## Worksheet 4 Output

## ✓ Importing Required Libraries import os import numpy as np import tensorflow as tf from tensorflow.keras.utils import to\_categorical from sklearn.model\_selection import train\_test\_split from sklearn.utils import shuffle import matplotlib.pyplot as plt from PIL import Image from tensorflow import keras from tensorflow.keras.models import Sequential from tensorflow.keras.layers import Dense, Flatten, BatchNormalization, Dropout from tensorflow.keras.optimizers import Adam from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping import matplotlib.pyplot as plt



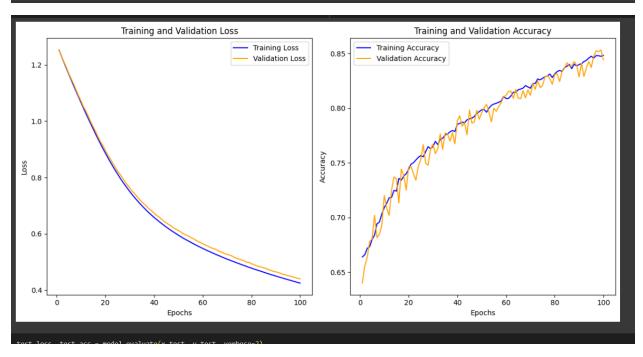
```
num classes = 10
    input_shape = (28*28, 1)
    model = keras.Sequential(
    keras.layers.Input(shape=input_shape),
    keras.layers.Flatten(),
    keras.layers.Dense(64, activation="sigmoid"),
    keras.layers.Dense(128, activation="sigmoid"),
    keras.layers.Dense(256, activation="sigmoid"),
    keras.layers.Dense(num_classes, activation="softmax"),
model.summary()

→ Model: "sequential_8"

      Layer (type)
                                             Output Shape
                                                                                   Param #
      flatten_8 (Flatten)
      dense_26 (Dense)
      dense_27 (Dense)
      dense 28 (Dense)
      dense_29 (Dense)
     Total params: 94,154 (367.79 KB)
     Trainable params:
                             (367.79 KB)
     Non-trainable params: 0 (0.00 B)
```

```
Epoch 1/100
                             1s 7ms/step - accuracy: 0.6654 - loss: 1.2685 - val_accuracy: 0.6400 - val_loss: 1.2532
113/113 -
Epoch 2/100
                            1s 5ms/step - accuracy: 0.6653 - loss: 1.2388 - val_accuracy: 0.6557 - val_loss: 1.2308
113/113 -
113/113
                            1s 6ms/step - accuracy: 0.6652 - loss: 1.2198 - val_accuracy: 0.6635 - val_loss: 1.2108
113/113
                             1s 6ms/step - accuracy: 0.6711 - loss: 1.1970 - val_accuracy: 0.6780 - val_loss: 1.1919
Epoch 5/100
113/113 -
                             1s 6ms/step - accuracy: 0.6748 - loss: 1.1754 - val_accuracy: 0.6804 - val_loss: 1.1698
Epoch 6/100
113/113
                             2s 9ms/step - accuracy: 0.6831 - loss: 1.1519 - val_accuracy: 0.7020 - val_loss: 1.1508
Epoch 7/100
113/113
                             1s 9ms/step - accuracy: 0.7022 - loss: 1.1257 - val_accuracy: 0.6816 - val_loss: 1.1325
Epoch 8/100
113/113
                             1s 9ms/step - accuracy: 0.6950 - loss: 1.1115 - val_accuracy: 0.6847 - val_loss: 1.1115
Epoch 9/100
113/113
                            1s 6ms/step - accuracy: 0.7025 - loss: 1.0855 - val accuracy: 0.6929 - val loss: 1.0912
Epoch 10/100
113/113
                            1s 6ms/step - accuracy: 0.7069 - loss: 1.0727 - val accuracy: 0.7200 - val loss: 1.0725
Epoch 11/100
                            1s 6ms/step - accuracy: 0.7100 - loss: 1.0556 - val accuracy: 0.7086 - val loss: 1.0521
113/113
Epoch 12/100
                            1s 6ms/step - accuracy: 0.7143 - loss: 1.0380 - val_accuracy: 0.7020 - val_loss: 1.0368
113/113
Epoch 13/100
                            1s 6ms/step - accuracy: 0.7191 - loss: 1.0172 - val_accuracy: 0.7251 - val_loss: 1.0180
113/113 -
Epoch 14/100
                            1s 7ms/step - accuracy: 0.7238 - loss: 0.9874 - val_accuracy: 0.7373 - val_loss: 0.9985
113/113 -
Epoch 15/100
                            1s 6ms/step - accuracy: 0.7247 - loss: 0.9759 - val_accuracy: 0.7353 - val_loss: 0.9811
113/113
Epoch 16/100
113/113
                            1s 6ms/step - accuracy: 0.7364 - loss: 0.9618 - val_accuracy: 0.7133 - val_loss: 0.9639
Epoch 17/100
113/113
                             1s 6ms/step - accuracy: 0.7285 - loss: 0.9345 - val_accuracy: 0.7443 - val_loss: 0.9449
Epoch 18/100
113/113 -
                             1s 6ms/step - accuracy: 0.7389 - loss: 0.9195 - val_accuracy: 0.7384 - val_loss: 0.9282
Epoch 19/100
113/113 -
                            1s 6ms/step - accuracy: 0.7388 - loss: 0.9055 - val_accuracy: 0.7251 - val_loss: 0.9141
Epoch 20/100
```

```
113/113
                            1s 6ms/step - accuracy: 0.8329 - loss: 0.4772 - val accuracy: 0.8365 - val loss: 0.4822
Epoch 85/100
                            1s 6ms/step - accuracy: 0.8409 - loss: 0.4585 - val_accuracy: 0.8416 - val_loss: 0.4796
113/113
Fnoch 86/100
                            1s 6ms/step - accuracy: 0.8382 - loss: 0.4651 - val_accuracy: 0.8384 - val_loss: 0.4770
113/113
Epoch 87/100
113/113
                            1s 10ms/step - accuracy: 0.8408 - loss: 0.4545 - val accuracy: 0.8388 - val loss: 0.4727
113/113
                            1s 9ms/step - accuracy: 0.8449 - loss: 0.4488 - val_accuracy: 0.8427 - val_loss: 0.4704
Epoch 89/100
113/113
                            1s 9ms/step - accuracy: 0.8372 - loss: 0.4509 - val_accuracy: 0.8392 - val_loss: 0.4669
Epoch 90/100
113/113
                             1s 6ms/step - accuracy: 0.8413 - loss: 0.4541 - val_accuracy: 0.8286 - val_loss: 0.4651
Epoch 91/100
113/113
                            1s 6ms/step - accuracy: 0.8382 - loss: 0.4506 - val_accuracy: 0.8412 - val_loss: 0.4628
Epoch 92/100
113/113
                            1s 6ms/step - accuracy: 0.8433 - loss: 0.4445 - val accuracy: 0.8290 - val loss: 0.4598
Epoch 93/100
113/113
                            1s 6ms/step - accuracy: 0.8411 - loss: 0.4479 - val accuracy: 0.8369 - val loss: 0.4562
Epoch 94/100
                            1s 6ms/step - accuracy: 0.8466 - loss: 0.4346 - val_accuracy: 0.8427 - val_loss: 0.4550
113/113
Epoch 95/100
                            1s 6ms/step - accuracy: 0.8500 - loss: 0.4386 - val_accuracy: 0.8373 - val_loss: 0.4511
113/113
Epoch 96/100
                            1s 6ms/step - accuracy: 0.8524 - loss: 0.4248 - val_accuracy: 0.8475 - val_loss: 0.4499
113/113
113/113
                            1s 6ms/step - accuracy: 0.8493 - loss: 0.4279 - val_accuracy: 0.8525 - val_loss: 0.4467
Epoch 98/100
113/113
                            1s 6ms/step - accuracy: 0.8442 - loss: 0.4394 - val_accuracy: 0.8518 - val_loss: 0.4446
Epoch 99/100
113/113
                            1s 6ms/step - accuracy: 0.8483 - loss: 0.4290 - val_accuracy: 0.8533 - val_loss: 0.4417
Epoch 100/100
                            1s 9ms/step - accuracy: 0.8466 - loss: 0.4299 - val_accuracy: 0.8443 - val_loss: 0.4401
113/113
```

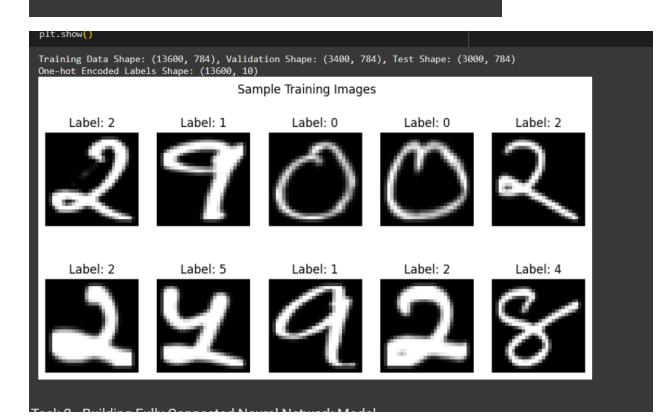


94/94 - 0s - 2ms/step - accuracy: 0.8493 - loss: 0.4373
Test Accuracy: 0.8493
model.save("devnagari digit classifier.h5")

MRNHUGALING we saring your wood as a MESS file via 'model.com()' or 'baras caring.come model(model)'. This file formst is considered legacy, we recommend using instead the native formst, e.g. 'model.com('my\_model.kerus')' or 'baras.coming.come\_model(model)'. This file formst is considered legacy, we recommend using instead the native formst e.g. 'model.com('my\_model.kerus')' or 'baras.coming.come\_model(model)'. This file formst is considered legacy. We recommend using instead the native formst formst, e.g. 'model.com('my\_model.kerus')' or 'baras.coming.come\_model(model)'. This file formst is considered legacy. We recommend using instead the native formst formst, e.g. 'model.com('my\_model.kerus')' or 'baras.coming.come\_model(model)'. This file formst is considered legacy. We recommend using instead the native formst formst, e.g. 'model.com('my\_model.kerus')' or 'baras.coming.come\_model(model)'. This file formst is considered legacy. We recommend using instead the native formst formst formst. The native formst formst

**94/94 Os** 2ms/step Predicted label for first image: 8

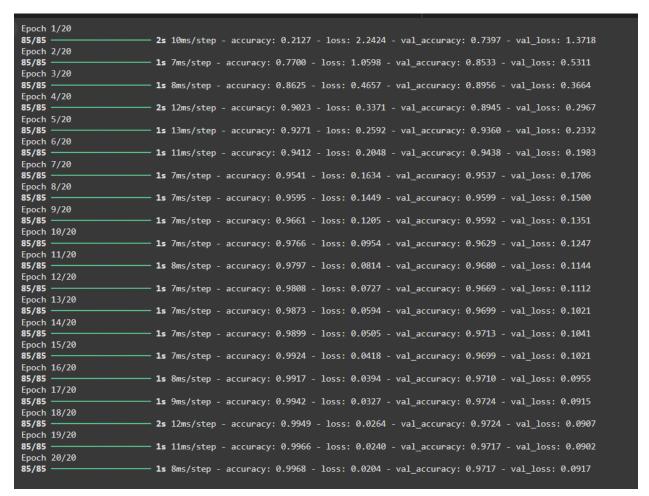
True label for first image: 0

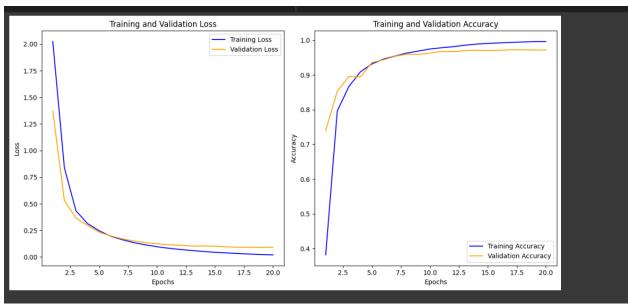


## Model: "sequential\_11"

Layer (type)	Output Shape	Param #
dense_38 (Dense)	(None, 64)	50,240
dense_39 (Dense)	(None, 128)	8,320
dense_40 (Dense)	(None, 256)	33,024
dense_41 (Dense)	(None, 10)	2,570

Total params: 94,154 (367.79 KB)
Trainable params: 94,154 (367.79 KB)
Non-trainable params: 0 (0.00 B)





94/94 - 0s - 2ms/step - accuracy: 0.9740 - loss: 0.0903

Test Accuracy: 0.9740 Test Loss: 0.0903

MMNDGrabl/You are saving your model as an 1695 file via 'model.sero')' or 'keros.saving.save\_model(model, 'my\_model.keros')'. Bits films tis considered legacy, ble recommend using instead the native feras format, e.g. 'model.seroe('my\_model.keros')' or 'keros.saving.save\_model(model, 'my\_model.keros')'. MMNDGrables\*(conservative) as 'formating.gitgit\_consists.note\_gitgit\_gitgit\_consists.note\_gitgit\_gitgit\_consists.note\_gitgit