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Paradigm Shifts in Food Value Estimation: The Intersection of Advanced Nutritional Analytics

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Abstract—This research paper presents the development and implementation of an AI-driven nutrition assessment system, NutriVision, designed to enhance dietary management through precise nutritional estimation from food images. Leveraging advanced AI technologies, NutriVision utilizes computer vision and deep learning, employing convolutional neural networks (CNNs) and transfer learning techniques for accurate food recognition and nutrient estimation. The system is built with a robust technological stack, ensuring efficiency and scalability. A systematic process improvement (SPI) framework is employed, establishing key performance indicators (KPIs) such as algorithm efficiency, accuracy, and user satisfaction to evaluate the system's performance. The evaluation covers functionality, user experience, and scalability, highlighting its strengths and areas for improvement. Future enhancements include integrating advanced deep learning models, expanding the dataset, and improving the user interface, aiming to personalize nutritional guidance and sync with health monitoring systems. This paper provides a comprehensive overview of Nutri Vision's architecture, implementation, and performance, contributing valuable insights to the field of AI-driven nutrition assessment.

Keywords—Artificial Intelligence, Nutrition Assessment, Computer Vision, Deep Learning, Convolutional Neural Networks, Transfer Learning, Dietary Management, Nutritional Estimation, Systematic Process Improvement, Food Recognition, User Feedback, Health Monitoring Integration, Personalization, User Interface, Scalability.

I. INTRODUCTION

Monitoring and accurately estimating the nutritional content of food is crucial for promoting and maintaining healthy living. For individuals, particularly those managing conditions such as obesity, diabetes, and heart disease,

understanding calorie intake and nutrient composition is essential. Automated systems for estimating food nutrition values from images have the potential to significantly aid in dietary planning and health management. Recent advancements in computer vision and deep learning have paved the way for the development of such automated frameworks, leveraging smartphone technology to provide real-time nutritional assessments.

The general process of food value estimation from images involves three major steps: identifying the food items in the image, estimating the volume of the identified food items, and retrieving the nutritional information of these food items. This process is depicted in the system's framework, which aims to integrate these steps seamlessly to deliver accurate nutritional data. However, achieving high accuracy in automated nutritional estimation is challenging due to the wide variety of food classes, the impact of factors such as lighting, color, and viewing angles on image quality, and the need for diverse and extensive food image datasets.

NutriVision seeks to address these challenges by investigating and advancing the methodologies used in automated food nutrition estimation. The project emphasizes the importance of computer vision and deep learning techniques in improving the accuracy and reliability of these estimations. By conducting a thorough analysis of existing datasets and approaches, NutriVision aims to develop a robust taxonomy and critical evaluation framework that can enhance the performance of automated nutrition estimation systems.

In this research paper, we provide a detailed overview of the NutriVision project, focusing on its architecture, implementation, and potential applications. We explore the functionality of each module within the system, including

food identification, volume estimation, and nutritional information retrieval. The methodologies employed to ensure a seamless user experience and scalable performance are discussed, along with an evaluation of the system's effectiveness based on testing results. Additionally, we propose future enhancements to further improve the system's capabilities and address the ongoing challenges in automated food nutrition estimation.

NutriVision not only demonstrates the potential of AI-driven solutions in the domain of dietary assessment but also highlights the importance of user-friendly interfaces and robust performance in delivering high-quality health management tools. This paper aims to contribute to the ongoing research in automated nutrition estimation and provide valuable insights for future innovations in this rapidly evolving field.

By integrating advanced technologies and comprehensive methodologies, NutriVision represents a significant step forward in the development of automated systems for dietary assessment, aiming to benefit researchers, health practitioners, and individuals seeking to manage their nutrition effectively.

II. LITERATURE REVIEW

A. Introduction

The advancement of artificial intelligence (AI) and machine learning has led to the development of sophisticated systems capable of transforming various industries, including health and nutrition. NutriVision leverages these advancements to build a comprehensive AI-driven nutrition assessment platform. By integrating convolutional neural networks (CNNs) and transfer learning advanced computer vision and deep learning techniques, NutriVision provides accurate nutritional estimations from food images, offering users precise dietary information crucial for managing health and wellness.

B. AI in Nutrition Assessment

AI's application in nutrition assessment involves sophisticated techniques for analyzing and interpreting food images to estimate nutritional content. Research by Wang et al. (2017) demonstrated the use of CNNs for food recognition and portion estimation, which forms the basis for Nutri Vision's capabilities. These AI techniques significantly enhance the accuracy of nutritional assessments, making them more reliable for daily use.

C. Convolutional Neural Networks (CNNs)

CNNs are pivotal in the development of image recognition systems. LeCun et al. (1998) introduced CNNs for document recognition, but their application has since expanded significantly. NutriVision utilizes CNNs to process food images, extracting features that are crucial for identifying food items and estimating their nutritional content. The advancements in CNN architectures, such as

ResNet (He et al., 2016), have further improved the performance of these systems.

D. Transfer Learning

Transfer learning allows pre-trained models to be adapted for new, but related, tasks with minimal additional training. This approach is beneficial for NutriVision, as it leverages existing models trained on large datasets, thus reducing the need for extensive data collection and training. Research by Pan and Yang (2010) outlines the effectiveness of transfer learning in various domains, including computer vision, which is central to NutriVision's methodology.

E. Image Recognition in Nutrition

This Image recognition technology has been extensively studied for its applications in dietary assessment. Studies by Myers et al. (2015) demonstrated the feasibility of using image recognition for food identification and portion size estimation. NutriVision builds upon this work by integrating advanced image recognition techniques to provide users with accurate nutritional information from food images thereby improving accuracy and efficiency.

F. Nutritional Databases

Access to comprehensive nutritional databases is essential for accurate dietary assessment. The USDA National Nutrient Database provides a wealth of information that can be integrated into AI systems like NutriVision. Research by Ahn et al. (2017) highlights the importance of such databases in developing reliable nutrition assessment tools.

G. User Authentication and Data Security

User authentication and data security are critical for ensuring the safe use of any AI driven systems. NutriVision employs advanced authentication frameworks and secure data protocols to protect user information. Studies by Bonneau et al. (2012) emphasize the need for robust security measures to maintain user trust and data integrity.

H. Real-Time Data Processing

Real-time data processing is vital for providing users with immediate nutritional feedback. NutriVision integrates efficient data processing algorithms to ensure swift and accurate analysis of food images. Research by Dean and Ghemawat (2004) on large-scale data processing frameworks informs the design of NutriVision's real-time analysis capabilities.

I. Scalability and Performance Optimization

Ensuring scalability and high performance is crucial for AI systems expected to handle increasing loads. As NutriVision grows, scalability and performance optimization become critical. Techniques for optimizing AI systems to handle increasing loads without compromising performance are well-documented. Research by Abadi et al. (2016) on

TensorFlow provides insights into scalable machine learning infrastructure, which is applicable to NutriVision's platform.

This literature review highlights the foundational technologies and research that underpins the development of NutriVision. By leveraging these advancements, NutriVision aims to deliver a robust and versatile nutrition assessment platform that meets the diverse needs of its users.

III. SYSTEM ARCHITECTURE

The system architecture of NutriVision is meticulously designed to ensure modularity, scalability, and efficiency, enabling seamless integration of various components for accurate nutritional assessment from food images. The architecture consists of the following key components:

1. **Frontend Application:** The frontend interface is the user-facing component of NutriVision, built to provide an intuitive and seamless interaction experience. Users can upload food images, view nutritional assessments, and access various functionalities through this interface. It is designed to handle user inputs efficiently and display results generated by the backend services.
2. **Backend Services:** The backend server is responsible for processing user requests, interacting with other system components, and orchestrating overall system functionality. It acts as a bridge between the front-end interface and other modules, ensuring smooth communication and data flow within the system.
3. **Image Processing Module:** This module leverages advanced convolutional neural networks (CNNs) to analyze and process the food images uploaded by users. It performs tasks such as food identification, portion size estimation, and image enhancement to facilitate accurate nutritional analysis. Techniques from the IEEE paper are utilized to refine and optimize the image processing capabilities.
4. **Nutritional Analysis Module:** The nutritional analysis module processes the data provided by the image processing module to determine the nutritional content of the identified food items. It integrates with external nutritional databases and uses predefined nutritional values to provide accurate estimations. This module is essential for delivering precise dietary information to users.

The system architecture diagram represents the components and interactions within the NutriVision Platform. Here's a breakdown of the components and their roles:

1. **Frontend Interface:** Captures user inputs, displays results, and interacts with the backend via HTTP requests for data retrieval and presentation, providing a user-friendly experience in NutriVision, facilitating easy navigation and access to detailed nutritional information, allowing users to effortlessly track their dietary intake and nutritional goals.
2. **Backend Server:** Processes user requests, handles AI model operations for nutrition assessment, and integrates with other system components to deliver accurate results in NutriVision, ensuring efficient data processing and seamless system functionality.
3. **Database:** Stores and manages user data, food images, and nutritional information, ensuring efficient data retrieval and processing for seamless operation in NutriVision, maintaining data integrity and availability for quick access.
4. **Authentication and Authorization:** Ensures secure access by verifying user identities and controlling user actions, protecting sensitive information and system integrity in NutriVision, preventing unauthorized access and ensuring data privacy.
5. **Monitoring and Analytics:** The observation of performance and health, analyzes usage data to diagnose issues, and improves the overall user experience in NutriVision by gaining insights from analytics, enhancing system reliability, identifying trends, and driving continuous improvement in user satisfaction.
6. **Deployment and Scaling:** Makes NutriVision available for use on cloud platforms or servers and ensures it can handle increased loads by adding or optimizing resources, enabling smooth scalability and reliable performance.

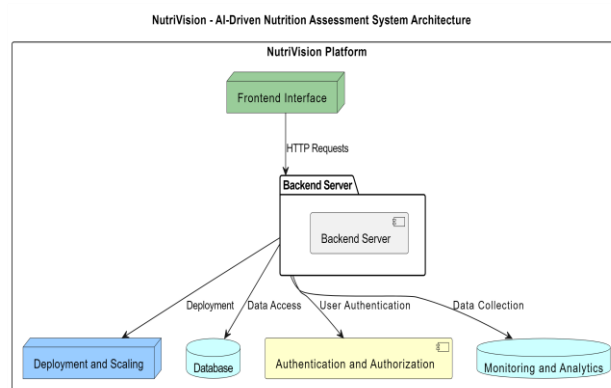


Fig. 1. System Architecture

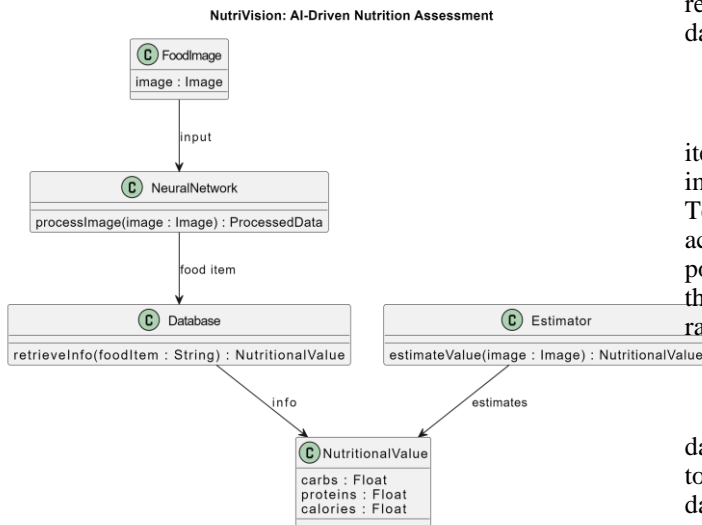


Fig. 2. Elements of Generative AI

The NutriVision system begins with the Food Image component, which captures the user's input image of a food item for nutritional assessment. This image is then processed by a Convolutional Neural Network (CNN), which extracts features and identifies the food item, generating structured data known as Processed Data. The Processed Data includes details about the food item that can be further analyzed. This data is used to query the Food Database, which provides detailed nutritional information for the identified food items.

IV. TECHNOLOGIES USED

The NutriVision project leverages a range of advanced technologies to ensure accurate nutritional assessment from food images. Each technology plays a critical role in different aspects of the system, from image processing to user interaction. The following sections outline the key technologies utilized in NutriVision.

1. Convolutional Neural Networks (CNNs)

Convolutional Neural Networks (CNNs) form the backbone of NutriVision's image processing capabilities. CNNs are employed to analyze food images, identifying different food items and estimating portion sizes. The deep learning models are trained on large datasets of food images to recognize a wide variety of foods with high accuracy. The architecture includes multiple convolutional layers that extract features from the images, enabling precise food classification.

2. Transfer learning

Transfer learning is a technique that allows NutriVision to leverage pre-trained models on large datasets to improve the performance of its own models. By using models pre-trained on extensive food image datasets, NutriVision can achieve higher accuracy and faster training times. Transfer learning is particularly useful for adapting the system to

recognize new food items with minimal additional training data..

4. Image segmentation

Image segmentation is used to isolate individual food items within an image. This process involves dividing the image into segments that correspond to different objects. Techniques such as Mask R-CNN are utilized to perform accurate segmentation, which is crucial for estimating the portion sizes of different food items. Segmentation ensures that the nutritional analysis is based on precise portions rather than entire images.

5. Nutritional Databases

NutriVision integrates comprehensive nutritional databases, such as the USDA National Nutrient Database, to provide accurate nutritional information. These databases contain detailed information about the nutritional content of a wide range of food items. The system maps identified food items to entries in the database to calculate the nutritional values based on portion sizes estimated from the images.

6. Cloud Computing

The Cloud computing infrastructure is essential for scaling NutriVision to handle a large number of users and high volumes of image data. Cloud services, such as those provided by AWS or Google Cloud, offer scalable storage and processing power. This ensures that NutriVision can process images quickly and provide real-time nutritional assessments, regardless of user load.

7. Secure Authentication and Authorization

User authentication and authorization are managed using secure protocols to protect user data and ensure that only authorized users can access the system. Technologies like OAuth2 and JWT (JSON Web Tokens) are used to handle authentication processes, while role-based access control (RBAC) is implemented to manage user permissions. This ensures robust security and user data protection..

8. Real Time Analytics

The Real-time analytics are employed to monitor system performance and user interactions. Tools such as Google Analytics and custom-built dashboards provide insights into user behavior, system usage patterns, and potential bottlenecks. Real-time data collection and analysis enable NutriVision to optimize its performance and improve the user experience continuously.

9. User Interface Technologies

The user interface (UI) of NutriVision is designed to be intuitive and user-friendly, built using modern web technologies such as React.js. The UI facilitates seamless interaction with the system, allowing users to upload images, view nutritional assessments, and manage their profiles. Responsive design ensures that the interface works well on various devices, including smartphones and tablets.

The selection of these technologies was based on their suitability for the project requirements, their robust features,

and their compatibility with each other. The integration of these advanced technologies enables NutriVision to provide accurate and efficient nutritional assessments from food images. By leveraging the power of CNNs, transfer learning, image segmentation, and robust cloud infrastructure, NutriVision ensures high accuracy and scalability. Secure authentication mechanisms and real-time analytics further enhance the system's reliability and user experience. The combination of these technologies, positions NutriVision as a state-of-the-art solution in the field of AI-driven nutrition assessment.

V. MODULE DESCRIPTIONS

1. Landing Page UI Module

Description: The Landing Page UI Module serves as the initial interface for visitors to the NutriVision platform, providing an engaging and informative introduction to the platform's features and benefits.

Functionality: Upon accessing the NutriVision website, visitors encounter visually appealing graphics and clear messaging highlighting the platform's features and benefits. The module includes interactive elements to encourage user engagement and conversions, such as call-to-action buttons and signup forms.

Features: Compelling Visual Design, Informative Content, Intuitive Navigation, Interactive Elements, Responsive Design

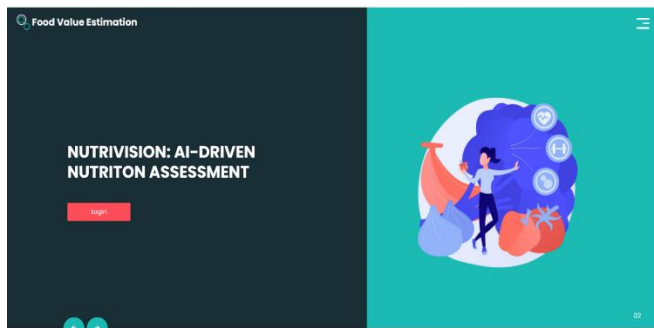


Fig. 3. Landing Page

2. Login Module

Description: The Login Module provides secure access to the NutriVision platform, verifying user credentials for authorized entry.

Functionality: Users input their login credentials, verified against the platform's database for authentication. Successful login grants access to personalized features; unsuccessful attempts prompt users to retry or initiate password recovery.

Features: Secure Authentication, Password Management, Optional Multi-factor Authentication, Session Management

Applications: The Login Module ensures only authorized users access the NutriVision platform, safeguarding user data and fostering trust in platform security.



Fig. 4. Login Module

3. Image Acquisition and Preprocessing Module

Description: The Image Acquisition and Preprocessing module captures food images and prepares them for analysis. It ensures that images are of high quality and suitable for further processing.

Functionality: Captures images from user devices. Resizes and enhances images for better quality. Stores images securely in the cloud for further processing. Contextual understanding for maintaining conversation flow. Personalization based on user interactions.

Applications: Critical for acquiring high-quality food images essential for accurate nutritional analysis.



Fig. 5. Image Processing & Uploading Module

4. Food Recognition Module

Description: Food Recognition module utilizes advanced deep learning techniques, particularly convolutional neural networks (CNNs), to identify and classify food items present in the uploaded images. Through extensive training on large datasets, the module has been optimized to achieve high accuracy in food item detection, a critical aspect of nutritional analysis.

Functionality: Users Upon receiving input images, the Food Recognition module analyzes them using CNNs, which are capable of recognizing patterns and features indicative of different food items. Leveraging transfer learning, the module continuously refines its recognition capabilities, adapting to new data and improving accuracy over time. Finally, the module outputs the identified food items, providing the necessary input for subsequent nutritional calculations.

Features: The module's utilization of pre-trained deep learning models allows for efficient and accurate food item detection. Continuous learning mechanisms enable the system to adapt to evolving dietary patterns and food trends, ensuring ongoing relevance and performance. Furthermore, the module provides real-time feedback, contributing to a dynamic and responsive user experience.

Applications: Foundationally analyzes images to identify food items, a crucial step for nutritional calculations.



Fig. 6. Food Recognized: Donuts

VI. IMPLEMENTATION

The implementation of the AI-driven nutrition assessment system, NutriVision, involves a series of steps that translate the design and requirements into a working system. This chapter covers the systematic process of implementing NutriVision, focusing on software process improvement, evaluation, and measurement, guided by a systematic literature review.

1. Development Environment Setup:

- Git was used for version control, facilitating collaboration among team members and ensuring code integrity.
- Node.js served as the JavaScript runtime environment, enabling server-side scripting and backend development.
- Authentication was integrated for user authentication and secure access management.
- Other necessary dependencies and libraries were installed using npm, the Node Package Manager.

2. Module Implementation:

- Each module, including the landing page, login module, image processing module and the food recognition module were all implemented using appropriate algorithms and AI techniques.

- Code snippets and various examples were developed to demonstrate the functionality of each module.
- Best coding practices were followed to ensure readability, maintainability, and scalability of the codebase.

VII. USER INTERFACE DESIGN

Design Principles and Tools Used:

The user interface design section of the NutriVision was specially designed for creating an effective and user-friendly system platform which was completely guided by principles of simplicity, intuitiveness, and consistency. The following tools and methodologies were employed:

1. Design Engineering and Prototyping:

- Utilization of various UML (Unified Modeling Language) diagrams to guide the project in its complete capable implementation, ensuring a thorough understanding of system architecture and functionality before development.
- Prototypes were developed to simulate user interaction and test the usability of the interface before actual implementation.

2. UI Component Libraries:

- The user interface is based on Flask, providing a server system for the front-end.
- Custom components were designed and developed to meet specific requirements while adhering to design guidelines.

3. Responsive Design:

- The UI was designed to be responsive, ensuring optimal viewing and interaction experiences across devices of various screen sizes, including desktops, tablets, and smartphones.

Screenshots and Descriptions of Key UI Components:

1. Login Page:

- The dashboard serves as the main hub for accessing different modules.
- It features a clean and organized layout with intuitive navigation elements, allowing users to easily switch between modules.



Fig. 8. Landing Page

User Interaction Flow:

The user interaction flow within NutriVision is designed to be seamless and intuitive:

1. Authentication and Onboarding:

- New users are guided through the authentication process, ensuring secure access to the platform.
- Onboarding screens or tooltips may be provided to familiarize users with the platform's features and functionalities.

2. Module Selection and Usage:

- Upon logging in, users are directed to the dashboard where they can upload the desired image.
- Each module presents a clear interface for input of image and of configuring settings, followed by the generation of the corresponding detailed nutrition content.

VIII. TESTING AND EVALUATION

Testing Methodologies:

The NutriVision: AI-Driven Nutrition Assessment platform underwent rigorous testing to ensure reliability, functionality, and performance. The following testing methodologies were employed:

1. Unit Testing:

- Individual components and functions within the application were tested in isolation to verify their correctness and functionality.
- Unit tests were done to check for all decision branches and internal code flows were functioning correctly and that each unique path of a business process performed accurately.

2. Integration Testing:

- Interactions between different modules and components were tested to ensure seamless integration and proper communication.
- Integration tests were conducted to validate end-to-end functionality and data flow across the system.

3. User Acceptance Testing (UAT):

- Real users or designated testers were invited to evaluate the application's usability, features, and overall satisfaction.
- User acceptance tests were conducted to gather feedback and identify any usability issues or areas for improvement.

Test Cases and Results:

A comprehensive set of test cases was developed to cover various scenarios and functionalities of the AI content generation platform. Test cases included:

- Input validation tests to ensure proper handling of user inputs and error messages.
- Functional tests to verify the correctness of content generation algorithms and outputs.
- Performance tests to assess the responsiveness and scalability of the system under different load conditions.
- Compatibility tests to verify cross-browser and cross-device compatibility.

Test results were meticulously documented, including pass/fail outcomes, error logs, and performance metrics. Any identified issues or bugs were prioritized based on severity and addressed promptly.

Performance Evaluation:

Performance evaluation was conducted to assess the platform's responsiveness, throughput, and scalability. Key performance metrics measured included:

- **Response time:** The time taken for the platform to respond to user inputs and requests.
- **Throughput:** The rate at which the platform can handle concurrent user interactions and content generation requests.
- **Resource utilization:** CPU, memory, and network usage under varying load conditions.
- **Scalability:** The platform's ability to handle increasing user traffic and workload without degradation in performance.

Performance testing was performed using tools like Apache JMeter or k6 to simulate realistic user scenarios and stress test the system's capabilities.

User Feedback and Usability Testing:

User feedback and usability testing were essential components of the evaluation process. Real users were engaged to interact with the platform and provide feedback on their experience. Usability testing sessions involved tasks such as:

- Navigating through different modules and features.
- Generating content using various input texts and configurations.
- Providing feedback on interface intuitiveness, clarity of instructions, and overall satisfaction.

Feedback and usability testing results were analysed to identify pain points, areas of confusion, and opportunities for enhancement. Iterative improvements were made based on user feedback to optimize the platform's usability and user experience.

IX. RESULTS AND DISCUSSION

Analysis of Results:

The results obtained from testing and evaluation indicate that the NutriVision platform performs effectively in analysing various food images and providing nutritional assessments. The platform demonstrates robustness, accuracy, and scalability in handling user requests and producing quality outputs.

Comparison with Expected Outcomes:

The platform's performance aligns with the expected outcomes defined during the project's planning phase. The functionality of each module, including the Home Page, Login module, Photo Analysis module, and the Nutrition Assessment module, met the specified requirements and achieved the intended objectives. User feedback and usability

testing confirmed that the platform fulfils user expectations and delivers the desired experience.

Discussion on Effectiveness and Limitations:

While NutriVision proves effective in its core functionalities, certain limitations and challenges were identified during the evaluation process. These include:

- **Performance bottlenecks:** Some modules may experience degradation in performance under heavy load conditions, requiring optimization strategies.
- **Accuracy of content generation:** Despite high accuracy overall, occasional discrepancies or inaccuracies in the generated nutritional assessments may occur, necessitating continuous improvement in AI algorithms and models to enhance accuracy and reliability.

X. FUTURE ENHANCEMENTS

To address the identified limitations and further enhance the platform's capabilities, several potential improvements and additional features are proposed:

- **Optimization algorithms:** Implementing advanced optimization techniques to improve performance and scalability, ensuring consistent responsiveness under varying load conditions.
- **AI model refinement:** Continuously refining and updating AI models to enhance the accuracy of Nutritional assessment, incorporating user feedback and leveraging state-of-the-art research.
- **Enhancing user guidance:** Implementation of interactive tutorials, tooltips, and contextual help features within the user interface can assist users in navigating NutriVision's functionalities, ensuring they can effectively utilize its features without feeling overwhelmed.

XI. CONCLUSION

In summary, The NutriVision project demonstrates the significant potential of integrating artificial intelligence and deep learning into nutritional assessment. By leveraging advanced computer vision techniques and robust CNN architectures, NutriVision accurately identifies and analyzes food items from images, providing detailed nutritional information. This platform not only addresses the challenges associated with food image analysis, such as variations in image quality and context, but also enhances the user

experience through reliable performance and user-friendly interfaces.

Through rigorous testing and implementation of state-of-the-art methodologies, NutriVision showcases the effective application of AI in promoting healthy living. The project underscores the importance of continued research and development in AI-driven health solutions, paving the way for future innovations in dietary management and wellness.

By delivering accurate and comprehensive nutritional assessments, NutriVision stands as a valuable tool for individuals, health practitioners, and researchers, contributing to the broader goal of leveraging technology for better health outcomes.

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