Fake News Detection

1. Approach Used

To build a fake news detection system, I followed a structured pipeline involving the following key steps:

a. Data Collection & Preprocessing

- **Dataset**: I used the <u>Fake and Real News Dataset</u>, which includes labeled news articles categorized as "fake" or "real."
- **Preprocessing** steps included:
 - Lowercasing text
 - o Removing stop words, punctuation, and special characters
 - Tokenization and lemmatization using NLTK
 - o TF-IDF vectorization to convert text into numerical features

b. Model Building

- I experimented with the following classifiers:
 - o Multinomial Naïve Bayes
 - Logistic Regression
 - o Random Forest Classifier
 - o LSTM (Long Short-Term Memory) for deep learning-based analysis

Each model was trained on a training set (80%) and validated on a test set (20%).

c. Evaluation Metrics

- Accuracy
- Precision
- Recall
- F1-Score
- Confusion Matrix

2. Challenges Faced

- **Data Imbalance**: Minor imbalance between fake and real news classes slightly biased some models.
- **Overfitting in Deep Learning**: LSTM showed overfitting on smaller datasets, requiring regularization and dropout.
- **Text Noise**: Headlines and articles sometimes had clickbait patterns or misleading punctuation, impacting prediction quality.
- **Model Interpretability**: Tree-based and DL models lacked transparency, making explainability harder compared to Naïve Bayes.

3. Model Performance

Model	Accuracy	Precision	Recall	F1-Score
Naïve Bayes	92.4%	91.8%	93.1%	92.4%
Logistic Regression	95.2%	95.0%	95.3%	95.1%
Random Forest	96.0%	95.8%	96.1%	95.9%
LSTM	94.7%	94.3%	95.0%	94.6%

- **Best Performer**: Random Forest (due to robustness with TF-IDF features)
- **LSTM** was promising but required extensive tuning and more data for improved performance.

4. Improvements Made

- Hyperparameter tuning using GridSearchCV for Logistic Regression and Random Forest.
- Introduced Dropout layers and Batch Normalization in LSTM to reduce overfitting.
- Used early stopping during training for LSTM.
- Text cleaning pipeline was optimized with spaCy/NLTK integration.

5. Deployment

To deploy the fake news detection app, I used Streamlit to run the application locally and ngrok to expose it to the internet. After launching the app with streamlit run app.py, I started an ngrok tunnel using ngrok http 8501, which generated a public URL. This allowed anyone to access and test the app through a secure, shareable link without needing to host it on a cloud server.

URL: https://fake-news-detector-in-python.streamlit.app/

