#### **Assignment -3**

Assignment Date	19 September 2022
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Maximum Marks	2 Marks

### 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.

### 2.IMAGE AUGUMENTATION

Solution:

train\_datagen =

 $Image Data Generator (rescale = 1./255, shear\_range = 0.2, zoom\_range = 0.2, hear\_range = 0.2, toom\_range = 0.2, hear\_range = 0.2, hear\_$ 

orizontal\_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}
```

### **3.CREATE MODEL**

Solution:

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

# ${\bf 4.ADD\ LAYERS (Convolution, MxPooling, Flatten, Dense-(Hidden))} \\$

# Layers), Output)

activation="rel u"))

```
Solution:
```

```
model.add(Convolution2D(32,(3,3),input_shape=(64,64,3),activation='relu
'))
model.add(MaxPooling2D(pool_size=(2,2)))
model.add(Flatten())
model.add(Dense(units=300,kernel_initializer="random_uniform",
activation="rel u"))
```

model.add(Dense(units=200,kernel\_initializer="random\_uniform",

model.add(Dense(units=4,kernel\_initializer="random\_uniform",activation = "softm ax"))

#### 5.COMPILE THE MODEL

Solution:

model.compile(loss="categorical\_crossentropy",optimizer="adam",metric
s=['accu racy'])

### **6.FIT THE MODEL Solution:**

model.fit\_generator(x\_train,steps\_per\_epoch=39,epochs=25,validation\_d ata=x\_te st,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.

0.8966 - accuracy: 0.6284 - val\_loss: 0.7700 - val\_accuracy: 0.6781 Epoch

```
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
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
795ms/step - loss: 0.6430 - accuracy: 0.7528 - val_loss: 0.5661 -
val_accuracy: 0.7250 Epoch 10/25 39/39
0.5744 - accuracy: 0.7617 - val_loss: 0.3414 - val_accuracy: 0.8875 Epoch
792ms/step - loss: 0.5035 - accuracy: 0.8013 - val_loss: 0.5984 -
val_accuracy: 0.7781 Epoch 12/25 39/39
0.4987 -accuracy: 0.8053 - val_loss: 0.3194 - val_accuracy: 0.8781 Epoch 13/25
accuracy: 0.8183 - val loss: 0.2687 - val accuracy: 0.8906 Epoch 14/25 39/39
accuracy: 0.8740 - val loss: 0.2047 - val accuracy: 0.9312 Epoch 15/25 39/39
accuracy: 0.8667 - val loss: 0.3596 - val accuracy: 0.8313 Epoch 16/25 39/39
```

```
accuracy: 0.8708 - val loss: 0.1499 - val accuracy: 0.9625 Epoch 17/25 39/39
accuracy: 0.8885 - val loss: 0.1655 - val accuracy: 0.9406 Epoch 18/25 39/39
accuracy: 0.8990 - val loss: 0.1121 - val accuracy: 0.9656 Epoch 19/25 39/39
accuracy: 0.9063 - val loss: 0.0975 - val accuracy: 0.9563 Epoch 20/25 39/39
accuracy: 0.9233 - val loss: 0.0822 - val accuracy: 0.9844 Epoch 21/25 39/39
accuracy: 0.9346 - val loss: 0.0978 - val accuracy: 0.9625 Epoch 22/25 39/39
accuracy: 0.9330 - val loss: 0.2019 - val accuracy: 0.9312 Epoch 23/25 39/39
accuracy: 0.9410 - val loss: 0.0647 - val accuracy: 0.9781 Epoch 24/25 39/39
accuracy: 0.9491 - val loss: 0.0550 - val accuracy: 0.9750 Epoch 25/25 39/39
accuracy: 0.9346 - val loss: 0.1726 - val accuracy: 0.9312
```

### 7.SAVE THE MODEL

Solution:

model.save("animal.h5")

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**



# **OUTPUT**

PIL.Image.Image

x=image.img\_to\_array(img)

Χ

### **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