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AI Cargo Recognition and Management at Busan Port Smart Logistics Center

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01. Introduction

“Automated Warehouse System”

- Quickly development in automated warehouse trend.
- The automated warehouse improves the operation efficiency.
- The expected development of the warehouse automation in the future.



Figure 1: Smart warehouse concept

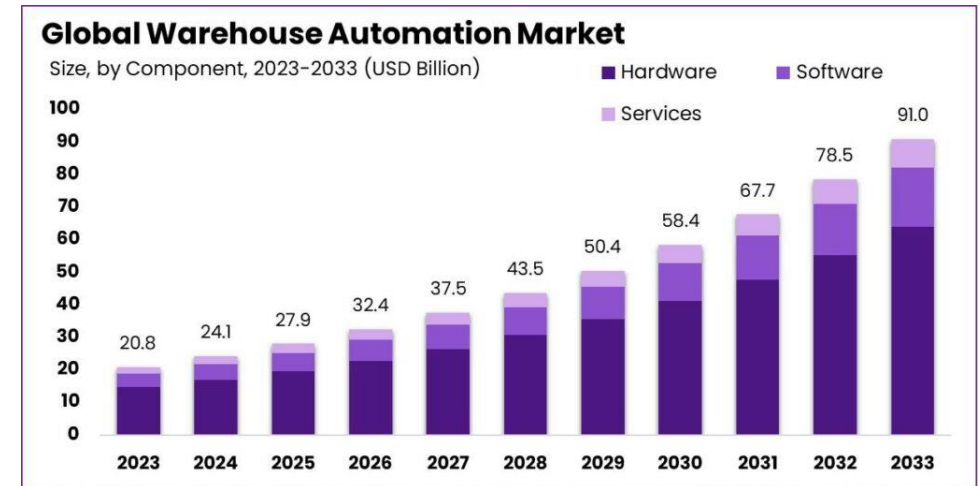


Figure 2: Smart warehouse concept (Source 1)

01. Introduction

“Smart Technology”

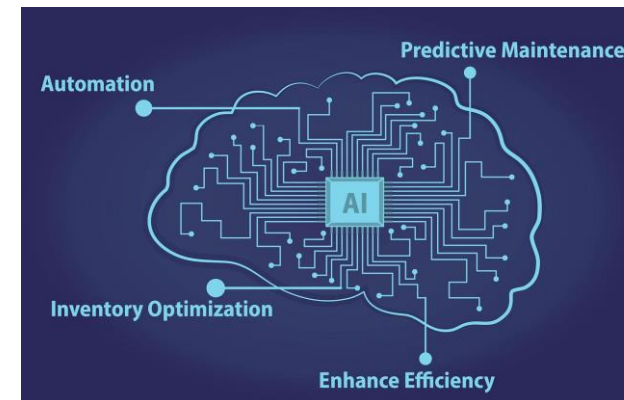
- Automated Guided Vehicle
- IoT technology
- AI and machine learning
- Computer Vision
- Navigation technology



AGV



IoT system



AI technology



Computer Vision

Figure 3: Smart technology

01. Introduction

“Computer Vision”

- Application in inventory management and fulfillment.
- Safety application to detect and avoid the object around the vehicle.

=> Based on the requirement and the quality of the computer vision in automated warehouse, this topic proposed the application of YOLOv8 for inventory management.

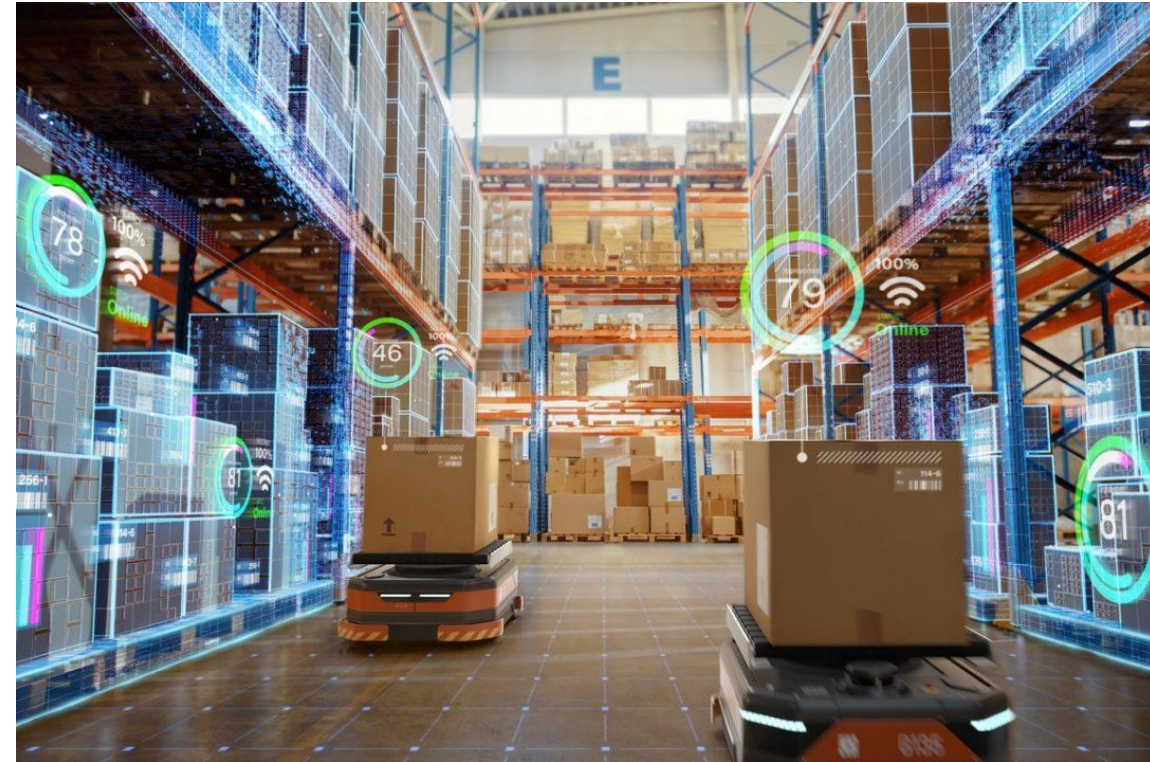


Figure 4: Computer vision application

02. Methodology

“Detection model”

- YOLO (You Only Look Once) is a popular deep learning model used for object detection in images and videos.
- YOLO's key advantage lies in its speed and efficiency, allowing it to achieve high accuracy while maintaining low latency.

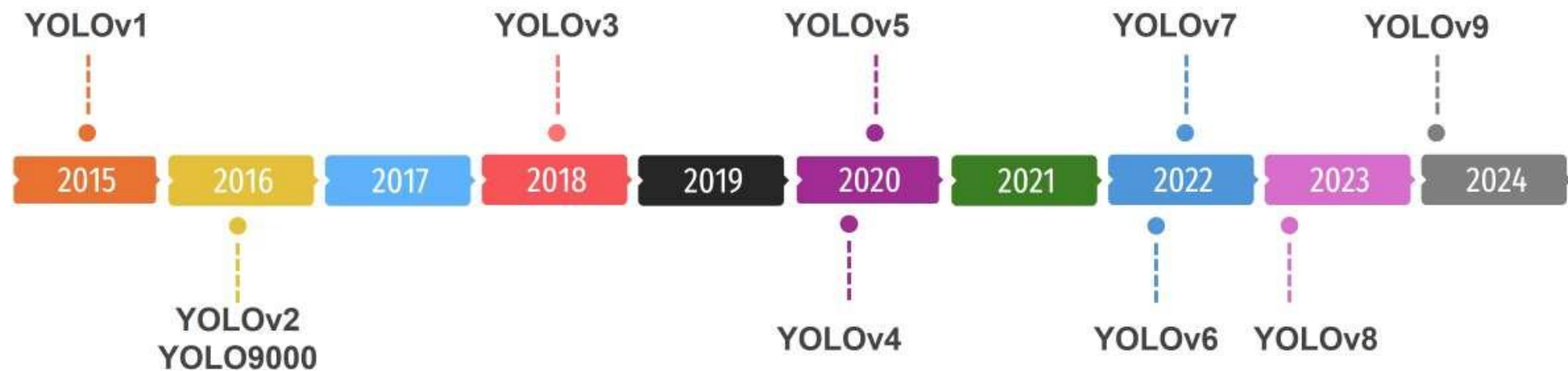


Figure 5: You Only Look Once (YOLO) model timeline

02. Methodology

“Detection model”

- In this application, YOLOv8 is proposed.
- The YOLOv8 architecture is introduced in Figure 6
- YOLOv8 is developed based on YOLOv5, which utilizes a single convolutional neural network (CNN) to detect objects and can run inferences in real time

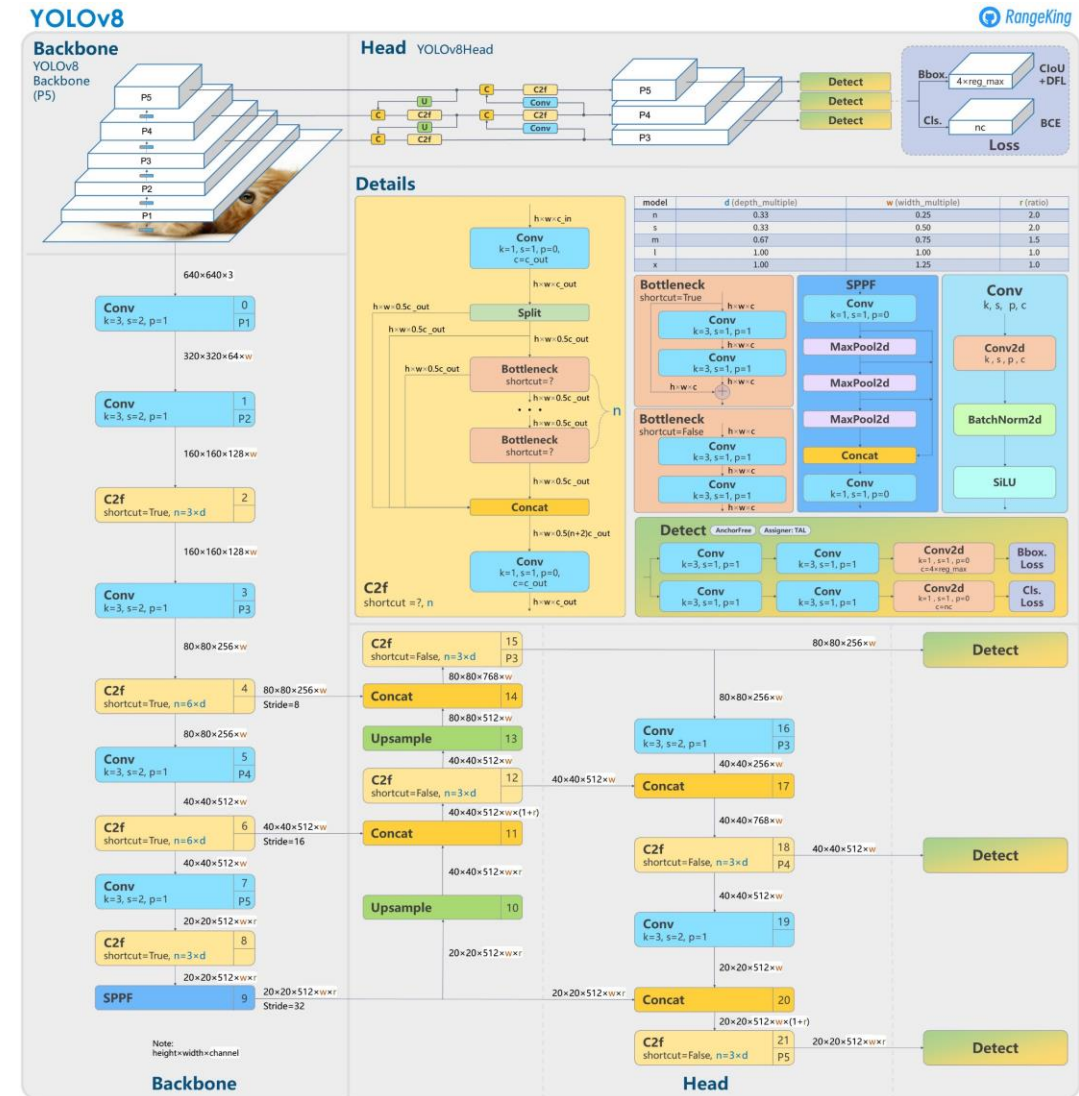


Figure 6: YOLOv8 outperformance architecture

02. Methodology

“Training workflow for object detection model”

- The training workflow for the object detection model involves six steps, which will be introduced sequentially.

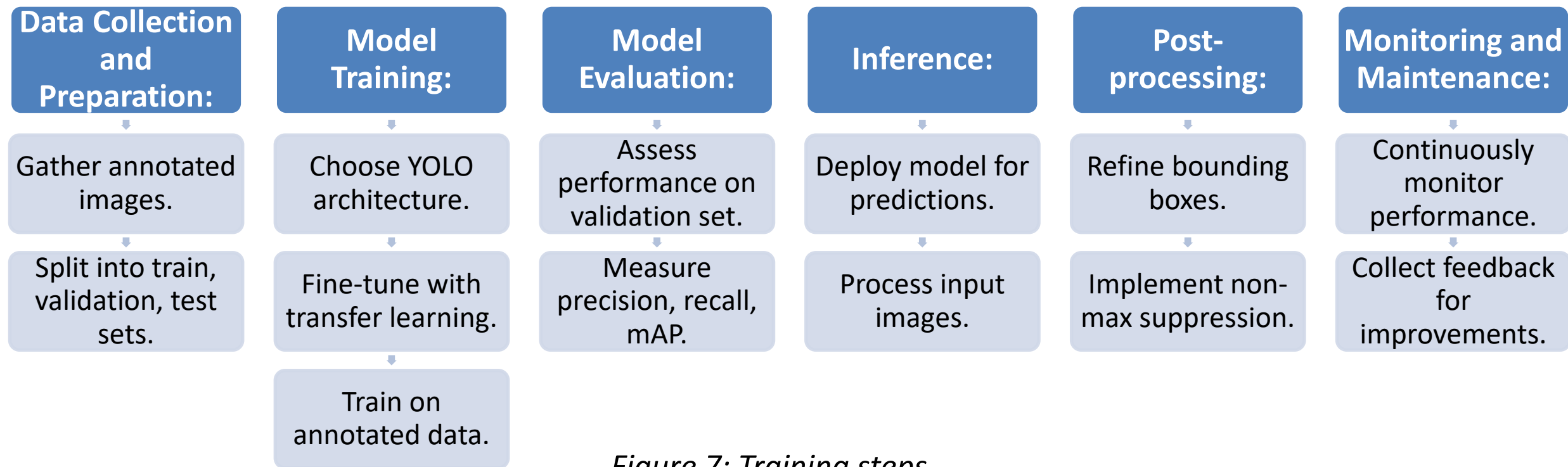


Figure 7: Training steps

02. Methodology

“Warehouse Management System Concept”

- Below system is proposed to make the application, which is the integration of the IoT sensor, Database and the Warehouse management system

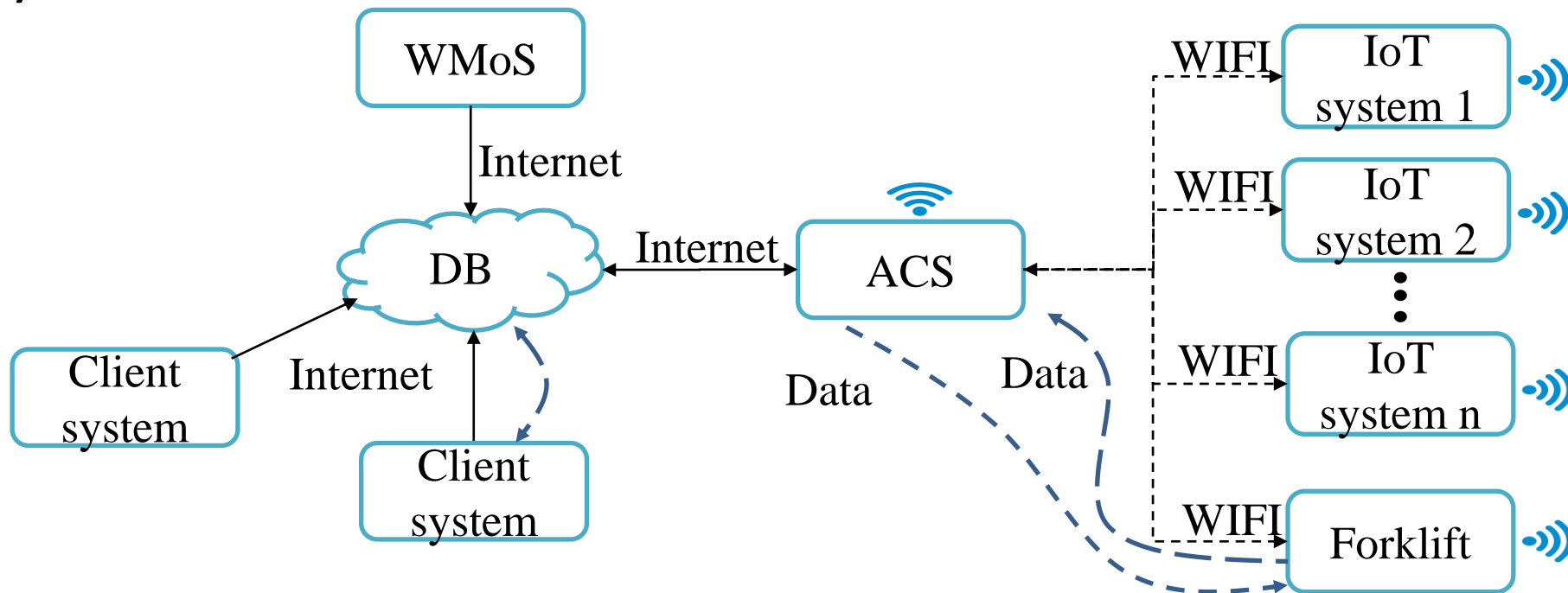


Figure 8: The general concept of the application in warehouse

03. Experiment

“Device Design”

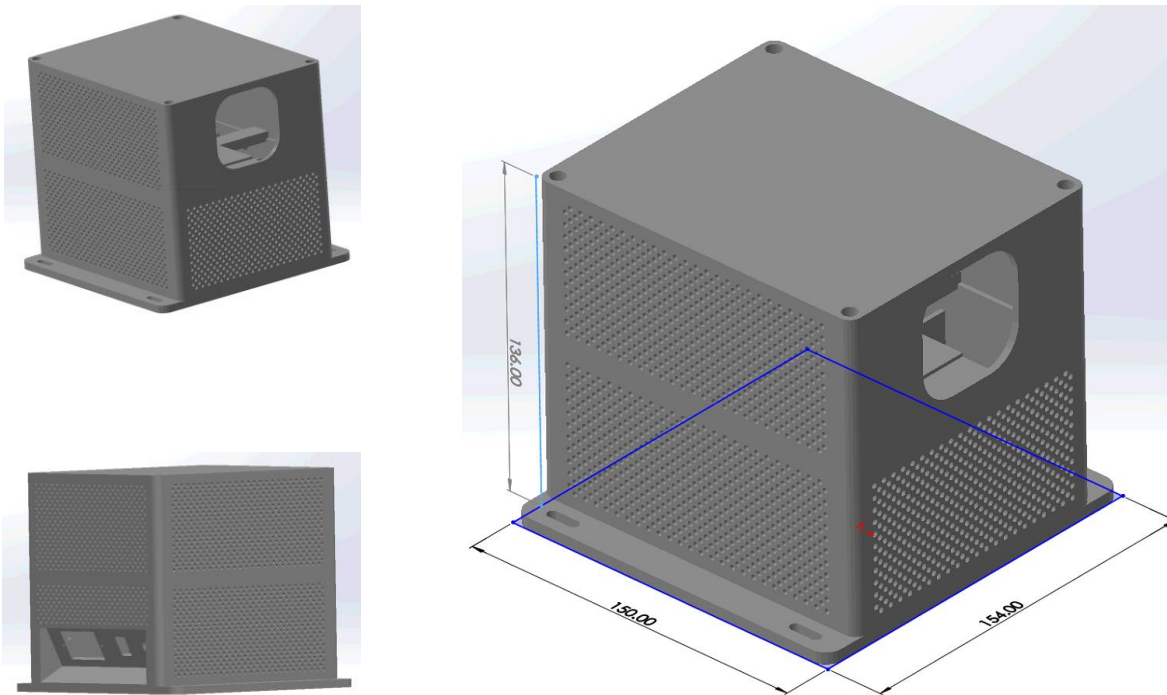


Figure 9. 3D design using Solidworks

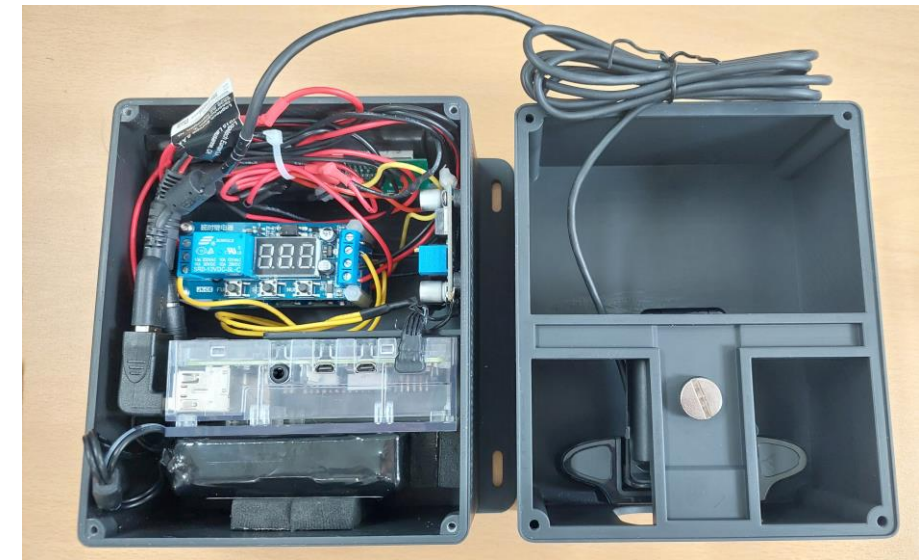


Figure 10. Actual device

03. Experiment

“Software design”

- To communicate between the system and device. WMoS software is developed. The concept of the software is proposed as:

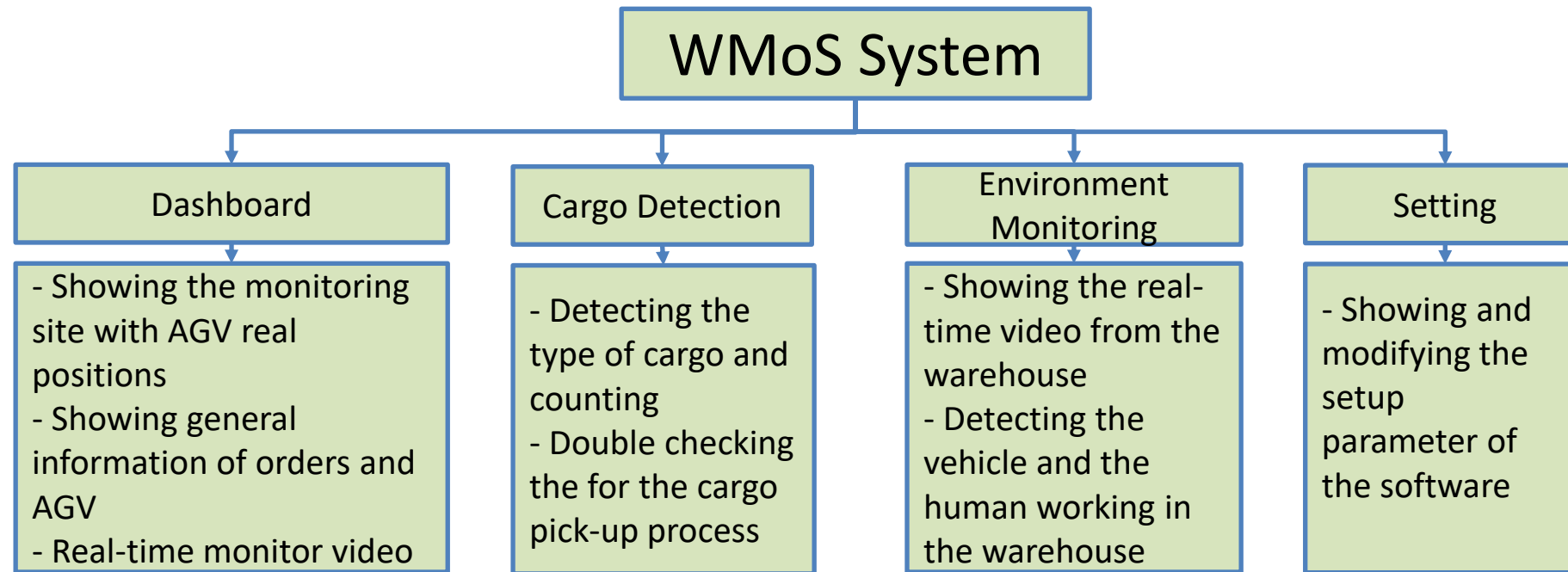


Figure 11. WMoS Software Function Overview

03. Experiment

“Software design”

- Cargo detection function in software

The screenshot displays the Warehouse Monitoring System interface. It features a central camera feed showing a warehouse interior with blue shelving units and white cargo bags. The interface is divided into several sections:

- Camera link checking:** A box on the left containing the text "Camera link checking" with a red arrow pointing to the "Play" button in the top left corner of the camera feed.
- Camera display:** A box on the left containing the text "Camera display" with a red arrow pointing to the camera feed itself.
- History:** A box on the right containing the text "History" with a red arrow pointing to the "History" tab in the top right corner of the interface.
- Result of detection:** A box on the right containing the text "Result of detection" with a red arrow pointing to the "Result of detection" tab in the top right corner of the interface.
- Search area:** A box on the right containing the text "Search area" with a red arrow pointing to the "Search area" tab in the top right corner of the interface.

The interface also includes a table with the following data:

| Time | Device | Position | Cargo A | Cargo B | Cargo C | Cargo D | Cargo E | Cargo F |
|----------------------|----------|----------|---------|---------|---------|---------|---------|---------|
| 4/29/2024 1:39:09 PM | Sample 1 | Sample | 0 | 0 | 0 | 0 | 0 | 0 |
| 4/29/2024 1:39:09 PM | Sample 2 | Sample | 0 | 0 | 0 | 0 | 0 | 0 |
| 4/29/2024 1:39:09 PM | Sample 3 | Sample | 0 | 0 | 0 | 0 | 0 | 0 |

At the bottom of the interface, there is a section for "Receiving from:" and "Number of Cargo:" with a "Connect" button. The "Number of Cargo:" section shows counts for Cargo A through F: Cargo A: 0, Cargo B: 1, Cargo C: 1, Cargo D: 0, Cargo E: 0, and Cargo F: 0.

Figure 12: Cargo Detection Tab

03. Experiment

“Experiment Setup”

- The experiment is carried out in KULS warehouse
- The layout out of the warehouse is introduced in Figure...

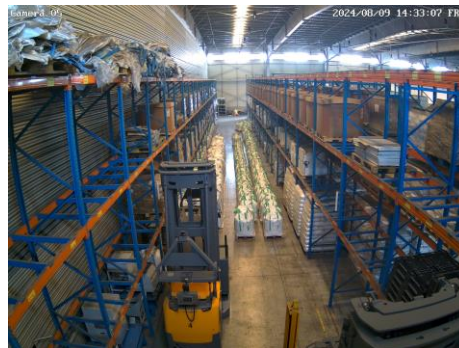


Figure 13. Warehouse CCTV

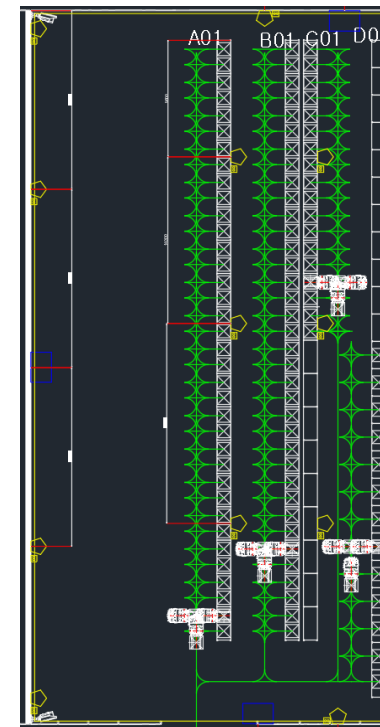


Figure 14. 2D Design

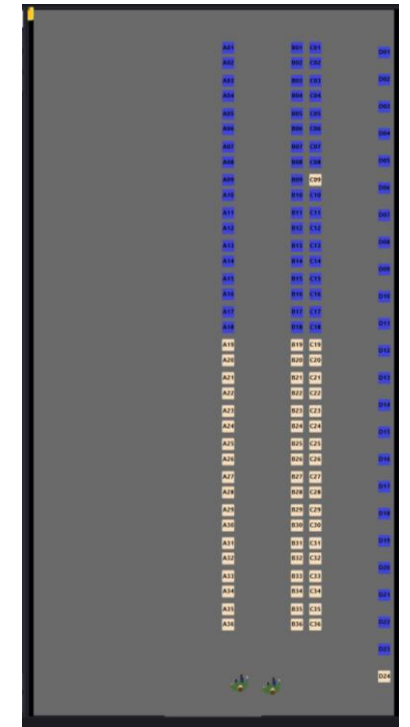


Figure 15. Simulation

03. Experiment

“Experiment Setup”

- Module setup



Figure 16: Camera setup on AGV



Cargo-A



Cargo-B



Cargo-C



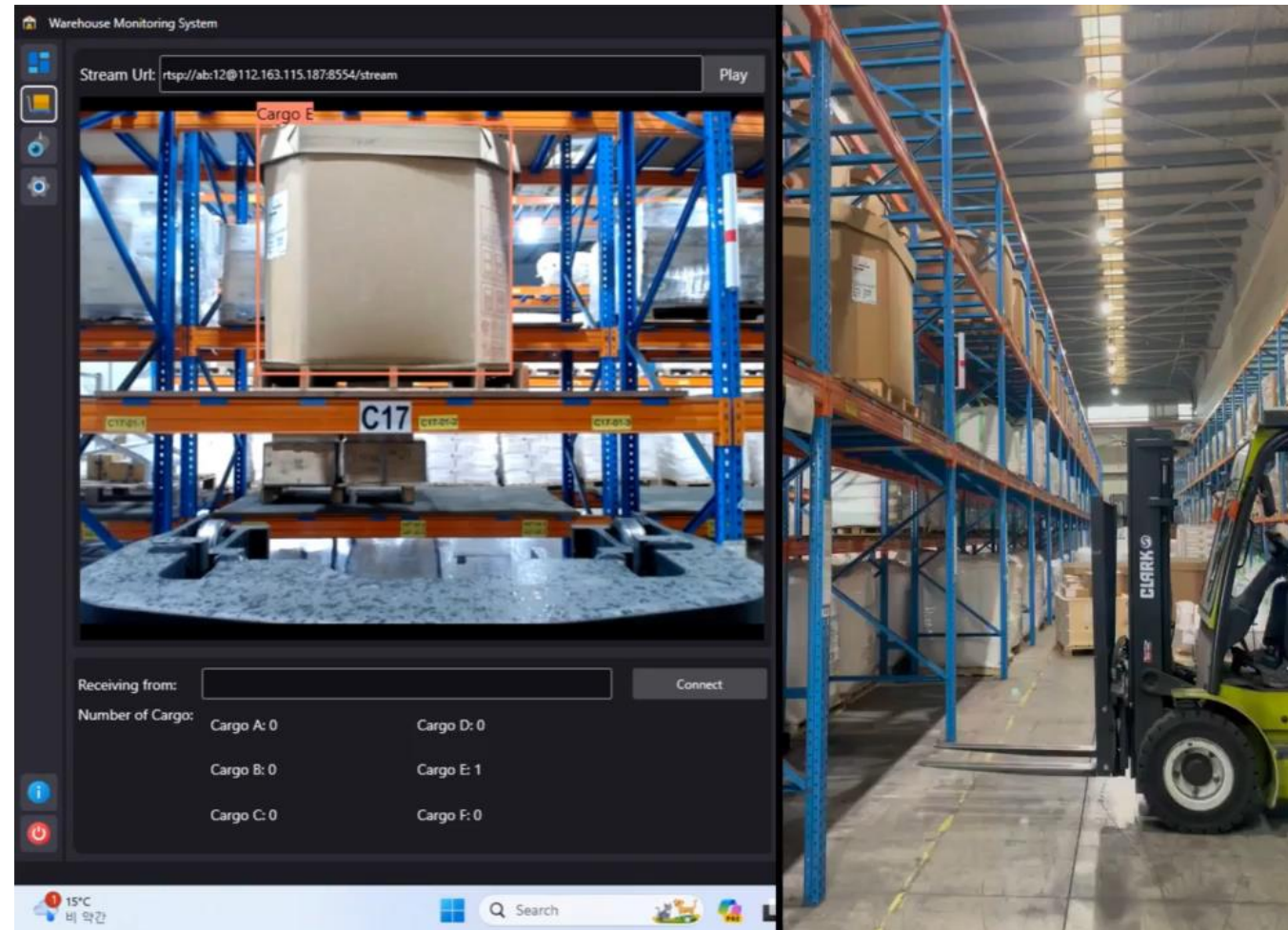
Cargo-D



Cargo-E

Figure 17: Types of cargo using for training

04. Results







Video 1. Experiment Carried out in Warehouse

04. Results

“Main results”

- Table 1 shows the summary of the detection experiment.







Table 1: Results of cargo detection and counting

| No. | Object detection | Detection quantity | Actual quantity | Pass /Fail | No. | Object detection | Detection quantity | Actual quantity | Pass /Fail |
|-----|---|--|--|------------|-----|---|--|--|------------|
| 1 |  | Cargo-A: 0 Cargo-B: 0 Cargo-C: 4 Cargo-D: 1 Cargo-E: 0 | Cargo-A: 0 Cargo-B: 0 Cargo-C: 4 Cargo-D: 1 Cargo-E: 0 | Pass | 6 |  | Cargo-A: 2 Cargo-B: 0 Cargo-C: 0 Cargo-D: 0 Cargo-E: 0 | Cargo-A: 2 Cargo-B: 0 Cargo-C: 0 Cargo-D: 0 Cargo-E: 0 | Pass |
| 2 |  | Cargo-A: 0 Cargo-B: 0 Cargo-C: 4 Cargo-D: 0 Cargo-E: 0 | Cargo-A: 0 Cargo-B: 0 Cargo-C: 4 Cargo-D: 0 Cargo-E: 0 | Pass | 7 |  | Cargo-A: 0 Cargo-B: 6 Cargo-C: 0 Cargo-D: 0 Cargo-E: 0 | Cargo-A: 0 Cargo-B: 6 Cargo-C: 0 Cargo-D: 0 Cargo-E: 0 | Pass |

04. Results

“Main results”

Table 1: Results of cargo detection and counting

| No. | Object detection | Detection quantity | Actual quantity | Pass /Fail | No. | Object detection | Detection quantity | Actual quantity | Pass /Fail |
|-----|---|--|--|------------|-----|---|---|---|------------|
| 3 |  | Cargo-A: 0 Cargo-B: 0 Cargo-C: 0 Cargo-D: 0 Cargo-E: 4 | Cargo-A: 0 Cargo-B: 0 Cargo-C: 0 Cargo-D: 0 Cargo-E: 4 | Pass | 8 |  | Cargo-A: 6 Cargo-B: 2 Cargo-C: 0 Cargo-D: 0 Cargo-E: 0 | Cargo-A: 6 Cargo-B: 2 Cargo-C: 0 Cargo-D: 0 Cargo-E: 0 | Pass |
| 4 |  | Cargo-A: 0 Cargo-B: 0 Cargo-C: 4 Cargo-D: 0 Cargo-E: 0 | Cargo-A: 0 Cargo-B: 0 Cargo-C: 4 Cargo-D: 0 Cargo-E: 0 | Pass | 9 |  | Cargo-A: 0 Cargo-B: 10 Cargo-C: 0 Cargo-D: 0 Cargo-E: 0 | Cargo-A: 0 Cargo-B: 10 Cargo-C: 0 Cargo-D: 0 Cargo-E: 0 | Pass |
| 5 |  | Cargo-A: 3 Cargo-B: 0 Cargo-C: 0 Cargo-D: 0 Cargo-E: 0 | Cargo-A: 3 Cargo-B: 0 Cargo-C: 0 Cargo-D: 0 Cargo-E: 0 | Pass | 10 |  | Cargo-A: 0 Cargo-B: 9 Cargo-C: 0 Cargo-D: 0 Cargo-E: 2 | Cargo-A: 0 Cargo-B: 9 Cargo-C: 0 Cargo-D: 0 Cargo-E: 2 | Pass |

05. Conclusion

- The integration of the YOLOv8 model into AGV system presents a significant advancement for warehouse automation. This approach not only enhances inventory control and order fulfillment but also boosts overall productivity and operational effectiveness.
- The successful implementation of this system represents a critical step towards realizing fully automated warehouse operations, thereby optimizing productivity and operational efficiency in modern logistics environments.

References

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- 6) YOLO model - <https://docs.ultralytics.com/>

QUESTIONS

Q & A

ANSWERS