import keras

from keras.preprocessing.image import ImageDataGenerator

In [2]:

#Define the parameters/arguments for ImageDataGenerator class train_datagen=ImageDataGenerator(rescale=1./255,shear_range=0.2,rotation_range=180,zoom_range=0.2,horizontal_flip=**True**)

test_datagen=ImageDataGenerator(rescale=1./255)

In [6]:

#Applying ImageDataGenerator functionality to trainset x_train=train_datagen.flow_from_directory('/content/Dataset/Dataset/train_set',t arget_size=(128,128),batch_size=32,class_mode='binary')

Found 436 images belonging to 2 classes.

In [7]:

#Applying ImageDataGenerator functionality to testset
x_test=test_datagen.flow_from_directory('/content/Dataset/Dataset/test_set',targ
et_size=(128,128),batch_size=32,class_mode='binary')

Found 121 images belonging to 2 classes.

In [8]:

#import model building libraries

#To define Linear initialisation import Sequential

from keras.models import Sequential

#To add layers import Dense

from keras.layers import Dense

#To create Convolution kernel import Convolution2D

from keras.layers import Convolution2D

#import Maxpooling layer

from keras.layers import MaxPooling2D

#import flatten layer

from keras.layers import Flatten

import warnings

warnings.filterwarnings('ignore')

In [9]:

#initializing the model model=Sequential()

In [10]:

#add convolutional layer

model.add(Convolution2D(32,(3,3),input_shape=(128,128,3),activation='relu')) #add maxpooling layer

model.add(MaxPooling2D(pool_size=(2,2)))

```
#add flatten layer
model.add(Flatten())
                                          In [11]:
#add hidden layer
model.add(Dense(150,activation='relu'))
#add output layer
model.add(Dense(1,activation='sigmoid'))
                                          In [12]:
#configure the learning process
model.compile(loss='binary_crossentropy',optimizer="adam",metrics=["accurac
y"])
                                          In [13]:
#Training the model
model.fit_generator(x_train,steps_per_epoch=14,epochs=10,validation_data=x_
test, validation_steps=4)
Epoch 1/10
accuracy: 0.7156 - val_loss: 0.3046 - val_accuracy: 0.9256
Epoch 2/10
accuracy: 0.8899 - val_loss: 0.0900 - val_accuracy: 0.9669
Epoch 3/10
accuracy: 0.8830 - val loss: 0.0665 - val accuracy: 0.9752
Epoch 4/10
accuracy: 0.9083 - val_loss: 0.0653 - val_accuracy: 0.9835
Epoch 5/10
accuracy: 0.9106 - val_loss: 0.0727 - val_accuracy: 0.9752
Epoch 6/10
accuracy: 0.9335 - val_loss: 0.0804 - val_accuracy: 0.9669
Epoch 7/10
accuracy: 0.9335 - val_loss: 0.0777 - val_accuracy: 0.9669
Epoch 8/10
accuracy: 0.9335 - val loss: 0.0795 - val accuracy: 0.9669
Epoch 9/10
accuracy: 0.9381 - val loss: 0.0851 - val accuracy: 0.9752
```

Epoch 10/10	
14/14 [====================================	- 24s 2s/step - loss: 0.1690 -
accuracy: 0.9289 - val_loss: 0.0647 - val_accuracy	y: 0.9752
	Out[13]:
	In [14]:
model.save("forest1.h5")	
	In []: