In [1]:

```
#stopword
'''Removing stop words with NLTK in Python.
The process of converting data to something a computer can understand is referred to as pre
One of the major forms of pre-processing is to filter out useless data.
In natural language processing, useless words (data), are referred to as stop words.'''
import nltk
nltk.download('stopwords')
from nltk.corpus import stopwords
stopwords.words('english')
[nltk_data] Downloading package stopwords to
                C:\Users\91814\AppData\Roaming\nltk_data...
[nltk_data]
[nltk_data]
              Package stopwords is already up-to-date!
In [2]:
```

```
#CMU wordlist : CMUdict is a versioned python wrapper package for The CMU Pronouncing Dict
#The main purpose is to expose the data with little or no assumption on how it is to be use
import nltk
nltk.download('cmudict')
import nltk
entries=nltk.corpus.cmudict.entries()
len(entries)
```

[nltk data] Downloading package cmudict to

C:\Users\91814\AppData\Roaming\nltk_data... [nltk_data]

[nltk_data] Package cmudict is already up-to-date!

Out[2]:

133737

```
In [3]:
```

```
from nltk.corpus import wordnet as wn
'''Look up a word using synsets();
this function has an optional pos argument which lets you constrain the part of speech of t
wn.synsets('motorcar')
Out[3]:
[Synset('car.n.01')]
In [4]:
wn.synset('car.n.01').lemma_names()
Out[4]:
['car', 'auto', 'automobile', 'machine', 'motorcar']
In [5]:
#TASK CLASSIFIER
def gender_features(word):
    return {'last_letter':word[-1]} #extracting the last letter of the word. Which can be
In [6]:
gender_features('obama')
Out[6]:
{'last_letter': 'a'}
In [7]:
import nltk
nltk.download('names')
from nltk.corpus import names
labeled_names = ([(name, 'male') for name in names.words('male.txt')]+
[(name, 'female') for name in names.words('female.txt')])
[nltk_data] Downloading package names to
[nltk data]
                C:\Users\91814\AppData\Roaming\nltk data...
[nltk_data]
              Package names is already up-to-date!
In [8]:
import random
random.shuffle(labeled names)
In [9]:
featuresets=[(gender_features(n),gender) for (n,gender) in labeled_names]
In [10]:
train_set,test_test=featuresets[500:],featuresets[:500] #seprating test and train data.
```

```
In [11]:
import nltk
classifier=nltk.NaiveBayesClassifier.train(train set) #using NaiveBayes classifier and trai
In [12]:
classifier.classify(gender_features('David'))
Out[12]:
'male'
In [13]:
classifier.classify(gender_features('Michelle'))
Out[13]:
'female'
In [14]:
classifier.classify(gender_features('obama'))
Out[14]:
'female'
In [15]:
classifier.classify(gender_features('Alex'))
Out[15]:
'female'
In [16]:
print(nltk.classify.accuracy(classifier,test_test)) # printing the classifier accuracy.
0.762
In [17]:
#Task 3 Vectoriser and cosine similarity
                                            vectoriser used for word to number
from sklearn.feature_extraction.text import CountVectorizer
#from sklearn.feature extraction.text import TfidVectorizer
```

```
In [18]:
vect=CountVectorizer(binary=True)
corpus = ["Tessaract is good optical character recognition engine ", "optical character re
vect.fit(corpus)
Out[18]:
CountVectorizer(analyzer='word', binary=True, decode_error='strict',
                dtype=<class 'numpy.int64'>, encoding='utf-8', input='conten
t',
                lowercase=True, max_df=1.0, max_features=None, min_df=1,
                ngram_range=(1, 1), preprocessor=None, stop_words=None,
                strip_accents=None, token_pattern='(?u)\\b\\w\\w+\\b',
                tokenizer=None, vocabulary=None)
In [19]:
vocab=vect.vocabulary_
In [20]:
for key in sorted(vocab.keys()):
    print("{}:{}".format(key,vocab[key]))
character:0
engine:1
good:2
is:3
optical:4
recognition:5
significant:6
tessaract:7
In [21]:
"this is a good optical illusion"]).toarray()) #converting into boolean single dimension ved
[[0 0 1 1 1 0 0 0]]
In [22]:
print(vect.transform(corpus).toarray())
[[1 1 1 1 1 1 1 0 1]
 [10011110]]
In [23]:
from sklearn.metrics.pairwise import cosine_similarity
In [24]:
#simalrity between two sentence from given corpus #using cosine similarity.
similarity = cosine_similarity(vect.transform(["Google Cloud Vision is a character recognit
```

In [25]: print(similarity) #this show how similar two input sentences are. [[0.89442719]] In []: