EMERGING METHODS FOR EARLY DETECTION OF FOREST

FIRE TEAM ID: PNT2022TMID25656

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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, maintainedby 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 ResearchPress.
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- 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 helpsidentify and describe the user's needs and pain points.

Empathy Map How to tackle the increase in count of forest fines? What do they THINK AND FEEL? nich affects. worres E and rations What do they What do they HEAR? SEE? the people around the place can be whig boss say Nunds years for the forest to be what influencers say what the murket offers Check weather and drought conditions. What do they that are at stake. SAY AND DO? behavior towards others

3.2 Ideation & Brainstorming

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

Becomes

difficult to put out forest fine if it has spread over a large GAIN

straic es

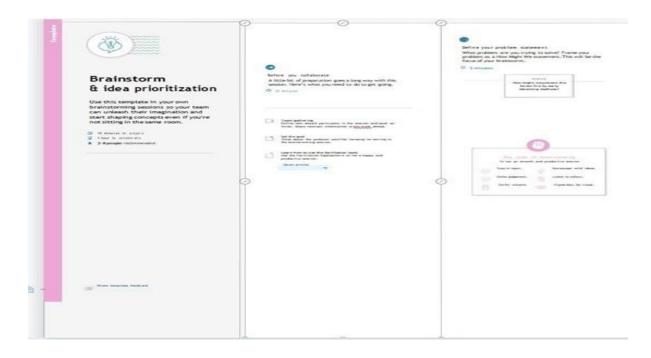
letected at an early stage and can be prevented before it

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

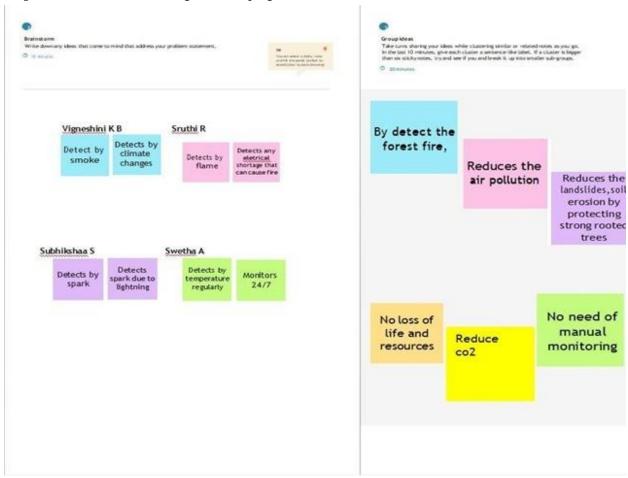
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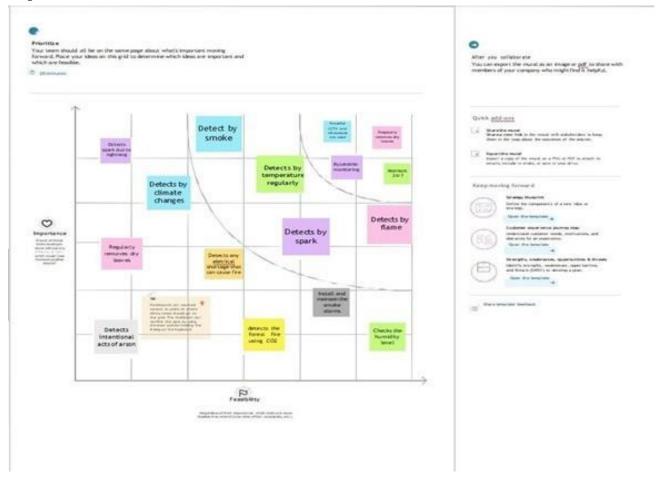
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Step-2: Brainstorm, Idea Listing and Grouping



Step-3: Idea Prioritization



3.3 Proposed Solution

S/no	Parameter	Description
1	Problem Statement (Problemto be solved)	A forest fire risk prediction algorithm, based on support vector machines, is presented. The algorithm depends on previous weatherconditions in order to predict the fire hazard level of a day.
2	Idea / Solution description	Use computer vision methodsfor 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 andland telephone lines cut, destruction of homes and industries.
5	Business Model (Revenue Model)	The proposed method was 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

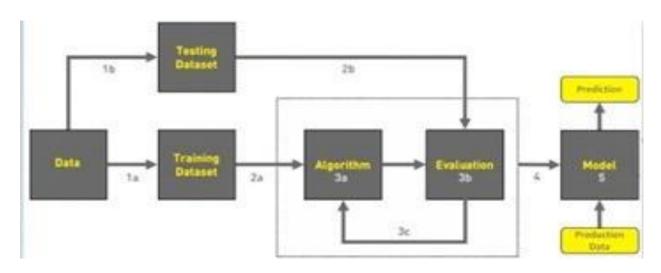
FR No.	Functional Requirement	Sub Requirement (Story /
	(Epic)	Sub-Task)
FR-1	User Registration	Registration through Gmail
FR-2	User Confirmation	Confirmation via Email
		Confirmation via OTP
FR-3	User Login	Login using credentials
FR-4	User Search	Search for Info on forest fire occurrence
FR-5	User 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

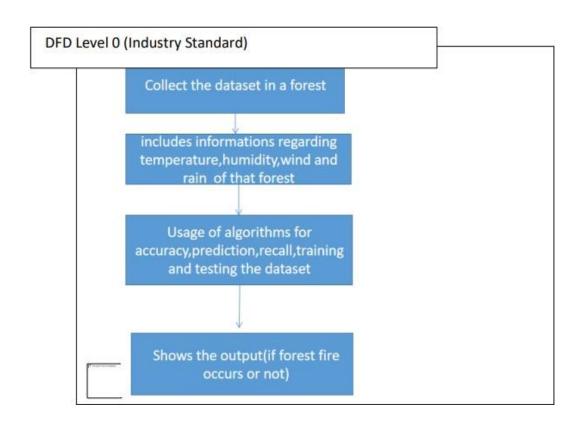
4.2 Non-Functional requirements

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

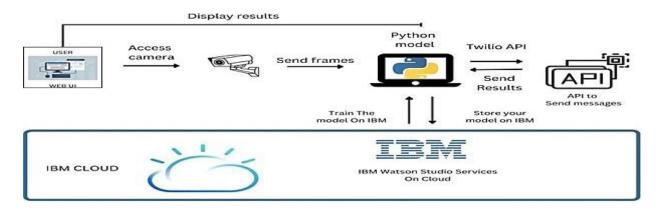
5. PROJECT DESIGN

5.1 Data Flow Diagrams





5.2 Solution & Technical Architecture



User Interface The user uses the console to access the interface Input Video Feed Video Feed Web Camera/Vide on a site Conversion Video inputted is converted into Frames Feeding the Model The Frames are sentto the Deep learning model Data set from Cloud train set, train the model Cloud Database The model is trained in the cloud more precise with detections more images can be added later on. Infrastructure (Server Application Python/HTML, CS Javascript and react. Js Web Camera/Vide on a site Web Camera/Vide on a site Thank Converter Our Model Data set from Cloud Storage, Database Data set from Cloud Storage, Database Application Java/python	
the interface react.Js Video Feed Web Camera/Vide on a site Conversion Video inputted is converted into Frames Feeding the Model The Frames are sentto the Deep learning model Dataset Using Test set and train set, train the model Cloud Database The model is trained in the cloud more precise with detections more images can be added later on. Teach Video Feed Web Camera/Vide on a site Web Camera/Vide on a site Our Model Data set from Cloud Storage, Database IBM Cloudant, Pythe Flask.	SS,
Video Feed Input Video Feed Web Camera/Video on a site Conversion Video inputted is converter into Frames Frame Converter Our Model The Frames are sentto the Deep learning model Using Test set and train set, train the model Cloud Database The model is trained in the cloud more precise with detections more images can be added later on. Web Camera/Video on a site Frame Converter Our Model Bata set from Cloud Storage, Database in the cloud more precise with detections more images can be added later on.	
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later on.	
7 Infrastructure (Server Application Java/python	
7 Infrastructure (Server Application Java/python	
/ Cloud), API Deployment on Local ,React.Js ,JavaScrij	pt
System / Cloud Local ,HTML ,CSS ,IBM	1
,Cloud Server Cloud ,OPEN CV	
Configuration, Twilio ,Anaconda Navigat	tor
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 Implementations	Mandatory Access	e.g. SHA-256,
		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 balancingto	IBM load balancer
		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

<u>6. PROJECT PLANNING & SCHEDULING</u>

6.1 Sprint Planning & Estimation

Sprint	Functional Requirement (Epic)	User story number	User story/Task	Story points	Priority	Team Members
Sprint-2	Saving the Model	USN-2	As a developer saving the model developed for estimation of fire	10	high	Vigneshini KB Sruthi R Subhikshaa S Swetha A
Sprint-3	Video Analysis	USN-3		10	Medium	Vigneshini KB Sruthi R Subhikshaa S Swetha A
Sprint-3	Twilio Message Service	USN-3		10	Low	Vigneshini KB Sruthi R Subhikshaa S Swetha A
Sprint-4	Alert Sound and Message	USN-4	Sending Alert text message using registered twilio account and produce output sound alert alarm	10	Low	Vigneshini KB Sruthi R Subhikshaa S Swetha A
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	Vigneshini KB Sruthi R Subhikshaa S Swetha A

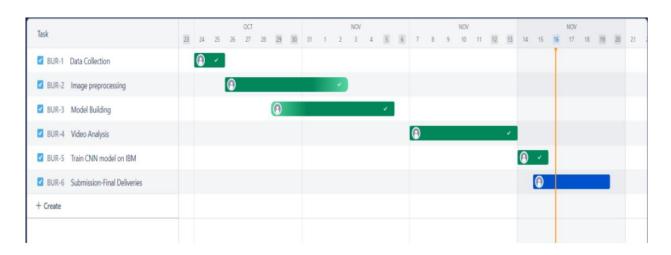
6.2 Sprint Delivery Schedule

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 2022	29 Oct 2022	20	29 Oct 2022
Sprint-2	20	6days	31 Oct 2022	05 Nov 2022	20	05 Nov 2022
Sprint-3	20	6days	07 Nov 2022	12 Nov 2022	20	12 Nov 2022
Sprint-4	20	6days	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.

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 Modeltraining.

importing Required Libraries:

import keras from keras.preprocessing.imageimport
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.layersimport

MaxPooling2D from keras.layersimport

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)	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 importClient 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='AC50d663c8a7c2d8b35b1fc09dfda93bda'
     auth_token='86f345babfa094d1015a0e1137dbb679'client
     =Client(account_sid,auth_token) message=client.messages \
    .create(body='------', reate(body='------',
            from_='+19457581434',to='+916369 659 356')
     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='AC50d663c8a7c2d8b35b1fc09dfda93bda'
    auth token='86f345babfa094d1015a0e1137dbb679' client
    =Client(account_sid,auth_token) message=client.messages \
    .create(body='------', reate(body='------',
          from_='+19457581434',to='+916369 659 356')
    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-31025-1660194638/blob/main/Final%20Deliverables/EntireModel.ipynb

GitHub & Project Demo Link

Github: https://github.com/IBM-EPBL/IBM-Project-31025-1660194638 Demo

Link:

https://drive.google.com/file/d/1MmxM2gr7TgNv_0eDZ6fgQSWnhPxCRcqA/view?usp=share_link