Hotel Review and Sentiment Analysis

CNN

Word2vec:

Pseudocode:

- 1. Connected the dataset to Colab using Google Drive.
- 2. Loaded the dataset, named reviews df, from text reviews.csv.
- 3. Performed text cleaning with the function def clean(text){}.
- 4. Imported TensorFlow and Keras libraries.
- 5. Created a dictionary with keys for text (reviews) and sentiments (labels).
- 6. Initialized a tokenizer.
- 7. Split the data into training and test sets with a 70:30 ratio.
- 8. Created a tokenizer using vocabsize for Word2Vec embedding.
- 9. Developed an embedding matrix using the tokenizer.
- 10. Trained a CNN model with a single layer.
- 11. Assessed the accuracy of the trained model on the testing dataset, stored as accuracy through model.evaluate().

Fasttext:

- 1. Established the connection between the dataset and Colab using Google Drive.
- 2. Loaded the dataset, reviews_df, from the file text_reviews.csv.
- 3. Applied text cleaning using the function def clean(text){}.
- 4. Imported the TensorFlow and Keras libraries.
- 5. Created a dictionary with keys 'text' for reviews and 'sentiments' for labels.
- 6. Initialized a tokenizer.
- 7. Divided the data into training and test sets with a 70:30 ratio.
- 8. Imported the wiki-news dataset for fasttext embedding.
- 9. Generated an embedding matrix using fasttext.
- 10. Trained a CNN model with a single layer.
- 11.Evaluated the accuracy of the trained model on the testing dataset, stored as accuracy using model.evaluate().

MCNN

Word2vec:

Pseudocode:

- 1. Connected the dataset to Colab via Google Drive.
- 2. Loaded the dataset, designated as reviews df, from text reviews.csv.
- 3. Implemented text cleaning through the function clean(text){}.
- 4. Imported TensorFlow and Keras libraries.
- 5. Established a dictionary with keys for text (reviews) and sentiments (labels).
- 6. Initialized a tokenizer.
- 7. Divided the data into training and test sets with a 70:30 ratio.
- 8. Created vocabulary size using the tokenizer.
- 9. Generated an embedding matrix using Word2Vec with the tokenizer.
- 10. Trained an MCNN model with multiple layers.
- 11. Evaluated the accuracy of the trained model on the testing dataset, stored as accuracy using model.evaluate().

Fasttext:

- 1. Connected the dataset to Colab through Google Drive.
- 2. Loaded the dataset, named reviews_df, from text_reviews.csv.
- 3. Implemented text cleaning through the function clean(text).
- 4. Brought in TensorFlow and Keras libraries.
- 5. Established a dictionary with keys for text (reviews) and sentiments (labels).
- 6. Initialized a tokenizer.
- 7. Divided the data into training and test sets with a 70:30 ratio.
- 8. Incorporated the wiki-news dataset for fasttext embedding.
- 9. Developed an embedding matrix utilizing fasttext.
- 10. Conducted training on an MCNN model with multiple layers.
- 11. Assessed the accuracy of the trained model on the testing dataset, stored as accuracy through model.evaluate().

BERT:

Pseudocode:

- 1. Established the connection between the dataset and Colab using Google Drive.
- 2. Installed BERT.
- 3. Loaded the dataset, reviews df, from the file text reviews.csv.
- 4. Implemented text cleaning using the function def clean(text){}.
- 5. Imported TensorFlow and Keras libraries.
- 6. Created a dictionary with keys 'text' for reviews and 'sentiments' for labels.
- 7. Initialized a BERT tokenizer.
- 8. Shuffled the dataset.
- 9. Created a batched Dataset for model training.
- 10. Divided the data into training and test sets with a 70:30 ratio.
- 11. Trained the MCNN model with multiple layers.
- 12. Evaluated the accuracy of the trained model on the testing dataset, stored as accuracy using model.evaluate().

RNN

Word2vec:

- 1. Connected the dataset to Colab via Google Drive.
- 2. Loaded the dataset, designated as reviews df, from text reviews.csv.
- 3. Implemented text cleaning through the function clean(text){}.
- 4. Imported TensorFlow and Keras libraries.
- 5. Established a dictionary with keys for text (reviews) and sentiments (labels).
- 6. Initialized a tokenizer.
- 7. Divided the data into training and test sets with a 70:30 ratio.
- 8. Created vocabulary size using the tokenizer.
- 9. Generated an embedding matrix using Word2Vec with the tokenizer.
- 10. Trained an RNN model with multiple layers.
- 11.Evaluated the accuracy of the trained model on the testing dataset, stored as accuracy using model.evaluate().

Fasttext:

Pseudocode:

- 1. Connected the dataset to Colab through Google Drive.
- 2. Loaded the dataset, named reviews_df, from text_reviews.csv.
- 3. Implemented text cleaning through the function clean(text).
- 4. Brought in TensorFlow and Keras libraries.
- 5. Established a dictionary with keys for text (reviews) and sentiments (labels).
- 6. Initialized a tokenizer.
- 7. Divided the data into training and test sets with a 70:30 ratio.
- 8. Incorporated the wiki-news dataset for fasttext embedding.
- 9. Developed an embedding matrix utilizing fasttext.
- 10. Conducted training on an RNN model with multiple layers.
- 11. Assessed the accuracy of the trained model on the testing dataset, stored as accuracy through model.evaluate().

HAN

Word2vec:

- 1. Connected the dataset to Colab via Google Drive.
- 2. Loaded the dataset, naming it reviews_df, from text_reviews.csv.
- 3. Implemented text cleaning with the function def clean(text){}.
- 4. Imported Theano and Keras libraries.
- 5. Created a dictionary with keys for text ("reviews") and sentiments ("labels").
- 6. Initialized a tokenizer using Word2Vec.
- 7. Split the data into training and test sets with a 70:30 ratio.
- 8. Created both a sentence tokenizer and a word tokenizer.
- 9. Generated an embedding matrix using both tokenizers.
- 10. Trained the Hierarchical Attention Network (HAN) model.
- 11.Evaluated the accuracy of the trained model on the testing dataset, stored as accuracy through model.evaluate().

Fasttext:

Pseudocode:

- 1. Connected the dataset to Colab using Google Drive.
- 2. Loaded the dataset, assigning it to reviews_df, from text_reviews.csv.
- 3. Implemented text cleaning using the function def clean(text){}.
- 4. Imported Theano and Keras libraries.
- 5. Created a dictionary with keys for text ("reviews") and sentiments ("labels").
- 6. Initialized a tokenizer.
- 7. Imported the wiki-news dataset for word embedding.
- 8. Split the data into training and test sets with a 70:30 ratio.
- 9. Created both a sentence tokenizer and a word tokenizer.
- 10.Generated an embedding matrix using both tokenizers.
- 11. Trained the Hierarchical Attention Network (HAN) model.
- 12. Evaluated the accuracy of the trained model on the testing dataset, stored as accuracy through model.evaluate().

Glove:

- 1. Mounted the dataset to Colab using Google Drive.
- 2. Read the dataset, assigning it to reviews_df, from the file text_reviews.csv.
- 3. Implemented text cleaning through the function def clean(text){}.
- 4. Imported Theano and Keras libraries.
- 5. Created a dictionary with keys 'text' for reviews and 'sentiments' for labels.
- 6. Established a tokenizer using Word2Vec.
- 7. Split the data into training and test sets with a 70:30 ratio.
- 8. Created both a sentence tokenizer and a word tokenizer.
- 9. Developed an embedding matrix using both tokenizers.
- 10. Trained the Hierarchical Attention Network (HAN) model.
- 11. Evaluated the accuracy of the trained model on the testing dataset, stored as accuracy through model.evaluate().

RMDL

BERT:

Pseudocode:

- 1. Established the connection between the dataset and Colab using Google Drive.
- 2. Installed BERT.
- 3. Loaded the dataset, reviews df, from the file text reviews.csv.
- 4. Implemented text cleaning using the function def clean(text){}.
- 5. Imported TensorFlow and Keras libraries.
- 6. Created a dictionary with keys 'text' for reviews and 'sentiments' for labels.
- 7. Initialized a BERT tokenizer.
- 8. Shuffled the dataset.
- 9. Created a batched Dataset for model training.
- 10. Divided the data into training and test sets with a 70:30 ratio.
- 11. Trained the RMDL model with multiple layers.
- 12. Evaluated the accuracy of the trained model on the testing dataset, stored as accuracy using model.evaluate().

Fasttext:

- 13. Connected the dataset to Colab using Google Drive.
- 14.Loaded the dataset, assigning it to reviews_df, from text_reviews.csv.
- 15.Implemented text cleaning using the function def clean(text){}.
- 16.Imported Theano and Keras libraries.
- 17. Created a dictionary with keys for text ("reviews") and sentiments ("labels").
- 18.Initialized a tokenizer.
- 19.Imported the wiki-news dataset for word embedding.
- 20. Split the data into training and test sets with a 70:30 ratio.
- 21. Created both a sentence tokenizer and a word tokenizer.
- 22. Generated an embedding matrix using both tokenizers.
- 23. Trained the RMDL model.
- 24. Evaluated the accuracy of the trained model on the testing dataset, stored as accuracy through model.evaluate().

Glove:

- 1. Established the dataset connection to Colab using Google Drive.
- 2. Installed RMDL.
- 3. Loaded the dataset, naming it reviews df, from text reviews.csv.
- 4. Implemented text cleaning with the function def clean(text){}.
- 5. Imported TensorFlow and Keras libraries.
- 6. Created a dictionary with keys for text ("reviews") and sentiments ("labels").
- 7. Imported the GloVe 6B dataset for word vectors.
- 8. Shuffled the dataset.
- 9. Created word vectors.
- 10.Generated an embedding matrix.
- 11. Split the data into training and test sets with a 70:30 ratio.
- 12. Trained the RMDL model with DNN, CNN, and RNN as three internal models.
- 13. Evaluated the accuracy of the trained model on the testing dataset, stored as accuracy using model.evaluate().