

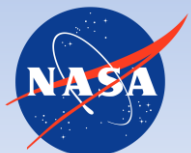
Software-In-Loop-Simulation Using VR/Game Environment

- Updates and Future Works -

Petros Voulgaris

AVIATE SEMINAR
04/21, 2023

Credits: NASA / Lillian Gipson



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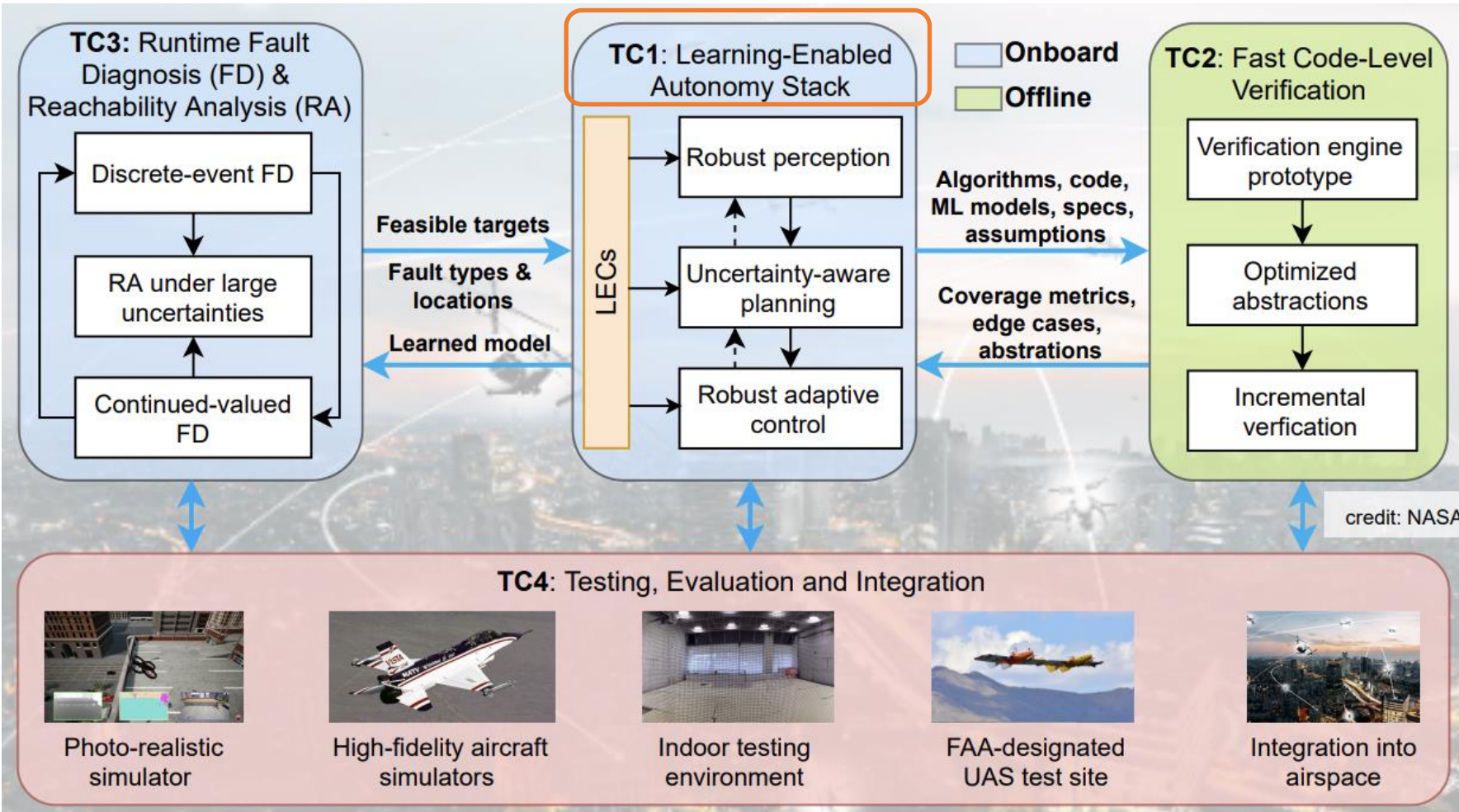
Outline

- **Introduction**
- **The Simulation Tool**
- **Updates:**
 - MATLAB Simulink Integration for NASA's vehicle simulation model.
 - 2D path planning with Lidar sensor.
- **Use Cases of the Game Environment and Future Works**
 - Adversarial Image Perturbations Against Autonomous Vehicles*
 - Robust Control in the UAV Object Detection Autonomy Pipeline
 - Future Works on the Simulator

*Yoon, Hyung-Jin, Hamidreza Jafarnejadsani, and Petros Voulgaris. "Learning When to Use Adaptive Adversarial Image Perturbations against Autonomous Vehicles." IEEE Robotics and Automation Letters (RA-L) *Accepted*.

Introduction

ENGINEERING FRAMEWORK TO BE TESTED



We need methods to implement the **Learning-Enabled Autonomy Stack (TC1)** for Testing (TC4), Verification (TC2), and Fault Diagnosis (TC3).

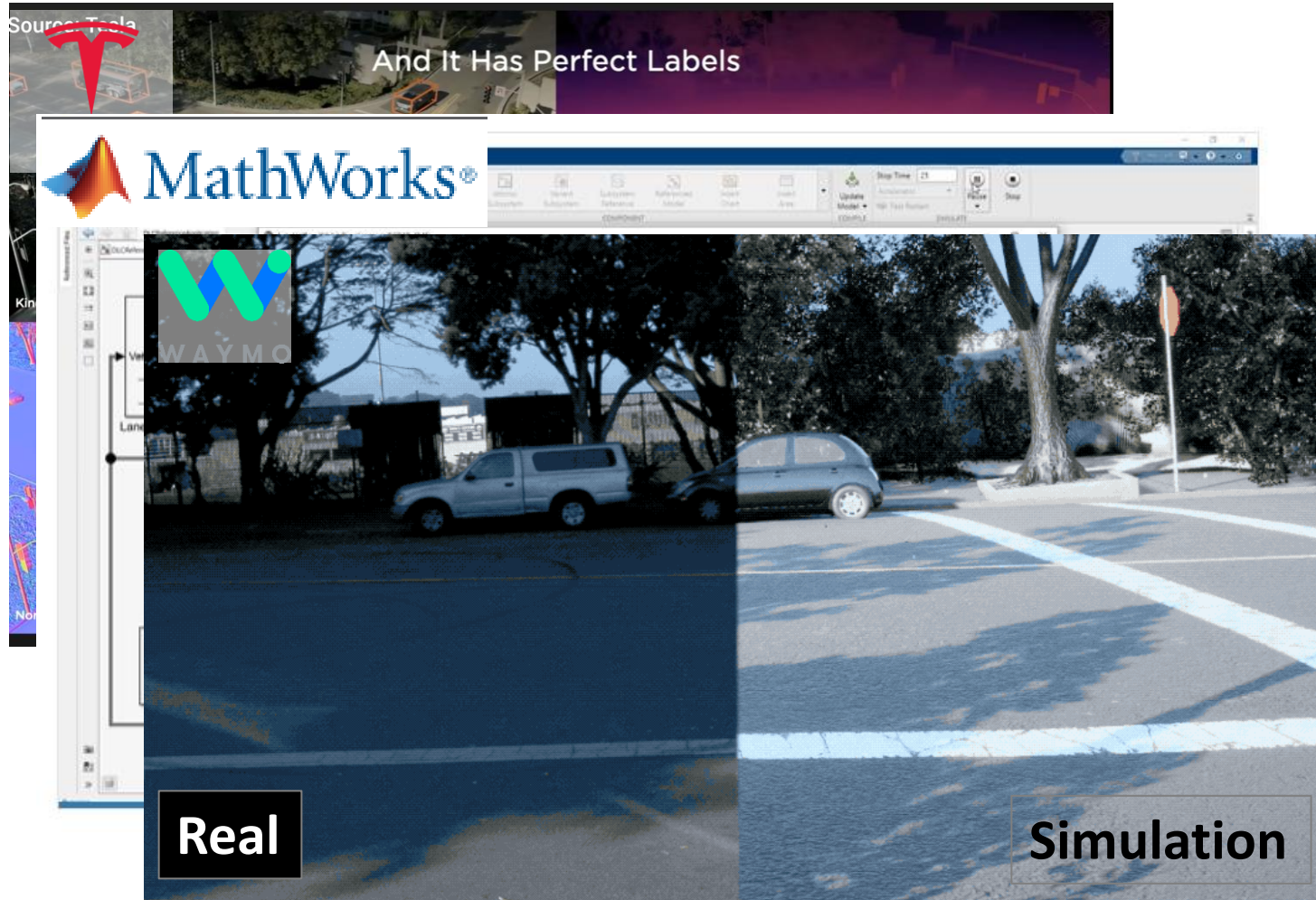
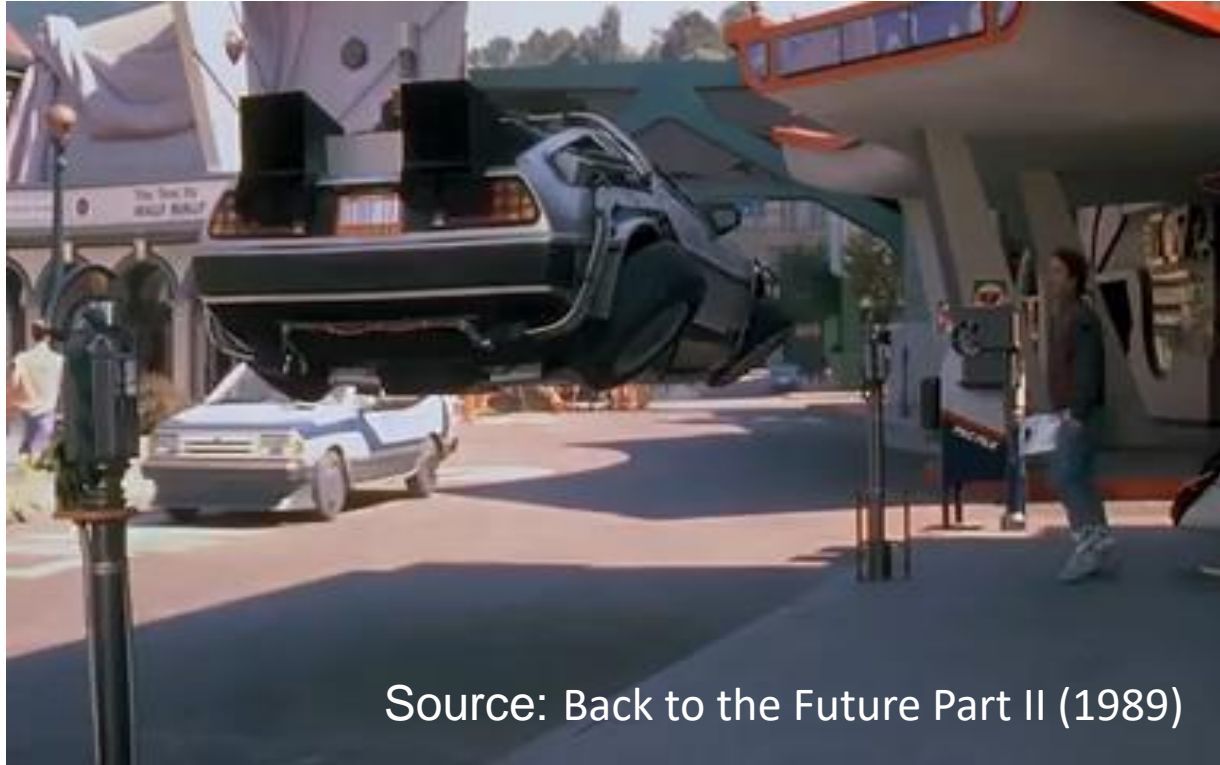


Photo Realistic Simulation using **VR/Game** engine has become an **industry standard** for developing and testing autonomous driving software stacks.

Therefore, we aim to implement the *Learning-Enabled Autonomy Stack (TC1)* with **VR/game environment**.



*Urban Air Mobility (UAM) Landing on **Street***

*Learning-Enabled Autonomy Stack (TC1) will be tested in simulation for landing and taking-off in **cluttered environments** (see the left) using perception, planning, and control methods.*

Therefore, we are using an urban environment developed by **CARLA**.



- How far **realistic** and **detailed** simulation do we need?

Figure 1.3 Deterministic vs **stochastic optimization**

	Deterministic	Stochastic
Models	System of equations	Complex functions, numerical simulations, physical systems
Objective	Minimize cost	Policy evaluation, risk measures
Searching for	Real-valued vectors	Functions (policies)
Goal	Finding optimal decision	Finding optimal policies
What is hard	Designing algorithms	Modeling

*See the **freely available** (community license?) **VR/Game project** in the next slide.*

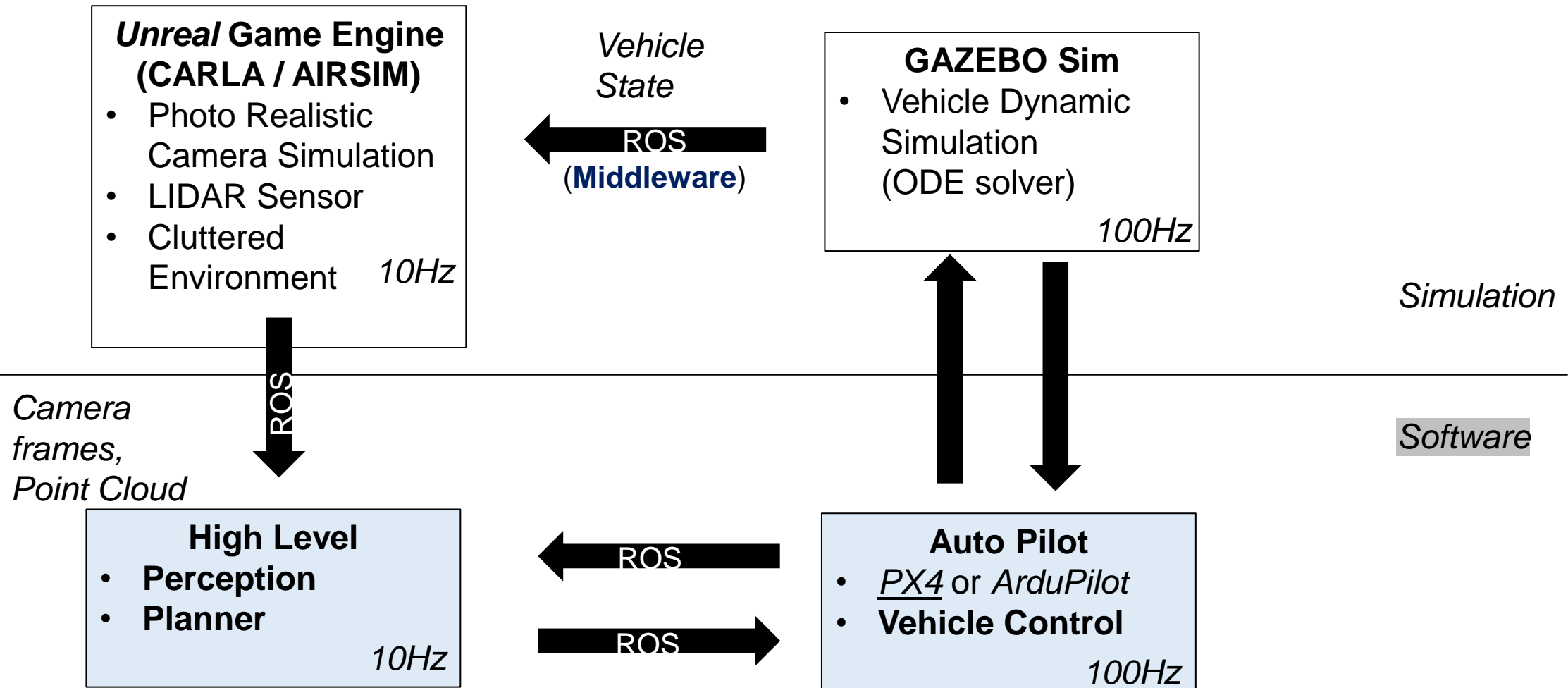
GAME/VR EDITOR

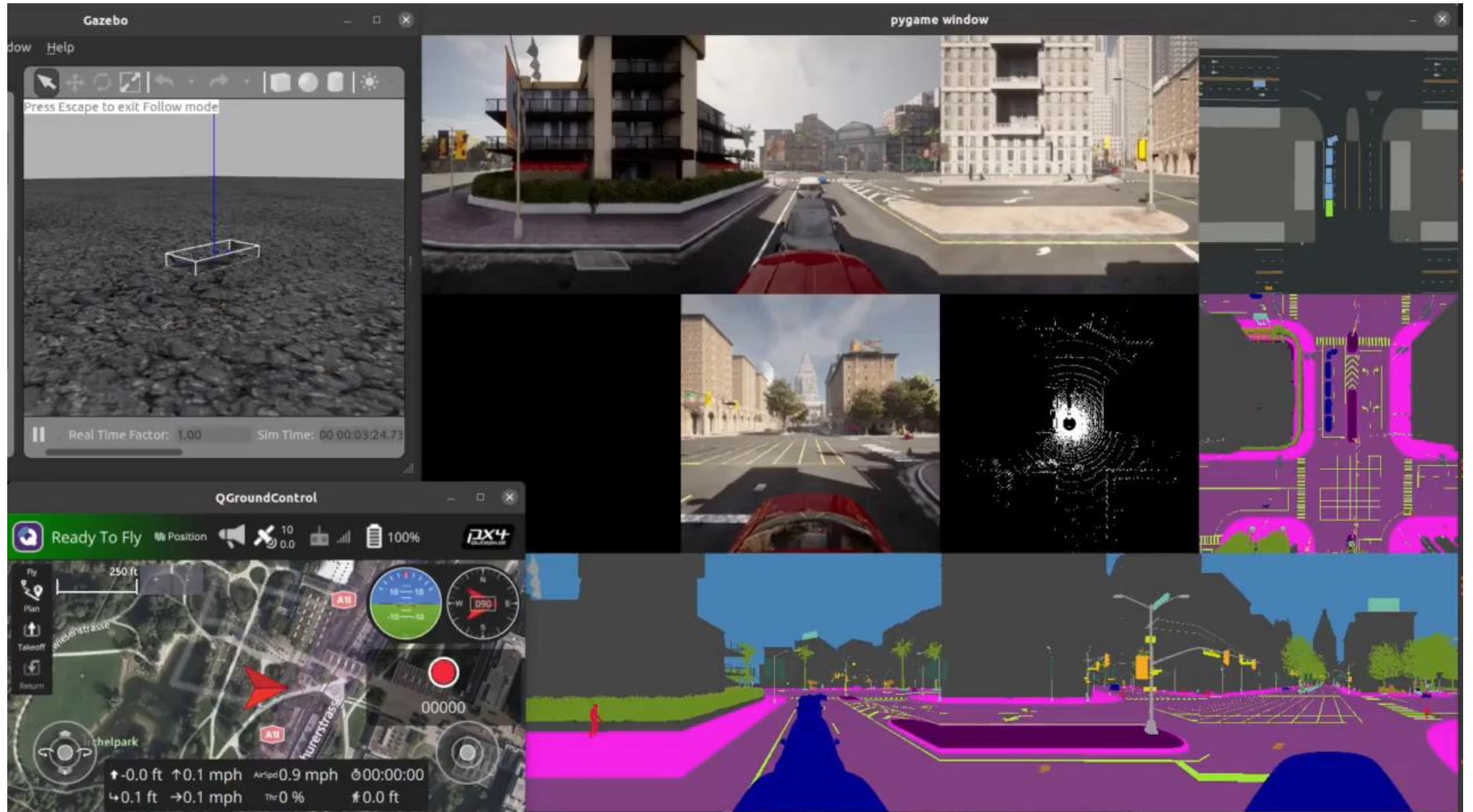


The Simulation Tool

MODULAR IMPLEMENTATION DIAGRAM

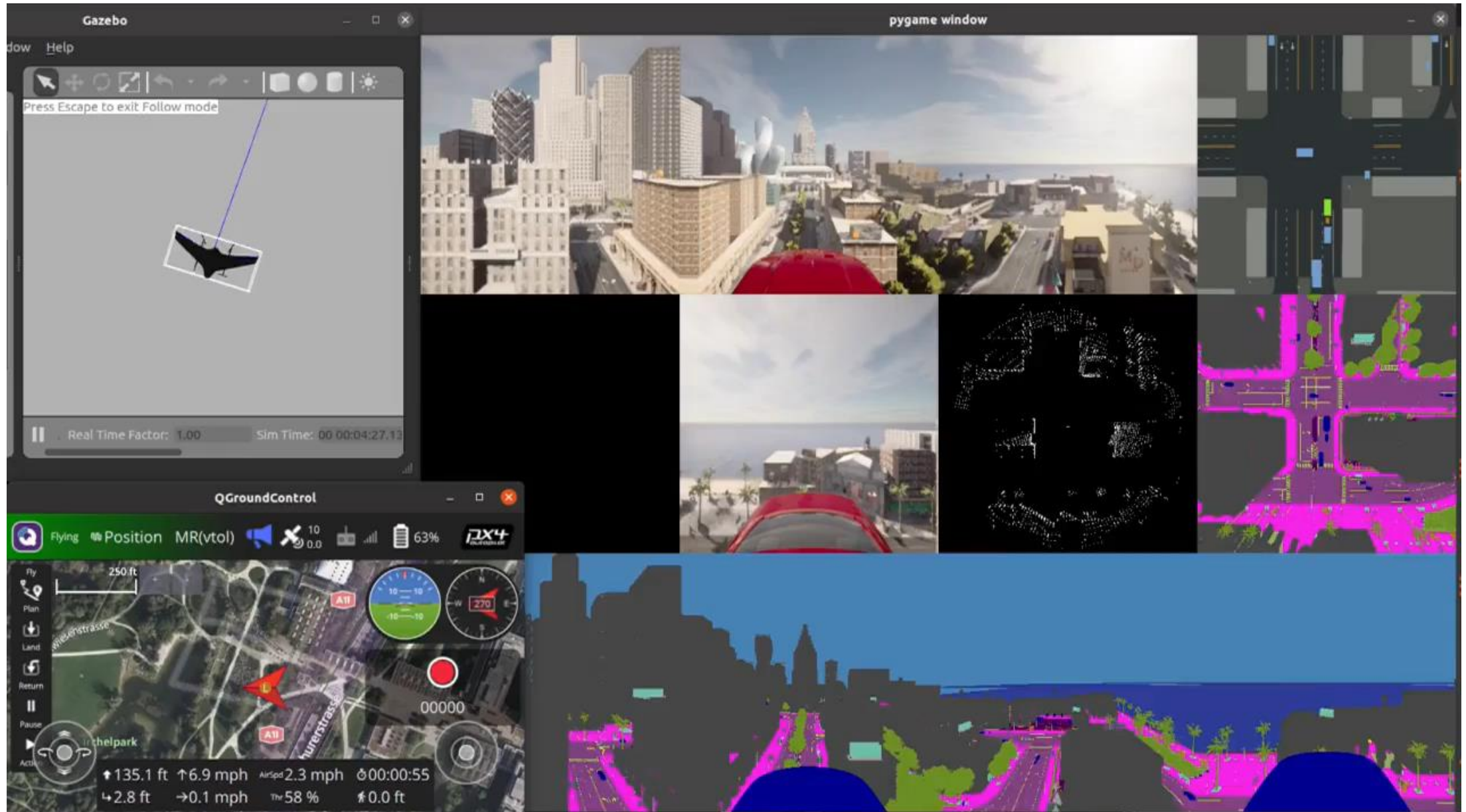
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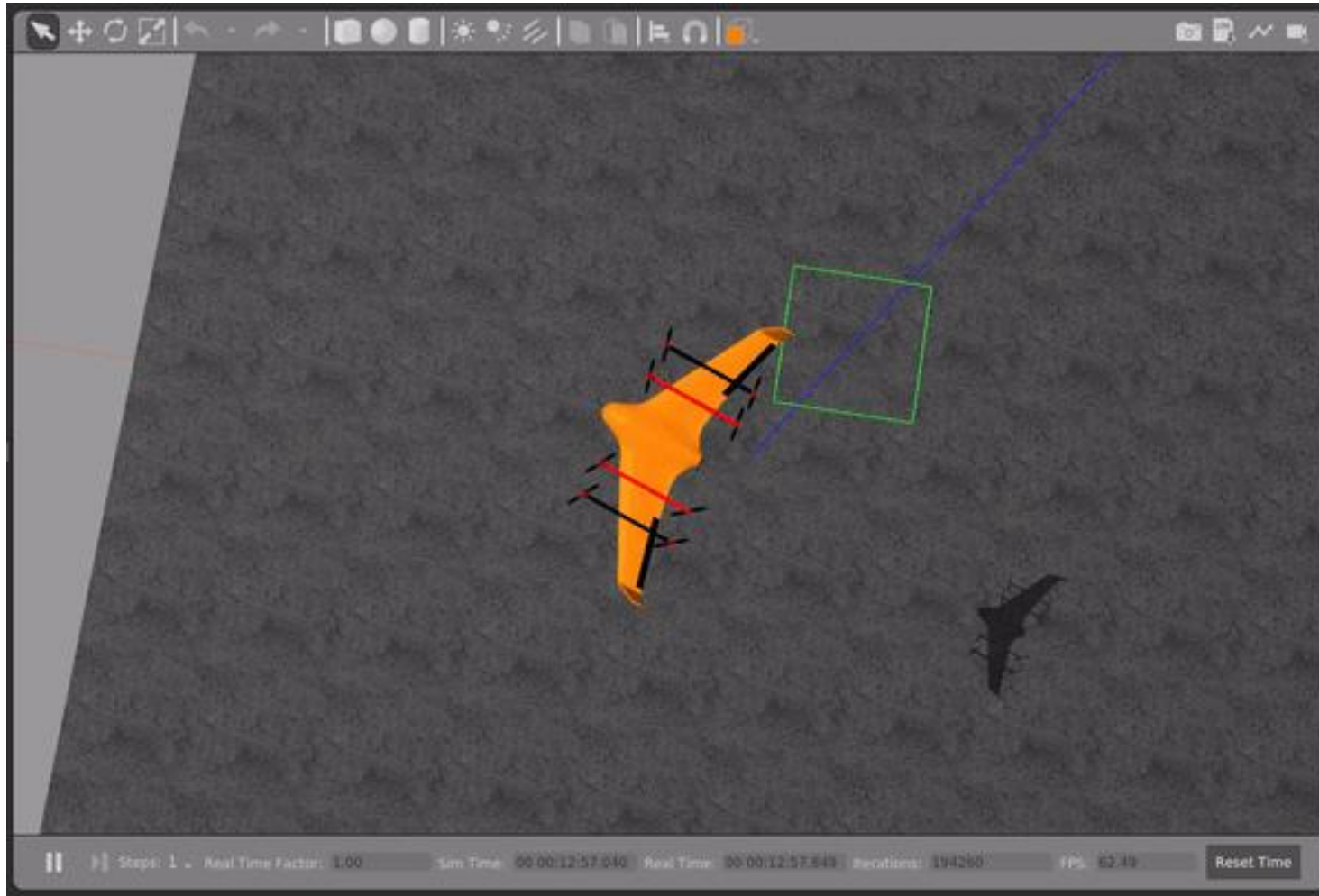




VTOL SIM – PX4 AUTOPILOT – UNREAL GAME ENGINE

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This customized vehicle has **4 more propellers** for vertical take-off.



The ***Lift+Cruise*** by **NASA** is a full-scale, VTOL, distributed propulsion aircraft concept with **eight fixed-pitch lifting rotors** and **one variable-pitch, rear-mounted pusher propeller**.



VOLY M20 Specification

Body length: 2.5m

Wingspan: 3.8m

Maximum takeoff weight: 50kg

Maximum payload: 15kg

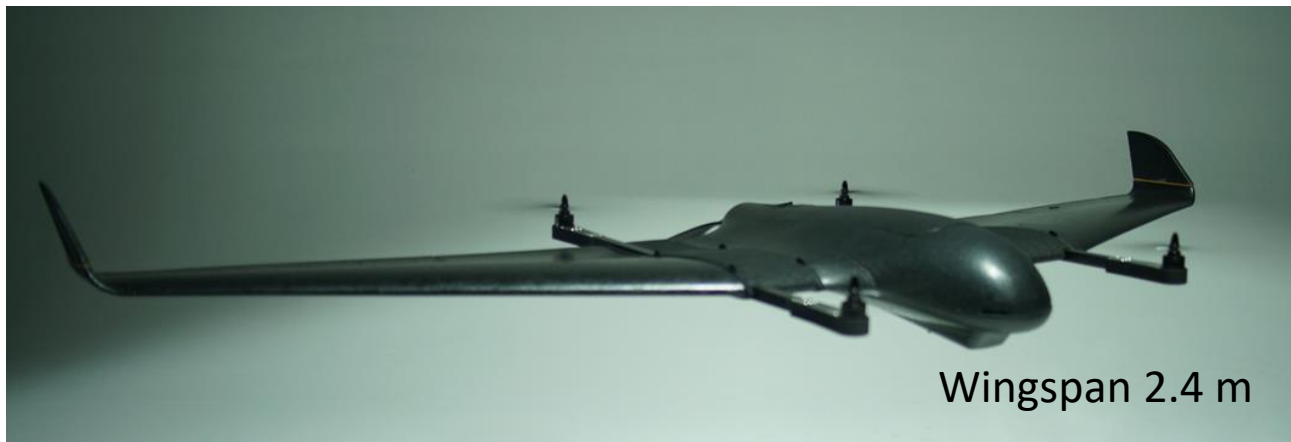
Flying radius: 200-400km

Maximum speed: 130km/h

Maximum oil load: 12L

Lift limit: 4000m

Maximum wind resistance: 12m/s (6 wind)



PX4 Autopilot S/W is used for **commercial VTOL** systems that have 2 to 4 meters wingspans.

Updates

Unreal Game Engine (CARLA / AIRSIM)

- Photo Realistic Camera Simulation
- LIDAR Sensor
- Cluttered Environment 10Hz

Vehicle State



MathWorks®
ROS in Simulink

Simulation

Software

Camera frames,
Point Cloud

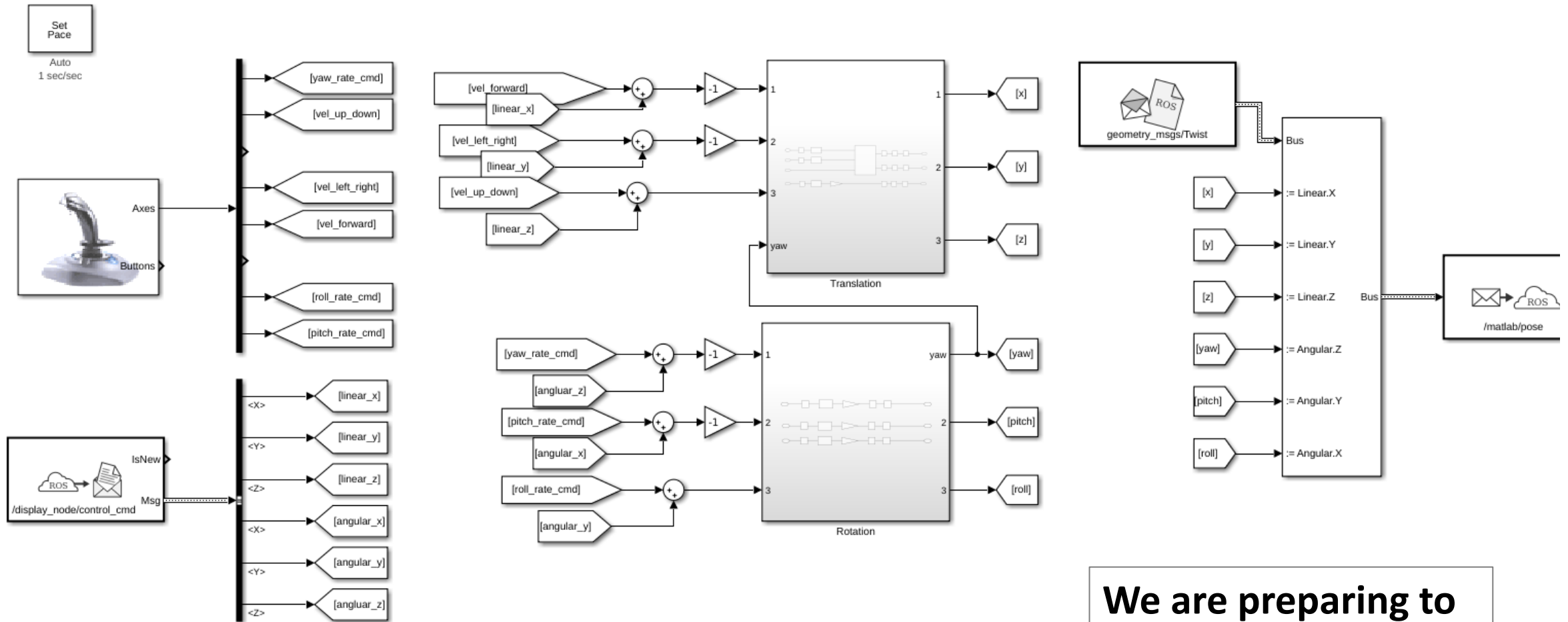


High Level

- Perception
- Planner

10Hz

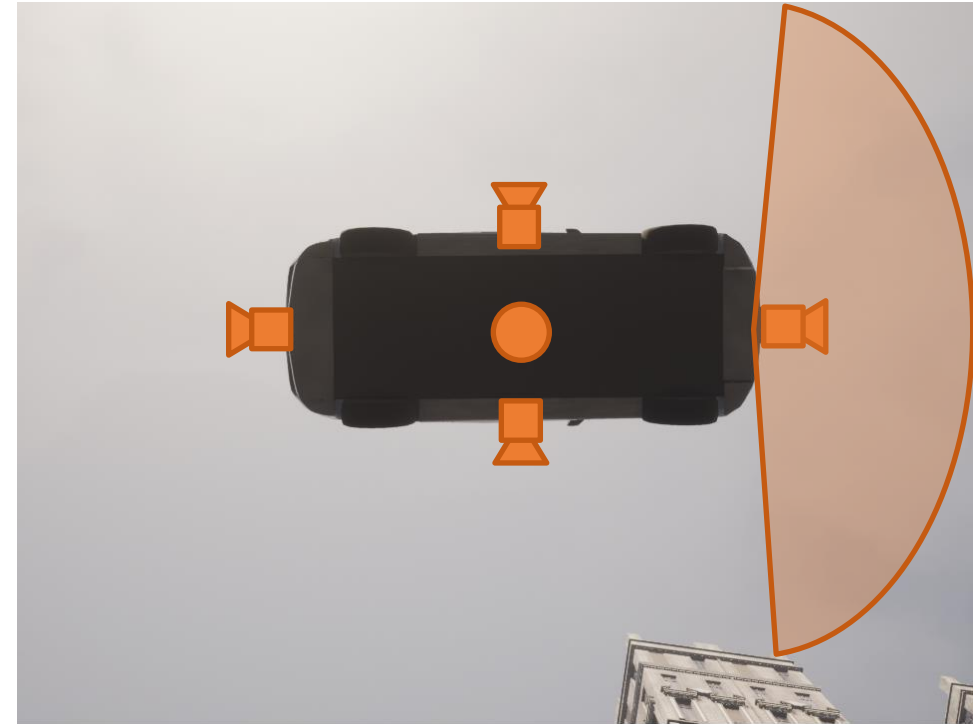
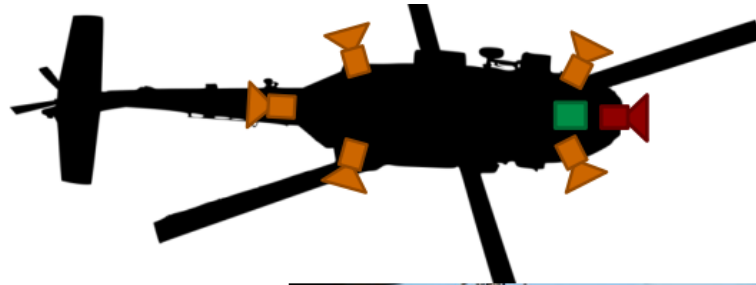




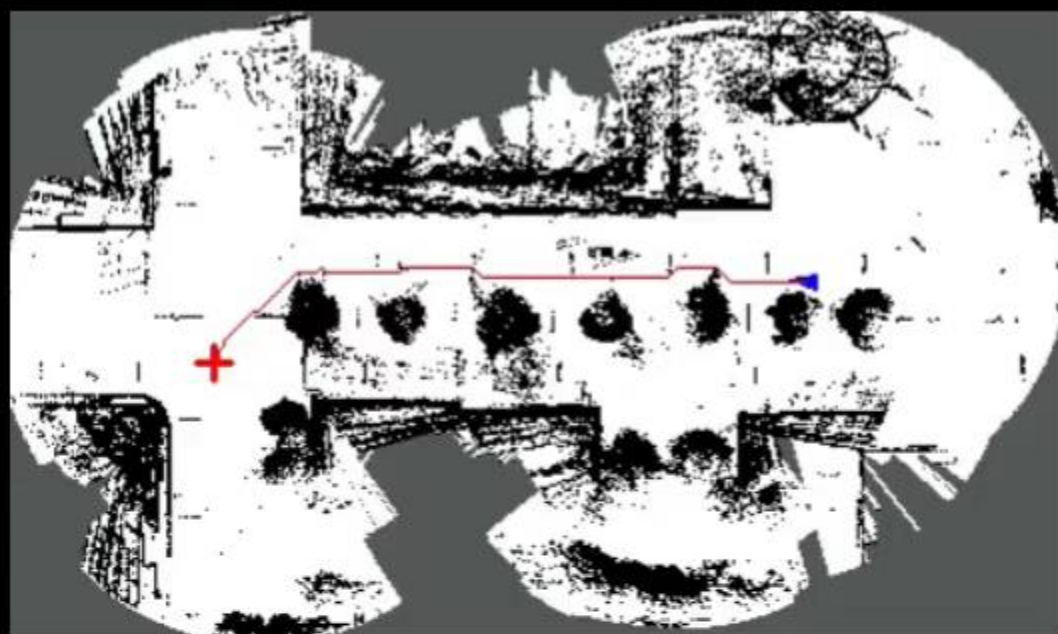
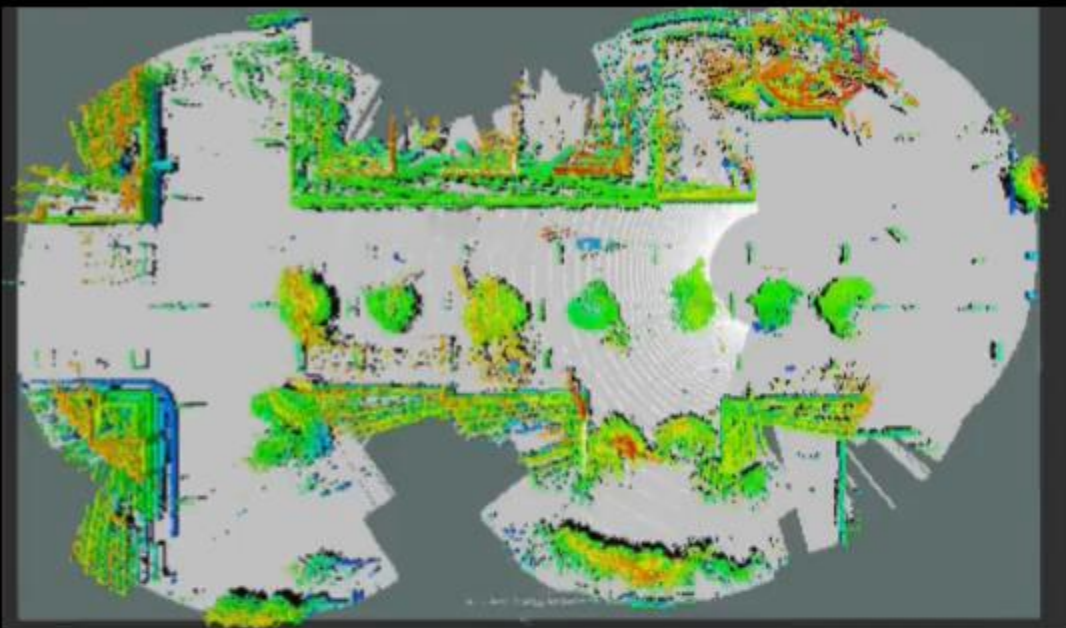
We are preparing to use **NASA's Simulink Model.**

Rough Overview of Helicopter

- Primary forward looking primary camera the pilot uses to look ahead
 - Also has forward looking lidar and radar
- Multiple situational awareness cameras for looking around
- Dual INS solution that provide position and orientation

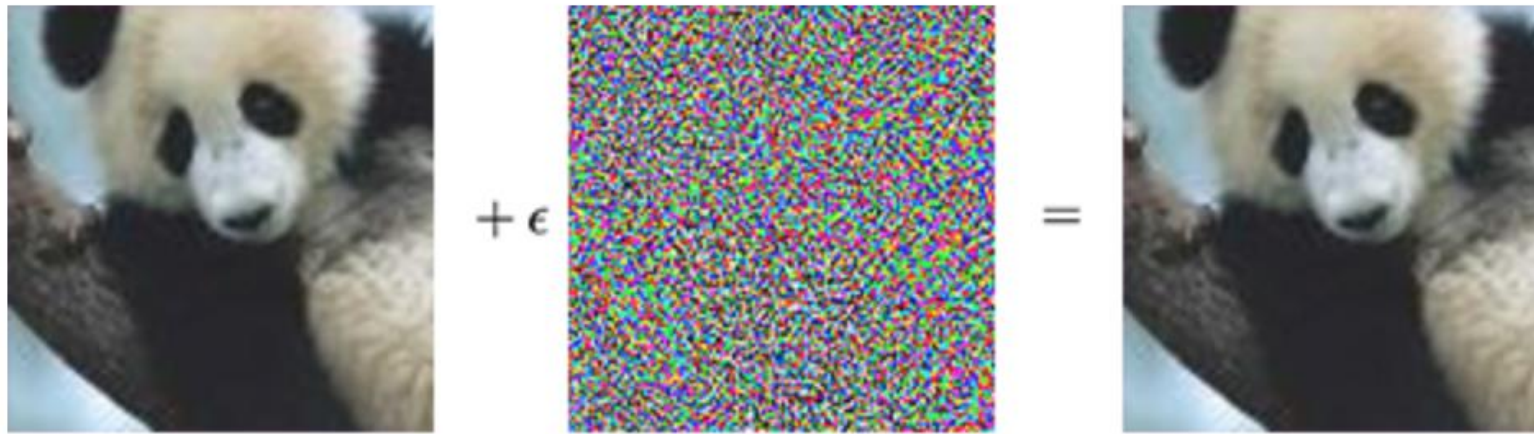


We mimicked **SNC's** sensor configuration.



Use Cases of the Game Environment and Future Works

Adversarial Machine Learning (Adv.ML)



"panda"

57.7% confidence

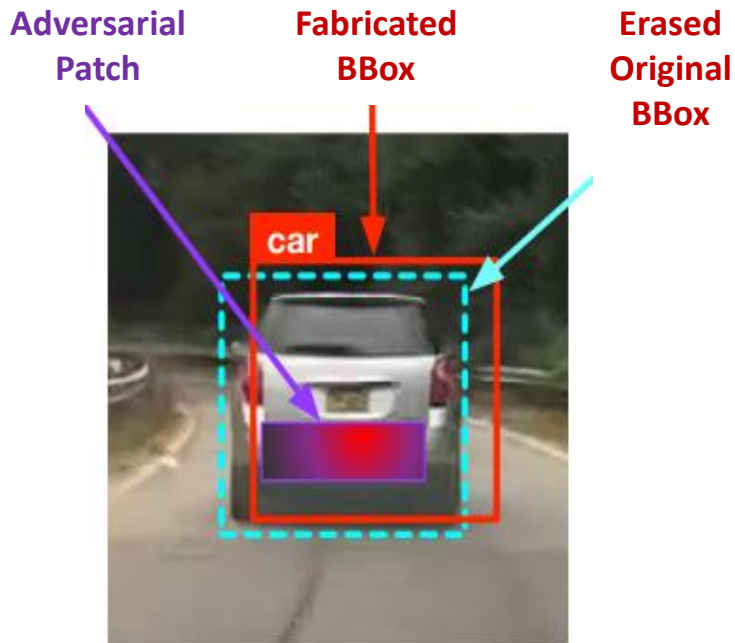
"gibbon"

99.3% confidence

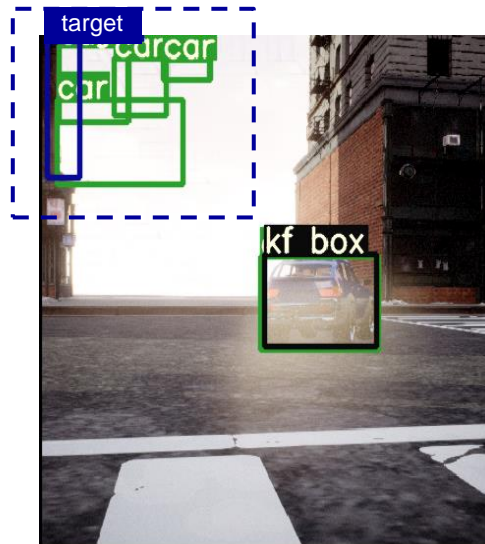
Adv. ML aims to find weakness of ML tool to **make the ML tool robust.**

*The deep learning classifier shows **vulnerability to mere noise**. The deep learning model can have a **weak point** in terms of performance, **depending on the class of the object**.*

We propose a stochastic optimization framework that monitors the attacker's capability of generating the adversarial perturbations.



Existing Method*



Use image attack

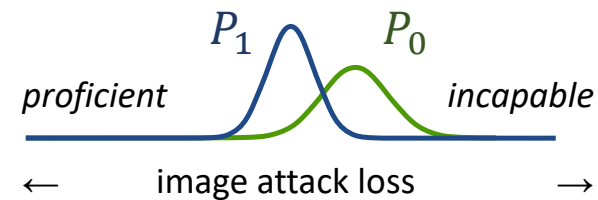
Stochastic Binary Decision

$$l_1 \sim P_1[\text{loss} | \text{state}, \text{attack}; \theta]$$

$$l_0 \sim P_0[\text{loss} | \text{state}; \theta]$$

if $l_1 < l_0$ then use attack

else do not use attack



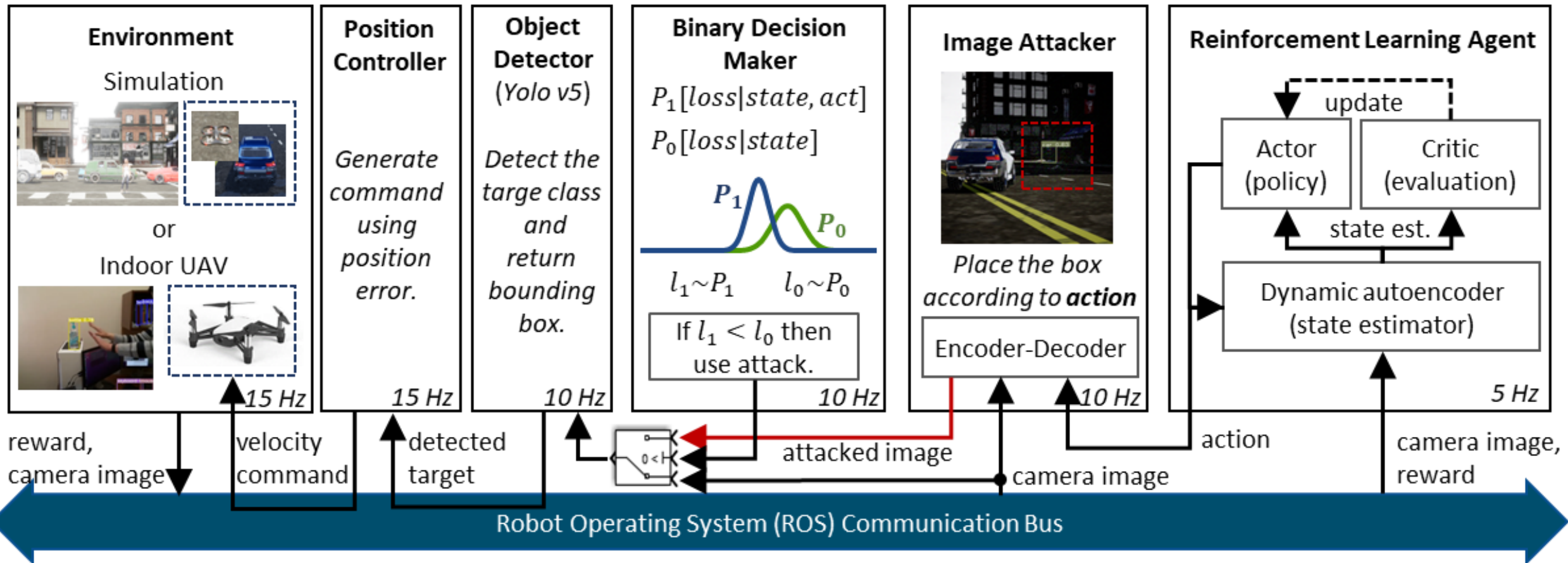
Do not use image attack

Proposed Adaptive Method

*Jia, Yunhan, et al. "Fooling Detection Alone is Not Enough: First Adversarial Attack against Multiple Object Tracking." International Conference on Learning Representations (ICLR). 2020.

Adversarial Image Perturbations Against Autonomous Vehicles

The proposed framework is implemented using ROS for online communication and simultaneous running of the following modules.

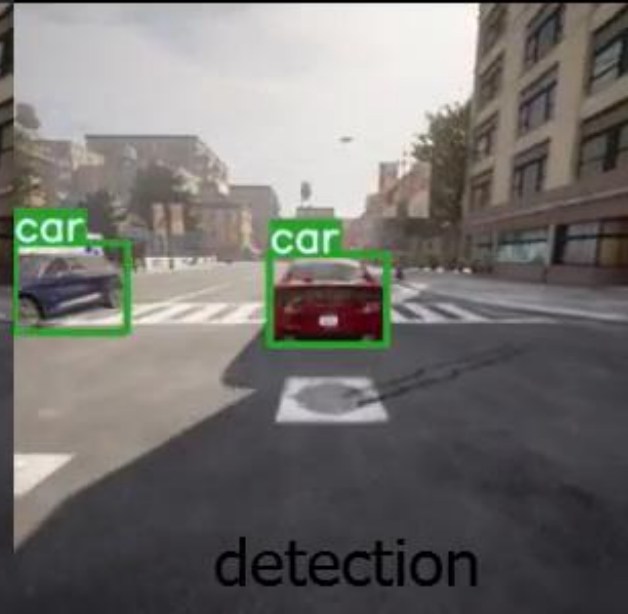
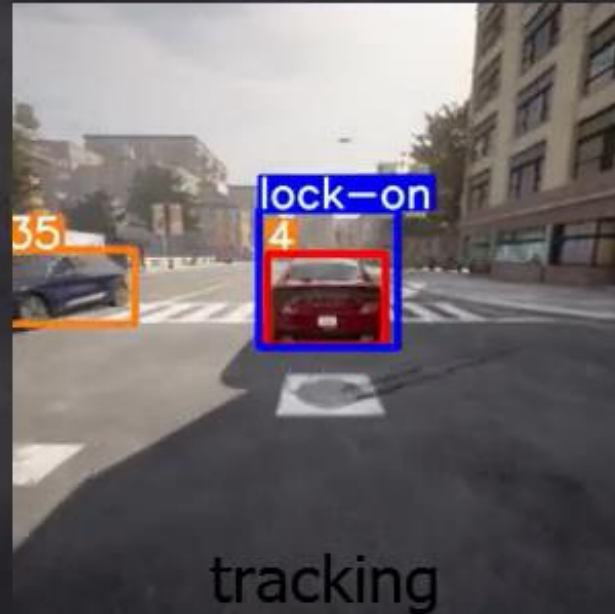
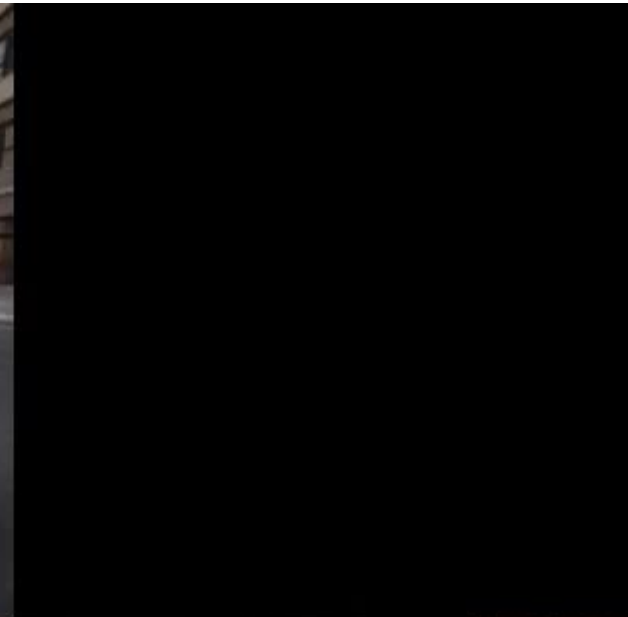
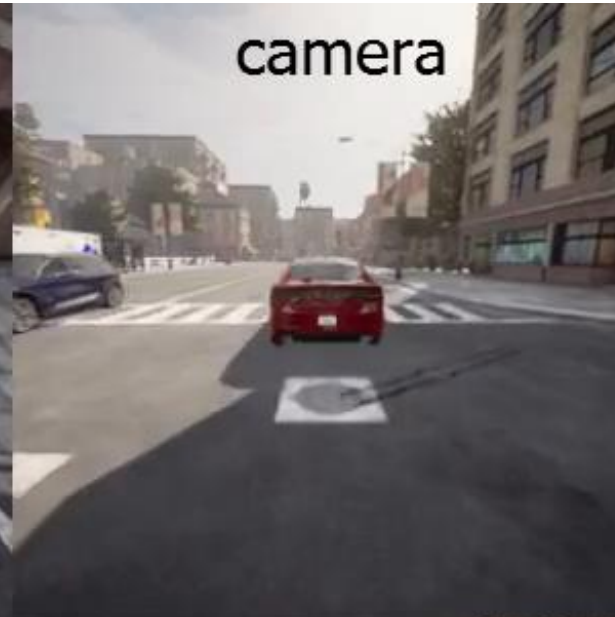
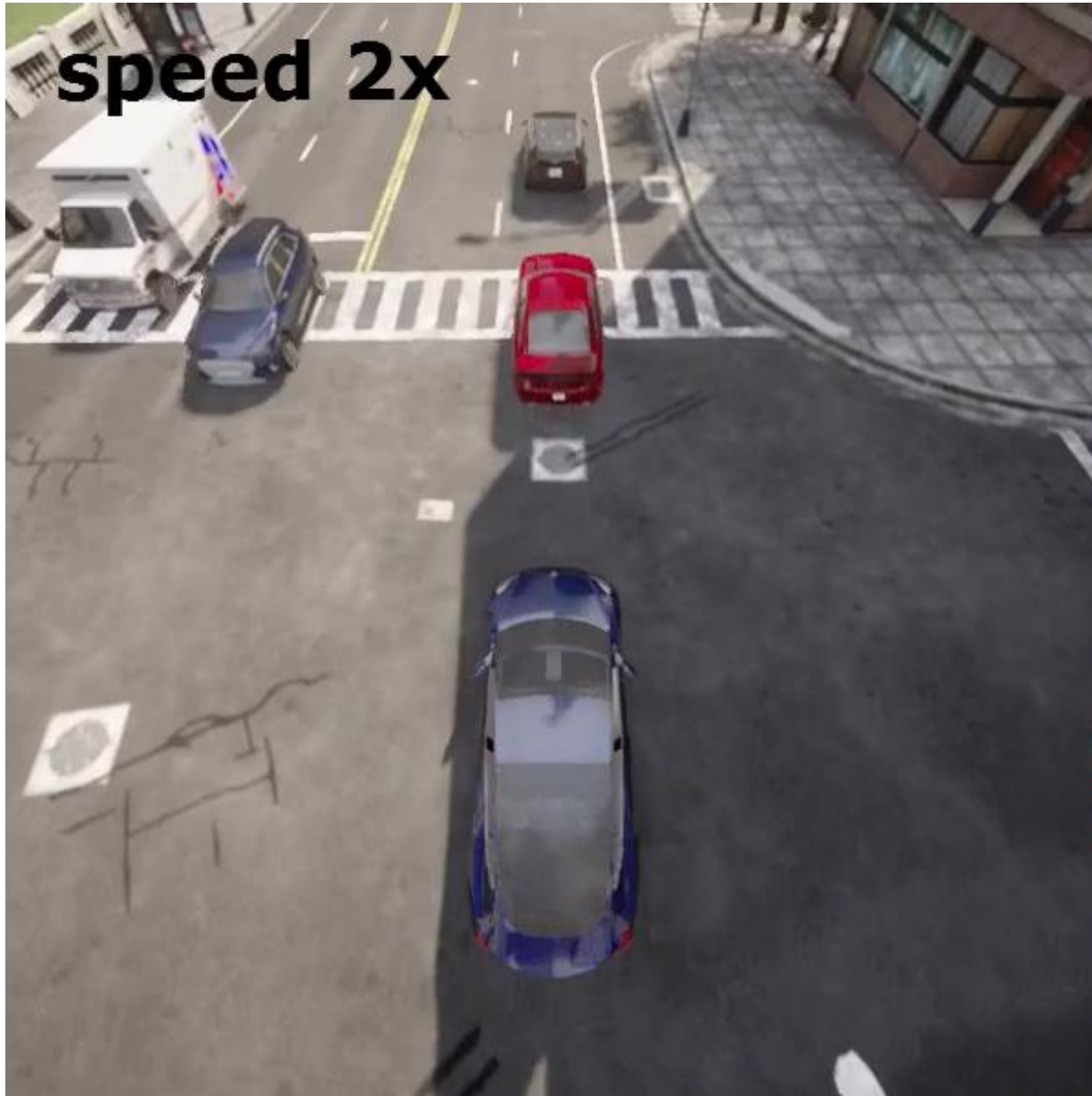


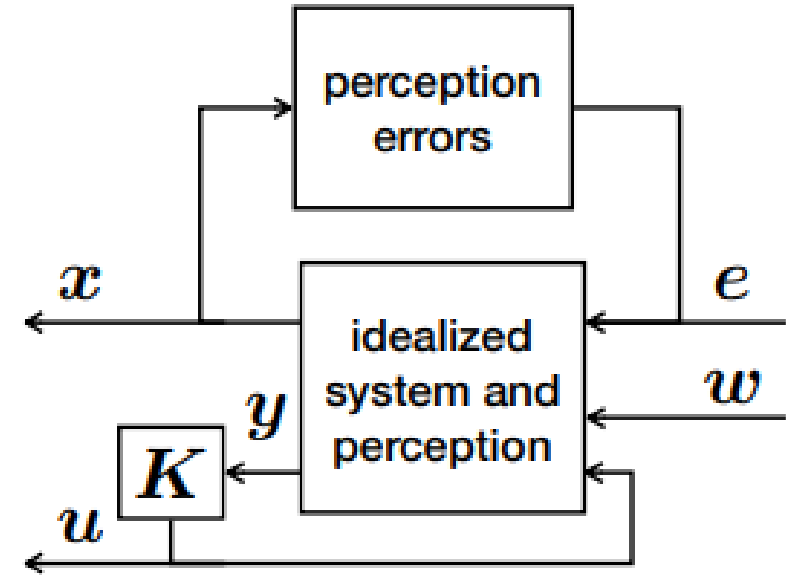
Learning When to Use Adaptive Adversarial Image Perturbations against Autonomous Vehicles

I. Real Time Image Attack with a Simulation

II. Learning Image Attack against an Indoor drone

Adversarial Image Perturbations Against Autonomous Vehicles





We quantify **robustness against the error** and the **performance** using

$$\mathcal{H}_\infty(\mathbf{G}_{xe}) \quad \text{vs.} \quad \mathcal{H}_\infty(\mathbf{G}_{xw})$$

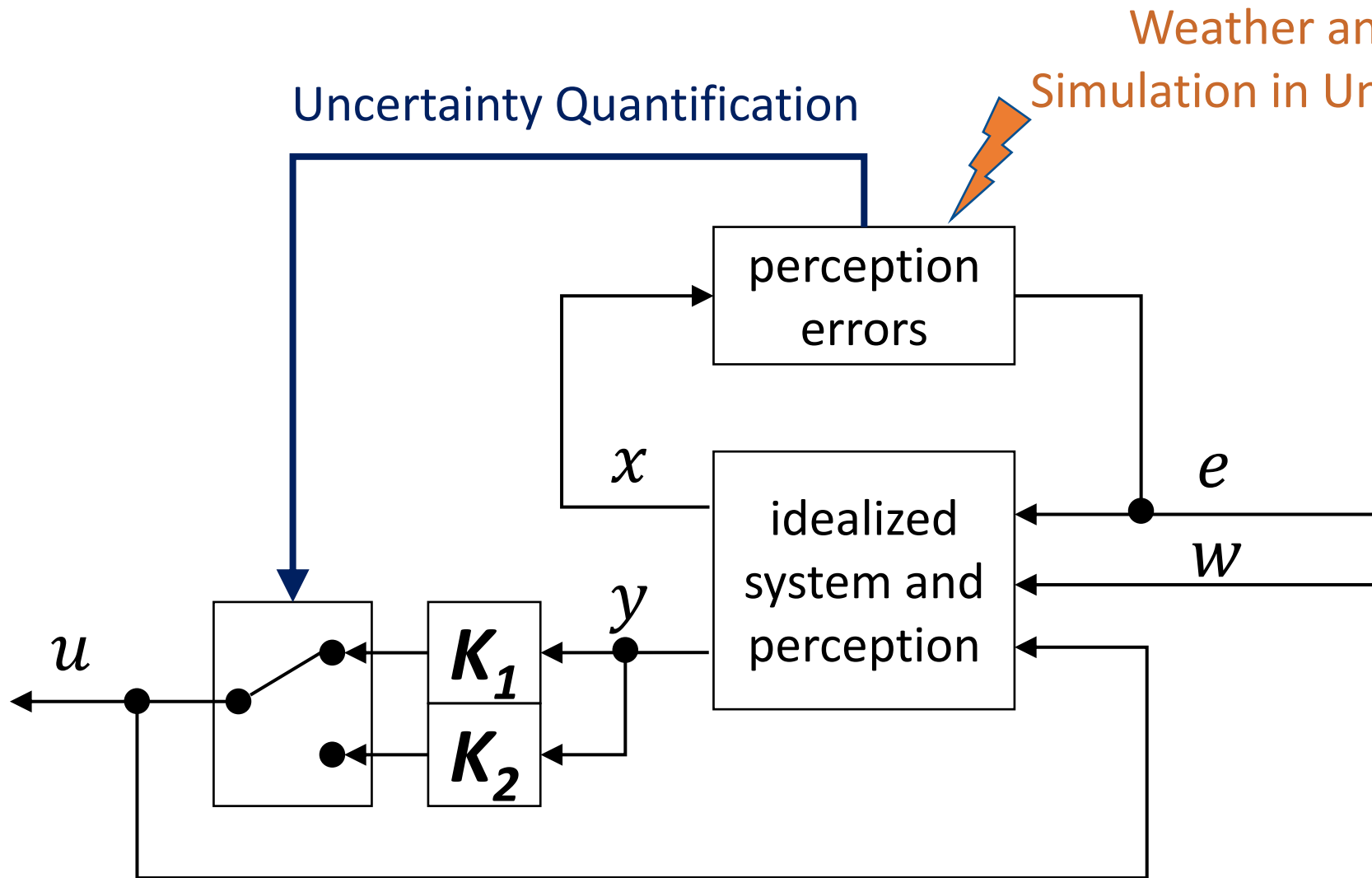
*Dean, Sarah, et al. "Robust guarantees for perception-based control." *Learning for Dynamics and Control*. PMLR, 2020.



(a) Prediction for fully visible pedestrian. (b) Prediction for heavily occluded pedestrian.

Fig. 4 – Comparing aleatoric uncertainties for **occluded** and fully visible pedestrians. Both Predictions were produced by prior 3. Note the drastically increased variance in width and height for the heavily occluded pedestrian.

There exist works on the **estimation of uncertainty in object detection.**



We aim to implement the **common sense** that we need to be cautious when the environment is uncertain using the switching logic.

- We need to further organize and clean the GitHub repository at <https://github.com/stargaze221/RRAAA-ULI>
- Replace the flying car with a better-looking VTOL in the Unreal Engine such as



www.turbosquid.com

- Devise a sensor fusion algorithm to integrate the Lidar sensor with the occupancy grid, the camera sensor with the object detection.
- Devise an estimation method to predict and track the moving obstacle in the environment.

Learning Autonomy and Control Systems Lab

Prof. Petros Voulgaris

MECHANICAL ENGINEERING



Petros Voulgaris



Hyung-Jin Yoon
(Postdoc)

- Decision making under uncertainties
- Autonomous System



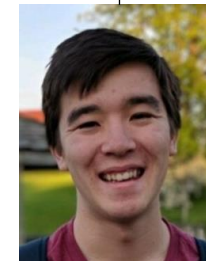
Yichuan Li
(Postdoc)

- Multiagent optimization for control and estimation



Antonio Castaño
(Grad. Student)

- UAV Control
- VR simulation



Caleb Patton
(Grad. Student)

- Machine learning
- VR simulation



Jessica Peterson
(Upcoming Grad.)

- UAV - Human Pilot collaboration
- UAV test and verification

Learning Autonomy and Control Systems Lab

Prof. Petros Voulgaris

MECHANICAL ENGINEERING



Thank you



University of Nevada, Reno

Systems Integration & Field Deployment Verification

AVIATE Seminars

Christos Papachristos
Assistant Professor
Department of Computer Science and Engineering
University of Nevada, Reno

April 20, 2023

Credits: NASA / Lillian Gipson



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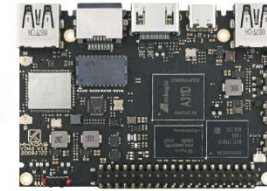


Small-scale Aerial System w/ Multi-Day Field Deployment Autonomy

Khadas VIM3: 4x 2.2Ghz Cortex-A73, 2x 1.8Ghz Cortex-A53, 5.0 TOPS NPU



ROS arm



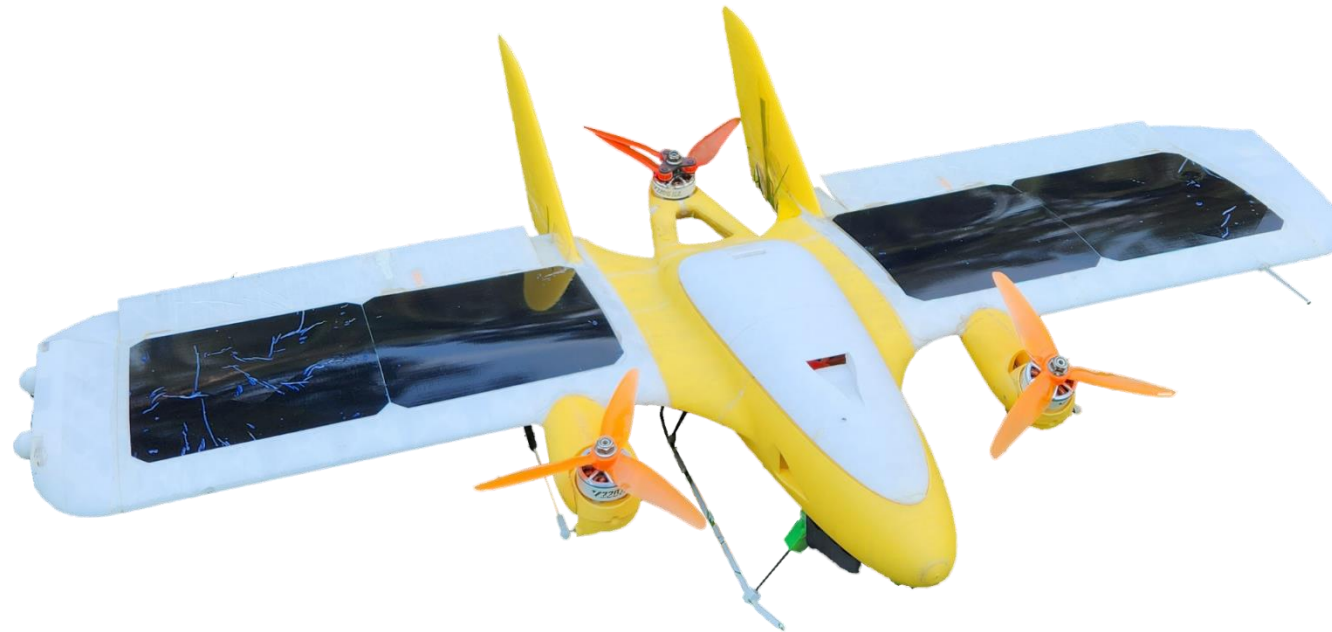
u-Blox Neo-M9N



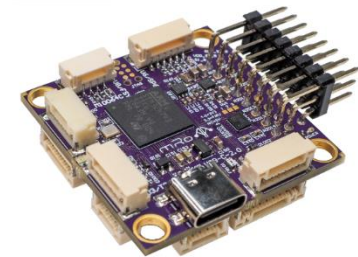
Intel Realsense T265



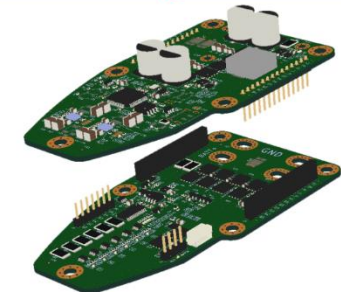
FLIR Chameleon 3



mRo PixRacer Pro



Power Management Board



Towards Multi-Day Field Deployment Autonomy: A Long-Term Self-Sustainable Micro Aerial Robot

Stephen J. Carlson, Prateek Arora, Tolga Karakurt,
Brandon Moore, and Christos Papachristos



This material is based upon work supported by the
NASA Award: "ULI: Robust and Resilient Autonomy for
Advanced Air Mobility" and the NSF Award: 2008904:
RI: Small: "Learning Resilient Autonomous Flight"


Robotic Workers Lab




University of Nevada,
Reno





 This material is based upon work supported by the NSF Award: 2008904
RI: Small: Learning Resilient Autonomous Flight.

 **Robotic Workers Lab**  University of Nevada,
Reno 

 This material is based upon work supported by the NSF Awards:
2148788: EPSCoR RII Track-1: "Harnessing the Data Revolution
for Fire Science", and IIS-2150394: REU Site: "Collaborative
Human-Robot Interaction for Robots in the Field"

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Reno 

Mid-scale Aerial System Development w/ Amphibious Capabilities

Firefly ROC-RK3588S-PC: 4x 2.4Ghz Cortex-A76, 4x 1.8Ghz Cortex-A55, 6.0 TOPS NPU



u-Blox Neo-M9N



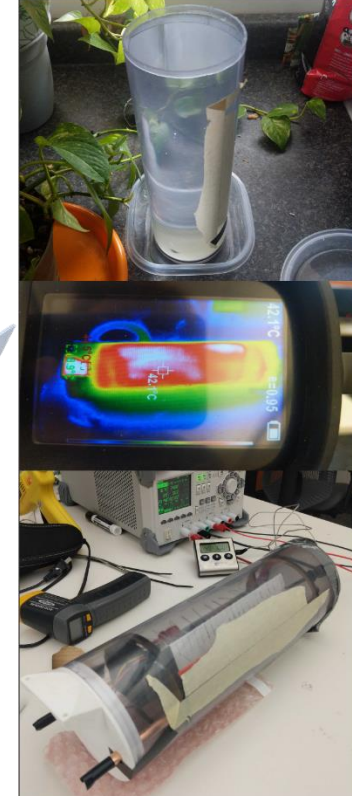
Intel Realsense T265



LiDAR / Vision



Watertight Compartment



mRo PixRacer Pro

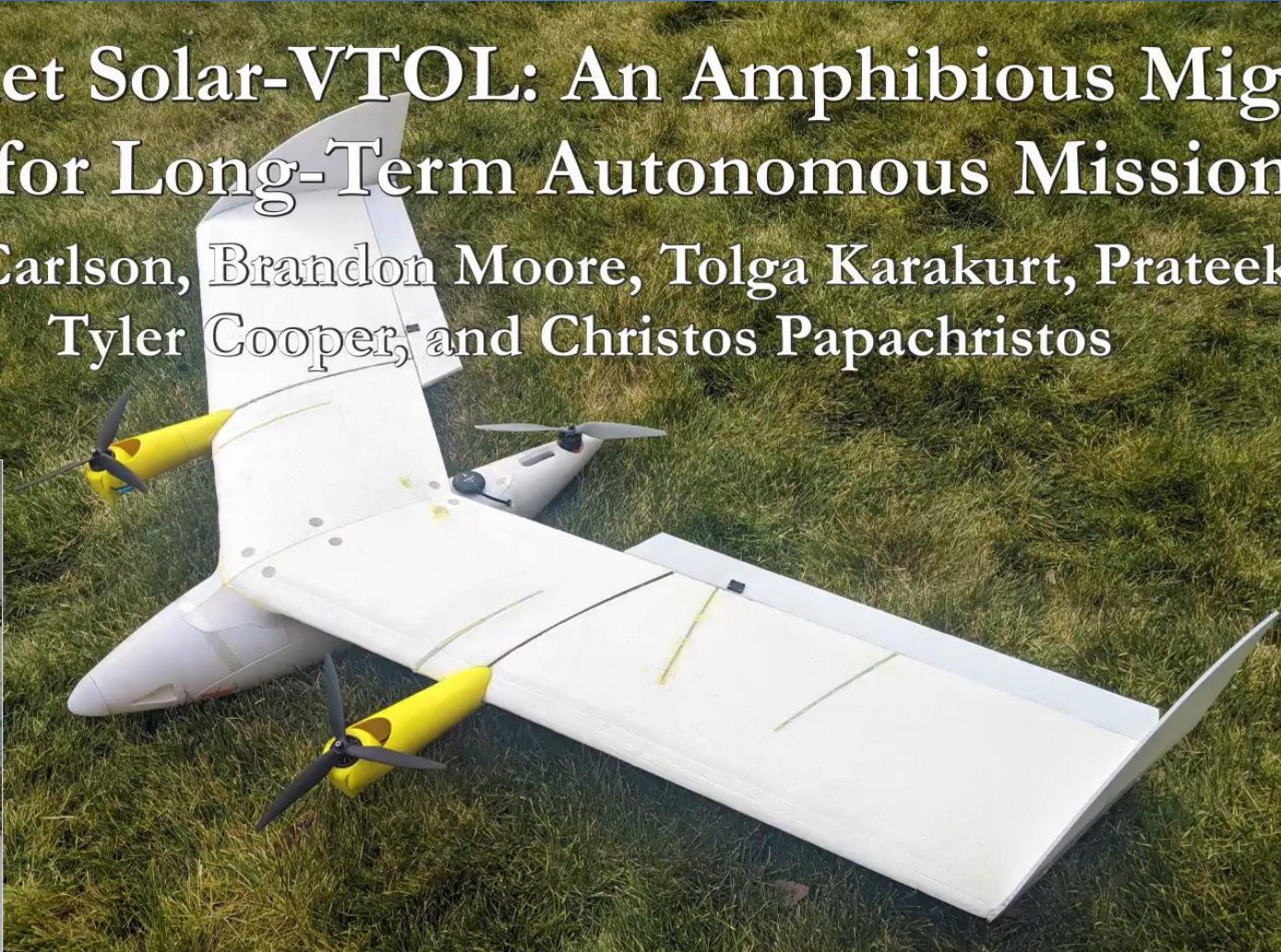


Power Management Board



The Gannet Solar-VTOL: An Amphibious Migratory UAV for Long-Term Autonomous Missions

Stephen J. Carlson, Brandon Moore, Tolga Karakurt, Prateek Arora, Tyler Cooper, and Christos Papachristos



This material is based upon work supported by the NASA Award: "ULI: Robust and Resilient Autonomy for Advanced Air Mobility" and the NSF Award: 2008904: RI: Small: "Learning Resilient Autonomous Flight"

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University of Nevada,
Reno



Heavy-Payload commercial Multicopter

Intel NUC 11 PAH i7: 4x 2.8Ghz i7-1165G7



DJI GPS



Intel Realsense T265



Ouster OS0-128



DJI Autopilot



Initial Field Testing with Heavy-Payload commercial Multicopter



Educational Content & Digital Twin Simulations

<https://www.roboticworkerslab.com/education/courses>



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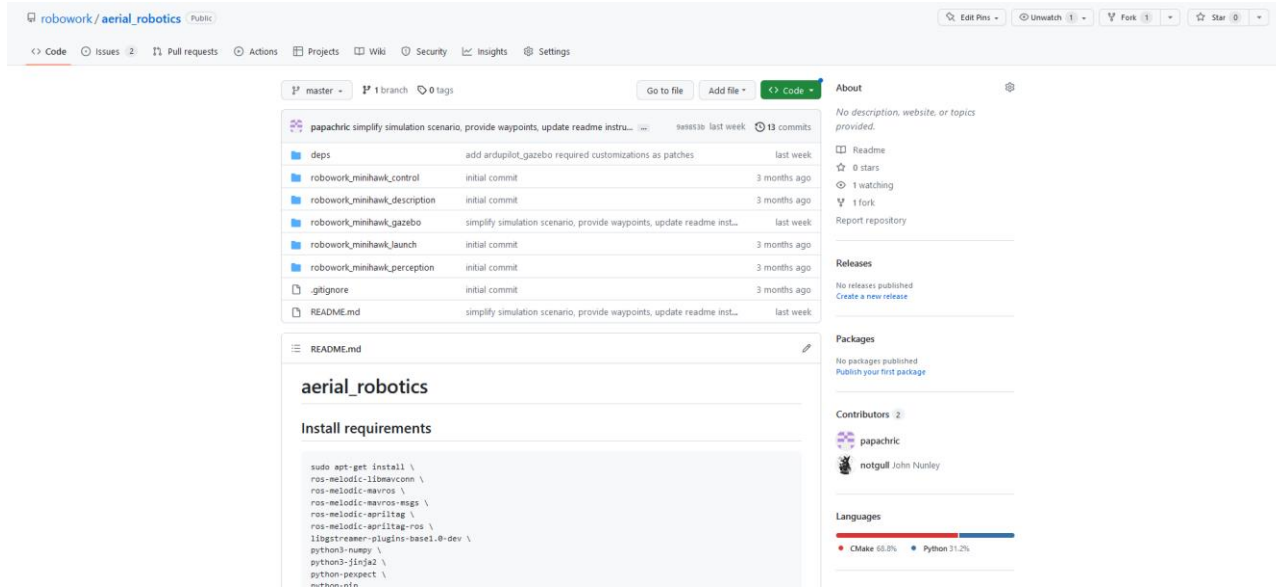
Home Research Education Resources Contact

Courses

Aerial Robotics (Robotics)

Offered as: CS-491/691

https://github.com/robwork/aerial_robotics



robwork / aerial_robotics Public

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1 watching

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No releases published

Create a new release

Packages

No packages published

Publish your first package

Contributors

papachric

notgull John Nunley

Languages

CMake 68.8%

Python 31.2%

aerial_robotics

Install requirements

```
sudo apt-get install \
ros-melodic-libmavconn \
ros-melodic-mavros \
ros-melodic-mavros-msgs \
ros-melodic-apriltag \
ros-melodic-apriltag-ros \
libgitreaper-plugins-base1.8-dev \
python3-numpy \
python3-jinja2 \
python-rospect \
python-nfa
```

