

## Import Libraries

```
import pandas as pd
import numpy as np
import re
import string
import tensorflow as tf
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Embedding, LSTM, Dense, Dropout
```

## 2. Load Dataset

```
df = pd.read_csv("Fake_New_data.csv")

print(df.head())
print(df.isnull().sum())
```

```
   Unnamed: 0      title \
0      8476      You Can Smell Hillary's Fear
1    10294  Watch The Exact Moment Paul Ryan Committed Pol...
2     3608      Kerry to go to Paris in gesture of sympathy
3    10142  Bernie supporters on Twitter erupt in anger ag...
4       875  The Battle of New York: Why This Primary Matters

      text label
0  Daniel Greenfield, a Shillman Journalism Fello...  FAKE
1  Google Pinterest Digg LinkedIn Reddit Stumbleu...  FAKE
2  U.S. Secretary of State John F. Kerry said Mon...  REAL
3  - Kaydee King (@KaydeeKing) November 9, 2016 T...  FAKE
4  It's primary day in New York and front-runners...  REAL
Unnamed: 0      0
title           0
text           0
label          0
dtype: int64
```

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print(df.isnull().sum())
```

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Unnamed: 0      0
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text           0
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```

## 3. Preprocessing

```
def clean_text(text):
    text = str(text).lower()
    text = re.sub(r"http\S+", "", text)          # remove URLs
    text = re.sub(r"[^a-zA-Z\s]", "", text)      # remove numbers, punctuation
    text = text.translate(str.maketrans("", "", string.punctuation))
    text = re.sub(r"\s+", " ", text).strip()     # remove extra spaces
    return text

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

# Convert labels (FAKE/REAL) to 0/1
df["label"] = df["label"].map({"FAKE": 0, "REAL": 1})
```

## 4. Tokenization + Sequencing

```
X = df["clean_text"].values
y = df["label"].values

tokenizer = Tokenizer(num_words=50000, oov_token="<OOV>")
tokenizer.fit_on_texts(X)
```

```
sequences = tokenizer.texts_to_sequences(X)
padded = pad_sequences(sequences, maxlen=300)
```

## 5. Train-Test Split

```
X_train, X_test, y_train, y_test = train_test_split(
    padded, y, test_size=0.2, random_state=42
)
```

## 6. LSTM Model

```
model = Sequential([
    Embedding(input_dim=50000, output_dim=128, input_length=300),
    LSTM(128, return_sequences=False),
    Dropout(0.3),
    Dense(64, activation='relu'),
    Dropout(0.3),
    Dense(1, activation='sigmoid')
])

model.compile(
    loss="binary_crossentropy",
    optimizer="adam",
    metrics=["accuracy"]
)

model.summary()
```

/usr/local/lib/python3.12/dist-packages/keras/src/layers/core/embedding.py:97: UserWarning: Argument `input\_length` is deprecated. Just rer  
warnings.warn(  
Model: "sequential"

Layer (type)	Output Shape	Param #
embedding ( <a href="#">Embedding</a> )	?	0 (unbuilt)
lstm ( <a href="#">LSTM</a> )	?	0 (unbuilt)
dropout ( <a href="#">Dropout</a> )	?	0
dense ( <a href="#">Dense</a> )	?	0 (unbuilt)
dropout_1 ( <a href="#">Dropout</a> )	?	0
dense_1 ( <a href="#">Dense</a> )	?	0 (unbuilt)

Total params: 0 (0.00 B)  
Trainable params: 0 (0.00 B)  
Non-trainable params: 0 (0.00 B)

## 7. Train Model

```
history = model.fit(
    X_train, y_train,
    validation_split=0.2,
    epochs=5,
    batch_size=64,
    verbose=1
)
```

```
Epoch 1/5
64/64 ————— 85s 1s/step - accuracy: 0.5911 - loss: 0.6668 - val_accuracy: 0.7633 - val_loss: 0.4650
Epoch 2/5
64/64 ————— 59s 905ms/step - accuracy: 0.8833 - loss: 0.2966 - val_accuracy: 0.8629 - val_loss: 0.3547
Epoch 3/5
64/64 ————— 46s 716ms/step - accuracy: 0.9690 - loss: 0.1085 - val_accuracy: 0.8462 - val_loss: 0.4242
Epoch 4/5
64/64 ————— 47s 741ms/step - accuracy: 0.9963 - loss: 0.0141 - val_accuracy: 0.8383 - val_loss: 0.6100
Epoch 5/5
64/64 ————— 48s 746ms/step - accuracy: 0.9994 - loss: 0.0042 - val_accuracy: 0.8442 - val_loss: 0.6027
```

## 8. Evaluation

```
pred = (model.predict(X_test) > 0.5).astype("int32")

acc = accuracy_score(y_test, pred)
prec = precision_score(y_test, pred)
rec = recall_score(y_test, pred)
f1 = f1_score(y_test, pred)

print("Accuracy:", acc)
print("Precision:", prec)
print("Recall:", rec)
print("F1 Score:", f1)
```

**40/40** ————— 5s 124ms/step  
Accuracy: 0.8492501973164956  
Precision: 0.82  
Recall: 0.8982785602503912  
F1 Score: 0.8573562359970127

## 9. Make Prediction

```
def predict_news(text):  
    cleaned = clean_text(text)  
    seq = tokenizer.texts_to_sequences([cleaned])  
    pad = pad_sequences(seq, maxlen=300)  
    prob = model.predict(pad)[0][0]  
    return "REAL" if prob > 0.5 else "FAKE"  
  
print(predict_news("Breaking! Scientists discover new planet."))
```

**1/1** ————— 0s 66ms/step  
FAKE