# EMERGING METHODS FOR EARLY DETECTION OF FOREST FIRES

## Video Analysis

# **Creating An Account In Twilio Service**

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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, rotation\_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/MyDrive/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 Dense 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
                                  =====] - 205s 15s/step - loss: 2.7344 -
accuracy: 0.7454 - val loss: 0.2016 - val accuracy: 0.9256
Epoch 2/10
                  14/14 [==
val loss: 0.2290 - val accuracy: 0.9339
Epoch 3/10
==] - 20s 1s/step - loss: accuracy: 0.8922 -
val loss: 0.0524 - val accuracy: 0.9835
Epoch 4/10
14/14 [======
                                      ====] - 20s 1s/step - loss: accuracy: 0.9174 -
val loss: 0.1570 - val accuracy: 0.9421
Epoch 5/10
                                       ===1 - 20s 1s/step - loss: accuracy: 0.9083 -
14/14 [=====
val loss: 0.0767 - val accuracy: 0.9752
Epoch 6/10
14/14 [====
                                      ====] - 20s 1s/step - loss: accuracy: 0.9335 -
val loss: 0.0749 - val accuracy: 0.9752
Epoch 7/10
14/14 [===
                                       ===] - 20s 1s/step - loss: accuracy: 0.9312 -
val loss: 0.1264 - val accuracy: 0.9421
Epoch 8/10
14/14 [==
                                        ===] - 20s 1s/step - loss: accuracy: 0.9266 -
val loss: 0.0652 - val accuracy: 0.9835
Epoch 9/10
=] - 20s 1s/step - loss: accuracy: 0.9358 -
val loss: 0.0567 - val accuracy: 0.9835
Epoch 10/10
14/14 [======
                                         =] - 20s 1s/step - loss: accuracy: 0.9404 -
val loss: 0.0448 - val accuracy: 0.9917
0.3267 -
0.2991 -
0.2418 -
0.1984 -
0.1643 -
0.1538 -
0.1732 -
0.1514 -
0.1445 -
<keras.callbacks.History at 0x7f51fdf33610>
```

```
model.save("forest1.h5")
```

#### **Predictions**

```
#import load model from keras.model
from keras.models import load model
#import image class from keras
from tensorflow.keras.preprocessing import image #import numpy import numpy as
np
#import cv2
import cv2
#load the saved model
model = load model("forest1.h5")
img=image.load img(r'/content/drive/MyDrive/Dataset/test set/forest/
0.48007200 1530881924 final forest.jpg')
x=image.img to array(img)
res = cv2.resize(x, dsize=(128, 128), interpolation=cv2.INTER CUBIC) #expand the
image shape
x=np.expand dims(res,axis=0)
pred= model.predict(x)
1/1 [==
                               =======] - 0s 94ms/step pred
array([[0.]], dtype=float32)
OpenCV For Video Processing
pip install twilio
Looking in indexes: https://pypi.org/simple, https://us-
python.pkg.dev/colab-wheels/public/simple/
Collecting twilio
Downloading twilio-7.15.1-py2.py3-none-any.whl (1.4 MB)
ent already satisfied: pytz in /usr/local/lib/python3.7/dist-packages (from twilio)
(2022.5)
Collecting PyJWT<3.0.0,>=2.0.0
Downloading PyJWT-2.6.0-py3-none-any.whl (20 kB)
Requirement already satisfied: requests>=2.0.0 in /usr/local/lib/python3.7/dist-
packages (from twilio) (2.23.0) Requirement already satisfied: chardet<4,>=3.0.2 in
/usr/local/lib/python3.7/dist-packages (from requests>=2.0.0->twilio) (3.0.4)
Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages
(from requests>=2.0.0->twilio)
(2.10)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.7/dist-
packages (from requests>=2.0.0->twilio) (2022.9.24)
```

Requirement already satisfied: urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1 in

```
/usr/local/lib/python3.7/dist-packages (from requests>=2.0.0->twilio) (1.24.3)
Installing collected packages: PyJWT, twilio
Successfully installed PyJWT-2.6.0 twilio-7.15.1
pip install playsound
Looking in indexes: https://pypi.org/simple, https://us-
python.pkg.dev/colab-wheels/public/simple/
Collecting playsound
Downloading playsound-1.3.0.tar.gz (7.7 kB) Building wheels for collected
packages: playsound
Building wheel for playsound (setup.py) ... e=playsound-1.3.0-py3- none-any.whl
size=7035
sha256=e7e96c774a98522e182b59b7b292f0f932097658d8bfce86c922c363f862b0e
Stored in directory:
/root/.cache/pip/wheels/ba/f8/bb/ea57c0146b664dca3a0ada4199b0ecb5f9dfc
b7b7e22b65ba2
Successfully built playsound
Installing collected packages: playsound
Successfully installed playsound-1.3.0
#import opency library
import cv2
#import numpy
import numpy as np
#import image function from keras
from keras.preprocessing import image
#import load model from keras
from keras.models import load model
#import client from twilio API
from twilio.rest import Client
#import playsound package
from playsound import playsound
WARNING:playsound:playsound is relying on another python subprocess. Please
use 'pip install pygobject' if you want playsound to run more efficiently.
#load the saved model
model=load model("forest1.h5") #define video video=cv2.VideoCapture(0) #define
the features name=['forest','with fire']
```

## Creating An Account In Twilio Service

```
account_sid='ACfb4e6d0e7b0d25def63044919f1b96e3'
auth_token='f9ae4fc4a617a527da8672e97eefb2d8'
client=Client(account_sid,auth_token)
message=client.messages \
.create(
    body='Forest Fire is detected, stay alert',
    from_='+1 302 248 4366',
    to='+91 99400 12164'
```

```
)
print(message.sid)
```

SM4aa5a4751b7bcec159dc4c695752293d