

PROJECT **REPORT**

Project Name: **EMERGING METHODS FOR EARLY DETECTION OF FOREST FIRES.**

Team id: **PNT2022TMID07050**

Team members: **DEVI SRAVANTI GUBBALA -TEAM LEAD.**

AKSHARA YUVARAJ

DIVYASRI.P

ESTHER JANET.S

1. INTRODUCTION

1.1 Project overview

Wildfire, also called forest fire, bush or vegetation fire, can be described as any uncontrolled and non-prescribed combustion or burning of plants in a natural setting such as a forest, grassland, brush land or tundra, which consumes the natural fuels and spreads based on environmental conditions (e.g., wind, topography). 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 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. This is a huge problem which needs to be tackled and thus through this project we provide away to tackle the issue.

1.2 Purpose

The purpose of the project is to detect the forest fire earlier.

2. LITERATURE SURVEY

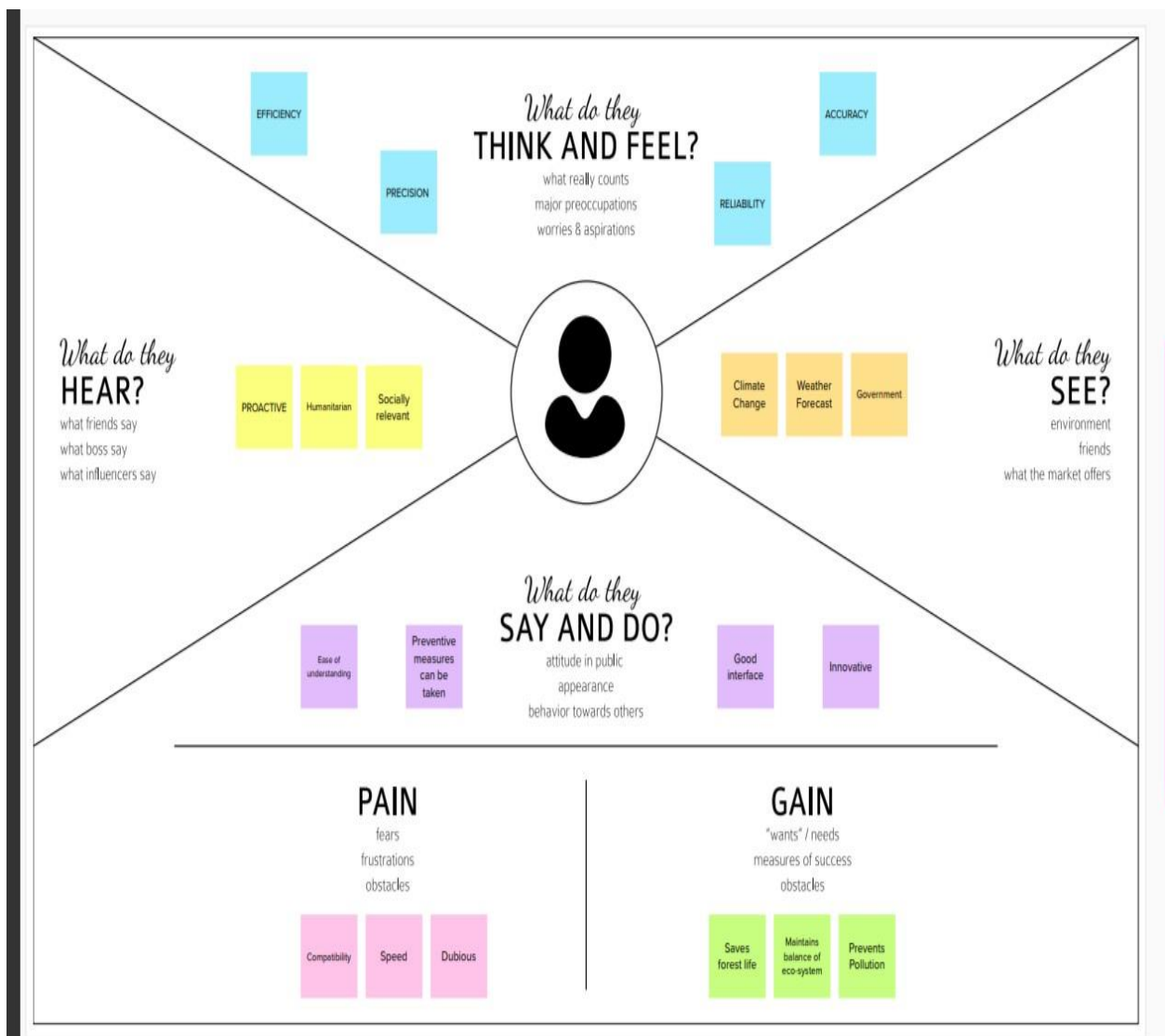
2.1 Reference

S. NO	TITL E	AUTHO R	YEAR
1.	Image Processing for Forest Fire Detection.	Priyadharshini	2016
2.	Forest fire prediction and detection system.	Faroudja Abid	2020


3.	systematic approaches in managing forest fires .	Aditya Dhall	2020
----	--	--------------	------

3. IDEATION & PROPOSED SOLUTION

3.1 Empathy map



3.2 Ideation & Brainstorming



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.

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.

- A Team gathering**
Define who should participate in the session and send an invite. Share relevant information or pre-work ahead.
- B Set the goal**
Think about the problem you'll be focusing on solving in the brainstorming session.
- C Learn how to use the facilitation tools**
Use the Facilitation Superpowers to run a happy and productive session.
[Open article](#) →

3 Group ideas
Take turns sharing your ideas while clustering similar or related notes as you go. In the last 10 minutes, give each cluster a sentence-like label. If a cluster is bigger than six sticky notes, try and see if you and break it up into smaller sub-groups.

Devi

- Alerting GPS, location to be shared to first responder
- Winging alarm, an audible alarm, sound, light and vibration
- Automatically alert the contact of the emergency situation
- Smoking the alarm, alerting the contact of the emergency situation
- Creating a dashboard to show the location of the emergency situation

Esther

- A mobile application which would alert the emergency situation
- NO cameras to detect the smoke during the emergency
- Smoke and gas sensors to detect early fire
- The emergency situation is a place to detect the emergency situation
- Warning, emergency situation, alerting the emergency situation

Divya

- Mobile app that would alert the emergency situation
- Mobile app that would alert the emergency situation
- The emergency situation is a place to detect the emergency situation
- Warning, emergency situation, alerting the emergency situation
- Information about the emergency situation

Aksahara

- NO cameras to detect the smoke during the emergency
- Smoke and gas sensors to detect early fire
- The emergency situation is a place to detect the emergency situation
- Warning, emergency situation, alerting the emergency situation
- Information about the emergency situation

Mobile applications for alerting

- A mobile application which would alert the emergency situation
- Winging alarm, an audible alarm, sound, light and vibration

Sensor based fire detection

- NO cameras to detect the smoke during the emergency
- Smoke and gas sensors to detect early fire

Preparedness measures for controlling fire

- Fire fighters should be alerted in advance and ready to go
- Mobile app that would alert the emergency situation
- Warning, emergency situation, alerting the emergency situation
- Information about the emergency situation

Data for future reference

- Information about the emergency situation
- Creating a dashboard to show the location of the emergency situation

Locating forest fires

- Alerting GPS, location to be shared to first responder
- Warning, emergency situation, alerting the emergency situation

Fire detection using cameras

- NO cameras to detect the smoke during the emergency
- Warning, emergency situation, alerting the emergency situation

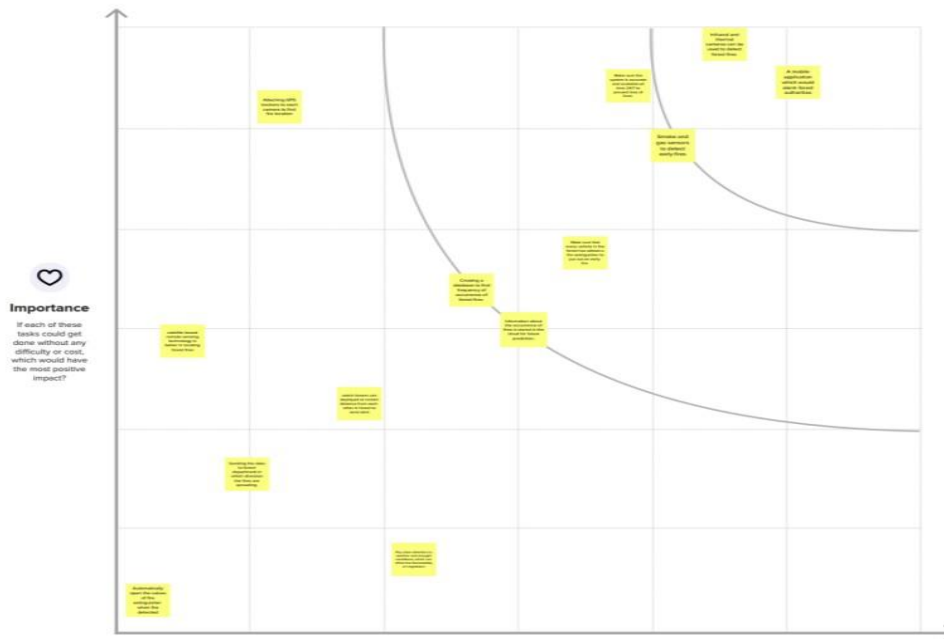
Miscellaneous ideas

- Mobile app that would alert the emergency situation
- Warning, emergency situation, alerting the emergency situation

4

Prioritize

Your team should all be on the same page about what's important moving forward. Place your ideas on this grid to determine which ideas are important and which are feasible.



3.3 Proposed solution

S.NO:	PARAMETERS	REPRESENTATION
1.	Problem Statement (Description of an issue to be addressed)	<ul style="list-style-type: none"> Fire was one of the first and greatest invention of man. But these days due to global warming and climate change, fires have become very violent and destructive. Forest fires are one such evil looming the Earth destroying all the flora and fauna with the devastating fumes and flares it carries with itself. Recent forest fires in California is an evident example of the intensity of the issue and the immediate action that needs to be taken.
2.	Plan of Design and Execution	<ul style="list-style-type: none"> The propose a platform that uses Unmanned Aerial Vehicles (UAVs), which constantly patrol over potentially threatened by fire areas. The UAVs also utilize the benefits from Artificial Intelligence(AI) and are equipped with on-board processing capabilities. This allows them to use computer vision methods for recognition and detection of smoke or fire, based on the still images or the video input from the drone cameras. The system is designed for monitor the causing factors of forest fires such as temperature, humidity , air pressure level,oxygen and Carbon dioxide on the surface of air. The user interacts with a web camera to read the video.

		<ul style="list-style-type: none"> 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. We classify images using a Convolutional Neural Network and use other open CV tools.
3.	Peculiarity/ Novelty	<ul style="list-style-type: none"> Makes use of real time monitoring and allows pre-cursors to potential issues (such as corrosion) to be flagged up and immediately be addressed before major issues occur.
4.	Social Outlook / Customer Friendly	<ul style="list-style-type: none"> Will warn the customers before any fire outbreak. Prevents any potential devastation and issues precautions. Protects the flora and fauna from any unfortunate accidents. Saves forest and human life prevents desertification.
5.	Business Model	<ul style="list-style-type: none"> Focuses more on sensor probes, wireless sensor networks and machine learning which makes the deployment more easier.
6.	Feasibility of Solution	<ul style="list-style-type: none"> Cost effective More performance measure Economical Accurate Effective Reliable Socially intact

3.4 Problem solution fit

Proposed solution fit.		
1.Customer Segment -To adopt a new technology. -For officers who works in forestry department.	2.Problems/Pains -Deterioration of air quality,loss of property ,resources and animal. -Sometimes devices may malfunction.	3.Triggers and emotions -To get prior information of forest fire -It would proceed the misinformation or late details about the forest fire.
4.Customer Limitations -Should have knowledge about the devices. -feature loaded device.	5.Problem Root/Cause -The forest fire starts from natural cause such as lightning. -Less humidity, high temperature may also cause forest fire	6.Your Solutions -We train the model with required algorithm like CNN,images of smoke,fire -Classifying the intensity of the flame using sensors.
7.Available Solution -satellite based system give high resolution image but it provides image of entire earth for every two days,that is long time for fire scanning.	8.Channels of Behavior -They should monitor and checj the device functionalitiy, to alert the smokejumpers. -They should be present at the fire spot with extinguisher and with all saftey precautions.	9.Behavior -It emits a large amount of CO2 which may lead to increase in global warming. -It measures the intensity,light,colour and defines according to its behaviour.

4. REQUIREMENT ANALYSIS

4.1 Functional requirement

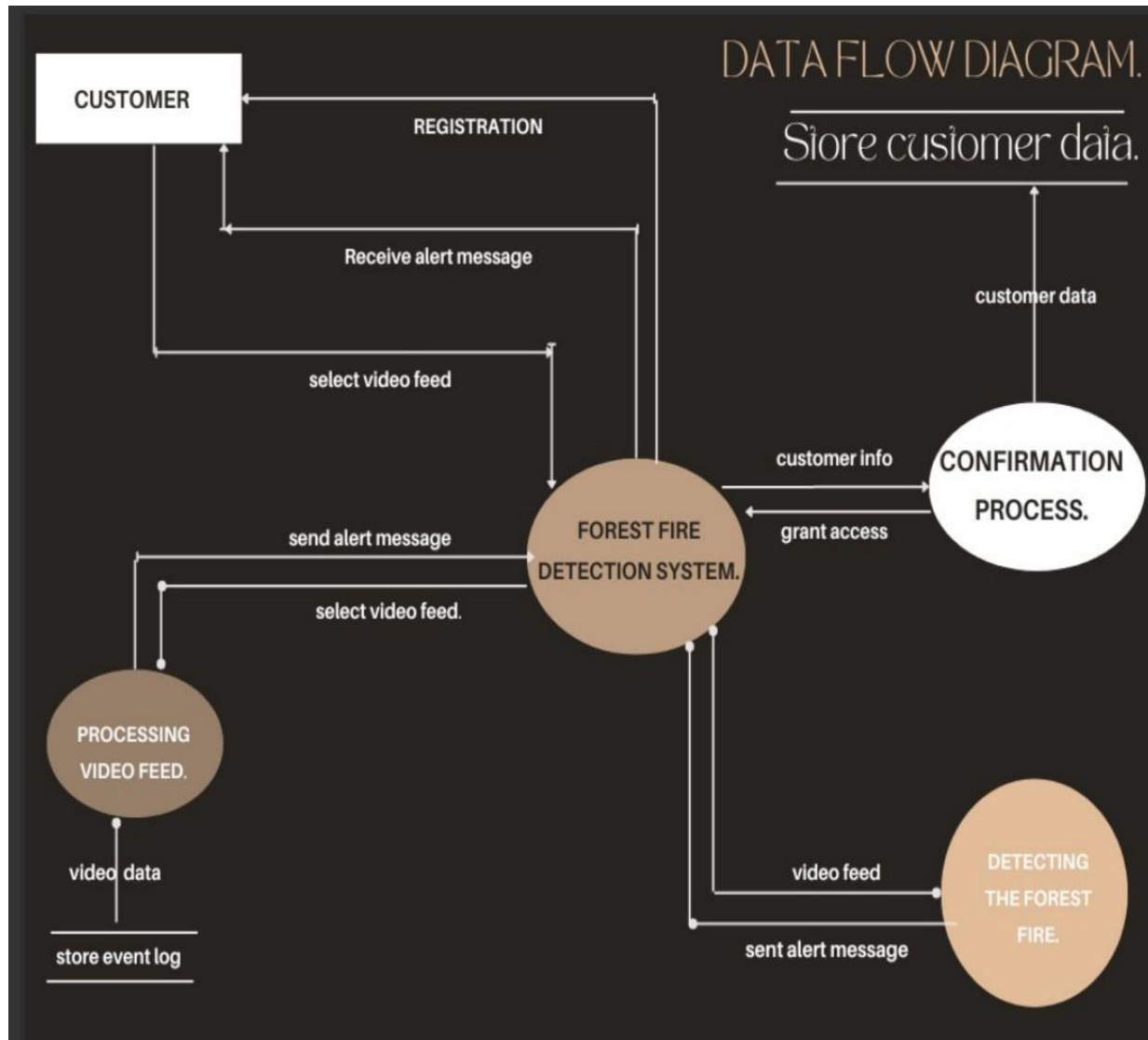
FR. NO.	Functional Requirement	Sub Requirement (Story / Sub-Task)
FR-1	User Registration	Registration through Form Registration through wildfire portal.
FR-2	User Confirmation	Confirmation via Email Confirmation via OTP
FR-3	Data Prediction	Scientists create computer models to predict wildfire potential under a range of potential climate futures. Using different projections of temperature and downfall, scientists predict where and when wildfires are likely to occur

4.2 Non-Functional requirement

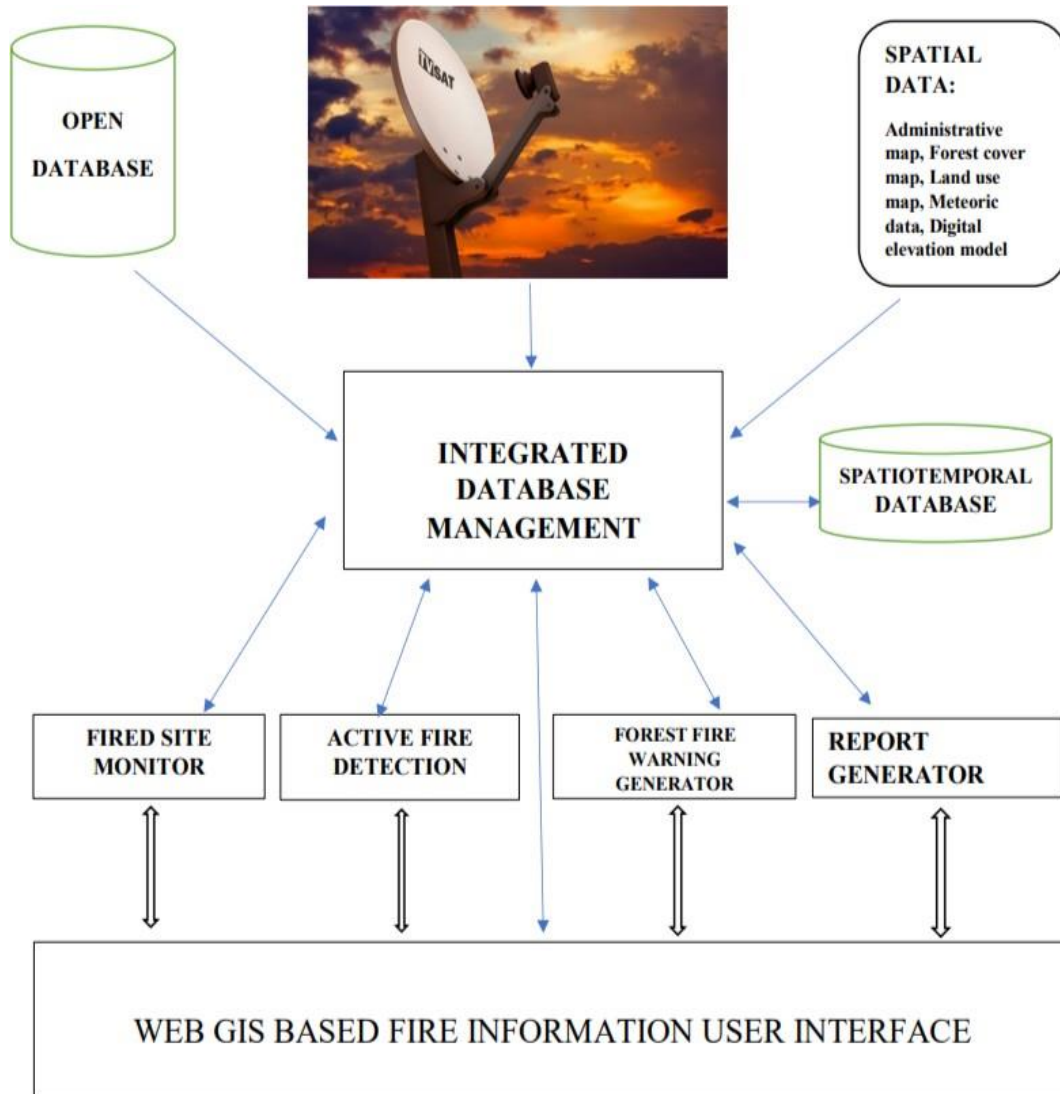
FR. NO.	Non-Functional Requirement	Description
NFR-1	Usability	Many methods have been proposed to detect forest fires, such as camera-based systems, WSN-based systems, and machine learning coating-based systems, with both positive and negative aspects and performance figures of detection.
NFR-2	Protection	We have designed this project to secure the forest from wild fires.
NFR-3	Performance	In the event of a fire, the primary objective of using drones is to gather situational consciousness, which can be used to direct the efforts of the firefighters in locating and controlling hot spots. Just like urban fires, forest fires to require monitoring so that firefighters know what they are dealing with.

5. PROJECT DESIGN

5.1 Data Flow Diagrams



5.2 Solution Architecture



6. PROJECT PLANNING & SCHEDULING

6.1 Sprint Planning & Estimation

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	Image Processing	USN-1	Processing the image to find the fire is detected or not.	1	Medium	1.Devi 2.Esther 3.Divya 4.Akshara

Sprint-1		USN-2	The output would have to give high accuracy.	2	High	1.Devi 2.Esther 3.Divya 4.Akshara
Sprint-2	Video Processing	USN-3	The drone videos will be split into frames to detect the fire.	3	High	1.Devi 2.Esther 3.Divya 4.Akshara
Sprint-3	Alerting	USN-4	After the fire is detected the alert message has to be sent.	2	High	1.Devi 2.Esther 3.Divya 4.Akshara

Sprint-4	Location tracking	USN-5	The exact location of the drone will be predicted and sent along with the alert message.	2	High	1.Devi 2.Esther 3.Divya 4.Akshara
----------	-------------------	-------	--	---	------	--

6.2 Sprint delivery schedule

Project Tracker, Velocity & Burndown Chart:

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	25 Oct 2022	30 Oct 2022	30	30 Oct 2022
Sprint-2	20	6 Days	1 Nov 2022	06 Nov 2022	20	06 Nov 2022
Sprint-3	20	6 Days	08 Nov 2022	13 Nov 2022	20	13 Nov 2022
Sprint-4	20	6 Days	15 Nov 2022	20 Nov 2022	20	20 Nov 2022

Velocity:

Imagine we have a 10-day sprint duration, and the velocity of the team is 20 (points per sprint). Let's now calculate the team's average velocity (AV) per iteration unit (story points per day)

$$AV = \text{Sprint duration} / \text{Velocity} \\ = 20 / 6 = 3$$


```

class_mode = 'binary')
Found 442 images belonging to 2 classes.
Found 121 images belonging to 2 classes.

test_dataset.class_indices

{'forest': 0, 'with fire': 1}

model = keras.Sequential()
model.add(keras.layers.Conv2D(32, (3,3), activation='relu', input_shape=(150,
150, 3)))
model.add(keras.layers.MaxPool2D(2,2))
model.add(keras.layers.Conv2D(64, (3,3), activation='relu'))
model.add(keras.layers.MaxPool2D(2,2))
model.add(keras.layers.Conv2D(128, (3,3), activation='relu'))
model.add(keras.layers.MaxPool2D(2,2))
model.add(keras.layers.Conv2D(128, (3,3), activation='relu'))
model.add(keras.layers.MaxPool2D(2,2))
model.add(keras.layers.Flatten())
model.add(keras.layers.Dense(512, activation='relu'))
model.add(keras.layers.Dense(1, activation='sigmoid'))
model.summary()
Model: "sequential"

```

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 148, 148, 32)	896
max_pooling2d (MaxPooling2D)	(None, 74, 74, 32)	0
conv2d_1 (Conv2D)	(None, 72, 72, 64)	18496
max_pooling2d_1 (MaxPooling2D)	(None, 36, 36, 64)	0
conv2d_2 (Conv2D)	(None, 34, 34, 128)	73856
max_pooling2d_2 (MaxPooling2D)	(None, 17, 17, 128)	0
conv2d_3 (Conv2D)	(None, 15, 15, 128)	147584

max_pooling2d_3 (MaxPooling 2D)	(None, 7, 7, 128)	0
flatten (Flatten)	(None, 6272)	0
dense (Dense)	(None, 512)	3211776
dense_1 (Dense)	(None, 1)	513

```

=====
Total params: 3,453,121
Trainable params: 3,453,121
Non-trainable params: 0

```

```
model.compile(optimizer='adam',loss='binary_crossentropy',metrics=['accuracy'])
```

```

r = model.fit(train_dataset,
              epochs = 10,
              validation_data = test_dataset)

```

```

Epoch 1/10
14/14 [=====] - 140s 10s/step - loss: 0.5568 -
accuracy: 0.7466 - val_loss: 0.2537 - val_accuracy: 0.9504
Epoch 2/10
14/14 [=====] - 37s 3s/step - loss: 0.2948 -
accuracy: 0.8914 - val_loss: 0.0443 - val_accuracy: 0.9835
Epoch 3/10
14/14 [=====] - 38s 3s/step - loss: 0.2028 -
accuracy: 0.9231 - val_loss: 0.1178 - val_accuracy: 0.9752
Epoch 4/10
14/14 [=====] - 37s 3s/step - loss: 0.1618 -
accuracy: 0.9389 - val_loss: 0.0174 - val_accuracy: 1.0000
Epoch 5/10
14/14 [=====] - 35s 3s/step - loss: 0.2101 -
accuracy: 0.9276 - val_loss: 0.0741 - val_accuracy: 0.9835
Epoch 6/10
14/14 [=====] - 37s 3s/step - loss: 0.1757 -
accuracy: 0.9367 - val_loss: 0.1567 - val_accuracy: 0.9174
Epoch 7/10
14/14 [=====] - 39s 3s/step - loss: 0.1656 -
accuracy: 0.9367 - val_loss: 0.0986 - val_accuracy: 0.9504
Epoch 8/10

```

```
14/14 [=====] - 38s 3s/step - loss: 0.1422 -  
accuracy: 0.9502 - val_loss: 0.0220 - val_accuracy: 1.0000
```

```
Epoch 9/10
```

```
14/14 [=====] - 38s 3s/step - loss: 0.1242 -  
accuracy: 0.9615 - val_loss: 0.0337 - val_accuracy: 1.0000
```

```
Epoch 10/10
```

```
14/14 [=====] - 38s 3s/step - loss: 0.0923 -  
accuracy: 0.9706 - val_loss: 0.0392 - val_accuracy: 0.9669
```

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

```
predictions = model.predict(test_dataset)
```

```
predictions = np.round(predictions)
```

```
4/4 [=====] - 6s 1s/step
```

```
predictions
```

```
array([[1.],  
       [0.],  
       [0.],  
       [0.],  
       [0.],  
       [0.],  
       [1.],  
       [0.],  
       [0.],  
       [0.],  
       [1.],  
       [1.],  
       [0.],  
       [1.],  
       [0.],  
       [0.],  
       [1.],  
       [0.],  
       [0.],  
       [0.],  
       [1.],  
       [1.],  
       [1.],  
       [0.],  
       [0.]])
```

[1.],
[0.],
[0.],
[1.],
[1.],
[0.],
[1.],
[1.],
[1.],
[0.],
[0.],
[0.],
[0.],
[1.],
[0.],
[0.],
[1.],
[0.],
[0.],
[1.],
[0.],
[0.],
[1.],
[0.],
[1.],
[1.],
[1.],
[1.],
[1.],
[1.],
[0.],
[0.],
[0.],
[0.],
[1.],
[0.],
[0.],
[0.],
[1.],
[0.],
[0.],
[1.],
[1.],
[0.],
[1.],
[0.],

[0.],
[1.],
[0.],
[1.],
[0.],
[0.],
[0.],
[1.],
[0.],
[0.],
[1.],
[0.],
[0.],
[1.],
[1.],
[0.],
[0.],
[1.],
[1.],
[1.],
[0.],
[0.],
[1.],
[1.],
[0.],
[1.],
[1.],
[1.],
[1.],
[1.],
[0.],
[1.],
[1.],
[0.],
[0.],
[0.],
[1.],
[1.],
[0.],
[1.],
[1.],
[1.],
[1.],
[0.],

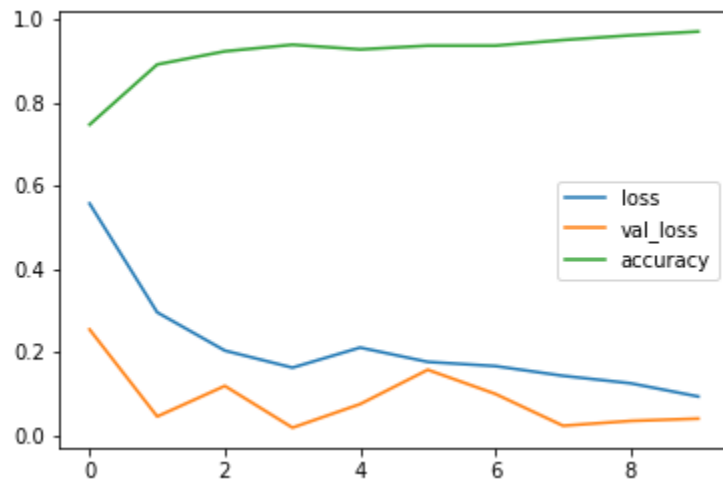

```

[0.],
[0.],
[0.],
[0.],
[0.],
[0.],
[0.],
[0.],
[0.],
[1.]], dtype=float32)

print(len(predictions))
121

import matplotlib.pyplot as plt
plt.plot(r.history['loss'], label='loss')
plt.plot(r.history['val_loss'], label='val_loss')
plt.plot(r.history['accuracy'], label='accuracy')
plt.legend()

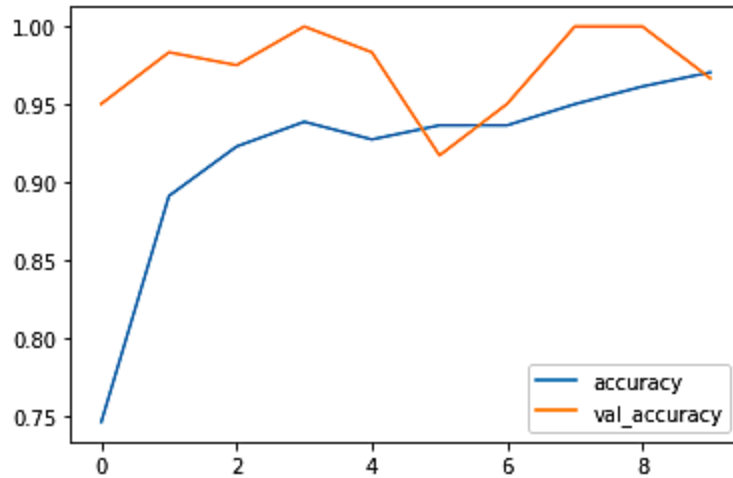
```



```

plt.plot(r.history['accuracy'], label='accuracy')
plt.plot(r.history['val_accuracy'], label='val_accuracy')
plt.legend()

```



```
def predictImage(filename):
    img1 = image.load_img(filename,target_size=(150,150))
    plt.imshow(img1)
    Y = image.img_to_array(img1)
    X = np.expand_dims(Y,axis=0)
    val = model.predict(X)
    print(val)
    if val == 1:
        plt.xlabel("Fire")
    elif val == 0:
        plt.xlabel("No Fire")
```

6.5 Sprint-3

```
{
  "nbformat": 4,
  "nbformat_minor": 0,
  "metadata": {
    "colab": {
      "provenance": []
    },
    "kernelspec": {
      "name": "python3",
      "display_name": "Python 3"
    },
    "language_info": {
      "name": "python"
    }
  }
```

```

},
"cells": [
  {
    "cell_type": "code",
    "execution_count": 1,
    "metadata": {
      "colab": {
        "base_uri": "https://localhost:8080/",
        "height": 35
      },
      "id": "cm0cXpbvyyBp",
      "outputId": "4bffc3ff-b763-4d6d-c12b-02b62f8c32fa"
    },
    "outputs": [
      {
        "output_type": "execute_result", "data": {
          "text/plain": [
            "'/content'"
          ],
          "application/vnd.google.colaboratory.intrinsic+json": { "type": "string" }
        },
        "metadata": {},
        "execution_count": 1
      }
    ],
    "source": [
      "pwd"
    ]
  },
  {
    "cell_type": "code", "source": [
      "!pip install keras\n",
      "!pip install tensorflow\n", "!pip\ninstall opencv-python"
    ],
    "metadata": {
      "colab": {

```

```

"base_uri": "https://localhost:8080/"
},
"id": "UnpPHFm0y4lm",
"outputId": "543ceb28-d9f5-4c1c-9934-02075c827323"
},
"execution_count": 2,
"outputs": [
{
"output_type": "stream",
"name": "stdout",
"text": [
"Looking in indexes: https:// pypi.org/simple, https:// us-python.pkg.dev/colab- wheels/public/simple\n",
"Requirement already satisfied: keras in /usr/local/lib/python3.7/dist-packages (2.9.0)\n", "Looking in
indexes: https:// pypi.org/simple, https:// us-python.pkg.dev/colab-
wheels/public/simple\n",
"Requirement already satisfied: tensorflow in /usr/local/lib/python3.7/dist-packages (2.9.2)\n",
"Requirement already satisfied: six>=1.12.0 in /usr/local/lib/python3.7/dist-packages (from
tensorflow) (1.15.0)\n",
"Requirement already satisfied: keras<2.10.0,>=2.9.0rc0 in /usr/local/lib/python3.7/dist- packages (from
tensorflow) (2.9.0)\n",
"Requirement already satisfied: libclang>=13.0.0 in /usr/local/lib/python3.7/dist- packages (from
tensorflow) (14.0.6)\n",
"Requirement already satisfied: gast<=0.4.0,>=0.2.1 in /usr/local/lib/python3.7/dist- packages (from
tensorflow) (0.4.0)\n",
"Requirement already satisfied: protobuf<3.20,>=3.9.2 in /usr/local/lib/python3.7/dist- packages (from
tensorflow) (3.19.6)\n",
"Requirement already satisfied: flatbuffers<2,>=1.12 in /usr/local/lib/python3.7/dist- packages
(from tensorflow) (1.12)\n",
"Requirement already satisfied: wrapt>=1.11.0 in /usr/local/lib/python3.7/dist-packages (from
tensorflow) (1.14.1)\n",
"Requirement already satisfied: packaging in /usr/local/lib/python3.7/dist-packages (from
tensorflow) (21.3)\n",
"Requirement already satisfied: tensorflow-estimator<2.10.0,>=2.9.0rc0 in
/usr/local/lib/python3.7/dist-packages (from tensorflow) (2.9.0)\n",
"Requirement already satisfied: h5py>=2.9.0 in /usr/local/lib/python3.7/dist-packages (from
tensorflow) (3.1.0)\n",
"Requirement already satisfied: absl-py>=1.0.0 in /usr/local/lib/python3.7/dist-packages (from
tensorflow) (1.3.0)\n",
"Requirement already satisfied: astunparse>=1.6.0 in /usr/local/lib/python3.7/dist-

```

packages (from tensorflow) (1.6.3)\n",

"Requirement already satisfied: typing-extensions>=3.6.6 in /usr/local/lib/python3.7/dist-packages (from tensorflow) (4.1.1)\n",

"Requirement already satisfied: termcolor>=1.1.0 in /usr/local/lib/python3.7/dist-packages (from tensorflow) (2.1.0)\n",

"Requirement already satisfied: grpcio<2.0,>=1.24.3 in /usr/local/lib/python3.7/dist-packages (from tensorflow) (1.50.0)\n",

"Requirement already satisfied: google-pasta>=0.1.1 in /usr/local/lib/python3.7/dist-packages (from tensorflow) (0.2.0)\n",

"Requirement already satisfied: opt-einsum>=2.3.2 in /usr/local/lib/python3.7/dist-packages (from tensorflow) (3.3.0)\n",

"Requirement already satisfied: keras-preprocessing>=1.1.1 in
/usr/local/lib/python3.7/dist-packages (from tensorflow) (1.1.2)\n",

"Requirement already satisfied: numpy>=1.20 in /usr/local/lib/python3.7/dist-packages (from tensorflow) (1.21.6)\n",

"Requirement already satisfied: setuptools in /usr/local/lib/python3.7/dist-packages (from tensorflow) (57.4.0)\n",

"Requirement already satisfied: tensorboard<2.10,>=2.9 in /usr/local/lib/python3.7/dist-packages (from tensorflow) (2.9.1)\n",

"Requirement already satisfied: tensorflow-io-gcs-filesystem>=0.23.1 in
/usr/local/lib/python3.7/dist-packages (from tensorflow) (0.27.0)\n",

"Requirement already satisfied: wheel<1.0,>=0.23.0 in /usr/local/lib/python3.7/dist-packages (from astunparse>=1.6.0->tensorflow) (0.38.3)\n",

"Requirement already satisfied: cached-property in /usr/local/lib/python3.7/dist-packages (from h5py>=2.9.0->tensorflow) (1.5.2)\n",

"Requirement already satisfied: werkzeug>=1.0.1 in /usr/local/lib/python3.7/dist-packages (from tensorboard<2.10,>=2.9->tensorflow) (1.0.1)\n",

"Requirement already satisfied: markdown>=2.6.8 in /usr/local/lib/python3.7/dist-packages (from tensorboard<2.10,>=2.9->tensorflow) (3.4.1)\n",

"Requirement already satisfied: requests<3,>=2.21.0 in /usr/local/lib/python3.7/dist-packages (from tensorboard<2.10,>=2.9->tensorflow) (2.23.0)\n",

"Requirement already satisfied: google-auth<3,>=1.6.3 in /usr/local/lib/python3.7/dist-packages (from tensorboard<2.10,>=2.9->tensorflow) (2.14.1)\n",

"Requirement already satisfied: tensorboard-plugin-wit>=1.6.0 in
/usr/local/lib/python3.7/dist-packages (from tensorboard<2.10,>=2.9->tensorflow) (1.8.1)\n", "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)\n", "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)\n", "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)\n",

```

    "Requirement already satisfied: rsa<5,>=3.1.4 in /usr/local/lib/python3.7/dist-packages(from google-
auth<3,>=1.6.3->tensorboard<2.10,>=2.9->tensorflow) (4.9)\n",
    "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)\n",
    "Requirement already satisfied: requests-oauthlib>=0.7.0 in /usr/local/lib/python3.7/dist- packages
(from google-auth-oauthlib<0.5,>=0.4.1->tensorboard<2.10,>=2.9->tensorflow) (1.3.1)\n",
    "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)\n",
    "Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.7/dist-packages (from importlib-
metadata>=4.4->markdown>=2.6.8->tensorboard<2.10,>=2.9->tensorflow) (3.10.0)\n",
    "Requirement already satisfied: pyasn1<0.5.0,>=0.4.6 in /usr/local/lib/python3.7/dist- packages (from
pyasn1-modules>=0.2.1->google-auth<3,>=1.6.3->tensorboard<2.10,>=2.9-
>tensorflow) (0.4.8)\n",
    "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)\n",
    "Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.7/dist- packages (from
requests<3,>=2.21.0->tensorboard<2.10,>=2.9->tensorflow) (2022.9.24)\n",
    "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)\n",
    "Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages (from
requests<3,>=2.21.0->tensorboard<2.10,>=2.9->tensorflow) (2.10)\n",
    "Requirement already satisfied: oauthlib>=3.0.0 in /usr/local/lib/python3.7/dist-packages(from
requests-oauthlib>=0.7.0->google-auth-oauthlib<0.5,>=0.4.1->tensorboard<2.10,>=2.9-
>tensorflow) (3.2.2)\n",
    "Requirement already satisfied: pyparsing!=3.0.5,>=2.0.2 in /usr/local/lib/python3.7/dist- packages
(from packaging->tensorflow) (3.0.9)\n",
    "Looking in indexes: https:/ pypi.org/simple, https:/ us-python.pkg.dev/colab- wheels/public/simple/\n",
    "Requirement already satisfied: opencv-python in /usr/local/lib/python3.7/dist-packages (4.6.0.66)\n",
    "Requirement already satisfied: numpy>=1.14.5 in /usr/local/lib/python3.7/dist-packages(from opencv-
python) (1.21.6)\n"
]
}
],
{
    "cell_type": "code","source":
[

```

```

"from keras.models import Sequential\n",
"from keras.layers import Dense\n",
"from keras.layers import Convolution2D\n",
"from keras.layers import MaxPooling2D\n",
"from keras.layers import Flatten"
],
"metadata": {
  "id": "LXQp5JUy8by"
},
"execution_count": 3,
"outputs": []
},
{
  "cell_type": "code", "source":
  [
    "from tensorflow.keras.preprocessing.image import ImageDataGenerator\n", "train =
    ImageDataGenerator(rescale=1/255)\n",
    "test = ImageDataGenerator(rescale=1/255)"
  ],
  "metadata": {
    "id": "-joU1JMNzBID"
  },
  "execution_count": 4,
  "outputs": []
},
{
  "cell_type": "code", "source":
  [
    "pwd"
  ],
  "metadata": {
    "colab": {
      "base_uri": "https://localhost:8080/",
      "height": 35
    },
    "id": "3zbBmApYzzob",
    "outputId": "fa8fb36a-473c-4662-dbf7-67598141fa83"
  },
  "execution_count": 8,
  "outputs": [
    {

```



```

"output_type": "execute_result", "data": {
  "text/plain": [
    ""/content""
  ],
  "application/vnd.google.colaboratory.intrinsic+json": { "type":
    "string"
  }
},
"metadata": {},
"execution_count": 8
}
]
},
{
  "cell_type": "code", "source":
  [
    "import os\n",
    "filenames = os.listdir('/content/drive/MyDrive/train_set')"
  ],
  "metadata": {
    "id": "hNu0gAxNz5wV"
  },
  "execution_count": 9,
  "outputs": []
},
{
  "cell_type": "code", "source":
  [
    "x_train = train_dataset = train.flow_from_directory(\"/content/drive/MyDrive/train_set\", \n",
    "                                                    target_size= (64,64),\n",
    "                                                    batch_size = 32,\n",
    "                                                    class_mode = 'binary')\n",
    "x_test = test_dataset = test.flow_from_directory(\"/content/drive/MyDrive/test_set\", \n",
    "                                                  target_size= (64,64),\n",
    "                                                  batch_size = 32,\n",
    "                                                  class_mode = 'binary')\n",
  ],
  "metadata": {
    "colab": {
      "base_uri": "https://localhost:8080/"
    }
  }
}

```

```

    },
    "id": "Ewd7ALq80I2h",
    "outputId": "707c55af-f9b0-4164-e2cf-d642448ec7d1"
  },
  "execution_count": 10,
  "outputs": [
    {
      "output_type": "stream",
      "name": "stdout",
      "text": [
        "Found 442 images belonging to 2 classes.\n",
        "Found 121 images belonging to 2 classes.\n"
      ]
    }
  ]
},
{
  "cell_type": "code",
  "source": [
    "x_test.class_indices"
  ],
  "metadata": {
    "colab": {
      "base_uri": "https://localhost:8080/"
    },
    "id": "Ypg9hbSD0VMb",
    "outputId": "ddf6fee2-231b-4b2a-fc45-156d2c968517"
  },
  "execution_count": 11,
  "outputs": [
    {
      "output_type": "execute_result", "data": {
        "text/plain": [
          "'{'forest': 0, 'with fire': 1}'"
        ]
      },
      "metadata": {},
      "execution_count": 11
    }
  ]
}
]

```

```
},
{
  "cell_type": "code", "source":
  [
    "model = Sequential()"
  ],
  "metadata": {
    "id": "qyQ20wPg0XZg"
  },
  "execution_count": 12,
  "outputs": []
},
{
  "cell_type": "code", "source":
  [
    "model.add(Convolution2D(32,(3,3),activation='relu',input_shape=(64,64,3)))"
  ],
  "metadata": {
    "id": "3CCe1wJK0dq6"
  },
  "execution_count": 13,
  "outputs": []
},
{
  "cell_type": "code", "source":
  [
    "model.add(MaxPooling2D(2,2))"
  ],
  "metadata": {
    "id": "KiB6bTwt0gYl"
  },
  "execution_count": 14,
  "outputs": []
},
{
  "cell_type": "code",
  "source": [
    "model.add(Flatten())"
  ],
  "metadata": {
    "id": "gezPUvME0ixZ"
```

```

    },
    "execution_count": 15,
    "outputs": []
  },
  {
    "cell_type": "code", "source":
    [
      "model.add(Dense(512,activation='relu'))\n",
      "model.add(Dense(1,activation='sigmoid'))"
    ],
    "metadata": {
      "id": "jITuRAuQ0l7j"
    },
    "execution_count": 16,
    "outputs": []
  },
  {
    "cell_type": "code", "source":
    [
      "model.compile(optimizer=\"adam\",loss=\"binary_crossentropy\",metrics=[\"accuracy\"])"
    ],
    "metadata": {
      "id": "3VRs-oXD0quq"
    },
    "execution_count": 17,
    "outputs": []
  },
  {
    "cell_type": "code", "source":
    [
      "model.fit(x_train,steps_per_epoch=14
,epochs=10,validation_data=x_test,validation_steps=4)"
    ],
    "metadata": {
      "colab": {
        "base_uri": "https://localhost:8080/"
      },
      "id": "7Ek-Gm6P0vpW",
      "outputId": "f6a1472f-709a-4f36-9d4f-fa1946838e10"
    },
    "execution_count": 18,

```

```

"outputs": [
  {
    "output_type": "stream",
    "name": "stdout",
    "text": [
      "Epoch 1/10\n",
      "14/14 [=====] - 157s 11s/step - loss: 3.7764 - accuracy:
0.5928 - val_loss: 0.3833 - val_accuracy: 0.8182\n", "Epoch
2/10\n",
      "14/14 [=====] - 22s 2s/step - loss: 1.0376 - accuracy:
0.6855 - val_loss: 0.1756 - val_accuracy: 0.9339\n", "Epoch
3/10\n",
      "14/14 [=====] - 21s 1s/step - loss: 0.2968 - accuracy:
0.8688 - val_loss: 0.1248 - val_accuracy: 0.9835\n", "Epoch
4/10\n",
      "14/14 [=====] - 21s 2s/step - loss: 0.2413 - accuracy:
0.9072 - val_loss: 0.1233 - val_accuracy: 0.9504\n", "Epoch
5/10\n",
      "14/14 [=====] - 21s 2s/step - loss: 0.1790 - accuracy:
0.9321 - val_loss: 0.0887 - val_accuracy: 0.9669\n", "Epoch
6/10\n",
      "14/14 [=====] - 21s 2s/step - loss: 0.1427 - accuracy:
0.9457 - val_loss: 0.0762 - val_accuracy: 0.9752\n", "Epoch
7/10\n",
      "14/14 [=====] - 21s 2s/step - loss: 0.1059 - accuracy:
0.9706 - val_loss: 0.0514 - val_accuracy: 0.9917\n", "Epoch
8/10\n",
      "14/14 [=====] - 21s 1s/step - loss: 0.0835 - accuracy:
0.9774 - val_loss: 0.0272 - val_accuracy: 1.0000\n", "Epoch
9/10\n",
      "14/14 [=====] - 21s 2s/step - loss: 0.0657 - accuracy:
0.9774 - val_loss: 0.0266 - val_accuracy: 0.9917\n", "Epoch
10/10\n",
      "14/14 [=====] - 23s 2s/step - loss: 0.0465 - accuracy:
0.9819 - val_loss: 0.0153 - val_accuracy: 1.0000\n"
    ]
  },
  {
    "output_type": "execute_result", "data": {
      "text/plain": [

```

```

    "<keras.callbacks.History at 0x7f2c0d8ec590>"
  ]
},
"metadata": {},
"execution_count": 18
}
]
},
{
  "cell_type": "code", "source":
  [
    "model.save(\"forest1.h5\")"
  ],
  "metadata": {
    "id": "wqMJztaF00Qh"
  },
  "execution_count": 19,
  "outputs": []
},
{
  "cell_type": "code", "source":
  [
    "!tar -zcvf image-classification-model_new.tgz forest1.h5"
  ],
  "metadata": {
    "colab": {
      "base_uri": "https://localhost:8080/"
    },
    "id": "3jah6H9-2Znl",
    "outputId": "ed7217d8-6994-4f98-b136-98b1ff5f6b8b"
  },
  "execution_count": 20,
  "outputs": [
    {
      "output_type": "stream",
      "name": "stdout",
      "text": [
        "forest1.h5\n"
      ]
    }
  ]
}
]

```

```

},
{
  "cell_type": "code", "source":
  [
    "ls -l"
  ],
  "metadata": {
    "colab": {
      "base_uri": "https://localhost:8080/"
    },
    "id": "V9oOmNUW2cWk",
    "outputId": "c706f1e1-1c8d-4b26-d8dd-a4bcdfe60e5c"
  },
  "execution_count": 21,
  "outputs": [
    {
      "output_type": "stream",
      "name": "stdout",
      "text": [
        "\u001b[0m\u001b[01;34mdrive\u001b[0m^\n",
        "forest1.h5\n",
        "image-classification-model_new.tgz\n",
        "\u001b[01;34msample_data\u001b[0m^\n"
      ]
    }
  ]
},
{
  "cell_type": "code", "source":
  [
    "!pip install watson-machine-learning-client --upgrade"
  ],
  "metadata": {
    "colab": {
      "base_uri": "https://localhost:8080/",
      "height": 1000
    },
    "id": "nQj_2bZ62ns3",
    "outputId": "103e599b-947e-46f0-eb02-0dd7142a2130"
  },
  "execution_count": 22,

```



```

"outputs": [
{
  "output_type": "stream",
  "name": "stdout",
  "text": [
    "Looking in indexes: https:// pypi.org/simple, https:// us-python.pkg.dev/colab-
wheels/public/simple/\n",
    "Collecting watson-machine-learning-client\n",
    "  Downloading watson_machine_learning_client-1.0.391-py3-none-any.whl (538 kB)\n",
    "\u001b[K [REDACTED] 538 kB 7.0 MB/s \n",
    "\u001b[?25hRequirement already satisfied: requests in /usr/local/lib/python3.7/dist-
packages (from watson-machine-learning-client) (2.23.0)\n",
    "Requirement already satisfied: tqdm in /usr/local/lib/python3.7/dist-packages (from watson-
machine-learning-client) (4.64.1)\n",
    "Requirement already satisfied: tabulate in /usr/local/lib/python3.7/dist-packages (from watson-
machine-learning-client) (0.8.10)\n",
    "Collecting lomond\n",
    "  Downloading lomond-0.3.3-py2.py3-none-any.whl (35 kB)\n",
    "Requirement already satisfied: certifi in /usr/local/lib/python3.7/dist-packages (from watson-
machine-learning-client) (2022.9.24)\n",
    "Requirement already satisfied: urllib3 in /usr/local/lib/python3.7/dist-packages (from watson-
machine-learning-client) (1.24.3)\n",
    "Collecting boto3\n",
    "  Downloading boto3-1.26.11-py3-none-any.whl (132 kB)\n",
    "\u001b[K [REDACTED] 132 kB 53.7 MB/s \n",
    "\u001b[?25hRequirement already satisfied: pandas in /usr/local/lib/python3.7/dist-
packages (from watson-machine-learning-client) (1.3.5)\n",
    "Collecting
ibm-cos-sdk\n",
    "  Downloading ibm-cos-sdk-2.12.0.tar.gz (55 kB)\n",
    "\u001b[K [REDACTED] 55 kB 3.9 MB/s \n",
    "\u001b[?25hCollecting jmespath<2.0.0,>=0.7.1\n",
    "  Downloading jmespath-1.0.1-py3-none-any.whl (20 kB)\n",
    "Collecting
s3transfer<0.7.0,>=0.6.0\n",
    "  Downloading s3transfer-0.6.0-py3-none-any.whl (79 kB)\n",
    "\u001b[K [REDACTED] 79 kB 9.5 MB/s \n",
    "\u001b[?25hCollecting botocore<1.30.0,>=1.29.11\n",
    "  Downloading botocore-1.29.11-py3-none-any.whl (9.9 MB)\n",
    "\u001b[K [REDACTED] 9.9 MB 45.4 MB/s \n",
    "\u001b[?25hRequirement already satisfied: python-dateutil<3.0.0,>=2.1 in
/usr/local/lib/python3.7/dist-packages (from botocore<1.30.0,>=1.29.11->boto3->watson- machine-
learning-client) (2.8.2)\n",

```

```
"Collecting urllib3\n",
" Downloading urllib3-1.26.12-py2.py3-none-any.whl (140 kB)\n",
"\u001b[K [REDACTED] 140 kB 42.1 MB/s \n",
"\u001b[?25hRequirement already satisfied: six>=1.5 in /usr/local/lib/python3.7/dist-
packages (from python-dateutil<3.0.0,>=2.1->botocore<1.30.0,>=1.29.11->boto3->watson- machine-
learning-client) (1.15.0)\n",
"Collecting ibm-cos-sdk-core==2.12.0\n",
" Downloading ibm-cos-sdk-core-2.12.0.tar.gz (956 kB)\n",
"\u001b[K [REDACTED] 956 kB 51.7 MB/s \n",
"\u001b[?25hCollecting ibm-cos-sdk-s3transfer==2.12.0\n",
" Downloading ibm-cos-sdk-s3transfer-2.12.0.tar.gz (135 kB)\n",
"\u001b[K [REDACTED] 135 kB 54.2 MB/s \n",
"\u001b[?25hCollecting jmespath<2.0.0,>=0.7.1\n",
" Downloading jmespath-0.10.0-py2.py3-none-any.whl (24 kB)\n", "Collecting
requests\n",
" Downloading requests-2.28.1-py3-none-any.whl (62 kB)\n",
"\u001b[K [REDACTED] 62 kB 1.6 MB/s \n",
"\u001b[?25hRequirement already satisfied: charset-normalizer<3,>=2 in
/usr/local/lib/python3.7/dist-packages (from requests->watson-machine-learning-client)(2.1.1)\n",
"Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.7/dist-packages (from
requests->watson-machine-learning-client) (2.10)\n",
"Requirement already satisfied: pytz>=2017.3 in /usr/local/lib/python3.7/dist-packages (from pandas-
>watson-machine-learning-client) (2022.6)\n",
"Requirement already satisfied: numpy>=1.17.3 in /usr/local/lib/python3.7/dist-packages(from pandas-
>watson-machine-learning-client) (1.21.6)\n",
"Building wheels for collected packages: ibm-cos-sdk, ibm-cos-sdk-core, ibm-cos-sdk-s3transfer\n",
" Building wheel for ibm-cos-sdk (setup.py) ... \u001b[?25l\u001b[?25hdone\n",
" Created wheel for ibm-cos-sdk: filename=ibm_cos_sdk-2.12.0-py3-none-any.whl size=73931
sha256=841189e9104158317d85f159529014a3c34da1db4455cc140ecfd657ba3ed2ef\n",
" Stored in directory:
/root/.cache/pip/wheels/ec/94/29/2b57327cf00664b6614304f7958abd29d77ea0e5bbece2ea57\n",
" Building wheel for ibm-cos-sdk-core (setup.py) ... \u001b[?25l\u001b[?25hdone\n", " Created
wheel for ibm-cos-sdk-core: filename=ibm_cos_sdk_core-2.12.0-py3-none-
any.whl size=562962 sha256=6dd5fd11a6eb4cc566eefe7e82e573055238fbc5bdafc2604c164f8a6fa02255\n",
" Stored in directory:
/root/.cache/pip/wheels/64/56/fb/5cd6f4f40406c828a5289b95b2752a4d142a9afb359244ed8
```

```

d\n",
    " Building wheel for ibm-cos-sdk-s3transfer (setup.py) ...
\u001b[?25l\u001b[?25hdone\n",
    " Created wheel for ibm-cos-sdk-s3transfer: filename=ibm_cos_sdk_s3transfer-2.12.0- py3-none-
any.whl size=89778
sha256=3c9215c3ddaa7fc31a8c3783a78b5e3aa7a4cb9ea8d7dc1178e709c1ccb392a8\n",
    " Stored in directory:
/root/.cache/pip/wheels/57/79/6a/ffe3370ed7ebc00604f9f76766e1e0348dcdcad2b2e32df9e1
\n",
    "Successfully built ibm-cos-sdk ibm-cos-sdk-core ibm-cos-sdk-s3transfer\n", "Installing collected
packages: urllib3, requests, jmespath, ibm-cos-sdk-core, botocore,
s3transfer, ibm-cos-sdk-s3transfer, lomond, ibm-cos-sdk, boto3, watson-machine-learning-client\n",
    " Attempting uninstall: urllib3\n",
    " Found existing installation: urllib3 1.24.3\n", "
    Uninstalling urllib3-1.24.3:\n",
    " Successfully uninstalled urllib3-1.24.3\n", "
    Attempting uninstall: requests\n",
    " Found existing installation: requests 2.23.0\n", "
    Uninstalling requests-2.23.0:\n",
    " Successfully uninstalled requests-2.23.0\n",
    "Successfully installed boto3-1.26.11 botocore-1.29.11 ibm-cos-sdk-2.12.0 ibm-cos-sdk- core-2.12.0
ibm-cos-sdk-s3transfer-2.12.0 jmespath-0.10.0 lomond-0.3.3 requests-2.28.1 s3transfer-0.6.0 urllib3-
1.26.12 watson-machine-learning-client-1.0.391\n"
]
},
{
    "output_type": "display_data",
    "data": {
        "application/vnd.colab-display-data+json": {
            "pip_warning": {
                "packages": [
                    "requests",
                    "urllib3"
                ]
            }
        }
    },
    "metadata": {}
}
]

```

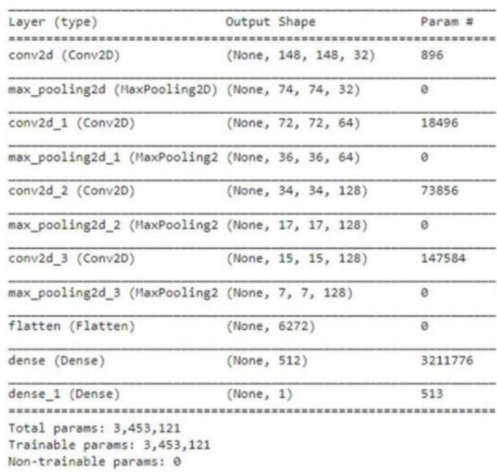
```

},
{
  "cell_type": "code", "source":
  [
    "def guid_from_space_name(client, space_name):\n", "
      space = client.spaces.get_details()\n",
    "  #print(space)\n",
    "  return(next(item for item in space['resources'] if item['entity']['name'] ==
space_name)['metadata']['id'])"
  ],
  "metadata": {
    "id": "QSDKfvy_3H8Q"
  },
  "execution_count": 25,
  "outputs": []
}
]
}

```

7. TESTING AND RESULTS

7.1 Performance Testing

S. No	Parameter	Values	Screenshot
1.	Model Summary	3,453,213	 <pre> Layer (type) Output Shape Param # ----- conv2d (Conv2D) (None, 148, 148, 32) 896 max_pooling2d (MaxPooling2D) (None, 74, 74, 32) 0 conv2d_1 (Conv2D) (None, 72, 72, 64) 18496 max_pooling2d_1 (MaxPooling2 (None, 36, 36, 64) 0 conv2d_2 (Conv2D) (None, 34, 34, 128) 73856 max_pooling2d_2 (MaxPooling2 (None, 17, 17, 128) 0 conv2d_3 (Conv2D) (None, 15, 15, 128) 147584 max_pooling2d_3 (MaxPooling2 (None, 7, 7, 128) 0 flatten (Flatten) (None, 6272) 0 dense (Dense) (None, 512) 3211776 dense_1 (Dense) (None, 1) 513 ----- Total params: 3,453,121 Trainable params: 3,453,121 Non-trainable params: 0 </pre>

2.	Accuracy	<p>Training Accuracy - 0.9663</p> <p>Validation Accuracy -0.9795</p>	<p>Epoch 1/10 14/14 [=====] - 96s 7s/step - loss: 0.5717 - accuracy: 0.6552 - val_loss: 0.2805 - val_accuracy: 0.8750</p> <p>Epoch 2/10 14/14 [=====] - 84s 6s/step - loss: 0.3266 - accuracy: 0.8434 - val_loss: 0.1193 - val_accuracy: 0.9667</p> <p>Epoch 3/10 14/14 [=====] - 74s 5s/step - loss: 0.2247 - accuracy: 0.9227 - val_loss: 0.1184 - val_accuracy: 0.9500</p> <p>Epoch 4/10 14/14 [=====] - 75s 5s/step - loss: 0.1682 - accuracy: 0.9425 - val_loss: 0.0206 - val_accuracy: 1.0000</p> <p>Epoch 5/10 14/14 [=====] - 82s 6s/step - loss: 0.1173 - accuracy: 0.9610 - val_loss: 0.0631 - val_accuracy: 0.9667</p> <p>Epoch 6/10 14/14 [=====] - 76s 6s/step - loss: 0.0925 - accuracy: 0.9741 - val_loss: 0.0169 - val_accuracy: 1.0000</p> <p>Epoch 7/10 14/14 [=====] - 80s 6s/step - loss: 0.0804 - accuracy: 0.9714 - val_loss: 0.0194 - val_accuracy: 1.0000</p> <p>Epoch 8/10 14/14 [=====] - 72s 5s/step - loss: 0.1030 - accuracy: 0.9661 - val_loss: 0.0435 - val_accuracy: 0.9833</p> <p>Epoch 9/10 14/14 [=====] - 70s 5s/step - loss: 0.1032 - accuracy: 0.9676 - val_loss: 0.0444 - val_accuracy: 1.0000</p> <p>Epoch 10/10 14/14 [=====] - 92s 6s/step - loss: 0.1144 - accuracy: 0.9665 - val_loss: 0.0807 - val_accuracy: 0.9833</p>
----	----------	--	--

7.2 User acceptance testing

Resolution	Severity 1	Severity 2	Severity 3	Severity 4	Subtotal
By Design	1	1	2	0	4
Duplicate	0	0	0	0	0
External	0	0	2	1	3
Fixed	4	2	4	1	11
Not Reproduced	0	0	0	0	0
Skipped	0	0	1	1	2
Won't Fix	0	0	0	1	1
Totals	5	3	9	4	21

7.3 Test case

				Date	08-Nov-2022			
				Team ID	PNT2022TMID07050			
				Project Name	Emerging Methods for Early Detection of Forest fire			
				Maximum Marks	4 marks			
Test case ID	Feature Type	Component	Test Scenario	Expected Result	Actual Result	Status	BUGID	Executed By
Home Page_TC_O01	UI	Home Page	Display the Emerging Methods of Forest Fires	Displaying the Home Page	Home Page displayed	Pass	-	Devi, Esther
Home Page_TC_O02	UI	Home Page	Displayed the Prediction Forest Fires	Displaying the content of home page	Content of Homepage is displayed	Pass	-	Divya, Akshara
Home Page_TC_O03	Functional	Home page	Checks whether the Drop the Image Here! Button is visible	Displays the Button	Drop the Image Here! Button is pops up.	Pass	-	Devi, Esther
PredictedPage_TC_O04	Functional	Predicted page	Display the Prediction Page and Choose Image Button	Displays the Prediction Page and the Choose Image Button	Prediction page displayed. Choose image button was clicked.	Pass	-	Divya, Akshara, Devi, Esther.
PredictedPage_TC_O04	Functional	Predicted page	Select the Image and Click the Predict Button	Display the selected Image	Displays the Selected Image	Pass	-	Devi, Esther

7.4 Test case analysis

Section	Total Cases	Not Tested	Fails	Passes
Client Application	10	0	0	10
Security	2	0	0	2
Performance	2	0	0	2
Exception Reporting	2	0	0	2
Final Report Output	3	0	0	3

8. ADVANTAGES & DISADVANTAGES

ADVANTAGES:

- > The proposed system detects the forest fire at a faster rate compared to existing system. It has enhanced data collection feature.
- > The major aspect is that it reduces false alarm and also has accuracy due to various sensors present.
- > It minimize the human effort as it works automatically. This is meagre -cost due to which can be easily accessed.
- > The main objective of our project is to receive an alert message through an app to the respective user.

DISADVANTAGES:

- > The electrical interference diminishes the potency of radio receiver.
- > The main drawback is that it has less coverage range areas

9. CONCLUSION

This type of system is the first of its kind to ensure no further damage is then to forests when there is a fire breakout and instantly a message is sent to the user through the App. Immediate response or early warning to a fire breakout is mostly the only way to avoid losses and biology, cultural heritage damages to a great extent. Therefore the most important goals in fire surveillance are quick and authentic detection of fire. It is so much easier to suppress fire while it is in its early stages. info about the progress of fire is highly valuable for managing fire during all its stages. Based on this data the firefighting staff can be guided on target to block fire before it reaches cultural heritage sites and to suppress it quickly by utilizing required firefighting equipment and vehicles. With further research and invention, this project can be implemented in various forest areas so that we can save our forests and maintain great environs.

10. FUTURE SCOPE

This project is far from complete and there is a lot of room for betterment. Some of the betterment that can be made to this project are as follows:

An Additional pump can be added so that it automatically sends water when there is a fire breakout. Also industrial sensors can be used for better ranging and accuracy.

→ This project has endless potential and can always be enhanced to become better. enforce this concept in the real world will benefit several industries and reduce the workload on many workers, enhancing overall work efficiency.

GitHub :

<https://github.com/IBM-EPBL/IBM-Project-19171-1659694051>