Assignment 4

Text and Sequence Data

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Objective

This is classification on the **IMDB** dataset for movie reviews, either positive negative. a binary The IMDB dataset consists of 50,000 reviews; however, only the top 10,000 words are considered in this dataset. The number of training samples ranges from 100 to 100,000, and 10,000 samples for validation. Before training, the data preparation includes an embedding laver and a pre-trained embedding model. A different approach is tried to see the performance.

Data Preprocessing

- Word Embeddings:

These are the methods of converting the text into word embeddings, representing words as fixed-size vectors. This transformation will be restricted to 10,000 samples. Another review conversion is into numerical sequences, where each number will correspond to a certain word. It is not possible to provide such sequences directly to the neural network.

- Tensor Construction:

Tensors are designed from these numerical sequences as the next step in the attempt to solve this. Each sample takes the form of a (samples, word indices) tensor. To maintain consistency in length across samples, methods like padding with placeholder words or numbers are applied.

Methodology

This study compares two approaches to generating word embeddings for the IMDB dataset:

- A. Custom-trained Embedding Layer
- B. Pretrained Embedding Layer (GloVe Model)

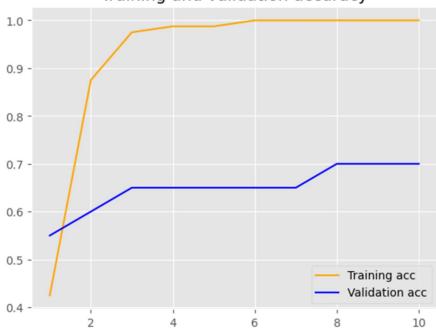
The GloVe model is a widely used pre-trained word embedding model trained on large text datasets. Performance testing was done for both approaches using various sizes of training samples: 100, 1,000, 5,000, and 10,000 samples.

- Custom-trained Embedding Layer: A custom embedding layer was trained on the IMDB dataset and tested its performance for various sample sizes.
- **Pre-trained Word Embedding Layer:** A setup similar to the present, with accuracy compared to a custom-trained layer under the same conditions, was done using the GloVe model.

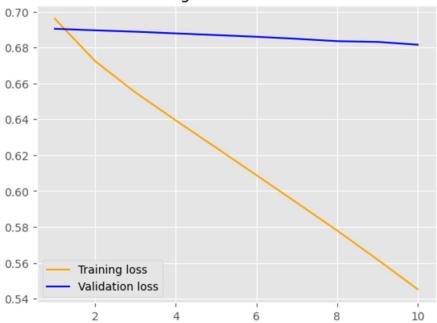
A. Custom-trained Embedding Layer

1) A custom embedding layer with a training sample size of 100.





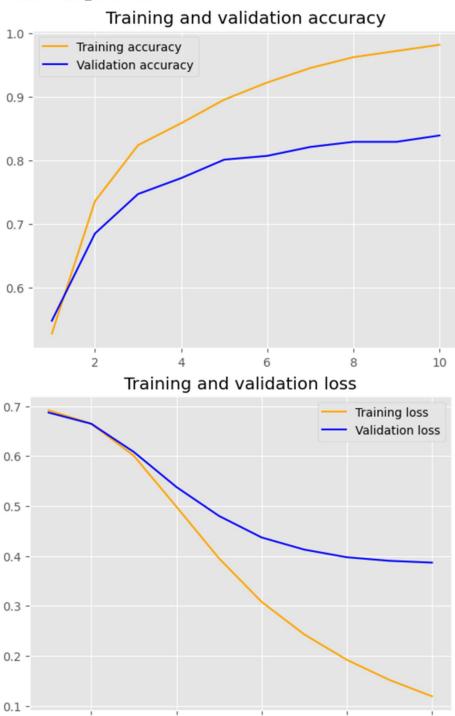
Training and validation loss



Test loss: 0.6938053369522095 Test accuracy: 0.5012400150299072

The model was trained using a custom-trained embedding layer with a sample size of 100, achieving a training accuracy of 100% and a test loss of 0.69.

2) A custom-trained embedding layer with a training sample size of 5000.



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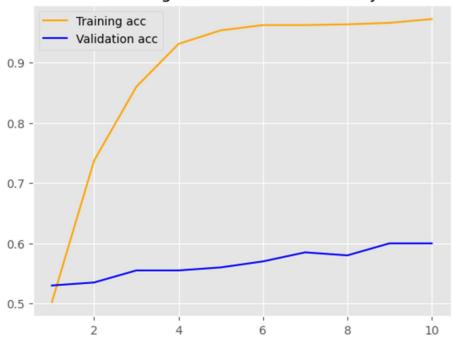
Test loss: 0.3757614493370056 Test accuracy: 0.8284000158309937

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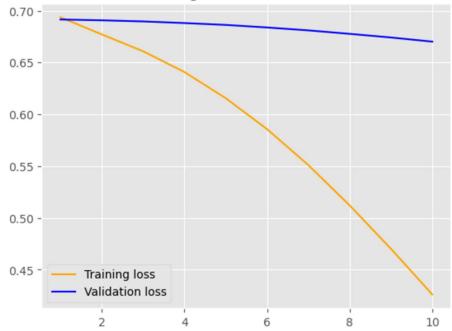
Using a sample size of 5,000, the model achieved a training accuracy of 98.15% and a test loss of 0.37.

3) A custom-trained embedding layer with a training sample size of 1000

Training and validation accuracy



Training and validation loss

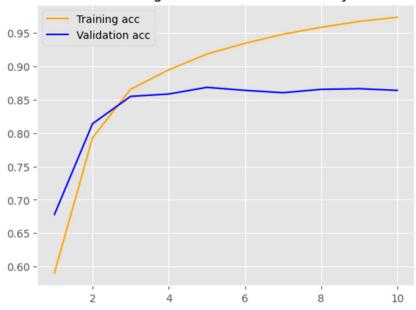


Test loss: 0.6815415620803833 Test accuracy: 0.564079999923706

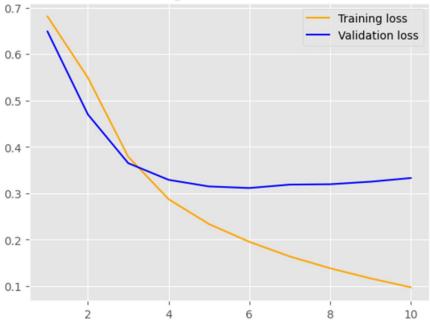
With a sample size of 1,000: Achieved 97.25% training accuracy and a test loss of 0.68.

4) A custom-trained embedding layer with a training sample size of 10,000.

Training and validation accuracy



Training and validation loss



Test loss: 0.336191326379776 Test accuracy: 0.8588799834251404

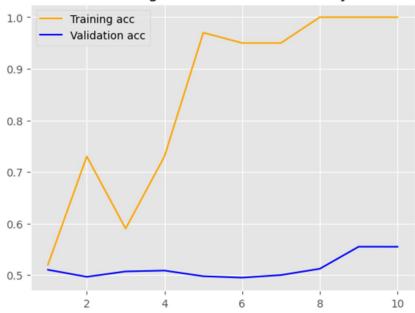
With a sample size of 10,000: Achieved 97.36% training accuracy and a test loss of 0.33.

Result: With training sample sizes of 100, 1,000, 5,000, and 10,000, the model's accuracy ranged from 97.25% to 100%, with the highest accuracy of 100% achieved using a sample size of 100.

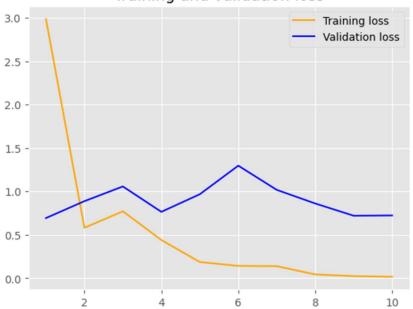
B) Pretrained Word Embedding Layer (GloVe)

1) A pre-trained word embedding layer with a training sample size of 100.

Training and validation accuracy



Training and validation loss



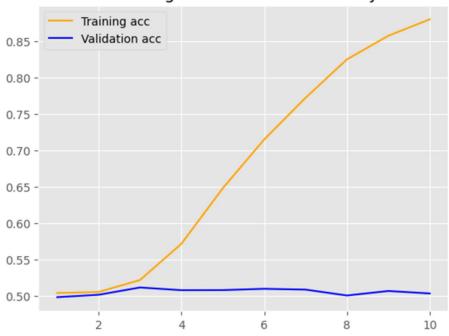
Test loss: 0.778832733631134

Test accuracy: 0.49803999066352844

The model was trained using pre-trained word embedding (GloVe) with a sample size of 100, achieving a training accuracy of 100% and a test loss of 0.77.

2) A pre-trained word embedding layer with a training sample size of 5000.

Training and validation accuracy



Training and validation loss

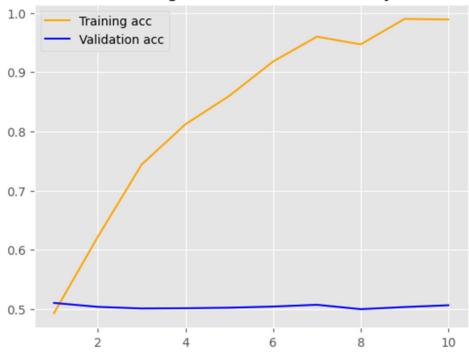


Test loss: 1.0668143033981323 Test accuracy: 0.49636000394821167

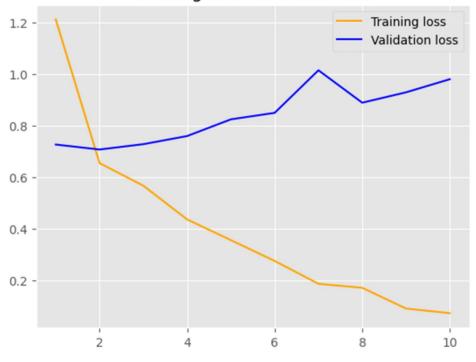
The model was trained using a sample size of 5,000, achieving a training accuracy of 87.96% and a test loss of 1.06.

3) A pre-trained word embedding layer with a training sample size of 1000.

Training and validation accuracy



Training and validation loss

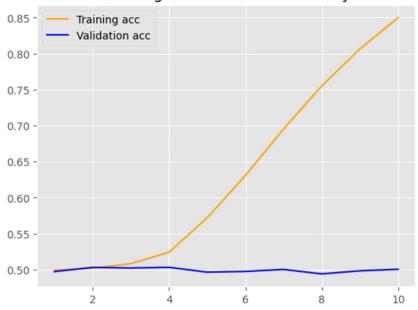


Test loss: 0.9887002110481262 Test accuracy: 0.495279997587204

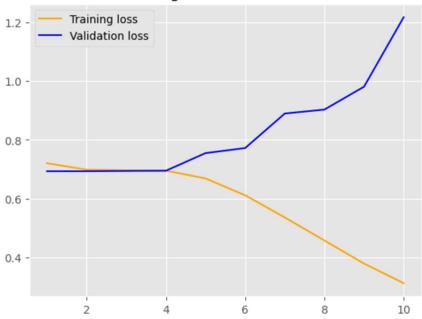
With a sample size of 1,000: Achieved 98.90% training accuracy and a test loss of 0.98.

4) A pre-trained word embedding layer with a training sample size of 10,000.

Training and validation accuracy



Training and validation loss



Test loss: 1.2223607301712036 Test accuracy: 0.4965200126171112

With a sample size of 10,000: Achieved 85% training accuracy and a test loss of 1.22.

Result: The model's accuracy ranged from 85% to 100%, depending on the sample size. While the highest accuracy of 100% was achieved with 100 samples, the model showed a tendency to overfit as the sample size increased, resulting in a decline in accuracy.

Conclusion

The analysis shows that custom-trained embedding layers are always more accurate and have lower test loss compared to pre-trained GloVe embeddings, especially as the training dataset size increases. Custom embeddings follow a smooth trend in performance improvement as dataset sizes grow larger, culminating in very high accuracy and low test loss. In contrast, pre-trained GloVe embeddings are more apt to be overfitted on small datasets and generalize poorly with larger datasets, leading to higher test losses and lower accuracy.

While custom embeddings work better on larger datasets, pre-trained embeddings are a valid choice when computation resources or training data are limited. However, the risk of overfitting increases significantly in such cases. As such, the choice of embedding methods presents a balancing act: while more flexible models perform better, they do so at great computational cost.