

Lab Exercise 1 - Train an RNN Model for Text Classification

Goal

Train your own deep-learning model (RNN) on a text dataset and test its accuracy.

What you will do

1. Load a dataset

Example: IMDB movie reviews (Positive/Negative).

2. Split the dataset

- o 80% Training
- o 20% Testing

3. Preprocess the text

- o Convert text to sequences of numbers
- o Pad all sequences to the same length

4. Build the RNN model

- o Embedding layer
- o LSTM or GRU layer
- o Final output layer

5. Train the model

- o Use the training set
- o Plot loss and accuracy

6. Test the model

- o Use the test set
- o Calculate accuracy

Lab Exercise 2 - Run LLAMA Locally Using Ollama & Compare With RNN

Goal

Run a pre-trained LLaMA model on your own laptop using **Ollama**, and compare its performance with the RNN you trained in Lab 1.

What you will do

Step 1 - Install Ollama

Download from: <https://ollama.com/download>

Run:

```
ollama run llama3
```

Step 2 - Test the LLaMA model on your dataset

For each test example:

Send the text to LLaMA with a prompt like:

Classify the sentiment of this review as Positive or Negative:

<review>

Answer with only one word.

Record:

- LLaMA's prediction
- Whether it matches the true label

Step 3 - Calculate accuracy

Just like the RNN model, compute:

$$\text{Accuracy} = \text{correct_predictions} / \text{total_test_samples}$$

Step 4 - Compare RNN vs LLaMA

Create a small table:

Model	Accuracy	Strengths	Weaknesses
RNN (your model)	?	simple, low cost	lower accuracy
LLaMA (pretrained)	?	high accuracy, understands text better	needs GPU/CPU

Lab Exercise 3 - Build a Simple RAG System (Retrieval + LLaMA)

Goal

Build a Question-Answer system that uses your documents + LLaMA.

What you will do

Step 1 - Prepare your documents

Collect 3–5 text files (notes, PDFs converted to text).

Break them into **small chunks** (e.g., 200–300 words).

Step 2 - Create embeddings

Use SentenceTransformers:

```
from sentence_transformers import SentenceTransformer
```

```
model = SentenceTransformer("all-MiniLM-L6-v2")
embeddings = model.encode(chunks, convert_to_numpy=True)
```

Step 3 - Build a simple similarity search

For a user question:

- Convert question into embedding
- Compare with chunk embeddings
- Retrieve top 3 relevant chunks

Step 4 - Build the RAG prompt

Send this to LLaMA:

Use only the following context to answer the question:

<context from top chunks>

Question: <user question>

Answer:

Step 5 - Test RAG vs No-RAG

Create 10 questions based on your documents.

Compare:

Method	Correct?	Notes
LLaMA alone	?	May hallucinate
LLaMA + RAG	?	More accurate