

## LAB 8 : Experiment Using LSTM

AIM: To build and implement a LSTM (Long short-Term-Memory) model for seq prediction.

Pseudo Code: . import required libraries

- Load and preprocess the sequential dataset
- Normalize the data
- Create input-output pairs
- Reshape X into samples
- Define LSTM model:
  - Initialize seq. model
  - Add LSTM layer with seq. units
  - Add dense output layer
- Compile the model with optimizer & loss
- Train the model using modifier.
- Evaluate model performance on test data
- Predict future or test samples
- Visualize predicted v/s actual output.

Observation:

- The training loss decreased gradually with each epoch, indicating that the model is learning the sequence pattern.
- LSTM performs ~~better~~ better than simple RNNs when dealing with long-term dependencies
- The predicted output closely follows the trend of actual data, demonstrating the model's ability to remember previous context.

→ However, training time is higher compared to standard RNN due to more complex computations

### Result:

→ The experiment was successfully carried out and LSTM model was implemented to learn and predict seq. pattern effectively

~~eff~~  
~~diag~~  
~~10/10~~

Accuracy

Group 1 LSTM

Group 2 LSTM

Group 3 LSTM

Group 4 LSTM

Comparison with LSTM

Group 1 LSTM

Group 2 LSTM

Group 3 LSTM

Group 4 LSTM

Group 1 LSTM

0.0

0.1

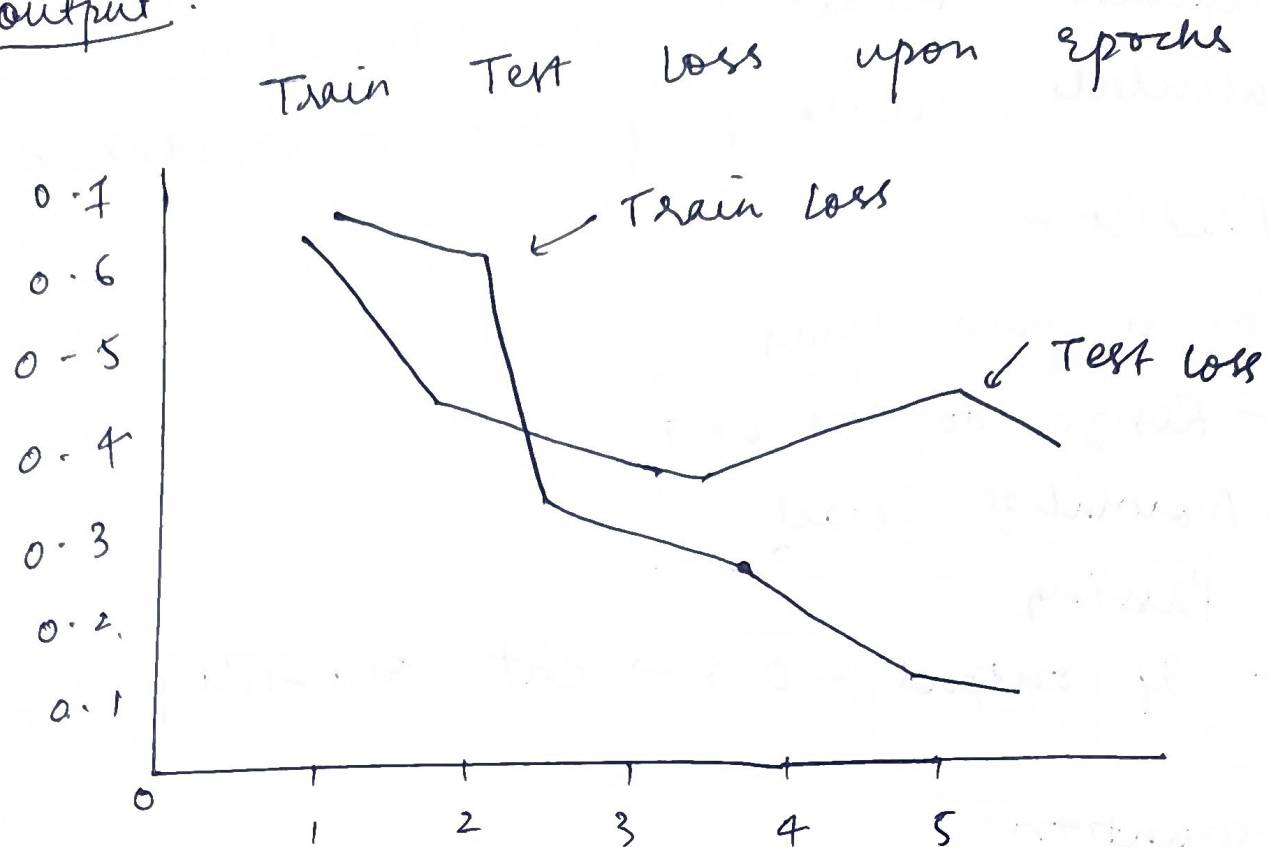
0.2

0.3

0.4

0.5

output:



Test accuracy = 89.77%

Predictions:

Input : The movie is good

O/P : Positive (probability : 0.9903)

~~I/P : The movie is bad~~

~~O/P : Negative (prob : 0.0013)~~



# LSTM architecture

