

## Build a Recurrent Neural Network (RNN)

30.09.25 O : features - OPa.O : stat - of / & stage  
OPa.O : features - P.Pa.O : stat - of / & stage  
Aim: To build a Recurrent Neural Network

08 08 (KNN).ma - page 0 : 08/08/2018

OP82.0 : *fruticosa* - OP82.0 : *solan* - OP12.0 *Wap*

moles  $\text{NaOH} = \frac{12.8 \text{ g}}{40 \text{ g/mol}} = 0.32 \text{ mol}$

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Initializ. RNN parameters. -  $\alpha_1$  to  $\alpha_n$

Quesd. o : frequent -  $\frac{1}{10} \cdot 100 = 10\%$  // need

APPENDIX B: INPUT SIZE - APPENDIX C: INPUT SIZE - APPENDIX D: INPUT SIZE

- Hidden state size:  $c = 200 - 400$  is enough

Output size ( $\text{cm}^3$ ) =  $0.001 \times 0.001 \times 0.001 = 1 \text{ cm}^3$

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08.22 0 : ~~classmate~~ - pg 23. 0 : 211 - 051 21 above

2. Define RNN forward pass

for each time step -  $\Delta t$  -  $\Delta x$  -  $\Delta y$

update hidden state using:

\* previous hidden state, output predicted based on final hidden state.

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

#### 4. Train RNN:

for each epoch :

for each batch:

- forward pass: compute outputs.

~~- compute less~~

- 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.

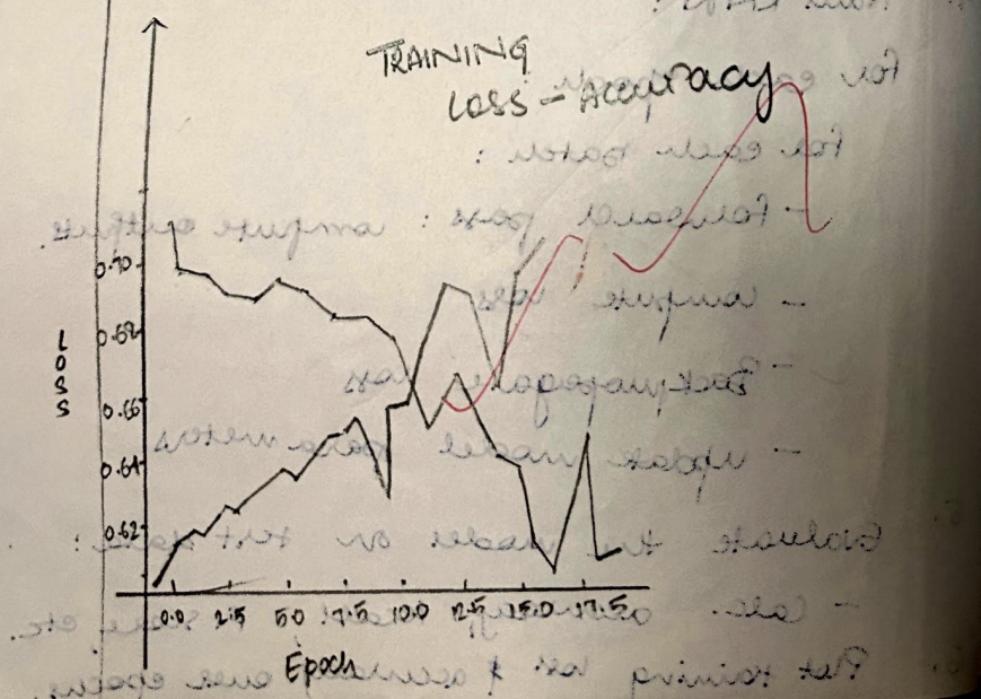


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Observation:

epoch 1 / 20 -	loss : 0.7046	Accuracy : 0.4801
epoch 2 / 20 -	loss : 0.6950	Accuracy : 0.5385
epoch 3 / 20 -	loss : 0.6938	Accuracy : 0.5490
epoch 4 / 20 -	loss : 0.6888	Accuracy : 0.5945
epoch 5 / 20 -	loss : 0.6869	Accuracy : 0.5611
epoch 6 / 20 -	loss : 0.6895	Accuracy : 0.5611
epoch 7 / 20 -	loss : 0.6867	Accuracy : 0.5550
epoch 8 / 20 -	loss : 0.6821	Accuracy : 0.5668
epoch 9 / 20 -	loss : 0.6820	Accuracy : 0.5668
epoch 10 / 20 -	loss : 0.6719	Accuracy : 0.5740
epoch 11 / 20 -	loss : 0.6675	Accuracy : 0.5668
epoch 12 / 20 -	loss : 0.6591	Accuracy : 0.5999
epoch 13 / 20 -	loss : 0.6710	Accuracy : 0.6170
epoch 14 / 20 -	loss : 0.6536	Accuracy : 0.6888
epoch 15 / 20 -	loss : 0.6512	Accuracy : 0.6230
epoch 16 / 20 -	loss : 0.6239	Accuracy : 0.6222
epoch 17 / 20 -	loss : 0.6163	Accuracy : 0.6668
epoch 18 / 20 -	loss : 0.6120	Accuracy : 0.6668
epoch 19 / 20 -	loss : 0.6131	Accuracy : 0.6388
epoch 20 / 20 -	loss : 0.6138	Accuracy : 0.6667

Final practice:  
Accuracy: 0.6790.8 recall using no  
recall: 0.6738  
(acc part 2) F1 Score: 0.6825



[17]

```
%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] ✓ 1s
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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Commands

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```
[17] 1s
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: 0.2620 | Val Loss: 0.2712
```

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Terminal

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

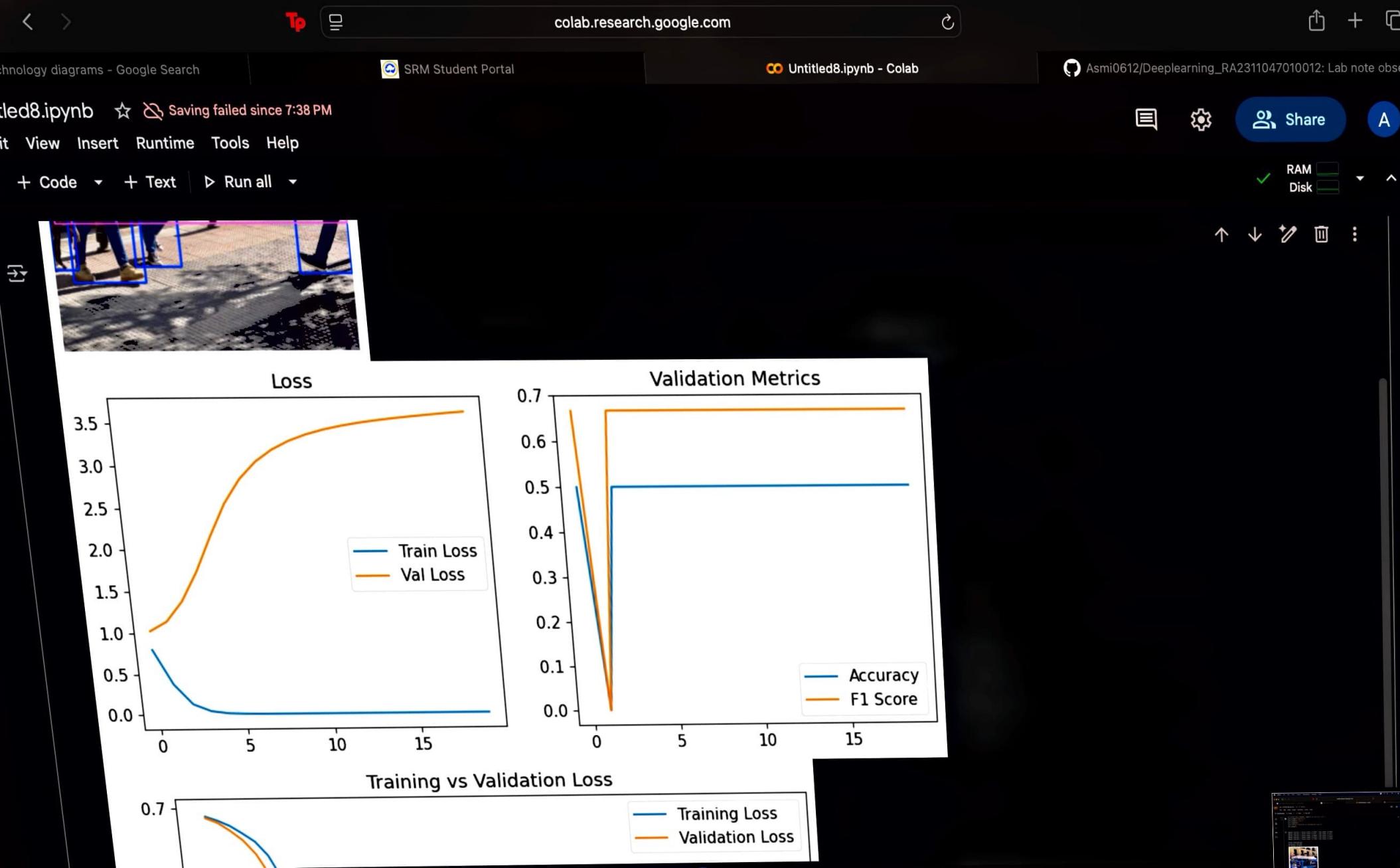
```
→ Epoch [10/50] | Train Loss: 0.6448 | Val Loss: 0.6290
Epoch [20/50] | Train Loss: 0.4371 | Val Loss: 0.3371
Epoch [30/50] | Train Loss: 0.3088 | Val Loss: 0.2713
Epoch [40/50] | Train Loss: 0.2222 | Val Loss: 0.2002
Epoch [50/50] | Train Loss: 0.1768 | Val Loss: 0.1569
```

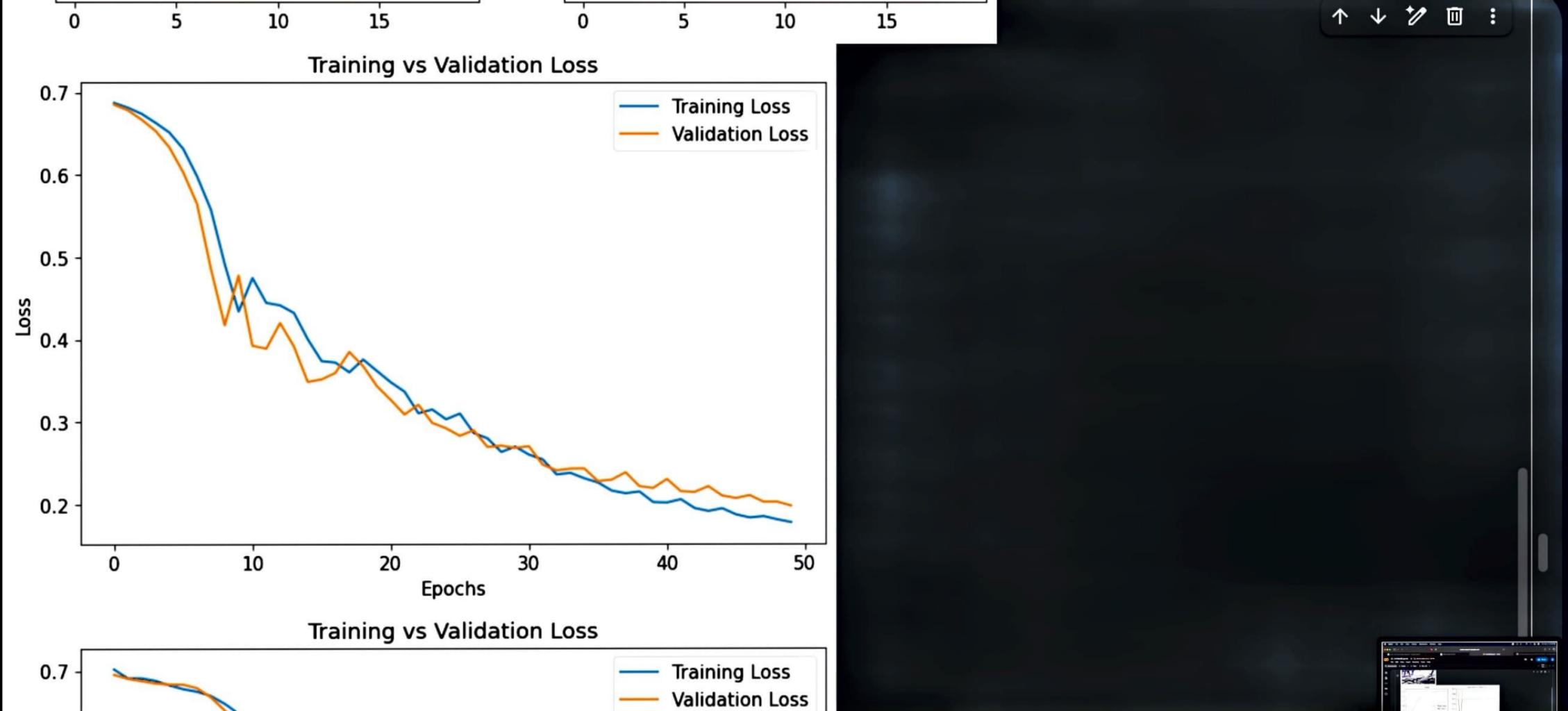
Final Evaluation:

Accuracy: 94.00%

F1 Score: 0.9375







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Training vs Validation Loss

Epochs

Loss

Training Loss Validation Loss

Epochs

[11] 0s

image\_files = ['/path/to/your/image1.jpg', '/other/path/image2.jpg']

Variables Terminal ✓ 8:23

Epochs	Training Loss	Validation Loss
0	0.70	0.70
10	0.58	0.55
20	0.38	0.35
30	0.30	0.28
40	0.22	0.20
50	0.18	0.17

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