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CHAPTER 1

INTRODUCTION

1.1 PROJECT OVERVIEW

A naturalist is someone who studies the patterns of nature, identifies kind of flora and fauna in nature. Being able to identify the flora and fauna around us often leads to an interest in protecting wild species, and collecting and sharing information about the species we see on our travels is very useful for conservation groups like NCC. We use artificial neural network to train these image and build a deep learning model. When venturing into the woods, field naturalists usually rely on common approaches like always carrying a guidebook around everywhere or seeking help from experienced ornithologists. There should be a handy tool for them to capture ,identify and share the beauty to the outside world.

1.2 PURPOSE

The project aims to create an application for the hikers to identify rare species of birds, flowers, mammals by giving a picture taken by them. Field naturalists can only use this web app from anywhere to identify the birds, flowers, mammals and other species they see on their hikes, canoe trips and other excursions. We use artificial neural network to train these image and build a deep learning model. In this project, we are creating a web application which uses a deep learning model, trained on different species of birds, flowers andmammals and get the prediction of the bird when an image is been given.

CHAPTER - 2

LITERATURE SURVEY

2.1 EXISTING PROBLEM

The Problem is to classify the type of species. The key relationship in this work is between field biologist and technologists, thus many of our activities will involves hybrid artistic and scientific examinations of the wildness surrounding us. For instance we may develop biological tools for studying nearby creatures, and then adapt these into artistic devices for continued exploration and sharing of this phenomena.

2.2 SURVEY WORK

2.2.1 UNGULATE DETECTION AND SPECIES CLASSIFICATION FROM CAMERA TRAP IMAGES USING RETINA NET AND FASTER R-CNN (2022)

[Gholamreza Anbarjafari, Ilja Pavlovs, Kadir Aktas, Egils Avots, Jevgenijs Filipovs, AgrisBrauns, Gundega Done, Dainis Jakovels, Gholamreza Anbarjafari]

This paper presents a new dataset of wild ungulates which was collected in Latvia. It demonstrate two methods, which use RetinaNet and Faster R-CNN as backbones respectively, to detect the animals in the images. Faster R-CNN–ResNet50 network and RetinaNet were trained for 34,850 iterations (10 epochs) on the training dataset with a batch size of 4, learning rate of 0.0001 and Adam optimizer for the weight update. The general structure of the detector

involves image embedding, object localization and classification. DNN consisting of convolutional layerswhich are used for the feature extraction from the input image. Usually, backbone networks which are pre- trained on a natural image dataset such as ImageNetare used. Common networks used as the backbone are ResNet50, VGG160, Inception-ResNetV2 and DarkNet-19. The neck network takes and processes inputs from the different layers of the backbone, harnessing advantages of data pattern distribution over different feature map scales by using FPN (Feature Pyramid Network). A feed-forward neuralnetwork which performs the classification or regression task.

2.2.2 CONVOLUTIONAL NETWORKBASED ANIMAL RECOGNITION USING YOLO AND DARKNET (2021) [B.Karthikeya Reddy,Shahana Bano, G.Greeshmanth Reddy, Rakesh Kommineni, P.Yaswanth Reddy]

This research work has developed a YOLOV3 model to identify the animal present in the image given by user. The algorithm used in YOLOV3 model is darknet, which has a pretrained dataset. Machine learning has been applied to image processing. The image of animal will be given as input, then it will display the name of the animal as output by using YOLOV3 model. The detection is done by using a pre-trained coco dataset from darknet. The image is broken into various lengths and widths based on the given input image. Here for the recognition of image, YOLOV3 model is using recognizer deep learning package. The overall performance of the model is based on the different training images and testing images of the dataset. The detection is done by using a pre-trained coco dataset from darknet.

2.2.3 RECOGNITION OF ENDEMIC BIRD SPECIES USING DEEP LEARNING MODELS (2021). [Yo-Ping Huang,Haobijam Basanta]

The objective of the paper is identifying the bird species from images. This study developed a transfer learning-based method using Inception-ResNet-v2 to detect and classifybird species. To validate the reliability of the model, it adopted a technique that involves swapping misclassified data between training and validation datasets. The swapped data are retrained until the most suitable result is obtained. Additionally, fivefold cross-validation was performed to verify the predictive performance of the model. The proposed model was tested using 760 images of birds belonging to 29 species that are endemicto Taiwan. The model has achieved an accuracy of 98.39% in the classification of 29 endemic bird species. The model achieved a precision, recall, and F1-score of 98.49%, 97.50%, and 97.90%, respectively, in classifying bird species endemicto Taiwan.

2.2.4 THE ANALYSIS OF PLANTS IMAGE RECOGNITION BASED ON DEEP LEARNING AND ARTIFICIAL NEURAL NETWORK(2020).[Jiang Huixian]

This paper aims to identify and classify the plant using the leaves of the plant. The approach is to extract plant leaf features and identify plant species based on image analysis. The plant leaf images are segmented and the feature extraction algorithm is used to extract leaf shape and texture features from leaf

sample images. An artificial neural network classification method based on backpropagation error algorithm (BP algorithm) is proposed to recognize plant leaves. This paper studies the existing plant image location and recognition technology, and introduces deep learning theory. After that, the high dimensional expression of image features by artificial neural network in deep learning theory is analyzed. The existing ANN model is improved and some new techniques and methods are introduced to construct a new ANN model. The model unifies the processes of image segmentation, target featureextraction and target classification

2.2.5 PLANT SPECIES RECOGNITION USING MORPHOLOGICAL FEATURES AND ADAPTIVE BOOSTING METHODOLOGY (2019).

[Munish Kumar, Surbhi Gupta, Xiao-Zhi Gao and Amitoj Singh]

The paper uses a novel plant species classifier that recognizes the plant species in the image. Out of many features, leaf shape is a conspicuous element that most algorithms rely on to perceive and describe a plant. The system extracts the morphological features of the plant leaf and classifies using Multilayer Perceptron and other classification algorithm along with AdaBoost methodology. Different classifiers, i.e., KNN, Decision Tree and Multilayer perceptron are employed to test the accuracy of the algorithm. The authors have observed that the maximum precision rate of 95.42% has been achieved for 32 kinds of plant leaves and the proposed system has performed better than the existing techniques for plant leaf recognition.

2.2.6 BIRD IMAGE RETRIEVAL AND RECOGNITION USING A DEEP LEARNING PLATFORM (2019). [Yo-Ping Huang, HaobijamBasanta]

The authors have developed a deep learning platform that helps users recognize various species of birds endemicto Taiwan. A mobile application named the Internetof Birds (IoB) is developed that recognizes 27 species of birds. The deep learning model for bird image classification using the CNN framework is described. Bird images were learned by a convolutional neural network (CNN) to localize prominent features in the images. The model established and generated a bounded region of interest to refine the shapes and colors of the object granularities and subsequently balanced the distribution of birds. Then, a skip connection method was used to linearly combine the outputs of the previous and current layers to improve feature extraction. Then it applied the softmax function to obtain a probability distribution of bird features. The platform uses cloud based deep learning for image processing to identify bird species from digital images. The proposed system could detect and differentiate uploaded images with an overall accuracy of 98.70%.

2.2.7 AN EFFICIENT FRAMEWORK FOR ANIMAL BREEDS CLASSIFICATION USING SEMI-SUPERVISED LEARNINGAND MULTI- PART CONVOLUTIONAL NEURAL NETWORK (MP-CNN) (2019).[S. Divya Meena, L. Agilandeeswari]

The paper focus on classifying 27 classes of animals with 35,992 trainingimages. The proposed model classifies the animals on both generic and

fine- grained level. It has built a semi- supervised learning based Multi-part Convolutional Neural Network (MP-CNN) with a hybrid feature extraction framework of Fisher Vector based Stacked Autoencoder. With Semi-supervised learning based pseudo-labels, the modelclassifies new classes of unlabeled images too. Hellinger Kernel classifier method has been modified and used to re-train the misclassified classes of animals which further enhance the accuracy. Semi-supervised learning based pseudo-labels, the model classifies new classes of unlabeled images too. The testing accuracy increases as the models get trained. The experimental results how that the overall accuracy is 99.6%.

2.3 PROBLEM STATEMENT DEFINITION

Problem Statement (PS)	I am (Customer)	I'm trying to	But	Because	Which make me feel
PS-1	Researchers	Scan the species to identify the behaviour	Unable to get the clear image	It contains low pixel value	Upset
PS-2	Student	Explore the species	Unable topredict	It is poisono us or danger	Panic

PS-3	Tourist	Capture	It	More	Anxious
		theimage	shows	details to	
			data not	be	
			found	updated	
PS-4	Public	Scan either flora or fauna in same time	It can't support	Both were separate and consume large amount of time to scan	Frustrated

CHAPTER -3 IDEATION & PROPOSED SOLUTION

3.1 EMPATHY MAP CANVAS

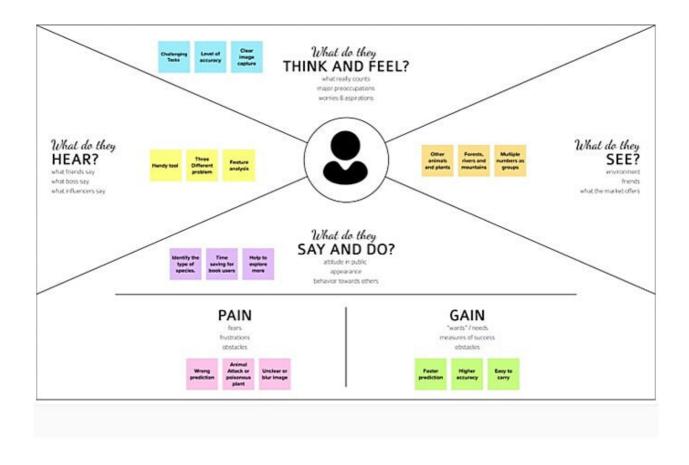


Fig 3.1 Empathy Map canvas

3.2 BRAINSTORMING AND IDEA PRIORITIZATION

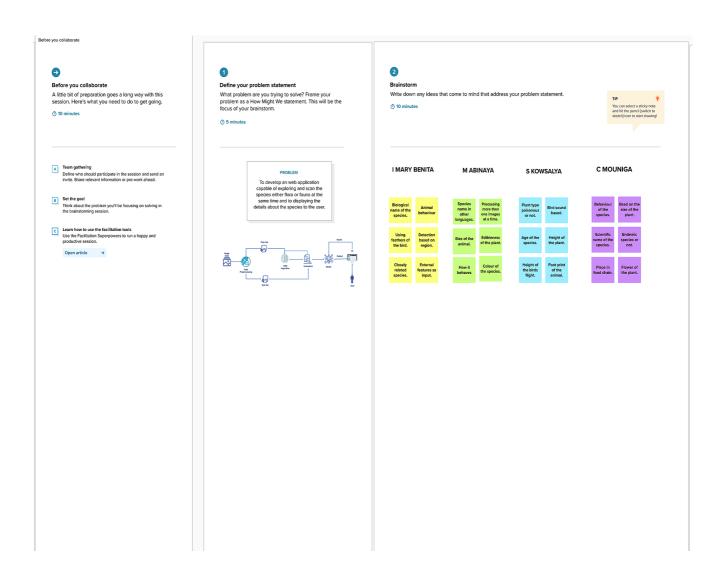


Fig 3.2 (A) Brainstorming and Idea Prioritization

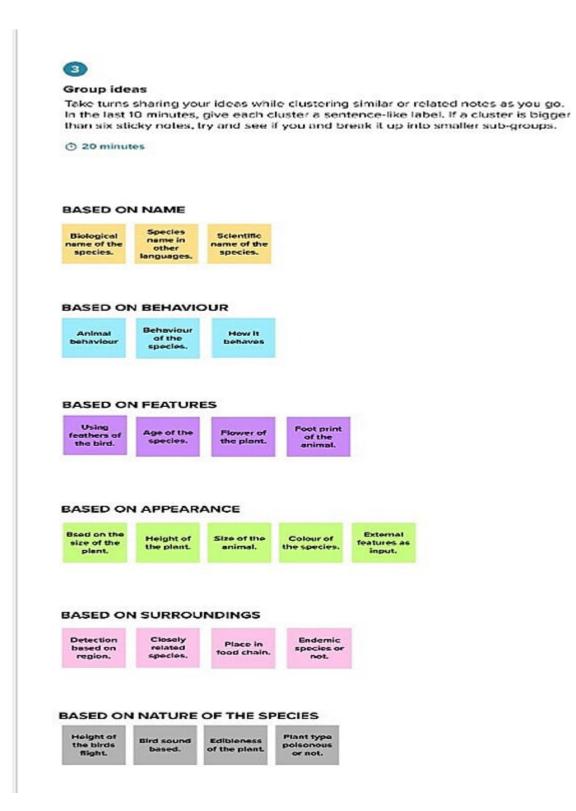
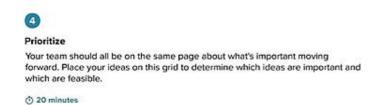


Fig 3.2 (B) Brainstorming and Idea Prioritization



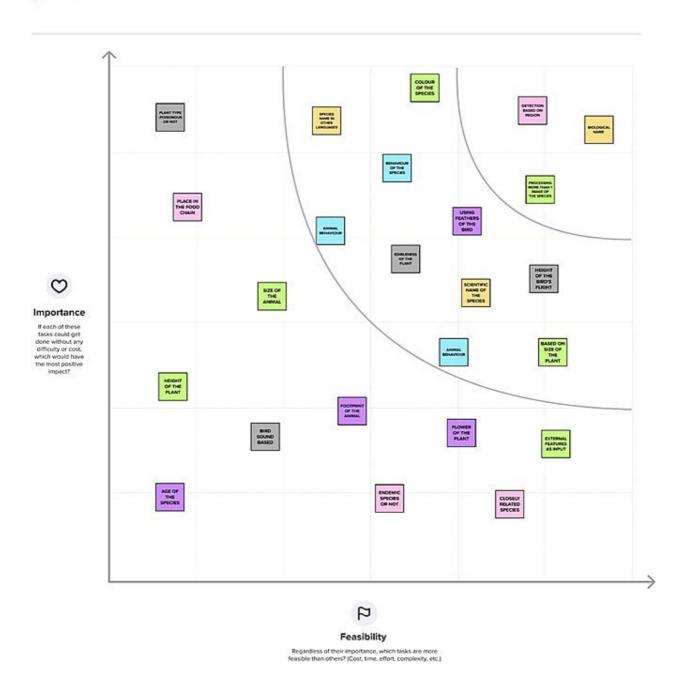


Fig 3.2 (C) Brainstorming and Idea Prioritization

3.3 PROPOSED SOLUTION

S No.	Parameter	Description
1.	Problem Statement (Problem	To identify a species in a forestor in
	to be solved)	any other place, we need to carry a heavy book or seek a professional.
2.	Idea/ Solution description	A Web Application that is trained with CNN using deep learning model on different species can replace such big books.
3.	Novelty / Uniqueness	The model can identify different species of plants, birds and animals.
4.	Social Impact /	Customer can identify the type of
	Customer	species faster and easier without
	Satisfaction	searching in books page by page.
5.	Business Model(Revenue	The model can differentiate the species
	Model)	at a faster rate with better accuracy.
6.	Scalability of the Solution	The web application apart from
		researchers can also be used by
		students and common
		people .

3.4 PROBLEM SOLUTION FIT

Explore AS, differentiat 1. CUSTOMER SEGMENT(S) 6. CUSTOMER CONSTRAINTS 5. AVAILABLE SOLUTIONS 1. Experience professionals and 1. Field naturalists always carry a 1.No knowledge about bio diversity. Inexperienced people who are willing to guidebook around everywhere or seeks 2.Cannot remember everything. learn about bio diversity. help from experienced Ornithologist. 3.Not able to identify the plants and 2. People who go for hikes or trips to the 2. Get help from experienced people. animals forest areas and mountains. 3.Internet and other apps 3. Amateurs or Students or people who like to learn more about the biodiversity. 9. PROBLEM ROOT CAUSE 7. BEHAVIOUR 2. JOBS-TO-BE-DONE / PROBLEMS 1. Need to depend on experts like 1. Fear of misidentification. 1. Carry guide books or other notes to Ornitologists, Zoolagists, Botanist. 2. Need to know about at least the identify species. 2. Users may not be a naturalist or just a Get help from experienced student who just started to learn so they professionals 3. No knowledge or experience about may not know any information. bio diversity as the user is just starting 3. Try to remember the species based on 3. Too much data cannot be stored by to learn which can lead to confusion its feature. any human or they may forget or other 4. Plant identifier. due to any other problems like age. 5. Animal identifier 3. TRIGGERS 10. YOUR SOLUTION 8.CHANNELS of BEHAVIOUR СН ONLINE 1. All information should be available in They hear about new app with best features by 1. All features are accessible during on application. Extraet online & offline CH of BE their friend or colleagues, or read about appir online 2. Display Botanical names . news, or search the internet on their own. 2. Search using the internet about the 3. Display alert messages for plants/animals 4. EMOTIONS: BEFORE / AFTER using different colors. species 4. Small description about them . 1.Before: Feel very sad to carry book always 5. Rarities of the species OFFLINE They may hesitate to task help from experts 1.Get help from friends or professionals. Sometime feels shame to ask help from experts 2. Guidebook or they even take their After: Feels happy no need carry heavy books own notes Feel proud because no need to seek help from experts.

CHAPTER-4

REQUIREMENTS ANALYSIS

4.1 FUNCTIONAL REQUIREMENTS

Following are the functional requirements of the proposed solution.

FR	Functional Requirement	Sub Requirement (Story/ Sub-Task)
No.	(Epic)	
FR-1	User Registration	Registration through Gmail
FR-2	User Confirmation	Confirmation via Email
		Confirmation via OTP
FR-3	Login	Enter the valid username and password
FR-4	capturing	Capturing the species image
FR-5	Uploading	Captured image can be upload in the
		application
FR-6	Processing and displaying	It process and shows the result about the
		species details

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 helps user to get information about
		the species and also user can access
		our site without any prior knowledge.
NFR-2	Security	It generate the OTP whether theuser
	Security	given phone number or email are valid
NFR-3	Reliability	Different and rare species information are
		availability in our sites.
NFR-4	Performance	Speed Response
		Bandwidth constraints
NFR-5	Availability	Focuses on the user's expectation and
		experience.
NFR-6	Scalability	Handle current and futureloads
		It works more efficiently.

CHAPTER-5

PROJECT DESIGN

5.1 DATA FLOW DIAGRAM

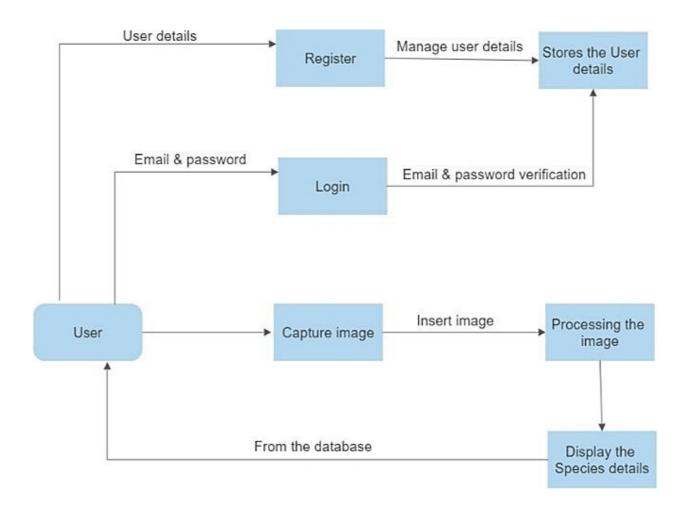


Fig 5.5 Data Flow Diagram

5.2 SOLUTION ARCHITECTURE

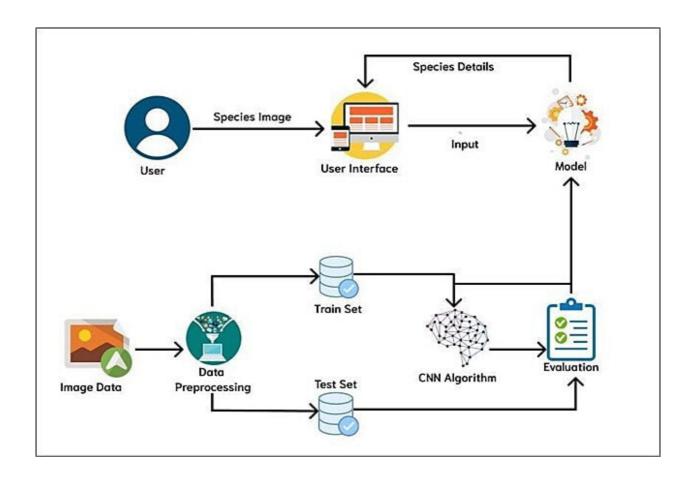


Fig 5.2 Solution Architecture

TECHNICAL ARCHITECTURE

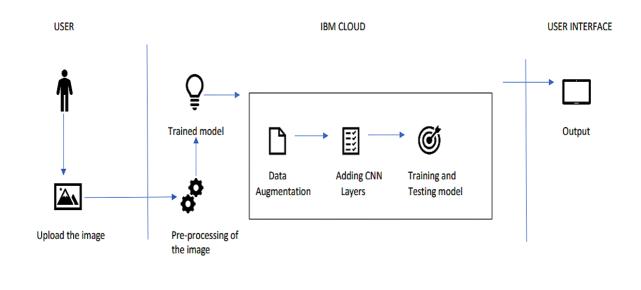


Fig 5.2 Technology Architecture

5.4 USER STORIES

		UserStory/ Task	_		Release
uirement ic)	Story Number		criteria	ty	
raictration	IISN 1	Acausor Lean	Lean	High	Sprint-1
gistiation	0311-1	· ·		mgn	Spriiit-1
		_			
		-			
		-			
		password.			
nfirmation	USN-2	As a user, I will		High	Sprint-1
		receive	I can receive		
		confirmation	confirmation		
		_	confirm		
Login	USN-3	· ·	I can access	Low	Sprint-2
		_			
		_			
Capturing	USN-4			High	Sprint-1
		<u> </u>			
			1 -		
Upload	IICN E	*		Uiah	Cpript 1
Opioau	0311-3		_	півіі	Sprint-1
		· ·			
		_	application.		
	egistration Infirmation Login Capturing Upload	egistration USN-1 Login USN-3 Capturing USN-4	egistration USN-1 As a user, I can register for the application by entering my email, password, and confirming my password. Infirmation USN-2 As a user, I will receive confirmation email once I have registered for the application Login USN-3 As a user, I can log into the application by entering Gmail and password. Capturing USN-4 As a user,I can capture the imageof the species.	egistration USN-1 As a user, I can register for the application by entering my email, password, and confirming my password. Infirmation USN-2 As a user, I will receive confirmation email once I have registered for the application by entering Gmail and password. Login USN-3 As a user, I can log into the application by entering Gmail and password. Capturing USN-4 As a user,I can capture the image of the species. Upload USN-5 As a user, I can ustorethe photo in the devices. Upload USN-5 As a user, I can capplication. I can access the dashboard with Gmail account. I can storethe photo in the devices. The image fed into the web application.	egistration USN-1 As a user, I can register for the application by entering my password, and confirming my password. Infirmation USN-2 As a user, I will receive confirmation email once I have registered for the application by entering Gmail and password. Login USN-3 As a user, I can log into the application by entering Gmail and password. Capturing USN-4 As a user, I can capture the image of the species. Upload USN-5 As a upon I can capture the image of the upload of the image

Admin	Processing and	USN-6	As a Admin, I	I can view	High	Sprint-2
	Display		can display	the details		
			thedetails of the	ofthe		
			species.	species.		

CHAPTER-6

PROJECT PLANNING & SCHEDULING

6.1 SPRINT PLANNING & ESTIMATION

Sprint	Functional	User	UserStory /	Story	Priority	
	Requireme	Story	Task	Points		Members
	nt	Numb				
	(Epic)	er				
Sprint-1	Registration	USN-1	As a User, I can register for the	2	High	Abinaya M
			application by			
			entering my			
			email,passwo			
			rd and			
			confirming by			
			password.	_	_	
Sprint-1		USN-2	As a user,I	2	Low	Mouniga C
			will receive			
			confirmation			
			email onceI			
			have			
			registered for			
			the			
			application.			
Sprint-1	Login	USN-3	As a user,I		Medium	Kowsalya
			can log into			S
			the			
			application by			
			entering			

			email& password.			
Sprint-1		USN-4	As a user,I can uploadthe image to identify thespecies.	3	High	Mary Benita I
Sprint-1	Dataset collection	USN-5	Datasets are collected to train the model.	2	High	Abinaya M
Sprint-2	Data Pre- processing	USN-6	The data is loaded andPre-processed to trainthe model.	4	High	Mouniga C
Sprint-2	Build and Train the model	USN-7	The modelis trained using Training dataset.	8	High	Kowsalya S
Sprint-2	Evaluate the model	USN-8	The model is evaluated.	6	High	Mary Benita I
Sprint-3	Create Applicati on	USN-9	Application is builtusing Python Flask.	8	Medium	Abinaya M
Sprint-3	Load the model	USN-10	The model is loaded into Python Flask.	6	High	Mouniga C

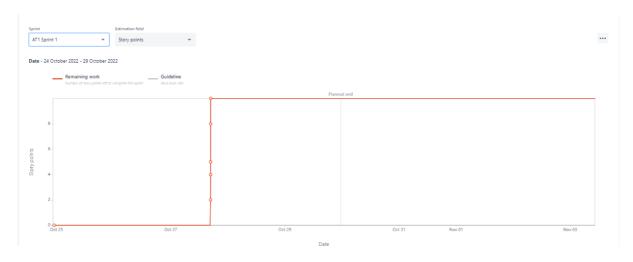
Sprint-4	Species	USN-11	As a user, I	6	Medium	Kowsalya
	identification		can view the			S
			species details.			
Sprint-4	Logout	USN-12	As a user,I can	2	Low	Mary
			logout of the			Benita I
			application.			

6.2 SPRINT DELIVERY SCHEDULE

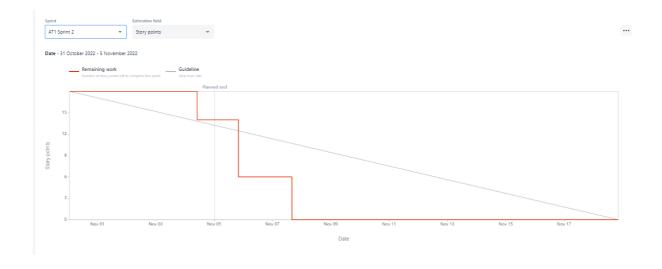
Sprint	Total	Duration	Sprint Start	Sprint End	•	Sprint
	Story		Date	Date	Completed (as on	Release
	Points			(Planned)	Planned End	Date
					Date)	(Actual)
Sprint-1	10	6 Days	24 Oct 2022	29 Oct 2022	10	30 Oct 2022
Sprint-2	18	6 Days	31 Oct 2022	05 Nov 2022	18	06 NOV 2022
Sprint-3	14	6 Days	07 Nov 2022	12 Nov 2022	14	13 NOV 2022
Sprint-4	8	6 Days	14 Nov 2022	19 Nov 2022	8	19 NOV 2022

6.3 REPORT FROM JIRA

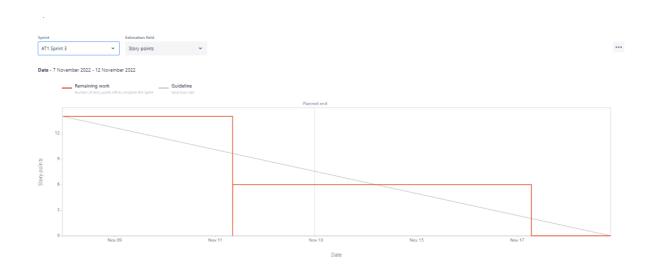
SPRINT 1:



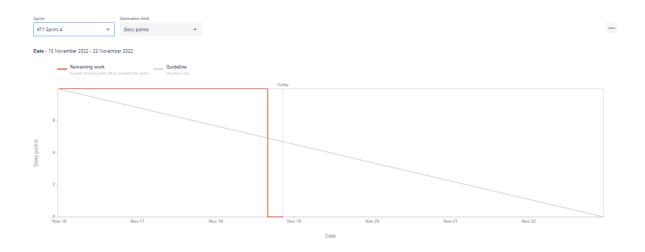
SPRINT 2



SPRINT 3



SPRINT 4



CHAPTER - 7

CODING & SOLUTIONING

7.1 FEATURE 1

The web application accepts an image as input and passes it to the CNN model . This model is built locally and deployed into python flask.

```
pred = np.argmax(loaded_model.predict(x), axis=-1)
```

7.2 FEATURE 2

The image passed by the app is classified by the CNN model. The name of the species is passed from the model to the application.

```
pred = np.argmax(loaded_model.predict(x), axis=-1)
```

CHAPTER-8 TESTING

8.1 TEST CASES

Test	Compo	Test Scenario	Steps To Execute	Test Data	Expected	Actual	Status
case ID	nent				Result	Result	
LoginPa	Home	Verify user is	1.Enter URL		Login/Sign in	Working as	Pass
ge_TC_	page	able to see the	2.Click on Explore		page should	expected	
01		Login/Sign in	now/login button		display		
		page when user	3. Verify login/Sign in				
		clicked on login	page displayed or not				
		button					

LoginPa	Login	Verify the login	1.Enter URL		Application	Working as	Pass
ge_TC_	page	elements	2. Click on Login		should show	expected	
02		Login/Sign in	button 3.Verify		below UI		
		appear	login/Sign in		elements:		
			elements:		a. email text		
			a. email text box		box		
			b. password text box		b. password		
			c. Login button		text box		
			d. Sign up link for		c. Login button		
			new user account.		d. Sign up link		
					for new user		
					account.		
LoginPa	Login	Verify user is	1.Enter URL	Username:	User should	Working as	Pass
ge_TC_	Page	able to log into	2. Click on Login	abinaya@gm	navigate to	expected	
03		application with	button	ail.com	upload page		
		Valid credentials	3.Enter Valid	password:			
			username/email in	abi@123			
			Email text box				
			4.Enter valid				
			password in password				
			text box				

			5.Click on login				
			button				
LoginPa	Login	Verify user is	1.Enter URL	Username:	Application	Working as	Pass
ge_TC_	Page	able to log into	2. Click on Login	abinaya@gm	should show	expected	
04		application with	button	ail.com	'Invalid details'		
		Invalid	3.Enter valid	password:	validation		
		credentials	username/email in	abi@321	message		
			Email text box				
			4.Enter Invalid				
			password in password				
			text box				
			5.Click on login				
			button				
LoginPa	Upload	Verify user is	1.Click choose file	Flower.png	The image	Working as	Pass
ge_TC_	Page	able to upload	2.Select image from		should be	expected	
05		image file	local directory		uploaded		

LoginPa	Upload	Verify user is	1.Click predict button	Flower.png	Flower name	Working as	Pass
ge_TC_	Page	able to view the			should be	expected	
06		species name			shown		
LoginPa	Upload	Verify user is	1.Click logout button		User should	Working as	pass
ge_TC_	page	able to logout			navigate to	expected	
07					upload page		

8.2 USER ACCEPTANCE TESTING

8.2.1. PURPOSE OF DOCUMENT

The purpose of this document is to briefly explain the test coverage and open issues of the Digital naturalist AI tool based on biodiversity resarchers project at the time of the release to User Acceptance Testing (UAT).

8.2.2. DEFECT ANALYSIS

Resolution	Severity	Severity	Severity	Severity	Sub
	1	2	3	4	total
By Design	9	5	3	2	19
Duplicate	0	1	2	1	4
External	3	4	1	2	10
Fixed	10	3	5	21	39
Not Reproduced	0	0	0	1	1
Skipped	1	1	0	0	2
Won't Fix	0	3	3	2	8
Totals	23	17	14	29	83

8.2.3. TEST CASE ANALYSIS

Section	Total Cases	Not Tested	Fa il	Pass
Print Engine	7	0	0	7
Client Application	50	0	1	50
Security	3	0	0	3
Outsource Shipping	3	0	0	3
Exception Reporting	9	0	0	9
Final Report Output	4	0	0	4
Version Control	2	0	0	2

CHAPTER - 9 RESULT

9.1 PERFORMANCE METRICS

Model Performance Testing:

S.No.	Parameter	Values	Screenshot
1.	Model Summary	Total params:22,704,9 66 Trainable params:22,704,9 66 Non-Trainable:0	## + 0 @ (@) (@) (@) (@) (@) (@) (@) (@) (@) (
2.	Accuracy	Training Accuracy - 92.55 Validation Accuracy - 78.69	mary() 730 - 38s - loss: 0.3923 - accuracy: 0.8628 - val_loss: 0.6718 - val_accur /30 - 34s - loss: 0.3251 - accuracy: 0.8920 - val_loss: 0.5334 - val_accur /30 - 31s - loss: 0.2748 - accuracy: 0.9095 - val_loss: 0.6757 - val_accur /30 - 38s - loss: 0.3206 - accuracy: 0.8993 - val_loss: 0.5967 - val_accur /30 - 32s - loss: 0.2370 - accuracy: 0.9328 - val_loss: 0.9432 - val_accur /30 - 29s - loss: 0.2524 - accuracy: 0.9182 - val_loss: 0.5927 - val_accur /30 - 29s - loss: 0.2306 - accuracy: 0.9080 - val_loss: 0.5708 - val_accur /30 - 28s - loss: 0.1621 - accuracy: 0.9489 - val_loss: 0.8260 - val_accur /30 - 34s - loss: 0.2478 - accuracy: 0.9255 - val_loss: 0.5331 - val_accur sequential"

CHAPTER - 10 ADVANTAGES & DISADVANTAGES

ADVANTAGES

Main advantage in our application is to provides three different species in one web application using the deep learning concepts. It does not require any special hardware because it can be implemented with low-cost devices such as cameras. An understanding of what species are and how to identify them is critical, both for biologists and for the general public. Biological diversity is

being lost as species go extinct, and it is only by understanding species.

DISADVANTAGES

The main disadvantage of our model is that the accuracy rate is low when the input image is not clear. Another disadvantage of this system is that it consumes time to and the system may not be able to recognize some numbers, such as possible.

CHAPTER - 11

CONCLUSION

Field naturalists can only use this web app from anywhare to identify the birds, flowers, mammals and other species they see on their hikes, canoe trips and other excursions. In this project, we are creating a web application which uses a deep learning model, trained on different species of birds, flowers and

mammals. There is great diversity among naturalists, but some common ground too. All naturalism begin with an admiring attitude towards science and its achievements. In many cases this admiring attitude is combined with a contempt or distrust for the way that philosophy has been or is conducted. This combination of views has a long history. Many of the advocates of first philosophy, Descartes, kant and Carnap, shared the same admiration of science or nascent science and distrust of philosophy. Descartes, for example. uses scepticism as a device to sweep away the old Aristotelian foundations of knowledge, so that he can build an entirely new philisophy that makes room for the new mathematical science.

CHAPTER - 12 FUTURE SCOPE

Essentially, the proposed guidelines treat statistical comparison of ML based quality estimators as a multi-dimentional problem. Accordingly, we seek to assess the predictors more holistically in terms of thier local performance on specific test conditions, their learning ability and the magnitude of treatment affect (to quantify the practical significance of the observed differences). In

constract, the current approach tends to reduce this task to binary and global statistical decision making and does not reveal systematic weakness of the predicators. In order to provide a tool for pratical use, software implementing the proposed guidelines is made publicly available.

CHAPTER - 13 APPENDIX

13.1 SOURCE CODE

import

os

import numpy as np
from keras.models import load_model
from keras.preprocessing import image

```
import tensorflow as tf
from flask import Flask, render template, request
from werkzeug.utils import secure_filename
from keras.models import model_from_json
from PIL import Image
app = Flask(__name__)
json_file = open('final_model.json', 'r')
loaded_model_json = json_file.read()
json_file.close()
loaded_model = model_from_json(loaded_model_json)
loaded_model.load_weights("final_model.h5")
# loaded_model=load_model('uploads/final_model.h5')
@app.route('/')
def index():
  return render_template("home.html")
@app.route('/login')
def index1():
  return render_template("login.html")
@app.route('/register')
def index2():
  return render_template("register.html")
```

```
@app.route('/upload')
def index3():
  return render_template("upload.html")
@app.route('/predict', methods=['GET', 'POST'])
def Upload():
  if request.method == 'POST':
    f = request.files['image']
    basepath = os.path.dirname(__file__)
    file path = os.path.join(basepath, ", secure filename(f.filename))
    f.save(file_path)
    img = image.load_img(file_path, target_size=(224, 224))
    x = image.img_to_array(img)
    x = np.expand dims(x, axis=0)
    pred = np.argmax(loaded_model.predict(x), axis=-1)
op = ['Great Indian Bustard Bird', 'Spoon Billed Sandpiper Bird', 'Corpse
Flower', 'Lady Slipper Orchid Flower',
        'Pangolin Mammal', 'Senenca White Deer Mammal']
    text = op[pred[0]]
    return render_template('upload.html',value=text)
if __name__ == '__main__':
  app.run(host='0.0.0.0',port=8000,debug=True)
```

13.2 GITHUB & PROJECT DEMO LINK

Github link: https://github.com/IBM-EPBL/IBM-Project-36795-1664177558

Project Demo: https://www.youtube.com/watch?v=EpHMxeWaaQw

CHAPTER - 14

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