

# Neural Networks & Deep Learning Assignment-6

Harshavardhan Reddy Pattepur  
700754833

## Repository Link :

<https://github.com/harshavardhanreddy27/Assignment---6>

## Video Link:

[https://drive.google.com/file/d/1zjrASy8DHu\\_a4HS3j16pS7lXlxCY9ds8/view?usp=share\\_link](https://drive.google.com/file/d/1zjrASy8DHu_a4HS3j16pS7lXlxCY9ds8/view?usp=share_link)

## Code Screenshots:

### Question 1b:

The screenshot shows a Google Colab notebook titled "Assignment - 6". The code cell contains Python code for loading a dataset, performing one-hot encoding on the 'diagnosis' column, and printing the first few rows of the DataFrame. The output shows the breast cancer dataset with columns: id, diagnosis, radius\_mean, texture\_mean, perimeter\_mean, area\_mean, smoothness\_mean, compactness\_mean, concavity\_mean, concave points\_mean, radius\_worst, texture\_worst, perimeter\_worst, area\_worst.

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense

df = pd.read_csv('/content/breastcancer.csv')

df = df.drop(columns=['Unnamed: 32'], errors='ignore')

df['diagnosis'] = df['diagnosis'].map({'M': 1, 'B': 0})

print(df.head())

   id  diagnosis  radius_mean  texture_mean  perimeter_mean  area_mean \
0   842302         1      17.99      10.38      122.80     1001.0
1   842517         1      20.57      17.77      132.90     1326.0
2   84300903        1      19.69      21.25      130.00     1203.0
3   84348301        1      11.42      20.38      77.58      386.1
4   84358402         1      20.29      14.34      135.10     1297.0

   smoothness_mean  compactness_mean  concavity_mean  concave points_mean \
0      0.11840       0.27760       0.3001       0.14710
1      0.08474       0.07864       0.0869       0.07017
2      0.10960       0.15990       0.1974       0.12790
3      0.14250       0.28390       0.2414       0.10520
4      0.10030       0.13280       0.1980       0.10430

   ...  radius_worst  texture_worst  perimeter_worst  area_worst \
0   ...      25.38      17.33      184.60     2019.0
1   ...      24.99      23.41      158.80     1956.0
2   ...      23.57      25.53      152.50     1709.0
3   ...      14.91      26.50      98.87      567.7
```

## Question 1a:

Assignment - 6

File Edit View Insert Runtime Tools Help All changes saved

Comment Share H

+ Code + Text

1(a) Added more density layers to the existing code and checked the accuracy level

```
{x} 2s X = df.drop(columns=['id', 'diagnosis'])
y = df['diagnosis']

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

model_task1 = Sequential()
model_task1.add(Dense(32, activation='relu', input_shape=(X_train.shape[1],)))
model_task1.add(Dense(64, activation='relu'))
model_task1.add(Dense(128, activation='relu'))
model_task1.add(Dense(1, activation='sigmoid'))

model_task1.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])

model_task1.fit(X_train, y_train, epochs=10, batch_size=32, validation_data=(X_test, y_test))

accuracy_task1 = model_task1.evaluate(X_test, y_test)[1]
print(f'Task 1 - Model Accuracy: {accuracy_task1}')

Epoch 1/10
15/15 [=====] - 1s 17ms/step - loss: 4.0996 - accuracy: 0.6440 - val_loss: 2.7071 - val_accuracy: 0.3860
Epoch 2/10
15/15 [=====] - 0s 4ms/step - loss: 1.5084 - accuracy: 0.6659 - val_loss: 0.6956 - val_accuracy: 0.8772
Epoch 3/10
15/15 [=====] - 0s 5ms/step - loss: 0.7544 - accuracy: 0.8396 - val_loss: 0.7175 - val_accuracy: 0.7456
Epoch 4/10
15/15 [=====] - 0s 5ms/step - loss: 0.5125 - accuracy: 0.8725 - val_loss: 0.2191 - val_accuracy: 0.9123
Epoch 5/10
15/15 [=====] - 0s 4ms/step - loss: 0.4241 - accuracy: 0.8835 - val_loss: 0.2806 - val_accuracy: 0.9035
Epoch 6/10
15/15 [=====] - 0s 5ms/step - loss: 0.6570 - accuracy: 0.7912 - val_loss: 1.9739 - val_accuracy: 0.7193
Epoch 7/10
15/15 [=====] - 0s 6ms/step - loss: 0.9945 - accuracy: 0.8044 - val_loss: 0.2126 - val_accuracy: 0.9386
Epoch 8/10
15/15 [=====] - 0s 5ms/step - loss: 0.4832 - accuracy: 0.8769 - val_loss: 0.1506 - val_accuracy: 0.9298
Epoch 9/10
15/15 [=====] - 0s 5ms/step - loss: 0.3390 - accuracy: 0.8967 - val_loss: 0.1378 - val_accuracy: 0.9386
```

## Question 1c:

Assignment - 6

File Edit View Insert Runtime Tools Help All changes saved

Comment Share H

+ Code + Text

1(c)

```
{x} 2s scaler = StandardScaler()
X_train_normalized = scaler.fit_transform(X_train)
X_test_normalized = scaler.transform(X_test)

model_task3 = Sequential()
model_task3.add(Dense(32, activation='relu', input_shape=(X_train_normalized.shape[1],)))
model_task3.add(Dense(64, activation='relu'))
model_task3.add(Dense(1, activation='sigmoid'))

model_task3.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])

model_task3.fit(X_train_normalized, y_train, epochs=10, batch_size=32, validation_data=(X_test_normalized, y_test))

accuracy_task3 = model_task3.evaluate(X_test_normalized, y_test)[1]
print(f'Task 3 - Model Accuracy with Normalization: {accuracy_task3}')

Epoch 1/10
15/15 [=====] - 1s 17ms/step - loss: 0.6160 - accuracy: 0.6352 - val_loss: 0.4231 - val_accuracy: 0.9211
Epoch 2/10
15/15 [=====] - 0s 5ms/step - loss: 0.3641 - accuracy: 0.9165 - val_loss: 0.2532 - val_accuracy: 0.9474
Epoch 3/10
15/15 [=====] - 0s 4ms/step - loss: 0.2372 - accuracy: 0.9385 - val_loss: 0.1656 - val_accuracy: 0.9649
Epoch 4/10
15/15 [=====] - 0s 4ms/step - loss: 0.1723 - accuracy: 0.9495 - val_loss: 0.1240 - val_accuracy: 0.9649
Epoch 5/10
15/15 [=====] - 0s 4ms/step - loss: 0.1372 - accuracy: 0.9582 - val_loss: 0.1024 - val_accuracy: 0.9649
Epoch 6/10
15/15 [=====] - 0s 5ms/step - loss: 0.1132 - accuracy: 0.9692 - val_loss: 0.0892 - val_accuracy: 0.9737
Epoch 7/10
15/15 [=====] - 0s 4ms/step - loss: 0.0960 - accuracy: 0.9824 - val_loss: 0.0817 - val_accuracy: 0.9649
Epoch 8/10
15/15 [=====] - 0s 5ms/step - loss: 0.0837 - accuracy: 0.9846 - val_loss: 0.0773 - val_accuracy: 0.9649
Epoch 9/10
15/15 [=====] - 0s 5ms/step - loss: 0.0747 - accuracy: 0.9846 - val_loss: 0.0753 - val_accuracy: 0.9737
Epoch 10/10
15/15 [=====] - 0s 4ms/step - loss: 0.0670 - accuracy: 0.9846 - val_loss: 0.0727 - val_accuracy: 0.9737
```

## Question 2a:

Assignment - 6

File Edit View Insert Runtime Tools Help All changes saved

Comment Share H

RAM Disk Colab AI

+ Code + Text

2(a) Plot the loss and accuracy for both training data and validation data using the history object in the source code

```
import tensorflow as tf
from tensorflow.keras import layers, models
from tensorflow.keras.datasets import mnist
import matplotlib.pyplot as plt

(train_images, train_labels), (test_images, test_labels) = mnist.load_data()

train_images, test_images = train_images / 255.0, test_images / 255.0

model = models.Sequential()
model.add(layers.Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 1)))
model.add(layers.MaxPooling2D((2, 2)))
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.MaxPooling2D((2, 2)))
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.Flatten())
model.add(layers.Dense(64, activation='relu'))
model.add(layers.Dense(10, activation='softmax'))

model.compile(optimizer='adam',
              loss='sparse_categorical_crossentropy',
              metrics=['accuracy'])

train_images = train_images[..., tf.newaxis]
test_images = test_images[..., tf.newaxis]

history = model.fit(train_images, train_labels, epochs=10, validation_data=(test_images, test_labels))

plt.figure(figsize=(12, 4))

plt.subplot(1, 2, 1)
plt.plot(history.history['loss'])
```

Assignment - 6

File Edit View Insert Runtime Tools Help All changes saved

Comment Share H

RAM Disk Colab AI

+ Code + Text

```
train_images = train_images[..., tf.newaxis]
test_images = test_images[..., tf.newaxis]

history = model.fit(train_images, train_labels, epochs=10, validation_data=(test_images, test_labels))

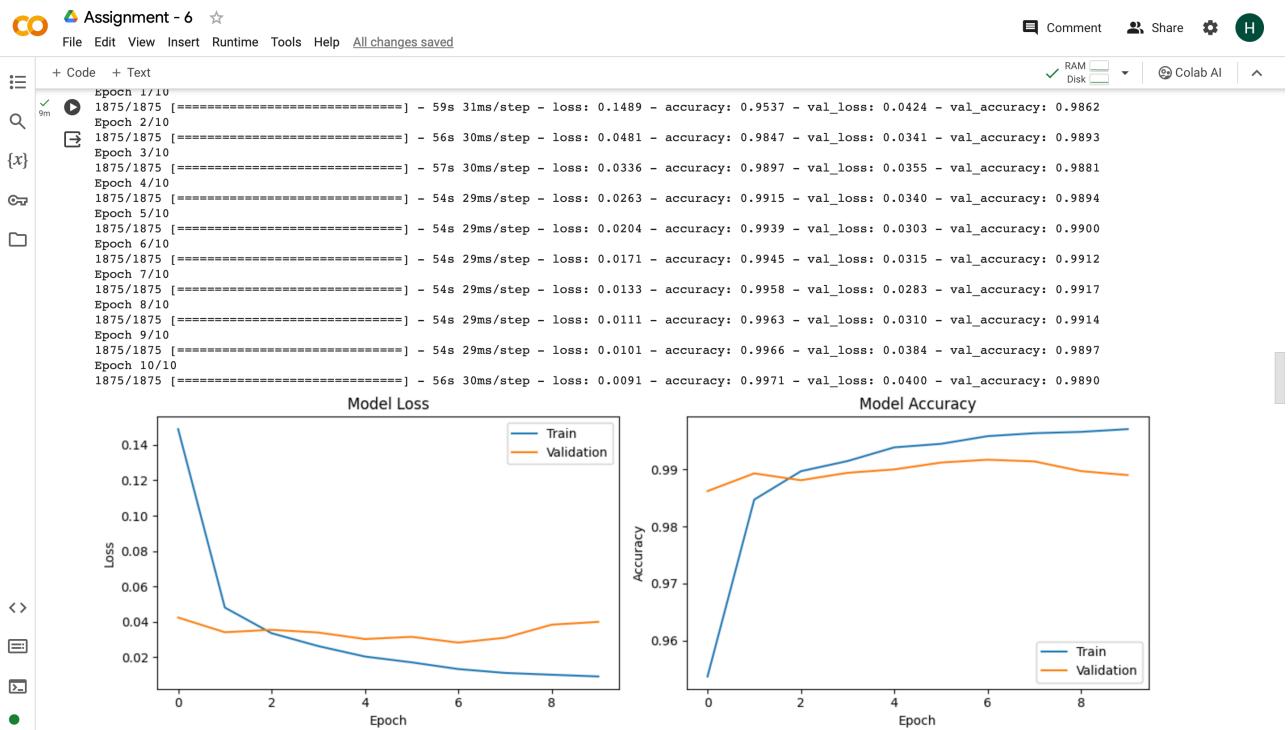
plt.figure(figsize=(12, 4))

plt.subplot(1, 2, 1)
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.title('Model Loss')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend(['Train', 'Validation'], loc='upper right')

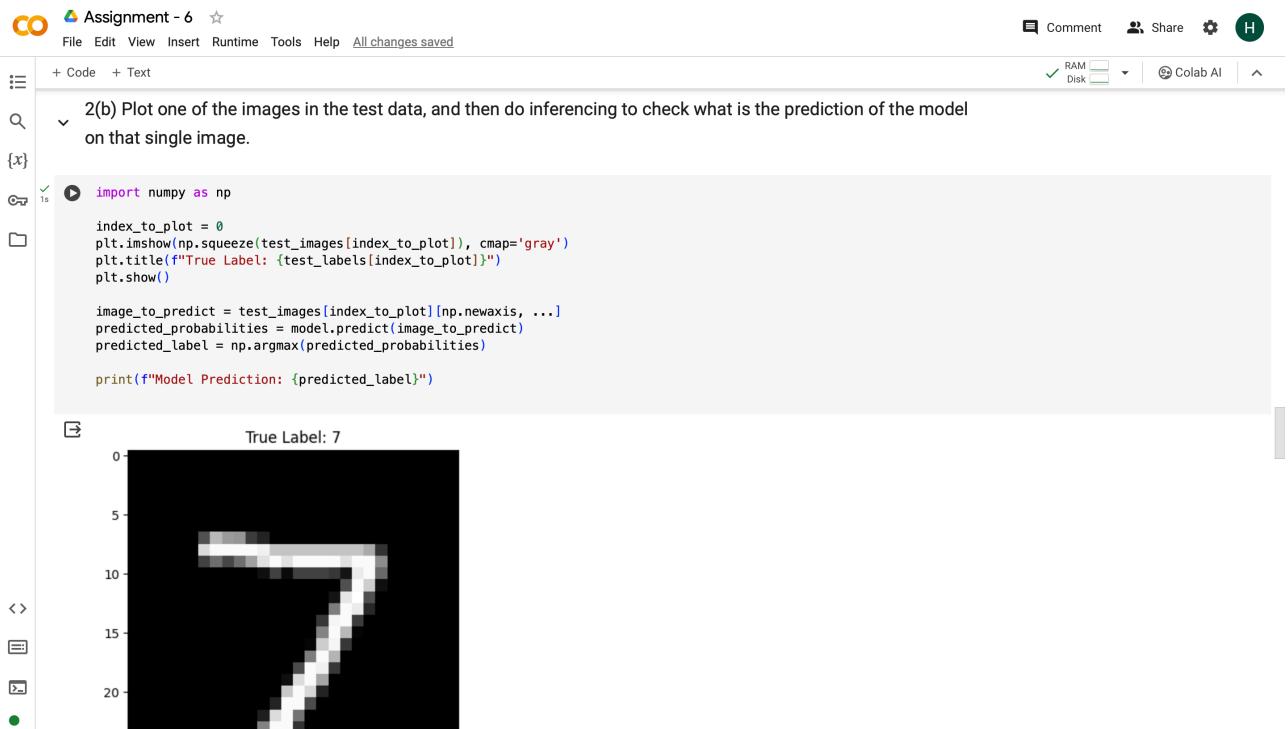
plt.subplot(1, 2, 2)
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('Model Accuracy')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend(['Train', 'Validation'], loc='lower right')

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

Epoch 1/10  
1875/1875 [=====] - 59s 31ms/step - loss: 0.1489 - accuracy: 0.9537 - val\_loss: 0.0424 - val\_accuracy: 0.9862  
Epoch 2/10  
1875/1875 [=====] - 56s 30ms/step - loss: 0.0481 - accuracy: 0.9847 - val\_loss: 0.0341 - val\_accuracy: 0.9893  
Epoch 3/10  
1875/1875 [=====] - 57s 30ms/step - loss: 0.0336 - accuracy: 0.9897 - val\_loss: 0.0355 - val\_accuracy: 0.9881  
Epoch 4/10  
1875/1875 [=====] - 54s 29ms/step - loss: 0.0263 - accuracy: 0.9915 - val\_loss: 0.0340 - val\_accuracy: 0.9894  
Epoch 5/10  
1875/1875 [=====] - 54s 29ms/step - loss: 0.0204 - accuracy: 0.9939 - val\_loss: 0.0303 - val\_accuracy: 0.9900  
Epoch 6/10  
1875/1875 [=====] - 54s 29ms/step - loss: 0.0171 - accuracy: 0.9945 - val\_loss: 0.0315 - val\_accuracy: 0.9912



## Question 2b:



Assignment - 6

```
+ Code + Text
index_to_plot = 0
plt.imshow(np.squeeze(test_images[index_to_plot]), cmap='gray')
plt.title(f"True Label: {test_labels[index_to_plot]}")
plt.show()

image_to_predict = test_images[index_to_plot][np.newaxis, ...]
predicted_probabilities = model.predict(image_to_predict)
predicted_label = np.argmax(predicted_probabilities)

print(f"Model Prediction: {predicted_label}")


1/1 [=====] ~ 0s 439ms/step
Model Prediction: 7
```

## Question 2c:

Assignment - 6

```
+ Code + Text
Model Prediction: 7

2(c) We had used 2 hidden layers and Relu activation. Try to change the number of hidden layer and the activation to tanh or sigmoid and see what happens.

modified_model = models.Sequential()
modified_model.add(layers.Conv2D(32, (3, 3), activation='tanh', input_shape=(28, 28, 1)))
modified_model.add(layers.MaxPooling2D((2, 2)))
modified_model.add(layers.Conv2D(64, (3, 3), activation='tanh'))
modified_model.add(layers.MaxPooling2D((2, 2)))
modified_model.add(layers.Flatten())
modified_model.add(layers.Dense(128, activation='tanh'))
modified_model.add(layers.Dense(10, activation='softmax'))

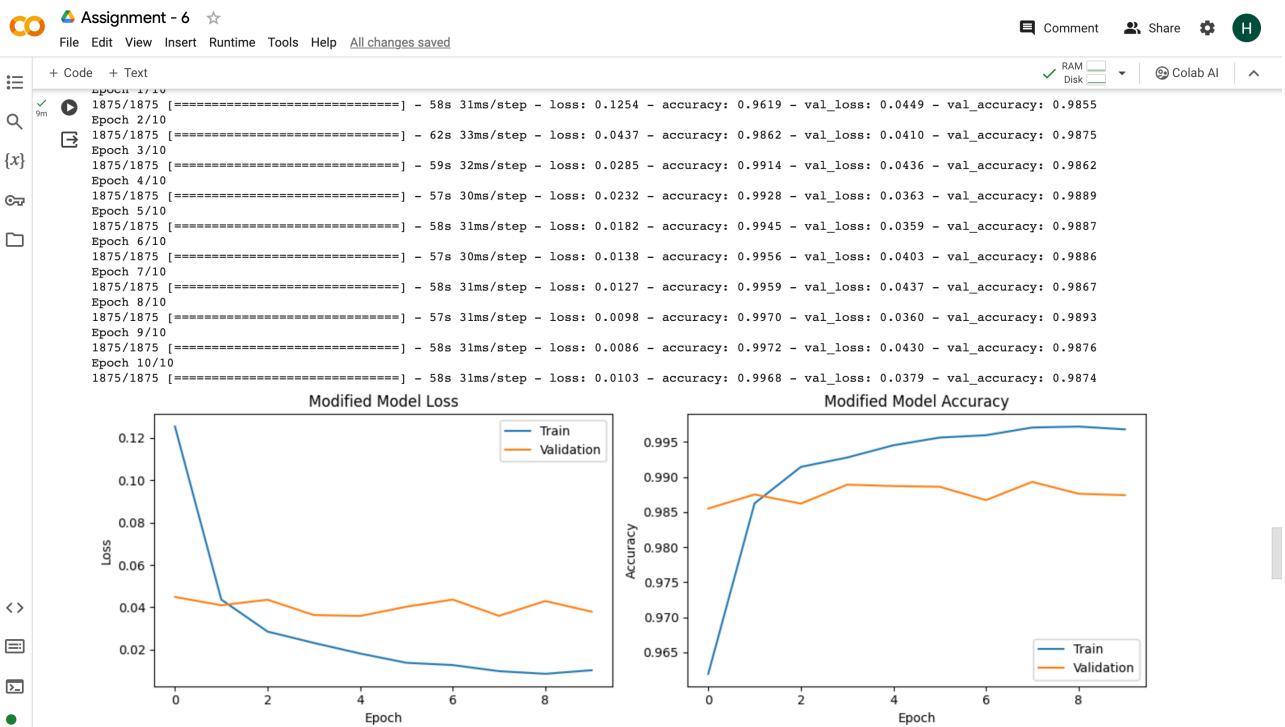
modified_model.compile(optimizer='adam',
                      loss='sparse_categorical_crossentropy',
                      metrics=['accuracy'])

modified_history = modified_model.fit(train_images, train_labels, epochs=10, validation_data=(test_images, test_labels))

plt.figure(figsize=(12, 4))

plt.subplot(1, 2, 1)
plt.plot(modified_history.history['loss'])
plt.plot(modified_history.history['val_loss'])
plt.title('Modified Model Loss')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend(['Train', 'Validation'], loc='upper right')

plt.subplot(1, 2, 2)
plt.plot(modified_history.history['accuracy'])
plt.plot(modified_history.history['val_accuracy'])
plt.title('Modified Model Accuracy')
plt.xlabel('Epoch')
```



## Question 2d:

The figure shows a Google Colab notebook interface. The code cell contains Python code for loading MNIST data, creating a sequential model, and fitting it without scaling the images. The plot shows 'Model Loss (No Scaling)' versus Epoch.

```

#(train_images, train_labels), (test_images, test_labels) = mnist.load_data()

model_no_scaling = models.Sequential()
model_no_scaling.add(layers.Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 1)))
model_no_scaling.add(layers.MaxPooling2D((2, 2)))
model_no_scaling.add(layers.Conv2D(64, (3, 3), activation='relu'))
model_no_scaling.add(layers.MaxPooling2D((2, 2)))
model_no_scaling.add(layers.Conv2D(64, (3, 3), activation='relu'))
model_no_scaling.add(layers.Flatten())
model_no_scaling.add(layers.Dense(64, activation='relu'))
model_no_scaling.add(layers.Dense(10, activation='softmax'))

model_no_scaling.compile(optimizer='adam',
                        loss='sparse_categorical_crossentropy',
                        metrics=['accuracy'])

train_images = train_images[..., tf.newaxis]
test_images = test_images[..., tf.newaxis]

history_no_scaling = model_no_scaling.fit(train_images, train_labels, epochs=10, validation_data=(test_images, test_labels))

plt.figure(figsize=(12, 4))

plt.subplot(1, 2, 1)
plt.plot(history_no_scaling.history['loss'])
plt.plot(history_no_scaling.history['val_loss'])
plt.title('Model Loss (No Scaling)')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend(['Train', 'Validation'], loc='upper right')

plt.subplot(1, 2, 2)
plt.plot(history_no_scaling.history['accuracy'])

```

**Assignment - 6**

File Edit View Insert Runtime Tools Help All changes saved

Comment Share H

```
+ Code + Text
9m 1875/1875 [=====] - 57s 30ms/step - loss: 0.2724 - accuracy: 0.9415 - val_loss: 0.0649 - val_accuracy: 0.9809
Epoch 2/10
1875/1875 [=====] - 55s 29ms/step - loss: 0.0676 - accuracy: 0.9797 - val_loss: 0.0650 - val_accuracy: 0.9809
Epoch 3/10
1875/1875 [=====] - 56s 30ms/step - loss: 0.0556 - accuracy: 0.9828 - val_loss: 0.0667 - val_accuracy: 0.9809
Epoch 4/10
1875/1875 [=====] - 54s 29ms/step - loss: 0.0440 - accuracy: 0.9868 - val_loss: 0.0420 - val_accuracy: 0.9874
Epoch 5/10
1875/1875 [=====] - 55s 29ms/step - loss: 0.0394 - accuracy: 0.9879 - val_loss: 0.0476 - val_accuracy: 0.9861
Epoch 6/10
1875/1875 [=====] - 58s 31ms/step - loss: 0.0343 - accuracy: 0.9894 - val_loss: 0.0569 - val_accuracy: 0.9839
Epoch 7/10
1875/1875 [=====] - 54s 29ms/step - loss: 0.0293 - accuracy: 0.9908 - val_loss: 0.0425 - val_accuracy: 0.9886
Epoch 8/10
1875/1875 [=====] - 57s 30ms/step - loss: 0.0260 - accuracy: 0.9924 - val_loss: 0.0630 - val_accuracy: 0.9838
Epoch 9/10
1875/1875 [=====] - 54s 29ms/step - loss: 0.0219 - accuracy: 0.9933 - val_loss: 0.0482 - val_accuracy: 0.9884
Epoch 10/10
1875/1875 [=====] - 54s 29ms/step - loss: 0.0226 - accuracy: 0.9929 - val_loss: 0.0484 - val_accuracy: 0.9880
```

**Model Loss (No Scaling)**

Epoch	Train Loss	Validation Loss
0	0.27	0.06
1	0.08	0.06
2	0.06	0.07
3	0.05	0.05
4	0.05	0.06
5	0.04	0.07
6	0.04	0.05
7	0.04	0.07
8	0.04	0.06
9	0.04	0.06
10	0.04	0.06

**Model Accuracy (No Scaling)**

Epoch	Train Accuracy	Validation Accuracy
0	0.94	0.98
1	0.98	0.98
2	0.98	0.98
3	0.99	0.99
4	0.99	0.99
5	0.99	0.98
6	0.99	0.99
7	0.99	0.98
8	0.99	0.99
9	0.99	0.99
10	0.99	0.99