# **Fake News Detection**

## Overview of Problem Statement

Fake news has become a critical issue with the rise of online platforms. The spread of misinform can mislead the public and create social, political, or economic consequences. Detecting and cla fake news is therefore a significant challenge in Natural Language Processing (NLP).

# Objective

The main objective of this project is to develop a deep learning-based classifier that can accurate distinguish between Fake and Real news articles based on their textual content.

## Import Libraries

```
In [1]: # Import Libraries
        import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        import seaborn as sns
        from wordcloud import WordCloud
        from sklearn.model selection import train test split
        from sklearn.preprocessing import LabelEncoder
        from sklearn.metrics import accuracy score, classification report
        from tensorflow.keras.preprocessing.text import Tokenizer
        from tensorflow.keras.preprocessing.sequence import pad_sequences
        from tensorflow.keras.models import Sequential
        from tensorflow.keras.layers import Embedding, LSTM, Dense, Dropout, Bidi
        from tensorflow.keras.callbacks import EarlyStopping
In [ ]:
In [2]: # Step 2: Load Dataset
        df = pd.read csv('news.csv')
        df
```

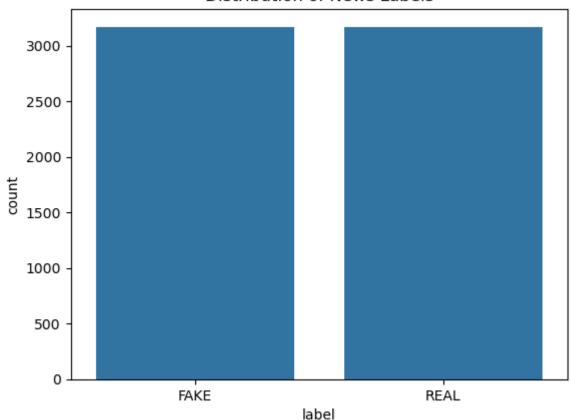
Out[2]:	Unnamed: 0	title	text	
0	8476	You Can Smell Hillary's Fear	Daniel Greenfield, a Shillman Journalism Fello	
1	10294	Watch The Exact Moment Paul Ryan Committed Pol	Google Pinterest Digg Linkedin Reddit Stumbleu	
2	3608	Kerry to go to Paris in gesture of sympathy	U.S. Secretary of State John F. Kerry said Mon	
3	10142	Bernie supporters on Twitter erupt in anger ag	— Kaydee King (@KaydeeKing) November 9, 2016 T	
4	875	The Battle of New York: Why This Primary Matters	It's primary day in New York and front- runners	
6330	4490	State Department says it can't find emails fro	The State Department told the Republican Natio	
6331	8062	The 'P' in PBS Should Stand for 'Plutocratic'	The 'P' in PBS Should Stand for 'Plutocratic'	
6332	8622	Anti-Trump Protesters Are Tools of the Oligarc	Anti-Trump Protesters Are Tools of the Oligar	
6333	4021	In Ethiopia, Obama seeks progress on peace, se	ADDIS ABABA, Ethiopia —President Obama convene	
6334	4330	Jeb Bush Is Suddenly Attacking Trump. Here's W	Jeb Bush Is Suddenly Attacking Trump. Here's W	

6335 rows × 4 columns

# Data Cleaning

Out[3]:		title	text	label	text_
	0	You Can Smell Hillary's Fear	Daniel Greenfield, a Shillman Journalism Fello	FAKE	
	1	Watch The Exact Moment Paul Ryan Committed Pol	Google Pinterest Digg Linkedin Reddit Stumbleu	FAKE	
	2	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	
	4	The Battle of New York: Why This Primary Matters	It's primary day in New York and front- runners	REAL	

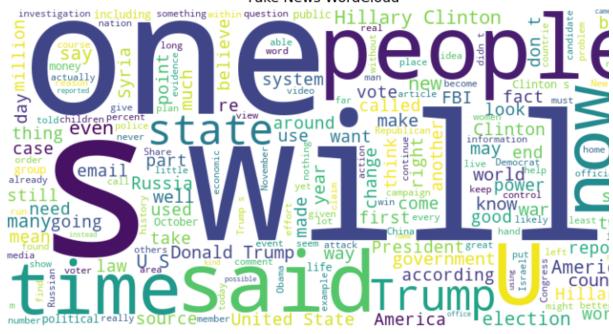
## Distribution of News Labels



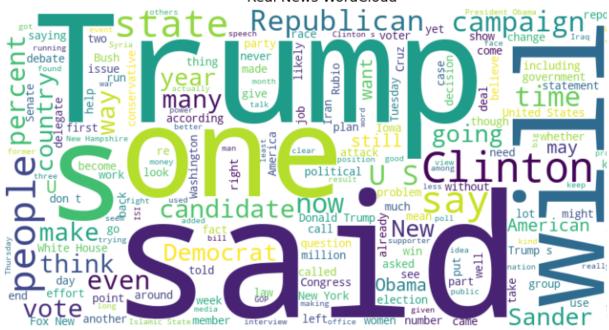
```
In [5]: # EDA - Word Cloud
    def generate_wordcloud(text, title):
        wordcloud = WordCloud(width=800, height=400, background_color='white'
        plt.figure(figsize=(10, 5))
        plt.imshow(wordcloud, interpolation='bilinear')
        plt.title(title)
        plt.axis('off')
        plt.show()

    generate_wordcloud(df[df['label'] == 'FAKE']['text'], 'Fake News WordCloudgenerate_wordcloud(df[df['label'] == 'REAL']['text'], 'Real News WordCloud
```

#### Fake News WordCloud



#### Real News WordCloud

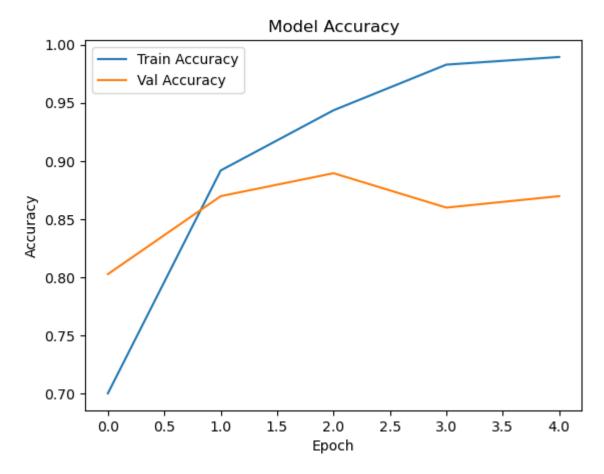


# **Encoding label**

```
In [8]: tokenizer = Tokenizer(num words=6000, oov token='<00V>')
        tokenizer.fit on texts(X train)
        X train seg = tokenizer.texts to sequences(X train)
        X test seq = tokenizer.texts to sequences(X test)
        X_train_pad = pad_sequences(X_train_seq, maxlen=500)
        X test pad = pad sequences(X test seq, maxlen=500)
In [9]: # Optimized LSTM Model
        model = Sequential([
            Embedding(input dim=6000, output dim=128, input length=500),
            SpatialDropout1D(0.3),
            Bidirectional(LSTM(64, return sequences=True)),
            Dropout (0.4),
            Bidirectional(LSTM(32)),
            Dense(64, activation='relu'),
            Dropout (0.5),
            Dense(1, activation='sigmoid')
        ])
        model.compile(loss='binary crossentropy', optimizer='adam', metrics=['acci
        early stop = EarlyStopping(monitor='val loss', patience=2, restore best we
        history = model.fit(X_train_pad, y_train, epochs=10, batch_size=64, validate)
       Epoch 1/10
       G:\Anaconda\Lib\site-packages\keras\src\layers\core\embedding.py:97: UserWarn
       Argument `input length` is deprecated. Just remove it.
         warnings.warn(
                              — 95s 1s/step - accuracy: 0.6403 - loss: 0.6455 -
       72/72 -
       val accuracy: 0.8028 - val loss: 0.4589
       Epoch 2/10
                               — 89s 1s/step - accuracy: 0.8727 - loss: 0.3258 -
       val_accuracy: 0.8698 - val_loss: 0.3187
       Epoch 3/10
                               — 130s 1s/step - accuracy: 0.9427 - loss: 0.1690 -
       72/72 -
       val_accuracy: 0.8895 - val_loss: 0.2949
       Epoch 4/10
       72/72 ———
                   82s 1s/step - accuracy: 0.9869 - loss: 0.0576 -
       val accuracy: 0.8600 - val loss: 0.3980
       Epoch 5/10
                              —— 136s 1s/step - accuracy: 0.9908 - loss: 0.0387 -
       72/72 -
       val_accuracy: 0.8698 - val_loss: 0.4606
```

```
In [10]: # Model Evaluation
         y pred prob = model.predict(X test pad)
         y_pred = (y_pred_prob > 0.5).astype(int)
         print('Accuracy:', accuracy_score(y_test, y_pred))
         print(classification_report(y_test, y_pred, target_names=le.classes_))
        40/40
                                   - 8s 179ms/step
        Accuracy: 0.8997632202052092
                       precision
                                    recall f1-score
                                                        support
                FAKE
                            0.91
                                      0.88
                                                 0.90
                                                            628
                                      0.92
                                                 0.90
                REAL
                            0.89
                                                            639
                                                 0.90
                                                           1267
            accuracy
                            0.90
                                      0.90
                                                 0.90
                                                           1267
           macro avg
        weighted avg
                            0.90
                                      0.90
                                                 0.90
                                                           1267
```

```
In [11]: # Accuracy Over Epochs
    plt.plot(history.history['accuracy'], label='Train Accuracy')
    plt.plot(history.history['val_accuracy'], label='Val Accuracy')
    plt.title('Model Accuracy')
    plt.xlabel('Epoch')
    plt.ylabel('Accuracy')
    plt.legend()
    plt.show()
```



```
In [12]: from tensorflow.keras.preprocessing.sequence import pad sequences
         # Input news
         X = "Karry to go to France in gesture of sympathy"
         # Define padding and truncation types
         padding type = 'post'
         trunc type = 'post'
         # Convert input to sequence
         sequences = tokenizer.texts_to_sequences([X])
         padded = pad sequences(sequences, maxlen=500, padding=padding type, trunca
         # Predict using the model
         prediction = model.predict(padded, verbose=0)[0][0]
         # Interpret result
         if prediction >= 0.5:
             print("This news is True (REAL)")
             print("This news is False (FAKE)")
        This news is False (FAKE)
```

## Results

Accuracy: ~90% (based on evaluation in notebook).

Classification Report: Balanced precision, recall, and F1-scores for both Fake and Real labels.

Training Curve: Accuracy improved steadily, validation stabilized after early stopping.

## Conclusion

The Bidirectional LSTM model effectively classifies Fake vs Real news with high accuracy. The p demonstrates the potential of deep learning for fake news detection in NLP

```
In [ ]:
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```