



Subject - NATURAL LANGUAGE PROCESSING

A word cloud centered around the acronym 'NLP' (Natural Language Processing). The words are arranged in a circular pattern, with 'NLP' being the largest and most central. Other prominent words include 'natural', 'language', 'processing', 'learning', 'interaction', 'text', 'linguistics', 'automatic', 'understanding', 'public', 'download', 'process', 'computer', 'retrieval', 'tag', 'typo', 'design', 'discourse', 'job', 'analysis', 'word', 'communicate', 'simulation', 'keywords', 'telecommunications', 'output', 'operating', 'typography', 'information', 'human', 'systems', 'coreference', 'programming', 'technology', 'automated', 'evaluation', 'statistical', 'artificial', 'connect', 'machine', 'networks', 'summarization', 'intelligence', 'cloud', 'science', 'evolution', 'data', 'layout', 'input', and 'output'.

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V Semester

1. Consider an example to find the similarity between two vectors – ‘**x**’ and ‘**y**’, using Cosine Similarity.

The ‘x’ vector has values, **x** = { **3, 2, 0, 5** }

The ‘y’ vector has values, **y** = { **1, 0, 0, 0** }

Design a program to compute the similarity between x and y.

2. Consider two sets $A = (0,1,2,5,6)$ and $B = (0, 2,3, 5, 7,9)$. How similar are A and B? Design a program to compute the similarity using Jaccard similarity.

```
[1] import math
```

```
▶ x=[3,2,0,5]
  y=[1,0,0,0]

  # Cosine Similarity

  def SOS(x):
      res=0
      for i in x:
          res+=(i*i)
      return res

  def dotProd(x,y):
      res=0
      for i in range(len(x)):
          res+=(x[i]*y[i])
      return res

  def CS(x,y):
      return dotProd(x,y)/math.sqrt(SOS(x)*SOS(y))

  print("Cosine Similarity between x & y is",CS(x,y))
```

```
📄 Cosine Similarity between x & y is 0.48666426339228763
```

```

▶ A=[0,1,2,5,6]
  B=[0,2,3,5,7,9]

#Jaccard Similarity

def JS(a,b):
    setA=set(a)
    setB=set(b)
    return len(setA.intersection(setB))/len(setA.union(setB))

print("Similarity between A & B is",JS(A,B))

```

➞ Similarity between A & B is 0.375

3. Consider Three documents:

d_1 : "Jack London traveled to Oakland"

d_2 : "Jack London traveled to the city of Oakland"

d_3 : "Jack traveled from Oakland to London"

Based on shingles of size 2 (2-grams or bigrams), what are the Jaccard coefficients $J(d_1, d_2)$ and $J(d_1, d_3)$?

```

from nltk.tokenize import word_tokenize
from nltk.util import bigrams

d1="Jack London traveled to Oakland"
d2="Jack London traveled to the city of Oakland"
d3="Jack traveled from Oakland to London"

# doc to Bigrams
def d2bigram(d):
    return set(bigrams(word_tokenize(d)))

# Jaccard Coefficient
def JC(d1,d2):
    bgram_d1=d2bigram(d1)
    bgram_d2=d2bigram(d2)
    return len(bgram_d1.intersection(bgram_d2))/len(bgram_d1.union(bgram_d2))

print("J(d1, d2) =",JC(d1,d2))
print("J(d1, d3) =",JC(d1,d3))

```

$J(d_1, d_2) = 0.375$
 $J(d_1, d_3) = 0.0$

4. Consider set of documents as

D1 : I am Sam.

D2 : Sam I am.

D3 : I do not like green eggs and ham.

D4 : I do not like them, Sam I am.

- Design a program for (k = 1)-shingles of D1 U D2 U D3 U D4 : U is UNION**
- Design a program for (k = 2)-shingles of D1 U D2 U D3 U D4 : U is UNION**
- Design a program for (k = 3)-Character shingles of D1 U D2**
- Design a program for (k = 4)-Character shingles of D1 U D2**

```
from nltk.tokenize import word_tokenize
from nltk.util import ngrams

d1="I am Sam."
d2="Sam I am."
d3="I do not like green eggs and ham."
d4="I do not like them, Sam I am."

# doc to ngrams
def d2ngram(d,n):
    return set(ngrams(word_tokenize(d),n))

# doc to k character shingles
def d2k_shingle(d,k):
    return set(ngrams(d,k))

# union of results
def d_s2union(docs):
    res=set()
    for i in docs:
        res=res.union(i)
    return res

print('a')
l=[]
n=1
l.append(d2ngram(d1,n))
l.append(d2ngram(d2,n))
l.append(d2ngram(d3,n))
l.append(d2ngram(d4,n))
print(d_s2union(l))

print('\nb')
l=[]
n=2
l.append(d2ngram(d1,n))
l.append(d2ngram(d2,n))
l.append(d2ngram(d3,n))
l.append(d2ngram(d4,n))
print(d_s2union(l))

print('\nc')
l=[]
k=3
l.append(d2k_shingle(d1,k))
l.append(d2k_shingle(d1,k))
print(d_s2union(l))

print('\nd')
l=[]
k=4
l.append(d2k_shingle(d1,k))
l.append(d2k_shingle(d1,k))
print(d_s2union(l))
```

Output:

- $\{('not',), ('and',), (','), (','), ('like',), ('I',), ('eggs',), ('am',), ('them',), ('green',), ('Sam',), ('ham',), ('do',))\}$
 - $\{('I', 'am'), ('I', 'do'), ('and', 'ham'), ('am', '.'), ('do', 'not'), ('not', 'like'), ('eggs', 'and'), ('Sam', '.'), ('like', 'green'), (',', 'Sam'), ('like', 'them'), ('am', 'Sam'), ('ham', '.'), ('Sam', 'I'), ('them', ','), ('green', 'eggs')\}$
 - $\{('I', ' ', 'a'), (' ', 'S', 'a'), ('a', 'm', '.'), (' ', 'a', 'm'), ('S', 'a', 'm'), ('a', 'm', ' '), ('m', ' ', 'S')\}$
 - $\{('I', ' ', 'a', 'm'), (' ', 'a', 'm', ' '), (' ', 'S', 'a', 'm'), ('a', 'm', ' ', 'S'), ('m', ' ', 'S', 'a'), ('S', 'a', 'm', ' '), (' ', 'S', 'a', 'm')\}$
5. Suppose our document D is the string abcdabd, and we pick $k = 2$. Then the set of 2-shingles for D is {ab, bc, cd, da, bd}. Note that the substring ab appears twice within D, but appears only once as a shingle. A variation of shingling produces a bag, rather than

a set, so each shingle would appear in the result as many times as it appears in the document. Design a program to model the above scenario.

```

▶ from nltk.util import ngrams

d="abcdabd"

# doc to k character shingles with repetition
def d2k_shingle(d,k):
    return list(ngrams(d,k))

print(d2k_shingle(d,2))

```

☞ [('a', 'b'), ('b', 'c'), ('c', 'd'), ('d', 'a'), ('a', 'b'), ('b', 'd')]

6. A database contains 80 records on a particular topic. A search was conducted on that topic and 60 records were retrieved. Of the 60 records retrieved, 45 were relevant. Design a Program to compute the precision and recall scores for the search.

```

▶ #precision
def precision(retrieved,relevant):
    return (relevant/retrieved)*100

#recall
def recall(total,relevant):
    return (relevant/total)*100

#func
def fun(total,retrieved,relevant):
    print("Total Records =",total)
    print("Retrieved Records =",retrieved)
    print("Relevant Records Retrieved =",relevant)
    print()
    print("Precision =",precision(retrieved,relevant),"%")
    print("Recall =",recall(total,relevant),"%")

total=80
retrieved=60
relevant=45

fun(total,retrieved,relevant)

```

☞ Total Records = 80
Retrieved Records = 60
Relevant Records Retrieved = 45

Precision = 75.0 %
Recall = 56.25 %

7. Consider this sentence: “a rose is a rose is a rose”. And represent this document as set of shingles (Word n gram). Design a program for extracting shingles for $n = 3$ and $n = 4$ for above sentence.

```

▶ from nltk.tokenize import word_tokenize
  from nltk.util import ngrams

  d="a rose is a rose is a rose"

  # doc to ngrams
  def d2ngram(d,n):
      | return set(ngrams(word_tokenize(d),n))

  print("Word ngram for n=3:")
  print(d2ngram(d,3))
  print()
  print("Word ngram for n=4:")
  print(d2ngram(d,4))

```

```

↳ Word ngram for n=3:
  {'a', 'rose', 'is'}, ('is', 'a', 'rose'), ('rose', 'is', 'a')}

  Word ngram for n=4:
  {'is', 'a', 'rose', 'is'}, ('rose', 'is', 'a', 'rose'), ('a', 'rose', 'is', 'a')}

```

8. Compute the similarity between:

- a) night and nacht
- b) Ashish and Aasheesh

```
[11] from nltk.util import ngrams
```

```

▶ # doc to k character shingles
  def d2k_shingle(d,k):
      | return set(ngrams(d,k))

  # Jaccard Similarity
  def JS(d1,d2,n):
      | nshingle_d1=d2k_shingle(d1,n)
      | nshingle_d2=d2k_shingle(d2,n)
      | return len(nshingle_d1.intersection(nshingle_d2))/len(nshingle_d1.union(nshingle_d2))

  print("a)")
  print("Similarity between night and nacht is",JS("night","nacht",2))

  print("\nb)")
  print("Similarity between Ashish and Aasheesh is",JS("Ashish","Aasheesh",2))

```

```

↳ a)
  Similarity between night and nacht is 0.14285714285714285

  b)
  Similarity between Ashish and Aasheesh is 0.1111111111111111

```

9. Consider a case insensitive query and document collection with a query Q and a document collection consisting of the following three documents:

Q: "gold silver truck"

D1: "Shipment of gold damaged in a fire"

D2: "Delivery of silver arrived in a silver truck"

D3: "Shipment of gold arrived in a truck"

Design a program to compute the similarity between

a) Q & D1

b) Q & D2

c) Q & D3

Find which document (D) matches best with the query (Q)

```
from nltk.tokenize import word_tokenize
from nltk.util import bigrams

q="gold silver truck"
d1="Shipment of gold damaged in a fire"
d2="Delivery of silver arrived in a silver truck"
d3="Shipment of gold arrived in a truck"

# doc to Bigrams
def d2bigram(d):
    return set(bigrams(word_tokenize(d)))

# Jaccard Coefficient
def JS(d1,d2):
    bgram_d1=d2bigram(d1)
    bgram_d2=d2bigram(d2)
    return len(bgram_d1.intersection(bgram_d2))/len(bgram_d1.union(bgram_d2))

#find best match
def findBestMatch(q,*d):
    res=0
    ans=""
    d_num=0
    for i in range(len(d)):
        doc=d[i]
        tmp=JS(q,doc)
        print("For JS(Q, D{}): {}".format(i+1,tmp))
        if tmp>res:
            res=tmp
            ans=doc
            d_num=i+1
    print("Document D{} matches best with the given Query.".format(d_num))

findBestMatch(q,d1,d2,d3)
```

```
For JS(Q, D1): 0.0
For JS(Q, D2): 0.125
For JS(Q, D3): 0.0
Document D2 matches best with the given Query.
```


Thank You!