In [1]:

```
1
2 %%html
3 <style>
4 table {display: block;}
5 td {
6 font-size: 18px
7 }
8 .rendered_html { font-size: 18px; }
9 *{ line-height: 200%; }
10 </style>
```

Welcome to the Natural Language

Processing and the Web WS22/23

course

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Practice Class

When:

- Group-A:Tuesday 16:15 17:45 D-125/129:
- Group-B: Wednesday 16:15 -- 17:45 G-203:

Who:

- 1. Dr. Seid Muhie Yimam
- 2. Dr. Abhik Jana

Exercises

- Exercises (some can be done in groups) should be submitted before the next tutorial
- Machine learning and final projects will have separate deadlines
- Machine learning projects and final projects can be done in groups

To pass the class:

- Obtain at least 50% in the exercise points and Machine learning projects AND
- Obtain at least 50% in the final project points
- You must use your own computer to do the exercises and projects

 All exercises will use Python as programming language with different NLP libraries(example NLTK, spaCy), visualization (example mtplotlib or seaborn), and machine learning (example scikit learn, fast.ai and PyTorch) frameworks

In [3]:

```
1 y = 6
2 print(y)
```

6

Assignment 1 solution

Course Outline

Week	Торіс	Lecturer	When
1 -	NLTK and different NLP processing components	Seid/Abhik	Oct 18/19 Nov 1/2
4	Introductions to ML using Scikit learn and Feature engineering for ML projects	Seid/Abhik	Nov 8/9, Distribute ML Project
5	Deep learning using PyTorch	Seid/Abhik	Nov 15/16
6	Crossmodal Learning using PyTorch	Seid/Abhik	Nov 22/23
7	ML project presentation	Seid/Abhik	Nov 29/20
8	Distribute Project idea	Seid/Abhik	Dec 6/7
9	Project proposal presentation/discussion	Seid/Abhik	Dec 13/14
10 - 13	Project development consultation	Seid/Abhik	Dec 20/21/ Jan 24/25
14	Final project presentation	Seid/Abhik	Jan 31/Feb 1

How to install Python

The Practice class will be conducted using Python 3. We will rely on Anaconda3 (https://www.anaconda.com/distribution/)), specifically version 3.7+, a Python (and R) distribution which include most of the NLP and machine learning components. It also includes Anaconda Navigator, which is desktop graphical user interface that allows to launch different applications (Jupyter notebook, Spider GUI, GraphViz...) easily.

How to install Anaconda

Choose Python 3.7+ and install it to a location you prefer. Currently, Anconda installs around 200 packages, which include NLTK and the scikit learn machine learning packages. See here (http://docs.anaconda.com/anaconda) for the list of included packages.

Accessing Jupyter Notebook

For most of the practice classes, we will use the Jupyter Notebook web application to run and demonstrate examples. Except in the case of developing complex assignments and projects, you will also use Jupyter notebook to submit your assignments and in-class exercises.

How to open Jupyter Notebook:

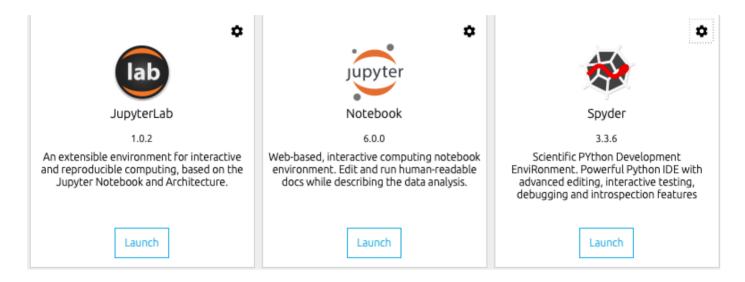
Using Anaconda Navigation

To open Jupyter Notebook using Anaconda navigator:

Once Anaconda is installed, open Anaconda Navigator from your program/application lists for MacOS and Windows.

For Linux, open a terminal and type anaconda-navigator

Then you can choose either JupyterLab or Jupyter Notebook to open Jupyter notebook. You can also open it via Spyder but you need to install a notebook plugin



Open Juypter using command line

From the command line, type:

```
jupyter notebook
```

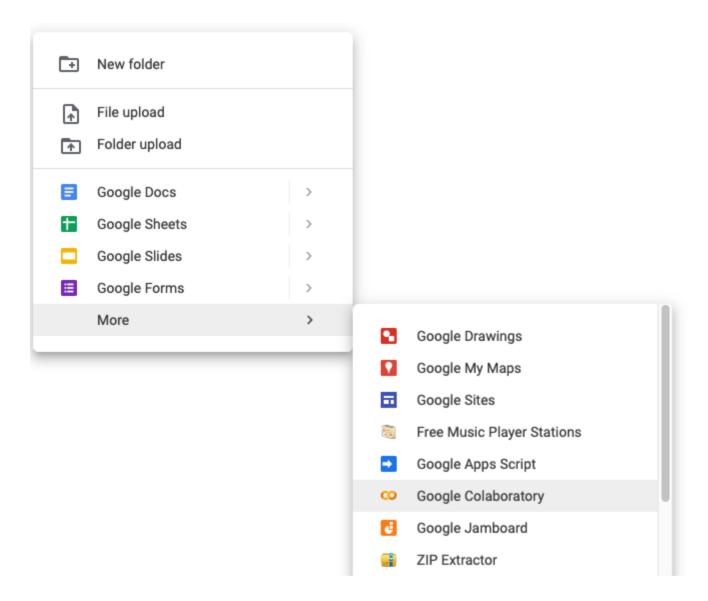
This will open the Jupyter Notebook in the browser.

Google Colab

Google provides a free cloud service based on Jupyter Notebooks that support free GPU, allows developing deep learning applications using PyTorch, TensorFlow, Keras, and so on. Using Colab, you can share your Jupyter notebooks, mount your Google Drive, etc.

Creating a folder for your notebooks

Create a folder in your Google Drive and create Google Colaboratory



Or you can directly go to Colab: https://colab.research.google.com/) and create a new notebook.

From there you can create a new notebook and set up your environment, such as selecting a GPU.

Mounting your Google Drive

Mounting your Google Drive allows you to persist models, files, and logs on your Google Drive folder as the Colab session is limited.

Once in Colab, you can run the following code to mount to your google drive

```
from google.colab import drive
drive.mount('/content/gdrive')
```

It will generate a link, click the link, copy the code and paste it in the text box, press Enter and your drive will be mounted.

To see the contents in your Google Drive

```
!ls "/content/gdrive/My Drive/"
```

How to open the associated notebooks

- 1. From Moodle, unzip the attachment ch1.zip
- Once Jupyter notebook is open, navigate to the extracted folder and open the file ch1.ipynb

Notebook cheatsheet

See https://www.cheatography.com/weidadeyue/cheat-sheets/jupyter-notebook/pdf bw/) to get the basic shortcuts and commands to execute a Jupyter notebook cell or display information

Example: Inside a cell, The following are important shortcuts:

- 1. shift-Tab-Tab -- Show documentation of methods or classes
- 2. shift-Enter --- Execute cell and move/create the next cell
- 3. Ctrl-Enter -- Execute a cell and remain the focus on the same cell

~ Hands on Session ~

Install Anaconda, Open Notebooks, Colab Environment 15 minutes

Python preprocessing and NLP pipeline

In this practice class, we will install and discuss two packages/libraries that are important for preprocessing and NLP pipeline, namely NLTK and spaCy. We will use these two libraries extensively for the practice classes and assignments throughout the course.

Installing NLTK components

NLTK is installed automatically with the Anaconda distribution. However, different modules, datasets, and packages are left for the user to download on demand. Try running the following code to install some or all the required resources. If you already have the required resources, the program should run without any problem. Otherwise, it will give an error with hints on how to install the missing resources.

Example of installing missed resources:

```
import nltk
nltk.download('punkt')
```

```
In [2]:
```

```
import nltk
nltk.download('punkt')

[nltk_data] Downloading package punkt to
[nltk_data] /Users/seidmuhieyimam/nltk_data...
[nltk_data] Package punkt is already up-to-date!

Out[2]:

True
```

If you want to choose from a graphical user interface, run the following from the command line

```
nlt.download()
```

Then you can choose what you like to download. You can download everything but it will take more time and storage.

To install all the packages/resources, run the following

```
nltk.download('all')
```

In [3]:

True

```
import nltk

#Download specific resources

nltk.download('gutenberg')

nltk.download('inaugural')

nltk.download('inps_chat')

nltk.download('webtext')

nltk.download('treebank')

# If you want to download everything from nltk, uncomment the following

# nltk.download('all')
```

```
[nltk data] Downloading package gutenberg to
[nltk data]
                /Users/seidmuhieyimam/nltk data...
             Package gutenberg is already up-to-date!
[nltk data]
[nltk data] Downloading package genesis to
[nltk data]
               /Users/seidmuhieyimam/nltk data...
[nltk data] Package genesis is already up-to-date!
[nltk_data] Downloading package inaugural to
[nltk data]
                /Users/seidmuhieyimam/nltk data...
[nltk data] Package inaugural is already up-to-date!
[nltk data] Downloading package nps chat to
                /Users/seidmuhieyimam/nltk data...
[nltk data]
[nltk data]
             Package nps chat is already up-to-date!
[nltk_data] Downloading package webtext to
                /Users/seidmuhieyimam/nltk data...
[nltk data]
[nltk data] Package webtext is already up-to-date!
[nltk data] Downloading package treebank to
                /Users/seidmuhieyimam/nltk data...
[nltk data]
[nltk data]
             Package treebank is already up-to-date!
Out[3]:
```

The NLTK resource contains dataset that are annotated for different linguistic annotation tasks such as POS tagging, dependency parsing, NER...

```
In [4]:
```

```
# Show all books from nltk
   from nltk.book import *
   #the books are available as text1---text9
   print (text1.name)
 4
    print("----")
    print (text1.concordance("monstrous"))
   print ("---")
*** Introductory Examples for the NLTK Book ***
Loading text1, ..., text9 and sent1, ..., sent9
Type the name of the text or sentence to view it.
Type: 'texts()' or 'sents()' to list the materials.
text1: Moby Dick by Herman Melville 1851
text2: Sense and Sensibility by Jane Austen 1811
text3: The Book of Genesis
text4: Inaugural Address Corpus
text5: Chat Corpus
text6: Monty Python and the Holy Grail
text7: Wall Street Journal
text8: Personals Corpus
text9: The Man Who Was Thursday by G . K . Chesterton 1908
Moby Dick by Herman Melville 1851
Displaying 11 of 11 matches:
ong the former , one was of a most monstrous size . ... This came towa
rds us ,
ON OF THE PSALMS . " Touching that monstrous bulk of the whale or ork
we have r
ll over with a heathenish array of monstrous clubs and spears . Some w
ere thick
d as you gazed , and wondered what monstrous cannibal and savage could
ever hav
that has survived the flood; most monstrous and most mountainous! Th
at Himmal
they might scout at Moby Dick as a monstrous fable , or still worse an
d more de
th of Radney .'" CHAPTER 55 Of the Monstrous Pictures of Whales . I sh
all ere 1
ing Scenes . In connexion with the monstrous pictures of whales , I am
strongly
ere to enter upon those still more monstrous stories of them which are
to be fo
ght have been rummaged out of this monstrous cabinet there is no telli
ng . But
of Whale - Bones; for Whales of a monstrous size are oftentimes cast
up dead u
None
____
```

For example, the brown corpus can be downloaded and imported as follows

```
nltk.download('brown')
```

```
from nltk.corpus import brown
```

The brown corpus includes different genres of texts such as fictions, reviews, hobbies, news, and so on.

To view the categories:

'romance',

'science fiction']

```
brown.categories()
```

```
In [5]:
   nltk.download('brown')
    from nltk.corpus import brown
   brown.categories()
[nltk_data] Downloading package brown to
[nltk data] /Users/seidmuhieyimam/nltk data...
[nltk data] Package brown is already up-to-date!
Out[5]:
['adventure',
 'belles lettres',
 'editorial',
 'fiction',
 'government',
 'hobbies',
 'humor',
 'learned',
 'lore',
 'mystery',
 'news',
 'religion',
 'reviews',
```

```
#List the first 10 words in the corpus

print(brown.words(categories='fiction')[:10])

# How many words are there in the corpus

print("All:", len(brown.words()))

# How many words in the adventure category

print("adventure:",len(brown.words(categories='adventure')))

# Print the last 10 sentences from the romance categories

print("sents from romance:",brown.sents(categories="romance")[-10:])

# How many sentences are there in total

print("All Sentences:", len(brown.sents()))
```

```
['Thirty-three', 'Scotty', 'did', 'not', 'go', 'back', 'to', 'school',
'.', 'His']
All: 1161192
adventure: 69342
sents from romance: [["I'm", 'not', 'giving', 'you', 'a', 'chance',
 ',', 'Bill', ',', 'but', 'availing', 'myself', 'of', 'your', 'generou
s', 'offer', 'of', 'assistance', '.'], ['Good', 'luck', 'to', 'you', "''", '.'], ['``', 'All', 'the', 'in-laws', 'have', 'got', 'to', 'have', 'their', 'day', "''", ', 'Adam', 'said', ',', 'and', 'glared',
'at', 'William', 'and', 'Freddy', 'in', 'turn', '.'], ['Sweat', 'start
ed', 'out', 'on', "William's", 'forehead', ',', 'whether', 'from', 're
lief', 'or', 'disquietude', 'he', 'could', 'not', 'tell', '.'], ['Across', 'the', 'table', ',', 'Hamrick', 'saluted', 'him', 'jubilantly', 'with', 'an', 'encircled', 'thumb', 'and', 'forefinger', '.'], ['Nobod
y', 'else', 'showed', 'pleasure', '.'], ['Spike-haired', ',', 'burly', ',', 'red-faced', ',', 'decked', 'with', 'horn-rimmed', 'glasses', 'an
d', 'an', 'Ivy', 'League', 'suit', ',', 'Jack', 'Hamrick', 'awaited',
'William', 'at', 'the', "officers'", 'club', '.'], ['``', 'Hello',
 william, at, the, biliters, club, .,, , said, ,',', 'and', 'grinned', '.'], ['`
', 'I', 'suppose', 'I', 'can', 'never', 'expect', 'to', 'call', 'yo u', "'", 'General', "'", 'after', 'that', 'Washington', 'episode',
"''", '.'], ['``', "I'm", 'afraid', 'not', "''", '.']]
All Sentences: 57340
```

NLTK also includes Web and chat texts from Firefox discussion forum, the movie script from Pirates of the Caribbean, wine reviews, ...

To import the Web and chat corpus:

```
from nltk.corpus import webtext
```

In [7]:

```
from nltk.corpus import webtext

for id in webtext.fileids():

#display some parts of the text in the document
print(id, webtext.raw(id)[:30],"...")
```

```
firefox.txt Cookie Manager: "Don't allow s ... grail.txt SCENE 1: [wind] [clop clop clo ... overheard.txt White guy: So, do you have any ... pirates.txt PIRATES OF THE CARRIBEAN: DEAD ... singles.txt 25 SEXY MALE, seeks attrac old ... wine.txt Lovely delicate, fragrant Rhon ...
```

NLTK also include inaugural speeches of US presidents.

You can import these corpora as follows

```
from nltk.corpus import inaugural
```

In [8]:

```
from nltk.corpus import inaugural

# print all inaugural

print("Inaugral Texts:",inaugural.fileids())

inaugural.raw("2017-Trump.txt")[:1000]
```

Inaugral Texts: ['1789-Washington.txt', '1793-Washington.txt', '1797-A dams.txt', '1801-Jefferson.txt', '1805-Jefferson.txt', '1809-Madison.t xt', '1813-Madison.txt', '1817-Monroe.txt', '1821-Monroe.txt', '1825-A dams.txt', '1829-Jackson.txt', '1833-Jackson.txt', '1837-VanBuren.tx t', '1841-Harrison.txt', '1845-Polk.txt', '1849-Taylor.txt', '1853-Pie rce.txt', '1857-Buchanan.txt', '1861-Lincoln.txt', '1865-Lincoln.txt', '1869-Grant.txt', '1873-Grant.txt', '1877-Hayes.txt', '1881-Garfield.t xt', '1885-Cleveland.txt', '1889-Harrison.txt', '1893-Cleveland.txt', '1897-McKinley.txt', '1901-McKinley.txt', '1905-Roosevelt.txt', '1909-Taft.txt', '1913-Wilson.txt', '1917-Wilson.txt', '1921-Harding.txt', '1925-Coolidge.txt', '1929-Hoover.txt', '1933-Roosevelt.txt', '1937-Ro osevelt.txt', '1941-Roosevelt.txt', '1945-Roosevelt.txt', '1949-Truma n.txt', '1953-Eisenhower.txt', '1957-Eisenhower.txt', '1961-Kennedy.tx t', '1965-Johnson.txt', '1969-Nixon.txt', '1973-Nixon.txt', '1977-Cart er.txt', '1981-Reagan.txt', '1985-Reagan.txt', '1989-Bush.txt', '1993-Clinton.txt', '1997-Clinton.txt', '2001-Bush.txt', '2005-Bush.txt', '2 009-Obama.txt', '2013-Obama.txt', '2017-Trump.txt']

Out[8]:

"Chief Justice Roberts, President Carter, President Clinton, President Bush, President Obama, fellow Americans, and people of the world: Than k you.\n\nWe, the citizens of America, are now joined in a great natio nal effort to rebuild our country and restore its promise for all of o ur people. Together, we will determine the course of America and the w orld for many, many years to come. We will face challenges, we will confront hardships, but we will get the job done.\n\nEvery 4 years, we gather on these steps to carry out the orderly and peaceful transfer of power, and we are grateful to President Obama and First Lady Michelle Obama for their gracious aid throughout this transition. They have been magnificent. Thank you.\n\nToday's ceremony, however, has very special meaning. Because today we are not merely transferring power from one administration to another or from one party to another, but we are transferring power from Washington, DC, and giving it back to you, the people.\n\nFor too long, a "

NLTK contains tagged and parsed corpora that we will discuss in the upcoming practice classes

In [9]:

```
import nltk
   nltk.download('maxent ne chunker')
   nltk.download('averaged perceptron tagger')
 3
   nltk.download('words')
   texts = [
 7
      ONCE UPON A TIME a girl named Cinderella lived with her stepmother and two st
 8
 9
      Poor Cinderella had to work hard all day long so the others could rest.
10
      It was she who had to wake up each morning when it was still dark and cold to
      It was she who cooked the meals. It was she who kept the fire going.
11
12
      The poor girl could not stay clean, from all the ashes and cinders by the fir
13
14
   for text in texts:
15
        sentences = nltk.sent tokenize(text)
16
        for sentence in sentences:
17
            words = nltk.word tokenize(sentence)
18
19
            tagged_words = nltk.pos_tag(words)
20
            ne_tagged_words = nltk.ne_chunk(tagged_words)
            print (ne_tagged_words)
[nltk data] Downloading package maxent ne chunker to
```

```
[nltk data]
                /Users/seidmuhieyimam/nltk data...
              Package maxent ne chunker is already up-to-date!
[nltk data]
[nltk_data] Downloading package averaged_perceptron_tagger to
[nltk data]
                /Users/seidmuhieyimam/nltk data...
[nltk data]
              Package averaged perceptron tagger is already up-to-
[nltk_data]
                  date!
[nltk data] Downloading package words to
                /Users/seidmuhieyimam/nltk data...
[nltk data]
[nltk data]
              Package words is already up-to-date!
(S
 ONCE/RB
 UPON/IN
 A/NNP
 TIME/NNP
 a/DT
 girl/NN
 named/VBN
  (PERSON Cinderella/NNP)
 lived/VBD
 with/IN
```

```
her/PRP$
  stepmother/NN
  and/CC
  two/CD
  stepsisters/NNS
  ./.)
(S
  (PERSON Poor/NNP)
  (PERSON Cinderella/NNP)
 had/VBD
  to/TO
 work/VB
 hard/RB
  all/DT
 day/NN
  long/RB
  so/RB
  the/DT
 others/NNS
 could/MD
 rest/VB
  ./.)
(S
  It/PRP
 was/VBD
  she/PRP
 who/WP
 had/VBD
  to/TO
 wake/VB
 up/RP
  each/DT
 morning/NN
 when/WRB
  it/PRP
 was/VBD
  still/RB
 dark/JJ
  and/CC
 cold/JJ
 to/TO
  start/VB
 the/DT
  fire/NN
  ./.)
(S It/PRP was/VBD she/PRP who/WP cooked/VBD the/DT meals/NNS ./.)
(S
  It/PRP
 was/VBD
  she/PRP
 who/WP
 kept/VBD
 the/DT
 fire/NN
  going/VBG
  ./.)
(S
 The/DT
 poor/JJ
  girl/NN
 could/MD
```

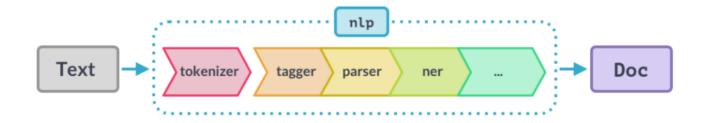
```
not/RB
stay/VB
clean/JJ
,/,
from/IN
all/PDT
the/DT
ashes/NNS
and/CC
cinders/NNS
by/IN
the/DT
fire/NN
./.)
```

~ Hands on Session ~

Experiment with NLTK 15 minutes

spaCy

spaCy is a free, open-source library for advanced Natural Language Processing (NLP) in Python. It is a package, with a support for a number of nlp tasks such as tokenization, lemmatisation, part-of-speech tagging, entity recognition and so on.



How to Install spaCy

To install spaCy, visit the documentation page here (https://spacy.io/usage).

Using pip

```
pip install -U spacy
```

If you want to install it from the notebook, append the ! before the command

```
!pip install -U spacy
```

To install some models, run the following from the command line (make sure to use the correct python version, that is python3)

```
python3 -m spacy download en_core_web_sm
```

or

```
python3 -m spacy download en
```

en_core_web_sm: (11 MB) English multi-task CNN trained on OntoNotes. Assigns context-specific token vectors, POS tags, dependency parse and named entities.

en_core_web_md: (91 MB) English multi-task CNN trained on OntoNotes, with GloVe vectors trained on Common Crawl. Assigns word vectors, context-specific token vectors, POS tags, dependency parse and named entities.

en_core_web_lg: (789 MB) English multi-task CNN trained on OntoNotes, with GloVe vectors trained on Common Crawl. Assigns word vectors, context-specific token vectors, POS tags, dependency parse and named entities.

```
!pip install -U spacy
Requirement already satisfied: spacy in /opt/anaconda3/lib/python3.8/s
ite-packages (3.1.3)
Requirement already satisfied: tqdm<5.0.0,>=4.38.0 in /opt/anaconda3/1
ib/python3.8/site-packages (from spacy) (4.59.0)
Requirement already satisfied: pathy>=0.3.5 in /opt/anaconda3/lib/pyth
on3.8/site-packages (from spacy) (0.6.0)
Requirement already satisfied: catalogue<2.1.0,>=2.0.6 in /opt/anacond
a3/lib/python3.8/site-packages (from spacy) (2.0.6)
Requirement already satisfied: jinja2 in /opt/anaconda3/lib/python3.8/
site-packages (from spacy) (2.11.3)
Requirement already satisfied: murmurhash<1.1.0,>=0.28.0 in /opt/anaco
nda3/lib/python3.8/site-packages (from spacy) (1.0.5)
Requirement already satisfied: blis<0.8.0,>=0.4.0 in /opt/anaconda3/li
b/python3.8/site-packages (from spacy) (0.7.4)
Requirement already satisfied: numpy>=1.15.0 in /opt/anaconda3/lib/pyt
hon3.8/site-packages (from spacy) (1.20.1)
Requirement already satisfied: cymem<2.1.0,>=2.0.2 in /opt/anaconda3/1
ib/python3.8/site-packages (from spacy) (2.0.5)
Requirement already satisfied: srsly<3.0.0,>=2.4.1 in /opt/anaconda3/1
ib/python3.8/site-packages (from spacy) (2.4.1)
Requirement already satisfied: packaging>=20.0 in /opt/anaconda3/lib/p
ython3.8/site-packages (from spacy) (20.9)
Requirement already satisfied: thinc<8.1.0,>=8.0.9 in /opt/anaconda3/1
ib/python3.8/site-packages (from spacy) (8.0.10)
Requirement already satisfied: spacy-legacy<3.1.0,>=3.0.8 in /opt/anac
onda3/lib/python3.8/site-packages (from spacy) (3.0.8)
Requirement already satisfied: wasabi<1.1.0,>=0.8.1 in /opt/anaconda3/
lib/python3.8/site-packages (from spacy) (0.8.2)
Requirement already satisfied: pydantic!=1.8,!=1.8.1,<1.9.0,>=1.7.4 in
/opt/anaconda3/lib/python3.8/site-packages (from spacy) (1.8.2)
Requirement already satisfied: typer<0.5.0,>=0.3.0 in /opt/anaconda3/1
ib/python3.8/site-packages (from spacy) (0.4.0)
Requirement already satisfied: requests<3.0.0,>=2.13.0 in /opt/anacond
a3/lib/python3.8/site-packages (from spacy) (2.25.1)
Requirement already satisfied: setuptools in /opt/anaconda3/lib/python
3.8/site-packages (from spacy) (52.0.0.post20210125)
Requirement already satisfied: preshed<3.1.0,>=3.0.2 in /opt/anaconda
3/lib/python3.8/site-packages (from spacy) (3.0.5)
Requirement already satisfied: pyparsing>=2.0.2 in /opt/anaconda3/lib/
python3.8/site-packages (from packaging>=20.0->spacy) (2.4.7)
Requirement already satisfied: smart-open<6.0.0,>=5.0.0 in /opt/anacon
da3/lib/python3.8/site-packages (from pathy>=0.3.5->spacy) (5.2.1)
Requirement already satisfied: typing-extensions>=3.7.4.3 in /opt/anac
onda3/lib/python3.8/site-packages (from pydantic!=1.8,!=1.8.1,<1.9.0,>
=1.7.4->spacy) (3.7.4.3)
Requirement already satisfied: idna<3,>=2.5 in /opt/anaconda3/lib/pyth
on3.8/site-packages (from requests<3.0.0,>=2.13.0->spacy) (2.10)
Requirement already satisfied: certifi>=2017.4.17 in /opt/anaconda3/li
b/python3.8/site-packages (from requests<3.0.0,>=2.13.0->spacy) (2020.
12.5)
Requirement already satisfied: chardet<5,>=3.0.2 in /opt/anaconda3/li
b/python3.8/site-packages (from requests<3.0.0,>=2.13.0->spacy) (4.0.
0)
```

Requirement already satisfied: urllib3<1.27,>=1.21.1 in /opt/anaconda

3/lib/python3.8/site-packages (from requests<3.0.0,>=2.13.0->spacy)
(1.26.4)

Requirement already satisfied: click<9.0.0,>=7.1.1 in /opt/anaconda3/1 ib/python3.8/site-packages (from typer<0.5.0,>=0.3.0->spacy) (7.1.2) Requirement already satisfied: MarkupSafe>=0.23 in /opt/anaconda3/lib/python3.8/site-packages (from jinja2->spacy) (1.1.1)

```
!python3 -m spacy download en core web sm
Collecting en-core-web-sm==3.1.0
  Downloading https://github.com/explosion/spacy-models/releases/downl
oad/en_core_web_sm-3.1.0/en_core_web_sm-3.1.0-py3-none-any.whl (http
s://github.com/explosion/spacy-models/releases/download/en core web sm
-3.1.0/en core web sm-3.1.0-py3-none-any.whl) (13.6 MB)
                                      | 13.6 MB 5.9 MB/s eta 0:00:01
Requirement already satisfied: spacy<3.2.0,>=3.1.0 in /opt/anaconda3/1
ib/python3.8/site-packages (from en-core-web-sm==3.1.0) (3.1.3)
Requirement already satisfied: murmurhash<1.1.0,>=0.28.0 in /opt/anaco
nda3/lib/python3.8/site-packages (from spacy<3.2.0,>=3.1.0->en-core-we
b-sm==3.1.0) (1.0.5)
Requirement already satisfied: spacy-legacy<3.1.0,>=3.0.8 in /opt/anac
onda3/lib/python3.8/site-packages (from spacy<3.2.0,>=3.1.0->en-core-w
eb-sm==3.1.0) (3.0.8)
Requirement already satisfied: jinja2 in /opt/anaconda3/lib/python3.8/
site-packages (from spacy<3.2.0,>=3.1.0->en-core-web-sm==3.1.0) (2.11.
3)
Requirement already satisfied: cymem<2.1.0,>=2.0.2 in /opt/anaconda3/1
ib/python3.8/site-packages (from spacy<3.2.0,>=3.1.0->en-core-web-sm==
3.1.0) (2.0.5)
Requirement already satisfied: wasabi<1.1.0,>=0.8.1 in /opt/anaconda3/
lib/python3.8/site-packages (from spacy<3.2.0,>=3.1.0->en-core-web-sm=
=3.1.0) (0.8.2)
Requirement already satisfied: srsly<3.0.0,>=2.4.1 in /opt/anaconda3/1
ib/python3.8/site-packages (from spacy<3.2.0,>=3.1.0->en-core-web-sm==
3.1.0) (2.4.1)
Requirement already satisfied: tqdm<5.0.0,>=4.38.0 in /opt/anaconda3/1
ib/python3.8/site-packages (from spacy<3.2.0,>=3.1.0->en-core-web-sm==
3.1.0) (4.59.0)
Requirement already satisfied: numpy>=1.15.0 in /opt/anaconda3/lib/pyt
hon3.8/site-packages (from spacy<3.2.0,>=3.1.0->en-core-web-sm==3.1.0)
(1.20.1)
Requirement already satisfied: pydantic!=1.8,!=1.8.1,<1.9.0,>=1.7.4 in
/opt/anaconda3/lib/python3.8/site-packages (from spacy<3.2.0,>=3.1.0->
en-core-web-sm==3.1.0) (1.8.2)
Requirement already satisfied: setuptools in /opt/anaconda3/lib/python
3.8/site-packages (from spacy<3.2.0,>=3.1.0->en-core-web-sm==3.1.0) (5
2.0.0.post20210125)
Requirement already satisfied: thinc<8.1.0,>=8.0.9 in /opt/anaconda3/1
ib/python3.8/site-packages (from spacy<3.2.0,>=3.1.0->en-core-web-sm==
3.1.0) (8.0.10)
Requirement already satisfied: pathy>=0.3.5 in /opt/anaconda3/lib/pyth
on3.8/site-packages (from spacy<3.2.0,>=3.1.0->en-core-web-sm==3.1.0)
Requirement already satisfied: packaging>=20.0 in /opt/anaconda3/lib/p
ython3.8/site-packages (from spacy<3.2.0,>=3.1.0->en-core-web-sm==3.1.
0) (20.9)
Requirement already satisfied: catalogue<2.1.0,>=2.0.6 in /opt/anacond
a3/lib/python3.8/site-packages (from spacy<3.2.0,>=3.1.0->en-core-web-
sm==3.1.0) (2.0.6)
Requirement already satisfied: typer<0.5.0,>=0.3.0 in /opt/anaconda3/1
ib/python3.8/site-packages (from spacy<3.2.0,>=3.1.0->en-core-web-sm==
3.1.0) (0.4.0)
```

```
Requirement already satisfied: preshed<3.1.0,>=3.0.2 in /opt/anaconda
3/lib/python3.8/site-packages (from spacy<3.2.0,>=3.1.0->en-core-web-s
m==3.1.0) (3.0.5)
Requirement already satisfied: requests<3.0.0,>=2.13.0 in /opt/anacond
a3/lib/python3.8/site-packages (from spacy<3.2.0,>=3.1.0->en-core-web-
sm==3.1.0) (2.25.1)
Requirement already satisfied: blis<0.8.0,>=0.4.0 in /opt/anaconda3/li
b/python3.8/site-packages (from spacy<3.2.0,>=3.1.0->en-core-web-sm==
3.1.0) (0.7.4)
Requirement already satisfied: pyparsing>=2.0.2 in /opt/anaconda3/lib/
python3.8/site-packages (from packaging>=20.0->spacy<3.2.0,>=3.1.0->en
-core-web-sm==3.1.0) (2.4.7)
Requirement already satisfied: smart-open<6.0.0,>=5.0.0 in /opt/anacon
da3/lib/python3.8/site-packages (from pathy>=0.3.5->spacy<3.2.0,>=3.1.
0 - en-core-web-sm = 3.1.0) (5.2.1)
Requirement already satisfied: typing-extensions>=3.7.4.3 in /opt/anac
onda3/lib/python3.8/site-packages (from pydantic!=1.8,!=1.8.1,<1.9.0,>
=1.7.4-spacy<3.2.0,>=3.1.0-en-core-web-sm==3.1.0) (3.7.4.3)
Requirement already satisfied: certifi>=2017.4.17 in /opt/anaconda3/li
b/python3.8/site-packages (from requests<3.0.0,>=2.13.0->spacy<3.2.0,>
=3.1.0->en-core-web-sm==3.1.0) (2020.12.5)
Requirement already satisfied: idna<3,>=2.5 in /opt/anaconda3/lib/pyth
on3.8/site-packages (from requests<3.0.0,>=2.13.0->spacy<3.2.0,>=3.1.0
->en-core-web-sm==3.1.0) (2.10)
Requirement already satisfied: chardet<5,>=3.0.2 in /opt/anaconda3/li
b/python3.8/site-packages (from requests<3.0.0,>=2.13.0->spacy<3.2.0,>
=3.1.0 - \text{en-core-web-sm} = 3.1.0) (4.0.0)
Requirement already satisfied: urllib3<1.27,>=1.21.1 in /opt/anaconda
```

Requirement already satisfied: urllib3<1.27,>=1.21.1 in /opt/anaconda 3/lib/python3.8/site-packages (from requests<3.0.0,>=2.13.0->spacy<3.2.0,>=3.1.0->en-core-web-sm==3.1.0) (1.26.4)

Requirement already satisfied: click<9.0.0,>=7.1.1 in /opt/anaconda3/1 ib/python3.8/site-packages (from typer<0.5.0,>=0.3.0->spacy<3.2.0,>=3.1.0->en-core-web-sm==3.1.0) (7.1.2)

Requirement already satisfied: MarkupSafe>=0.23 in /opt/anaconda3/lib/python3.8/site-packages (from jinja2->spacy<3.2.0,>=3.1.0->en-core-web-sm==3.1.0) (1.1.1)

✓ Download and installation successful

You can now load the package via spacy.load('en_core_web_sm')

```
# start spaCy's pipeline, which can load differnt spacy models
import spacy
nlp = spacy.load("en_core_web_sm")
doc = nlp("The big grey dog ate all of the chocolate, but fortunately the he was
#Tokenization
print("Using split", doc.text.split())
#Using spaCy, only orth_ method
print("Using spaCy",[token.orth_ for token in doc])
# using spacy, including the integer representation of tokens
print("spaCy integer",[(token.orth_, token.orth) for token in doc])
```

```
Using split ['The', 'big', 'grey', 'dog', 'ate', 'all', 'of', 'the', 'chocolate,', 'but', 'fortunately', 'the', 'he', "wasn't", 'sick!']
Using spaCy ['The', 'big', 'grey', 'dog', 'ate', 'all', 'of', 'the', 'chocolate', ',', 'but', 'fortunately', 'the', 'he', 'was', "n't", 'sick', '!']
spaCy integer [(The, 'The', 5059648917813135842), (big, 'big', 1551163 2813958231649), (grey, 'grey', 10475807793332549289), (dog, 'dog', 756 2983679033046312), (ate, 'ate', 10806788082624814911), (all, 'all', 13 409319323822384369), (of, 'of', 886050111519832510), (the, 'the', 7425 985699627899538), (chocolate, 'chocolate', 10946593968795032542), (,, ', 2593208677638477497), (but, 'but', 14560795576765492085), (fortunately, 'fortunately', 13851269277375979931), (the, 'the', 742598569962 7899538), (he, 'he', 1655312771067108281), (was, 'was', 99216865133789 12864), (n't, "n't", 2043519015752540944), (sick, 'sick', 148415976098 57081305), (!, '!', 17494803046312582752)]
```

Important functions for language processing

1. Referencing, Equality, Conditionals, Sequences, Operations on sequences, Rearranging, Zipping, enumerating

```
In [13]:
```

```
1  a = 'English'
2  b = a
3  a = 'French'
4  # b is Still English
5  print (b)
6  print("---")
```

English

In [14]:

```
1  a = ["English", "Spanish"]
2  b = a
3  a[0] = "French"
4  # Now b is changed ==> access by reference
5  print (b)
6  print("---")
```

['French', 'Spanish']

In [15]:

```
# the same reference for all 5 entries

a = ["German"] *5

print ([a[i] for i,_ in enumerate(a)])

# Are they the same? use the 'is' keyword

print (a[0] is a[1] is a[2] is a[3] is a[4])

print("---")
```

['German', 'German', 'German', 'German'] True

```
In [16]:
```

```
# change value of one of the entry (the forth item)

a[3] = "French"

print ([a[i] for i,_ in enumerate(a)])

print (a[0] is a[1] is a[2] is a[3] is a[4])

print("---")
```

```
['German', 'German', 'French', 'German']
False
---
```

In [17]:

```
# conditioning
words = ["the", "quick", "brown", "fox",", "jumps", "over", "the", "lazy", "do
punct ="!.,"
for word in words:
    if word in punct:
        print (word + "is a punctuation", end="; ")# print single line until end
```

, is a punctuation; !is a punctuation;

In [18]:

```
# Zip (concatenate lists). Example words, frequencies, POS Tag

print()
print("---")

words = ['I', 'turned', 'off', 'the', 'spectroroute']

freq = [15, 3,6,23,1]

tags = ['noun', 'verb', 'prep', 'det', 'noun']

print ("Zipped:", list(zip(words, tags, freq)))

print("---")
```

```
Zipped: [('I', 'noun', 15), ('turned', 'verb', 3), ('off', 'prep', 6),
('the', 'det', 23), ('spectroroute', 'noun', 1)]
```

```
In [19]:
```

```
# Sorting, reversing, differences
words_on = ['I', 'turned', 'on', 'the', 'light']
print ("sorted:", sorted(words))

print ("reveresed:",list(reversed(words)))

print ("difference:", set(words).difference(words_on))

print ("difference:",set(words_on).difference(words))

print("---")

sorted: ['I', 'off', 'spectroroute', 'the', 'turned']
```

```
reveresed: ['spectroroute', 'the', 'off', 'turned', 'I']
difference: {'off', 'spectroroute'}
difference: {'light', 'on'}
---
```

In [20]:

```
# rearrenging lists
words_on[2], words_on[3], words_on[4] = words_on[3], words_on[4], words_on[2]
print("rearranged:", words_on)
```

```
rearranged: ['I', 'turned', 'the', 'light', 'on']
```

2. Splitting datasets:

We will discuss in upcoming practice class on how to properly split datasets into training, development, and test set. There are frameworks that help splitting and shuffling.

In [21]:

```
text = """

text = """

This is a simple training example.

We split by the text into training and test sets.

Most of the data goes to training.

Small parts for testing

"""

split = int(0.9 * len(text.split()))

train, test = text.split()[:split], text.split()[split:]

print ("train set", train)

print("test set", test)
```

```
train set ['This', 'is', 'a', 'simple', 'training', 'example.', 'We',
'split', 'by', 'the', 'text', 'into', 'training', 'and', 'test', 'set
s.', 'Most', 'of', 'the', 'data', 'goes', 'to', 'training.', 'Small']
test set ['parts', 'for', 'testing']
```

3. Generator

Assume you are searching a given word in a corpus

```
def search1(substring, words):
    result = []
    for word in words:
        if substring in word:
            result.append(word)
    return result
```

Here in search1, the list is accumulated and the final result is returned once the search is over. What if the file is very large...?

```
def search2(substring, words):
    for word in words:
       if substring in word:
          yield word
```

Here, search2 uses a yield statement, which is a generator. It will pause when it get the first word and the caller function will do whatever with the search word before search continues.

```
In [22]:
```

```
import nltk
   def search1(substring, words):
       result = []
 3
       for word in words:
 4
            if substring in word:
                result.append(word)
 7
       return result
 8
   def search2(substring, words):
       for word in words:
10
11
            if substring in word:
                yield word
12
13
14
   for item in search1('zzo', nltk.corpus.brown.words()):
       print(item, end=" ")
16
   print("\n----")
17
18
19
   for item in search2('zzo', nltk.corpus.brown.words()):
       print(item, end=" ")
20
```

Palazzo Palazz

Palazzo Palazz

4. Procedural and declarative style

Example, count the average length of a token

```
In [23]:
```

```
#Procedural - average word length
tokens = nltk.corpus.brown.words(categories='news')
count = 0
total = 0
for token in tokens:
    count += 1
    total += len(token)
print(total / count)
```

4.401545438271973

In [24]:

```
#Declarative - average word length
tokens = nltk.corpus.brown.words(categories='news')
print (sum(len(t) for t in tokens)/len(tokens))
```

4.401545438271973

In [25]:

```
# Procedural - longest word
text = nltk.corpus.gutenberg.words('milton-paradise.txt')
longest = ''
for word in text:
   if len(word) > len(longest):
        longest = word
print(longest) # Print the longest word (ONLY one).
```

unextinguishable

In [26]:

```
# Procedural - longest word
maxlen = max(len(word) for word in text)# find longest word length
print([word for word in text if len(word) == maxlen]) # print those words
```

['unextinguishable', 'transubstantiate', 'inextinguishable', 'incompre hensible']

5. Using the map functions

```
In [27]:
```

```
# maps - applies a function to every item in a sequence
# Example - average lenghts of a sentence in the news Brown corpus

lengths = list(map(len, nltk.corpus.brown.sents(categories='news')))

print (sum(lengths) / len(lengths))

# the same as the following
lengths = [len(sent) for sent in nltk.corpus.brown.sents(categories='news')]

print(sum(lengths) / len(lengths))
```

21.75081116158339 21.75081116158339

6. Using lambda functions in maps

```
In [28]:
```

```
# count the number of vowels in each word
sent = ['Take', 'care', 'of', 'the', 'sense', ',', 'and', 'the', 'sounds', 'will
list(map(lambda w: len(list(filter(lambda c: c.lower() in "aeiou", w))), sent))
```

```
Out[28]:
[2, 2, 1, 1, 2, 0, 1, 1, 2, 1, 2, 2, 1, 3, 0]
```

7. Named arguments

When there are a lot of parameters for a function, it is possible to refer them by name, even using default values.

```
In [29]:
```

```
def pos tag(word, tag="UNK"):
2
       return {word: tag}
   print(pos tag("man", "NN"))
   print(pos tag("mann"))
   # *args, **kwargs ==> variable number of unamed and named/keywords arguments
   def freq words(file, min=1, num=10):
       text = open(file).read()
7
       tokens = nltk.word tokenize(text)
8
9
       freql = nltk.FreqDist(t for t in tokens if len(t) >= min)
       return list(freql.keys())[:num]
10
   print(freq words('data/news/news1.txt', 14, 5))
11
   print(freq words('data/news/news1.txt', min=14, num=5))
12
   print(freq words('data/news/news1.txt', num=5, min=14))
```

```
{'man': 'NN'}
{'mann': 'UNK'}
['ethnic-majority', 'characteristic', 'underdeveloped', 'Party-approve
d', 'popularization']
['ethnic-majority', 'characteristic', 'underdeveloped', 'Party-approve
d', 'popularization']
['ethnic-majority', 'characteristic', 'underdeveloped', 'Party-approve
d', 'popularization']
```

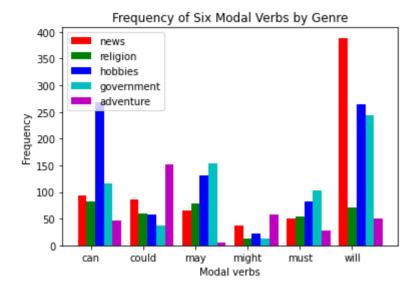
8. Matplotlib

Matplotlib is a Python 2D plotting library which produces high quality figures in a variety of formats and interactive environments across platforms. It can be used in Python scripts, the Python and IPython shells, the Jupyter notebook, web application servers.

```
# Draw a barchart for the 6 modal verbs from the brown corpus
   from numpy import arange
 3
   from matplotlib import pyplot
   %matplotlib inline
 4
   colors = 'rgbcmyk' # red, green, blue, cyan, magenta, yellow, black
   def bar chart(categories, words, counts):
       "Plot a bar chart showing counts for each word by category"
 7
       ind = arange(len(words))
8
9
       width = 1 / (len(categories) + 1)
10
       bar groups = []
       for c in range(len(categories)):
11
12
           bars = pyplot.bar(ind+c*width, counts[categories[c]], width,
                             color=colors(c % len(colors)))
13
14
           bar groups.append(bars)
       pyplot.xticks(ind+width, words)
15
       pyplot.legend([b[0] for b in bar_groups], categories, loc='upper left')
16
       pyplot.ylabel('Frequency');
17
       pyplot.xlabel('Modal verbs');
18
19
       pyplot.title('Frequency of Six Modal Verbs by Genre');
20
       pyplot.show();
   genres = ['news', 'religion', 'hobbies', 'government', 'adventure']
   modals = ['can', 'could', 'may', 'might', 'must', 'will']
2.2
   cfdist = nltk.ConditionalFreqDist(
23
       (genre, word)
2.4
25
       for genre in genres
26
       for word in nltk.corpus.brown.words(categories=genre)
27
       if word in modals)
   counts = {}
2.8
   for genre in genres:
       counts[genre] = [cfdist[genre][word] for word in modals]
30
   print(genres, modals, counts)
32
   bar chart(genres, modals, counts)
```

```
['news', 'religion', 'hobbies', 'government', 'adventure'] ['can', 'c
ould', 'may', 'might', 'must', 'will'] {'news': [93, 86, 66, 38, 50,
389], 'religion': [82, 59, 78, 12, 54, 71], 'hobbies': [268, 58, 131,
```

22, 83, 264], 'government': [117, 38, 153, 13, 102, 244], 'adventur e': [46, 151, 5, 58, 27, 50]}



~ Hands on Session ~

Experimenting with installing Spacy and important functions 10 minutes

Exercise 1 (5 points)

Submit befor the practice class of next week

- Count the number of tokens and types in data/news/news.txt. Use the split() method to split the text with white space
- 2. Count and show the words that are common between data/news/news.txt and data/news/news1.txt.
- 3. Print the largest word (s) for each letter (a,b,c ...z) from the nltk books corpus using declarative style. Example ... i-> incomprehensible, inextinguishable; t->transubstantiate, ...
- 4. Print the most frequent starting and ending characters of words in the nltk books.
- 5. Design an algorithm to find the statistically improbable phrases (SIP) of a document collection. You might just come up with a Pseudocode. SIPs are phrases that are less probably in a universal collection but common to a particular topic. For example, dependency parsing might be a common phrase that occur in computational linguistic journals but less frequent when compared with a global science journal indexes.

my name

solution 1

```
In [ ]:
```

```
In [ ]:
1
```

Materials

- spaCy (https://course.spacy.io/)
- NLTK book (http://www.nltk.org/book/).
- Handbook of NLP

(https://karczmarczuk.users.greyc.fr/TEACH/TAL/Doc/Handbook%20Of%20Natura