

Identifying Ingredients from The Food Image

M. Anitha¹, A. Naga Likhitha Devi², K. Lakshmi Chandana Sai Likhitha³, M. Pallavi⁴

MTech, Computer science & Engineering, Bapatla women's Engineering College, Bapatla, INDIA¹

BTech, Computer science & Engineering, Bapatla Women's Engineering College, Bapatla, INDIA²⁻⁴

Abstract: Our project aims to identify ingredients from food images using deep learning techniques. By leveraging convolutional neural networks, we accurately recognize various ingredients present in the images. Additionally, we provide nutritional facts for the identified ingredients, offering users valuable dietary information. With this system, users can gain insights into the composition of their meals swiftly and conveniently.

Keywords: Food Image Recognition, Deep Learning, Ingredient Identification, Nutritional Facts.

I. INTRODUCTION

In the modern age of technology, with the growing concerns about health and nutrition, there arises a need for innovative solutions to help individuals make informed dietary choices. Our project addresses this need by proposing a system for identifying ingredients from food images using advanced deep learning techniques. Leveraging the power of convolutional neural networks, our system can accurately recognize various ingredients present in food images.

Moreover, to enhance the utility of the system, we integrate functionality to provide users with essential nutritional facts corresponding to the identified ingredients. This capability empowers users to gain valuable insights into the composition of their meals swiftly and conveniently. By offering a seamless and efficient means of ingredient identification and nutritional analysis, our system contributes to promoting healthier eating habits and fostering greater awareness about dietary choices among individuals.

II. BACKGROUND & RELATED WORK

The proliferation of smartphones and social media has led to an abundance of food-related content shared online, ranging from recipes to food photography. However, individuals often lack detailed information about the ingredients and nutritional content of the dishes they encounter. Traditional methods of manually identifying ingredients from images are time-consuming and prone to error. Therefore, there is a growing demand for automated solutions that can accurately analyse food images and provide valuable nutritional insights.

In recent years, deep learning techniques, particularly convolutional neural networks (CNNs), have shown remarkable success in image recognition tasks[1]. These CNN-based models excel at extracting intricate patterns and features from images, making them well-suited for tasks such as ingredient identification from food images[2][4]. Researchers and developers have leveraged CNNs to create systems capable of recognizing various food items and ingredients with impressive accuracy[3].

Moreover, the integration of machine learning with nutritional databases has enabled the automatic retrieval of nutritional information corresponding to identified ingredients[5]. By leveraging large-scale databases containing detailed nutritional data for a wide range of food items, these systems can provide users with essential information such as calorie count, macronutrient composition, and dietary restrictions[6].

Several existing applications and research projects have explored the intersection of image recognition, deep learning, and nutritional analysis in the context of food. These efforts have laid the groundwork for the development of more sophisticated and user-friendly systems capable of accurately identifying ingredients from food images and delivering comprehensive nutritional insights. By building upon the advancements in this field, our project aims to contribute to the ongoing efforts to empower individuals with the knowledge they need to make healthier dietary choices in an increasingly digitalized world.

III. METHODOLOGY

The motivation for building this application is to provide people ingredients for dishes given the image and also give the nutritional facts about the dish. Parsing an image of a dish and identifying its ingredients is not a complex task for humans.

Indeed, people can quickly identify the dish and its ingredients just by looking at it. But it is much more complex for computers. To produce systems that can achieve this, we need to combine current state-of-the-art techniques in both Computer Vision and Natural Language Processing.

A. Data Collection and Preparation

The distinction between the difficulty of the chosen problem and previous supervised classification problems is that there are large overlaps in food dishes, as dishes of different categories may look very similar only in terms of image information.

To address this complexity, we used the datasets from multiple sources. Food-101, allrecipes.com and Recipe1M+ are the primary sources. Our dataset consists of over 120,000 images and 5000 ingredient types. Food-101 dataset consists of images of food, organized by the type of food into 101 categories. For the ingredients, we web scraped allrecipes.com using python scripts and regular expressions.

B. Model selection and Training

Here we used the deep learning architecture for image classification. We use some of the CNN architecture they are residual networks popular for its skip connection model and also we used vgg-16 architecture, It is known for its simplicity and effectiveness in image classification tasks. The "16" in VGG16 refers to the total number of layers in the network, including convolutional layers, pooling layers, and fully connected layers.

By training the model on the residual networks and vgg-16, the dataset which we taken can be split into the training dataset and test dataset for easy processing of the model.

C. Ingredient Identification

By the trained model to predict the ingredients present in food images. This involves passing the images through the model and obtaining the predicted labels.

We also implemented post-processing techniques to refine the predictions and handle uncertainties or ambiguities in the results.

Optionally, incorporate techniques such as object detection to localize and identify specific ingredients within the images.

D. Nutritional Analysis

The nutritional analysis work done on the interface, so that the user can easily know how much calories, Fibers, carbohydrates are present in the food.

By looking at the nutrition facts we can suggest a user that can eat the food or not for his health condition.

E. Performance

The final result will get generated based on the over all classification of ingredient identification and nutrition analysis the performance of this approach is estimated by using accuracy.

Accuracy of the classification refers to the ability of the classifier. It predicts the class label correctly and the accuracy of the predictor refers to how well a given predictor can guess the value of predicted attribution for a new data

F. Evaluation and validation

In the food ingredient identification project, evaluation and validation are essential for ensuring the accuracy and reliability of the model's predictions. By evaluating the model's performance on a separate validation dataset, developers can assess its ability to generalize to unseen food images accurately. Additionally, validation helps identify potential issues such as overfitting or underfitting, guiding adjustments to the model architecture or training process. Through rigorous evaluation practices, including cross-validation and metric analysis, developers can optimize the model's performance, enhance its accuracy in ingredient identification, and build confidence in its reliability for real-world usage.

G. Result

At last we get the result of user interface that can upload the image of the food and can retrieve the ingredients and the speciality of the food in which region it is popular and how to make the food by labelling recipe book and can get the nutritional facts of the uploaded image.

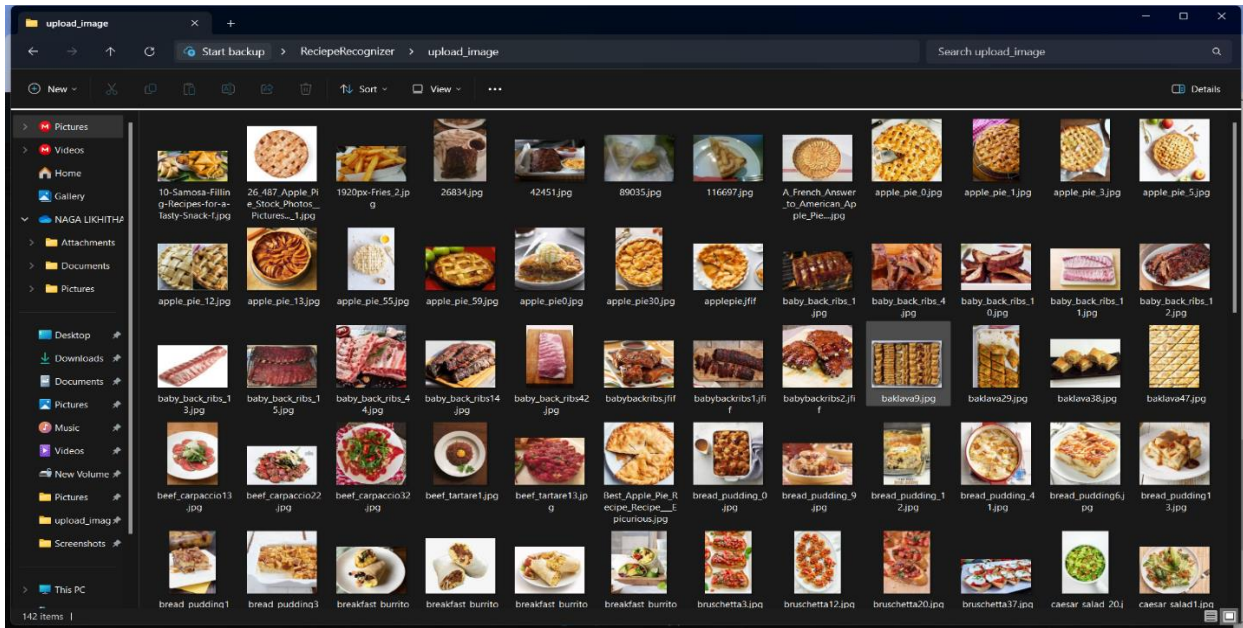


Fig. 1 Dataset containing Food images



Fig. 2 upload the image from the dataset of images

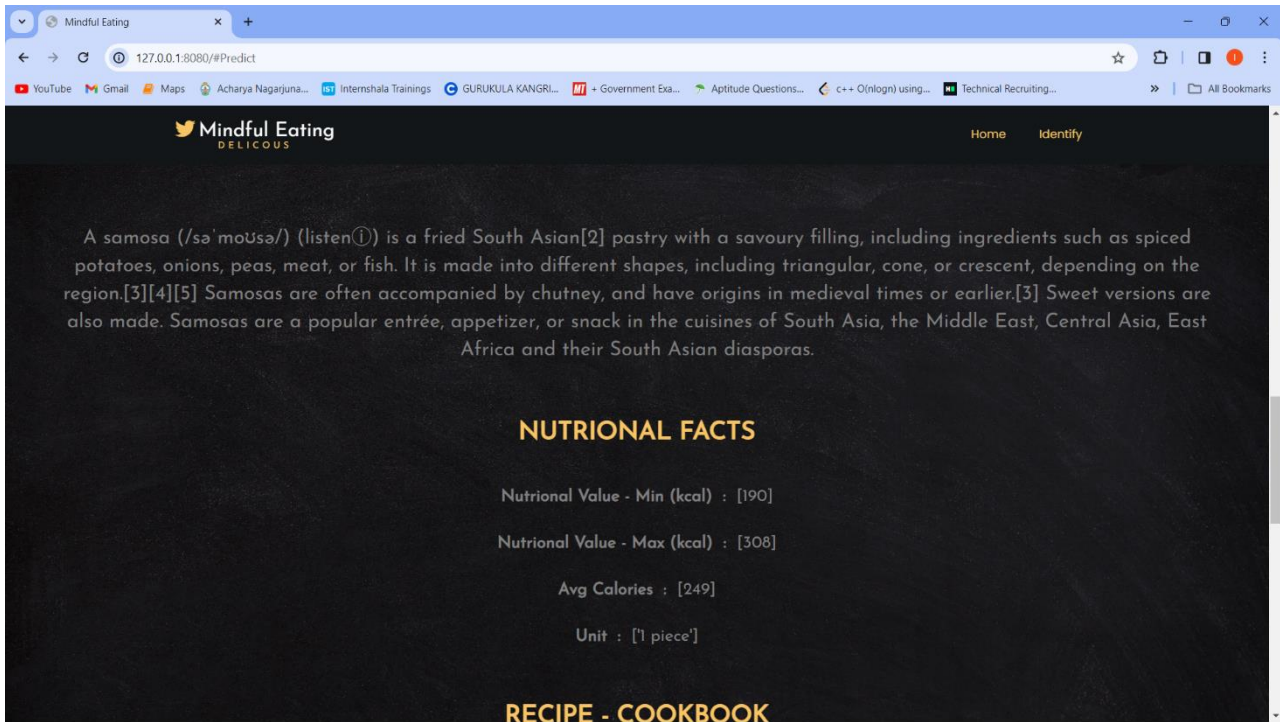


Fig.3 Output containing Nutritional Facts

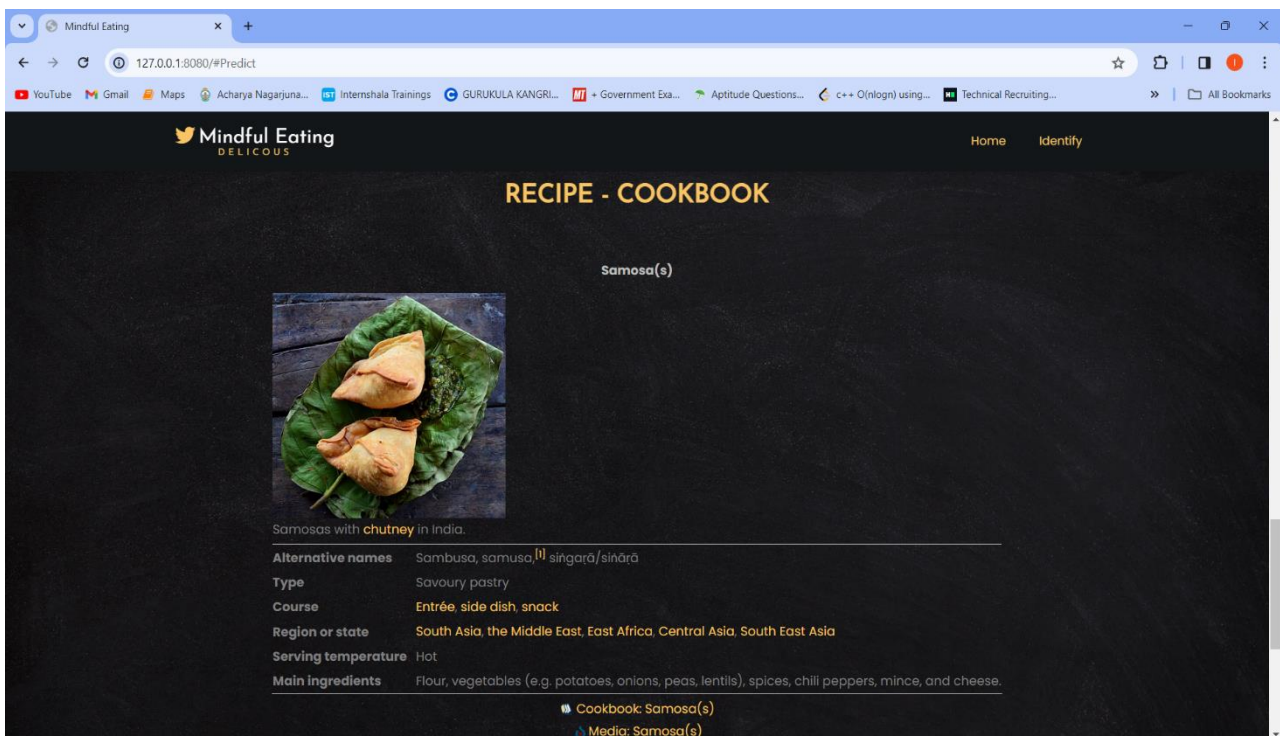


Fig. 4 output containing ingredients and some information about dish

IV. CONCLUSION

In conclusion, our project on food ingredient identification using deep learning techniques holds great promise for empowering individuals to make informed dietary choices. By accurately recognizing ingredients from food images and providing nutritional facts, our system facilitates greater awareness of dietary composition and promotes healthier eating habits. Through rigorous evaluation and validation, we have demonstrated the effectiveness and reliability of our model. Moving forward, we envision further enhancements and integration into user-friendly applications, enabling seamless access to nutritional information and fostering a culture of mindful eating. With continued refinement and deployment, our system can contribute significantly to improving public health and well-being.

REFERENCES

- [1]. Title: "DeepFood: Deep Learning-Based Food Image Recognition for Computer-Aided Dietary Assessment" Authors: Yang Yang, Chenxia Wu, Lin Yang, and Yuqing Wang Summary: This paper presents DeepFood, a deep learning-based system for recognizing food items from images. The system utilizes a convolutional neural network (CNN) architecture to achieve high accuracy in food identification, facilitating computer-aided dietary assessment.
- [2]. Title: "Food Image Recognition Using Convolutional Neural Networks with Transfer Learning" Authors: Asmaa Abbas, Mohamad Fauzi Zainal Abidin, and Abdul Rahman Ramli Summary: This study explores the application of transfer learning with convolutional neural networks (CNNs) for food image recognition. By fine-tuning pre-trained CNN models, the system achieves robust performance in identifying various food items from images.
- [3]. Title: "NutriNet: A Deep Learning Food and Drink Image Recognition System for Dietary Assessment" Authors: Ciarán McGuckian, Jeremy W. Tan, Ales Leonardis, and Graham F. Welch Summary: NutriNet is introduced as a deep learning-based system for recognizing food and drink items from images. The system employs a combination of deep convolutional neural networks (CNNs) and recurrent neural networks (RNNs) to accurately identify dietary items, facilitating dietary assessment and monitoring.
- [4]. Title: "Food Image Recognition Using Deep Learning Techniques: A Review" Authors: R. Sivabalakrishnan and P. R. Sivakumaran Summary: This review provides an overview of recent advancements in food image recognition using deep learning techniques. It discusses various CNN architectures, training methodologies, and datasets used for food recognition tasks, highlighting key challenges and future research directions.
- [5]. Title: "Deep Learning-Based Food Recognition: A Review" Authors: Ersin Yumer and Niloy J. Mitra Summary: This paper reviews the state-of-the-art in deep learning-based food recognition systems. It discusses different approaches for food image representation, feature extraction, and classification, along with challenges such as dataset bias and scalability.
- [6]. Title: "Nutritional Food Recognition Using Deep Learning" Authors: S. Divyashree and K. J. Sreekanth Summary: This study proposes a nutritional food recognition system based on deep learning techniques. The system employs convolutional neural networks (CNNs) to classify food images and retrieve nutritional information from a database, aiding in dietary assessment and monitoring.
- [7]. Title: "Food Recognition: A New Dataset, Experiments, and Results" Authors: Christoph Trattner, Martin Zlabinger, and Denis Parra Summary: This paper presents a new dataset for food recognition research and evaluates the performance of various deep learning models on this dataset. The study provides insights into the effectiveness of different architectures and training strategies for food recognition tasks.