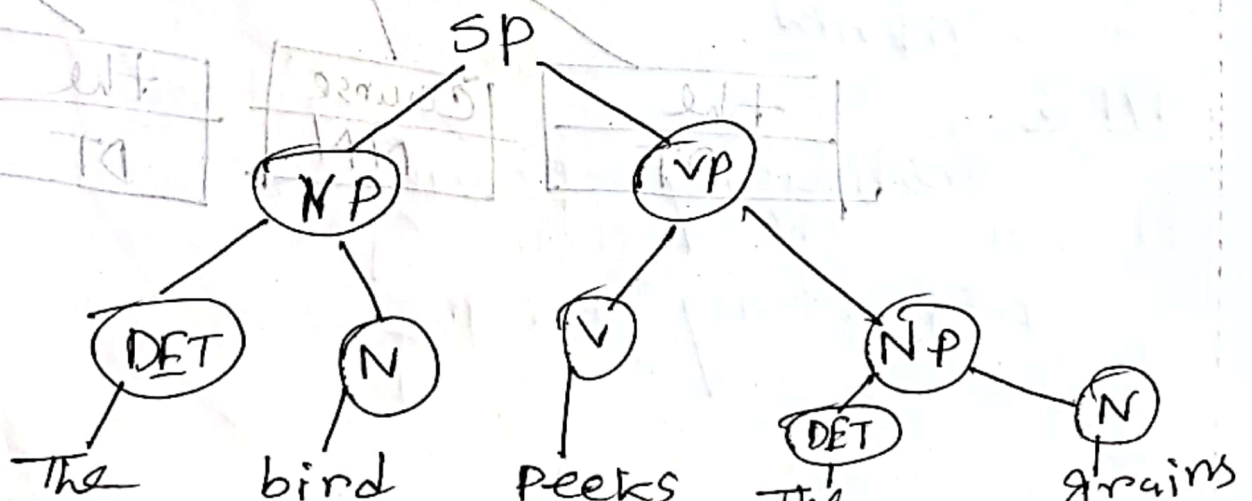
$DET \rightarrow a \mid the$

$ADJ \rightarrow beautiful \mid perching$

$N \rightarrow bird \mid birds \mid grain \mid grains$

$V \rightarrow peck \mid pecks \mid pecking$

→ "The bird pecks the grain."



Parsing

→ determine the symmetric structure of sentence.

"~~she enrolled~~"

"The quick brown fox jump over the lazy dog"

