

Student Name and Roll Number: RAHUL DHAWAN (17CSU146)
Semester /Section: VIII-C
Date: 30 January 2021
Faculty Signature:
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Grade:

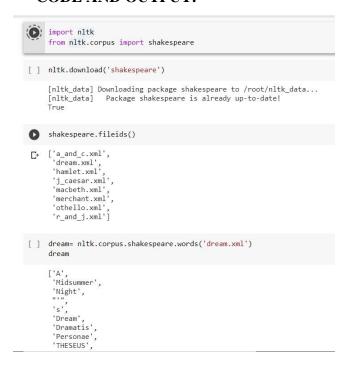
# **Objectives:**

- 1. Import the corpus Shakespeare and find the frequency of each word in the file dream.xml.
- 2. Find 5 most frequently occurring words from the file dream.xml.
- 3. Import wordnet corpus from the available nltk corpus list and find out the sysnset of word bank. Also find the definition and example of first sysnset in the list.

**Outcome:** Students will be able to learn the concepts of nltk.freqdist(), collections in python and sysnets in wordnet library.

### **Problem Statement:**

1. Import the corpus Shakespeare and find the frequency of each word in the file dream.xml.



```
[ ] len(dream)
    21538
[] #cleaning data of punctuations
    import string
    l=string.punctuation.split()
    no_punct_dream=[words *or words in dream if words not in string.punctuations
    no punct dream
     'Midsummer',
     'Night',
     'Dreaxi',
     Dr amatis,
     Personae,
     'THESEUR,'
     'Duke, '
     'of',
     'Athens',
     'EGEUS',
     '*ather',
     'Hermia',
     'LYSANDER',
     'DEMETRIUS',
     'in',
     'love',
     'with',
'Hermia',
     'PHILOSTRATE',
     'master',
     'of',
[ ] #Finding frequency
     fdist=nltk.FreqDist(w.lower() for w in no punct dream)
     fdist
     FreqDist(('a' : 273,
                 'midsummer': 2,
                'night': 52,
                's': 133,
                'dream': 16,
                'dramatis': 1,
                'personae': 1,
                'theseus': 67,
                'duke': 14,
                 'of': 272,
                'athens': 27,
                'egeus': 17,
                 'father': 14,
                'to': 3d0,
                'hermia': 103,
                'lysander': 103,
                'demetrius': 101,
                'in': 2d3,
                'love': 117,
                'with' : 177,
                'philostrate': ld,
                'master': 8,
                'the': 563,
                'revels': 5,
                'quince': 55,
                'carpenter': 1,
                'snug': 10,
```

'joiner': d,
'bottom': 69,
'weaver': 3,
'flute': 19,
'bellows': 3,

2. Find 5 most frequently occurring words from the file dream.xml.

# **CODE AND OUTPUT:**

Q2. Finding most frequently occuring words from the file dream.xml.

```
[ ] Dictionary=dict(fdist)

[ ] from collections import Counter
    dict(Counter(Dictionary).most_common(5))

{'and': 574, 'i': 470, 'the': 563, 'to': 340, 'you': 274}
```

3. Import wordnet corpus from the available nltk corpus list and find out the sysnset of word bank. Also find the definition and example of first sysnset in the list.

```
nltk.download('wordnet')
from nltk.corpus import wordnet
syns=wordnet.synsets("Bank")

[nltk_data] Downloading package wordnet to /root/nltk_data...
[nltk_data] Package wordnet is already up-to-date!

print("Definition of the word Bank:")
print(syns[@].definition())
print("\nExamples of the word Bank:")
print(syns[@].examples())

Definition of the word Bank:
sloping land (especially the slope beside a body of water)

Examples of the word Bank:
['they pulled the canoe up on the bank', 'he sat on the bank of the river and watched the currents']
```

TWIA Lab Manual (CSL 554) 2020-21

# **EXPERIMENT NO. 2**

Student Name and Roll Number: RAHUL DHAWAN (17CSU146)
Semester /Section: VIII-C
Date: 6 February 2021
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Faculty Signature:
Grade:

# **Objective:**

- 1. Print all the Arabic Stopwords.
- 2. Omit a given list of stop words from the total stopwords list of English language.

**Outcome:** Students will be able to understand the concept of stopwords in nltk and list comprehensions in python.

# **Problem Statement:**

1. Print all the Arabic Stopwords. CODE AND OUTPUT:

```
[1] import nltk
    nltk.download('stopwords')

[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data] Unzipping corpora/stopwords.zip.
True

Arabic_Stopwords = set(nltk.corpus.stopwords.words("arabic"))
Arabic_Stopwords

[2 {'ai',
    'lai',
    'lai',
```

2. Omit a given list of stop words from the total stopwords list of English language.

```
English_Stopwords = list(nltk.corpus.stopwords.words("english"))
      English_Stopwords
 'myself',
       'we',
'our',
'ours',
        'ourselves',
        'you',
       "you're",
       "you've",
       "you'll",
        "you'd",
        'your',
'yours',
        'yourself',
        'yourselves',
       'ĥe',
'him',
        'his',
        'himself',
        'she',
l=['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'Mahima', 'Munjal']
print(1)
['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'Mahima', 'Munjal']
for i in 1:
  if i in English_Stopwords:
    English_Stopwords.remove(i)
English_Stopwords
['you',
"you're",
"you're",
"you've",
"you'll",
"you'd",
'your',
 'yourself',
 'yourselves',
 'he',
'him',
 'his',
 'himself',
'she',
"she's",
 'her',
```



Student Name and Roll Number: RAHUL DHAWAN (17CSU146)
Semester /Section: VIII-C
Date: 13 February 2021
Faculty Signature:
Grade:

# **Objectives:**

- 1. Print the total number of male and female names in the names corpus. Then, Print the first 15 male and female names.
- 2. From the names corpus, combine all the labelled male and female names and print any 20.
- 3. Print the definition and examples of any one English language word using WordNet corpus.

**Outcome:** Students will be able to explore names corpus, understand the concept of labelling the data and learn about sysnets in Wordnet corpus.

## **Problem Statement:**

1. Print the total number of male and female names in the names corpus. Then, Print the first 15 male and female names.

```
[ ] nltk.download('names')
from nltk.corpus import names
print("\nNumber of male names:")
print (len(names.words('male.txt')))
print("Number of female names:")
print (len(names.words('female.txt')))

[nltk_data] Downloading package names to /root/nltk_data...
[nltk_data] Package names is already up-to-date!

Number of male names:
2943
Number of female names:
5001
```

# 2. From the names corpus, combine all the labelled male and female names and print any 20.

```
Male_names = names.words('male.txt')
             Female_names = names.words('female.txt')
            print("\nFirst 15 male names:")
            print (male_names[0:15])
             print("\nFirst 15 female names:")
            print (female_names[0:15])
           First 15 male names:
           ['Aamir', 'Aaron', 'Abbey', 'Abbie', 'Abbot', 'Abbott', 'Abby', 'Abdel', 'Abdul', 'Abdulkarim', 'Abdullah', 'Abe', 'Abel', 'Abelard', 'Abner']
           First 15 female names:
           ['Abagael', 'Abagail', 'Abbe', 'Abbe', 'Abbe', 'Abbi', 'Abbie', 'Abbi', 'Abigael', 'Abigail', 'Abigale', 'Abra', 'Acacia', 'Ada', 'Adah', 'Adaline']
                 Male= names.words('male.txt')
                  Female = names.words('female.txt')
                  Label_Male= [(str(name), 'male') for name in Male]
                 Label_Female = [(str(name), 'female') for name in Female]
                 print("Male :",Label_Male)
                  print("Female :",Label_Female)
                 Male : [('Aamir', 'male'), ('Abboy', 'male'), ('Abbey', 'male'), ('Abbie', 'male'), ('Abbot', 'male'), ('Abb
                 Female : [('Abagael', 'female'), ('Abagael', 'female'), ('Abbe', 'female'), ('Abbe', 'female'), ('Abbi', '
                 4
[ ] import random
                 Label_All = Label_Male + Label_Female
                  random.shuffle(Label_All)
                  print("First 20 random labeled combined names:")
                 Label_All[:20]
                  First 20 random labeled combined names:
                  [('Averil', 'female'),
                    [('Averil', 'female'),
('Ashely', 'female'),
('Abbot', 'male'),
('Connor', 'male'),
('Kara', 'female'),
('Nissa', 'female'),
('Belia', 'female'),
('Bren', 'female'),
('Alison', 'female'),
('Hermia' 'female')
                      ('Hermia', 'female'),
('Vera', 'female'),
('Ralina', 'female'),
('Clari', 'female'),
```

# 3. Print the definition and examples of any one English language word using WordNet corpus.

# **CODE AND OUTPUT:**

```
nltk.download('wordnet')
from nltk.corpus import wordnet
syns=wordnet.synsets("Telephone")

[nltk_data] Downloading package wordnet to /root/nltk_data...
[nltk_data] Package wordnet is already up-to-date!

print("Definition of the word Telephone:")
print(syns[0].definition())
print("\nExamples of the word Telephone:")
print(syns[0].examples())

Definition of the word Telephone:
electronic equipment that converts sound into electrical signals that can be transmitted over distances and then converts received signals back into sounds
```

Examples of the word Telephone: ['I talked to him on the telephone']



Student Name and Roll Number: RAHUL DHAWAN (17CSU146)
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**Objective:** To implement Levenshtein Edit Distance , Jaccard similarity , Cosine Similarity using both TF-IDF and count vectorizer

**Outcome:** Students will be able to demonstrate Levenshtein Edit Distance , Jaccard similarity , Cosine Similarity using both TF-IDF and count vectorizer

### **Problem Statement:**

1. Demonstrate the computation of Similarity Metrics such as Jaccard, Levenshtein and Cosine.

```
#Jaccard Similarity - Method 1

def jaccard_similarity(list1, list2):
    intersection = len(list(set(list1).intersection(list2)))
    union = (len(list1) + len(list2)) - intersection
    return float(intersection)/union

data1=input()
data2=input()
list1 = data1.split(" ")
list2 = data2.split(" ")
print("List 1 ",list1)
print("List 2 ",list2)

Mahima Munjal is a good girl
Mahima Munjal studies at NCU
List 1 ['Mahima', 'Munjal', 'is', 'a', 'good', 'girl']
List 2 ['Mahima', 'Munjal', 'studies', 'at', 'NCU']

jaccard_similarity(list1, list2)
```

```
#Jaccard Similarity - Method Z
def jaccard similarities(1ist1, list2):
    s1 = set(list1)
    s2 = set(list2)
    return float(len(s1.intersection(s2)) / len(s1.union(sZ)))
3accard_similarities(1ist1, list2)
0.222Z22ZZ2ZZZZZZZ
LEVENSHTEIN DISTANCE
   ttLe vensht e1n D1 stance
     impo | enc hant
     datal - "Mahima Munjal is a good girl."
     data2 = "Mahima Munjal"
    enchant.utils.levenshtein(datal,data2)
COSINE SIMILARITY
    1np a °t nltk
     nltk.down load ( ' punkt ')
     nltk.download('stopwords')
     [nltk_data) Downloading package punkt to /root/nltk_data...
     [nltk_data) Package punkt is already up-to-date!
     nltk data) Downloading package stopwords to /root/nltk data...
     [nltk_data) Package stopwords is already uo-to-date!
     True
I rom nltk . corpus in paat st opwords
from nltk.tokenzze import word_tokenize
I = input ("Enter first string: ").lower()
II = input ("Enter second string: ").lower()
Enter first string: Mahima Munjal is a good girl
Enter second string: Mahima Munjal studies at NCU and works at Nagarro
# tokenization
I list = uord tokenize(I)
II list = word tokenize(II)
print('First List', I_list)
print('Second Lzst', II list)
First List ['mahima', 'munjal', 'is', 'a', 'good', 'girl']
Second List ['mahima', 'munjal', 'studies', 'at', 'ncu', 'and', 'works', 'at', 'uagarro']
#remov1ng st opwords
sw = stopuords.words('english')
11 =[];12 =[1
I set = (w for w in I list if not w insw}
II_set = {w for w in II_list if not w in su}
pr1nt ('F1rst Set,' I set)
pr1nt('5econd Set,' II_set)
First Set {'good', 'mahima', 'girl', 'munjal'}
Second Set {'nagaono', 'ncu' 'stud1es', 'munjaJ,'
                                                  'monks','mahima'}
```

```
# Forming a set containing keywords of both strings
rvector = I set.union(II set)
for w in rvector:
   if w in I set: 11.append(1)
   else: 11.append(0)
   if w in II set: 12.append(1)
   else: 12.append(0)
c = 0
# cosine formula
for i in range(len(rvector)):
      c+= 11[i]'12[i]
cosine = c / float((sum(11)*sum(1Z))**0.5)
print('Cosine Similarity Value :',cosine)
Cosine Similarity Value' 0.A882A8290A638631
COSINE SIMILARITY MATRIX
[] I = input("Enter first string: ").lower()
    II = input ("Enter second string: ").lower()
     Enter first string: Mahima Munjal is a student Df Text and Web Analytics
     Enter second string: Mahima Munjal is a student of The Northcap University
[ ] documents = [I,II]
[ ] from sklearn.feature extraction.text import CountVectorizer
     import pandas as pd
[ ] count vectorizer = CountVectorizer(stop_words='english')
     count vectorizer = CountVectorizer()
     sparse matrix = count vectorZzer.fit transform(documents)
    doc term matrix = sparse matrix.tDdense()
     df = pd.DataFrame(doc term matrix,
                     columns=count vectorizer.get feature names(),
                    index=['I', 'II'])
     df
            ana1yt1cs and is Inah1ma munjal northcap a-I- student text I he univers1ty web
               1 1 1
                             1
                                   1
                                             0 1
      ı
                                                        1 1 0
                                                                                 1
               0 0 1 1 1 1 1 1 0 1
[ ] from sklearn.metrics.pairwise import cosine similarity
      print(cosine similarity(df, df))
                      0.589255651
      [[1.
       [&.&g 2s&6s1.
```

2. Calculate the TF-IDF vectorizer on 2 documents.

### **CODE AND OUTPUT:**

### 02. CALCULATE THE TFIDF VECTORIZOR ON 2 DOCUMENTS.

```
[ ] I = input("Enter first string: ").lower()
    II = input("Enter second string: ").lower()

Enter first string: Mahima Munjal loves to watch movies
    Enter second string: Mahima Munjal loves to do yoga

[ ] from sklearn.feature_extraction.text import TfidfVectorizer
    corpus = [I,II]
    vectorizer = TfidfVectorizer()
    X = vectorizer.fit_transform(corpus)
    print('Vectorizer Features :',vectorizer.get_feature_names())
    print('Vectorizer Shape: ',X.shape)

Vectorizer Features : ['do', 'loves', 'mahima', 'movies', 'munjal', 'to', 'watch', 'yoga']
    Vectorizer Shape: (2, 8)
```

3. Apply the max-df, min-df param in the TF-IDF function.

## **CODE AND OUTPUT:**

Q3. Apply the max-df, min-df param in the TF-IDF function.

```
[ ] data = [I,II]
    count_vec = CountVectorizer(stop_words="english", analyzer='word',ngram_range=(1, 1), max_df=0.50, min_df=1, max_features=None)

count_train = count_vec.fit(data)
    bag_of_words = count_vec.transform(data)

print('Features :',count_vec.get_feature_names())
```

Features : ['movies', 'watch', 'yoga']

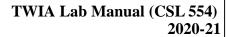
4. Compute Cosine Similarity using both TF-IDF and Count vectorizer

```
[26] from sklearn.feature_extraction.text import CountVectorizer
    from sklearn.feature_extraction.text import TfidfTransformer
    from nltk.corpus import stopwords
    import numpy as np
    import numpy.linalg as LA
    import nltk
    nltk.download('stopwords')

[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data] Package stopwords is already up-to-date!
True

[29] train_set = ["Mahima Munjal is a good girl", "Mahima Munjal studies at The Northcap University."] # Documents
    test_set = ["A good girl named Mahima Munjal studies at the Northcap University."] # Query
    stopWords = stopwords.words('english')
```

```
vectorizer = CountVectorizer(stop words = stopWords)
print('Vectorizer : ', vectorizer)
transformer = TfidfTransformer()
print('\n\nTF-IDF Transformer : ',transformer)
                           CountVectorizer(analyzer='wo d', bina y=False, decode_error='strict',
Vectorizer
                                dtype=<class 'numpy.int64'», encoding='utf-8', input='content',
                                lowercase=True, max_df=1.0, max features=None, min df=1,
                               ngram range=(1, 1), preprocessor=None,
                                "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he', 'him', 'his', 'himself', 'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its',
                                                       'itself', ...],
                                strip accents=None, token pattern='(?u)\b\,
                                tokenizer=None, vocabu]ary=None)
TF-IDF Transformer TfidfTransformer(norm='12', smooth_idf=True, sublinear_t*=False, use_idf=True)
  #Using Count Vectorizer
            trainVectorlzerArray = vectorizer.fit transform(train set).toarray()
            testVectorizerArray vectorizer.transform(test_set).toarray()
            print ('Fit Vectorizer to train set \n', trainVectorizerArray)
            print ('\n\nTransform Vectorizer to test set\n', testUectorizerArray)
            Fit vectorizer to train set
            [[1 1 1 1 0 0 0]
             [0 0 1 1 1 1 1]]
           Transform Vectorizer to test set
             [[1 1 1 1 1 1 1]]
         #Using
            transformer.fit(trainVectorizerArray)
            print('Fit transformer to train set \n', transformer.transform(trainVectorizerArray).toarray())
            transformer.fit(test4ectorizerArray)
            tfidf = transformer.transform(testVectorizerArray)
            print('\n\nFit transformer to test set\n', tfidf.todense())
  Fit transformer to train set
            [[0.57615236 0.57615236 0.40993715 0.40993715 0.
                                                                                                                                          0.
               0.
             [0.
                                      0.
                                                               0.35520009 0.35520009 0.49922133 0.ñ9922133
                0.49922133]]
            Fit transformer to test set
              [[0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.37796447 \ 0.3779647 \ 0.3779647 \ 0.3779647 \ 0.3779647 \ 0.3779647 \ 0.3779647 \ 0.3779647 \ 0.3779647 \ 0.3779647 \ 0.3779647
               0.3779644711
```





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**Objective:** Implementation of the Lesk algorithm for Word Sense Disambiguation.

**Outcome:** Students will be able to demonstrate how to Lesk algorithm works.

**Problem Statement:** Implement Lesk algorithm for Word Sense Disambiguation.

```
import nltk
     nltk.download('averaged_perceptron_tagger')
     nltk.download('wordnet')
     nltk.download('punkt')
 [nltk_data] Downloading package averaged_perceptron_tagger to
     [nltk_data] /root/nltk_data...
     [nltk_data] Package averaged_perceptron_tagger is already up-to-
     [nltk_data]
                      date!
     [nltk_data] Downloading package wordnet to /root/nltk_data...
     [nltk data] Package wordnet is already up-to-date!
     [nltk_data] Downloading package punkt to /root/nltk_data...
     [nltk_data] Unzipping tokenizers/punkt.zip.
     True
[12] from pywsd.lesk import simple_lesk
[16] sentences = ['I went to the bank to deposit my money', 'The river bank had a lot of fishes and crocodiles.']
     #ambiguous word - Bank
```

# · LESK WORKS CORRECTLY

```
}13| # Context 1 - Financial institution
    print ("Context-1:", sentences[0])
    answer = simple lesk(sentences[0], 'bank')
     print ("Sense:", answer)
    print ("Definition ", answer.definition())
    # Correct Output - Financial Institution printed
     # No disambiguity
    Context-1: I went to the bank to deposit my money
    Sense: Synset('depository financial institution.n.01')
    Definition: a financial institution that accepts deposits and channels the money into lending activities
     # Context 2 - River Bank
     print ("Context-2:", sentences[1])
    answer = simple_lesk(sentences[1],'bank')
     print ("Sense:", answer)
    print ("Definition ", answer.definition())
     # Correct Output - River Bank or sloping land printed
     # No disambiguity
    Context-2: The river bank had a lot of fishes and crocodiles.
    Sense: Syn set( 'bank. n. 01')
    Definition' sloping land (especially the slope beside a body of water)
```

### LESK WORKS INCORRECTLY

#### [19] 4 Context 2 — TnBe/SeBdf1ng/Sapf1ng

```
print ("Context-2:', new sentences[1j)
    answer = simple lesk(new sentences[1j,'plant')
    print ("Sense:', answer)
    print ("Definition : ", answer.definition())
    # Correct output - Plant bearing flower sense printed
    4 No d1samb1qu1ty
   Context-3: The plant was no longer bearing flowers
    Sense: Synset('plant.v.01')
    Definition: put or set (seeds, seedlings, or plants) into the ground
# Context 3 - Industrial plant (Added the word industrial before plant in
    # Context 1)
    print ("Context-3:', new sentences[2j)
    answer = simple lesk(new sentences[2j,'plant')
    print ("Sense:', answer)
    print ("Definition : ", answer.definition())
    # Correct output - Industrial plantprinted
    # (Disambiguity resolved by adding the mord 'industrial' in the sentence.)
Context-3: The workers at the industrial plant were overworked
    Sense: Synset('plant.n.01')
    Definition: buildings for carrying on industrial labor
```