

# Research Design II

## Fruit Classification using Convolutional Neural Networks

Russell Cachia, Mario Mallia Milanes  
Institute of Information & Communication Technology  
Malta College of Arts, Science & Technology  
Corradino Hill  
Paola PLA 9032

{russell.cachia.g53991, mario.mallia.milanes, ictar}@mcast.edu.mt

### I. INTRODUCTION, POSITIONING, RESEARCH UNION

#### A. Description of Theme and Topic Rational

In the field of computer vision and image processing, the theme of fruit classification is a very interesting area of research, with the increasing advancements in technology and its agricultural importance. Nowadays, the importance of fruit classification is present in multiple different fields, such as dietary monitoring, food processing, and agricultural automation.

With the advancements in Computer Vision and Deep Learning [1], Convolutional Neural Networks (CNNs) have emerged as a game changer in this context, offering a powerful and automated solution for fruit quality control, sorting, and also grading. By making use of the capabilities of CNNs, fruit industry stakeholders can streamline their operations, minimize manual labor, reduce post-harvest losses, and also ensure consistent product quality, ultimately leading to profitability and customer satisfaction.

In today's world, where the visual complexity and uniqueness of fruits present significant challenges, harnessing the power of Convolutional Neural Networks (CNNs) [2] emerges as a promising solution.

#### B. Positioning and Research Union

This study positions itself and the researcher at the intersection of computer vision techniques and practical applications which in this case is fruit classification. With the industry ever-expanding, the aim is to improve fruit classification accuracy.

Considering the given nature of the research problem and objectives, a positivist philosophy is best suited. This is due to the study primarily being quantitative, relying on a dataset of 60,000 images for fruit classification. The large-scale quantitative approach aligns with positivism's emphasis on empirical evidence and the scientific method. By adopting a positivist stance, our study aims to uncover objective truths about fruit classification using Convolutional Neural Networks (CNNs), contributing to the empirical knowledge base in this field.

The research approach taken for this study was a deductive one, starting from established theories and then extending the knowledge to classify fruit using CNNs. This deductive approach allows the researcher to formulate hypotheses based on existing knowledge and test them using quantitative analysis of the dataset.

The research strategy employed was experimental research. By forming a hypothesis, experimental research enables the researcher to systematically test the hypothesis through controlled experiments. Experimental research allowed for the control of variables and conditions during the classification process. Since the study relies on quantitative data and aims to evaluate the performance of CNN models, the experimental research provides a suitable framework for conducting rigorous quantitative analysis.

For the time horizon of this study, a cross-sectional approach was adopted, collecting data from a sample of fruit images. This approach offers valuable insights into the performance and applicability of CNNs in real-world fruit classification scenarios.

In terms of data collection methodology, the researcher adhered to a mono method. The use of the mono method ensures consistency and reliability in data collection, maintaining uniformity throughout the study's execution. Employing such techniques enhances the credibility of the dataset, finally leading to strengthening the research outcomes without repeating information unnecessarily.

#### C. Background to this Research Theme

Diving into the background of Fruit Classification by using Convolutional Neural Networks is of utmost importance to highlight the evolution of computer vision techniques and their application in real-world situations. Fruit classification using Convolutional Neural Networks has gained significant attention in recent years due to its potential applications in agriculture, food processing, and computer vision. CNNs have proven to be effective in image classification tasks, leveraging their ability to automatically learn and extract meaningful features from raw image data. The architecture of CNNs,

consisting of convolutional, pooling, and fully connected layers, allows them to capture spatial hierarchies and patterns present in fruit images [2]. Various studies have explored the use of CNNs for fruit classification, employing different methodologies and data sets. Previous studies have shown high levels of accuracy when it comes to testing these CNNs.

In this study the researcher intends to provide an overview of some existing studies on fruit classification using CNNs. By analysing the strengths and limitations of these studies, the researcher seeks to identify areas for further improvement and future directions in fruit classification using CNNs. These studies are as follows:

- 1) Xiong et al. (2018) developed an algorithm for fruit-picking robots using artificial illumination to recognize grapes. Their system's accuracy was 92% [3].
- 2) Another study conducted by Katarzyna and Pawel performed fruit classification for supermarket retail sales systems. They suggested using a 9-layered deep neural network to classify six different apple cultivars. They claimed 99.78% accuracy [4].
- 3) In another study, Kumari and V. Gomathy have classified fruits using color and texture characteristics. To extract the region of interest, they used HSV color space thresholding. After using a three-level discrete wavelet transform, they then retrieved color data from the hue and saturation channels and texture information from the luminance channel. They used an SVM classifier to categorize 10 fruit types from the supermarket produce data set. Their method was 95.3% accurate [5].
- 4) A CNN based classifier was developed by Sakib, Ashrafi and Siddique which was able to recognize 25 classes of fruits on Fruits 360 data set with 100% test and 99.79% train accuracies [6].

TABLE I  
STATE OF THE ART RESULTS

#	Study	Citation	Metrics	Results
1	AI Robots recognizing grapes	[3]	Accuracy	92%
2	Fruit Classification in Retail Systems	[4]	Accuracy	99.78%
3	HSV Thresholding for Fruit Classification	[5]	Accuracy	95.3%
4	CNN based Classifier	[6]	Accuracy	99.79%

#### D. Hypothesis

The objective of this study is to explore the application of computer vision and machine learning techniques in fruit classification. The primary focus is on assessing the accuracy of fruit evaluation, with the choice of datasets and algorithms serving as independent variables to analyze their influence on the accuracy of classification. Therefore, the hypothesis posits that through the utilization of computer vision technologies, it is viable to achieve precise identification of various types of fruits.

#### E. Research Aim and Purpose Statement

The primary aim of the study is to explore the potential of computer vision techniques in improving fruit classification

accuracy. Monitoring the impact of different datasets and algorithms on the performance of CNN models for fruit classification. By evaluating these variables the study will assess the feasibility and effectiveness of these computer vision techniques.

The feature extraction process is important in order to obtain information in data. This process is to determine the performance of the machine learning [7]. By focusing on the accuracy of fruit evaluation as the main dependent variable, the study aims to understand how the choice of datasets and algorithms influences classification performance. Through rigorous analysis, the study seeks to validate the hypothesis. This study addresses two specific research questions:

- 1) How does the choice of data set impact the performance of CNN models for fruit classification?
- 2) How much of a difference do layers create with regard to accuracy?

By addressing these research questions the researcher aims to contribute valuable insights to the field of automated fruit classification and help inform future research and projects in of this domain.

## II. REVIEW OF RESEARCH METHODOLOGIES AND MAP

### III. REFLECTION OF THE CHOSEN METHODOLOGY

### IV. RESULTS, ANALYSIS AND DISCUSSION

### V. CONCLUSION

### REFERENCES

- [1] M. Hassaballah and K. M. Hosny, "Recent advances in computer vision," *Studies in computational intelligence*, vol. 804, pp. 1–84, 2019.
- [2] D. Bhatt, C. Patel, H. Talsania, J. Patel, R. Vaghela, S. Pandya, K. Modi, and H. Ghayvat, "Cnn variants for computer vision: history, architecture, application, challenges and future scope," *Electronics*, vol. 10, no. 20, p. 2470, 2021.
- [3] J. Xiong, Z. Liu, R. Lin, R. Bu, Z. He, Z. Yang, and C. Liang, "Green grape detection and picking-point calculation in a night-time natural environment using a charge-coupled device (ccd) vision sensor with artificial illumination," *Sensors*, vol. 18, no. 4, p. 969, 2018.
- [4] R. Katarzyna and M. Pawel, "A vision-based method utilizing deep convolutional neural networks for fruit variety classification in uncertainty conditions of retail sales," *Applied Sciences*, vol. 9, no. 19, p. 3971, 2019.
- [5] R. S. S. Kumari and V. Gomathy, "Fruit classification using statistical features in svm classifier," in *2018 4th International Conference on Electrical Energy Systems (ICEES)*. IEEE, 2018, pp. 526–529.
- [6] S. Sakib, Z. Ashrafi, M. Siddique, and A. Bakr, "Implementation of fruits recognition classifier using convolutional neural network algorithm for observation of accuracies for various hidden layers," *arXiv preprint arXiv:1904.00783*, 2019.
- [7] S. I. C. Myongkyoon Yang, "Fruit classification using convolutional neural network(cnn)," *Precision Agriculture Science and Technology*, pp. 1–8, 2021.