



Progress Report:

Digital Twin Technology & Automated Ground-based Predatory Mite Releaser

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Machine learning and
field robotics for
precision agriculture

Smart Agriculture Lab

Where Agriculture & Technology Meet to Build Future

Digital Twin for Strawberry Farm

Challenges in Ag Automation & Robotics



Time-Consuming

Testing is limited to few months



Special Expertise

Installation
Maintenance



Costly

Data collection
Travel to farm site



Risk

Bad data
Crop is lost due to disease or frost



How to reduce the turnaround time for autonomous systems?

Synthetic Data

Definition:

- Data generated artificially, not from real-world observations

Purpose:

- Can replicate the characteristics of real data
- Useful for enhancing datasets, privacy, and testing

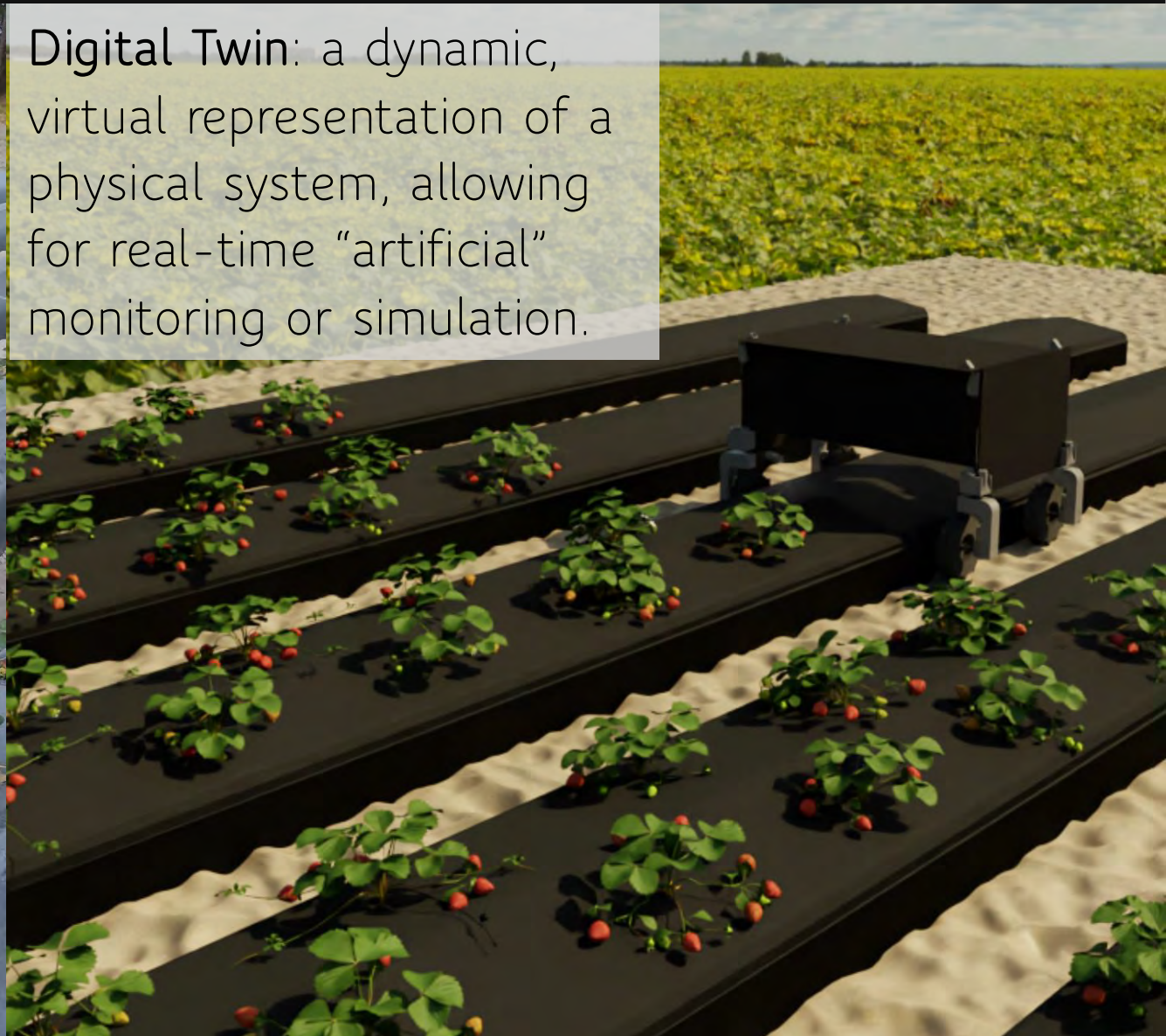
Generation Techniques:

- Statistical methods (e.g., bootstrapping, synthetic regression)
- Generative models (e.g., GANs)
- Simulations & 3D graphics

Exploring the Virtual Strawberry Farm



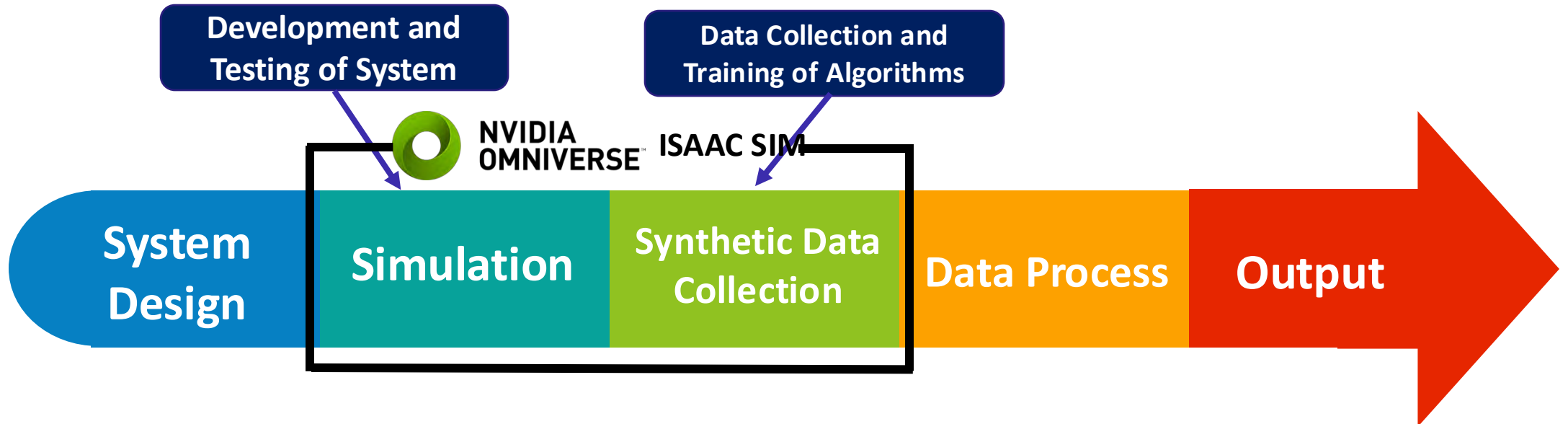
Digital Twin: a dynamic, virtual representation of a physical system, allowing for real-time “artificial” monitoring or simulation.



Benefits of Digital Twin

Speed up the development process of robots and AI

- Data augmentation for limited datasets.
- Training robust machine learning models.
- Testing new agricultural tools without risk.
 - Ensuring farmers' data privacy.



Creating a Scene



Procedural Modeling

- Rules-based approach for modeling rather than manual design of components
- Flexibility: Easier to modify model components instead of recreating the entire model again
- Randomization: Can randomize model



Clump radius	<input type="range"/>	0.504347
Clump angle	<input type="range"/>	0.634783
Leaf stalk length	<input type="range"/>	0.226086
Leaf stalk angle	<input type="range"/>	0.930435
Leaf stalk curving	<input type="range"/>	0.32826
Leaf scale	<input type="range"/>	0.417391
Leaf width	<input type="range"/>	0.617391
Leaf stalk axial rotation	<input type="range"/>	0.521739
Flower stem length	<input type="range"/>	0.730435
Flower stem angle	<input type="range"/>	0.391305
Sepal length	<input type="range"/>	1
Sepal width	<input type="range"/>	1
Petal length	<input type="range"/>	0.965217
Petal width	<input type="range"/>	0.660871
Fruit scale	<input type="range"/>	0.53913
Fruit length	<input type="range"/>	0
Fruit shape change	<input type="range"/>	0
Fruits warp intensity	<input type="range"/>	0.608696
Flower and fruit growth	<input type="range"/>	0.626087
▼ MOTHER PLANT DENSITY		
Plant unit (1 to 10)	<input type="range"/>	0.017391
Leaf petiole (3 to 13 +/- 1)	<input type="range"/>	0.373913
Flower or fruit stem (1 to 5 +/- 1)	<input type="range"/>	0.66087
Fruit warped ratio	<input type="range"/>	0

More realistic strawberry Plant - Runner



Hardware Setup in the Field

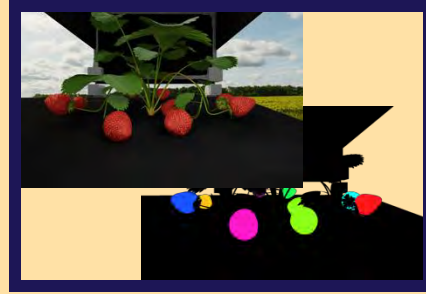


Hardware Setup in Isaac Sim

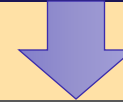
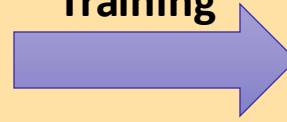


Software Setup

Synthetic Data & Auto Labeling



Model Training



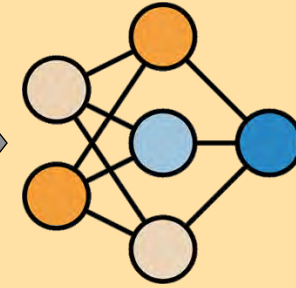
Data Collection



Simulated Camera



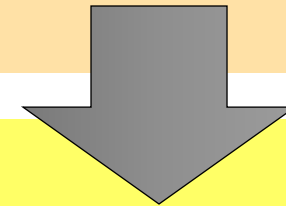
Mask R-CNN



Fruit Detection



VIRTUAL



REAL



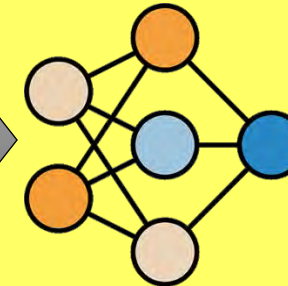
Data Collection



Real Camera



Mask R-CNN



Fruit Detection



Field



Simulation



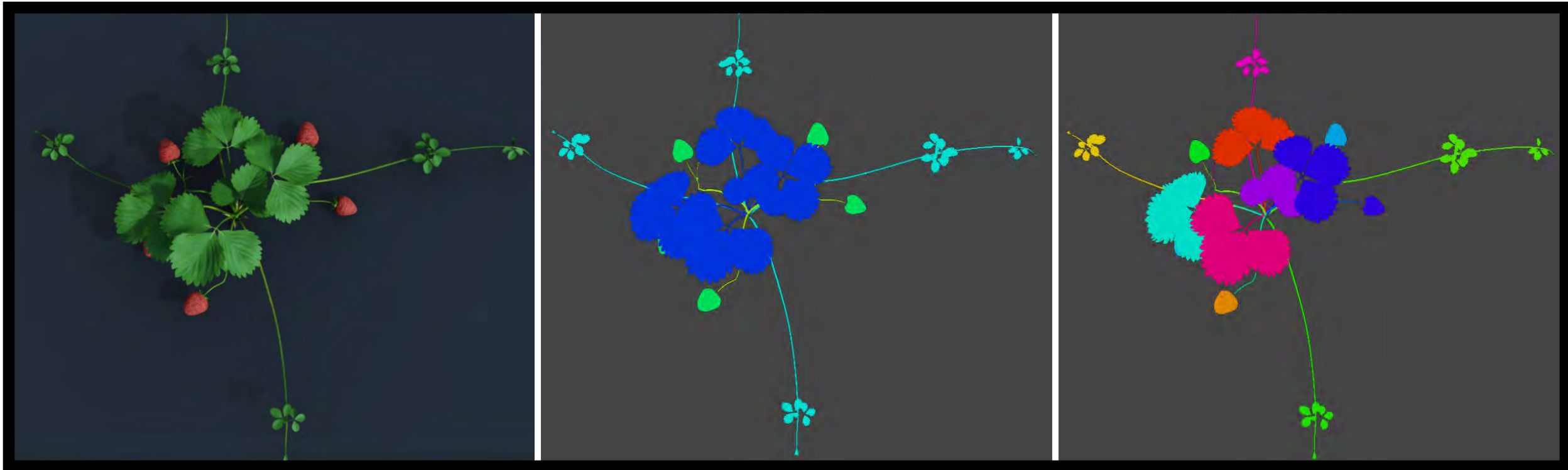
Camera Perspectives

Automated Labeling

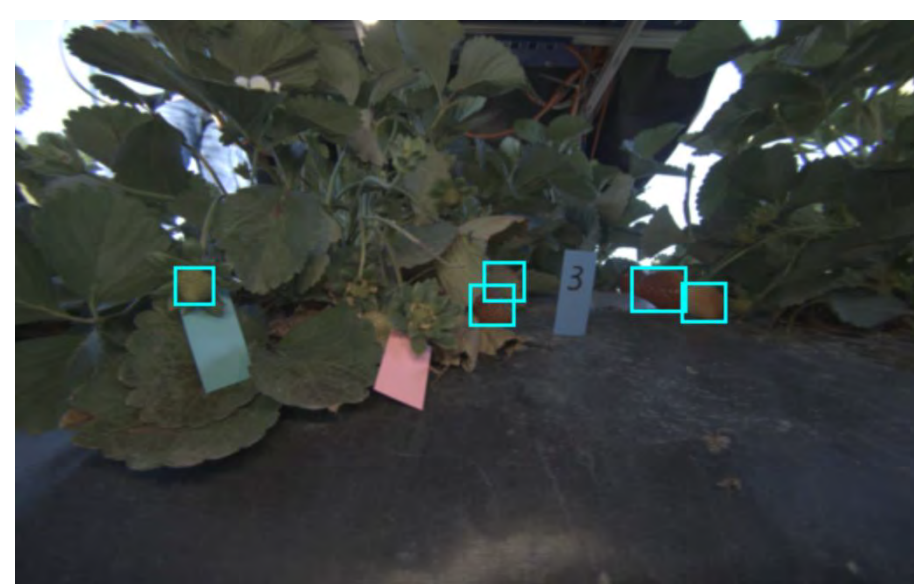
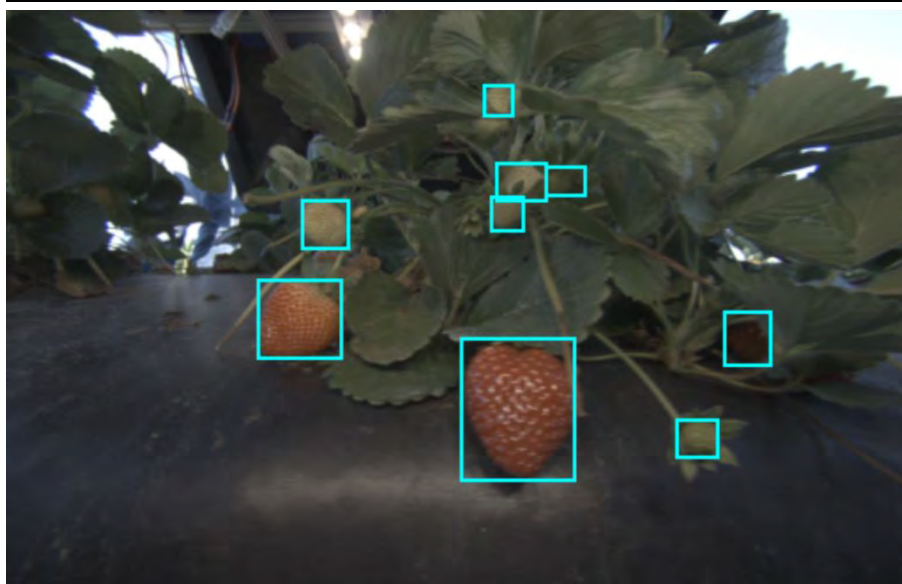
RGB

Semantic Segmentation

Instance Segmentation



Results: Fruit Detection on Field Images



Digital Twin for Runner Cutting

Bidirectional communication between ROS and the hardware, as well as its reflection in the virtual environment of Isaac Sim



Current Challenge: need for innovative 3D Modeling

- Automatic rendering based on AI
- Speedy performance



Results: Fruit Detection

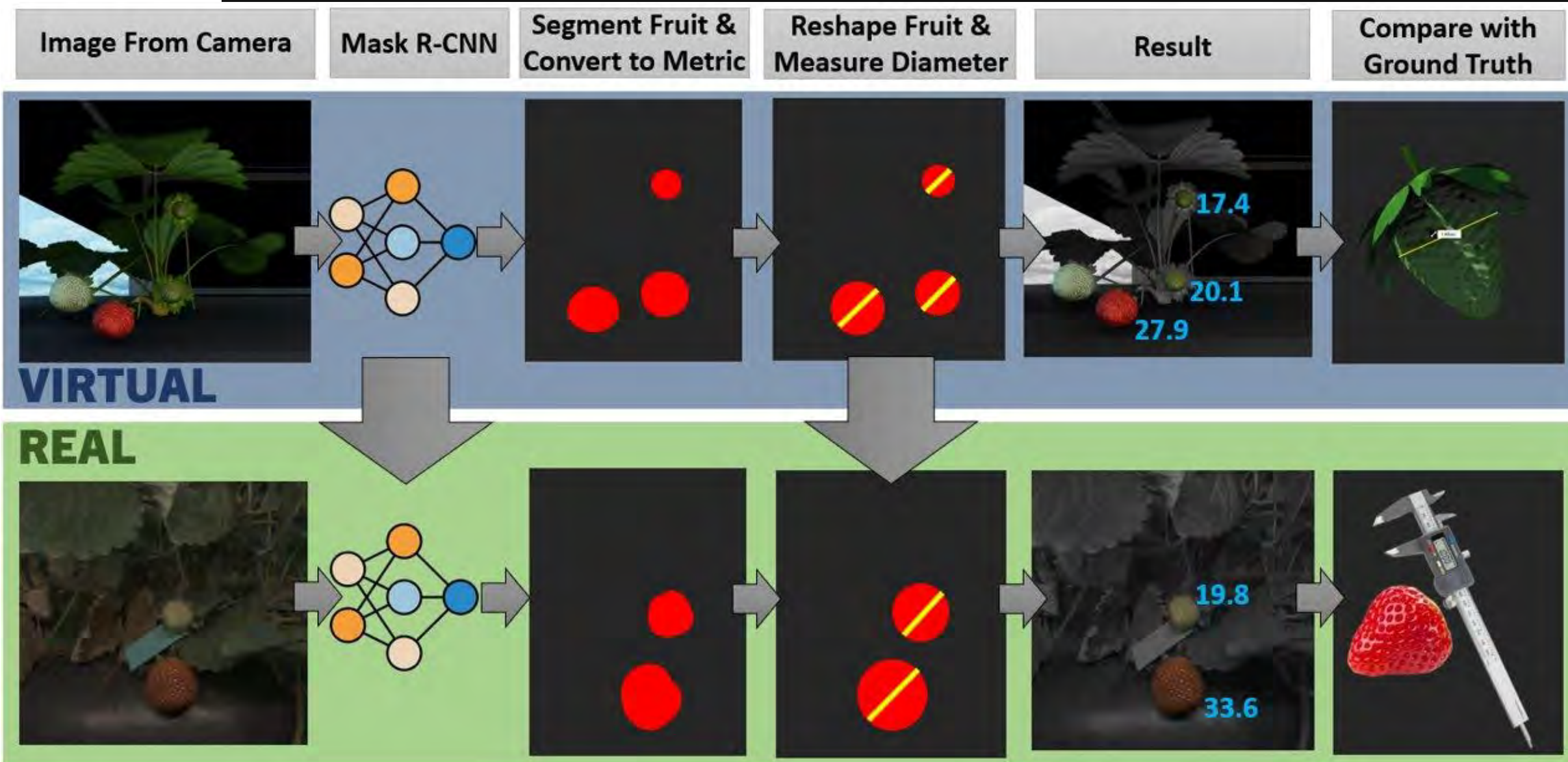
Training using Synthetic Data only

	TP	FP	FN	Precision	Recall	F1-Score
 Synthetic Data Gen-1	56	2	86	0.96	0.39	0.56
 Synthetic Data Gen-2	127	17	15	0.88	0.89	0.89
Gen-1 + Gen 2	96	3	44	0.97	0.69	0.80
Real Images	129	2	12	0.98	0.91	0.95

Results: Fruit Detection

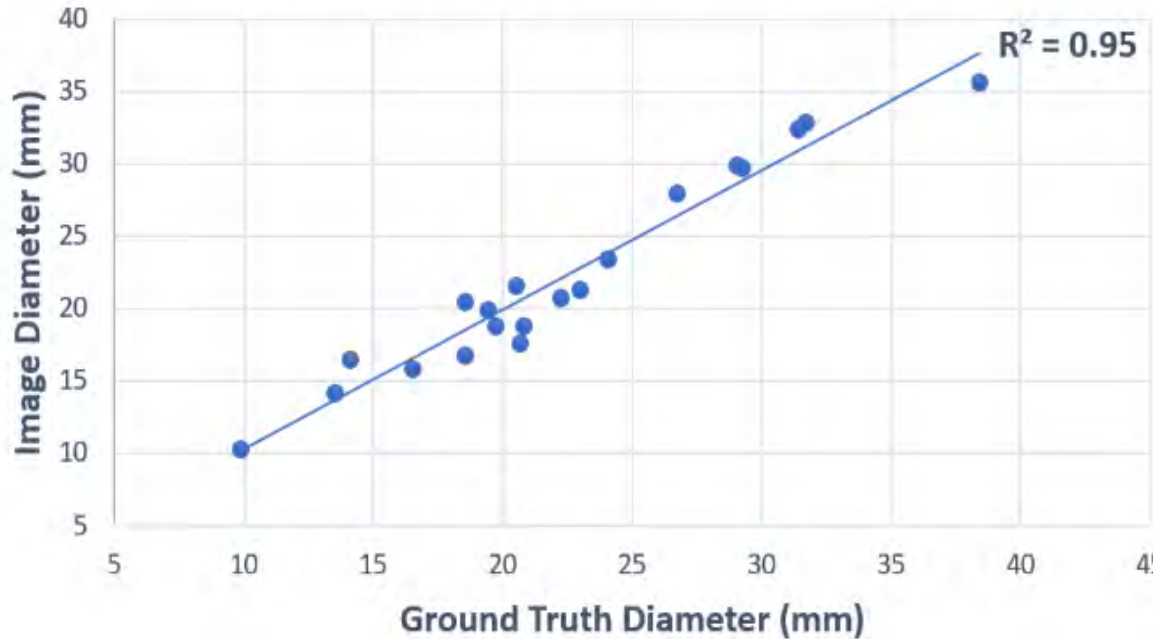
Trial 1				
	Fruit Count	Precision	Recall	F1-Score
All Fruit	295	0.95	0.89	0.92
Red Fruit	167	0.95	0.99	0.97
White Fruit	61	1.00	0.87	0.93
Green Fruit	67	0.94	0.72	0.82
Trial 2				
All Fruit	134	0.89	0.75	0.81
Red Fruit	39	0.87	0.95	0.91
White Fruit	52	0.98	0.81	0.89
Green Fruit	43	0.83	0.58	0.68

Results: Fruit Sizing



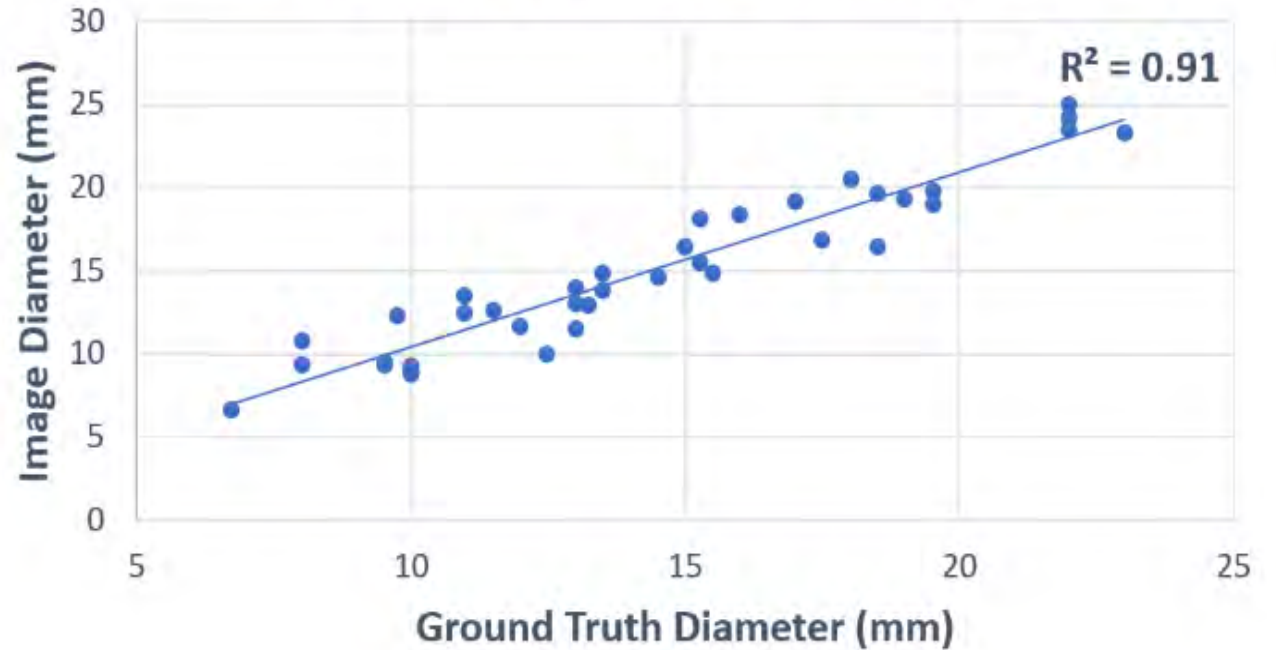
Results: Fruit Sizing

Simulation Based Diameter Estimation



Average Diameter Error: 1.5 mm

Field Based Diameter Estimation



Average Diameter Error: 1.6 mm

Ground-based Predatory Mite Releaser

How to combat resistance to pesticides?

- Use of predatory mites like *Amblyseius swirskii* can help in this endeavor
- However, this is labor and time intensive and alternatives like aerial releases with drones are expensive and not very precise

OBJECTIVES

Develop a computer vision algorithm to identify and locate strawberry plants in a test field

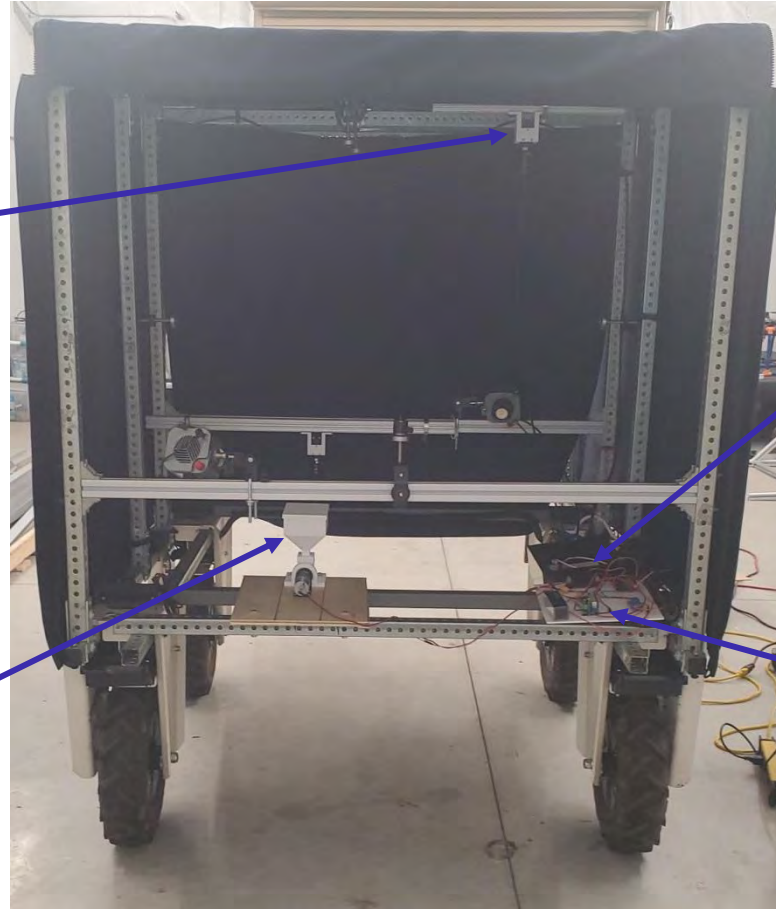
Design and implement a mite dispensing system to release predatory mites on strawberry plants

Materials Used for System

Flir Blackfly S camera



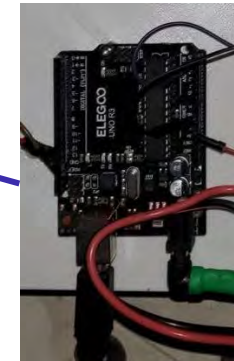
Mite Releaser Prototype



Intel Edge Computer

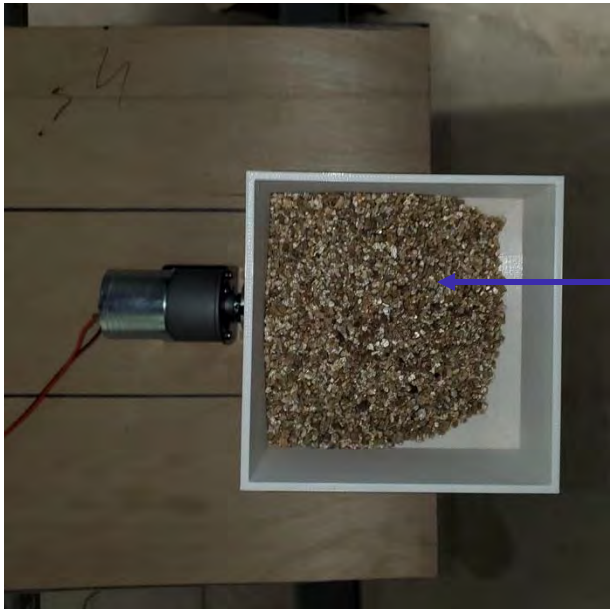
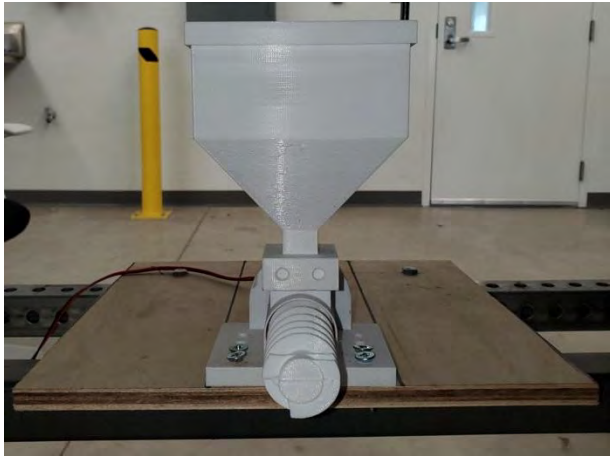


Arduino Uno Microcontroller



Ground Vehicle

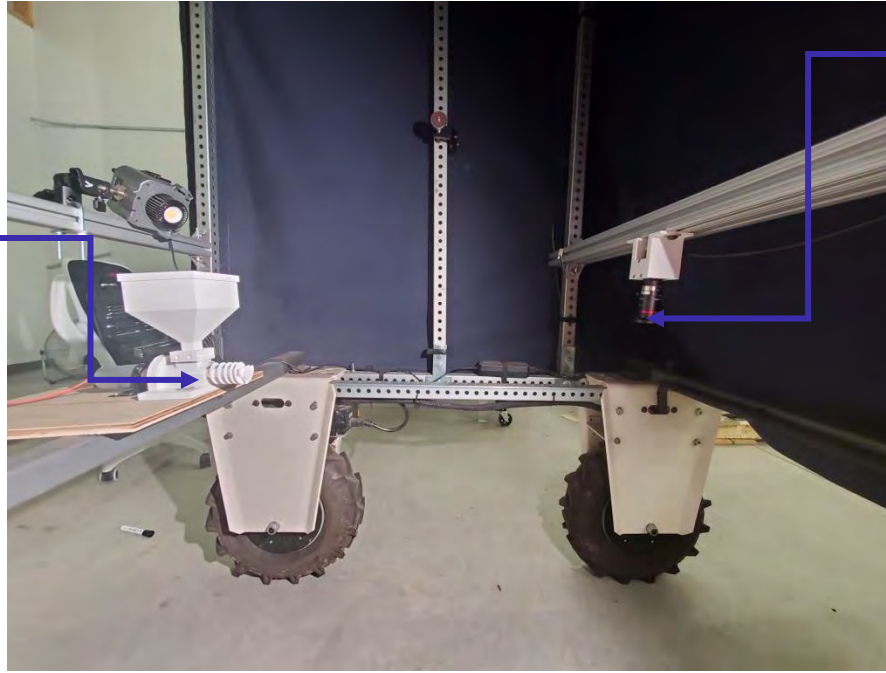
Mite Releaser Prototype



Releaser

Vermiculite

- Testing with Vermiculite substrate
- Rotates at a speed of 30 RPM for 0.2 seconds (avg 1.8 ml/plant)
- Releases Uniform amount of Vermiculite per rotation

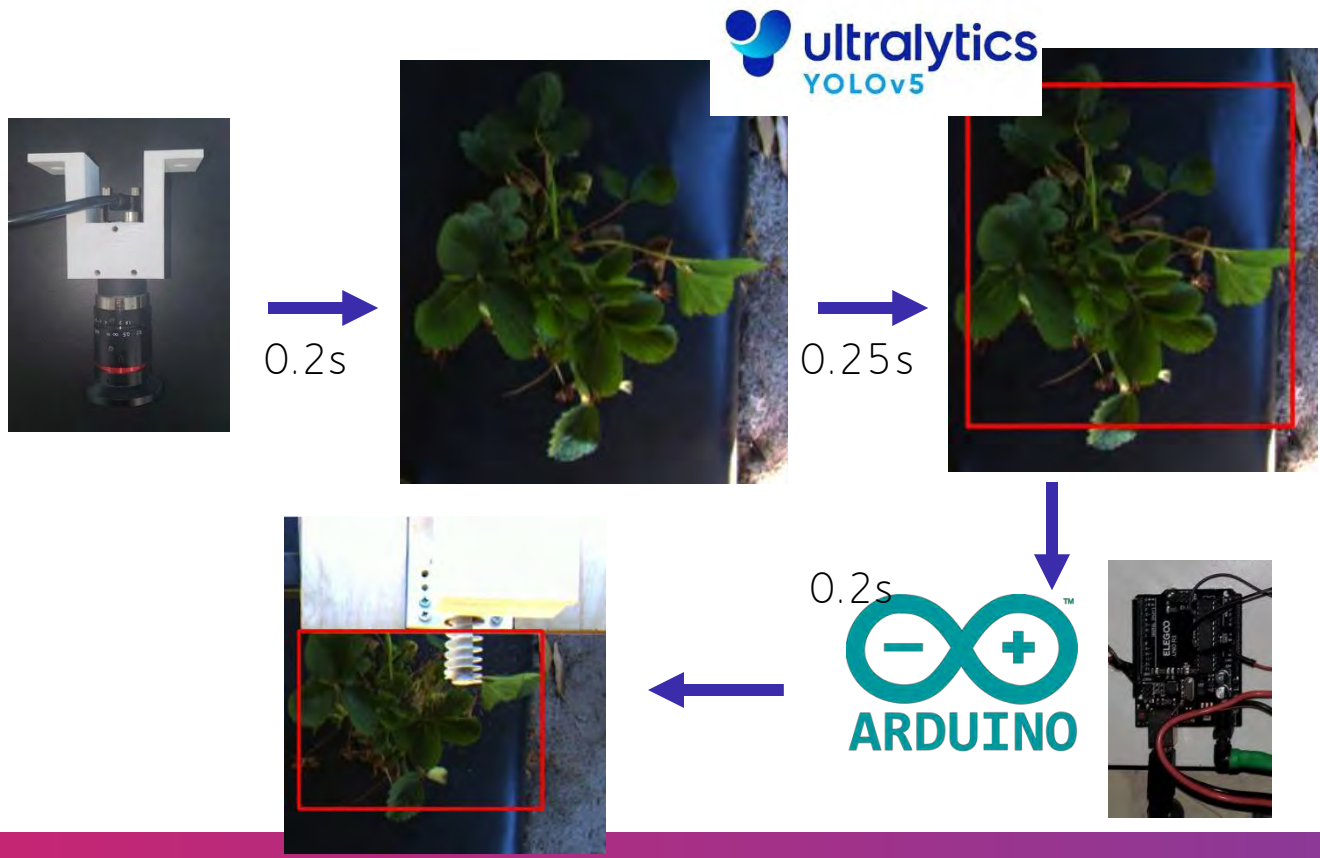


Camera

Screw Conveyor release mechanism

- Camera placed 28 inches from releaser
- Vehicle speed = 0.5 mile/hr
- Camera sends signal to the releaser when it reaches the intended plant.

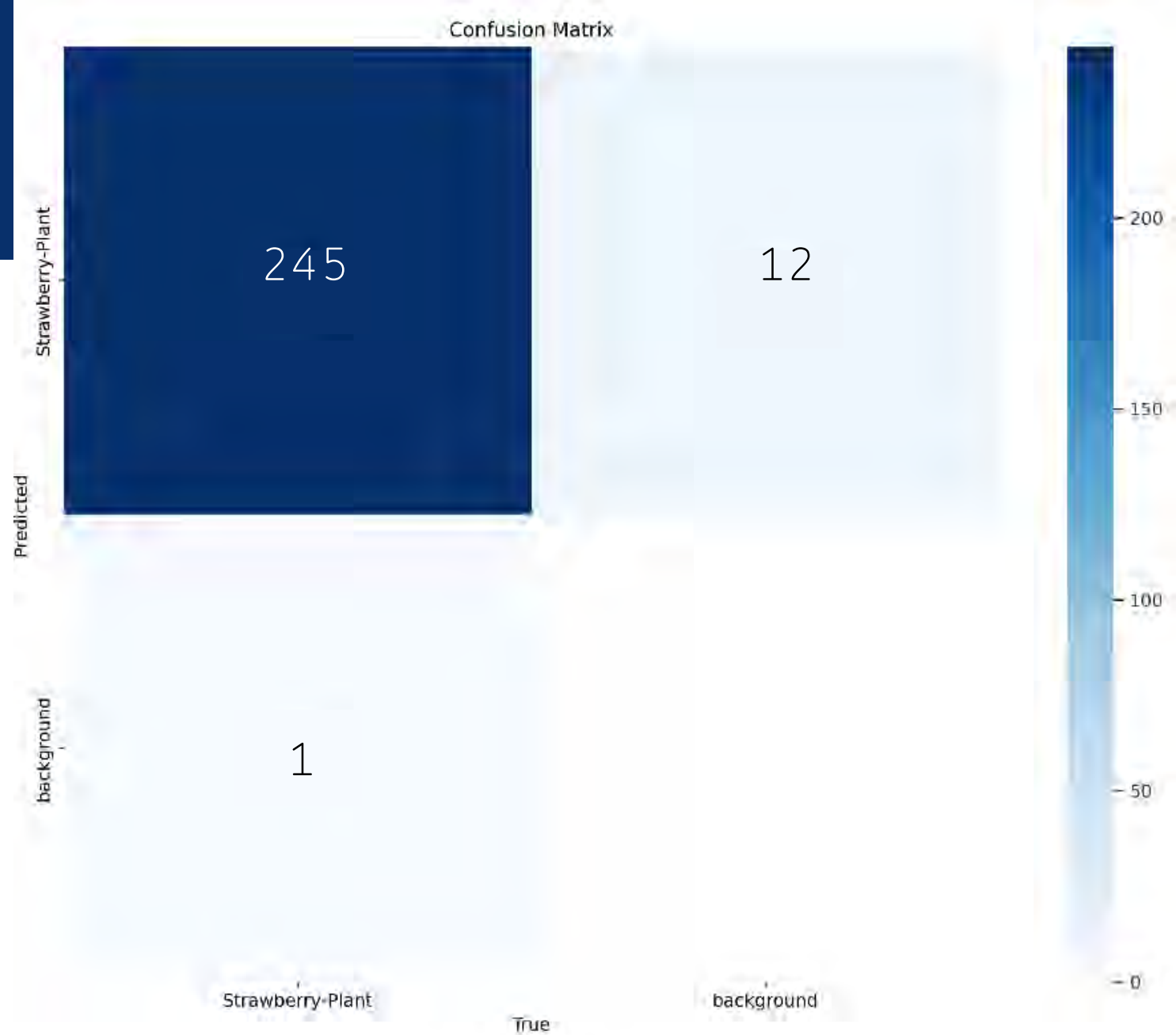
System Workflow



Object	Count
Plant	325
Gap	11
Weed	22
Total	358

Strawberry plants detection model

Parameter	Value
Images	48
Instances	246
Precision	0.97
Recall	0.99
mAP* 0.5	0.96
mAP 0.5:0.95	0.99



Results

Case A

248
(69%)

Success –
Plant is sprayed

When ground vehicle crosses a strawberry plant, actuator turns when area is 80% of the original bounding area

Case B

77
(21%)

Failure
Plant missed

When ground vehicle crosses a strawberry plant, actuator does not turn-on when area is 80% of the original bounding area

Case C

33
(9%)

Failure
non-target
sprayed

Actuator turns on when ground vehicle drives through weeds or a gap in the strawberry bed

Results



What We Will Improve

Hardware Limitations

Hardware Delay &
Code Optimization



Variable Rate Application



Environmental Limitations

Windy Conditions



Vehicle/Plant Row
Misalignment



Future Research Area

- Improved computation with faster computer processors
- Plant specific chilli thrips symptoms detection
- Automated vehicle navigation and predatory mite release
- Double plant row functionality
- Wind blocker

Acknowledgement



Uchechukwu Ilodibe
PhD Student



Namrata Dutt
PhD Student

ANY QUESTIONS?

YOU CAN FIND ME AT:

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THANK YOU!
