Instruction Documentation for Data Scientist Assignment

Objective

The objective of this assignment is to extract textual data (only the main article from the web page) from articles available at given URLs and perform text analysis to compute specific linguistic and sentiment-based variables. The results are written into a structured CSV file for further analysis. Below, I explain my approach, the techniques used, and the step-by-step implementation process.

Approach and Solution

Step 1: Understanding the Variables to Compute

The task requires computing the following variable:

POSITIVE SCORE, NEGATIVE SCORE, POLARITY SCORE, SUBJECTIVITY SCORE, AVG SENTENCE LENGTH, PERCENTAGE OF COMPLEX WORDS, FOG INDEX, AVG NUMBER OF WORDS PER SENTENCE, COMPLEX WORD COUNT, WORD COUNT, SYLLABLE PER WORD, PERSONAL PRONOUNS, AVG WORD LENGTH

Step 2: Approach to Extract and Analyze Text

The workflow was broken down into several logical steps:

1. Input and Initialization:

Read the input Input.csv containing URLs and associated IDs using pandas.

Initialize the output CSV file Output Data Structure.csv with the specified headers(given variables which we studied above).

2. Web Scraping:

- For web scraping, I used the newspaper3k library for extracting textual content from the given URLs. This
 library efficiently downloads and parses web articles, which includes only their titles and main content
 which we need in this task.
- If newspaper3k failed due to timeouts, the requests-html and readability libraries served as fallback options.

(There are other more reliable and advance libraries like BeautifulSoup, Selenium, etc. but they fall behind in extracting cleaner text data and also requires more advance knowledge of html and the library itself, but in this case we only need the article data for that newspaper3k or requests-html and readability are very useful as they provide the required data much easier in only couple of lines of code)

3. Text Preprocessing:

- **Text Cleaning**: Removed HTML tags, URLs, punctuation, emojis, and stop words to focus on meaningful content using regex expressions and NLTK.
- Tokenization: Used NLTK's sent_tokenize and word_tokenize functions to split the text into sentences and words.

4. Linguistic Feature Computation:

- **Sentiment Scores**:Matched words against a predefined dictionary of positive and negative words. Calculated positive and negative scores by counting occurrences of respective words.
- **Complex Words**: Defined complex words as those with more than two syllables. Applied rules to handle edge cases like silent "e" and suffixes like "es" and "ed."
- **Readability and Structural Metrics**: Calculated metrics such as fog index, average sentence length, and percentage of complex words using formula-based approaches.
- **Pronoun Counting**: Used regex to identify and count first-person pronouns like *I, we, my, ours, us* while avoiding false positives (e.g., "US" as a country name).
- 5. Writing Output: Iteratively appended computed variables to the output CSV file.

How to Run the Script

Dependencies: Ensure the following Python libraries are installed:

- o requests-html
- readability-lxml
- o newspaper3k
- o nltk
- o pandas
- o beautifulsoup4
- o lxml

Install any missing dependencies using the install_package function provided in the script or with pip

Input Files:

- Place the Input.csv file in the working directory. The file should contain at least two columns: URL_ID and URL.
- Include directories for:
 - Stop Words: A folder named StopWords containing files with stop words.
 - Master Dictionary: A folder named MasterDictionary containing positive-words.txt and negative-words.txt.

Run the Script: Execute the script from the command line below-

```
python web_scrap.py
```

Output:

- The script generates two outputs:
 - o Individual .txt files containing cleaned article content, named by URL_ID.
 - A CSV file named Output Data Structure.csv containing the computed variables for each URL.

Key Implementation Highlights

• Error Handling:

 Incorporated fallback mechanisms for downloading articles and robust handling for timeouts or unavailable URLs.

Modular Design:

 Functions for tasks such as removing HTML tags, stop words, emojis, and punctuation ensure reusability and clarity.

• Optimized Processing:

 Used dictionary lookups for positive and negative word matching to improve performance on large texts.

Challenges and Solutions

- 1. Challenge: Some URLs timed out during scraping.
 - Solution: Increased timeout settings and used fallback libraries (readability and requests-html).
- 2. **Challenge**: Handling different types of punctuation and stop words across multiple files.
 - Solution: Consolidated all stop words into a single list dynamically during runtime.
- 3. **Challenge**: Calculating complex metrics like fog index and syllables accurately.
 - **Solution**: Wrote custom functions to handle specific linguistic rules.