# WELCOME

# IBM NAAN MUTHALVAN PROJECT

# ARTIFICIAL INTELLIGENCE(AI)

### **Project Title:** Fake News Detection Using NLP

#### **BATCH MEMBERS:**

AKILA J -951321106001

DAYANA S -951321106007

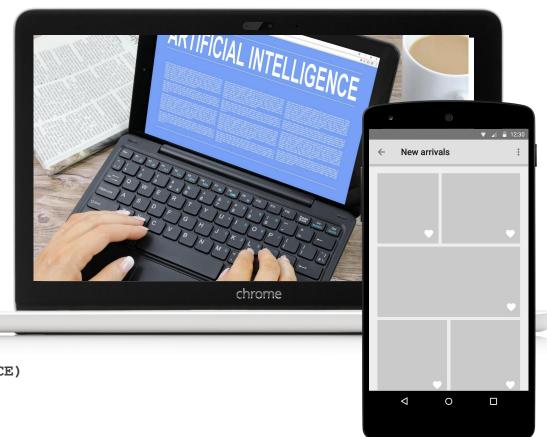
DIVYA DHARSHINI R -951321106008

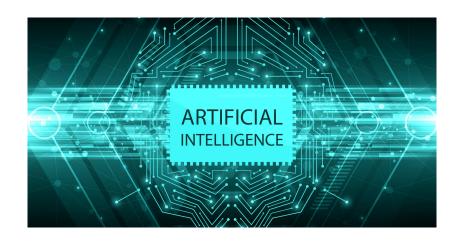
JERLIN PRABHA J -951321106018

RATHINA MALA B -951321106039

#### MENTOR:

Dr.M.RUBAN GLADWIN, M.E, Ph.d, (AP/ECE)





### **INTRODUCTION OF AI:**

Intelligence: "The capacity to learn and solve problems'.

Artificial intelligence is the simulation of humans intelligence by machines.



#### **INTRODUCTION OF FAKE NEWS DETECTION:**

Fake news is the intentional broadcasting of false or misleading claims as news, where the statements are purposely deceitful.

### **Problem Statement:**

The fake news dataset is one of the classic text analytics datasets available on Kaggle. It consists of genuine and fake articles' titles and text from different authors. Our job is to create a model which predicts whether a given news is real or fake.

### **Problem Definition:**

The problem is to develop a fake news detection model using a Kaggle dataset. The goal is to distinguish between genuine and fake news articles based on their titles and text. This project involves using natural language processing (NLP) techniques to preprocess the text data, building a machine learning model for classification, and evaluating the model's performance.

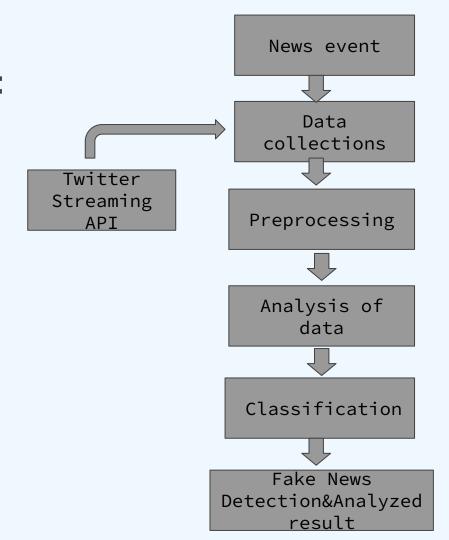
- Input: News items, social contexts and associated side information
- Output: One of two labels: 'fake' or 'real'.

# **Design Thinking:**

- 1. Data Source: Choose the fake news dataset available on Kaggle, containing articles titles and text, along with their labels (genuine or fake).
- 2. Data Preprocessing: Clean and preprocess the textual data to prepare it for analysis.
- 3. Feature Extraction: Utilize techniques like TF-IDF (Term Frequency-Inverse Document Frequency) or word embeddings to convert text into numerical features.
- 4. Model Selection: Select a suitable classification algorithm (e.g., Logistic Regression, Random Forest, or Neural Networks) for the fake news detection task.
- 5. Model Training: Train the selected model using the preprocessed data.
- 6. Evaluation: Evaluate the model's performance using metrics like accuracy, precision, recall, F1-score, and ROC-AUC.

### **BLOCK DIAGRAM:**

\_\_\_\_



## **Software Tools & Program Languages:**

- 1) pandas (gateway into the world of data science)
- 2) numby (fundamental Python package for scientific computations)
- 3) sklearn (the most useful and robust library for machine learning in Python and classification of the function)
- 4)TfidfVectorizer

### Programming language:

1) Python

## Innovation techniques:

#### IT'S SOLUTION:

It is vital to recognize and differentiate between false and accurate news. One method is to have an expert decide, and fact checks every piece of information, but this takes time and needs expertise that cannot be shared. Secondly, we can use machine learning and artificial intelligence tools to automate the identification of fake news.

Online news information includes various unstructured format data (such as documents, videos, and audio), but we will concentrate on text format news here. With the progress of machine learning and Natural language processing, we can now recognize the misleading and false character of an article or statement.

### **Data Exploration**

In this work, we utilized the fake news dataset from Kaggle to classify untrustworthy news articles as fake news. We have a complete training dataset containing the following characteristics:

### **Data Cleaning for Analysis**

In this section, we will clean our dataset to do some analysis:

- Drop unused rows and columns.
- Perform null value imputation.
- Remove special characters.
- Remove stop words.

# Data cleaning for program:

```
1.# Constants that are used to sanitize the datasets
2.column_n = ['id', 'title', 'author', 'text', 'label']
3.remove_c = ['id', 'author']
4.categorical_features = []
5.target_col = ['label']
6.text_f = ['title', 'text']
```

```
# Clean Datasets
import nltk
from nltk.corpus import stopwords
import re
from nltk.stem.porter import PorterStemmer
from collections import Counter
ps = PorterStemmer()
wnl = nltk.stem.WordNetLemmatizer()
stop_words = stopwords.words('english')
stopwords_dict = Counter(stop_words)
# Removed unused clumns
def remove_unused_c(df,column_n=remove_c):
    df = df.drop(column_n,axis=1)
    return df
```

```
# Cleaning text from unused characters
def clean_text(text):
   text = str(text).replace(r'http[\w:/\.]+', ' ') # removing urls
   text = str(text).replace(r'[^\.\w\s]', '') # remove everything but characters and punctuation
   text = str(text).replace('[^a-zA-Z]', ' ')
   text = str(text).replace(r'\s\s+', ' ')
   text = text.lower().strip()
   #text = ' '.join(text)
   return text
## Nltk Preprocessing include:
# Stop words, Stemming and Lemmetization
# For our project we use only Stop word removal
def nltk_preprocess(text):
   text = clean text(text)
   wordlist = re.sub(r'[^\w\s]', '', text).split()
   #text = ' '.join([word for word in wordlist if word not in stopwords_dict])
   #text = [ps.stem(word) for word in wordlist if not word in stopwords_dict]
   text = ' '.join([wnl.lemmatize(word) for word in wordlist if word not in stopwords_dict])
   return text
```

### **Preprocessing:**

```
# Perform data cleaning on train and test dataset by calling clean_dataset function

df = clean_dataset(news_d)

# apply preprocessing on text through apply method by calling the function nltk_preprocess

df ["text"] = df.text.apply(nltk_preprocess)

# apply preprocessing on title through apply method by calling the function nltk_preprocess

df ["title"] = df.title.apply(nltk_preprocess)

Copy
# Dataset after cleaning and preprocessing step

df.head()
```

#### Output:

```
title
                 label
        text
        house dem aide didnt even see comeys letter ja . .
                                                             house dem aide didnt even see comeys letter ja . .
0
1
        flynn hillary clinton big woman campus breitbart
                                                             ever get feeling life circle roundabout rather . .
2
        truth might get fired
                                                    truth might get fired october 29 2016 tension ... 1
3
        15 civilian killed single u airstrike identified
                                                             video 15 civilian killed single u airstrikeid...
        iranian woman jailed fictional unpublished sto . .
                                                             print iranian woman sentenced six year pri
```

### Let's make a function that accepts the article text as an argument and return whether it's fake or not:

def get prediction(text, convert to label=False):

```
# prepare our text into tokenized sequence
inputs = tokenizer(text, padding=True, truncation=True, max length=max length, return tensors="pt").to("cuda")
# perform inference to our model
outputs = model(**inputs)
# get output probabilities by doing softmax
probs = outputs[0].softmax(1)
# executing argmax function to get the candidate label
d = {
   0: "reliable",
   1: "fake"
if convert to label:
  return d[int(probs.argmax())]
else:
    return int(probs.argmax())
```

```
real_news = """
```

Tim Tebow Will Attempt Another Comeback, This Time in Baseball - The New York Times", Daniel Victor, "If at first you don't succeed, try a different sport. Tim Tebow, who was a Heisman quarterback at the University of Florida but was unable to hold an N. F. L. job, is pursuing a career in Major League Baseball. <SNIPPED>

11 11 11

get prediction(real news, convert to label=True)

### **OUTPUT:**

reliable

### Steps For Detecting Fake News with Python:

Follow the below steps for detecting fake news and complete your first advanced Python Project –

#### **Step 1**:Make necessary imports:

- 1. import numpy as np
- 2.import pandas as pd
- 3.import itertools
- 4.from sklearn.model\_selection import train\_test\_split
- 5.from sklearn.feature\_extraction.text import TfidfVectorizer
- 6.from sklearn.linear\_model import PassiveAggressiveClassifier
- 7.from sklearn.metrics import accuracy\_score, confusion\_matrix

# Output screenshot:

**step2**:. Now, let's read the data into a DataFrame, and get the shape of the data and the first 5 records

```
1. #Read the data
 2.df=pd.read_csv('D:\\DataFlair\\news.csv')
 3.#Get shape and head
 4.df.shape
 5.df.head()
 output:
                      [2]: #Read the data
                             df=pd.read csv('D:\\DataFlair\\news.csv')
                             #Get shape and head
                             df.shape
                             df.head()
                                                                                      title
                                Unnamed: 0
                                                                                                                                     text label
                                       8476
                                                                  You Can Smell Hillary's Fear
                                                                                                 Daniel Greenfield, a Shillman Journalism Fello...
                             0
                                                                                               Google Pinterest Digg Linkedin Reddit Stumbleu...
                                            Watch The Exact Moment Paul Ryan Committed Pol...
                             2
                                       3608
                                                     Kerry to go to Paris in gesture of sympathy
                                                                                                 U.S. Secretary of State John F. Kerry said Mon... REAL
                             3
                                                 Bernie supporters on Twitter erupt in anger ag... - Kaydee King (@KaydeeKing) November 9, 2016 T... FAKE
                                      10142
                                        875
                                               The Battle of New York: Why This Primary Matters
                                                                                                It's primary day in New York and front-runners... REAL
```

# Output for data Frame:

**step3:** And get the labels from the DataFrame. 1.#DataFlair - Get the labels 2.labels=df.label 3.labels.head() output: [3]: #DataFlair - Get the labels labels=df.label labels.head() [3]: 0 FAKE FAKE REAL FAKE REAL Name: label, dtype: object

# Split the dataset:

**step4:** Split the dataset into training and testing sets.

```
1.#DataFlair - Split the dataset
2.x_train,x_test,y_train,y_test=train_test_split(df['text'], labels, test_size=0.2, random_state=7)
```

**Step5:** Let's initialize a <u>TfidfVectorizer</u> with stop words from the English language and a maximum document frequency of 0.7 (terms with a higher document frequency will be discarded). Stop words are the most common words in a language that are to be filtered out before processing the natural language data. And a TfidfVectorizer turns a collection of raw documents into a matrix of TF-IDF features.

### Classified the dataset:

```
1.#DataFlair - Initialize a TfidfVectorizer
2.tfidf_vectorizer=TfidfVectorizer(stop_words='english', max_df=0.7)
3.#DataFlair - Fit and transform train set, transform test set
4.tfidf_train=tfidf_vectorizer.fit_transform(x_train)
5.tfidf_test=tfidf_vectorizer.transform(x_test)
```

**step6:** Next, we'll initialize a PassiveAggressiveClassifier. This is. We'll fit this on tfidf\_train and y\_train.

Then, we'll predict on the <u>test set</u> from the TfidfVectorizer and calculate the accuracy with accuracy\_score() from sklearn.metrics. <u>output:</u>

```
[6]: #DataFlair - Initialize a PassiveAggressiveClassifier
pac=PassiveAggressiveClassifier(max_iter=50)
pac.fit(tfidf_train,y_train)

#DataFlair - Predict on the test set and calculate accuracy
y_pred=pac.predict(tfidf_test)
score=accuracy_score(y_test,y_pred)
print(f'Accuracy: {round(score*100,2)}%')
Accuracy: 92.82%
```

# Fake News Detection & Analysed result:

**Step7:** We got an accuracy of 92.82% with this model. Finally, let's print out a confusion matrix to gain insight into the number of false and true negatives and positives.

```
1.#DataFlair - Build confusion matrix
2.confusion_matrix(y_test,y_pred, labels=['FAKE','REAL'])
```

#### **Output screenshot:**

#### Source code:

fake

```
# This Python 3 environment comes with many helpful analytics libraries installed
# It is defined by the kaggle/python Docker image: https://github.com/kaggle/docker-python
# For example, here's several helpful packages to load
import warnings
warnings.filterwarnings('ignore')
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import nltk
from nltk.sentiment import SentimentIntensityAnalyzer
import warnings
warnings.filterwarnings("ignore")
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score, classification_report
# Input data files are available in the read-only "../input/" directory
# For example, running this (by clicking run or pressing Shift+Enter) will list all files under the input directory
import os
for dirname, _, filenames in os.walk('/kaggle/input'):
   for filename in filenames:
       print(os.path.join(dirname, filename))
# You can write up to 20GB to the current directory (/kaggle/working/) that gets preserved as output when you create a version using "Save & Run All"
# You can also write temporary files to /kaggle/temp/, but they won't be saved outside of the current session
/kaggle/input/fake-and-real-news-dataset/True.csv
  /kaggle/input/fake-and-real-news-dataset/Fake.csv
 Loading Data:
 true = pd.read_csv('/kaggle/input/fake-and-real-news-dataset/True.csv')
 fake = pd.read_csv('/kaggle/input/fake-and-real-news-dataset/Fake.csv')In [3]:
 linkcode
 fake['Category'] = 'fake'
```

### output:

s.n	title	text	subje t	date	cate gory
0	Donald Trump Sends Out Embarrassing New Year'	Donald Trump just couldn t wish all Americans	News	December 31, 2017	fake
1	Drunk Bragging Trump Staffer Started Russian	House Intelligence Committee Chairman Devin Nu	News	December 31, 2017	fake
2	Sheriff David Clarke Becomes An Internet Joke	On Friday, it was revealed that former Milwauk	News	December 30, 2017	fake
3	Trump Is So Obsessed He Even Has Obama's Name	On Christmas day, Donald Trump announced that	News	December 29, 2017	fake
4	Pope Francis Just Called Out Donald Trump Dur	Pope Francis used his annual Christmas Day mes	News	December 25, 2017	fake

23476	McPain: John McCain Furious That Iran Treated	21st Century Wire says As 21WIRE reported earl	Middle-east	January 16, 2016	fake
23477	JUSTICE? Yahoo Settles E-mail Privacy Class-ac	21st Century Wire says It s a familiar theme	Middle-east	January 16, 2016	fake
23478	Sunnistan: US and Allied 'Safe Zone' Plan to T	Patrick Henningsen 21st Century WireRemember	Middle-east	January 15, 2016	fake
23479	How to Blow \$700 Million: Al Jazeera America F	21st Century Wire says Al Jazeera America will	Middle-east	January 14, 2016	fake
23480	10 U.S. Navy Sailors Held by Iranian Military	21st Century Wire says As 21WIRE predicted in	Middle-east	January 12, 2016	fake

```
true['Category'] = 'true'
True
```

#### Outout:

s.n o	Title	text		date	Cat eg ory
0	As U.S. budget fight looms, Republicans flip t	WASHINGTON (Reuters) - The head of a conservat	politicsNews	December 31, 2017	true
1	U.S. military to accept transgender recruits o	WASHINGTON (Reuters) - Transgender people will	politicsNews	December 29, 2017	true
2	Senior U.S. Republican senator: 'Let Mr. Muell	WASHINGTON (Reuters) - The special counsel inv	politicsNews	December 31, 2017	true
3	FBI Russia probe helped by Australian diplomat	WASHINGTON (Reuters) - Trump campaign adviser	politicsNews	December 30, 2017	true
4	Trump wants Postal Service to charge 'much mor	SEATTLE/WASHINGTON (Reuters) - President Donal	politicsNews	December 29, 2017	true

21112	Tany committee Taltie Backe flew C.C. approach	Brooders (readers) 19110 amos on raceday we	worldnews	7.09001.22, 2017	uuc
21413	LexisNexis withdrew two products from Chinese	LONDON (Reuters) - LexisNexis, a provider of I	worldnews	August 22, 2017	true
21414	Minsk cultural hub becomes haven from authorities	MINSK (Reuters) - In the shadow of disused Sov	worldnews	August 22, 2017	true
21415	Vatican upbeat on possibility of Pope Francis	MOSCOW (Reuters) - Vatican Secretary of State	worldnews	August 22, 2017	true
21416	Indonesia to buy \$1.14 billion worth of Russia	JAKARTA (Reuters) - Indonesia will buy 11 Sukh	worldnews	August 22, 2017	true

21412 'Fully committed' NATO backs new U.S. approach... | BRUSSELS (Reuters) - NATO allies on Tuesday we... | worldnews | August 22, 2017 | true

#Now let's combine the whole dataset into one
data = pd.concat([fake, true], ignore\_index = True)
data

s.n o	title	text	subje ct	date	Cat ego ry
0	Donald Trump Sends Out Embarrassing New Year'	Donald Trump just couldn t wish all Americans	News	December 31, 2017	fake
1	Drunk Bragging Trump Staffer Started Russian	House Intelligence Committee Chairman Devin Nu	News	December 31, 2017	fake
2	Sheriff David Clarke Becomes An Internet Joke	On Friday, it was revealed that former Milwauk	News	December 30, 2017	fake
3	Trump Is So Obsessed He Even Has Obama's Name	On Christmas day, Donald Trump announced that	News	December 29, 2017	fake
4	Pope Francis Just Called Out Donald Trump Dur	Pope Francis used his annual Christmas Day mes	News	December 25, 2017	fake

\_

44893	'Fully committed' NATO backs new U.S. approach	BRUSSELS (Reuters) - NATO allies on Tuesday we	worldnews	August 22, 2017	true
44894	LexisNexis withdrew two products from Chinese	LONDON (Reuters) - LexisNexis, a provider of I	worldnews	August 22, 2017	true
44895	Minsk cultural hub becomes haven from authorities	MINSK (Reuters) - In the shadow of disused Sov	worldnews	August 22, 2017	true
44896	Vatican upbeat on possibility of Pope Francis	MOSCOW (Reuters) - Vatican Secretary of State	worldnews	August 22, 2017	true
44897	Indonesia to buy \$1.14 billion worth of Russia	JAKARTA (Reuters) - Indonesia will buy 11 Sukh	worldnews	August 22, 2017	true

### Data.shape

Out[6]:

(44898, 5)

### Preprocessing:

44897

Name: Category, Length: 44898, dtype: int64

```
data['Category'].value_counts()
```

```
Category
fake
       23481
true
       21417
    Name: count, dtype: int64
                                                                                                                                                    In [8]:
#Transforming category values to numerical
from sklearn.preprocessing import LabelEncoder
encoder = LabelEncoder()
data['Category'] = encoder.fit_transform(data['Category'])
                                                                                                                                                    In [9]:
data['Category']
                                                                                                                                                     Out[9]:
        0
44893
44894
44895
44896
```

Out[7]:

```
Out[10]:
<44898x20896 sparse matrix of type '<class 'numpy.float64'>'
     with 546512 stored elements in Compressed Sparse Row format>
 Modeling:
 from sklearn.model_selection import train_test_split
 X = title
 y = data['Category']
 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2,
 random_state = 42)
```

In [12]:

```
model = SVC()
model.fit(X_train, y_train)

SVC
SVC()
```

title

title = vectorizer.fit\_transform(data['title'])

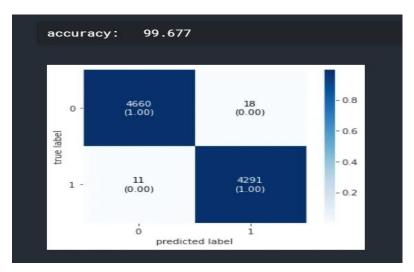
Out[12]:

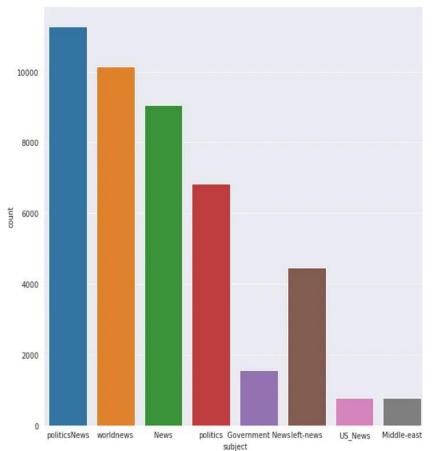
Out[12]

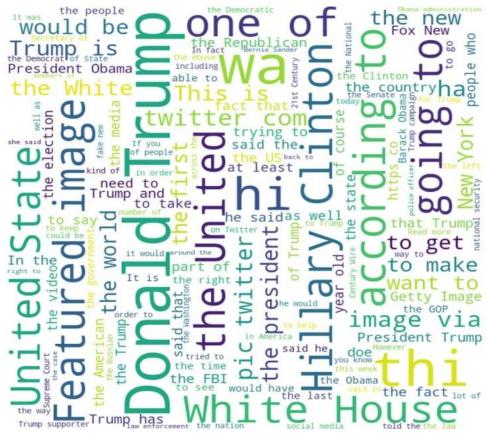
```
y_pred = model.predict(X_test)
print('Classification Report: ')
print(classification_report(y_test, y_pred))
In[13]:
```

#### Classification Report:

	precision	recall	f1-score	support
0	0.97	0.96	0.96	4733
1	0.95	0.97	0.96	4247
accuracy			0.96	8980
macro avg	0.96	0.96	0.96	8980
weighted av	g 0.9	6 0.9	0.9	6 8980







### Reference:

1)Artificial Intelligence for Humans by Jeff Heaton <a href="https://youtu.be/f7ckMaEJvh4?si=0EYjBfRCUjoKahgf">https://youtu.be/f7ckMaEJvh4?si=0EYjBfRCUjoKahgf</a>

- 2)https://youtu.be/2o-\_bMKGi\_o?si=y\_L8U6EVfIU\_tc93
- 3) <a href="https://youtu.be/DdgymahD7Zg?si=MTWrP0W1s7QvSQid">https://youtu.be/DdgymahD7Zg?si=MTWrP0W1s7QvSQid</a>
- 4) Combating Fake News with Computational Intelligence Techniques

#### books.googl

Mohamed Lahby, Al-Sakib Khan Pathan, Yassine Maleh · 2021

5)https://books.google.co.in/books?id=GfFUEAAAQBAJ&printsec=frontcover&dq=fake+news+detection+using+python+and+nlp+reference&hl=en&newbks=1&newbks\_redir=1&sa=X&ved=2ahUKEwj7h0iU3u0BAxWsbWwGHdD7BFYQ6AF6BAgFEAI

# Fake news Detection using NLP

THANKING YOU

