

Ex: 9.

Build a Recurrent Neural Network (RNN)

30.09.25

Aim:

To build a Recurrent Neural Network (RNN).

Pseudo code:

1. Initialize RNN parameters:

- Input size

- Hidden state size

- Output size (no. of classes)

2. Define RNN forward pass:

for each time step t in sequence length:

- update hidden state using input at t & previous hidden state, output pred. based on final hidden state.

3. Define loss function (eg: cross entropy loss)

4. Train RNN:

for each epoch:

for each batch:

- forward pass: compute outputs.

- compute loss

- Backpropagate loss

- update model parameters.

5. Evaluate the model on test data:

- Calc. accuracy, recall, F1 score, etc.

6. Plot training loss & accuracy over epochs.

Justification:

RNN: Chosen because it can process sequential data by maintaining hidden state.

Result:

RNN successfully built & trained on sequential data.

Successfully Implemented.

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Observation:

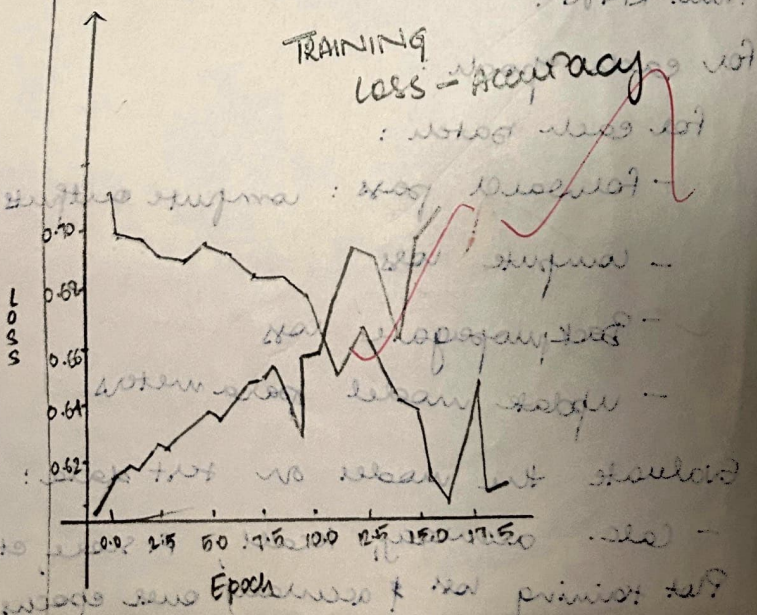
epoch 1/20 - loss : 0.7046 - Accuracy: 0.480
 epoch 2/20 - loss : 0.6950 - Accuracy: 0.535
 epoch 3/20 - loss : 0.6938 - Accuracy: 0.549
 epoch 4/20 - loss : 0.6888 - Accuracy: 0.549
 epoch 5/20 - loss : 0.6869 - Accuracy: 0.561
 epoch 6/20 - loss : 0.6895 - Accuracy: 0.561
 epoch 7/20 - loss : 0.6867 - Accuracy: 0.561
 epoch 8/20 - loss : 0.6821 - Accuracy: 0.561
 epoch 9/20 - loss : 0.6820 - Accuracy: 0.561
 epoch 10/20 - loss : 0.6779 - Accuracy: 0.561
 epoch 11/20 - loss : 0.6675 - Accuracy: 0.561
 epoch 12/20 - loss : 0.6591 - Accuracy: 0.561
 epoch 13/20 - loss : 0.6710 - Accuracy: 0.561
 epoch 14/20 - loss : 0.6536 - Accuracy: 0.561
 epoch 15/20 - loss : 0.6512 - Accuracy: 0.561
 epoch 16/20 - loss : 0.6239 - Accuracy: 0.561
 epoch 17/20 - loss : 0.6163 - Accuracy: 0.561
 epoch 18/20 - loss : 0.6120 - Accuracy: 0.561
 epoch 19/20 - loss : 0.6131 - Accuracy: 0.561
 epoch 20/20 - loss : 0.6138 - Accuracy: 0.561

Final Metrics:

Accuracy: 0.6790

Recall: 0.6738

F1 Score: 0.6825



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```
%matplotlib inline
import torch
import torch.nn as nn
import torch.optim as optim
from sklearn.metrics import f1_score, accuracy_score
import matplotlib.pyplot as plt
import numpy as np

def generate_data(n_samples=1000, seq_len=10):
    X = np.random.rand(n_samples, seq_len, 1)
    y = (X.sum(axis=1) > 5).astype(int)
    return torch.tensor(X, dtype=torch.float32), torch.tensor(y, dtype=torch.long).squeeze()

X, y = generate_data()
train_size = int(0.8 * len(X))
X_train, X_val = X[:train_size], X[train_size:]
y_train, y_val = y[:train_size], y[train_size:]

class SimpleRNN(nn.Module):
    def __init__(self, input_size, hidden_size, num_classes):
        super(SimpleRNN, self).__init__()
        self.rnn = nn.RNN(input_size, hidden_size, batch_first=True)
        self.fc = nn.Linear(hidden_size, num_classes)
    def forward(self, x):
        out, _ = self.rnn(x)
        out = out[:, -1, :]
        out = self.fc(out)
        return out

model = SimpleRNN(input_size=1, hidden_size=16, num_classes=2)
```



[17]

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```
self.fc = nn.Linear(hidden_size, num_classes)
def forward(self, x):
    out, _ = self.rnn(x)
    out = out[:, -1, :]
    out = self.fc(out)
    return out
```

```
model = SimpleRNN(input_size=1, hidden_size=16, num_classes=2)
criterion = nn.CrossEntropyLoss()
optimizer = optim.Adam(model.parameters(), lr=0.01)
```

```
epochs = 50
train_losses = []
val_losses = []
```

```
for epoch in range(epochs):
    model.train()
    optimizer.zero_grad()
    outputs = model(X_train)
    loss = criterion(outputs, y_train)
    loss.backward()
    optimizer.step()
    train_losses.append(loss.item())
    model.eval()
    with torch.no_grad():
        val_outputs = model(X_val)
        val_loss = criterion(val_outputs, y_val)
        val_losses.append(val_loss.item())
    if (epoch+1) % 10 == 0:
        print(f"Epoch [{epoch+1}/{epochs}] | Train Loss: {loss.item():.4f} | Val Loss: {val_loss.item():.4f}")
```

```
model.eval()
```



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Untitled8.ipynb - Colab

Asmi0612/Deeplearning_RA2311047010012: Lab note obs

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[17]

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```
model.eval()
with torch.no_grad():
    val_outputs = model(X_val)
    val_loss = criterion(val_outputs, y_val)
    val_losses.append(val_loss.item())
    if (epoch+1) % 10 == 0:
        print(f"Epoch [{epoch+1}/{epochs}] | Train Loss: {loss.item():.4f} | Val Loss: {val_loss.item():.4f}")

model.eval()
with torch.no_grad():
    y_pred = model(X_val).argmax(dim=1).numpy()
    y_true = y_val.numpy()

acc = accuracy_score(y_true, y_pred)
f1 = f1_score(y_true, y_pred)

print("\nFinal Evaluation:")
print(f"Accuracy: {acc*100:.2f}%")
print(f"F1 Score: {f1:.4f}")

plt.figure(figsize=(8,5))
plt.plot(train_losses, label='Training Loss')
plt.plot(val_losses, label='Validation Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()
plt.title('Training vs Validation Loss')
plt.show()
```

Epoch [10/50] | Train Loss: 0.6448 | Val Loss: 0.6290

Epoch [20/50] | Train Loss: 0.4371 | Val Loss: 0.3371

Epoch [30/50] | Train Loss: 0.3333 | Val Loss: 0.3313

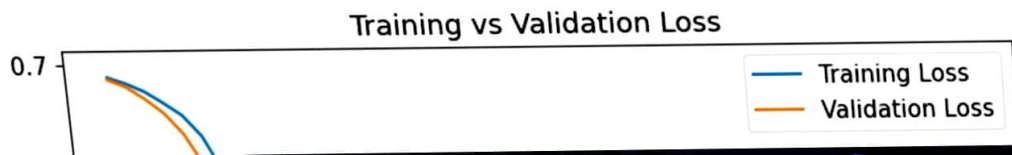
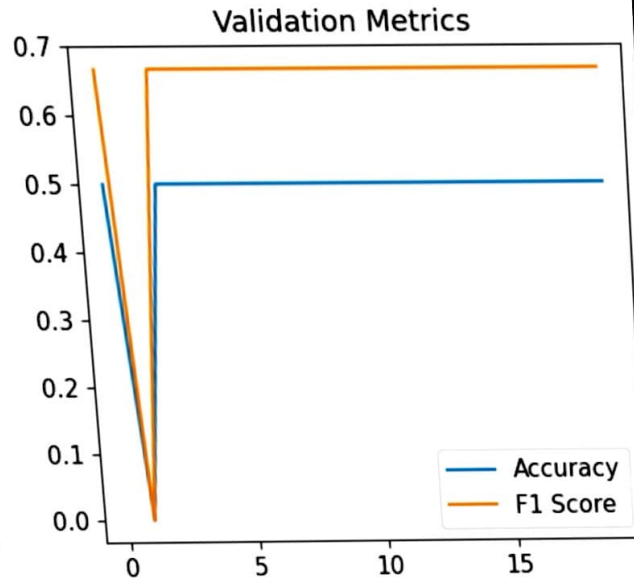
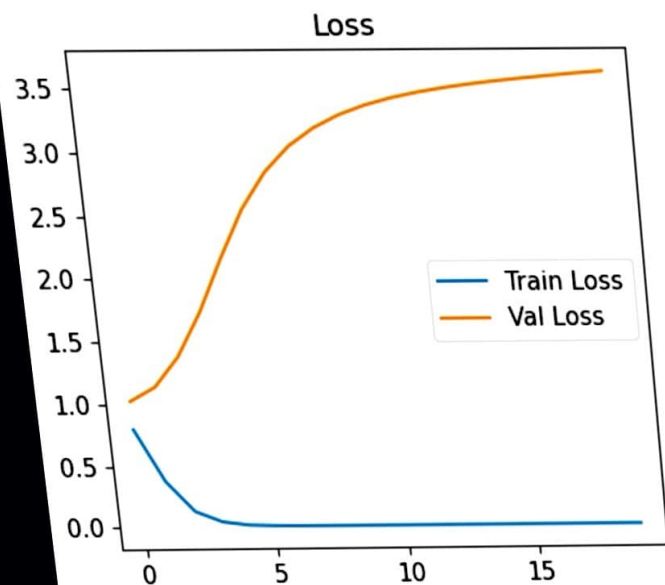
Variables Terminal

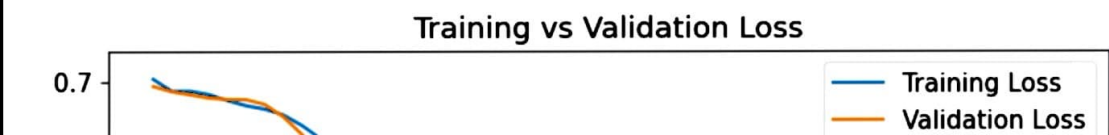
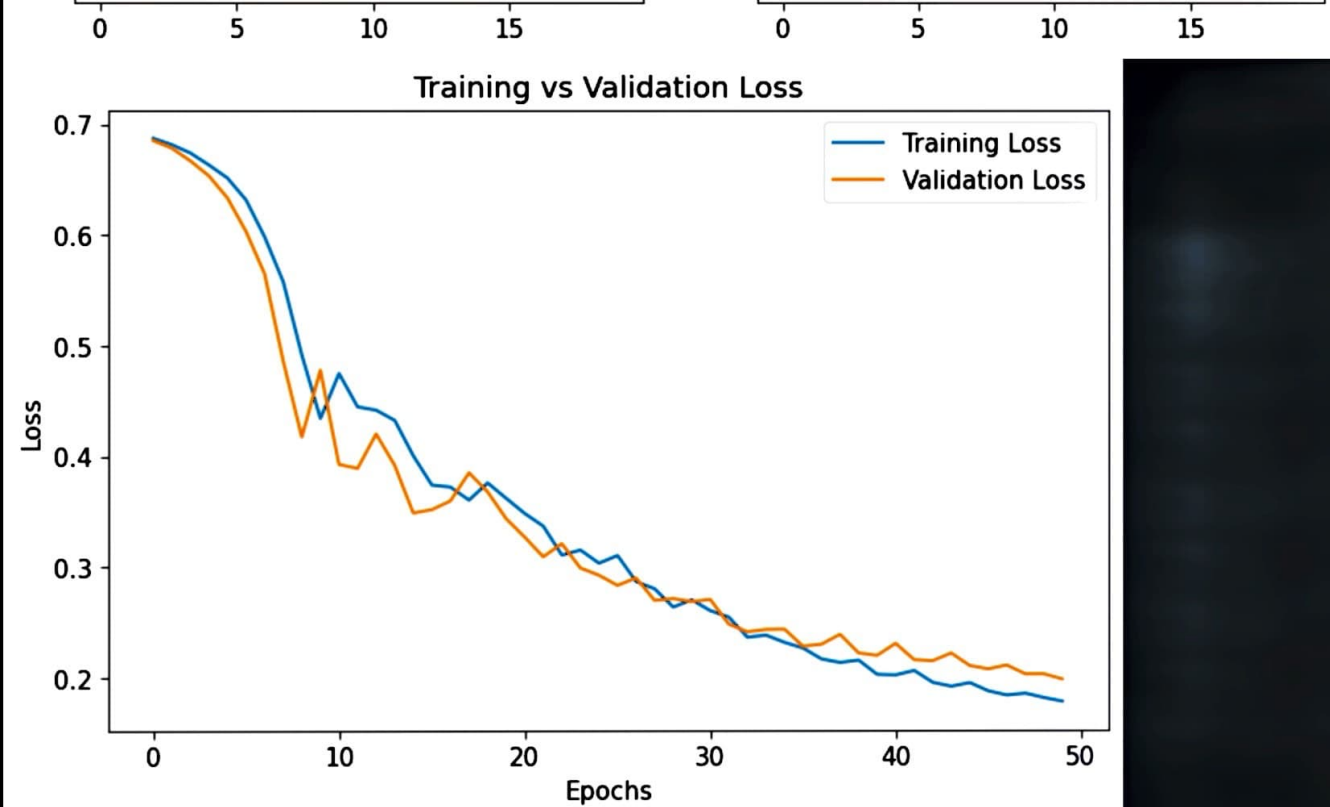
```
[17] plt.plot(val_losses, label='validation Loss')  
plt.xlabel('Epochs')  
plt.ylabel('Loss')  
plt.legend()  
plt.title('Training vs Validation Loss')  
plt.show()
```

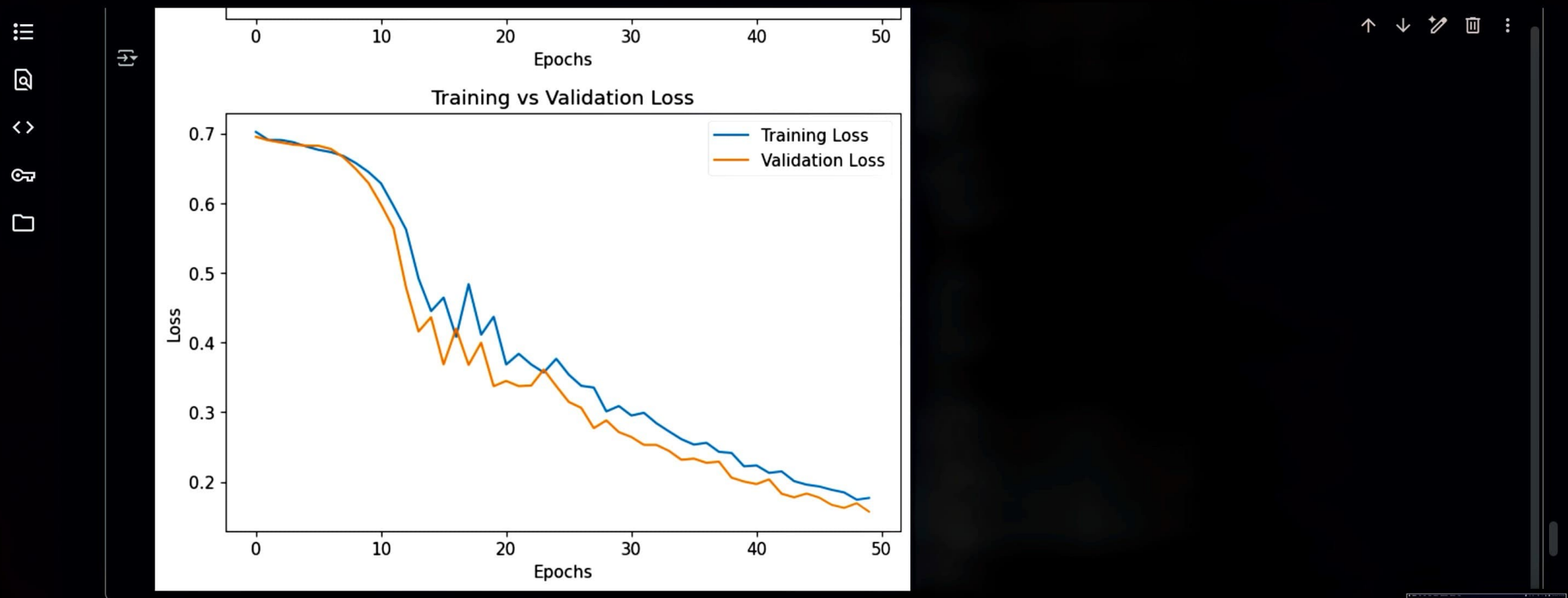
```
Epoch [10/50] | Train Loss: 0.6448 | Val Loss: 0.6290  
Epoch [20/50] | Train Loss: 0.4371 | Val Loss: 0.3371  
Epoch [30/50] | Train Loss: 0.3088 | Val Loss: 0.2713  
Epoch [40/50] | Train Loss: 0.2222 | Val Loss: 0.2002  
Epoch [50/50] | Train Loss: 0.1768 | Val Loss: 0.1569
```

Final Evaluation:
Accuracy: 94.00%
F1 Score: 0.9375









```
[11] ✓ 0s image_files = ['/path/to/your/image1.jpg', '/other/path/image2.jpg']
```