

Practical No:01

Name:Anuj Shailendra Naikodi

Roll No:41

```
import re

from nltk.corpus import stopwords

from nltk.stem import PorterStemmer

import nltk

nltk.download('stopwords')

text = """Natural language processing (NLP) is a subfield of linguistics, computer science, and artificial intelligence concerned with the interactions between computers and human language."""

print("Original Text:\n", text)

text = text.lower()

tokens = re.findall(r'\b[a-z]+\b', text)

print("\nTokens:\n", tokens)

stop_words = set(stopwords.words('english'))

filtered_tokens = [w for w in tokens if w not in stop_words]

print("\nAfter Stopword Removal:\n", filtered_tokens)

stemmer = PorterStemmer()

stemmed_tokens = [stemmer.stem(w) for w in filtered_tokens]

print("\nAfter Stemming:\n", stemmed_tokens)
```

Output:

```
Original Text:
Natural language processing (NLP) is a subfield of linguistics, computer science, and artificial intelligence concerned with the interactions between computers and human language.

Tokens:
['natural', 'language', 'processing', 'nlp', 'is', 'a', 'subfield', 'of', 'linguistics', 'computer', 'science', 'and', 'artificial', 'intelligence', 'concerned', 'with', 't

After Stopword Removal:
['natural', 'language', 'processing', 'nlp', 'subfield', 'linguistics', 'computer', 'science', 'artificial', 'intelligence', 'concerned', 'interactions', 'computers', 'huma

After Stemming:
['natur', 'languag', 'process', 'nlp', 'subfield', 'linguist', 'comput', 'scienc', 'artifici', 'intellig', 'concern', 'interact', 'comput', 'human', 'languag']
[nltk_data]  Downloading package stopwords to /root/nltk_data...
[nltk_data]  Unzipping corpora/stopwords.zip.
```

Practical No:02

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```
from collections import defaultdict
import re
```

```
documents = {
```

```
    1: "Natural language processing enables computers to understand human language.",
    2: "Machine learning and artificial intelligence are key components of NLP.",
    3: "Text mining involves analyzing large volumes of text data.",
    4: "Deep learning methods are used in computer vision and NLP."
```

```
}
```

```
inverted_index = defaultdict(set)
```

```
for doc_id, text in documents.items():
```

```
    words = re.findall(r'\b\w+\b', text.lower())
```

```
    for word in words:
```

```
        inverted_index[word].add(doc_id)
```

```
print("Inverted Index:")
```

```
for word, doc_ids in sorted(inverted_index.items()):
```

```
    print(f'{word}: {sorted(doc_ids)}')
```

```
def search(query):
```

```
    query_word = query.lower()
```

```
    if query_word in inverted_index:
```

```
        print(f'\nDocuments containing \'{query_word}\':\n{sorted(inverted_index[query_word])}')
```

```
    else:
```

```
        print(f'\nNo documents found containing \'{query_word}\'')
```

```
search("NLP")
```

```
search("learning")
```

```
search("python")
```

Output:

```
Inverted Index:  
analyzing: [3]  
and: [2, 4]  
are: [2, 4]  
artificial: [2]  
components: [2]  
computer: [4]  
computers: [1]  
data: [3]  
deep: [4]  
enables: [1]  
human: [1]  
in: [4]  
intelligence: [2]  
involves: [3]  
key: [2]  
language: [1]  
large: [3]  
learning: [2, 4]  
machine: [2]  
methods: [4]  
mining: [3]  
natural: [1]  
nlp: [2, 4]  
of: [2, 3]  
processing: [1]  
text: [3]  
to: [1]  
understand: [1]  
used: [4]  
vision: [4]  
volumes: [3]
```

```
Documents containing 'nlp': [2, 4]
```

```
Documents containing 'learning': [2, 4]
```

```
No documents found containing 'python'.
```

Practical No:03

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Roll No:41

```
import pandas as pd
```

```
data = pd.DataFrame({  
    'Age': [67, 57, 43, 71, 36],  
    'Sex': [0, 1, 0, 1, 0],  
    'HeartDisease': [1, 1, 0, 0, 1]  
})  
print("Sample Dataset:\n", data)
```

```
p_hd1 = data['HeartDisease'].mean()  
p_hd0 = 1 - p_hd1  
print(f"\nP(HeartDisease=1) = {p_hd1:.2f}, P(HeartDisease=0) = {p_hd0:.2f}")
```

```
age_hd1 = data[data['HeartDisease']==1]['Age'].value_counts(normalize=True)  
age_hd0 = data[data['HeartDisease']==0]['Age'].value_counts(normalize=True)  
print("\nP(Age|HeartDisease=1):\n", age_hd1)  
print("\nP(Age|HeartDisease=0):\n", age_hd0)
```

```
def predict_heart_disease(age):  
    ph1 = p_hd1 * age_hd1.get(age, 0.01)  
    ph0 = p_hd0 * age_hd0.get(age, 0.01)  
    prob_hd1 = ph1 / (ph1 + ph0)  
    return prob_hd1
```

```
age_test = 57  
prob = predict_heart_disease(age_test)  
print(f"\nPredicted probability for HeartDisease given Age={age_test}: {prob:.2f}")
```

Output:

Sample Dataset:

	Age	Sex	HeartDisease
0	67	0	1
1	57	1	1
2	43	0	0
3	71	1	0
4	36	0	1

$P(\text{HeartDisease}=1) = 0.60$, $P(\text{HeartDisease}=0) = 0.40$

$P(\text{Age}|\text{HeartDisease}=1)$:

Age
67
57
36

Name: proportion, dtype: float64

$P(\text{Age}|\text{HeartDisease}=0)$:

Age
43
71

Name: proportion, dtype: float64

Predicted probability for HeartDisease given Age=57: 0.98

Practical No:04

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```
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.datasets import load_iris
from sklearn.preprocessing import StandardScaler
from sklearn.cluster import AgglomerativeClustering
from scipy.cluster.hierarchy import dendrogram, linkage
```

```
iris = load_iris()
```

```
X = iris.data
```

```
feature_names = iris.feature_names
```

```
scaler = StandardScaler()
```

```
X_scaled = scaler.fit_transform(X)
```

```
linked = linkage(X_scaled, method='ward')
```

```
plt.figure(figsize=(10, 6))
```

```
dendrogram(linked, orientation='top', distance_sort='descending', show_leaf_counts=True)
```

```
plt.title("Dendrogram for Agglomerative Hierarchical Clustering")
```

```
plt.xlabel("Samples")
```

```
plt.ylabel("Euclidean Distance")
```

```
plt.show()
```

```
cluster = AgglomerativeClustering(n_clusters=3, linkage='ward')
```

```
labels = cluster.fit_predict(X_scaled)
```

```
df = pd.DataFrame(X, columns=feature_names)
```

```

df['Cluster'] = labels

print("\nClustered Data (first 10 rows):\n", df.head(10))

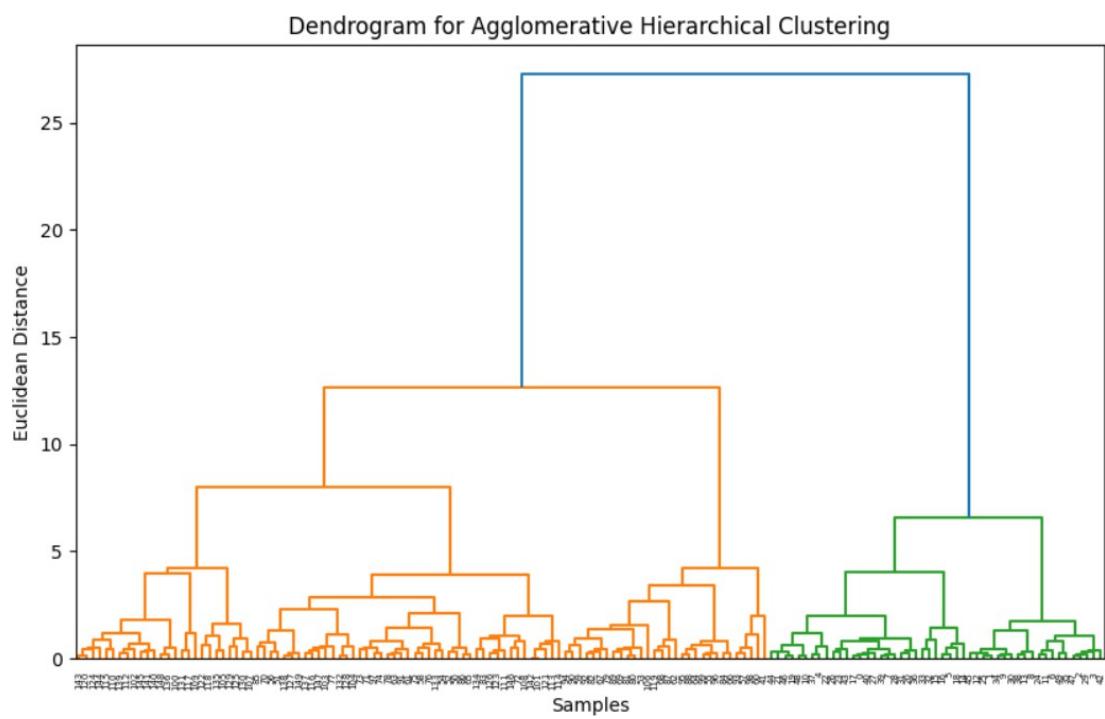
plt.figure(figsize=(8,6))

plt.scatter(df.iloc[:,0], df.iloc[:,1], c=df['Cluster'], cmap='rainbow')

plt.xlabel(feature_names[0])
plt.ylabel(feature_names[1])
plt.title("Agglomerative Clustering Result")
plt.show()

```

Output:



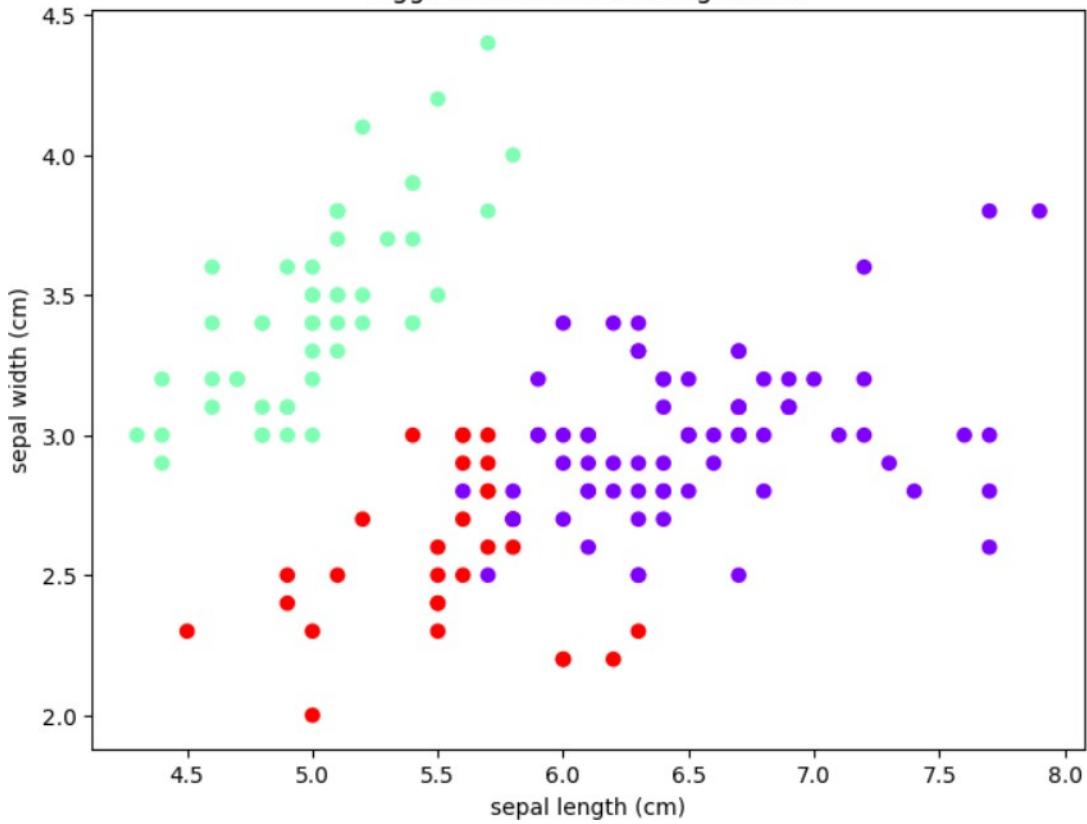
Clustered Data (first 10 rows):

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	\
0	5.1	3.5	1.4	0.2	
1	4.9	3.0	1.4	0.2	
2	4.7	3.2	1.3	0.2	
3	4.6	3.1	1.5	0.2	
4	5.0	3.6	1.4	0.2	
5	5.4	3.9	1.7	0.4	
6	4.6	3.4	1.4	0.3	
7	5.0	3.4	1.5	0.2	
8	4.4	2.9	1.4	0.2	
9	4.9	3.1	1.5	0.1	

Cluster

0	1
1	1
2	1
3	1
4	1
5	1
6	1
7	1
8	1
9	1

Agglomerative Clustering Result



Practical No:05

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Roll No:41

```
import numpy as np
```

```
import pandas as pd
```

```
graph = {
```

```
    "A": ["B", "C"],
```

```
    "B": ["C"],
```

```
    "C": ["A"],
```

```
    "D": ["C"]
```

```
}
```

```
pages = list(graph.keys())
```

```
N = len(pages)
```

```
damping = 0.85
```

```
pr = {page: 1/N for page in pages}
```

```
iterations = 100
```

```
for i in range(iterations):
```

```
    new_pr = {}
```

```
    for page in pages:
```

```
        rank_sum = 0
```

```
        for p in pages:
```

```
            if page in graph[p]:
```

```
                rank_sum += pr[p] / len(graph[p])
```

```
                new_pr[page] = (1 - damping) / N + damping * rank_sum
```

```
    pr = new_pr
```

```
pr_df = pd.DataFrame(list(pr.items()), columns=["Page", "PageRank"])
```

```
pr_df=pr_df.sort_values(by="PageRank", ascending=False)
print(pr_df)
```

Output:

	Page	PageRank
2	C	0.394149
0	A	0.372527
1	B	0.195824
3	D	0.037500