# EMERGING METHODS FOR EARLY DETECTION OF FOREST FIRES

## MODEL BUILDING

#### **SAVE THE MODEL**

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	D) YEAR AREA (A)
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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
warnings.filterwarnings('ignore')
```

# Initializing the model

```
model=Sequential()

Add 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
14/14 [=======] - 26s 2s/step - loss:
```

```
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
0.173 -
accuracy: 0.9174 - val loss: 0.0537 - val accuracy: 0.9835
Epoch 10/10
14/14 [=======] - 26s
                                                0.154 -
2s/step - loss:
accuracy: 0.9312 - val_loss: 0.0573 - val_accuracy: 0.9835
<keras.callbacks.History at 0x7f05d66a9c90>
```

#### Save The Model

model.save("forest1.h5")