

# Assignment – 4

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Github Link :

<https://github.com/ganeshkonagalla123/Neural-Networks/tree/main/Assignment4>

## 1<sup>st</sup> Program Code :

```
# Imports
import numpy as np
import pandas as pd
from keras.models import Sequential
from keras.layers import Dense
from sklearn.datasets import load_breast_cancer
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler

# Load the Breast Cancer dataset
data = load_breast_cancer()
X = data.data
y = data.target

# Normalize the data
scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)

# Split the dataset into training and testing sets
X_train, X_test, Y_train, Y_test = train_test_split(X_scaled, y, test_size=0.25, random_state=87)

# Neural network model with additional dense layers
model = Sequential()
model.add(Dense(20, input_dim=X_train.shape[1], activation='relu')) # Adjusted input_dim for Breast Cancer dataset
model.add(Dense(64, activation='relu')) # Additional dense layer
model.add(Dense(64, activation='relu')) # Additional dense layer
model.add(Dense(1, activation='sigmoid')) # Output layer for binary classification

# Compile the model
model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])

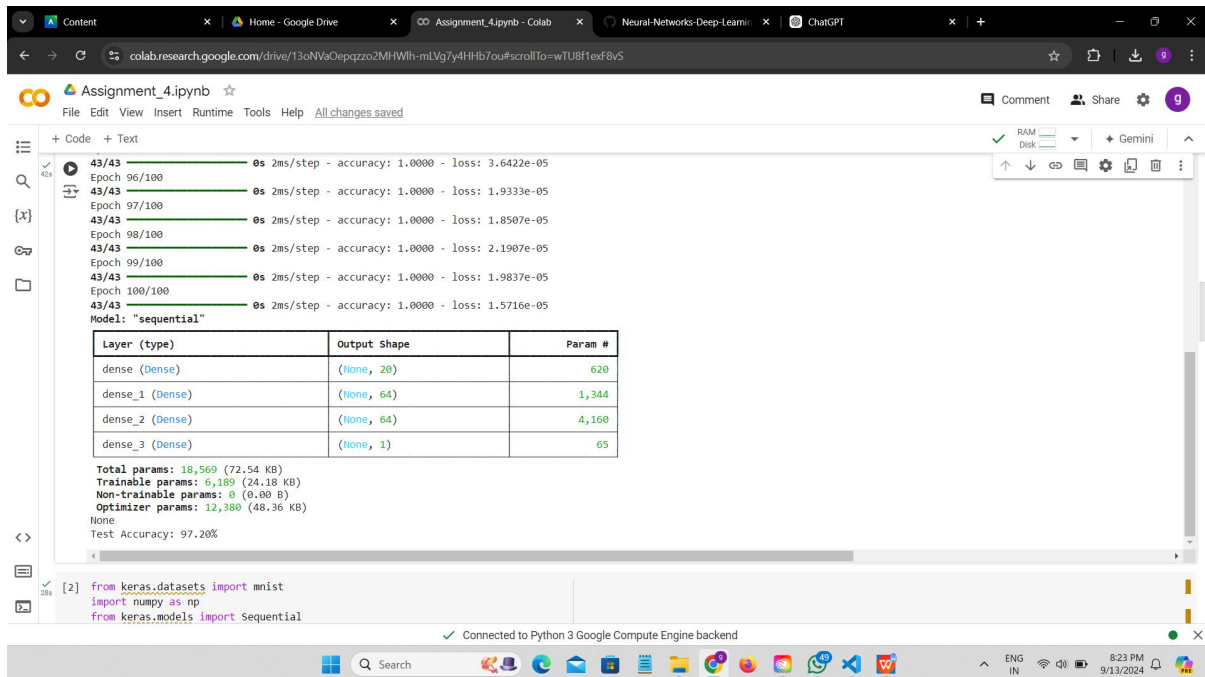
# Train the model
model.fit(X_train, Y_train, epochs=100, batch_size=10, verbose=1)

# Model summary
print(model.summary())

# Evaluate the model on the test data
loss, accuracy = model.evaluate(X_test, Y_test, verbose=0)
```

```
print(f'Test Accuracy: {accuracy*100:.2f}%')
```

## Output :



Assignment\_4.ipynb

43/43 — 0s 2ms/step - accuracy: 1.0000 - loss: 3.6422e-05  
Epoch 96/100  
43/43 — 0s 2ms/step - accuracy: 1.0000 - loss: 1.9333e-05  
Epoch 97/100  
43/43 — 0s 2ms/step - accuracy: 1.0000 - loss: 1.8507e-05  
Epoch 98/100  
43/43 — 0s 2ms/step - accuracy: 1.0000 - loss: 2.1907e-05  
Epoch 99/100  
43/43 — 0s 2ms/step - accuracy: 1.0000 - loss: 1.9837e-05  
Epoch 100/100  
43/43 — 0s 2ms/step - accuracy: 1.0000 - loss: 1.5716e-05

Model: "sequential"

| Layer (type)    | Output Shape | Param # |
|-----------------|--------------|---------|
| dense (Dense)   | (None, 20)   | 620     |
| dense_1 (Dense) | (None, 64)   | 1,344   |
| dense_2 (Dense) | (None, 64)   | 4,160   |
| dense_3 (Dense) | (None, 1)    | 65      |

Total params: 18,569 (72.54 KB)  
Trainable params: 6,189 (24.18 KB)  
Non-trainable params: 0 (0.00 B)  
Optimizer params: 12,380 (48.36 KB)  
None  
Test Accuracy: 97.20%

```
[2] from keras.datasets import mnist
import numpy as np
from keras.models import Sequential
```

Connected to Python 3 Google Compute Engine backend

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## 2<sup>nd</sup> Program Code :

```
from keras.datasets import mnist
import numpy as np
from keras.models import Sequential
from keras.layers import Dense
from keras.utils import to_categorical
import matplotlib.pyplot as plt

# Load the MNIST dataset
(train_images, train_labels), (test_images, test_labels) = mnist.load_data()

# Preprocess the data
dimData = np.prod(train_images.shape[1:])
train_data = train_images.reshape(train_images.shape[0], dimData).astype('float32') / 255
test_data = test_images.reshape(test_images.shape[0], dimData).astype('float32') / 255
train_labels_one_hot = to_categorical(train_labels)
test_labels_one_hot = to_categorical(test_labels)

# Creating the network
model = Sequential([
    Dense(512, activation='relu', input_shape=(dimData,)),
    Dense(512, activation='relu'),
    Dense(10, activation='softmax')
])

model.compile(optimizer='rmsprop', loss='categorical_crossentropy', metrics=['accuracy'])

# Train the model
history = model.fit(train_data, train_labels_one_hot, batch_size=256, epochs=10, verbose=1,
                    validation_data=(test_data, test_labels_one_hot))

# Plotting the accuracy and loss
plt.figure(figsize=(12, 5))
plt.subplot(1, 2, 1)
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Accuracy')
plt.legend()

plt.subplot(1, 2, 2)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Loss')
plt.legend()
plt.show()

# Predicting a single image
def predict_single_image(image_data):
    image_data = image_data.reshape(1, dimData).astype('float32') / 255
    prediction = model.predict(image_data)
```

```
predicted_class = np.argmax(prediction)
return predicted_class
```

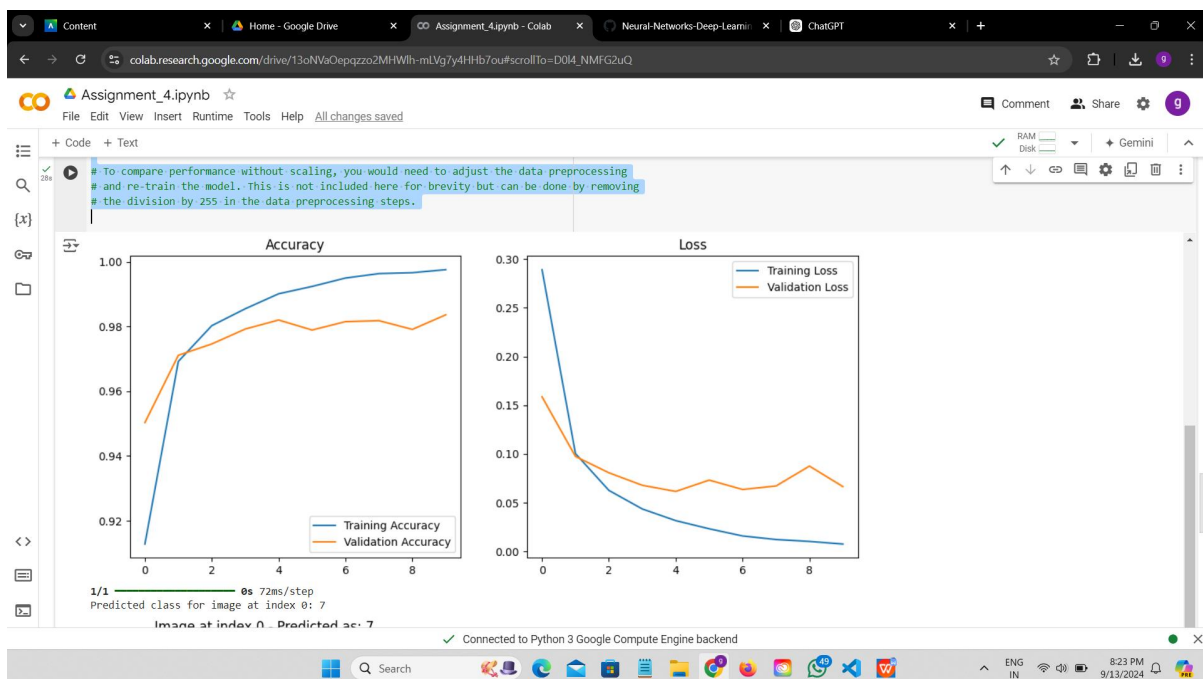
```
# Choose an image from the test set
image_index = 0 # Change this to see different predictions
predicted_class = predict_single_image(test_images[image_index])
print(f'Predicted class for image at index {image_index}: {predicted_class}')
plt.imshow(test_images[image_index], cmap='gray')
plt.title(f'Image at index {image_index} - Predicted as: {predicted_class}')
plt.show()
```

```
# Modify the model to use different activation functions and fewer layers for comparison
# You can adjust the following model architecture as needed
model_tanh = Sequential([
    Dense(512, activation='tanh', input_shape=(dimData,)),
    Dense(10, activation='softmax')
])
```

```
model_tanh.compile(optimizer='rmsprop', loss='categorical_crossentropy', metrics=['accuracy'])
model_tanh.fit(train_data, train_labels_one_hot, batch_size=256, epochs=10, verbose=1,
               validation_data=(test_data, test_labels_one_hot))
```

```
# To compare performance without scaling, you would need to adjust the data preprocessing
# and re-train the model. This is not included here for brevity but can be done by removing
# the division by 255 in the data preprocessing steps.
```

## Output :



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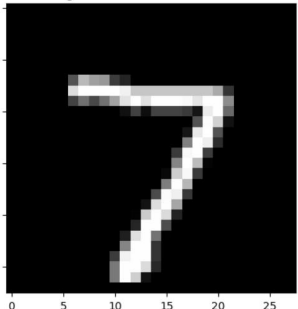
colab.research.google.com/drive/13oNvaOepqzzo2MhWlh-mLVg7y4Hh67ou#scrollTo=D0l4\_NMFG2uQ

Assignment\_4.ipynb

File Edit View Insert Runtime Tools Help All changes saved

+ Code + Text

1/1 0s 72ms/step  
Predicted class for image at index 0: 7  
Image at index 0 - Predicted as: 7



Epoch 1/10  
235/235 5s 17ms/step - accuracy: 0.8157 - loss: 0.6131 - val\_accuracy: 0.9029 - val\_loss: 0.3202  
Epoch 2/10  
235/235 4s 18ms/step - accuracy: 0.9274 - loss: 0.2513 - val\_accuracy: 0.9343 - val\_loss: 0.2150

RAM Disk

+ Gemini

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