# A PROJECT REPORT <u>EMERGING METHODS FOR EARLYDETECTION OF</u> <u>FOREST FIRE</u>

Submitted by

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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.

## 1.1 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.

## 1.2 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

## 2.1 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) [1], which will soon become part of the European Emergency

Management Service, maintained by the Copernicus Earth Observation Programme [2]. 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 [3], 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 [3]. 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 [4]. 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.[5]

#### 2.2 References

- 1. Georgi Hristov; Jordan Raychev; Diyana Kinaneva; Plamen Zahariev, Published in: 2018 28th EAEEIE Annual Conference (EAEEIE).
- 2. Chi Yuan, Youmin Zhang, and Zhixiang Liu , Published in: 2015, Published by NRC Research Press.
- 3. Mohamed Hefeeda and Majid Bagheri , Published in: June 26, 2008.
- 4. PRIYADARSHINI M HANAMARADDI, Published in: January 2016.
- 5. Dr. Panagiotis Barmpoutis, Periklis Papaioannou, Dr. Kosmas Dimitropoulos, Dr. Nikos GRAMMALIDIS, Published in: 11 November 2020.
- 6. Vinay Chowdary, Mukul Kumar Gupta, Rajesh Singh, Published in:2018
- 7. Majid Bahrepour, Nirvana Meratnia, Paul Havinga, Published in: January 2008.
- 8. Dr.L.Latha , Published in: January 2015 9. P. Piccinini, S. Calderara, and R. Cucchiara , Published in: September, 2006.

#### 2.3 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

## 3.1 Empathy Map Canvas

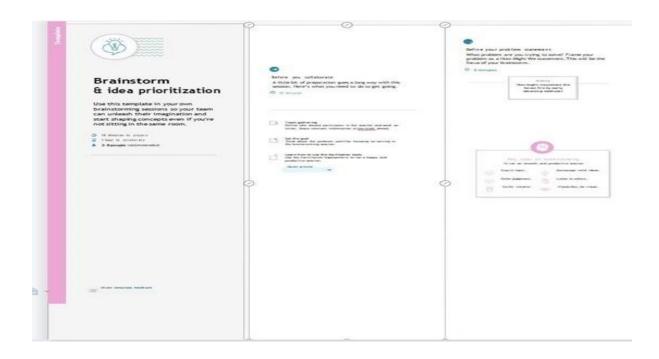
An empathy map canvas is a more in-depth version of the original empathy map, which helps identify and describe the user's needs and pain points.

## **Empathy Map** How to tackle the increase in count of forest fires? What do they THINK AND FEEL? major preoccupations nich affects. worres. E and rations What do they What do they HEAR? SEE? the people around the place can be whig boss say years for the forest to be regrown. what the murket offers white influencers too Check weather and drought conditions. What do they that are at stake. SAY AND DO? behavior towards others PAIN GAIN Becomes fors fractions difficult to put out forest fire if it has spread over a large detected at an early stage and can be prevented before it mosades straic es

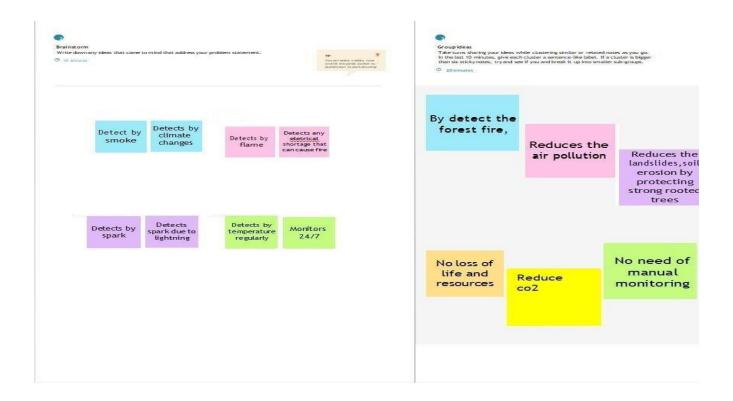
## 3.2 Ideation & Brainstorming

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

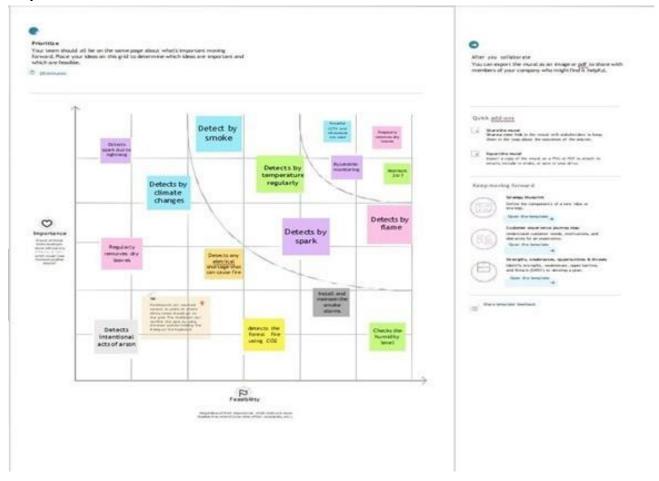
**Step-1**: Team Gathering, Collaboration and Select the Problem Statement



Step-2: Brainstorm, Idea Listing and Grouping



## Step-3: Idea Prioritization



## 3.3 Proposed Solution

S/no	Parameter	Description
1	Problem Statement (Problem	A forest fire risk prediction
	to be solved)	algorithm, based on support
		vector machines, is
		presented. The algorithm
		depends on previous weather
		conditions in order to predict
		the fire hazard level of a day.
2	Idea / Solution description	Use computer vision methods
		for recognitionand detection
		of smoke or fire.
3	Novelty / Uniqueness	Real time computer program
		detect forest fire in earliest
		before itspread to larger area.
4	Impact on society	Blocked roads and railway
		lines, electricity, mobile and
		land telephone lines cut,
		destruction of homes and
		industries.
5	Business Model (Revenue	The proposed method was
	Model)	implemented using the
		Python programming
		language on a Core i3 or
		greater ( CPU and 4GB RAM.)
6	Scalability of the Solution	Computer vision models
		enable land cover
		classification and smoke
		detection from satellite and
		ground cameras

## 4. REQUIREMENT ANALYSIS

# 4.1 Functional requirement

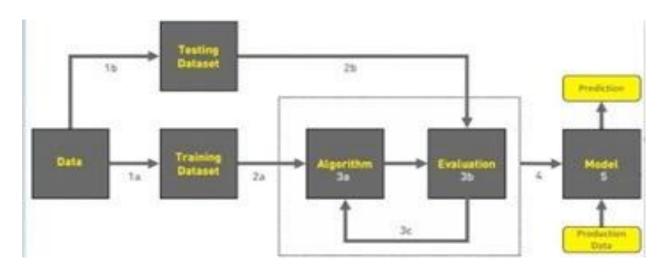
FR No.	Functional Requirement	Sub Requirement (Story /
	(Epic)	Sub-Task)
FR-1	User Registration	Registration through Gmail
FR-2	User Confirmation	Confirmation via Email
		Confirmation via OTP
FR-3	User Login	Login using credentials
FR-4	User Search	Search for Info on forest fire occurrence
FR-5	User Profile	User shall be given a live feed of the forest
FR-6	User Application	User is alerted if there is a forest fire occurrence intheir surroundings

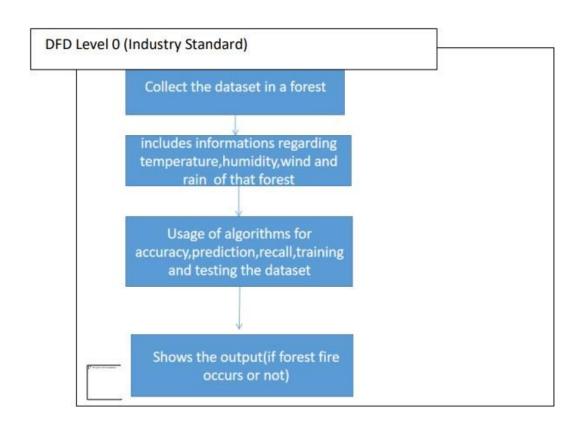
# 4.2 Non-Functional requirements

FR No	Non-Functional Requirement	Description
NFR-1	Usability	Alerts according to the user
		location
NFR-2	Security	Instant live feed with alert of
		the situation
NFR-3	Reliability	The prediction of the forest
		fire is 87% accurate
NFR-4	Performance	The feed and the alert
		message an immediate
		action without a lag
NFR-5	Availability	The application gives alerts
		and live feeds 24/7
NFR-6		Early detection and alerting
		users are done efficiently and
		in a faster means

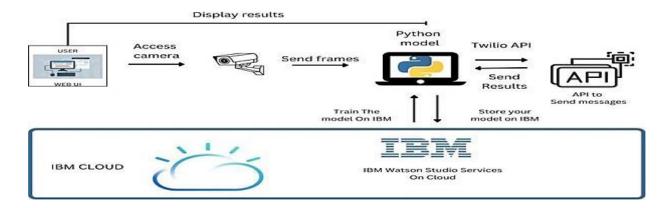
## **5. PROJECT DESIGN**

## **5.1 Data Flow Diagrams**





## **5.2 Solution & Technical Architecture**



S.No	Component	Description	Technology
1	User Interface	The user uses the	Python/HTML,CSS,
		console to access	Javascript and
		the interface	react.Js
2		Video Feed	Web Camera/Video
	Input		on a site
3	Conversion	Video inputted is	Frame Converter
		converted into	
		Frames	
4	Feeding the Model	The Frames are sent	Our Model
		to the Deep learning	
		model	
5	Dataset	Using Test set and	Data set from Cloud
		train set , train the	Storage , Database
		model	
6	Cloud Database	The model is trained	IBM Cloudant ,Python
		in the cloud more	Flask.
		precise with	
		detections more	
		images can be added	
		later on.	
7	Infrastructure (Server	Application	Java/python
	/ Cloud), API	Deployment on Local	,React.Js ,JavaScript
		System / Cloud Local	,HTML ,CSS ,IBM
		,Cloud Server	Cloud ,OPEN CV
		Configuration, Twilio	,Anaconda Navigator
		API to send	,Local.
		messages	

**Table-2: Application Characteristics:** 

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

## 5.3 User Stories

User Type	Functional	User	User Story /	Acceptan		
	Requireme	Story	Task	ce criteria	Priority	Release
	nt (Epic)	Number				
Environme	Collect the	USN-1	As an	It is	High	Sprint-1
ntalist	data		Environmentali	necessary		
			st,it is	to collect		
			necessary to	the right		
			collect the	data else		
			data of the	the		
			forest which	prediction		
			includes	may		
			temperature,hu	become		
			midity,wind	wrong		
			and rain of the			
			forest			

USN-2	Identify algorithms that can be used for prediction	To collect the algorithm to identify the accuracy level of each algorithms	Medium	Sprint-2
USN-3	Identify the accuracy of each algorithms	Accuracy of each algorithm- calculated so that it is easy to obtain the most accurate output	High	Sprint-2
USN-4	Evaluate the Dataset	Data is evaluated before processing	Medium	Sprint-1
USN-5	Identify accuracy,preci sion,recall of each algorithms	These values are important for obtaining the right output	High	Sprint-3
USN-6	Outputs from each algorithm are obtained	It is highly used to predict the effect and to take precaution ary measures	High	Sprint-4

# 6. PROJECT PLANNING & SCHEDULING

# 6.1 Sprint Planning & Estimation

Sprint-2   Sprint-3   Video   Analysis   Sprint-3   Twilio   Message   Service   Sprint-3   Sprint-3   Alert Sound and Message   Sprint-3   Sprint-3   Sprint-3   Sprint-3   Sprint-3   Sprint-3   Sprint-3   Sprint-3   Twilio   Message   Service   Sprint-3   Sprint-3   Sprint-3   Sprint-3   Twilio   Message   Service   Sprint-3   Sprint-3   Sprint-3   Sprint-3   Twilio   Message   Service   Sprint-3   Sp	Sprint	Functional	User	User story/Task	Story	Priority	Team
Sprint-2       Saving the Model       USN-2       As a developer saving the model developed for estimation of fire       10       high       Nalubolu Naveen Kumar Avula Hari Krishna Reddy         Sprint-3       Video Analysis       USN-3       10       Medium Naveen Kumar Nalubolu Naveen Kumar Avula Hari Krishna Reddy         Sprint-3       Twilio Message Service       USN-3       10       Low Nalubolu Naveen Kumar Avula Hari Krishna Reddy         Sprint-3       Twilio Message Service       USN-3       10       Low Nalubolu Naveen Kumar Avula Hari Krishna Reddy         Sprint-4       Alert Sound and Message       USN-4       Sending Alert text message using registered twilio account and produce output       10       Low Nalubolu Naveen Kumar Avula Hari Krishna Reddy		_	_		points		Members
Model   Saving the model developed for estimation of fire		, - ,					
estimation of fire    Avula Hari Krishna Reddy   Chinthala Mounish	Sprint-2	=	USN-2	saving the model	10	high	
Sprint-3 Video Analysis USN-3 IO Medium Nalubolu Naveen Kumar Avula Hari Krishna Reddy Chinthala Mounish Narayana Vasanth Kumar  Sprint-3 Twilio Message Service USN-3 IO Low Nalubolu Naveen Kumar Avula Hari Krishna Reddy Chinthala Mounish Narayana Vasanth Kumar Avula Hari Krishna Reddy Chinthala Mounish Narayana Vasanth Kumar Avula Hari Krishna Reddy Chinthala Mounish Narayana Vasanth Kumar Avula Hari Krishna Reddy Chinthala Mounish Narayana Vasanth Kumar Avula Hari Krishna Reddy Registered twilio account and produce output Avula Hari Krishna Reddy				=			
Sprint-3 Video Analysis USN-3 Message Service USN-3 Message Service USN-3 Avula Hari Krishna Reddy Avula Hari Krishna Reddy Chinthala Mounish Narayana Avula Hari Krishna Reddy Chinthala Mounish Narayana Vasanth Kumar Avula Hari Krishna Reddy Chinthala Mounish Narayana Vasanth Kumar Avula Hari Krishna Reddy Chinthala Mounish Narayana Vasanth Kumar Avula Hari Krishna Reddy							
Analysis  Analysis  Analysis  Naveen Kumar  Avula Hari Krishna Reddy  Chinthala Mounish  Narayana Vasanth Kumar  Sprint-3  Twilio Message Service  In the service Service  To the service Service  Avula Hari Krishna Reddy  Chinthala Mounish  Narayana Vasanth Krishna Reddy  Chinthala Mounish  Narayana Vasanth Kumar  Sprint-4  Alert Sound and Message  Alert text message using registered twilio account and produce output  Avula Hari Krishna Reddy			110110		10		Vasanth Kumar
Sprint-3  Sprint-3  Twilio Message Service  Message Mounish Narayana Vasanth Kumar  Narayana Vasanth Kumar  Sprint-4  Alert Sound and Message Message using registered twilio account and produce output  Message  Krishna Reddy  Chinthala Mounish Narayana Vasanth Kumar  Avula Hari Krishna Reddy	Sprint-3		USN-3		10	Medium	
Sprint-3 Twilio USN-3 10 Low Nalubolu Naveen Kumar Avula Hari Krishna Reddy  Sprint-4 Alert Sound and Message  Sprint-4 VSP Alert Sound and Message USN-4 Sending Alert text message using registered twilio account and produce output  Mounish Narayana Vasanth Kumar  Sprint-4 Low Nalubolu Naveen Kumar  Avula Hari Krishna Reddy							
Sprint-3 Twilio USN-3 Message Service 10 Low Nalubolu Naveen Kumar Avula Hari Krishna Reddy Chinthala Mounish Narayana Vasanth Kumar  Sprint-4 Alert Sound and Message registered twilio account and produce output 10 Low Nalubolu Naveen Kumar							
Sprint-3  Twilio Message Service  10  Low Nalubolu Naveen Kumar  Avula Hari Krishna Reddy  Chinthala Mounish  Narayana Vasanth Kumar  Sprint-4  Alert Sound and Message  Service  10  Low Nalubolu Naveen Kumar  Narayana Vasanth Kumar  Sending Alert text message using registered twilio account and produce output  10  Low Nalubolu Naveen Kumar  Avula Hari Krishna Reddy							
Message Service  Naveen Kumar  Avula Hari Krishna Reddy  Chinthala Mounish  Narayana Vasanth Kumar  Sprint-4  Alert Sound and Message  USN-4  Sending Alert text message using registered twilio account and produce output  Naveen Kumar  Avula Hari Krishna Reddy							Kumar
Avula Hari Krishna Reddy  Chinthala Mounish  Narayana Vasanth Kumar  Sprint-4 Alert Sound and Message USN-4 Sending Alert text message using registered twilio account and produce output  Avula Hari Krishna Reddy	Sprint-3	Message	USN-3		10	Low	
Sprint-4 Alert Sound and Message USN-4 Sending Alert text message using registered twilio account and produce output Mounish  Narayana Vasanth Kumar  Nalubolu Naveen Kumar  Avula Hari Krishna Reddy		Service					
Sprint-4 Alert Sound and Message USN-4 Sending Alert text message using registered twilio account and produce output Vasanth Kumar  Vasanth Kumar  Nalubolu Naveen Kumar  Avula Hari Krishna Reddy							
Sprint-4 Alert Sound and Message USN-4 Sending Alert text message using registered twilio account and produce output Kumar  Kumar  Low Nalubolu Naveen Kumar  Avula Hari Krishna Reddy							Narayana
Sprint-4 Alert Sound and Message USN-4 Sending Alert text message using registered twilio account and produce output 10 Low Nalubolu Naveen Kumar Avula Hari Krishna Reddy							
and Message message using registered twilio account and produce output  Naveen Kumar  Avula Hari Krishna Reddy							
account and produce output  Avuia Hari Krishna Reddy	Sprint-4		USN-4	message using	10	Low	
				account and			
, I Dillittidid				produce output sound alert alarm			Chinthala

						Mounish  Narayana  Vasanth  Kumar
Sprint-4	Train Model on cloud	USN-5	Application Deployment on Local System/ Cloud Local Server Congiguration: Cloud Server Configuration:and to train the deep learning model in IBM Cloud	10	Medium	Nalubolu Naveen Kumar Avula Hari Krishna Reddy Chinthala Mounish Narayana Vasanth Kumar

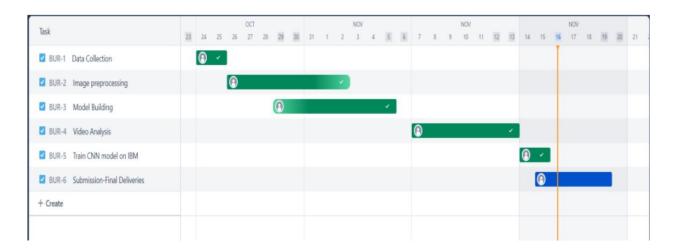
## **6.2 Sprint Delivery Schedule**

Sprint	Total Story Points	Duration	Sprint Start Date	Sprint End Date(Plann ed)	Story Points Completed (as on planned end date)	Sprint Release Date (Actual)
Sprint-1	20	6 days	24 Oct	29 Oct	20	29 Oct
			2022	2022		2022
Sprint-2	20	6days	31 Oct	05 Nov	20	05 Nov
			2022	2022		2022
Sprint-3	20	6days	07 Nov	12 Nov	20	12 Nov
			2022	2022		2022
Sprint-4	20	6days	14 Nov	19 Nov	20	19 Nov
			2022	2022		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.

## 6.3 Reports from JIRA

**Burndown Chart:** 



## 7. CODING & SOLUTION

## 7.1 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

batch\_size=batch\_size, class\_mode='binary' )

## 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),
```

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.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)	0

Total params: 26,285,997 Trainable params: 26,285,997 Non-trainable params: 0

## 7.2 Feature 2

```
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('train_set/forest/NoFire
 (1).bmp',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 sid=' AC805dad0282a5eaebfa3149365f1c605d
auth_token='a27b535245ca70f9c7b4875865a620e1
 'client =Client(account_sid,auth_token)
     message=client.messages \
     .create(body='-----Fire is detected, Stay Alert !!!------',
from_='+ 12232174143
            ',to='+916300780739')
     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('q'):
         break
video.release()
 cv2.destroyAllWindows()
```

#### 8.TESTING

## 8.1 Test Cases & User Acceptance Testing

## Testing with input video recording from user end:

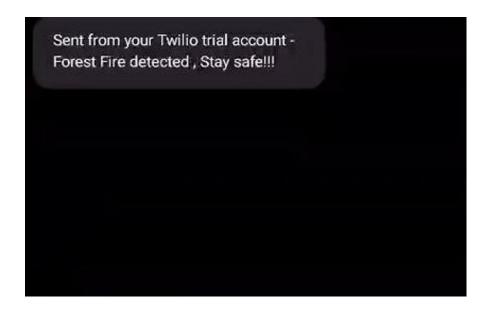
```
import cv2
 import numpy as np
 from keras.preprocessing
 import image from keras.models
 import load_model from twilio.rest
 import Client from playsound
 import playsound
 model=load_model('forest1.h5')
video=cv2.VideoCapture(0)
 name=['forest','with fire']
while(True):
   ret,frame=video.read()
   cv2.imshow('frame',frame)
   cv2.imwrite('image.jpg',frame)
   img=image.load_img('train_set/forest/NoFire (1).bmp',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_sid=' AC805dad0282a5eaebfa3149365f1c605d
auth_token='a27b535245ca70f9c7b4875865a620e1
 'client =Client(account_sid,auth_token)
     message=client.messages \
     .create(body='-----Fire is detected,Stay Alert !!!------',
from_='+ 12232174143
            ',to='+916300780739')
     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('q'):
        break
```

video.release() cv2.destroyAllWindows()

# OUTPUT:







## 9.Result

#### 9.1 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
```

## 10. 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

## 11.CONCLUSION:

Thus we have constructed a model 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 was made with the studies we have done.

## **12.FUTURE SCOPES:**

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

## 13.APPENDIX:

Source Code: https://github.com/IBM-EPBL/IBM-Project-54046-1661588193/blob/main/Final%20Deliverables/EntireModel.ipynb

## **GitHub & Project Demo Link**

Github: https://github.com/IBM-EPBL/IBM-Project-54046-1661588193