

New Lec

ch-2

Gutenberg corpus

- refers to Project Gutenberg

↳ largest collection of books  
whose copyright has ended

↳  
good for text processing

- a subset of it is in nltk

Brown corpus

↳  
each doc only 1 category

Reuters corpus

↳  
each doc can have multiple categories

↳  
so there will be overlaps

↳ CoNLL dataset

- de facto std dataset for NLP

↳ Bhāṣini project by Indian govt

- to collect text data in Indian languages

↳ If most of data is English then it will only benefit English speakers

↳ 4 type of corpus

e.g.

|             |                    |
|-------------|--------------------|
| Isolated    | gutenberg          |
| Categorized | brown              |
| overlapping | Reuters            |
| temporal    | inaugural speeches |

↳ Generating random text with Bigrams

- You want to predict next word
- find all bigrams of that word
- o/p most repeating word in bigrams with that word

↳ Wordlist corpus → Unusual words

↳ Pronouncing dictionary → for text to sound

↳ WordNet

- Dictionary
- provides sense

↳ Identical words can have diff senses based on the context

- Its not language but meaning based



### ch-3 Processing Raw Text

↳ Accessing text from web and from disk

import re → regular expression

↓  
technique to find patterns of text

import pprint → pp pretty print

↓  
library for

a = "I wouldn't go to class"

a.split() = ['I', 'wouldn't', 'go', 'to', 'class']

word\_tokenize(a) = ['I', 'would', 'not', 'go', 'to', 'class']

from urllib → allows you to get url data from internet

↳ Text on internet is encoded e.g. on ASCII or UTF-8

so you'll have to decode them

Most pages of internet are actually html pages

→ merely decoding does not work

→ you need to parse it from html structure

BeautifulSoup uses to parse html



## New Lec

Mitterm Quiz (1 hour or longer)

↳ diff from earlier quizzes

↳ were direct ques

↳ Will have actual reasoning ques

- scenario based questions

- won't have coding ques

- but will have conceptual tech ques

(like what algo takes less time)

especially ML part

- more descriptive stuff

eg. que what was reason behind split of cybernetics & AI

- google won't help you much

- you must be literate in concepts

- handwritten & open everything (except open chatbot)

## Couple of papers to read

↳ Assigned readings are components of this course

- means will come in exams

④ The global landscape of AI ethics guidelines

- Paper which covers all ethical guidelines that exist

⑤ Ethics as an escape from regulation by Ben Wagner

- Ethics is excellent but normative not legal

- more we talk abt ethics, less we talk about regulations

- hence companies only talk abt ethics not regulations



## ⑥ Governance with Teeth by Marja

- Ethics is something which <sup>means Human rights</sup> does not have legal analogy or universal acceptance
- Human rights are not normative but globally defined
- There is no legal tool to enforce ethics but human rights have to be enforced on every country
- Across borders regulations based on human rights can be imposed

## ⑦ How to recognize AI snake oil by Arvind Narayanan

- This is not paper but slides by a MIT prof
- snake oil means fraud
- Most of AI claims are snake oil
- snake oil is of diff types
- this slide talks about how to recognize such AI snake oils, some are very sophisticated like morphed datasets, exaggerated claims etc
- This ~~book~~ <sup>prof</sup> is writing a book, too

## Let's continue with NLTK ch-3

↳ Prev: how to download url from internet

- problem when HTML page

- hence pass the page through HTML parser

(BeautifulSoup)



## ↳ Reading Local files

`f = open ('<file-path>', 'w')`

(means to write)

`f.write ("<whatever you want to write>")`

`f.close`

- whenever 'w' used, it overwrites

means it will delete prev data

- to append data, use 'a' instead of 'w'

- 'r' used to read data

If you open in 'r' mode & run a for loop

for `i` in `f`:

} then `i` will go sentence  
by sentence

## ↳ NLP pipeline

- basic pipeline for all kind of research you'll be doing

HTML → ASCII → Text → Vocab

- nltk - wordpunc Tokenise

↳ this takes care of punctuations

↳ Next part of chapter tells about strings

↳ Then directly jump to "Regular Expressions" topic

- tool to find patterns from text

eg, you want to find all words ending with 'ing'

- every prog lang like python have their own regular expression tool.



netk.corpus.words.words

↳ gives every single word in netk

netk.corpus.words.words ('en')

↳ gives every single english word in netk

~~netk.corpus.words.words~~ ('ing\$', w)

- ee. search

\$ means word ending

will give all words ending with ing

- ee. search ('^win', w)

^ means word starting

will give all words starting with win

- ee. search ('aa+', w)

will give all words containing 'aa' or multiple a's

↳ Then chap goes into deeper in regular expressions

↳ Normalizing text

- changing everything in lower case

↳ Stemmers → smallest possible word in given word

e.g. Tables  $\xrightarrow[\text{word}]{\text{stem}}$  Table

listen  $\xrightarrow[\text{word}]{\text{stem}}$  list

↳ Lemmatization

- process of stemming is called lemmatization



↳ Segmentation

- breaking text in sentences

↳ Tokenization or word segmentation

- breaking text in words

↳ list to string: `join()`

↳ string to list: `split()`

## chapter 4

↳ Assignment of operators

Comparison of operators

↳ list, Tuples, sequence types

↳ No need to know generator expression

↳ Python coding style

every language has one

e.g. oop in c++

- Procedural vs declarative style

- variable scope

- checking parameter type

- Python does not allow to declare type of variable

- it only allows us to check type of variable

↳ Not much imp in context of this course or NLTK

- this chap is more abt learning coding disciplin in python.

- To learn more python, go through python book on module



- ↳ Algorithm design
  - how to code program s.t. it consumes less resources or less time
  - basically code in an efficient way
- e.g. divide and conquer in sorting

## ↳ Python libraries

## chapter 5

- ↳ starts to talk abt ML in NLP
- ↳ POS tagging
  - classified learning task
  - given text, find pos for every word
  - pos tag is grammatical role of that word e.g.  
noun, adverb, adjective, ...
  - In pos tag, position & tag of one word affects others
  - so words are not indep
- ↳ Dictionaries in Python
- ↳ How to make pos tagger?
  - one of the easiest ML task
  - default tagging: tag Noun to every word  
(but majority words are Noun in English)
  - then use regular tagging: e.g. everything which  
expression end with "ould" is modals
  - then use lookup tagger: lookup table of majority tags
  - find a dataset with pos tags, find out which tag  
is associated most of the times with a word  
(majority)



- till now you used only single words
- now increase size of data
- Increasing words of dataset will give upto 90% accuracy
- but 90 is still bad
- This 90 to 98-99, you'll need context
- So you now move to bigrams
- this word followed by that word is noun (or verb or ...)
- lookup table = unigram tagger
- bigram tagger → you'll get 95-96% accuracy
- This is something, you'll must be able to code