# **Emerging Methods for Early Detection of Forest**

A Project report submitted in partial fulfilment of 7th semester in degree **BACHELOR OF ENGINEERING** IN

## COMPUTER SCIENCE AND ENGINEERING

# Submitted by

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**NOV-2022** 

# JAYALAKSHMI INSTITUTE OF TECHNOLOGY, THOPPUR (A Constituent College of Anna University, Chennai)



# **BONAFIDE CERTIFICATE**

Certified that this project report "Emerging Methods for Early Detection of Forest Fires" is the bonafide work done APOORVA.C (610919104009), ILAKIYADHARSHINI.G(610919104029), JAYANTHI.M (610919104031), JAYANTHI.S (610919104032) for Nalayathiran in VII semesterof B.E., degree course in Computer Science and Engineering branch during the academic year of 2022-2023.

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Evaluator R.KARTHIK.ME

Head of the Department MR SATHYA.ME

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APOORVA.C ILAKIYADHARSHINI.G JAYANTHI.M JAYANTHI.S

#### Abstract

A wildland fire is an uncontrolled fire that occurs mainly in forest areas, although it can also invade urban or agricultural areas. Among the main causes of wildfires, human factors, either intentional or accidental, are the most usual ones. The number and impact of forest fires are expected to grow as a consequence of the global warming. In order to fight against these disasters, it is necessary to adopt a comprehensive, multifaceted approach that enables a continuous situational awareness and instant responsiveness. This paper describes a hierarchical wireless sensor network aimed at early fire detection in risky areas, integrated with the fire fighting command centres, geographical information systems, and fire simulators. This configuration has been successfully tested in two fire simulations involving all the key players in fire fighting operations: fire brigades, communication systems, and aerial, coordination, and land means.

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#### 1. INTRODUCTION

Forest fires are a major environmental issue, creating economic and ecological damage while endangering human lives. There are typically about 100,000 wildfires in the United States every year. Over 9 million acres of land have been destroyed due to treacherous wildfires. It is difficult to predict and detect Forest Fire in a sparsely populated forest area and it is more difficult if the prediction is done using ground-based methods like Camera or Video-Based approach. Satellites can be an important source of data prior to and also during the Fire due to its reliability and efficiency. The various real-time forest fire detection and prediction approaches, with the goal of informing the local fire authorities

#### **Project Overview**

The idea is to create aand develop a system that can identify the effects of the forest fire and it can analyse the forest fire by advanced AI techniques and CNN Algorithm then the Prediction model is Checked and then the model is connected with Twilio account credentials of the Developer consisting of phone numbers of the persons in the surroundings of the people in the area of easy forest fire zone then an security sound alert system is developed to make a alert sound which is downloaded from internet then the entire model is deployed to the IBM Cloud account that we have created.

#### **Purpose**

The forest fires destroys the wildlife habitat, damages the environment, affects the climate, spoils the biological properties of the soil, etc. So the forest fire detection is a major issue in the present decade. At the same time the forest fire have to be detected as fast as possible.

#### 2. LITERATURE SURVEY

#### **Existing problem**

Forest fires have been and still are serious problem for the European Union and for all other countries in Europe. In the year 2000, the EU has established the European Forest Fire Information system (EFFIS), which will soon become part of the European Emergency Management Service, maintained by the Copernicus Earth Observation Programme. This system provides valuable near real-time and also historical data on the forest fires in Europe, the Middle East and North Africa. Currently EFFIS is being used and supported with data by 25 EU member states and by numerous other countries. According to the annual report of EFFIS for 2016, more than 54 000 forest fires have occurred all around Europe and they have led to nearly 376 thousand hectares of burnt areas. If we compare these values to the average values from the EFFIS reports for the period 2006-2015, the number of forest fires have decreased by 13327 or by nearly 20%. This decrease can be explained with the more severe actions and sanctions towards the arsonists and with the introduction of more advanced technical solutions for early detection of the fires. Even though their number is decreasing, the forest fires continue to be extremely devastating events and they have destroyed just 27 thousand hectares (or 6.6 %) less than the average burnt areas for the period 2006-2015, according to . Confirmation for this are the devastating forest fires form 2018, which took place in the Attica region of Greece and led to

more than 90 fatalities and to more than 200 injured people, as well as to the destruction to thousands of buildings. Forest Fires can be divided into 4 categories in the forests of Hungary based on tree and other vegetation species: • underground burning, peat fire; • fire in undergrowth or dead fallen leaves; • fire in seedlings and saplings; • fire in trunks and shrouds.

#### References

[1] Official webpage of the European Forest Fire Information System at:

http://effis.jrc.ec.europa.eu/

[2] Official webpage of the Copernicus Earth Observation Programme at:

http://www.copernicus.eu

[3] Forest Fires in Europe, Middle East and North Africa 2016, JRC Science for policy report,

BN 978-92-79-71292-0, ISSN 1831-9424, doi:10.2760/17690, availabe at:

http://effis.jrc.ec.europa.eu/media/cms\_page\_media/40/Forest\_fires\_i

n\_Europe\_Middle\_east\_and\_North\_Africa\_2016\_final\_pdf\_JZU7He L.pdf

[4] The 2018 Attica wildfires Wikipedia webpage available at

https://en.wikipedia.org/wiki/2018\_Attica\_wildfires

[5] Rajmund Kuti,"Characterstic of forest fire and its impact on environment",(2016)

#### **Problem Statement Definition**

The user interacts with a web camera to read the video.

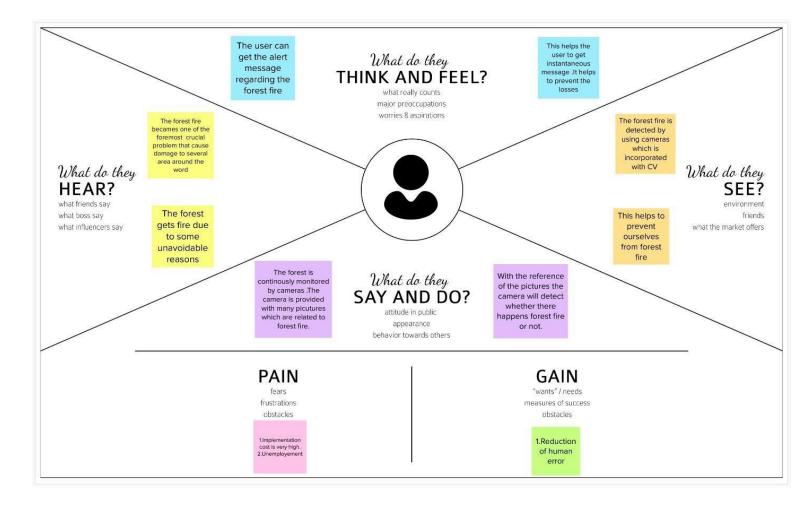
Once the input image from the video frame is sent to the model, if the fire is detected it is showcased on the console, and alerting sound will be generated and an alert message will be sent to the Authorities.

- Data Collection.
  - Collect the dataset or create the dataset.
- Image Preprocessing.
  - Import ImageDataGenerator Library.
  - Define the parameters /arguments for ImageDataGenerator class
  - Applying ImageDataGenerator on trainset and test set.
- Model Building
  - Import the model building Libraries
  - Initializing the model
  - Adding CNN Layers
  - Adding Hidden Layer
  - Adding Output Layer
  - Configure the Learning Process
  - Training and testing the model
  - Optimize the Model
  - Save the Model
- Video Streaming and alerting
  - OpenCV for video processing
  - Creating an account in Twilio service
  - Use Twilio API to send messages.

#### 3. IDEATION AND PROPOSED SOLUTION

## **Empathy Map Canvas**

An empathy map canvas is a more in-depth version of the original empathy map, which helps identify and describe the user's needs and pain points.



#### **Ideation & Brainstorming**

organizing the brainstorming session and prioritize the top 3 ideas based on the feasibility & importance.

Step-1: Team Gathering, Collaboration and Select the Problem Statement



# Brainstorm & idea prioritization

Use this template in your own brainstorming sessions so your team can unleash their imagination and start shaping concepts even if you're not sitting in the same room.

(9 10 minutes to prepare

J; 1 hour to collabora

2-8 people recommended

0

#### Before you collaborate

A little bit of preparation goes a long way with this session. Here's what you need to do to get going.

(!) 10 minutes

O Team gathering

Define who should participate in the session and send an invite. Share relevant information or pre-work ahead

O Set the go

Think about the problem you'll be focusing on solving in the brainstorming session.

Learn how to use the facilitation tools
 Use the Facilitation Superpowers to run a happy and productive session

Open article

0

#### Define your problem statement

What problem are you trying to solve? Frame your problem as a How Might We statement. This will be the focus of your brainstorm.

5 minutes

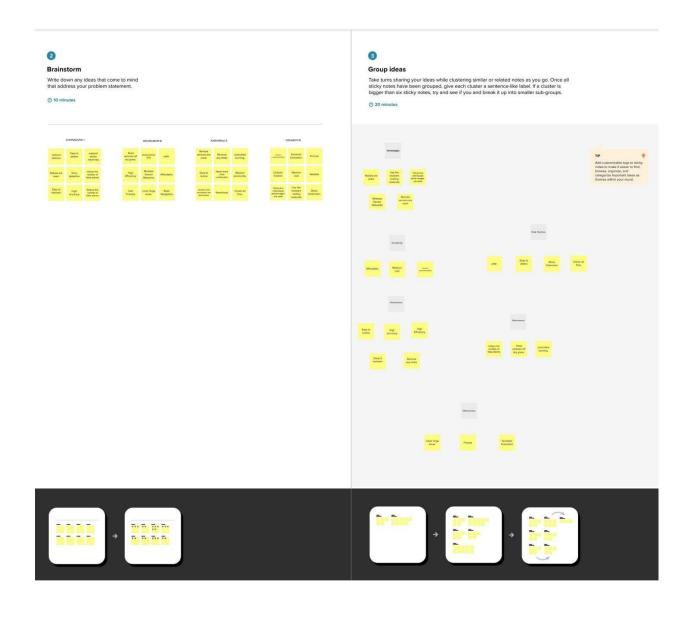
How might we detect the Forest Fires Early to prevent the loss of valuable timber resources?



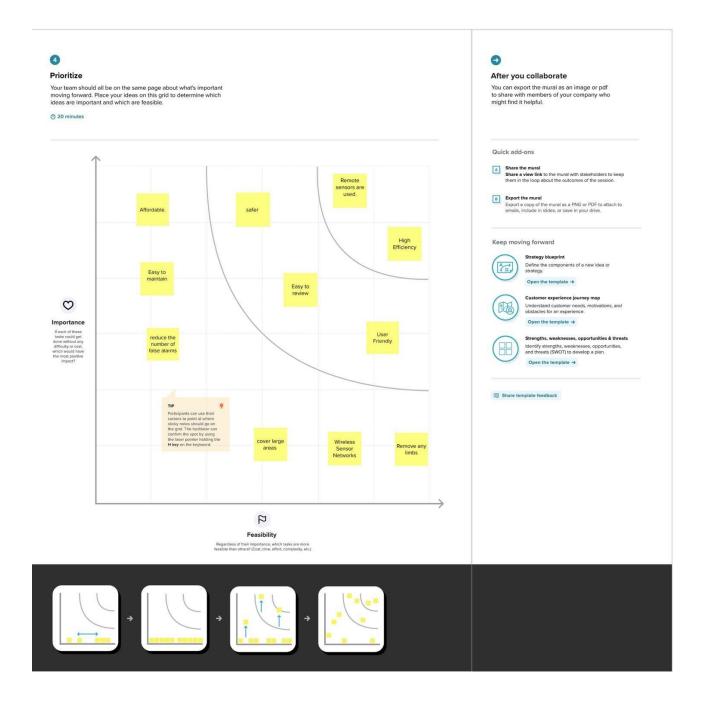
Share template feedback



Step-2: Brainstorm, Idea Listing and Grouping



# Step-3: Idea Prioritization



## **Proposed Solution:**

S.No.	Parameter	Description
1.	Problem Statement (Problem to be solved)	Emerging methods for early detection of forests fires
2.	Idea / Solution description	Forest surveillance video cameras can be used to monitor the forest areas and they can alert the forest department if there is any symptoms of forest fire or any other suspicious activities.
3.	Novelty / Uniqueness	The digital image processing technique, pattern-recognition technology and reinforcement learning can greatly improve the sensing of forest fire and they are much more effective as they improve forecast and reaction time is much less
4.	Social Impact / Customer Satisfaction	This product has huge social and biological impact as prevention of forest fire can save countless acres of forest land and wild lives. Forest fire also increases the amount of CO2 in the atmosphere. So prevention of forest fire can also reduce global warming.
5.	Business Model (Revenue Model)	This product can be only used by a giant corporation or a government to monitor huge reserve forests. This can be considered as a profitable and useful product as government spends millions of dollars for detection of forest fires and millions more if there is any actual forest fire like rescuing and stopping the fire.
6.	Scalability of the Solution	It is highly challenging to implement this method as we have to install hundreds of cameras to cover a respectable amount of area. The cameras need to be connected to electricity and they also need to be connected to internet to process the image and analyze them. They need to be connected to local server which should be located at the middle of the forest.

# 4. REQUIREMENT ANALYSIS

# **Functional requirement**

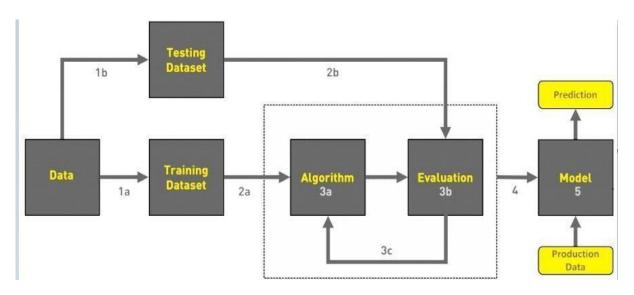
S.No	Characteristics	Description	Technology
1.	Open-Source Frameworks	Python Flask framework is used	Technology of Opensource framework
2.	Security Implementations	Mandatory Access Control (MAC) and	e.g. SHA-256, Encryptions, IAM
		Preventative Security Control is used	Controls, OWASP etc.
3.	Scalable Architecture	High scalability with 3-tier architecture	Web server - HTML ,CSS ,JavaScript
			Application server – Python , Anaconda
			Database server –IBM DB2
4.	Availability	Use of load balancing to distribute traffic across	IBM load balancer
		servers	
5.	Performance	Enhance the performance by using IBM CDN	IBM Content Delivery Network

# **Non-Functional requirements**

S.No	Characteristics	Description	Technology
1.	Open-Source Frameworks	Python Flask framework is used	Technology of Opensource framework
2.	Security Implementations	Mandatory Access Control (MAC) and	e.g. SHA-256, Encryptions, IAM
		Preventative Security Control is used	Controls, OWASP etc.
3.	Scalable Architecture	High scalability with 3-tier architecture	Web server - HTML ,CSS ,JavaScript
			Application server - Python , Anaconda
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4.	Availability	Use of load balancing to distribute traffic across	IBM load balancer
	*	servers	
5.	Performance	Enhance the performance by using IBM CDN	IBM Content Delivery Network

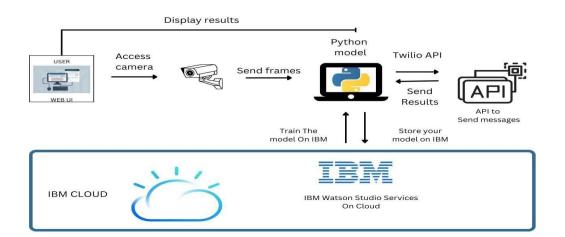
#### 5. PROJECT DESIGN

## **Data Flow Diagrams**



- 1. COLLECT DATA
- 2. EVALUATE DATA SET
- 3. IMPLEMENT ALGORITHMS
- 4. EVALUATE THE ACCURACY OF EACH ALGORITHMS
- 5. DISPLAY RESULTS

## **Solution & Technical Architecture**



## **User Stories**

User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance criteria	Priority	Release
Developer	Data Collection	USN-1	Collecting and Analysing the raw Image Data.	Through my Jupyter Notebook / google colab.	High	Sprint-1
Developer	Image Preprocessing	USN-1	Converting and correcting the image to make image quality and resolution high by rotation images in all possible directions and gaining knowledge.	Through my Jupyter Notebook / google colab.& click Run	High	Sprint-1
Developer	Trainset and Testset Image Data generation	USN-1	Converting and correcting the image to make image quality and resolution high by rotation images in all possible directions and gaining knowledge for test and train data.	Through my Jupyter Notebook / google colab.	Medium	Sprint-1
Developer	Model Building	USN-2	Logic for Model by some Algorithms /Activation Functions.	Through my Jupyter Notebook / google colab.	High	Sprint-2
ē	Saving the Model	USN-2	As a Developer saving the model developed for estimation of fire	Through my Jupyter Notebook / google colab.	High	Sprint-2
	Video Analysis	USN-3		Through my Dashboard	Medium	Sprint-3
Customer	Twilio Message service	USN-3		Twilio message service	Low	Sprint-3
Customer	Alert Sound and Message	USN-4	Sending Alert text message using registered twilio account and produce output sound alert alarm .	Playsound package	Low	Sprint-4
Administrator	Train Model on Cloud	USN-5	Application Deployment on Local System / Cloud Local Server Configuration: Cloud Server Configuration : and to train the deep learning model in IBM Cloud.	IBM Cloud deployment service	Medium	Sprint-4

## 6.PROJECT PLANNING & SCHEDULING

# **Sprint Planning & Estimation**

Sprint	Functional Requiremen t (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	Registration	USN-1	As a user, I can register for the application by entering my email, password, and confirming my password.	20	High	APOORVA.C ILAKIYADHARSHINI.G JAYANTHI.M JAYANTHI.S
		USN-2	As a user, I will receive confirmation email once I have registered for the application usage.	20	High	APOORVA.C ILAKIYADHARSHINI.G JAYANTHI.M JAYANTHI.S
Sprint-2	Input	USN-3	Whenever the fire is detected, the information is given to the database.	20	High	APOORVA.C ILAKIYADHARSHINI.G JAYANTHI.M JAYANTHI.S

Sprint-2		USN-4	When it is the wildfire then the alarming system is activated.	20	High	APOORVA.C ILAKIYADHARSHINI.G JAYANTHI.M JAYANTHI.S
Sprint-3	Output	USN-5	And the alarm also sent to the corresponding departments and made them know that the wildfire is erupted.	20	High	APOORVA.C ILAKIYADHARSHINI.G JAYANTHI.M JAYANTHI.S
Sprint-4	Action	USN-6	Required actions will be taken in order to control erupted wildfire by reaching as early as possible to the destination with thehelp of detecting systems.	20	High	APOORVA.C ILAKIYADHARSHINI.G JAYANTHI.M JAYANTHI.S

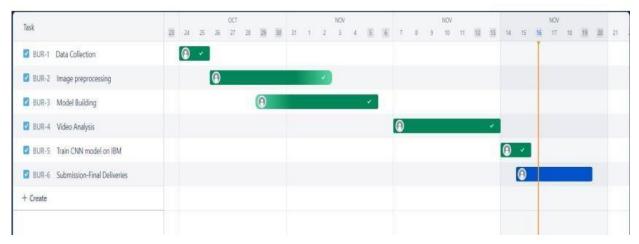
## **Sprint Delivery Schedule**

Sprint	Total Story Points	Duration	Sprint Start Date	Sprint End Date(Planned)	Story Points Completed (as on Planned End Date)	Sprint Release Date(Actual)
Sprint-1	20	6 Days	24 Oct 2022	29 Oct 2022	20	29 Oct 2022
Sprint-2	20	6 Days	31 Oct 2022	05 Nov 2022	20	05 Nov 2022
Sprint-3	20	6 Days	07 Nov 2022	12 Nov 2022	20	12 Nov 2022
Sprint-4	20	6 Days	14 Nov 2022	19 Nov 2022	20	19 Nov 2022

The following table shows the sprint works assigned to the members along with the priority and story points assigned with the functional requirements with regards to user story.

## Reports from JIRA

Burndown Chart:



## 6. CODING & SOLUTION

#### Feature 1

In Feature 1 module we have made data collection and Image preprocessing for and Model training.

importing Required Libraries:

import keras from keras.preprocessing.image import ImageDataGenerator import matplotlib.pyplot as plt import numpy as np batch\_size = 32

image resizing and preprocessing :
train\_datagen = ImageDataGenerator(

```
shear range=0.2,
  rotation_range=180,
  zoom_range=0.2,
  horizontal_flip=True,
val_datagen = ImageDataGenerator(
  rescale=1./255
train_generator = train_datagen.flow_from_directory(
  'train_set/',
  target_size=(150, 150),
  batch_size=batch_size,
  class_mode='binary'
)
val_generator = val_datagen.flow_from_directory(
  'test set/',
  target_size=(150, 150),
  batch_size=batch_size,
  class mode='binary'
)
Creating the sequential model:
from keras.models import Sequential
from keras.layers import Convolution2D
from keras.layers import MaxPooling2D
from keras.layers import Activation
from keras.layers import Dropout
from keras.layers import Flatten
from keras.layers import Dense
model=Sequential()
model.add(Convolution2D(32,(3,3),input_shape=(150,150,3))) #Convolutional 2D Layer
model.add(Activation('relu'))
model.add(MaxPooling2D(pool_size=(2,2))) # MaxPooling Layer
model.add(Flatten())
                         #Flatten Layer to make a array
model.add(Dense(150))
model.add(Activation('relu'))
model.add(Dropout(0.5))
model.add(Dense(1))
model.add(Activation('sigmoid'))
model.compile(
  loss='binary_crossentropy',
  optimizer='adam',
  metrics=['accuracy']
Model summary:
```

## model.summary()

Model: "sequential"

Layer (type)	Output	Shape	Param #
conv2d (Conv2D)	(None,	148, 148, 32)	896
activation (Activation)	(None,	148, 148, 32)	0
max_pooling2d (MaxPooling2D)	(None,	74, 74, 32)	0
flatten (Flatten)	(None,	175232)	0
dense (Dense)	(None,	150)	26284950
activation_1 (Activation)	(None,	150)	0
dropout (Dropout)	(None,	150)	0
dense_1 (Dense)	(None,	1)	151
activation_2 (Activation)	(None,	1)	Ø
Total params: 26,285,997 Trainable params: 26,285,997 Non-trainable params: 0			

#### Feature 2

from keras.models import load\_model from keras.preprocessing import image import numpy as np import cv2 from PIL import Image, ImageOps model=load\_model("forest1.h5") from twilio.rest import Client from playsound import playsound model=load\_model('forest1.h5') video=cv2.VideoCapture(0) name=['forest','with fire'] account\_durai='ACca0e8bb11699d2957d67c979ca84b68a' auth\_token='bcb5f3850ef4b7ed263f60efc9acecdb' client =Client(account\_durai,auth\_token) message=client.messages \ .create( body='-----', from\_='+19457581434',to='+919043062227')

```
print(message.sid), print("Alert Message sent")
SMb8a51eaeb987fbc8d5eced2dab56300a
Alert Message sent
```

#### 7. TESTING

#### **Test Cases & User Acceptance Testing**

Testing with input video recording from user end:

```
import cv2
import numpy as np
from keras.preprocessing import image
from keras.models import load model
from twilio.rest import Client
from playsound import playsound
model=load_model('forest1.h5')
video=cv2.VideoCapture(0)
name=['forest','with fire']
while(True):
  ret,frame=video.read()
  cv2.imshow('frame',frame)
  cv2.imwrite('image.jpg',frame)
  img=image.load_img('image.jpg',target_size=(64,64))
  x=image.img_to_array(img)
  x=np.expand\_dims(x,axis=0)
  pred=model.predict(x)
  index=np.argmax(pred)
  if index==0:
    account_durao='ACca0e8bb11699d2957d67c979ca84b68a'
    auth token='bcb5f3850ef4b7ed263f60efc9acecdb'
    client
    =Client(account_durai,auth_token)
    message=client.messages \
    .create(body='-----', reate(body='-----',
         from_='+19457581434',to='+919043062227')
    print(message.sid)
    print('Fire detected')
    print("Alert Message sent!")
    playsound('tornado-siren.mp3')
  else:
    print('No Danger')
    cv2.imshow("image.jpg",frame)
    if cv2.waitKey(2) & 0xff == ord('a'):
       break
video.release()
cv2.destroyAllWindows()
```

```
account_durai = 'AC04fd8c4ea21f7599b004db5c72066eef' auth_token = '48a23af63a81a9fe85bf6e600f3668f4' client = Client(account_durai, auth_token) message = client.messages \ .create( body='Forest fire is detected , stay alert', from_='+16075363954', to='+919043062227' ) print(message.durai)
```

#### 8. RESULTS

#### **Performance Metrics**

```
loss: 0.3438 - accuracy: 0.8483 - val_loss: 0.2485 - val_accuracy: 0.958

loss: 0.3816 - accuracy: 0.8483 - val_loss: 0.2569 - val_accuracy: 0.958

loss: 0.4068 - accuracy: 0.8391 - val_loss: 0.2547 - val_accuracy: 0.958

loss: 0.3312 - accuracy: 0.8437 - val_loss: 0.2601 - val_accuracy: 0.950

loss: 0.5621 - accuracy: 0.8368 - val_loss: 0.2679 - val_accuracy: 0.958
```

## 9. ADVANTAGES & DISADVANTAGES

#### **Advantages**

- Easily detect and Estimate the Forest Fire.
- Most Accurate
- Flexible Model which can give maximized outcome
- No Specific Requirements needed to implement the model

#### **Disadvanatges**

- Training model is time consuming process.
- Error in Cv can cause damage to camera
- Access of camera are prohibited due to personal issues

#### 10. CONCLUSION

Thus we have constructed a model that can can identify the effects of the forest fire and it can analyse the forest fire by advanced AI techniques and CNN Algorithm then the Prediction model is Checked and then the model is connected with Twilio account credentials of the Developer consisting of phone numbers of the persons in the surroundings of the people in the area of easy forest fire zone then an security sound alert system is developed to make a alert sound which is downloaded from internet then the entire model is deployed to the IBM Cloud account that we have created was made with the studies we have done.

#### 11. FUTURE SCOPES

12. APPENDIX

- It can be developed as a Web or Android Application.
- In future Alternate Advanced technologies can be Implemented.
- The Identification and tracking system can be implemented if possible.

Source Code link		

## **Source Code**

om tensorflow) (0.4.0)

```
from google.colab import drive
drive.mount('/content/drive')
Mounted at /content/drive
                                                                                              In [ ]:
!pip install tensorflow
!pip install opency-python
!pip install opency-contrib-python
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simpl
Requirement already satisfied: tensorflow in /usr/local/lib/python3.7/dist-packages (2.9.2)
Requirement already satisfied: libclang>=13.0.0 in /usr/local/lib/python3.7/dist-packages (from
tensorflow) (14.0.6)
Requirement already satisfied: tensorflow-estimator<2.10.0,>=2.9.0rc0 in /usr/local/lib/python3.
7/dist-packages (from tensorflow) (2.9.0)
Requirement already satisfied: keras-preprocessing>=1.1.1 in /usr/local/lib/python3.7/dist-packa
ges (from tensorflow) (1.1.2)
Requirement already satisfied: six>=1.12.0 in /usr/local/lib/python3.7/dist-packages (from tenso
rflow) (1.15.0)
Requirement already satisfied: tensorboard<2.10,>=2.9 in /usr/local/lib/python3.7/dist-packages
(from tensorflow) (2.9.1)
Requirement already satisfied: h5py>=2.9.0 in /usr/local/lib/python3.7/dist-packages (from tenso
rflow) (3.1.0)
Requirement already satisfied: numpy>=1.20 in /usr/local/lib/python3.7/dist-packages (from tenso
rflow) (1.21.6)
Requirement already satisfied: wrapt>=1.11.0 in /usr/local/lib/python3.7/dist-packages (from ten
sorflow) (1.14.1)
Requirement already satisfied: flatbuffers<2,>=1.12 in /usr/local/lib/python3.7/dist-packages (f
rom tensorflow) (1.12)
Requirement already satisfied: protobuf<3.20,>=3.9.2 in /usr/local/lib/python3.7/dist-packages (
from tensorflow) (3.19.6)
Requirement already satisfied: tensorflow-io-gcs-filesystem>=0.23.1 in /usr/local/lib/python3.7/
dist-packages (from tensorflow) (0.27.0)
Requirement already satisfied: absl-py>=1.0.0 in /usr/local/lib/python3.7/dist-packages (from te
nsorflow) (1.3.0)
Requirement already satisfied: packaging in /usr/local/lib/python3.7/dist-packages (from tensorf
low) (21.3)
Requirement already satisfied: opt-einsum>=2.3.2 in /usr/local/lib/python3.7/dist-packages (from
tensorflow) (3.3.0)
Requirement already satisfied: termcolor>=1.1.0 in /usr/local/lib/python3.7/dist-packages (from
tensorflow) (2.1.0)
Requirement already satisfied: setuptools in /usr/local/lib/python3.7/dist-packages (from tensor
flow) (57.4.0)
Requirement already satisfied: keras<2.10.0,>=2.9.0rc0 in /usr/local/lib/python3.7/dist-packages
 (from tensorflow) (2.9.0)
Requirement already satisfied: google-pasta>=0.1.1 in /usr/local/lib/python3.7/dist-packages (fr
om tensorflow) (0.2.0)
Requirement already satisfied: grpcio<2.0,>=1.24.3 in /usr/local/lib/python3.7/dist-packages (fr
om tensorflow) (1.50.0)
Requirement already satisfied: astunparse>=1.6.0 in /usr/local/lib/python3.7/dist-packages (from
 tensorflow) (1.6.3)
Requirement already satisfied: gast<=0.4.0,>=0.2.1 in /usr/local/lib/python3.7/dist-packages (fr
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Requirement already satisfied: typing-extensions>=3.6.6 in /usr/local/lib/python3.7/dist-package
s (from tensorflow) (4.1.1)
Requirement already satisfied: wheel<1.0,>=0.23.0 in /usr/local/lib/python3.7/dist-packages (fro
m astunparse>=1.6.0->tensorflow) (0.38.3)
Requirement already satisfied: cached-property in /usr/local/lib/python3.7/dist-packages (from h
5py >= 2.9.0 - tensorflow) (1.5.2)
Requirement already satisfied: markdown>=2.6.8 in /usr/local/lib/python3.7/dist-packages (from t
ensorboard\langle 2.10, \rangle = 2.9 - \lambda  tensorflow) (3.4.1)
Requirement already satisfied: tensorboard-data-server<0.7.0,>=0.6.0 in /usr/local/lib/python3.7
/dist-packages (from tensorboard<2.10,>=2.9->tensorflow) (0.6.1)
Requirement already satisfied: tensorboard-plugin-wit>=1.6.0 in /usr/local/lib/python3.7/dist-pa
ckages (from tensorboard<2.10,>=2.9->tensorflow) (1.8.1)
Requirement already satisfied: google-auth-oauthlib<0.5,>=0.4.1 in /usr/local/lib/python3.7/dist
-packages (from tensorboard<2.10,>=2.9->tensorflow) (0.4.6)
Requirement already satisfied: requests<3,>=2.21.0 in /usr/local/lib/python3.7/dist-packages (fr
om tensorboard<2.10,>=2.9->tensorflow) (2.23.0)
Requirement already satisfied: werkzeug>=1.0.1 in /usr/local/lib/python3.7/dist-packages (from t
ensorboard<2.10,>=2.9->tensorflow) (1.0.1)
Requirement already satisfied: google-auth<3,>=1.6.3 in /usr/local/lib/python3.7/dist-packages (
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Requirement already satisfied: pyasn1-modules>=0.2.1 in /usr/local/lib/python3.7/dist-packages (
from google-auth<3,>=1.6.3->tensorboard<2.10,>=2.9->tensorflow) (0.2.8)
Requirement already satisfied: cachetools<6.0,>=2.0.0 in /usr/local/lib/python3.7/dist-packages
(from google-auth<3,>=1.6.3->tensorboard<2.10,>=2.9->tensorflow) (5.2.0)
Requirement already satisfied: rsa<5,>=3.1.4 in /usr/local/lib/python3.7/dist-packages (from goo
qle-auth<3,>=1.6.3->tensorboard<2.10,>=2.9->tensorflow) (4.9)
Requirement already satisfied: requests-oauthlib>=0.7.0 in /usr/local/lib/python3.7/dist-package
s (from google-auth-oauthlib<0.5,>=0.4.1->tensorboard<2.10,>=2.9->tensorflow) (1.3.1)
Requirement already satisfied: importlib-metadata>=4.4 in /usr/local/lib/python3.7/dist-packages
 (from markdown>=2.6.8->tensorboard<2.10,>=2.9->tensorflow) (4.13.0)
Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.7/dist-packages (from import1
ib-metadata \ge 4.4-markdown \ge 2.6.8- tensorboard < 2.10,>=2.9- tensorflow) (3.10.0)
Requirement already satisfied: pyasn1<0.5.0,>=0.4.6 in /usr/local/lib/python3.7/dist-packages (f
rom pyasn1-modules>=0.2.1-yoogle-auth<3,>=1.6.3-ytensorboard<2.10,>=2.9-ytensorflow) (0.4.8)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.7/dist-packages (fro
m requests < 3, >= 2.21.0 - tensorboard < 2.10, >= 2.9 - tensorflow) (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<3,>=2.21.0->tensorboard<2.10,>=2.9->tensorflow) (1.24.3)
Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.7/dist-packages (from
requests<3,>=2.21.0->tensorboard<2.10,>=2.9->tensorflow) (3.0.4)
Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages (from requ
ests<3,>=2.21.0->tensorboard<2.10,>=2.9->tensorflow) (2.10)
Requirement already satisfied: oauthlib>=3.0.0 in /usr/local/lib/python3.7/dist-packages (from r
equests-oauthlib>=0.7.0->google-auth-oauthlib<0.5,>=0.4.1->tensorboard<2.10,>=2.9->tensorflow)
3.2.2)
Requirement already satisfied: pyparsing!=3.0.5,>=2.0.2 in /usr/local/lib/python3.7/dist-package
s (from packaging->tensorflow) (3.0.9)
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simpl
Requirement already satisfied: opency-python in /usr/local/lib/python3.7/dist-packages (4.6.0.66
Requirement already satisfied: numpy>=1.14.5 in /usr/local/lib/python3.7/dist-packages (from ope
ncv-python) (1.21.6)
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simpl
Requirement already satisfied: opency-contrib-python in /usr/local/lib/python3.7/dist-packages (
4.6.0.66)
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```
Requirement already satisfied: numpy>=1.14.5 in /usr/local/lib/python3.7/dist-packages (from ope
ncv-contrib-python) (1.21.6)
                                                                                                 In []:
import tensorflow as tf
import numpy as np
from tensorflow import keras
import os
import cv2
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.preprocessing import image
                                                                                                 In []:
train=ImageDataGenerator(rescale=1./255,
                                  shear range=0.2,
                                  rotation range=180,
                                  zoom range=0.2,
                                  horizontal flip=True)
train = ImageDataGenerator(rescale=1/255)
test = ImageDataGenerator(rescale=1/255)
                                                                                                 In [ ]:
train dataset =
train.flow from directory("/content/drive/MyDrive/ibm/Dataset/Dataset/train set",
                                            target_size=(128,128),
                                           batch size = 32,
                                           class mode = 'binary' )
Found 436 images belonging to 2 classes.
                                                                                                 In [ ]:
test dataset = test.flow from directory("/content/drive/MyDrive/ibm/Dataset/Dataset/test set",
                                           target_size=(128,128),
                                           batch size = 32,
                                           class mode = 'binary' )
Found 320 images belonging to 2 classes.
                                                                                                 In []:
test dataset.class indices
                                                                                                Out[]:
{'forest': 0, 'with fire': 1}
                                                                                                 In [ ]:
#to define linear initialisation import sequential
from keras.models import Sequential
#to add layer 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')
                                                                                                 In [ ]:
model = keras.Sequential()
model.add(Convolution2D(32,(3,3),input shape=(128,128,3),activation='relu'))
model.add(MaxPooling2D(pool size=(2,2)))
model.add(Convolution2D(32,(3,3),activation='relu'))
model.add(MaxPooling2D(pool size=(2,2)))
model.add(Convolution2D(32, (3, 3), activation='relu'))
model.add(MaxPooling2D(pool size=(2,2)))
```

```
model.add(Convolution2D(32,(3,3),activation='relu'))
model.add(MaxPooling2D(pool size=(2,2)))
model.add(Flatten())
                                                             In [ ]:
model.add(Dense(150,activation='relu'))
model.add(Dense(1,activation='sigmoid'))
                                                             In []:
model.compile(loss = 'binary crossentropy',
         optimizer = "adam",
         metrics = ["accuracy"])
                                                             In []:
r = model.fit(train dataset, epochs = 5, validation data = test dataset)
Epoch 1/5
oss: 0.2729 - val accuracy: 0.9219
Epoch 2/5
s: 0.1726 - val accuracy: 0.9187
Epoch 3/5
s: 0.2235 - val accuracy: 0.8938
Epoch 4/5
s: 0.1444 - val accuracy: 0.9406
Epoch 5/5
s: 0.1118 - val accuracy: 0.9688
                                                             In []:
predictions = model.predict(test dataset)
predictions = np.round(predictions)
10/10 [======== ] - 8s 839ms/step
                                                             In [ ]:
predictions
                                                             Out[]:
array([[1.],
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       [1.]], dtype=float32)
                                                                                              In []:
print(len(predictions))
320
                                                                                              In []:
model.save("/content/drive/MyDrive/ibm/music/forest1.h5")
                                                                                              In []:
#import load model from keras.model
from keras.models import load model
#import image class from keras
import tensorflow as tf
from tensorflow.keras.preprocessing import image
#import numpy
import numpy as np
#import cv2
import cv2
                                                                                              In []:
model = load model("/content/drive/MyDrive/ibm/music/forest1.h5")
                                                                                              In []:
def predictImage(filename):
  img1 = image.load img(filename, target size=(128,128))
  Y = image.img_to_array(img1)
  X = np.expand dims(Y,axis=0)
  val = model.predict(X)
  print(val)
  if val == 1:
    print(" fire")
  elif val == 0:
      print("no fire")
                                                                                              In [ ]:
predictImage("/content/drive/MyDrive/ibm/Dataset/Dataset/test set/with
fire/Forest fire MNRF esize IMG 6743.jpg")
1/1 [=======] - 0s 118ms/step
[[1.]]
fire
                                                                                              In []:
!pip install twilio
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simpl
Collecting twilio
  Downloading twilio-7.15.2-py2.py3-none-any.whl (1.4 MB)
            | 1.4 MB 4.2 MB/s
```

```
Requirement already satisfied: requests>=2.0.0 in /usr/local/lib/python3.7/dist-packages (from t
wilio) (2.23.0)
Collecting PyJWT<3.0.0,>=2.0.0
  Downloading PyJWT-2.6.0-py3-none-any.whl (20 kB)
Requirement already satisfied: pytz in /usr/local/lib/python3.7/dist-packages (from twilio) (202
Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages (from requ
ests >= 2.0.0 -> twilio) (2.10)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.7/dist-packages (fro
m requests>=2.0.0->twilio) (2022.9.24)
Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.7/dist-packages (from
 reguests>=2.0.0->twilio) (3.0.4)
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-7.15.2
                                                                                               In [ ]:
!pip install playsound
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simpl
Requirement already satisfied: playsound in /usr/local/lib/python3.7/dist-packages (1.3.0)
                                                                                               In [ ]:
#import opency librariy
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
#imort playsound package
from playsound import playsound
                                                                                               In [ ]:
#load the saved model
model = load model(r'/content/drive/MyDrive/ibm/music/forest1.h5')
#define video
video = cv2.VideoCapture('/content/Fighting Fire with Fire _ Explained in 30 Seconds.mp4')
#define the features
name = ['forest','with forest']
                                                                                               In []:
account durai = 'AC153cd036f84d1a7a72baf185e662b651'
auth token = '*********************
client = Client(account durai, auth token)
message = client.messages \
    .create(
         body='Forest fire go to saved places',
         from ='+13605838876',
         to='+919043062227'
     )
print (message.sid)
SM35132eb94a5fb50e21844149b2b85005
```

```
import cv2
             import threading
             import playsound
             from twilio.rest import Client
             fire_cascade = cv2.CascadeClassifier('fire_detection.xml')
             vid = cv2.VideoCapture(0)
             runOnce = False
             Font = cv2.FONT HERSHEY SIMPLEX
             def play_alarm_sound_function():
                  playsound.playsound('fire_Alarm.mp3',True)
                  print("Fire alarm end")
             def send_Twilio_function():
                 SID = 'AC153cd036f84d1a7a72baf185e662b651'
                 Auth_Token = '*****************
                 cl = Client(SID, Auth_Token)
                 cl.messages.create(body="Forest Fire Go To Saved Places", from_= '+13605838875',to='+919043062227')
             while(True):
                 Alarm Status = False
                 ret, Forest = vid.read()
                  gray = cv2.cvtColor(Forest, cv2.COLOR_BGR2GRAY)
                 fire = fire_cascade.detectMultiScale(Forest, 1.2, 5)
                 for (x,y,w,h) in fire:
                     cv2.rectangle(Forest,(x-20,y-20),(x+w+20,y+h+20),(0,255,0),2)
                     roi_gray = gray[y:y+h, x:x+w]
                      roi_color = Forest[y:y+h, x:x+w]
                     cv2.putText(Forest, "Fire", (x,w), Font, 2, (0,0,225), 2, cv2.LINE_AA)
                     print("Fire alarm initiated")
                     threading.Thread(target=play_alarm_sound_function).start() # To call alarm thread
                      if runOnce == False:
                          print("Twilio send initiated")
```

```
threading.Thread(target=send_Twilio_function).start() # To call Twilio thread
                        runOnce = True
                   if runOnce == True:
                        print("Twilio is already sent once")
                        runOnce = True
               cv2.imshow('Forest', Forest)
               if cv2.waitKey(1) & 0xFF == ord('q'):
                   Break
           vid.release
           cv2.destroyAllWindows
import cv2
             import threading
             import playsound
             import smtplib
             fire_cascade = cv2.CascadeClassifier('train/fire_detection.xml')
             vid = cv2.VideoCapture(0)
             runOnce = False
             Font = cv2.FONT_HERSHEY_SIMPLEX
             def play_alarm_sound_function():
                 playsound.playsound('music/fire_Alarm.mp3',True)
                 print("Fire alarm end")
             def send_mail_function():
                 recipientmail = "19161681@saec.ac.in"
                 recipientmail = recipientmail.lower()
                 try:
                     server = smtplib.SMTP('smtp.gmail.com', 587)
                     server.ehlo()
                     server.starttls()
                     server.login("cdurai1321@gmail.com", 'zchgwcublcrobkqu')
                     server.sendmail('1916168l@saec.ac.in', recipientmail, "WildFire forest ")
                     print("Alert mail sent sucesfully to {}".format(recipientmail))
                     server.close()
                 except Exception as e:
                     print(e)
             while(True):
                 Alarm_Status = False
```

```
ret, frame = vid.read()
   gray = cv2.cvtColor(frame, cv2.COLOR_BGR2GRAY)
   fire = fire_cascade.detectMultiScale(frame, 1.2, 5)
   for (x,y,w,h) in fire:
        cv2.rectangle(frame,(x-20,y-20),(x+w+20,y+h+20),(0,255,0),2)
        roi_gray = gray[y:y+h, x:x+w]
        roi_color = frame[y:y+h, x:x+w]
       cv2.putText(frame, "Fire",(x,w),Font,2,(0,0,225),2,cv2.LINE_AA)
        print("Fire alarm initiated")
        threading.Thread(target=play_alarm_sound_function).start() # To call alarm thread
        if runOnce == False:
            print("Mail send initiated")
            threading.Thread(target=send_mail_function).start() # To call Email thread
            runOnce = True
        if runOnce == True:
            print("Mail is already sent once")
            runOnce = True
   cv2.imshow('Forest', frame)
   if cv2.waitKey(1) & 0xFF == ord('q'):
        break
vid.release
cv2.destroyAllWindows
```