Machine learning

Q-3. Imagine you have a dataset where you have different categories of data, Now you need to find the most similar data to the given data by using any 4 different similarity algorithms. Now you have to build a model which can find the most similar data to the given data.

Dataset This is the Dataset You can use this dataset for this question.

In [1]:

```
## Import the necessary libraries:-
import pandas as pd
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.metrics.pairwise import cosine_similarity, euclidean_distances, manhat
```

In [3]:

```
# Load the dataset
data = pd.read_json("Downloads/archive (3)/News_Category_Dataset_v3.json", lines=Tr
```

In [4]:

```
1 ## Checking Top 5 Rows
2 data.head()
```

Out[4]:

1 https://www.huffpost.com/entry/covid-boosters Roll Up Sleeves For O Americans Roll Up Sleeves For O American Airlines Flyer Charged, Banned For Li 1 https://www.huffpost.com/entry/funniest-tweets 1 https://www.huffpost.com/entry/funniest-tweets 1 https://www.huffpost.com/entry/funniest-tweets 1 https://www.huffpost.com/entry/funniest-parent 23 Of The Funniest Tweets About Cats And Dogs The Funniest Tweets From Parents This Week (Se Woman Who Called Cops On Black Bird-Watcher Lo 4 https://www.huffpost.com/entry/amy-cooper-lose Americans Roll Up U.S. NEWS Health experts said it is too early to predict Af Ohnson AF Carla K Johnson AF Wan, Papenfus: PARENTING "Until you have a dog you don't understand wha Papenfus: PARENTING grown-up toothpaste on my toothpaste on my to Papenfus: PARENTING Papenfus Investment firm Franklin Te Papenfus: Papenfus Investment firm Golgowsk		link	headline	category	short_description	authors
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4 https://www.huffpost.com/entry/amy-coper-lose Cops On cooper-lose Black Bird-Watcher Who Called Amy Cooper accused Nina investment firm Golgowsk Franklin Te	3	·	Funniest Tweets From Parents This Week	PARENTING	grown-up toothpaste on my	
4	4		Who Called Cops On Black Bird- Watcher	U.S. NEWS	accused investment firm	
	4					>

In [5]:

```
# Select relevant columns for analysis
data = data[['category', 'headline', 'short_description']]
```

In [6]:

```
1 # Preprocess the data
2 data['text'] = data['headline'] + ' ' + data['short_description']
```

In [7]:

```
# Vectorize the text data
vectorizer = TfidfVectorizer()
X = vectorizer.fit_transform(data['text'])
```

In [8]:

```
# Function to find the most similar data using different similarity algorithms
 1
 2
   def find_similar_data(query, top_n=5):
 3
       # Vectorize the query
 4
       query_vector = vectorizer.transform([query])
 5
       # Calculate similarities using different algorithms
 6
 7
       cosine sim = cosine similarity(X, query vector).flatten()
 8
       euclidean_sim = euclidean_distances(X, query_vector).flatten()
 9
       manhattan_sim = manhattan_distances(X, query_vector).flatten()
10
11
       # Combine similarities from different algorithms
12
       similarity_scores = (cosine_sim + euclidean_sim + manhattan_sim ) / 4
13
14
       # Find the indices of top similar data points
15
16
       top_indices = similarity_scores.argsort()[-top_n:][::-1]
17
       # Return the top similar data points
18
       similar_data = data.iloc[top_indices]
19
20
21
       return similar_data
```

In [9]:

```
# Example usage
query = "New research on climate change"
similar_data = find_similar_data(query)
print(similar_data)
```

```
category
                                           headline
109802 WORLDPOST
                  Weekend Roundup: Laughing at God
66816
        POLITICS
                                     Sunday Roundup
                                     Sunday Roundup
63109
        POLITICS
107893
        POLITICS
                                     Sunday Roundup
                                     Sunday Roundup
64398
        POLITICS
                                        short_description \
109802 The first principle of an open society is not ...
        This week the nation watched as the #NeverTrum...
66816
        This week, the nation was reminded, in ways bo...
63109
        This week began with "The Horrible Call" final...
107893
        This week started off with the horror in Orlan...
64398
                                                     text
109802 Weekend Roundup: Laughing at God The first pri...
        Sunday Roundup This week the nation watched as...
66816
        Sunday Roundup This week, the nation was remin...
63109
        Sunday Roundup This week began with "The Horri...
107893
64398
        Sunday Roundup This week started off with the ...
```

1