

# **Detecting Parkinson's Disease using Machine Learning**

**TEAM ID : PNT2022TMID15817**

## **CHAPTER – 1**

### **INTRODUCTION**

#### **1.1.PROJECT OVERVIEW:**

Parkinson's disease is one of the most common neurodegenerative diseases with a prevalence rate of 1% in the population above 60 years old. The diagnosis of PD is traditionally based on motor symptoms. The symptoms of the disease will occur slowly, the symptoms include shaking, rigidity, slowness of movement and difficulty with walking, thinking and behavior change are common symptoms of this disease. This disease severely affects patients quality of life(QoL), social functions and family relationships, and places heavy economic burdens at individual and society levels. There is no defined test for early diagnosis of Parkinson's patient and medical decisions are provided based on the medical history of the patient and hence the possibility of misdiagnosis. Several researches were made to predict this disease in early stage. But they didn't predict properly and cannot be able to give better results.

#### **1.2 .PURPOSE:**

The aim of the project is vocal dysphonia analysis of Parkinson's patient from voice dataset with different machine learning algorithms with a goal to achieve better performance with less number of attributes. For addressing these difficulties and to refine the diagnosis and assessment procedures of PD, machine learning methods have been implemented for the classification of PD and healthy controls or patients with similar clinical presentations which can help to predict the disease earlier. Machine learning techniques are being increasingly applied in the healthcare sector. As its name implies, machine learning allows for a computer program to learn and extract meaningful representation from data.

## CHAPTER – 2

### LITERATURE SURVEY

#### 2.1.EXISTING PROBLEM:

In existing system, the comparative study of various machine learning algorithms is carried out. For analysis and prediction of Parkinson's PPMI data sets and six different classification algorithms are used. The results show that the multiclass classifier and logistic regression better performed than the other algorithms for the data sets. In future, more number of biomarker features are to be included for the prediction of progression of P. This study proposed regression, decision tree and neural network analysis to analyse the databank of Parkinson disease for error probability calculated. The result was logistic regression, classification and neural network analysis error probability by 5.15%, 8.47% and 23.73% respectively.

#### 2.2.REFERENCES:

S.no	Year	Researcher	Title	Methodology	Remarks
01	2019	S. Kanagaraj, M.S. Hema, M. Nageswara Gupta	Machine Learning Techniques for Prediction of Parkinson's Disease using BigData	Progression Marker Initiative (PPMI)	Predicts Parkinson's disease at an early stage from the formerly available public database

02	2019	F.M. Javed Mehedi Shamrat, Md. Asaduzzaman, A.K.M. Sazzadur Rahman, Raja Tariqul Hasan Tusher, Zarrin Tasnim	A Comparative Analysis Of Parkinson Disease Prediction Using Machine Learning	machine learning techniques	Thus, different experiments to assess the three machine learning supervised algorithms for recognition of Parkinson's disease
03	2014	Ma, C. et al	Identifying Parkinson disease using machine learning	SVM, KNN, and extreme learning machine (ELM)	discriminate healthy people from those with Parkinson's disease
04	2014	Ma, Ouyang, Chen, & Zhao	analysis of Parkinson disease	SCFW-KELM	Hybrid method is used to analysis Parkinson's disease
05	2014	Yahia A. et al	classification algorithm based on Naïve Bayes and K-Nearest Neighbours (KNN)	Parkinson speech dataset	Thus, Parkinson's Disease is detected through voice signal
06	2013	Chen et al.,	demonstrative precision for the identification of Parkinson Disease	fuzzy-based KNN model, a hybrid model	Identified the Parkinson's disease using these methods
07	2013	Sriram, Rao, Narayana, Kaladhar, Vital	detection of Parkinson diseases using machine learning algorithms	voice data	Thus, the analysis of voice data to understand the presence of Parkinson disease

08	2011	Rusz J	measurements to differentiate Parkinson disease from healthy subjects	vector machine	Vector machine is used to differentiate Parkinson disease
09	2011	OzciftA. et al	detection of Parkinson diseases using machine learning algorithms	computer-aided diagnosis (CADx) systems	Thus, Parkinson's Disease is detected
10	2011	Wu, S et al	Analysing the databank of Parkinson disease	regression, decision tree and neural network	Thus databank of Parkinson's disease is analysed

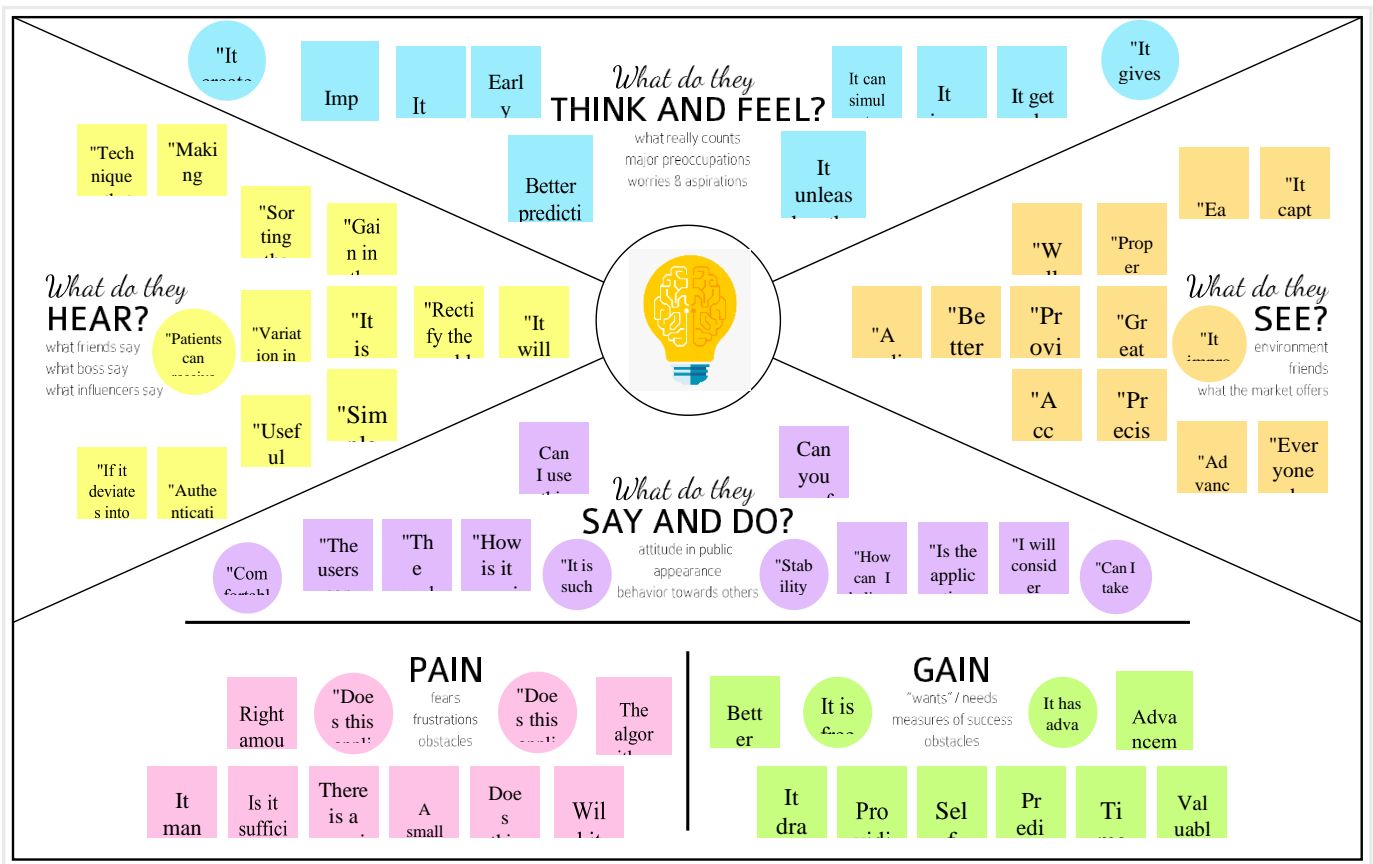
### 2.3.PROBLEM STATEMENT DEFINITION:

Who does the problem affect?	People who are men with minimization of nerve cells in primarily of village areas.
What are the boundaries of the problem?	People who are men with weak nerve cells and age over 50
What is the issue?	<p>In real time life of human, if the person is affected by Parkinson disease then it produces the side effect problems like dry skin and dandruff which majorly affects the quality of the life.</p> <p>As the age gets progresses, it causes the people to face major problem with the nerve cells in the brain.</p>
When does the issue occur?	<p>During the age excess of over 50</p> <p>as they will affect the people with loss of nerve cells in the brain.</p>
Where is the issue coming?	It majorly occurs due to the age getting over 50 and as maximum in village areas.
What methodology used to solve the issue?	Supervised and Un-supervised machine learning, Data mining , Computer vision with OpenCV, Python web application interface – Flask , IBM Cloud.

## CHAPTER – 3

### IDEATION & PROPOSED SOLUTION

#### 3.1. EMPATHY MAP CANVAS:



### 3.2. IDEATION AND BRAINSTORMING:

**Brainstorm & idea prioritization**

Use this template in your own brainstorming sessions so your team can visualize their imagination and start shaping concepts even if you're not sitting in the same room.

- 10 minutes template
- 10 minutes template
- 10 minutes template

**Define your problem statement**

What problem are you trying to solve? Frame your problem as a How might We statement. This will be the focus of your brainstorm.

**Brainstorm**

Write down any ideas that come to mind that address your problem statement.

**Group ideas**

Use this template to group your ideas into categories or related ideas as you go.

**Prioritize**

Use this template to group your ideas into categories or related ideas as you go.

**Group ideas based on logic/speech changes**

Change in speech and emotions expressed

**Group ideas based on walking mobility**

Potentially based walking difficulty

**Group ideas based on rigidity in hands**

predicts the disease by measuring the drawing speed

**Group ideas based on genes**

Genetic information can be used to predict the risk of Parkinson's disease

**Prioritize**

Use this template to group your ideas into categories or related ideas as you go.

### 3.3.PROPOSED SOLUTION:

S.No	Parameter	Description
1.	Problem Statement (Problem to be solved)	Parkinson's disease disorder is a brain disorder that causes unintended or uncontrollable movements, such as shaking, stiffness, and difficulty with balance and coordination. Symptoms usually begin gradually and worsen over time. As the disease progresses, people may have difficulty walking and talking.
2.	Idea / Solution description	Studies investigate signals from sustained phonation and text dependent speech modalities for Parkinson's disease screening. Phonation corresponds to the vowel voicing task and speech to the pronunciation of a short sentence, signal will be recorded through channel simultaneously through mobile phone or microphone. Parkinson disease affects vocal cord so the motion of speech is detected and evaluated.
3.	Novelty / Uniqueness	Testing 25 non impulsive patients with Parkinson's disease (PD) and 27 PD patients with impulsive compulsive behaviors (ICBs). Both patient groups were examined "on" and "off" dopaminergic medication in a counterbalanced order and their behavior was compared with 24 healthy controls. We found that PD patients with ICBs were significantly more prone to choose novel options than either non impulsive PD patients or controls, regardless of medication status. Our findings suggest that attraction to novelty is a personality trait in all PD patients with ICBs which is independent of medication status.
4.	Social Impact/ Customer Satisfaction	Since it is based on the voice based detection it is very convenient to use. As it helps the people to detect the Parkinson's disease in early stage, the loss of life is prevented. It detects without cost and helps



		to avoid travelling and time.
5.	Business Model (Revenue Model)	A free platform with useful feature. Any adult and young people can use it and suggest it to others to increase the value
6.	Scalability of the Solution	Additional features can be added anytime anywhere. Any number of users can access it all at once.

### 3.4. PROBLEM SOLUTION FIT:

Problem-Solution fit canvas 2.0		Team No: PNT2022TMD15817		
Define CS, fit into	<b>1. CUSTOMER SEGMENT(S)</b> <span>CS</span> Who is your customer? People who sense nervous disorder or symptoms related to Parkinson's disease. Symptoms includes rigidity of muscles, illusions and so on.	<b>6. CUSTOMER</b> <span>CC</span> What constraints prevent your customers from taking action or limit their choices of solutions? i.e. spending power, budget, no cash, network connection, available devices. Network connection may cause trouble as the website will be deployed. Laptops or mobile phones need to be available.	<b>5. AVAILABLE SOLUTIONS</b> <span>AS</span> Which solutions are available to the customers when they face the problem or need to get the job done? What have they tried in the past? What pros & cons do these solutions have? i.e. pen and paper is an alternative to digital notetaking. Disease can be detected using existing methods but the accuracy of the disease may not be accurate.	
	<b>2. JOBS-TO-BE-DONE / PROBLEMS</b> <span>J&amp;P</span> Which jobs-to-be-done (or problems) do you address for your customers? There could be more than one; explore different sides. To be aware of the disease among the minds of the people to be safe and live a longer life happily.	<b>9. PROBLEM ROOT CAUSE</b> <span>RC</span> What is the real reason that this problem exists? What is the back story behind the need to do this job? i.e. customers have to do it because of the change in legislations. People are unaware that Parkinson's disease exists. This may cost their life at danger. Early prediction of this disease can prevent one's life from danger.	<b>7. BEHAVIOUR</b> <span>BE</span> What does your customer do to address the problem and get the job done? i.e. directly related: find the right solution/installer, calculate usage and benefits; indirectly associated: customers spend free time on volunteering work (i.e. Greenpeace) Seek for device such as Laptop or mobile phones to detect the disease.	Explore AS. Focus on J&P, tap into BE, understand
Identify strong TR & EM	<b>3. TRIGGERS</b> <span>TR</span> What triggers customers to act? i.e. seeing their neighbour installing solar panels, reading about a more efficient solution in the news. Customers need to use electronic device and know the usage of the device up to an extent.	<b>10. YOUR SOLUTION</b> <span>SL</span> If you are working on an existing business, write down your current solution first, fill in the canvas, and check how much it fits ideally. If you are working on a new business proposition, then keep it blank until you fill in the canvas and come up with a solution that fits within customer limitations, solves a problem and matches customer behaviour. Solution aims to detect the Parkinson's disease in hand-drawn images of spirals and waves using Machine Learning.	<b>8. CHANNELS of BEHAVIOUR</b> <span>CH</span> <b>8.1 ONLINE:</b> What kind of actions do customers take online? Extract online channels from 7 # Have a stable internet connection and use Google Chrome, Mozilla Firefox. <b>8.2 OFFLINE:</b> What kind of actions do customers take offline? Extract offline channels from 7 and tag them for customer development. Make sure to have an electronic device such as mobile phone or laptops.	Extract online & offline CH of BE
	<b>4. EMOTIONS: BEFORE / AFTER</b> <span>EM</span> How do customers feel when they face a problem of a job and afterwards? i.e. lost, insecure > confident, in control - use it to your communication strategy & design. Customer will take proper precautions and lead a peaceful life.			

## CHAPTER-4

### REQUIREMENT ANALYSIS

#### 4.1. FUNCTIONAL REQUIREMENT:

FR NO:	FUNCTIONAL REQUIREMENT	SUB REQUIREMENT
FR-1	User Registration	Registration through Gmail Registration through Form Registration through LinkedIN
FR-2	User Confirmation	Confirmation via Email Confirmation via OTP
FR-3	User Authentication	Verify the user.
FR-4	Provide hand drawn images	Give input to the application to check for the disease
FR-5	Detection of the disease	Accuracy of the figure is shown with future precautions

#### 4.2. NON-FUNCTIONAL REQUIREMENT:

FR.NO:	NON-FUNCTIONAL REQUIREMENT	DESCRIPTION
NFR-1	Usability	The application is user friendly.
NFR-2	Security	Data is secured and confidential.
NFR-3	Reliability	Prevent the model from moving into production
NFR-4	Performance	Detection of the disease is accurate
NFR-5	Availability	Deployed in cloud so it is accessible
NFR-6	Scalability	Application performs well under an increased workload

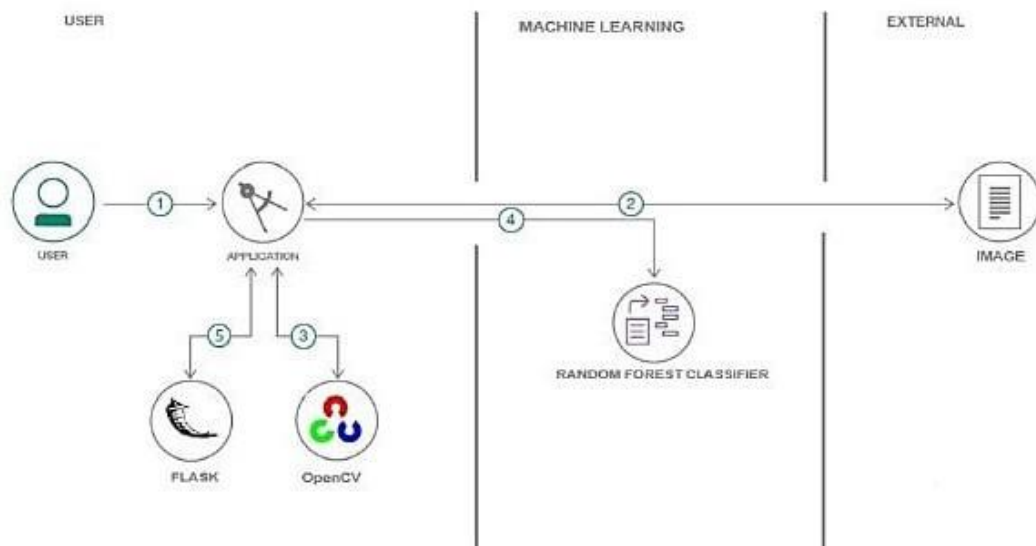
## CHAPTER-5

### PROJECT DESIGN

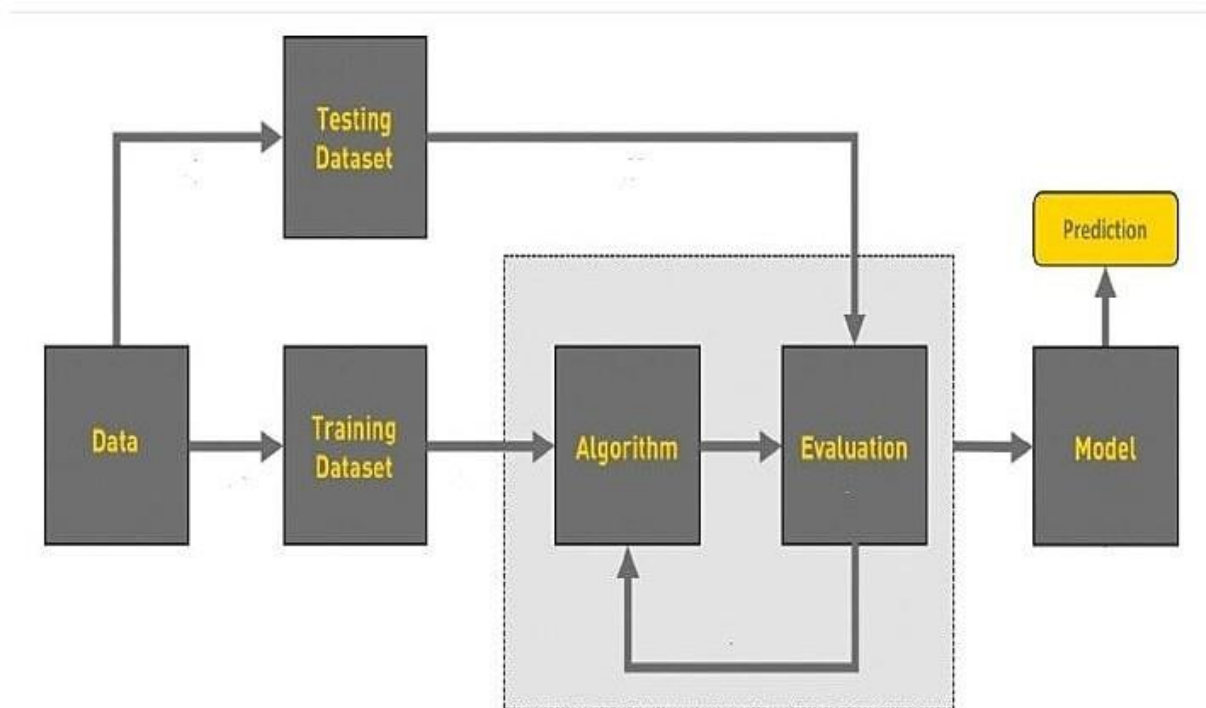
#### 5.1. DATA FLOW DIAGRAM:

A Data Flow Diagram (DFD) is a traditional visual representation of the information flows within a system. A neat and clear DFD can depict the right amount of the system requirement graphically. It shows how data enters and leaves the system, what changes the information, and where data is stored.

**Flow:**



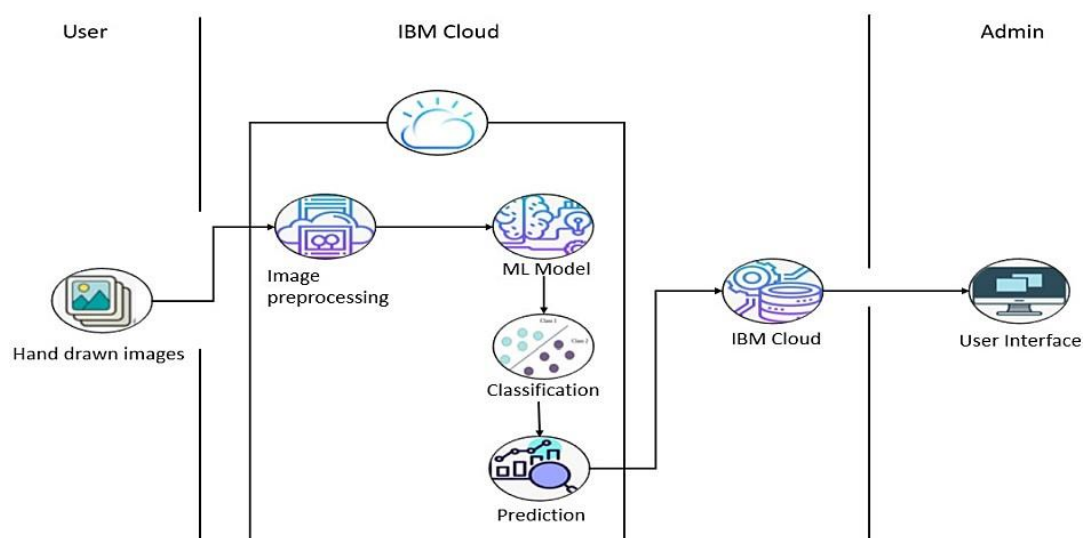
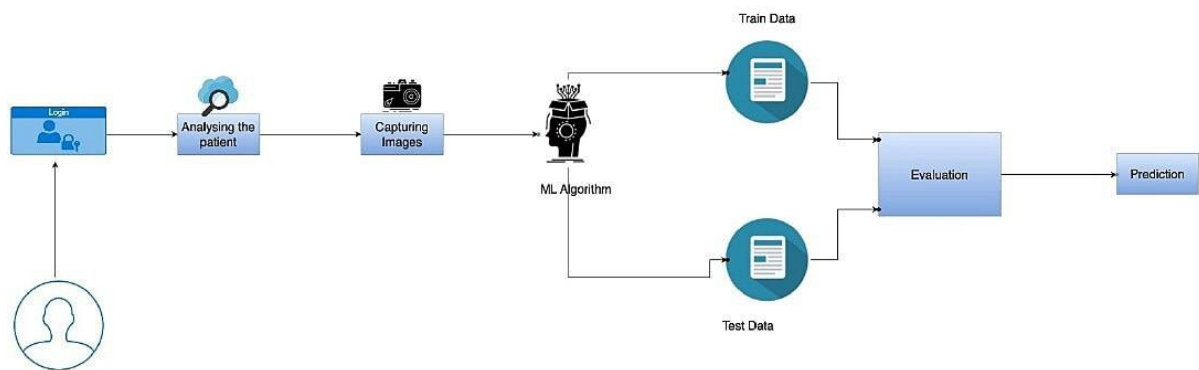
### Data Flow Diagram:



1. User configures credentials and starts the app.
2. User selects hand drawn images to process and load.
3. OpenCV does the image pre-processing.
4. The processed image is sent to the Random Forest Classifier.
5. The predicted output is visualised using Flask.

## 5.2.SOLUTION AND TECHNICAL ARCHITECTURE:

- 1.Create and login to the IBM Credentials.
- 2.Link the GitHub accountwith the IBM.
- 3.Notebook downloads from the dataset and imports data to analyses the patients.
- 4.After analyzing the affected patients we have to capture the images of them.
- 5.By using Machine Learning Algorithm, we have train and test the data for the further evaluation process.
- 6.After getting out the evaluation process we have to predict the given model by using Machine Learning.



### 5.3. USER STORIES:

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 register & access the dashboard with Gmail Login	Medium	Sprint-1
	Login	USN-5	As a user, I can log into the application by entering email & password	I can login & access my account with my registered credentials	High	Sprint-1
	Dashboard	USN-6	As a user, I can access the services and information provided in the dashboard	I can upload the hand drawn images, I can view the result, I can edit my profile and I can view my history	High	Sprint-1
Customer (Web user)	Login	USN-7	As a user, I can log into the web application and access the dashboard	I can login with the same registered credentials and access my account through web application	High	Sprint-1
Customer Care Executive	Help Desk	USN-8	As a user, I can get the guidance from the customer care	I can get help from the customer care for carrying out my tasks	High	Sprint-2



Administrator	Management	USN-9	As an administrator, I can collect new datasets and keep the model trained	I can collect and train the model with new dataset frequently	High	Sprint-2
		USN-10	As an administrator, I can update other features of the application	I can update and tune the features of application if needed	Medium	Sprint-1
		USN-11	As an administrator, I can maintain the information about the user	I can maintain information like user type and other such information	Medium	Sprint-1
		USN-12	As an administrator, I can maintain third-party services	I can support and maintain any third-party services	Low	Sprint-2

## CHAPTER – 6

### PROJECT PLANNING AND SCHEDULING

#### 6.1.SPRINT DELIVERY SCHEDULE:

TITLE	DESCRIPTION	DATE
Ideation Phase	<ul style="list-style-type: none"><li>• Literature Survey</li><li>• Empathy Map</li><li>• Brainstorming</li><li>• Problem Statement</li></ul>	29 August 2022 – 17 September 2022
Project Design Phase1	<ul style="list-style-type: none"><li>• Problem Solution Fit</li><li>• Proposed Solution</li><li>• Solution Architecture</li></ul>	19 September 2022- 01 October 2022
Project Design Phase2	<ul style="list-style-type: none"><li>• Requirement Analysis</li><li>• Customer Journey</li><li>• Data Flow Diagrams</li><li>• Technical Architecture</li></ul>	03 October 2022- 15 October 2022
Project Planning Phase	<ul style="list-style-type: none"><li>• Sprint Delivery Plan</li><li>• JIRA files</li></ul>	17 October 2022- 22 October 2022
Project Development Phase	<ul style="list-style-type: none"><li>• Sprint 1</li><li>• Sprint 2</li><li>• Sprint 3</li><li>• Sprint 4</li></ul>	24 October 2022- 19 November 2022

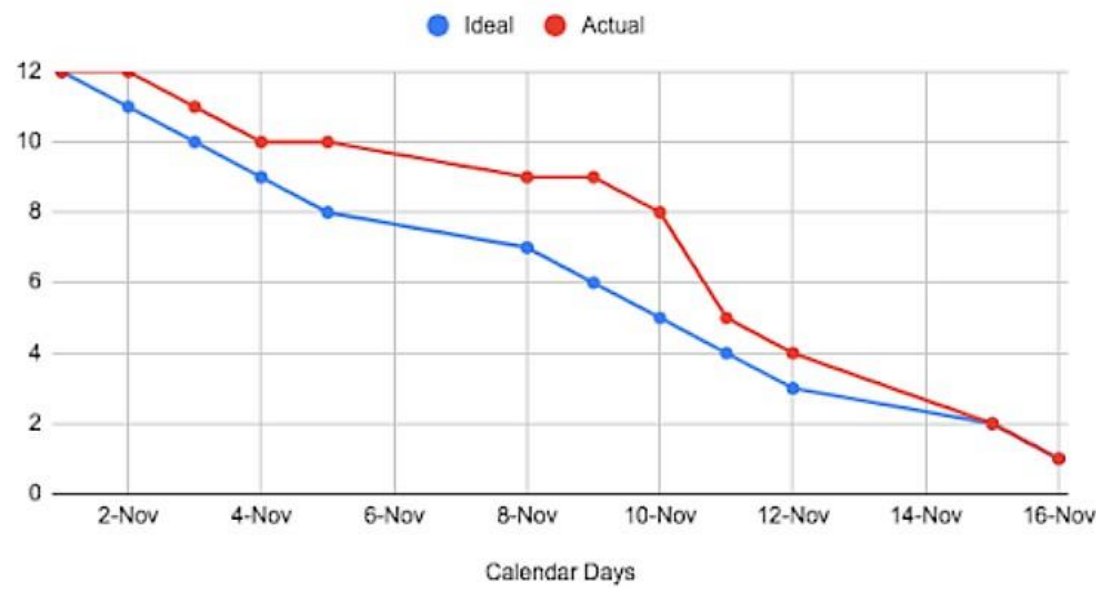
#### 6.2.SPRINT PLANNINGAND ESTIMATION:

SPRI NT	FUNCTIONAL REQUIREME NT	USER STORY NUMB ER	USER STORY/TA SK	STO RY POIN TS	PRIORI TY	TEAM MEMBE RS
Sprint- 1	Pre-processing data	USN-1	Collect Dataset	5	High	Keerthana
Sprint- 1		USN-2	Import therequired libraries, Read&Clean the datasets.	5	High	Hemalatha

Sprint-2	Building the model	USN-1	Split the data into dependent and independent variables.	4	High	Janani P
Sprint-2		USN-2	Apply using regression model.	2	Medium	Janani S
Sprint-3	Application Building	USN-1	Build python flask application and HTML page.	5	High	Janani P
Sprint-3		USN-2	Execute and test the application.	2	Medium	Janani S
Sprint-4	Training the model	USN-1	Train machine learning model.	5	High	Hemalatha
		USN-2	Integrate flask.	5	High	Keerthana

6.3.REPORTS FROM JIRA:

Burndown Chart



## CHAPTER – 7

### CODING AND SOLUTION

#### 7.1 FEATURE 1 (Decision Tree Classifier)

Decision Tree Classifier is used to train and test the model for detecting the phishing website with the help of collected and preprocessed dataset collections. NumPy is a library for the Python programming language, adding support for large, multi-dimensional arrays and matrices, along with a large collection of high-level mathematical functions to operate on these arrays. Moreover, NumPy forms the foundation of the Machine Learning stack. Pandas is an open-source Python package that is most widely used for data science/data analysis and machine learning tasks. Seaborn is a Python data visualization library based on matplotlib. It provides a high-level interface for drawing attractive and informative statistical graphics. For a brief introduction to the ideas behind the library, you can read the introductory notes or the paper.

Matplotlib is a comprehensive library for creating static, animated, and interactive visualizations in Python. Matplotlib makes easy things easy and hard things possible. Create publication quality plots. Make interactive figures that can zoom, pan, update. EDA is applied to investigate the data and summarize the key insights. It will give you the basic understanding of your data, its distribution, null values and much more. You can either explore data using graphs or through some python functions. There will be two types of analysis. Descriptive statistics are brief informational coefficients that summarize a given data set, which can be either a representation of the entire population or a sample of a population. Descriptive statistics are broken down into measures of central tendency and measures of variability.

Label Encoding refers to converting the labels into a numeric form to convert them into the machine-readable form. Machine learning algorithms can then decide in a better way how those labels must be operated. It is an important pre-processing step for the structured dataset in supervised learning. “Pickling” is the process whereby a Python object hierarchy is converted into a byte stream, and “unpickling” is the inverse operation, whereby a byte stream is converted back into an object hierarchy. 19

XGBoost is an optimized distributed gradient boosting library designed to be highly efficient, flexible, and portable. It implements machine learning algorithms under the Gradient Boosting framework.

## **7.2 FEATURE 2 (Flask Connection)**

The framework is the basis upon which software programs are built. It serves as a foundation for software developers, allowing them to create a variety of applications for certain platforms. It is a set of functions and predefined classes used to connect with the system software and handle inputs and outputs. It simplifies the life of a developer while giving them the ability to use certain extensions and makes the online applications scalable and maintainable. Flask is a web application framework written in Python. A Web Application Framework or simply a Web Framework represents a collection of libraries and modules that enable web application developers to write applications without worrying about lowlevel details such as protocol, thread management, among other examples.

Flask is a web application framework written in Python. Flask is based on the Werkzeug WSGI toolkit and the Jinja2 template engine. Both are Pocco projects. The Web Server Gateway Interface (Web Server Gateway Interface, WSGI) has been used as a standard for Python web application development. WSGI is the specification of a common interface between web servers and web applications. Flask is often referred to as a micro-framework. It is designed to keep the core of the application simple and scalable. Instead of an abstraction layer for database support, Flask supports extensions to add such capabilities to the application. Unlike the Django framework, Flask is very Pythonic. It's easy to get started with Flask, because it doesn't have a huge learning curve. HTML stands for Hyper Text Markup Language. HTML is the standard markup language for creating Web pages. HTML describes the structure of a Web page. HTML consists of a series of elements. HTML elements tell the browser how to display the content. Flask is used for developing web applications using python, implemented on Werkzeug and Jinja2. Advantages of using Flask framework are: There is a built-in development server and a fast debugger provided. The model deployed using Flask is used to predict the Chronic Kidney Disease. Hypertext markup language (HTML) is the basic language used to create documents for the Web and, along 20 with HTTP

(hypertext transfer protocol) and URLs (universal resource locators), is one of the three main protocols of the Web. Hypertext is text that contains hyperlinks. A hyperlink is an automated cross-reference to another location on the same document or to another document which, when selected by a user, causes the computer to display the linked location or document within a concise period.

A markup language is a set of tags that can be embedded in digital text to provide additional information about it, including its content, structure and appearance. This information facilitates automated operations on the text, including formatting it for display, searching it and even modifying it. Some type of markup language is employed by every word processing program and by nearly every other program that displays text, although such languages and their tags are typically hidden from the user. HTML consists of a set of predefined tags that can be embedded in text by web site designers in order to indicate the details of how web pages are rendered (i.e., converted into a final, easily usable, form) by web browsers. These details include paragraphing, margins, fonts (including style and size), columns, colors (background and text), links, the location of images, text flow around images, tables, and user input form elements (such as spaces for adding text and submit buttons).

## CHAPTER – 8

### TESTING

#### 8.1. TEST CASES:

Test case ID	Feature Type	Component	Test Scenario	Expected Result	Actual Result	Status
HP_TC_001	UI	Home Page	Verify UI elements in the Home Page	The Home page must be displayed properly	Working as expected	Pass
HP_TC_002	UI	Home Page	Check if the UI elements are displayed properly in different screen sizes	The Home page must be displayed properly in all sizes	Working as expected	Pass
HP_TC_003	Functional	Home page	Check if user can upload their file	The input image should be uploaded to the application successfully	Working as expected	Pass
HP_TC_004	Functional	Home page	Check if user cannot upload unsupported files	The application should not allow user to select a non image file	upload any file	Pass

Test case ID	Feature Type	Component	Test Scenario	Expected Result	Actual Result	Status
HP_TC_005	Functional	Home page	Check if the page redirects to the result page once the input is given	The page should redirect to the results page	Working as expected	Pass
BE_TC_001	Functional	Backend	Check if all the routes are working properly	All the routes should properly work	Working as expected	Pass
M_TC_001	Functional	Model	Check if the model can handle various image sizes	The model should rescale the image and predict the results	Working as expected	Pass
M_TC_002	Functional	Model	Check if the model predicts the image	The model should predict the image	Working as expected	Pass



Test case ID	Feature Type	Component	Test Scenario	Expected Result	Actual Result	Status
M_TC_003	Functional	Model	Check if the model can handle complex input image	The model should predict the number in the complex image	Working as expected	Pass
RP_TC_001	UI	Result Page	Verify UI elements in the Result Page	The Result page must be displayed properly	Working as expected	Pass
RP_TC_002	UI	Result Page	Check if the input image is displayed properly	The input image should be displayed properly	The size of the input image exceeds the display container	Fail
RP_TC_003	UI	Result Page	Check if the result is displayed properly	The result should be displayed properly	Working as expected	Pass

18	RP_TC_004	UI	Result Page	Check if the other predictions are displayed properly	The other predictions should be displayed properly	Working as expected	Pass
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## 8.2. USER ACCEPTANCE TESTING:

### 8.2.1. DEFECT ANALYSIS:

Resolution	Severity 1	Severity 2	Severity 3	Severity 4	Severity 5
By Design	1	0	1	0	2
Duplicate	0	0	0	0	0
External	0	0	2	0	2
Fixed	4	1	0	1	6
Not Reproduced	0	0	0	1	1
Skipped	0	0	0	1	1
Won't Fix	1	0	1	0	2
Total	6	1	4	3	14

### 8.2.2. TEST CASE ANALYSIS:

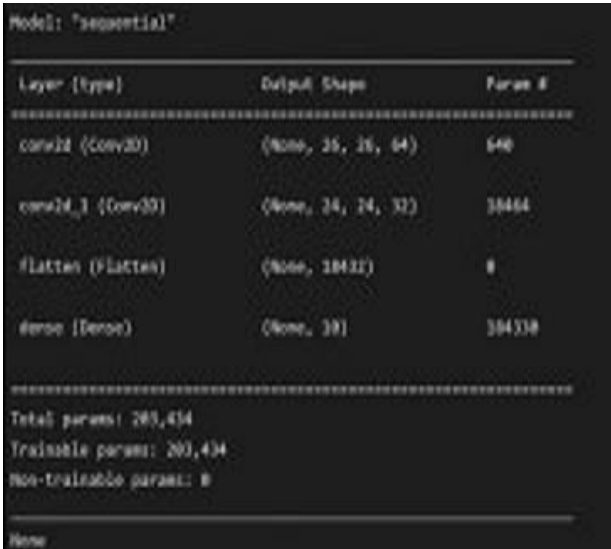
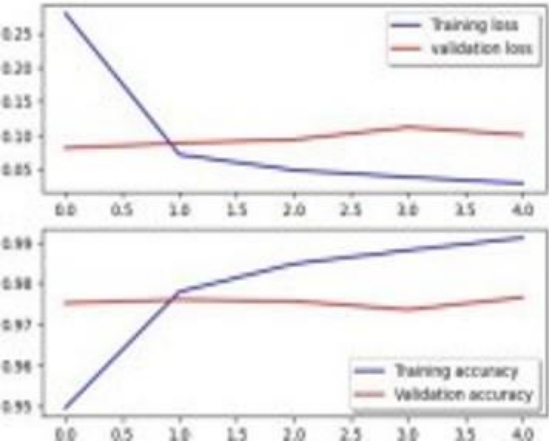
SECTION	TOTAL CASES	NOT TESTED	FAIL	PASS
Client Application	10	0	3	7
Security	2	0	1	1
Performance	3	0	1	2

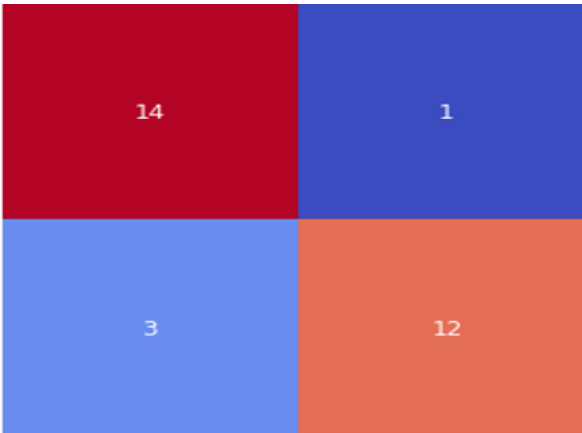
Exception Reporting	2	0	0	2
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## CHAPTER – 9

### RESULT

#### 9.1.PERFORMANCE METRICES:

S.NO	PARAMETER	VALUES	SCREENSHOT
1.	MODEL SUMMA RY		
2.	ACCURACY	Training Accurac y-98% Validati on Accurac y-97%	

3.	CONFUSION MATRIX		 <p>A 2x2 confusion matrix visualization. The top row is labeled '0' on the left. The bottom row is labeled '1' on the left. The bottom-left cell is labeled '0' below it, and the bottom-right cell is labeled '1' below it. The cells contain the following counts: top-left (red) is 14, top-right (blue) is 1, bottom-left (light blue) is 3, and bottom-right (orange) is 12.</p> <table><tr><th></th><th>0</th><th>1</th></tr><tr><th>0</th><td>14</td><td>1</td></tr><tr><th>1</th><td>3</td><td>12</td></tr></table>		0	1	0	14	1	1	3	12
	0	1										
0	14	1										
1	3	12										

## **CHAPTER – 10**

### **ADVANTAGES AND DISADVANTAGES**

#### **ADVANTAGES:**

- Reduces manual work
- More accurate than average human
- Capable of handling a lot of data
- Can be used anywhere from any device

#### **DISADVANTAGES:**

- Cannot handle complex data
- All the data must be in image format
- Requires a high performance server for faster predictions
- Prone to occasional errors

## **CHAPTER – 11**

### **CONCLUSION**

We have evaluated machine learning method for predicting Parkinson's Disease using Classification Method like Logistic Regression and Decision Tree Method which will be similar to Classification. In my project, I have exploited and evaluated the ability of motor functions of the person with the help of given measures who is healthier and affected. I collected 195 voice recording of an individual. I trained and tested the person's body condition with the help of given measures. The model trained with Decision Tree gives better accuracy than other model for predicting the person who is affected with the disease or not. It can be useful in early stage detection of this disease and can easily able to give treatment for the persons. In the future, we will continue research to develop advanced techniques for predicting Parkinson's Disease in large database.

## **CHAPTER – 12**

### **FUTURE SCOPE**

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

- a. Add support to detect from multiple images and save the results
- b. Add support to detect multiple images
- c. Improve model to detect from complex images

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

## CHAPTER-13

### APPENDIX

#### SOURCE CODE:

##### Importing the Necessary Libraries

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import zipfile as zf
import os
import random
import cv2
import pickle
from imutils import build_montages
from imutils import paths
from sklearn.metrics import classification_report, confusion_matrix
from sklearn import metrics
from sklearn.preprocessing import LabelEncoder, LabelBinarizer
from sklearn.model_selection import train_test_split
from sklearn.ensemble import
RandomForestClassifier, GradientBoostingClassifier, ExtraTreesClassifier
from skimage import feature
from google.colab.patches import cv2_imshow
```

##### Loading the training and testing dataset

```
handle_spiral = zf.ZipFile(r'dataset1.zip')
handle_spiral.extractall('dataset1')
handle_spiral.close()

spiral_train_healthy = os.listdir('dataset1/dataset/spiral/training/healthy')
```



```
spiral_train_park = os.listdir('#dataset1/dataset/spiral/training/parkinson#;')
```

```
fp_spiral_train_healthy = '#dataset1/dataset/spiral/training/healthy#;
```

```
fp_spiral_train_park = '#dataset1/dataset/spiral/training/parkinson#;
```

```
spiral_test_healthy = os.listdir('#dataset1/dataset/spiral/testing/healthy#;')
```

```
spiral_test_park = os.listdir('#dataset1/dataset/spiral/testing/parkinson#;')
```

```
fp_spiral_test_healthy = '#dataset1/dataset/spiral/testing/healthy#;
```

```
fp_spiral_test_park = '#dataset1/dataset/spiral/testing/parkinson#;
```

### **Quantifying Images**

```
def quantify_image(image):
```

```
    features = feature.hog(image,orientations=9,
```

```
    pixels_per_cell=(10,10),cells_per_block=(2,2),transform
```

```
    _sqrt=True,block_norm='L1') return features
```

### **Splitting up of training and testing data**

```
trainX = []
```

```
testX = [] outputs = []
```

```
trainY = []
```

```
testY = []
```

```
for i in spiral_train_healthy:
```

```
    image = cv2.imread(fp_spiral_train_healthy+i)
```

```
    image = cv2.cvtColor(image , cv2.COLOR_BGR2GRAY) image= cv2.resize(image ,  
(200,200))
```

```
    image =cv2.threshold(image, 0, 255,cv2.THRESH_BINARY_INV | cv2.THRESH  
_OTSU)[1]
```

```
    features = quantify_image(image) trainX.append(features)
```

```
trainY.append('#healthy#') for i in spiral_train_park:
```

```
    image = cv2.imread(fp_spiral_train_park+i)
```

```
    image = cv2.cvtColor(image , cv2.COLOR_BGR2GRAY) image= cv2.resize(image ,  
(200,200))
```

```
    image = cv2.threshold(image ,0,255,cv2.THRESH_BINARY_INV | cv2.THRESH  
_OTSU)[1]
```

```
    features = quantify_image(image) trainX.append(features)
```

```

trainY.append('#parkinson#') for i in spiral_test_healthy:
image = cv2.imread(fp_spiral_test_healthy+i) outputs.append(image)
image = cv2.cvtColor(image , cv2.COLOR_BGR2GRAY) image= cv2.resize(image ,
(200,200))
image = cv2.threshold(image ,0,255,cv2.THRESH_BINARY_INV | cv2.THRESH
_OTSU)[1]
features = quantify_image(image) testX.append(features) testY.append('#healthy#')
for i in spiral_test_park:
image = cv2.imread(fp_spiral_test_park+i) outputs.append(image)
image = cv2.cvtColor(image , cv2.COLOR_BGR2GRAY)

image = cv2.resize(image , (200,200))
image = cv2.threshold(image ,0,255,cv2.THRESH_BINARY_INV| cv2.THRESH
_OTSU)[1]
features = quantify_image(image) testX.append(features)
testY.append('#parkinson#') trainX= np.array(trainX)
testX = np.array(testX) trainY = np.array(trainY) testY = np.array(testY) trainX
trainY

```

## Label Encoding

```

le = LabelEncoder()
trainY = le.fit_transform(trainY) testY = le.transform(testY) print(trainX.shape,trainY.shape)
trainY
testY

```

## Model Building

```

Training the model print('Training model. ')
model = RandomForestClassifier(n_estimators=100) model.fit(trainX,trainY)
preds = model.predict(testX) preds

```

## Model Evaluation

```

cnf = confusion_matrix(testY,preds) cnf
array([[14,1], [ 3, 12]]) plt.figure(figsize=(5,5))
sns.heatmap(cnf , annot=True , cmap='coolwarm' , cbar=False) plt.show()

```

```
acc = metrics.accuracy_score(testY,preds)acc
```

```
indexes = np.random.randint(0,30,25)indexes
```

### Testing Model

```
testpath=list(paths.list_images(fp_spiral_train_healthy))idxs=np.arange(0,len(testpath))
```

```
idxs=np.random.choice(idxs,size=(25,),replace=False) images=[]
```

```
for i in idxs: image=cv2.imread(testpath[i]) output=image.copy()
```

```
output=cv2.resize(output,(128,128))
```

```
image=cv2.cvtColor(image,cv2.COLOR_BGR2GRAY) image=cv2.resize(image,(200,200))
```

```
image=cv2.threshold(image,0,255,cv2.THRESH_BINARY_INV | cv2.THRESH_OTSU)[1]
```

```
features= quantify_image(image) preds=model.predict([features])
```

```
label=le.inverse_transform(preds)[0]if label=="healthy":
```

```
color=(0,255,0)else:
```

```
(0,0,255)
```

```
cv2.putText(output,label, (3,20),cv2.FONT_HERSHEY_SIMPLEX,0.5,color,2)
```

```
images.append(output)
```

```
montage = build_montages(images,(128,128),(5,5))[0]
```

```
cv2.imshow(montage)
```

```
cv2.waitKey(0)montage=build_montages(images,(128,128),(5,5))[0]
```

```
cv2.imshow(montage)
```

```
cv2.waitKey(0)
```

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

```
cm = confusion_matrix(testY, predictions).flatten()print(cm)
```

```
(tn, fp, fn, tp) = cm
```

```
accuracy = (tp + tn) / float(cm.sum())print(accuracy)
```

## Flask App

```
from flask import Flask, request, render_template
import pickle
from skimage import feature
import os.path
# from werkzeug.utils import secure_filename
# from model import model

app = Flask(__name__)

@app.route("/")
def about():
    return render_template("home.html")

@app.route("/home")
def home():
    return render_template("home.html")

@app.route("/upload")
def test():
    return render_template("pred.html")

@app.route("/logout")
def log():
    return render_template("home.html")

@app.route('/predict', methods=['GET', 'POST'])
def upload():
    if request.method == 'POST':
        f = request.files['file'] # requesting the file
        filename_secure = secure_filename(f.filename)
        basepath = os.path.dirname(__file__) # storing the file in uploads folder
        filepath = os.path.join(basepath, "uploads", filename_secure)
        f.save(filepath) # saving the file

        # Loading the saved model
        print("[INFO] loading model...")
        model = pickle.loads(open('parkinson.pkl', "rb").read())
        local_filename = f"./uploads/{filename_secure}"
        local_filename = local_filename.replace(' ', '_')

        # Pre-process the image in the same manner we did earlier
        image = cv2.imread(filepath)
        output = image.copy()

        # Load the input image, convert it to grayscale, and resize
        output = cv2.resize(output, (128, 128))
```

```
image = cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)image= cv2.resize(image, (200, 200))
image = cv2.threshold(image, 0, 255,
cv2.THRESH_BINARY_INV | cv2.THRESH_OTSU)[1]
```

```
# Quantify the image and make predictions based on the extracted features using the last
trained RandomForest
features = feature.hog(image, orientations=9,
pixels_per_cell=(10, 10), cells_per_block=(2, 2), transform_sqrt=True, block_norm="L1")
preds = model.predict([features])print(preds)
ls = ["healthy", "parkinson"]result = ls[preds[0]]
```

```
""color = (0, 255, 0) if result == "healthy" else (0, 0, 255)
cv2.putText(output,result, (3, 20),
cv2.FONT_HERSHEY_SIMPLEX, 0.5, color, 2)
cv2.imshow("Output", output)cv2.waitKey(0)""
return resultreturn None
```

```
if __name__ == '__main__':app.run()
```

## HOME PAGE(HTML)

```
<!DOCTYPE
html>

<html lang="en">
<head>
<meta charset="UTF-8" />
<meta name="viewport" content="width=device-width, initial-scale=1.0"
/>
<meta http-equiv="X-UA-Compatible" content="ie=edge" />
<title>HomePage</title>
<style>
body {
background: linear-gradient(to right,#33ccff 0%, #99ffcc100%);
background-size: cover;
background-position: relative;
background-repeat: no-repeat;
height: 100%;
width: 100%;
}
h3 {
text-align: center;
color: white;
}
```

```
.main {  
margin-top: 100px;  
}  
p {  
color: black;  
text-indent: 10px;  
margin: 10px;  
font-size: 20px;  
}
```

```
  a{  
color: grey;  
float: right;  
text-decoration: none;  
font-style: normal;  
padding-right: 20px;  
}
```

```
  a:hover {  
background-color:  
black;  
color: white;  
font-size: 30px;  
padding-left: 10px;  
border-radius: 5px;  
}
```

```
  ul {  
align-items: center;  
display: flex;  
list-style-type: none;  
width: 100%;  
gap: 3rem;  
justify-content: center;  
font-size: 2rem;  
position: fixed;  
top: 0;  
margin: 0;  
padding: 1rem;  
background-color:  
white;  
}
```

```

li {
  cursor: pointer;
}
li a {
  text-decoration: none; color:
  inherit;
}
li.active {
  font-weight: bold;
  color: orangered;
}

img {
  width: 450px;
  height: 400px;
  padding: 25px;
}
img:hover {
  border-color: grey;
}
#im {
  width: 1450px;
  height: 700px;
  padding: 25px;
}
</style>
</head>
<body>
<nav>
<ul>
<li class="active"><a href="/home">Home</a></li>
<li class="active"><a href="/upload">Predict-
Results</a></li>
</ul>
</nav>
<br /><br /><br />
<h1>
<center>
<b class="pd"
><font color="black" size="15" font-family="Comic Sans MS"
>Detection of Parkinson's Disease using ML</font
></b
>

```

</center>  
</h1>  
<div>  
<center>  
<p style="text-align: left">  
Parkinson disease(PD) is a progressive neurodegenerative disorder that impacts more than 6 million people around the world.Parkinson's disease is non-communicable, early-stage detection of Parkinson's can prevent furtherdamages in humansuffering from it. However,Nonetheless, non-specialist physicians still do not have a definitive testfor PD, similarly in the earlystage of the diseased person wherethe signs maybe intermittent andbadly characterized. It

resulted in a high rate of misdiagnosis (up to 25% among non-specialists) and many years before treatment, patients can have the disorder. A more accurate, unbiased means of earlydetection is required, preferably one that individuals can use in their home setting.However, it has beenobserved that PD's presence in a humanis related to its hand-writing as well as hand-drawn subjects. From that perspective, several techniques have beenproposed by researchers to detect Parkinson's disease from hand-drawn images of suspected people. But the previous methods have their constraints.

</p>  
</center>  
<h4>  
<center>  
<b class="pd"  
><font color="black" size="12" font-family="Comic SansMS"  
>Causes andSymptoms of Parkinson's Disease</font  
>  
</b>  
</center>  
</h4>  
<span>





</span>

<span>

</span>

<span

></span>

<span

```
><img
title="Effect"
/></span>
<span
></span>
<span
></span>
<h3>
<center>
<font color="black" size="12" font-family="Comic Sans MS"
>Treatment for parkinson disease</font
>
</center>
</h3>
<span
></span>
<span
></span>
<span
></span>

<h3>

<center>

<font color="black" size="12" font-family="Comic Sans MS"

>How brains looks during PD?</font

>

</center>

</h3>

<span

></span>

<span

></span>

<br /><br />

</div>

</body>

</html>

**BASE PAGE(HTML)**

<html

```
lang="en"
>
<head>
<meta charset="UTF-8" />
<meta name="viewport" content="width=device-width, initial-scale=1.0" />
<meta http-equiv="X-UA-Compatible" content="ie=edge" />
<title>Predict</title>
<link
href="https://cdn.bootcss.com/bootstrap/4.0.0/css/bootstrap.min.css"
rel="stylesheet"
/>
<script src="https://cdn.bootcss.com/popper.js/1.12.9/umd/popper.min.js"></script>
<script src="https://cdn.bootcss.com/jquery/3.3.1/jquery.min.js"></script>
<script src="https://cdn.bootcss.com/bootstrap/4.0.0/js/bootstrap.min.js"></script>
<link
href="{ { url_for('static', filename='css/main.css') } }"
rel="stylesheet"
/>
<style>
body {
background-image: url("https://img.freepik.com/free-vector/clean-medical-patterned-background-vector_53876-140867.jpg?w=1060&t=st=1667911964~exp=1667912564~hmac=4298568f384f42cfc60423d63ac6a8c806e4fe025c1bed2f32ae68b3f15b2139");
background-position: center;
background-repeat: no-repeat;
background-size: cover;
height: 100%;
width: 100%;
}
h1 {
font-size: 40px;
text-align: center;
color: black;
font-style: italic;
```

```
font-weight: bolder;  
}
```

```
h2 {  
font-size: 35px;  
text-align: center;  
color: black;
```

```
font-style: italic;  
font-weight: bolder;  
}
```

```
h5 {  
font-size: 25px;  
text-align: center;  
color: black;  
font-weight: bolder;  
}
```

```
a {  
color: grey;  
float: right;  
text-decoration: none;  
font-style: normal;  
padding-right: 20px;  
}
```

```
a:hover {  
background-color: black;  
color: white;  
font-size: 30px;  
padding-left: 10px;  
border-radius: 5px;  
}
```

```
ul {  
align-items: center;  
display: flex;  
list-style-type: none;  
width: 100%;  
gap: 3rem;
```

```
justify-content: center;
font-size: 2rem;
position: fixed;
top: 0;
margin: 0;
padding: 1rem;
background-color: white;
}
li {
cursor: pointer;
}
```

```
li a {
text-decoration: none;
color: inherit;
}
li.active {
font-weight: bold;
color: orangered;
}
```

```
</style>
```

```
</head>
```

```
<body>
```

```
<nav>
```

```
<ul>
```

```
<li class="active"><a href="/home">Home</a></li>
```

```
<li class="active"><a href="/upload">Predict-Results</a></li>
```

```
</ul>
```

```
</nav>
```

```
<br />
```

```
<h1><b>Prevention is better than cure!</b></h1>
```

```
<br />
```

```
<h2>
```

```
<center>
```

```
♥Diagnosis is not the end, but the beginning of practice.
```

```
</center>
```

```

</h2>
<br />
<h2><center>♥Detect the disease and take measures wisely</center></h2>
<br />
<h5>
NOTE: Upload an spiral or wave page drawn by the patient/user in a white
sheet
</h5>
<div class="container">
<center>
<div id="content" style="margin-top: 2em">
{% block content %}{% endblock %}
</div>
</center>
</div>
</body>
<footer>
<script
src="{ { url_for('static', filename='js/main.js') } }"

```

## PREDICTION PAGE(HTML)

```

{% extends
"base.html"
%} {%
block
content %}
<div>
<form id="upload-file" method="post" enctype="multipart/form-data">
<center>
<label for="imageUpload" class="upload-label">
Choose...
</label>
<input type="file" name="file" id="imageUpload" accept=".png, .jpg, .jpeg">
</center>

```

```

</form>
<center> <div class="image-section" style="display:none;">
<div class="img-preview">
<div id="imagePreview">
</div></center>
</div>
<center>
<div>
<button type="button" class="btn btn-primary btn-lg " id="btn-predict">Predict!</button>
</div>
</center>
</div>
<div class="loader" style="display:none;"></div>
<h3 id="result">
<span> </span>
</h3>
</div>
type="text/javascript"
></script>
</footer>
</html>
47
{ % endblock % }

```

## HOME PAGE(CSS)

```

.imgpreview
{
width: 256px;
height: 256px;
position: relative;
border: 5px solid #F8F8F8;
box-shadow: 0px 2px 4px 0px rgba(0, 0, 0, 0.1);
margin-top: 1em;
margin-bottom: 1em;
}

```



```
.img-preview>div {
  width: 100%;
  height: 100%;
  background-size: 256px 256px;
  background-repeat: no-repeat;
  background-position: center;
}
input[type="file"] {
  display: none;
}
.upload-label {
  display: inline-block;
  padding: 12px 30px;
  background: #fe2727;
  color: #fff;
  font-size: 1em;
  transition: all .4s;
  cursor: pointer;
}
.upload-label:hover {
  background: #34495E;
  color: #39D2B4;
}
48
.loader {
  border: 8px solid #f3f3f3;
  /* Light grey */
  border-top: 8px solid #3498db;
  /* Blue */
  border-radius: 50%;
  width: 50px;
  height: 50px;
  animation: spin 1s linear infinite;
}
@keyframes spin {
  0% {
```

```
transform: rotate(0deg);
}
100% {
transform: rotate(360deg);
}
}
```

## **HOME PAGE(JS)**

```
.imgpreview
{
width: 256px;
height: 256px;
position: relative;
border: 5px solid #F8F8F8;
box-shadow: 0px 2px 4px 0px rgba(0, 0, 0, 0.1);
margin-top: 1em;
margin-bottom: 1em;
}
.img-preview>div {
width: 100%;
height: 100%;
background-size: 256px 256px;
background-repeat: no-repeat;
background-position: center;
}
input[type="file"] {
49
display: none;
}
.upload-label {
display: inline-block;
padding: 12px 30px;
background: #fe2727;
color: #fff;
font-size: 1em;
transition: all .4s;
cursor: pointer;
```

```
}  
.upload-label:hover {  
  background: #34495E;  
  color: #39D2B4;  
}  
.loader {  
  border: 8px solid #f3f3f3;  
  /* Light grey */  
  border-top: 8px solid #3498db;  
  /* Blue */  
  border-radius: 50%;  
  width: 50px;  
  height: 50px;  
  animation: spin 1s linear infinite;  
}  
@keyframes spin {  
  0% {  
    transform: rotate(0deg);  
  }  
  100% {  
    transform: rotate(360deg);  
  }  
}
```

#### **GITHUB**

<https://github.com/IBM-EPBL/IBM-Project-14972-1659592860>

#### **PROJECT DEMO LINK**

<https://drive.google.com/file/d/1Itoc0fK8vPaYUEtDLPIK3kGhI3RvxRU3/view?usp=sharing>