



DETECTION OF ADULTERATION IN FRUITS USING MACHINE LEARNING

Bellapukonda Sudarshan¹, Bhavya S D², Dr. S. Bhargavi³, Bhavyashree N⁴

Department of Electronics and Communication Engineering, SJC Institution of Technology, Chickballapur, Karnataka¹

Department of Electronics and Communication Engineering, SJC Institution of Technology,

Chickballapur, Karnataka²

Professor, Department of Electronics and Communication Engineering, SJC Institution of Technology, Chickballapur,

Karnataka³

Department of Electronics and Communication Engineering, SJC Institution of Technology, Chickballapur,

Karnataka⁴

Abstract— Fruit adulteration may seriously harm human health and cause substantial financial losses for both growers and consumers. Traditional methods for fruit adulteration detection can be time-consuming and sometimes call for specialised equipment. A rapid and accurate method of spotting fruit adulteration is offered by machine learning. Using this technique, information on both healthy and tainted fruit samples is gathered, including information on their chemical makeup. Then, utilising these data and attributes taken from the chemical composition data, machine learning models are developed. The models may then be used to precisely determine whether a fruit sample has been tampered with or not, assisting in the reduction of dangerous or fraudulent items that are consumed. Machine learning's application to the detection of fruit adulteration has the potential to increase food safety, shield consumers from dishonest business practises, and lessen financial losses for farmers and other food industry participants.

I. INTRODUCTION

Fruit adulteration has become a big issue in the food sector since it may put customers' health at risk and cause growers and farmers to lose a lot of money. The time-consuming and sometimes expensive nature of conventional procedures for fruit adulteration detection makes them unsuitable for everyday usage. A rapid and accurate method of spotting fruit adulteration is offered by machine learning. These models can correctly determine whether a fruit sample has been contaminated or not by being trained on data acquired from both normal and adulterated fruit samples. This can lessen financial losses in the food sector by preventing the consumption of dangerous or counterfeit products. In this research, we investigate the data collecting procedure, feature extraction methods, model training and testing, and machine learning for fruit adulteration detection. We will also go through the possible advantages of using machine learning to the detection of fruit adulteration as well as the difficulties in putting this strategy into practise. In the end, using machine learning to the detection of fruit adulteration has the potential to increase food safety, safeguard consumers, and help the food sector in a big way.

II. OBJECTIVES

1. To build an automated system to identify the adulteration in fruit.
2. To give healthy fruit to people.
3. To build the system which provides safe food items.
4. To improve the accuracy in formalin detection in fruit.
5. To implement the proposed system using machine learning.

III. PROPOSED SYSTEM

A proposed system for detecting adulteration in fruits using machine learning could involve the following components:

1. **Data Collection:** The first step in developing a machine learning model for detecting fruit adulteration is to collect a large dataset of PPM values of both adulterated and unadulterated fruits.
2. **Data Preparation:** This step involves cleaning the data, reducing noise, and normalizing the features.



3. **Model Building:** The next step is to build a machine learning model that can detect fruit adulteration. Various algorithms such as SVM and KNN can be used to develop the model. The model is trained using the preprocessed dataset.
4. **Deployment:** Once the model is developed and evaluated, it can be deployed to detect fruit adulteration in real-world scenarios. The system could be integrated sensors to automate the detection process.
5. **Continuous Improvement:** The model can be updated and improved over time as more data becomes available. This would involve retraining the model on new data and incorporating new features into the model to improve its accuracy and performance.

The proposed system would provide a fast, efficient, and cost-effective solution for detecting fruit adulteration, ensuring the safety and quality of fruits for consumers. which uses fuzzy logic algorithms to determine the appropriate amount of water to be applied to each plant based on the current environmental conditions.

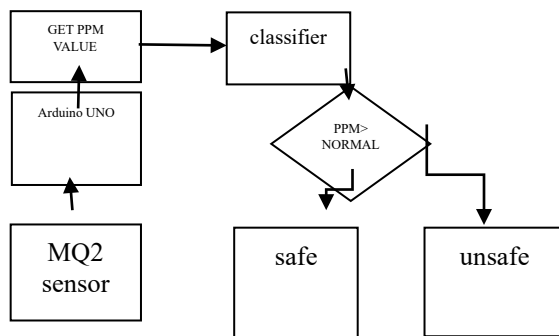


Fig1: Block Diagram of proposed system

IV. METHODOLOGY

The methodology for detecting fruit adulteration using machine learning involves several key steps. The first step is to collect a dataset of PPM values of both adulterated and unadulterated fruits. The next step is to select the most relevant features that have a strong correlation with the presence of adulteration. Once the features are selected, a machine learning model is developed using various algorithms such as SVM, KNN, and SVM linear kernel. The developed model is trained using the preprocessed dataset and evaluated using a test dataset to assess its performance. The model is then optimized by tuning the parameters and hyperparameters of the algorithms to improve its performance. Finally, the model is deployed to detect fruit adulteration in real-world scenarios, and it is continuously improved by retraining it on new data and incorporating new features into the model to improve its accuracy and performance. This methodology provides a comprehensive approach to detecting fruit adulteration using machine learning, which can help ensure the safety and quality of fruits for consumers and enable the food industry to comply with food safety regulations.

V. HARDWARE AND SOFTWARE REQUIREMENT

HARDWARE

1. Arduino UNO:



Fig2: Arduino UNO

The Arduino family's UNO board is the most popular and well-documented model. A microcontroller board called



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Arduino UNO is based on the ATmega328P. It contains 6 analogue inputs, a 16 MHz ceramic resonator, 14 digital input/output pins (six of which may be used as PWM outputs), a USB port, a power connector, an ICSP header, and a reset button.

2. LED:



Fig3: 16*2 led display

You may choose between any of the 16 Pin LCD displays described above because the programming method is the same for all of them. The 16x2 LCD module's pinout and pin description are shown below.

3. MQ2 SENSOR:



Fig4: MQ2 sensor

An electronic sensor called the MQ2 gas sensor measures the amount of gases in the air, including LPG, propane, methane, hydrogen, alcohol, smoke, and carbon monoxide.

Chemiresistor is another name for the MQ2 gas sensor. When the sensing component comes into touch with the gas, the resistance of the component changes. The detection of gas uses this variation in resistance value.

SOFTWARE

1.python:



Fig5: python

Python is straightforward to understand and write because of its syntax. It is especially suited for novices who are starting to write since it employs indentation rather than curly braces or other symbols to separate blocks of code. Additionally, Python offers a sizable standard library that contains a variety of modules and methods that may be used to carry out everyday operations like working with files, processing data, and connecting to databases.

VI. RESULTS



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Machine learning algorithms work at identifying fruit adulteration and imply that the proposed models may identify adulteration in fruits with high rates of accuracy. The amount and quality of the dataset, the features chosen, and the technique chosen may all affect how well the model performs. Consequently, it is crucial to thoroughly assess the effectiveness of the created model using suitable assessment criteria.

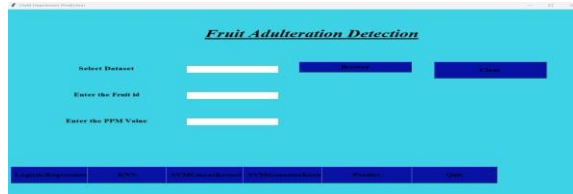


Fig6: Home page



Fig7: values enter in the home page



Fig8: values show in the graph

```
C:\WINDOWS\system32\cmd
MSE VALUE FOR Logistic Regression IS 0.012500
MAE VALUE FOR Logistic Regression IS 0.012500
P-SQUARED VALUE FOR Logistic Regression IS 0.245599
RMSE VALUE FOR Logistic Regression IS 0.111883
ACCURACY VALUE Logistic Regression IS 0.987500

predicted
[1 0 0 0 1 0 1 0 1 1 0 0 1 1 1 1 0 0 0 0 0 1 1 0 0 0 0 1 0 0 0 0 0 0 1
 0 0 1 1 0 0 0 1 0 0 0 0 0 1 0 0 1 1 0 0 1 0 0 0 1 1 1 0 1 0 0 1 0 0 1 1 1
 0 0 1 0 1]
179 1
115 0
96 0
46 0
9 0
..
111 0
109 0
197 1
141 0
216 1
Name: state, Length: 88, dtype: int64

MSE VALUE FOR Logistic Regression IS 0.012500
MAE VALUE FOR Logistic Regression IS 0.012500
P-SQUARED VALUE FOR Logistic Regression IS 0.245599
RMSE VALUE FOR Logistic Regression IS 0.111883
ACCURACY VALUE Logistic Regression IS 0.987500
```

Fig9: values show in command window



Fig10: Software Connect with Hardware

In this project is developed a model using Arduino UNO and machine learning algorithm to implement to detect of fruit adulteration efficiently.

Fruit and Formaline detection technique is developed to detect the presence of formaline using machine learning approaches.

Based on the collected data set system will apply and detect the fruit adulterated or not.

VII. FUTURE SCOPE

The detection of fruit adulteration using machine learning algorithms has a significant potential for future research and development. Here are some of the potential areas of future research:

Exploration of Deep Learning Techniques: While the performance of machine learning algorithms such as SVM, CNN, and DT has shown promising results in detecting fruit adulteration, further exploration of deep learning techniques such as Convolutional Neural Networks (CNNs) and Recurrent Neural Networks (RNNs) can potentially improve the accuracy and performance of the developed models.

Development of a Portable Device: Developing a portable device that can detect fruit adulteration in real-time using machine learning algorithms can be a potential future direction for this research. Such a device could be used by food safety regulators to quickly detect adulteration in fruits in real-world scenarios.

Integration with IoT Devices: Integration of machine learning algorithms for detecting fruit adulteration with IoT devices such as sensors, cameras, and drones can potentially provide a comprehensive solution for monitoring the quality and safety of fruits.

Extension to Other Food Products: While this paper focuses on the detection of fruit adulteration, the proposed methodology can potentially be extended to other food products such as vegetables, grains, and dairy products, to ensure the safety and quality of these products for consumers.

Evaluation of Model Interpretability: While machine learning algorithms have shown promising results in detecting fruit adulteration, further research is needed to evaluate the interpretability of the developed models. Understanding how the models make decisions can potentially provide insights into the factors contributing to fruit adulteration and help improve food safety regulations.

VIII. CONCLUSION

Machine learning algorithms may be used to identify fruit adulteration, which is a crucial duty that can assist guarantee the consumer's fruit's safety and quality. Using machine learning, this research suggested a thorough technique for identifying adulterated fruit. This methodology entails numerous crucial processes, including data collection, data preparation, feature selection, model development, model assessment, model optimisation, deployment, and continual improvement. The suggested technique offers a thorough way for identifying fruit adulteration using machine learning, which may assist secure the consumer's safety and the quality of fruits as well as the food industry's ability to adhere to



food safety rules. To increase the precision and performance of the established models and to study the efficacy of the suggested approach in identifying adulteration in fruits in real-world settings, more research is required.

REFERENCES

1. Anusha K, Anusha MG, Archana Deshpande, Sanjana K, "Detection of adulteration in fruits using machine 'learnings," vol 9, no 6, pp 2393-8374, 2022.
2. Mahanth S, Hemalath G, Swathi H S, Mrs, Madhu G, "Detection of adulteration in fruits and vegetables using machine learning," vol 3, no 5, pp 3478-3479, May 2022.
3. J SampathKumar, M Rajaganapathy, M Praveen and S Socrates, "Food adulteration detection using machine learning," vol 08, no 5, May 2021.
4. Anjana. M. P and Pradeepa. K, "Detection food using the concepts of machine learning," vol 14, no 5, pp 0973-4562, 2019.
5. M. Samad Khan, F. Nowshad, and M. d. N. Islam, "Concentration and formation behavior of naturally occurring formaldehyde in foods," vol 16, no 4, pp 0758-6518.