

## 1.DOWNLOAD THE DATASET

```
In [ ]: from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

```
In [ ]: from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Convolution2D,MaxPool2D,Flatten,Dense
```

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

## 2.IMAGE AUGUMENTATION

```
In [ ]: train_datagen = ImageDataGenerator(rescale=1./255,shear_range=0.2,zoom_range=0.2,horizontal_fli
```

```
In [ ]: test_datagen=ImageDataGenerator(rescale=1./255)
```

```
In [ ]: x_train = train_datagen.flow_from_directory(r"/content/drive/MyDrive/dataset/Training",target_s
```

Found 1238 images belonging to 4 classes.

```
In [ ]: #load your images data

x_test = test_datagen.flow_from_directory(r"/content/drive/MyDrive/dataset/Testing",target_size
```

Found 326 images belonging to 4 classes.

```
In [ ]: x_train.class_indices
```

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

## 3.CREATE MODEL

```
In [ ]: #initialize the model
model=Sequential()
```

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

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

```
In [ ]: #add max pooling layer
model.add(MaxPooling2D(pool_size=(2,2)))
```

```
In [ ]: model.add(Flatten())
```

```
In [ ]: #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"))
```

## 5.COMPILE THE MODEL

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

## 6.FIT THE MODEL

```
In [ ]: model.fit_generator(x_train, steps_per_epoch=39, epochs=25, validation_data=x_test, validation_steps=1)
```

/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: 0.3596 - val\_accuracy: 0.8313

```

_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
Out[ ]: <keras.callbacks.History at 0x7f42189f8dd0>

```

## 7.SAVE THE MODEL

```
In [ ]: model.save("animal.h5")
```

## 8.TEST THE MODEL

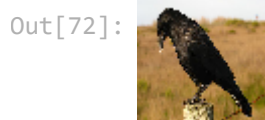
```
In [85]: #CNN prediction
from tensorflow.keras.models import load_model
```

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

```
In [ ]: import numpy as np
```

```
In [71]: model = load_model('animal.h5')
img = image.load_img('/content/drive/MyDrive/dataset/Testing/crows/Z1 (28).jpg',target_size=(64,64))
```

```
In [72]: img
```



```
In [73]: type(img)
```

```
Out[73]: PIL.Image.Image
```

```
In [75]: x=image.img_to_array(img)
```

```
In [76]: x
```

```
Out[76]: 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)
```

```
In [77]: x.shape
```

```
Out[77]: (64, 64, 3)
```

```
In [78]: x=np.expand_dims(x,axis=0)
```

```
In [79]: pred_prob=model.predict(x)
```

```
In [80]: pred_prob
```

```
Out[80]: array([[0., 1., 0., 0.]], dtype=float32)
```

```
In [81]: class_name=['Bear','Crow','Elephant','Rat']
```

```
In [82]: pred_id=pred_prob.argmax(axis=1)[0]
```

```
In [83]: pred_id
```

```
Out[83]: 1
```

```
In [84]: print('Predicted animal is',str(class_name[pred_id]))
```

```
Predicted animal is Crow
```

```
In [ ]:
```