

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/Trai  
ning", 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/Testi  
ng",  
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",metrics=['accuracy'])
```

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/ipynb_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



```
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
```