

¹ CAWSR: Carla-AutoWare Scenario Runner

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⁵ Summary

⁶ CAWSR (**CARLA-AutoWare-Scenario Runner**) facilitates the simulation-based testing of the
⁷ open-source autonomous driving system, Autoware, within CARLA, the state-of-the-art open-
⁸ source driving simulator. Building on existing tools, this project introduces a research-oriented
⁹ testing framework for the execution of complex driving scenarios, as well as supporting the
¹⁰ implementation of a wide range of verification strategies.

¹¹ Statement of Need

¹² Verifying Autonomous Driving Systems (ADS) is a critical step before they can be deployed.
¹³ However, relying only on real-world testing is too expensive, inefficient, and potentially
¹⁴ dangerous. Consequently, simulation-based testing has become essential, allowing researchers
¹⁵ to safely test driving agents against critical situations at scale. Among these tools, CARLA
¹⁶ ([Dosovitskiy et al., 2017](#)) has become the de-facto standard in the research community due to
¹⁷ its rich ecosystem of open-source tools, benchmarks, and documentation.

¹⁸ Currently, the standard for evaluating ADS in CARLA is the CARLA Leaderboard and its
¹⁹ engine, Scenario Runner (SR) ([CARLA, 2025](#)). This framework is typically used to test
²⁰ “black-box” driving agents, such as ML-based systems which expose only sensor-level inputs
²¹ and driving control outputs. By running a set of predefined, challenging driving scenarios,
²² researchers can systematically assess agent performance using common metrics like driving
²³ score, infractions, and route completion. However, applying this testing framework to industry-
²⁴ grade ADS, such as Autoware ([Kato et al., 2018](#)) or Apollo ([Baidu, 2017](#)), remains difficult.
²⁵ Although communication bridges exist between CARLA and these systems ([Guardstrikelab,](#)
²⁶ [2023; Kaljavesi et al., 2024](#)), they lack native support for scenario execution engines, which
²⁷ limits their utility for scenario-based testing.

²⁸ This gap has created a significant bottleneck for the research community. Previously, researchers
²⁹ developing scenario generation algorithms mainly relied on combining Apollo with the LGSVL
³⁰ simulator ([Rong et al., 2020](#)). However, LGSVL is now outdated, with official support ending
³¹ in January 2022. This leaves many researchers without a suitable industry-grade “subject”
³² for evaluating their algorithms. While recent tools like PCLA ([Tehrani et al., 2025](#)) attempt
³³ to simplify deploying Autoware (and other ADS implementations) into CARLA, they focus
³⁴ primarily on simplifying the ADS implementations and abstracting the setup process across
³⁵ different CARLA versions. They lack the deep integration required between the agent and
³⁶ simulator to execute complex, route-based scenarios.

³⁷ CAWSR aims to bridge this gap by enabling the evaluation of Autoware in complex driving
³⁸ scenarios within CARLA. By building on the established CARLA platform, this work provides a
³⁹ modern replacement for the outdated Apollo/LGSVL workflow. It also allows Autoware to be
⁴⁰ directly compared with state-of-the-art research agents on the CARLA Leaderboard.

⁴¹ Effective ADS verification requires the ability to systematically explore the operational design
⁴² domain. To support this, CAWSR provides a flexible interface for algorithmic scenario generation.
⁴³ This facilitates a wide range of verification strategies based on common metrics, such as the
⁴⁴ CARLA Leaderboard's driving score (CARLA Team, 2024).
⁴⁵ Lastly, it is worth noting that simulators can often introduce unintended nondeterminism,
⁴⁶ which leads to inconsistent test results (Chance et al., 2022; Osikowicz et al., 2025). Therefore,
⁴⁷ CAWSR is designed to minimise such nondeterminism throughout the evaluation pipeline.

⁴⁸ Research Impact

⁴⁹ CAWSR provides a significant research impact by bridging the gap between the widely-used
⁵⁰ CARLA simulator and the industry-grade Autoware ADS. It addresses a critical bottleneck
⁵¹ in the ADS testing research community by offering a modern replacement for the outdated
⁵² Apollo/LGSVL workflow, which has lacked official support since 2022.
⁵³ It enhances research reproducibility by implementing a fully synchronous evaluation pipeline
⁵⁴ that minimises unintended nondeterminism in simulation-based testing. To ensure community
⁵⁵ readiness and ease of use, it is distributed as a Docker container.
⁵⁶ Furthermore, by adopting CARLA Leaderboard metrics, CAWSR enables researchers to directly
⁵⁷ compare Autoware with other state-of-the-art driving agents in the CARLA environment. It
⁵⁸ provides an essential foundation for the systematic evaluation of various testing approaches for
⁵⁹ ADS.

⁶⁰ Software Design

⁶¹ CAWSR is a fully synchronous testing framework that directly integrates the CARLA simulator,
⁶² Scenario Runner (as the scenario executor), and Autoware (as the System Under Test) to
⁶³ facilitate autonomous driving testing research. The tool is distributed as a containerized
⁶⁴ deployment using Docker to manage complex dependencies and simplify the setup process.
⁶⁵ Currently, two modes of operation are supported:

- ⁶⁶ 1. *Scenario Generation Mode*: Enables the dynamic generation and execution of scenarios
⁶⁷ (e.g. iterative scenario generation) provided by a user-defined algorithm. This is
⁶⁸ particularly useful for assessing the performance of new simulation-based ADS testing
⁶⁹ techniques.
- ⁷⁰ 2. *Benchmark Mode*: Allows the execution of a predefined set of scenario definitions provided
⁷¹ by the user. This is useful for standardised evaluations and comparisons between different
⁷² driving agents.

⁷³ The evaluation pipeline is engineered to be fully synchronous, minimising unintentional non-
⁷⁴ determinism to facilitate reproducible results. However, it is noted that minor variations may
⁷⁵ still persist due to inherent non-determinism in upstream dependencies, such as the driving
⁷⁶ simulator or the driving agent itself (Chance et al., 2022; Osikowicz et al., 2025).

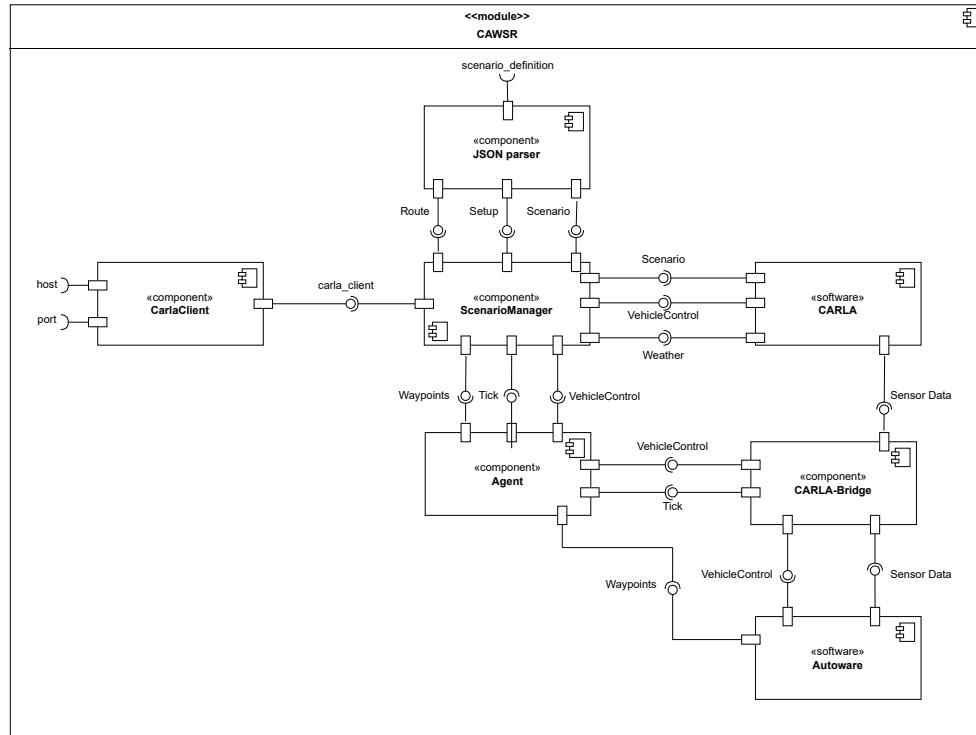


Figure 1: Internal component diagram of CAWSR.

77 Figure 1 illustrates the CAWSR architecture and its fundamental components. The framework
 78 operates through four primary modules:

- 79 ▪ CarlaClient: A native CARLA PythonAPI class that establishes a TCP connection
 80 (via host IP and port). It serves as the framework's exclusive interface for extracting
 81 simulation data and spawning entities.
- 82 ▪ JSON Parser: Translates the *scenario_definition* (see Figure 2) into a Behavior Tree
 83 (BT). It utilises Scenario Runner's *Atomic Behaviours* and *Atomic Conditions* as modular
 84 primitives to define discrete actions (e.g., spawning pedestrians) and logic triggers.
- 85 ▪ ScenarioManager (CARLA, 2025): Orchestrates the simulation loop by evaluating the
 86 BT to update actor states and triggering CARLA simulation ticks. Execution terminates
 87 based on CARLA Leaderboard criteria (CARLA Team, 2024), as summarised in Table 1.
 88 Post-execution, the module calculates the Driving Score (DS) according to the official
 89 leaderboard metrics.
- 90 ▪ Agent and CarlaBridge: The Agent manages the ROS2 connection to Autoware. At
 91 each timestep, the CarlaBridge (Kaljavesi et al., 2024) transforms CARLA snapshots
 92 and sensor data into the Autoware coordinate system. Autoware processes these inputs
 93 to issue control commands, which the Agent then applies to the ego vehicle.

Table 1: Termination Criteria of each scenario within CAWSR.

Termination Criteria	Description
Route_Completion	Agent reached the end of the route.
Actor_Blocked	Agent is blocked, not moving for 180s.
Simulation_Timeout	No client-server communication established (30s).

94 To facilitate development, we introduce a new domain model for the definition of route-based
 95 scenarios within CARLA, described in [Figure 2](#), alongside a JSON implementation. This model
 96 is based on the format introduced by Scenario Runner, facilitating support between both
 97 frameworks.

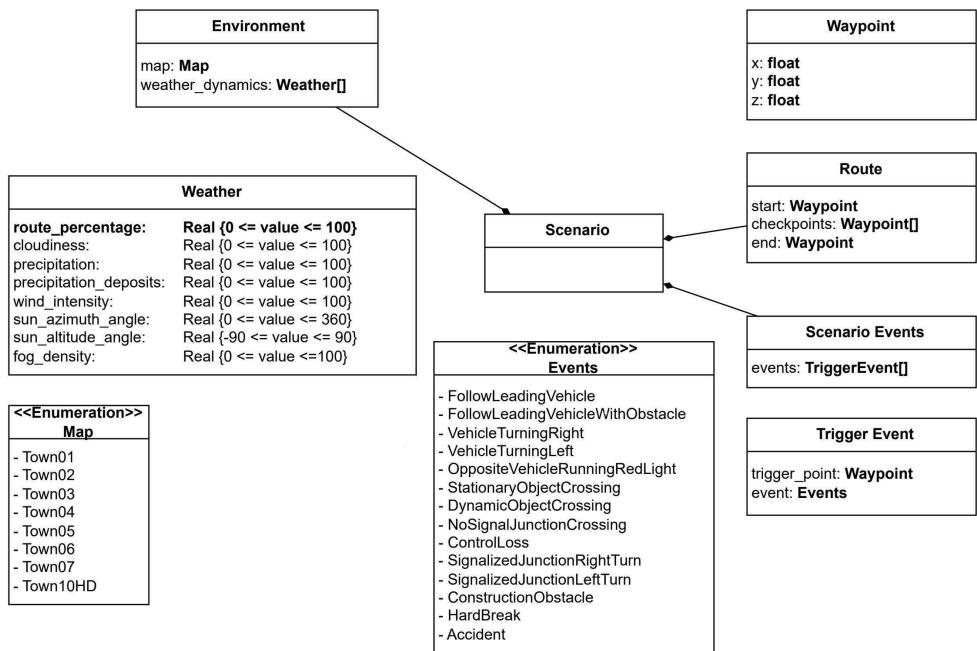


Figure 2: Scenario definition domain model.

98 Conclusion

99 To summarise, CAWSR provides ADS testing research community an easy to use Autoware
 100 evaluation pipeline. We hope that this work can facilitate the evaluation of new testing
 101 approaches on a state of the art driving system.

102 AI Usage Disclosure

103 Generative AI tools were used in this work solely to support high-level research concepts and
 104 structural ideas. All software implementation, including the source code, architecture, and
 105 deployment scripts, was authored entirely by the researchers without AI assistance.

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