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Magnum Opus Task II

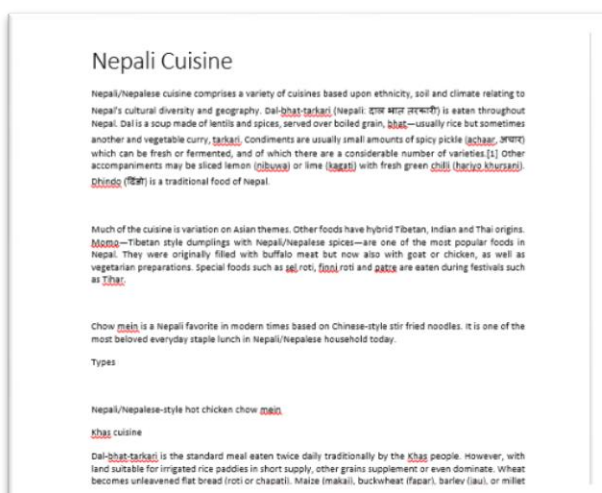
Cosine Distance/Similarity between two documents before and after stemming

Documents:

1. Indian Cuisine.docx



2. Nepali Cuisine.docx



Code Snippet:

Magnum Opus Task II - Cosine Distance/similarity between two documents

(Two documents used are Indian cuisine & Nepali cuisine obtained from wikipedia) Ayush Karn - 17BCE2381

```
In [95]: import nltk

In [96]: from nltk.corpus import stopwords
         from nltk.tokenize import word_tokenize

In [97]: import docx #to read docx files

In [98]: from docx import Document
         document1 = Document('Indian Cuisine.docx') #Document for indian cuisine
         doc1=""
         for para in document1.paragraphs:
             temp=para.text
             doc1=doc1+temp

In [99]: document2 = Document('Nepali Cuisine.docx') #Document for nepali cuisine
         doc2=""
         for para in document2.paragraphs:
             temp=para.text
             doc2=doc2+temp
```

Before Stemming

```
In [100]: # tokenization
         Doc1_list = word_tokenize(doc1)
         Doc2_list = word_tokenize(doc2)

         # sw contains the list of stopwords
         sw = stopwords.words('english')
         l1=[];l2=[]

         # remove stop words from string
         Doc1_set = {w for w in Doc1_list if not w in sw} #excluding the stop words from doc1 tokens
         Doc2_set = {w for w in Doc2_list if not w in sw} #excluding the stop words from doc2 tokens

         # form a set containing keywords of both strings
         rvector = Doc1_set.union(Doc2_set)
         for w in rvector:
             if w in Doc1_set: l1.append(1) # create a vector
             else: l1.append(0)
             if w in Doc2_set: l2.append(1)
             else: l2.append(0)
         c = 0

         # cosine formula
         for i in range(len(rvector)):
             c+= l1[i]*l2[i]
         cosine = c / float((sum(l1)*sum(l2))**0.5)
         print("Cosine similarity: ", cosine)

Cosine similarity: 0.2570134622843176
```

Hence cosine similarity obtained before stemming is approximately 0.257 between the two documents

After Stemming

```
In [101]: sno = nltk.stem.SnowballStemmer('english')
         Doc1_stemmed = [] #Stemming Doc1 tokens
         for sentence in Doc1_list:
             Doc1_stemmed.append(" ".join([sno.stem(i) for i in sentence.split()]))

         for item in Doc1_stemmed [0:20]: #display 0-20 stemmed words for example output
             print(item)

indian
cuisin
indian
cuisin
consist
of
a
wide
variety
of
region
and
tradit
cuisin
nativ
to
the
indian
subcontin
.
```

```
In [102]: Doc2_stemmed = [] #Stemming Doc 2 tokens
for sentence in Doc2_list:
    Doc2_stemmed.append(" ".join([sno.stem(i) for i in sentence.split()]))

for item in Doc2_stemmed [0:20]:
    print(item)
```

```
nepali
cuisinenepali/nepales
cuisin
compris
a
varietati
of
cuisin
base
upon
ethnic
,
soil
and
climat
relat
to
nepal
's
cultur
```

```
In [103]: # sw contains the list of stopwords
sw = stopwords.words('english')
l1 = []; l2 = []

# remove stop words from string
Doc1_set = {w for w in Doc1_stemmed if not w in sw} #excluding the stop words from doc1 tokens
Doc2_set = {w for w in Doc2_stemmed if not w in sw} #excluding the stop words from doc2 tokens

# form a set containing keywords of both strings
rvector = Doc1_set.union(Doc2_set)
for w in rvector:
    if w in Doc1_set: l1.append(1) # create a vector
    else: l1.append(0)
    if w in Doc2_set: l2.append(1)
    else: l2.append(0)
c = 0

# cosine formula
for i in range(len(rvector)):
    c += l1[i]*l2[i]
cosine = c / float((sum(l1)*sum(l2))**0.5)
print("Cosine similarity: ", cosine)

Cosine similarity: 0.28832231219045157
```

Hence cosine similarity obtained after stemming is approximately 0.288 between the two documents

Errors faced:

1. docx files weren't readable directly. Python-docx had to be imported for reading documents while txt files can be read directly.
2. Hyperlinks (as copied from Wikipedia) were not read as a normal text by the 'python-docx' extension.
So, all the hyperlinks had to be manually be removed from the docx file.
3. Inefficient Stemming algorithm, internet sources say snowball stemmer as best for general use but lots of incorrect results were obtained on stemming.

Eg: cuisine -> cuisine
varieties -> varietati
relations -> relat

Conclusion

Hence, the cosine distance between two documents was found to be increased after stemming than before. With a better stemming algorithm, results would have been more accurate. Also for observance, to exactly same documents had a cosine similarity of 1.

References

- <https://www.machinelearningplus.com/nlp/cosine-similarity/>
- <https://www.geeksforgeeks.org/python-measure-similarity-between-two-sentences-using-cosine-similarity/>
- <https://xapian.org/docs/stemming.html>
- <https://www.nltk.org/howto/stem.html>