

Fruit Disease Detection Android App

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Abstract: *The development of the app involved several stages, including dataset collection, pre-processing, feature extraction, and disease classification. A comprehensive dataset comprising both healthy and diseased fruit images is curated and used for training the machine learning model. Pre-processing techniques are applied to enhance the captured fruit images, while feature extraction algorithms extract meaningful information to feed into the classification model. The trained model is then integrated into the Android app, allowing for real-time disease detection and diagnosis. The user interface of the app is designed to be intuitive and user-friendly, catering to farmers and agricultural professionals with varying levels of technical expertise. Users can simply capture an image of a fruit using their smartphone's camera and initiate the disease detection process. The app provides instantaneous feedback on the presence of diseases, along with detailed information regarding the specific disease type and severity.*

Keywords: Fruit disease detection, Android app, image processing, machine learning, and crop health monitoring

I. INTRODUCTION

In recent years, advancements in image processing techniques and machine learning algorithms have shown great promise in the field of disease detection in agriculture. These technologies can be harnessed to develop innovative solutions that provide accurate and timely diagnoses of fruit diseases, enabling farmers to take proactive measures to mitigate the impact on their crops.

In response to this need, we present the Fruit Disease Detection Android App, a mobile application specifically designed to assist farmers and agricultural professionals in detecting and identifying diseases in fruits. The app harnesses the power of smartphones and leverages their built-in cameras to capture images of fruit samples. These images are then processed using sophisticated image processing algorithms to enhance their quality and extract relevant features.

The trained model is integrated into the Android app, enabling real-time disease detection and diagnosis. The user-friendly interface of the app allows farmers to easily capture fruit images and receive instant feedback on the presence and type of disease affecting their crops. The development of this app involved careful dataset collection, preprocessing of fruit images, feature extraction, and disease classification. A comprehensive dataset comprising both healthy and diseased fruit images was curated and used to train the machine learning model, ensuring its accuracy and robustness.

II. PROBLEM STATEMENT

There is an urgent need for a comprehensive fruit disease detection Android app that leverages the power of image processing and machine learning techniques to provide farmers with an intuitive, reliable, and accessible tool for early detection and accurate diagnosis of fruit diseases. Such an app should be capable of seamlessly integrating with smartphone cameras, enabling real-time image analysis and instant feedback on disease presence and classification. By addressing these challenges, the app would empower farmers to make informed decisions, implement timely interventions, and effectively manage crop health, leading to increased productivity, reduced losses, and sustainable agricultural practices.

III. TECHNOLOGY STACK

What is an Android Application-?

Android Studio and Firebase are used for developing our project which are available everywhere. It provides the technical guarantee of accuracy, reliable and security. The current system develop is technically feasible with all the resources need for development of the apps as well as the maintenance of the same is easy.

Android Studio is the official integrated development environment (IDE) for Google's Android operating system, built on JetBrains' IntelliJ IDEA software and designed specifically for Android development. It is available for download on Windows, macOS and Linux based operating systems. It is a replacement for the Eclipse Android Development Tools (ADT) as primary IDE for native Android application development.

Android Studio was announced on May 16, 2013 at the Google I/O conference. It was in early access preview stage starting from version 0.1 in May 2013, then entered beta stage starting from version 0.8 which was released in June 2014. The first stable build was released in December 2014, starting from version 1.0. The current stable version is 3.2, which was released in September 2018.

Features:

The following features are provided in the current stable version:

- Android-specific refactoring and quick fixes.
- Lint tools to catch performance, usability, version compatibility and other problems.
- Pro-Guard integration and app-signing capabilities.
- Template-based wizards to create common Android designs and components
- Support for building Android Wear apps.
- Android Virtual Device (Emulator) to run and debug apps in the Android studio.
- Gradle-based build support.

Java Language

Java is the name of a programming language created by Sun Microsystems in 1995. This company was bought out by Oracle Corporation, which continues to keep it up to date. The latest version is Java SE 9, which came out in 2017.

Java, which was called Oak when it was still being developed, is object oriented, meaning it is based on objects that work together to make programs do their jobs. Java code looks like C, C++, or C#, but code written in those languages will not work in Java in most cases without being changed.

Java runs on many different operating systems, including Android, the world's most popular mobile operating system. This makes Java platform independent. It does this by making the Java compiler turn code into Java bytecode instead of machine code. This means that when the program is executed, the Java Virtual Machine interprets the bytecode and translates it into machine code.

Java Concepts

Java was developed to achieve 5 main goals. These are:

- It should be simple, object-oriented, distributed and easy to learn.
- It should be robust and secure.
- It should be independent of a given computer architecture or platform.
- It should be very performant.
- It should be possible to write an interpreter for the language. The language should also support parallelism and use dynamic typing.

XML Language

In computing, Extensible Mark-up Language (XML) is a mark-up language that defines a set of rules for encoding documents in a format that is both human-readable and machine readable. The W3C's XML 1.0 Specification and several other related specifications all of them free open standards define XML.

The design goals of XML emphasize simplicity, generality, and usability across the Internet. It is a textual data format with strong support via Unicode for different human languages. Although, the design of XML focuses on documents, the language is widely used for the representation of arbitrary data structures such as those used in web services. Several schema systems exist to aid in the definition of XML-based languages, while programmers have developed many application programming interfaces (APIs) to aid the processing of XML data.

Firestore -

- Firestore is a Backend-as-a-Service that started as a YC11 start up and grew up into a next-generation app-development platform on Google Cloud Platform.
- Firestore frees developers to focus crafting fantastic user experiences. You don't need to manage servers. You don't need to write APIs. Firestore is your server, your API and your data store, all written so generically that you can modify it to suit most needs. Yeah, you'll occasionally need to use other bits of the Google Cloud for your advanced applications. Firestore can't be everything to everybody. But it gets pretty close.
- Real-time data is the way of the future. Nothing compares to it.
- Most databases require you to make HTTP calls to get and sync your data. Most databases give you data only when you ask for it.
- When you connect your app to Firestore, you're not connecting through normal HTTP. You're connecting through a Web-Socket. Web-Sockets are much, much faster than HTTP. You don't have to make individual Web-Socket calls, because one socket connection is plenty. All of your data syncs automatically through that single Web-Socket as fast as your client's network can carry it.
- Firestore sends you, new data as soon as it's updated. When your client saves a change to the data, all connected clients receive the updated data almost instantly.
- Firestore Storage provides a simple way to save binary files—most often images, but it could be anything—to Google Cloud Storage directly from the client!!!
- Firestore Storage has its own system of security rules to protect your G-Cloud bucket from the masses, while granting detailed write privileges to your authenticated clients.

IV. ALGORITHM

Haar Cascade Classifiers

- Haar cascades are widely used for object detection tasks.
- They utilize a series of rectangular features and a trained model to detect objects of interest.
- Haar cascades are commonly used for face detection but can be adapted for other objects as well.

Convolutional Neural Networks (CNN)

- CNNs are deep learning models specifically designed for image analysis tasks.
- They learn hierarchical representations from data and are highly effective for image detection and recognition.
- CNNs have achieved state-of-the-art performance in various object detection challenges, such as the popular Faster R-CNN and YOLO frameworks.

1. Image Acquisition:

- Utilize the smartphone's camera functionality to capture an image of the fruit.
- Ensure adequate lighting conditions and minimize any possible camera shake.

2. Image Preprocessing:

- Apply noise reduction techniques, such as Gaussian filtering or median filtering, to remove unwanted noise from the image.
- Enhance the image quality using methods like histogram equalization or contrast stretching to improve visibility.

3. Image Segmentation:

- Employ segmentation algorithms, such as thresholding or clustering, to separate the fruit region from the background.
- Extract the fruit region based on color, texture, or other distinctive features.

4. Feature Extraction:

- Extract relevant features from the segmented fruit region to represent disease-related characteristics.
- Common features include color histograms, texture descriptors (e.g., Gabor filters, Local Binary Patterns), shape descriptors (e.g., Hu moments, Zernike moments), and statistical features.

5. Disease Classification:

- Utilize machine learning algorithms, such as Support Vector Machines (SVM), Random Forest, or Convolutional Neural Networks (CNN), for disease classification.
- Train a model using a labeled dataset consisting of healthy and diseased fruit samples to learn patterns and characteristics associated with different diseases.
- Extract the features from the segmented fruit region and input them into the trained model for disease classification.
- The model assigns a probability or label to indicate the presence and type of disease.

6. Decision Making and Display Results:

- Determine a threshold for disease probability or confidence level to make a decision about disease presence.
- Display the results to the user, indicating whether the fruit is healthy or affected by a specific disease.
- Provide additional information about the detected disease, including its name, severity, and recommended treatment options.

7. Real-Time Processing and User Interaction:

- Implement the algorithm to process images in real-time, enabling immediate feedback to the user.
- Design a user-friendly interface allowing users to capture fruit images, initiate the disease detection process, and view the results effortlessly.
- Enable functionalities such as image storage, disease history tracking, and access to a disease database for further reference.

V. MODULE DESCRIPTION

1. Image Acquisition Module:

- This module captures images of fruits using the smartphone's camera functionality.
- It ensures optimal lighting conditions and provides options for image stabilization to minimize blur.

2. Image Preprocessing Module:

- The preprocessing module applies techniques like noise reduction to remove unwanted artifacts or noise from the captured fruit images.
- It enhances image quality through methods such as histogram equalization or contrast stretching, improving visibility and reducing inconsistencies.

3. Image Segmentation Module:

- The segmentation module separates the fruit region from the background or other objects in the image.
- It employs algorithms like thresholding or clustering to extract the fruit region based on color, texture, or other distinctive features.

4. Feature Extraction Module:

- This module extracts relevant features from the segmented fruit region to represent disease-related characteristics.
- Common features include color histograms, texture descriptors (e.g., Gabor filters, Local Binary Patterns), shape descriptors (e.g., Hu moments, Zernike moments), and statistical features.

5. Disease Classification Module:

- The classification module utilizes machine learning algorithms to classify the fruit image based on the extracted features.
- It trains a model using a labeled dataset consisting of healthy and diseased fruit samples, learning.
- The module employs algorithms such as Support Vector Machines (SVM), Random Forest, or Convolutional Neural Networks (CNN) to classify the fruit image and determine the presence and type of disease.

6. Decision Making and Result Display Module:

- Based on the disease classification, this module makes a decision about the presence of a disease and its type.
- It sets a threshold for disease probability or confidence level to make accurate determinations.
- The module presents the results to the user, indicating whether the fruit is healthy or affected by a specific disease, along with additional information such as disease severity and recommended treatments.

7. Real-Time Processing and User Interaction Module:

- The real-time processing module ensures efficient and prompt image analysis, providing immediate feedback to the user.
- It integrates with the smartphone's interface, allowing users to capture fruit images, initiate the disease detection process, and view the results seamlessly.
- The module may also support functionalities such as image storage, disease history tracking, and access to a comprehensive disease database for further reference and information.

Each module plays a crucial role in the overall functionality of the Fruit Disease Detection Android App, contributing to accurate disease detection, user-friendly interaction, and timely intervention for improved crop health management.

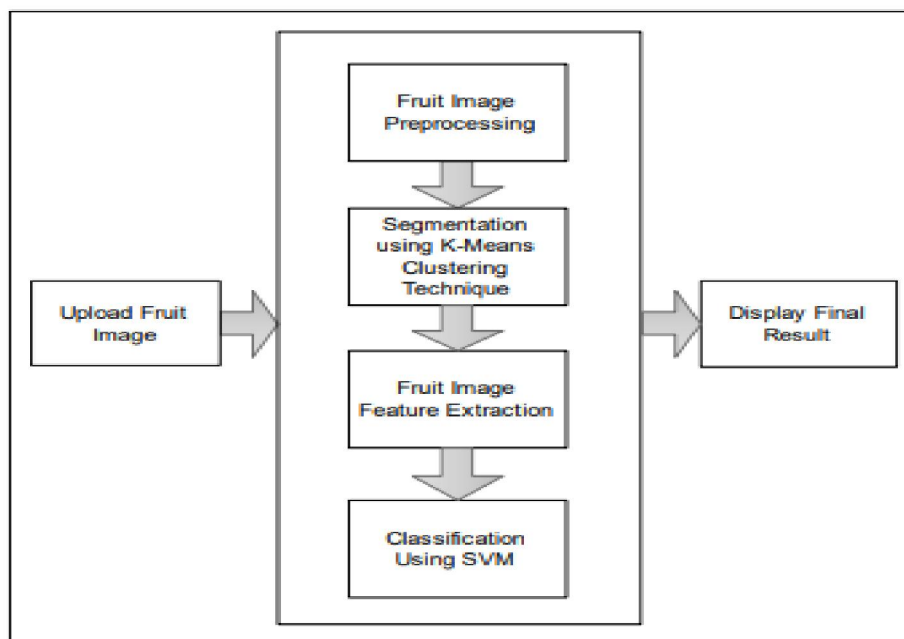


Fig: System Architecture

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VI. CONCLUSION

The app's integration with Android devices ensures widespread adoption and scalability. Farmers from diverse regions and backgrounds can easily access and utilize the app, promoting its potential impact in the agricultural sector. The app's ability to track disease history and access a comprehensive disease database enhances its usefulness as a resource for farmers to make informed decisions and adopt effective disease management strategies.

In conclusion, the Fruit Disease Detection Android App revolutionizes fruit disease detection and monitoring, empowering farmers to proactively manage crop health and optimize agricultural productivity. By providing accurate and timely diagnoses, the app contributes to reducing crop losses, improving yields, and promoting sustainable agricultural practices. As technology continues to evolve, this app represents a significant step towards revolutionizing the way diseases in fruits are detected, leading to a more secure and productive agricultural industry.

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