Emerging methods for early detection of forest fire

MODEL BUILDING

TRAINING THE MODEL

Team ID	PNT2022TMID16960
Project Name	Emerging methods for early detection of forest fire

TRAINING THE MODEL:

At this point, we have training data and a fully configured neural network to train. All that is left is to pass the data to the model for the training process to commence, a process that is completed by iterating on the training data. Training begins by calling the fit () method.

The arguments are the batch size as you are using "adam" (bath gradient descent and epochs: no: of times the model should get trained). steps_per_epoch:

- It specifies the total number of steps taken from the generator as soon as one epoch is finished and the next epoch has started. We can calculate the value of steps_per_epoch as the total number of samples in your training folder divided by the batch size.
- Epochs: an integer and number of epochs we want to train our model for.
- Validation_datacan be either input and targets list generator inputs, targets, and sample_weights list which can be used to evaluate.

The loss and metrics for any model after any epoch has ended.

• Validation_steps:

Only if the validation_data is a generator then only this argument can be used. It specifies the total number of steps taken from the generator before it is stopped at every epoch and its value is calculated as the total number of validation data points in your dataset divided by the validation batch size.

IMPORT LIBRARIES:

•	Importing Keras libraries	
	import keras	

Importing ImageDataGenerator from Keras

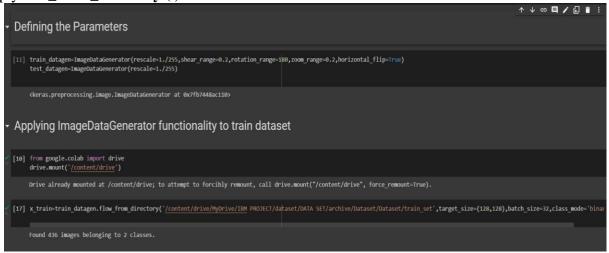
from keras.preprocessing.image import ImageDataGenerator

IMPORT ImageDataGenerator FROM KERAS:



APPLYING ImageDataGenerator to train dataset:

plyflow_from_directory ()methodfor Train folder.



APPLYING ImageDataGenerator to test dataset:

Applying the **flow_from_directory** () methodfortest folder.



IMPORTING MODEL BUILDING LIBRARIES:

Importing Model Building Libraries

```
#to define the linear Initialisation import sequential
from keras.models import Sequential
#to add layers import Dense
from keras.layers import Dense
#to create Convolutional 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')
```

INITIALIZING THE MODEL:

Initializing the model

```
model=Sequential()
```

ADDING CNN LAYERS:

→ Adding CNN Layers

```
model.add(Convolution2D(32,(3,3),input_shape=(128,128,3),activation='relu'))
#add maxpooling layers
model.add(MaxPooling2D(pool_size=(2,2)))
#add faltten layer
model.add(Flatten())
```

ADDING DENSE LAYERS:

Add Dense layers

```
#add hidden layers
model.add(Dense(150,activation='relu'))
#add output layer
model.add(Dense(1,activation='sigmoid'))
```

CONFIGURING THE LEARNING PROCESS:

configuring the learning process

```
model.compile(loss='binary_crossentropy',optimizer="adam",metrics=["accuracy"])
```

TRAINING THE MODEL:

Training the model

```
model.fit_generator(x_train,steps_per_epoch=14,epochs=10,validation_data=x_test,validation_
  Epoch 1/10
  Epoch 2/10
  Epoch 3/10
  14/14 [============= ] - 32s 2s/step - loss: 0.2979 - accuracy: 0.885
  Epoch 5/10
  Epoch 7/10
  Epoch 8/10
  14/14 [==================== ] - 30s 2s/step - loss: 0.1796 - accuracy: 0.924
  Epoch 9/10
  Epoch 10/10
  14/14 [============== ] - 27s 2s/step - loss: 0.2593 - accuracy: 0.889
  <keras.callbacks.History at 0x7fd537101390>
```