#### **Text normalisation**

- 1. Process of transforming text into a single canonical form
- 2. Before text normalization we should be aware of
  - i) what type of text is to be normalized and,
  - ii) how it is to be processed afterwards

## Why text normalisation?

- 1) In string searching e.g. 'john' and 'John' (Case based matching)
- 2) American or British English spelling
- 3) Multiple form of single word e.g USA or US
- 4) frequently used in converting text to speech

Numbers, dates, acronyms, and abbreviations are non-standard "words" that need to be pronounced differently depending on context

# Sample example of text normalisation

```
1 import numpy as np
2 import string
3 import re
1 data_list = ['The Patient! is s wai@ting, for5 you in room Number','johN', 'J
2 print(data list)
1 preprocessed_text = []
1
2 for i in range(len(data list)):
3
         data = data_list[i]
4
         # Tokenize i.e. split on white spaces
5
         data = data.split()
6
         # Convert to lowercase
7
          data = [word.lower() for word in data]
```

```
✓ 0s completed at 10:36 AM
                                                                              X
          data = [re.sub(r"[@?\(^)+\) 0-9]", "", word) for word in data ] # No
15
16
          # Store as string
          data = ' '.join(data)
17
          preprocessed_text.append(data)
18
1 print("*****Text before preprocessing*****")
2 for text in data_list:
    print(text)
4
5 print("*****Text after preprocessing*****")
6 for text in preprocessed_text:
    print(text)
```

## 1. Parsing

## 2. Morpheme

## 3. Stemming

- chopping affies from a word
- may or may not has dictionary meaning
- Used in **information retrieval** for searching, Sentiment Analysis, document clustring. e.g search for party (search engine will show parties)
- Mostly used stemmer Porter stemmer
- other stemmers:
  - 1. Lovins Stemmer
  - 2. Dawson Stemmer
  - 3. Krovetz Stemmer
  - 4. Xerox Stemmer
  - 5. N-Gram Stemmer

#### Steps to use Porter stemmer

- Step 1 Import the NLTK library and from NLTK import PorterStemmer
- Step 2 Creat a variable and store PorterStemmer into it
- Step 3 use PorterStemmer

# Stemming of words

```
1 import nltk
2 from nltk.stem import PorterStemmer

1 ps = PorterStemmer()

1 print(ps.stem('bat'))
2 print(ps.stem('batting'))
```

# Stemming of a sentence

```
1 from nltk.tokenize import word_tokenize
2 nltk.download('punkt')

1 text = "This was not the map we found in Billy Bones's chest, but an accurate

1 words = word_tokenize(text) # or you can use above approach of spliting based
2 print(words)

1 stemed_words = []
2 for w in words:
3 stemed_words.append(ps.stem(w))

1 print(stemed_words)
```

He is reading detective stories

**T T T** 

He be read detective story NOTE: am, are, and is have the shared lemma be

#### 5. Different lemmatizers:

- WordnetLemmatizer
- 2. spaCy
- 3. TexxtBlob Lemmatizer
- 4. Pattern Lemmatizer
- 5. Standford CoreNLP Lemmatizer
- 6. Gensim lemmatizer

#### Lemmatization of words

```
1 from nltk.stem import WordNetLemmatizer
2 nltk.download('wordnet') #wordnet last tutorial
3 nltk.download('averaged_perceptron_tagger') # will download pos tag
1 lemmatizer = WordNetLemmatizer()
1 print(lemmatizer.lemmatize('sang'))
```

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6 print(lemmatized\_text)

```
1 text = "This was not the map we found in Billy Bones's chest, but an accurate
1 words = word_tokenize(text) # or you can use above approach of spliting based
2 print(words)

1 lemma_words = []
2 for w in words:
3 lemma_words.append(lemmatizer.lemmatize(w, pos='v'))

1 print(lemma_words)

1 # converting list into text stream
2 print("***** Text before lemmatization*****\n")
3 print(text)
4 lemmatized_text = ' '.join(lemma_words)
5 print("\n***** Text after lemmatization*****\n")
```