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#Import all necessary packages
import requests
import os
import bs4
from apiclient.discovery import build
import string
from youtube_transcript_api import YouTubeTranscriptApi
import csv
import json
import pandas as pd
from sentence transformers import SentenceTransformer
from sklearn import preprocessing
from textblob import TextBlob
from scipy.spatial import distance
import pickle
#Load the text embedding model
vectorize = SentenceTransformer('paraphrase-multilingual-mpnet-base-v2')
#Initialize the parameters for the Youtube API
DEVELOPER KEY = "AIzaSyDqlrLX0prKppP3eG02uqprQJk080XoKrc"
YOUTUBE API_SERVICE_NAME = "youtube"
YOUTUBE_API_VERSION = "v3"
#Build the Youtube API Object
youtube object = build(YOUTUBE API SERVICE NAME, YOUTUBE API VERSION, developerKey =
DEVELOPER KEY)
#Function to classify a video as Clickbait or Not Clickbait
#vID is the Video ID of the Youtube Video
def isclickbait(vID):
  #Loading the model
  model = pickle.load(open('/content/isclickbait.sav','rb'))
  #Creating the object used for normalizing the values
  mms = preprocessing.MinMaxScaler()
  #Words often associated with clickbait stored as set to find intersection
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"misinformation", "credible", "crap", "tosh", "dumb",
                         "life", "time", "unreliable", "lack", "bogus", "fraud", "phony", "scam", "sham", "trick",
"cheat", "hoax", "rumour", "rumor",
                         "gossip"}
 #Range of data of the features used in training, required for normalizing the
data
 ranges = {'likes':[[-1],[10051603]],
            'views': [[367],[213578965]],
            'dislikes': [[-1],[386489]],
            'noOfComments' : [[0],[100]],
            'count_fake_comment' : [[0],[20]],
            'neg_comments' : [[0],[37]]
           }
 #Extract video related data / meta-data
video stat=youtube object.videos().list(part='snippet,statistics,contentDetails',id
=vID).execute()
  likes = video_stat['items'][0]['statistics']['likeCount'] if 'likeCount' in
video_stat['items'][0]['statistics'] else -1
 views = video stat['items'][0]['statistics']['viewCount']
  dislikes = video_stat['items'][0]['statistics']['dislikeCount']
 ldRatio = (int(likes) - int(dislikes))/(int(likes) + int(dislikes))
 #Extract the comments of the video
 comment data = []
 try:
video response=youtube object.commentThreads().list(part='snippet,replies',videoId=
vID,maxResults=100).execute()
    for item in video response['items']:
                   comment =
item['snippet']['topLevelComment']['snippet']['textDisplay']
                   comment data.append(comment)
 except:
    print("Disabled Comments, get search URLs")
    url = 'https://google.com/search?q='
    request_result=requests.get( url + meta_data[i]['title'] + " fake")
    soup = bs4.BeautifulSoup(request_result.text,
                           "html.parser")
    comment_data = [i.getText() for i in soup.find_all( 'h3' )]
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#Extract the audio transcript of the video
 try:
   audioTranscript = YouTubeTranscriptApi.get transcript(vID)
   transcript = ""
   for i in audioTranscript:
     transcript += i['text'] + " "
 except Exception as e:
     transcript = " ".join(comment_data)
 noOfComments = len(comment data)
 #Count the no of comments which claim the video to be 'fake' or 'Clickbait'
 count_fake_comment = 0
 neg_comments = 0
 for comment in comment_data:
      processed comment = comment.translate(str.maketrans('', '',
string.punctuation))
     if TextBlob(comment).sentiment.polarity < 0:</pre>
       neg comments += 1
     unique_words_comment = set(processed_comment.split(" "))
      if len(clickbait words corpus.intersection(unique words comment)) >= 1:
          count fake comment += 1
 try:
     #Calculate Fake Comment Ratio and Positive comment ratio
     fake_comment_ratio = count_fake_comment / len(comment_data)
     pos ratio = (noOfComments - neg comments) / len(comment data)
 except ZeroDivisionError:
     fake_comment_ratio = -1
     pos ratio = -1
 #Get the title of the video
 title = video_stat['items'][0]['snippet']['title']
 #Vectorize the title of the video
 title = vectorize.encode(title)
 #Vectorize the audio transcript of the video
 transcript = vectorize.encode(transcript)
 #Find the cosine similarity of the above vectors
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titleVtranscript = distance.cosine(title,transcript)
 #Normalizing the data using the initialized ranges
 ranges['likes'].insert(1,[likes])
 ranges['views'].insert(1,[views])
 ranges['dislikes'].insert(1,[dislikes])
 ranges['noOfComments'].insert(1,[noOfComments])
 ranges['count_fake_comment'].insert(1,[count_fake_comment])
 ranges['neg comments'].insert(1,[neg comments])
 likes = mms.fit_transform(ranges['likes'])[1][0]
 views = mms.fit transform(ranges['views'])[1][0]
 dislikes = mms.fit transform(ranges['dislikes'])[1][0]
 noOfComments = mms.fit transform(ranges['noOfComments'])[1][0]
  count_fake_comment = mms.fit_transform(ranges['count_fake_comment'])[1][0]
 neg_comments = mms.fit_transform(ranges['neg_comments'])[1][0]
 #Storing all the data collected in a single list for prediction
  data =
[likes,views,dislikes,ldRatio,noOfComments,count fake comment,fake comment ratio,ti
tleVtranscript, neg comments, pos ratio]
 #Prediction
 res = (model.predict_proba([data])[::,1]*100)[0]
 #Print the results
 print("Our Model predicted that this video is {:.2f}% clickbait".format(res))
```