

# **VALIDATION AND VERIFICATION OF SAFETY-CRITICAL ASPECTS OF AUTONOMY IN ORBITAL ROBOTICS**

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# Orbital Robotics

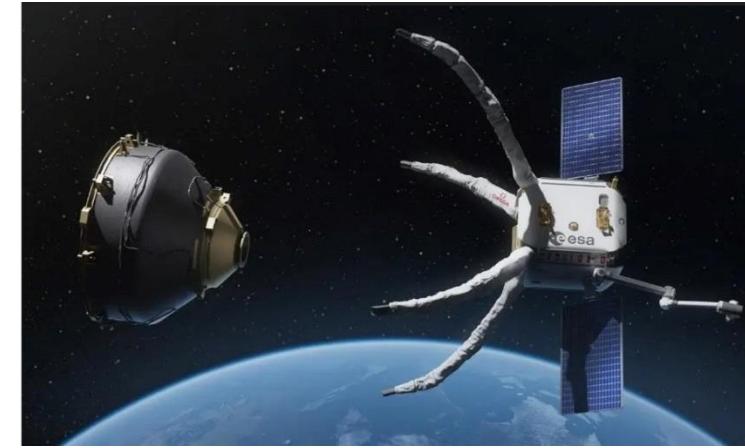
## Scenarios and Software Architecture

- In-Orbit Servicing
  - Life extension (MEV), Relocation
  - Upgrade, Repair (RSGS, EROSS SC)
- Active Debris Removal
  - Astroscale
  - ESA ADRIOS / Clearspace-1
- Assembly and Manufacturing
  - EU STARFAB automated orbitalwarehouse

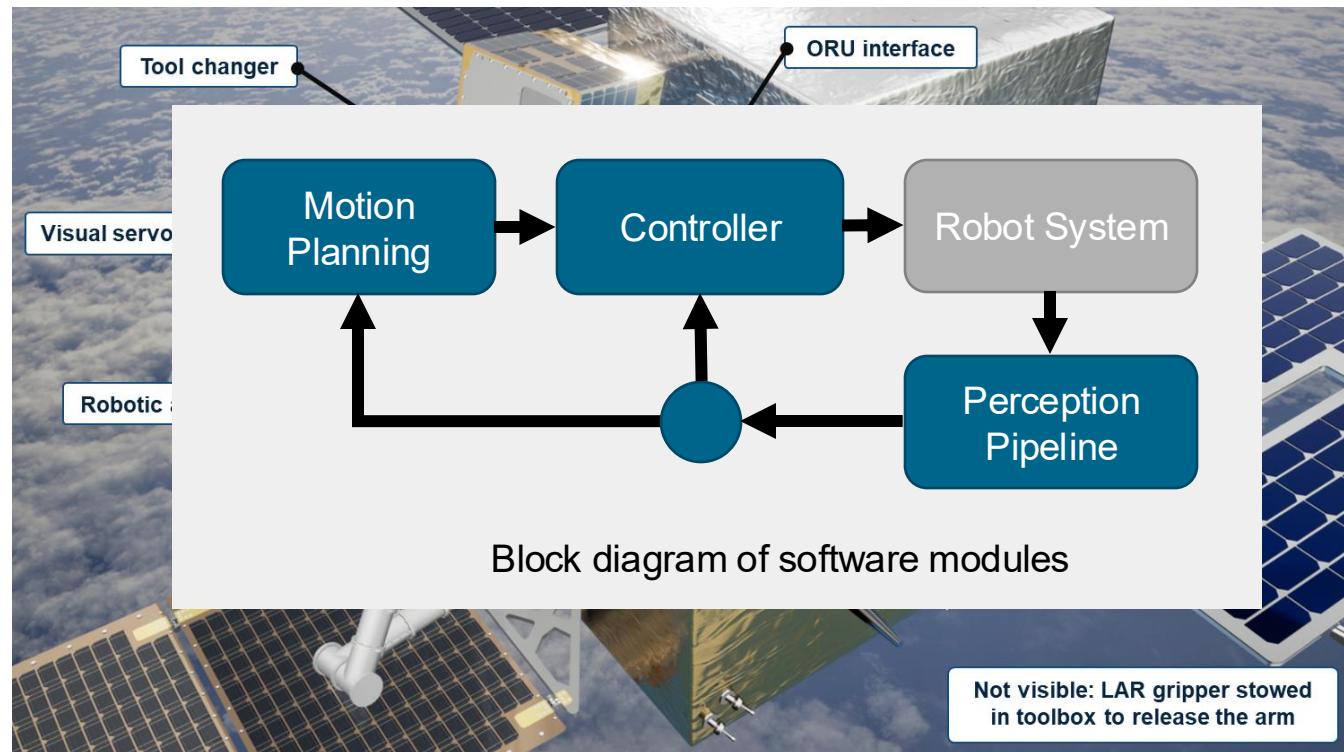
ERoss SC (EU)  
European robotic orbital support services -  
servicing component



ADRAS-J Mission (JAXA)



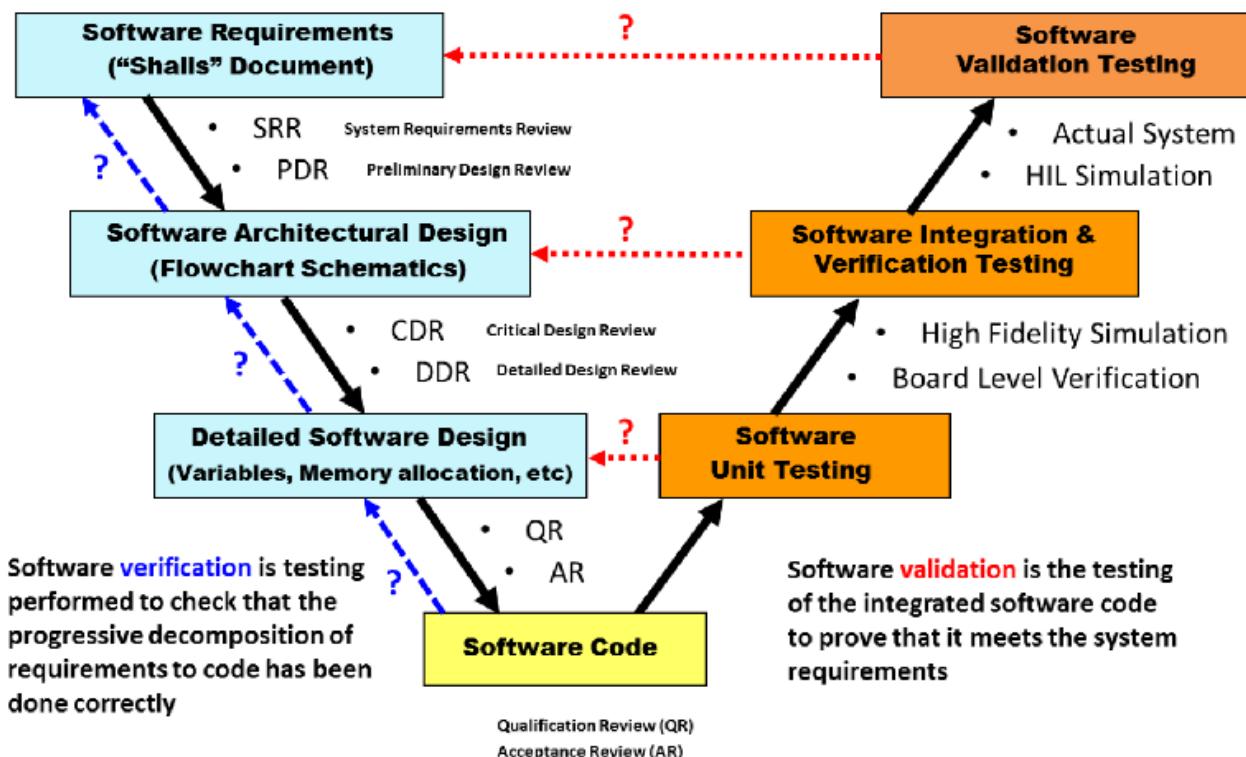
Clearspace-1 (ESA)



# V&V for On-Orbit Tasks



- Verification: does the code perform as it was designed to?
- Validation: does the integrated code perform as it should in the target environment (e.g. PC/OBC, on-ground/on-orbit)?
- No formal standard for V&V of Optimal Control and DL-based Perception for on-orbit tasks



Waterfall software development lifecycle, verification, and validation.  
Credit: NASA Marshall Space Flight Center, ECSS-M-ST-10C

# Optimal Control | Task



- Robots involved in highly complex tasks on-orbit
  - Highly nonlinear
  - Highly constrained
  - Often not suitable for convexification
- Formulate task as a parametric NLP( $p$ ):

- $p \in \mathbb{R}^{n_p}$
- $t_0 \leq t \leq t_f$
- $\delta t = \frac{t_f - t_0}{n-1}$

$$\begin{aligned} & \min_{\mathbf{z} \in \mathbb{R}^{n_z}} J(\mathbf{z}, p) \\ \text{s.t. } & G_i(\mathbf{z}, p) \leq 0, \quad i = 0, \dots, n_G \\ & H_j(\mathbf{z}, p) = 0, \quad j = 0, \dots, n_H \end{aligned}$$

- V&V of non-convex optimal control-based methods
  - Not codified in literature
  - Can follow a similar procedure used in V&V of SW for small satellite
- Models are not perfect → **uncertainty!**

# Optimal Control | Method

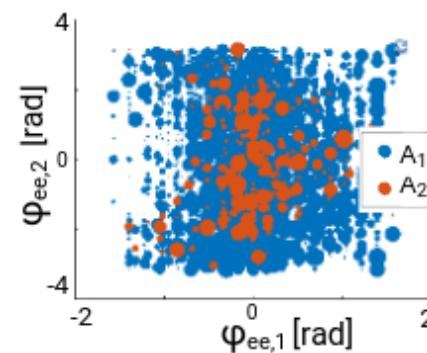
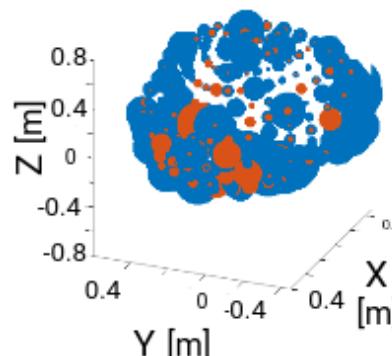
## Task Workspace

- A discrete grid on the task parameter space with:
  - Admissible solution on the grid points
  - Provable estimate of neighborhoods of validity of the sensitivity-based update for each grid point

$$\bar{z}(\mathbf{p}) = \hat{z}(\hat{\mathbf{p}}) + \frac{dz}{d\mathbf{p}}(\mathbf{p} - \hat{\mathbf{p}})$$

- Result:
  - Indication of robustness distribution on the task parameter space
  - Simple onboard computation

Task workspace  
for robot arm  
approach task



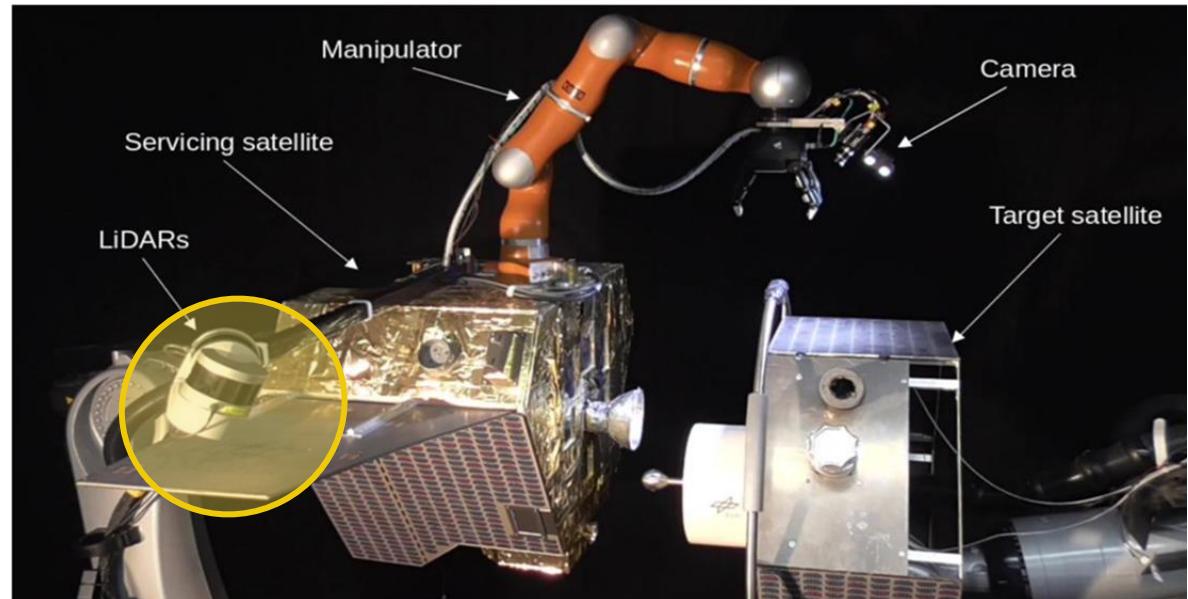
## The Task Workspace in V&V

- Admissibility → Satisfied mission operational requirements in the presence of uncertainty
- Mission planning:
  - Identify the safest corridors within which to operate
  - Targeted, early iteration on mission requirements
- Mission operations:
  - Assure robot operators that a generated or updated motion plan is feasible
  - Guide on whether another trajectory should be considered

# DL-based Perception | Method



- We developed a DL method for **on-board 6D pose estimation** of a known target satellite
  - **Lightweight** architecture for Point cloud 2 Pose Regression (**P2PReg**), encoder based on PointNet layers
  - Provides **robust** initial pose estimate of a known client satellite, for initializing the visual tracker
  - Processes unordered point sets and regresses pose parameters **adapted to client object symmetry**

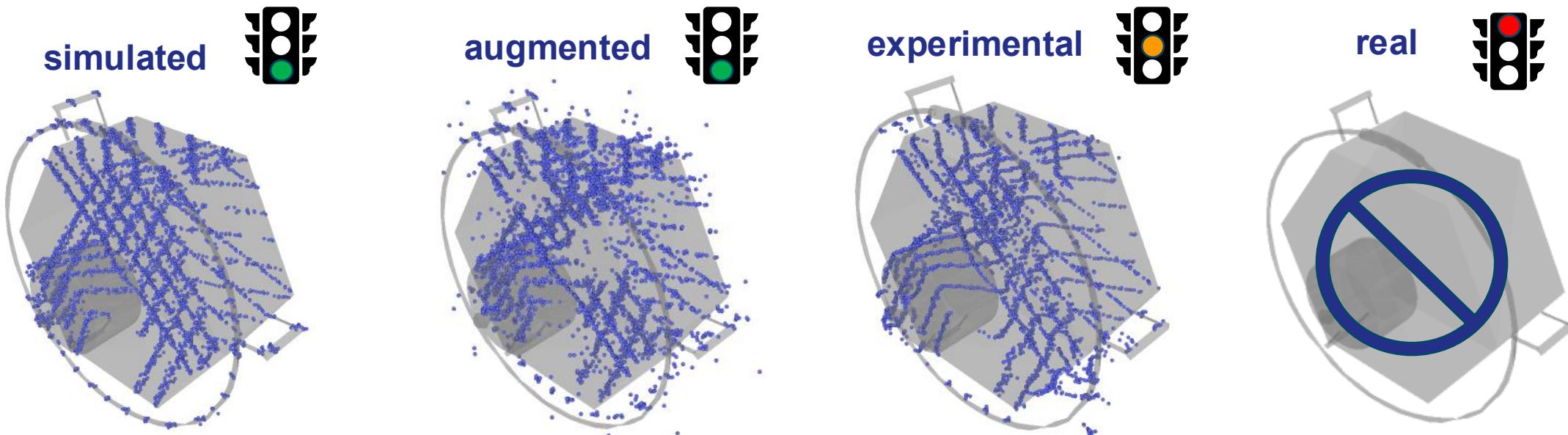


OOS-SIM (On Orbit Servicing – SIMulator), DLR

[M. Piccinin, U. Hillenbrand, “Deep Learning-based Pose Regression for Satellites: Handling Orientation Ambiguities in LiDAR Data”. Journal Of Image and Graphics, Vol. 13, No. 2, 2025.]

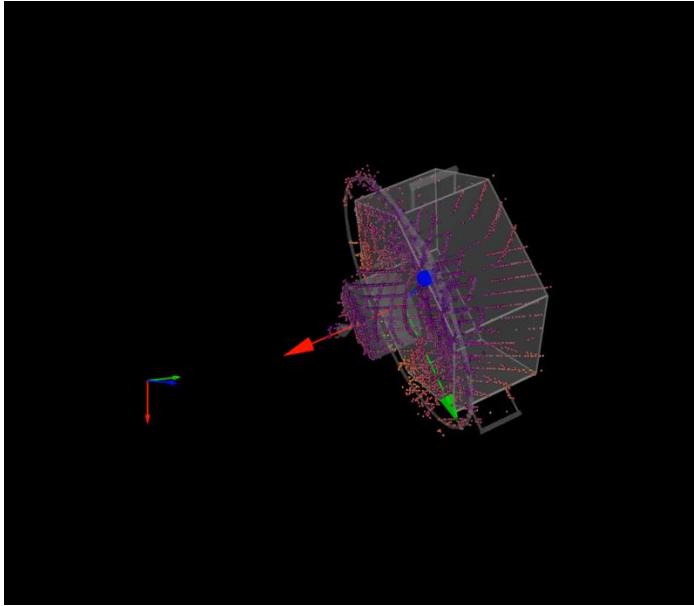
- V&V for DL methods in space safety-critical applications → need of a W-shaped iterative process

## 1. Data quality



- Representativeness
- Consistency
- Representativeness
- Fitness
- Metadata quality

## 2. Model development and testing

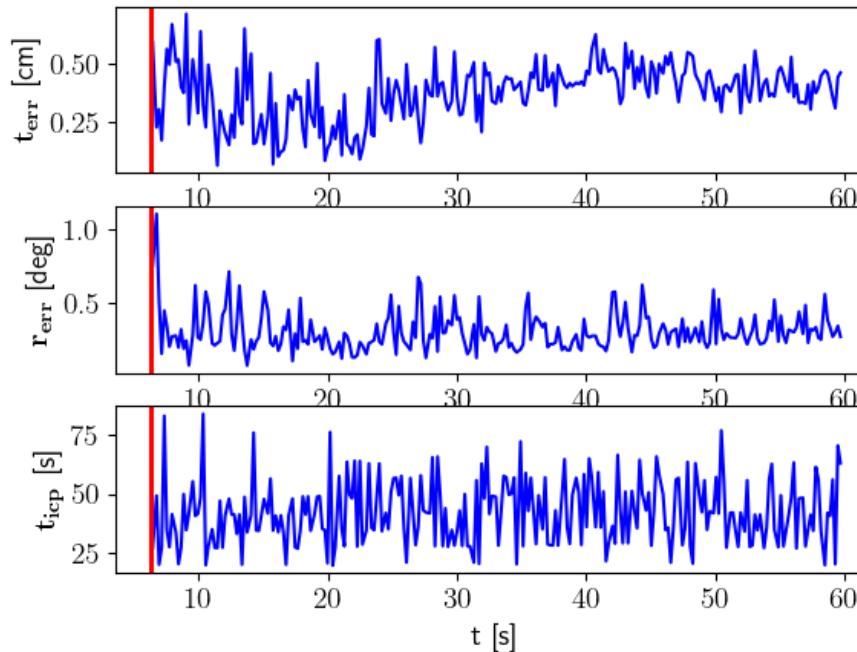


Evaluation of P2PReg pose initialization on single point clouds from on OOS-SIM.

- Outperforms classical and other DL methods
- Trained for **robustness** against data artifacts and model deviations
- Trained on solely synthetic data, it achieves excellent **sim2real transfer**

## 3. System testing

- Pose estimation running on space-relevant HW (ARM v7 processor)
- P2PReg successfully initializing the ICP tracker
- Requirements on translation error ( $t_{err}$ ), rotation error ( $r_{err}$ ) error and compute time ( $t_{icp}$ ) are met



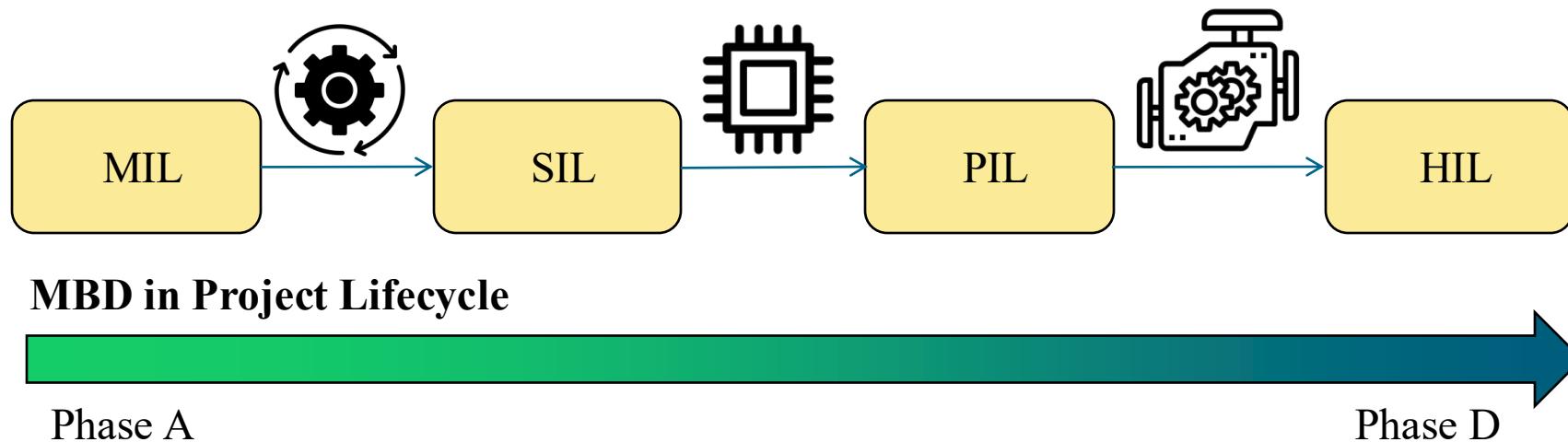
Pose estimation error and compute time for an example trajectory in open-loop PIL tests.

# Model-based Design (MBD)



**DDVV: Design, Development, Validation & Verification**

- From **Phase A/B1** to incremental **Phase C/D**
    - Incremental **TRL** improvement

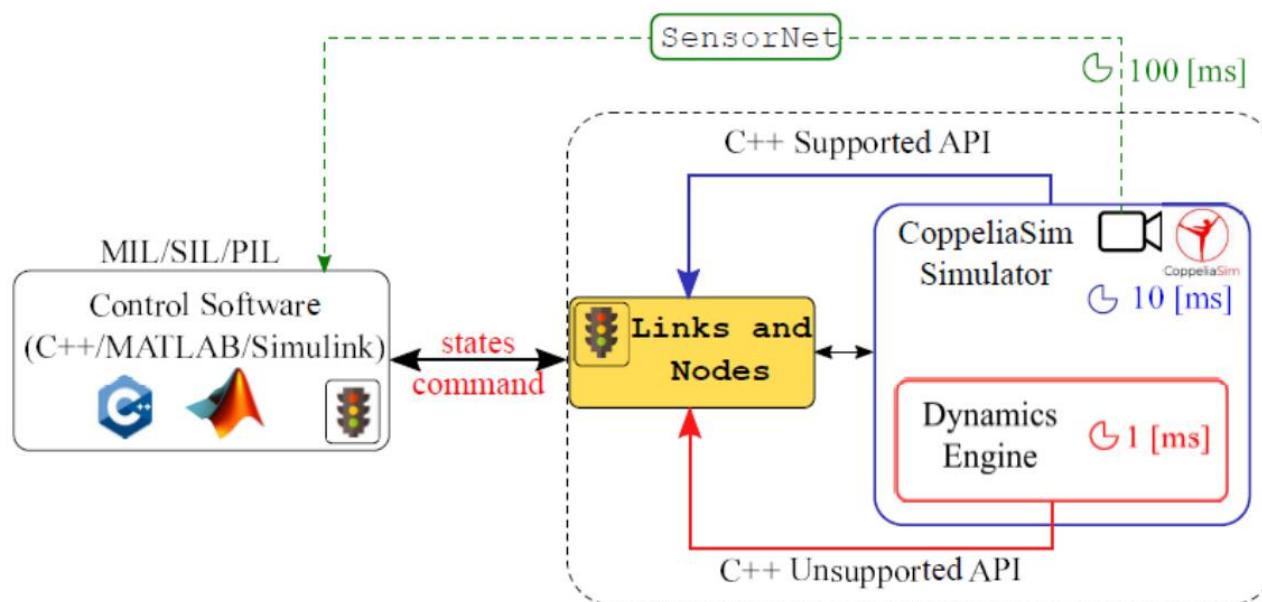
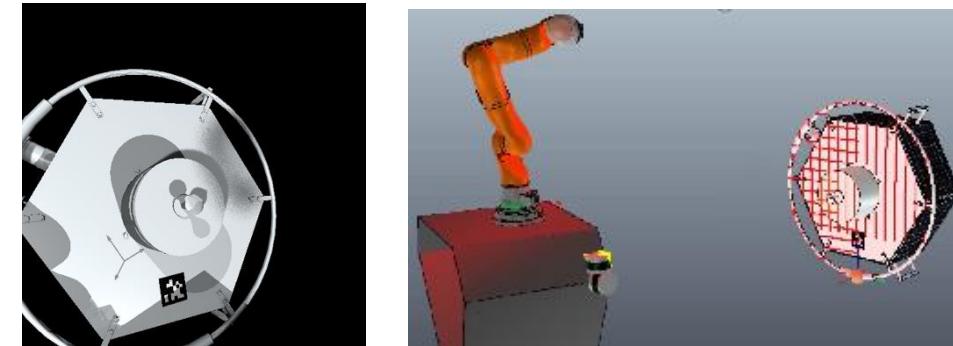


- **Different requirements:** Models, Sensors/Actuators, On-board computer, Math libraries
  - **Key objective:** Provide a domain-specific DDVV implementation for Orbital Robotics

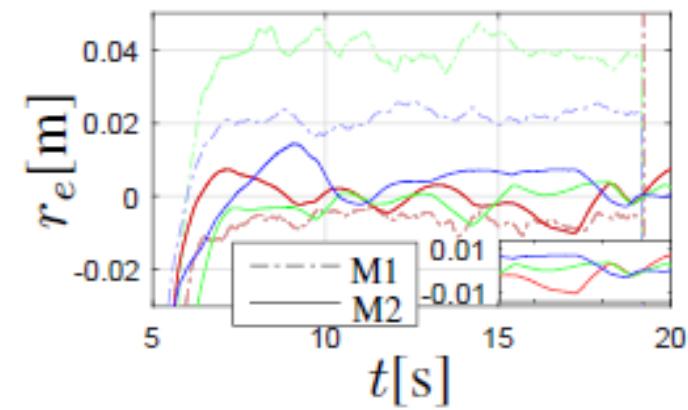
# Model-in-the-Loop (MIL)



- Model-in-the-Loop:
  - Components:
    - Sensor Performance Models (Camera, LiDAR)
    - Actuator Performance Models (joint flexibility)
    - Disturbance models (e.g., sloshing, flexible appendages)



**Perception:** Synthetic Sensor data from MIL Framework



**Control:** Free-floating (M1) vs Free-flying (M2) End-effector controller (pos. error) in MIL Framework

# Conclusion



- High-performance perception and control methods need dedicated V&V standards.
- In optimal control for space, common approaches include convexification, which is not suitable for robotics. New approaches are proposed for provable treatment of task uncertainty and for substantial reduction of V&V complexity.
- In ML-based pose estimation, some V&V guidelines are available. They were successfully implemented in our orbital scenario within MIL/SIL/HIL.
- The Model-based design approach was developed for rapid prototyping in orbital robotics.