Project Title: Recipe Generation with LSTM Networks

Objective:

Develop a model to generate cooking recipes using the RecipeNLG dataset. Students will implement an LSTM-based neural network to learn from the dataset and generate coherent and contextually relevant recipes.

Dataset:

• Name: RecipeNLG

• Content: 2,231,142 cooking recipes

• **File:** RecipeNLG_dataset.csv (2.29 GB)

 Columns: Title, Ingredients, Directions, Link, Source, NER (Named Entity Recognition)

Project Outline:

1. Data Exploration and Preprocessing:

- Load the dataset: Use appropriate libraries (e.g., Pandas) to load and explore the CSV file.
- Data Cleaning: Handle missing values, duplicate entries, and any inconsistencies.
- Text Processing: Tokenize the text, remove stop words, and perform stemming/lemmatization if needed.
- Data Formatting: Convert recipes into sequences suitable for LSTM input. Separate ingredients and directions for targeted training.

2. Data Preparation:

- Sequence Creation: Create sequences of tokens from recipe directions for training the LSTM model.
- Train-Test Split: Divide the dataset into training and testing sets.

3. Model Building:

- Define the LSTM Model:
 - Input Layer: Embed the sequences using an embedding layer.
 - **LSTM Layers**: Implement one or more LSTM layers with dropout for regularization.
 - Dense Layer: Add a dense layer to output predictions.
 - Output Layer: Use a softmax activation function to predict the next token in the sequence.

 Compile the Model: Choose an appropriate loss function (e.g., categorical cross-entropy) and optimizer (e.g., Adam).

4. Training:

- o **Fit the Model**: Train the LSTM model on the training data.
- Monitor Performance: Track the loss and accuracy metrics during training. Use validation data to monitor overfitting.

5. Evaluation:

- Generate Recipes: Use the trained model to generate new recipes from a seed input.
- Evaluate Output: Assess the quality and coherence of the generated recipes. Consider using human evaluation or similarity metrics with existing recipes.

6. Documentation and Reporting:

- Code Documentation: Ensure all code is well-commented and organized.
- Final Report: Prepare a comprehensive report detailing the methodology, model performance, challenges faced, and results. Include code snippets, visualizations, and generated recipes.

7. Submission:

• **Deliverables**: Submit the code, a detailed report, and any supplementary materials (e.g., trained models, scripts).

Evaluation Criteria:

- Data Preprocessing and Cleaning: Effectiveness in handling and preparing data.
- Model Implementation: Accuracy and performance of the LSTM model.
- Recipe Generation Quality: Coherence and relevance of generated recipes.
- Documentation: Clarity and thoroughness of the final report and code documentation.

Additional Resources:

RecipeNLG Dataset