Report – Assignment 1

Course: - Advanced NLP

Name – Dhaval Taunk | Roll No. - 2021701028 | Date-11-09-2021

1. Code and model details:-

- a. Code files:
 - i. SVD_co-occurence.ipynb
 - ii. word2vec-cbow.ipynb
- b. Models:
 - i. Models.zip
 - 1. tokens list.txt
 - 2. co_occurence.npz
 - 3. svd matrix.txt
 - 4. tokenizer.pickle
 - 5. weights.best.hdf5
 - 6. cbow_vecs.txt

2. Download models.zip:-

a. Link:- https://drive.google.com/file/d/1ICRS5sgw6xlf6AKIHk0o5RxD-qVoK gU/view?usp=sharing

3. Unique tokens used:-

- a. SVD_co-occurrence.ipynb:- 100679
- b. Word2vec-cbow.ipynb:- 100715

4. How to run the code:-

- a. First download the models.zip from the above given link. Extract it to the same folder where the code is present. The directory structure should look like below:
 - i. SVD_co-occurence.ipynb
 - ii. word2vec-cbow.ipynb
 - iii. models folder after extracting models.zip

b. SVD_co-occurrence.ipynb:-

- i. This notebook uses the following files:-
 - 1. tokens_list.txt :- contains unique tokens used
 - 2. **co_occurence.npz** :- contains co_occurrence matrix
 - 3. **svd_matrix.txt** :- contains trained word vectors
- ii. This file contains code to calculate word embeddings using Co-Occurrence Matrix by applying Singular Value Decomposition (SVD).
- iii. There are 2 sections in the file:-
 - 1. **Training:** Contains code to train the embeddings.

- 2. Inference:- Contains code to test the trained model.
- iv. To check the model performance, directly section 'Inference' can be run.
- v. The notebook already contains the output shown for question 3 and 4 I.e. "Display the top-10 word vectors for 5 different words" and "the top 10 closest words for the word 'camera'".
- vi. Output can be seen just by loading the script to jupyter notebooks.

c. Word2vec-cbow.ipynb:-

- i. This notebook uses the following files:-
 - 1. **tokenizer.pickle**:-contains tokenizer
 - 2. weights.best.hdf5:- contained trained choe model
 - 3. **cbow_vecs.txt** :- contains generated word vectors
- ii. This file contains code to calculate word embeddings using CBOW model.
- iii. There are 2 sections in the file:-
 - 1. **Training**:- Contains code to train the embeddings.
 - 2. Inference:- Contains code to test the trained model.
- iv. To check the model performance, directly section 'Inference' can be run.
- v. The notebook already contains the output shown for question 3 and 4 l.e. "Display the top-10 word vectors for 5 different words" and "the top 10 closest words for the word 'camera'".
- vi. Output can be seen just by loading the script to jupyter notebooks.

5. Results:-

a. The results for both the models (t-SNE plots) are saved in their respective notebooks. They can be visualized by loading the notebooks to jupyter notebooks.

6. Challenges faced:-

a. SVD_co-occurence.ipynb:-

- i. First tried making co-occurene matrix by using simple for loop over the entire dump. Google colab keeps on crashing.
- ii. Then tried removing the words with frequency less than 5 and applying the above method. Google colab keeps on crashing.
- iii. Then removed stop-words as well. Google colab crashed again.
- iv. Then tried scipy.sparse.csr matrix. But colab still keeps crashing.
- v. Tried some optimizations like removing punctuations and other. Still colab crahsed.
- vi. Then tried sklearn CountVectorizer to make word count matrix on words with frequency >=5. This time colab didn't crashed.
- vii. Then built co-occurrence matrix from count matrix. Saved it in a file. So that crashing colab doesn't lose the data.
- viii. Then tried making SVD matrix from the co-occurrence matrix using scipy.linalg.svd. Colab crashed.
- ix. Then tried making SVD using sklearn's TruncatedSVD. This time colab didn't get crashed.

x. The output embeddings are of shape (256, 1) which are saved in svd_mat.txt.

b. word2vec-cbow.ipynb:-

- i. First tried trained the entire dump by writing model structure in tensorflow (using model.fit()). Kaggle kernel crashed.
- ii. Then tried the words whose frequecy >=5. Kaggle kernel again crashed.
- iii. Then used generator's concept to iteratively generate the training data. So that kernel doesn't get crashed. It worked by using model.fit_generator(). Trained on the entire dataset. But the results were not good.
- iv. Then trained using on the words whose frequency >= 5. Model's performance improved than the previous one.
- v. The output embeddings are of shape (256, 1) which are saved in cbow_vecs.txt.