

Mobile app for detecting of potential bud fruitfulness: A tool for predicting

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Abstract: The budding process in grape cultivation is a crucial stage that significantly influences the quality and yield of grape harvests. This abstract introduces an innovative approach that combines the power of the YOLOv8 algorithm with a mobile application to enhance the detection and monitoring of grape buds during the budding process.

High-resolution images of grapevines in various budding stages are captured using imaging tools and these images are processed through the YOLOv8 algorithm. This state-of-the-art object detection system excels at accurately identifying grape buds and categorizing them based on their developmental stages, including dormant, swelling, green tip, and advanced growth phases.

Here we are doing the integration of a user-friendly mobile app. The application provides Grape growers and vineyard managers with real-time access to the YOLOv8-based detection system. This empowers users to view, analyze, and track grape bud development directly from their mobile devices. It also allows for immediate decision-making, as users can time vital tasks such as pruning, irrigation, and disease management with precision.

In conclusion, the fusion of the YOLOv8 algorithm with a user-friendly mobile application for grape bud detection in the budding process is a groundbreaking advancement in precision viticulture. This innovative solution not only improves the accuracy of bud detection but also offers accessible and data-driven vineyard management. The research presented here opens new horizons for grape cultivation, benefiting both the viticulture industry and the broader agricultural community.

Keywords: Segmentation, Bud's Detection, Final work on App, YOLOv8 algorithm.

I. INTRODUCTION

In the field of viticulture, a successful harvest depends on careful grape bud detection. Cutting of defected bud from accurate point is most important part in this Farming. This introduction offers a novel way to maximize grape bud recognition during blossoming by fusing the YOLOv8 algorithm with an easy-to-use mobile app. This system allows real-time access to grape bud data by combining mobile accessibility with computer vision. With the app, viticulturists may make well-informed decisions, track historical data, gain predictive insights, and revolutionize vineyard management to achieve greater production and quality.

This app will help to predict the healthy and unhealthy bud of grapes. The main purpose of this app is to detecting of potential and bud fruitfulness. In addition to promising productive harvests and data-driven, sustainable vineyard management, this article investigates the integration of YOLOv8 with a mobile app for accurate grape bud recognition, which will elevate the farmer growth in grapes farming.

This paper explores the integration of YOLOv8 and a mobile app for precise grape bud detection, promising not only fruitful harvests but also data-driven and sustainable vineyard management, elevating the wine industry.

II. LITERATURE SURVEY

Mukhamadiev S. (2023) [1] In their work, Mukhamadiev S.; Nesteruk S.; Illarionova S.; Somov, A. introduced "Enabling Multi- Part Plant Segmentation with Instance-Level Augmentation Using Weak Annotations". The work shows that segmenting plant parts using models trained without exact mask annotations is still effective.

When weakly supervised techniques are used instead of just bounding box annotations, instance segmentation is greatly improved. Complex object structures are better understood by the model through the combination of spatial and semantic information. The Spatial Attention Module (SAM) demonstrates that models can perform better than optimal conditions in certain situations and handle segmentation tasks efficiently with weak labels. The method can extract useful semantic information and performs exceptionally well at segmenting complex plant structures. Using pseudo-masks, instance-level augmentation enhances model performance as well. For even more advancement, future studies can investigate specialized multi-part augmentations.

Nur-E-Aznin Mimma (2022) [2] In their work, authors Nur-E-Aznin Mimma, Sumon Ahmed, Tahsin Rahman, and Riasat Khan explore the “Fruits Classification and Detection Application Using Deep Learning”. This work presents a deep learning based automated fruit detection and classification system that uses conventionally augmented datasets. It is integrated into an Android app and website, using VGG16/ResNet50 for classification and YOLOv3 for fruit detection. Larger datasets, more sophisticated models, and the ability to detect faulty fruits are among the upcoming advancements that will help modern farming.

Ferrara, G. (2021) [3] In their work, authors Ferrara, G. and Mazzeo, A. conducted a comprehensive study on “Potential and Actual Bud Fruitfulness: A Tool for Predicting and Managing the Yield of Table Grape Varieties”. Before winter pruning and long before real bud fruitfulness can be determined in the vineyard the following spring, grapevine bud fruitfulness can be evaluated through microscopic bud dissection. Because inflorescence primordia might be difficult to discern in certain kinds, bud dissections should only be done by qualified and trained individuals. Bud dissection was used to determine the proportion of fertile buds, the number of inflorescence primordia (IP) per node, and the incidence of primary bud necrosis (PBN). These data might be used to forecast yield in the following season or to manage winter pruning. The information was gradually matched to fruitful buds really seen in the vineyard throughout spring and fertile buds.

Monteiro, A.I. (2021) [4] In their study, Monteiro, A.I.; Malheiro, A.C. and Bacelar, E.A. investigate “Morphology, Physiology and Analysis Techniques of Grapevine Bud Fruitfulness: A Review”. In order for inflorescences and clusters to fully form, grapevine reproductive development takes place over the course of two growing seasons, or vegetative cycles. The process of induction and floral differentiation, which results in the development of reproductive components within dormant buds, is a multifaceted process with three distinct phases: the formation of anlagen, inflorescence primordia, and flowers.

This series of phases is made up of morphological, biochemical, and physiological processes that are impacted by a variety of endogenous and exogenous variables. The number of clusters that could potentially emerge in the upcoming growing season is determined by the establishment of inflorescence primordia. Thus, by determining bud fruitfulness during bud dormancy, viticulturists and winemakers can get an estimate of the initial crop. With the use of this information, bud load can be modified to promote a balanced yield and higher-quality fruit.

Cain Hickey and Tremain Hatch (2021) [5] In their study, authors Cain Hickey and Tremain Hatch explore “Dormant Cane and Spur Pruning in Bunch Grape Vineyard”. Dormant pruning establishes the crop level and canopy density prior to the emergence of green tissues, making it a crucial option in vineyard management. When selecting a pruning technique, growers need to take a number of factors into account, such as the design of the vineyard, the cultivar, and the labor force available throughout the year.

To properly manage their diverse cultivars and vineyard blocks, some farmers could decide to implement multiple pruning procedures. Creating a plan that specifies when and how each vineyard block will be pruned during the dormant season is crucial, regardless of the pruning technique. The foundation for efficient vineyard management throughout the upcoming growing season is laid by effective dormant pruning.

Dr. Belinda Rawnsley, Dr. Cassandra Collins (2020) [6] In their research, authors Dr. Belinda Rawnsley, Dr. Cassandra Collins conducted a comprehensive study on “Improving vineyard productivity through assessment of bud fruitfulness and bud necrosis”. Three or more buds make up a grapevine compound bud, with the main bud yielding the fruiting branches for the upcoming growing season. PBN, or primary bud necrosis, is a physiological condition that kills the main bud. Bud dissection is a tool used to forecast and evaluate bud fruitfulness.

The incidence of PBN has been brought to light by vineyard yield potential. PBN lacks bud dissection can go unnoticed because secondary buds generate fewer bunches of shoots, and these are often smaller. As a result, vineyards impacted by PBN do not reach their full potential yield. The project’s objectives were to: evaluate PBN’s spread and extent; decide on its timing and development of PBN and suggest suitable management strategies for regulation.

III. SYSTEM DEVELOPMENT**App Design:**

Mobile App: Located at the center of the system, the main interface is the mobile app, which lets users take pictures, process images, and see the findings of analyses.

Camera Input & Image Display: This part manages image capture using the camera on the mobile device and presents the user with the photos that have been taken, making sure that the user experience is flawless.

Image Processing (Segmentation - YOLOv8): Grape plants are precisely segmented using the YOLOv8 algorithm, which separates the plant into its main parts—the cane and cordons—and establishes the groundwork for further investigation.

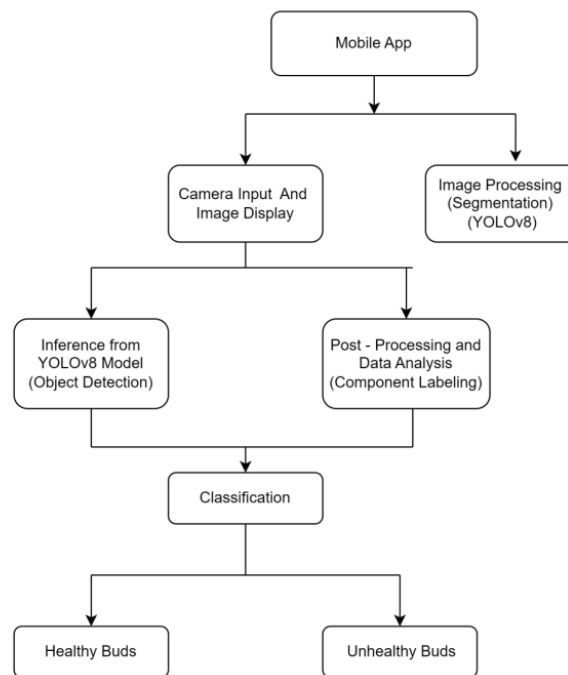


Figure 1: System App Diagram

Inference from YOLOv8 Model (Object Detection): Using YOLOv8, the system performs object detection with a particular focus on cane bud identification, an important step in the grape plant analysis process.

Post-Processing and Data Analysis (Component Labeling): Following segmentation and object detection, the system proceeds with data analysis and post-processing, which entails fine-tuning and evaluating the obtained results as well as labeling the components that have been identified.

Classification (Healthy/Unhealthy): This essential component plays a crucial role in determining the health status of the grape plant by categorizing the identified buds as either healthy or unhealthy. The specific classification model is not explicitly labeled in the diagram.

Visualization & Result Display: The user experience is improved when the user is presented with the processed and classified results in an easily navigable and visually appealing format.

IV. FLOWCHART OF METHODOLOGY**Image Acquisition:**

This block uses a camera to take a picture of the grape bud. The particular characteristics of the grape buds that are being used to differentiate between healthy and unhealthy buds will determine the type of camera that is utilized. For instance, spectral data about the grape bud can be obtained using a hyperspectral camera, and features like the bud's chlorophyll content can then be extracted.

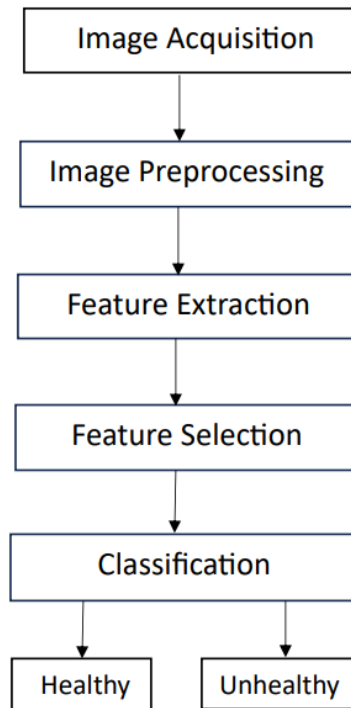


Figure 2: System Flowchart

Image Preprocessing: In order to enhance the acquired image's quality and facilitate the execution of the ensuing processing steps, this block preprocesses it. This could involve actions like image resizing, contrast improvement, and noise reduction.

Feature Extraction: This block takes characteristics that can identify between healthy and unhealthy grape buds from the preprocessed image. These are a few instances of features that could be utilized.

Color features: Grape buds can be identified as healthy or unhealthy based on their color. For instance, whereas unhealthy buds can be brown, yellow, or black, healthy buds are usually green.

Texture features: You can also tell the difference between healthy and unhealthy grape buds by their texture. For instance, the texture of healthy buds is smooth, whereas that of unhealthy buds may be bumpy or scaly.

Shape features: The shape of the grape bud can also be used to distinguish between healthy and unhealthy buds. For example, healthy buds are typically round or oval, while unhealthy buds may be deformed or misshapen.

Feature Selection: This block selects a subset of the extracted features that are most relevant for distinguishing between healthy and unhealthy grape buds. This can be done using a variety of feature selection methods, such as correlation analysis or information gain.

Classification: This block classifies the grape bud as healthy or unhealthy using a machine learning algorithm. The machine learning algorithm has been trained on a dataset of labeled grape bud images. Some common machine learning algorithms used for classification include support vector machines, decision trees, and neural networks.

User Interface: Create an intuitive user interface for the app that shows the user the results. The detected buds' health should be clearly indicated by this interface.

Data Storage and Analysis: Save the pictures and results so that users can monitor the buds' condition over time. In vineyard management, this data can also be utilized for analysis and decision-making.

Real-time Updates: Incorporate a real-time feature to track and evaluate the health of the buds as they change over time.

V. CONCLUSION

In conclusion, there are several drawbacks to the labor-intensive, subjective, and ineffective traditional grape bud fruitfulness evaluation techniques used in viticulture. In computer vision, the YOLO algorithm is presented as a scalable, objective solution that provides real-time data and accuracy for certain problems.

Reliability in grape bud detection is essential for vineyard management and grape quality. Even with the existence of specialized devices, accessibility is still an issue, especially in places like India. The creation of a Mobile App seeks to close this gap and transform grape farming methods by offering grape growers an accessible and reasonably priced alternative.

To put it briefly, the goal of this project is to create a mobile application that makes use of the Yolov8 machine learning technique for bud identification, the Yolov8 approach for grapevine segmentation, and automated grape bud health evaluation. Accurately categorizing grape buds as healthy or unhealthy is the main goal in order to improve grape quality and streamline viticulture procedures. This initiative is a promising first step toward the modernization of grape bud evaluation for vineyard management purposes.

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