

Instructions:

- The assignment is to be attempted in a group of max 2. You have to clearly mention the contribution of each member in README.pdf.
- Language allowed: Python
- You are allowed to use libraries such as NLTK for data preprocessing.
- For Plagiarism, institute policy will be followed.
- You need to submit README.pdf, Code files (it should include both .py files/.ipynb files), and Output.pdf.
- Mention methodology, preprocessing steps and assumptions you may have in README.pdf.
- Mention your sample outputs in the output.pdf.
- You are advised to prepare a well-documented code file.
- Submit code, readme and output files in ZIP format with the following name: A3_<roll_no1>_<roll_no2>.zip
- Use classroom discussion for any doubt.

Dataset: Download [here](#)

- Each sentence represents one sentence and is separated by a fullstop(.) at the end.
- Words in each sentence have the format **word / tag**.
Note: In some cases, the word itself contains a slash (/), e.g., **a/b / tag**. In such cases, treat a/b as a single word unit.
- Sentences are already tokenized.

Task: Part-of-Speech Tagging

1. Design and implement Hidden Markov Model (HMM) based Part-of-Speech (POS) tagger implementing Viterbi algorithm with the following assumptions. **[20]**
 - a. **Markov assumption length 1** - Probability of any state s_k depends on its previous state only, i.e., $P(s_k | s_{k-1})$
 - b. **[Optional] Markov assumption length 2** - Probability of any state s_k depends on its previous two states only, i.e., $P(s_k | s_{k-2} s_{k-1})$
Note: Only if your system permits. It is expected that it will increase the performance of the tagger.
2. Design and implement a MLP-based Part-of-Speech (POS) tagger implementing the following. **[15]**
 - a. Create Word2Vec & Glove representations of each word. Compare their performances.
 - b. Use the representation of each to perform the task of POS tagging using multi-layer perceptrons(MLP).
Note: You need to think of a suitable architecture.

Experiments:

1. Perform 3-fold cross validation on the dataset performing both Task 1 and 2 and show a comparative study (tabular) on following basis: **[5]**
 - a. Precision, recall and F1-score.
 - b. Tag-wise precision, recall and F1-score
 - c. Confusion matrix (Each element A_{ij} of matrix A denotes the number of times tag i classified as tag j).
 - d. Statistics of tag set.
2. Which word types are most frequently tagged incorrectly by HMM and MLP? Please provide a most likely justification for the same. **[10]**