

Diet Alchemist

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ABSTRACT

Food is an essential aspect of human existence, and individuals are constantly exploring new and delectable dishes. Often, people choose food items from grocery stores without knowing their names or recognizing them immediately. Therefore, it is crucial to comprehend the elements that can be combined to create delightful recipes. Selecting the appropriate recipe from a list of ingredients poses a significant challenge for both novice and expert chefs. Machine learning plays a prominent role in our daily lives, such as in object recognition through image processing. However, traditional methods employed in this process, which involve numerous food items, present a higher risk of error. To address these challenges, we developed a model that identifies food ingredients and formulated an algorithm to recommend recipes based on the identified ingredients. Our research involved constructing a unique dataset comprising 9,856 photos, categorized into 32 types of food items. We utilized a Convolutional Neural Network (CNN) model for food item recognition and employed machine learning techniques to generate recipes. Our approach achieved an impressive accuracy rate of 94 percent, which holds significant practical value..

Keywords— food ingredients, recipe recommendation, machine learning, deep learning, Convolutional Neural Network (CNN), object recognition, image processing.

INTRODUCTION

In With the rapid advancements in sensor technology and GPU computing speed, deep learning algorithms have experienced significant acceleration in recent years. Within the field of computer vision, recipe detection of food images using deep neural networks has emerged as a prominent area of research. This technology involves automatically identifying the ingredients and recipe associated with a given food image, and it finds applications in food recognition, dietary tracking, recommendation systems, and food marketing.

The process of recipe detection begins by inputting a large dataset of food images into a deep neural network or utilizing a pre-existing trained model. The model extracts meaningful features from the images and learns representations that aid in ingredient identification. It then maps the identified ingredients to corresponding recipes, providing the most probable recipe associated with the input image. However, this process poses several challenges, including image noise, ingredient variability based on cooking style, composition, lighting conditions, and the complexity of analyzing a vast amount of images.

Despite these challenges, researchers have applied various approaches to enhance the efficiency of recipe detection. These include leveraging Convolutional Neural Network (CNN) architectures, employing transfer learning techniques, and utilizing ensemble methods. Such techniques have led to significant improvements in both the accuracy and speed of recipe detection, enabling more effective applications in health and wellness, recipe recommendation, and nutrition management.

While humans effortlessly recognize their surroundings and objects in their daily lives, computers require significant computational power, complex algorithms, and dedicated efforts to accurately identify patterns and regions where objects may be present. Object detection and recognition have been extensively studied and will implement over the years, forming the foundation of computer vision systems.

Overall, the field of recipe detection using deep learning techniques has shown promising advancements, offering valuable contributions to various domains, including health, wellness, and culinary experiences.



LITERATURE SURVEY

Paper Title: Food Category Representatives: Extracting Categories from Meal Names in Food Recordings In this paper, we propose a method for compressing a meal name into a shorter representation. First, we collect similar in this paper, Food Log is a multimedia recording tool for producing food records for many individuals. In one year of operation, Food Log has produced more than one million food records for meals eaten by users. We found nearly 70,000 unique food records among these data. In analyzing them, one of the challenges are to extract meal categories from such many records. In meal names using a k-nearest neighbor search. Next, we construct a word graph to model the relationship between the meal names and items in the database. We select representative words by identifying minimal paths in the word graph. Finally, we obtain a few words that represent categorical information about the original meal name. We applied the method to data in food records for both Food Log and the Rakuten Recipe database.

Paper Title: Suggestion Analysis for Food Recipe Improve Suggestion analysis for food recipe improvement is to identify helpful suggestions from user comments to improve the recipes. Consequently, user comments about food recipes are classified into two groups that are comments with suggestions or without suggestions. The word information from modified lexicons and created rules for interpreting meaning are applied to analyze those comments or opinions. Natural language processing and text analysis are included in the proposed analysis technique. The automate comment analysis can help both users to choose the preferred food recipes and recipe authors to develop their own creative recipes. To summarize food recipe improvement, the user comments are collected and grouped into suggestion comments and other comments. The evaluation of proposed suggestion analysis shows that the accuracy and precision of comment classification are more than 70.

Food Classification from Images Using Convolution Neural Networks The process of identifying food items from an image is quite an interesting field with various applications. Since food monitoring plays a leading role in health-related problems, it is becoming more essential in our day-to-day lives. In this paper, an approach has been presented to classify images of food using convolutional neural networks. Unlike the traditional artificial neural networks, convolutional neural networks have the capability of estimating the score function directly from image pixels. A 2D convolution layer has been utilised which creates a convolution kernel that is convolved with the layer input to produce a tensor of outputs. There are multiple such layers, and the outputs are concatenated at parts to form the final tensor of outputs. We also use the Max-Pooling function for the data, and the features extracted from this function are used to train the network. An accuracy of 86.97% implementation.

Paper Title: Food Image Classification and Nutrition Detection Using CNN Deep learning and the availability of greater datasets and computational resources have made classification more straightforward. In recent years, the convolutional neural network has become the most commonly used and popular image categorization approach. In this paper, various transfer learning approaches are used to classify images from an Indian cuisine dataset. Food plays a significant role in human life because it supplies us with a variety of nutrients, so it is crucial for everyone to keep track of their eating habits. As a result, food classification is a must for a better way of living. Pretrained models are employed in this project instead of typical ways of developing a model from the ground up, which saves computing time and money while also producing superior results.

PROBLEM DEFINATION

The project tackles various challenges and problems associated with meal planning, cooking, and food waste. The primary problem it addresses is the difficulty individuals encounter when deciding what to cook for dinner despite having ingredients in their fridge. This issue is particularly relevant for those who aspire to cook at home, maintain healthier eating habits, and minimize food waste.

One problem is the lack of recipe inspiration, where individuals struggle to come up with creative and enjoyable meal ideas using the ingredients they have on hand. This often leads to repetitive and uninspiring meals or the temptation to opt for take-out instead. Additionally, the time and effort required for meal planning pose a significant challenge. The process of searching for recipes, creating shopping lists, and ensuring ingredient availability can be time-consuming and overwhelming, discouraging people from cooking at home and resulting in frequent dining out or reliance on convenience foods.

Food waste and inefficient utilization of ingredients present another problem. Many individuals dispose of ingredients because they are unsure how to incorporate them effectively into meals. This contributes to the larger issue of food waste, where edible resources are needlessly discarded. Furthermore, individuals with dietary restrictions or food allergies face challenges in finding

suitable recipes that meet their specific needs. Without tailored recommendations based on available ingredients, it can be difficult for them to prepare safe and nutritious meals.

PROPOSED METHODOLOGY

This project makes significant contributions in several key areas. Firstly, it enhances the convenience of meal planning by providing users with recipe recommendations based on the ingredients available in their refrigerators. By eliminating the need to visit a grocery store or search for recipes, users can save time and effort in deciding what to cook for dinner.

Secondly, the project encourages and promotes home cooking, contributing to healthier eating habits. By empowering individuals to utilize the ingredients they already have, it motivates them to prepare meals at home instead of relying on restaurant or take-out options. This shift towards home cooking can lead to improved nutrition and overall well-being.

Additionally, the project addresses the critical issue of food waste. By suggesting recipes that utilize the ingredients users already possess, it helps reduce the amount of discarded food. This not only has environmental benefits but also promotes sustainability by ensuring efficient use of food resources.

Furthermore, the system takes into account individuals with specific dietary needs or restrictions. By providing recipe recommendations tailored to their available ingredients, it ensures that users can make safe and suitable food choices that align with their dietary requirements.

The project also offers financial benefits, as users can make the most of their existing ingredients, potentially reducing the frequency of grocery store visits and resulting in cost savings over time.

Lastly, this project showcases the integration of advanced technologies, such as image recognition, deep learning, and recommendation algorithms, in the culinary domain. By employing these cutting-edge techniques, it demonstrates the potential for technological innovation in everyday cooking experiences and serves as a foundation for future research and development in recipe recommendation systems.

In summary, this project contributes to convenience, healthier eating, food waste reduction, customization for dietary needs, cost savings, and technological advancements, making it a valuable and impactful endeavor

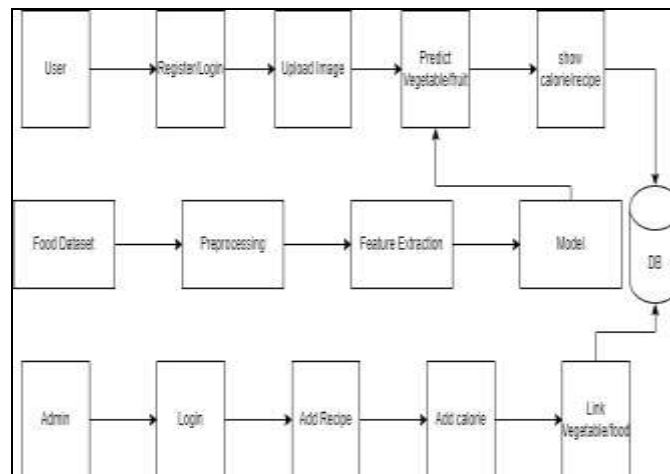


Fig 1. System Architecture



Fig 2. Upload Page



Fig 3. Vegetable detection

CONCLUSIONS

In conclusion, this project aimed to develop a recipe recommendation system with ingredient-based search and meal planning capabilities. Through extensive research and implementation, several key findings and outcomes have been achieved.

The project successfully will be implemented using a deep learning model for food image recognition and ingredient extraction, allowing users to input available ingredients and receive personalized recipe recommendations. By incorporating user preferences, dietary restrictions, and ingredient availability, the system provides relevant and tailored recipe suggestions, promoting user engagement and satisfaction.

The integration of external services, such as grocery store APIs and delivery services, enhances the user experience by providing real-time ingredient availability and facilitating the shopping process. This feature contributes to reducing food waste and encouraging efficient ingredient utilization.

Through thorough testing and evaluation, the system's performance, responsiveness, and accuracy have been assessed. User feedback and iterative improvements have played a crucial role in refining the system and addressing user needs and preferences.

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