

# SynCity: Synchronized Urban Traffic via AV-Infrastructure Synergy

**by**

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## Outline

- 1 Introduction
  - 2 Literature Review
  - 3 Problem Statement
  - 4 Proposed Methodology
  - 5 System Architecture
  - 6 Experimental Setup
  - 7 Conclusion
  - 8 References

# Introduction

- Rapid urbanization has led to unprecedented traffic congestion, posing significant challenges to urban mobility and sustainability.
- This project proposes a novel platform for seamless integration of Autonomous Vehicles (AVs) with smart city infrastructure to create an intelligent traffic management system.
- At the core is Vehicle-to-Infrastructure (V2I) communication, enabling real-time dialogue between AVs and central traffic control networks.
- This bidirectional data flow enables dynamic signal control, cooperative intersection management, and predictive traffic routing.
- Primary objectives: optimize traffic flow, reduce congestion, minimize travel times, enhance road safety, and decrease emissions.

# Literature Review (1/2)

| S.No | Author(s)                    | Year | Title  | Methodology / Dataset  |
|------|------------------------------|------|--|--|
| 1    | Lopez, P. A., et al.         | 2018 | Microscopic Traffic Simulation using SUMO          | SUMO architecture, network importation, TraCI interface      |
| 2    | Alvarez Lopez, F. M., et al. | 2018 | The SUMO-RL framework                              | SUMO + OpenAI Gym for Reinforcement Learning traffic control |
| 3    | Krajzewicz, D., et al.       | 2012 | Recent Development and Applications of SUMO        | SUMO applications overview, TraCI for online interaction     |
| 4    | Walraven, E., et al.         | 2016 | Multi-Agent Deep RL for Traffic Signal Control     | Deep RL for traffic lights, custom Python with SUMO          |
| 5    | Gatfor, V., et al.           | 2021 | Review on traffic datasets for ML applications     | Survey of real-world vs synthetic datasets                   |
| 6    | Every, A., et al.            | 2020 | T-Intersection: RL Environment for Traffic Control | SUMO environment for T-intersection benchmarking             |

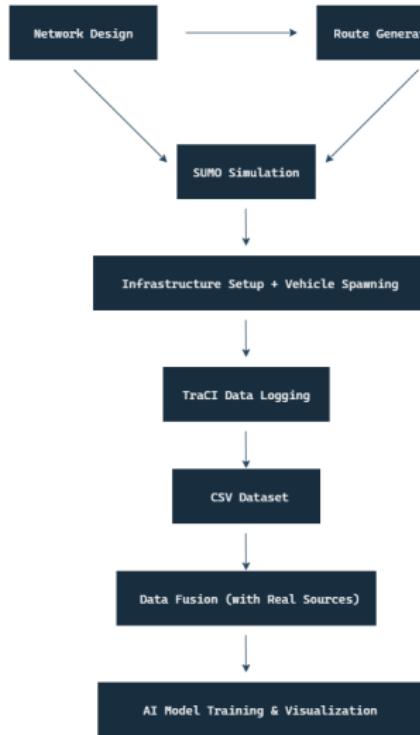
## Literature Review (2/2)

| S.No | Author(s)          | Year | Title   | Methodology / Dataset                                    |
|------|--------------------|------|---|--|
| 7    | Hayat, A., et al.  | 2023 | SUMO-based dataset for traffic signal control | Public dataset generation using SUMO for traffic control |
| 8    | Codeca, L., et al. | 2017 | Simulating Autonomous and Connected Vehicles  | SUMO with network simulators for V2X communication       |
| 9    | Mo, Y., et al.     | 2022 | Review of traffic data collection methods     | Traffic data methods review, simulation data context     |
| 10   | Polyak, B. T.      | 1964 | Speeding up convergence of iteration methods  | Foundational optimization methods for ML training        |

# Problem Statement

- Traditional traffic management systems cannot adapt dynamically to real-time traffic conditions
- Lack of synchronization between autonomous vehicles and urban infrastructure
- Inefficient traffic signal timing leads to unnecessary congestion and delays
- Limited data integration between vehicle movements and infrastructure control
- High emissions and fuel consumption due to stop-and-go traffic patterns
- Need for scalable solutions that can handle increasing urban vehicle density

# Proposed Methodology – Diagram



# Proposed Methodology – Explanation

- **Simulation Setup:** Network Design → Route Generation → SUMO Simulation
- **Simulation Execution:** Infrastructure Setup and Vehicle Spawning within simulation
- **Data Collection:** TraCI Data Logging → CSV Dataset generation
- **Enhancement & Utilization:** Data Fusion with real sources → AI Model Training & Visualization
- Creates ML-ready datasets with synchronized vehicle-infrastructure data
- Enables training of predictive models and reinforcement learning agents

# System Architecture – Explanation

- **Simulation Environment:** SUMO (Simulation of Urban Mobility) as core traffic simulator
- **Control & Data Interface:** TraCI API in Python for real-time simulation control
- **Data Storage:** CSV/JSON files storing time-series traffic data
- **Data Processing:** Preprocessing stage for data cleaning and normalization
- **ML Models:** For prediction and reinforcement learning tasks
- **Visualization:** Matplotlib, Seaborn, Grafana, or Plotly Dash for analytics

# System Architecture – Diagram



# Experimental Setup

- **Hardware:** Standard computer setup (SUMO and Python are lightweight and scalable)
- **Software:**
  - SUMO (Simulation of Urban Mobility)
  - Python with TraCI API
  - Matplotlib/Seaborn for visualization
  - Grafana or Plotly Dash for advanced dashboards
- **Datasets:**
  - Primary output: `synCity_dataset.csv`
  - ML-ready time-series data with vehicle-infrastructure synchronization
- **Evaluation Metrics:**
  - Traffic flow improvement (delay reduction, throughput)
  - ML model accuracy and RL performance
  - Dataset completeness and real-world pattern similarity

# Conclusion

- Developed SynCity framework for synchronized urban traffic management using AV-infrastructure synergy
- Created comprehensive simulation environment using SUMO with real-time data collection via TraCI
- Generated ML-ready datasets enabling training of predictive models and reinforcement learning agents
- Established foundation for dynamic traffic signal control and cooperative intersection management
- Future work: Integration with real-world traffic data, multi-city scalability testing, and enhanced V2X communication protocols
- Potential impact: Reduced congestion, improved traffic flow, lower emissions, and enhanced urban mobility

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# Thank You!

## Team SynCity