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**Master's Thesis**

# **Mixed Reality Environment for Robot-in-the-Loop-Testing using Digital Twins**

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# **Abstract**

Hier sollte in einem Abstract kurz der Inhalt der Arbeit erläutert werden.  
Zuerst auf deutsch.

Then an abstract of the thesis in english should follow.



# Contents

<b>1</b>	<b>Introduction</b>	<b>1</b>
<b>2</b>	<b>Background and Related Work</b>	<b>5</b>
2.1	Robot-in-the-Loop-Testing . . . . .	5
2.2	Digital Twins . . . . .	8
2.3	Mixed Reality . . . . .	9
2.4	The VERA Framework and EMARO (Seperates Kapitel?) . . . . .	11
2.4.1	The EMARO Test Platform (detaillierte Beschreibung?) . . . . .	11
2.4.2	Original VERA System Architecture . . . . .	11
2.4.3	Platform Changes and Identified Limitations . . . . .	11
<b>3</b>	<b>Requirements</b>	<b>13</b>
<b>4</b>	<b>Concept and Implementation</b>	<b>15</b>
4.1	System Architecture . . . . .	15
4.1.1	Integration of Unity, ROS 2, and EMAROS . . . . .	15
4.1.2	Communication Concept . . . . .	15
4.2	Development Milestones . . . . .	15
4.2.1	Base Integration . . . . .	15
4.2.2	Realization of the Digital Twin . . . . .	15
4.2.3	Interaction Scenarios . . . . .	15
4.2.4	VR Integration . . . . .	15
4.2.5	Pure Virtual Model . . . . .	16
<b>5</b>	<b>Evaluation</b>	<b>17</b>
5.1	Methodology . . . . .	17
5.1.1	Test Scenarios . . . . .	17
5.1.2	Metrics . . . . .	17
5.2	Execution and Results . . . . .	17
5.2.1	Base Integration Performance . . . . .	17
5.2.2	Scenario-dependent Performance . . . . .	17

## *Contents*

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5.3 Discussion . . . . .	17
<b>6 Conclusion</b>	<b>19</b>
<b>Bibliography</b>	<b>25</b>

# 1 Introduction

Robotic and Autonomous Systems (RAS) combine multiple disciplines, including control engineering and robotics, mechanical engineering, electronics, and software engineering. Testing these systems is not as straightforward as using traditional methods, and they require more because they span multiple disciplines. For researchers and engineers specializing in software testing, adapting existing techniques for RAS presents a significant challenge. This is why there is extensive literature dedicated to proposing and evaluating different testing techniques and processes, showing how essential it is to establish structured methods for ensuring the reliability of these complex systems. [AMV23]

This validation challenge highlights a fundamental tension between simulation-only testing and full-scale physical experimentation, which are the two established methods in robotics evaluation. Simulation is central to the robotic development, offering a way to test control logic and experiment with system configurations before committing them to hardware [Hu05; Mic04]. Simulations are often easier to set up, less expensive, and can run faster than real-world tests, allowing for rapid design exploration and to get a better grasp of complex algorithms [Mic04; DRC+17; BM18]. However, a gap between the virtual and physical worlds emerges when simulation models are abandoned during the implementation phase. Virtual models, by their nature, cannot perfectly represent every characteristic of physical hardware, such as the exact effects of friction, sensor noise, or small differences in motors [Hu05; BM18]. On the other hand, testing with real hardware provides the highest fidelity and allows for controls to be refined based on the presence of all the unpredictable physics and variability that simulations can't replicate [Hu05; CCC+21]. But this approach has its own severe drawbacks. Conducting physical experiments can be resource-intensive and time-consuming and require significant funds, manpower, and infrastructure [Hu05; DRC+17; Mic04]. Furthermore, some critical scenarios, such as emergencies, are too dangerous to be recreated in the physical world [Hu05]. For complex, large cooperative systems, testing the entire system with only real components may not even be feasible due to the issues of complexity and scale [Hu05]. Since neither approach on its own is

sufficient, there is a need for an intermediate solution that can connect the simulation with real-world experimentation [Hu05].

Hybrid methods such as Hardware-in-the-Loop Simulation (HILS) have emerged to bridge the validation gap by replacing segments of pure simulations with actual hardware components [Hu05]. This research focuses on a specific application of this concept known as "robot-in-the-loop" (RitL) simulation, a more recent approach that allows physical robots and digital robot models to work together for comprehensive system testing and evaluation [Hu05]. RitL allows researchers to work with actual robots inside a simulated environment, even when a full physical setup is unavailable [Hu05]. Central to this approach is the Digital Twin [AA23]. The digital twin is a live, virtual replica of a physical robot that tracks its real-world counterpart in real time [AA23]. With a continuous connection and rapid data exchange, the digital twin duplicates the robot's actions, ensuring both remain in sync [AA23]. To manage this hybrid test environment, Augmented Reality (AR) is increasingly utilized as a way for interaction and information exchange with autonomous systems, enhancing the efficiency of Human-Robot Interaction (HRI) [MV20]. The combination of RitL testing [Hu05], synchronized Digital Twins [AA23], and AR interfaces [MV20] paves the way for the development and validation of future robotic systems.

The "Virtual Environment for mobile Robotic Applications" (VERA) framework integrates digital twins, AR, and vehicle-in-the-loop testing together into one system [Geh24]. VERA provides a modular platform for creating, managing, and visualizing synchronized virtual environments, which are presented both in simulations and as real-world projections [Geh24]. This master's thesis builds directly upon the foundation laid by this original framework.

The VERA framework provides a good foundation, but its initial version has several limitations that this thesis seeks to address [Geh24]. Firstly, the custom environment manager does not feature a full physics simulation, with the integration of enhanced physics noted as a direction for future work [Geh24]. The 2D visualizer that is created with Pygame faced performance problems while handling a higher number of dynamic objects [Geh24]. If there are too many simultaneous updates, its capacity was exceeded, which led to delayed and incomplete visualizations [Geh24]. Additionally, while the framework supports AR projections, a more immersive Virtual Reality (VR) interaction was not implemented and was identified as potential future work [Geh24]. Therefore, the main problem is that although VERA's concepts are promising, its custom components have technical constraints that limit it, particularly regarding physical realism, scalability, and user interaction [Geh24].



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This thesis proposes replacing VERA's custom simulation and visualization components [Geh24] with the Unity Engine [Uni23]. This implementation will use the Unity Engine [Uni23], which was selected for its graphics, integrated physics, and VR support, because these features directly address the technical limitations previously restricting VERA.

The core goal is to create a real-time digital twin [AA23] of the EMAROS robot [Geh24], integrating its complete model, live sensor data, and physical properties such as mass, inertia, and collision models. This digital twin will serve as the foundation for "robot-in-the-loop" [Hu05] testing scenarios. Additionally, this project will implement interaction by extending the original AR projections [Geh24] to include Virtual Reality (VR) [EM21] and desktop interfaces for scenario modification, robot control, and data visualization. This also includes reimplementing and enhancing VERA's AR floor projection feature by using Unity's rendering tools to improve visual quality and system capabilities [Uni23].

[**NOT FINAL**]The remainder of this thesis is structured as follows: Chapter 2 reviews the foundational concepts of Robot-in-the-Loop testing, Digital Twins, and Mixed Reality, and includes a review of the original VERA framework. Chapter 3 defines the new system requirements based on VERA's limitations. Chapter 4 details the implementation of the new architecture, including the Unity/ROS 2 integration, the EMAROS digital twin and the implemented scenarios. Chapter 5 presents the evaluation and discusses the results. Finally, Chapter 6 summarizes the thesis contributions and provides an outlook on future research.[**NOT FINAL**]



## 2 Background and Related Work

Einführung in das kapitel [NOT FINAL]

### 2.1 Robot-in-the-Loop-Testing

The validation and verification of modern Robotic and Autonomous Systems (RAS) is a significant challenge due to their complexity [AMV23]. This is because they integrate software, mechanical, and electrical engineering all at once [AMV23]. A central problem is ensuring that the software and hardware work together seamlessly, especially when testing both simultaneously [AMV23]. The X-in-the-Loop (XitL) simulation paradigm addresses this challenge [AAR+19]. It offers a way to combine the flexibility of software simulation with the realism of physical experiments [AAR+19].

A foundational XitL method is Hardware-in-the-Loop (HIL) simulation, which is a technique for testing mechatronic systems [MTH22; AAR+19]. The main principle of HIL is creating a closed loop between the real hardware that is being tested and a simulation that represents the rest of the system or its operational environment [MTH22]. This setup effectively tricks the hardware into behaving as if it were operating in a real system, which allows for testing across a wide range of virtual scenarios [AAR+19]. The main reason for using HIL is to shorten development cycles and prevent costly or dangerous failures by making exhaustive testing possible before the system is actually completely implemented [MTH22]. This is why HIL is essential in industries like automotive, aerospace, and robotics, where real-world testing can be too expensive, dangerous, or even impossible [AAR+19].

Robot-in-the-Loop (RitL) [MTH22] simulation extends the Hardware-in-the-Loop (HIL) concept. Instead of just a component, the hardware under test is a complete robotic system, such as an uncrewed vehicle [MTH22]. RitL replaces components of an pure simulated setup with the actual robot, increasing the realism of testing [Hu05]. [NOT FINAL]As illustrated in Figure ??[NOT FINAL], a typical RitL configuration has the robot's real actuators operating in the physical world, while while its sensors

interact with a simulated environment instead [Hu05; MTH22]. This creates a hybrid setup where, for example, the robot might use virtual sensors to see objects in the simulation but use its real motors to move physically [Hu05]. To keep the physical and virtual worlds synchronized, the real robot often has a virtual counterpart in the simulation which state is updated as the physical robot acts and moves [Hu05].

Figure 2.1: [NOT FINAL]The real robot’s actions affect its virtual counterpart, and the virtual environment provides sensor data back to the real robot.[NOT FINAL]

The main benefit of RitL is that it uses the dynamics of actual hardware for repeatable testing of high-level software, such as navigation algorithms [MTH22]. This method avoids the expense, complexity, and risk of full physical testbeds while being a secure and useful substitute [MTH22]. RitL becomes therefore an tool for safely evaluating system performance in the lab [Hu05; MTH22]. This is especially important when working on projects that are hard to replicate, such as large-scale robot swarms or Mars rover missions, or when safety is at risk [Hu05; MTH22].

The RitL paradigm has been widely applied to validate complex autonomous systems, particularly in the automotive field using Vehicle-in-the-Loop (ViTL) testing. For example, the Dynamic Vehicle-in-the-Loop (DynViL) architecture integrates a real test vehicle with the high-fidelity CARLA simulator, which operates on the Unreal Engine [DSR+22]. As shown in Figure 2.2, this approach allows a vehicle on an empty track to be stimulated by sensor data from a virtual world [DSR+22]. This lets automated driving functions to be safely and reliably tested in situations that would be hazardous to physically replicate [DSR+22]. CARLA and the Unreal Engine are also used in a similar Vehicle-in-Virtual-Environment (VVE) method which establishes a closed loop in which the motion of the real vehicle is tracked and reflected in the virtual world, making it possible its control systems to respond to simulated events [CCG+23].

These examples show a trend towards using game engine based simulators to evaluate autonomous vehicles. This method falls somewhere between physical testing, where safety and consistency can be problematic, and pure software simulation, which frequently lacks vehicle dynamics fidelity [CCG+23]. Some systems even stimulate the vehicle’s actual sensors. For instance, the Radar Target Simulator (RTS) can feed artificial radar echoes from a virtual scene to the vehicle’s actual radar sensor [DKK+21]. This allows for end-to-end validation of the entire perception and control pipeline [DKK+21].

The X-in-the-Loop approach is found in many different areas, not just a single field. In marine robotics, a VIL framework was developed to test the long term autonomy of



Figure 2.2: An example of a Vehicle-in-the-Loop (ViTL) setup. A real car on a test track connected to a high-fidelity simulator like CARLA that generates virtual traffic and sensor data. [DSR+22]

a robot swarm. In this system, an Autonomous Surface Vehicle (ASV) interacts with multiple simulated underwater sensor nodes to test cooperative behaviors without the logistical cost of deploying a full swarm [BVM20]. In the aerospace domain, the RFlySim platform uses an FPGA-based HIL system to create a high fidelity simulation for testing UAV autopilot systems in a lab, which reduces the need for expensive and risky outdoor flights [DKQ+21].

These approaches continue to move toward deeper integration. This includes the introduction of Scenario-in-the-Loop (SciL) frameworks that aim to completely blur the lines between real and virtual testing [VTS21]. These systems rely on creating detailed digital twins of the entire test environment and combining real and virtual components to run complex, mixed reality test scenarios [VTS21].

### 2.2 Digital Twins

The Digital Twin (DT) is an important concept in many current X-in-the-Loop frameworks. The DT serves as the virtual counterpart to a physical system, acting as a virtual copy that allows real-time monitoring and simulation [AA23]. The idea is to create a digital information model of a physical system that stays linked to it for the duration of its lifecycle [GV17].

This concept was initially known as the Information Mirroring Model and consists of three core elements: the physical product, its corresponding virtual model, and the data connection that connects them [GV17; AA23]. A diagram of this structure is shown in Figure 2.3. The virtual model is more than a basic geometric shape, it is often a detailed simulation that can model mechanical, electrical, and software properties of a system [LWS+21]. Information moves in both directions, so sensor data can flow from the real world to update the virtual model [GV17; LWS+21]. In turn, the virtual model can also send commands back to control or optimize the physical system [GV17; LWS+21]. This continuous, bidirectional data exchange characterizes a Digital Twin [AA23].



Figure 2.3: The foundational concept of a Digital Twin, illustrating the three core components: the physical entity, the virtual model, and the bi-directional data connection that links them [GV17; LWS+21]. [Gri15]

Digital Twins in robotics are frequently created using simulation software and the middleware that connects the virtual simulation to the actual hardware is often the Robot Operating System (ROS) [MFG+22]. Stączek et al. used the Gazebo simulator with ROS to create a DT of a factory floor, which they used to test and optimize the

navigation algorithms of a mobile robot in narrow corridors [SPD+21]. To validate a deep learning-based harvesting robot, R. Singh et al. adopted a similar strategy and created a DT of a greenhouse in Gazebo [BSH24]. Other simulators like CoppeliaSim and Webots are also common. Magrin et al. used CoppeliaSim and ROS to create a DT as a learning tool for mobile robot design [MCT21], while Marques et al. used Webots for an Automated Guided Vehicle (AGV), synchronizing it via an OPC-UA server [MRS24]. These examples show a common approach of using traditional simulators to simulate robot kinematics and sensor feedback for closed-loop testing.

More recently, game engines have become a popular choice for creating Digital Twins, as they offer more realistic graphics and physics. Unity [Uni23], for instance, is used by developers to build virtual environments that are highly realistic. [NOT FINAL] This can be seen in Figure ??[NOT FINAL]. Pérez et al. used Unity to develop a DT of a multi-robot cell with a Virtual Reality (VR) interface for virtual commissioning and operator training [PRR+20].

Figure 2.4: [NOT FINAL] A comparison of robotic Digital Twin environments, showing a view from a traditional simulator (left) versus a high-fidelity visualization from a modern game engine like Unity (right).[NOT FINAL]

Another important consideration are the technical capabilities these engines. Yang et al. [YMZ20] simulated the physics of a UAV using Unity and its built-in NVIDIA PhysX engine. They also demonstrated the creation of virtual sensors, such as a LiDAR, directly within the game engine using its raycasting API [YMZ20]. Research has also been done on the performance and reliability of these game engine-based frameworks. Kwon et al. developed a safety-critical DT in Unity and ROS 2, reaching a data-transmission latency of 13 ms for predicting collisions [KYS+25]. Similarly, M. Singh et al. created a DT using ROS and Unity and conducted performance validation with a communication latency of 77.67 ms and a positional accuracy of 99.99% [SKG+24].

## 2.3 Mixed Reality

Beyond the testing paradigm and the digital twin, the way a user interacts with the system is another part of a modern robotics framework. Virtual, Augmented, and Mixed Reality (VAM) have become promising technologies for improving the information exchange between humans and robots [WPC+22]. In robotics, Augmented Reality (AR) is an especially useful tool for enhancing Human-Robot Interaction (HRI) by integrating 3D virtual objects into a real-world environment in real-time [MV20].

The relationship between these technologies is formally described by the foundational Reality-Virtuality (RV) Continuum concept, first introduced by Milgram and Kishino [MK94]. As illustrated in Figure 2.5, this continuum is a scale that is anchored by a purely real environment at one end and a completely virtual one at the other [MK94; SSW21; MV20].



Figure 2.5: The Reality-Virtuality Continuum, illustrating the spectrum from a real environment to a completely virtual one. [WPC+22]

A Virtual Reality (VR) environment is an endpoint of the continuum, where the user is totally immersed in and can interact with a fully synthetic world [MK94]. This approach is useful for HRI research, as it allows for testing interactions with virtual robots in scenarios where it might be unsafe or too expensive for physical hardware [WPC+22]. The general term Mixed Reality (MR) describes the entire spectrum between the two extremes, where real and virtual worlds are combined into a single display [MK94]. MR is made up of two main categories: Augmented Reality (AR) and Augmented Virtuality (AV) [MK94; MV20]. AR is the process of adding virtual objects to a real environment [MK94]. This lets for example, HRI researchers to place 3D data and intentions of a robot directly into the physical space of an user [WPC+22]. The opposite is Augmented Virtuality (AV), where a primarily virtual world is enhanced with elements from the real world, like live video feeds [MK94].



## **2.4 The VERA Framework and EMARO (Seperates Kapitel?)**

### **2.4.1 The EMARO Test Platform (detaillierte Beschreibung?)**

### **2.4.2 Original VERA System Architecture**

#### **2.4.2.1 Virtual Environment Positioning System (VEPS)**

#### **2.4.2.2 Virtual Environment Manager**

#### **2.4.2.3 Visualization System (RViz and Pygame)**

### **2.4.3 Platform Changes and Identified Limitations**

#### **2.4.3.1 Changes of the Positioning System**

#### **2.4.3.2 Identified Limitations of the Original Software**



## **3 Requirements**



# **4 Concept and Implementation**

## **4.1 System Architecture**

### **4.1.1 Integration of Unity, ROS 2, and EMAROS**

components, their responsibilities, and deployment setup.

### **4.1.2 Communication Concept**

Topics, message types, timing, synchronization ...

## **4.2 Development Milestones**

### **4.2.1 Base Integration**

Initial data flow, messaging bridge, and minimal scenes.

### **4.2.2 Realization of the Digital Twin**

Robot modeling, environment setup, and synchronization.

### **4.2.3 Interaction Scenarios**

Scenarios and scripting.

### **4.2.4 VR Integration**

Devices and implementation, maybe before Interaction Scenarios?

#### **4.2.5 Pure Virtual Model**

Implementation and use cases without the physical counterpart.

# **5 Evaluation**

## **5.1 Methodology**

### **5.1.1 Test Scenarios**

Define representative scenarios and justification.

### **5.1.2 Metrics**

Latency, frame rate, synchronization accuracy, and resource usage.

## **5.2 Execution and Results**

### **5.2.1 Base Integration Performance**

baseline measurements and analysis

### **5.2.2 Scenario-dependent Performance**

performance under different interaction and load conditions.

## **5.3 Discussion**

results, limitations, and implications.





## **6 Conclusion**

summary of the work and discussion of future research directions.



## List of Figures

2.2	An example of a Vehicle-in-the-Loop (VitL) setup. A real car on a test track connected to a high-fidelity simulator like CARLA that generates virtual traffic and sensor data. [DSR+22] . . . . .	7
2.3	The foundational concept of a Digital Twin, illustrating the three core components: the physical entity, the virtual model, and the bi-directional data connection that links them [GV17; LWS+21]. [Gri15] . . . . .	8
2.5	The Reality-Virtuality Continuum, illustrating the spectrum from a real environment to a completely virtual one. [WPC+22] . . . . .	10



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## **Declaration of Authorship**

I hereby declare that I have written this thesis independently and without unauthorized assistance. I have not used any sources or aids other than those indicated. All passages taken verbatim or in spirit from the works of other authors have been properly acknowledged and cited.

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