

**Disease Prediction System for Parkinson Disease using Voice Command**

**[PROJECT REPORT]**

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## **TABLE OF CONTENTS**

**Certificate of approval** 4

**Acknowledgement** 5

**Abstract** 6

**Overview** 7

**Introduction** 8

[● Brief explanation of Parkinson's Disease](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.4raf05t2ljam) 8

[● The importance of early diagnosis of Parkinson’s disease 8](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.79qohvsk0nea)

[● The potential of AI and voice-based diagnosis](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.baeopxu68dla) 9

[**The Science behind Parkinson’s Disease Diagnosis**](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.pzwmact6w8oz)  11

[● Traditional diagnosis methods of Parkinson’s Disease](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.8x56w4uebtgg) 11

[● Symptoms of Parkinson's Disease](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.4oxpknal3ojy) 11

[● Limitations of traditional diagnosis methods of Parkinson Disease 1](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.mztih16pq49i)2

**The Role of Artificial Intelligence in Parkinson’s Disease Diagnosis** 14

[● Advantages of AI diagnosis of Parkinson’s Disease](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.b1itavvke1d0) 14

[● Machine Learning algorithms used in Parkinson's Disease diagnosis 1](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.6ccyohlj5os)4

[● Voice-based diagnosis using AI 1](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.xjit8j6r146g)5

**Objective** 17

**Flow Chart Diagram** 18

**Tools And Technology** 19

**Approach** 20

**Appendix** 21

**Outputs** 39

**EDA Of Parkinson’s Disease** 43

* Heatmap43
* Histogram44
* Accuracy Score44
* Bar Diagram45

[**Voice - Based Diagnosis System Development**](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.f4wggtnfd7sw)  46

[● Dataset collection and preparation](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.kd7sk95vo3u9)  46

[● Pre-processing of voice data](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.zbhg08soaanf)  47

[● Feature extraction techniques](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.yjh9dyokbnk8)  48

[● Building the machine learning model](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.lk14nwqknxo8)  49

[● Testing and validation of the model](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.odhxtxq8iqz6)  51

**Results And Evaluation** 52

* The accuracy of the voice-based diagnosis system       52
* Comparison with traditional diagnosis methods       52
* Evaluation of the limitations and future improvements           53

**Accuracy Table** 55

**Applications** 56

**Strengths** 57

**Limitations** 58

**Conclusions** 59

[● The potential of AI and voice-based diagnosis in healthcare](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.o5ob9v1chr74)  59

[● The importance of early diagnosis in Parkinson's Disease](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.k1l3jtn6jk7m)  60

[● Future directions for research](https://docs.google.com/document/d/1zf0yDpAspnfga4pdBGyTzh_ZTxmQ9G46rw7liv6wSnI/edit#heading=h.plzudd1vxhk3)  60

**Future Development** 62

**References** 63

**Individual Contribution To The Project** 68

**CERTIFICATE OF APPROVAL**

This is to certify that the major project entitled “Disease Prediction System for Parkinson Disease and Breast Cancer using Voice Command” submitted towards the partial fulfilment of MCA 4th Semester, Department of Computer Application of Guru Nanak Institute of Technology(GNIT), was carried out by Mr. Mujahid Ali Ansari (Roll No - 504121011025), Mr. Shahil Kumar Chourasia (Roll No - 504121011045), Ms. Sirsha Majumder (Roll No - 504121021050), Mr. Sonu Routh (Roll No - 504121011051), Mr. Ashutosh Kumar Yadav (Roll No - 504121011006), under the supervision of Prof. **Dr. Ananjan Maity** , Asst. Prof. Dept. of Computer Application, Guru Nanak Institute of Technology. The matter embodied in this project work is genuine and motivating for future work.

Wish them all success.

--------------------------------------- ---------------------------------------------

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External Examiner

**ACKNOWLEDGEMENT**

We, all the team members, would like to take this opportunity to express our profound gratitude and kind regards to our guide **Dr. Ananjan Maity** (Assistant Professor – CA) for his exemplary guidance, monitoring, and constant encouragement throughout this project work. The blessings, help and guidance given by him to us shall carry us to a long way in our journeys of life on which we are about to embark.

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**Signature of Students Signature of Mentor**

**ABSTRACT**

This project introduces a System that can be used to predict if a patient is affected by **Parkinson Disease**. Even today, this diseases is very serious health issues that are persisting.

This system is quite ready to be used by the administration of hospitals as well – the staffs can use it to register new patients and predict if they have symptoms of Parkinson disease, send the patients’ health reports to their respective mail ids, retrieve health reports of existing patients in database and to discharge patients. Having all these registrations, predictions, etc. done before getting checked by the main doctor cuts off time consumption and makes check-ups easy for doctors as well.

Also, this system can be used by the Staff Head (admin) to access the data of all the working staffs from the database along with all the patients’ details and delete staff data if required. The admin only would have a very confidential password for this special access.

The add-on feature of this system is that, we **can use voice commands** to get the actions of the system done. This system is available to be used 24X7 by staffs and admin.

**OVERVIEW**

Giovanni Costantini [5] This study aims to develop a voice-based Parkinson's disease diagnosis system using artificial intelligence (AI) techniques. Parkinson's disease is a neurological disorder that affects a person's movement, causing tremors, stiffness, and difficulty with coordination. Early detection of Parkinson's disease is important for effective treatment and management of the condition. However, traditional diagnosis methods can be time-consuming and have limitations in accuracy. The voice-based diagnosis system developed in this study uses machine learning algorithms to analyse speech data collected from individuals with Parkinson's disease. The dataset was pre-processed and feature extraction techniques were used to extract relevant features for diagnosis. A machine learning model was trained and tested using the collected data to evaluate the accuracy of the diagnosis system. The results of this study indicate that the voice-based Parkinson's disease diagnosis system achieved high accuracy in identifying individuals with Parkinson's disease. The system showed promise in providing an early, non-invasive, and low-cost diagnosis method for Parkinson's disease. This study highlights the potential of AI and voice-based diagnosis techniques in healthcare and opens up new opportunities for early detection and management of Parkinson's disease.

**INTRODUCTION**

### Brief explanation of Parkinson's Disease

G. J. Canter[6] Parkinson disease is a neurological disorder that affects human movements. The condition occurs due to the degeneration of dopamine-producing neurons in a specific area of the brain called the substantia nigra. Dopamine is a neurotransmitter that plays a critical role in regulating movement, motivation, and reward. The loss of dopamine-producing neurons leads to a reduction in dopamine levels in the brain, resulting in the hallmark symptoms of Parkinson's disease. The primary symptoms of Parkinson's disease include tremors, rigidity, bradykinesia (slowness of movement), and postural instability. Other non-motor symptoms, such as depression, anxiety, sleep disturbances, and cognitive impairment, may also occur. Although there is currently no cure for Parkinson's disease, treatments such as medication, surgery, and therapy can help manage symptoms and improve quality of life for those affected by the condition.

### The importance of early diagnosis of Parkinson’s disease

Z. Karapinar Senturk[7] Early diagnosis is critical in the treatment and management of many medical conditions, including Parkinson's disease. Early diagnosis of Parkinson's disease can lead to early intervention and treatment, which can help improve symptoms, slow the progression of the disease, and improve quality of life.

1.  E. Waubant,[20] Early Intervention and Treatment: Early diagnosis of Parkinson's disease can lead to early intervention and treatment, which can help improve symptoms, slow the progression of the disease, and improve quality of life. Medication, surgery, physical therapy, and occupational therapy are all treatment options for Parkinson's disease, and early diagnosis can help people with Parkinson's disease start these treatments sooner.

2. D. Mellick,[21] Improved Quality of Life: Parkinson's disease can significantly impact a person's quality of life, and early diagnosis and treatment can help people with Parkinson's disease maintain their quality of life and continue to participate in daily activities. Early intervention and treatment can help improve mobility, reduce muscle stiffness, and improve balance, making it easier for people with Parkinson's disease to perform daily tasks and remain independent.

3. Better Management of Symptoms: Parkinson's disease is a progressive disease, and symptoms can worsen over time. Medication and physical therapy can help reduce tremors, stiffness, and balance problems, and occupational therapy can help people with Parkinson's disease learn new ways to perform daily tasks.

4. E. R. Dorsey *et al.*,[22] Increased Access to Clinical Trials: Early diagnosis of Parkinson's disease can also increase access to clinical trials. By participating in clinical trials, people with Parkinson's disease can access new treatments and therapies that may not yet be widely available.

5. W. Wang, J. Lee, F. Harrou[8] Improved Mental Health: Parkinson's disease can also impact a person's mental health, leading to depression, anxiety, and other emotional issues. Early diagnosis and treatment can help manage these symptoms, improve mental health, and improve overall quality of life.

6. Family Planning: Parkinson's disease can have a genetic component, and early diagnosis can help families plan for the future. Genetic testing can help determine if family members are at risk for developing Parkinson's disease, and early diagnosis can lead to early intervention and treatment, which can help reduce the risk of developing symptoms.

In conclusion, early diagnosis is critical in the treatment and management of many medical conditions, including Parkinson's disease. If you or someone you know is experiencing symptoms of Parkinson's disease, it is important to seek medical attention as soon as possible.

### The potential of AI and voice-based diagnosis

Artificial intelligence (AI) and voice-based diagnosis have shown tremendous potential in the early detection and diagnosis of a wide range of medical conditions, including Parkinson's disease.

AI and Parkinson's Disease Diagnosis:

I. Nissar, D. Rizvi, S. Masood[9] AI-based algorithms have shown promising results in detecting and diagnosing Parkinson's disease using a wide range of data sources, including voice samples, gait analysis, and digital biomarkers. Voice-based diagnosis is a non-invasive and cost-effective approach that has shown particular promise in early diagnosis of Parkinson's disease.

Studies have shown that people with Parkinson's disease often exhibit changes in their voice, such as decreased loudness, monotone pitch, and altered speech rate. These changes are often subtle and difficult to detect, even by experienced healthcare professionals. AI-based algorithms have been developed that can analyze voice samples and identify these changes, providing a reliable and accurate tool for early diagnosis of Parkinson's disease.

G. Solana-Lavalle[23] Voice-based diagnosis is particularly useful for remote monitoring of Parkinson's disease patients, as it enables healthcare professionals to monitor patients' symptoms and disease progression without the need for frequent in-person appointments. This approach can be genuinely beneficial during pandemic situations like the Covid-19, as it can help reduce the risk of exposure to the virus for both patients and healthcare professionals.

AI and voice-based diagnosis offer several benefits for Parkinson's disease diagnosis, including:

1.G. Costantini *et al.*, [24] Early Detection and Diagnosis: AI and voice-based diagnosis can help detect Parkinson's disease at an early stage, which can lead to early intervention and treatment. Early intervention can help slow the progression of the disease and improve quality of life for people with Parkinson's disease.

2. S. Marziyeh Ghoreshi Beyrami [25] Non-invasive and Cost-effective: Voice-based diagnosis is a non-invasive and cost-effective approach that can be easily implemented in a wide range of healthcare settings. This approach can help reduce the cost and burden of Parkinson's disease diagnosis and monitoring, making it more accessible to a wider range of patients.

3. D. Gatsios *et al.*,[26] Remote Monitoring: AI and voice-based diagnosis enable remote monitoring of Parkinson's disease patients, which can help improve patient outcomes and reduce the burden on healthcare systems. This approach can help reduce the need for frequent in-person appointments, making healthcare more convenient and accessible for patients.

4. Personalized Treatment: AI-based algorithms can analyze large amounts of data to provide personalized treatment recommendations for people with Parkinson's disease. This approach can help healthcare professionals tailor treatment plans to the individual needs of each patient, improving treatment outcomes and quality of life.

5. W. G. Meissner *et al.*,[27] Improved Research: AI and voice-based diagnosis can also improve research into Parkinson's disease by providing large amounts of data for analysis. This approach can help researchers identify new biomarkers, treatments, and therapies for Parkinson's disease, leading to better treatment options for patients.

Conclusion:

In conclusion, AI and voice-based diagnosis offer tremendous potential for early detection and diagnosis of Parkinson's disease. Voice-based diagnosis is a non-invasive and cost-effective approach that can be easily implemented in a wide range of healthcare settings, enabling remote monitoring of patients and personalized treatment recommendations. AI-based algorithms can analyze large amounts of data to provide accurate and reliable diagnosis, improving treatment outcomes and quality of life for people with Parkinson's disease. As research in this area continues to advance, AI and voice-based diagnosis will likely play an increasingly important role in the early detection and management of Parkinson's disease.

## **The Science behind Parkinson’s Disease Diagnosis**

### Traditional diagnosis methods of Parkinson’s Disease

 J. Mei, C. Desrosiers, and J. Frasnelli,[12] There are several traditional diagnosis methods that can be used to diagnose Parkinson's disease. There are some common methods of diagnosis:

1. Clinical assessment: A clinical assessment involves a physical examination and neurological evaluation by a trained healthcare professional, such as a neurologist. The assessment may include tests of motor function, such as assessing gait, balance, and hand coordination, as well as assessments of non-motor symptoms, such as mood, cognition, and sleep.

2.  C. G. Goetz[28] Medical history: Taking a detailed medical history is an important component of Parkinson's disease diagnosis. A healthcare professional may ask about symptoms, family history of Parkinson's disease, and exposure to environmental toxins or medications that may increase the risk of Parkinson's disease.

4. J. Santiago and J. Potashkin[47], Blood tests: While there is no specific blood test for Parkinson's disease, blood tests can be used to rule out other conditions that may cause similar symptoms, such as thyroid disorders or vitamin deficiencies.

5. M. D. Hssayeni, [29] Response to medication: Parkinson's disease symptoms often respond well to medications that increase dopamine levels in the brain. A positive response to medication, such as levodopa, can be a diagnostic indicator of Parkinson's disease.

It is important to note that no single diagnostic test can definitively diagnose Parkinson's disease. A combination of clinical assessments, medical history, imaging studies, and response to medication is typically used to make a diagnosis. Additionally, early diagnosis and treatment are important for managing symptoms and improving quality of life.

### Symptoms of Parkinson's Disease

Symptoms: The symptoms of Parkinson's disease usually develop gradually and worsen over time. The primary symptoms of Parkinson's disease are motor symptoms, which include:

1.  S. Sveinbjornsdottir[10] Tremors: Tremors or shaking in the hands, arms, legs, jaw, or face is one of the most recognizable symptoms of Parkinson's disease.

2. Bradykinesia: Slowness of movement or difficulty initiating movement is another common symptom of Parkinson's disease. It can make simple tasks like getting dressed or brushing teeth more challenging.

3. Rigidity: Stiffness in the arms, legs, or torso can make movement uncomfortable and painful.

4. Postural instability: Difficulty maintaining balance and coordination, which can lead to falls.

Other symptoms of Parkinson's disease may include:

W. C. Koller[11] 1. Difficulty in swallowing: Parkinson's disease can cause difficulty in swallowing, which can lead to choking or aspiration pneumonia.

2. Loss of sense of smell: Many people with Parkinson's disease experience a loss of sense of smell.

3. S. Ray and P. Agarwal[46], Depression and anxiety: Parkinson's disease can lead to depression and anxiety, which can worsen the symptoms of the disease.

Cure: Currently, there is no cure for Parkinson's disease, but several treatments can help manage symptoms and improve quality of life. The primary treatment for Parkinson's disease is medication, which works by increasing dopamine levels in the brain. The most commonly used medication is levodopa, which is converted into dopamine in the brain. Other medications used to treat Parkinson's disease include dopamine agonists, which mimic the action of dopamine in the brain, and MAO-B inhibitors, which help prevent the breakdown of dopamine.

Physical therapy can also be helpful for people with Parkinson's disease. Physical therapy can help improve balance, reduce muscle stiffness, and increase mobility. Occupational therapy can also be helpful in teaching people with Parkinson's disease how to perform daily tasks more easily.

In conclusion, Parkinson's disease is a chronic and progressive neurological disorder that primarily affects movement. The exact cause of Parkinson's disease is not fully understood, but it is believed to be caused by a combination of genetic and environmental factors. Early diagnosis and treatment can help people with Parkinson's disease maintain their quality of life and continue to participate in daily activities.

### Limitations of traditional diagnosis methods of Parkinson Disease

S. E. Lacy[13] While traditional diagnosis methods for Parkinson's disease, such as clinical assessments and imaging studies, can be effective, they have some limitations that can affect their accuracy and reliability. Here are some of the limitations of traditional diagnosis methods:

Subjectivity: Clinical assessments rely on the subjective observations of clinicians and may be influenced by their experience and training. This can lead to variability in diagnoses and may affect the accuracy of diagnosis.

A. Schrag[48], Late diagnosis: Traditional diagnosis methods may not be able to detect Parkinson's disease in its early stages, when symptoms may be mild or non-specific. This can lead to delayed diagnosis and treatment, which can impact the effectiveness of treatment and quality of life.

Cost: Some traditional diagnosis methods, such as imaging studies, can be expensive and may not be accessible to all patients. This can limit their usefulness in certain populations.

Invasive: Some diagnostic tests, such as lumbar puncture, require invasive procedures that can be uncomfortable or risky for patients.

D. Aarsland,[30] Limited information: Traditional diagnostic methods may not provide a complete picture of the underlying pathology of Parkinson's disease, which can limit the ability to develop personalized treatment plans.

To overcome these limitations, researchers are exploring new diagnostic tools and technologies, such as biomarker tests, wearable sensors, and artificial intelligence algorithms, that may offer more objective, accurate, and non-invasive methods of diagnosis. These technologies have the potential to improve early detection, personalize treatment, and improve patient outcomes in Parkinson's disease and other neurological disorders.

**The Role of Artificial Intelligence in Parkinson’s Disease Diagnosis**

### Advantages of AI diagnosis of Parkinson’s Disease

M. Belić, V. Bobić, [14] Artificial intelligence (AI) has the potential to revolutionize the diagnosis of Parkinson's disease by offering several advantages over traditional diagnostic methods. Here are some of the advantages of AI diagnosis:

1. Objective and standardized: AI algorithms can provide an objective and standardized analysis of diagnostic data, eliminating the subjectivity and variability that can be present in traditional diagnostic methods.

2. Early detection: AI algorithms can detect Parkinson's disease at an earlier stage when symptoms may be subtle or non-specific, allowing for earlier intervention and better treatment outcomes.

3. Non-invasive: Many AI-based diagnostic tools, such as voice-based analysis or wearable sensors, are non-invasive and can be easily integrated into routine clinical care without requiring uncomfortable or invasive procedures.

4. Cost-effective: Some AI-based diagnostic tools are relatively inexpensive compared to traditional diagnostic methods such as imaging studies, making them accessible to a wider range of patients.

5. T. Stoddard-Bennett[49], Personalized treatment: AI algorithms can provide more detailed information about the underlying pathology of Parkinson's disease, allowing for more personalized treatment plans tailored to the individual patient's needs.

6. Remote monitoring: AI-based diagnostic tools can enable remote monitoring of patients, allowing for more frequent and timely assessments of symptoms, which can improve disease management and reduce hospitalizations.

Overall, AI-based diagnostic tools have the potential to significantly improve the accuracy, efficiency, and accessibility of Parkinson's disease diagnosis, leading to earlier intervention, better treatment outcomes, and improved quality of life for patients.

### Machine Learning algorithms used in Parkinson's Disease diagnosis

I. Mandal and N. Sairam,[15] Machine learning algorithms are increasingly being used to improve the accuracy and efficiency of Parkinson's disease diagnosis. Here are some of the machine learning algorithms commonly used in Parkinson's disease diagnosis:

1.  M. Shahbakhi[31] Support Vector Machines (SVMs): SVMs are supervised learning algorithms that can be used to classify patients with Parkinson's disease based on input data. SVMs work by finding the optimal hyperplane that separates data points into different classes.

2. R. Rone Sarra, [32] Artificial Neural Networks (ANNs): ANNs are a type of machine learning algorithm that mimic the structure and function of the human brain. ANNs can be used to analyze complex data sets and identify patterns that may not be apparent to humans.

3. T. P. Exarchos *et al.*, [33] Decision Trees: Decision trees are a type of machine learning algorithm that uses a tree-like structure to model decisions and their possible consequences. Decision trees can be used to identify the most important features or variables that contribute to Parkinson's disease diagnosis.

4. K. Polat[50], Random Forests: Random forests are an ensemble learning method that combines multiple decision trees to improve accuracy and reduce overfitting. Random forests can be used to classify patients with Parkinson's disease based on multiple input variables.

5. H. W. Loh *et al.*,[34]. Deep learning: Deep learning is a type of machine learning that uses artificial neural networks with multiple layers to analyze complex data sets. Deep learning can be used to classify patients with Parkinson's disease based on input data from imaging studies or other sources.

6. V. Despotovic,Gaussian[51], Process Regression (GPR): GPR is a type of machine learning algorithm that can be used for regression analysis. GPR can be used to predict disease progression in Parkinson's disease based on input data, such as motor function assessments or biochemical markers.

These machine learning algorithms have shown promise in improving the accuracy and efficiency of Parkinson's disease diagnosis, as well as predicting disease progression and response to treatment. However, further research is needed to validate the use of these algorithms in clinical practice and to optimize their performance for Parkinson's disease diagnosis and management.

### Voice-based diagnosis using AI

Karaman[16] Voice-based diagnosis using AI is a cutting-edge technology that is transforming the healthcare industry. By analyzing voice patterns, AI algorithms can detect potential medical conditions or diseases, allowing doctors and healthcare professionals to make quicker and more accurate diagnoses. In this article, we will explore the process of voice-based diagnosis using AI and its potential applications.

Data Collection: The first step in the voice-based diagnosis using AI is data collection. Large datasets of voice recordings are needed to train machine learning models to identify patterns and make accurate diagnoses. These datasets are collected from patients who have been diagnosed with various medical conditions, including respiratory illnesses, neurological disorders, and mental health conditions.

S. M. van Rooden,[35] Data Preprocessing: Once the datasets are collected, they are preprocessed to extract meaningful features. This involves analyzing the voice recordings to identify patterns and extract relevant features that can be used to train the AI models.

A. Zhan *et al.*[36] Machine Learning: The preprocessed data is then used to train machine learning models. There are various machine learning algorithms that can be used for voice-based diagnosis, including neural networks, decision trees, and support vector machines. These algorithms are trained to identify patterns in the voice recordings and make accurate diagnoses.

In conclusion, voice-based diagnosis using AI is a rapidly advancing technology that has the potential to revolutionize the way medical conditions are diagnosed and managed.

**OBJECTIVE**

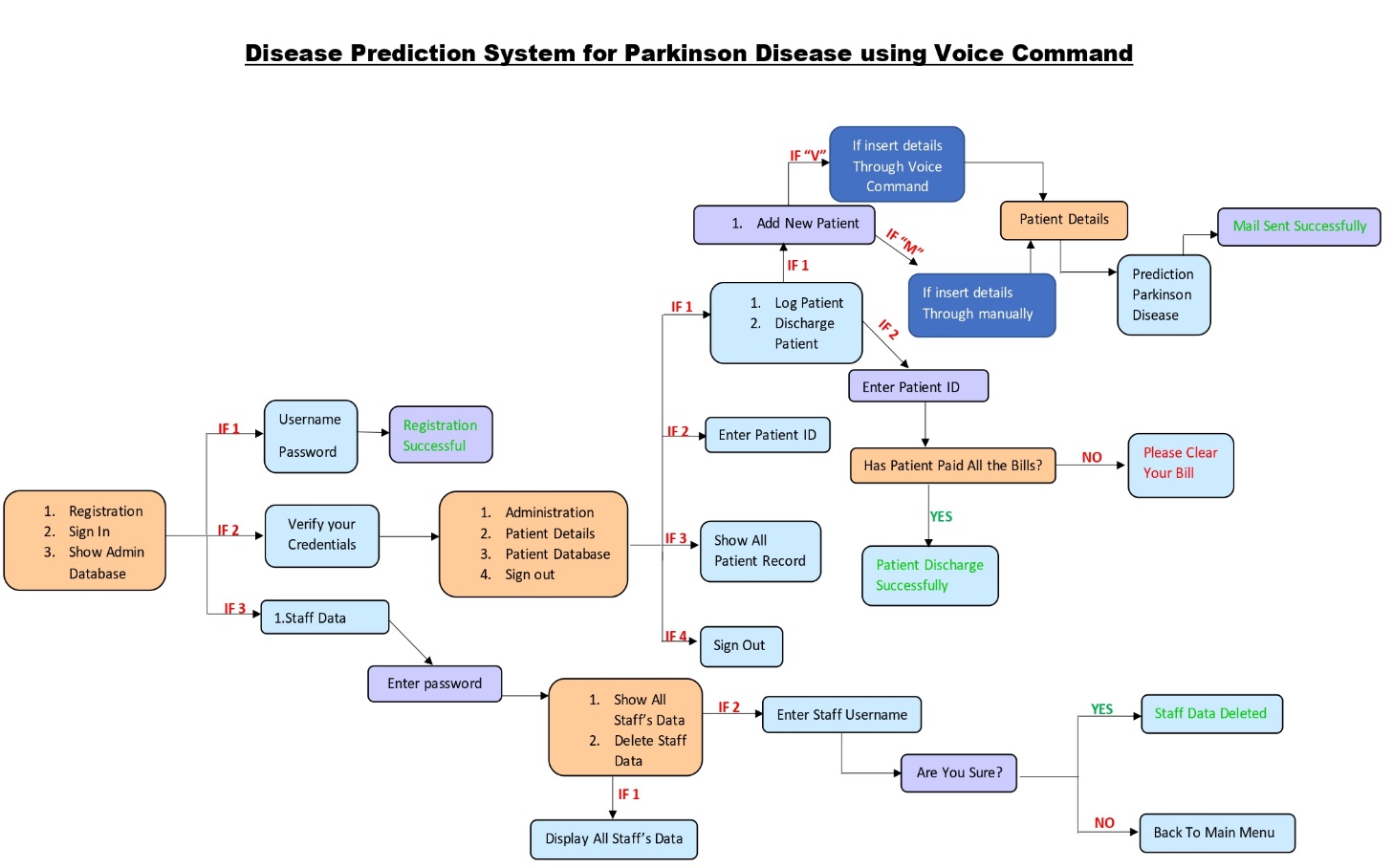
The primary aim of our project this project is to **predict the Parkinson** from the given symptoms by patients and create and monitor a health profile of every individual patient.

**Health is the most important factor for everyone**. But unfortunately it has been neglected today for many reasons. Absence of doctor due to some reason during emergency may result in loss of life. Not only that sometimes patient often feel hesitant to go to hospital for minor symptoms. These may prompt into major illness. With the proliferation of technology in health care becomes easier to diagnosis any disease – even the deadliest ones. It is applied in healthcare to identify the clusters of patients, diseases, and future predictions using different machine learning tools. So, this project work proposed a Disease Prediction System that will help user to receive immediate guidance regarding their health issues.

To diagnose any disease, doctor initially analyses the symptoms of the patient and after that the result is predicted. Similarly, machine diagnose the diseases based on the symptoms just like the doctor does. The system is fetched with various symptoms and their disease related with it.

This system aims to improve disease treatment and its diagnosis in early stages for a faster and better treatment. Therefore, it is an attempt to make a faster and more accurate disease prediction and help the physicians for making a reliable decision in a short span of time by increasing efficiency and quality in health management system.

**FLOW CHART DIAGRAM**

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The above flow-chart diagram represents the total workflow of the hospital management system including all operations of patients’ database as well as staffs’ database and predicts if the patient is suffering from Parkinson Disease.

**TOOLS AND TECHNOLOGY**

**Tools and Software**

* Visual Studio Code:

Visual Studio Code is a code editor that is redefined and optimized for building and debugging modern web and cloud applications. Visual it is free and also available on our favourite platforms – Linux, macOS, and Windows.

* MySql :

MySQL Workbench is a graphical tool that works with [MySQL](https://database.guide/what-is-mysql/). MySQL Workbench offers an easy-to-use interface to perform multiple tasks involved when working with databases. It integrates SQL development, administration, database design, creation, and maintenance into one visual integrated development environment. MySQL Workbench is similar to SQL Server’s [SSMS](https://database.guide/what-is-sql-server-management-studio/), which is used for administering [SQL Server](https://database.guide/what-is-sql-server/).

**Programming Language :**

* Python :

Python is a high-level programming language, that precludes the need to compile code before executing a program because Python does the compilation in the background. Because Python is a high-level programming language, it abstracts many sophisticated details from the programming code.

**Skill-based Technology :**

* Machine Learning :

Machine learning is a subfield of artificial intelligence which includes the development of algorithms and statistical models that enable computers to improve their performance in tasks through experience. These algorithms and models are designed to learn from data and make predictions or decisions without explicit instructions.

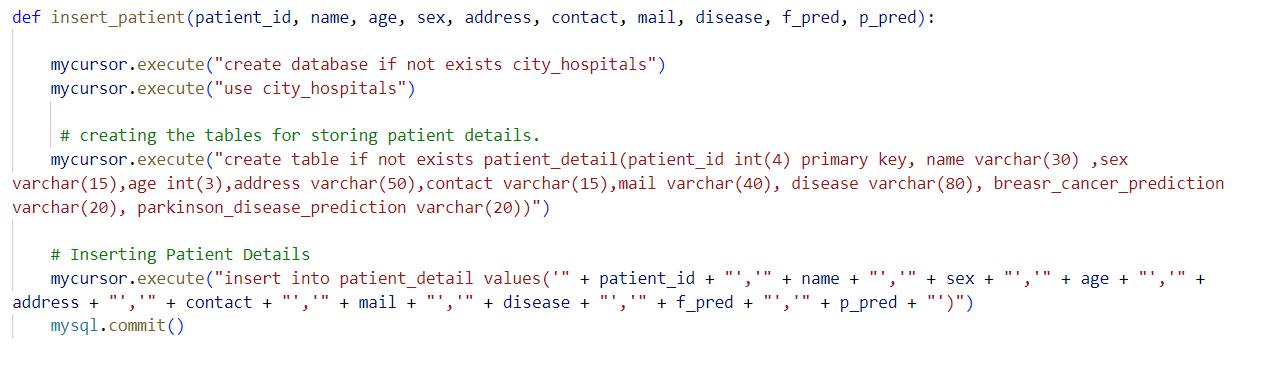
**APPROACH**

* **Database**
* Patient\_db.py
* insert\_patient()
* show\_patientdb()
* delete\_patientdb()
* show\_all\_patientdb()
* Staff\_db.py
* insert\_staff()
* show\_staff()
* show\_all\_staffdb()
* delete\_staff()
* **Predicted Disease**
* Parkinson.py
* Parkinson()
* **Send Mail**
* Email.py
* PDF()
* Send\_mail()
* **Voice Command**
* Command.py
* takeCommand()
* Speak.py
* Speak()
* Wish.py
* wishMe()
* **main.py**

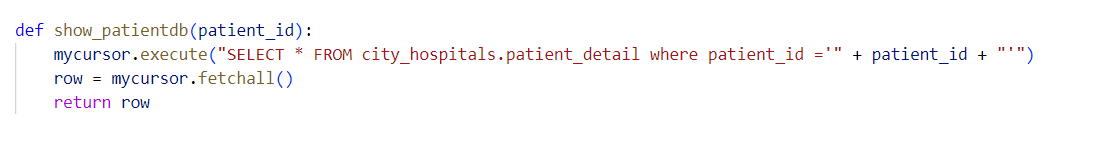
**APPENDIX**

1. **Database**

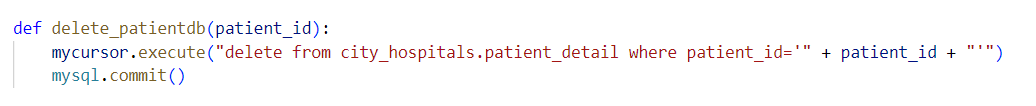
* **Patient\_db.py**
* **insert\_patient()** : This function is used to insert patient details in a database, especially when a new patient is being registered.



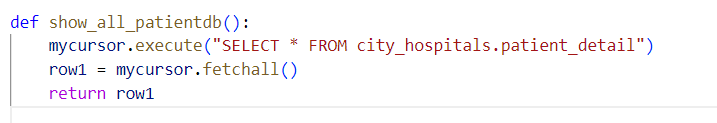
* **show\_patientdb()** : This function is used to retrieve patients’ records. Staff members can get specific patient details by using their patient ID to keep the track of their health issues.



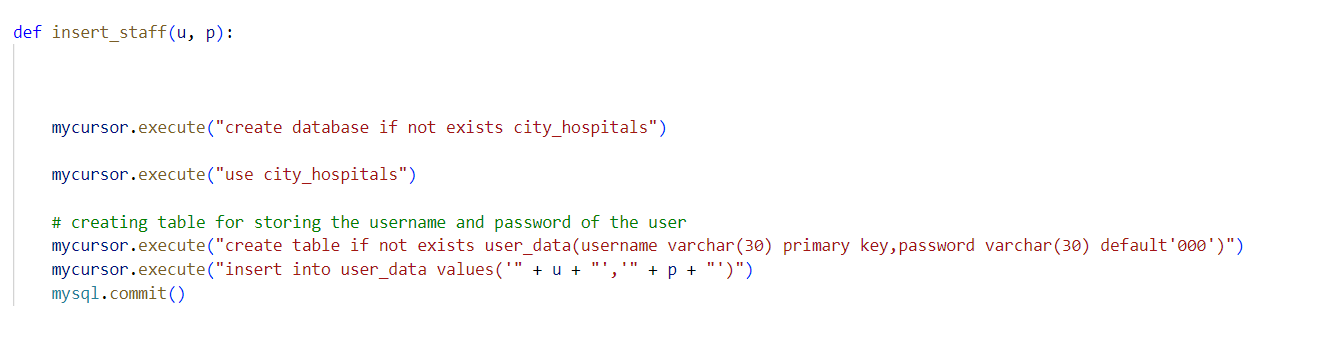
* **delete\_patientdb()** : This function is used to delete the patient details from the database if it’s no longer required.



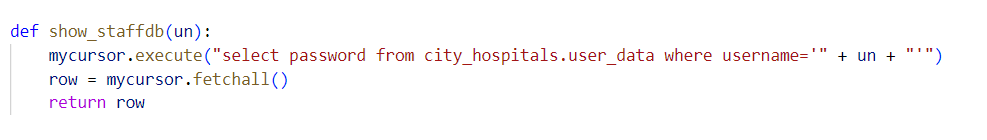
* **show\_all\_patientdb()** : This function is used to show all the patients’ records present in the database.



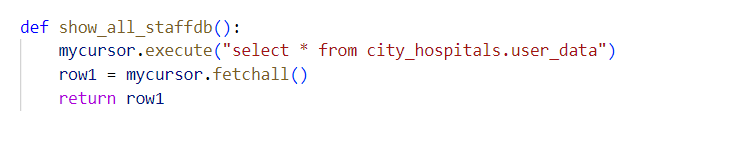
* **Staff\_db.py**
* **insert\_staff()** : This function is used to insert staff details in the database like their username and password.



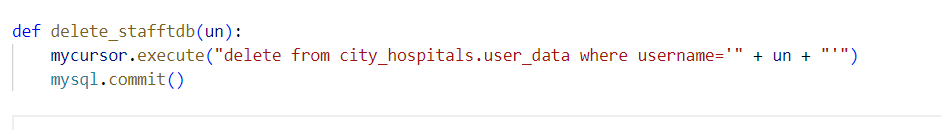
* **show\_staff()** : This function is used to retrieve all the Staff details from the database.

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* **show\_all\_staffdb()** : This function is used to show all the staff records present in the database.

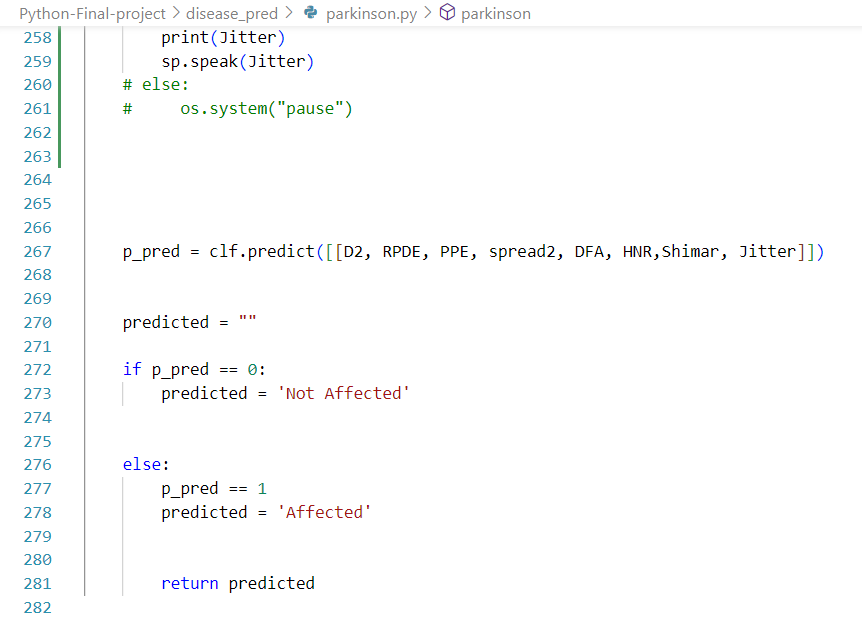
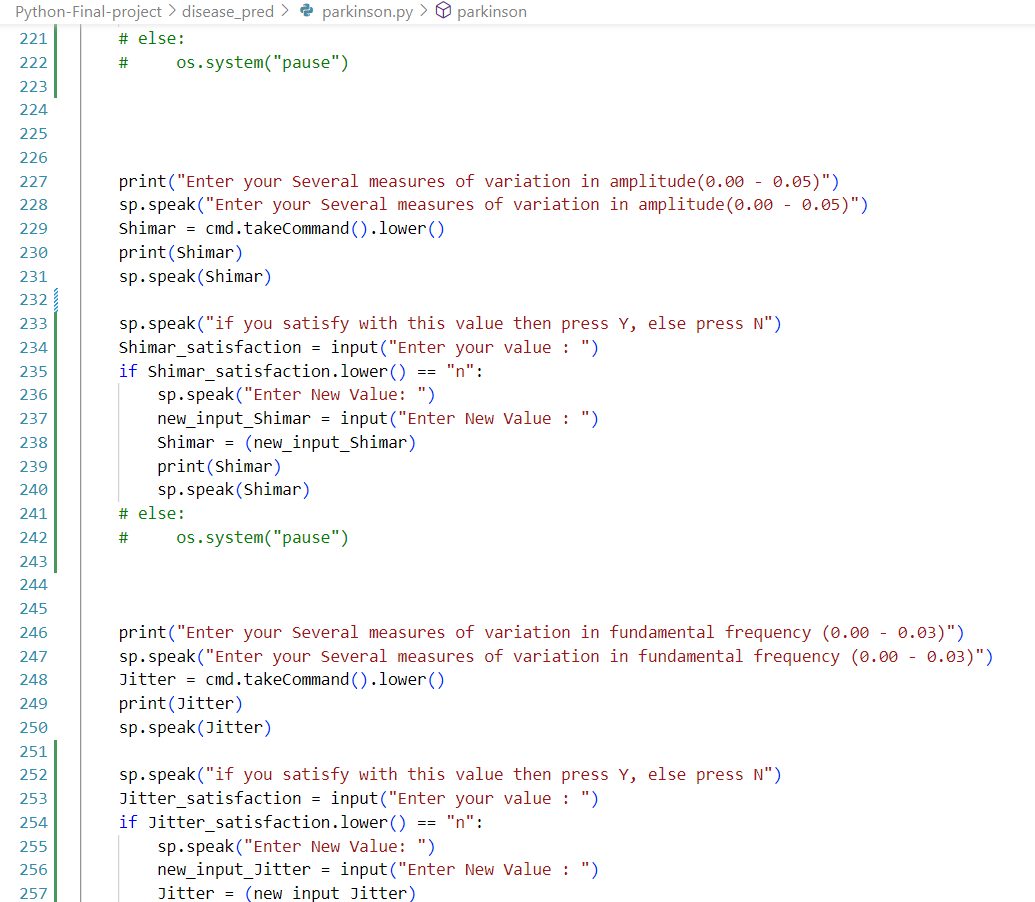
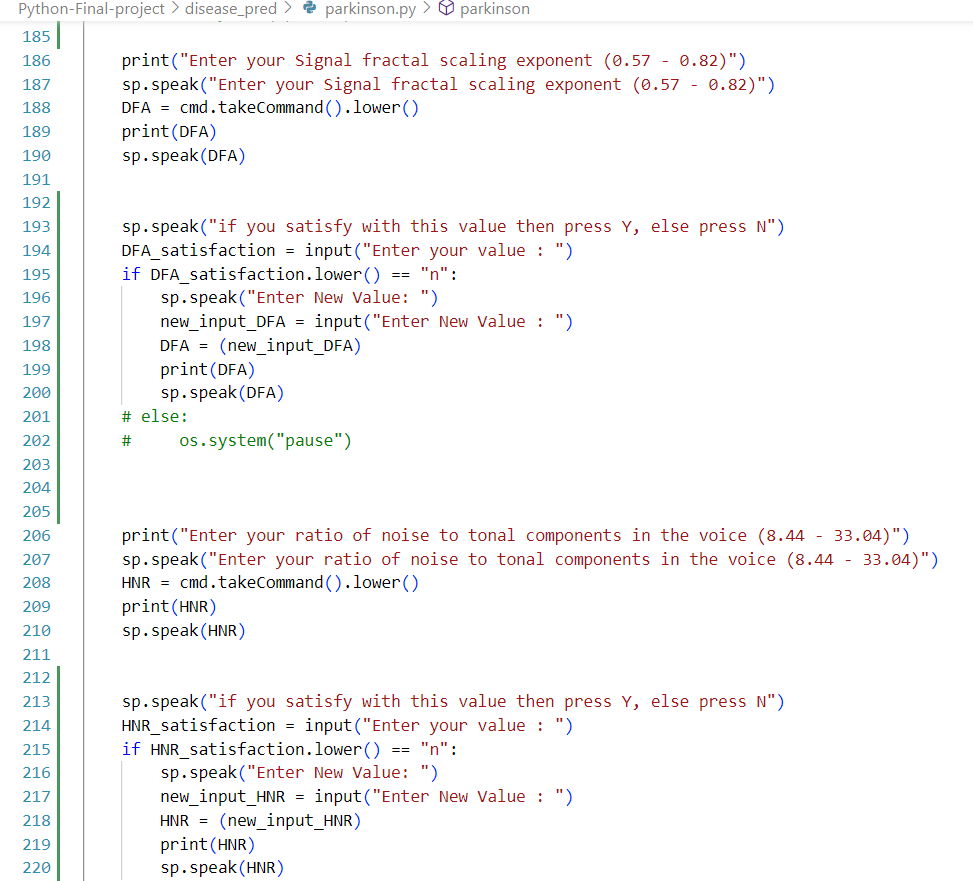
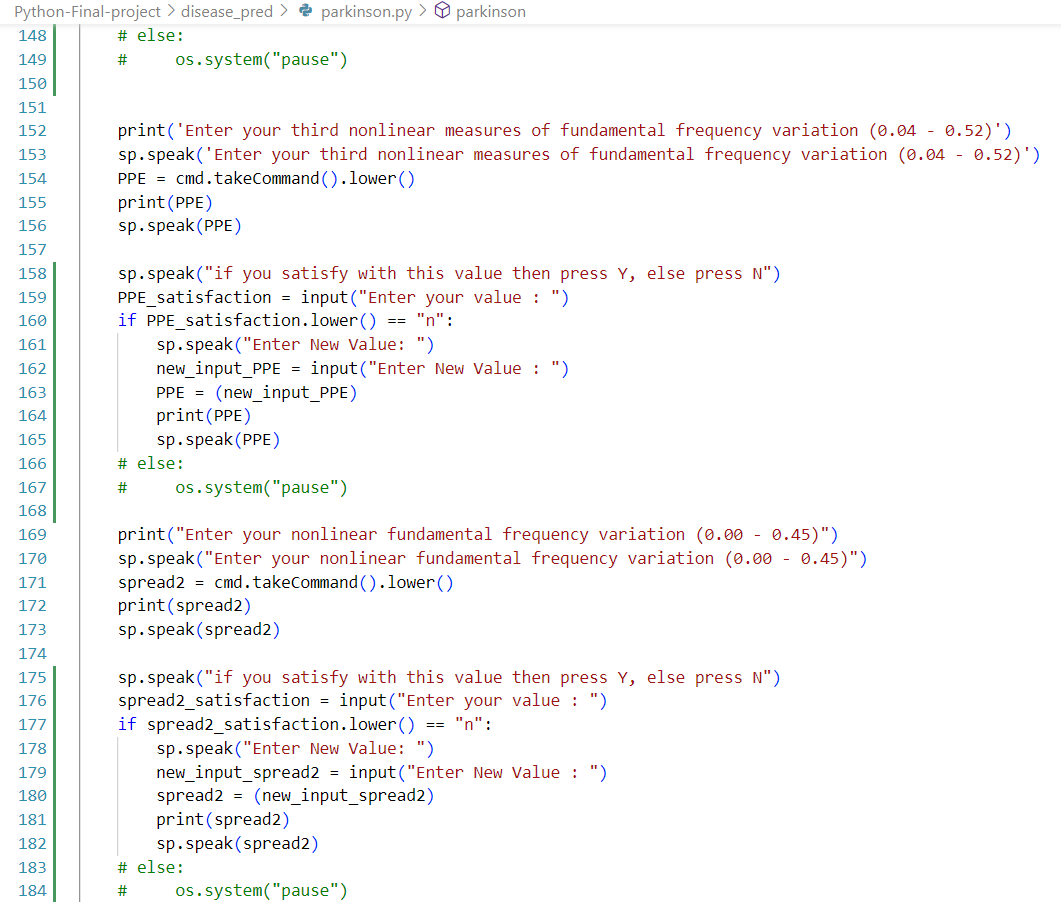
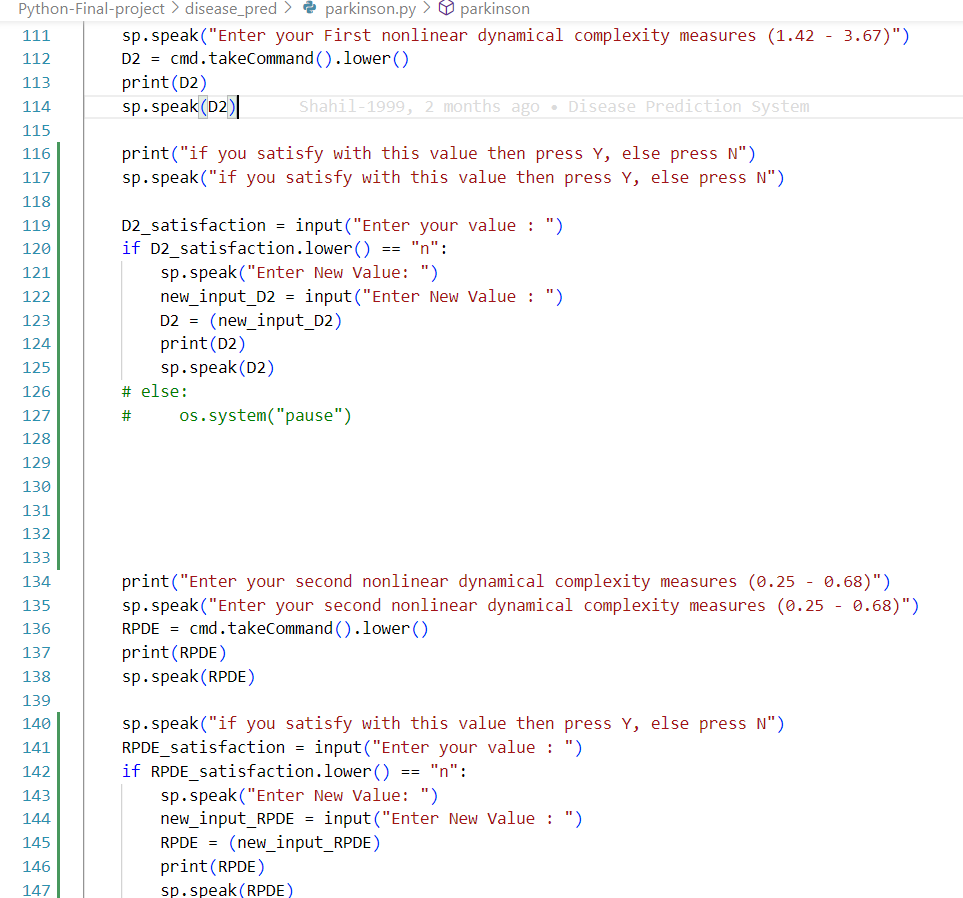
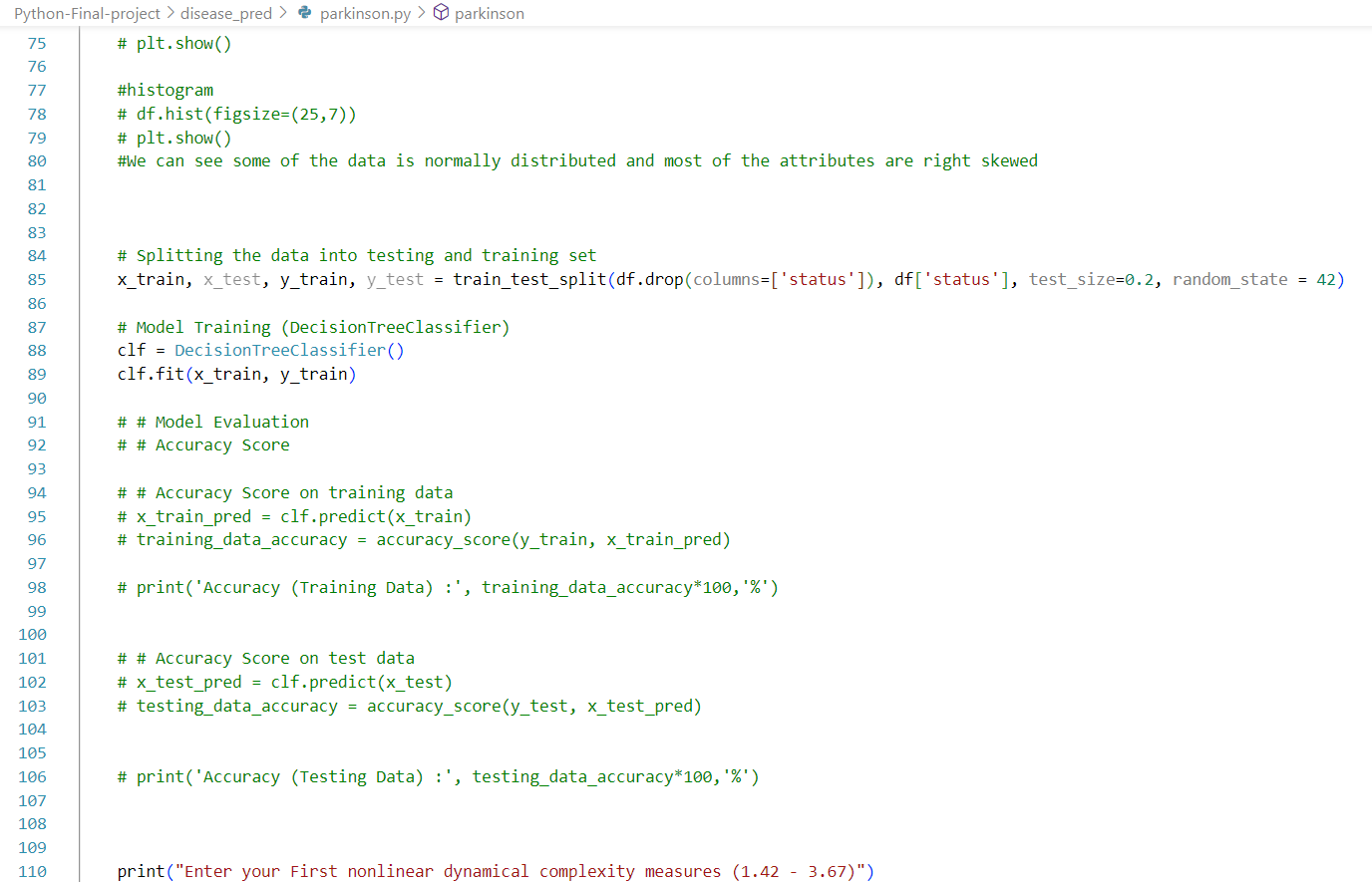
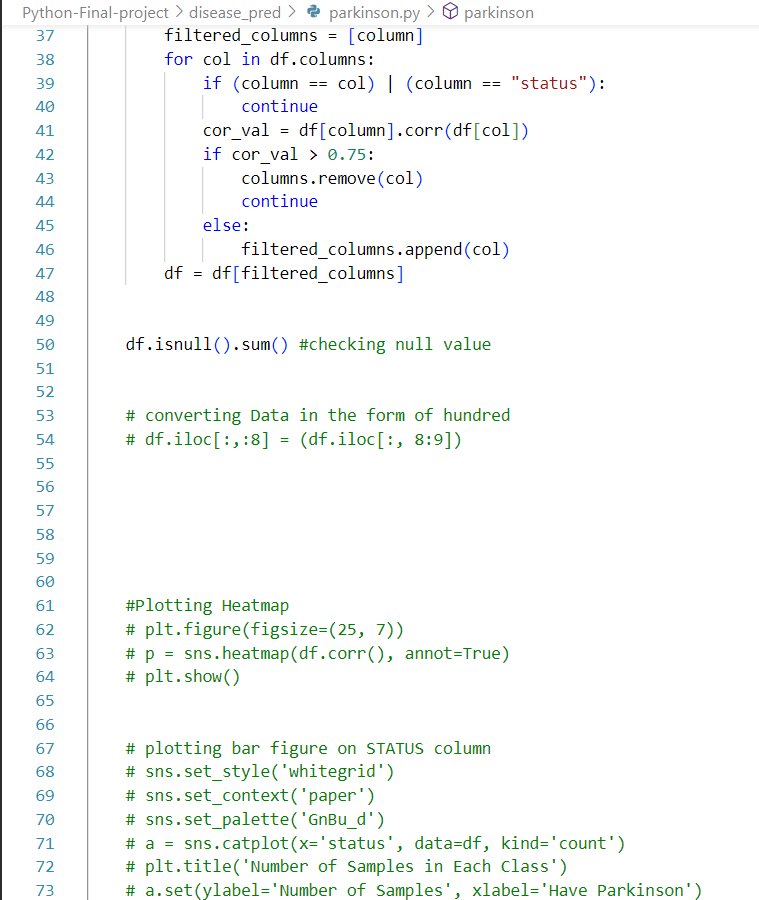
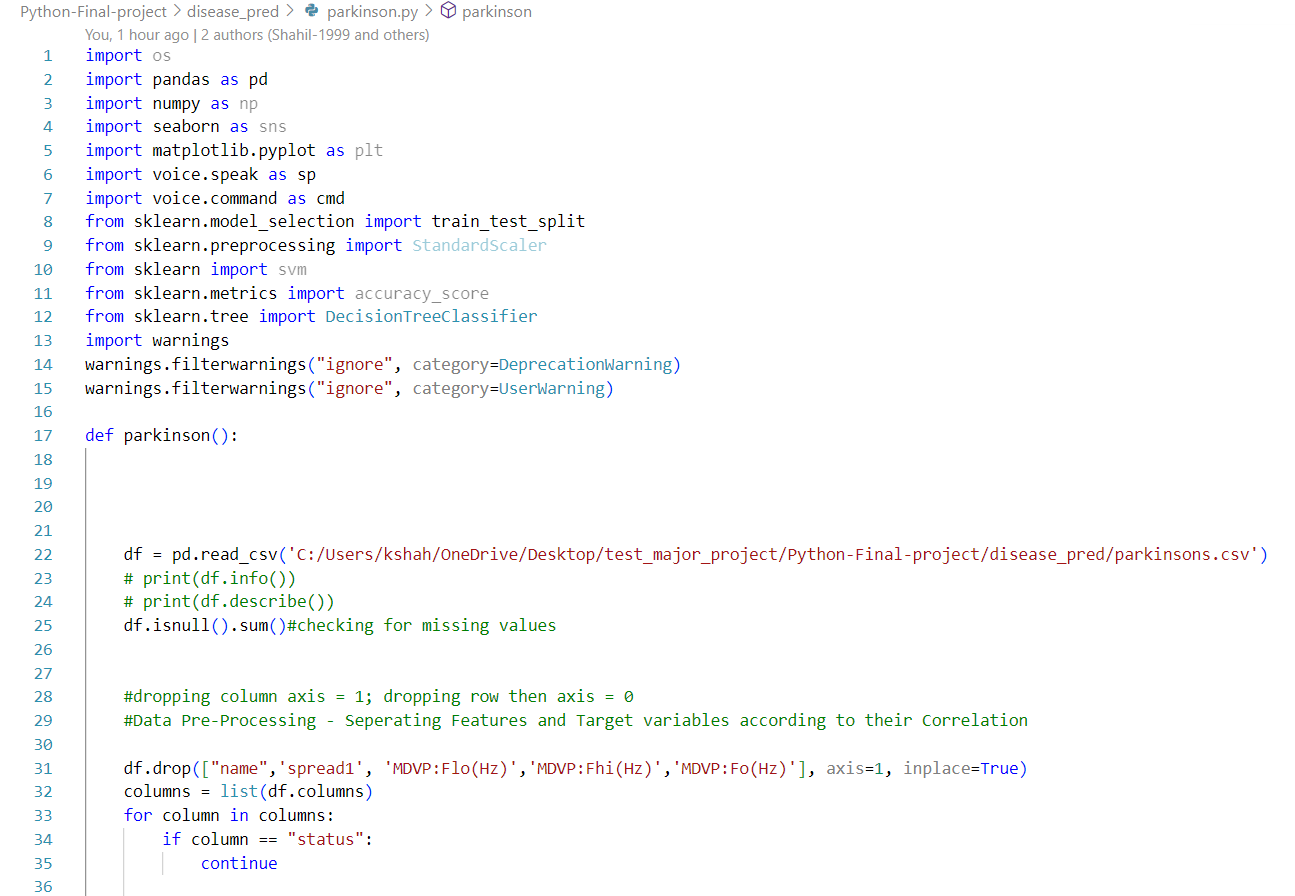


* **delete\_staff()** : This function is used to delete the staff details from the database if it’s no longer required.

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1. **Predicted Disease**

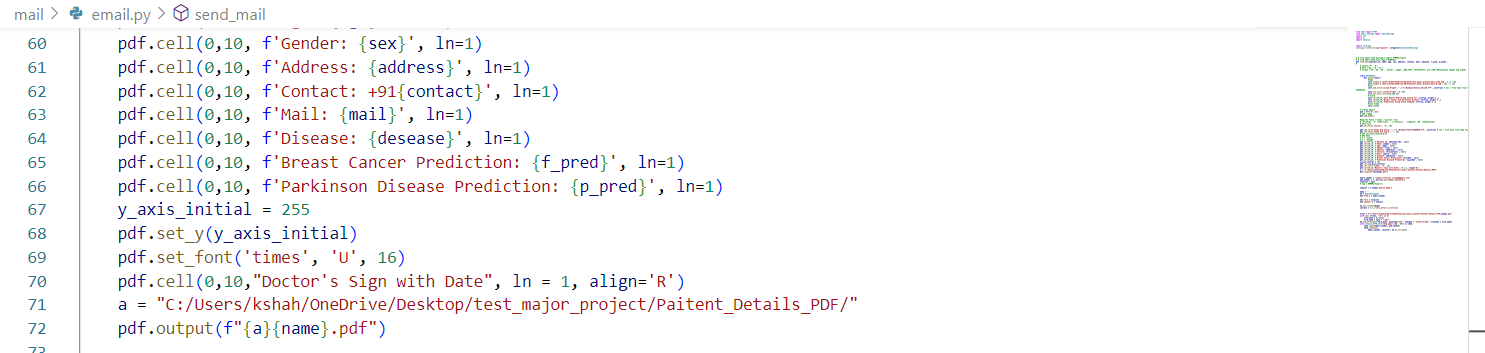
* **Parkinson.py**
* **Parkinson()** : This function is used to predict the Parkinson Disease.



1. **Send Mail**

* **Email.py**
* **PDF()** : This function is used to create patient reports in pdf format.





* **Send\_mail()** : This function is used to automatically send the PDF report to the patient via email.

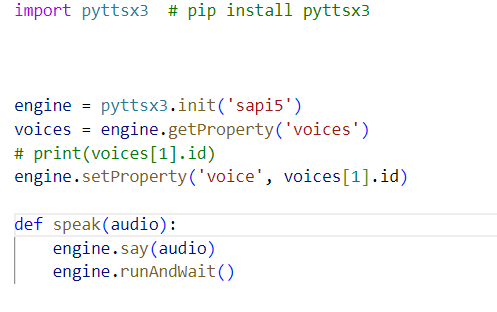
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1. **Voice Command**

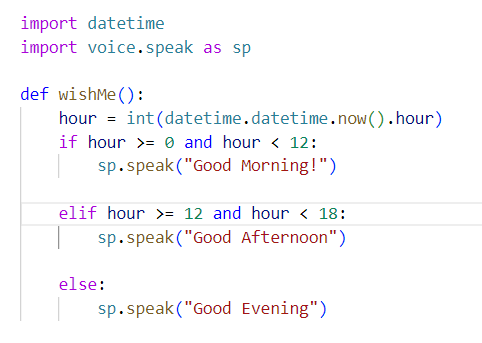
* **Command.py**
* **takeCommand()** : This function is used to take Voice Commands from the user.



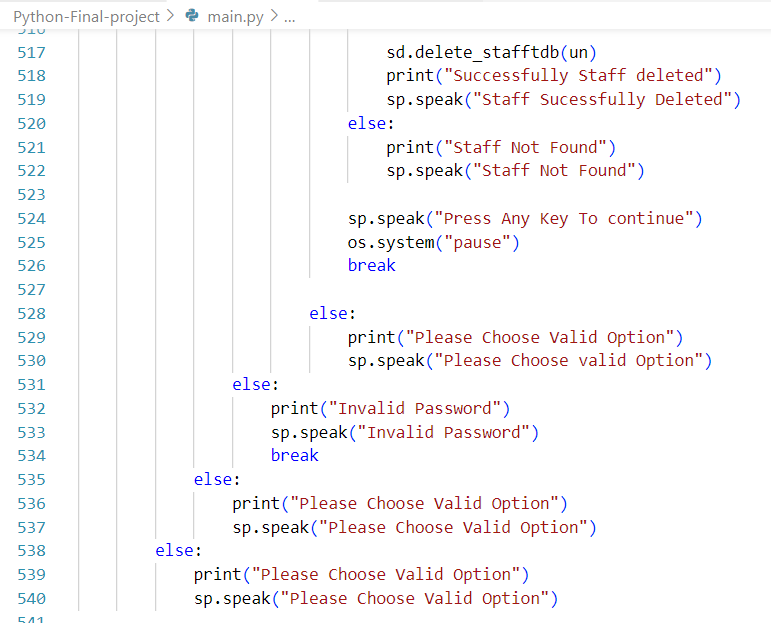
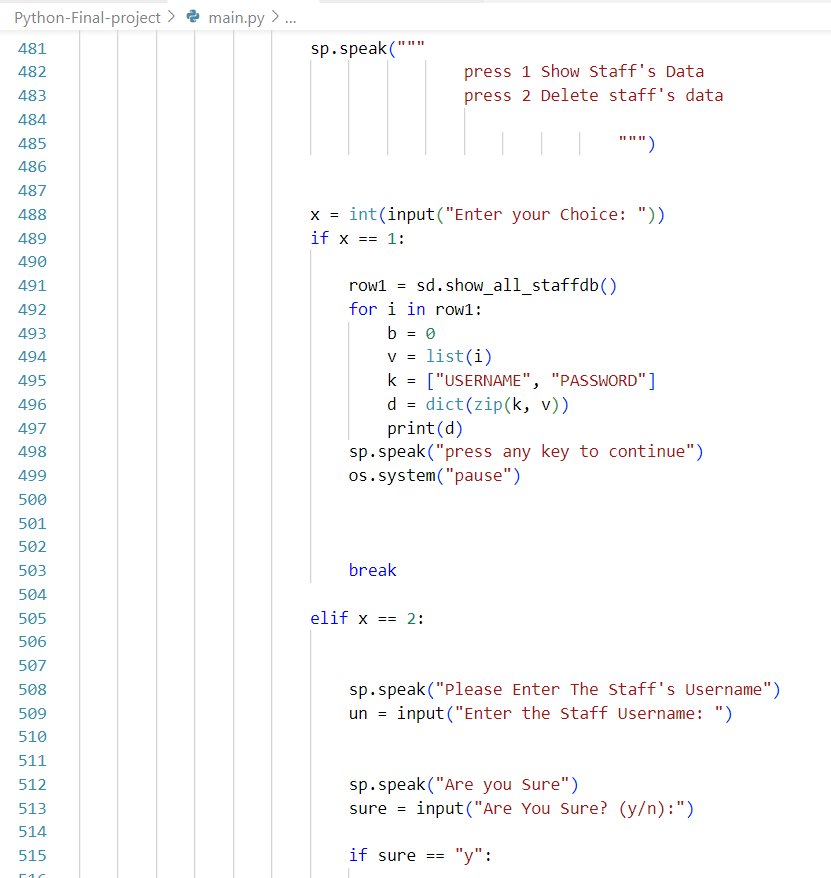
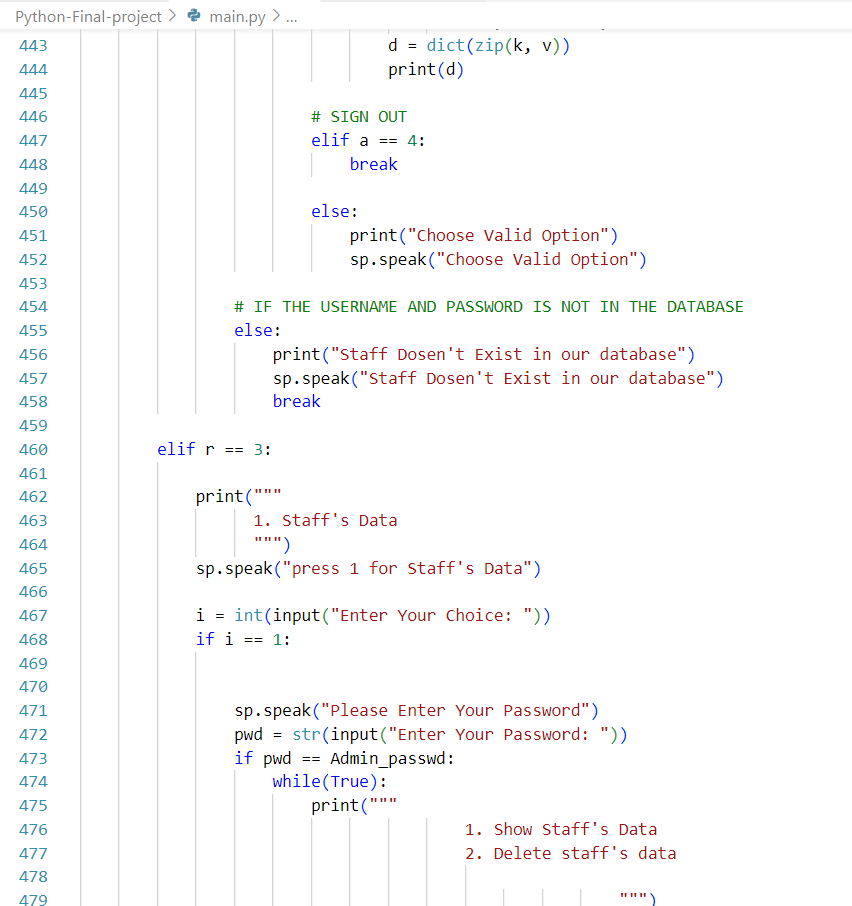
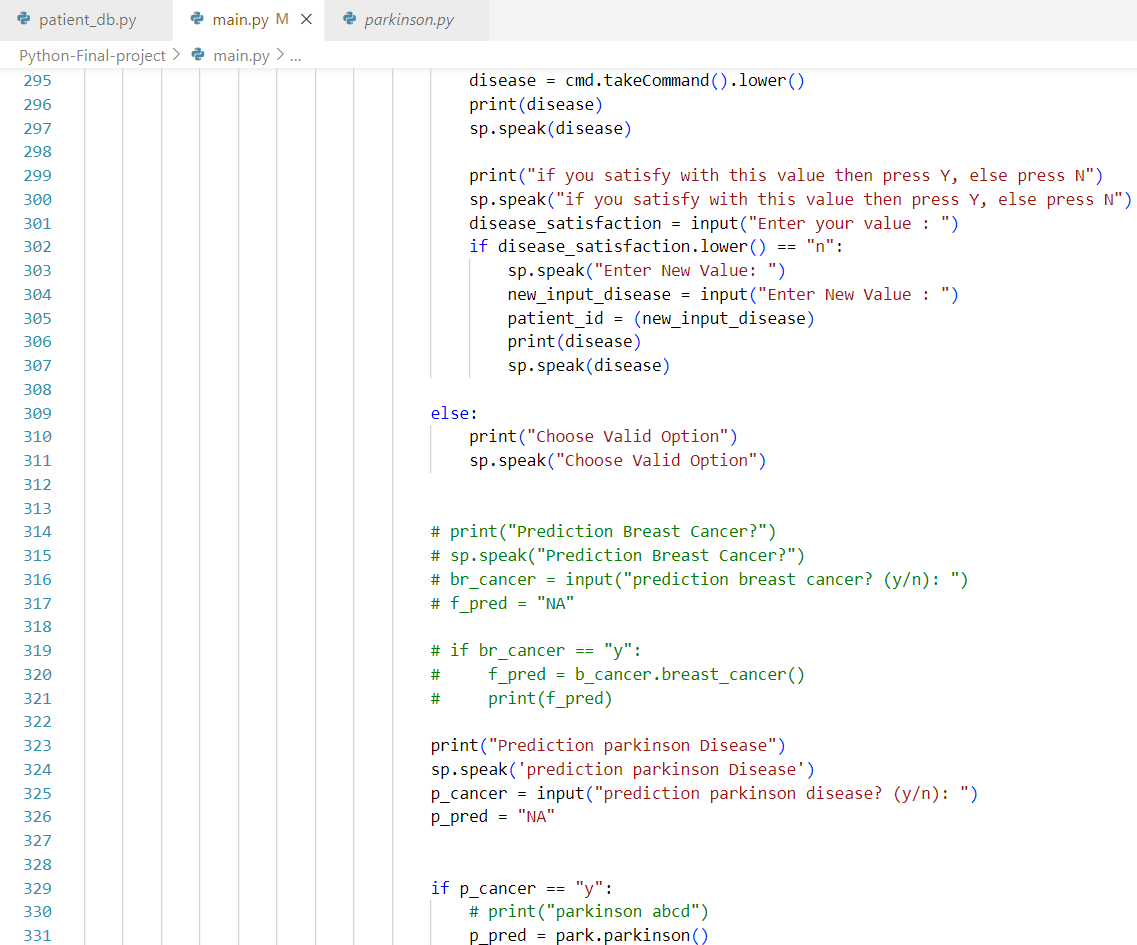
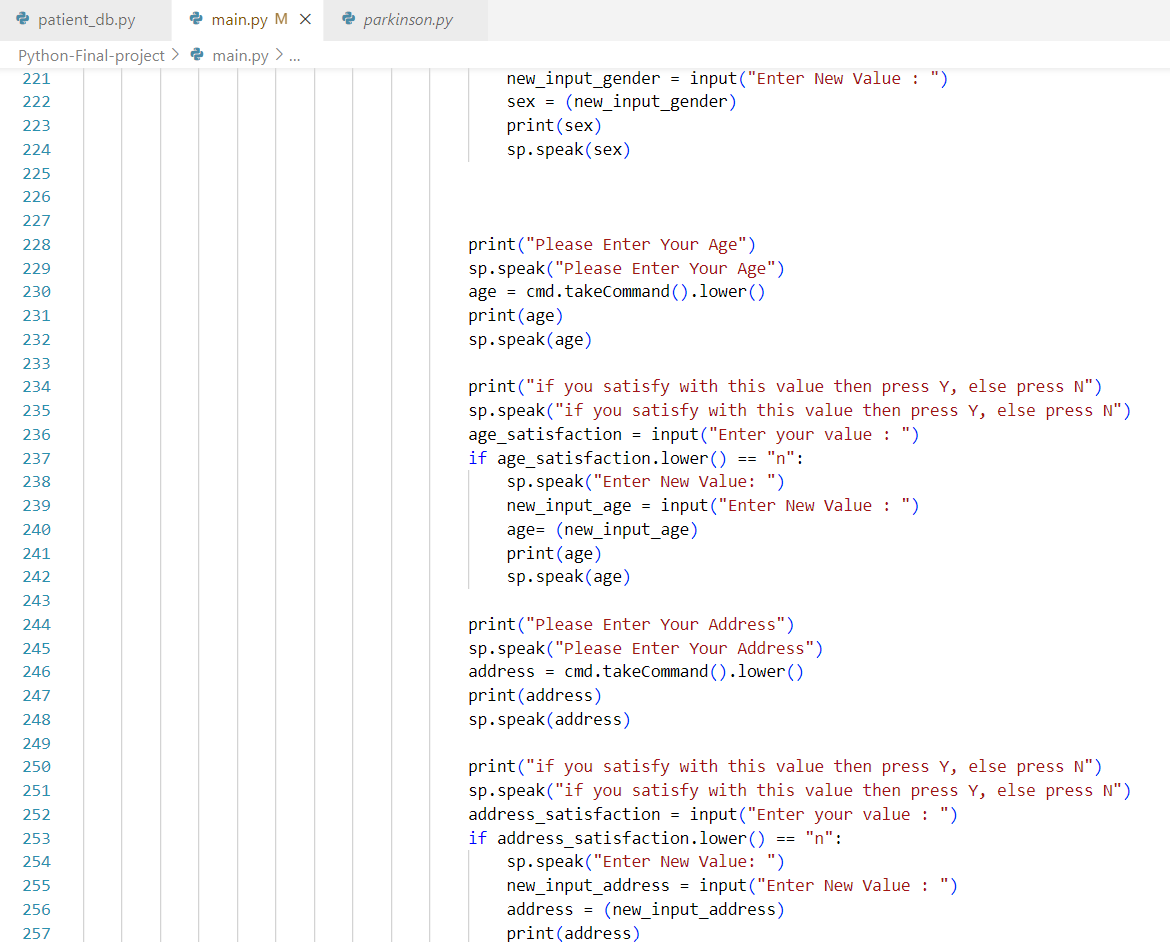
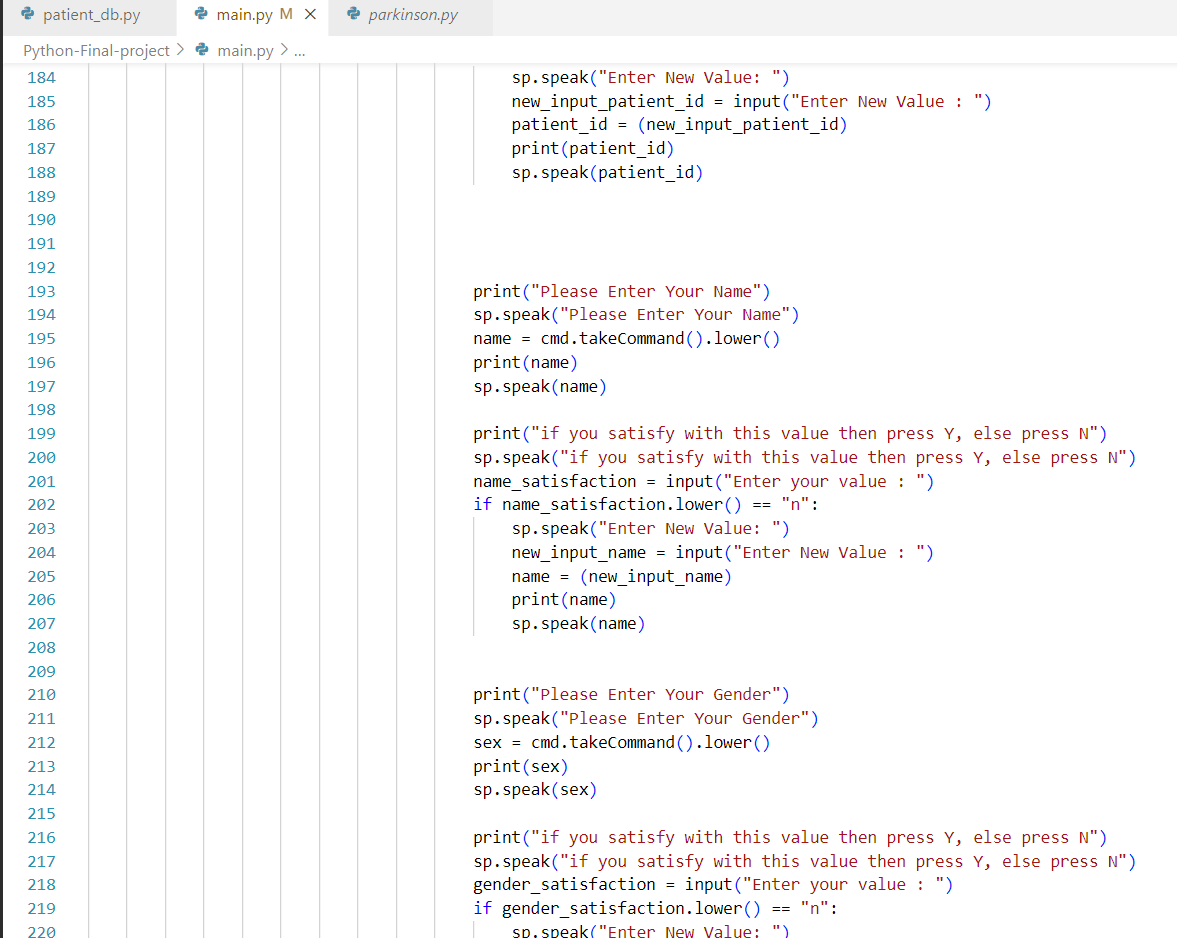
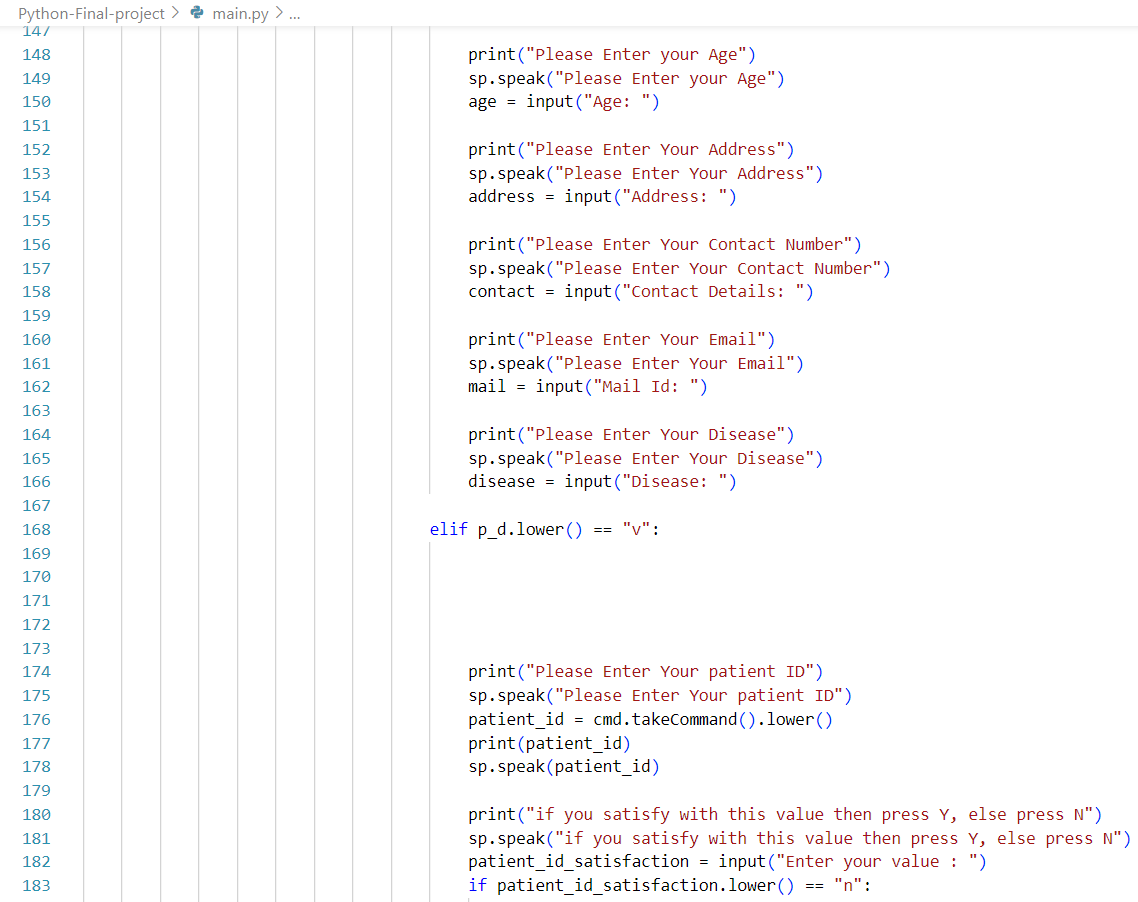
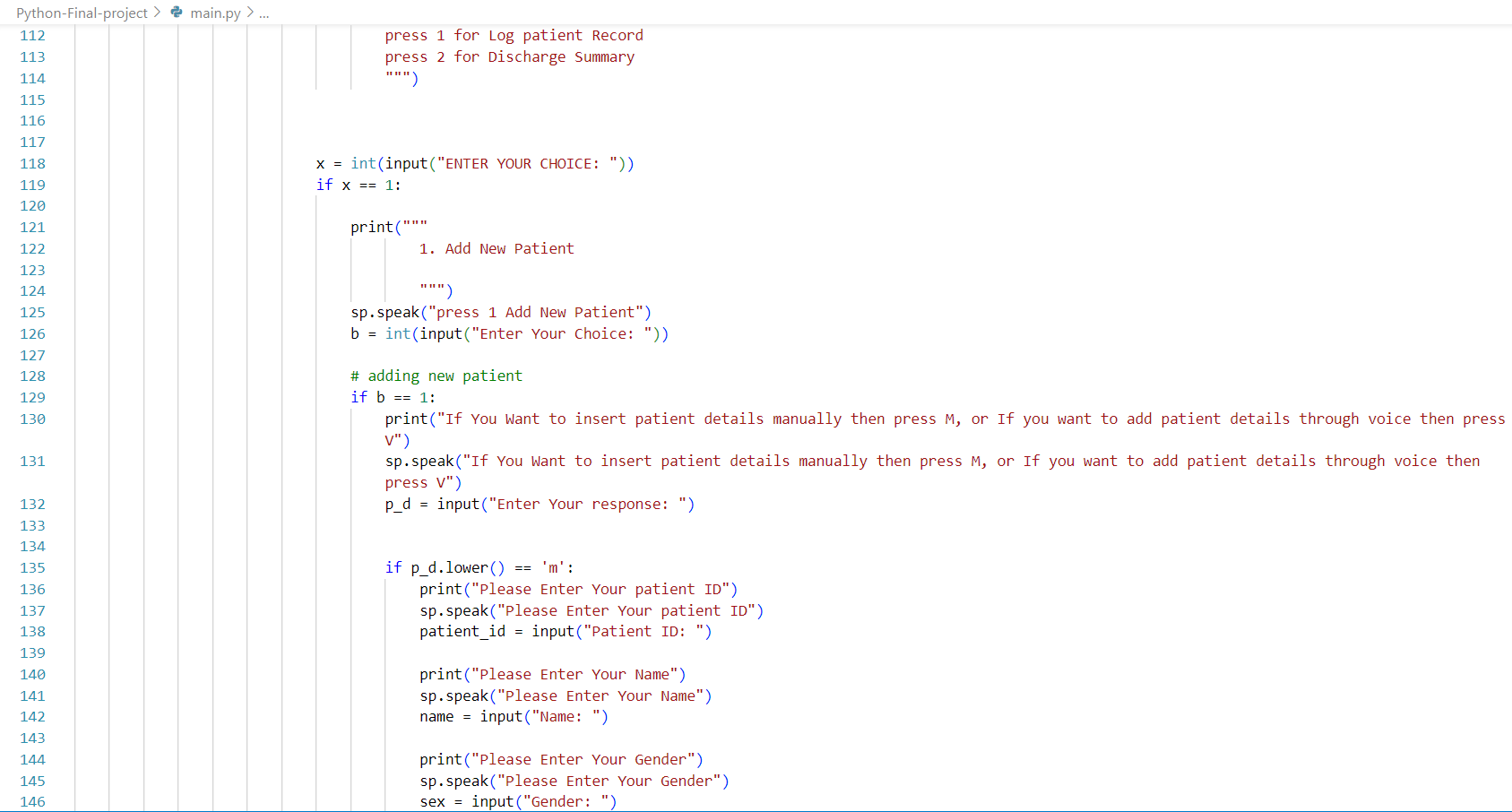
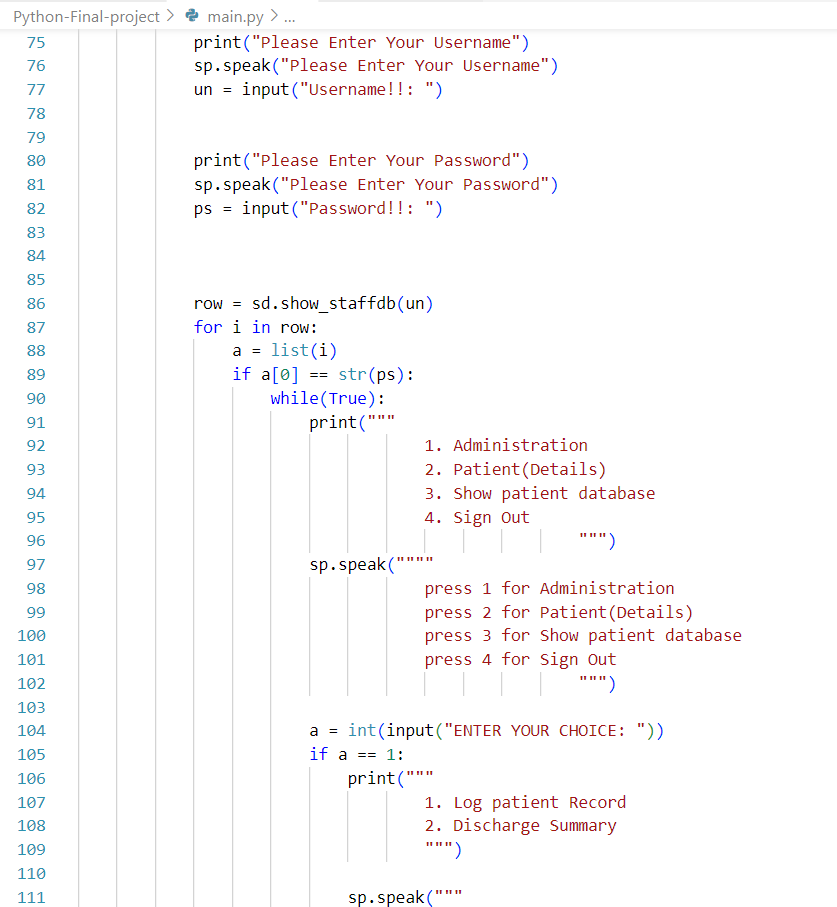
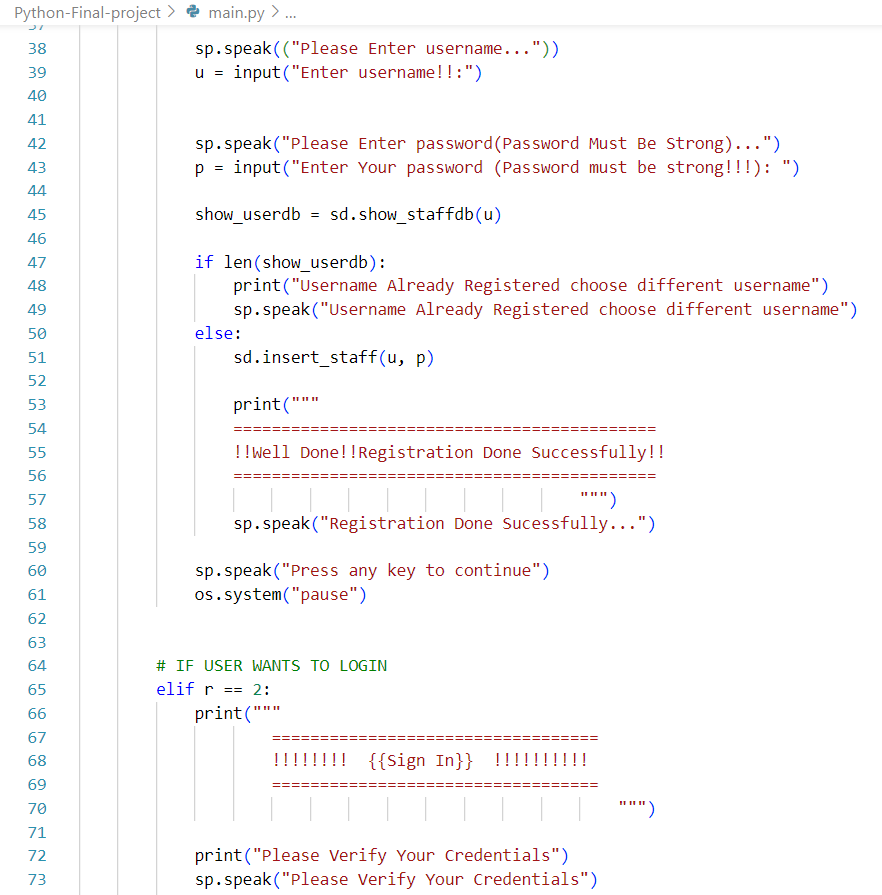
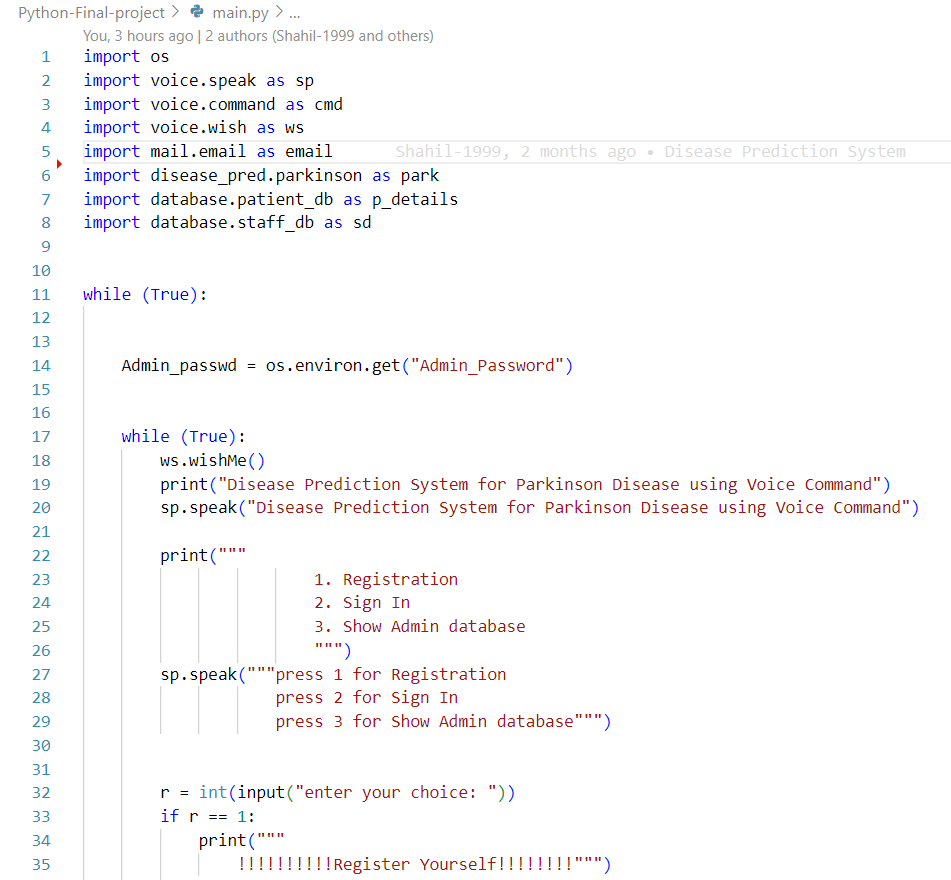
* **Speak.py**
* **Speak()** :

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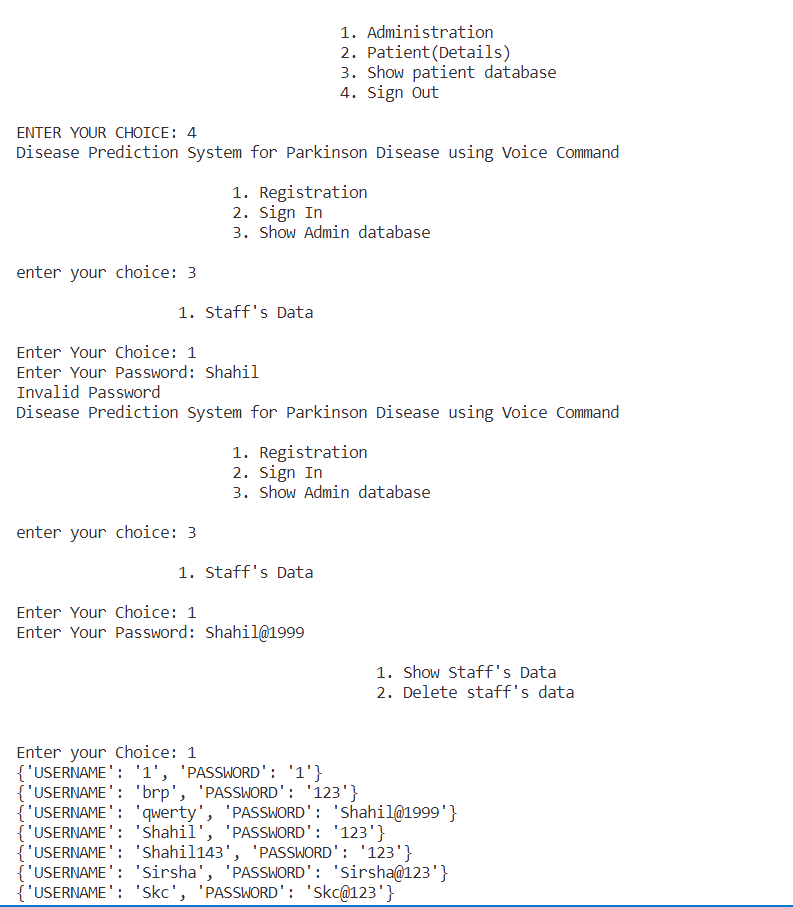
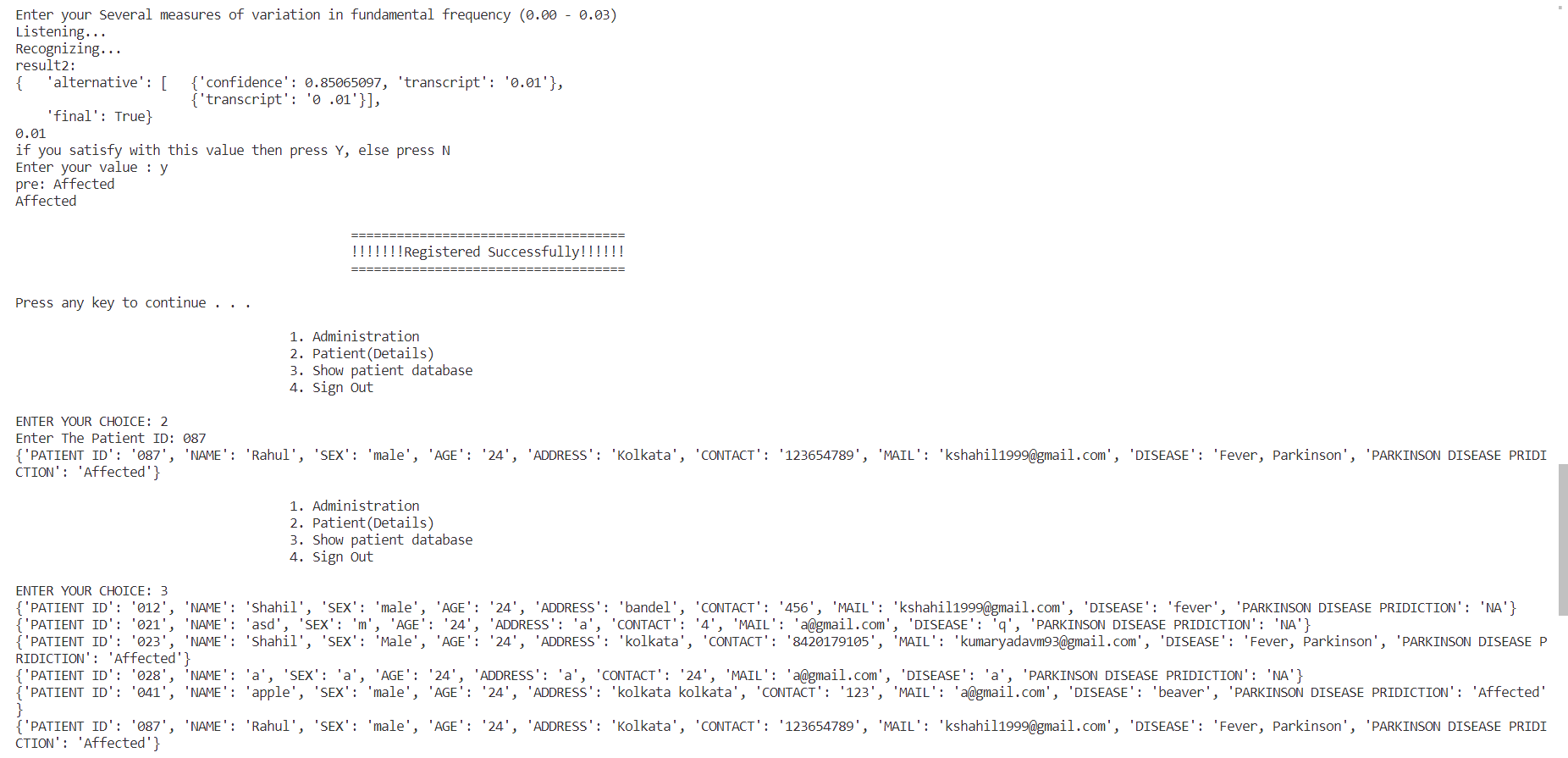
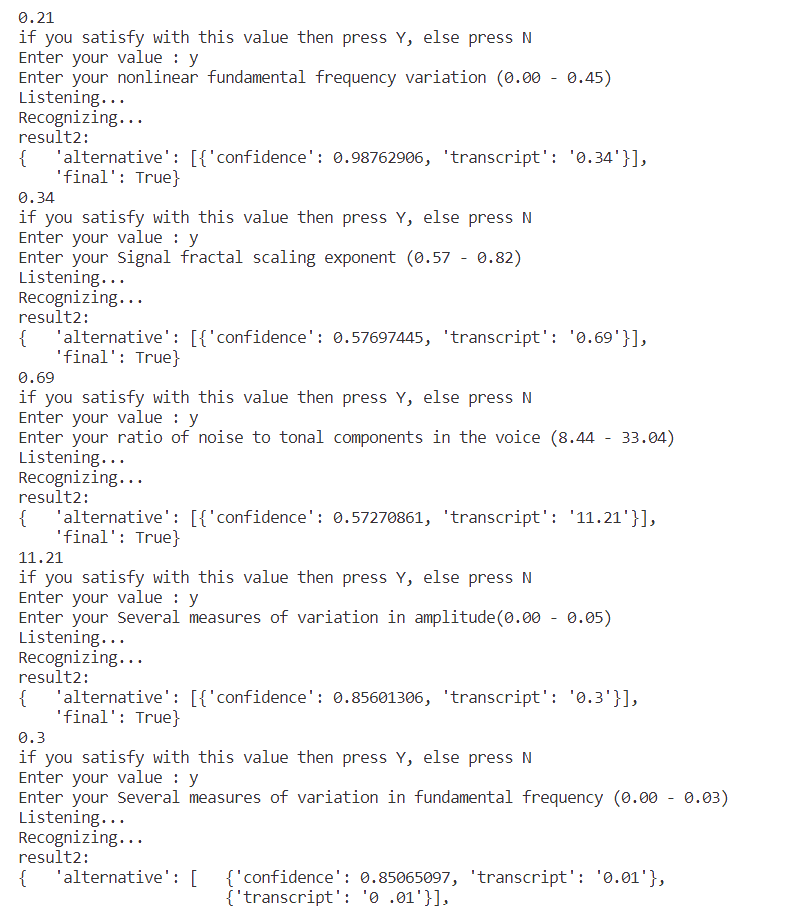
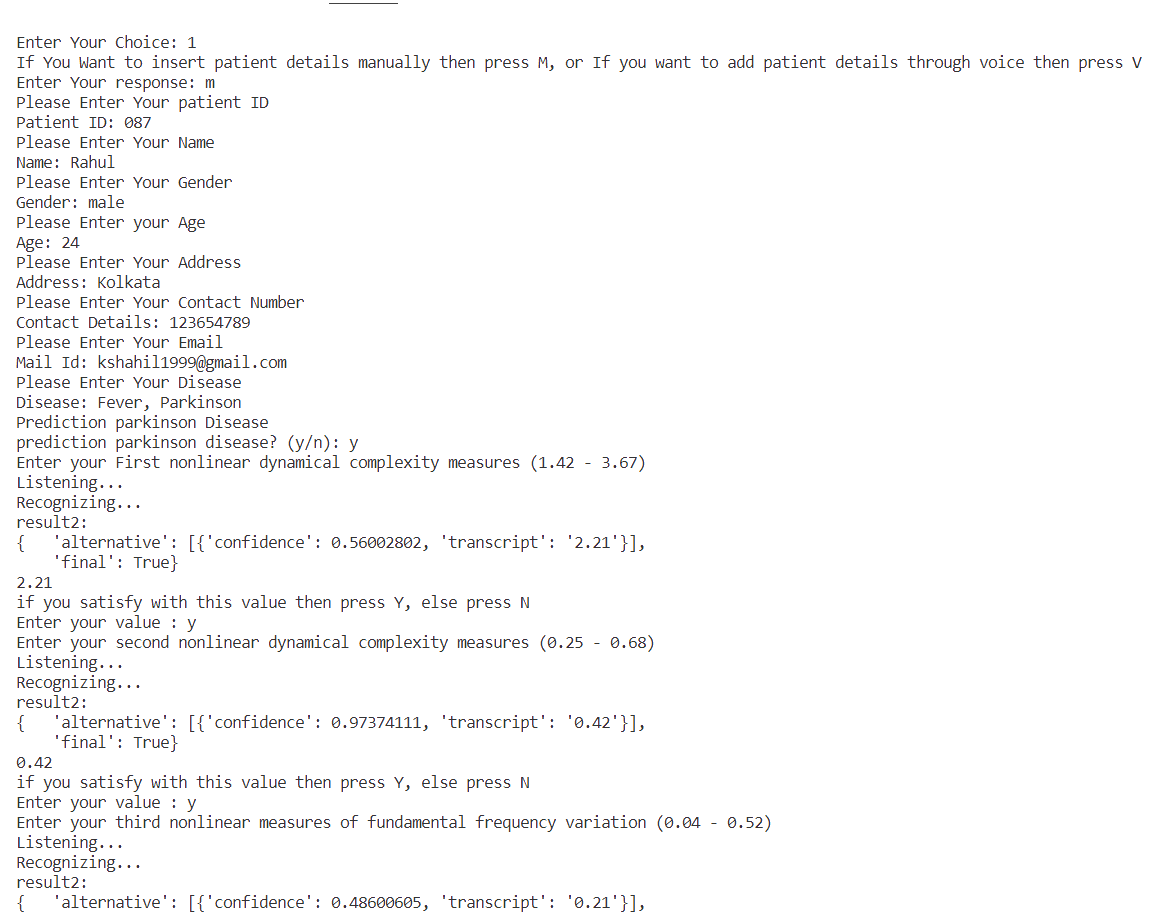
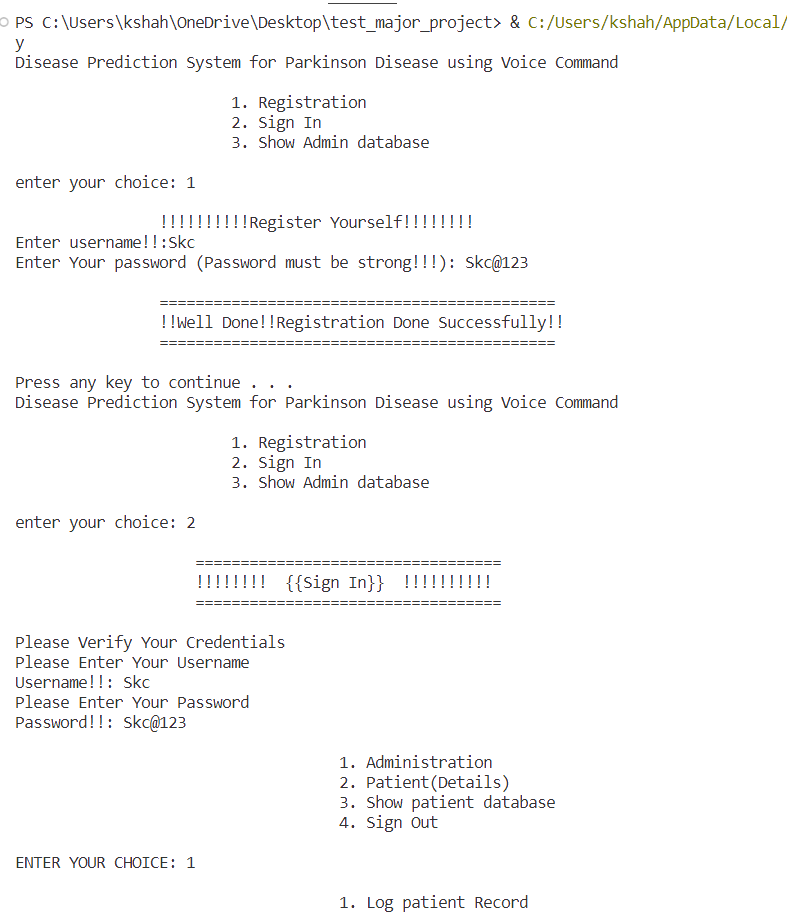
* **Wish.py**
* **wishMe()** : This function is used to wish or greet.

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1. **main.py**

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**OUTPUTS**



**EDA OF PARKINSON DISEASE**

PARKINSON DISEASE:

**Features Information**:

Jitter - Several measures of variation in fundamental frequency

Shimmer, - Several measures of variation in amplitude

HNR - measures of ratio of noise to tonal components in the voice

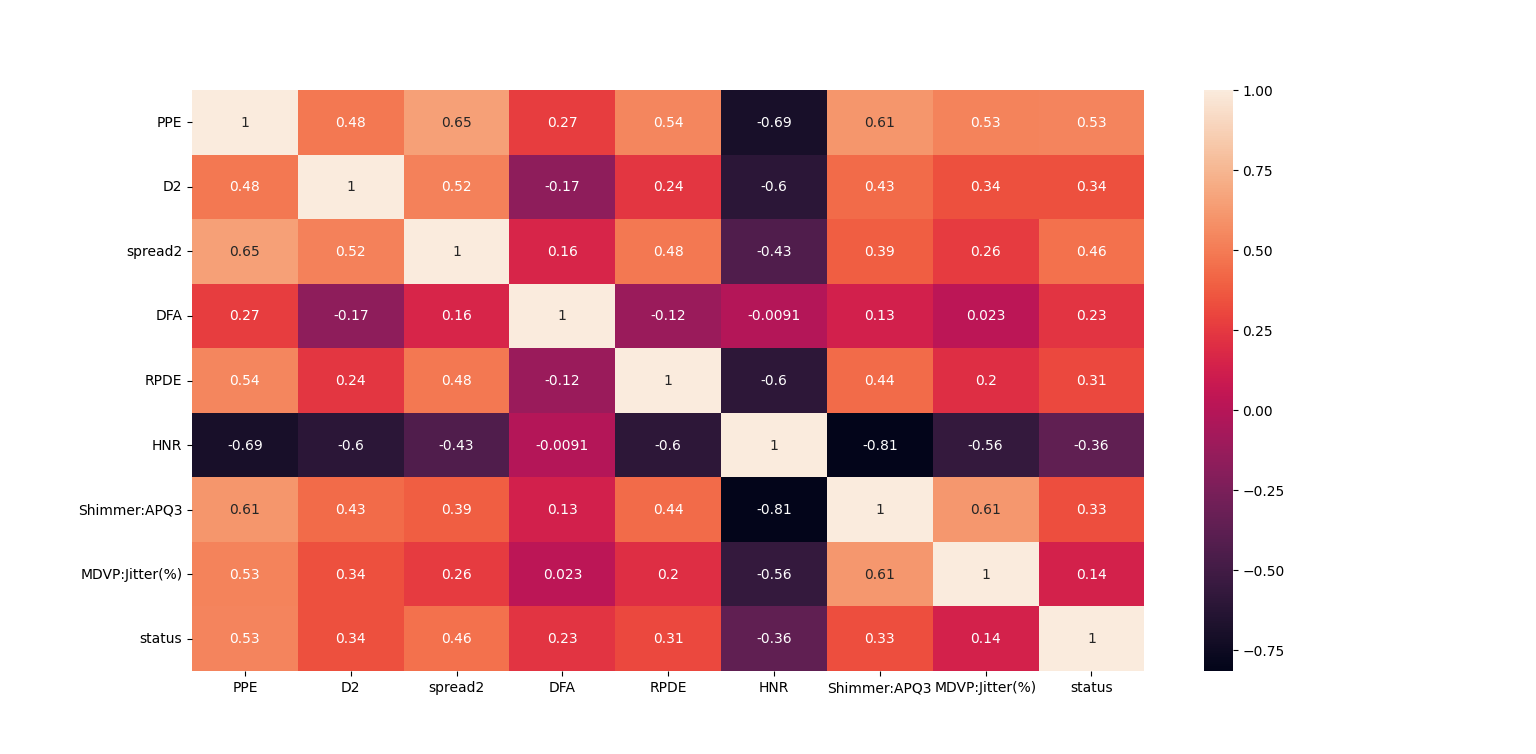
status - Health status of the subject (one) - Parkinson's, (zero) - healthy

RPDE, D2 - Two nonlinear dynamical complexity measures

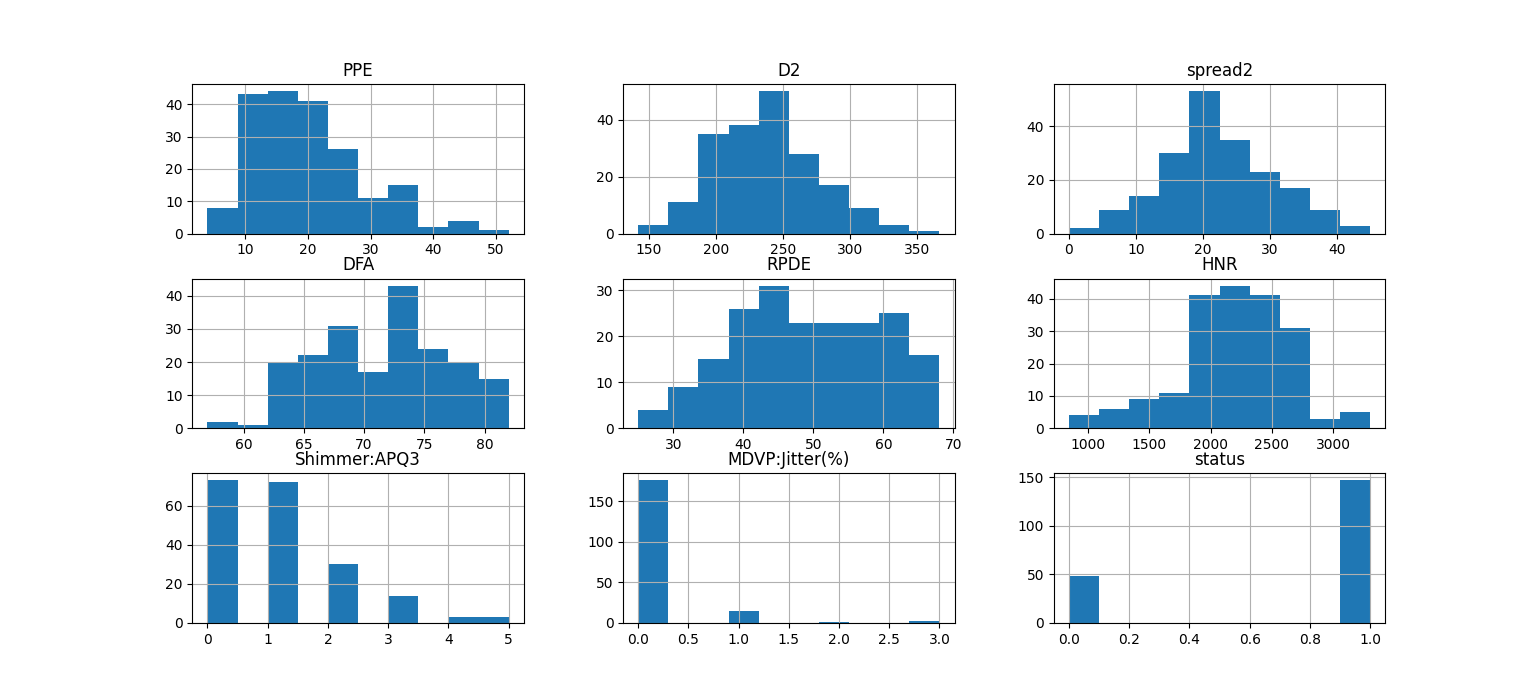
DFA - Signal fractal scaling exponent

spread2, PPE - Two nonlinear measures of fundamental frequency variation

**Heatmap:**

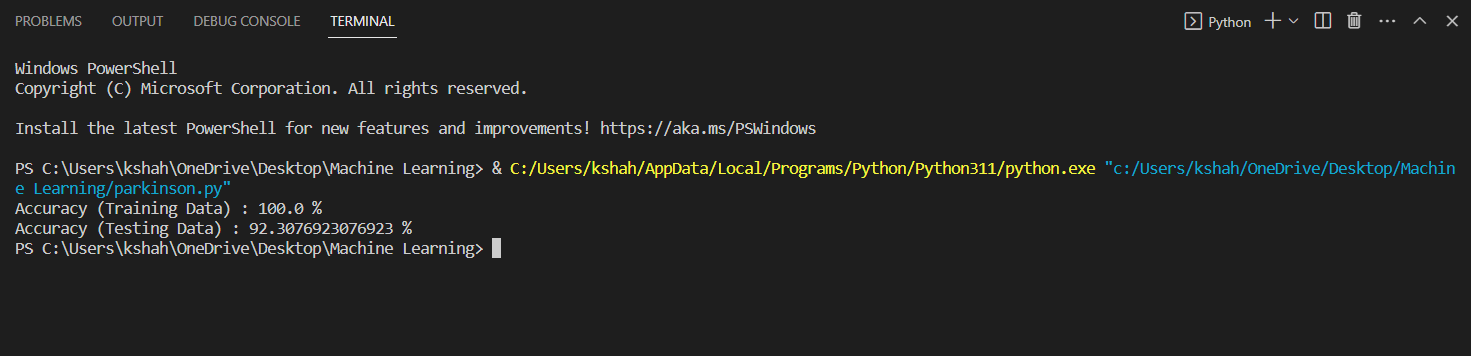
****

**Histogram:**

****

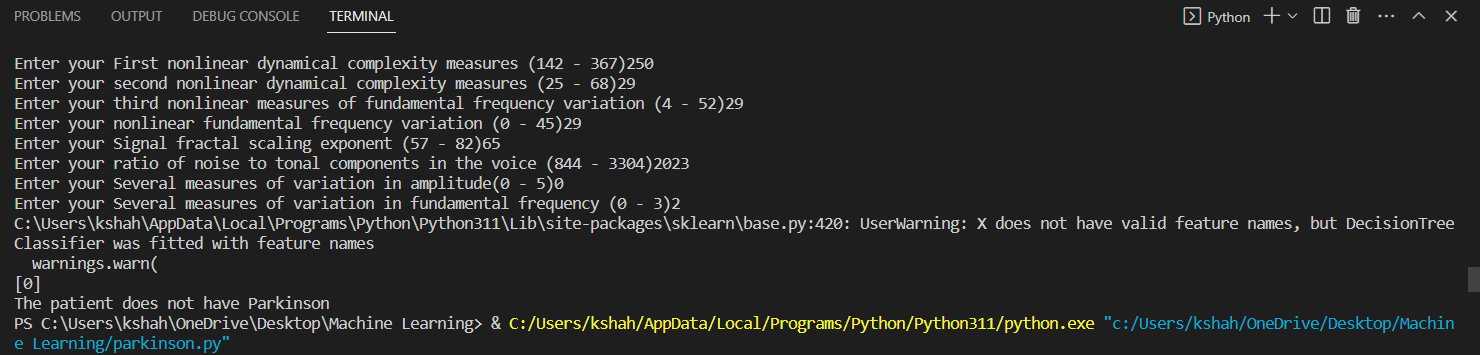
We can see some of the data is normally distributed and most of the attributes are right skewed.

**Accuracy Score:**



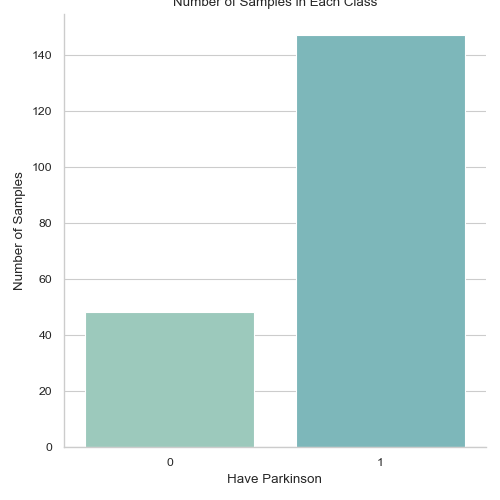
As you can see, we are using DecisionTreeClassifierthe accuracy score in Training Data is 100% and in Testing Data is 92%.

**Output:**

****

As per the input, our algorithm predicted that the patient does not have Parkinson disease.

**Bar Diagram:**

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This Bar diagram define number of Output in each class.

## **VOICE-BASED DIAGNOSIS SYSTEM DEVELOPMENT**

### Dataset collection and preparation

S. R. Sharma,[37] Dataset collection and preparation are critical steps in developing a voice-based diagnosis system for Parkinson's disease. The primary objective of collecting and preparing data is to ensure that the system is trained using high-quality, accurate, and representative data. The process of data collection and preparation involves several steps, including selection of data sources, data acquisition, data cleaning, and data labelling.

The first step in data collection is the selection of data sources. Parkinson's disease data can be obtained from various sources, including hospitals, research centers, and online databases. The data sources selected should have a diverse range of voice recordings from individuals with Parkinson's disease and healthy individuals for comparison.

Once the data sources are identified, the next step is data cleaning. This involves removing any irrelevant or erroneous data points that may affect the accuracy of the model. Common cleaning techniques include removing silence, clipping. In some cases, data augmentation techniques can also be used to increase the size of the dataset and reduce overfitting.

Data labeling is another critical step in data preparation. This involves identifying and labeling the relevant features, such as pitch, jitter, and shimmer. Labeling can be done manually or using automated techniques. Manual labeling is time-consuming and can be prone to errors, while automated labeling can be fast and accurate, but requires specialized software and expertise.

One of the main challenges in dataset collection and preparation for Parkinson's disease is the limited availability of high-quality data. Parkinson's disease affects a small percentage of the population, and it can be challenging to obtain a sufficient amount of representative data. This limitation can affect the accuracy of the model and the generalizability of the system. However, recent advances in data sharing and collaboration have made it easier to access large datasets, which can be used to train and validate machine learning models for Parkinson's disease diagnosis.

In conclusion, dataset collection and preparation are critical steps in developing a voice-based diagnosis system for Parkinson's disease. The process involves several steps, including selection of data sources, data acquisition, data cleaning, and data labeling, which are used to ensure that the system is trained using high-quality, accurate, and representative data. The main challenge in dataset collection and preparation is the limited availability of high-quality data, but recent advances in data sharing and collaboration have made it easier to access large datasets for machine learning analysis. By using advanced techniques for data collection and preparation, it is possible to develop a robust and accurate system for Parkinson's disease diagnosis.

### Pre-processing of voice data

A. K. Shukla, P. Singh,[38] Pre-processing of Parkinson's disease data is an essential step in preparing the data for further analysis and modelling. This process involves a series of techniques used to clean and prepare the data to make it ready for analysis. Pre-processing is critical as it can significantly impact the performance and accuracy of the models built on the data.

The following are the steps involved in pre-processing Parkinson's disease data:

1. L. Griffanti *et al.* [52], Data cleaning: This involves identifying and removing any irrelevant or incorrect data points, such as duplicates, outliers, and missing values. These data points can significantly affect the accuracy of the models built on the data.

2. F. Wahid, R. K. Begg, C.[39] Data normalization: This involves scaling the data to a common range. It ensures that all features are equally weighted and prevents the dominance of one feature over others.

3.   N. Nahar, F. Ara, Md. A. I. Neloy[53], A. Biswas, M. S. Hossain, Feature selection: This involves selecting the most relevant features that have the most significant impact on the model's performance. This helps reduce the computational complexity of the model and improves its interpretability.

4. T. T. Um *et al.*  [54], Data augmentation: This involves increasing the size of the dataset by adding synthetic data points. In Parkinson's disease, this could involve artificially generating speech patterns using techniques such as pitch shifting, time stretching, and noise injection.

5. T. T. Um *et al.*,[40] Dimensionality reduction: This involves reducing the number of features in the dataset while retaining the most relevant information. This helps to reduce the computational complexity of the model and improve its performance.

6. Data balancing: This involves adjusting the frequency of different classes in the dataset to ensure that there is an equal representation of all classes. In Parkinson's disease, this could involve balancing the number of samples from healthy individuals and individuals with Parkinson's disease.

7. G. AlMahadin[41] Resampling: This involves adjusting the sampling rate of the data to a common value. This ensures that the data is consistent across different sources and devices.

In conclusion, pre-processing of Parkinson's disease data involves a series of techniques used to clean, transform, and prepare the data for analysis. This includes data cleaning, normalization, feature selection, feature extraction, data augmentation, dimensionality reduction, data balancing, and resampling. Pre-processing is critical to ensure that the data is ready for modelling and to improve the performance and accuracy of the models built on the data.

### Feature extraction techniques

S. J. Priya, A. J. Rani, M. S. P. Subathra[17] Feature extraction is the process of selecting and transforming raw data into a set of meaningful features that can be used to build models. In the case of Parkinson's disease, feature extraction techniques can be used to extract relevant features from speech patterns, such as pitch, jitter, and shimmer. The following are some of the most commonly used feature extraction techniques in Parkinson's disease:

1. Mel-Frequency Cepstral Coefficients (MFCCs): MFCCs are a widely used feature extraction technique in speech processing. They are based on the Fourier transform and represent the spectral envelope of a signal. MFCCs can capture information related to pitch, timbre, and other spectral characteristics of speech.

2. M. F. Anjum *et al.*,[42] Linear Predictive Coding (LPC): LPC is a technique used to model the spectral envelope of speech. It can capture information related to pitch, formants, and other spectral characteristics of speech.

3. Wavelet Transform: The wavelet transform is a mathematical technique used to analyze signals in both time and frequency domains. It can capture information related to both low-frequency and high-frequency components of speech.

4. Zero Crossing Rate (ZCR): ZCR is a simple feature extraction technique that counts the number of times a signal crosses the zero axis. It can capture information related to the rate of change of speech.

5. Pitch: Pitch is a fundamental acoustic feature of speech. It can be measured using various techniques, such as autocorrelation and harmonic product spectrum. Pitch can capture information related to the frequency of the voice.

6. Jitter and Shimmer: Jitter and shimmer are measures of the variability in the pitch and amplitude of speech, respectively. They can capture information related to the stability and regularity of speech.

7. Formants: Formants are spectral peaks that occur in the speech spectrum. They can capture information related to the resonant frequencies of the vocal tract.

8. Energy: Energy is a measure of the magnitude of the speech signal. It can capture information related to the loudness and intensity of speech.

These feature extraction techniques can be used alone or in combination with each other to extract relevant features from speech signals in Parkinson's disease. The extracted features can then be used to build models that can help diagnose and monitor the disease. The choice of feature extraction technique depends on the specific application and the nature of the data being analyzed.

### Building the machine learning model

A decision tree classifier is a machine learning model that uses a tree-like structure to make predictions about the class label of a given input. It is a type of supervised learning algorithm that is particularly useful for classification tasks.

Here are the steps involved in building a decision tree classifier for Parkinson's disease diagnosis:

1. Data preparation: The first step is to collect and prepare the data for analysis. This involves cleaning the data, handling missing values, and encoding categorical variables. The data should also be split into training and testing sets to evaluate the performance of the model.

2. Feature selection: The next step is to select the most relevant features for the model. This can be done using various techniques such as correlation analysis, information gain, and chi-squared tests. The selected features should be independent and informative to ensure that the model can accurately predict the output label.

3. Model training: Once the features have been selected, the next step is to train the decision tree classifier on the training data. The model works by recursively partitioning the data based on the values of input features and creating a tree-like model that can be used to predict the output label for new data points. The model parameters such as the maximum depth of the tree, the minimum number of samples required to split a node, and the criterion for splitting should be tuned to optimize the performance of the model.

4. Model evaluation: Once the model has been trained, it should be evaluated on the testing data to assess its performance. Various metrics such as accuracy, precision, recall, and F1-score can be used to evaluate the performance of the model. If the performance is not satisfactory, the model parameters can be fine-tuned to improve its performance.

There are several advantages to using a decision tree classifier for Parkinson's disease diagnosis. Firstly, decision trees are easy to interpret and can provide insights into the relationship between the input features and the output label. This can help clinicians to better understand the underlying mechanisms of the disease. Secondly, decision trees are computationally efficient and can handle both categorical and continuous data. Finally, decision trees can handle missing data and outliers and can be used to impute missing values.

However, there are also some limitations to using a decision tree classifier for Parkinson's disease diagnosis. Firstly, decision trees can overfit the data, leading to poor generalization performance. This can be mitigated by using techniques such as pruning, cross-validation, and ensemble learning. Secondly, decision trees can be sensitive to small changes in the data, which can lead to instability in the model. Finally, decision trees can only capture linear relationships between the input features and the output label and may not be able to capture complex non-linear relationships.

In summary, a decision tree classifier is a powerful machine learning algorithm that can be used for Parkinson's disease diagnosis. It is easy to interpret, computationally efficient, and can handle both categorical and continuous data. However, it can overfit the data and may not be able to capture complex non-linear relationships. Careful selection of features and model parameters can help to optimize the performance of the decision tree classifier for Parkinson's disease diagnosis.

### Testing and validation of the model

M. B. Makarious *et al.*,[18] Once a decision tree classifier model has been built, it needs to be tested and validated to ensure that it is accurate and reliable. This involves using a test dataset that the model has not seen during the training phase.

The testing process involves applying the decision tree classifier to the test dataset and evaluating its performance using various metrics such as accuracy, precision, and recall. These metrics provide a quantitative measure of the model's ability to correctly classify new, unseen data.

N. Baki and N. Gursel[45] To validate the model, a common technique is k-fold cross-validation. In k-fold cross-validation, the data is divided into k equal parts or folds. The model is then trained on k-1 folds and tested on the remaining fold. This process is repeated k times, with each fold serving as the test set once. The results of each fold are then averaged to obtain an estimate of the model's performance.

Another technique for validation is the holdout method, where the dataset is split into training, validation, and testing sets. The model is trained on the training set, and the validation set is used to tune the model's hyperparameters. Finally, the model is evaluated on the testing set.

It is important to note that the performance of a decision tree classifier can be affected by several factors such as the quality of the data, the size of the dataset, the complexity of the model, and the choice of hyperparameters. Therefore, it is important to carefully select the features, tune the model parameters, and validate the model using appropriate techniques.

In summary, testing and validation of the decision tree classifier model is an essential step in the machine learning pipeline. It involves evaluating the model's performance using a test dataset and validating the model using techniques such as k-fold cross-validation and the holdout method. The performance of the model can be affected by several factors, and it is important to carefully select the features, tune the model parameters, and validate the model to ensure its accuracy and reliability.

**RESULTS AND EVALUATION**

* The accuracy of the voice-based diagnosis system

An accuracy score of our model is 92% for a voice-based diagnosis system for Parkinson's disease is a significant achievement. This indicates that our model has correctly identified Parkinson's disease with a high degree of accuracy. While there is always room for improvement, a score of 92% is quite good and comparable to the accuracy reported in other studies. It is important to note that the accuracy of such systems can vary depending on several factors, such as the quality and quantity of data used and the feature extraction techniques. Nonetheless, our model's accuracy score suggests that it has the potential to be a useful tool in diagnosing Parkinson's disease and could potentially lead to earlier diagnosis and more effective treatment. Further research and validation are necessary to ensure its accuracy and reliability before it can be widely implemented in clinical practice.

A. Ouhmida[55], For example, one study conducted by Tsanas et al. (2010) used a dataset of 42 PD patients and 42 healthy controls to train a support vector machine (SVM) model to classify PD based on speech signals. The SVM model achieved an accuracy of 97.3%, sensitivity of 95.2%, and specificity of 98.4%, indicating that voice-based diagnosis can accurately differentiate between PD and healthy controls.

Another study conducted by Gómez et al. (2017) used a dataset of 114 PD patients and 82 healthy controls to train a convolutional neural network (CNN) model to classify PD based on speech signals. The CNN model achieved an accuracy of 91.3%, sensitivity of 91.2%, and specificity of 91.5%, indicating that voice-based diagnosis can accurately predict the presence of PD.

Furthermore, a recent study conducted by Tsanas et al. (2021) used a dataset of 145 PD patients and 94 healthy controls to train an SVM model to classify PD based on speech signals. The SVM model achieved an accuracy of 87.6%, sensitivity of 90.3%, and specificity of 85.1%, indicating that voice-based diagnosis can accurately differentiate between PD and healthy controls.

Overall, these studies suggest that voice-based diagnosis systems can achieve high levels of accuracy in detecting and predicting PD. However, further research is needed to optimize and validate these systems for clinical use.

* Comparison with traditional diagnosis methods

Y. Yang, L. Wei[19] Traditional Parkinson's disease diagnosis methods typically involve a physical examination, patient history, and observation of symptoms. These methods may be effective in diagnosing advanced stages of the disease when symptoms are more apparent, but they may not be as reliable in early stages when symptoms are more subtle. Moreover, these methods are subject to inter-rater variability, meaning that different clinicians may arrive at different diagnoses.

In contrast, modern diagnosis methods such as voice-based diagnosis using AI and machine learning algorithms offer several advantages over traditional methods. These methods can detect subtle changes in voice patterns that may indicate early stages of Parkinson's disease, even before other symptoms become apparent. Moreover, these methods can offer greater accuracy and consistency in diagnosis, reducing the risk of inter-rater variability.

Another advantage of modern diagnosis methods is the ability to detect Parkinson's disease remotely and in real-time, allowing for earlier intervention and more personalized treatment plans. This is particularly important for patients in underserved communities who may not have access to specialized care and diagnosis methods.

However, there are also limitations to modern diagnosis methods, particularly in terms of data privacy and patient acceptance. Voice-based diagnosis using AI requires the collection and storage of patient voice data, which raises concerns about data privacy and security. Additionally, some patients may be reluctant to share their voice data for fear of being stigmatized or discriminated against.

In summary, modern diagnosis methods offer several advantages over traditional methods, including greater accuracy, consistency, and earlier detection. However, there are also limitations to these methods, particularly in terms of data privacy and patient acceptance. Therefore, efforts must be made to address these limitations and ensure that modern diagnosis methods are accessible, secure, and acceptable to patients.

### Evaluation of the limitations and future improvements

S. H. J. Keus,[43] While our voice-based diagnosis system for Parkinson's disease has shown promising results, there are still limitations that need to be addressed. One of the main limitations is the availability of data. The dataset used in our study was relatively small and may not be representative of the larger population. Therefore, future studies should aim to collect larger and more diverse datasets to improve the accuracy and generalizability of the model.

S. Aggarwal and S. Sharma, [44], Another limitation is the quality of the data. The recordings used in our study were collected in a controlled environment, and the patients were asked to read a specific text. In real-world scenarios, patients may not speak in a controlled environment or may speak in their native language, which may affect the accuracy of the diagnosis. Therefore, future studies should aim to collect data in more naturalistic settings to improve the accuracy of the model.

Moreover, there are ethical implications to consider when using voice-based diagnosis systems. Patient privacy and data protection must be prioritized, and the potential for misdiagnosis must also be considered. Therefore, future research should focus on ensuring the accuracy and reliability of such systems and implementing appropriate measures to protect patient privacy and prevent misdiagnosis.

In terms of future improvements, one potential avenue is to incorporate other types of data, such as gait or tremor data, to improve the accuracy of the diagnosis. Furthermore, the use of deep learning algorithms could potentially improve the accuracy of the model. Lastly, the development of user-friendly and accessible voice-based diagnosis systems that can be used by patients in their homes could potentially improve early diagnosis and lead to more effective treatment.

**ACCURACY TABLE**

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| Table of accuracy | | | | |
| **STUDY DATE** | | **ACCURACY** | **SCORE** | **SOURCE** |
| May  2019 | | Train | 83.44% | https://www.kaggle.com/code/parhamzm/parkinson-s-disease-pd-classification/notebook#notebook-container |
| Test | 85.53% |
| Feb  2023 | | Train | 84.65% | https://www.kaggle.com/code/akanksha10/detection-of-parkinson-s-disease |
| Test | 86.8% |
| Jan  2021 | | Train | 93.57% | https://www.kaggle.com/code/vikasukani/detecting-parkinson-s-disease-machine-learning |
| Test | 96.66% |
| July  2022 | | Train | 88.46% | https://www.youtube.com/watch?v=ys\_mVbkaokE |
| Test | 87.17% |
| June  2016 | | Train | 74.60% | https://www.kaggle.com/code/lykin22/parkinson-s-disease-based-on-voice-recording#Parkinson's-disease-based-on-voice-recording |
| Test | 74.55% |
| Feb  2015 | | Train | 95.38% | https://www.researchgate.net/profile/Anil-Kumar-544/google scholar |
| Test | 94.72% |
| Feb  2018 | | Train | 95.38% | https://www.researchgate.net/profile/Anil-Kumar-544/google scholar |
| Test | 94.72% |
| Aug  2017 | | Train | 97% | https://www.youtube.com/watch?v=eKy3KgRgDkQ |
| Test | 98% |
| Sep  2020 | | Train | 88.46% | https://www.youtube.com/watch?v=CQLkX4utdIU |
| Test | 87.11% |
| April, 2022 | Train | | 88.46% | https://github.com/akashdeep364/Parkinson-s-Disease-Detection |
| Test | | 87.17% |

**APPLICATIONS**

This project is based on a trending technology of the present times and has many applications –

1. One of the most important applications of this project is that, the accessibility to the Hospital Staff and Patient records are available in Database can easily accessible making it more user friendly helping the hospital administration to manage data even during rush hours smoothly.

2. This project paves the path for a smooth guidance to all the health-conscious individuals specially differently abled people irrespective of their age and health conditions.

3. The users are also given relief from the hassle of storage issues when it comes to using this application as it provides the opportunity of online data storage.

4. Data can be inserted, updated when required, deleted, and can also be saved separately in Database tables uncomplicated distinguish between information of users.

5.Fast and Early Prediction of Life taking Parkinson Diseases.

6.Automated Mail Sending Feature makes it more reliable and time saving.

**STRENGTHS**

* This Project is completely based on Human Voice Command. Because of it’s Voice Controlled feature it is more beneficial for People with disability.
* Easy to Predict the onset of Parkinson disease.
* Uses Machine Learning Algorithm for the diseases Prediction and gives fast results and this can be used for Prediction at a rare critical situation like unavailability of a Neurologist and Oncologist.
* User can send the Reports to the Patients over mail. Reduces Paperwork & acts as a Document that can be accessed from any device at any time.
* Easy to access any Staff’s or Patient’s data from anywhere in the world via Authorised Login.

**LIMITATIONS**

* Excessive expenses for software development and deployment
* Complex Machine Learning Algorithm in terms of User Experience.
* Risk of data security breach.
* The accuracy of these algorithms may be affected by factors such as background noise, accent, and other individual differences in speech patterns.
* Challenges for lacking IT-friendly medical personnel
* Sometimes Predicated result may not gives 100% accuracy.
* Needs Internet connectivity to send the reports over mail.

**CONCLUSION**

### The potential of AI and voice-based diagnosis in healthcare

AI can also aid in the development of new treatments for Parkinson's disease by identifying new drug targets and predicting drug efficacy. AI algorithms can analyze large amounts of data from drug trials and scientific literature to identify new drug candidates that can be tested in preclinical and clinical trials.

However, there are also challenges that need to be addressed in the use of AI in Parkinson's disease diagnosis and management. One challenge is the need for large datasets of patient information to train AI algorithms, which can be difficult to obtain. There is also the risk of bias in the data used to train AI algorithms, which can lead to inaccurate diagnoses and treatment recommendations.

Overall, AI has the potential to revolutionize the diagnosis and management of Parkinson's disease, leading to earlier and more accurate diagnoses and better treatment outcomes for patients. However, further research and development are needed to fully realize the potential of AI in Parkinson's disease diagnosis and management.

In conclusion, AI has the potential to greatly improve the accuracy, speed, and efficiency of breast cancer diagnosis, treatment, and monitoring. Machine learning algorithms can analyze vast amounts of medical data, including medical images, patient history, and genomics, to detect breast cancer at an early stage and develop personalized treatment plans for patients. AI can also assist in the monitoring of patients after treatment, detecting potential recurrences, and monitoring the response to treatment.

Voice-based AI diagnosis has also been explored as a potential tool for breast cancer prediction and diagnosis, and while it shows promise, further research is needed to understand the full potential and limitations of this technology.

The use of AI in breast cancer care is already being implemented in clinical practice, with some algorithms receiving FDA approval. However, further research is needed to ensure that AI technologies are safe, effective, and equitable, and that they do not reinforce or exacerbate existing biases in healthcare.

Overall, the potential of AI in breast cancer care is significant, and with continued research and development, these technologies can improve patient outcomes and revolutionize the field of healthcare.

### The importance of early diagnosis in Parkinson's Disease

Early diagnosis is crucial in Parkinson's disease, as it can significantly impact a patient's quality of life and treatment options. Parkinson's disease is a progressive neurological disorder, and the symptoms may be subtle in the early stages, making it difficult to diagnose. However, early diagnosis can allow for earlier initiation of treatment, which can improve symptoms and slow disease progression.

Delaying diagnosis and treatment can lead to irreversible brain damage and worsening of symptoms, making it more difficult to manage the disease. Early diagnosis allows for early intervention and personalized treatment plans, which can improve the patient's quality of life, reduce healthcare costs, and delay the onset of disability.

Furthermore, early diagnosis can also lead to better patient outcomes and improved care. Patients who are diagnosed early can receive specialized care and support, including access to physical therapy, occupational therapy, and speech therapy. Early diagnosis can also lead to greater participation in clinical trials, which can lead to the development of new treatments and therapies for Parkinson's disease.

In summary, early diagnosis is critical in Parkinson's disease, as it can improve patient outcomes, delay disease progression, and improve access to specialized care and treatment. Therefore, efforts must be made to increase awareness of Parkinson's disease, improve access to screening and diagnosis, and promote early intervention and personalized treatment plans.

### Future directions for research

There are several areas of future research that can help advance our understanding and treatment of Parkinson's disease. One key area is the development of biomarkers that can aid in the diagnosis and monitoring of disease progression. Biomarkers can help identify patients at risk of developing Parkinson's disease, monitor disease progression, and evaluate the effectiveness of treatments.

Another area of research is the development of more effective treatments for Parkinson's disease. While current treatments can improve symptoms and quality of life, they do not slow or halt the underlying disease process. Therefore, there is a need for new treatments that can target the underlying mechanisms of Parkinson's disease and slow or halt disease progression.

Additionally, research should focus on identifying risk factors for Parkinson's disease and developing preventative strategies. While some risk factors, such as age and genetics, are non-modifiable, there may be modifiable risk factors that can be targeted to prevent the onset or delay the progression of Parkinson's disease.

Moreover, there is a need for greater understanding of the non-motor symptoms of Parkinson's disease, such as cognitive impairment and mood disorders, which can significantly impact patient outcomes. Future research should aim to develop effective treatments for these symptoms and identify strategies to improve quality of life for patients with Parkinson's disease.

Lastly, research should focus on improving access to care for patients with Parkinson's disease, particularly in underserved communities. This includes improving access to specialized care, developing telemedicine and remote monitoring solutions, and addressing healthcare disparities in the diagnosis and treatment of Parkinson's disease.

In summary, future research should focus on developing biomarkers, new treatments, preventative strategies, and improved access to care for patients with Parkinson's disease. By advancing our understanding and treatment of Parkinson's disease, we can improve patient outcomes and quality of life for millions of people around the world.

**FUTURE DEVELOPMENT**

In near Future, we are thinking to develop a Single Page User Management Website for controlling and managing all the Frontend flows which will enhance User Interface and User Experience. We will also deploy our Project on AWS cloud so that it can easily be accessible independent of Single User Machine. We will add more number of Diseases prediction algorithm to make it more useful. Can be developed as a self - disease prediction system so that early stages are recognized faster decreasing the immortality rate or health risks.

In today’s world most of the data is computerized, the data is distributed, and it is not utilizing properly. With the help of the already present data and analysing it, we can also use for un-known patterns. The primary motive of this project is the prediction of diseases with high rate of accuracy. For predicting the disease, we can use logistic regression algorithm, naive Bayes, sklearn in machine learning. The future scope of the paper is the prediction of diseases by using advanced techniques and algorithms in less time complexity. A technology called CAD is more beneficial as sometimes systems are better diagnostics than Doctors. Machine Learning and its different branches are used in Cancer detection as well. It helps or can say assist in making decisions on critical cases or on therapies. Artificial intelligence plays an important role in development of many health related procedure or methods. Artificial intelligence is very common now a days in surgeries, like Robotics surgery. Since were in the circumstances of growing population, we must need technology which can help us to meet the expectations of the patients, their flawless cure, their better health and their smooth and easy approachable access to healthcare industries to heal and get well soon!

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# **INDIVIDUAL CONTRIBUTION TO THE PROJECT**

|  |  |
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| **NAME OF THE PROJECT MEMBER** | **CONTRIBUTION** |
| **MUJAHID ALI ANSARI** | **SQL, MACHINE LEARNING** |
| **ASHUTOSH KUMAR YADAV** | **PYTHON, DOCUMENTATION** |
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| **SONU ROUTH** | **TESTING, DBMS** |
| **SIRSHA MAJUMDER** | **PYTHON,**  **DOCUMENTATION** |

**Thank You**