

PARTICULARS OF EXPERIMENTS PERFORMED

1. NLP MODELS

Tokenization :

```
>>> import nltk
>>> from nltk.tokenize import sent_tokenize, word_tokenize
>>> string = """ hi ! hello, how are you . """
>>> sent_tokenize(string)
o/p: ['hi!', 'hello, how are you.']

>>> string = """ hi ! how are you ? ? """
>>> sent_tokenize(string)
o/p: ['hi! how are you? ?']

>>> word_tokenize(string)
o/p: ['hi', '!', 'how', 'are', 'you', '?', '?']
```

Stopwords :

```
>>> nltk.download("stopwords")
o/p: [nltk_data] Downloading package stopwords to
[nltk_data] c:\users\y20cbu\appdata\roaming\nltk_data...
[nltk_data] Package stopwords is already up-to-date!
```

True

```
>>> from nltk.corpus import stopwords
>>> from nltk.tokenize import word_tokenize
>>> string = "ramu ate the apple !!"
>>> string1 = word_tokenize(string)
>>> string1
o/p: ['ramu', 'ate', 'the', 'apple', '!', '!']

>>> stop_words = set(stopwords.words("english"))
```

Regd. No. 490C3004

```

>>> filtered_list = []
>>> for word in words_in_quote:
....     if word.casefold() not in stop_words:
....         filtered_list.append(word)
>>> filtered_list = [
....     word for word in strings if word.casefold() not in stop_words
.... ]
>>> filtered_list
O/P: ['ramu', 'ate', 'apple', '!', '!']

```

Stemming :

```

>>> from nltk.stem import PorterStemmer
>>> from nltk.tokenize import word_tokenize
>>> stemmer = PorterStemmer()
>>> string = """ helping duplicating waiting vanishing washing
    blocker"""
>>> words1 = word_tokenize(string)
>>> words1
O/P: ['helping', 'duplicating', 'waiting', 'vanishing', 'washing', 'blocker']
>>> stemmed_words1 = [stemmer.stem(word) for word in words1]
>>> stemmed_words1
O/P: ['help', 'duPlic', 'wait', 'vanish', 'wash', 'blocker']

```

Tagging parts of speech:

```

>>> from nltk.tokenize import word_tokenize
>>> string = """ you wish to make an apple pie from scratch."""
>>> words = word_tokenize(string)
>>> import nltk

```

```
>>> nltk.pos_tag(words)
```

O/P: [('you', 'PRP'),

('wish', 'VBP'),

('to', 'TO'),

('make', 'VB'),

('an', 'DT'),

('apple', 'NN'),

('pie', 'NN'),

('from', 'IN'),

('scratch', 'NN'),

('.', '.'),

Lemmatizing:

```
>>> from nltk.stem import WordNetLemmatizer
```

```
>>> lemmatizer = WordNetLemmatizer()
```

```
>>> lemmatizer.lemmatize("scarves")
```

O/P: 'Scarf'

```
>>> string = "Aksh loves apples"
```

```
>>> words = word_tokenize(string)
```

```
>>> words
```

O/P: ['Aksh', 'loves', 'apples'] ✓

```
>>> lemmatized_words = [lemmatizer.lemmatize(word) for word in words]
```

```
>>> lemmatized_words
```

O/P: ['Aksh', 'love', 'apple'] ✓

2. Spell Checking:

Text Blob : (standard spell checker)

Cmd :

Pip install Textblob

Idle :

```
>>> from textblob import TextBlob
```

```
>>> fb = TextBlob("i want fotball")
```

```
>>> fb.correct()
```

O/p: TextBlob (" i want football")

```
>>> fb = TextBlob(" i want to paly fotball")
```

```
>>> fb.correct()
```

O/p: TextBlob (" i want to pale football")

```
>>> fb = TextBlob(" Python is eays to learn")
```

```
>>> fb.correct()
```

O/p: TextBlob (" platon is days to learn")

Custom Spell checker: (spello)

Cmd :

Pip install spello

Idle :

```
>>> from spello.model import SpellCorrectionModel
```

```
>>> sp = SpellCorrectionModel(language = "en")
```

→ Create a sample.txt file and save it in a notepad and copy the path of that file.

```
>>> with open("c:/users/y20cb004/Documents/sample.txt", "r") as file:  
    data = file.readlines()
```

```
data = [i.strip() for i in data]
```

data

O/P: [spell checking in nltk.', 'nlp experience some ambiguity at each level.', 'open cmd prompt.', 'open python and type:]

```
>>> Sp.train(data)
```

O/P: Spell Training started...

Context model Training started...

Symspell Training started...

Phoneme Training started...

Spell Training completed successfully.....

```
>>> Sp.spell_correct("spell chekcing in nltk")
```

O/P: {original-Text': 'spell chekcing in nltk', 'spell-corrected-text': 'spell checking in nltk', 'correction-dict': {'sepli': 'spell', 'chekcing': 'checking', 'nltk': 'nltk'}}

✓

3. Word Segmentation:

Cmd: Pip install spacy

Idle:

>>> import nltk

>>> from nltk.probability import FreqDist

>>> text = "Guntur District was formed on the 1st october, 1904 with Head Quarters at Guntur after bifurcating Krishna and Nellore districts. Prior to 1859 there was 'Guntur District' with Head Quarters at Guntur but with a different jurisdiction".

>>> text

O/P: 'Guntur District was formed on the 1st october, 1904 with Head Quarters at Guntur after bifurcating Krishna and Nellore districts. Prior to 1859 there was 'Guntur District' with Head Quarters at Guntur but with a different jurisdiction'.

>>> from nltk.tokenize import word_tokenize

>>> words = word_tokenize(text)

>>> words

O/P: ['Guntur', 'District', 'was', 'formed', 'on', 'the', '1st', 'october',
'1904', 'with', 'Head', 'Quarters', 'at', 'Guntur', 'after',
'bifurcating', 'Krishna', 'and', 'Nellore', 'districts', '.', 'Prior', 'to',
'1859', 'there', 'was', ' ', 'Guntur', 'District', ' ', 'with', 'Head',
'Quarters', 'at', 'Guntur', 'but', 'with', 'a', 'different',
'jurisdiction', '.']

with respect to

Prop. Melius et al.

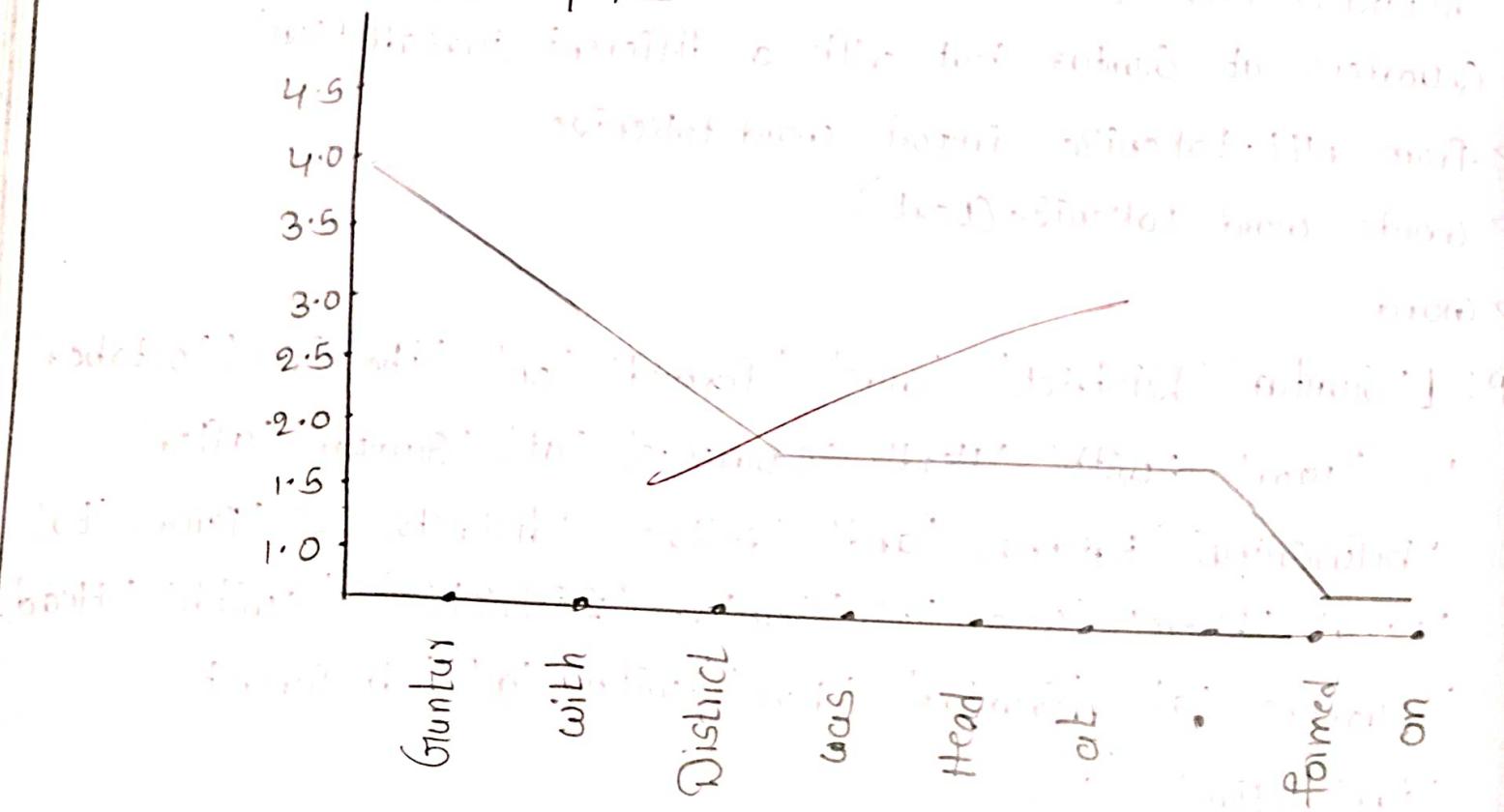
101

The frequency

of word endings for different words are plotted in Fig. 10.1
with the number of syllables often added. It can be seen from
this figure that words with more syllables tend to have
multiple forms in the last column. In contrast both

the very short words have one ending. Is there any relationship between the number of syllables often added to the ending before

Word Segmentation Using plot: our word was segmented by



```
>>> len(words)
```

O/P: 42

Word Frequency Distribution:

```
>>> FreqDist(words)
```

O/P: FreqDist({'Guntur': 4, 'with': 3, 'District': 2, 'was': 2, 'Head': 2, 'Quarters': 2, 'at': 2, '.': 2, 'formed': 1, 'on': 1, ...})

```
>>> word1 = FreqDist(words)
```

```
>>> word1
```

O/P: FreqDist({'Guntur': 4, 'with': 3, 'District': 2, 'was': 2, 'Head': 2, 'Quarters': 2, 'at': 2, '.': 2, 'formed': 1, 'on': 1, ...})

```
>>> print(word1)
```

O/P: <FreqDist with 31 samples and 42 outcomes>

```
>>> len(FreqDist(words))
```

O/P: 31

```
>>> word1.most_common(5)
```

O/P: [('Guntur', 4), ('with', 3), ('District', 2), ('was', 2), ('Head', 2)]

```
>>> word1.freq("at")
```

O/P: 0.0476190

```
>>> word1.freq("mlt")
```

O/P: 0.0

Word Segmentation using plot:

```
>>> import matplotlib.pyplot as plt
```

```
>>> fig, ax = plt.subplots(figsize=(10, 5))
```

```
>>> word1.plot(10)
```

Sentence Segmentation:

Cmd:

pip install spacy

python -m spacy download en-core-web-lg

Idle:

>>> import spacy

>>> import nltk

>>> n = spacy.load("en-core-web-lg")

>>> n

O/p: <spacy.lang.en.English object at 0x00000147D1BF89D0>

>>> doc = n("This is First line. This is second line")

>>> doc

O/p: This is first line. This is second line

>>> for sent in doc.sents:

.... print(sent.text)

O/p: This is First line

This is second line

4. Count Vector Class

Cmd:

Pip install scikit-learn Scipy matplotlib numpy
(or)

Pip install sklearn

Idle:

>>> import sklearn
>>> from sklearn.feature_extraction.text import CountVectorizer

>>> word1 = ["This is first line, this is second line, this is third line"]

>>> word1

O/P: ['This is first line, this is second line, This is third line']

>>> vectorizer = CountVectorizer()

>>> vectorizer.fit(word1)

O/P: CountVectorizer()

>>> X = vectorizer.fit_transform(word1)

>>> X

O/P: <1x6 sparse matrix of type '<class 'numpy.int64'>' with 6 stored elements in Compressed Sparse Row format>

>>> vectorizer.vocabulary-

O/P: {'this': 5, 'is': 1, 'first': 0, 'line': 2, 'second': 3, 'third': 4}

>>> vectorizer.vocabulary_.get("this")

O/P: 5

>>> vectorizer.get_feature_names()

O/P: ['first', 'is', 'line', 'second', 'third', 'this']

```
>>> x.toarray()
```

O/p: array([[1, 3, 3, 1, 1, 3]], dtype=int64)

cmd:

Pip install pandas

idle:

```
>>> import pandas as pd
```

```
>>> matrix = pd.DataFrame(x.toarray(), columns=Vectorizer.get_feature_names())
```

```
>>> matrix
```

O/p: first is line second third this

0	1	3	3	1	1	3
---	---	---	---	---	---	---

de

5.

N-Gram Model :-

Idle:

```
>>> import sklearn
>>> from sklearn.feature_extraction.text import CountVectorizer
>>> text = ["this is my new sentence"]
>>> n_gram = CountVectorizer(ngram_range=(2,2))
>>> trans = n_gram.fit_transform(text)
>>> trans
```

O/P: <1x4 sparse matrix of type '<class 'numpy.int64'>' with
4 stored elements in Compressed Sparse Row format>

>>> n_gram.vocabulary-

O/P: {'this': 0, 'is': 1, 'my': 2, 'new': 3}

>>> trans.toarray()

O/P: array([[1, 1, 1, 0]], dtype=int64)

```
>>> n_gram = CountVectorizer(ngram_range=(3,3))
```

```
>>> trans = n_gram.fit_transform(text)
```

>>> trans

O/P: <1x3 sparse matrix of type '<class 'numpy.int64'>' with
3 stored elements in Compressed Sparse Row format>

>>> n_gram.vocabulary-

O/P: {'this': 0, 'is': 1, 'my': 2, 'new': 3, 'sentence': 4}

>>> trans.toarray()

O/P: array([[1, 1, 0]], dtype=int64)

```
>>> n_gram = CountVectorizer(ngram_range=(1,3))
```

```
>>> trans = n_gram.fit_transform(text)
```

>>> Trans

O/p: <1x12 Sparse matrix of type '<class 'numpy.int64'>' with
12 stored elements in Compressed Sparse Row format>

>>> n-gram. vocabulary-

O/p: { 'this': 9, 'is': 0, 'my': 3, 'new': 6, 'sentence': 8, 'this is': 10,
'is my': 1, 'my new': 4, 'new sentence': 7, 'this is my': 11, 'is
my new': 2, 'my new sentence': 5 }

>>> Trans.toarray()

O/p: array([[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1]], dtype=int64)

>>> n-gram. vocabulary_.get("this is")

O/p: 10

✓

Bag of Words :

>>> trans.shape

o/p: (1, 12)

>>> t = (" i am learning nlp")

>>> t

o/p: [' i am learning nlp']

>>> n-gram.transform(t).toarray()

o/p: array([0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0]), dtype=int64

>>> t = [" i am my learning nlp"]

>>> n-gram.transform(t).toarray()

o/p: array([[0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0]], dtype=int64)

>>> trans = n-gram.fit_transform(text+t)

>>> trans

o/p: < 9x20 sparse matrix of type '<class 'numpy.int64'>' with
91 stored elements in Compressed Sparse Row format >

>>> n-gram.vocabulary-

o/p: {'This':17, 'is':3, 'my': 8, 'new':13, 'sentence':16, 'this is':15,
'is my':4, 'my new':11, 'new sentence':14, 'this is my':19, 'is my
new':5, 'my new sentence':12, 'am':0, 'learning':6, 'nlp':15,
'am my':1, 'my learning':9, 'learning nlp':7, 'am my learning':2,
'my learning nlp':10}

>>> n-gram.transform(text+t).toarray()

o/p: array([0, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1], [1, 1, 1, 0, 0, 0,
1, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0]), dtype=int64

Inverse Transformations:

Idle:
 >>> import sklearn

>>> from sklearn.feature-extraction.text import CountVectorizer
 >>> word1 = ["Welcome to CSBS, Welcome to third year, Welcome to Natural Language Processing"]
 >>> word1

O/P: ["Welcome to CSBS, Welcome to third year, Welcome to Natural Language Processing"]

>>> vectorizer = CountVectorizer()

>>> vectorizer.fit(word1)

O/P: CountVectorizer()

>>> X = vectorizer.fit_transform(word1)

>>> X

O/P: <1x8 sparse matrix of type '<class 'numpy.int64'>' with 8 stored elements in Compressed Sparse Row format>

>>> vectorizer.vocabulary-

O/P: {'welcome': 6, 'to': 5, 'csbs': 0, 'third': 4, 'year': 7, 'natural': 2, 'language': 1, 'processing': 3}

>>> vectorizer.inverse_transform(X)

O/P: [array(['Welcome', 'to', 'csbs', 'third', 'year', 'natural', 'language', 'processing'], dtype='|<U10'])]

>>> from sklearn.feature-extraction.text import TfidfVectorizer

>>> import pandas as pd

>>> Corpus = ["This is first sentence", "This is second sentence", "This is third sentence"]

>>> Corpus

o/p: ['This is first sentence', 'This is second sentence', 'This is third sentence']

>>> Corpus[0]

o/p: 'This is first sentence'

>>> Corpus[1]

o/p: 'This is second sentence'

>>> Vectorizer = TfidfVectorizer()

>>> X = vectorizer.fit_transform(Corpus)

>>> X

o/p: <3x6 sparse matrix of type '<class 'numpy.float64'>' with 12 stored elements in Compressed Sparse Row format>

>>> vectorizer.vocabulary_-

o/p: {'this': 5, 'is': 1, 'first': 0, 'sentence': 3, 'second': 2, 'third': 4}

>>> vectorizer.get_feature_names()

o/p: ['first', 'is', 'second', 'sentence', 'third', 'this']

>>> vectorizer.idf_-

o/p: array([1.69314718, 1., 1.69314718, 1., 1.69314718, 1.])

>>> matrix = pd.DataFrame(X.toarray(), columns=vectorizer.get_feature_names())

>>> matrix

o/p:

	first	is	Second	sentence	third	this
0	0.69903	0.412859	0.00000	0.412859	0.00000	0.412859
1	0.00000	0.412859	0.69903	0.412859	0.00000	0.412859
2	0.00000	0.412859	0.00000	0.412859	0.69903	0.412859

X

6. Pos Tagging :

Idle :

```
>>> import nltk
>>> nltk.help.upenn_tagset("MD")
```

O/p: MD: modal auxiliary

Can Cannot Could Couldn't dare may might must need
 ought shall should shouldn't will would.

```
>>> text = "The Python programming language provides a wide range  

  of tools and libraries for attacking specific NLP tasks"
```

```
>>> text
```

O/p: 'The python programming language provides a wide range of tools
 and libraries for attacking specific NLP tasks.'

```
>>> words = nltk.word_tokenize(text)
```

```
>>> words
```

O/p: ['The', 'python', 'programming', 'language', 'provides', 'a', 'wide',
 'range', 'of', 'tools', 'and', 'libraries', 'for', 'attacking', 'specific',
 'NLP', 'tasks', ':']

```
>>> len(words)
```

O/p: 18

```
>>> nltk.pos_tag(words)
```

O/p: [('The', 'DT'), ('Python', 'NNP'), ('programming', 'NN'), ('language',
 'NN'), ('provides', 'VBZ'), ('a', 'DT'), ('wide', 'JJ'), ('range', 'NN'),
 ('of', 'IN'), ('tools', 'NNS'), ('and', 'CC'), ('libraries', 'NNS'), ('for',
 'IN'), ('attacking', 'VBG'), ('specific', 'JJ'), ('NLP', 'NNP'), ('tasks',
 'NNS'), ('.', '.')]

```
>>> nltk.help.brown_tagset()
```

O/P:

(: opening parenthesis
)): closing parenthesis
)*: negator
not n't

,: Comma

--: dash

.: Sentence Terminator

:: colon ? ; ! :

ABN: determiner / pronoun, pre-qualifier

all half many many

WRB + MD: WH-adverb + modal auxiliary
where'd

>>> nltk.help.upenn_tagset("NNP")

O/p: NNP: noun, proper, singular

McLawn Venneboerger (zestochwa Ranjei Conchita Triumplane
Christas oceanside Escobar Kreisler Sawyer Cougar Yvette Eirin
ODI Darryl CTCA shannon A.K.C. Heltex Liverpool....

>>> nltk.help.upenn_tagset("CC")

O/p: CC: Conjunction, Coordinating

& 'n and both but either etc for less minus
neither nor or plus so therefore times v.
versus vs. whether yet

ofc

Multiword Expressions

```
>>> import nltk  
>>> nltk.download('punkt')  
[nltk-data] Downloading package punkt to  
[nltk-data] C:\Users\y90cbyl\AppData\Roaming\nltk_data....  
[nltk-data] Package punkt is already up-to-date!  
True  
>>> from nltk.tokenize import MWETokenizer  
>>> from nltk.tokenize import word_tokenize  
>>> mwe = MWETokenizer([('new', 'york'), ('Hang', 'Kong'), ('takes', 'off')],  
separator = "-")  
>>> text1 = "the flight is going from new york to Hang Kong"  
>>> text2 = "The plane takes off from new york at 10 am"  
>>> mwe1 = mwe.tokenize(word_tokenize(text1))  
>>> mwe2 = mwe.tokenize(word_tokenize(text2))  
>>> mwe1  
[('the', 'flight', 'is', 'going', 'from', 'new', 'york', 'to', 'Hang',  
'Kong')]  
>>> mwe2  
[('the', 'plane', 'takes-off', 'from', 'new', 'york', 'at', '10am')]
```

✓ 9/0

Cleaning Text Data

cmd:

Pip install cleantext

Idle:

```
>>> import cleantext
>>> from cleantext import clean
```

Ascii Code:

```
>>> s1 = "Ko!u017euluo161\ud101dek"
```

```
>>> clean(s1, to_ascii='True')
```

O/p: 'kozusadek'

LowerCase:

```
>>> s2 = " my name is Chandu"
```

```
>>> clean(s2, lower=True)
```

O/p: 'my name is chandu'

URL Replacement:

```
>>> s3 = " https://www.google.com has surpassed https://www.Bing.com in
      search volume"
```

```
>>> clean(s3, no_urls=True; replace_with_url="URL")
```

O/p: 'URL has surpassed URL in search volume'

Replace Currency:

```
>>> s4 = " i want $40"
```

```
>>> clean(s4, no_currency_symbols=True)
```

O/p: 'i want <cur>40'

```
>>> clean(s4, no_currency_symbols=True; replace_with_currency_symbol
      ="Rupees")
```

O/p: 'i want rupees40'

No punctuations?
 >>> S5 = "40,000 is more than 30,000"

>>> clean(S5, no-punct = 'True')

O/p: '40000 is more than 30000'

>>> clean(S5, no-punct = 'True', replace-with-punct = "@")

O/p: '40@000 is more than 30@000'

Digit Replacement:

>>> S6 = "hleghly34345676jyhtuiop"

>>> clean(S6, no-digits = 'True')

O/p: 'hleghly000000000jyhtuiop'

>>> clean(S6, no-digits = 'True', replace-with-digit = '')

O/p: 'hleghly jyhtuiop'

Unicode:

>>> S7 = "z\u00fcrick"

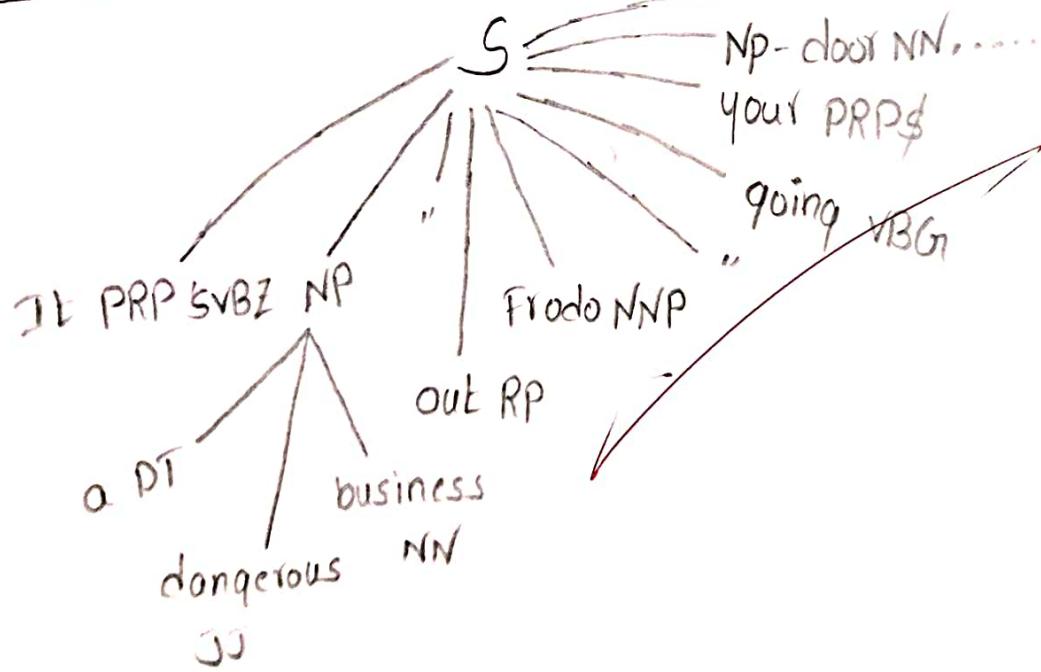
>>> clean(S7, fix-unicode = True)

O/p: 'zuwick'

• probably
• fix-unicode

✓
✓

q. chunking:
output -



q. Chunking:

Idle:

>>> import nltk

>>> from nltk.tokenize import word_tokenize

>>> string = "It's a dangerous business, Frodo, going out your door."

>>> words = word_tokenize(string)

>>> words

O/p: ['it', "'s", 'a', 'dangerous', 'business', "'", 'Frodo', "'going',
'out', 'your', 'door', '.']

>>> pos = nltk.pos_tag(words)

>>> pos

O/p: [('it', 'PRP'), ("'s", 'VBZ'), ('a', 'DT'), ('dangerous', 'JJ'),
(('business', 'NN'), (',', ','), ('Frodo', 'NNP'), (',', ','),
(('going', 'VBG'), ('out', 'RP'), ('your', 'PRP') ('door', 'NN'),
(',', ','))]

>>> grammar = "NP:{<DT>? <JJ*> <NN>}"

>>> chunk_Parser = nltk.RegexpParser(grammar)

>>> tree = chunk_Parser.parse(pos)

>>> tree.draw()

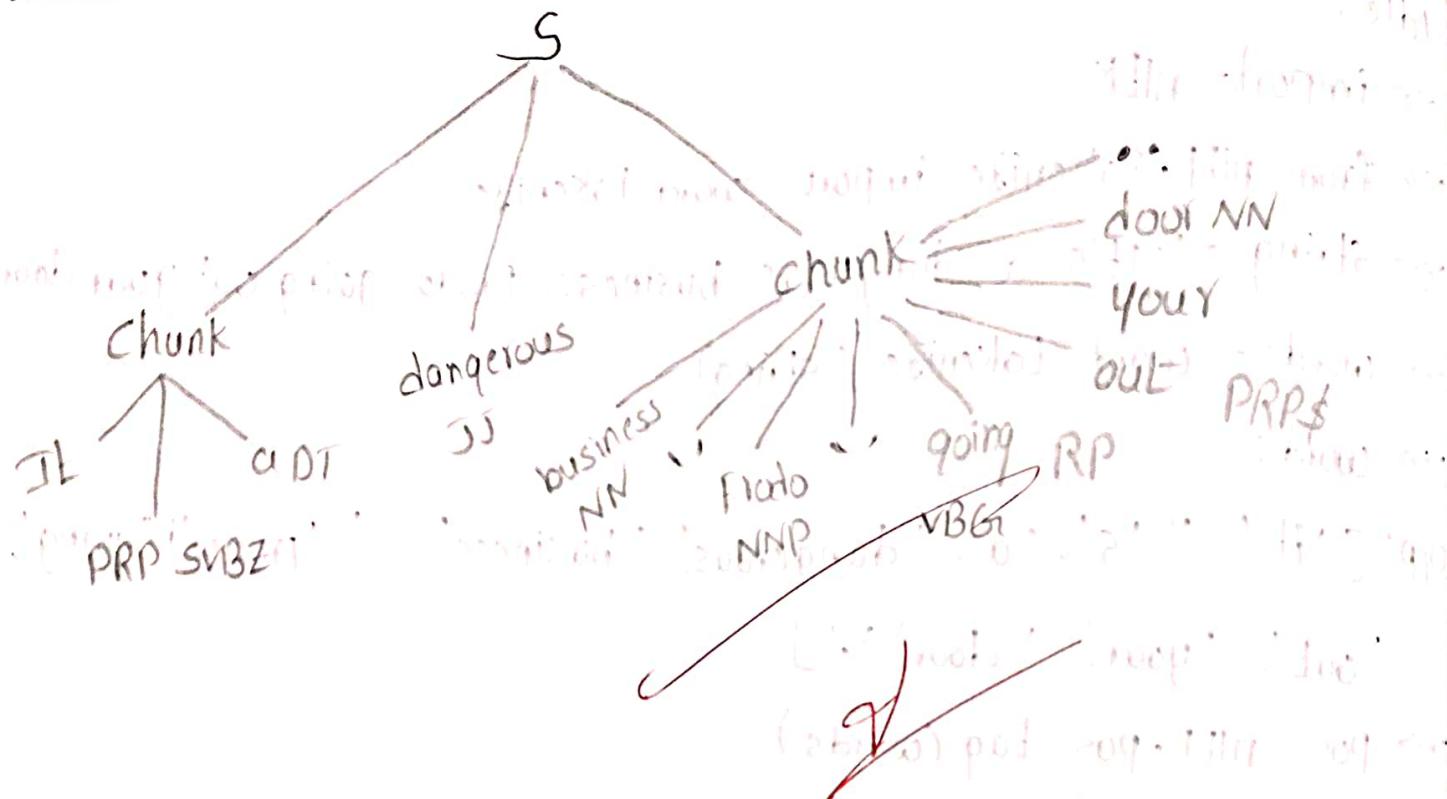
Chinking:

>>> pos

>>> grammar = " " "

Chinking :

output -



chunk : { <.*> + }

 |
 y <JJ| " "

>>> chunk - parser = nltk. RegexParser(grammer)

>>> tree = chunk - parser . Parse(pos)

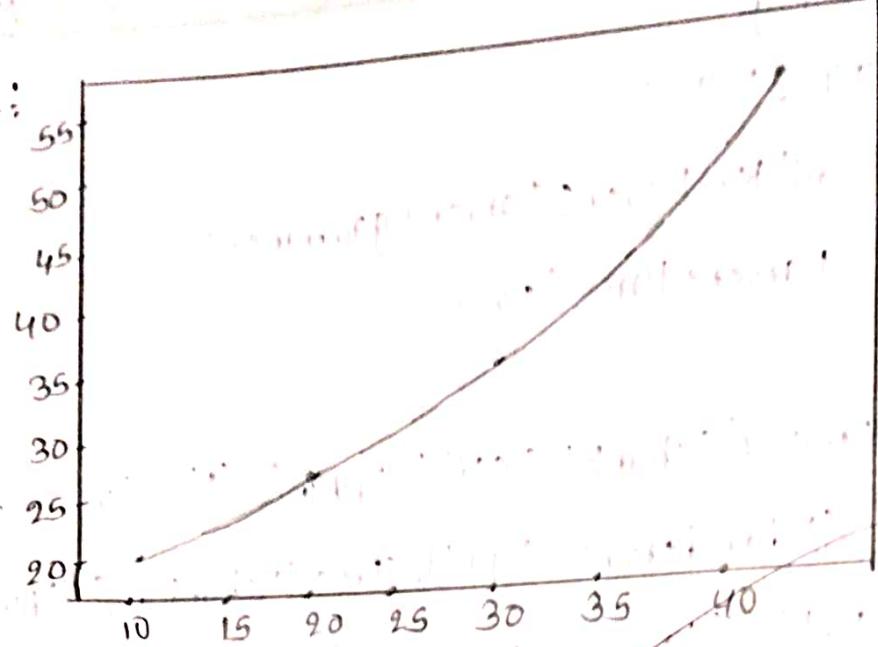
>>> tree

O/p: Tree ('s', [Tree ('chunk', [(('iE', 'PRP'), ('s', 'VBZ'),
 ('a', 'DT')), ('dangerous', 'JJ'), Tree ('chunk', [('business',
 'NN'), ('.', ','), ('Frodo', 'NNP'), ('.', ','), ('going', 'VBD'),
 ('out', 'RP'), ('your', 'PRP\$'), ('door', 'NN'), ('.', '.')])])])

>>> tree.draw()

10.

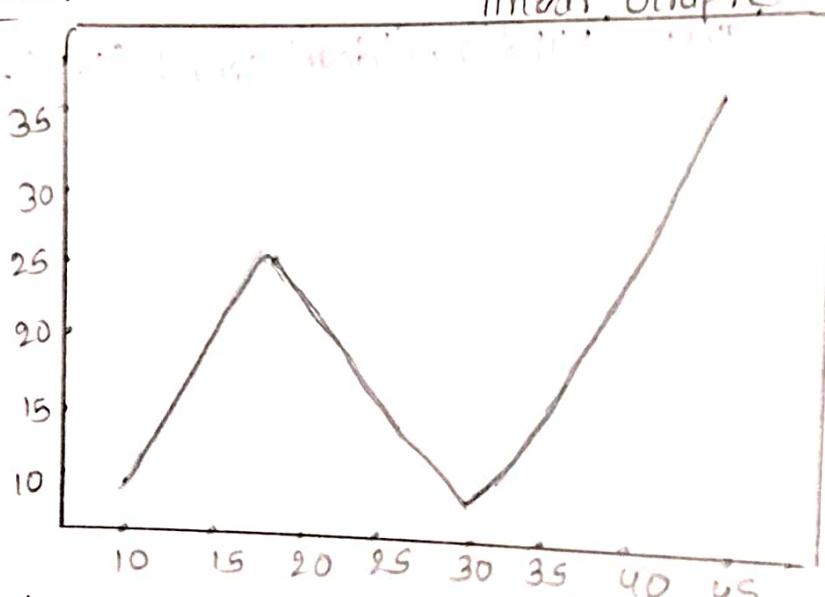
output:



Adding Title:

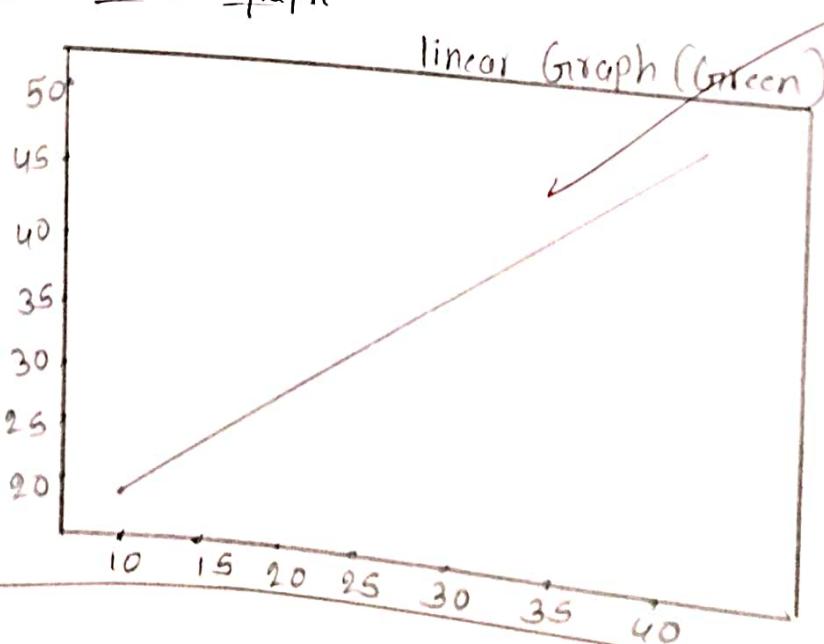
linear Graph

output:



Adding color to the graph

output:



Matplot Library:

Cmd:

Pip install matplotlib

Idle:

>>> import matplotlib.pyplot as plt

>>> x = [10, 20, 30, 40]

>>> y = [20, 25, 35, 55]

>>> plt.plot(x, y)

O/p: [<matplotlib.lines.Line2D object at 0x00016>]

>>> plt.show()

Adding title:

>>> x = [10, 20, 30, 40]

>>> y = [20, 25, 30, 35]

>>> plt.plot(x, y)

O/p: [<matplotlib.lines.Line2D object at 0x002>]

>>> plt.title("linear Graph")

O/p: Text(0.5, 1.0, 'linear Graph')

>>> plt.show()

Adding color to the graph:

>>> x = [10, 20, 30, 40]

>>> y = [20, 30, 40, 50]

>>> plt.plot(x, y)

O/p: [<matplotlib.lines.Line2D object at 0x0002>]

>>> plt.title("linear Graph", fontsize=25, color="green")

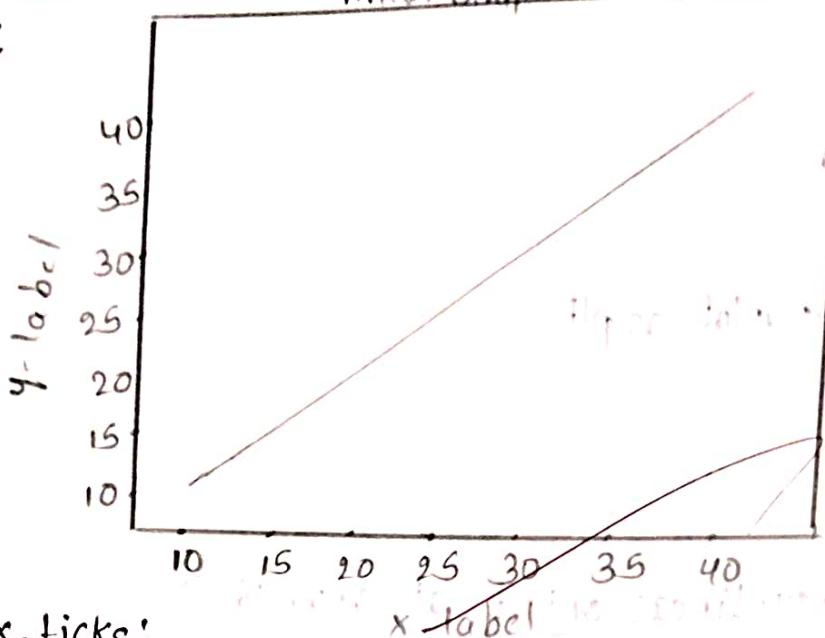
O/p: Text(0.5, 1.0, 'linear Graph')

>>> plt.show()

Adding labels:

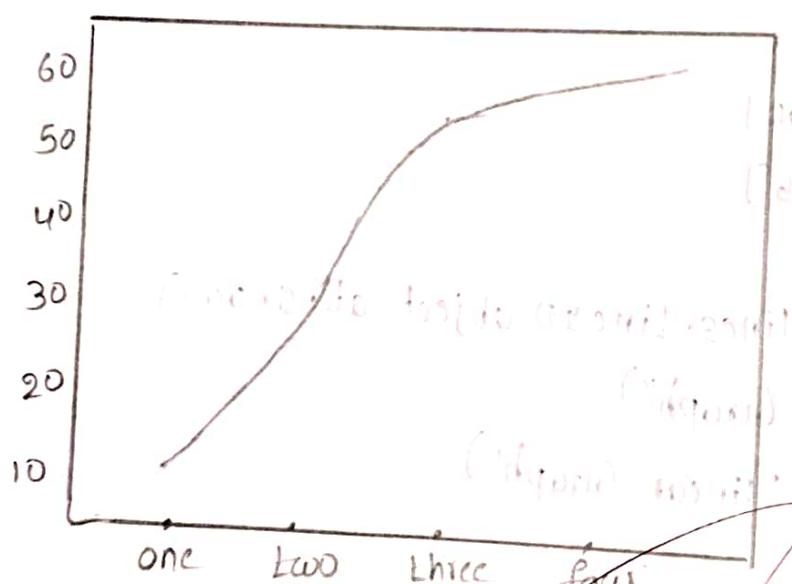
linear Graph

output:



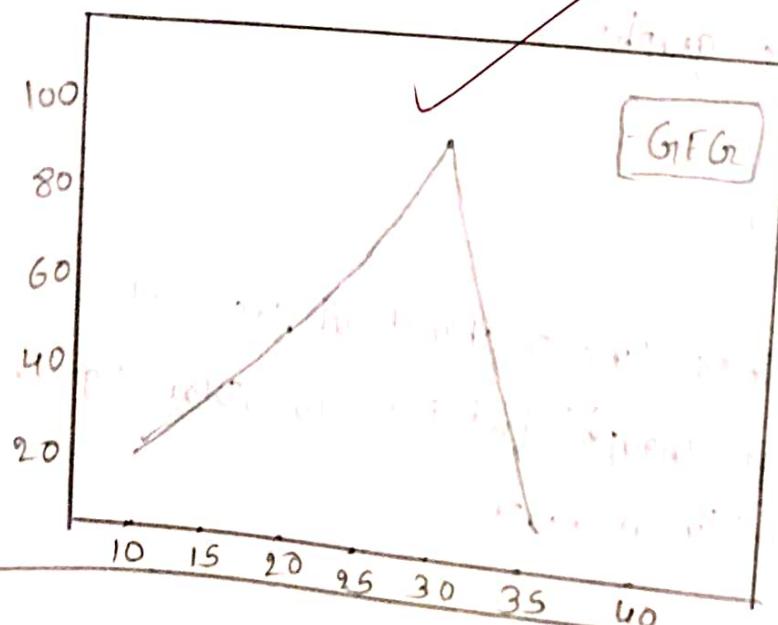
ylim and x-ticks:

output:



legend:

output:



Adding labels:

```
>>> x = [10, 20, 30, 40]
>>> y = [20, 30, 40, 50]
>>> plt.plot(x, y)
>>> plt.title('linear Graph', fontsize=25, color="green")
>>> plt.xlabel('x-label')
>>> plt.ylabel('y-label')
>>> plt.show()
```

ylim and xticks:

```
>>> x = [10, 20, 30, 40]
>>> y = [20, 30, 40, 50]
>>> plt.ylim(0, 60)
>>> plt.xticks(x, labels=["one", "two", "three", "four"])
>>> plt.plot(x, y)
>>> plt.show()
```

legend:

```
>>> x = [10, 20, 30, 40]
>>> y = [20, 34, 65, 34]
>>> plt.plot(x, y)
>>> plt.title('Linear Graph')
>>> plt.xlabel('x-label')
>>> plt.ylabel('y-label')
>>> plt.legend(['G1/G2'])
>>> plt.show()
```

Figure class:

```
>>> import matplotlib.pyplot as plt
>>> from matplotlib.figure import Figure
>>> x = [10, 20, 30, 40]
>>> y = [20, 25, 35, 55]
# Creating a new figure with width=7, inches and height=5 inches
```

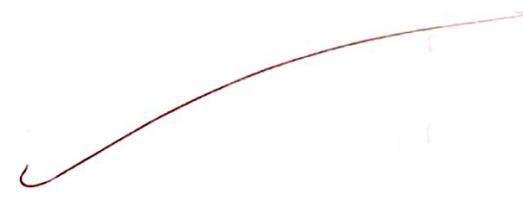
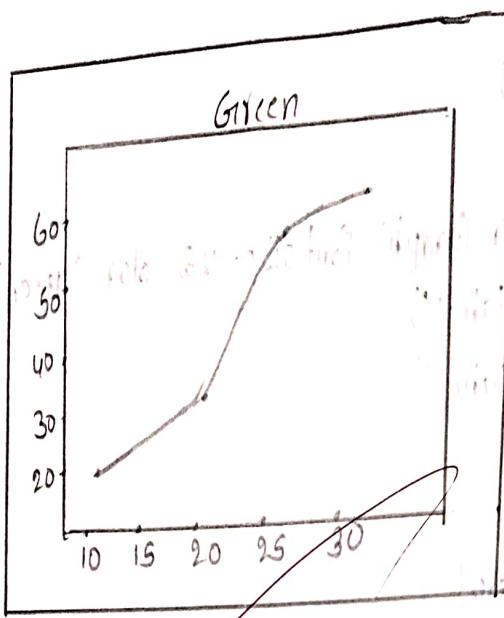


Figure class:

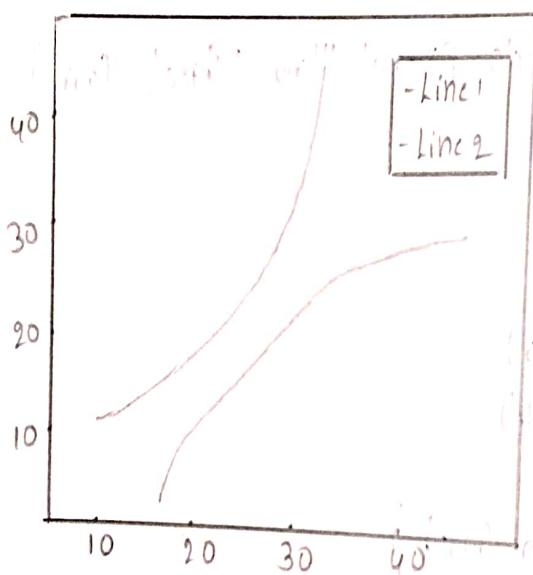
Output:



→ Blue color
x-axis label

Multiple plots:

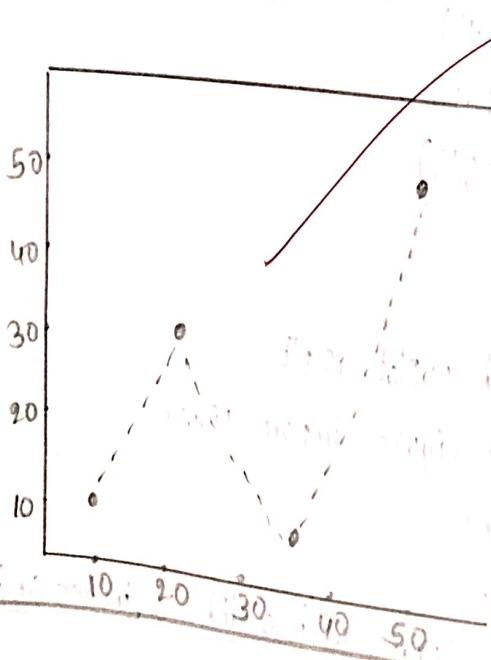
Output:



color of set 1
color of set 2

Linechart:

Output:



color of set 1
color of set 2

with face color as Green, edgecolor as red and line width as 7.

```
>>> fig = plt.figure(figsize=(7,5), facecolor='g', edgecolor='r'; linewidth=7)
```

Creating new axes for the figure

```
>>> ax = fig.add_axes([1,1,1,1])
```

Adding the data to be plotted

```
>>> ax.plot(x,y)
```

Multiple plots :

```
>>> x = [10, 20, 30, 40]
```

```
>>> y = [10, 20, 20, 30]
```

```
>>> fig = plt.figure(figsize=(5,4))
```

```
>>> ax = fig.add_axes([1, 1, 1, 1])
```

```
>>> ax1 = ax.plot(x,y)
```

```
>>> ax2 = ax.plot(y,x)
```

```
>>> ax.set_title("Multiple Graphs")
```

```
>>> ax.set_xlabel("x-Axis")
```

```
>>> ax.set_ylabel("y-Axis")
```

```
>>> ax.legend(labels=['Line 1', 'Line 2'])
```

```
>>> plt.show()
```

Line chart :

```
>>> import matplotlib.pyplot as plt
```

```
>>> x = [10, 20, 30, 40]
```

```
>>> y = [20, 25, 30, 35]
```

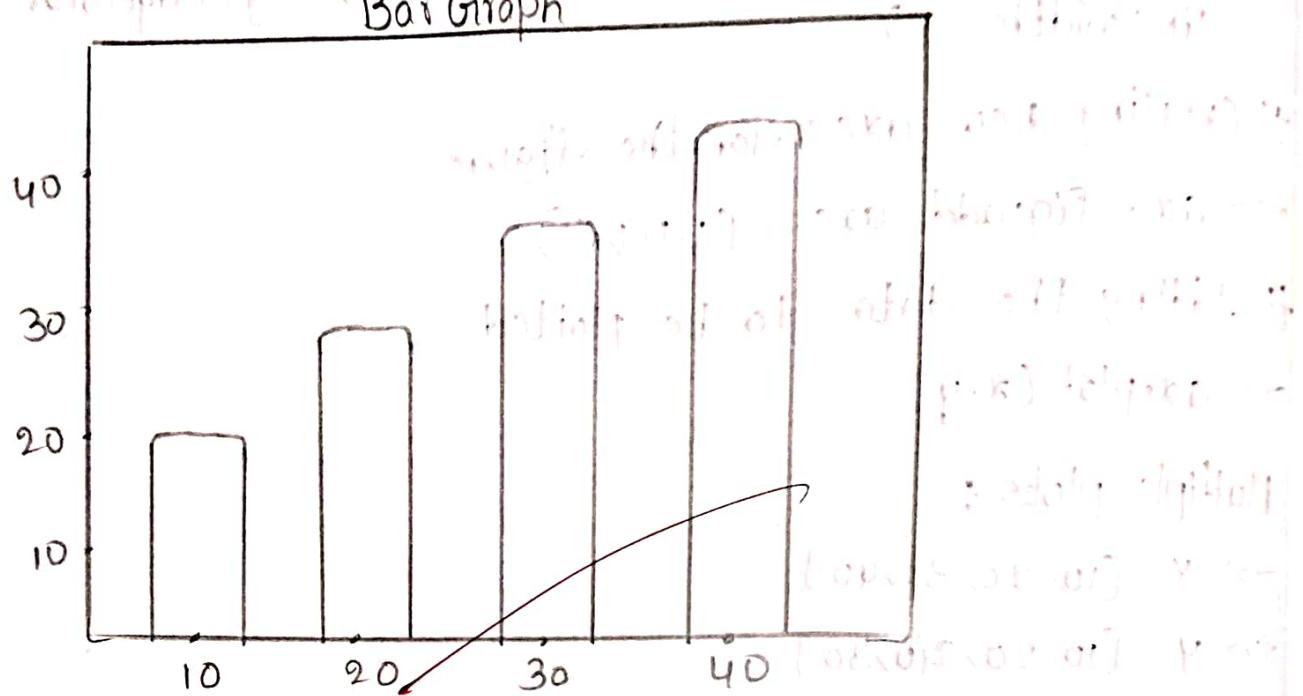
 , color='green', linewidth=3, marker='o', markersize=15,

```
>>> plt.plot(x,y, linestyle='--')
```

```
>>> plt.title("Line chart")
```

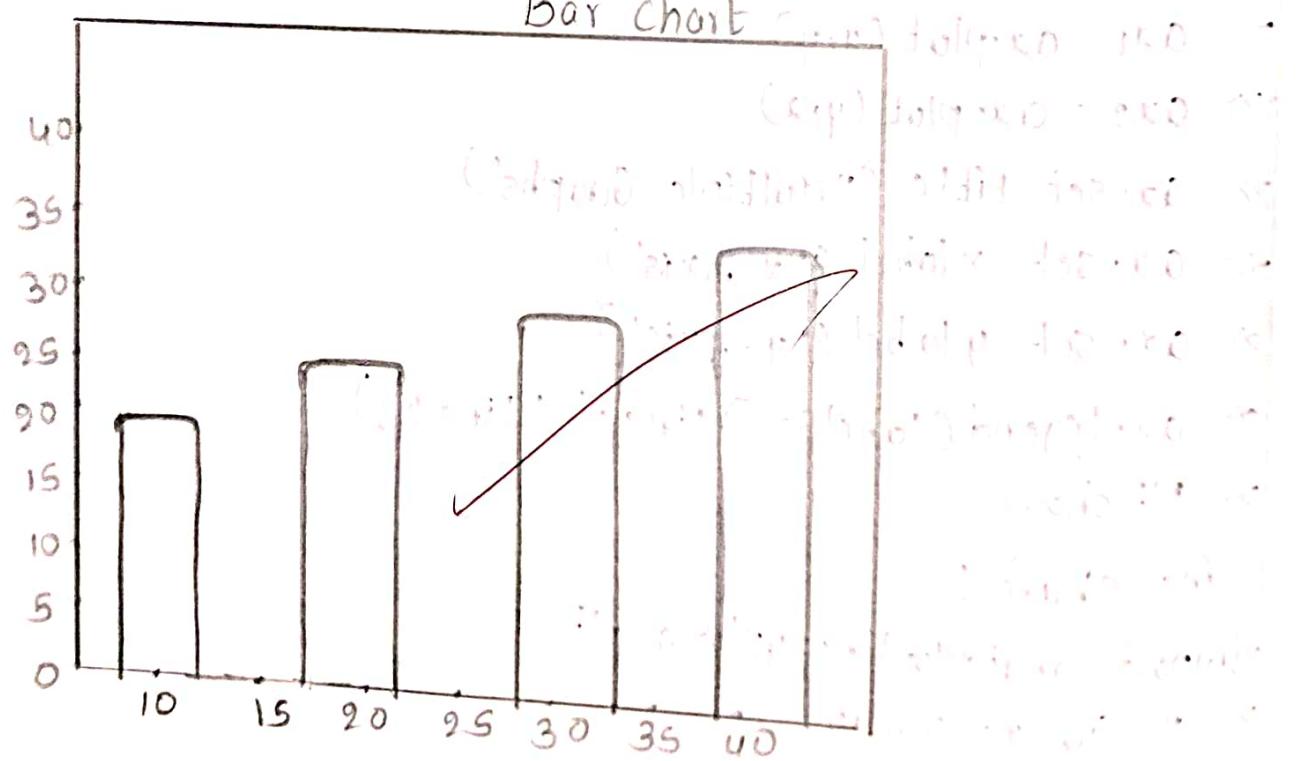
```
>>> plt.xlabel("x-axis")
```

Bar Graph: will have bars on one side and axis on other side.
Output: velocities [m/s] vs time [s] output from graph



Bar chart:

Output:



```
plt.ylabel("y-axis")  
plt.show()
```

Bar Graph:

```
>>> import matplotlib.pyplot as plt  
>>> x = [10, 20, 30, 40]  
>>> y = [20, 30, 40, 50]  
>>> plt.bar(x, y, color='green', width=3)  
>>> plt.title("Bar Graph")  
>>> plt.show()
```

BarChart:

```
>>> import matplotlib.pyplot as plt  
>>> x = [10, 20, 30, 40]  
>>> y = [20, 30, 40, 50]  
>>> plt.bar(x, y, color="green", width=3, edgecolor="blue", linewidth=2)  
>>> plt.title("Barchart")  
>>> plt.show()
```

✓ ✓

HMM Learn:

Cmd:

pip install hmmLearn

Idle:

```
>>> import nltk
>>> import hmmLearn
>>> import numpy as np
>>> from hmmLearn.hmm import CategoricalHMM
>>> startprob = np.array([0.5, 0.5])
>>> Transmat = np.array([[0.7, 0.3], [0.3, 0.7]])
>>> Covar = np.array([[0.9, 0.1], [0.2, 0.8]])
>>> model = CategoricalHMM(n_components=2, startprob=startprob,
   Transmat=Transmat)
>>> X = [[0, 1, 0, 1], [0, 0, 0, 0], [1, 1, 1, 1], [1, 0, 0, 0]]
>>> Model.fit(X)

o/p: CategoricalHMM(n_components=2, random_state=RandomState
(MT 19937) at 0x19E11586640,
startprob=prior = array([0.5, 0.5])
Transmat=prior = array([[0.7, 0.3], [0.3, 0.7]]))

>>> print(model.Transmat_)

o/p: [[ 0.2800  0.7199]
      [ 0.594   0.4053]]

>>> prob = model.decode(np.array([0, 1, 0, 1]).reshape(4, 1), lengths
= None, algorithm=None)
```

```
>>> print(prob)
```

O/P: [-9.941, array([1, 0, 1, 0], dtype=int64)]

```
>>> print(np.exp(prob[0]))
```

O/P: 0.0527

```
>>> x, z = model.sample(10)
```

```
>>> print(x)
```

O/P: [[0]

[0]

[1]

[0]

[1]

[0]

[0]

[1]

[0]

[0]

[0])



```
>>> print(z)
```

O/P: [1 0 1 0 0 1 0 1]



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12. Word Embedding in NLP (Word2Vec)

Cmd:

PIP install gensim

Idle:

```
>>> import gensim  
>>> from gensim.models import Word2Vec  
>>> import warnings  
>>> warnings.filterwarnings('ignore')  
>>> st = [['this', 'is', 'sample1'], ['this', 'is', 'sample2'], ['this', 'is',  
         'sample3']]  
>>> print(st)
```

O/p: [[['this', 'is', 'sample1'],
 ['this', 'is', 'sample2'],
 ['this', 'is', 'sample3']]]

```
>>> model = Word2Vec(st, min_count=1)
```

```
>>> print(model)
```

O/p: Word2Vec(vocab=5, vector_size=100, alpha=0.025)

```
>>> words = list(model.wv.key_to_index)
```

```
>>> print(words)
```

O/p: ['is', 'this', 'sample3', 'sample2', 'sample1']

✓