#### **Natural Language Processing**

## Assignment 1: Basic-Level - Text Preprocessing for Al Solution Platform

**Overview:** One Hour Al Solutions needs to process client problem descriptions to make them ready for analysis. Your task is to implement basic text preprocessing techniques covered in the educational material.

# Requirements:

- 1. Use the provided "client\_problems.csv" dataset containing AI problem descriptions from clients
- 2. Implement a complete text preprocessing pipeline that includes:
  - Tokenization (splitting text into words)
  - Converting text to lowercase
  - Removing punctuation and special characters
  - Removing stop words
  - Performing stemming or lemmatization
- 3. Compare the results of stemming vs. lemmatization on 5 example sentences
- 4. Calculate and report basic statistics about the processed text (average word count before and after, most common words)

# **Expected Output:**

- Processed text data with all preprocessing steps applied
- Comparison table showing original text, text after stemming, and text after lemmatization
- Summary statistics about the text corpus before and after preprocessing

**Dataset Description:** The "client\_problems.csv" will contain columns: problem\_id, problem description, difficulty level

# Assignment 2: Intermediate-Level - Bag of Words and TF-IDF for AI Problem Classification

**Overview:** One Hour AI Solutions wants to categorize incoming client problems to route them to appropriate engineers. Your task is to implement text representation techniques (Bag of Words and TF-IDF) and use them for simple classification.

#### Requirements:

- 1. Use the provided "categorized\_problems.csv" dataset containing problem descriptions and their categories
- 2. Implement:
  - A Bag of Words representation for the problem descriptions
  - o A TF-IDF representation for the same descriptions
- 3. Use a simple Naive Bayes classifier to:
  - Train a model using both representations
  - Predict the category of new problem descriptions
- 4. Compare the performance of BoW vs. TF-IDF approaches
- 5. Identify the most influential words for each category using TF-IDF scores

# **Expected Output:**

- Comparison of classification accuracy between BoW and TF-IDF representations
- List of top 5 most important words for each problem category
- Analysis of when each representation performs better or worse

**Dataset Description:** The "categorized\_problems.csv" will contain columns: problem\_id, problem\_description, problem\_category

## Assignment 3: Moderately Advanced - Rule-Based Sentiment Analysis for Client Feedback

**Overview:** One Hour AI Solutions collects feedback after each session with an AI engineer. Your task is to build a rule-based sentiment analysis system to analyze these feedback comments.

#### Requirements:

- 1. Use the provided "session\_feedback.csv" dataset containing client feedback after AI solution sessions
- 2. Implement a rule-based sentiment analyzer that:
  - Uses dictionaries of positive and negative words
  - o Handles negation (e.g., "not bad" should be interpreted as positive)
  - Assigns sentiment scores to feedback text
  - Classifies feedback as positive, negative, or neutral
- 3. Create a function to identify specific aspects being mentioned (e.g., engineer knowledge, solution quality, time efficiency)
- 4. Evaluate your analyzer by comparing its predictions to human-labeled sentiments provided in the dataset

## Expected Output:

- Overall sentiment distribution across all feedback
- Aspect-specific sentiment analysis (which aspects receive more positive/negative feedback)
- Evaluation metrics (accuracy, precision, recall) for your sentiment analyzer
- Discussion of limitations of the rule-based approach

**Dataset Description:** The "session\_feedback.csv" will contain columns: feedback\_id, feedback\_text, human\_labeled\_sentiment