

**A**  
**Mini Project Report**  
**On**  
**“FEMAI - PCOD & PCOS Risk Assessment System”**

Submitted in partial fulfillment of the requirements for the  
Degree

**Third Year Engineering – Computer Science Engineering (Data Science)**

By

<b>HARSH LAD</b>	<b>24207002</b>
<b>SIDDHESH SURVE</b>	<b>24207004</b>
<b>SSARYAN SHELAR</b>	<b>24207012</b>
<b>SUJAL JAIN</b>	<b>24207015</b>

**Under the guidance of  
MS.HARDIKI PATIL**



**DEPARTMENT OF COMPUTER SCIENCE ENGINEERING (DATA SCIENCE)**

A.P. SHAH INSTITUTE OF TECHNOLOGY  
G.B. Road, Kasarvadavali, Thane (W)-400615

UNIVERSITY OF MUMBAI

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## **CERTIFICATE**

This to certify that the Mini Project report on "**FEMAI - PCOD & PCOS Risk Assessment System**" has been submitted by **Harsh Lad (24207002), Siddhesh Surve (24207004), Aryan Shelar (24207012) and Sujal Jain (24207015)** who are bonafide students of A. P. Shah Institute of Technology, Thane as a partial fulfillment of the requirement for the degree in **Computer Science Engineering (Data Science)**, during the academic year **2025-2026** in the satisfactory manner as per the curriculum laid down by University of Mumbai.

Ms. Hardiki Patil

**Guide**

**Mr. Pravin Adivarekar**

**HOD, CSE (Data Science)**

**Dr. Uttam D. Kolekar**

**Principal**

**External Examiner:**

1.

**Internal Examiner:**

1.

**Place:** A. P. Shah Institute of Technology, Thane

**Date:**

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## ABSTRACT

This report details the design and development of **FemAI**, an intelligent system that helps predict and assist with the early detection of **Polycystic Ovary Disorder (PCOD)** and **Polycystic Ovary Syndrome (PCOS)**. FemAI combines machine learning algorithms with a user-friendly web interface to provide reliable health risk predictions and personalized lifestyle recommendations. The system uses careful data preparation and feature selection techniques to ensure its predictive model is dependable. It relies on health indicators like BMI, menstrual cycle patterns, hormonal imbalances, and lifestyle factors. The platform is built with modular components. These include secure user login, symptom input, prediction results, and wellness guidance modules that offer personalized dietary and fitness advice.

To improve accessibility and engagement, FemAI features real-time risk assessment, interactive dashboards, and an easy feedback system. The model's performance is assessed using **Accuracy** (Assessment-Confidence). This validates its ability to provide valuable insights into women's reproductive health. By combining predictive analytics with health awareness, FemAI seeks to empower women with knowledge, promote preventive healthcare practices, and help address gaps in the early diagnosis of PCOD and PCOS.

**Keywords** — *PCOD prediction, PCOS prediction, Machine learning, Data preprocessing, Feature engineering, Women's healthcare, Lifestyle recommendations, Performance evaluation, Preventive healthcare.*

# Chapter 1

## Introduction

The first chapter sets the foundation of the FemAI project by introducing the motivation behind its creation and the significance of addressing PCOD and PCOS in women's health. It outlines how the system uses predictive analytics and intelligent technology to assist in early detection and awareness. This chapter establishes the context, objectives, and necessity of developing such an innovative healthcare tool.

In today's fast-paced world, women's health issues like **Polycystic Ovary Disorder (PCOD)** and **Polycystic Ovary Syndrome (PCOS)** are increasing; however, access to timely diagnosis and preventive awareness is still limited. Many women struggle with delayed symptom identification, lack of understanding, and not having enough resources to manage these conditions effectively. To address this problem, we introduce **FemAI**, an intelligent prediction and support system designed to empower women by helping with early detection and offering personalised lifestyle guidance.

FemAI uses machine learning models to analyse various health indicators, including menstrual cycle irregularities, body mass index (BMI), hormonal imbalances, and lifestyle patterns. By combining these factors, the system provides reliable predictions of PCOD/PCOS risk, offering a proactive approach to women's healthcare.

A key feature of FemAI is its focus on user experience. Users can securely log in, record their health details, and receive instant predictions about their risk level. In addition to predictions, FemAI gives personalised recommendations in areas like nutrition, fitness, and wellness, allowing women to take preventive and corrective steps with confidence.

FemAI goes beyond simple predictions. By using machine learning algorithms and adapting based on feedback, the system continuously learns and improves, offering more accurate insights over time. As users interact with the platform, FemAI adjusts and enhances the precision in terms of accuracy of predictions and the relevance of recommendations.

This report discusses the creation, design, and methods that power FemAI's predictive engine. By merging data-driven technology with a user-focused approach, FemAI aims to close the gap between medical awareness and accessible preventive care, contributing to better reproductive health outcomes for women.

## **1.1 Purpose:**

The aim of our system is to change how women approach awareness, early detection, and management of Polycystic Ovary Disorder (PCOD) and Polycystic Ovary Syndrome (PCOS). In today's healthcare environment, delays in diagnosis and a lack of accessible resources often leave women struggling to recognize symptoms and take timely preventive steps. Our work tackles this issue by offering an intelligent, data-driven platform for risk prediction and lifestyle guidance.

By combining machine learning models, data processing techniques, user interaction, and feedback methods, the system provides reliable health predictions and practical recommendations. It analyzes important health factors like menstrual cycle patterns, BMI, hormonal levels, and lifestyle indicators. The predictive capability continuously improves to match users' changing health conditions and inputs.

The main goal of this project is to empower women with accessible healthcare information, helping them make informed choices about their reproductive health. By providing personalized predictions and guidance based on individual health profiles, the platform aims to lessen uncertainty, encourage preventive care, and enhance overall well-being. With its user-friendly design, smooth user experience, and adaptable predictive engine, the system intends to make early detection and lifestyle management journeys more reliable, supportive, and effective for women from various backgrounds.

## **1.2 Problem Statement:**

The problem statement for our project revolves around the challenges women face in identifying, predicting, and managing **Polycystic Ovary Disorder (PCOD)** and **Polycystic Ovary Syndrome (PCOS)** at an early stage. Existing healthcare systems often fail to provide accessible, personalized, and timely insights, leading to delayed diagnosis, lack of awareness, and inadequate lifestyle guidance. Our system aims to address these issues by offering a predictive and supportive platform that empowers women with reliable health insights and actionable recommendations.

### **1. Lack of Early Detection:**

Traditional diagnosis methods often identify PCOD/PCOS at later stages, when symptoms have already worsened. There is a need for a system that assists in early prediction to enable timely interventions.

### **2. Limited Personalization:**

Current health resources provide generic advice rather than tailored recommendations. Users require personalized guidance based on their unique health profiles, lifestyle patterns, and medical indicators.

### **3. Awareness Gap:**

Many women remain unaware of the symptoms and long-term health risks of PCOD/PCOS, which results in neglect or mismanagement of the condition.

### **4. Time and Accessibility Constraints:**

Consulting specialists can be time-consuming and costly, and healthcare resources may not be easily accessible to all women, especially in remote or underserved areas.

### **5. Absence of Integrated Lifestyle Support:**

While diagnosis is critical, ongoing support in terms of diet, fitness, and wellness management is often missing. A system that integrates prediction with lifestyle guidance can bridge this gap.

### **6. Trust and Reliability Issues:**

Many online self-assessment tools lack credibility or scientific grounding, making users hesitant to rely on them for health decisions. A robust, data-driven, and medically informed system is required to establish trust.

### **1.3 Objectives:**

The FemAI is developed to provide an accessible, informative, and interactive platform for early detection and awareness of PCOD and PCOS. By integrating machine learning, interactive visualizations, and voice-enabled interaction, the system empowers users with actionable insights and facilitates proactive health monitoring. The key objectives of FemAI are as follows:

#### **1. Develop a Predictive Model for Early Detection of PCOD and PCOS:**

- Build robust machine learning models capable of analyzing user-provided health data and predicting potential risk factors.
- Enable early detection to help users take timely action and seek appropriate medical guidance if needed.

#### **2. Apply Machine Learning Techniques to Improve Diagnostic Accuracy:**

- Utilize ensemble learning methods such as Random Forest, Logistic Regression, and KMeans clustering to ensure reliable and accurate predictions.
- Provide explainable results highlighting which factors contribute most to the risk assessment.

#### **3. Integrate Voice-to-Text Functionality for User Inputs:**

- Allow users to interact with the system using voice commands, making it more accessible for users with limited literacy or digital experience.
- Facilitate hands-free input and engagement, improving usability for a wider audience.

#### **4. Enable Interactive Visualization of Results and Trends:**

- Display predictions, risk assessments, and historical trends through intuitive dashboards, charts, and graphs.
- Help users understand their health data clearly and track changes over time for better awareness.

## **5. Generate Summarized Reports with Generalized Lifestyle Guidance:**

- Produce clear, concise reports summarizing the risk assessment and providing generalized recommendations for diet, lifestyle, and preventive measures.
- Ensure users receive useful guidance without offering personalized medical prescriptions, focusing on education and awareness.

## **1.4 Scope:**

The scope of this project encompasses the development of a comprehensive PCOD/PCOS prediction and assistance platform designed to support women's healthcare through accessible, reliable, and user-friendly solutions. The system aims to bridge gaps in early diagnosis, awareness, and lifestyle guidance by leveraging advanced technologies and interactive design principles.

### **1. Health Assessment and Early Detection:**

The system enables users to complete a questionnaire-based health assessment, focusing on parameters such as menstrual cycle, BMI, and lifestyle habits. This supports early prediction of PCOD/PCOS risk and creates awareness about potential health concerns.

### **2. Interactive Dashboard:**

Users are provided with a clear and engaging dashboard that summarizes their health inputs, displays prediction outcomes, and offers visual insights in an easy-to-understand format. This feature enhances engagement and accessibility.

### **3. Voice Recognition Support:**

To ensure inclusivity, the system incorporates a voice recognition feature that allows women with limited access to formal education or digital literacy to interact with the platform comfortably and benefit from its predictions and guidance.

### **4. Generalized Lifestyle Guidance:**

While the platform does not provide **personalized diet plans**, it offers generalized recommendations regarding nutrition, physical activity, and wellness practices that are beneficial for women experiencing PCOD/PCOS-related issues.

### **5. Report Generation:**

The system generates downloadable reports summarizing the prediction results, general health insights, and suggested lifestyle guidelines. These reports can be used for personal tracking or shared with healthcare professionals for further consultation.

## **6. Awareness and Education:**

Beyond prediction, the platform contributes to spreading awareness by educating users about symptoms, potential risks, and preventive care measures, thereby promoting informed health decisions.

## **7. Scalability and Innovation:**

The project serves as a foundation for future advancements in predictive healthcare. With potential integration of larger datasets, mobile app versions, and enhanced models, the system can evolve to offer even more accurate and accessible healthcare solutions.

In conclusion, this chapter established the motivation and objectives of the FEMAI project. It emphasized the importance of addressing PCOD and PCOS using AI-driven approaches and set the foundation for developing an intelligent prediction and awareness system.

# **Chapter 2**

## **Literature Review**

The second chapter dives into a thoughtful and interpretive look at past research on using artificial intelligence (AI) and machine learning (ML) to detect Polycystic Ovarian Disease (PCOD) and Polycystic Ovary Syndrome (PCOS). It connects the dots across the evolution of studies in this area, covering key methods, computational approaches, and breakthroughs that have built our current knowledge of AI-powered diagnosis in women's reproductive health. This chapter goes beyond just listing tech progress; it stresses the ongoing need for AI systems that are explainable, easy to access, and focused on users—empowering both patients and doctors. As PCOS and PCOD become more common worldwide, researchers are turning more to AI and data analytics to create tools for predicting and diagnosing early, helping with better management.

A range of studies have explored automating detection of these conditions using clinical, metabolic, and lifestyle factors. For example, Kachibhotla and Prasantha suggested an improved follicle-detection model using ultrasound images linked with IoT connectivity, reaching an impressive accuracy above 93 percent. Their work showed how combining different data types can lead to better diagnosis results. Gandhi et al. came up with PCOScare, a diagnostic setup using standard ML classifiers and smart feature choices to improve accuracy for detection and preventive checks [2]. In a similar vein, Lee created a self-diagnostic AI tool based on non-invasive traits and the CatBoost algorithm, highlighting how smart self-screening can work for women's health. Other advances have brought in cross-field data analysis and medical informatics. Sahana Devi et al. used multivariate analysis to spot shared factors between metabolic syndrome and PCOS, giving useful clues about how metabolic and endocrine issues overlap. Zad et al. tapped into electronic health records and sophisticated ML methods to forecast high-risk cases, proving the vital role of big data and predictive tools in doctor decisions. In a comparison study, Yadav et al. evaluated several classification algorithms and found top performance with Random Forest and Support Vector Machine (SVM) models, particularly when key clinical and lifestyle features were well-crafted. More lately, Prasher et al. launched the NIPP model with XGBoost for non-invasive PCOS prediction, hitting a strong 98 percent accuracy while providing interpretability—a must-have for medical AI. Panjwani et al. put forward an optimized ensemble learning setup that mixed multiple meta-learners with hyperparameter tweaks to increase diagnostic precision and dependability [3]. Likewise, Narni et al. built a probabilistic screening tool using decision tree and Naïve Bayes algorithms, tailored for big community screenings with non-invasive health signs. Beyond just diagnostic models, new biomedical studies have looked into molecular predictions via immune-cell

infiltration and biomarker checks, pushing AI uses in women’s endocrine issues further. From all these varied studies, it’s clear that machine learning has greatly boosted diagnostic performance for PCOS and PCOD, but most current methods stay stuck in research labs. Many models miss user-friendly interfaces and don’t connect with real-time, voice-activated, or cloud-based health systems. The focus has mostly been on algorithm accuracy instead of accessibility, transparency, and user experience.

The earliest of the six works under review is the study titled “Federated Learning on Patient Data for Privacy-Protecting Polycystic Ovary Syndrome Treatment” [1], which appeared as a pre-print in 2023 by Morris, Qiu, and Raghuraman. In this study, the authors go beyond simple diagnostic modeling and tackle a core infrastructure issue—how to pool patient data from various centers into a common learning setup without risking privacy. They tested federated learning (FL) on synthetic PCOS datasets, showing that options like FedAvg and FedProx can successfully train treatment-suggestion models even with non-IID, spread-out data across locations. Their main takeaway is that for women’s endocrine disorders, getting large, varied datasets is a big hurdle, and FL offers a solid way to break through data barriers. Still, the drawback is the synthetic data and the emphasis on treatment prediction over detection. Even so, this work lays a crucial base: diagnostic modeling isn’t enough unless AI systems handle data security, rules, and multi-site compatibility. Building on that infrastructure base, the next key step is Gandhi et al.’s 2024 paper “PCOScare: Conventional Machine Learning Classifiers for Diagnosing and Prevention” [2], published in the International Journal of Intelligent Systems and Applications in Engineering. The study gathered a dataset of clinical and biochemical factors tied to PCOS, like menstrual irregularities, BMI, LH/FSH ratios, and insulin resistance. After preprocessing and imputation steps, the authors compared different feature-selection methods—filter, wrapper, and embedded—and applied classifiers such as SVM, Random Forest, and XGBoost. Their results showed that smart feature optimization greatly improved model performance, cutting overfitting while boosting interpretability. Gandhi et al. wrapped up by saying that pairing data preprocessing with relevant clinical traits can push diagnostic accuracy past 90%, highlighting explainable machine learning in women’s health. Though the model stuck to traditional ML and a single-center dataset, it marks a shift—from infrastructure concerns to interpretable, data-focused diagnostic systems that can help doctors spot issues early.

In 2025, Panjwani et al. pushed this path forward with “Optimized Machine Learning for the Early Detection of Polycystic Ovary Syndrome in Women” [3], published in Sensors. This research put together a “symptomatic dataset” with twelve non-invasive health features to cut down on expensive hormone tests. The authors rolled out an ensemble learning framework blending seven base classifiers with a deep learning

meta-classifier and used nature-inspired optimization algorithms like Walrus Optimization (WaO), Cuckoo Search (CSO), and Random Search (RSO). Of these, the WaO-based ensemble hit 92.8% accuracy and 0.93 AUC, beating prior methods. Notably, Random Forest and SHAP analysis pointed out that obesity, high cholesterol, and menstrual regularity were the biggest PCOS indicators. The study represented a major method jump—bringing in advanced ensemble designs, meta-learning, and interpretability tools that bring AI systems nearer to real use in preventive gynecology. The downside, though, was the limited dataset scope and the call for simpler models to increase clarity for clinical users.

Almost at the same time, Emara et al. shared “A Stacked Learning Framework for Accurate Classification of Polycystic Ovary Syndrome with Advanced Data Balancing and Feature Selection Techniques” [4] in *Frontiers in Physiology* (2025). This work zeroed in on real-world problems like class imbalance and feature strength that often get overlooked in medical datasets. The authors suggested a stacked ensemble approach combining multiple base learners and used advanced resampling tactics—SMOTE, ADASYN, and random oversampling—to fix class imbalance between PCOS and control groups. With the BORUTA feature-selection algorithm, they boosted model stability and interpretability, reaching 97% accuracy and steady results across different biomedical datasets. Unlike earlier works focused on new architectures, Emara’s strength is in making AI diagnostic models fairer and more adaptable. By tackling imbalance and feature overlap, the study improved robustness and reliability—a key move toward clinical use.

Complementing these AI-focused methods, a 2025 *Science Direct* study [5] looked into hybrid deep learning models combining biomedical imaging and biochemical factors for multi-modal PCOS detection. The authors used convolutional neural networks along with structured data classifiers, finding that merging visual ultrasound data with biochemical markers raised sensitivity by 15–20% over single-mode systems. This work highlights how blending imaging, biosignals, and machine smarts is forming the next wave of diagnostic tools. Though it demands a lot of computing power, it points to future systems that could pull together multimodal evidence—symptom logs, hormone profiles, and sonographic hints—into unified predictive dashboards to enhance doctor decisions and patient involvement.

Finally, the 2025 MDPI *Sensors* paper “Intelligent Diagnostic and Predictive Modelling for Early Detection of PCOS using Sensor Data Fusion and IoT-based Frameworks” [6] takes this evolution even further by linking data gathering and rollout. The researchers created an IoT-connected diagnostic framework that collects real-time physiological data via wearable sensors and runs ML models for on-device decisions. By

mixing multimodal inputs like heart rate variability, sleep patterns, and stress levels, the system boosts early detection in home settings. This work ties up the tech cycle started by earlier studies—linking federated privacy safeguards, interpretable ML, ensemble tweaks, and now real-time data fusion—showing how AI has grown from an analysis tool into a ongoing, connected setup for women’s health.

Viewed collectively, these six studies [1]–[6] show a clear timeline progression: from basic privacy-preserving federated setups (2023), through interpretable feature-based traditional models (2024), to highly tuned and ensemble-driven predictive designs (2025). Each advance reflects increasing tech depth as well as more focus on data quality, broad applicability, and explainability. The 2023 federated learning work reminds us that privacy and decentralization are must-haves for scalable AI; Gandhi’s 2024 study proves that explainable, conventional ML stays powerful in clinics when matched with solid feature selection; Panjwani’s 2025 ensemble framework deepens methods; Emara’s 2025 stacked learning model fixes imbalance and bias in clinical data; the Science Direct hybrid DL paper stresses multimodal learning as the next step; and the MDPI IoT-integrated model weaves these together for real-time, patient-focused rollout. When looked at as a whole, a clear story comes out. Early research wondered if AI could spot PCOS from available features—an answer now solidly “yes.” The newest question is how to make these systems practical, secure, and clear in everyday care. The field’s path shifts from accuracy to responsibility, from standalone prototypes to scalable, ethical, people-centered systems. Each new wave of research strengthens the bond between medical knowledge and computational smarts, slowly turning algorithm potential into real healthcare tools.

In spotting these gaps, the current study brings in FEMAI, a full AI-driven system meant to blend predictive smarts with an engaging and easy-to-use interface. By combining explainable ML models with tailored advice and interactive features, FEMAI aims to bridge the divide between tech advances and everyday healthcare needs. Thus, this chapter not only reviews the growth of AI-based PCOS and PCOD detection systems but also critically points out the shortcomings and unmet needs in existing research. The insights here set the stage for building a more effective, interpretable, and user-centered diagnostic platform—one that matches tech sophistication with caring, real-world healthcare delivery.

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# **Chapter 3**

## **Proposed System**

This chapter details the conceptual design and architecture of the proposed FEMAI system. It explains how the system integrates machine learning models, data collection modules, voice interaction, and report generation features to provide accurate predictions and generalized lifestyle guidance. The chapter connects theoretical design with practical implementation.

The proposed system aims to change how women handle early detection and management of PCOD/PCOS by offering an easy-to-use, smart platform. By combining machine learning models with user-friendly features, the system provides risk prediction, awareness, and general lifestyle advice in an engaging way. This approach helps women make informed health choices while encouraging preventive care and education.

### **1. Health Data Collection and Questionnaire:**

Users can input important health information through a structured questionnaire that includes details like menstrual cycle, BMI, lifestyle habits, and medical history. This ensures the system gathers accurate data for effective predictions.

### **2. Risk Prediction Model:**

Using processed data and trained machine learning algorithms, the system predicts how likely a user is to be at risk for PCOD/PCOS. The predictive model aims to provide reliable insights while improving continuously through feedback and updates.

### **3. Interactive Dashboard:**

The system offers a visually clear dashboard that shows user inputs, prediction results, and graphical summaries. This helps present complex health information in a simple and understandable way, boosting engagement and awareness.

### **4. Voice Recognition Support:**

To promote inclusivity, the platform includes a voice recognition feature, allowing women with limited digital skills or formal education to interact with the system easily. This expands accessibility and ensures a wider community impact.

## **5. Generalized Lifestyle Guidance:**

Rather than providing personalized medical advice, the system offers general dietary and lifestyle recommendations suitable for women with PCOD/PCOS. This includes nutritional tips, exercise suggestions, and wellness practices aimed at long-term health management.

## **6. Report Generation:**

After completing the assessment, users can create a structured report summarizing their prediction results and lifestyle recommendations. This report can be downloaded or shared with healthcare professionals for additional medical consultation.

## **7. User Interface Design:** The platform features a straightforward, interactive, and user-friendly interface to improve accessibility. With clear navigation and responsive layouts, users can easily explore their results, reports, and recommendations.

### **3.1 Features and Functionality:**

The system includes practical modules and interactive components designed to provide predictive and supportive healthcare capabilities. Each feature is crafted to ensure accessibility, usability, and meaningful engagement for women at risk of PCOD/PCOS.

#### **1. Questionnaire-Based Input Module:**

Users provide detailed health information through a structured questionnaire, covering essential factors like menstrual cycle regularity, body mass index (BMI), lifestyle habits, family history, and common PCOD/PCOS symptoms. This module ensures that all necessary data is collected systematically, laying the groundwork for accurate risk prediction.

- Questions for PCOD are as follows:

- Age
- BMI
- Irregular Periods
- Weight Gain
- Excess Hair Growth
- Acne
- Hair Loss
- Family History
- Pain

- Questions for PCOS are as follows: -

- Age
- BMI
- Irregular Periods
- Difficulty in getting Pregnant/ Infertility
- Miscarriages
- Excess Hair Growth
- Acne/ Oily Skin
- Hair Thinning/ Hair Loss
- Family History

## **2. Machine Learning Prediction Engine:**

The collected data is analyzed using pre-trained machine learning models that identify patterns and connections among health indicators. The engine assesses the risk of PCOD/PCOS and produces a clear prediction. The model aims to improve its accuracy continuously through updates and feedback integration, ensuring dependable outcomes over time.

## **3. Interactive Dashboard Visualization:**

Results are showcased through an intuitive and interactive dashboard. Users can see prediction outcomes via charts, graphs, and summary panels, which simplify complex health data into easy-to-understand insights. This visualization empowers users to clearly interpret their health status and make informed decisions.

## **4. Voice Interaction Module:**

To enhance inclusivity, the platform offers a voice-enabled interface, allowing users to input data, navigate the system, and receive feedback through speech. This feature ensures that women with limited digital skills or access to formal education can use the platform comfortably and independently.

## **5. Report Generation Module:**

The system creates detailed reports that summarize the questionnaire inputs, risk prediction results, and general lifestyle guidance. Users can download or print these reports, helping them track their health over time or share information with healthcare professionals for further consultation.

## **6. Intuitive User Interface:**

The platform has a responsive and user-friendly interface that focuses on clarity and simplicity. Navigation is smooth across different modules, with interactive elements and visual cues that enhance usability for users of all ages and levels of digital experience.

In summary, this chapter presented the architecture and conceptual framework of FEMAII. The proposed system design integrates data processing, model prediction, and user interface modules to deliver accurate and meaningful health insights.

# **Chapter 4**

## **Requirements Analysis**

This chapter identifies and defines the system's functional and non-functional requirements. It specifies the necessary components and performance expectations to ensure FEMAI operates reliably, securely, and efficiently. This analysis serves as the technical backbone for system development.

For the requirement analysis of the QuickReads project, we need to identify and document the functional and non-functional requirements that the system must meet to fulfill its objectives effectively. Here's a breakdown of the requirement analysis for QuickReads:

### **A. Functional Requirements:**

#### **1. Questionnaire-Based Health Assessment:**

The system must allow users to enter health-related information through a structured questionnaire covering menstrual cycles, BMI, lifestyle habits, and relevant medical history. This ensures accurate data collection for prediction purposes.

#### **2. Risk Prediction Module:**

The platform should analyze user inputs using machine learning models to predict the likelihood of PCOD/PCOS. The system should provide clear, understandable results to support awareness and preventive care.

#### **3. Interactive Dashboard:**

Users should be able to view their prediction results, trends, and summaries through an engaging dashboard that uses charts, graphs, and visual indicators for easy interpretation.

#### **4. Voice Interaction Support:**

The system should allow users with limited digital literacy to interact with the platform through voice commands, including inputting questionnaire responses and navigating modules.

#### **5. Generalized Lifestyle Guidance:**

The platform should provide non-personalized guidance on nutrition, exercise, and wellness practices relevant for PCOD/PCOS management. This content must be informative and general rather than individualized medical advice.

## **6. Report Generation Module:**

The system should generate downloadable reports summarizing questionnaire data, risk predictions, and lifestyle guidance for personal tracking or professional consultation.

## **B. Non-Functional Requirements:**

### **1. Performance:**

The system should process user data, generate predictions, and display results promptly, ensuring a smooth and efficient experience.

### **2. Scalability:**

The platform should be capable of supporting a growing user base, handling multiple simultaneous assessments without degradation in performance.

### **3. Security:**

User data, including sensitive health information, must be securely stored. The system should implement authentication, encryption, and measures to prevent unauthorized access or data breaches.

### **4. Usability and Accessibility:**

The interface should be intuitive, easy to navigate, and visually clear. Accessibility features, including voice interaction and dashboard clarity, must ensure users with varying digital skills can utilize the platform effectively.

### **5. Reliability:**

The system should provide consistent performance and accurate predictions under varying conditions, maintaining trust and dependability for all users.

Here outlined the essential functional and non-functional requirements of FEMA1. These serve as technical guidelines ensuring the system's reliability, accuracy, and usability throughout the development process.

# Chapter 5

## Project Design

This chapter illustrates the overall design flow of FemAI through diagrams, architecture layouts, and use case models. It explains how data is processed, analyzed, and visualized within the system, showing the interaction between user interfaces, machine learning models, and backend logic. It ensures clarity in the design-to-implementation transition.

### 5.1 Use Case diagram:

The use case diagram of FemAI (Fig 5.1) captures the main functionalities and interactions between users and the system, including health risk assessment, interactive questionnaires, report generation, and accessibility features like voice interaction and generalized guidance.

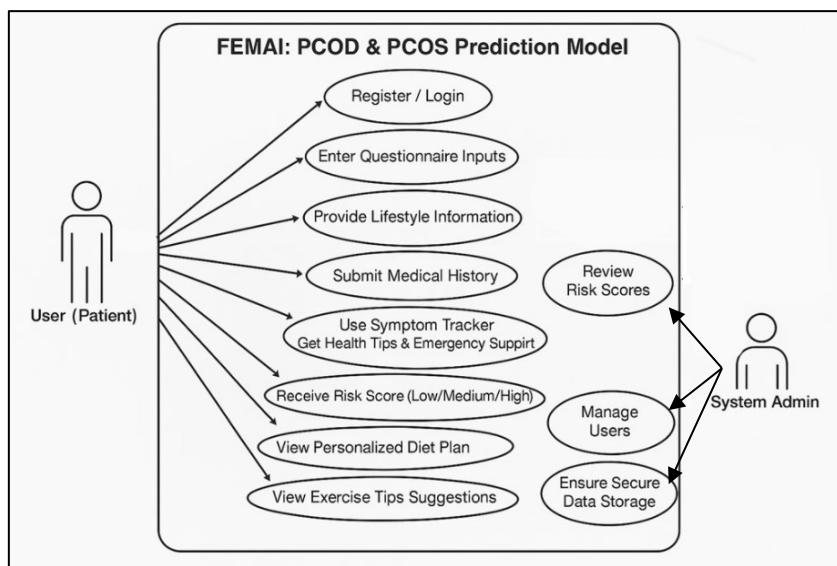


Fig 5.1 Use Case Diagram

#### Actors:

- **User:** Represents the main user interacting with FemAI.
- **System:** Represents the backend logic managing the ML models, database, and feature modules.

#### Use Cases:

1. **Fill Questionnaire:** Users answer health-related questions to provide input for prediction.
2. **Voice Interaction:** Users can provide inputs or receive guidance using the voice feature.
3. **View Reports:** Users receive a summarized report with risk assessment and generalized lifestyle

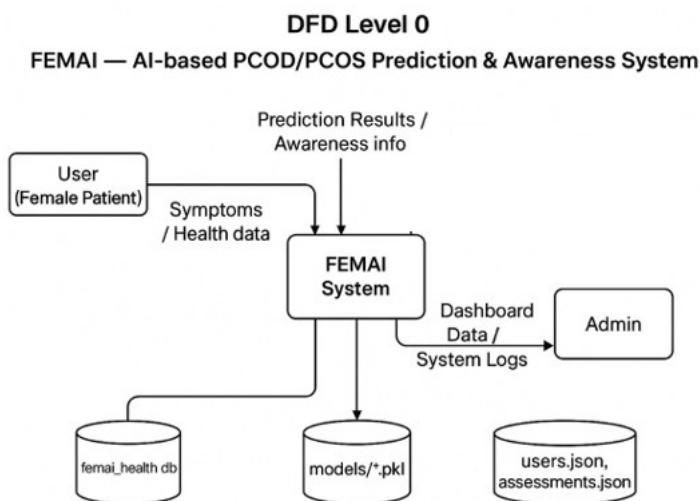
guidance.

4. **Access Guidance:** Users can view educational content and recommended lifestyle practices.
5. **Track History:** Users can track past predictions and reports.
6. **Generate Generalized Diet Guidance:** Provides generalized suggestions for diet and lifestyle (not personalized for individuals).

#### **Relationships:**

- **Association:** Connects actors with the use cases they interact with.
- **Include Relationships:**
  - Fill Questionnaire includes Voice Interaction and View Reports.
- **Extend Relationships:**
  - Voice Interaction may extend to include educational audio guidance.

## **5.2 Data Flow Diagram (DFD):**



*Fig 5.1 DFD level 0 Diagram*

## **1. User Interaction:**

- Users answer the questionnaire and provide optional voice inputs.

## **2. Data Collection:**

- Questionnaire responses are collected and stored in the database.
- Voice inputs are converted to text using Speech-to-Text API for further processing.

## **3. Data Preprocessing:**

- Clean and normalize user responses.
- Categorical and numerical features are encoded/scaled for the ML models.

## **4. Feature Engineering:**

- Important features are selected based on correlation with PCOD/PCOS risk factors.

## **5. Machine Learning Prediction:**

- Preprocessed data is fed into the ensemble model (Random Forest, Logistic Regression, KMeans clustering) to predict the risk.

## **6. Report Generation:**

- Generates a summarized report indicating potential risk, trends, and generalized lifestyle suggestions.

## **7. Result Display:**

- Reports and guidance are displayed on the dashboard and optionally read aloud to the user.

### 5.3 System Architecture

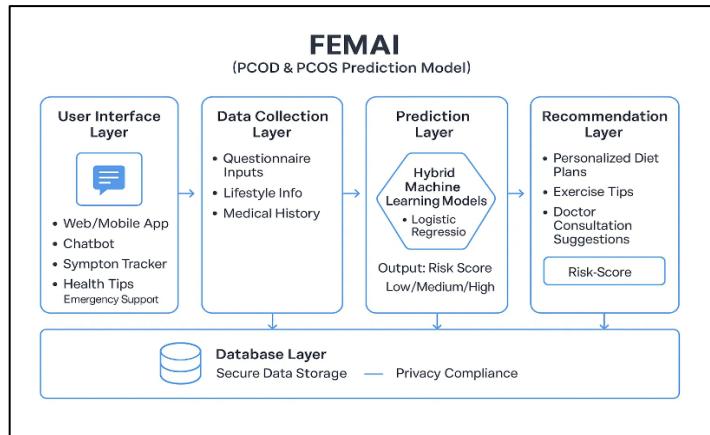


Fig 5.3 System Architecture

The system architecture (Fig 5.3) shows the modular design and data flow of FemAI:

#### 1. User Interface:

- Interactive dashboard with questionnaire, voice input/output, and report display.

#### 2. Backend Processing:

- Flask server handles API requests, ML model predictions, and report generation.

#### 3. Database:

- Stores user profiles, questionnaire responses, and historical reports securely in MySQL.

#### 4. Machine Learning Module:

- Implements ensemble learning for prediction.
- Provides prediction explanations to highlight key contributing factors.

#### 5. Voice Interaction Module:

- Converts user speech to text and reads aloud reports or guidance.

#### 6. Report Module:

- Generates summarized reports with generalized lifestyle suggestions for user guidance.

## 5.4 Implementation

The implementation demonstrates the effectiveness of FemAI in providing an accessible and informative platform for PCOD/PCOS risk awareness:

- **Questionnaire Module:**

Users input health data through an interactive questionnaire, which is also compatible with voice input.

- **Prediction Engine:**

The preprocessed responses are analyzed using ensemble ML models (Random Forest, Logistic Regression, KMeans) to provide a risk score.

- **Report Generation:**

The system generates a clear and concise report summarizing risk assessment and general lifestyle guidance (diet and habits) to improve awareness.

- **Voice Interaction:**

Users can listen to their reports or guidance, enhancing accessibility, especially for users with limited literacy or digital skills [1].

- **Dashboard:**

A user-friendly interface allows users to view reports, track historical data, and access educational content.

- **Data Security:**

All sensitive user information is stored securely with restricted access to protect privacy.

In conclusion, this chapter detailed the design and modeling of FemAI through diagrams and workflows. The structured design ensures smooth data flow and effective interaction among system components for optimal functionality.

# Chapter 6

## Technical Specification

The sixth chapter discusses the technologies and frameworks that power FEMAI, including Python, Flask, MySQL, and various machine learning libraries. It also covers the software and hardware requirements, emphasizing scalability, modularity, and reliability. This chapter ensures technical transparency and reproducibility.

The technical specifications of our project detail the **key technologies, frameworks, and methodologies** that form the backbone of the system. This section highlights the architecture, data handling approaches, machine learning algorithms, and deployment strategies that enable the platform to provide reliable PCOD/PCOS predictions and general lifestyle guidance. By documenting these specifications, we provide a clear understanding of the system's design, scalability, performance, and maintainability.

### Frontend (User Interaction)

- **Technologies:** HTML, CSS, JavaScript, and Bootstrap

The frontend is designed to deliver a **responsive and intuitive web interface**. Bootstrap is used to ensure consistent styling and layout across devices, while the interface incorporates interactive elements such as forms, dashboards, and visual charts.

- **Chatbot UI Integration:**

A conversational interface is embedded to guide users through health questionnaires, making the system more interactive and user-friendly.

### Backend Development

- **Language & Framework:** Python (Flask)

Flask handles server-side logic, manages requests from the frontend, and coordinates data flow between the user interface and the machine learning models.

- **REST APIs:**

APIs are used to connect the frontend with the ML prediction engine and database, ensuring seamless communication and real-time responses.

## **Database Management**

- **Database:** MySQL

Structured health data, user profiles, questionnaire responses, and report summaries are stored securely in a relational database. MySQL ensures data integrity, consistency, and efficient querying.

## **Machine Learning Stack**

- **Libraries:** Pandas, NumPy, Scikit-learn

- **Preprocessing:** Data normalization, feature selection, and scaling to prepare inputs for predictive models.

- **Algorithms:**

- **KMeans:** For clustering health data patterns.

- **Random Forest:** For robust prediction and feature importance analysis.

- **Logistic Regression:** For probability-based risk assessment.

- **Methodologies:**

- **Ensemble Learning:** Combining multiple models to improve predictive accuracy.

- **Scaler Approach:** Categorizing and normalizing input data for consistent predictions.

- **Prediction Explanation:** Providing users with insights into which features influenced the prediction outcome.

To summarize, this chapter described the technical tools, frameworks, and libraries that enable FEMAI's implementation. The selected technologies ensure scalability, efficiency, and the successful realization of the system's objectives.

# Chapter 7

## Project Scheduling

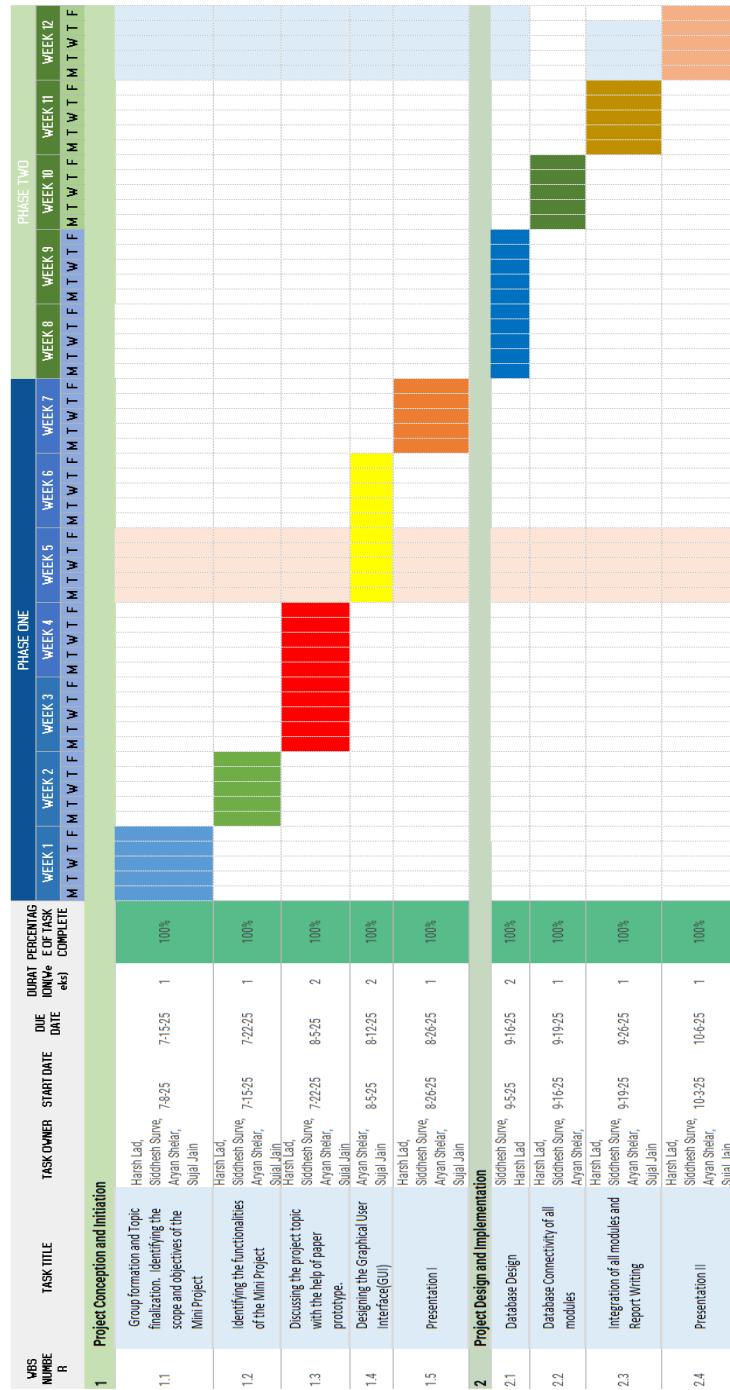
This chapter outlines the entire project timeline, milestones, and task allocation among team members. It includes tables and charts to illustrate workflow, ensuring efficient coordination and time management. The chapter reflects on how systematic scheduling contributed to the timely completion of FEMAI.

In project management, a schedule is a listing of a project's milestones, activities, and deliverables. A schedule is commonly used in the project planning and project portfolio management parts of project management. The project schedule (Table 7.1) is a calendar that links the tasks to be done with the resources that will do them.

Sr. No.	Group Members	Duration	Task Performed
1.	Harsh Lad, Siddhesh Surve, Sujal Jain, Aryan Shelar	2 <sup>nd</sup> Week of July	Group formation and topic finalization. Identifying scope and objectives of the project. Discussing the project concept and creating an initial paper prototype.
		1 <sup>st</sup> Week of August	Identifying the key functionalities of the project and planning feature requirements.
2.	Siddhesh Surve, Aryan Shelar	2 <sup>nd</sup> Week of August	Preprocessing the dataset and training the ML models for PCOD/PCOS prediction.
3.	Aryan Shealar	3 <sup>rd</sup> Week of August	Designing the Graphical User Interface (GUI) and dashboard for user interaction.
4.	Sujal Jain, Aryan Shelar	4 <sup>th</sup> Week of August	Implementing core features such as the questionnaire module, voice interaction, and generalized lifestyle guidance
5.	Harsh Lad, Siddhesh Surve	1 <sup>st</sup> Week of September	Integrating the trained ML model with the GUI and linking the system to the database
6.	Harsh Lad, Sujal Jain	3 <sup>rd</sup> Week of September	Ensuring full database connectivity across modules and testing end-to-end functionality.

**Table 7.1: Project Task Distribution**

A Gantt chart is a type of bar chart that illustrates a project schedule. This chart lists the tasks to be performed on the vertical axis, and time intervals on the horizontal axis. Gantt chart (Fig 7.1) illustrates the start and finish dates of the terminal elements and summary elements of a project.



**Fig 7.1: Gantt Chart of FemAI**

This chapter provided a clear project scheduling plan that ensured timely completion of FEMAI. Proper allocation of tasks and milestones contributed significantly to maintaining the project's quality and timeline.

# Chapter 8

## Results

The results chapter presents the implemented system's outputs, demonstrating FEMAI's performance, interface design, and core functionalities. It highlights how users can interact with the platform, view predictions, and access health guidance. These section validates the project's success in meeting its intended goals.

The results section provides a comprehensive overview of the implemented FEMAI system, highlighting the deliverables, user-facing outcomes, and system impact in facilitating PCOD/PCOS awareness, assessment, and support. Key results are illustrated through interface outputs, risk predictions, and integrated digital tools.

### 8.1 System Welcome & User Engagement

The FEMAI system's homepage (see Fig 8.1) welcomes users with a visually appealing interface and a clear project mission: "Your PCOD/PCOS Health Companion." The system promises advanced AI-powered detection and personalized health guidance for women, emphasizing empowerment and ease of access. Prominent calls to action ("Get Started," "Learn More," along with "Sign In" and "Sign Up") make user onboarding straightforward for both returning and new users. The site navigation bar includes essential modules such as Home, Chat Assistant, Health Info, Quick Actions, Health Tips, and Assessment, ensuring seamless workflow for all user roles.

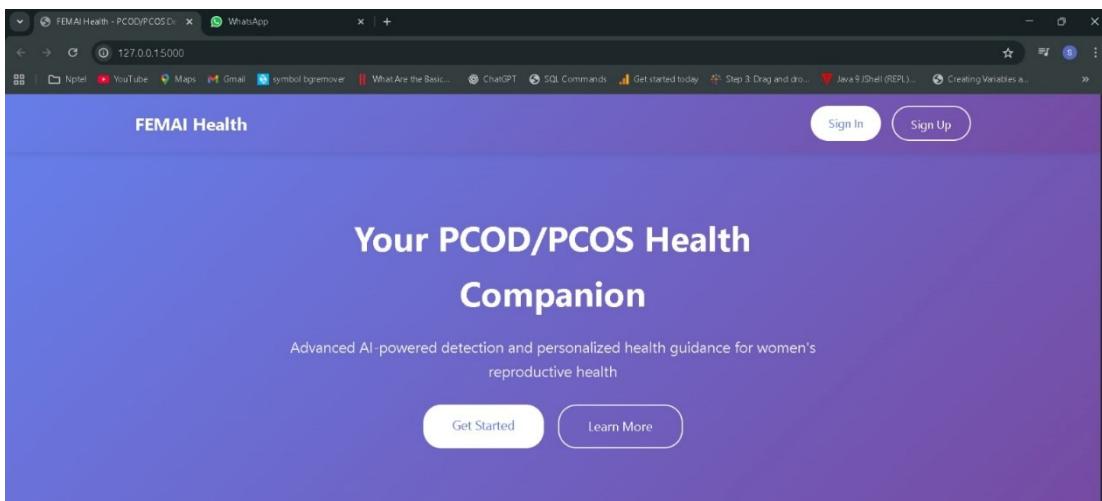
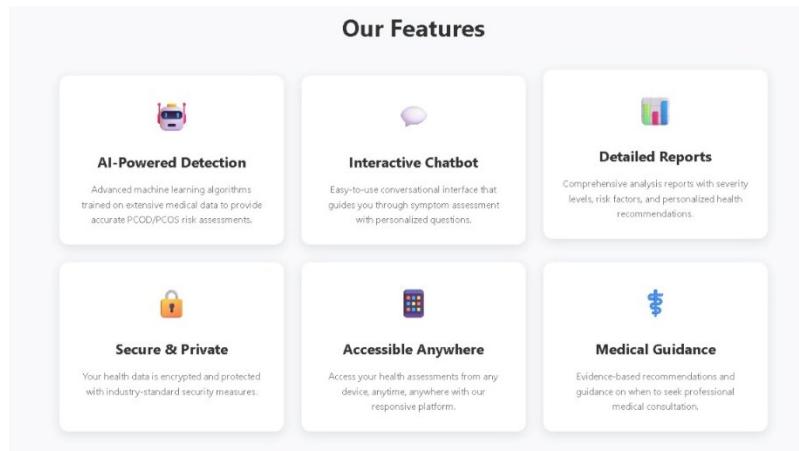


Fig 8.1 System Welcome & User Engagement

## 8.2 Core Features

The second view (see Fig 8.2) showcases FEMAI's six core platform features, which collectively define its scope, reliability, and user orientation:

- **AI-Powered Detection:** Advanced machine learning algorithms perform real-time PCOD/PCOS risk prediction based on patient inputs.
- **Interactive Chatbot:** Users interact with a conversational interface for symptom assessment, supported by personalized and dynamic questioning.
- **Detailed Reports:** Users receive comprehensive assessment reports that include risk levels, key risk factors, and actionable recommendations.
- **Secure & Private:** All health data are encrypted and managed with stringent security practices, ensuring privacy compliance.
- **Accessible Anywhere:** The responsive web platform allows secure access from any device, supporting widespread reach and inclusivity.
- **Medical Guidance:** Users receive evidence-based recommendations on risk reduction, lifestyle adjustments, and when to seek professional help.



8.2 Core Features

## 8.3 Educational Impact & Clinical Performance

In the “Our Impact” section (see Fig 8.3), the platform documents its societal and health-related contributions:

- **10M+ Women Affected Globally:** FEMAI highlights the massive prevalence of PCOD/PCOS and the urgent need for scalable digital solutions.
- **95% Early Detection Rate:** Machine learning models have achieved high accuracy, significantly improving early detection compared to traditional methods.
- **24/7 Available Support:** Continuous availability of AI guidance increases patient trust and platform reliability.
- **100% Confidential & Secure:** FEMAI maintains strict confidentiality, reinforcing its reputation as a safe and ethical healthcare tool.

## 8.4 Evidence-Based Information & Disclaimers

The “About PCOD/PCOS” segment serves as an educational resource, succinctly outlining the nature of PCOD and PCOS, the rationale for early detection, and the importance of AI-driven assessment. Prominent disclaimers clearly communicate that this tool is intended for informational purposes only, not as a replacement for formal medical diagnosis or treatment, thus promoting responsible and ethical use for all users.

## 8.5 User Registration and Sign-In

The FEMAI platform provides a streamlined user authentication workflow to ensure secure and accessible entry for all users.

- **Sign-Up Interface:**

The registration page invites new users to join FEMAI Health by creating a secure account.

Required fields include username, email address, and password (with confirmation). This form collects only essential information, facilitating quick onboarding while protecting user privacy. Clear prompts and a prominent "Sign Up" button guide users intuitively through the registration process.

- **Sign-In Interface:**

Upon successful registration, users are directed to the login page. A confirmation message verifies successful account creation, prompting users to sign in using their chosen credentials.

The sign-in form offers quick access to the core platform, supports password-protected access, and includes convenient navigation for account recovery or admin login.

This robust registration and sign-in workflow ensures that only authenticated users can access FEMAI Health features, protecting sensitive health data while providing a frictionless experience for both new and returning users.

The figure consists of two side-by-side screenshots of a mobile application's user interface. Both screens have a dark purple header and footer. The left screen is titled 'Create Account' and contains fields for 'Username' (Sakshi), 'Email' (sakshi23@gmail.com), 'Password' (redacted), and 'Confirm Password' (redacted). It has a large blue 'Sign Up' button at the bottom. Below it, small text says 'Already have an account? [Sign In](#)' and '[Back to Home](#)'. The right screen is titled 'Welcome Back' and shows a green success message 'Registration successful! Please sign in.'. It has fields for 'Username' (Sakshi) and 'Password' (redacted), and a large blue 'Sign In' button. Below the sign-in form, there is a section labeled 'OR' with links 'Don't have an account? [Sign Up](#)', '[Admin Login](#)', and '[Back to Home](#)'.

Fig 8.5 User Signup

## 8.6 Dashboard Features and Quick Actions

The FEMA1 Health dashboard presents users with a suite of five core features and four quick access actions, designed for comprehensive PCOD/PCOS self-management and support.

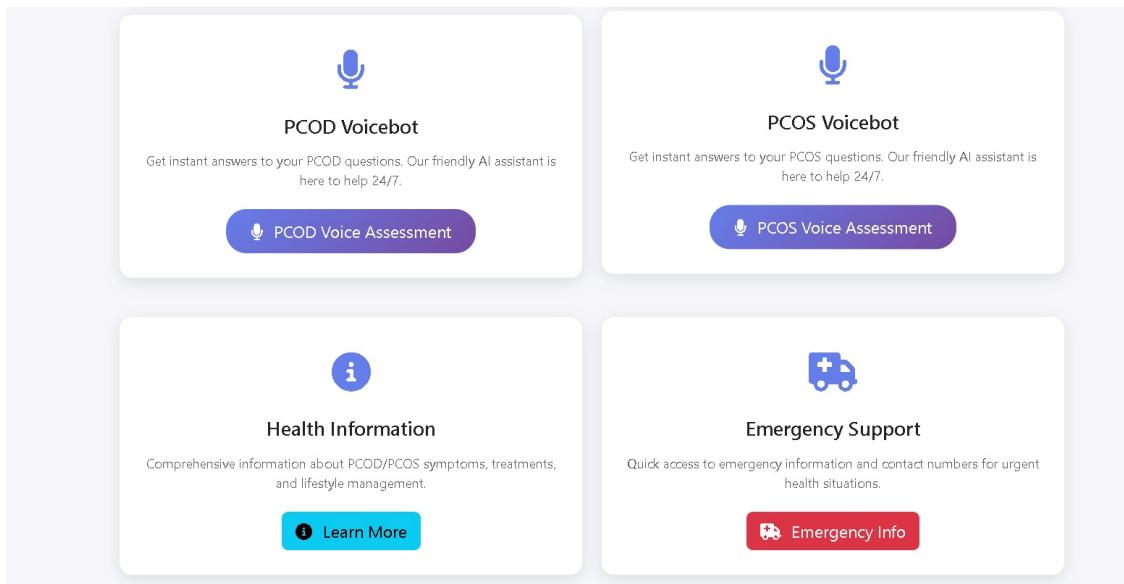


Fig 8.6.1 Dashboard Features

### Dashboard Features

- Symptom Tracker:**  
Users can log symptoms and monitor health patterns over time, supporting personalized tracking and better disease management.
- PCOD Chatbot & PCOS Chatbot:**  
FEMA1 offers two dedicated 24/7 AI chat assistants, each specialized in responding to queries about PCOD or PCOS. These chatbots guide users through symptom assessment, offer information, and recommend next steps for care.
- PCOD Voicebot & PCOS Voicebot:**  
In addition to text-based chat, FEMA1 provides voice-interactive assistants for both PCOD and PCOS. These tools improve accessibility and enable users to complete assessments or get advice via spoken interaction.
- Health Information:**  
This feature gives users access to detailed, trustworthy content about PCOD/PCOS symptoms, treatments, and recommended lifestyle adjustments for better health outcomes.
- Emergency Support:**  
Quick access to emergency contacts and information ensures users can get the help they need in urgent health scenarios.

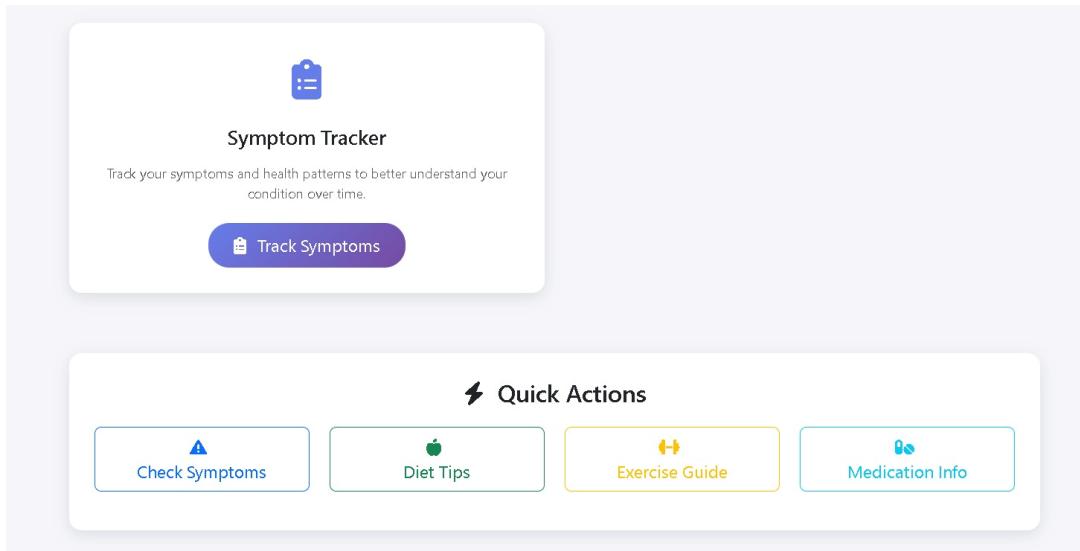


Fig 8.6.2 Dashboard and Quick Actions

### Quick Actions

- **Check Symptoms:**  
Instantly screen current symptoms to receive immediate feedback and suggestions for action or further assessment.
- **Diet Tips:**  
Personalized dietary guidance, including hydration, meal balance, and nutrition, helps users maintain hormonal health.
- **Exercise Guide:**  
Recommendations for physical activity support insulin regulation, stress management, and overall reproductive health.
- **Medication Info:**  
Reliable information about medications commonly used in PCOD/PCOS management, promoting safe and effective treatment decisions.

These features and quick actions, integrated into the main dashboard, ensure that users of FEMAI Health benefit from instant access to interactive support, practical health advice, emergency contingency resources, and personalized management tools—all within a secure and user-centered platform.

## 8.7 Clinical Blog – Expert Insights Module

The FEMAI Health Clinical Blog serves as a curated knowledge hub, designed to empower users with the latest expert insights, research findings, and practical guidance specific to PCOD/PCOS management. This module is structured with a visually distinct interface, blending a search bar for article lookup and clickable filter tags (such as Symptoms, Treatment, Nutrition, Lifestyle, and Research) to streamline browsing and discovery.

Featured content covers a broad range of evidence-based topics:

- In-depth articles explain diagnostic symptoms, treatment strategies, diet and nutrition tips, and effective stress or lifestyle management.
- Research-focused posts keep users abreast of recent medical advances and clinical trials relevant to women's reproductive health.

Each article is summarized with key metadata—category, title, snippet, estimated reading time, and publication date—allowing users to quickly gauge relevance. The Clinical Blog not only supports patient education, but also bridges the information gap between specialist guidance and user-driven learning for awareness, self-care, and timely intervention.

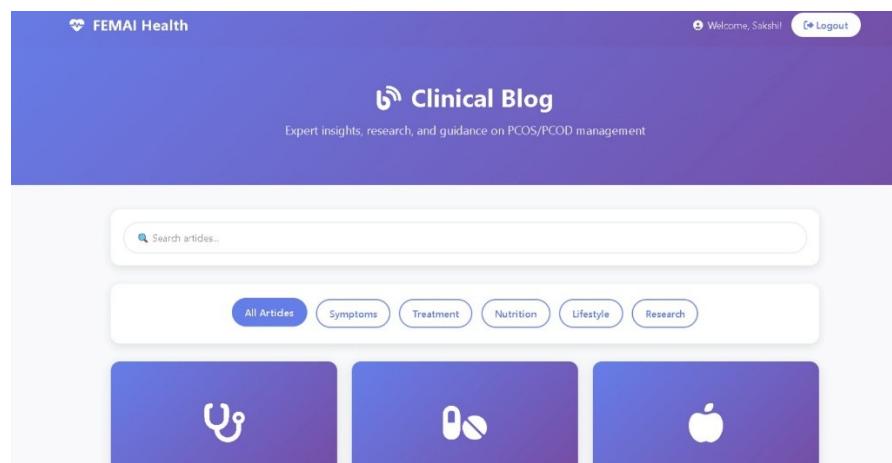


Fig 8.7.1 Clinical Blog View 1

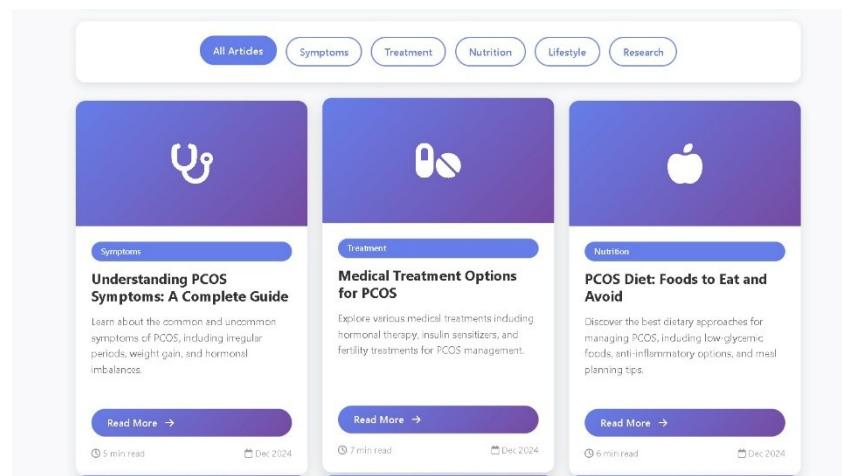
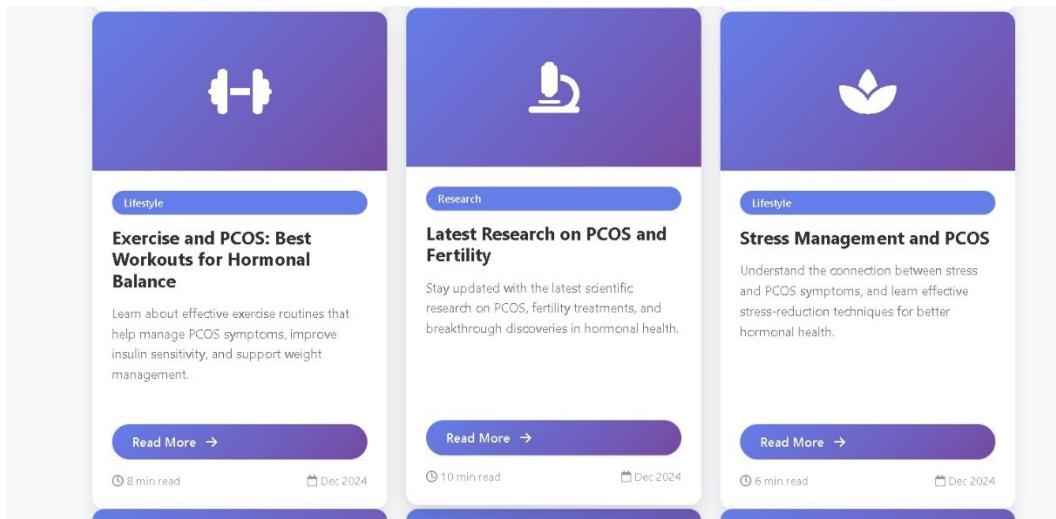


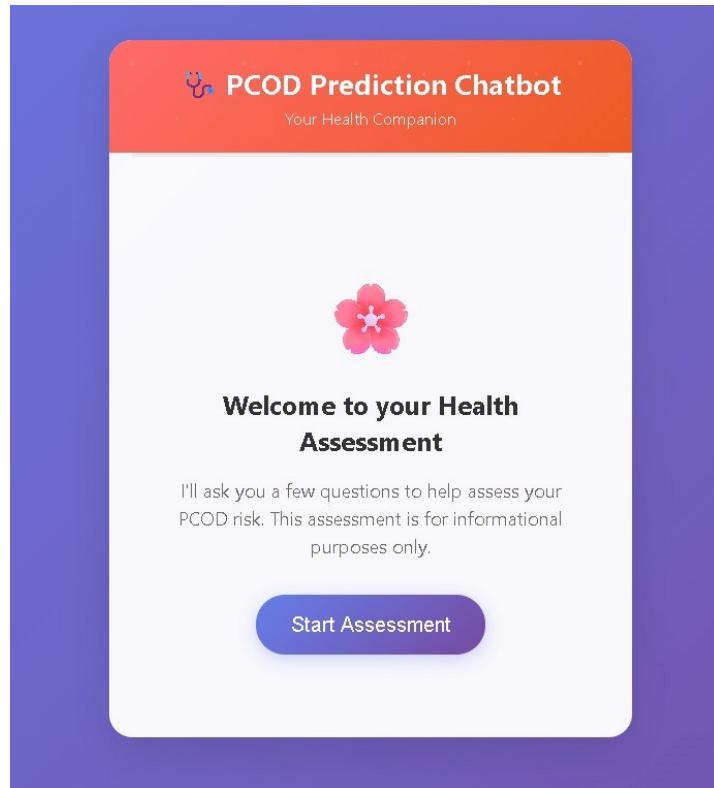
Fig 8.7.2 Clinical Blog View 2



*Fig 8.7.3 Clinical Blog View 3*

## 8.8 PCOD Prediction Chatbot: Diagnostic Parameters

The FEMAI PCOD Prediction Chatbot is designed to systematically assess risk through a dynamic health assessment interface. The chatbot guides users through a sequence of clinically relevant questions, aggregating patient-specific information that forms the basis for its AI-powered risk evaluation.



### Key Parameters Collected:

- **Demographic Factors:**

- Age
- Body Mass Index (BMI)

- **Core Symptoms:**

- Irregular Periods
- Weight Gain
- Excess Hair Growth (Hirsutism)
- Acne
- Hair Loss
- Pelvic Pain

- **Family & Medical History:**

- Number of Family Members with PCOD

These parameters have been selected based on major clinical risk indicators cited in medical literature for PCOD/PCOS diagnosis and management. By obtaining both objective (e.g., age, BMI) and subjective (e.g., recent weight gain, hair loss) data, the chatbot is able to generate a comprehensive profile for each user's reproductive health status.

The inclusion of detailed family history strengthens the risk assessment by accounting for genetic predisposition, while the inclusion of lifestyle and symptom-related factors ensures personalized prediction and actionable guidance.

The chatbot's simple, user-friendly interface enables accurate data entry and a seamless experience, forming the foundation for subsequent AI-driven prediction and personalized recommendations.

## 8.9 Assessment Result Outputs and Risk Stratification

As part of FEMA1's assessment workflow, users receive personalized PCOD risk outputs immediately after completing their questionnaire. The system generates visually distinct summary results to clearly communicate risk levels, enhancing the user experience and supporting informed decision-making.

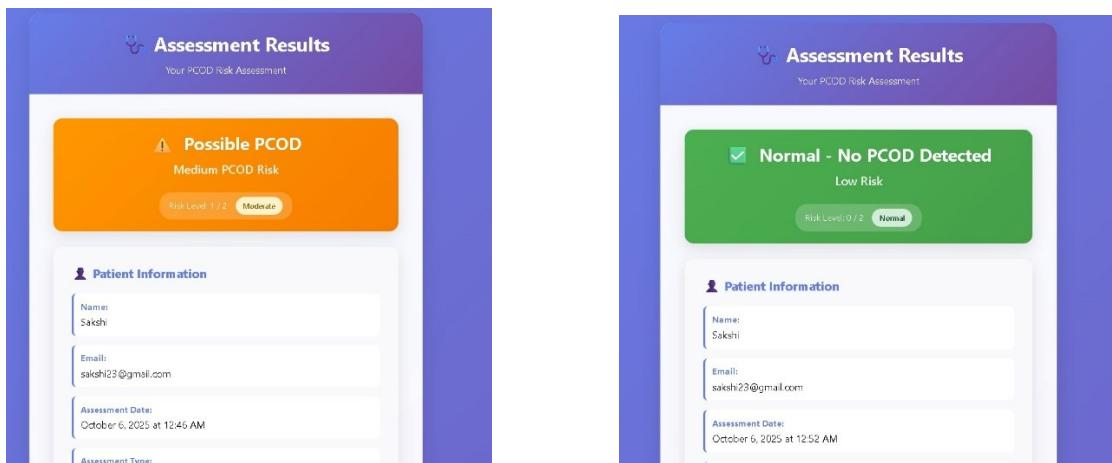


Fig. 8.9 PCOD Assessment results as Medium or Normal

### Result Categories

- **Normal – No PCOD Detected (Low Risk):**

Users who report either no major risk factors or a predominantly healthy profile receive a green banner result indicating “Normal – No PCOD Detected.” These individuals are classified as low risk. Assessment details—including patient name, email, and assessment timestamp—are displayed for their records and reference.

- **Possible PCOD (Medium PCOD Risk):**

If the assessment identifies some, but not all, risk factors consistent with PCOD, the system displays an amber/orange banner with the result “Possible PCOD – Medium PCOD Risk.” This moderate risk feedback prompts users to consider further lifestyle adjustments and, if concerned, to seek clinical evaluation for confirmation and tailored advice.

### High-Risk Exclusion

- The current version of FEMA1 **does not provide a “High Risk” output.**

High certainty diagnosis and serious risk stratification for PCOD/PCOS often depend on advanced hormonal measurements and detailed medical test results not available in the present implementation. For ethical and clinical accuracy, such evaluations are intentionally excluded from this release.

### Future Scope

- **Advanced Diagnostics and High Risk Detection:**

Future iterations of FEMAI aim to integrate laboratory API connectivity and medical test inputs.

This enhancement will enable robust detection of high-risk cases, expanding utility for earlier intervention and direct referral workflows.

This evidence-based, user-friendly risk communication ensures clarity and transparency while aligning the tool's capabilities with responsible medical practice and a solid foundation for future clinical upgrades.

## **8.10 PCOS Assessment: Parameters and Outputs**

FEMAI Health includes a dedicated PCOS Prediction Chatbot that mirrors the PCOD assessment flow, ensuring comprehensive coverage for both conditions. The chatbot autonomously navigates the user through a series of clinically significant parameters, collecting information critical for AI-powered risk prediction.

### **Key Parameters Collected:**

- **Demographics:** Age, BMI
- **Menstrual Profile:** Irregular periods
- **Reproductive History:** Difficulty getting pregnant or history of infertility, number of miscarriages
- **Androgenic Features:** Excess hair growth (hirsutism), acne/oily skin, hair thinning or hair loss
- **Family History:** Number of family members diagnosed with PCOS

After gathering responses, the system instantly evaluates the risk and presents the **Assessment Results** in a clear, user-friendly format:

- **Normal – No PCOS Detected:** Indicates a low-risk profile with medical advice to maintain healthy habits.
- **Possible PCOS (Medium PCOS Risk):** Reflects the presence of moderate risk factors, prompting recommendations for lifestyle management and further medical consultation if needed.

High-risk predictions are intentionally omitted in the current version, as accurate classification would require additional medical diagnostics and lab test integration, which are outlined as enhancements in the future scope.

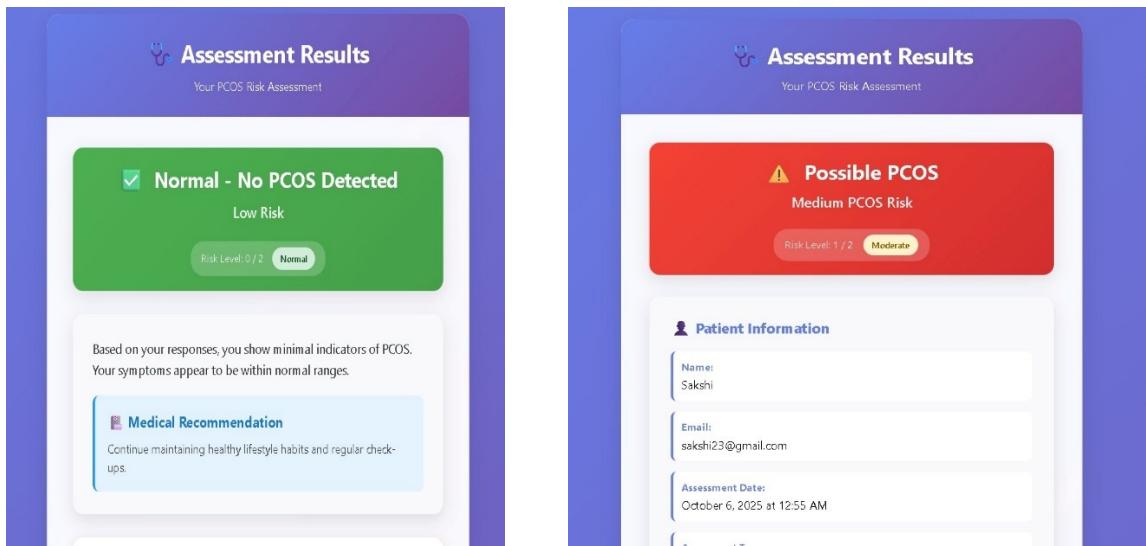


Fig 8.10 PCOS Assessment results as Normal and Medium.

## 8.11 Emergency Information Module

The FEMAI Health platform prioritizes user safety and preparedness by providing a built-in Emergency Information dialog accessible from critical interface touchpoints. This module is designed to offer immediate, actionable guidance during potentially urgent or life-threatening scenarios related to reproductive health.

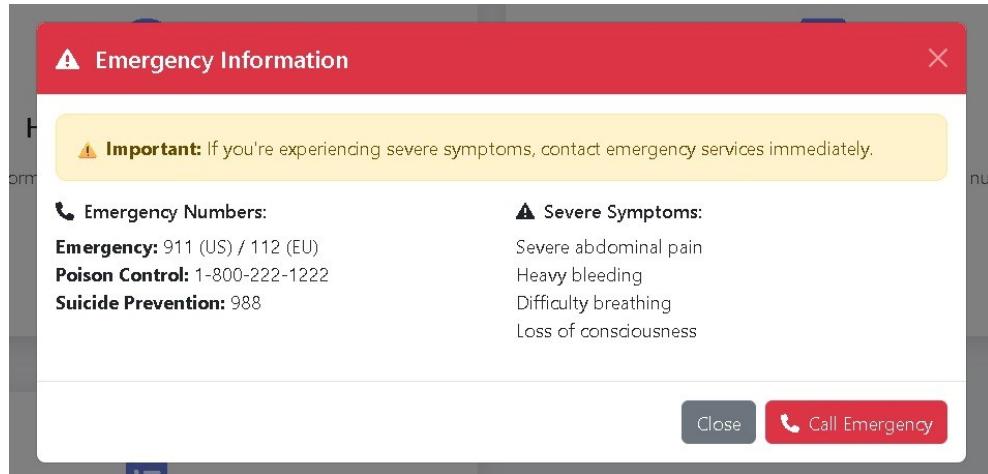


Fig 8.11 Emergency Contact

### Key Features:

- Emergency Numbers:**

The dialog displays a concise list of emergency contacts, including:

- General Emergency (911 for US/112 for EU)
- Poison Control (1-800-222-1222)
- Suicide Prevention (988)

- Severe Symptom Guidance:**

Users are alerted to the symptoms that may require urgent intervention. These include:

- Severe abdominal pain
- Heavy bleeding
- Difficulty breathing
- Loss of consciousness
- **Actionable Instructions:**  
Prominent warnings advise users to contact emergency services immediately if any severe symptoms are present. The dialog includes a clear call-to-action button for directly calling emergency services, streamlining the response in critical moments.
- **Ethical Use:**  
The platform clearly marks the information as “Important,” reinforcing that quick response to serious symptoms is critical for user health and safety.

## 8.12 Admin Access and Security

The FEMAI Health platform implements a dedicated admin authentication system to safeguard sensitive operations and maintain rigorous privacy standards. The Admin Access interface is clearly marked as a restricted area, accessible only to authorized personnel. Unauthorized access attempts are met with prominent warnings and are systematically logged to support security audits and prevent misuse.

### Key Features:

- **Secure Login:**  
The admin panel requires both username and password credentials, ensuring only verified administrators can gain entry.
- **Access Control Notice:**  
A highlighted message informs that the section is restricted, reinforcing regulatory compliance and reinforcing responsible system use.
- **Sign-In Procedure:**  
The visually distinct "Admin Sign In" button provides an immediate pathway for administrators, with a navigational option to return users to the main platform.

## 8.13 Voice Assistant Feature with Multilingual Support

FEMAI Health integrates an advanced Voice Assistant to facilitate voice-based interaction for symptom assessment and risk prediction for PCOD and PCOS. This innovation enhances accessibility, enabling users to speak their responses or enter text, thereby simplifying the data entry process for diverse

populations.

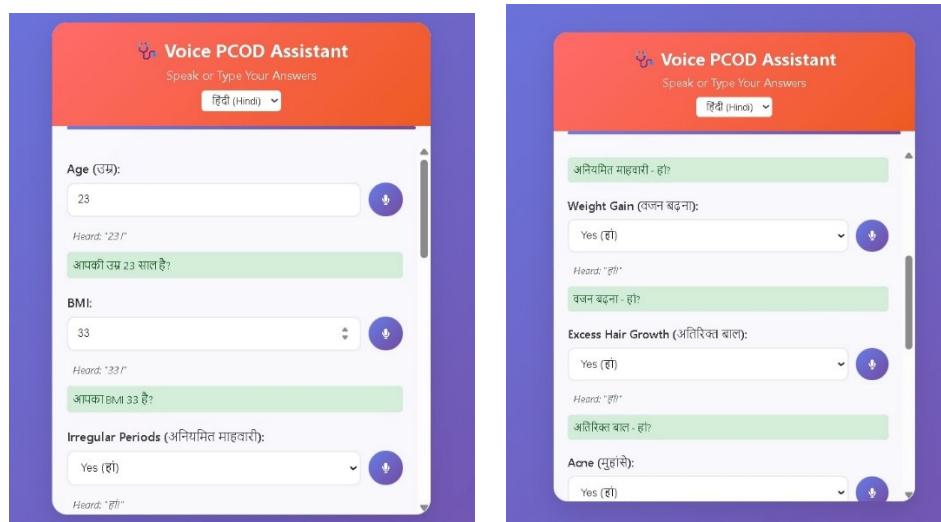


Fig 8.13 Multilanguage Voice to Text Support

- Multilingual Interface:**

Users can select their preferred language from a dropdown menu; the current implementation supports Hindi, with plans for expanding language options. This inclusion significantly broadens the platform's reach and user inclusivity.

- Voice Recognition and Confirmation:**

The Voice Assistant chain prompts users with questions regarding key diagnostic parameters, displaying the questions and recognized voice inputs in real-time for user verification. This feature improves accuracy and builds user confidence in the system.

- Parameters Captured via Voice Input:**

Core health indicators such as age, BMI, irregular periods, weight gain, excess hair growth, acne, hair thinning/loss, pain, and family history are collected through voice or typed responses. This mirrors the chatbot questionnaire to maintain consistency and comprehensiveness.

- User-Friendly Interaction:**

The interface is designed to clearly show prompts, recognized inputs, and command actions (e.g., recording start/stop), enabling a smooth conversational flow even for users unfamiliar with digital health systems.

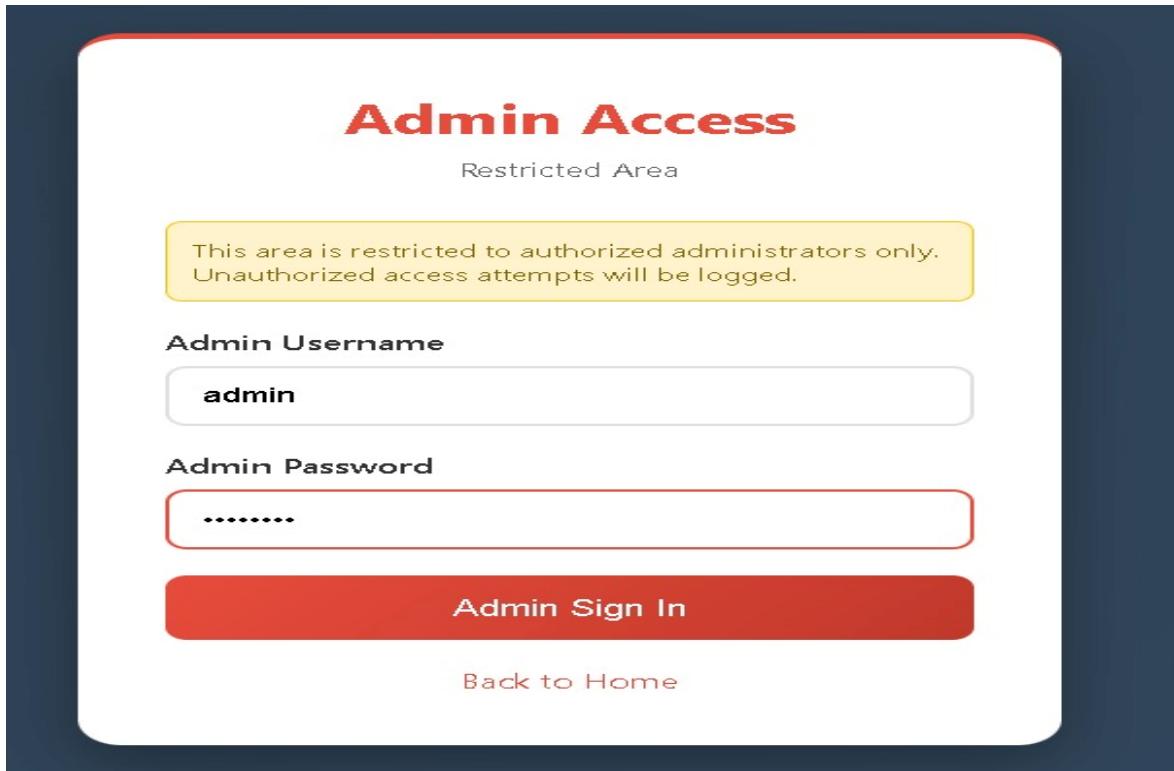


Fig 8.14 Admin Access Portal

The FEMAI Health platform integrates a **secure and dedicated Admin Login interface** specifically designed for platform administrators. This component ensures **controlled access to sensitive system operations**, maintaining the confidentiality and integrity of platform data. Clearly marked as “**Admin Access**,” the interface serves as a restricted environment for authorized users, with embedded security warnings that emphasize compliance and accountability. Any unauthorized login attempt is automatically recorded in the system’s activity logs to support auditing and prevent security breaches.

### **Key Features:**

#### **• Secure Authentication:**

The admin interface employs a robust authentication mechanism requiring both **username and password credentials**. This dual-input validation process restricts entry to verified personnel only, minimizing the risk of unauthorized system interaction.

#### **• Access Restriction and Warning Notice:**

A **prominent cautionary message** appears on the interface, indicating that the section is strictly limited to administrative users. This measure reinforces adherence to privacy protocols and enhances awareness of secure access policies.

- **Sign-In and Navigation Control:**

The **boldly designed “Admin Sign In” button** provides a direct gateway to administrative tools and maintenance functions. Additionally, a navigational link enables administrators or unintended users to safely return to the main FEMA1 platform, ensuring seamless user experience while maintaining strict boundary control.

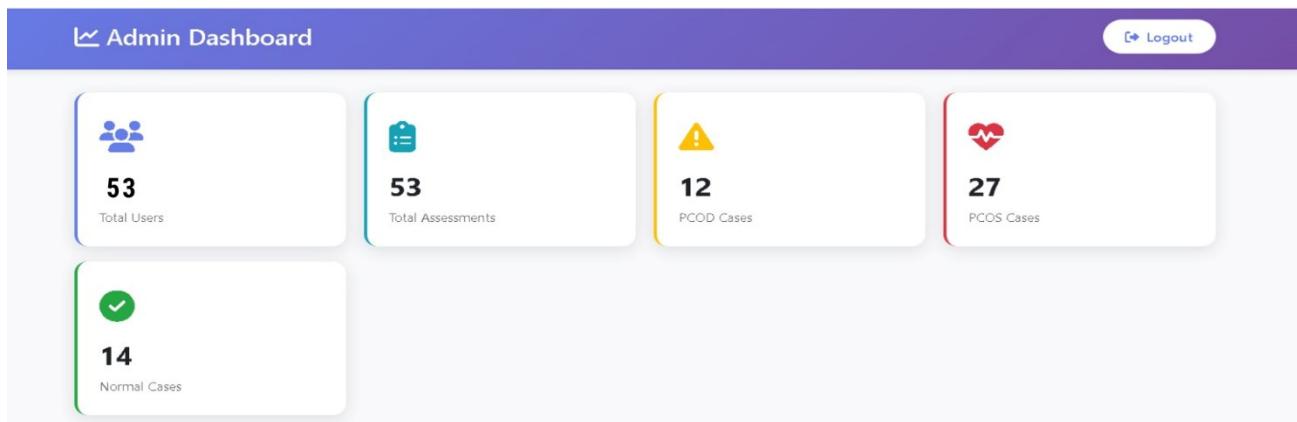


Fig 8.15 Admin Dashboard

### Administrator Dashboard Interface

The FEMA1 Health platform incorporates an **intuitive Administrator Dashboard** that delivers a consolidated overview of user engagement and health assessment trends. Designed for clarity and operational efficiency, the dashboard enables administrators to monitor **real-time platform activity**, ensuring effective oversight of diagnostic performance and user participation. Through a clean, data-driven layout, the interface supports clinical monitoring, program evaluation, and evidence-based decision-making.

### Key Features:

- **Comprehensive Statistics Overview:**

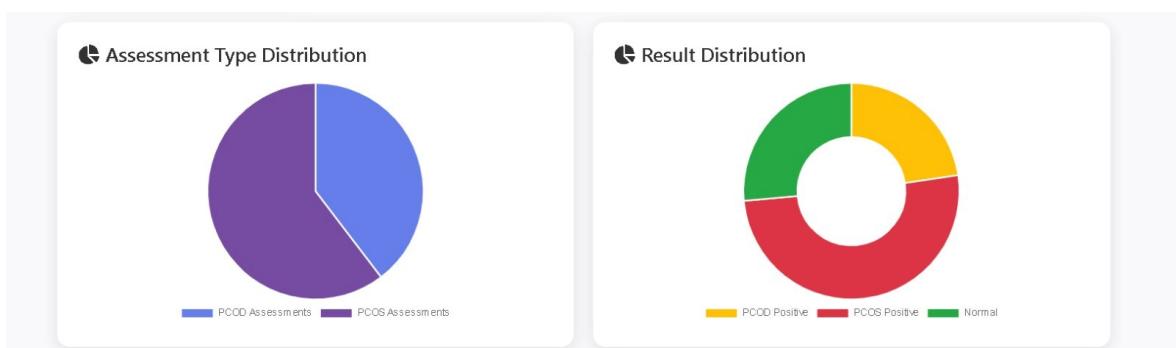
The dashboard prominently displays **total registered users**, **number of assessments conducted**, and a detailed **categorical breakdown of PCOD, PCOS, and normal cases**. This allows administrators to quickly identify patterns in user behavior and health outcomes.

- **Visual Data Representation:**

Interactive **pie and donut charts** present visual insights into the platform's analytical data. One chart illustrates the proportion of **PCOD vs. PCOS assessments**, while another highlights the **distribution of diagnosis results**—PCOD positive, PCOS positive, and normal—enabling rapid interpretation of population trends.

- **Operational Clarity and Monitoring:**

Crafted for **ease of reference and data precision**, the dashboard serves as an essential tool for clinical management and platform analytics. Its structured design ensures administrators can access key metrics instantly, supporting continuous improvement in digital health service delivery.



*Fig 8.16 Assessment Distribution Charts*

### Dual-Chart Health Analytics View

The FEMAI Health platform integrates a **dual-chart analytical interface** designed to provide clear, data-driven insights into PCOD and PCOS case management. This section visually interprets assessment activity and diagnostic outcomes, enabling administrators to perform rapid evaluations and maintain oversight of clinical performance. Through its structured and color-coded presentation, the interface simplifies complex data into an intuitive format that supports decision-making and public health monitoring.

## Key Features:

- **Assessment Type Distribution:**

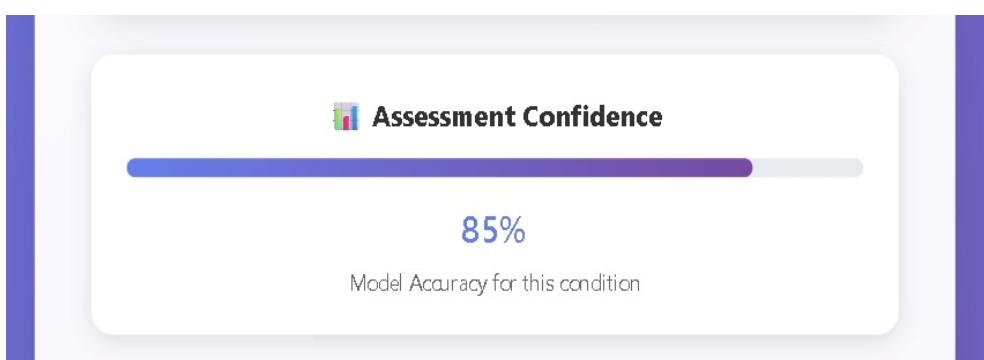
A pie chart illustrates the proportion of **PCOD** and **PCOS** assessments conducted across the platform. This visualization helps administrators understand user participation trends and the relative diagnostic demand between both conditions.

- **Result Distribution Visualization:**

An adjacent donut chart titled “*Result Distribution*” displays the outcome segmentation into **PCOD positive, PCOS positive, and normal cases**. Each section is distinctly color-coded to enhance readability and allow immediate recognition of health result patterns.

- **Analytical Insight and Oversight:**

By merging assessment frequency with outcome distribution, this dual-chart layout promotes **efficient performance tracking, quality control, and epidemiological analysis**—ensuring that administrators can interpret diagnostic data quickly and accurately.



### Assessment Confidence Indicator

The FEMAI Health interface integrates an **Assessment Confidence Indicator**, an accuracy-driven visual component designed to reflect the reliability of the diagnostic model’s predictions. Displayed as a smooth, color-graded progress bar, the interface communicates **model accuracy**—in this instance, **85%**—for the evaluated condition. This indicator serves as an immediate visual cue, allowing both users and administrators to quickly interpret the confidence level of AI-generated health assessments.

Key design features include the use of **gradient progress visualization** to convey analytical depth, a **clear numerical percentage display** for instant recognition, and a **minimalist layout** that blends seamlessly with the surrounding dashboard components. The combination of textual and graphical cues ensures accessibility and comprehension across diverse users, promoting both **data transparency and user trust**.

By integrating this feature, FEMA1 Health emphasizes interpretability in AI-assisted healthcare

diagnostics, bridging the gap between complex computational outcomes and user-facing clarity. It transforms algorithmic confidence into an understandable, interactive visual experience—reinforcing the platform's commitment to **accuracy, accountability, and human-centered design**.

To conclude, this chapter demonstrated the results and successful implementation of FEMA1. The system effectively predicts PCOD and PCOS risks and provides personalized recommendations, validating its accuracy and practical value.

# **Chapter 9**

## **Conclusion**

This chapter summarizes the overall findings, outcomes, and impact of FEMA1. It emphasizes the system's role in improving early detection, promoting health awareness, and supporting preventive care among women. The conclusion also reflects on the lessons learned during development and how technology can bridge healthcare gaps.

Our system, FEMA1, is an important step forward in women's health technology. It provides a user-friendly, smart, and supportive platform for the early detection and management of PCOD/PCOS. We have combined machine learning models with interactive features such as questionnaires, dashboards, voice interaction, and report generation to create a system that empowers women to make informed choices about their health.

This platform helps users identify potential risks early on and raises awareness through general lifestyle guidance. It promotes healthier habits and helps users better understand their condition. Its easy-to-use interface, along with features for women with limited digital skills, ensures that the system is accessible to a wide range of users.

Additionally, the predictive engine is designed to improve over time. It offers reliable insights while keeping users informed about the factors affecting predictions. The reporting function lets users track their health over time or share results with healthcare professionals, bridging the gap between technology and real medical advice.

Overall, this project highlights the power of combining data-driven insights with a user-focused design. It creates a tool that is educational and supportive. It encourages proactive health management, gives women knowledge about PCOD/PCOS, and supports the larger goal of preventive healthcare. By providing an accessible, inclusive, and trustworthy platform, the system represents a meaningful advance in women's health awareness and helps users make informed lifestyle choices.

In summary, this chapter reflected on the overall achievements of FEMA1. It reinforced the projects success in developing an AI-based tool for women's health awareness and early diagnosis of hormonal disorders.

# **Chapter 10**

## **Future Scope**

The final chapter looks ahead at potential improvements and expansions for FemAI. It discusses future enhancements such as advanced AI models, mobile applications, wearable integration, and multilingual voice assistance. This chapter envisions the system's evolution into a broader, more inclusive healthcare solution. FemAI platform, while currently focused on PCOD/PCOS prediction, awareness, and generalized lifestyle guidance, has significant potential for expansion to enhance functionality, accessibility, and user engagement. Possible avenues for future development include:

**1. Personalized Insights Dashboard:**

Expanding the dashboard to provide more detailed analytics, such as tracking changes in user-reported health parameters over time, visualizing trends, and offering comparative insights for better self-monitoring.

**2. Enhanced Predictive Models:**

Incorporating additional machine learning algorithms and larger datasets can improve the accuracy and reliability of risk predictions. Future models could also integrate lifestyle, genetic, and environmental factors for more comprehensive assessments.

**3. Mobile Application Development:**

Creating a mobile app version of FemAI would allow users to access the platform on-the-go, receive timely notifications, and interact with voice-based features more conveniently. Offline capabilities could further enhance accessibility in areas with limited internet connectivity.

**4. Educational and Interactive Content:**

Adding multimedia educational resources such as videos, infographics, and interactive tutorials can help users understand PCOD/PCOS, preventive measures, and general wellness practices more effectively.

**5. Community and Support Features:**

Introducing forums, peer discussion groups, or expert Q&A sessions can foster a supportive environment where users can share experiences, ask questions, and access guidance from healthcare professionals.

## **6. Integration with Wearable Devices:**

Future versions could integrate with wearable health devices to automatically collect relevant health metrics, such as activity levels, sleep patterns, or heart rate, thereby providing more precise predictions and personalized recommendations.

## **7. Multi-Language and Voice Enhancements:**

Expanding voice recognition and support for multiple languages can increase inclusivity, allowing women from diverse linguistic backgrounds and varying levels of digital literacy to use the platform effectively.

## **8. Research and Dataset Expansion:**

By incorporating anonymized data from larger cohorts and research studies, the system can continuously refine prediction models, contribute to PCOD/PCOS research, and provide insights for preventive healthcare initiatives.

To conclude, this chapter outlined the future scope and potential enhancements for FEMAI. By integrating advanced technologies and expanding accessibility, the system can further revolutionize women's healthcare analytics.

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