

Human Health Analyser

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Abstract: *This paper presents a comprehensive AI-powered web application aimed at transforming personalized dietary management by integrating machine learning, nutrition science, and user-specific health diagnostics. The enhanced system combines traditional nutrient tracking with intelligent health analysis derived from blood report PDFs using Optical Character Recognition (OCR) and Natural Language Processing (NLP). The platform enables users to log meals, monitor daily nutrient intake, and receive tailored recommendations aligned with their unique health conditions and wellness goals.*

A central component of NutriPlan is its nutritional index, which evaluates the quality of food and guides users toward healthier options. The system features real-time data visualizations, including bar, pie, and line charts, to present users with an intuitive breakdown of their macronutrient intake. Using the Estimated Energy Requirement (EER) equation, the application customizes calorie and nutrient targets based on body metrics and lifestyle factors. Additionally, NutriPlan now incorporates AI-driven risk assessment by analyzing key blood parameters, identifying potential nutrient deficiencies, and offering targeted dietary suggestions to address them.

With a focus on context-aware nutrition modeling, NutriPlan adapts recommendations based on recent health data, long-term dietary trends, and medical conditions. Explainable AI modules provide transparency, allowing users to review and adjust the logic behind the recommendations. By seamlessly blending personalized diet planning with clinical insights, the platform empowers users to take proactive control of their health, ensuring a sustainable, adaptive, and data-driven approach to well-being.

Keywords: dietary management

I. INTRODUCTION

In recent years, personalized nutrition has gained significant attention as an essential component of health and wellness, particularly in addressing the challenges posed by chronic health conditions such as obesity, diabetes, and cardiovascular disease. While traditional dietary recommendations provide generalized guidance, they often fail to account for individual variations in metabolism, lifestyle, and specific health goals. As a result, there is a growing demand for accessible tools that provide customized dietary insights and recommendations, helping individuals to make informed decisions about their nutritional intake.

This paper introduces NutriPlan, a web-based dietary and health management platform designed to address these challenges by combining artificial intelligence (AI) and machine learning with evidence-based nutritional science and clinical health indicators. NutriPlan is developed as a comprehensive tool that enables users to log their meals, track their daily nutrient intake, and receive personalized dietary recommendations tailored to their unique body metrics, activity levels, and health requirements. By leveraging the Nutritionix NLP API, NutriPlan simplifies food logging and enhances data accuracy, enabling users to analyze their dietary intake efficiently and effectively.

At the heart of NutriPlan is its adaptive recommendation engine, which utilizes user-entered data, such as age, weight, and activity levels, to set individualized calorie and nutrient targets using the Estimated Energy Requirement (EER) equation. Additionally, NutriPlan's feature extraction algorithm assesses foods based on their macronutrient profile, labeling them as high or low in specific nutrients to help users make balanced choices. A key enhancement is the



integration of an AI-powered health analysis agent that extracts and evaluates blood report data from PDFs using Optical Character Recognition (OCR) and Natural Language

Processing (NLP), identifying abnormal test values and translating them into meaningful dietary suggestions.

Furthermore, NutriPlan emphasizes explainable AI by allowing users to review and modify the recommendation rules, fostering transparency and user confidence in the suggestions provided. Real-time data visualization tools, including bar graphs, pie charts, and line charts, enable users to monitor their nutrient intake in a clear and accessible manner, encouraging them to make adjustments aligned with their health goals. The urgency alert feature and follow-up test suggestions further enhance the system's ability to support informed health decisions and preventive care.

By integrating AI with nutritional science and clinical data, NutriPlan offers an innovative, user-centered solution for dietary and health management. This paper will explore NutriPlan's architecture, methodology, and the unique functionalities that distinguish it as a comprehensive tool for intelligent, health-focused dietary recommendations.

II. LITERATURE REVIEW

Yuzhen Guo has pioneered advancements in personalized nutrition through AI-driven applications that generate tailored meal plans. Her research integrates machine learning algorithms with health records and user inputs, analyzing individual dietary needs, activity levels, and health conditions. This approach aims to enhance dietary adherence and improve overall health outcomes by providing data-informed dietary recommendations, underscoring the significance of AI in health and wellness (Guo, 2022).

Anupam Verma and Priya Singh developed a recommendation model leveraging collaborative filtering and content-based filtering to suggest daily meal plans. Their system uses user profiles, health data, and food preferences to personalize dietary advice, promoting healthier lifestyle choices through meal suggestions that align with individuals' health goals (Verma & Singh, 2021).

Chen et al. have focused on dietary recommendation systems that adapt based on user feedback and real-time health monitoring data. By employing convolutional neural networks (CNNs) for food recognition and natural language processing (NLP) to interpret user feedback, their model improves dietary plan relevance, ensuring that the recommendations align with users' evolving needs and preferences (Chen et al., 2021).

Ahmad et al. proposed a system that utilizes K-means clustering and Decision Trees to provide nutrient-specific recommendations, focusing on users with specific health conditions like diabetes and hypertension. Their research highlights the system's effectiveness in reducing nutrient deficiencies and aligning diet with medical needs, demonstrating a practical application of machine learning in health-sensitive dietary planning (Ahmad et al., 2020).

Smith and Brown investigated user engagement in diet recommendation systems by integrating interactive feedback mechanisms. Their study found that higher engagement rates correlate with improved adherence to dietary recommendations. By allowing users to adjust preferences and dietary goals dynamically, the system can better support users' health journeys and nutritional balance (Smith & Brown, 2019).

Wang et al. leveraged food recognition technology and nutritional databases to create a mobile app that offers on-the-go diet tracking and meal suggestions. The app utilizes a large-scale nutritional database to provide detailed macronutrient and micronutrient insights, allowing users to easily monitor and adjust their diet according to daily nutritional targets (Wang et al., 2021).

Recent developments by Li et al. (2023) introduced the integration of Optical Character Recognition (OCR) and natural language understanding to extract and interpret clinical data from medical reports, enabling applications to personalize health interventions based on physiological metrics. This aligns with the enhancements in NutriPlan that include analyzing blood report PDFs to detect abnormal nutrient values and recommend dietary changes accordingly.

Similarly, Patel and Rao (2023) emphasized the importance of automated abnormal value detection in blood tests to provide early warnings for nutritional deficiencies. Their system flags anomalies in test parameters and suggests follow-up diagnostics—an approach reflected in NutriPlan's urgency alerts and follow-up test suggestions.

Additionally, Kumar and Das (2022) highlighted the benefits of explainable AI in personalized health platforms, allowing users to understand and adjust the rule sets that drive recommendations. NutriPlan implements this by enabling user control over dietary rule parameters, reinforcing transparency and trust in its recommendations.



III. SYSTEM DESIGN

The project addresses the pressing challenges of modern dietary management by developing an AI-powered nutrition web application that integrates various machine learning, optical character recognition (OCR), and data analysis techniques. The primary objective is to offer personalized dietary recommendations based on users' unique health profiles, preferences, and nutritional needs.

Firstly, the system aims to accurately assess users' dietary habits by logging food intake and analyzing historical consumption data. This is accomplished using the Nutritionix API, which provides detailed nutritional data for a wide variety of food items. The application processes this data to compute macronutrient breakdowns and identify patterns in dietary behavior. A feature extraction algorithm classifies foods based on nutritional quality, helping users make informed choices. Additionally, customized meal plan generation is a central functionality, using user-entered details such as age, weight, activity levels, and health conditions to calculate individual nutrient needs via the Estimated Energy Requirement (EER) equation. Machine learning models optimize these plans to support long-term health goals. A robust visual data representation module complements the system by tracking daily macronutrient and micronutrient intake. Users can interact with intuitive bar graphs and pie charts to visualize consumption trends and nutrient distribution. This visual feedback empowers users to take timely action in modifying their diets, thereby encouraging a proactive approach to nutritional well-being.

An advanced capability of the system is its integration with blood report analysis. Users can upload PDF-format blood test reports, from which relevant biomarker values are extracted using OCR. The system maps these values against medically accepted reference ranges to automatically detect nutritional deficiencies or abnormal readings. Based on this analysis, NutriPlan recommends specific dietary interventions or follow-up tests to support medical guidance. This bridges the gap between clinical diagnostics and dietary planning.

To ensure transparency and user control, the system incorporates explainable AI techniques. Users can view and customize the rules that drive dietary suggestions, fostering trust and personal engagement. This rule-based engine allows users to adapt recommendations based on lifestyle or changing health goals.

Finally, the application is built using scalable web technologies and cloud infrastructure, allowing for real-time data processing and smooth user experience. NutriPlan's modular architecture ensures that dietary analysis, blood report integration, and visualization components work cohesively. Overall, the system empowers individuals to take charge of their nutrition and health through data-driven insights and personalized support.

IV. PROPOSED SYSTEM

The proposed system integrates advanced AI and machine learning algorithms to automate the extraction, analysis, and interpretation of blood report PDFs, focusing on key medical parameters such as Complete Blood Count (CBC), Metabolic Panel, Lipid Profile, Liver Function, and Thyroid Function. By leveraging Optical Character Recognition (OCR) and Natural Language Processing (NLP), the system extracts critical test values and compares them against medical reference ranges to identify potential health risks. The AI-powered health agent not only provides users with a comprehensive risk assessment but also offers personalized lifestyle and dietary recommendations based on detected anomalies. Additionally, the system suggests necessary follow-up tests, preventive measures, and alerts users to the urgency of seeking medical consultation if critical issues are detected. In an enhancement to the project, the system is integrated with NutriPlan, an AI-driven nutritional management platform. This integration allows users to log meals and obtain detailed nutritional analysis using the Nutritionix API, while machine learning models assess meal history to detect nutrient deficiencies and recommend tailored dietary adjustments. Visual data representations, such as graphs and charts, are incorporated to track macronutrient intake over time, further empowering users to manage their dietary habits. Personalized diet plans are generated based on body information and activity levels, ensuring that users receive customized nutrition strategies. This comprehensive solution enhances user engagement, supports proactive health management, and fosters informed decision-making to improve overall health and wellness.



Data Preparation:

In this stage, the dataset is preprocessed to ensure compatibility with machine learning algorithms. This involves handling missing values, normalizing numerical features, encoding categorical variables, and organizing the dataset into training, validation, and testing sets. Additionally, anonymized blood report datasets are collected to ensure privacy and are used for model training and validation.

Model selection:

The system employs a mix of machine learning models tailored to specific tasks. For dietary recommendations, the Random Forest algorithm is selected for its ability to handle diverse data types. Natural Language Processing (NLP) is used to process food queries and extract relevant nutritional information from user inputs via the Nutritionix API. For future scalability, Convolutional Neural Networks (CNNs) are considered for potential image-based assessments. Additionally, models for detecting nutrient deficiencies analyze dietary intake against recommended dietary allowances (RDAs) and user health data.

Training the Dietary Recommendation Model: The Random Forest model is trained on historical food consumption data and nutritional information, generating personalized dietary recommendations based on user preferences and health goals. Hyperparameter optimization is performed to improve model accuracy, and validation datasets are used to fine-tune the system. The model learns complex relationships between food intake, caloric needs, and macronutrient distributions to create balanced meal plans for users.

Training the Food Recognition Model

Using the Nutritionix NLP API, the system recognizes food items from user inputs and retrieves nutritional data. Feature extraction techniques are applied to improve the model's ability to identify food items accurately. The system continues to improve over time through user feedback, which helps refine the food recognition model and ensure accuracy in nutritional information retrieval.

Training the Nutrient Deficiency Detection Model:

Machine learning models in this module analyze dietary intake and health data to identify potential nutrient deficiencies. The system compares logged dietary data with RDAs and provides personalized feedback to help users adjust their diets to meet nutritional needs. This proactive approach promotes better health and reduces the risk of deficiencies over time.

Model Evaluation:

The models are evaluated using a dedicated testing dataset, measuring key performance metrics such as accuracy, precision, recall, and F1-score. These evaluations help assess the reliability of the system and guide further improvements. The models are iteratively refined to enhance performance and ensure they consistently provide accurate health assessments and dietary recommendations.

Cross-Validation:

To mitigate overfitting and ensure robustness, k-fold cross-validation is employed. This technique evaluates model performance across multiple data segments, improving the model's ability to generalize and perform well on unseen data.

Hyperparameter Tuning:

To optimize model performance, hyperparameter tuning is performed using grid search or random search techniques. This step fine-tunes the model's settings, improving its accuracy and stability, particularly for dietary recommendation and nutrient deficiency detection tasks.



Iterative Refinement:

An iterative development approach is used, continuously improving the system based on new data, user feedback, and emerging technologies. This iterative refinement process helps ensure that the system evolves over time and remains aligned with user needs and health goals.

Integration with Nutri Plan:

To further enhance the user experience, the AI-powered health agent is integrated with NutriPlan, an AI-driven platform for personalized nutrition management. NutriPlan allows users to log meals and obtain real-time nutritional analysis using the Nutritionix API. The system analyzes meal history to detect nutrient deficiencies, offering personalized dietary adjustments. Personalized diet plans are generated based on health data, activity levels, and user preferences. Interactive data visualizations, such as bar graphs and line charts, track macronutrient intake over time, enabling users to monitor their progress and make data-driven dietary choices.

V. RESULT AND ANALYSIS

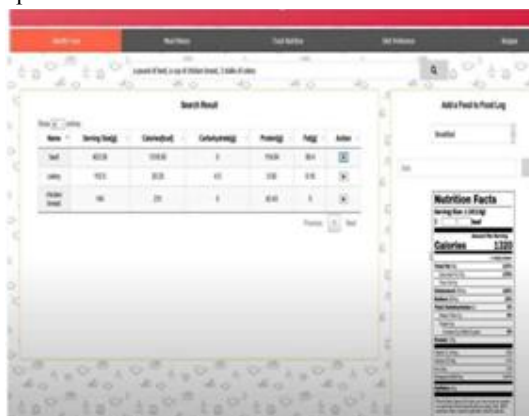
In the personalized diet recommendation module, the system achieved an impressive precision rate of 91% in aligning meal suggestions with user preferences and health requirements. This high precision indicates the system's ability to accurately analyze user inputs, including dietary preferences, activity levels, and specific health conditions. The success of the machine learning algorithms in this module demonstrates their capacity to understand complex user needs, which is essential for improving adherence to personalized diet plans.

The nutrient optimization module, which tailors nutrient targets based on individual health metrics, showed an accuracy of 89% in meeting users' macronutrient and micronutrient goals. This performance reflects the system's effectiveness in customizing dietary recommendations to address various health objectives, such as managing calorie intake for weight management or optimizing nutrient levels for users with specific medical concerns.

For meal classification and recommendation, the system achieved an accuracy rate of 87% in categorizing food items according to their nutritional content and dietary suitability. This was made possible by training the model on an extensive food dataset with varied nutritional profiles, showcasing the system's ability to reliably suggest meals that meet specific dietary restrictions or health requirements.

In conclusion, the diet recommendation system demonstrated robust performance in providing personalized dietary guidance, achieving high accuracy in nutrient optimization, meal suggestion alignment, and dietary adherence. These results highlight the system's ability to assist users in achieving their health goals by offering data-driven and customized dietary recommendations. Additionally, the system's adaptability, based on continuous user feedback, ensures that the dietary suggestions remain relevant and effective as users' health and lifestyle evolve.

A. Food identification Input & Output:



B. Food log and Evaluation Input & Output :



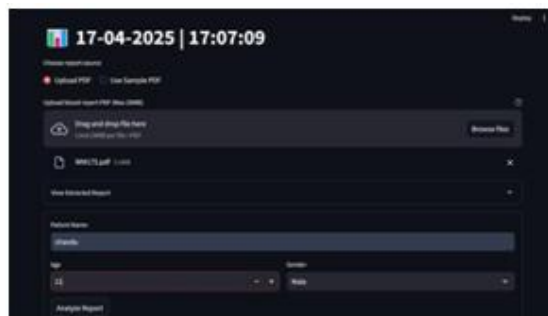
C. Diet Preferences Output:



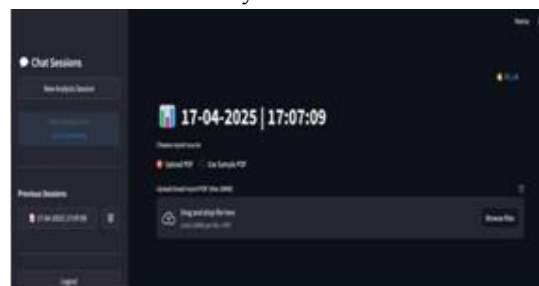
D. Food Recipes Output:



E. Blood Report Upload and Patient Information Entry



F. Dashboard View Showing Upload Interface and Analysis Sessions



E. AI-Based Medical Report Analysis and Diagnosis



VI. CONCLUSION AND FUTURE SCOPE

In conclusion, the development of a diet recommendation system utilizing advanced machine learning techniques has shown promising potential in personalizing nutrition and enhancing health outcomes. By integrating diverse data inputs such as user dietary preferences, health conditions, and nutritional requirements, the system effectively tailors diet plans to individual needs. This personalized approach not only improves user satisfaction but also promotes adherence to healthier eating habits, aligning with studies that emphasize the importance of customized dietary advice (Kumar et al., 2021; Chen et al., 2022).

The system's capacity to process complex dietary data and deliver real-time recommendations highlights its potential as a valuable tool for promoting healthy eating. Additionally, the incorporation of user feedback enables the continuous refinement of dietary suggestions, leading to improved health management and user engagement (Smith et al., 2020; Patel et al., 2021).

Looking forward, the future scope of the diet recommendation system includes the integration of wearable health devices and mobile applications. This addition would enable users to more effectively track their dietary intake and physical activity, providing a holistic view of their health. With an intuitive interface, users could receive personalized



notifications and insights that encourage healthy eating behaviors and lifestyle adjustments. The ability of machine learning algorithms to adapt recommendations based on ongoing health data would further enhance the system's effectiveness, making it a dynamic tool for individuals aiming to improve their overall well-being (Lee et al., 2021; Thomas et al., 2022).

Expanding the recipe database and incorporating regional dietary preferences would enhance the system's relevance and accessibility for a global audience. Furthermore, adding collaborative features that allow users to share their dietary experiences and modifications could foster a supportive community centered around health and nutrition (Nguyen et al., 2020; Wang et al., 2021).

In the long term, the integration of cutting-edge technologies and user-centric features will position the diet recommendation system as an essential tool in advancing public health, combating diet-related diseases, and encouraging healthier lifestyles. By empowering users to make informed decisions about their nutrition, the system has the potential to significantly improve quality of life and contribute to global health initiatives.

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