

A Method for Accelerated Simulations of Reinforcement Learning Tasks of UAVs in AirSim

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About me



Alberto Musa

Researcher assistant in the Department of Electrical, Electronic, and Information of the Alma Mater Università di Bologna.

Research topic:

 Optimization of autonomous Cyber Physical Systems (CPS), whit more focus on Unmanned Aerial Vehicles (UAVs), applied to Reinforcement Learning (RL) algorithms in real and simulated environments.

Introduction – UAV and Autonomous Navigation Results Conclusion



Mavic 2 PRO from DJI

Micro UAV from BitCraze

Unmanned Aerial Vehicles (**UAVs**) are complex robotic platforms (or Cyber-Physical Systems - **CPS**) designed for <u>flight without a human pilot</u>.

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Reinforcement Learning (*RL*) algorithms enable *UAVs* to perform <u>autonomous control tasks</u> such as obstacle avoidance.



Introduction – UAV and Autonomous Navigation



The RL loop requires to <u>simulate both the environment</u> and the CPS agent. RL <u>extracts knowledge from the</u> <u>interaction</u> between the environment and the agent. Intro

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RL training requires *performing the task* (called game) *repeatedly until the task is learned*. This can easily require *thousands or millions of games*.



Introduction – UAV and Autonomous Navigation

Simulators *replicate the agent capabilities* (e.g., the **UAV** flight) *and real environments*

- real-world physics rules, and perception capabilities.
- forces that act in the simulated scenario (gravity, rotors actuation, collisions, etc.)

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Most suitable simulators for UAV applications						
	AirSim	Flightmare	Gazebo	Webots		
Photorealism	X	x				
Co-Simulation	X		X	x		

Relevant simulator features are:

- **Co-Simulation:** rapid flight control and RL-trained solution rapid prototyping.
- **Photorealism:** if the task to be performed by the agent leverages camera sensors, accurate and photorealistic rendering of the scene becomes mandatory.

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Challenges



RQ2: What is the implication in terms of performance of the simulated drone flight, interaction with the environment, and trained RL algorithm?



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AirSim

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Client APIs

Used to interact with the built-in autopilot.

- Impose the desired waypoint through an **asynchronous command**
 - Enforcing synchronous control
- Gather the UAV agent state

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AirSim

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Built-in Flight Control Systems (**Simple Flight** Controller)

• <u>stoppable clock</u> to pause the simulation at any point.



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ClockSpeed setting

changes the ratio between the simulation and the wall clock time



AirSim Default Synchronous Command



Client *gathers multiple UAV states during the command* <u>execution</u> by issuing a concurrent <u>thread that pauses the</u> <u>simulation at specific time intervals.</u>

AirSim Default Synchronous Command (ADSC).

Mission Computer Application (MCA)

- **Deep RL** aims at <u>solving complex robotic tasks by</u> <u>mimicking human training behavior</u> with the use of **Neural Networks.**
 - Uses Client APIs to issue commands and observe the UAV state

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AirSim Setup Profiling

•Callback timing of three consecutive commands with **ADSC**.

- <u>Accelerated simulation is subject to command delays</u> due to communication between simulation, Simple flight, and Client API.
- Delay in the move command increases with the accelerator factor \Rightarrow during the command delay, <u>the simulator continues to</u> <u>evaluate the **UAV**</u>.



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- In an accelerated game <u>the simulator rendering time</u> <u>is lowered</u>, making <u>hard to complete the images.</u>
- •The result images may be perturbed.

Speed-up Effects Mitigation – Latencies

Time-Controlled Simulation Command (TCSC).

- uses the *stoppable clock* of the Simple Flight to control the simulation until the command expires.
- is implemented by *periodic stimulation interrupts* interleaved with command expired-time checks.

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Pause the simulation

Simulation pause

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- T = imposed time;

Simulation pause

1. Launch asynchronous command for time *T*



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- 2. Continue simulation for time *T*



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Simulation pause

- 1. Launch asynchronous command for time *T*
- 2. Continue simulation for time *T*
- 3. Collect simulation time T'



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Simulation pause

- 1. Launch asynchronous command for time *T*
- 2. Continue simulation for time *T*
- 3. Collect simulation time T'

4. If *T*′≤*T*: *T*=*T*-*T*′

return to 1. *Otherwise:* Collect State *or* End Command



Characterization Methodology

1. Accuracy of trajectories (**Deterministic Path 1**).

- Spatiotemporal Linear Combined (STLC) distance: score between 0 and 2, where <u>2 is the maximum value indicating</u> <u>that the trajectories are entirely overlapped</u>.
- *reference is the ADSC real-time trajectory*



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Characterization Methodology

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 - *reference is the ADSC real-time trajectory*

- 2. Visual perception accuracy (**Deterministic Path 2**).
 - Is the quality of the image related to the error in UAV gathering coordinates?
 - *Euclidean distance* between the coordinates (reference and accelerated simulation);
 - The <u>difference in percentage of image pixels</u> between the camera image taken in the accelerated and real-time simulation ⇒ Event Dissimilarity (ED).



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Characterization Methodology

3. Impact of the agent performance trained with RL.

Replicate an obstacle avoidance task in a lane (AirSim RL Environment) using event camera images

- **RL** training procedure of 5000 games.
- **RL** inference in real-time with **ADSC**.
- Evaluation of the UAV <u>success ratio</u> (number of times the UAV reaches the end of the environment without collisions) in inference over 100 games.
 - interchangeability of commands when the simulation happens in real-time.
 - comparison of the UAV success ratio computed with different AirSim acceleration factors.



I - Trajectory accuracy

- •For a given AirSim acceleration factor, the trajectory accuracy worsens with the command length \Rightarrow <u>the error</u> <u>accumulates between consecutive commands</u>. This is expected as the source of the trajectory error is the command latency.
- •The proposed **TCSC** command significantly <u>*outperforms*</u> the **ADSC** trajectory accuracy.
 - The proposed **TCSC** provides a *distance higher than 1.8 in the median case.*
 - The **ADSC** reaches it for acceleration lower than 5x.



II - Visual Perception accuracy

• 5x 0.35 10x 50x Έ 0.30 100x Distance 0.25 0.20 Euclidian 0.15 0.10 0.05 0.00 . 15.0% 0.0% 5.0% 10.0% 20.0% 25.0% **Event Dissimilarity**

Relation between acquisition points and quality of the images

Increasing the acceleration factor to slightly increases the **ED** and the Euclidean distance.

If the **ED** depended on the position error, the points would be *along the oblique arrow*.

For each acceleration factor, *horizontal clusters are* generated.

The difference in the images seen from the **UAV's** camera does not depend only on the position error induced by the simulation acceleration but that the latency in command can lead to a different camera orientation while the **UAV** is in a relatively similar position.

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III - RL Agent Performance

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Simulation acceleration with **TCSC** applied to **RL** algorithms does not alter the simulation environment.

By increasing the acceleration of the simulation, the trajectories and perception perturbations increase, impacting the performance of the **RL**.



ClockSpeed 5x 10x 50x 100x 1x**ADSC** 23 1 0 0 0 TCSC 21 30 23 14 13

- By accelerating the simulation to 5.2x the **UAV** success ratio is the same as the inference with training performed in real-time
- Training with *ClockSpeed* of 50x and 100x achieved effective speed-up of 14.8x and 15.4x and replicated more than 62% of the **UAV** success ratio in real-time.

Accelerated simulation training with **ADSC** failed to complete games in real-time inference.

With **TCSC**, training that in real-time simulation requires one week of simulations has been replicated in less than two days.

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Intro Background Methodology Results **Conclusion**

A method to accelerate the training of **UAV** agents trained in **RL** by reducing the simulation time.

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Thanks for your attention



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