

# Interim Report

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## Abstract & Plagiarism Declaration

*To address the issue of electronic waste and cluttered workspaces in the Level 1 Labs, this project aims to develop an automated system for identifying and sorting electronic components. Utilizing Computer Vision, the system will classify components like resistors, capacitors, ICs, and MOSFETs, and sort them into designated bins. The project will be executed in three stages: initial development of a Computer Vision system for component identification, integration of a semi-automated sorting mechanism, and potentially, full automation of the sorting process or additional features like IC testing. Supervised by Dr. Edward Stott, this project aims to practically solve the problems faced by the Level 1 Labs.*

*I affirm that I have provided explicit references for all the material in my Interim Report that is not authored by me, but is represented as my own work. I have used GPTv4, as an aid in the preparation of my report. It was used for LaTeX formatting, however all technical content is original.*

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## 1 Project Specification

**1** This project aims to solve the problem of electronic waste and cluttered workspaces in the Level 1 Labs by developing an automated system for identifying and sorting electronic components.

**2** It aims to do so by employing state of the art Computer Vision techniques to classify various electronic components, and then sort them into designated bins. The project's deliverables are as follows:

- 3** • A Computer Vision system for component identification
- 5** • An interface from which to observe and control the system's state
- 6** • A semi-automated sorting mechanism
- 7** • Potentially, full automation of the sorting process or additional features like IC testing

**8** For this initial stage in the project, the focus is on developing the foundation for the system to be built upon so that the development of the various systems can be done independently, in parallel, and in such a way that each system is modular and easily extensible. For this reason, the project is split into three stages:

- 9** • Initial development of a Computer Vision system for component identification
- Integration of a semi-automated sorting mechanism
- Potentially, full automation of the sorting process or additional features like IC testing

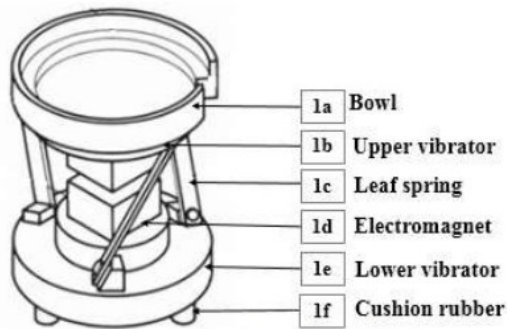


Fig. 1. Automatic vibratory bowl feeder.

Figure 1: VBF Nam et al.<sup>16</sup>

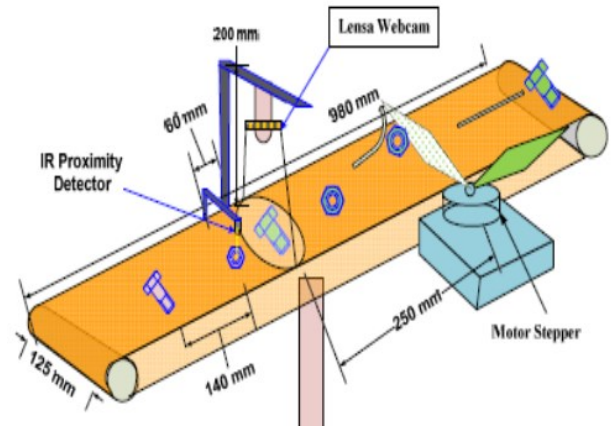


Figure 2: Conveyor Belt Dhenge et al.<sup>5</sup>

This approach was taken to allow me to set defined goals to ensure that the project is always working towards a deliverable product. Each stage extends the previous stage, and so the project can be considered as a series of smaller projects.

## 2 Background

For the background of this project, it is necessary to consider literature related to the following:

- Mechanical design of existing sorting machines
- Existing Computer Vision systems for component identification
- Real-time Computer Vision architectures

### 2.1 Mechanical Design

#### Bowl Feeders

In my research, I have discovered that most industrial sorting machines use a Vibratory Bowl Feeder (VBF) to feed components into the system; as shown in Figure 1, the VBF consists of a bowl that vibrates coupled with a spring and electromagnet to feed components into the system. The paper by Nam et al.<sup>16</sup> explores the optimal design of a VBF for USB keycaps, by attempting to identify the ideal parameters for the structure of the bowl, sorting track, mounting adapter, and suspension system. The paper also uses modal analysis to determine the natural frequencies of the system, and uses this to avoid resonant conditions that might cause inefficient or erratic operation.

This paper is useful as it provides a comprehensive overview of the design of a VBF, and provides a good starting point for the design of the VBF for in the future when the project is extended to include a VBF for fully autonomous sorting. The paper also provides a good overview of the design

considerations for the VBF, and so can be used as a reference during the design process.

Additionally, Reinhart and Loy<sup>23</sup> delves into the mathematical modelling of a VBF, focusing more on the overall performance of the VBF rather than efficiency, and Silversides et al.<sup>24</sup> provides a good overview of the forces involved in the operation of a VBF, strengthening the basis for its design and viability.

The paper by Zhang and of Technology. Department of Mechanical Engineering<sup>29</sup> outlines a sorting system for vials, and does not make use of a VBF, instead opting for a turntable design that mechanically orientates the vials. It primarily operates by using a design that is specific to the geometry of the vials, and so is not applicable to this project, however it does provide a good insight into the design of a sorting system.

It seems conclusive that a VBF is the most viable option for the feeding mechanism of the sorting system, and so the design of the VBF will be based on the research outlined above.

#### Transport Mechanisms

The paper by Dhenge et al.<sup>5</sup> depicts a conveyor belt system that transports Nuts and Bolts to a computer vision system for identification, which aligns with the goals of the project. The hardware prototype is shown in Figure 2, and a down-facing webcam is used to capture images of the components as they pass by. The webcam uses Principle Component Analysis (PCA) to identify the components, which will be discussed in the next section.

The components then separate into two chutes, one for Nuts and one for Bolts, using a stepper motor, which may be useful for the future design of the sorting system. Additionally, the approach taken by the paper Quilloy et al.<sup>22</sup> to

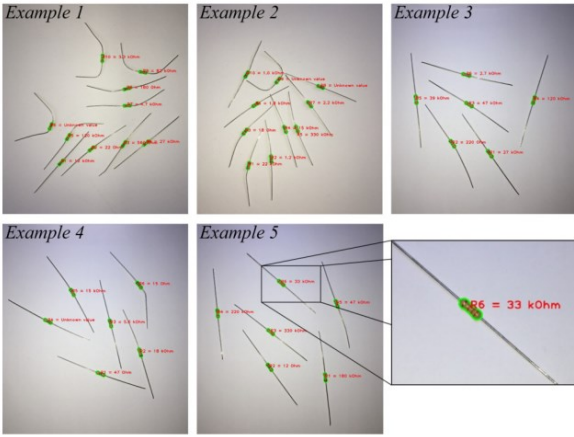


Figure 3: Resistor Test Set<sup>14</sup>

sort Philippine table eggs also features a similar conveyor belt system, down facing camera and an arm to sort the eggs into different categories.

The approaches taken here are similar to industrial sorting systems that I have come across in videos during my research; and this approach seems to answer two major design questions for the project, namely how to transport the components to the computer vision system, and how to transport the components from the computer vision system to the sorting system.

The use of a conveyor belt system is obvious, but the concept of having the camera facing down at the components is not one that I had considered initially, as for the first stage I wanted the system to be semi-autonomous; having the camera face-up allows the user to simply place the component on a plate that was directly above the camera, and the component would be immediately identified. Having the camera face down would block the user's view of the component, which would be a major inconvenience. However, it seems that this would be the most viable approach for the future of the project.

## 2.2 Existing Computer Vision Systems

A range of Computer Vision systems have been explored in the literature, ranging from PCA (Principle Component Analysis) Dhenge et al.<sup>5</sup> to more modern Computer Vision techniques Chand and Lal<sup>1</sup>, Xu et al.<sup>28</sup>. However, given the rate of advancement of Computer Vision techniques, the techniques used in the literature reflect the state of the art at the time of writing, and so are not necessarily the most viable current techniques for the project, but they inspire novel approaches to

the problem of component identification.

PCA with ANN (Artificial Neural Network) as used by Dhenge et al.<sup>5</sup> is a relatively simple but outdated statistical technique that was successful in identifying nuts and bolts, which are very distinct components. In this paper, PCA is used as a feature extractor and then fed directly into, though not explicitly mentioned, an FCL (Fully Connected Layer), as the classifier. Although a valid approach, CNNs (Convolutional Neural Networks) have been shown to be much more effective at feature extraction, and do not "have the problem of low recall and accuracy", as mentioned by Xu et al.<sup>28</sup>, that PCA may suffer from, and so this approach is not viable for the project.

The paper by Xu et al.<sup>28</sup> use the SqueezeNet CNN architecture to identify 22 different subcategories of electronic components, specifically resistors, capacitors, and inductors, and achieves a True Positive Rate (TPR) of 99.999% with only a 2.67ms average inference time on a GTX 1050 2GB GPU, released in 2016, an impressive result. This informs the design of the Computer Vision system, as it shows that a CNN is a viable approach to the problem of component identification.

The paper by Muminovic and Sokic<sup>14</sup> discuss the use of SVM to characterise resistors, by identifying the resistor's centroid (centre of mass), determining the resistor's orientation, and then analysing the bands to determine the resistor's value. This is directly applicable to the project, as it is a viable approach to identifying resistors, which is very distinct from other components, given the colour bands. The paper's novel approach to identifying the resistor values allows it to achieve high accuracy (86%, however the

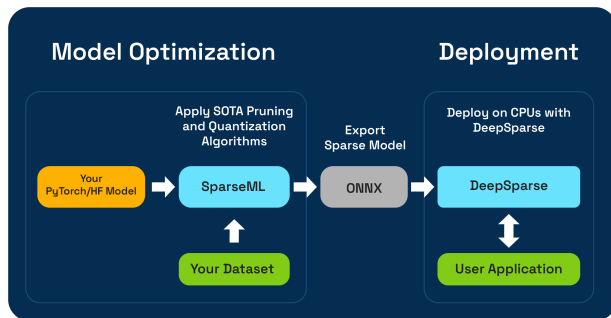


Figure 4: SparseML Pipeline<sup>13</sup>

test images used are resistors that are positioned far away from the camera, as shown in Figure 3, and so the bands are very small, which will not be the case in the project's vision system, so the accuracy will likely be higher) which makes it very promising for the project, and will inform the design of the Computer Vision system.

## 2.3 Real-time Computer Vision Architectures

For the problem of component identification, the Computer Vision system must be able to identify components in real-time, as the components will be moving on a conveyor belt, captured using a camera facing down at the components. This means that the Computer Vision system must be able to identify components in a single frame, and not require multiple frames to identify a component. As I envision the system to be self-contained, it must also be able to run on the Raspberry Pi 4, which has a 1.5GHz quad-core 64-bit ARM Cortex-A72 CPU, and 2GB of RAM. The system will also display a camera feed to the user, and so the Computer Vision system must be able to run in real-time alongside the camera feed.

This requires the Computer Vision system to be both computationally efficient and accurate, which is a difficult balance to strike.

For the explicit purpose of identifying electronic components, the paper by Chand and Lal<sup>1</sup> compares a range of different object detection architectures, including YOLOv3, YOLOv4 and Faster SqueezeNet, with YOLOv4 achieving a mAP (mean Average Precision) of 98.6%.

YOLO (You Only Look Once) is a real-time object detection architecture, that is used very commonly in Computer Vision applications, and has been shown to be very effective at

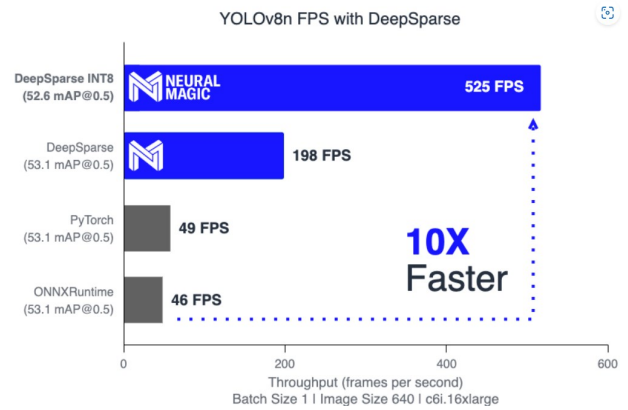


Figure 5: DeepSparse Performance<sup>12</sup>

identifying objects in real-time Terven and Cordova-Esparza<sup>26</sup>. It makes use of a single CNN that takes in an image and outputs a list of bounding boxes and class probabilities for each bounding box. This is in contrast to other object detection architectures, such as R-CNN, which uses a CNN to propose regions of interest, and then uses a second CNN to classify the regions of interest.

Other papers like Guo et al.<sup>8</sup> also comment on YOLOv4's effectiveness at identifying electronic components, achieving 93.94% mAP on a dataset of 20 different components, and the paper also comments on YOLOv4's ability to run in real-time, achieving 67 FPS, albeit on a powerful NVIDIA TITAN Xp GPU. For the Raspberry Pi 3B, the paper by Sismananda et al.<sup>25</sup> has shown to run YOLOv3 at a very low 1 FPS, with an IoU (Intersection over Union) accuracy of 86.7%, which is very low.

However, efforts made by Neural Magic<sup>12</sup> in the optimisation of YOLOv5 and YOLOv8 show performance improvements of up to 10x on CPUs, through their open-source optimisation toolkit SparseML<sup>13</sup> and CPU inferencing runtime, DeepSparse<sup>11</sup>; a very promising result.

Additionally, the YOLO architecture is one that I will likely come across in my modules Deep Learning COMP70010 and Machine Learning COMP70014, which will serve to deepen my understanding of the architecture and allow me to make more informed design decisions.

As such, the purposes of this project, I will be using the YOLOv8 architecture.

## 3 Implementation

I have identified four systems that compromise the project:

- The Hardware, Mechanics and Electronics

- The software that coordinates the various components
- The Computer Vision system
- The tools used to develop the system

### 3.1 Hardware, Mechanics and Electronics

The hardware, mechanics and electronics (HME) of the system are the physical components that make up the system. It is necessary to first consider the HME of the system as they are the foundation upon which the rest of the system is built; for example, when designing the Computer Vision system, it should utilise training data that was collected using the HME that the system will be deployed on, ensuring that the system is robust to the conditions it will be used in. As such, the main focus of this stage has been the HME.

#### 3.1.1 Hardware

For the initial stage of the project, the main hardware components of the HME are:

- Raspberry Pi 4 Model B 2GB
- 7" Touchscreen Display
- 24V dc, 6.25A, 150W Power Supply
- Okdo Adjustable Focus OV5647 Camera
- LED Light Strip

#### Raspberry Pi 4 Model B 2GB

The Raspberry Pi 4 Model B is the main component of the system, and is the central hub that all other components are connected to. This model was chosen as it is regarded as a reliable SoC and is widely used in the industry. It has a large amount of software and driver support, so I can be confident that I will be able to find a solution to any issues that may arise. Additionally, it has GPIO pins that can be used to control other components which will be necessary in future stages. It also has a dedicated camera port which allows for a camera to be connected directly to the SoC, which is necessary for the Computer Vision system.

It also has WiFi support for SSH and remote development, as well as HDMI port for the display. Originally, the 4GB model was ordered however an issue with the EE Stores resulted in receiving the 2GB model instead. This was not an issue as the 2GB model is sufficient for the initial stage of the project, however if it is found to be a bottleneck in future stages, a more powerful model may be

used. The decision to not return the 2GB model is because the EEStores are not available over the Christmas break.

#### 7" Touchscreen Display

The DFRobot 7" Touchscreen Display was chosen as it is a relatively cheap display that is compatible with the Raspberry Pi. It has touchscreen support and has a Raspberry Pi 4 mount on its back, as well as HDMI adapter boards to connect to the Pi. This means I do not have to design a physical mount for the Pi, and only need to design a mount for the display.

#### 24V dc, 6.25A, 150W Power Supply

There are two parameters that were taken into consideration when choosing a power supply: the power output and the voltage.

##### • Power Rating

The power supply must be able to drive all components in the system with overhead in case of spikes in power consumption. In this first stage, the system only consists of the Raspberry Pi, the display, LED lights and camera, making for a power consumption of under 20W, recorded using a smart plug with a power meter. However, in the future, the system will likely need to drive additional motors and sensors, so the total power consumption will be higher. Even with accounting for this, the power consumption will be under 100W, so a 150W power supply will be sufficient, allowing a 50W overhead. In hindsight, this may be considered excessive.

##### • Voltage Rating

The voltage of the power supply must be greater or equal than the highest voltage of any component in the system; this is necessary as I have only step-down converters available on hand, so choosing a higher voltage allows me to step-down the voltage to the required voltage for each component. 24V was chosen as it is a common voltage for stepper motors, which may be used in future stages of the project. It is also used for some LED strips, which are used in this stage. The Pi 4 uses 5V and the LED strip uses 12V, which the 24V power supply can be stepped down to.

#### Okdo Adjustable Focus OV5647 Camera

For the Computer Vision system, a camera is required to capture images of the components. The Okdo Adjustable Focus OV5647 Camera was



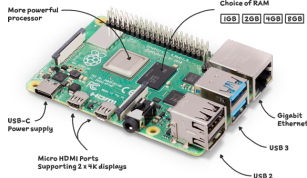


Figure 6: Raspberry Pi 4 Model B 2GB<sup>6</sup>



Figure 7: DFRobot 7" Touchscreen Display<sup>4</sup>



Figure 8: Power consumption of the system

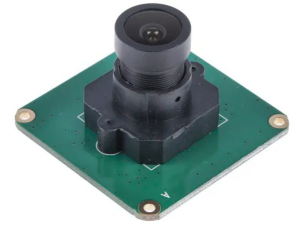


Figure 9: Okdo Adjustable Focus OV5647 Camera<sup>17</sup>

chosen as it is CSI compatible, meaning it can be connected directly to the Raspberry Pi, and has a manual focus ring, allowing it to be used as a macro camera to capture images of small components placed directly above it. The manual focus ring allows the camera to be specifically tuned to the design of the system, allowing for the best possible image quality. It also has a 5MP sensor<sup>18</sup>, which is sufficient for the Computer Vision system, as high resolution images would be preprocessed and reduced.

### LED Light Strip

An LED strip is required to illuminate the components so that the camera can capture images of them. Initially, a 12V LED Ring was chosen as it would allow for a uniform light source and can be mounted directly below the camera; however for reasons explained in the next section, it was replaced with a 5V WS2812B LED strip as it solves the issues faced with the LED Ring.

### 3.1.2 Mechanics

To house the components, a 3D printed housing was designed. It features a modular design, allowing for an iterative design process where I am able to work on different components of the system independently. It also allows for easy replacement of components should my design prove to be inadequate. The housing was designed in FreeCAD<sup>7</sup>, a free and open source parametric CAD software that I have personal experience with.

The parametric design allows for easy modification of the design should I need to make changes - with good design practices, I can ensure that changes can be done as I please by simply modifying a few numerical parameters. The design was then printed using my own personal 3D printer, a heavily modified Voxelab Aquila C2. For this stage on the project, I must

design 3 major components for the housing:

#### PSU Housing

The PSU housing contains the power supply and ensures that all high voltage components are safely enclosed. It also houses the power switch, the power socket and the terminals. As I will require step down converters for the Pi and LED strip, the PSU housing also contains mounting points for the step down converters, as well as a mounting plate to ensure that the step down converters are mounted flat to the PSU housing.

#### Camera Housing

The camera housing contains the camera and the LED strip. It also contains the mounting points for the camera and LED strip. The camera housing will also mount an acrylic plate above the camera, so components can be placed on the plate and be imaged by the camera.

From bottom to top in Figure 11, is the bottom casing, LED Ring mount, camera mount, light diffuser, middle casing and top casing.

#### LCD Display Housing

The LCD display housing contains the LCD display and the Raspberry Pi. As the 7" DFRobot LCD display has a Raspberry Pi 4 mount on its back, an explicit mount for the Pi is not required. The mount will position the LCD display at an angle, so it can be viewed from above. The LCD display simply slides into the mount, allowing for easy removal and also contains mounting points to secure the LCD display to the camera housing.

The LCD housing is split into two parts; the LCD cover, where the LCD display slides into, and a base plate, which mounts to the camera housing. The LCD cover is then mounted onto the base plate, securing the LCD display in place.

The components are secured using brass M3 heat-set inserts and M3 screws. The heat-set inserts are inserted into the 3D printed parts using a soldering iron, and the components are then

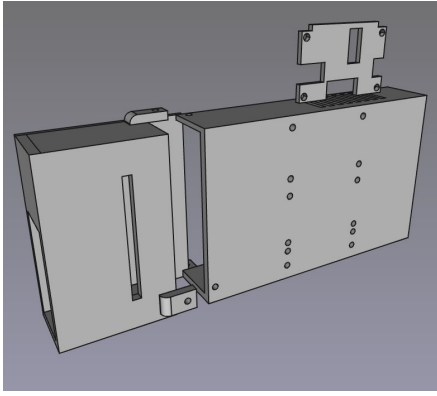


Figure 10: PSU Housing

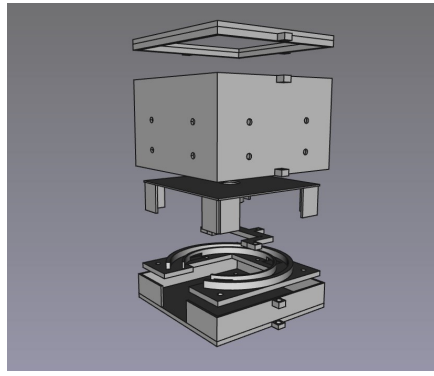


Figure 11: Camera Housing

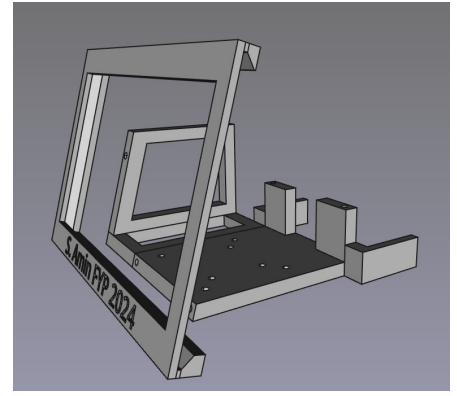


Figure 12: LCD Display Housing

screwed into the inserts. This allows for easy removal of components should I need to make changes to the design, following my modular design principle.

### 3.1.3 Electronics and Safety

As my project is a physical system, there are several electronic components that are required to make the system work.

#### Power Supply

As mentioned in Sections 3.1.1 and 3.1.2, the system is powered by a 24V, 6.25A, 150W power supply, and must be stepped down to 5V for the Pi and 12V for the LED strip. This is achieved using a XL4015 Step Down Converter<sup>10</sup>, a variable step down converter that can output up to 75W. The high power rating of the step down converter allows for a large overhead, which is useful for future stages of the project where additional components may be added to the system.

For powering the system, I have chosen a standard IEC C14 power connector and a RS Pro Snap-In Fused Rocker Switch<sup>3</sup>. As the power supply is a 6.25A power supply, I have chosen a 6A fuse for the switch, which is the closest fuse rating to the power supply's current rating. This ensures that the fuse will blow before the power supply is damaged, should there be a short circuit in the system.

For wiring the power supply and connecting the step down converters, I have chosen 18AWG wire, which is rated for 17A<sup>2</sup>, more than enough for the system's power consumption. The wire is securely crimped to the power supply using a crimping tool.

#### Raspberry Pi 4

To power the Raspberry Pi, I have chosen a standard USB-A to USB-C cable and connected it to a female USB-A breakout board connected to

the 5V step down converter. This allows for easy removal of the Pi should I need to make changes to the system, and prevents the unnecessary cutting of cables.

As the Pi is mounted to the LCD display, a USB-A to Micro USB cable is used to connect the Pi to the display, which is then connected to the Pi's HDMI port, using a supplied HDMI adapter board.

#### LED Lighting

As mentioned in Section 3.1.1, initially, a 12V LED Light Ring was used and was changed to be a WS2812B 5V LED strip, for issues described later. As shown in Figure 13, the LED Ring was connected to a 12V supply (achieved using the 12V step down converter) and parallel to a 100 $\mu$ F capacitor, which is used to smooth out the voltage spikes.

The LED Ring was then connected to an N-channel MOSFET IRLZ44N<sup>9</sup>, which is controlled by a PWM-enabled GPIO pin on the Pi. This GPIO pin is connected to a 1k $\Omega$  resistor, which prevents potential overcurrent from damaging the Pi. The MOSFET is then connected to ground, completing the circuit.

The IRLZ44N is a logic level MOSFET, meaning it can be fully turned on with a 3.3V gate voltage, which is the voltage of the Pi's GPIO pins, and the high switching speed (up to 1mHz) of the MOSFET allows for the LED Ring to be PWM controlled, allowing for the brightness of the LED Ring to be controlled.

## 3.2 Software

The entire system is controlled by a Raspberry Pi 4B running a very light weight Linux distribution called DietPi. I aim for the entire system to

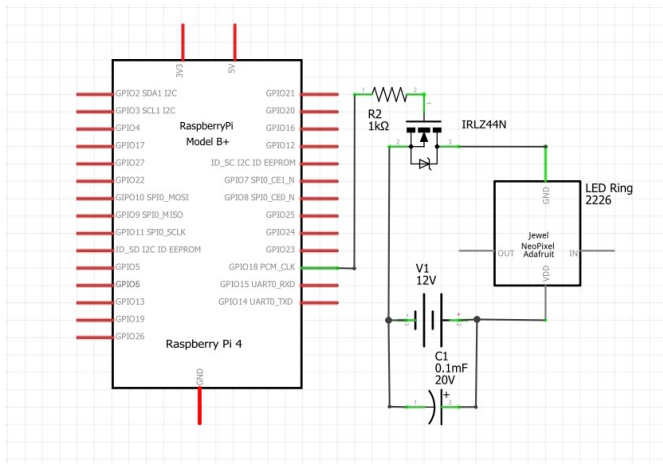


Figure 13: Wiring Schematic

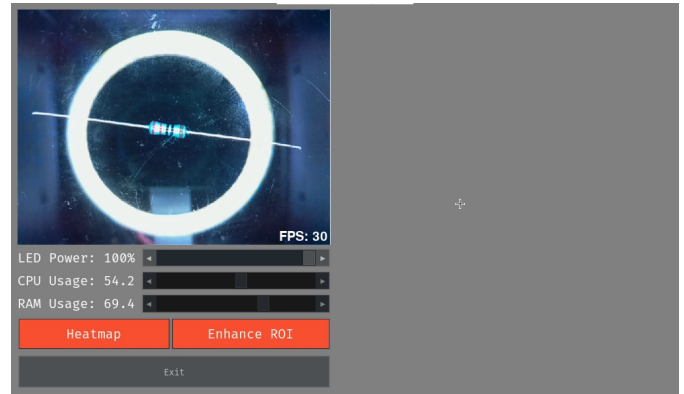


Figure 14: Main UI

run locally on the Pi, and so the Pi will be responsible for running the Computer Vision system, controlling the hardware, and running the user interface. In order to achieve this, I have identified the following software components:

- The User Interface
- The Computer Vision system
- The Hardware Control system

The system's software is written in Python<sup>21</sup>, as it is a language I am very competent with and is well suited for the project, given the availability of libraries for Computer Vision and hardware control. Again, the concept of modularity underpins the design of the software, and so each component is designed to be independent of the others, and can be developed in parallel.

All written Python code adheres to a style defined in a .pylintrc file, which is a configuration file for the pylint<sup>20</sup> linter. This ensures that the code is consistent and readable, and also helps to identify potential bugs and errors. The code is also documented using docstrings for readability and maintainability. Each software component is designed to be self-contained so they can be run independently for testing purposes.

All software constants are defined in a constants.py file, which is imported by all other Python files, allowing for easy modification of constants that modify the behaviour of the system which is useful for testing and debugging.

### User Interface

As the system is intended to be self-contained, it is important to have a user interface that displays all relevant information to the user and allows them to interact with the system. The UI is displayed on the 7" touchscreen display, and is written using Pygame<sup>19</sup>;

This design decision was made after experimentation with other UI libraries like Tkinter<sup>27</sup>, however it was found that Pygame was the most viable for several reasons; it contains a module for camera streaming, which is used to display the camera feed on the UI, and can also be used to feed the camera frames to the Computer Vision system; there is a very large amount of control provided by Pygame over the UI, especially in terms of performance. This is crucial as the UI is only there to provide information to the user, and so it is important that it does not interfere with the Computer Vision system or the hardware control system. With an additional library named pygame\_gui<sup>15</sup>, it is possible to create a UI with GUI elements like buttons, text boxes, and drop down menus, making for a fully comprehensive UI library.

In Figure 14, the current version of the UI is shown. On the left, the camera feed and camera frame rate is displayed, with system CPU and RAM usage displayed, for monitoring purposes. There is a slider to adjust the power of the connected LED Ring light, as well as two buttons related to the Computer Vision system.

The right side is blank for now, but will be used to display information about the Computer Vision system and the hardware control system.

## 3.3 Computer Vision

### 3.4 Design Problems and Solutions

3D printing is typically an iterative process, where the design is printed, tested, and then modified based on the results. This further reinforces the need for a modular design, as it allows for easy modification of each component. After a



few iterations with errors due to tolerances and slightly incorrect measurements, (not relevant to the technical content of this report) I was able to produce a design that physically fit together and mounted the components as intended, however when combined with the software and electronics, I encountered a few problems that I had to overcome by making some design decisions.

## 4 Project Plan

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## 5 Evaluation Plan

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## 6 Safety Plan

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## 7 References

### References

- [1] Praneel Chand and Sunil Lal. Vision-based detection and classification of used electronic parts. *Sensors*, 22(23), 2022. ISSN 1424-8220. doi: 10.3390/s22239079. URL <https://www.mdpi.com/1424-8220/22/23/9079>.
- [2] RS Components. Rs pro 18 awg hook up wire, pvc insulation. <https://docs.rs-online.com/770d/A700000007035534.pdf>, . Accessed: 22-01-2024.
- [3] RS Components. Rs pro c14 snap-in iec connector. <https://docs.rs-online.com/c6ad/0900766b815867b2.pdf>, . Accessed: 22-01-2024.
- [4] DFRobot. 7 inch hdmi display with usb touchscreen. <https://www.farnell.com/datasheets/3162025.pdf>. Accessed: 19-01-2024.
- [5] Amol I. Dhenge, Nāgpur, and A. S. Khobragade. Mechanical nut-bolt sorting using principle component analysis and artificial neural network. 2013. URL <https://api.semanticscholar.org/CorpusID:201742258>.
- [6] Raspberry Pi Foundation. Raspberry pi 4 model b. <https://www.raspberrypi.co>

- m/products/raspberry-pi-4-model-b/. Accessed: 19-01-2024.
- [7] FreeCAD. Freecad. <https://www.freecadweb.org/>. Accessed: 19-01-2024.
- [8] Ce Guo, Xiao ling Lv, Yan Zhang, and Ming lu Zhang. Improved yolov4-tiny network for real-time electronic component detection. *Scientific Reports*, 11(1):22744, 2021. doi: 10.1038/s41598-021-02225-y. URL <https://doi.org/10.1038/s41598-021-02225-y>.
- [9] Infineon. Irlz44n datasheet. [https://www.infineon.com/dgdl/Infineon-IRLZ44N-Datasheet-v01\\_01-EN.pdf?fileId=5546d462533600a40153567217c32725](https://www.infineon.com/dgdl/Infineon-IRLZ44N-Datasheet-v01_01-EN.pdf?fileId=5546d462533600a40153567217c32725). Accessed: 22-01-2024.
- [10] Katranji. Xl4015 datasheet. <https://www.katranji.com/tocimages/files/457120-274359.pdf>. Accessed: 22-01-2024.
- [11] Neural Magic. Deepspase. <https://github.com/neuralmagic/deepspase>,.
- [12] Neural Magic. Yolov8 detection 10x faster with deepspase—over 500 fps on a cpu. <https://neuralmagic.com/blog/yolov8-detection-10x-faster-with-deepspase-500-fps-on-a-cpu/>,.
- [13] Neural Magic. Sparseml. <https://github.com/neuralmagic/sparseml>,.
- [14] Mia Muminovic and Emir Sokic. Automatic segmentation and classification of resistors in digital images. In *2019 XXVII International Conference on Information, Communication and Automation Technologies (ICAT)*, pages 1–6, 2019. doi: 10.1109/ICAT47117.2019.8939034.
- [15] MyreMylar. Pygame gui. [https://github.com/MyreMylar/pygame\\_gui/tree/0cbf7056518377b455d51a8d20167f4029756ad9](https://github.com/MyreMylar/pygame_gui/tree/0cbf7056518377b455d51a8d20167f4029756ad9). Accessed: 22-01-2024.
- [16] Le Nam, Nguyen Mui, and Dang Tu. A method to design vibratory bowl feeder by using fem modal analysis. *Vietnam Journal of Science and Technology*, 57:102, 02 2019. doi: 10.15625/2525-2518/57/1/12859.
- [17] Okdo. Okdo, camera module, csi-2 with 2592 x 1944 resolution. <https://uk.rs-online.com/web/p/raspberry-pi-cameras/2020456/>,. Accessed: 19-01-2024.
- [18] Okdo. Okdo, camera module, csi-2 with 2592 x 1944 resolution specification. <https://docs.rs-online.com/b064/A7000000006917308.pdf>,. Accessed: 19-01-2024.
- [19] Pygame. Pygame documentation. <https://www.pygame.org/docs/>. Accessed: 22-01-2024.
- [20] Pylint. Pylint. <https://github.com/pylint-dev/pylint>. Accessed: 19-01-2024.
- [21] Python. Python. <https://www.python.org/>. Accessed: 19-01-2024.
- [22] Erwin Quilloy, C. Delfin, and M. Pepito. Single-line automated sorter using mechatronics and machine vision system for philippine table eggs. *African Journal of Agricultural Research*, 13:918–926, 04 2018. doi: 10.5897/AJAR2018.13113.
- [23] Gunther Reinhart and Michael Loy. Design of a modular feeder for optimal operating performance. *CIRP Journal of Manufacturing Science and Technology*, 3(3):191–195, 2010. ISSN 1755-5817. doi: <https://doi.org/10.1016/j.cirpj.2010.09.003>. URL <https://www.sciencedirect.com/science/article/pii/S1755581710000908>.
- [24] Richard Silversides, Jian S Dai, and Lakmal Seneviratne. Force Analysis of a Vibratory Bowl Feeder for Automatic Assembly. *Journal of Mechanical Design*, 127(4):637–645, 08 2004. ISSN 1050-0472. doi: 10.1115/1.1897407. URL <https://doi.org/10.1115/1.1897407>.
- [25] Pertiwang Sismananda, Maman Abdurohman, and Aji Gautama Putrada. Performance comparison of yolo-lite and yolov3 using raspberry pi and motioneyeos. In *2020 8th International Conference on Information and Communication Technology (ICoICT)*, pages 1–7, 2020. doi: 10.1109/ICoICT49345.2020.9166199.
- [26] Juan Terven and Diana-Margarita Cordova-Esparza. A comprehensive review

of yolo: From yolov1 to yolov8 and beyond.  
04 2023.

- [27] Tkinter. Tkinter documentation. <https://docs.python.org/3/library/tkinter.html>. Accessed: 22-01-2024.
- [28] Yuanyuan Xu, Genke Yang, Jiliang Luo, Jianan He, and Chenxi Huang. An electronic component recognition algorithm based on deep learning with a faster squeezenet. *Mathematical Problems in Engineering*, 2020: 2940286, 2020. doi: 10.1155/2020/2940286. URL <https://doi.org/10.1155/2020/2940286>.
- [29] Z. Zhang and Massachusetts Institute of Technology. Department of Mechanical Engineering. *Design and Development of an Automated Sorting and Orienting Machine for Vials*. Massachusetts Institute of Technology, Department of Mechanical Engineering, 2019. URL <https://books.google.co.uk/books?id=Mt5WzQEACAAJ>.