

# I HEALTH RECOMMENDATION SYSTEM

Disease prediction

# Our Team

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## • **Introduction**

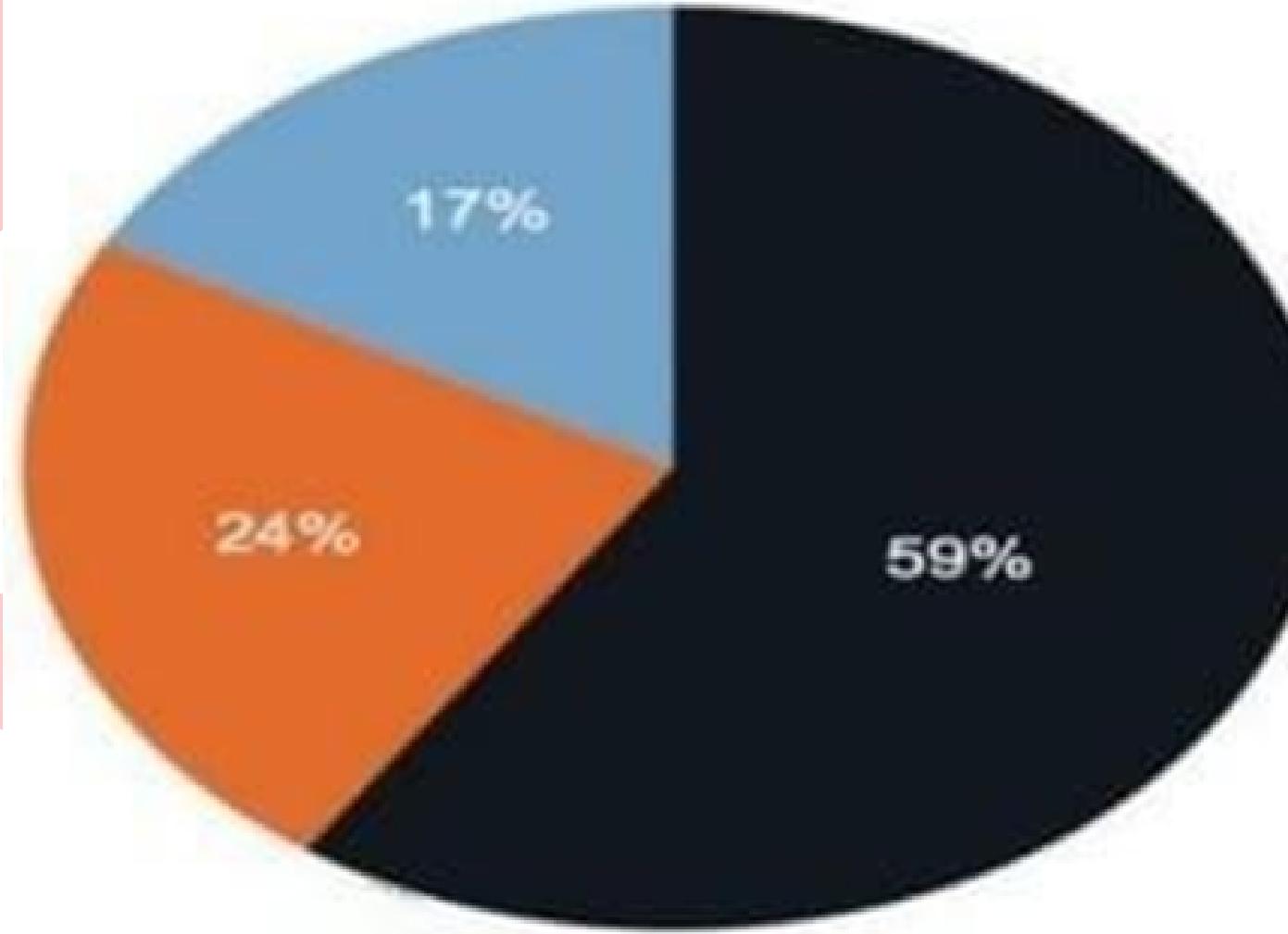
This project focuses on developing an intelligent healthcare chatbot. By leveraging artificial intelligence and natural language processing, the chatbot interacts with users to analyze their reported symptoms and identify possible diseases.

It provides instant medical guidance and tailored advice, especially useful in areas with limited healthcare services or outside regular clinic hours. The chatbot serves as an accessible and efficient tool to enhance health awareness, support early diagnosis, and improve overall public health.

# Background and Motivation

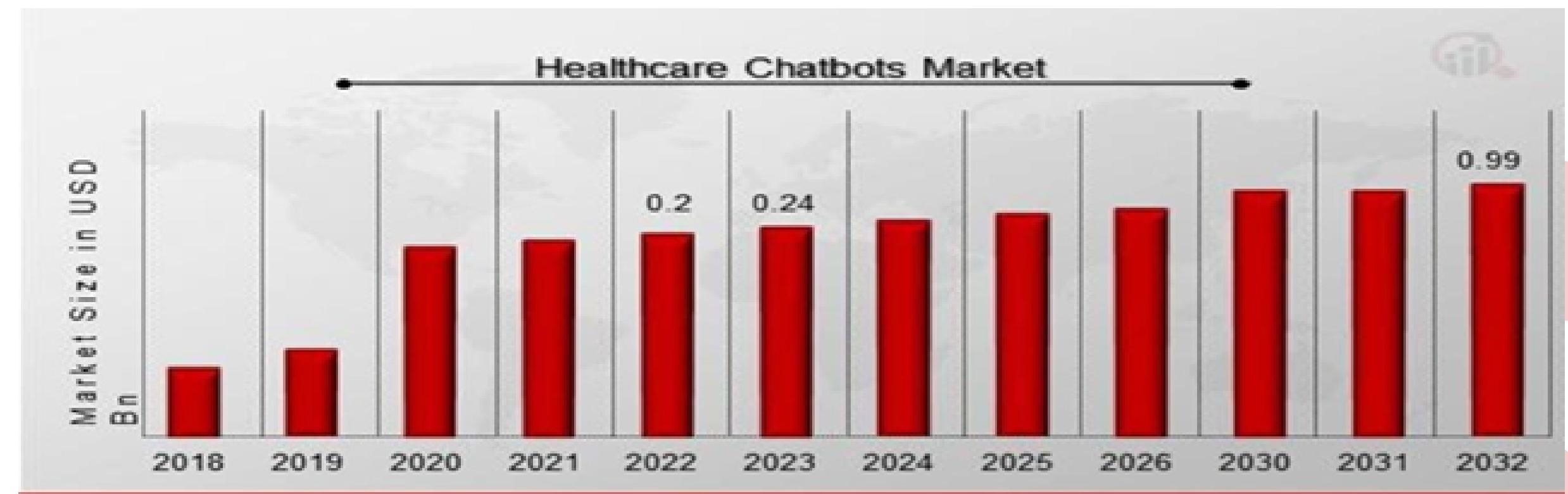
Over the past decade, global healthcare systems have faced growing challenges due to increasing patient loads, limited resources, and unequal access to medical services—especially in remote areas where timely consultations are often unavailable. Advances in artificial intelligence, particularly in machine learning and natural language processing (NLP), have made it possible to analyze medical data and simulate diagnostic reasoning. These developments inspired our team to explore how AI can serve as a reliable, automated first point of contact to bridge the healthcare accessibility gap.

Our motivation stems from the rising trust in AI tools and the urgent digital health solutions—an area still largely underserved. Key goals include providing 24/7 access to healthcare advice, empowering users with health education, promoting early detection through symptom monitoring, and supporting healthcare professionals by managing simple inquiries digitally. Notably, the healthcare chatbot market was valued at USD 0.2 billion in 2022, highlighting the growing relevance of this technology.



The chart indicates that 59% of the population supports AI's role in human assistance, while 24% lack knowledge about AI's functions and 17% are clueless.

# Statistics



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# Problem Statement

If the patient delays finding out about the disease he has, regardless of the reasons for his delay, and does not take the necessary precautions as soon as possible, his health condition will worsen before he is able to go to the doctor.

01

The user enters linguistic words that do not affect the understanding of symptoms, which leads to an incorrect classification of the disease.

02

Possible inaccurate classification of diseases and this may put the patient's life at risk.

03

Patients want a Suitable and Interactive GUI to easily deal with chatbot.

04

The patient does not have enough knowledge of the necessary precautions and which hospitals are closest to the patient's location.

# OBJECTIVE

To develop an AI-powered healthcare chatbot that analyzes user-reported symptoms, provides instant medical guidance, and improves accessibility to basic healthcare services especially in underserved or remote areas—while supporting early detection and empowering users with health awareness.

# solve problem

- 1- In order to understand the description of the patient's symptoms,**
- 2- In order to correctly diagnose the disease and achieve the best possible accuracy from the ML model, we applied hyperparameter tuning using Grid Search method on Different ML models.**

**3- Comprehensive Medical Dictionary:**  
The chatbot includes a built-in medical dictionary that not only provides the predicted disease but also offers relevant information such as treatment options and recommended lifestyle habits.

**4- Provide patients with some precautions and a Google Map to guide them to the closest hospitals to their location.**

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# limitations

**Low accessibility** in remote or underserved areas where medical consultation is limited.

**Limited adaptability:** Many current chatbots are rule-based and do not incorporate advanced ML models for disease classification.

Difficulty in understanding which treatment steps to follow based on their symptoms

**Limited access to immediate medical advice.**

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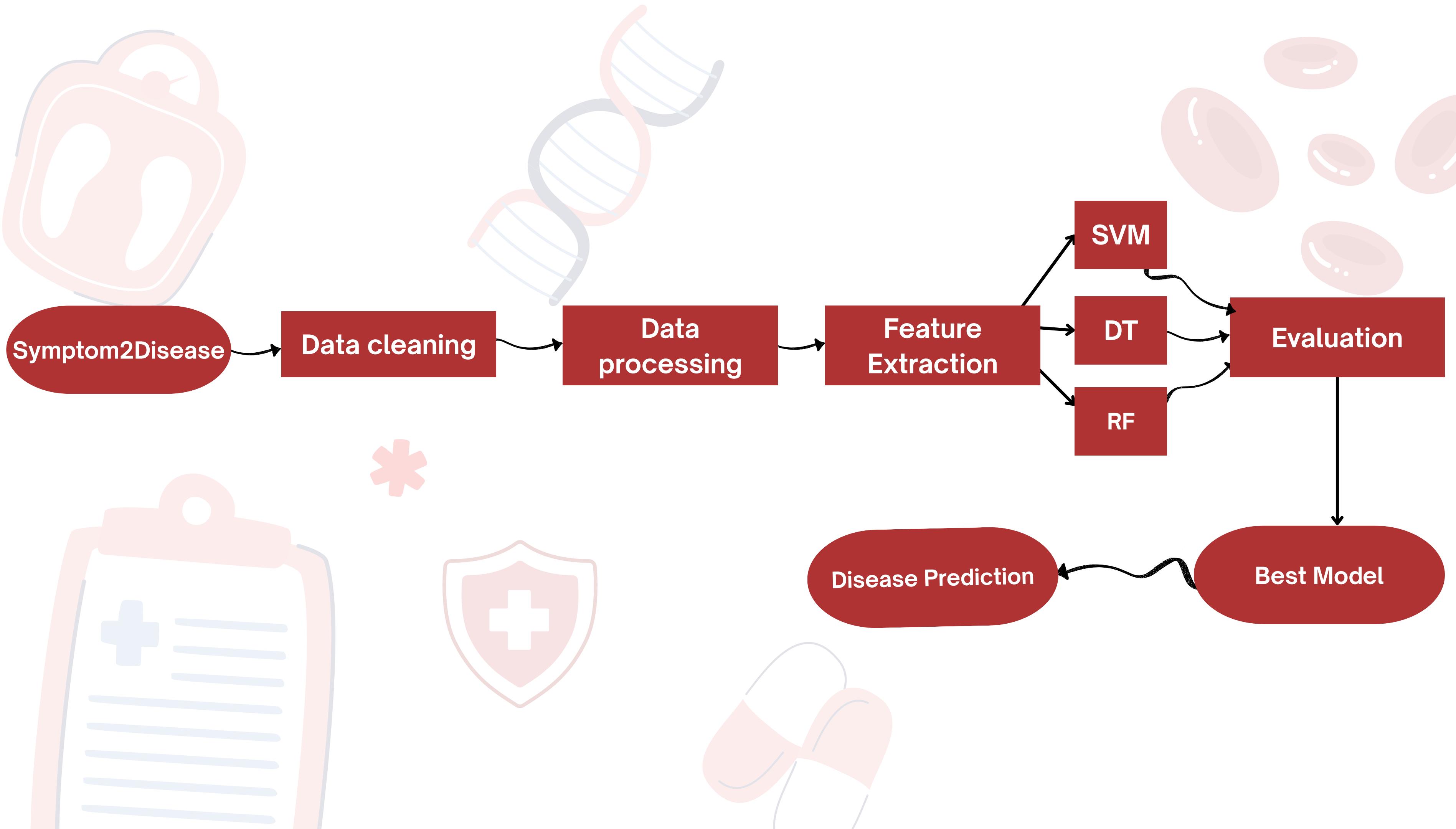
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# METHODOLOGY

## Symptom2Disease

### DESCRIPTION

THE SYMPTOM2DISEASE DATASET IS A COLLECTION OF DISEASES AND EACH DISEASE HAS DIFFERENT SYMPTOMS.

Disease	24
Sympotoms	1153

### DATASET SIZE

- ROWS : 1200
- COLUMNS : 2

### CONTAINS TWO COLUMNS :

- LABLE
- TEXT

lable	Text
Common Cold	My eyes are usually red and runny and my nose is always stuffy.
Acne	My skin has developed a terrible rash It has a lot of blackheads and pus-filled pimples.
GERD	I often vomit whatever I eat and I have a nagging pain in my upper abdomen

# PROPOSED APPROACH

Our healthcare chatbot development followed a systematic approach that combined data preparation, machine learning model development, back-end API building, and mobile app integration. The steps were as follows:

## NLP PREPROCESSING

- APPLIED LOWERCASING, PUNCTUATION/STOPWORD REMOVAL, TOKENIZATION (NLTK), AND LEMMATIZATION.
- CONVERTED TEXT INTO TF-IDF VECTORS USING TFIDFVECTORIZER.

## DATA PREPARATION

- USED THE SYMPTOM2DISEASE DATASET (INITIALLY 1,200 ENTRIES) AND EXPANDED IT TO 3,600 USING DATA AUGMENTATION (QUILLBOT PARAPHRASING).
- APPLIED DATA CLEANING, DUPLICATE REMOVAL, AND ENSURED SEMANTIC CONSISTENCY.

## MODEL TRAINING & EVALUATION

- TRAINED 3 MODELS: DECISION TREE, RANDOM FOREST, SVM (BEST: 97%), .
- USED GRIDSEARCHCV FOR HYPERPARAMETER TUNING.
- EVALUATED USING ACCURACY, CONFUSION MATRIX, AND CLASSIFICATION REPORT.

# PROPOSED APPROACH

## MODEL DEPLOYMENT (FLASK API)

- DEVELOPED /PREDICT ENDPOINT.
- RETURNED DISEASE PREDICTION AND CONFIDENCE SCORE.

## MOBILE APP (FLUTTER)

- BUILT WITH SUPPORT FOR ENGLISH, TEXT/VOICE INPUT, DISEASE DISPLAY,LIFESTYLE,RECOMMENDED SPECIALIST AND MEDICATIONS.
- IT ALSO INCLUDES A MEDICAL DICTIONARY SHOWING:DESCRIPTION, SYMPTOMS, CAUSES, DIAGNOSIS, MEDICATIONS, TREATMENT DURATION, PREVENTION, COMPLICATIONS, AND IMAGE REFERENCES.
- INTEGRATED GOOGLE MAPS FOR HOSPITAL AND PHARMACY LOCATIONS .

# Tools and Technologies

PYTHON

FLUTTER

DART

FLASK

Git, GitHub

## Machine Learning

SUPPORT VECTOR  
MACHINE(SVM)

RANDOM FOREST (RF)

DECISION TREE

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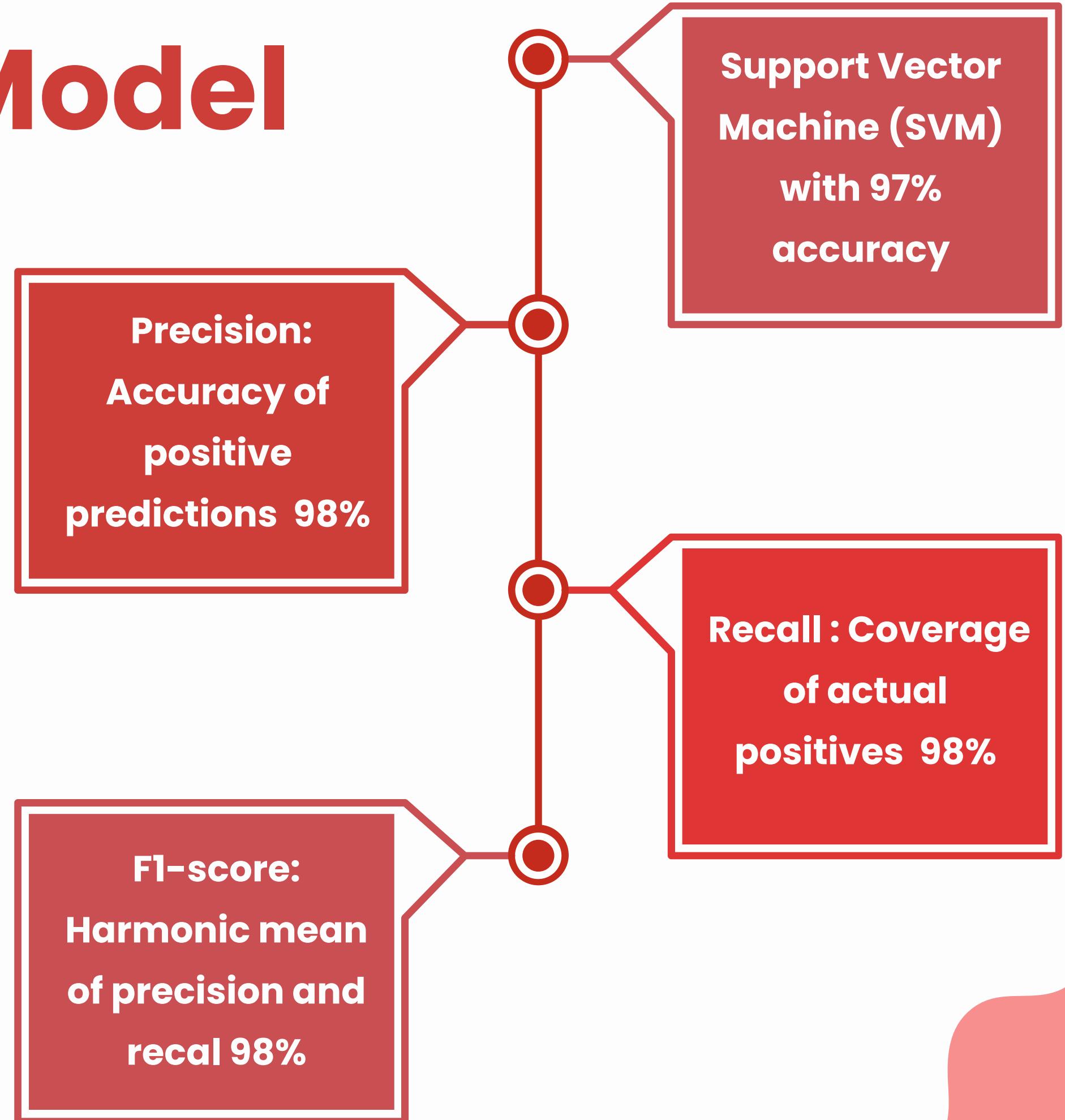
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# Model Evaluation

## Best Model



# TESTING

## Unit Testing

Each function (e.g., text preprocessing, TF-IDF vectorization, prediction) was tested individually, including handling of edge cases like empty or invalid inputs.

## Integration Testing

The communication between the Flutter app and Flask backend was tested to ensure proper API interaction.

## User Testing

Real users tested the mobile app by submitting symptoms via text/voice, switching languages, and viewing predictions and hospital, pharmacy locations. Feedback was used to enhance user experience.

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# Results

## Symptom2Disease

	SVM	DT	RF
Accuracy	97%	76%	95%
Precision	98%	82%	96%
Recall	98%	76%	96%
F1-score	98%	77%	96%

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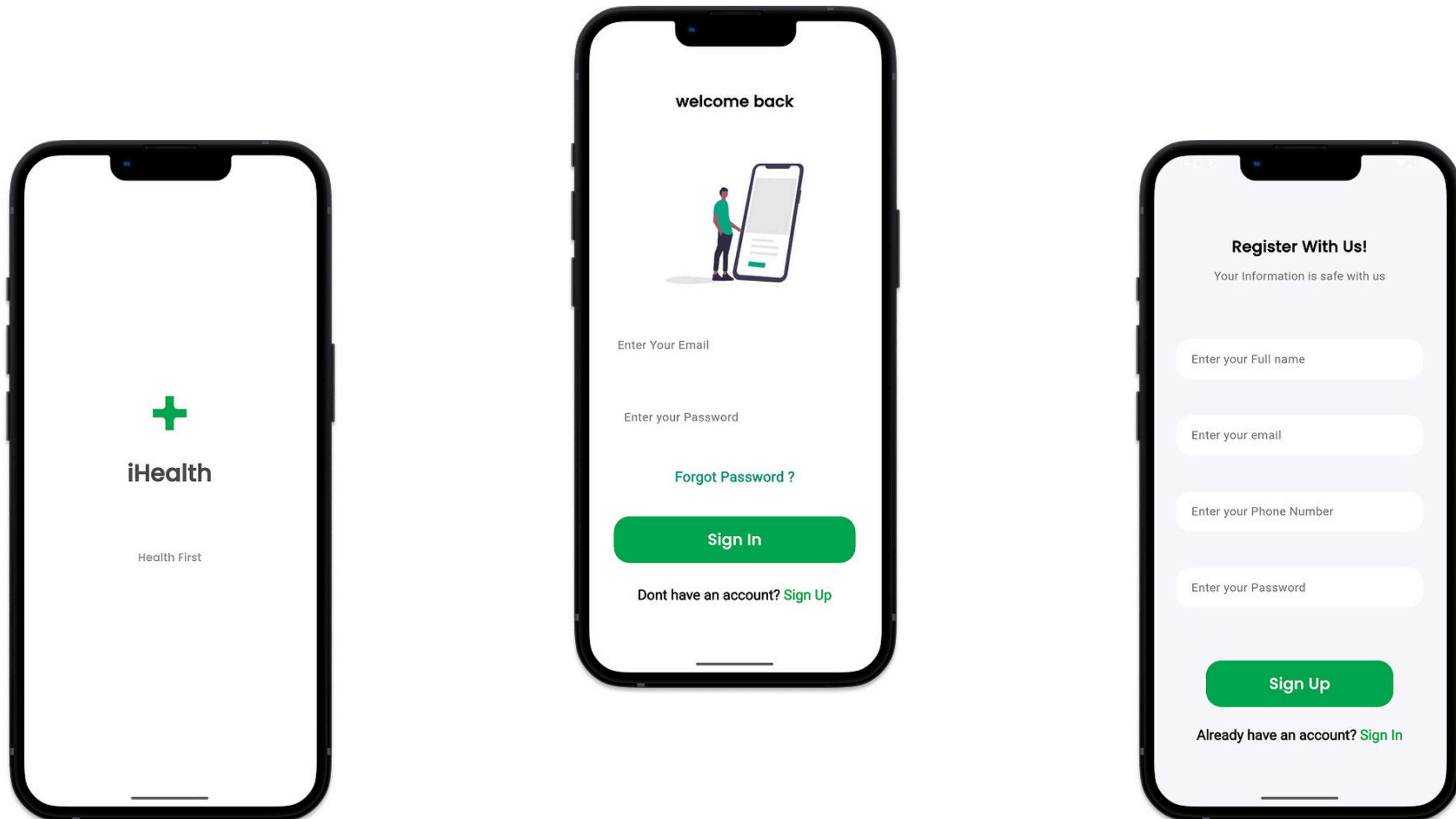
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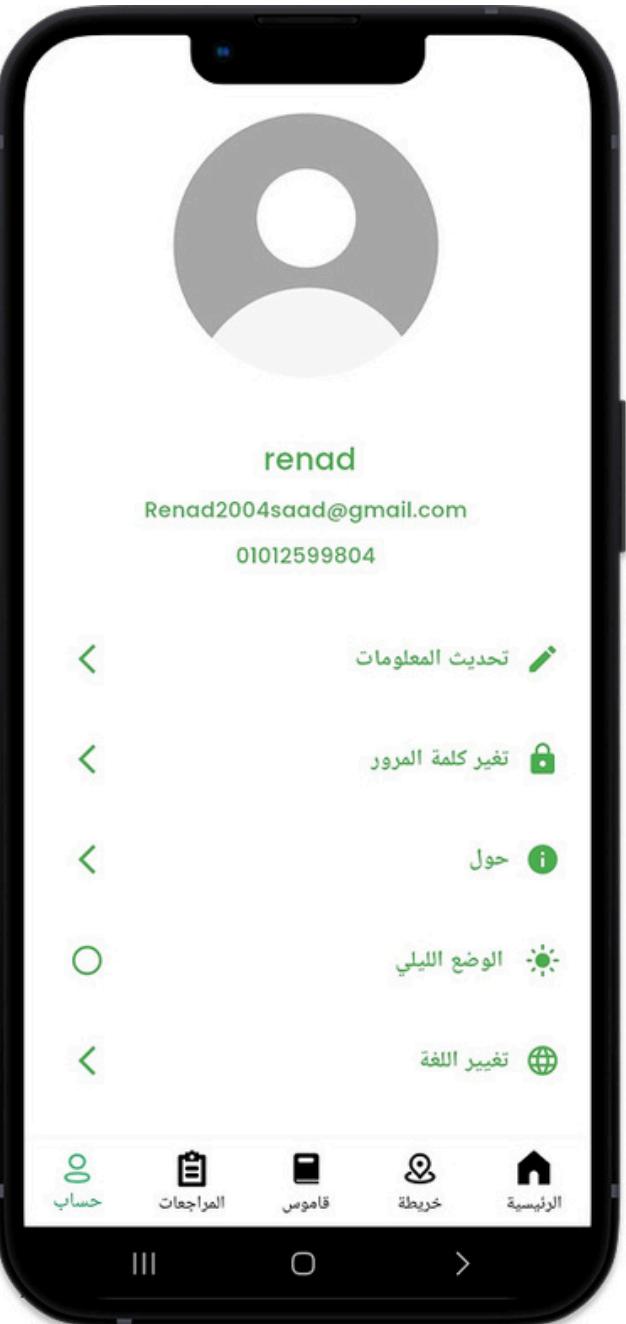
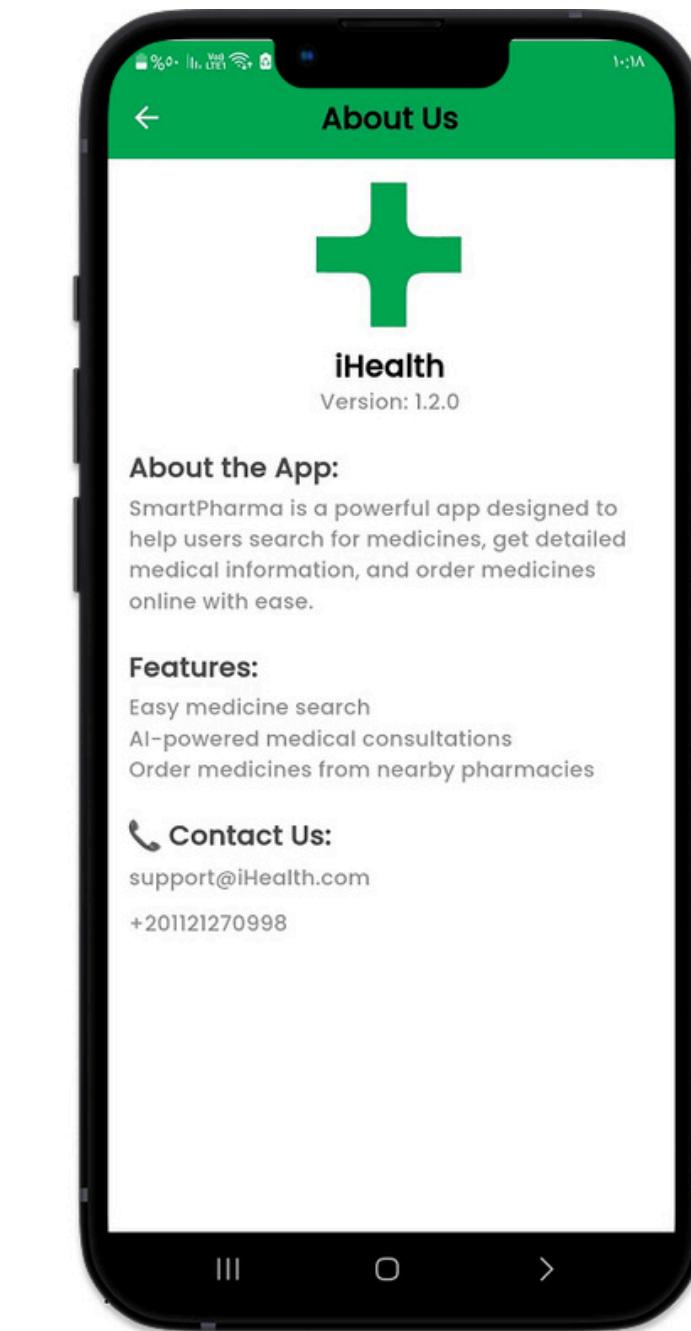
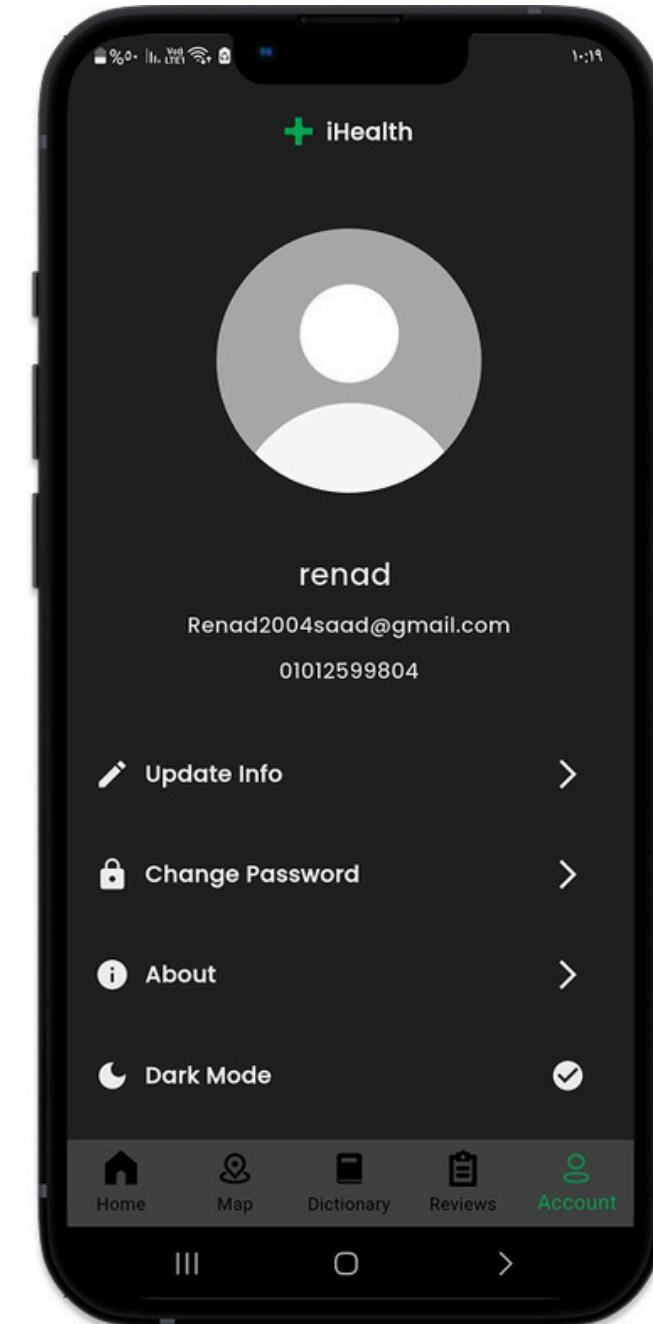
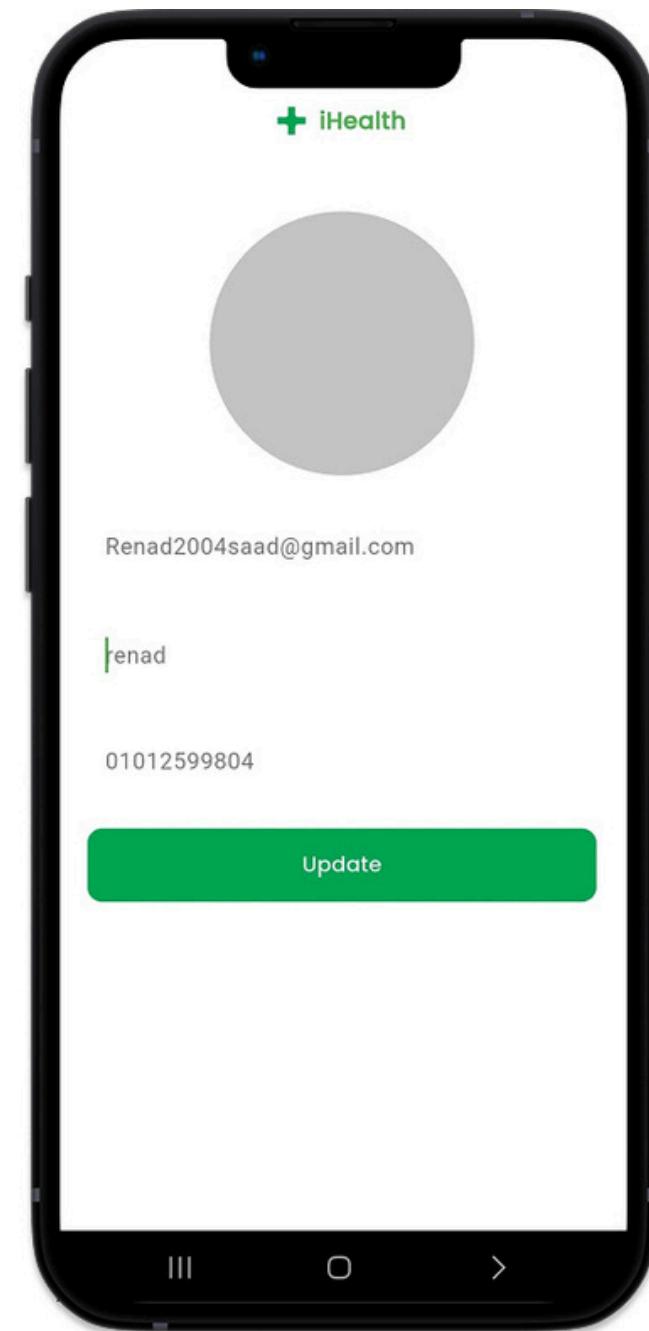
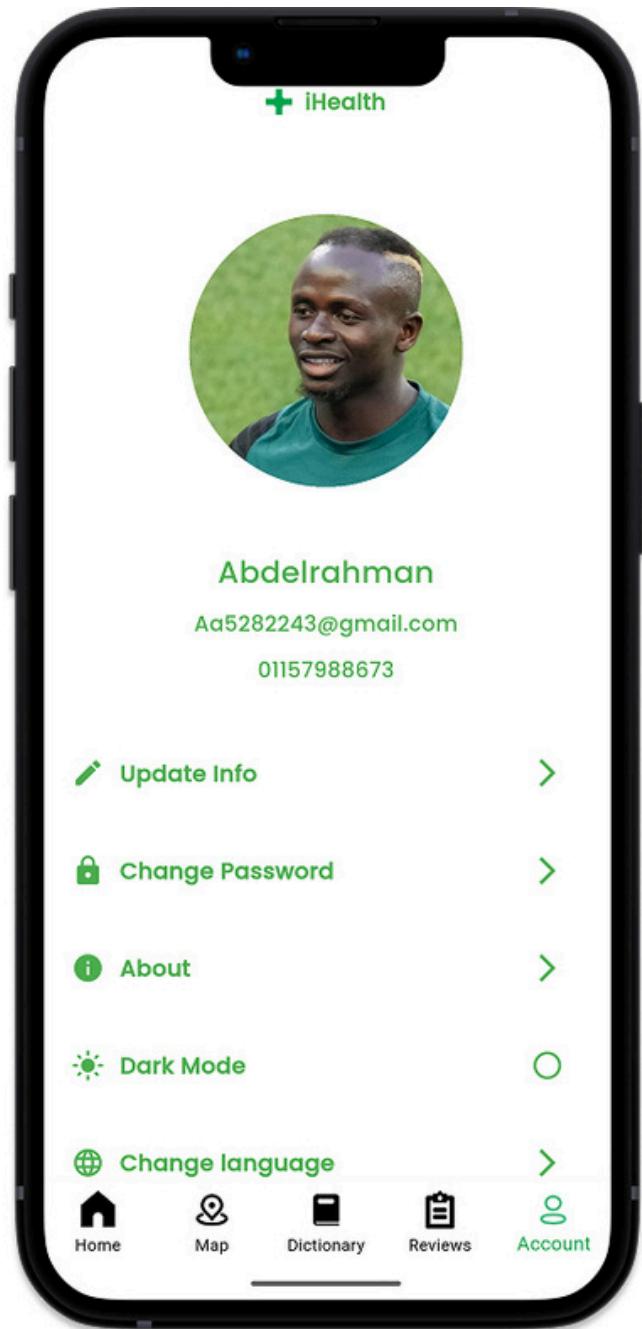
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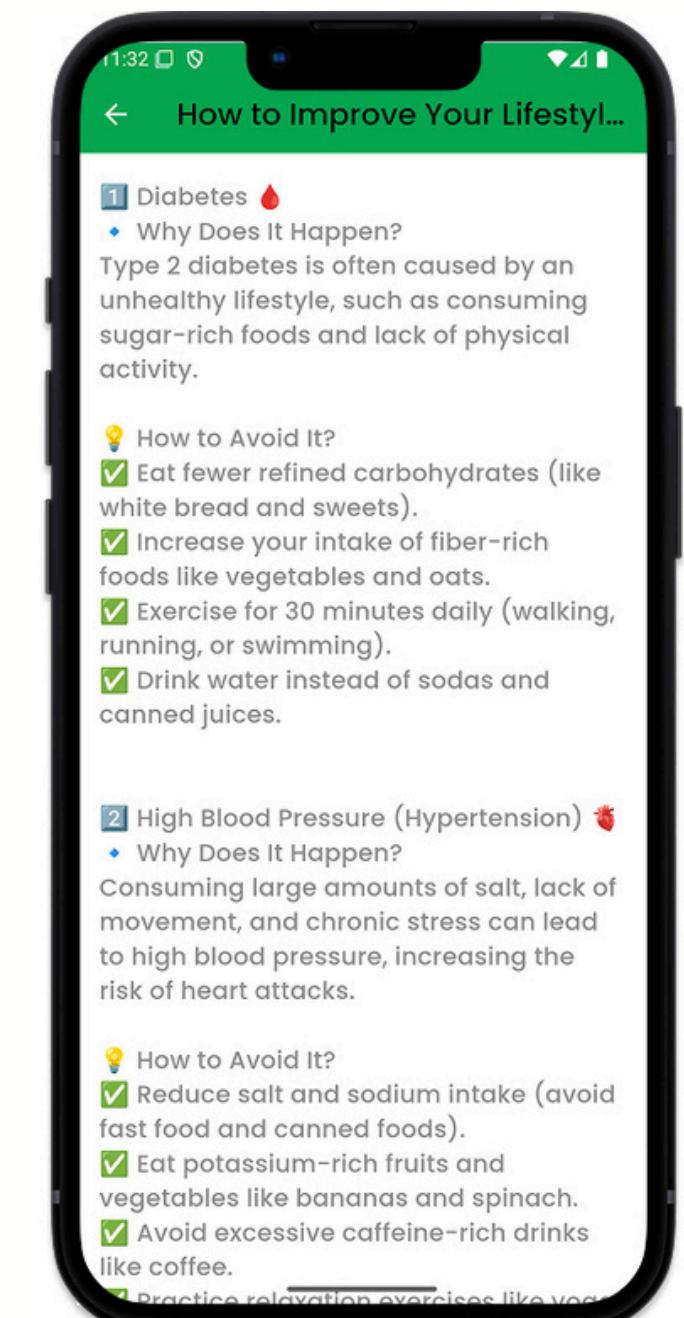
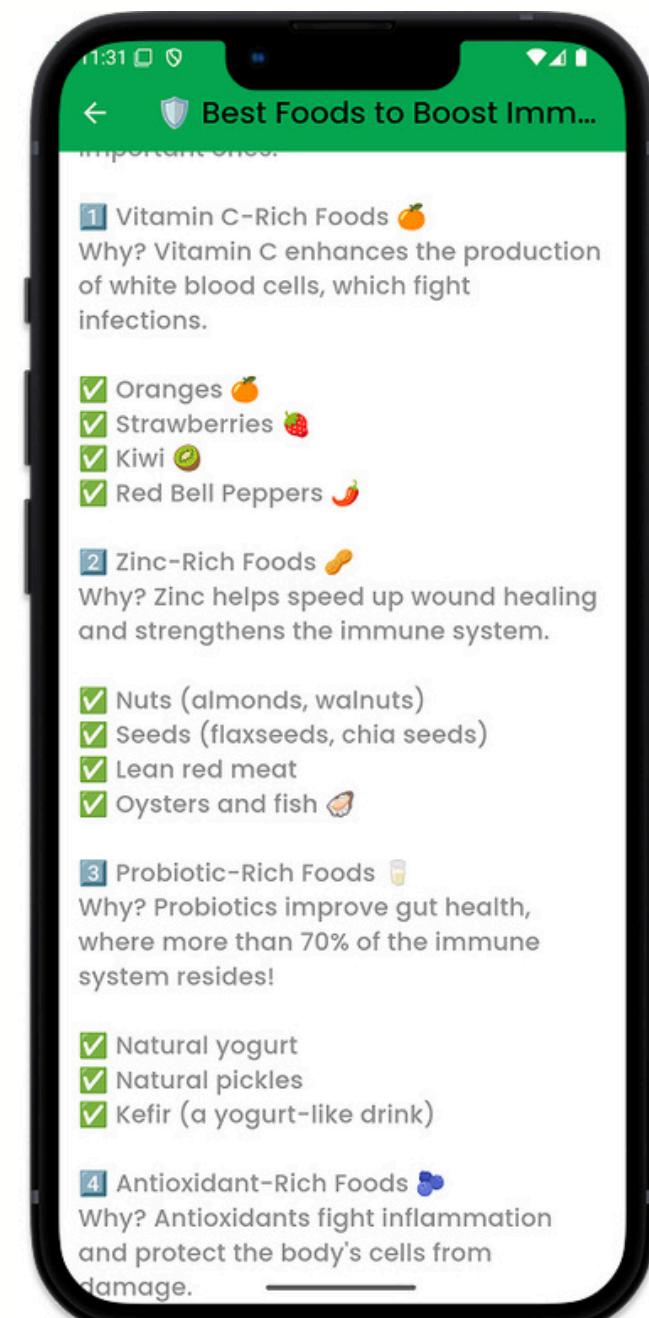
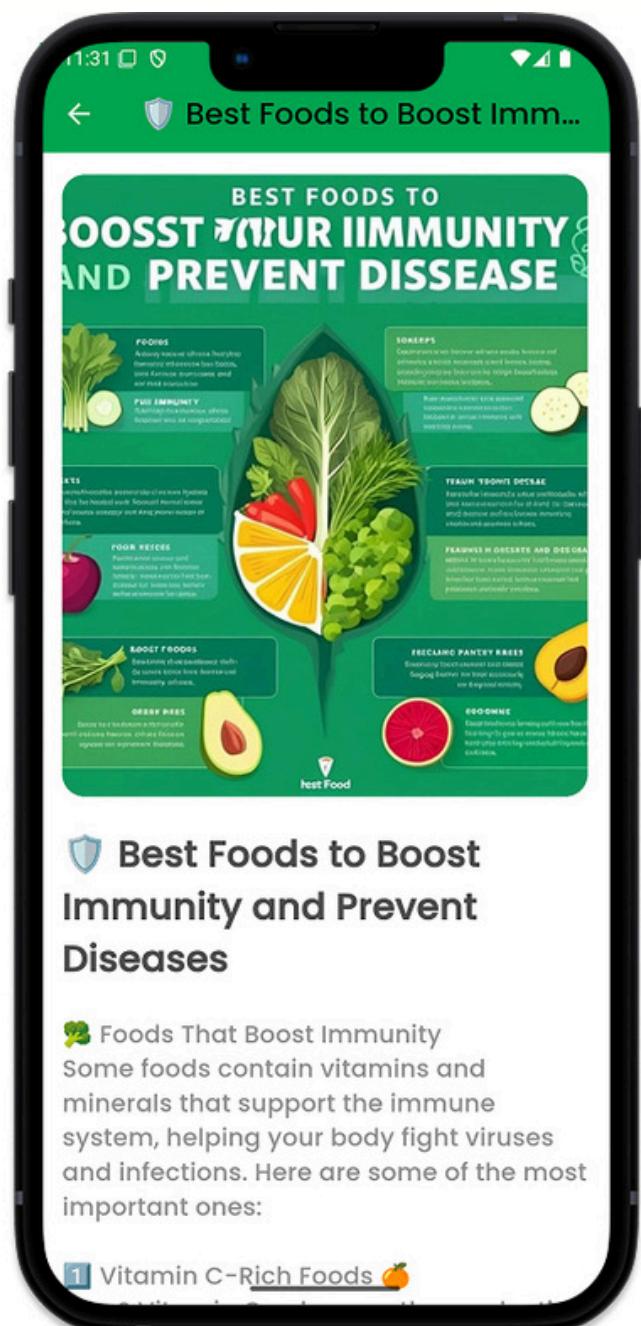
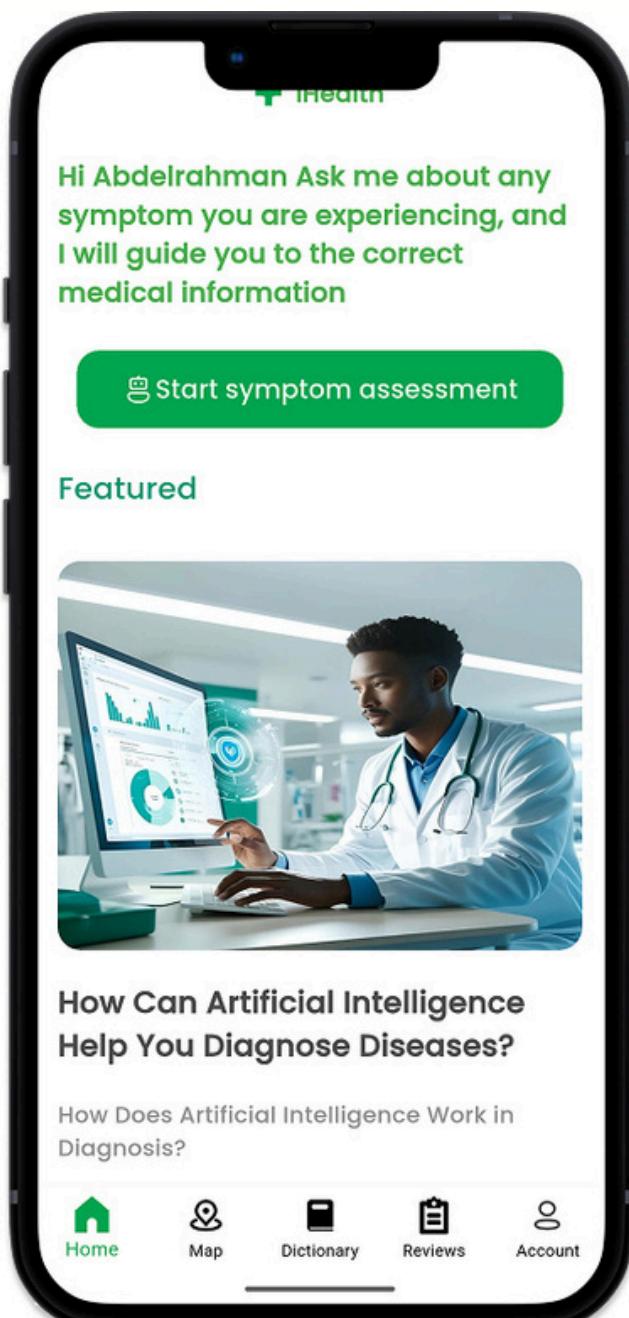
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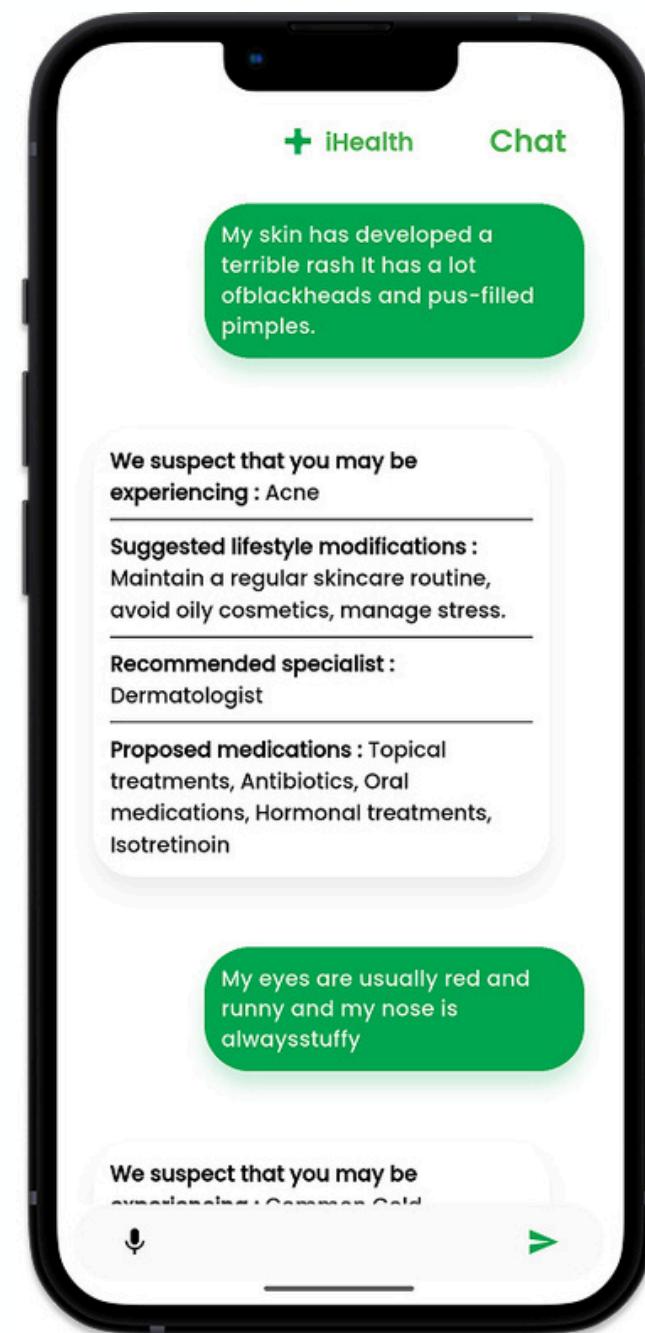
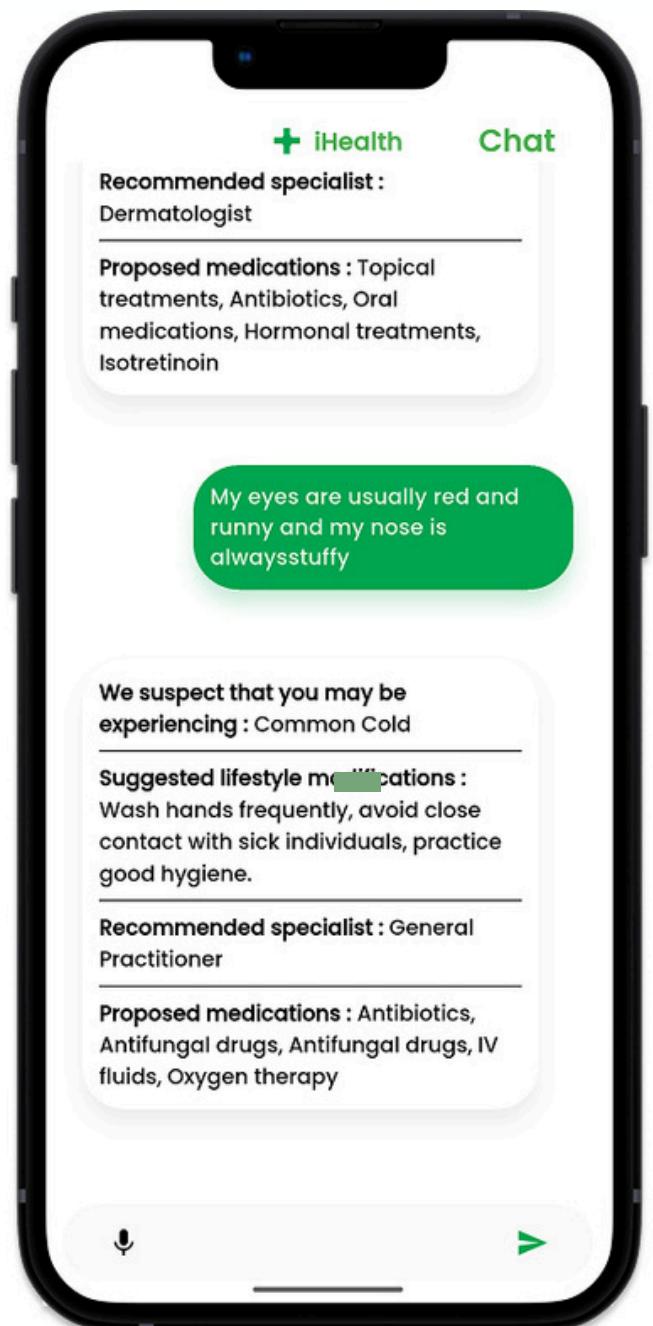
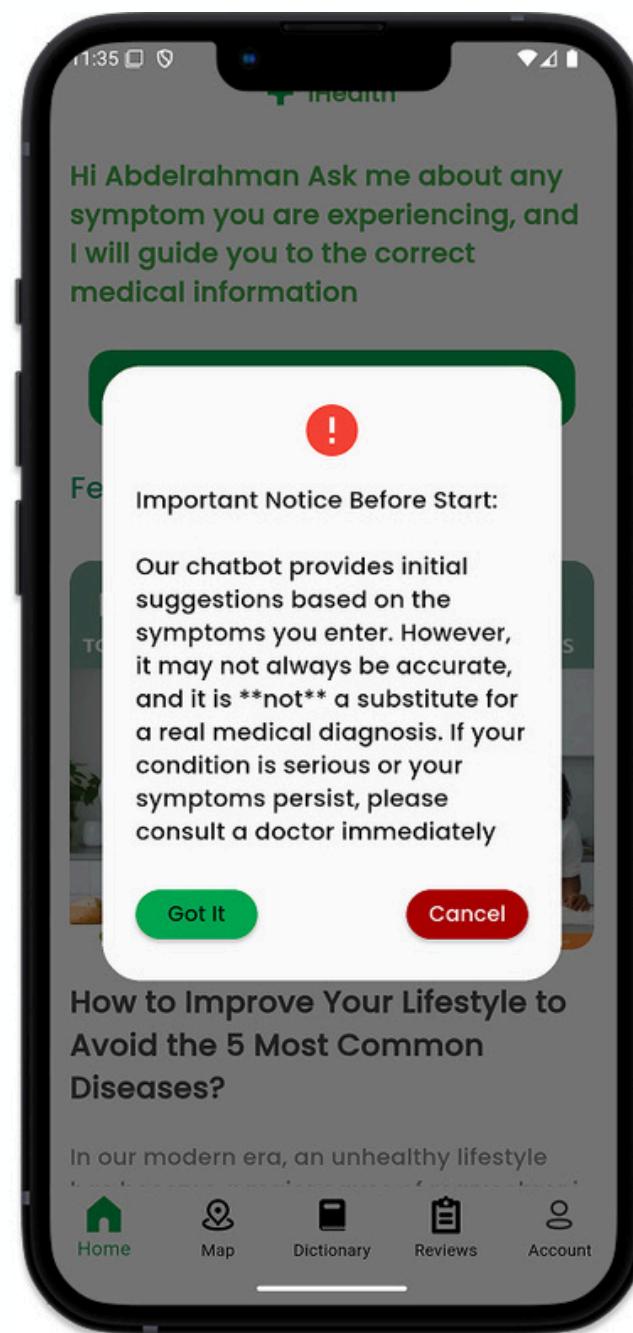
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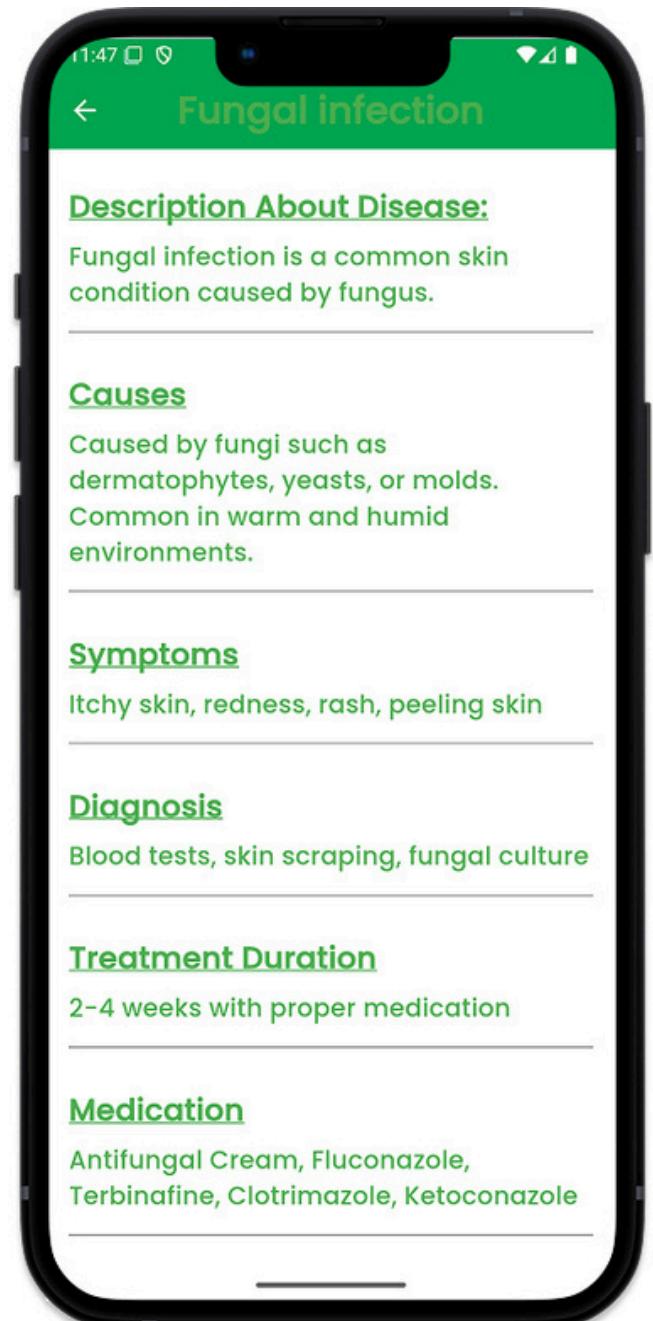
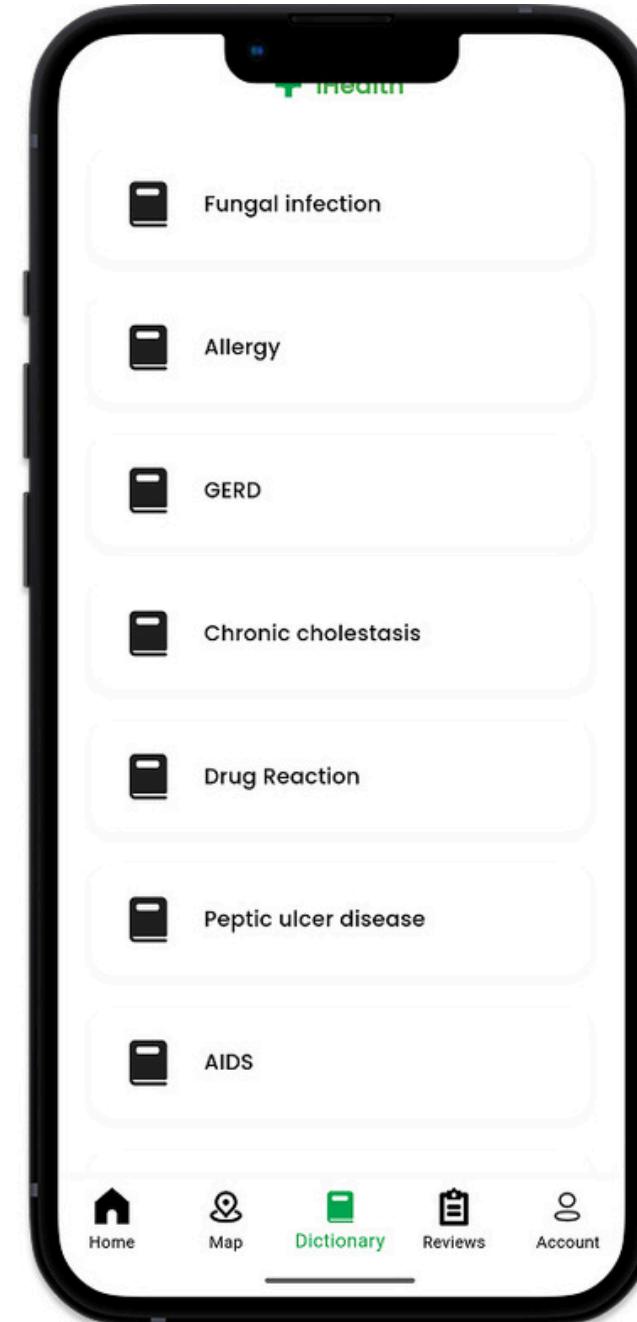
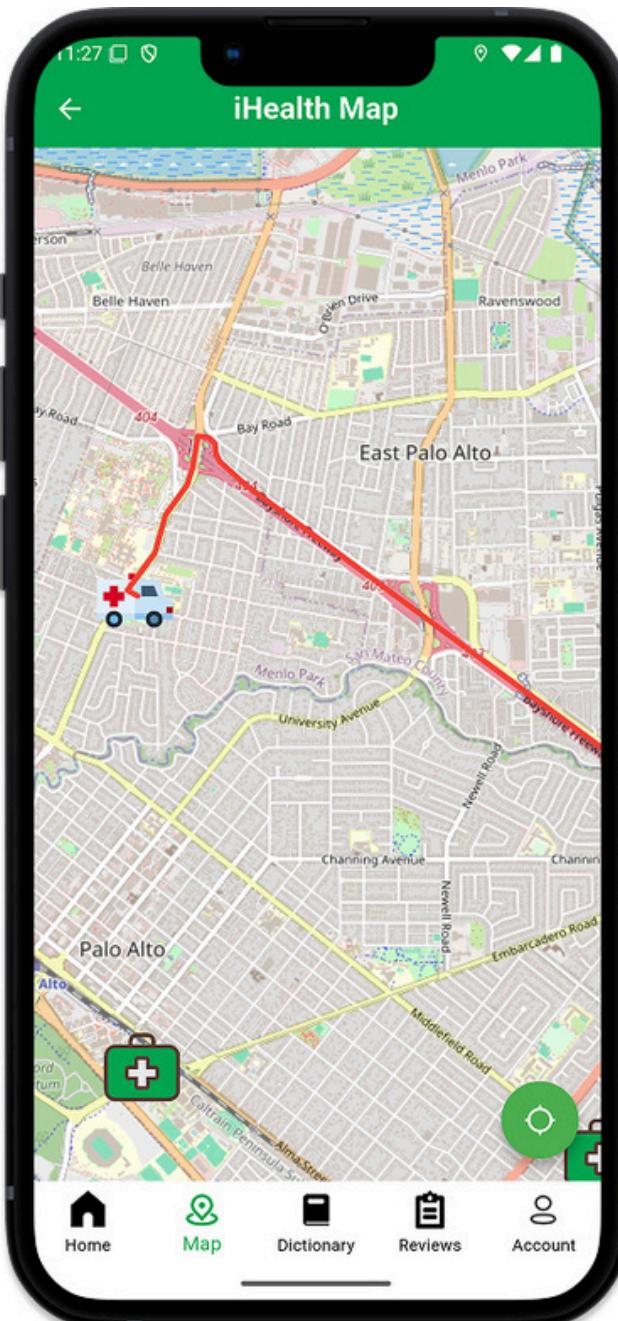
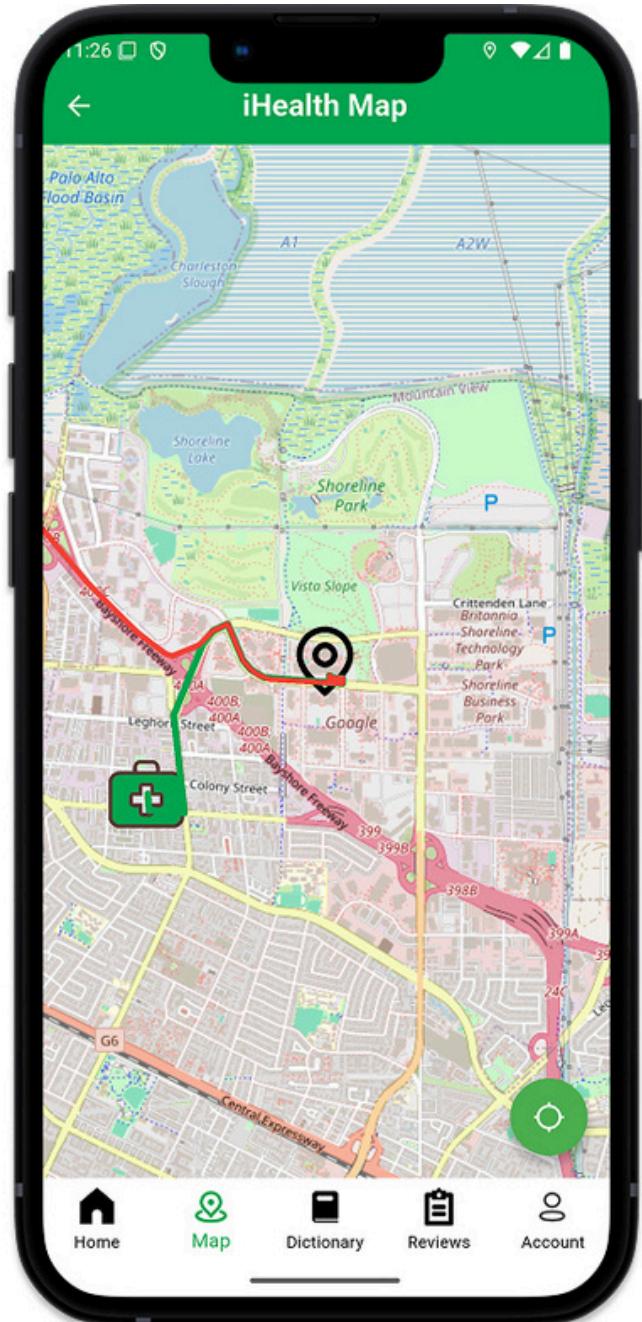
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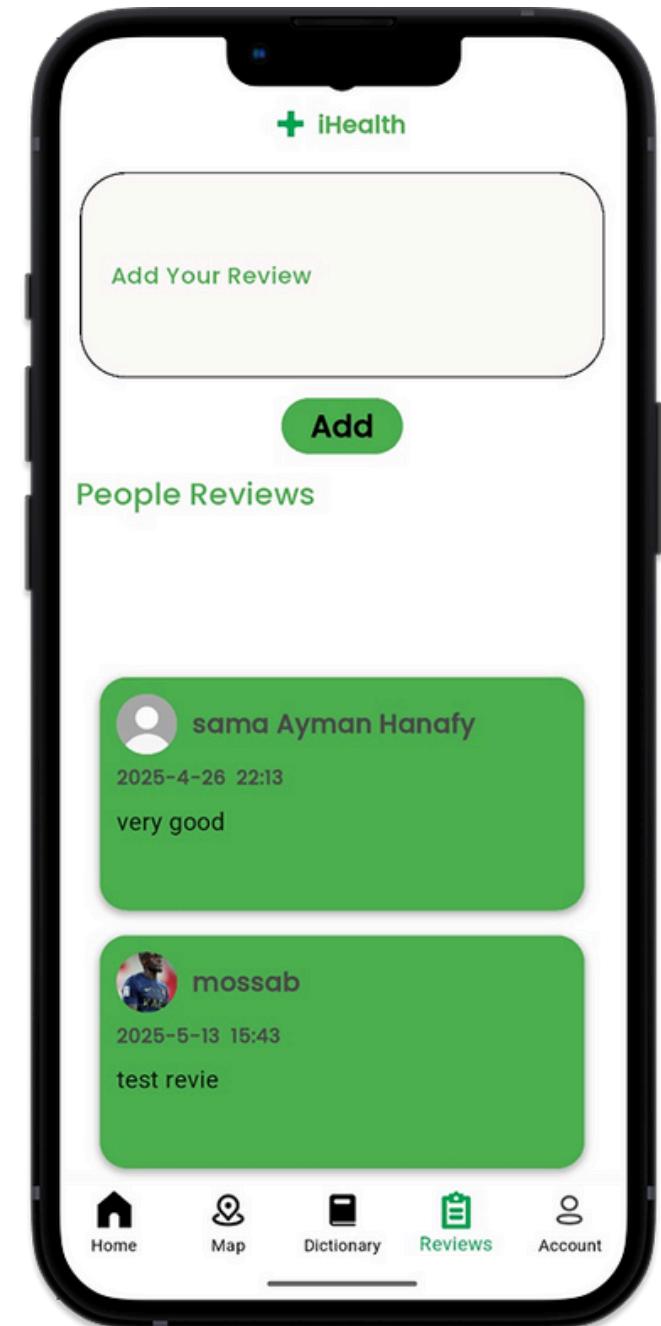
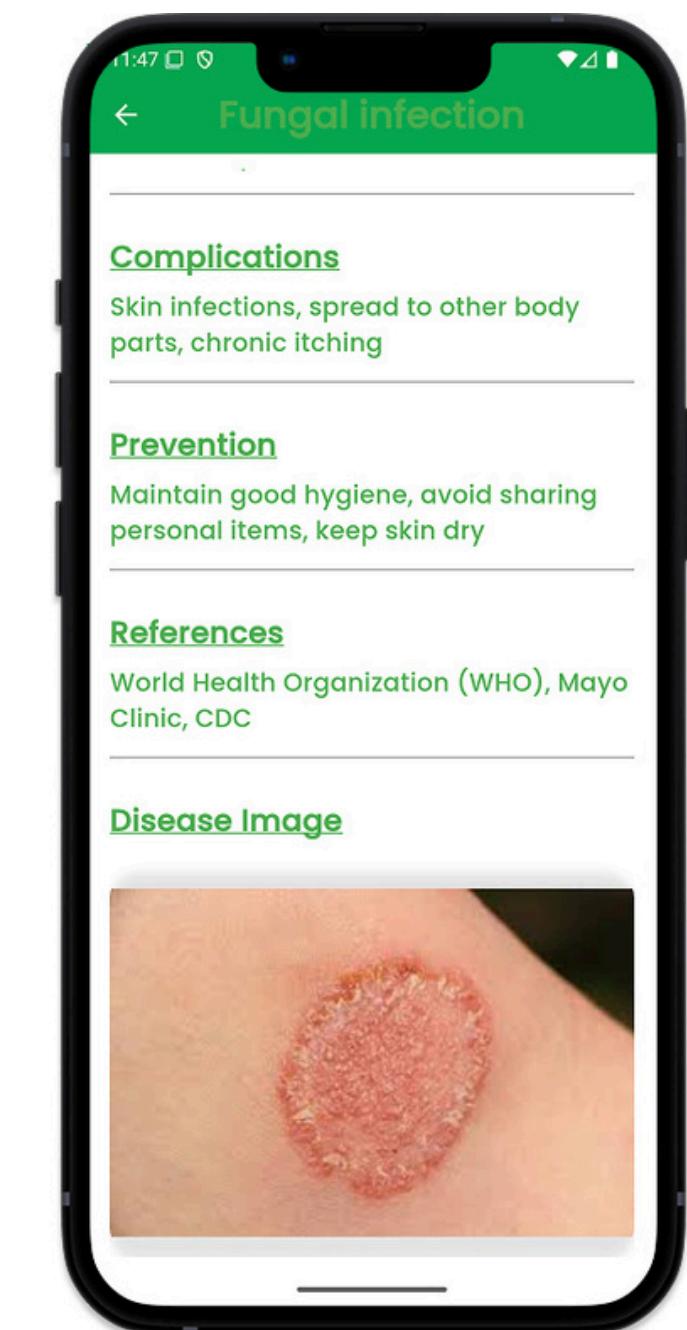
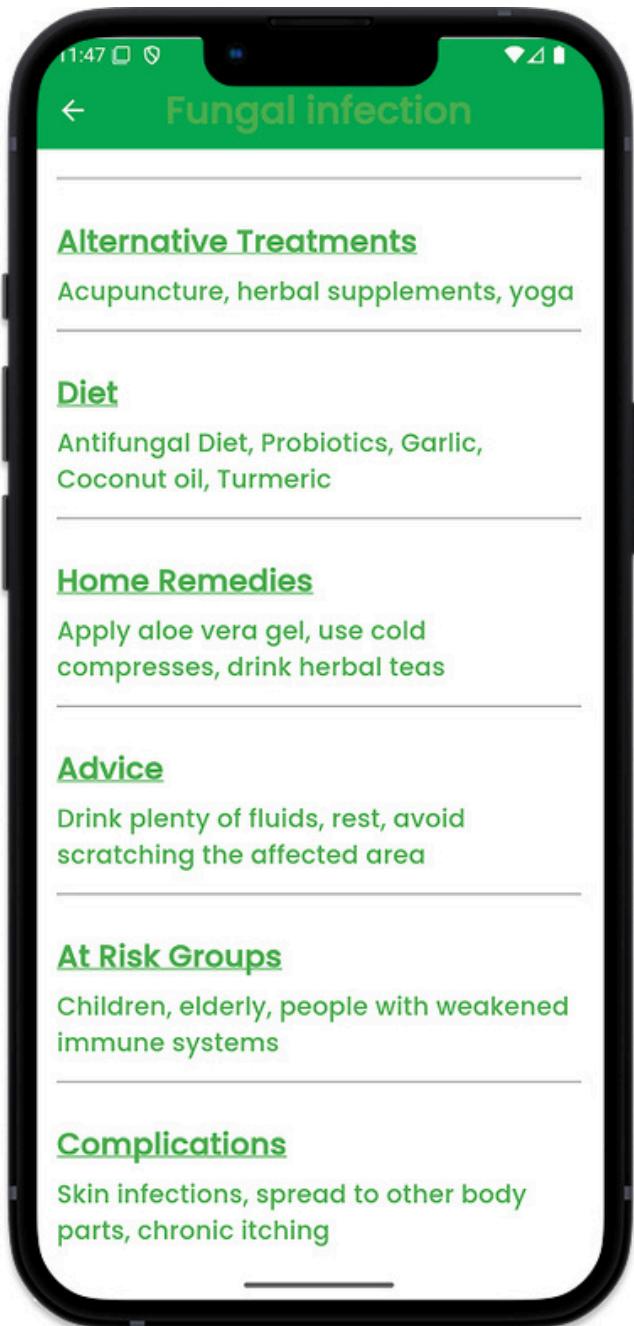
# Chatbot



# Application



# Application



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# Future work

**In the future, we will search for more Datasets in order to collect different and larger number of observations and utilize advanced Deep learning techniques to help us effectively in the natural language processing part to make chatbot even more interactive with the patient.**

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# REFERENCES

## • Papers

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[HL=EN&AS\\_SDTP=0%2C5&Q=MULTILINGUAL+HEALTHCARE+CHATBOT+USING+MACHINE+LEARNING&BTN\\_G=](https://scholar.google.com/scholar?hl=en&as_sdtp=0%2C5&q=MULTILINGUAL+HEALTHCARE+CHATBOT+USING+MACHINE+LEARNING&btnG=)

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[HTTPS://IJCRT.ORG/PAPERS/IJCRT\\_H020011.PDF](https://ijcrt.org/papers/IJCRT_H020011.PDF)

## Dataset

[4] Symptom2Disease Dataset

[https://www.kaggle.com/datasets/niyarrbarman/  
symptom2disease](https://www.kaggle.com/datasets/niyarrbarman/symptom2disease)

## Dart & Flutter

[6] Dart docs <https://dart.dev/guides>

[7] Flutter docs <https://docs.flutter.dev/>

## Selected book for ML & NLP

[5] Hands-On Machine Learning with Scikit-Learn,  
Keras, and TensorFlow. (2nd ed.). Geron Aurelien.  
(2019). O'Reilly Media, Inc.

## Selected book for Flutter

[8] Hands-On Flutter with Frank Zammetti, in  
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