# EMERGING METHODS FOR EARLY DETECTION OF FOREST FIRES

## MODEL BUILDING

## TRAINING THE MODEL

Date	06 November 2022
Team ID	PNT2022TMID08411
Project Name	Emerging Methods for Early Detection of Forest Fires

Importing The ImageDataGenerator Library import keras from keras.preprocessing.image import ImageDataGenerator Define the parameters/arguments for ImageDataGenerator class train\_datagen=ImageDataGenerator(rescale=1./255,shear\_range=0.2,rot ati on\_range=180,zoom\_range=0.2, horizontal\_flip=True) test\_datagen=ImageDataGenerator(rescale=1./255) Applying ImageDataGenerator functionality to trainset x\_train=train\_datagen.flow\_from\_directory(r'/content/drive/MyDriv e/Dataset/train\_set',target\_size=(128,128),batch\_size=32,

class\_mode='binary')

Found 436 images belonging to 2 classes.

## Applying ImageDataGenerator functionality to testset

x\_test=test\_datagen.flow\_from\_directory(r'/content/drive/MyDrive / Dataset/test\_set',target\_size=(128,128),batch\_size=32, class\_mode='binary')

Found 121 images belonging to 2 classes.

#### 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

## **Initializing the model**

model=Sequential() Add

warnings.filterwarnings('ignore')

## **CNN 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())
```

## **Add Hidden Layer**

```
#add hidden layer model.add(Dense(150,activation='relu'))
#add output layer
model.add(Dense(1,activation='sigmoid')
Configure the learning process
model.compile(loss='binary_crossentropy',optimizer="adam",metrics=[
"ac curacy"]) Train the model
model.fit_generator(x_train,steps_per_epoch=14,epochs=10,validation
_ da ta=x_test,validation_steps=4)
Epoch 1/10
14/14 [======] - 97s 7s/step - loss:
1.3060 - accuracy: 0.7775 - val_loss: 0.5513 -
val_accuracy: 0.8512
Epoch 2/10
14/14 [=======] - 26s 2s/step - loss:
0.3178 - accuracy: 0.8807 - val_loss: 0.1299 -
val_accuracy: 0.9421
Epoch 3/10
14/14 [=======] - 26s 2s/step - loss:
0.2226 - accuracy: 0.9106 - val loss: 0.1311 -
val_accuracy: 0.9421
Epoch 4/10
0.1836 - accuracy: 0.9174 - val_loss: 0.1129 -
val_accuracy: 0.9339
Epoch 5/10
14/14 [=======] - 30s 2s/step - loss:
0.1675 -
accuracy: 0.9243 - val_loss: 0.0925 - val_accuracy: 0.9669
Epoch 6/10
```

```
0.1884 - accuracy: 0.9289 - val_loss: 0.1287 -
val_accuracy: 0.9339
Epoch 7/10
14/14 [=======] - 28s 2s/step - loss:
0.1724 - accuracy: 0.9335 - val_loss: 0.0926 -
val_accuracy: 0.9752
Epoch 8/10
14/14 [=======] - 26s 2s/step - loss:
0.1510 - accuracy: 0.9404 - val loss: 0.0757 -
val_accuracy: 0.9752 Epoch 9/10
14/14 [=======] - 26s
                                                   0.173 -
2s/step - loss:
                                                   2
accuracy: 0.9174 - val_loss: 0.0537 - val_accuracy: 0.9835
Epoch 10/10
                 - 26s
        0.154 -
2s/step - loss:
                                                   6
accuracy: 0.9312 - val_loss: 0.0573 - val_accuracy: 0.9835
```

<keras.callbacks.History at 0x7f05d66a9c90>