Fruit Name Detection Based on Automated Robot

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Abstract- The deep convolutional neural network method for fruit detection presented in this study is outstanding. The objective is to create a fruit detection system that is accurate, rapid, and reliable using machine learning data. The method proposed uses convolutional neural network (CNN) technology to identify fruit in visual input. The wide range of fruit species makes image identification of fruit products normally very challenging. However, CNN is the most sophisticated use of deep learning, and recent research has shown that it is an incredibly effective way for detecting images. A dataset is created using the most frequent fruit items in a fruit-logging system that is open to the public, and detection performance is then evaluated using this dataset. CNN showed more accuracy compared to conventional support-vector machine-based methods using hand-crafted features. Additionally, CNN showed considerably improved fruit photo detection accuracy in comparison to a conventional method. This method is also a great deal simpler to apply to new fruits.

I. INTRODUCTION

Since the Stone Age, fruits have been an essential part of the human diet. They have a substantial nutritional impact on human health because of their high nutritional value. Wherever it occurs, fruit consumption must be checked for quality. To do this, it is possible to create a fruit identification system that can recognise various fruit varieties from images captured by any digital camera or smartphone in various locations.

We can use this method to evaluate the fruit's quality and create a robotic orchard harvesting system. In developing the system, machine learning methods were used. Fruit detection must be precise and efficient for a machine. Fruit detection systems are challenging to employ for a number of reasons, including the fact that fruits can occasionally blend in with the backdrop and be concealed by other objects or occur in situations with varying levels of illumination. A fruit identification system should be precise, able to learn from easily available data sets, generate predictions in real time, be able to adapt to diverse fruit kinds, and work with a number of modalities, such as infrared and colour images.

II. LITERATURE SURVEY

Fruit Classification and Recognization with Deep Learning Support on Embedded System (fruitnet)

The natural product arrangement process is significant for exchange. Organic product creation at gather time is exceptionally high. The grouping of natural products by their attributes and qualities is generally done physically and outwardly. This approach can bring about critical misfortunes in time, cost and work. In the proposed study, organic product acknowledgment was performed utilizing picture handling techniques. Calculation utilized and it's Accuracy CNN Algorithm utilized in this framework with exactness of 93.33 rate

AI Based Robotic Systems for the quality control of Date Palm Fruits - A Review

The date palm has acquired critical noticeable quality in worldwide business sectors lately and over the most recent twenty years. These techniques, in any case, have made the errand of performing post-collect work like self-evaluation, quality checking, and sifting of palm natural product a significant test, as most creation units actually utilize individuals to run self-appraisal undertakings. Calculation utilized and it's Accuracy CNN Algorithm utilized in this framework with exactness of 94.23 rate

Improved Kiwifruit Identification Using Pre-Trained VGG16 With RGB and NIR Information Fusion

This study shows how the original purposes the RGB-D (Red Green Blue - Depth) sensors and the adjusted wire of RGB and NIR pictures with profound convolutional brain (CNN) networks for organic product. It intends to foster a more exact, quicker, more solid natural product distinguishing proof framework, which is a significant piece of estimating organic product yield and programmed collecting.

IJSART - Volume 8 Issue 7 – JULY 2022

Calculation utilized and it's Accuracy CNN Algorithm utilized in this framework with precision of 91.37 rate

Transfer Learning based CNN Model for Classification of Mango Leaves Infected by Anthracnose

Anthracnose is a plant parasitic illness that produces acervulus spore in depressed leaves, harming the leaves and foliage of mango natural product. Anthracnose illness influences the nature of mango leafy foods. Calculation utilized and it's Accuracy CNN Algorithm utilized in this framework with precision of 92.57 rate

Autonomous Mobile Robot for Bananna Plant Disease Detection based on CNN and Multi-Spectral Vision System

This paper presents a free arrangement of field testing of apple plantations and early finding. Different sensors including hyperspectral, multispectral and visual outputs are utilized for diagnostics. Calculation utilized and it's Accuracy CNN Algorithm utilized in this framework with exactness of 90.57 rate

III. EXISTING SYSTEM

Machine learning is now frequently used to identify fruits. Only a few of the several researchers who utilised this methodology in their studies experienced any meaningful success. Despite the fact that many researchers are now aware of the problem, as demonstrated by the papers submitted in, the challenge of constructing a speedy and precise fruit detecting system still exists. It's due to the fruits' diverse exterior characteristics, such as their colour, size, texture, shape, and reflectivity levels. In the majority of these setups, the fruits are likewise largely abstracted and subjected to continually changing lighting and shadow circumstances.

IV. PROPOSED SYSTEM

Compared to standard neural networks, convolutional neural networks feature a unique architecture. An input is changed in conventional neural networks by being routed through a number of hidden layers.

In this system, each layer is made up of a collection of neurons that are all intimately connected to the neurons in the layer beneath it. The predictions are represented by the architecture, or last fully-connected layer, also called the output layer.

Other neural networks and convolutional neural networks are slightly different. First, the layers are organised

into three groups: width, height, and depth. Here, only some of the neurons in the subsequent layer are interconnected with those in the preceding layer. The output of the system is then compressed into a single probability score vector that is organised along the depth dimension.

V. SYSTEM IMPLEMENTATION

The process through which a theoretical idea is turned into a practical tool is called implementation. The consumer department is now bearing the brunt of resistance and the impact on existing practises. The most important stage in the creation of a new system, and the user must have faith that the new system will function and be effective, is the implementation process. If this process is not planned and managed, it can cause confusion.

The process of utilising the created system is known as programme implementation. This covers all activities related to using the new programme. Once the planning is finished, the organization's first priority is to make sure that the systems' procedures are functioning properly. Before the implementation process can begin, several requirements must be satisfied.



Figure1: In this figure show the fruit identification

VI. CONCLUSION

A particularly successful machine learning methodology that correctly recognises the output of images for the proposed model is the CNN method, which is being used. The deployed CNN algorithm is effective at classifying images and recognising objects. The model's high accuracy illustrates how well CNN meets the criteria for this kind of fruit recognition and that CNN has been effective in developing a top-notch algorithm for fruit recognition

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