

Exp 8:

Experiment using LSTM.

Aim:

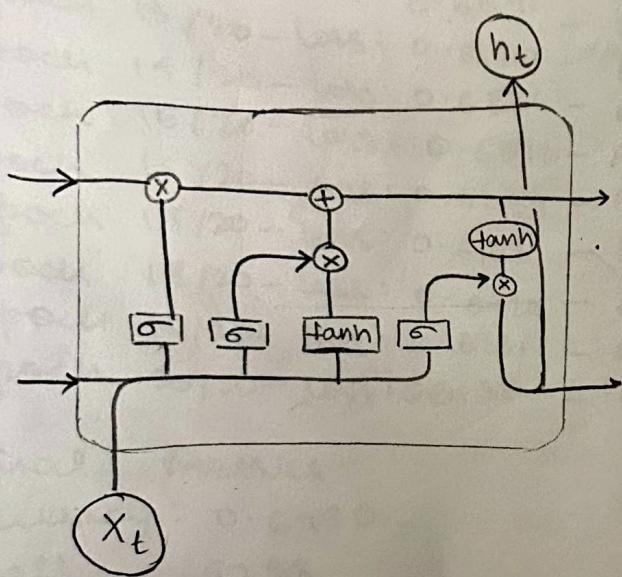
To implement and understand LSTM model in deep learning. (duped)

Pseudo code:

1. Import req. libraries = [0.01, 0.02] Numpy
2. Load and preprocess text data
 - Tokenize text [0.01, 0.02] Numpy
 - Convert to numerical sequences. [0.01, 0.02] Numpy
 - Pad sequences for uniform length. [0.01, 0.02] Numpy
 - split into train & test sets. [0.01, 0.02] Numpy
3. Define LSTM model:
 - Embedding layer
 - LSTM layers
 - Fully connected (linear) layers.
 - Sigmoid activation.
4. Define loss function & optimizer.
5. Train the model:
 - loop through epochs
 - Forward pass
 - compute loss
 - Backpropagation
 - update weights
6. Evaluate on test data.
7. Print accuracy & loss.

The only no minimizer is linear model: this
is due to the unique
that result

because of this property this
due to the no
differentiable



LSTM

diagram.

Justification:

LSTM (Long-term Term memory) networks are a special type of Recurrent Neural Network (RNN) capable of learning long-term dependencies.

- They are widely used in NLP tasks like,
- Sentiment Analysis
- Text generation
- Speech Recognition.

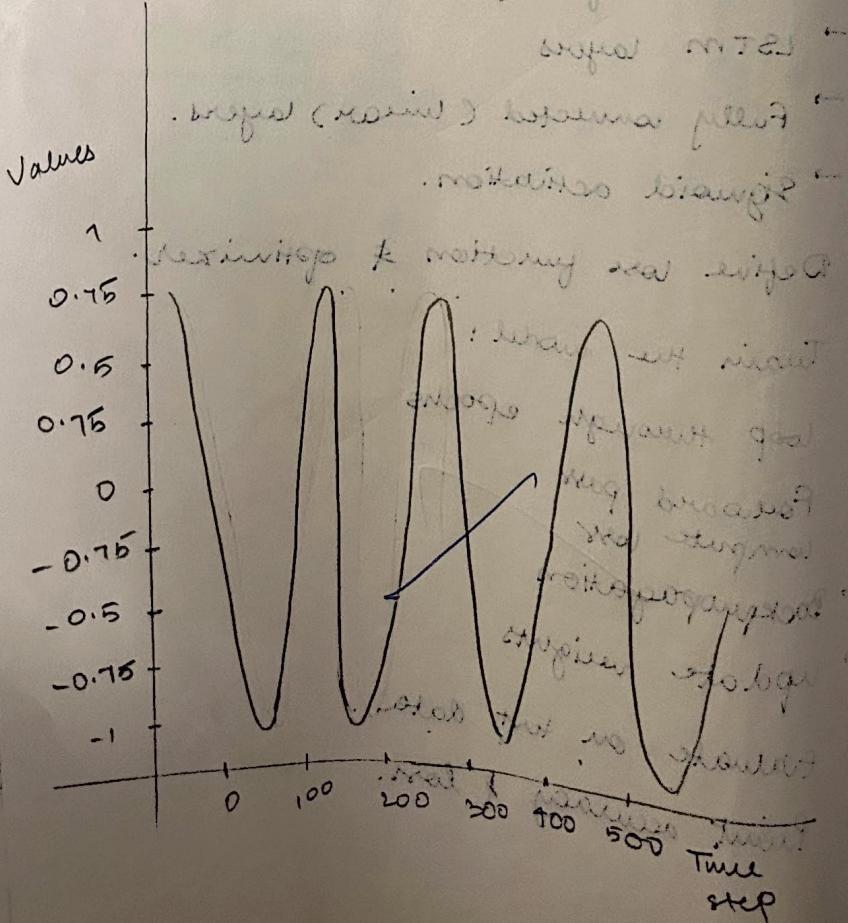
Result:

Program implemented successfully.

mixed noise from input

Output:

- below are the various loss functions of different epochs
- Epoch [10, 100], loss = 0.032186
 - Epoch [20, 100], loss = 0.017898
 - Epoch [30, 100], loss = 0.018660
 - Epoch [40, 100], loss = 0.016763
 - Epoch [50, 100], loss = 0.015243
 - Epoch [60, 100], loss = 0.014493
 - Epoch [70, 100], loss = 0.013756
 - Epoch [80, 100], loss = 0.016046
 - Epoch [90, 100], loss = 0.013142
 - Epoch [100, 100], loss = 0.012629



Untitled8.ipynb ⚡ Saving failed since 2:39 PM

File Edit View Insert Runtime Tools Help

Commands + Code + Text ▶ Run all

RAM Disk

```
[2] import torch
import torch.nn as nn
import torch.optim as optim
from sklearn.metrics import accuracy_score, f1_score, confusion_matrix
import matplotlib.pyplot as plt
import numpy as np

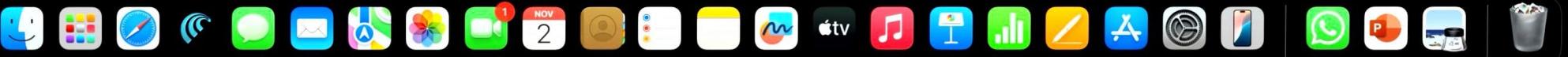
np.random.seed(42)
X = np.random.rand(1000, 10, 1)
y = (X.sum(axis=1) > 5).astype(int).flatten()
X_train, X_test = torch.tensor(X[:800], dtype=torch.float32), torch.tensor(X[800:], dtype=torch.float32)
y_train, y_test = torch.tensor(y[:800], dtype=torch.long), torch.tensor(y[800:], dtype=torch.long)

class LSTMModel(nn.Module):
    def __init__(self, input_size=1, hidden_size=32, num_layers=1, num_classes=2):
        super(LSTMModel, self).__init__()
        self.lstm = nn.LSTM(input_size, hidden_size, num_layers, batch_first=True)
        self.fc = nn.Linear(hidden_size, num_classes)
    def forward(self, x):
        out, (hn, cn) = self.lstm(x)
        out = self.fc(hn[-1])
        return out

model = LSTMModel()
criterion = nn.CrossEntropyLoss()
optimizer = optim.Adam(model.parameters(), lr=0.001)
num_epochs = 6
train_losses, test_accuracies = [], []

for epoch in range(num_epochs):
    # Forward
```

Variables Terminal ✨ 2:41 PM Python 3





```
[2]  optimizer = optim.Adam(model.parameters(), lr=0.001)
    num_epochs = 6
    train_losses, test_accuracies = [], []

    for epoch in range(num_epochs):
        # Forward
        outputs = model(X_train)
        loss = criterion(outputs, y_train)

        # Backward
        optimizer.zero_grad()
        loss.backward()
        optimizer.step()

        # Evaluate on test set
        with torch.no_grad():
            test_outputs = model(X_test)
            _, predicted = torch.max(test_outputs, 1)
            acc = accuracy_score(y_test, predicted)

            train_losses.append(loss.item())
            test_accuracies.append(acc)
            print(f"Epoch [{epoch+1}/{num_epochs}] Loss: {loss.item():.4f}, Test Acc: {acc:.4f}")

        with torch.no_grad():
            y_pred = torch.argmax(model(X_test), dim=1)
            acc = accuracy_score(y_test, y_pred)
            f1 = f1_score(y_test, y_pred)
            cm = confusion_matrix(y_test, y_pred)

            print("\n✓ Final Metrics:")
            print(f"Accuracy: {acc:.4f}")
            print(f"F1 Score: {f1:.4f}")


```





Commands + Code + Text ▶ Run all

RAM Disk

```
[2]   print("\n✓ Final Metrics:")
      print(f"Accuracy: {acc:.4f}")
      print(f"F1 Score: {f1:.4f}")
      print("Confusion Matrix:\n", cm)
      plt.figure(figsize=(10,4))

      plt.subplot(1,2,1)
      plt.plot(train_losses, label="Train Loss")
      plt.title("Training Loss")
      plt.xlabel("Epochs")
      plt.ylabel("Loss")
      plt.legend()

      plt.subplot(1,2,2)
      plt.plot(test_accuracies, label="Test Accuracy", color='orange')
      plt.title("Test Accuracy")
      plt.xlabel("Epochs")
      plt.ylabel("Accuracy")
      plt.legend()

      plt.tight_layout()
      plt.show()
```

```
Epoch [1/6] Loss: 0.6975, Test Acc: 0.4300
Epoch [2/6] Loss: 0.6967, Test Acc: 0.4300
Epoch [3/6] Loss: 0.6959, Test Acc: 0.4300
Epoch [4/6] Loss: 0.6952, Test Acc: 0.4300
Epoch [5/6] Loss: 0.6945, Test Acc: 0.4300
Epoch [6/6] Loss: 0.6939, Test Acc: 0.4300
```

✓ Final Metrics:
Accuracy: 0.4300

{ } Variables Terminal



✓ 2:41 PM





Commands + Code + Text ▶ Run all



RAM Disk

```
Epoch [1/6] Loss: 0.6975, Test Acc: 0.4300
Epoch [2/6] Loss: 0.6967, Test Acc: 0.4300
Epoch [3/6] Loss: 0.6959, Test Acc: 0.4300
Epoch [4/6] Loss: 0.6952, Test Acc: 0.4300
Epoch [5/6] Loss: 0.6945, Test Acc: 0.4300
Epoch [6/6] Loss: 0.6939, Test Acc: 0.4300
```

 Final Metrics:

Accuracy: 0.4300

F1 Score: 0.6014

Confusion Matrix:

```
[[ 0 114]
 [ 0 86]]
```

