

ASSIGNMENT 3

ASSIGNMENT DATE	12 OCTOBER 2022
TEAM ID	PNT2022TMID17273
PROJECT NAME	A NOVEL METHOD FOR HANDWRITTEN DIGIT RECOGNITION SYSTEM
MAXIMUM MARKS	2 marks

QUESTION: WRITE A PYTHON CODE FOR BLINKING LED AND TRAFFIC LIGHT FOR RASPBERRY PI.

PROJECT :

HANDWRITTEN DIGIT RECOGNITION SYSTEM

#Unzipping

#Data Augmentation

```
[ ]: from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

```
train_gen = ImageDataGenerator(rescale = 1. / 255 ,  
                                zoom_range = 0.2 ,  
                                horizontal_flip = True)  
test_gen = ImageDataGenerator(rescale = 1. / 255 )
```

```
[ ]: xtrain = train_gen . flow_from_directory( ' /content/flowers ' ,  
                                              target_size = ( 64 , 64 ) ,  
                                              class_mode = ' categorical ' ,  
                                              batch_size = 100 )
```

Found 4317 images belonging to 5 classes.

#Train

```
[ ]: from tensorflow.keras.models import Sequential from tensorflow.keras.layers  
import Convolution2D, MaxPooling2D, Dense, Flatten from keras.callbacks import  
EarlyStopping, ReduceLROnPlateau
```

```
[ ]: model = Sequential() model.add(Convolution2D(32,(3,3),activation='relu',input_shape=(64,64,3)))  
model.add(MaxPooling2D((2,2))) model.add(Flatten())  
  
model.add(Dense(400,activation='relu')) model.add(Dense(200,activation='relu'))  
model.add(Dense(100,activation='relu')) model.add(Dense(5,activation='softmax'))
```

```
[ ]: model . compile(optimizer = ' adam' ,loss = ' categorical_crossentropy ' ,metrics = [ ' accuracy ' ] )
```

```
[ ]: early_stopping = EarlyStopping(monitor = ' accuracy ' ,  
                                patience = 3)  
reduce_lr = ReduceLROnPlateau(monitor = ' accuracy ' ,  
                              patience = 5,  
                              factor = 0.5 ,min_lr = 0.00001 )  
  
callback = [ reduce_lr,early_stopping ]
```

```
[ ]: model . fit_generator(xtrain,  
                        steps_per_epoch = len ( xtrain),  
                        callbacks =callback,  
                        epochs =100 )
```

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:4: UserWarning: `Model.fit_generator` is deprecated and will be removed in a future version.

Please use `Model.fit`, which supports generators. after removing the
cwd from sys.path.

Epoch 1/100

44/44 [=====] - 31s 703ms/step - loss: 0.6074 accuracy: 0.7751 - lr: 0.0010

Epoch 2/100

44/44 [=====] - 31s 702ms/step - loss: 0.5491 accuracy: 0.7973 - lr: 0.0010

Epoch 3/100

44/44 [=====] - 31s 696ms/step - loss: 0.5417 accuracy: 0.8043 - lr: 0.0010

Epoch 4/100

44/44 [=====] - 31s 693ms/step - loss: 0.4930 accuracy: 0.8156 - lr: 0.0010

Epoch 5/100

44/44 [=====] - 31s 692ms/step - loss: 0.4616 accuracy: 0.8293 - lr: 0.0010

Epoch 6/100

44/44 [=====] - 31s 695ms/step - loss: 0.4350 accuracy: 0.8392 - lr: 0.0010

Epoch 7/100

44/44 [=====] - 31s 700ms/step - loss: 0.4190 accuracy: 0.8469 - lr: 0.0010

Epoch 8/100

44/44 [=====] - 31s 692ms/step - loss: 0.3975 accuracy: 0.8568 - lr: 0.0010

Epoch 9/100

44/44 [=====] - 31s 689ms/step - loss: 0.4207 accuracy: 0.8432 - lr: 0.0010

Epoch 10/100

44/44 [=====] - 31s 696ms/step - loss: 0.3674 accuracy: 0.8687 - lr: 0.0010

Epoch 11/100

44/44 [=====] - 31s 689ms/step - loss: 0.3267 accuracy: 0.8888 - lr:

0.0010
Epoch 12/100
44/44 [=====] - 31s 699ms/step - loss: 0.3255 accuracy: 0.8877 - lr:
0.0010
Epoch 13/100
44/44 [=====] - 31s 688ms/step - loss: 0.2975 accuracy: 0.8999 - lr:
0.0010
Epoch 14/100
44/44 [=====] - 31s 690ms/step - loss: 0.3144 accuracy: 0.8927 - lr:
0.0010
Epoch 15/100
44/44 [=====] - 31s 697ms/step - loss: 0.2544 accuracy: 0.9120 - lr:
0.0010
Epoch 16/100
44/44 [=====] - 31s 694ms/step - loss: 0.2986 accuracy: 0.8939 - lr:
0.0010
Epoch 17/100
44/44 [=====] - 31s 699ms/step - loss: 0.2570 accuracy: 0.9097 - lr:
0.0010
Epoch 18/100
44/44 [=====] - 31s 691ms/step - loss: 0.2282 accuracy: 0.9205 - lr:
0.0010
Epoch 19/100
44/44 [=====] - 31s 697ms/step - loss: 0.2180 accuracy: 0.9226 - lr:
0.0010
Epoch 20/100
44/44 [=====] - 31s 694ms/step - loss: 0.1995 accuracy: 0.9314 - lr:
0.0010
Epoch 21/100
44/44 [=====] - 31s 690ms/step - loss: 0.1805 accuracy: 0.9340 - lr:
0.0010
Epoch 22/100
44/44 [=====] - 31s 697ms/step - loss: 0.1986 accuracy: 0.9307 - lr:
0.0010
Epoch 23/100
44/44 [=====] - 31s 693ms/step - loss: 0.2023 accuracy: 0.9287 - lr:
0.0010
Epoch 24/100
44/44 [=====] - 31s 695ms/step - loss: 0.1646 accuracy: 0.9465 - lr:
0.0010
Epoch 25/100
44/44 [=====] - 31s 693ms/step - loss: 0.1727 accuracy: 0.9407 - lr:
0.0010
Epoch 26/100
44/44 [=====] - 32s 719ms/step - loss: 0.1285 accuracy: 0.9569 - lr:
0.0010
Epoch 27/100
44/44 [=====] - 31s 703ms/step - loss: 0.1705 accuracy: 0.9400 - lr:
0.0010
Epoch 28/100
44/44 [=====] - 31s 709ms/step - loss: 0.1531 accuracy: 0.9439 - lr:
0.0010

Epoch 29/100
44/44 [=====] - 32s 711ms/step - loss: 0.1413 accuracy: 0.9539 - lr: 0.0010

[]: <keras.callbacks.History at 0x7efc457f6e50>

```
[ ]: model . save( ' flower_cnn.h5 ' )
```

#Test

```
[ ]: import numpy as np  
from tensorflow.keras.preprocessing import image
```

```
[ ]: img = image . load_img( ' /content/testimage.jpg ' ,target_size =( 64 , 64 ))  
img
```

[]:



```
[ ]: h = image . img_to_array(img)  
h
```

```
[ ]: array([[1., 1., 1.], [1., 1., 1.],  
        [1., 1., 1.],  
        ...,  
        [1., 1., 1.],  
        [1., 1., 1.],  
        [1., 1., 1.]],  
       [[1., 1., 1.],  
        [1., 1., 1.],  
        [1., 1., 1.],  
        ...,  
        [1., 1., 1.],  
        [1., 1., 1.],  
        [1., 1., 1.]],  
       [[1., 1., 1.],  
        [1., 1., 1.],  
        [1., 1., 1.],  
        ...,  
        [1., 1., 1.],  
        [1., 1., 1.],  
        [1., 1., 1.]],  
       ...,  
       [[1., 1., 1.],
```

```

        [1., 1., 1.],
        [1., 1., 1.],
        ...,
        [1., 1., 1.],
        [1., 1., 1.],
        [1., 1., 1.]],

[[[1., 1., 1.],
    [1., 1., 1.],
    [1., 1., 1.],
    ...,
    [1., 1., 1.],
    [1., 1., 1.],
    [1., 1., 1.]],

[[[1., 1., 1.],
    [1., 1., 1.],
    [1., 1., 1.],
    ...,
    [1., 1., 1.],
    [1., 1., 1.],
    [1., 1., 1.]]], dtype=float32)

```

```

[]: h= np.expand_dims(h,axis      = 0)
    h

```

```

[]: array([[[[1., 1., 1.], [1., 1., 1.],
             [1., 1., 1.],
             ...,
             [1., 1., 1.],
             [1., 1., 1.],
             [1., 1., 1.]],

[[[1., 1., 1.],
    [1., 1., 1.],
    [1., 1., 1.],
    ...,
    [1., 1., 1.],
    [1., 1., 1.],
    [1., 1., 1.]],

[[[1., 1., 1.],
    [1., 1., 1.],
    [1., 1., 1.],
    ...,
    [1., 1., 1.],
    [1., 1., 1.],
    [1., 1., 1.]]],

...,

```

```
[[1., 1., 1.],  
 [1., 1., 1.],  
 [1., 1., 1.],  
 ...,  
 [1., 1., 1.],  
 [1., 1., 1.],  
 [1., 1., 1.]],
```

```
[[1., 1., 1.],  
 [1., 1., 1.],  
 [1., 1., 1.],  
 ...,  
 [1., 1., 1.],  
 [1., 1., 1.],  
 [1., 1., 1.]],
```

```
[[1., 1., 1.],  
 [1., 1., 1.],  
 [1., 1., 1.],  
 ...,  
 [1., 1., 1.],  
 [1., 1., 1.],  
 [1., 1., 1.]], dtype=float32)
```

```
[ ]: val = list ( xtrain . class_indices . keys()  
val
```

```
[ ]: ['daisy', 'dandelion', 'rose', 'sunflower', 'tulip']
```

```
[ ]: val[np . argmax(model . predict(h))]
```

```
[ ]: 'rose'
```