

# SPPU-TE-COMP-CONTENT – KSKA Git

Q1: Explain the Basic concept of Text Analysis?

- ANS.
- Text Analysis is a process of Analyzing and understanding the written or spoken language. It employs computer Algorithms and techniques to extract valuable information, pattern, and insight from extensive textual data.
- In simple terms, text Analytics empowers computers to understand and interpret Human Language.
  - Text Analytics has become a crucial tool in todays information Age for two main reasons.
    - i) Massive growth of Text Data.
    - ii) Extract valuable insights hidden within data.
  - Text Analytics is a powerful tool that finds the meaning and value hidden within mountains of the text data.
  - Text Analysis process typically involves/includes the several key steps such as:-
    - Identification
    - Tokenization.
    - Sentence Breaking.
    - Part-of-speech tagging.
    - Chunking.
    - Syntax Parsing.
    - Sentence Chaining.
    - Keyword Extraction.
    - Entity Recognition.

Q2: Explain Inverse Document Frequency in Detail.

- ANS.
- Inverse Document Frequency (IDF) is a key-concept in the terms Frequency Inverse - Document Frequency (IF-IDF) weighting scheme, which is commonly used in text Analytics and information retrieval.
- IDF measures how important a word is across a

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set of documents. It down-weights common words and highlights rare, informative ones.

D FORMULA :-

The standard formula for calculating the Inverse Document Frequency (IDF) of term  $t$  is:-

$$\text{IDF}(t) = \log\left(\frac{N}{dft}\right)$$

where,

$N$  → Total No. of Documents in the corpus.

$dft$  → Number of Documents that contain the term  $t$ .

- If the IDF of the term is high, it is considered to be more informative.

D APPLICATIONS:-

- ① It is used to search Engines to rank documents by Relevance.
- ② It is used for text classification, feature selection and vectorization.

Q3) Perform stemming for text = "studies studying cries cry". Compare the results generated with lemmatization. Comment on your answer how stemming and lemmatization differ from each other.

ANS. Given text : "studies studying cries cry"

- ① The output of stemming is:-

studies ⇒ studi

studying ⇒ study

cries ⇒ cri

cry ⇒ cry

- ② Lemmatization:- Output:-

studies ⇒ study

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- studying  $\Rightarrow$  study
- cries  $\Rightarrow$  cry
- cry  $\rightarrow$  cry

Compare: stemming and Lemmatization:-

- stemming chops the words to its root form
- . It uses simple rules and often leads to non-dictionary forms of words like.  
studies  $\rightarrow$  studi } ... (non - dictionary  
cries - cri } root form)
- . It is a faster technique but less accurate.
- Lemmatization uses vocabulary and Morphological Analysis, returning valid dictionary words like.  
studies  $\rightarrow$  study } ..(non - root forms)  
cries  $\rightarrow$  cry }  
. It gives more accurate results.

Q4) Write a Python code for removing stop words from the below documents, convert the document into lowercase and calculate the TF, IDF and TFIDF score for each document.

document A = "Jupiter is the largest Planet"

document B = "Mars is the fourth planet from the Sun."

Ans.

Python code:-

```
import nltk
from sklearn.feature_extraction.text import TfidfVectorizer
from nltk.corpus
import stopwords.
```

```
nltk.download('stopwords')
```

```
doc_a = doc_a.lower()
```

```
doc_b = doc_b.lower()
```

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```
doc_a = "Jupiter is the Largest Planet"
```

```
doc_b = "Mars is the fourth planet from the sun"
```

```
doc_a = doc_a.lower()
```

```
doc_b = doc_b.lower()
```

```
stopwords = set(stopwords.words('English'))
```

```
def remove_stopwords(text):
```

```
    return ' '.join([word for word in text.split() if word not in stop_words])
```

```
clean_doc_a = remove_stopwords(doc_a)
```

```
clean_doc_b = remove_stopwords(doc_b)
```

```
vectorizer = TfidfVectorizer()
```

```
tfidf_matrix = vectorizer.fit_transform([clean_doc_a, clean_doc_b])
```

```
words = vectorizer.get_feature_names_out()
```

```
for i, doc in enumerate([clean_doc_a, clean_doc_b]):
```

```
    print(f"\nTF-IDF For Document {['A', 'B'][i]}")
```

```
    for j, word in enumerate(words):
```

```
        print(f"\t{word}: {tfidf_matrix[i, j]:.4f}")
```

## OUTPUT:-

TF-IDF For Document A:-

jupiter : 0.7071

largest : 0.7071

Mars : 0.0000

fourth : 0.0000

planet : 0.0000

sun : 0.0000

TF-IDF For Document B:-

jupiter : 0.0000

largest : 0.0000

Mars : 0.5

fourth : 0.5

planet : 0.5

sun : 0.5