In	[1]:	<pre>import cv2 import numpy as np from keras.preprocessing import image_dataset_from_directory from keras.models import Sequential from keras.layers import Dense, Flatten, Dropout, GlobalAveragePooling2D from keras.layers import Conv2D from keras.optimizers.legacy import Adam from keras.layers import MaxPooling2D import os from keras.applications import DenseNet121</pre>			
In	[2]:	<pre>train_dir = 'C:/Users/VISHNU VARDHAN/Downloads/AIML PROJECT/Train' test_dir = 'C:/Users/VISHNU VARDHAN/Downloads/AIML PROJECT/Test'</pre>			
In	[3]:	<pre>crime_types = os.listdir(train_dir) n = len(crime_types) print("Number of crime categories: ", n)</pre>			
		Number of crime categories: 14	Į.		
In	[ ]:				
In	[ ]:				
In	[ ]:				
In	[4]:	<pre>train_gen = image_dataset_from_directory(     train_dir, image_size = (64,64), batch_size = 128, label_mode = "categorical", shu:     validation_split=0.2, subset = 'training', )</pre>			
		Found 1266345 files belonging to 14 classes. Using 1013076 files for training.			
In	[5]:	<pre>val_set = image_dataset_from_directory(     train_dir, image_size = (64,64), batch_size = 128, label_mode = "categorical", shu:     validation_split=0.2, subset = 'validation', )</pre>			
		Found 1266345 files belonging to 14 classes. Using 253269 files for validation.			
In	[6]:	<pre>test_gen = image_dataset_from_directory(     test_dir, image_size = (64,64), batch_size = 128,label_mode = "categorical",shuff] )</pre>			
In	[7]:	Found 111308 files belonging to	o 14 classes.		
<u> </u>	[/] *	<pre>def transfer_learning():     base_model = DenseNet121(include_top=False, input_shape = (64,64,3), weights='imag</pre>			
		<pre>thr = 149 for layers in base_model.layers[:thr]:     layers.trainable=False for layers in base_model.layers[:thr]:     layers.trainable=False return base_model</pre>			
In	[8]:	<pre>def create_model():</pre>			
		<pre>model = Sequential()  base_model = transfer_learning()</pre>			
		<pre>model.add(base_model) model.add(GlobalAveragePod</pre>	ling2D())		
		<pre>model.add(Dense(256, activ model.add(Dropout(0.2))</pre>	ration='relu'))		
		<pre>model.add(Dense(512, activ model.add(Dropout(0.2))</pre>	ration='relu'))		
		model.add(Dense(1024, acti			
		model.summary	Ton- Softmax //		
		return model			
In	[9]:	<pre>model = create_model() model.compile(optimizer='adam'</pre>	, loss='categoric	al_crossentropy', me	etrics = ['accuracy
In [	10]:	model.summary()			
		Model: "sequential"  Laver (type) O	itput Shape	Param #	
		Layer (type) Ou	Itput Shape 	Param # 7037504	
		Layer (type) Oudensenet121 (Functional) (Name of the property	Jone, 2, 2, 1024)	7037504	
		Layer (type) Oudensenet121 (Functional) (National)	None, 2, 2, 1024)	7037504	
		Layer (type)  densenet121 (Functional) (National) (National)  global_average_pooling2d (National) (National)  dense (Dense) (National) (National)  dense (Dense) (National) (National) (National)	Jone, 256)	7037504	
		Layer (type)  densenet121 (Functional) (National) (Nati	Jone, 256) Jone, 256) Jone, 512) Jone, 512) Jone, 1024)	7037504 0 262400 0 131584 0 525312	
		Layer (type)  densenet121 (Functional) (Note: The content of the c	Jone, 2, 2, 1024) Jone, 1024) Jone, 256) Jone, 256) Jone, 512) Jone, 512) Jone, 1024) Jone, 1024) Jone, 14)	7037504 0 262400 0 131584 0 525312 14350	
		Layer (type)  densenet121 (Functional) (National) (Nati	Jone, 256) Jone, 512) Jone, 1024) Jone, 512) Jone, 1024) Jone, 1024) Jone, 1024) Jone, 14)	7037504 0 262400 0 131584 0 525312 14350	
In	[]:	Layer (type) Outer densent 121 (Functional) (Note the property of the params: 7971150 (30.41 MET Trainable params: 6386894 (24.3)	Jone, 2, 2, 1024) Jone, 256) Jone, 256) Jone, 512) Jone, 512) Jone, 1024) Jone, 1024) Jone, 104) Jone, 14) Jone, 14)	7037504 0 262400 0 131584 0 525312 14350	
	[]:	Layer (type)  densenet121 (Functional) (National) (Nati	Jone, 2, 2, 1024) Jone, 256) Jone, 256) Jone, 512) Jone, 512) Jone, 512) Jone, 1024) Jone, 14) J	7037504 0 262400 0 131584 0 525312 14350	- accuracy: 0.9779
In	[]:	Layer (type) Outer densent 121 (Functional) (Name of the proposition o	Jone, 2, 2, 1024) Jone, 256) Jone, 256) Jone, 512) Jone, 512) Jone, 512) Jone, 1024) Jone, 14) J	7037504 0 262400 0 131584 0 525312 14350	- accuracy: 0.9779
In In	[]:	Layer (type)  densenet121 (Functional) (National) (Nati	Jone, 2, 2, 1024) Jone, 256) Jone, 256) Jone, 512) Jone, 512) Jone, 512) Jone, 1024) Jone, 14) J	7037504 0 262400 0 131584 0 525312 14350	- accuracy: 0.9779
In In	[]:	Layer (type)  densenet121 (Functional) (National) (Nati	Jone, 2, 2, 1024) Jone, 256) Jone, 256) Jone, 512) Jone, 512) Jone, 1024) Jone, 1024) Jone, 14) Jone, 16, MB) Jone, 16, MB) Jone, 16, MB Jone, 17, MB Jone, 17, MB Jone, 17, MB Jone, 1024) Jone, 1024) Jone, 512) Jone, 512) Jone, 512) Jone, 512) Jone, 1024) Jone, 512) Jone, 1024) Jone, 512) Jone, 1024) Jone, 1024 Jone	7037504 0 262400 0 131584 0 525312 14350 epochs = 1) A: 0s - loss: 0.0846	- accuracy: 0.9779
In In	[]:	Layer (type)  densenet121 (Functional) (National) (Nati	Jone, 2, 2, 1024) Jone, 1024) Jone, 256) Jone, 256) Jone, 512) Jone, 1024) Jone, 1024) Jone, 14) Jone, 16 Jone, 17	7037504 0 262400 0 131584 0 525312 14350 epochs = 1) A: 0s - loss: 0.0846	
In In In	[]:	Layer (type) One densenet121 (Functional) (Note of the property of the propert	Jone, 2, 2, 1024) Jone, 1024) Jone, 256) Jone, 256) Jone, 512) Jone, 512) Jone, 1024) Jone, 14) Jone, 1024) Jone, 1024) Jone, 1024) Jone, 512) Jone, 1024) Jone, 1024 Jone,	7037504 0 262400 0 131584 0 525312 14350 epochs = 1) A: 0s - loss: 0.0846	",compile=False)
In In In	[]:	Layer (type)  densenet121 (Functional) (National) (Nati	Jone, 2, 2, 1024) Jone, 1024) Jone, 256) Jone, 256) Jone, 512) Jone, 512) Jone, 1024) Jone, 14) Jone, 1024) Jone, 1024) Jone, 1024) Jone, 512) Jone, 1024) Jone, 1024 Jone,	7037504 0 262400 0 131584 0 525312 14350 epochs = 1) A: 0s - loss: 0.0846	",compile=False)
In In In In	[ ]: [ ]: [ ]:	Layer (type)  densenet121 (Functional) (Reconstructional) (Reconstruct	Jone, 2, 2, 1024) Jone, 1024) Jone, 256) Jone, 256) Jone, 512) Jone, 512) Jone, 1024) Jone, 14) Jone, 1024) Jone, 1024) Jone, 1024) Jone, 512) Jone, 1024) Jone, 1024 Jone,	7037504 0 262400 0 131584 0 525312 14350 epochs = 1) A: 0s - loss: 0.0846	",compile=False)
In In In In In	[ ]: [ ]: [ ]:	Layer (type) One densenet121 (Functional) (Note of the property of the propert	Jone, 2, 2, 1024) Jone, 1024) Jone, 256) Jone, 256) Jone, 512) Jone, 512) Jone, 1024) Jone, 14) Jone, 1024) Jone, 1024) Jone, 1024) Jone, 512) Jone, 1024) Jone, 1024 Jone,	7037504 0 262400 0 131584 0 525312 14350 epochs = 1) A: 0s - loss: 0.0846	",compile=False)
In In In In In In	[ ]: [ ]: [ ]:	Layer (type)  dense===================================	Jone, 2, 2, 1024) Jone, 1024) Jone, 256) Jone, 256) Jone, 512) Jone, 512) Jone, 1024) Jone, 14) Jone, 1024) Jone, 1024) Jone, 1024) Jone, 512) Jone, 1024) Jone, 1024 Jone,	7037504 0 262400 0 131584 0 525312 14350 epochs = 1) A: 0s - loss: 0.0846	",compile=False)
In In In In In In In		Layer (type)  densenet121 (Functional) (Recomposed to the second of the	Jone, 2, 2, 1024) Jone, 1024) Jone, 256) Jone, 256) Jone, 512) Jone, 512) Jone, 1024) Jone, 14) Jone, 1024) Jone, 1024) Jone, 1024) Jone, 512) Jone, 1024) Jone, 1024 Jone,	7037504 0 262400 0 131584 0 525312 14350 epochs = 1) A: 0s - loss: 0.0846	",compile=False)
In In In In In In In In In		Layer (type)  densenet121 (Functional) (Notes and the properties of the properties o	Jone, 2, 2, 1024) Jone, 1024) Jone, 256) Jone, 256) Jone, 512) Jone, 512) Jone, 1024) Jone, 14) Jone, 1024) Jone, 1024) Jone, 1024) Jone, 512) Jone, 1024) Jone, 1024 Jone,	7037504 0 262400 0 131584 0 525312 14350 epochs = 1) A: 0s - loss: 0.0846	",compile=False)
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In		Layer (type)  densenet121 (Functional) (Notes and the property of the property	Jone, 2, 2, 1024) Jone, 1024) Jone, 256) Jone, 256) Jone, 512) Jone, 512) Jone, 1024) Jone, 14) Jone, 1024) Jone, 1024) Jone, 1024) Jone, 1024) Jone, 512) Jone, 512 J	7037504  0  262400  0  131584  0  525312  14350  epochs = 1)  1: 0s - loss: 0.0846  eray  HNU VARDHAN/crime.h5  wnloads/Fighting.png	osion':5, 'Fighting
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