NNDL ICP6

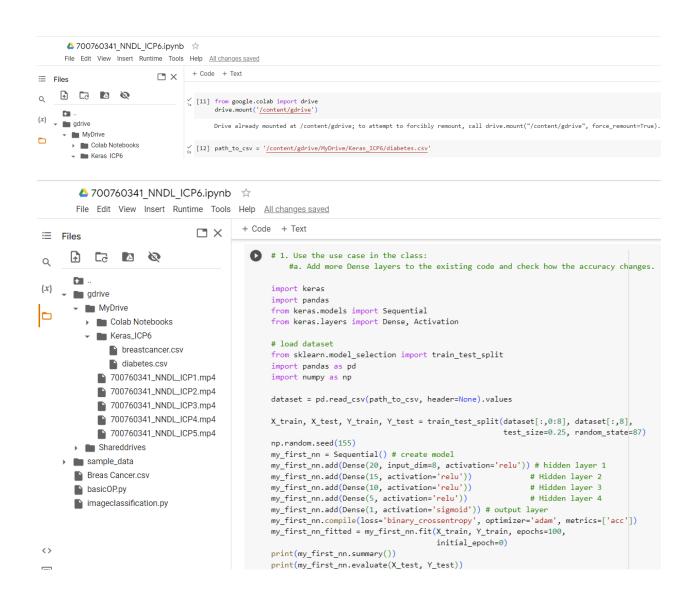
Student Name: Neeraj Kumar Barigela

Student Number:700760431

GitHub Link: https://github.com/neeraj4944/NNDL ICP6

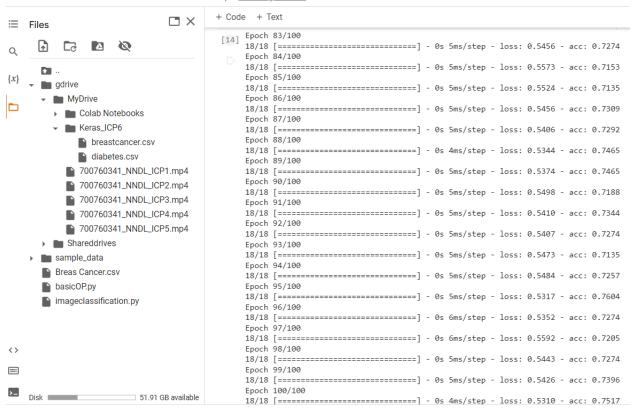
Video Link:

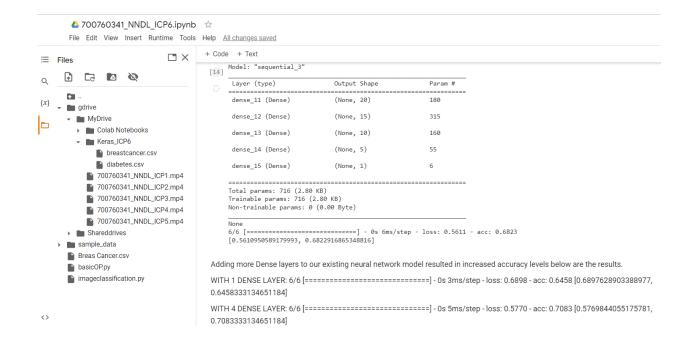
https://drive.google.com/file/d/1DSYIGnHxMTZUmFhqa6u1QH9zIWyqAYDx/view?usp=sharing

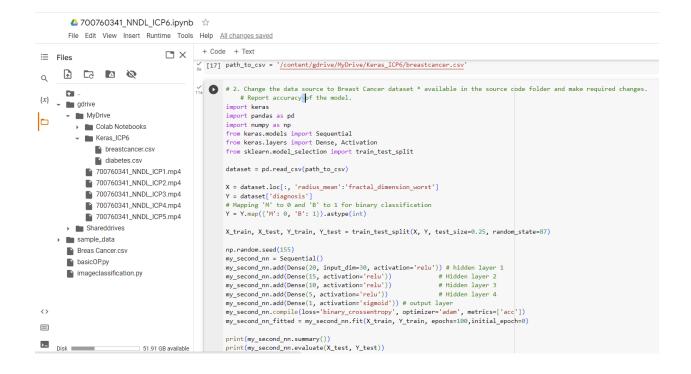


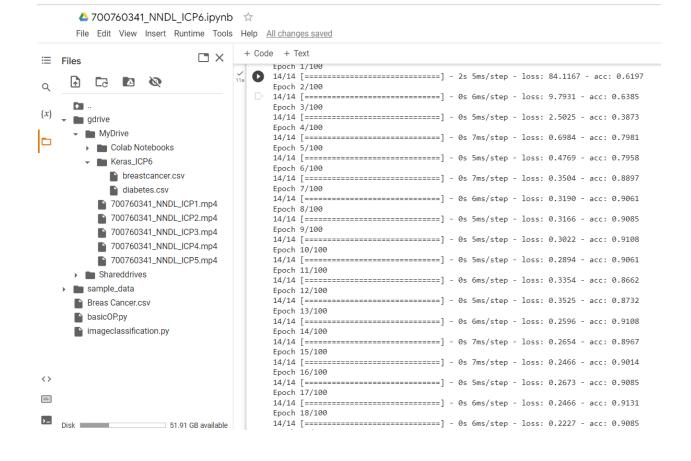
△ 700760341 NNDL ICP6.ipynb ☆

File Edit View Insert Runtime Tools Help All changes saved

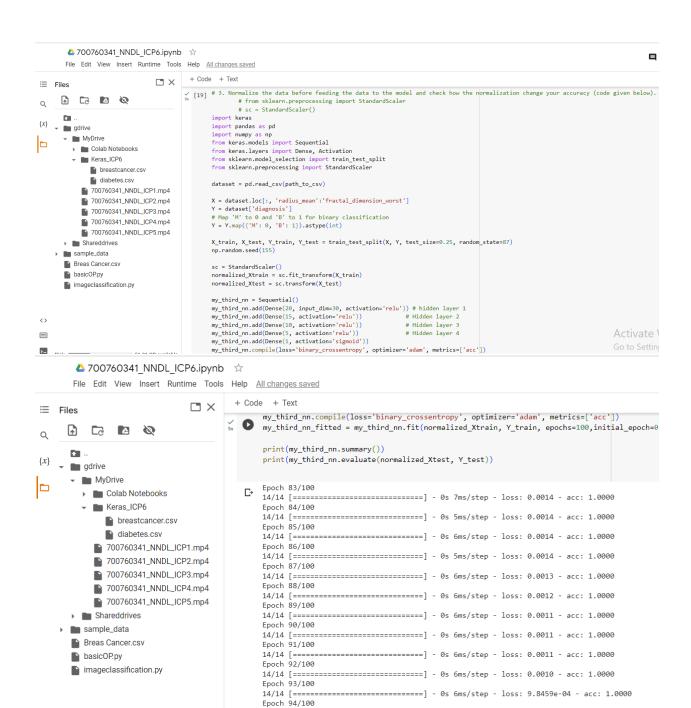








△ 700760341 NNDL ICP6.ipynb ☆ File Edit View Insert Runtime Tools Help All changes saved + Code + Text **≔** Files Epoch 21/100 lack14/14 [=============] - 0s 6ms/step - loss: 0.2155 - acc: 0.9272 Epoch 22/100 {*x*} Epoch 23/100 14/14 [=============] - 0s 7ms/step - loss: 0.2081 - acc: 0.9178 Epoch 24/100 Colab Notebooks 14/14 [============] - Os 6ms/step - loss: 0.2098 - acc: 0.9249 Epoch 25/100 ▼ Meras_ICP6 breastcancer.csv Epoch 26/100 diabetes.csv 14/14 [=============] - Os 6ms/step - loss: 0.2100 - acc: 0.9225 Epoch 27/100 ■ 700760341_NNDL_ICP1.mp4 14/14 [============] - 0s 7ms/step - loss: 0.2031 - acc: 0.9296 ■ 700760341_NNDL_ICP2.mp4 Epoch 28/100 ■ 700760341_NNDL_ICP3.mp4 Epoch 29/100 ■ 700760341_NNDL_ICP4.mp4 ■ 700760341_NNDL_ICP5.mp4 Epoch 30/100 Shareddrives Epoch 31/100 sample_data 14/14 [====== Breas Cancer.csv Epoch 32/100 14/14 [=============] - 0s 7ms/step - loss: 0.1973 - acc: 0.9296 basicOP.py Epoch 33/100 imageclassification.py 14/14 [==============] - 0s 6ms/step - loss: 0.2136 - acc: 0.9131 Fnoch 34/100 Epoch 35/100 <> Epoch 36/100 14/14 [===============] - 0s 6ms/step - loss: 0.2605 - acc: 0.9202 = Epoch 37/100 14/14 [=============] - 0s 7ms/step - loss: 0.2332 - acc: 0.9155 Disk 51.91 GB available Epoch 38/100 ♣ 700760341_NNDL_ICP6.ipynb ☆ File Edit View Insert Runtime Tools Help All changes saved + Code + Text \square \times Epoch 96/100 Q Epoch 97/100 • 14/14 [===== =========] - Os 4ms/step - loss: 0.1748 - acc: 0.9319 $\{x\}$ Epoch 98/100 qdrive Epoch 99/100 Colab Notebooks 14/14 [=============] - 0s 4ms/step - loss: 0.1639 - acc: 0.9319 Fnoch 100/100 ▼ Meras_ICP6 14/14 [=============] - 0s 4ms/step - loss: 0.1646 - acc: 0.9366 breastcancer.csv Model: "sequential 4" diabetes.csv Layer (type) Output Shape ■ 700760341 NNDL ICP1.mp4 ■ 700760341 NNDL ICP2.mp4 dense_16 (Dense) (None, 20) 620 ■ 700760341_NNDL_ICP3.mp4 dense 17 (Dense) (None, 15) 315 ■ 700760341_NNDL_ICP4.mp4 ■ 700760341_NNDL_ICP5.mp4 dense_18 (Dense) (None, 10) 160 Shareddrives dense_19 (Dense) (None, 5) sample_data (None, 1) Breas Cancer.csv dense_20 (Dense) 6 basicOP.py ----imageclassification.py Total params: 1156 (4.52 KB) Trainable params: 1156 (4.52 KB) Non-trainable params: 0 (0.00 Byte) <> 5/5 [==============] - 0s 4ms/step - loss: 0.3036 - acc: 0.8881 [0.3035851716995239, 0.8881118893623352] \equiv



Epoch 95/100

Epoch 96/100 14/14 [=====

Epoch 97/100

<>

=

>_

Disk

51.91 GB available

14/14 [==============] - 0s 6ms/step - loss: 9.4301e-04 - acc: 1.0000

==========] - Os 6ms/step - loss: 8.8701e-04 - acc: 1.0000

△ 700760341 NNDL ICP6.ipynb ☆



#change the labels frominteger to one-hot encoding. to_categorical is doing the same thing as LabelEncoder()

train_data /=255.0 test data /=255.0

train_labels_one_hot = to_categorical(train_labels)
test_labels_one_hot = to_categorical(test_labels)

<>

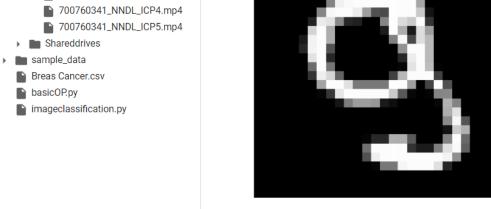
==

Disk 51.91 GB available

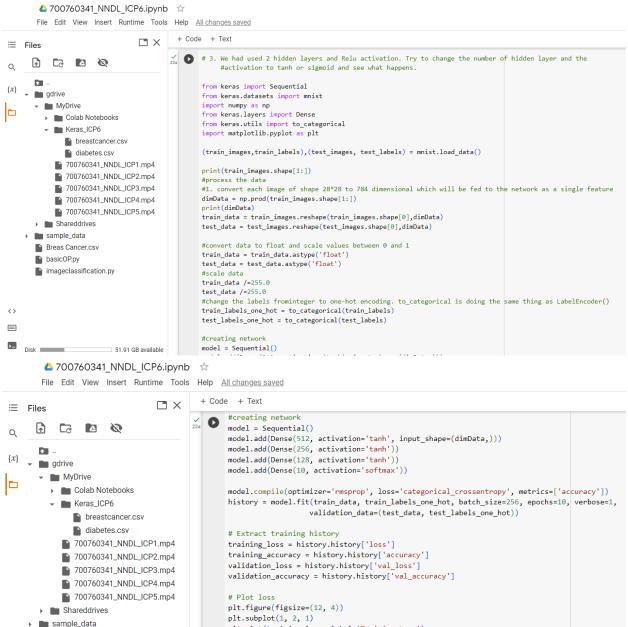
4 700760341_NNDL_ICP6.ipynb ☆

File Edit View Insert Runtime Tools Help All changes saved

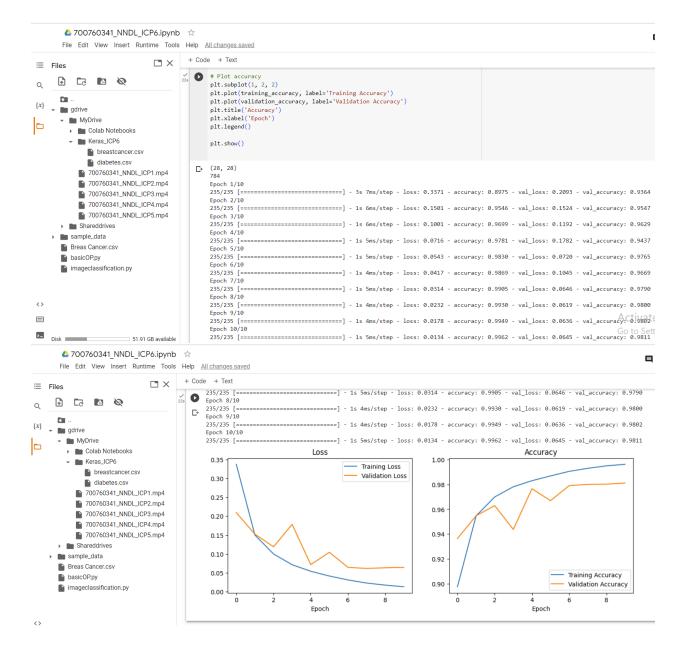
```
+ Code + Text
                                     \square \times
≔ Files
                                                        #creating network
                                               √
23s D
       model = Sequential()
Q
                                                        model.add(Dense(512, activation='relu', input_shape=(dimData,)))
                                                        model.add(Dense(512, activation='relu'))
        Can ...
\{x\}
                                                        model.add(Dense(10, activation='softmax'))
       gdrive
        MyDrive
                                                        model.compile(optimizer='rmsprop', loss='categorical_crossentropy', metrics=['accuracy'])
Colab Notebooks
                                                        history = model.fit(train_data, train_labels_one_hot, batch_size=256, epochs=10, verbose=1,
                                                                            validation_data=(test_data, test_labels_one_hot))
           ▼ ■ Keras_ICP6
                 breastcancer.csv
                                                        # Extract training history
                 diabetes.csv
                                                        training_loss = history.history['loss']
              ■ 700760341_NNDL_ICP1.mp4
                                                        training_accuracy = history.history['accuracy']
                                                        validation_loss = history.history['val_loss']
              ■ 700760341_NNDL_ICP2.mp4
                                                        validation_accuracy = history.history['val_accuracy']
              ■ 700760341_NNDL_ICP3.mp4
              ■ 700760341_NNDL_ICP4.mp4
                                                        # Plot loss
                                                        plt.figure(figsize=(12, 4))
              ■ 700760341_NNDL_ICP5.mp4
                                                        plt.subplot(1, 2, 1)
        Shareddrives
                                                        plt.plot(training_loss, label='Training Loss')
     sample_data
                                                        plt.plot(validation_loss, label='Validation Loss')
        Breas Cancer.csv
                                                        plt.title('Loss')
                                                        plt.xlabel('Epoch')
        basicOP.py
                                                        plt.legend()
        imageclassification.pv
                                                        # Plot accuracy
                                                        plt.subplot(1, 2, 2)
                                                        plt.plot(training_accuracy, label='Training Accuracy')
<>
                                                        plt.plot(validation_accuracy, label='Validation Accuracy')
                                                        plt.title('Accuracy')
\equiv
                                                        plt.xlabel('Epoch')
                                                        plt.legend()
                         51.91 GB available
       ♣ 700760341_NNDL_ICP6.ipynb ☆
                                                                                                                                                           File Edit View Insert Runtime Tools Help All changes saved
                                     + Code + Text
                             \square \times
≔ Files
                                     plt.show()
     Q
                                            # select a random image from the test data
idx = np.random.randint(test_data.shape[0])
      Dia .
\{x\} \longrightarrow gdrive
                                            image = test_data[idx].reshape(28, 28)
      # plot the selected image
        ▶ Colab Notebooks
                                           plt.figure()
plt.imshow(image, cmap='gray')
        breastcancer.csv
                                            plt.axis('off'
                                            plt.title('Selected Image')
             diabetes.csv
           ■ 700760341_NNDL_ICP1.mp4
                                           # do inferencing to check the model prediction on the selected image
prediction = model.predict(image.reshape(1, 784))
prediction = np.argmax(prediction)
           ■ 700760341_NNDL_ICP2.mp4
           ■ 700760341 NNDL ICP3.mp4
           ■ 700760341_NNDL_ICP4.mp4
           ■ 700760341 NNDL ICP5.mp4
                                            print('Predicted label:', prediction)
      Shareddrives
    sample data
                                        [→ (28, 28)
      Breas Cancer.csv
      basicOP.py
                                                                 =======] - 2s 6ms/step - loss: 0.2961 - accuracy: 0.9094 - val_loss: 0.1331 - val_accuracy: 0.9584
      imageclassification.pv
                                            235/235 [==
                                            Fnoch 2/10
                                            235/235 [==
                                                                 =======] - 1s 5ms/step - loss: 0.1008 - accuracy: 0.9687 - val_loss: 0.0966 - val_accuracy: 0.9693
                                            Epoch 3/10
                                                                   ========] - 1s 4ms/step - loss: 0.0636 - accuracy: 0.9803 - val loss: 0.1044 - val accuracy: 0.9672
                                            235/235 [==
<>
                                            Epoch 4/10
                                                        =============================== ] - 1s 4ms/step - loss: 0.0440 - accuracy: 0.9862 - val_loss: 0.0936 - val_accuracy: 0.9700 -
\equiv
                                            >_
   Disk Disk
                   51.91 GB available
```



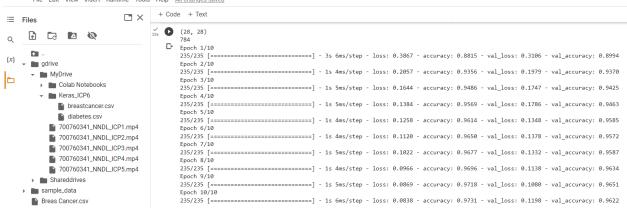
<> **■**



```
sample_data
                                                   plt.plot(training_loss, label='Training Loss')
       Breas Cancer.csv
                                                   plt.plot(validation_loss, label='Validation Loss')
       basicOP.py
                                                   plt.title('Loss'
                                                   plt.xlabel('Epoch')
       imageclassification.py
                                                   plt.legend()
                                                   # Plot accuracy
                                                   plt.subplot(1, 2, 2)
                                                   plt.plot(training_accuracy, label='Training Accuracy')
                                                   plt.plot(validation_accuracy, label='Validation Accuracy')
=:
                                                   plt.title('Accuracy')
                                                   plt.xlahel('Fpoch')
```



△ 700760341 NNDL ICP6.ipynb ☆ File Edit View Insert Runtime Tools Help All changes saved + Code + Text \square \times # 4. Kun the same code without scaling the images and check the performance? √ 230 **D** Q import matplotlib.pvplot as plt from keras import Sequential CH .. from keras.datasets import mnist $\{x\}$ import numpy as np from keras.layers import Dense from keras.utils import to_categorical Colab Notebooks ▼ Meras_ICP6 (train_images, train_labels), (test_images, test_labels) = mnist.load_data() breastcancer.csv diabetes.csv print(train images.shape[1:]) 700760341_NNDL_ICP1.mp4 # 1. Convert each image of shape 28*28 to 784 dimensional which will be fed to the network as a single feature dimData = np.prod(train_images.shape[1:]) ■ 700760341_NNDL_ICP2.mp4 ■ 700760341_NNDL_ICP3.mp4 ■ 700760341_NNDL_ICP4.mp4 train_data = train_images.reshape(train_images.shape[0], dimData) test_data = test_images.reshape(test_images.shape[0], dimData) 700760341_NNDL_ICP5.mp4 Shareddrives # Convert data to float (no scaling) sample_data train data = train data.astvpe('float') test_data = test_data.astype('float') Breas Cancer.csv asicOP.py # Change the labels from integer to one-hot encoding. to_categorical is doing the same thing as LabelEncoder() imageclassification.py train_labels_one_hot = to_categorical(train_labels) test_labels_one_hot = to_categorical(test_labels) # Creating network <> model = Sequential() model.add(Dense(512, activation='tanh', input_shape=(dimData,))) model.add(Dense(256, activation='tanh')) model.add(Dense(128, activation='tanh')) model.add(Dense(10, activation='softmax')) 51.91 GB available △ 700760341 NNDL ICP6.ipynb ☆ File Edit View Insert Runtime Tools Help All changes saved + Code + Text $\square \times$ **≔** Files model.compile(optimizer='rmsprop', loss='categorical_crossentropy', metrics=['accuracy']) [c 🖪 🕲 history = model.fit(train_data, train_labels_one_hot, batch_size=256, epochs=10, verbose=1, 0 validation_data=(test_data, test_labels_one_hot)) $\{X\}$ # Extract training history training_loss = history.history['loss'] training_accuracy = history.history['accuracy'] Colab Notebooks validation_loss = history.history['val_loss'] validation_accuracy = history.history['val_accuracy'] ▼ Meras_ICP6 breastcancer.csv # Plot loss diabetes.csv plt.figure(figsize=(12, 4)) ■ 700760341_NNDL_ICP1.mp4 plt.subplot(1, 2, 1) plt.plot(training_loss, label='Training Loss') ■ 700760341_NNDL_ICP2.mp4 plt.plot(validation_loss, label='Validation Loss') ■ 700760341_NNDL_ICP3.mp4 plt.title('Loss') ■ 700760341_NNDL_ICP4.mp4 plt.xlabel('Epoch') ■ 700760341_NNDL_ICP5.mp4 plt.legend() Shareddrives # Plot accuracy sample_data plt.subplot(1, 2, 2) Breas Cancer.csv plt.plot(training_accuracy, label='Training Accuracy') asicOP.py plt.plot(validation_accuracy, label='Validation Accuracy') plt.title('Accuracy') imageclassification.py plt.xlabel('Epoch') plt.legend() plt.show()

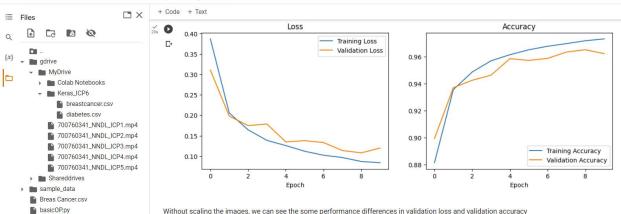


E

△ 700760341_NNDL_ICP6.ipynb 🕏

imageclassification.py

File Edit View Insert Runtime Tools Help All changes saved



Without scaling the images, we can see the some performance differences in validation loss and validation accuracy