

### Assignment -3

Assignment Date	9 October 2022
Student Name	SANDHIEEP RAAJHAN G P
Student Roll Number	717819P331
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 =  
ImageDataGenerator(rescale=1./255, shear_range=0.2, zoom_range=0.2, h  
orizontal_flip=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()
```

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

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="relu"))  
model.add(Dense(units=200,kernel_initializer="random_uniform",  
activation="relu"))
```

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

## 5.COMPILE THE MODEL

Solution:

```
model.compile(loss="categorical_crossentropy",optimizer="adam",metric
s=['accuracy'])
```

## 6.FIT THE MODEL Solution:

```
model.fit_generator(x_train,steps_per_epoch=39,epochs=25,validation_d
ata=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

## 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)

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
```