



Investigation into the correlation between AI in a simulated environment and using a real low-cost UAV



Project Github

Harkeerat Singh

Applied Artificial Intelligence - MSC

Aims & Objectives

AIMS:

The aim of this thesis is to gain an understanding through an investigative experimentation the effectiveness of an object detector under varying illumination conditions and altitudes on its ability to classify and detect trained targets in a simulated and real-world environment by using a camera sensor from a UAV to determine if there are any differences.

OBJECTIVES:

1. Devise an Experiment that is applicable in a simulation and the real world that utilises key “object targets” that the UAV can classify.

1. Accurately construct the Real-World experimental space within the simulation environment.

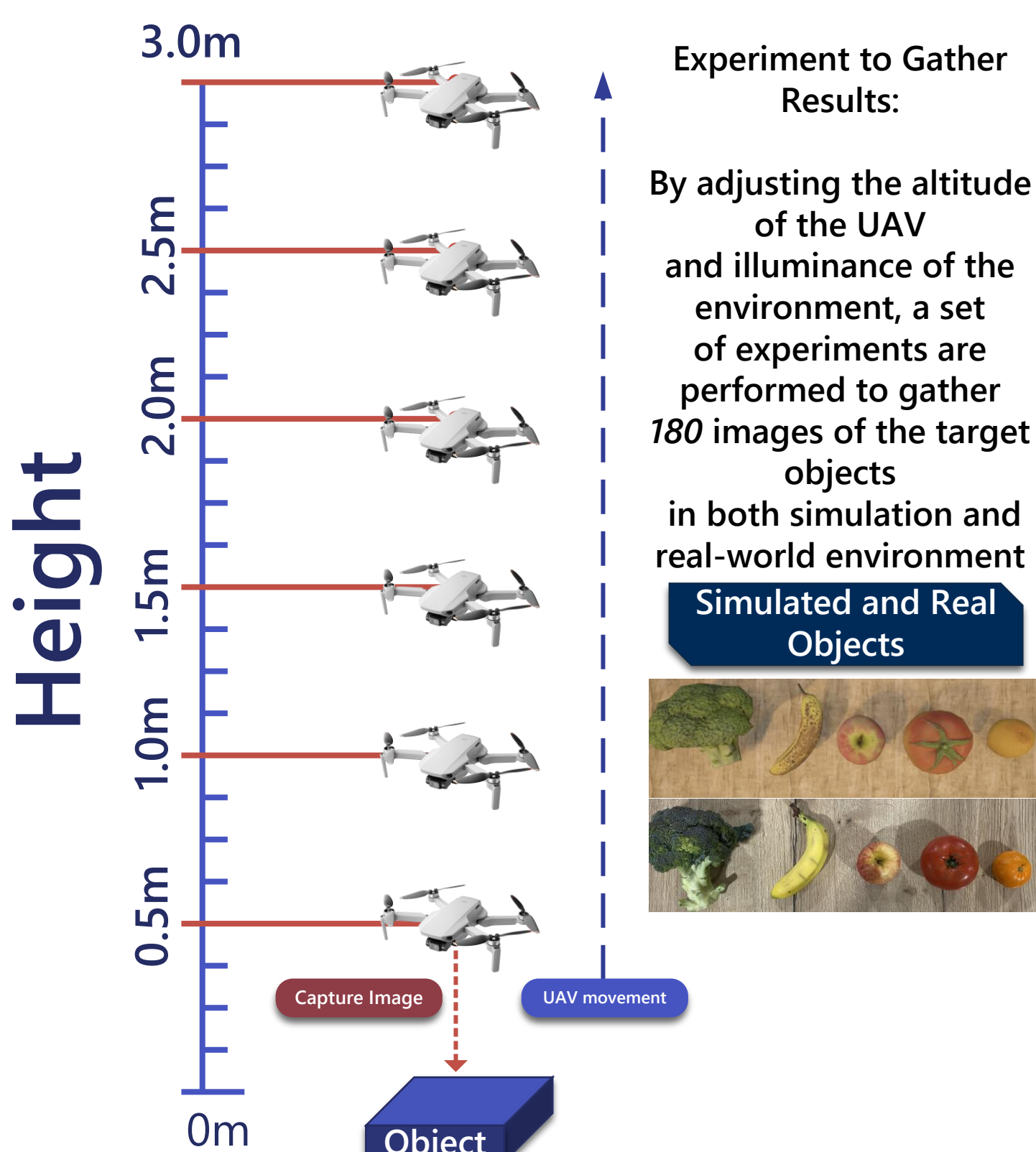
2. Synthesise & Train an AI algorithm within the realm of object detection using current methods.

3. Integrate the trained object detector algorithm into a UAV within a simulated and real-world environment.

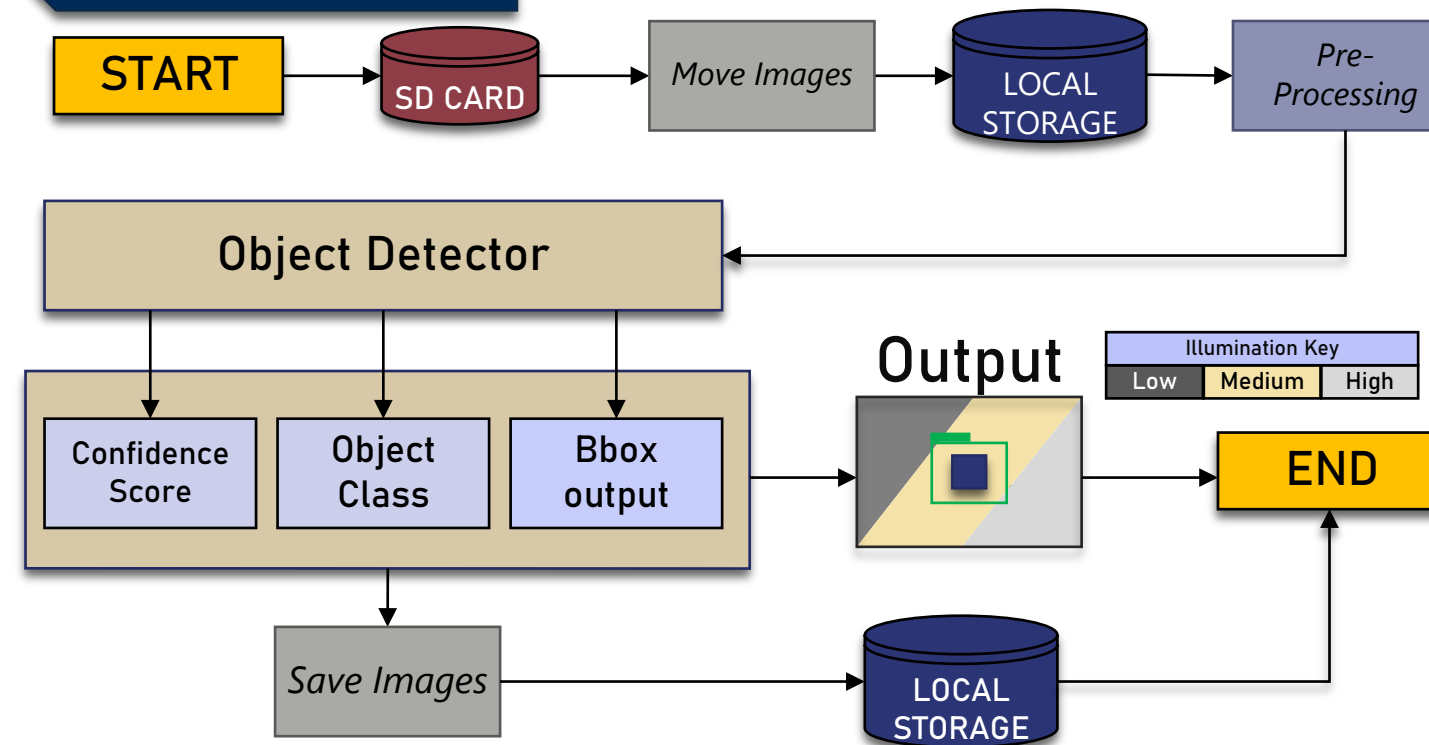
4. Conduct both experiments within the environments to obtain valid results.

5. Evaluate results by comparing gathered data and establish reasonings using evidence behind any differences or similarities.

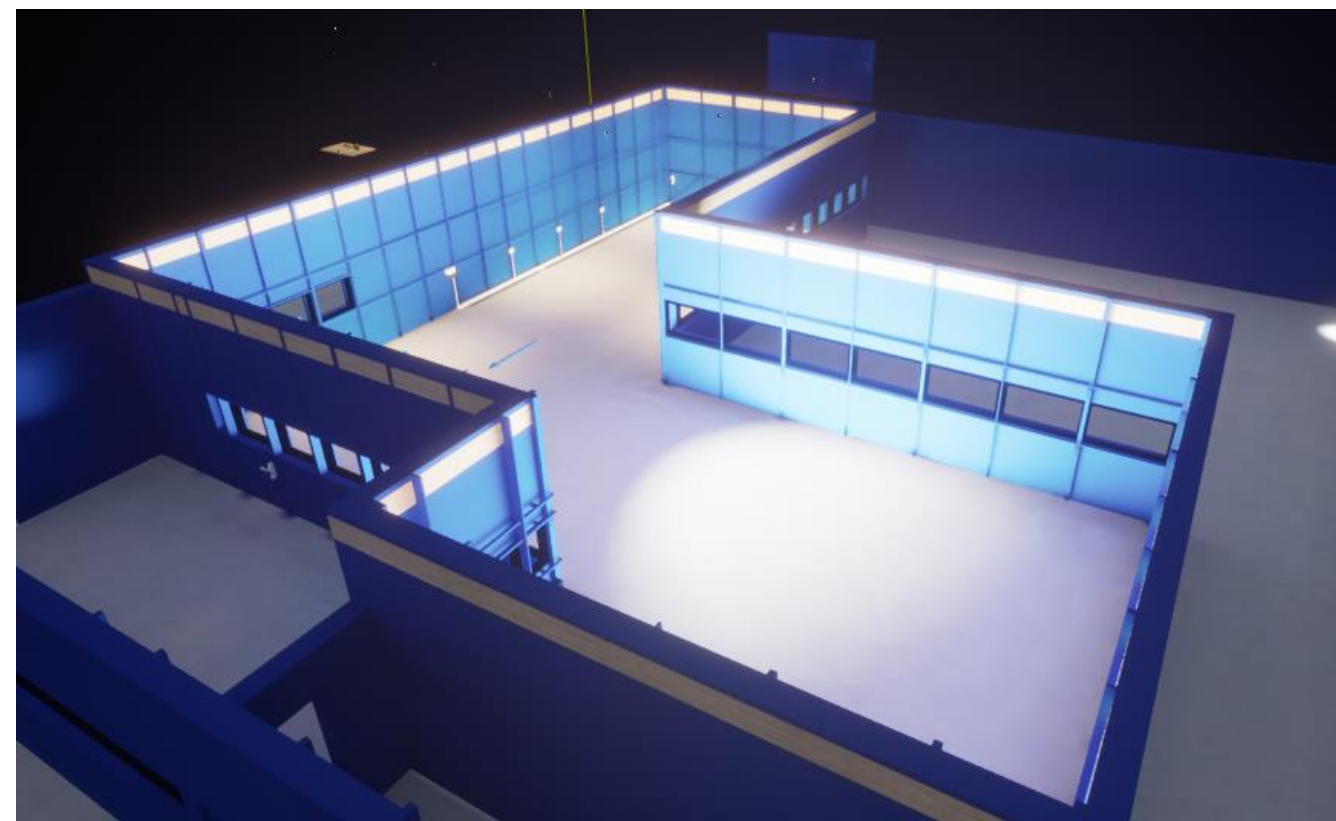
Methodology



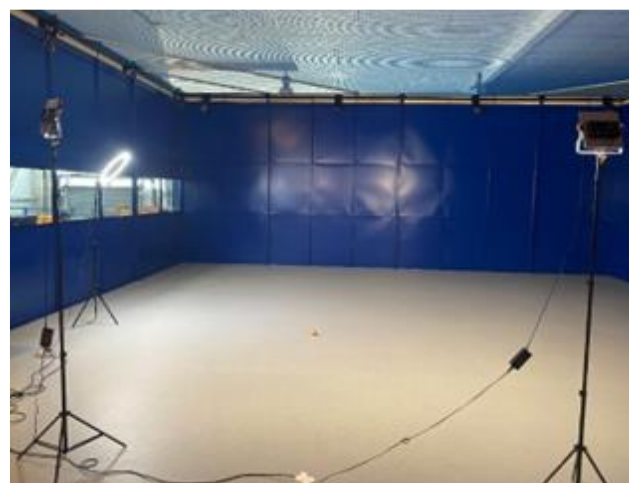
High-Level Overview



Implementation



Cranfield Flight Arena Developed within Unreal Engine For AirSim

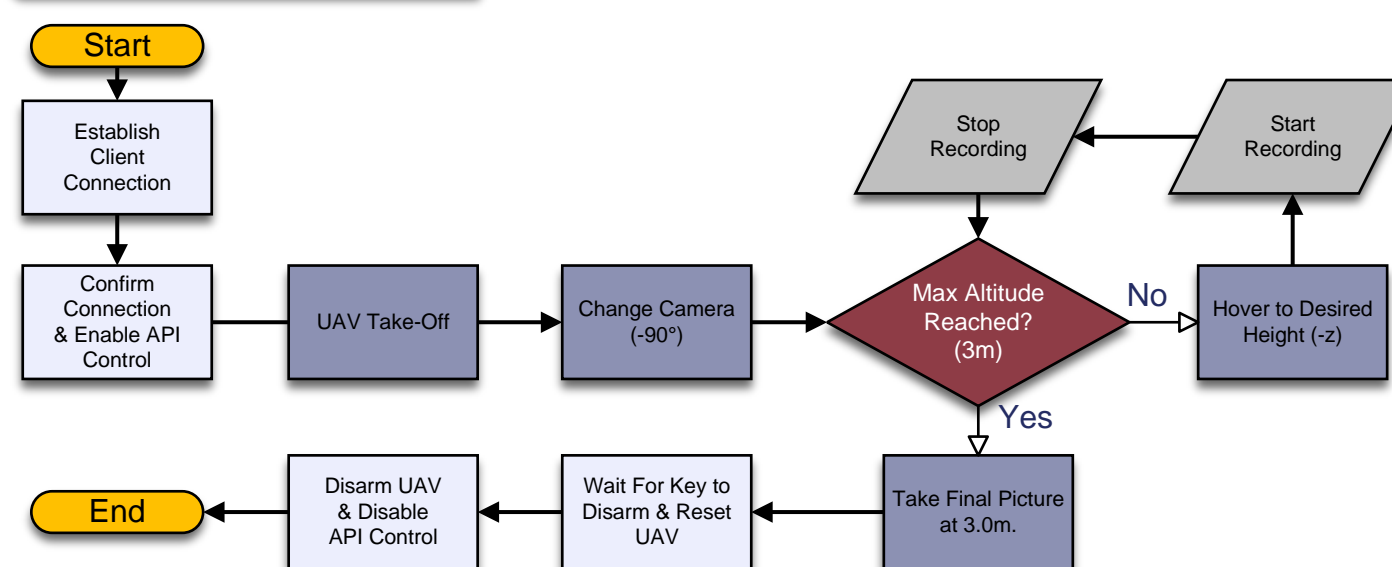


Illumination was controlled through the real world through LED Lights and Spotlights.

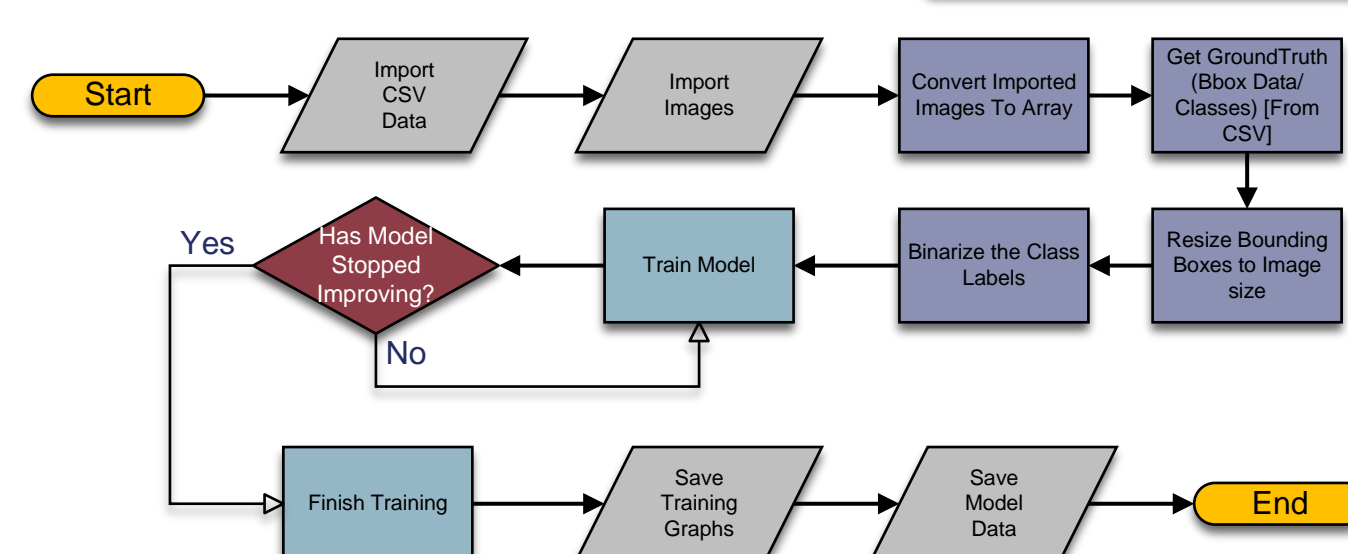
AirSim Altitude (-z(NED))	Real-World Altitude (m)
-3.05	0.5
-6.10	1.0
-8.13	1.5
-12.20	2.0
-15.25	2.5
-18.30	3.0

AirSim uses NED coordinates, which need a corresponding z value equivalence, this was found in the real-world and manipulated to work within the simulation.

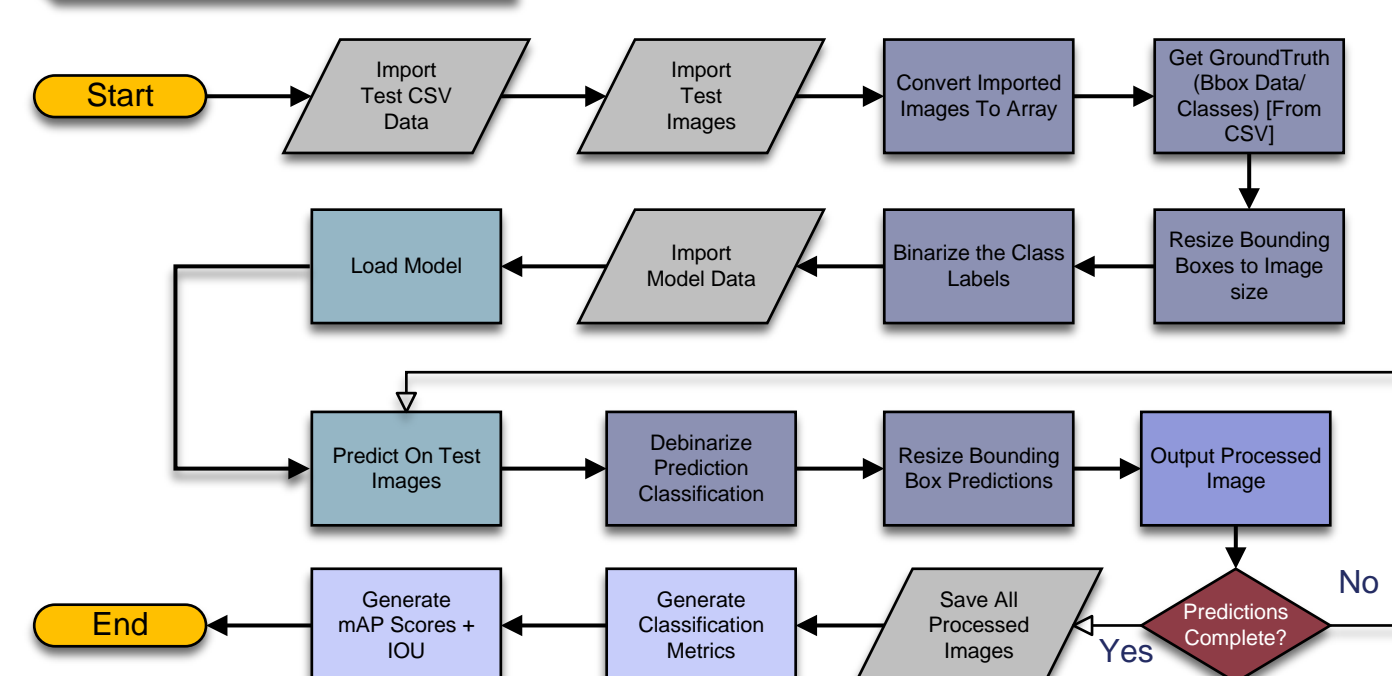
AirSim Algorithm Flowchart



Training Algorithm Flowchart

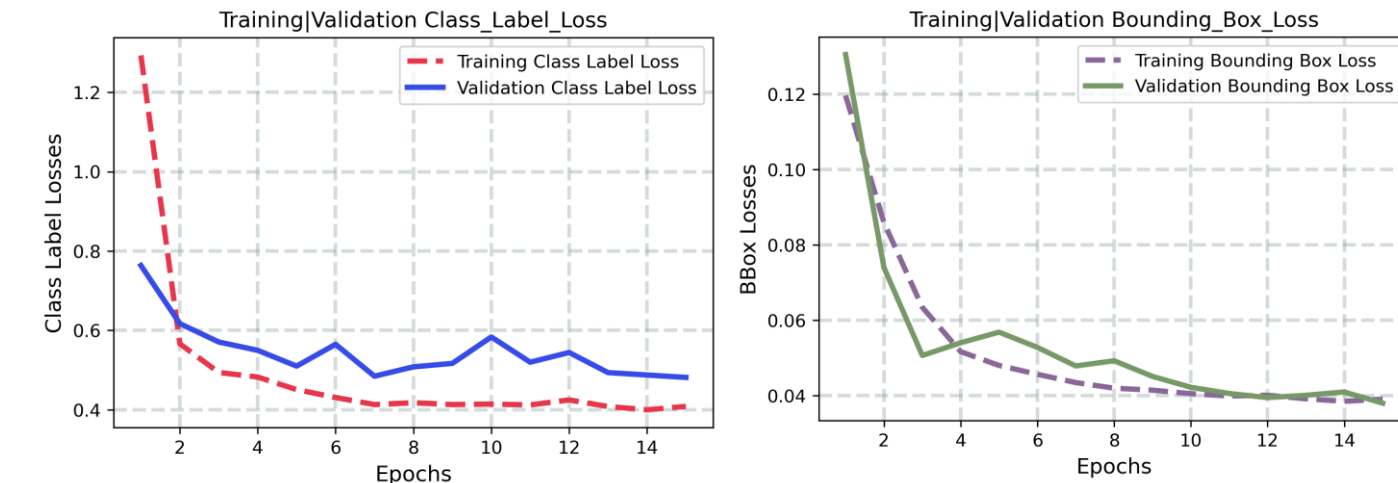


Testing Algorithm Flowchart

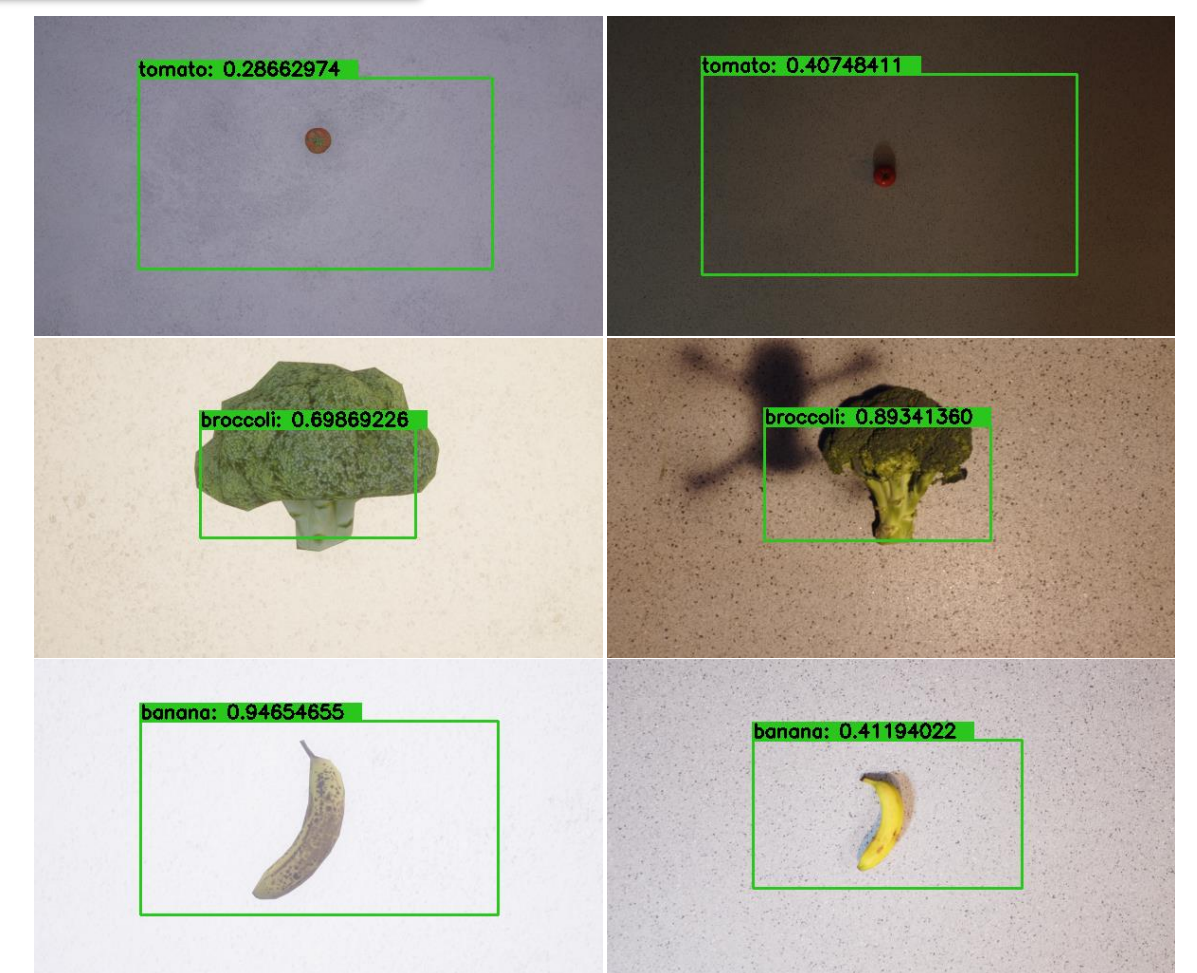


Results

Training Loss Curves



System Output



Classification - Metrics

Environment	Metrics	Object Classes					
		Apple	Banana	Broccoli	Tomato	Orange	Average
Real-World	Mean Accuracy	0.833	0.777	0.833	0.666	0.333	0.610
Simulator	Mean Accuracy	0.722	0.777	0.666	0.610	0.277	0.688

Environment	Metrics	Object Classes					
		Apple	Banana	Broccoli	Tomato	Orange	Average
Real-World	Mean F1-Score	0.833	0.777	0.775	0.660	0.333	0.595
Simulator	Mean F1-Score	0.667	0.777	0.607	0.590	0.333	0.675

Environment	Metrics	Object Classes					
		Apple	Banana	Broccoli	Tomato	Orange	Average
Real-World	Mean F1-Score	0.889	0.777	0.873	0.833	0.500	0.709
Simulator	Mean F1-Score	0.775	0.777	0.753	0.753	0.487	0.774

Regression - Metrics

Environment	Table of mean IoUs				
	IoU _{Apple}	IoU _{Banana}	IoU _{Broccoli}	IoU _{Tomato}	IoU _{Orange}
Real - World	0.399	0.424	0.472	0.398	0.396
Simulation	0.416	0.427	0.517	0.406	0.437

Environment	Table of mAP at IoU > 0.4				
	mAP _{Apple}	mAP _{Banana}	mAP _{Broccoli}	mAP _{Tomato}	mAP _{Orange}
Real - World	34.1	54.4	62.1	26.5	26.3
Simulation	24.5	40.3	69.7	22.8	39.2

Conclusion

Overall, through perseverance despite the limitations has provided an investigation into the correlation of an AI algorithm performing in a simulated and a real-world environment using a low-cost UAV to capture palpable results.

In terms of the object detection algorithm performing in both environments, the results correlate well, however this correlation is constricted through the model's ability to accurately predict the correct class and bounding box coordinates. The two manipulated values of altitude and illuminance do have an effect on the model's robustness although its very small.

The next step in refining the experiment would perform the experiments in a live scenario utilising a proposed system to access the closed architecture of the UAV through RTMP Streaming.