



AI-Driven Optimization For Personalized Health Metric Analysis

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Abstract - AI-Driven Optimization For Personalized Health Metric Analysis paper presents an AI driven therapeutic optimization framework for Personalized Health Metric Analysis, aimed at bridging the gap between fragmented health data and comprehensive wellness management. The proposed system integrates Natural Language Processing (NLP) to interpret medical text, Computer Vision (CV) to analyze meal images, and a machine learning based optimization engine to generate customized diet and workout recommendations. Unlike conventional fitness or nutrition apps, the framework incorporates region-specific food habits and individual health parameters such as BMI, glucose level, and activity data to deliver culturally relevant and adaptive insights. By unifying multimodal health data, the system enables early disease-risk detection and personalized therapeutic planning. This approach enhances preventive healthcare, user engagement, and real-time decision support, contributing to the evolution of intelligent and data driven health ecosystems.

Index Terms - Personalized Healthcare, Reinforcement Learning, Health Metric Analysis, HER

I.INTRODUCTION

Healthcare today is rapidly evolving with the integration of digital technologies, yet one of its persistent challenges is providing personalized therapeutic guidance that accounts for an individual's unique health conditions, lifestyle, and environment. Traditional healthcare systems often focus on generalized treatment plans, which may overlook critical variations in diet, activity level, and medical history. With chronic diseases such as diabetes, hypertension, and cardiovascular disorders on the rise, the need for personalized healthcare interventions has become more urgent than ever. There are many health apps and gadgets, but they often do just one thing. Some count your steps, while others track calories. Very few of them connect all the dots to give you a complete picture. No app really takes your personal health history, what you like to eat, your physical activity, and your long-term health risks into account all at once. This shows a big need for a smarter, all-in-one health tool.

Our project aims to fill that gap. We've built a system that works like a personal health coach, combining several powerful features. It creates a custom meal plan just for you, based on your health goals, medical history, and even your local food culture. It also lets you take a quick photo of your food and get instant feedback on whether it fits your diet.

To provide a truly complete plan, the system also recommends specific workouts for the users. Most

importantly, it uses all your information to give you a risk score for certain diseases, so you can make healthier choices before problems even start. Despite the increasing availability of electronic health records (EHRs), wearable devices, and fitness applications, health data remains fragmented and unstructured. Medical test results, dietary intake, lifestyle habits, and regional food preferences are typically recorded in isolated formats, making it difficult for clinicians and patients to obtain a holistic view. This lack of integration results in missed opportunities for early intervention, inefficient therapeutic recommendations, and suboptimal patient outcomes.

To address this challenge, we propose a system for Therapeutic Optimization for Personalized Health Metric Analysis, which integrates diverse data streams into a unified framework. The system leverages Natural Language Processing (NLP) to analyze textual health records, Computer Vision (CV) to extract calorie and nutrient information from meal images, and an AI-based recommendation engine that generates personalized diet and exercise plans. Uniquely, the framework also incorporates region specific food preferences (eg. South vs. North Indian diets), along with user-specific factors such as age, gender, weight, and medical history, ensuring recommendations are both accurate and culturally relevant.

II. LITERATURE REVIEW

Therapeutic optimization in healthcare has gained prominence in recent years due to the exponential growth of biomedical data and the increasing availability of intelligent health monitoring systems. Several researchers and organizations have developed computational models and optimization frameworks aimed at improving clinical decision making, enhancing patient outcomes, and reducing treatment costs. The following literature review highlights notable contributions in this field. Therapeutic optimization for health metric analysis has become an increasingly vital area of research due to the growing prevalence of chronic lifestyle related diseases. Various studies have explored how artificial intelligence and machine learning can personalize health recommendations, track progress, and provide real-time predictive analytics.

The research paper "Dynamic AI-Enhanced Therapeutic Framework for Precision Medicine" by R. Gayathri, S. K. B. Sangeetha, and Leena Rosalind, published in IEEE Access (2025), proposes an adaptive AI-based system that personalizes treatment using multi-modal data such as genomic, clinical, and wearable sensor inputs. It introduces a patient in the loop reinforcement learning model that continuously refines therapeutic decisions based on patient feedback and real-time health changes. The framework dynamically adjusts biomarker importance, ensuring accurate and personalized recommendations. Tested on the dataset, it achieved high precision, recall, and consistency across diverse patient profiles. Overall, the study presents a promising step toward realtime, data-driven, and adaptive precision medicine [1].

The study of "The Application of Artificial Intelligence in the Field of Mental Health A Systematic Review" by Raziye Dehbozorgi systematically reviews the use of AI in mental health diagnosis, prediction, and treatment. The study analyzed narrowing down to 15 key studies that met quality and inclusion criteria. It highlights how AI technologies, such as chatbots, predictive algorithms, and sentiment analysis tools, are enhancing early detection and personalized care in mental health. The authors also discuss challenges including ethical concerns, data privacy, and lack of standardized evaluation methods. Overall, the paper concludes that AI holds great promise for improving accessibility and precision in mental health care but requires more rigorous, transparent, and ethically guided research for real-world implementation [2].

The study of "Model Optimization for Personalized Health Metrics Analysis" by Madhuranga Perera, Amandi Pathirana, Lakshan Gamaathige, Anuradha Jayakody provide a crucial epidemiological insight into the role of Body Mass Index (BMI) in predicting complex multimorbidity [3] "AI Powered Personalized Healthcare Recommender" published by Hirushit S, Mr. S. Raja, Suwetha S, Yazhini J provides multicohort servational study reveals a strong association between BMI and the risk of developing multiple chronic diseases, underscoring BMI as a central health metric in therapeutic planning. Further supporting this, Taieb conducted a systematic review exploring how different BMI categories relate to health complications [4].

The review of "Personalized Health Assistant with Reinforcement Learning" identifies threshold based health risks, suggesting that personalization must consider such stratified risk levels for optimal therapy design. On the fitness side, authores Jennifer Jin, Mira Kim, Soo Dong Kim developed a machine learning model to predict customized workout plans [5].

By training supervised models on user activity and fitness goals, the system recommends exercise routines tailored to the individual, enhancing safety and efficiency in physical training. Expanding on this concept, Yadav and Jadhav designed Workout Whiz, an AI-based fitness assistant that employs ensemble learning and feedback loops to continuously adapt workout routines based on user performance[16]. This innovation ensures that users stay engaged and aligned with their fitness goals through realtime personalization.

Anusari proposed SriHealth, a unified mobile health platform tailored to Sri Lankan lifestyle patterns [17]. It integrates APIs to deliver personalized diet plans, workout schedules, and yoga routines. The system showcases how therapeutic optimization can be packaged into an accessible, culturesensitive application with dynamic updates. Lastly, Gondocs explored the synergy between “AI prediction models and human judgment” in medical diagnosis [18].

By combining XGBoost with clinician input, the study demonstrates improved diagnostic outcomes, highlighting the role of machine learning as a decision support tool in therapeutic care. The main objective of this project is to design a framework capable of analyzing a variety of health metrics and applying optimization models to recommend therapy strategies. These could include medication adjustments, physical activity suggestions, dietary changes, sleep improvements, or mental health interventions. The system utilizes a combination of artificial intelligence, data mining, and optimization algorithms to process the input metrics and evaluate the most beneficial treatment path.

Health metrics such as heart rate, oxygen saturation, glucose levels, sleep cycles, physical activity, and stress indicators can provide deep insights into a person’s wellness. However, raw data alone is not sufficient. What’s needed is a system that not only collects data but interprets it meaningfully and recommends optimized therapeutic actions, tailored specifically to the individual.

III. METHODOLOGY

The methodology adopted in this research is centered on the integration of multimodal health data with advanced machine learning techniques to provide personalized therapeutic optimization. The process begins with the collection of diverse data sources, which include structured medical reports such as blood pressure, blood sugar, and cholesterol levels, as well as demographic factors like age, gender, height, weight, and region. Lifestyle attributes, including hydration and daily activity levels, are also incorporated to capture a holistic view of the individual’s health. In addition, dietary information is collected through two channels: manual input of preferences, allergies, and restrictions, and computer vision based meal analysis, where images of meals are processed to estimate caloric and nutrient content.

NLP techniques are further employed to extract relevant health indicators and instructions from unstructured text such as doctor’s notes or prescriptions. Feature extraction is carried out across modalities: convolutional neural networks extract nutritional values from food images, NLP models detect key terms from medical text, and structured data is transformed into meaningful health metrics such as BMI, body fat ratio, and hydration efficiency. This standardized dataset becomes the foundation for model training and optimization.

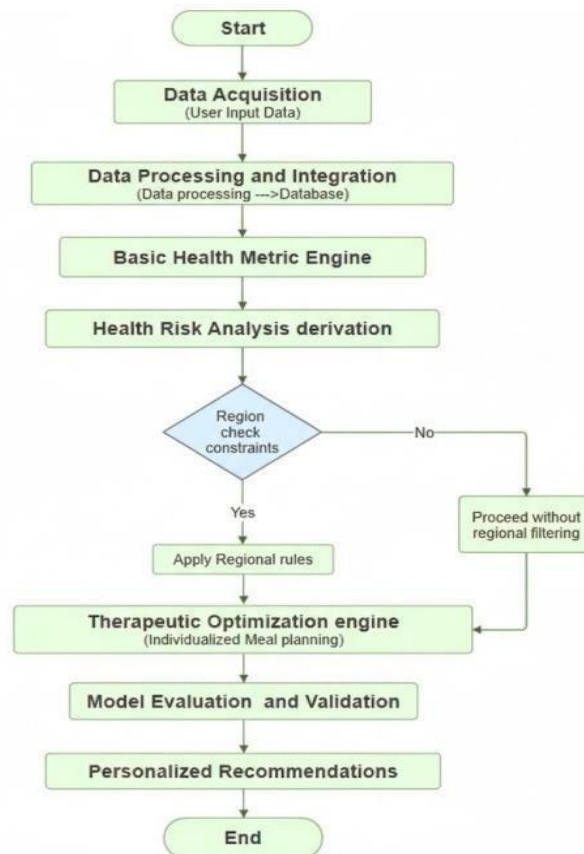


Fig : Flowchart Diagram of Therapeutic Optimization Model

It commences with Data Acquisition, where user input, including medical history and detailed lifestyle data, is gathered. This raw information is then processed, cleaned, and integrated into a central Database. The system then utilizes the Basic Health Metric Engine to calculate fundamental indicators like BMI and estimated caloric needs, which are then analyzed by the Health Risk Analysis derivation to establish the user's specific therapeutic goals and risk profile e.g., high risk for hypertension. A critical Region check constraint follows, determining if local dietary laws, ingredient availability, or cultural rules must be applied to ensure the plan is practical. All these refined parameters feed into the Therapeutic Optimization engine, which runs the algorithms necessary to formulate the Individualized Meal planning. Finally, the proposed plan undergoes rigorous Model Evaluation and Validation to ensure it is nutritionally sound and safe before the delivery of the final Personalized Recommendations to the user.

IV. SYSTEM DESIGN

The proposed AI-Driven Therapeutic Optimization framework integrates diverse health data sources, intelligent processing pipelines, and adaptive algorithms to deliver personalized healthcare recommendations. The architecture is modular, consisting of data processing, treatment optimization, functional modules, and algorithmic components.

1. Data Processing : Data processing transforms raw health data into structured, usable formats. The system collects information from EHRs, lab reports, lifestyle logs, and genetic profiles. Preprocessing techniques such as cleaning, normalization, and missing-value imputation ensure data quality. Feature extraction identifies key health indicators, while data integration combines multimodal datasets for a holistic view. Real-time IoT data is processed through scalable pipelines, and filtering algorithms remove noise and outliers, supporting biomarker analysis and predictive modeling.

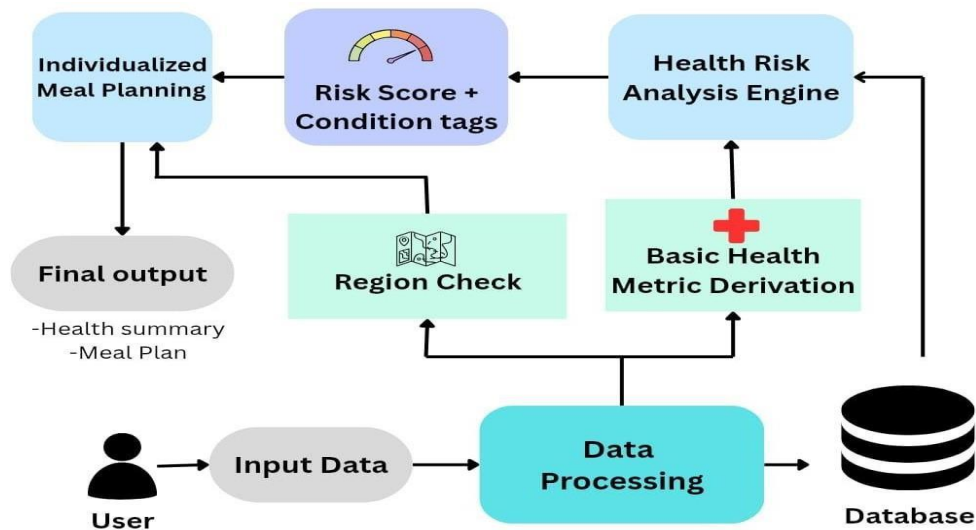


Fig : System Architecture of Therapeutic Optimization Model

2. **Treatment Optimization** : Treatment optimization bridges patient specific insights and actionable interventions. Machine learning models predict responses to therapies, while optimization algorithms generate personalized recommendations considering physiological, genetic, and lifestyle factors. Continuous feedback from devices allows dynamic updates to treatment plans, supporting safe dosing, scheduling, and lifestyle guidance. Cost effectiveness and resource efficiency are also incorporated to improve clinical outcomes.

3. Algorithms Used :

- a. **Rule-Based Algorithm** : Maps patient conditions to dietary and lifestyle interventions for transparency and baseline recommendations.
- b. **Natural Language Processing** : Processes medical text to extract symptoms, diagnoses, and treatment history, supporting multilingual data and structured outputs for analysis.
- c. **Image Recognition** : Scans food images to identify items, classify categories, and retrieve nutritional data, enabling accurate diet tracking.
- d. **Nutrition Mapping** : Aligns dietary intake with health metrics using standardized nutrition databases, generating precise macronutrient and micronutrient profiles for personalized recommendations.

V. DISCUSSION

Therapeutic optimization system significantly advances the personalized nutrition field by establishing a therapeutic core, moving beyond generic calorie counting to a comprehensive, goal- oriented risk management system. Our methodology is distinctive because it integrates static medical and lifestyle input with a dynamic analysis system. Specifically, the system utilizes the Health Risk Analysis derivation to define a specific therapeutic objective eg.blood sugar management and then employs the Therapeutic Optimization Engine to meet that objective.

The crucial element for real-world efficacy is the inclusion of the Regional Constraint Check, which is often overlooked in existing models. Its proactive solution to the "last mile problem" in diet recommendation through the Regional Constraint Check. By integrating localized food databases, seasonal data, and cultural norms, the system ensures that the personalized meal plan is not just theoretically sound but is practically actionable by the user in their specific geographic location.

VI. CONCLUSION

Therapeutic optimization for personalized health metric analysis survey shows that while AI is making big pace in healthcare, most of the attention is going to areas like mental health and precision medicine. These are important, but lifestyle related diseases like diabetes, obesity, and heart conditions are rising fast and still don't get enough focus from AI projects.

Addressing this gap is crucial and are largely preventable with timely intervention. We found that many existing systems are built to do just one thing, like tracking calories or analyzing medical scans. What's missing is a complete, personalized tool that connects the dots: your health history, your habits, your food choices, and your long-term risks. To tackle this issue, we've designed a unified framework that brings together nutrition, risk prediction, and user-specific insights.

VII. FUTURE SCOPE

1. **Wearable Tech Integration** : Connect with smartwatches and fitness trackers to pull real-time activity, sleep, and health data, enabling the optimization engine to make dynamic, day-to-day meal adjustments.
2. **Advanced Ingredient and Sourcing Integration** : Link the system to local grocery store APIs or meal kit services to generate personalized shopping lists or facilitate the direct purchase of plan- aligned ingredients.
3. **Telehealth/Expert Consultation Bridge** : Create a secure portal to share the generated reports and progress data directly with a registered dietitian or physician, facilitating expert review and collaborative care.
4. **Emotional and Stress Driven Eating Analysis** : Integrate mood and stress tracking with food logging to identify unhealthy coping patterns. The system can then offer timely, behavioral nudges or alternative, therapeutic food suggestions.
5. **Instant Nutritional Fact Lookup Barcode Scanner** : Integrate the image feature to work as a barcode scanner for packaged foods. Users can quickly scan an item to see its exact nutritional facts and instantly verify it against their dietary constraints.

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