



IJRASET

International Journal For Research in
Applied Science and Engineering Technology



INTERNATIONAL JOURNAL FOR RESEARCH

IN APPLIED SCIENCE & ENGINEERING TECHNOLOGY

Volume: 14 **Issue:** V **Month of publication:** May 2026

DOI: <https://doi.org/10.22214/ijraset.2026.81972>

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EvolveMe: AI Life Optimiser-Predictive Health & Personal Life Optimization System

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Abstract: *The EvolveMe is a multi-domain predictive health and personal life optimization framework. It is powered by Artificial Intelligence. Most fitness applications in the current market function separately focussing only on fitness or nutrition or skin care and thus our insights are limited and less personalized. EvolveMe provides interconnected and intelligent wellness insights across multiple domains by using FoodAI, SkinAI and Fitness Dashboards to offer integrated real-time data. The system analyses meals and skin conditions and predicts behavioural trends using machine learning and computer vision to provide helpful health predictions. A smart chatbot called EvolveMe Assistant recommends some lifestyle changes as per the evolving data of the user. The framework focuses on scalability and privacy. We implement session-based data management. In the future, we will also migrate to the cloud infrastructure so we can deploy globally. EvolveMe changes the way we monitor our health passively and turns it into intelligently optimising our lives for the better actively.*

Keywords: *AI wellness, predictive analysis, health optimization, integration across domains, machine learning, personalisation.*

I. INTRODUCTION

The emergence of AI and data analytics have benefitted in digital health and wellness management activities in recent years. Most health and fitness apps don't integrate with each other. Each focuses on only one component – fitness, nutrition, skin, etc. When UX specialists and brand designers don't use the same language, it confuses clients. With the rise of users relying on data to drive lifestyle decisions, we need a holistic system that unifies all our wellness indicators into one established intelligent system.

To tackle the issue, EvolveMe: AI Life Optimizer has been designed as a framework for predictive health and optimization of one's personal life. EvolveMe uses multiple wellness domains including baselines of FoodAI (intelligent meal recognition), SkinAI (definitive skin analysis on-the-go) and Fitness Dashboards (physical activity and biometric tracking). If we use machine learning, computer vision, cross-domain data analytics, the system takes isolated health data and turns it into actionable insights that is predictive. A context-aware assistant improves personalization by delivering evolving recommendations based on user behaviour. EvolveMe aims to shift the focus from just monitoring health reactively to proactively optimising life. In other words, this startup empowers people to act intelligently and proactively and make data-driven health and fitness choices.

II. EASE OF USE

The EvolveMe: AI Life Optimizer is built on the belief that managing your health should be effortless, centering its entire design on ease of use and everyday efficiency. By stripping away technical complexity, the platform offers a streamlined, intuitive interface that empowers anyone to track, understand, and improve their daily habits without needing to be a data expert. Instead of forcing users to jump between different apps for different needs, EvolveMe brings everything together into one cohesive dashboard, housing FoodAI, SkinAI, and Fitness tools under a single digital roof. This unified approach ensures that all your wellness information works in harmony, making the journey toward self-improvement feel like a natural part of your routine rather than a complicated tech project.

A. Maintaining the Integrity of the Specifications

Maintaining data integrity and system consistency is fundamental to the *EvolveMe: AI Life Optimizer* framework. As the platform integrates multiple AI-driven modules—each handling distinct data types such as images, biometric values, and behavioural inputs—it becomes essential to ensure uniform data validation, synchronization, and secure handling throughout the ecosystem.

Each data transaction in *EvolveMe* follows a specification-driven validation pipeline to prevent anomalies, duplication, or corruption.

For instance, image data from the SkinAI module and nutritional data from FoodAI are standardized before analysis to ensure compatibility with the central AI engine. The system employs a modular API schema, enforcing strict communication protocols between the backend services and the AI inference layers to maintain functional consistency across modules.

Furthermore, session-based data handling ensures that health records remain temporally bound to a user's active session, reducing latency while maintaining data coherence. Encryption and structured data formatting standards are applied to preserve privacy and prevent data leakage. Planned migration to cloud infrastructure in future versions will further enhance data integrity by introducing secure replication, real-time backups, and version control across distributed systems

In essence, EvolveMe maintains the integrity of its specifications through standardized data governance, structured communication protocols, and secure AI orchestration, ensuring reliable, precise, and ethically compliant system behaviour across all wellness domains.

III. SYSTEM ARCHITECTURE

The *EvolveMe: AI Life Optimizer* framework is designed with a modular and scalable architecture that enables seamless integration of multiple AI-driven wellness components into a unified system. The architecture ensures interoperability, data integrity, and high-performance analytics across diverse domains such as nutrition, skincare, fitness, and behavioral monitoring. Each subsystem operates autonomously while communicating through standardized APIs to the central AI engine, which handles data aggregation, prediction, and personalization.

The overall architecture consists of four primary modules: FoodAI, SkinAI, Fitness Dashboards, and the EvolveMe Assistant. These modules are supported by a secure backend infrastructure and session-based data handling, with planned migration to cloud storage for future scalability. The core workflow begins with data collection from user inputs or captured media, followed by preprocessing, AI-based analysis, and cross-domain correlation to generate predictive health insights.

A. FoodAI – Intelligent Meal Recognition and Nutritional Analysis

FoodAI leverages image recognition and deep learning models to identify food items from images uploaded by the user. The system extracts nutritional data such as calorie count, macronutrient distribution, and portion estimation. This data is then mapped to user-specific health goals stored in the backend, allowing EvolveMe to suggest dietary modifications and trend-based predictions for future wellness optimization. FoodAI's adaptive learning improves accuracy as users upload more diverse meal images over time.

B. SkinAI

SkinAI serves as the high-tech heart of the platform's skincare analysis, transforming your smartphone camera into a professional-grade diagnostic tool through the use of advanced computer vision and convolutional neural networks (CNNs). By performing a real-time digital analysis of your complexion, the module moves beyond surface-level observation to identify specific concerns like active acne, dehydration, and hyperpigmentation with clinical precision. What truly sets SkinAI apart is its holistic approach; it functions as a predictive engine that correlates your skin's health with your broader lifestyle, linking flare-ups or dryness to dietary choices, sleep patterns, and environmental stressors like UV levels or local humidity. Rather than offering generic solutions, it delivers dynamic, metric-driven recommendations that address the root causes of skin issues, ensuring that your skincare routine is as adaptive and unique as you are.

C. Fitness Dashboards and Manual Insights

The EvolveMe Fitness Dashboard serves as a centralized command center for your physical well-being, gathering exercise data, sleep patterns, and vital body metrics into one clear, comprehensive view. Whether you prefer the ease of automatic syncing from your favorite wearable devices or the precision of logging details manually, the platform bridges the gap to ensure no part of your progress is overlooked. By applying advanced predictive modeling, the system looks far beyond simple step counts; it cross-references your activity levels with your diet and rest quality to flag potential issues like overtraining, hidden fatigue, or nutritional gaps before they become problems. Instead of overwhelming you with raw numbers, the dashboard uses intuitive charts and visual trend maps to tell the story of your health, making it easy to understand exactly how your lifestyle choices are impacting your energy and long-term vitality.



Fig 1: Proposed Framework of EvolveMe

D. The EvolveMe Assistant

Acts as the intuitive "brain" of the platform, serving as a dedicated digital companion that grows with you by offering personalized, real-time guidance exactly when it's needed. By blending advanced natural language processing (NLP) with a human-centric approach, the assistant functions as a smart mediator that translates complex data and historical progress into simple, actionable steps for your daily life. It moves beyond generic health tips to focus on true optimization; whether it's suggesting a lighter workout based on your recovery speed, recommending meals tailored to your daily energy needs, or adjusting skincare routines to match environmental shifts, the assistant removes the guesswork from self-improvement. Ultimately, it transforms the journey toward your best self from a chore into a seamless, supportive conversation, bridging the gap between sophisticated analytics and your personal wellness goals.

E. Backend Infrastructure and Data Flow

The backend system is built using Flask and Node.js APIs to handle modular AI services efficiently. Data collected from users is processed through a session-based local storage model to ensure responsiveness during operation. Each module communicates with the central AI engine for inference, and the results are displayed through the React.js frontend. Future versions of EvolveMe will integrate cloud-based storage and computing to enhance data scalability, real-time synchronization, and security compliance.

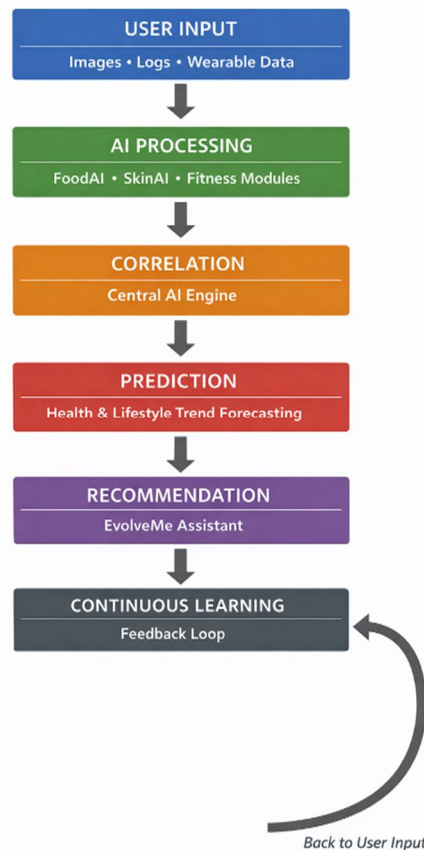
F. Integration Workflow

The system's workflow follows a cyclical model:

- 1) Data Input → user uploads images, health logs, or wearable data.
- 2) Processing → AI modules analyse individual datasets (FoodAI, SkinAI, Fitness).
- 3) Correlation → the central AI engine cross-analyses patterns between domains.
- 4) Prediction → system forecasts health trends and lifestyle outcomes.
- 5) Recommendation → the EvolveMe Assistant provides real-time, adaptive suggestions.

This architecture ensures that EvolveMe functions as a self-learning, predictive wellness ecosystem, capable of continuous evolution and user-specific adaptation.

EvolveMe Integration Workflow



The EvolveMe Integration Workflow operates as a continuous, self-improving loop designed to turn personal data into a roadmap for better living. It begins when a user shares information—whether that is a quick meal photo, a skin scan, or health stats from a wearable device—which is then instantly processed by specialized AI modules focused on nutrition, fitness, and skin health. Rather than looking at these areas in silos, the system’s central engine cross-references the data to find hidden patterns, such as how a specific diet might be impacting workout recovery or skin clarity. This allows the platform to move beyond simple tracking and into "predictive wellness," where it can actually forecast future health trends and potential outcomes based on your current habits. Finally, the EvolveMe Assistant delivers these findings as friendly, real-time suggestions that adapt to your changing needs, creating a personalized ecosystem that learns and evolves alongside you every single day.

G. Units

EvolveMe operates across multiple wellness domains where data is measured in diverse scientific units. For accuracy and consistency, SI (International System of Units) is adopted as the primary standard. Examples include:

- Energy and nutrition are measured in kilocalories (kcal).
- Weight and mass are expressed in kilograms (kg).
- Height and distance metrics are measured in meters (m) or centimetres (cm).
- Skin analysis metrics, such as texture variance or hydration level, are expressed in percentage (%).
- Sleep and activity durations are represented in hours (h) or minutes (min).
- Data volume and computational performance are expressed in megabytes (MB), gigabytes (GB), and milliseconds (ms) for response time.

All computations and analytical results maintain consistent unit standards to avoid dimensional inconsistencies across FoodAI, SkinAI, and Fitness modules.

H. Equations

EvolveMe utilizes compact mathematical formulations to normalize and interpret multimodal wellness data in real time. These equations support efficient aggregation of fitness, nutrition, skin, and laboratory insights while maintaining computational simplicity and clarity.

The overall fitness score is calculated by averaging step count, calorie expenditure, and heart rate:

where S denotes steps, C represents calories burned, and H indicates average heart rate.

Nutritional impact in the FoodAI module is estimated using confidence-weighted analysis of detected meals:

$$N = M \cdot \alpha, \quad (2)$$

where M is the estimated meal calorie value and α is the AI confidence score.

Skin health risk is assessed by combining detection confidence and severity estimation:

$$R_s = \beta \cdot \delta, \quad (3)$$

where β represents detection confidence and δ denotes severity.

Laboratory deviations are evaluated using a normalized risk index:

$$R_l = \frac{|V_o - V_n|}{V_n}, \quad (4)$$

where V_o is the observed biomarker value and V_n is the reference value.

A unified wellness score is derived by aggregating normalized module outputs:

$$W = \frac{F + N + R_s + R_l}{4}. \quad (5)$$

This formulation enables EvolveMe to deliver holistic, interpretable, and actionable wellness insights.

IV. CONCLUSION

EvolveMe represents a fundamental shift in how we approach personal health, moving away from fragmented tracking and toward a fully integrated, intelligent lifestyle optimization system. By unifying the often-isolated worlds of nutrition, fitness, and skincare into a single, cohesive framework, the platform eliminates the confusion caused by disconnected apps and generic advice. Through the power of FoodAI, SkinAI, and advanced fitness analytics, the system doesn't just record what has already happened; it uses complex cross-domain correlation and predictive modeling to anticipate your future needs. This allows the EvolveMe Assistant to provide hyper-personalized, real-time guidance that adapts as your body and environment change, turning passive data into active, life-improving choices. With a robust backend designed for data integrity and a clear path toward global cloud scalability, EvolveMe empowers individuals to take proactive control of their well-being. Ultimately, it bridges the gap between sophisticated machine learning and daily human habits, transforming the pursuit of health from a series of chores into an evolving, data-driven journey toward one's best self.

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International Journal for Research in Applied Science & Engineering Technology (IJRASET)

ISSN: 2321-9653; IC Value: 45.98; SJ Impact Factor: 7.538

Volume 14 Issue V May 2026- Available at www.ijraset.com



10.22214/IJRASET



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