FOOD RECOGNITION AND CALORIE PREDICTION

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Abstract:

The With the onset of lockdown in the COVID-19 scenario, people were forced to confine themselves within the four walls of their rooms which in the meantime invited mood disorders like depression, anxiety etc. Music has proven to be a potential empathetic companion in this tough time for all. The functions of playing music and multimedia have become essential in one device as a smart phone since the smart phone appeared. It is very convenient, but it contains controversial arguments about sound quality, so many smart phone users use the music player application. By using these music applications, people start to think about the relationship between music playing and sound quality. However, those applications are not perfect, so it is hard to choose a good application. The application will enable administrator to add a new in stance to the music database.

I. Introduction

Food is the key to the human body. So, a diet plan always needs to consider the total number of calories consumed to maintain a fit and healthy life.

But, in most cases, unfortunately, people face difficulties in estimating and measuring the amount of food intake due to the central lack of nutritional information. It will be useful if there is a system to keep track of and maintain the calorie intake.

Project Scope:

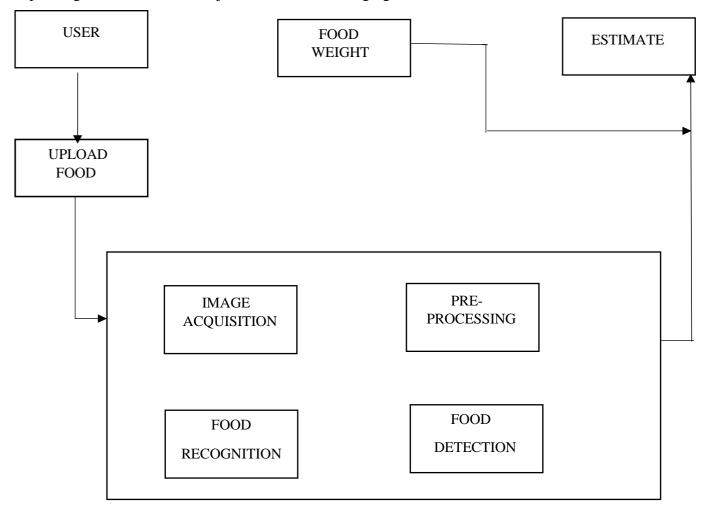
Project overview:

The scope of a project focused on food recognition and calorie prediction in computer vision would involve developing algorithms and models that can accurately identify different types of food from images or videos. The algorithms would then estimate the calorie content of the recognized food based on its visual characteristics and known nutritional information. **Project objectives:**

People are becoming more health conscious in today's fast-paced environment, which is driving up demand for technologies that assistthem in keeping track of their diet and nutrient intake.

Particularly when eating out or consuming homemade meals with variable components and serving quantities.

Since correct calorie predictions are frequently the result of this lack of specific knowledge, and for accomplishing fitness and health objectives can be challenging.



II. Related Work

Multi-Modal Food Recognition:

Some recent works combine visual information from images with textual data such as recipe ingredients or dish names to enhance food recognition accuracy. Multi-modal approaches, including both image and text data, have shown promising results in improving recognition performance.

Calorie Prediction Models:

Researchers have explored different machine learning techniques, including regression models and neural networks, to predict the calorie content of food items based on their visual features.

Transfer Learning:

Transfer learning techniques, where pre-trained models are fine-tuned on food-related datasets, have been employed to leverage knowledge learned from large-scale image datasets like ImageNet.

User-Generated Content Analysis:

Some studies focus on analyzing food images shared on social media platforms. By extracting information from these user-generated images, researchers gain insights into eating habits.

III. Proposed Work

The efficiency of the proposed system for food recognition and calorie prediction would depend on several factors, including the accuracy of the underlying algorithms, the quality and size of the food database, the hardware capabilities of the devices running the application, and the real-time processing requirements.

Our proposed system not only detects varieties of fruits & vegetables but also provides per serving calories of each food detected in a single image

IV. Methodology

Introduction:

The project consists of two steps, identifying food from an image and converting the food identified into a calorie estimation. We performed food image classification using CNN (convolutional Neural Network). Steps followed:

1. Pre-processing: Some basic pre-processing has been performed to clean the dataset where the irrelevant and noisy image of 15 categories have been removed. Also, data augmentation has been performed –Pixel values re-scaled in the range of [0,1].

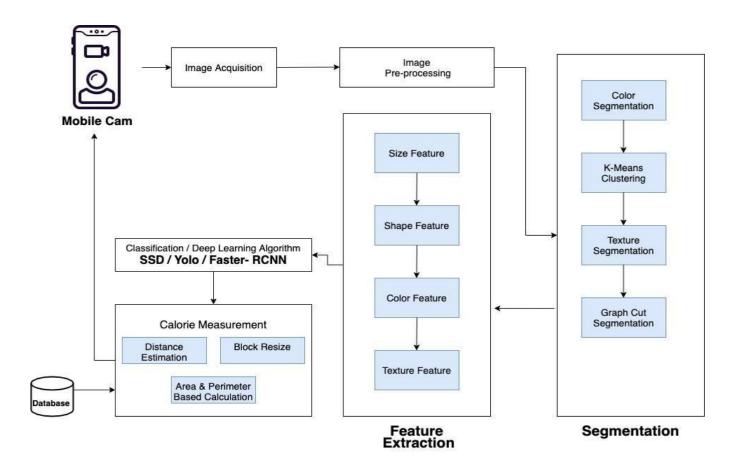
Random rotations max 40 degrees.

Random zoom applied.

Shear angle in counter-clockwise direction in degrees

2. Trained the model: We trained the model with images of 15 categories using the classifier CNN (convolutional Neural Network) which is a class of deep, feed forward artificial neural networks that has successfully been applied to analyzing visual imaginary.

Data Flow Diagram:



Novel idea:

Real-Time Food Recognition: Implement a deep learning model optimized for real-time object detection to recognize various food items in the camera feed. Utilize state-of-the-art object detection architectures like YOLO (You Only Look Once) or EfficientDet for faster and more accurate food recognition.

Augmented Reality Interface: Integrate AR technology to overlay information directly onto the live camera feed. Display recognized food names, nutritional information, and calorie counts in real-time, enhancing the user's experience and providing instant feedback.

User Interaction and Feedback: Allow users to interact with recognized food items on the screen. Provide additional details on the food item, such as ingredients, cooking methods, and potential health benefits. Implement user feedback mechanisms, enabling users to confirm or correct the recognized food, enhancing the system's accuracy over time.

V. Modules

1. Data Collection:

Data Sources: Collect food-related data from various sources such as food databases, user-contributed images, or restaurant menus. You can also use APIs like nutrition databases or web scraping to gather food information.

Data Types: Gather diverse data types, including images of food items, textual descriptions, nutritional information (calories, macros), and servingsizes. Ensure that the data is well-structured and categorized.

2. Data Preprocessing:

Image Preprocessing: Resize and standardize food images to a consistent format. Apply image augmentation techniques like rotation, flipping, and brightness adjustment to increase model generalization.

Text Preprocessing: Tokenize and clean textual descriptions of food items. Extract relevant information such as ingredient lists, portion sizes, and nutritional data.

Data Labeling: Annotate images or text with ground truth labels, including the food item's name, category, and, if possible, calorie content.

Normalization: Normalize nutritional data and portion sizes to a common standard (e.g., per 100 grams) for consistency

Data Splitting: Divide the dataset into training, validation, and test sets to evaluate the model's performance accurately.

3.Object Detection Model:

Choice of Model: Select an object detection model architecture that suits your project's needs. Popular choices include Faster R-CNN, YOLO (You Only Look Once), and SSD (Single Shot MultiBox Detector). These models are pre-trained on large datasets and can be fine-tuned for food recognition.

Transfer Learning: Initialize the chosen model with pre-trained weights on alarge dataset (e.g., COCO dataset) to leverage pre-learned features and improve convergence speed.

Dataset Preparation: Annotated Dataset: Create or acquire a dataset of foodimages with bounding box annotations that specify the location of each food

item within the image. Ensure that the dataset is diverse and representative of the types of food you want to recognize

Data Augmentation: Apply data augmentation techniques to the dataset to increase its size and diversity. Common augmentations include random rotations, flips, and changes in brightness and contrast.

4. Feature Extraction:

This is the main feature, here they identify the different shapes and patterns that are associated with a variety of foods. Deep learning is mainly used for image processing because this is used to perform hard identification that can be done in an image. discussed the identification of hidden features in the faceto perform face recognition.

The conversion of colored images to grayscale will reduce the image size while training the model and also this gray scaling will give a good view in model building. So we adopted this concept of Food identification in our model building. CNN isimplemented in this research with some of the techniques such as Max Pooling, and Convolution 2d, to extract more features In the image To avoid overfitting the model is trained with more Data and also provides a saturated amount of epochs which will build a perfectly trained Model. This approach is discussed for this optical evaluation with a 3D Convolution Neural Network (CNN).

5. Calorie prediction:

Calorie prediction for food recognition typically involves using computer vision and machine learning techniques to identify and quantify the food inan image and then estimate its calorie content.

Image Recognition: The first step is to use a convolutional neural network(CNN) or similar model to recognize and classify the food items in the image. This model should be trained on a large dataset of food images toaccurately identify various dishes.

Calorie Estimation: With both the food item and portion size known, the system can consult a nutritional database or dataset to estimate the calorie content. This may involve looking up the average calorie content for a specific portion of the recognized food.

6. Testing and validation:

There are critical steps in the development of any food recognition and calorie prediction system to ensure its accuracy, reliability, and effectiveness. Here's a comprehensive approach to testing and validation:

Unit Testing: Objective: Validate individual components (e.g., image preprocessing, deep learning model, database queries) to ensure they work as intended. Methods: Use

testing frameworks like unit tests in Python to createunit tests for functions and classes. Mock objects and data to isolate

components during testing.

1. Integration Testing: Objective: Verify that different components work together seamlessly as a system. Methods: Test interactions between modules(e.g., image recognition,

database queries, user interface). Ensure proper dataflow and error handling.

2. Functional Testing: Objective: Validate system functions against specified requirements.

Methods: Create test cases based on functional requirements.

Test various scenarios, such as different food items, input formats, and userinteractions, to ensure the system behaves as expected.

3. Performance Testing:

Objective: Evaluate the system's response time, resource usage, and throughput under various conditions.

Methods: Use tools like Apache JMeter or custom scripts to simulate different loads and measure system performance. Monitor CPU and memory usage during intensive tasks.

VI. Algorithm

1. Calorie Estimation:

How to predict the caloric content of a meal based on the food items.

2. Convolutional Neural Networks:

How they work to identify images and extract features.

3. Image preprocessing:

How to normalize images, crop unnecessary background, and improve recognition accuracy Steps to follow the algorithm

- Step1: Data collection and preparation
- Step2: food recognition model(convolution neural network)
- Step3: calorie prediction model(CNN)
- Step4:integrating food recognition and calorie prediction
- Step5:user interface (i/p)Step6: deployment (web/mobile application)Step7:continuous improvement (monitor)

VII. Conclusion

In conclusion, the integration of food recognition and calorie prediction technologies represents a promising avenue for revolutionizing how individuals manage their diets and make healthier lifestyle choices. By harnessing the power of deep learning, computer vision, and augmented reality, this innovative system can offer a range of benefits to users, such as instant food identification, accurate nutritional information, and personalized calorie predictions. A dynamic ecosystem for people who care about their health is also created by the potential interaction with fitness apps, cloud databases, and continuous learning methods. People may make educated decisions about their nutrition thanks to the ability to effortlessly log recognized foods, track dietary intake, and receive individualized health insights.

VII. References

ImageNet: A large dataset with millions of labeled images, widely used for training deep learning models including those used in food recognition.

Food-101: Contains 101 food categories, with 101,000 images in total, commonly used for food recognition tasks.

Software Tools and Libraries:

TensorFlow: An open-source machine learning framework developed by Google that provides tools and resources for building deep learning models, including those for food recognition.

PyTorch: Another popular open-source machine learning framework that is widely used in research for tasks like image recognition.

OpenCV: An open-source computer vision and machine learning software library that provides various tools for image processing and analysis, useful in food recognition applications.