



**NEW HORIZON
COLLEGE OF ENGINEERING**

Autonomous College, Affiliated to VTU | Approved by AICTE New Delhi & UGC
Accredited by NAAC with 'A' Grade & Accredited by NBA

DEPARTMENT OF INFORMATION SCIENCE AND ENGINEERING

A MINI PROJECT REPORT ON

“PREDICTION AND CLASSIFICATION OF THYROID DETECTION”

Submitted in the partial fulfillment of the requirements in the 4th semester of

**BACHELOR OF ENGINEERING
IN**

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FOR

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22ISE48 :COURSE CODE**

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CERTIFICATE

Certified that the project work entitled “PREDICTION AND CLASSIFICATION OF THYROID DETECTION” carried out by G.Shasi vardhan Reddy and K.Akash bearing usn 1NH22IS048 and 1NH22IS063 respectively, bonafide students of IV semester in partial fulfillment for the award of Bachelor of Engineering in Information Science & Engineering of New Horizon College of Engineering, an autonomous institute affiliated to the Visvesvaraya Technological University, Belagavi during the year 2023-24. It is certified that all corrections / suggestions indicated for Internal Assessment have been incorporated. The project report has been approved as it satisfies the academic requirements in respect of Mini Project work prescribed for the said Degree.

Name & Signature of Guide

Mrs.A. Shalini

Name & Signature of HOD

.Dr Vandana C.P

Examiners :

Name

Signature

1.

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2.

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ACKNOWLEDGEMENT

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-1NH22IS048

K.Akash

-1NH22IS063

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Abstract

Thyroid gland diseases are among the most prevalent endocrine disorders in the world, affecting people of all age groups from children to the elderly. It is the general concept for a medical problem that prevents one's thyroid from producing enough hormones. Early diagnosis and care of such diseases are thus vital in preventing disease advancement and even death in patients. Machine Learning (ML) techniques play a major role in such cases for making correct decisions, proper disease diagnosis and saving cost and time of patients.

The purpose of this study is prediction of thyroid disease using classification Predictive Modelling. Data Collection, Data Preprocessing and Several classification models including K- Nearest Neighbours (k-NN), Naive Bayes, Decision Tree models are fit to the data to perform predictions.

We believe that this work helps in providing a better classification model for prediction and diagnosis of thyroid gland diseases. The result of this proposed work can be used for the forthcoming identification to keep track on important factors affecting thyroid gland related diseases. The various algorithm results are based on speed, accuracy and performance of the model and cost for the treatment. Also, the classification of effective data are helps to find the treatment to the thyroid patients with better cost and facilitates the management.

CHAPTER 1

INTRODUCTION

Introduction

Thyroid illnesses and disorders are widespread hormonal problems that affect most of the world's population. There are two general types of thyroid disease, each with their own symptoms. one is hyperthyroid and the other is hypothyroid⁵. So, for example , Random Forests, KNN, Support Vector Machine (SVM), and Decision Trees are all machine learning methods widely used for predicting thyroid diseases. Machine Learning (ML) is an artificial intelligence field based on the idea that computers can learn from data, spot patterns, and make judgments with little or no human intervention.

Problem Definition

A lot of people find it difficult to maintain new habits or break old ones. Typical difficulties consist of:

1. According to statistics, thyroid disorders are on the rise in India. Approximately 1 in 10 Indian adults suffer from thyroid problem.
2. It has been estimated that around 42 million peoples suffer from thyroid disease.
3. Predicting thyroid disorder by doctor is a tedious process which might lead to negative prediction, only experienced doctor can examine the case properly.

Motivations

The following reasons drove the development of an habit monitoring website:

1. **Prevalence of Thyroid Disorders:** Thyroid disorders, including hypothyroidism and hyperthyroidism, are common globally, affecting millions of people. Early detection can significantly improve patient outcomes.

2. **Impact on Quality of Life:** Thyroid conditions can severely impact a person's quality of life, leading to symptoms such as fatigue, weight changes, depression, and cognitive impairments. Accurate prediction and detection can help mitigate these effects.
3. **Prevention of Complications:** Untreated thyroid disorders can lead to serious health complications, including cardiovascular diseases, infertility, and osteoporosis. Early detection allows for timely treatment to prevent these issues.
4. **Cost-Effective Healthcare:** Early and accurate detection of thyroid disorders can reduce healthcare costs by minimizing the need for extensive diagnostic procedures and preventing costly complications.
5. **improvement in Treatment Outcomes:** Accurate prediction models can help tailor treatment plans to individual patients, leading to better management of the disease and improved treatment outcomes.
6. **Public Health Significance:** Thyroid disorders have significant public health implications. A reliable prediction and detection system can support public health initiatives by identifying high-risk populations and enabling targeted interventions.

CHAPTER 2

LITERATURE SURVEY

1. Existing System

Thyroid disorders are prevalent, and early detection and accurate diagnosis are crucial for effective management. There are several existing systems and methods for the prediction, detection, and diagnosis of thyroid conditions. These systems can be categorized into diagnostic techniques, predictive analytics, and monitoring systems. Here's a comprehensive overview of the main systems and methods currently in use:

- **Blood Tests**

Thyroid-Stimulating Hormone (TSH) Test: Measures the amount of TSH in the blood, which indicates how well the thyroid gland is functioning.

Free T4 (Thyroxine) Test: Measures the amount of free thyroxine in the blood, which helps in assessing thyroid function.

Free T3 (Triiodothyronine) Test: Measures the amount of free triiodothyronine in the blood.

Thyroid Antibody Tests: Includes TPOAb (Thyroid Peroxidase Antibodies) and TgAb (Thyroglobulin Antibodies) tests for autoimmune thyroid conditions.

- **Imaging Techniques**

Ultrasound: Used to detect abnormalities in the thyroid gland, such as nodules or cysts.

Radioactive Iodine Uptake Test: Measures how much radioactive iodine the thyroid gland absorbs, which can indicate thyroid function.

Thyroid Scintigraphy (Scan): Provides images of the thyroid gland to identify nodules and assess their activity.

- **Machine Learning Models**

Risk Prediction Models: Algorithms such as logistic regression, random forests, and support vector machines are used to predict the likelihood of thyroid disorders based on patient data.

Deep Learning Algorithms: Convolutional Neural Networks (CNNs) and other deep learning models are used to analyze medical images and improve the detection of thyroid abnormalities.

2. Proposed System

Developing a new system for thyroid prediction and detection involves creating a comprehensive solution that integrates advanced technologies and methodologies to improve accuracy, accessibility, and efficiency in identifying thyroid disorders. Below is a detailed outline of a proposed system for thyroid prediction and detection, including its components, features, technologies, and implementation strategies.

A. Data Collection

1.1 Clinical Data Input

- **Data Sources:** Electronic Health Records (EHR), patient self-reported symptoms, and clinical examination data.
- **Features:** Interface for inputting lab results (TSH, Free T4, Free T3, Thyroid Antibodies), medical history, and physical examination findings.

1.2 Imaging Data Integration

- **Data Sources:** Ultrasound, Thyroid Scintigraphy, and other imaging techniques.
- **Features:** Integration with PACS (Picture Archiving and Communication System) for retrieving and displaying thyroid imaging results.

1.3 Home Testing Data Collection

- **Data Sources:** At-home thyroid test kits.
- **Features:** Interface for uploading and analyzing home test results, with secure transmission of data to the system.

1.4 Symptom Tracking and Wearable Devices

- **Data Sources:** Wearable health monitors and mobile apps.
- **Features:** Integration with wearable devices that track symptoms like fatigue, weight changes, and heart rate.

B. Diagnostic and Predictive Analytics

2.1 AI-Based Analysis Engine

- **Technologies:** Machine Learning and Deep Learning algorithm
- **Features:**

- **Predictive Models:** AI models such as logistic regression, random forests, and neural networks for risk assessment and disorder prediction.
- **Image Analysis:** Convolutional Neural Networks (CNNs) for analyzing thyroid ultrasound images and detecting abnormalities.
- **Natural Language Processing (NLP):** To extract relevant information from unstructured clinical notes and patient-reported data.

2.2 Decision Support System

- **Technologies:** Rule-based expert systems and AI decision support.
- **Features:**
 - **Clinical Decision Support:** Provides diagnostic recommendations based on input data and AI analysis.
 - **Treatment Suggestions:** Offers treatment options based on established clinical guidelines and patient-specific data.

C. User Interface

3.1 Patient Portal

- **Features:**
 - **Self-Assessment Tools:** Interactive questionnaires for symptom assessment.
 - **Test Result Access:** View and interpret lab results and imaging reports.
 - **Appointment Scheduling:** Schedule consultations with endocrinologists.

3.2 Healthcare Provider Dashboard

- **Features:**
 - **Data Management:** Access patient data, test results, and imaging.
 - **Diagnostic Tools:** View AI-generated risk assessments and diagnostic suggestions.
 - **Reporting:** Generate and export patient reports and treatment plans.

D. Monitoring and Feedback

4.1 Real-Time Monitoring

- **Technologies:** IoT (Internet of Things) for wearable devices and symptom tracking.
- **Features:** Continuous monitoring of patient symptoms and health metrics, with alerts for abnormal patterns.

3. Objectives of the Proposed System

1. Enhance Early Detection of Thyroid Disorders

Objective: To leverage advanced artificial intelligence (AI) algorithms to identify thyroid disorders at an early stage.

Explanation: The system employs machine learning models to analyze patient data, including lab results and symptoms, to predict the risk of conditions like hypothyroidism, hyperthyroidism, and thyroid cancer.

2. Automate Thyroid Ultrasound Image Analysis

Objective: To improve the accuracy and efficiency of thyroid ultrasound image interpretation through automated analysis.

Explanation: Convolutional Neural Networks (CNNs) are used to automatically detect and classify thyroid abnormalities in ultrasound images, such as nodules or cysts.

3. Integrate Multi-Source Clinical Data into a Unified Platform

Objective: To consolidate various sources of thyroid health data into a single, comprehensive platform for better management and analysis.

Explanation: The system integrates clinical lab results, imaging data, and home test results into one interface for seamless access and evaluation by healthcare professionals.

4. Provide Real-Time Monitoring of Thyroid-Related Symptoms

Objective: To offer tools for continuous, real-time monitoring of thyroid-related symptoms through wearable devices and mobile applications.

Explanation: Wearable technology and mobile apps are used to track symptoms such as fatigue, weight changes, and temperature fluctuations, which are then analyzed for trends and deviations.

5. Develop a User-Friendly Patient Portal for Health Management

Objective: To create an intuitive and accessible portal for patients to manage their thyroid health, including viewing test results and scheduling appointments.

Explanation: The patient portal provides a user-friendly interface for patients to access their health data, complete self-assessments, and manage their care.

CHAPTER 3

SYSTEM REQUIREMENTS SPECIFICATION

1. Hardware Requirements

The following are needed to efficiently use the application.

| | | |
|-----------|---|-------------------------|
| Processor | - | Intel Core i3 and above |
| Speed | - | GHz 2.5 |
| RAM | - | 8 GB (min) |
| Hard Disk | - | GB 50 |

2. Software Requirements

Software requirements define software resource fundamentals that need to be installed on a workstation to provide optimum working of a software. The following are required for optimal development and usage of the application.

| | | |
|----------------------|---|---------------------|
| Operating System | - | and 7 Windows above |
| Programming Language | - | Python |
| Compiler | - | Google Colab |

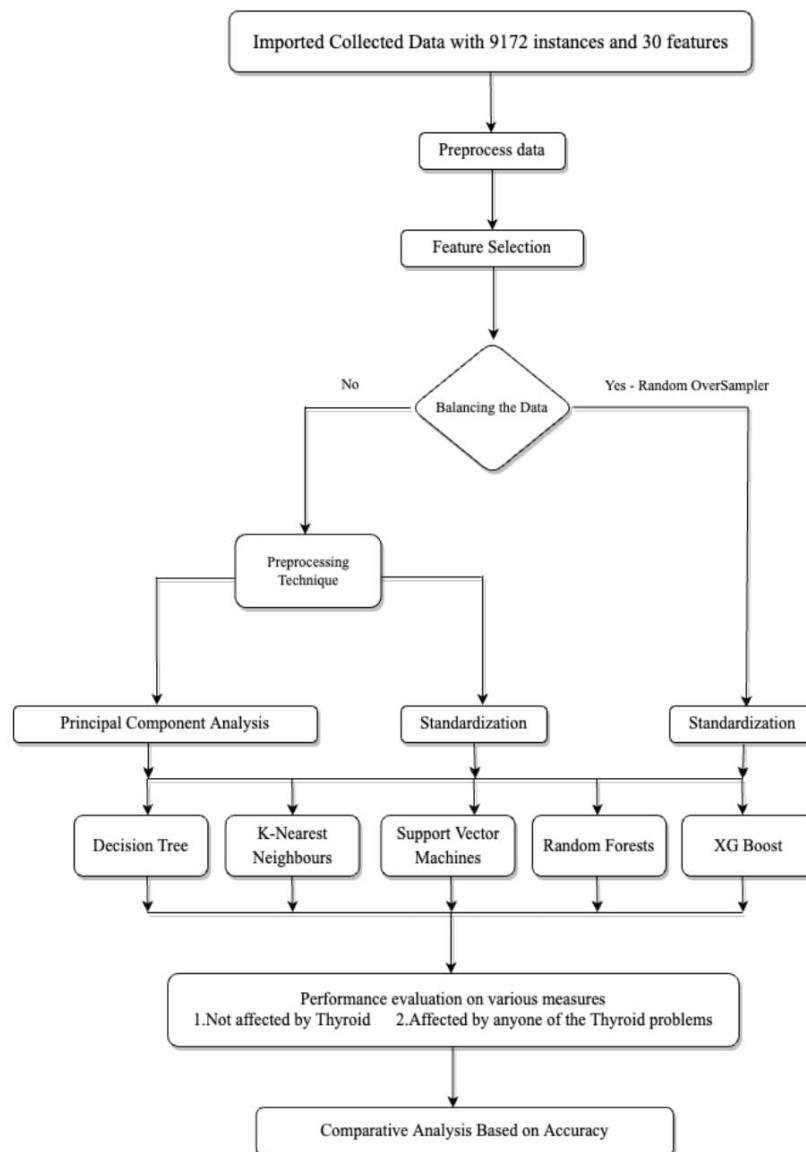
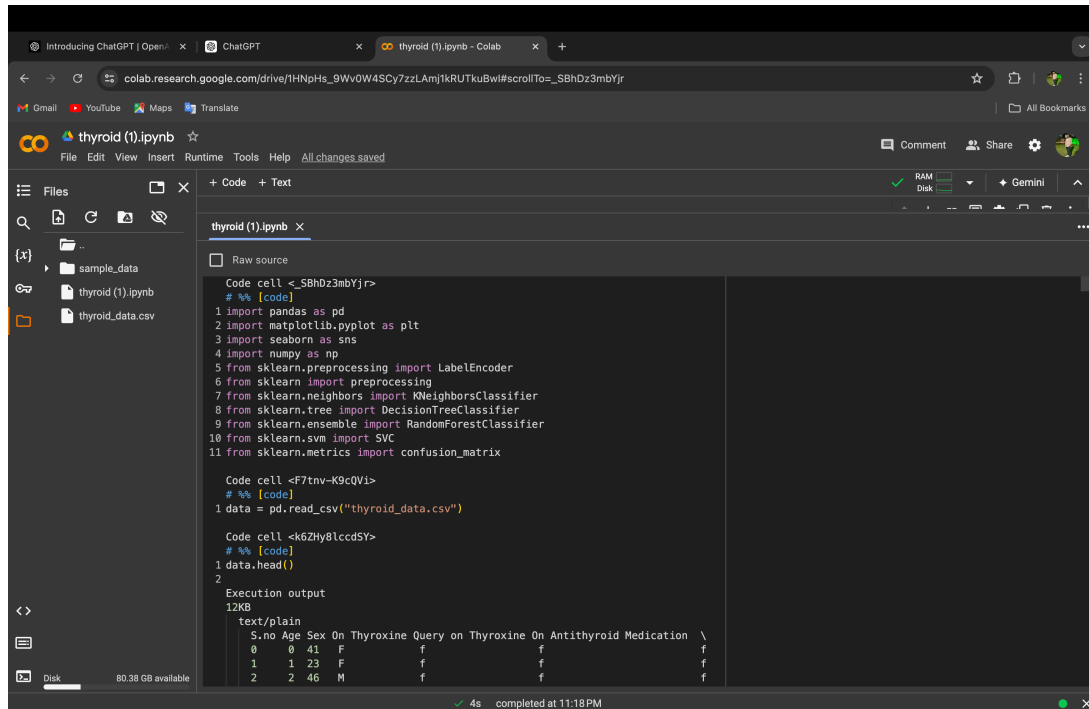
Chapter 4:**1. Flowchart of Proposed System****Methodology of our proposed thyroid Diagnosis**

Fig Flowchart :4.2 of proposed system

CHAPTER 5

IMPLEMENTATION

1. Code Implementation

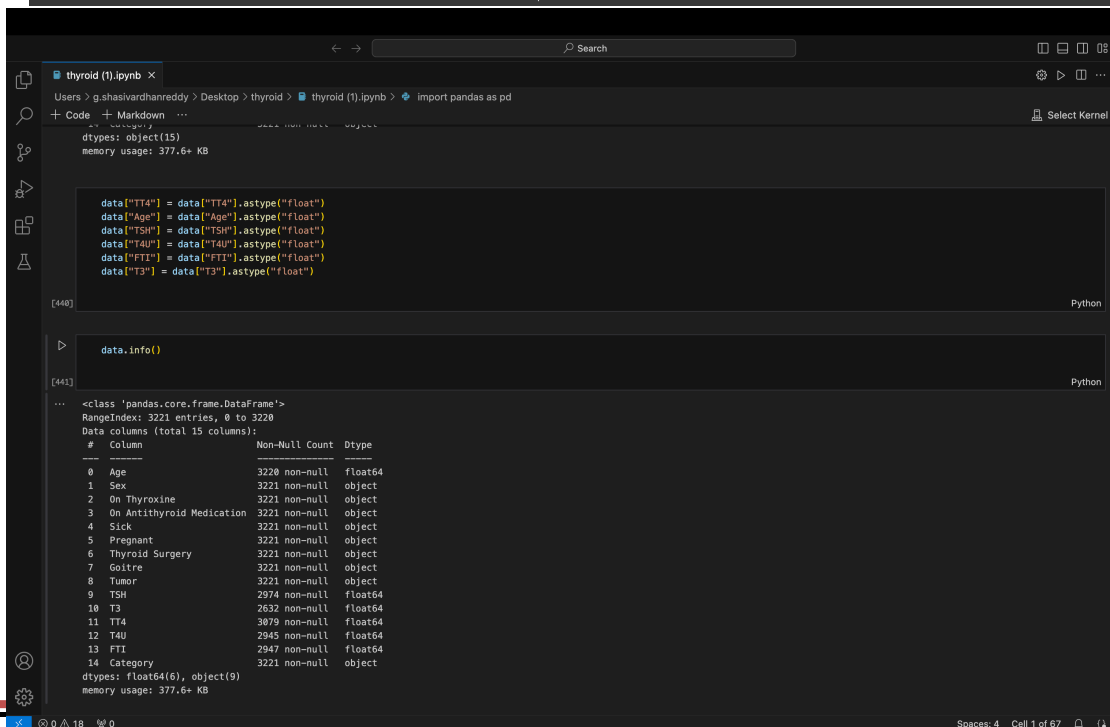


```
Code cell <_SBhdz3mbYjr>
# %% [code]
1 import pandas as pd
2 import matplotlib.pyplot as plt
3 import seaborn as sns
4 import numpy as np
5 from sklearn.preprocessing import LabelEncoder
6 from sklearn import preprocessing
7 from sklearn.neighbors import KNeighborsClassifier
8 from sklearn.tree import DecisionTreeClassifier
9 from sklearn.ensemble import RandomForestClassifier
10 from sklearn.svm import SVC
11 from sklearn.metrics import confusion_matrix

Code cell <F7tnv-K9cQVi>
# %% [code]
1 data = pd.read_csv("thyroid_data.csv")

Code cell <k6ZHy8lccdSY>
# %% [code]
1 data.head()
2

Execution output
12KB
text/plain
S.no Age Sex On Thyroxine Query on Thyroxine On Antithyroid Medication \
0 0 41 F f f f
1 1 23 F f f f
2 2 46 M f f f
```



```
Code cell <_SBhdz3mbYjr>
# %% [code]
1 import pandas as pd
2 import matplotlib.pyplot as plt
3 import seaborn as sns
4 import numpy as np
5 from sklearn.preprocessing import LabelEncoder
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7 from sklearn.neighbors import KNeighborsClassifier
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Execution output
12KB
text/plain
S.no Age Sex On Thyroxine Query on Thyroxine On Antithyroid Medication \
0 0 41 F f f f
1 1 23 F f f f
2 2 46 M f f f
```

```
Code cell <_SBhdz3mbYjr>
# %% [code]
1 data = pd.read_csv("thyroid_data.csv")
2 data["TT4"] = data["TT4"].astype("float")
3 data["Age"] = data["Age"].astype("float")
4 data["TSH"] = data["TSH"].astype("float")
5 data["T4U"] = data["T4U"].astype("float")
6 data["FTI"] = data["FTI"].astype("float")
7 data["T3"] = data["T3"].astype("float")

Code cell <k6ZHy8lccdSY>
# %% [code]
1 data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 3221 entries, 0 to 3220
Data columns (total 15 columns):
# Column Non-Null Count Dtype
---
0 Age 3221 non-null float64
1 Sex 3221 non-null object
2 On Thyroxine 3221 non-null object
3 On Antithyroid Medication 3221 non-null object
4 Sick 3221 non-null object
5 Pregnant 3221 non-null object
6 Thyroid Surgery 3221 non-null object
7 Goitre 3221 non-null object
8 Tumor 3221 non-null object
9 TSH 2974 non-null float64
10 T3 2632 non-null float64
11 TT4 3879 non-null float64
12 T4U 2945 non-null float64
13 FTI 2947 non-null float64
14 Category 3221 non-null object
dtypes: float64(6), object(9)
memory usage: 377.6+ KB
```

The screenshot shows a Jupyter Notebook with the following content:

- Cell [444]:

```
data['Age'] = np.where(data['Age'] > 100, np.nan, data['Age'])
```
- Cell [445]:

```
data["Age"] = data["Age"].fillna(data["Age"].median())
```
- Section: **TSH**
- Cell [446]:

```
TSH = data[(data["TSH"]>8)]
data['TSH'] = np.where(data['TSH'] > 8, np.nan, data['TSH'])
```
- Cell [447]:

```
TSH
```

The output of cell [447] shows a subset of the data with TSH values greater than 8:

| Age | Sex | On Thyroxine | On Antithyroid Medication | Sick | Pregnant | Thyroid Surgery | Goitre | Tumor | TSH | T3 | TT4 | T4U | FTI | Category |
|------|------|--------------|---------------------------|------|----------|-----------------|--------|-------|------|-----|-------|------|-------|----------|
| 15 | 65.0 | F | f | f | f | f | f | f | 12.0 | NaN | 99.0 | 1.14 | 87.0 | negative |
| 26 | 60.0 | M | t | f | f | f | f | f | 13.0 | 1.4 | 57.0 | 0.62 | 92.0 | negative |
| 40 | 44.0 | M | f | f | f | f | f | f | 45.0 | 1.4 | 39.0 | 1.16 | 33.0 | negative |
| 57 | 65.0 | M | f | f | f | f | f | f | 14.8 | 1.5 | 61.0 | 0.85 | 72.0 | negative |
| 58 | 27.0 | F | f | f | f | f | f | f | 15.0 | 1.6 | 82.0 | 0.82 | 100.0 | negative |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 3166 | 61.0 | F | f | f | t | f | f | f | 9.8 | 0.4 | 81.0 | 0.64 | 126.0 | sick |
| 3171 | 37.0 | F | f | f | f | f | f | f | 27.0 | 0.9 | 87.0 | 0.84 | 105.0 | sick |
| 3185 | 72.0 | F | f | f | f | f | f | f | 9.7 | 1.1 | 77.0 | 0.79 | 97.0 | sick |
| 3186 | 49.0 | F | f | f | f | f | f | f | 8.3 | 0.9 | 103.0 | 1.11 | 93.0 | sick |

The screenshot shows a Jupyter Notebook with the following content:

- Cell [442]:

```
data.describe()
```
- Cell [443]:

```
high_age = data[data["Age"]>100]
high_age
```
- Cell [444]:

```
data['Age'] = np.where(data['Age'] > 100, np.nan, data['Age'])
```
- Cell [445]:

```
data["Age"] = data["Age"].fillna(data["Age"].median())
```

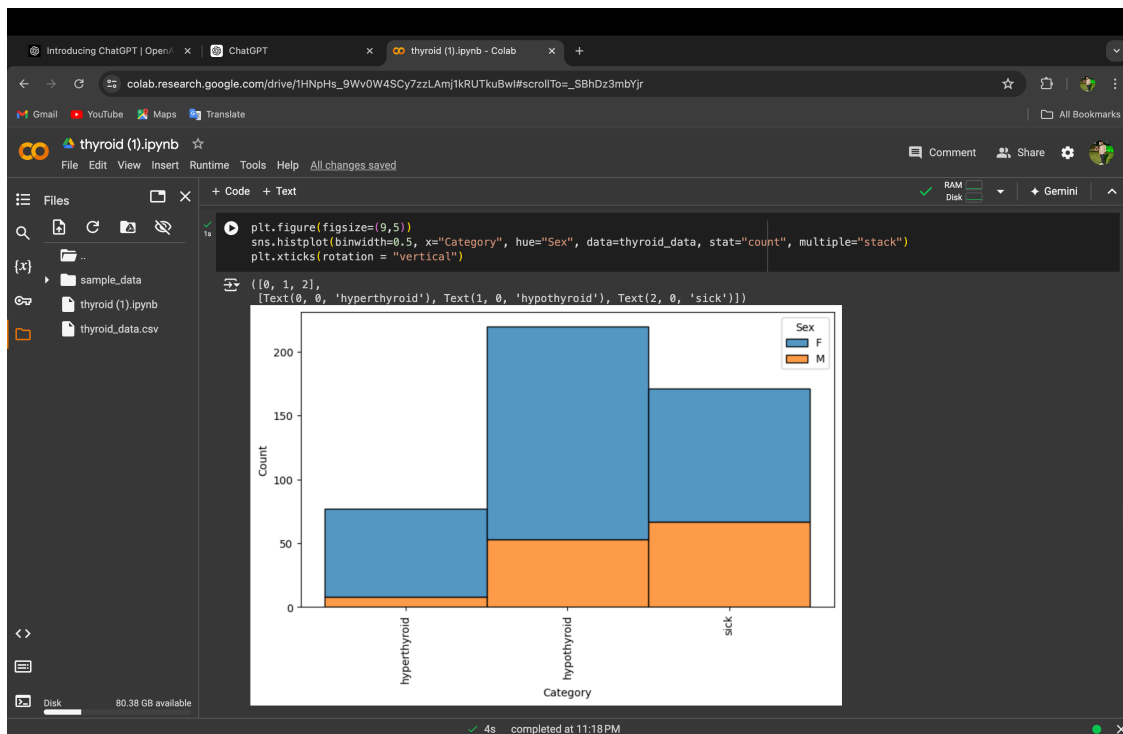
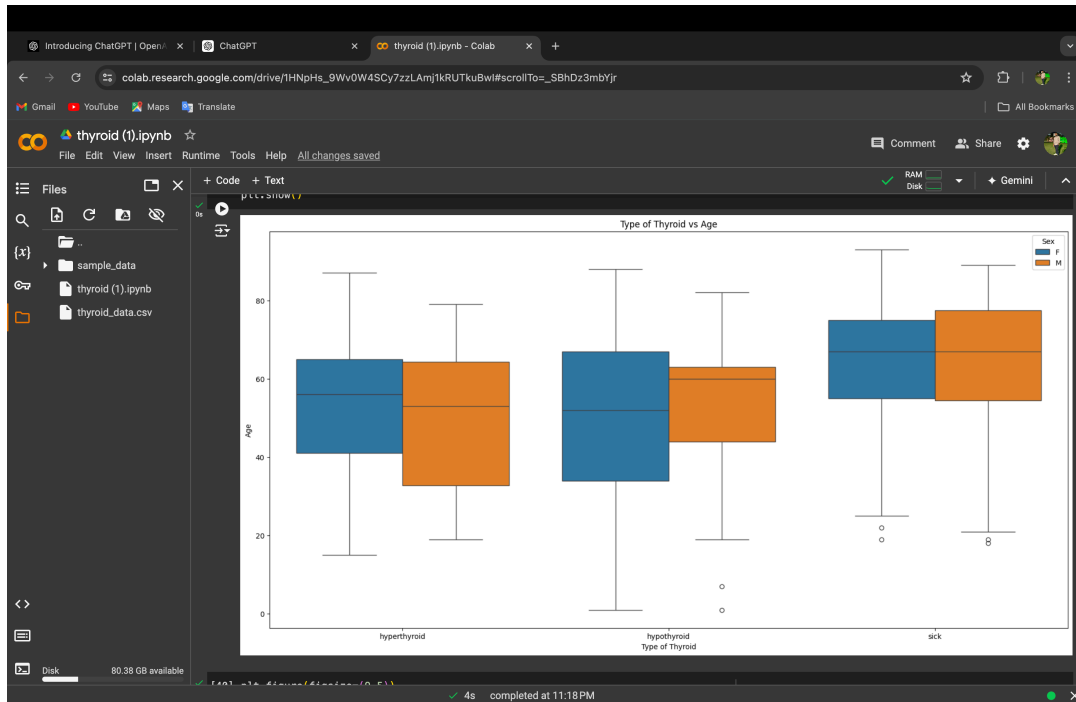
The output of cell [442] shows the summary statistics for the data:

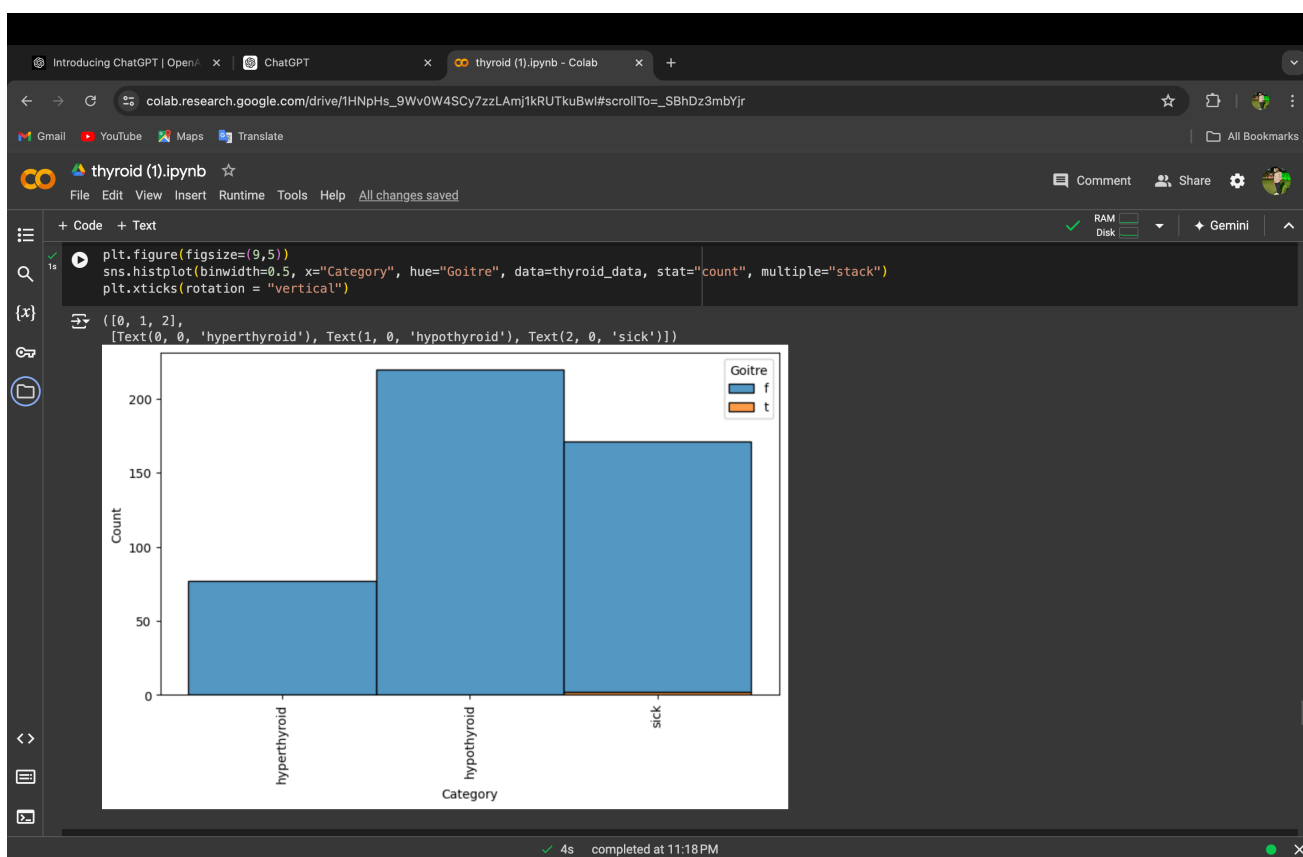
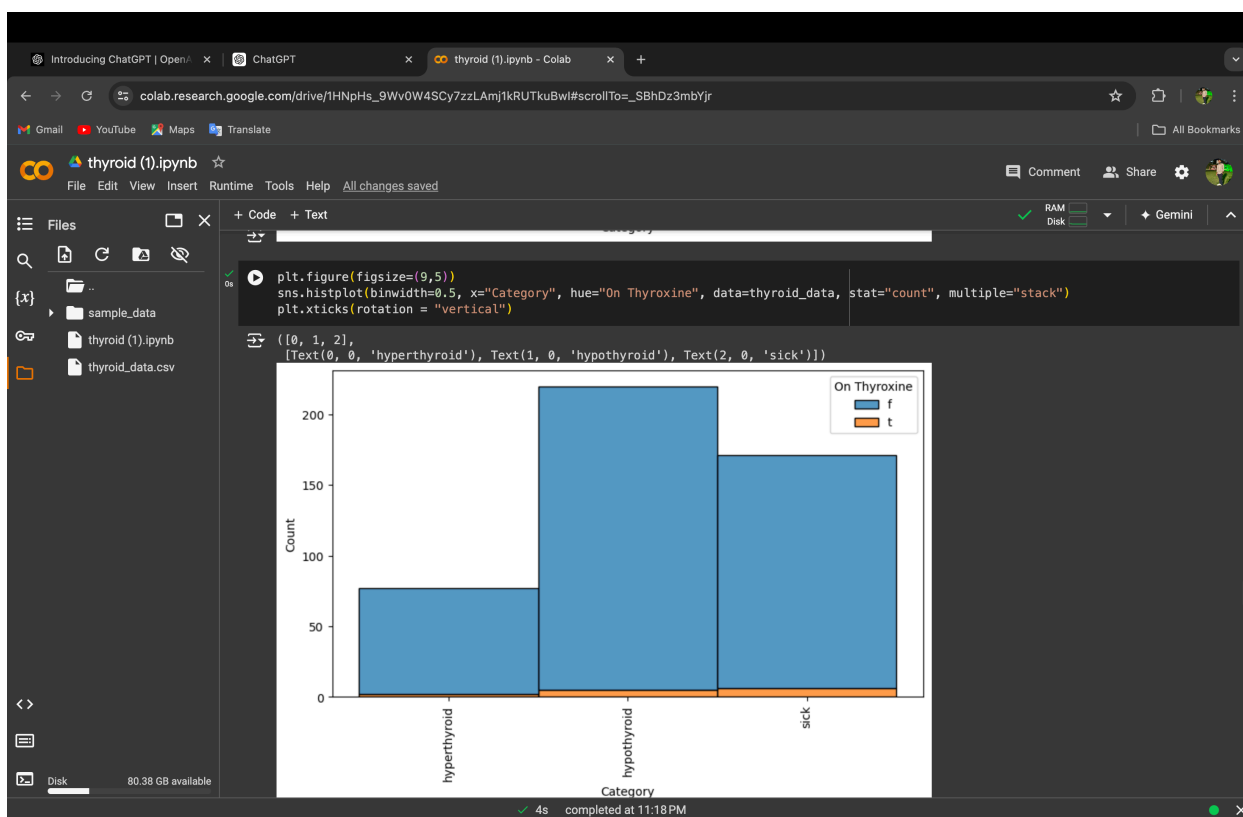
| | Age | TSH | T3 | TT4 | T4U | FTI |
|-------|-------------|-------------|-------------|-------------|-------------|-------------|
| count | 3220.000000 | 2974.000000 | 2632.000000 | 3079.000000 | 2945.000000 | 2947.000000 |
| mean | 52.532609 | 6.722840 | 1.963355 | 107.807665 | 0.989937 | 110.657550 |
| std | 20.381326 | 27.585818 | 0.928773 | 38.941172 | 0.194417 | 37.578197 |
| min | 1.000000 | 0.005000 | 0.050000 | 2.000000 | 0.310000 | 2.000000 |
| 25% | 37.000000 | 0.450000 | 1.500000 | 86.000000 | 0.870000 | 92.000000 |
| 50% | 55.000000 | 1.500000 | 1.900000 | 102.000000 | 0.970000 | 106.000000 |
| 75% | 68.000000 | 3.300000 | 2.300000 | 125.000000 | 1.080000 | 125.000000 |
| max | 455.000000 | 478.000000 | 10.600000 | 430.000000 | 2.120000 | 395.000000 |

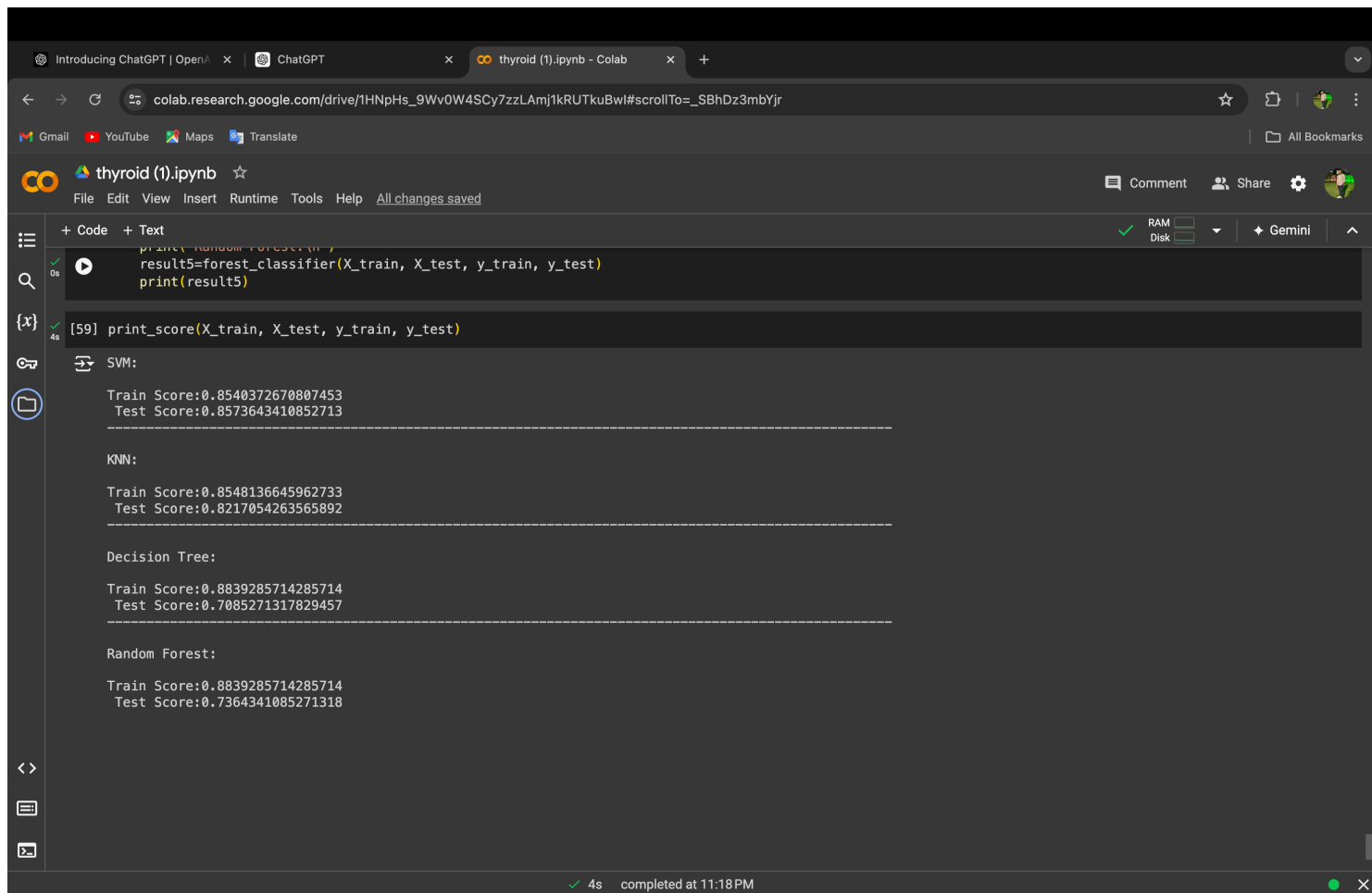
CHAPTER 6

EXPERIMENTAL RESULTS

1. Outcome of Proposed System







The screenshot displays a Google Colab notebook interface. The browser tabs at the top include 'Introducing ChatGPT | OpenAI', 'ChatGPT', and 'thyroid (1).ipynb - Colab'. The address bar shows the Colab URL. The notebook title is 'thyroid (1).ipynb'. The left sidebar contains icons for file management, search, and other notebook functions. The top right of the notebook area shows 'Comment', 'Share', and 'Gemin' options, along with RAM and Disk usage indicators.

The code editor shows the following Python code:

```
print(random_forest_classifier(X_train, X_test, y_train, y_test))
result5=forest_classifier(X_train, X_test, y_train, y_test)
print(result5)
```

The output of the code execution is displayed below the code cells:

```
[59] print_score(X_train, X_test, y_train, y_test)
```

The output shows the training and testing scores for four different models:

SVM:

```
Train Score:0.8540372670807453
Test Score:0.8573643410852713
```

KNN:

```
Train Score:0.8548136645962733
Test Score:0.8217054263565892
```

Decision Tree:

```
Train Score:0.8839285714285714
Test Score:0.7085271317829457
```

Random Forest:

```
Train Score:0.8839285714285714
Test Score:0.7364341085271318
```

The bottom status bar indicates that the execution was successful (green checkmark) and completed at 11:18 PM.

CHAPTER

7

CONCLUSION AND FUTURE ENHANCEMENT

Final Thoughts

In this project We looked into different techniques for the thyroid classification methods like SVM,KNN, random forest and SVM tree in which resulted: -test 20% and train 80% train score :85%and test score:85%.From this we can say that desicion SVM tree is the good classification algorithm with accuracy 85%and SVM is the best classification algorithm for thyroid prediction

Thyroid Detection using Machine Learning is a project idea that aims a smart and precise way to predict thyroid disease. We have made use of logistic regression algorithm to train our dataset and to predict thyroid disease with more accuracy. Here the machine is trained to detect whether the person normal, hyper-hypo thyroid ism based on the user's input. So when user enters data in web app the data will be processed in backend (model) and the result will be displayed onthe screen. Our objective was to give society an efficient and precise way of machine learning which can be used in applications aiming to perform disease detection.

Further development can be do by using image processing of ultrasonic scanning of thyroid images top redict thyroid nodules and cancer, which cannot be recognized in blood test report.By combining both the results, thyroid diseaseprediction can cover all thyroid related diseases.

REFERENCES

1. Research Papers
2. Web Links
3. Books