Table of contents

Chapter	Tittle	Pg.no
no		
1.	INTRODUCTION	2
	1.1 Project Overview	2 3 4
•	1.2 Purpose	3
2.	LITERATURE SURVEY	4
	2.1 Existing problem	4 5
	2.2 References	
	2.3 Problem Statement Definition	6
3.	IDEATION & PROPOSED SOLUTION	7
	3.1 Empathy Map Canvas	7
	3.2 Ideation & Brainstorming	8
	3.3 Proposed Solution	10
	3.4 Problem Solution fit	11
4.	REQUIREMENT ANALYSIS	12
	4.1 Functional requirement	12
	4.2 Non-Functional requirements	13
5.	PROJECT DESIGN	14
	5.1 Data Flow Diagrams	14
	5.2 Solution & Technical Architecture	16
	5.3 User Stories	17
6.	PROJECT PLANNING & SCHEDULING	19
	6.1 Sprint Planning & Estimation	19
	6.2 Sprint Delivery Schedule	20
	6.3 Reports from JIRA	20
7.	CODING & SOLUTIONING	21
	7.1 Feature 1	21
	7.2 Feature 2	21
8.	TESTING	22
	8.1 Test Cases	23
	8.2 User Acceptance Testing	24
9.	RESULTS	24
	9.1 Performance Metrics	24
10.	ADVANTAGES & DISADVANTAGES	25
11.	CONCLUSION	26
12.	FUTURE SCOPE	27
13.	APPENDIX	28
	13.1 Source Code	28
	13.2 GitHub & Project Demo Link	37
	$\boldsymbol{\mathcal{J}}$	

INTRODUCTION

1.1Project Overview

Natural disasters not only disturb the human ecological system but also destroy the properties and critical infrastructures of human societies and even lead to permanent change in the ecosystem. Disaster can be caused by naturally occurring events such as earthquakes, cyclones, floods, and wildfires. Many deep learning techniques have been applied by various researchers to detect and classify natural disasters to overcome losses in ecosystems, but detection of natural disasters still faces issues due to the complex and imbalanced structures of images. To tackle this problem, we developed a multilayered deep convolutional neural network model that classifies the natural disaster and tells the intensity of disaster of natural. The model uses an integrated webcam to capture the video frame and the video frame is compared with the Pre-trained model and the type of disaster is identified and showcased on the OpenCV window.

1.2 Purpose

The main purpose of this model is to detect and classify the type of disaster with a high accuracy rate. To prevent natural disasters in the future, said model can be used to predict future disasters and take some action against heavy loss of human ecological systems and property. The proposed model works in two blocks: Block-I convolutional neural network (B-I CNN), for detection and occurrence of disasters, and Block-II convolutional neural network (B-II CNN), for classification of natural disaster intensity types with different filters and parameters

The model is tested on 4428 natural images and performance is calculated and expressed as different statistical values: sensitivity (SE), 97.54%; specificity (SP), 98.22%; accuracy rate (AR), 99.92%; precision (PRE), 97.79%; and F1-score (F1), 97.97%. The overall accuracy for the whole model is 99.92%, which is competitive and comparable with state-of-the-art algorithms.

Natural disasters not only disturb the human ecological system but also destroy the properties and critical infrastructures of human societies and even lead to permanent change in the ecosystem. Disaster can be caused by naturally occurring events such as earthquakes, cyclones, floods, and wildfires. Many deep learning techniques have been applied by various researchers to detect and classify natural disasters to overcome losses in ecosystems, but detection of natural disasters still faces issues due to the complex and imbalanced structures of images. To tackle this problem, we propose a multilayered deep convolutional neural network.

LITERATURE SURVEY

2.1 Existing Problem

The detection of natural disasters by using deep learning techniques still faces various issues due to noise and serious class imbalance problems. But for a model to have sustained accuracy, the conditions above are necessary. ML techniques have the advantage of immediately filtering images, which would have required months to be sorted manually. Temporary settlements can also be detected, indicating areas of survivors in need. A number of researchers have worked on this type of disaster classification either on twitter datasets or on some particular website of news.

The input disaster information is converted into T-CAP so that it can be linked to the currently operating UHD broadcasting system. It also introduced the disaster information delivery process issued by the emulator and introduces the CAP element values specifically defined in the T-CAP. The risk of earthquake disaster in six western provinces of China is evaluated by analytic hierarchy process (AHP), and the evaluation results are compared. Due to climate changes, many natural disasters have become more serious, such as the intensity of typhoons that are getting higher every year. The characteristics of such a disaster is high-intensity but low-probability event (HILP).

We proposed multilayered deep convolutional neural network algorithm. We also need to develop better ways of communicating the results of models, including their limitations and uncertainties, so that decision makers better understand and appreciate both the predictions and the limitations of these prediction. Number of disasters and pandemics has caused a strain on the emergency services, and this is where ML algorithms are required to work efficiently and make the best use of existing resources.

2.2 References

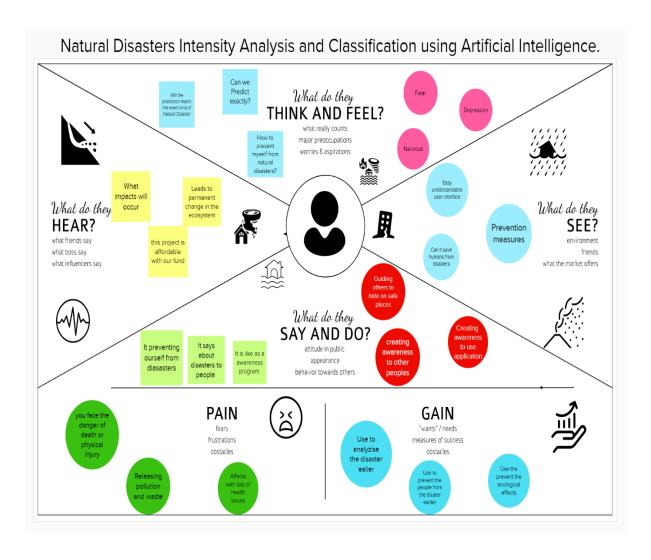
- 1. Mignan A., Broccardo M. Neural network applications in earthquake prediction (1994–2019): Meta-analytic and statistical insights on their limitations. Seism. Res. Lett. 2020;91:2330–2342. doi: 10.1785/0220200021.
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2.3Problem Statement Definition

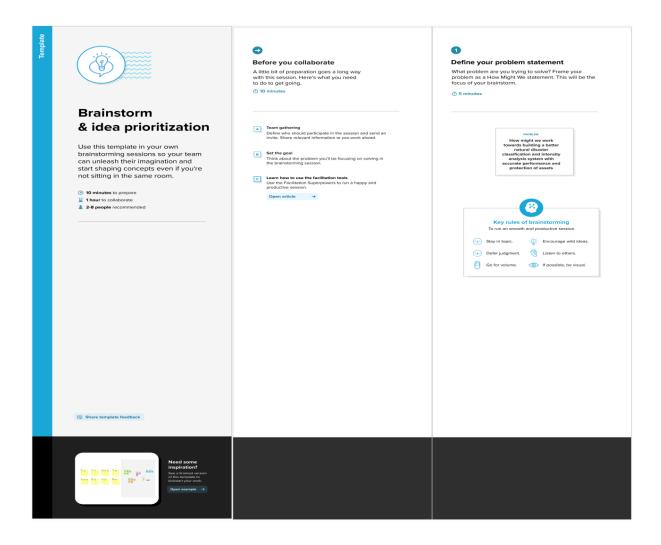
- Natural Disaster one of most inevitable disasters, it can be caused by naturally occurring events such as earthquakes, cyclones, floods, and wildfires.
- A natural disaster can cause loss of life or damage property like buildings will
 collapse due to seismological effects, and typically leaves some economic
 damage in its wake. diseases/viruses spread and sometimes natural disasters
 can devastate nations.
- Many deep learning techniques have been applied by various researchers to
 detect and classify natural disasters to overcome losses in ecosystems, but
 detection of natural disasters still faces issues due to the complex and
 imbalanced structures of images.
- The defined problem statements lead to focus on building a multilayered deep Convolutional Neural Network (CNN) model that classifies the natural disaster and monitors the intensity of its occurrence.
- For detection, an integrated webcam is used to capture the video frame and is compared with the pre-trained model and the type of disaster is identified and showcased on the user interface.

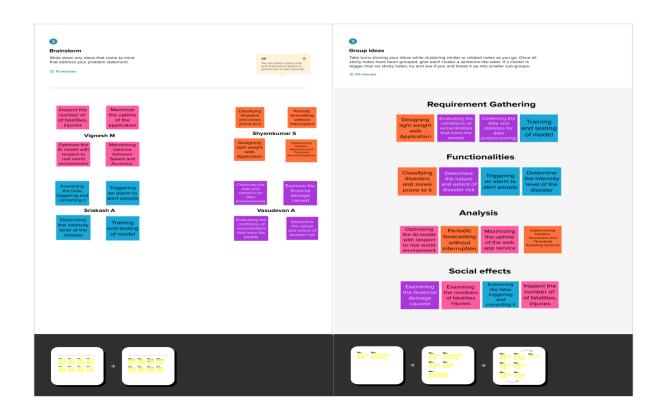
IDEATION & PROPOSED SOLUTION

3.1 Empathy Map Canvas



3.2 Ideation & Brainstorming



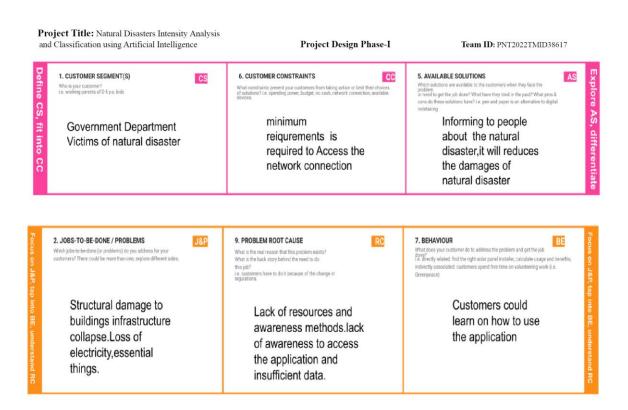


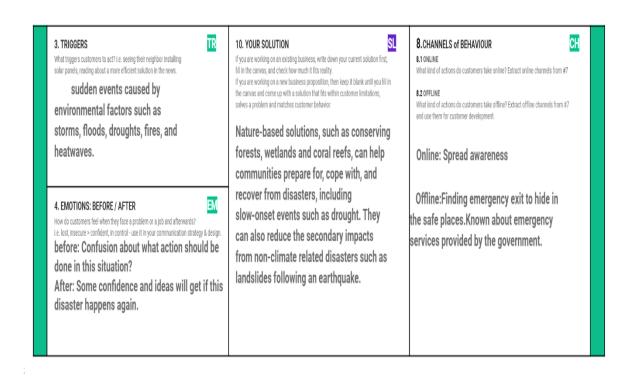


3.3 Proposed Solution

S. No	Parameter	Description					
1.	Problem Statement (Problem to	humans need a way to describe and					
	be solved)	analyse the disaster early so that they ca					
		protect themselves from the damages due					
		to the natural disaster.					
2.	Idea / Solution description	This project uses Multi-layered Deep					
		Convolutional Neural Network (pre-					
		trained) model to classify Natural Disaster					
		and calculate the intensity of the Disaster.					
3.	Novelty / Uniqueness	To overcome the non-clarity image					
		issues, the project uses the integrated					
		webcam to capture the video frame and					
		compare the data with pre-trained data.					
4.	Social Impact / Customer	By the application humans can do the					
	Satisfaction	safety precautions to avoid the damages					
		from the natural disasters, reduces the					
		damages and use of Deep CNN algorithm					
		with video frames accuracy improved.					
5.	Business Model (Revenue	The software requirements are affordable					
	Model)	and it is reliable one.					
6.	Scalability of the Solution	Highly expendable, dependable, reliable,					
		scalable and has robustness.					

3.4 Problem Solution fit





REQUIREMENT ANALYSIS

4.1 Functional requirement

Following are the functional requirements of the proposed solution.

FR No.	Functional Requirement (Epic)	Sub Requirement (Story / Sub-Task)
FR-1	User Registration	 Registering via Google Accounts Registering via Product's own user
FR-2	User Authentication	 management system Verification through OTP Verification through Email Link
FR-3	Designation of Region	Ease of selection of necessary areas to be monitored.Versatile and Flexible operations on
FR-4	Analysis of Required Phenomenon	 designated areas Simple and easy analysis on the specific phenomenon to be observed
FR-5	Accumulation of required Data	 Fast and Efficient data gathering capabilities. regarding past event analysis and future prediction
FR-6	Organizing Unstructured data	Processing of raw and clustered data into clear and refined data which is useful for analysis and prediction tasks
FR-7	Algorithm selection	 The freedom to choose from several classes of algorithm to be used in the process. Customization of algorithm to suit the needs of a specific purpose
FR-8	Prediction and analysis of data	 Accurate results of the analysis provided by the process. Advanced visualization techniques to help visualize the processed data for effective observation
FR-9	Report generation	Restructuring of obtained results into clear and detailed report for future studies

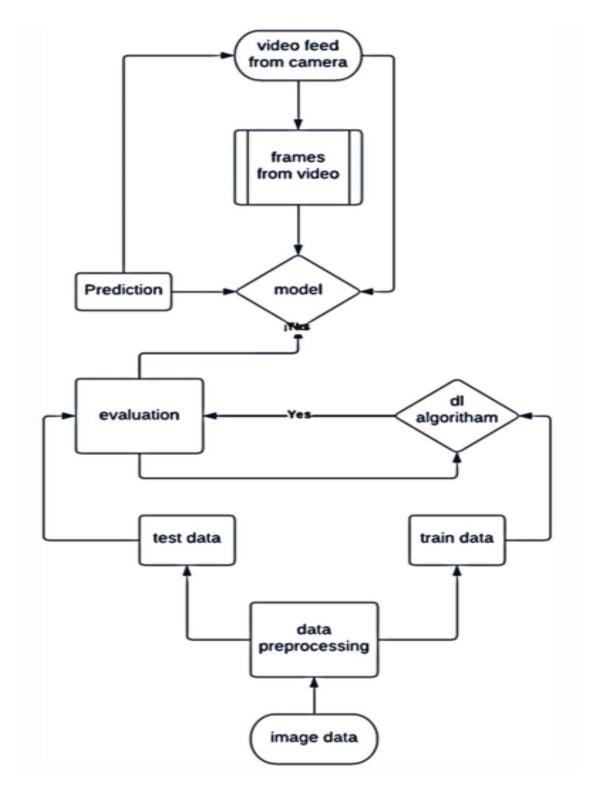
4.2 Non-Functional requirements

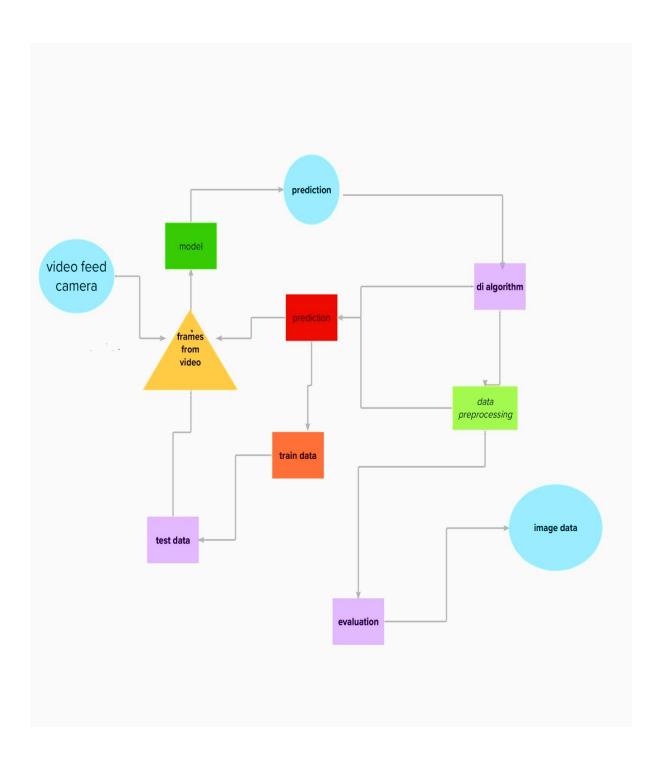
Following are the non-functional requirements of the proposed solution.

FR No.	Non-Functional	Description
	Requirement	
NFR-1	Usability	It is easy and quick method to predict the
		disasters.
NFR-2	Security	The secure pattern shares components with
		monitor and control for logging and control
		access and for providing audit trails
NFR-3	Reliability	it should be highly reliable
NFR-4	Performance	It deals with the measure of the system's
		response time.
NFR-5	Availability	t can be available at the any time and we can
		access during any disasters.
NFR-6	Scalability	Disaster damages are measured involves
		examining the number of fatalities, of
		injuries, of people affected.

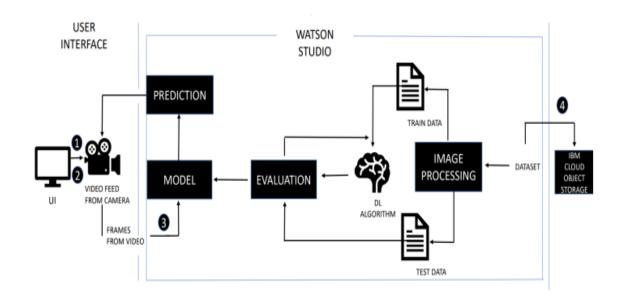
CHAPTER 5 PROJECT DESIGN

5.1 Data Flow Diagrams





5.2 Solution & Technical Architecture



5.3 User Stories

User Stories Use the below template to list all the user stories for the product.

User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance criteria	Priority	Release
Customer (Mobile user)	Registration	USN-1	As a user, I can register for the application by entering my email, password, and confirming my password.	I can access my account / dashboard	High	Sprint-1
		USN-2	As a user, I will receive confirmation email once I have registered for the Application	I can receive confirmation email & click confirm	High	Sprint-1
		USN-3	As a user, I can register for the application through Facebook	I can register & access the dashboard with Facebook Login	Low	Sprint-2
		USN-4	As a user, I can register for the application through Gmail	I can login with my password	Medium	Sprint-1
	Login	USN-5	As a user, I can log into the application by entering email & password	I can see the dashboard now	High	Sprint-1
	Dashboard	USN-6	As a user, I can update Disaster incidents.	I can update now.	Medium	Sprint-1
Customer (Web user)		USN-7	As a user, I can view Map Data.	I can see Map Data.	Medium	Sprint-1
Customer Care Executive	Authentication	USN-8	As a Community Leader, I can log into the application using my password	I can access my account.	High	Sprint-1
		LIENI O	As a Community Looden Loop and	Loon apply	Lich	Sprint 1

		USN-9	As a Community Leader, I can apply for membership.	I can apply membership	High	Sprint-1
		USN-10	As a Community Leader, I can verify Disaster.	Disaster verification	High	Sprint-1
System Administrator	Membership Approval	USN-11	As a administrator, I can approve the Membership application.	I can approve membership.	High	Sprint-1
	Update Disaster information	USN-12	As a administrator, I can update information about Disaster.	I can update disaster information.	High	Sprint-1

	Disaster verification	USN-13	As a administrator, I can verify disaster.	I can verify Disaster	High	Sprint-1
Community Leader and System Administrator	Disaster Queries	USN-14	Both are can able to ask disaster queries.	We can ask Queries about disaster.	Low	Sprint-2
	Disaster	USN-15	Both are can able to give	Both will give the	Low	Sprint-2
	Reports		disaster reports.	disaster reports		

PROJECT PLANNING & SCHEDULING

6.1 Sprint Planning & Estimation

Use the below template to create product backlog and sprint schedule

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	Registration	USN-1	As a user, I can register for the application by entering my email, password, and confirming my password.	5	High	M.VIGNESH A. SRIAKASH
Sprint-1		USN-2	As a user, I will receive confirmation email once I have registered for the application	5	High	S. SHYAM KUMAR A. VASUDEVAN
Sprint-1	Login	USN-3	As a user, I can log into the application by entering email & password	2	Low	M.VIGNESH A. SRIAKASH
Sprint-1	Designation of Region	USN-4	As a user, I can collect the dataset and select the region of interest to be monitored and analysed.	3	Medium	S. SHYAM KUMAR A. VASUDEVAN
Sprint-2	Algorithm selection	USN-5	As a user, I can choose the required algorithm for specific analysis.	7	Medium	M.VIGNESH A. SRIAKASH
Sprint-2	Training and Testing	USN-6	As a user, I can train and test the model using the algorithm.	7	Medium	S. SHYAM KUMAR A. VASUDEVAN
Sprint-3	Prediction and analysis of data	USN-7	As a user, I can predict and visualise the data effectively.	10	High	M.VIGNESH A. SRIAKASH
Sprint-3	Model building	USN-8	As a user, I can build with the web application	10	High	M.VIGNESH A. SRIAKASH
Sprint-4	Report generation	USN-9	As a user, I can generate detailed report on product data analysis.	10	High	S. SHYAM KUMAR A. VASUDEVAN
Sprint-4	Model deployment	USN-10	As an administrator, I can maintain third party services	5	Low	M.VIGNESH A. SRIAKASH

6.2 Sprint Delivery Schedule

Sprint	Total Story Points	Duration	Sprint Start Date	Sprint End Date (Planned)	Story Points Completed (as on Planned End Date)	Sprint Release Date (Actual)
Sprint-1	15	6 Days	24 Oct 2022	29 Oct 2022	15	30 Oct 2022
Sprint-2	14	6 Days	31 Oct 2022	05 Nov 2022	14	06 N0v 2022
Sprint-3	20	6 Days	07 Nov 2022	12 Nov 2022	20	08 Nov 2022
Sprint-4	15	6 Days	14 Nov 2022	19 Nov 2022	15	20 Nov 2022

6.3 Reports from JIRA



CODING & SOLUTIONING

7.1 Feature 1

A convolutional neural network is a class of artificial neural networks. It is a Deep Learning algorithm that can take in an input image, assign importance to various objects in the image and be able to differentiate one from the other. The pre-processing required in a Convent is much lower as compared to other classification algorithms.

The advantage of CNNs is to provide an efficient dense network which performs the predictionor identification efficiently.

Code is attached below

7.2 Feature 2

We developed a multilayered deep convolutional neural network model that classifies the natural disaster and tells the intensity of disaster of natural the model uses an integrated webcam to capture the video frame and the video frame is compared with the Pre-trained model and the typeof disaster is identified and showcased on the OpenCV window. A multilayer neural network with appropriate weights has been shown to be able to approximate any input-output function making it an attractive tool for modeling and forecasting.

Code is attached below.

TESTING

8.1 Test Cases

				Date	11-Nov-22								
				Team ID	PNT2022TMID38617								
				Project Name	Natural Disaster IntensityAnalysis and Classification using Artificial Intelligence								
				Maximum Marks	4 marks								
Test case ID	Feature Type	Component	Test Scenario	Pre-Requisite	Steps to Execute	Test Data	Expected Result	Actual Result	Status	Comments	TC for Automation(Y/N)	BUG ID	Executed By
HomePage_TC_00	UI	Home Page	Verify user is able to see the home page and other tabs, when user entered into the website	internet and device	Enter URL and click go click the tabs in the Navigation Bar	URL FOR THE WEBSITE	Website should be visible	Working as expected	Pass	NA	N	NA	VIGNESH M SHYAM KUMAR S SRIAKASH A VASUDEVAN A
HomePage_TC_OO	UI	Home Page	verify user is able to see the results tab		Enter URL and click go Click on results tab and check whether the user is able to see the flag card with open button	URL FOR THE WEBSITE	Application should show below UI elements: a. Header with live stream b. A camera icon c. A button named open	Working as expected	Pass	NA	N	NA	VIGNESH M SHYAM KUMAR S SRIAKASH A VASUDEVAN A
HomePage_TC_00	Functional	Home page	Verify user is able to click the button on the results tab		Emter URL and click go Click on results tab and check whether the user is able to click the button named open	URL FOR THE WEBSITE	User should click the button named open	Working as expected	Pass	NA	N	NA	VIGNESH M SHYAM KUMAR S SRIAKASH A VASUDEVAN A
HomePage_TC_00	Functional	access camera	Verify user is able to see that the camera is accessible and open when the button is clicked		1.Enter URL and click go 2.click on results tab 3. click open button	URL FOR THE WEBSITE	Application should able to access the camera and see the livestream	Working as expected	Pass	NA	N	NA	VIGNESH M SHYAM KUMAR S SRIAKASH A VASUDEVAN A
Camera_TC_004	Functional	camera	Verify user is able to capture the image from live stream		1 Enter URL and click go 2. click on results tab 3. click open button 4. camera is opened 5. click button to capture image	URL FOR THE WEBSITE	Application should able to capture image from livestream	Working as expected	Pass	NA	N	NA	VIGNESH M SHYAM KUMAR S SRIAKASH A VASUDEVAN A
Prediction_TC_005	Functional	output window	Verify user is able to see the predicted results in the window		when the image is captured again click button to see the results	URL FOR THE WEBSITE	Application should show the predicted results from the image captured	Working as expected	Pass	NA	N	NA	VIGNESH M SHYAM KUMAR S SRIAKASH A VASUDEVAN A

8.2 User Acceptance Testing

This document serves as a quick reference for the Deep Learning Fundus Image Analysis for Early Detection of Diabetic Retinopathy project's test coverage and open issues as of the project's release for user acceptance testing.

DEFECT ANALYSIS:

This shows how many bugs were fixed or closed at each severity level and how theywere fixed.

Resolution	Severity 1	Severity 2	Severity 3	Severity 4	Subtotal
By Design	4	5	2	3	14
Duplicate	1	0	3	1	5
External	2	3	0	1	6
Fixed	9	2	4	15	30
Not Reproduced	0	0	1	0	1
Skipped	1	0	1	1	3
Won't Fix	0	5	2	1	8
Totals	17	14	13	22	64

TEST-CASE ANALYSIS

This report shows the number of test cases that have passed, failed, and untested

Section	Total Cases	Not Tested	Fai l	Pass
Print Engine	9	0	0	9
Client Application	40	0	0	40
Security	3	0	0	3
Out-source Shipping	3	0	0	3
Exception Reporting	8	0	0	8
Final Report Output	4	0	0	4
Version Control	2	0	0	2

RESULTS

9.1 Performance Metrics

S.No.	Parameter	Values (Percentage)
1.	Model Summary	-96%
2.	Accuracy	Training Accuracy - 96.5% Validation Accuracy -92.3%
3.	Confidence Score (Only YoloProjects)	Class Detected - Nil Confidence Score - Nil

Our Project marks the successive performance by implementing in order to be cost effective andmore reliable to use and to predict the future from the natural disaster that we are ahead of. The successive way includes the objectives, activities and the approaches for the project. It mainly includes the trained dataset which gives an excessive measure of success which helps to overcome the future from this natural disaster.

ADVANTAGES & DISADVANTAGES

Advantages

Many deep learning techniques have been applied by various researchers to detect and classify natural disasters to overcome losses in ecosystems, but detection of natural disasters still faces issues due to the complex and imbalanced structures of images. To tackle this problem, we propose a multilayered deep convolutional neural network.

The proposed model works in two blocks: Block-I convolutional neural network (B-I CNN), for detection and occurrence of disasters, and Block-II convolutional neural network (B-II CNN), for classification of natural disaster intensity types with different filters and parameters.

The proposed model achieved the highest accuracy as compared to other state-of-the-art methods due to its multilayered structure. The proposed model performs significantly better for natural disaster detection and classification, but in the future the model can be used for various natural disaster detection processes.

The results were calculated as average statistical values: sensitivity, 97.54%; specificity, 98.22%; accuracy rate, 99.92%; precision, 97.79%; and F1-score, 97.97% for the proposed model. The proposed model achieved the highest accuracy as compared to other state-of-the-art methods due to its multilayered structure.

CONCLUSION

Many researchers have attempted to use different deep learning methods for detection of natural disasters. However, the detection of natural disasters by using deep learning techniques still faces various issues due to noise and serious class imbalance problems. To address these problems, we proposed a multilayered deep convolutional neural network for detection and intensity classification of natural disasters. The proposed method works in two blocks—one for detection of natural disaster occurrence and the second block is used to remove imbalanced class issues. The results were calculated as average statistical values: sensitivity, 97.54%; specificity, 98.22%; accuracy rate, 99.92%; precision, 97.79%; and F1-score, 97.97% for the proposed model. The proposed model achieved the highest accuracy as compared to other state-of-the-art methods due to its multilayered structure. The proposed model performs significantly better for natural disaster detection and classification, but in the future the model can be used for various natural disaster detection processes.

FUTURE SCOPE

The proposed method efficiently differentiated normal and abnormal behaviors of people during a natural disaster. The accuracy of the proposed system is improved by feeding previous encoding outputs to the decoding layers and combining them. Several data mining application were implemented using contents of social media; user generated content helps in disastrous events to gain vast amount of information. The CNN model is used to extract flood images from raw images and color filters are used to refine the desired detection, the proposed system's efficiency and accuracy were tested on several datasets and it outperformed other methods to give the highest results.

The proposed model shows better accuracy as compared to the recently developed techniques. The reason for this is that the proposed technique works in two parts: one for natural disaster occurrence detection and the second one for natural disaster classifications. The overall proposed model works on an image dataset to detect and classify the natural disasters.

Our project accuracy was very well compared to other recently developed natural disaster intensity analysis and classification models. So, our project will help in future to detect and analyze disaster. Due to detection before it happens everyone can aware from the disasters and it saves many lives, properties and environment from disasters.

APPENDIX

13.1 Source Code

Login.html:

```
<!DOCTYPE html>
<html lang="en">
<head>
  <meta charset="UTF-8">
  <meta name="viewport" content="width=device-width, initial-scale=1.0">
  <title>Login Form</title>
  <link rel="stylesheet" href="style.css">
  <link rel="stylesheet" href="https://cdnjs.cloudflare.com/ajax/libs/font-</pre>
awesome/5.15.3/css/all.min.css" />
</head>
<body>
  <div class="wrapper">
    <header>Login Form</header>
     <form action="#">
       <div class="field email">
         <div class="input-area">
            <input type="text" placeholder="Email Address">
            <i class="icon fas fa-envelope"></i>
            <i class="error error-icon fas fa-exclamation-circle"></i>
         </div>
```

```
<div class="error error-txt">Email can't be blank</div>
       </div>
       <div class="field password">
          <div class="input-area">
            <input type="password" placeholder="Password">
             <i class="icon fas fa-lock"></i>
             <i class="error error-icon fas fa-exclamation-circle"></i>
          </div>
          <div class="error error-txt">Password can't be blank</div>
       </div>
       <div class="pass-txt"><a href="#">Forgot password? </a></div>
       <input type="submit" value="Login">
     </form>
     <div class="sign-txt">Not yet member? <a href="#">Signup now</a></div>
  </div>
  <script src="script.js"></script>
</body>
</html>
Home.html
    <!DOCTYPE html>
<html lang="en">
<head>
 <a href="mailto://cdn.jsdelivr.net/npm/bootstrap@5.2.2/dist/css/bootstrap.min.css">https://cdn.jsdelivr.net/npm/bootstrap@5.2.2/dist/css/bootstrap.min.css</a>
  <meta charset="UTF-8">
  <meta http-equiv="X-UA-Compatible" content="IE=edge">
```

```
<meta name="viewport" content="width=, initial-scale=1.0">
  k rel="stylesheet" type="text/css" href="style.css" >
  link
href="https://cdn.jsdelivr.net/npm/bootstrap@5.0.2/dist/css/bootstrap.min.css"
rel="stylesheet" integrity="sha384-
EVSTQN3/azprG1Anm3QDgpJLIm9Nao0Yz1ztcQTwFspd3yD65VohhpuuCOm
LASjC" crossorigin="anonymous">
  <title>Natural Disaster</title></title>
</head>
<br/><body style="background-color: #34A0A4">
 <div class="card text-center" >
   <div class="card-header navbar-dark">
    cli class="nav-item">
      <a class="nav-link active" aria-current="true" href="home.html"
style="font-size: 24px;">Home</a>
     cli class="nav-item">
      <a class="nav-link" href="intro.html" style="font-size:
24px;">Introduction</a>
     cli class="nav-item">
      <a class="nav-link" href="upload.html" style="font-size: 24px;">Open
Web Cam</a>
     <h3 style="float: right;">AI based Natural Disaster Analysis</h3>
   </div>
  </div>
  <div class = "container" style="text-align: center;">
  <div class="card" style="width: 20rem; padding: 10px; margin: 40px; margin-</pre>
left: 40px; display: inline-block">
```

```
<img class="card-img-top" src="images/cyclone.jpg" alt="Card image cap">
   <div class="card-body" >
    <h5 class="card-title">Cyclone</h5>
    cyclone, large system of winds that circulates
counterclockwise direction north of the Equator and clockwise direction to the
south. 
    <a href="https://en.wikipedia.org/wiki/Cyclone" class="btn btn-
primary">Know more</a>
   </div>
  </div>
  <div class="card" style="width: 20rem; padding: 10px; margin: 40px; margin-</pre>
left: 40px; display: inline-block">
   <img class="card-img-top" src="images/earthquake.jpg" alt="Card image"</pre>
cap">
   <div class="card-body" >
    <h5 class="card-title">Earthquake</h5>
    A sudden violent shaking of the ground, causing great
destruction, as a result of movements within the earth's crust. 
    <a href="https://en.wikipedia.org/wiki/Earthquake" class="btn btn-
primary">Know more</a>
   </div>
   </div>
  </div>
  <div class = "container" style="text-align: center;">
   <div class="card" style="width: 20rem; padding: 10px; margin: 40px; margin-</pre>
left: 40px; display: inline-block">
    <img class="card-img-top" src="images/flood.jpg" alt="Card image cap">
    <div class="card-body" >
     <h5 class="card-title">Flood</h5>
     An overflow of a large amount of water beyond its
normal limits, especially over what is normally dry land. 
     <a href="https://en.wikipedia.org/wiki/Flood" class="btn btn-
primary">know more</a>
```

```
</div>
   </div>
   <div class="card" style="width: 20rem; padding: 10px; margin: 40px; margin-</pre>
left: 40px; display: inline-block">
    <img class="card-img-top" src="images/wildefire.jpg" alt="Card image cap">
    <div class="card-body" >
     <h5 class="card-title">Wild Fire</h5>
     A wildfire is an unplanned, uncontrolled and
unpredictable fire in an area of combustible vegetation starting in rural and urban
areas. 
     <a href="https://en.wikipedia.org/wiki/Wildfire" class="btn btn-
primary">Know more</a>
    </div>
    </div>
   </div>
</body>
</html>
Intro.html:
   <!DOCTYPE html>
<html lang="en">
<head>
  <meta charset="UTF-8">
  <meta http-equiv="X-UA-Compatible" content="IE=edge">
  <meta name="viewport" content="width=device-width, initial-scale=1.0">
  k rel="stylesheet" type="text/css" href="style.css" >
  link
href = "https://cdn.jsdelivr.net/npm/bootstrap@5.2.2/dist/css/bootstrap.min.css" \\
rel="stylesheet" integrity="sha384-
Zenh87qX5JnK2Jl0vWa8Ck2rdkQ2Bzep5IDxbcnCeuOxjzrPF/et3URy9Bv1WTRi
" crossorigin="anonymous">
  <title>Natural Disaster</title>
```

```
</head>
<body style="background-color: #34A0A4">
  <div class="card text-center">
    <div class="card-header">
     cli class="nav-item">
       <a class="nav-link" aria-current="true" href="home.html" style="font-
size: 24px;">Home</a>
      class="nav-item">
       <a class="nav-link active" href="intro.html" style="font-size:
24px;">Introduction</a>
      cli class="nav-item">
       <a class="nav-link" href="upload.html" style="font-size: 24px;">Open
Web Cam</a>
      <h3 style="float: right;">AI based Natural Disaster Analysis</h3>
    </div>
   </div>
   <h2 style="padding: 50px; margin: 50px; word-spacing: 15px; text-align:</pre>
center; line-height: 1.6;">
```

China, India and the United States are among the countries in the world most affected by natural disasters.

Natural disasters have the potential to wreck and even end the lives of those people,

who stand in their way.

affected by a natural disaster dramatically depends on where in the world you live,
The objective of

the project is to human build a web application to detect the type of disaster. The input

is taken from the in-built webcam, which in turn is given to the pre-trained model.

The model predicts the type of disaster and displayed on UI.

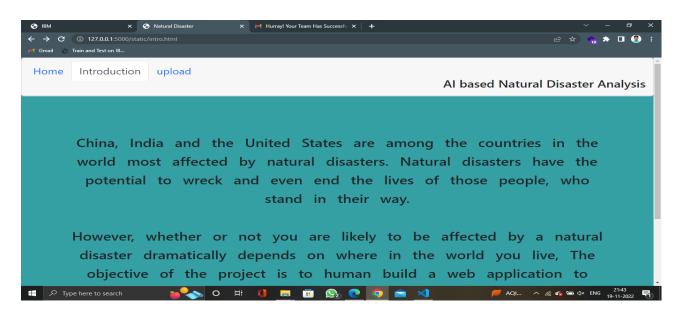
```
</h2>
</body>
</html>
```

Upload.html

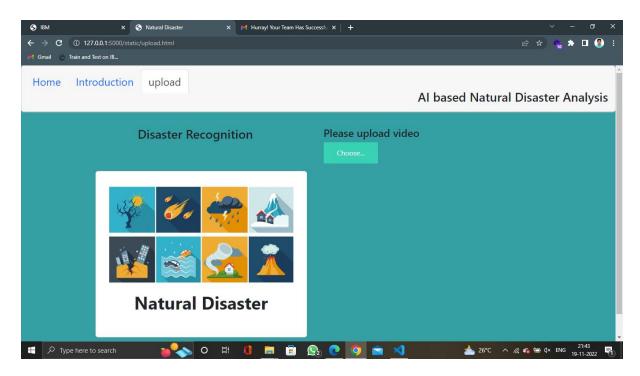
```
<!DOCTYPE html>
<html lang="en">
<head>
  <meta charset="UTF-8">
  <meta http-equiv="X-UA-Compatible" content="IE=edge">
  <meta name="viewport" content="width=device-width, initial-scale=1.0">
  k rel="stylesheet" type="text/css" href="style.css" >
  link
href="https://cdn.jsdelivr.net/npm/bootstrap@5.2.2/dist/css/bootstrap.min.css"
rel="stylesheet" integrity="sha384-
Zenh87qX5JnK2Jl0vWa8Ck2rdkQ2Bzep5IDxbcnCeuOxjzrPF/et3URy9Bv1WTRi
" crossorigin="anonymous">
  <title>Natural Disaster</title>
</head>
<body style="background-color:#34A0A4">
  <div class="card text-center">
    <div class="card-header">
     cli class="nav-item">
       <a class="nav-link" aria-current="true" href="home.html" style="font-
size: 24px;">Home</a>
      class="nav-item">
        <a class="nav-link" href="intro.html" style="font-size:
24px;">Introduction</a>
      cli class="nav-item">
        <a class="nav-link active" href="upload.html" style="font-size:
24px;">Open Web Cam</a>
      <h3 style="float: right;">AI based Natural Disaster Analysis</h3>
    </div>
   </div>
```

```
<center>
    <div class="card" style="width: 30rem; padding: 10px; margin: 40px;</pre>
margin-left: 40px;display:inline-block">
     <img class="card-img-top" src="images/1.png" alt="Card image cap">
     <div class="card-body" >
      <h1 class="card-title">Natural Disaster</h1>
     </div> </center> <center>
    <form action = "upload.html" method = "POST"
      enctype = "multipart/form-data">
      <input type = "file" name = "filename" />
      <input type = "submit" value="Submit"/>
      </form>
   </center>
   <script
src="https://cdn.jsdelivr.net/npm/@popperjs/core@2.11.6/dist/umd/popper.min.js"
integrity="sha384-
oBqDVmMz9ATKxIep9tiCxS/Z9fNfEXiDAYTujMAeBAsjFuCZSmKbSSUnQlm
h/jp3" crossorigin="anonymous"></script>
  <script
src="https://cdn.jsdelivr.net/npm/bootstrap@5.2.2/dist/js/bootstrap.min.js"
integrity="sha384-
IDwe1+LCz02ROU9k972gdyvl+AESN10+x7tBKgc9I5HFtuNz0wWnPclzo6p9vx
nk" crossorigin="anonymous"></script>
</body>
</html>
```

intro.html



upload.html



Home.html

