

### Assignment -3

#### Python Programming

|                     |                 |
|---------------------|-----------------|
| Assignment Date     | 06 October 2022 |
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| Student Roll Number | 812419104706    |
| Maximum Marks       | 2 Marks         |

#### Question-1:

##### 1.DOWNLOAD THE DATASET

##### Solution:

```
from google.colab import drive
drive.mount('/content/drive')
```

##### OUTPUT

Mounted at /content/drive

```
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Convolution2D,MaxPool2D,Flatten,Dense

from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

#### Question-2:

##### 2.IMAGE AUGUMENTATION

##### Solution:

```
train_datagen =
ImageDataGenerator(rescale=1./255,shear_range=0.2,zoom_range=0.2,horizontal_f
lip=True,vertical_flip=True)

test_datagen=ImageDataGenerator(rescale=1./255)

x_train =
train_datagen.flow_from_directory(r"/content/drive/MyDrive/dataset/Training",
target_size=(64,64),batch_size=32,class_mode="categorical")
```

##### OUTPUT

Found 1238 images belonging to 4 classes.

*#Load your images data*

x\_test =

```
test_datagen.flow_from_directory(r"/content/drive/MyDrive/dataset/Testing",target_size=(64,64),batch_size=32,class_mode="categorical")
```

### OUTPUT

Found 326 images belonging to 4 classes.

```
x_train.class_indices
```

### OUTPUT

```
{'bears': 0, 'crows': 1, 'elephants': 2, 'rats': 3}
```

### Question-3:

#### 3.CREATE MODEL

##### **Solution:**

```
#initialize the model
model=Sequential()
```

### Question-4:

#### 4.ADD LAYERS(Convolution,MxPooling,Flatten,Dense-(Hidden Layers),Output)

##### **Solution:**

```
#add convolution layer
model.add(Convolution2D(32,(3,3),input_shape=(64,64,3),activation='relu'))

#add max pooling layer
model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Flatten())

#hidden layers
model.add(Dense(units=300,kernel_initializer="random_uniform",activation="relu"))
model.add(Dense(units=200,kernel_initializer="random_uniform",activation="relu"))

#output layer
model.add(Dense(units=4,kernel_initializer="random_uniform",activation="softmax"))
```

### Question-5:

#### 5.COMPILE THE MODEL

##### **Solution:**

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

## Question-6:

### 6.FIT THE MODEL

#### Solution:

```
model.fit_generator(x_train, steps_per_epoch=39, epochs=25, validation_data=x_test, validation_steps=10)
```

#### OUTPUT

```
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:1: UserWarning:
`Model.fit_generator` is deprecated and will be removed in a future version.
Please use `Model.fit`, which supports generators.
```

```
"""Entry point for launching an IPython kernel.
```

```
Epoch 1/25
```

```
39/39 [=====] - 213s 5s/step - loss: 1.3571 - accuracy: 0.3086 - val_loss: 1.2797 - val_accuracy: 0.3844
```

```
Epoch 2/25
```

```
39/39 [=====] - 31s 796ms/step - loss: 1.2132 - accuracy: 0.4338 - val_loss: 0.9831 - val_accuracy: 0.5469
```

```
Epoch 3/25
```

```
39/39 [=====] - 31s 794ms/step - loss: 0.9853 - accuracy: 0.5792 - val_loss: 0.8243 - val_accuracy: 0.6500
```

```
Epoch 4/25
```

```
39/39 [=====] - 31s 790ms/step - loss: 0.8966 - accuracy: 0.6284 - val_loss: 0.7700 - val_accuracy: 0.6781
```

```
Epoch 5/25
```

```
39/39 [=====] - 31s 793ms/step - loss: 0.8226 - accuracy: 0.6656 - val_loss: 0.6223 - val_accuracy: 0.7656
```

```
Epoch 6/25
```

```
39/39 [=====] - 31s 800ms/step - loss: 0.7507 - accuracy: 0.6922 - val_loss: 0.5325 - val_accuracy: 0.8344
```

```
Epoch 7/25
```

```
39/39 [=====] - 31s 796ms/step - loss: 0.7334 - accuracy: 0.6931 - val_loss: 0.6391 - val_accuracy: 0.7563
```

```
Epoch 8/25
```

```
39/39 [=====] - 31s 800ms/step - loss: 0.6739 - accuracy: 0.7246 - val_loss: 0.4539 - val_accuracy: 0.8188
```

```
Epoch 9/25
```

```
39/39 [=====] - 31s 795ms/step - loss: 0.6430 - accuracy: 0.7528 - val_loss: 0.5661 - val_accuracy: 0.7250
```

```
Epoch 10/25
```

```
39/39 [=====] - 31s 793ms/step - loss: 0.5744 - accuracy: 0.7617 - val_loss: 0.3414 - val_accuracy: 0.8875
```

```
Epoch 11/25
```

```
39/39 [=====] - 31s 792ms/step - loss: 0.5035 - accuracy: 0.8013 - val_loss: 0.5984 - val_accuracy: 0.7781
```

```
Epoch 12/25
```

```
39/39 [=====] - 31s 790ms/step - loss: 0.4987 -
```

```
accuracy: 0.8053 - val_loss: 0.3194 - val_accuracy: 0.8781
Epoch 13/25
39/39 [=====] - 31s 794ms/step - loss: 0.4479 -
accuracy: 0.8183 - val_loss: 0.2687 - val_accuracy: 0.8906
Epoch 14/25
39/39 [=====] - 31s 793ms/step - loss: 0.3554 -
accuracy: 0.8740 - val_loss: 0.2047 - val_accuracy: 0.9312
Epoch 15/25
39/39 [=====] - 31s 796ms/step - loss: 0.3572 -
accuracy: 0.8667 - val_loss: 0.3596 - val_accuracy: 0.8313
Epoch 16/25
39/39 [=====] - 31s 791ms/step - loss: 0.3545 -
accuracy: 0.8708 - val_loss: 0.1499 - val_accuracy: 0.9625
Epoch 17/25
39/39 [=====] - 31s 794ms/step - loss: 0.3031 -
accuracy: 0.8885 - val_loss: 0.1655 - val_accuracy: 0.9406
Epoch 18/25
39/39 [=====] - 31s 794ms/step - loss: 0.3006 -
accuracy: 0.8990 - val_loss: 0.1121 - val_accuracy: 0.9656
Epoch 19/25
39/39 [=====] - 31s 796ms/step - loss: 0.2436 -
accuracy: 0.9063 - val_loss: 0.0975 - val_accuracy: 0.9563
Epoch 20/25
39/39 [=====] - 31s 793ms/step - loss: 0.2332 -
accuracy: 0.9233 - val_loss: 0.0822 - val_accuracy: 0.9844
Epoch 21/25
39/39 [=====] - 31s 788ms/step - loss: 0.1828 -
accuracy: 0.9346 - val_loss: 0.0978 - val_accuracy: 0.9625
Epoch 22/25
39/39 [=====] - 31s 791ms/step - loss: 0.2079 -
accuracy: 0.9330 - val_loss: 0.2019 - val_accuracy: 0.9312
Epoch 23/25
39/39 [=====] - 31s 796ms/step - loss: 0.1691 -
accuracy: 0.9410 - val_loss: 0.0647 - val_accuracy: 0.9781
Epoch 24/25
39/39 [=====] - 31s 798ms/step - loss: 0.1361 -
accuracy: 0.9491 - val_loss: 0.0550 - val_accuracy: 0.9750
Epoch 25/25
39/39 [=====] - 31s 795ms/step - loss: 0.1839 -
accuracy: 0.9346 - val_loss: 0.1726 - val_accuracy: 0.9312

<keras.callbacks.History at 0x7f42189f8dd0>
```

### Question-7:

#### 7.SAVE THE MODEL

#### Solution:

```
model.save("animal.h5")
```

## Question-8

### 8.TEST THE MODEL

#### Solution:

*#CNN prediction*

```
from tensorflow.keras.models import load_model

from tensorflow.keras.preprocessing import image

import numpy as np

model = load_model('animal.h5')
img = image.load_img('/content/drive/MyDrive/dataset/Testing/crows/Z1
(28).jpg',target_size=(64,64))

img
```

#### OUTPUT



```
type(img)
```

#### OUTPUT

```
PIL.Image.Image
```

```
x=image.img_to_array(img)
```

```
x
```

#### OUTPUT

```
array([[230., 238., 240.],
       [235., 239., 242.],
       [235., 239., 242.],
       ...,
       [241., 242., 244.],
       [242., 241., 246.],
       [242., 241., 246.]],

      [[234., 238., 241.],
       [235., 239., 242.],
       [235., 239., 242.],
       ...,
       [240., 241., 243.],
       [241., 240., 245.]])
```

```

        [242., 241., 246.]],

[[234., 238., 241.],
 [234., 238., 241.],
 [234., 238., 241.],
 ...,
 [242., 241., 246.],
 [242., 242., 244.],
 [242., 242., 244.]],

...,

[[136., 97., 30.],
 [147., 112., 56.],
 [168., 128., 59.],
 ...,
 [161., 122., 53.],
 [159., 124., 58.],
 [171., 132., 63.]],

[[136., 99., 29.],
 [147., 112., 44.],
 [176., 132., 71.],
 ...,
 [166., 128., 65.],
 [164., 126., 53.],
 [176., 131., 64.]],

[[148., 109., 50.],
 [151., 115., 55.],
 [191., 143., 79.],
 ...,
 [168., 130., 67.],
 [156., 122., 48.],
 [160., 121., 46.]]], dtype=float32)

```

x.shape

### **OUTPUT**

(64, 64, 3)

x=np.expand\_dims(x,axis=0)

pred\_prob=model.predict(x)

pred\_prob

### **OUTPUT**

```
array([[0., 1., 0., 0.]], dtype=float32)
class_name=['Bear', 'Crow', 'Elephant', 'Rat']
pred_id=pred_prob.argmax(axis=1)[0]
pred_id
```

#### **OUTPUT**

```
1
```

```
print('Predicted animal is',str(class_name[pred_id]))
```

#### **OUTPUT**

```
Predicted animal is Crow
```