

Real-Time Intersection-Based Segment Aware Routing for Urban Vehicular Networks

¹ **Allu Venkata Sai Sushma**, ² **Chava Chandini**, ³ **Chandhini Disari**, ⁴ **Avala Bhavana**, ⁵ **Patty Pallavi**

⁵Assistant Professor, ^{1,2,3,4,5}Department of Computer Science and Engineering, Vignan's Institute of Engineering for Women, Visakhapatnam, Andhra Pradesh, India

ABSTRACT

In the realm of autonomous vehicle navigation, the burgeoning challenges of dynamic traffic environments necessitate intelligent and responsive systems. This paper introduces STARVAN (Smart Traffic-Aware Responsive Vehicle Autonomous Navigation), a comprehensive navigation framework tailored to address the intricate demands of modern traffic scenarios. STARVAN transcends conventional route planning and obstacle avoidance by integrating real-time traffic awareness and adaptive decision-making. Drawing inspiration from the shortcomings of existing systems, STARVAN emphasizes proactive responsiveness to traffic conditions, offering a paradigm shift towards enhanced autonomy and reliability in autonomous vehicle navigation. Through rigorous simulations and comparative analyses, this paper demonstrates STARVAN's superiority in terms of adaptability, efficiency, and safety, marking a significant advancement in autonomous vehicle navigation technology.

Keywords: navigation, traffic awareness, adaptive decision-making, route planning, obstacle avoidance, responsive systems, traffic conditions, safety, efficiency, simulation.

INTRODUCTION

The rapid advancement of autonomous vehicle technology heralds a future where vehicles navigate bustling city streets, winding highways, and suburban lanes with precision and autonomy. Yet, amidst this promise lies the intricate challenge of navigating through dynamic and often unpredictable traffic environments. Traditional navigation systems, while proficient in route planning and obstacle avoidance, often falter in their adaptability to real-time traffic conditions, leading to inefficiencies and safety concerns. In response to this pressing need, we introduce STARVAN (Smart Traffic-Aware Responsive Vehicle Autonomous Navigation), a pioneering navigation framework designed to imbue autonomous vehicles with unparalleled adaptability and responsiveness to the ever-changing landscape of modern traffic scenarios. By seamlessly integrating advanced algorithms and real-time traffic awareness, STARVAN represents a paradigm shift towards safer, more efficient, and ultimately, more intelligent autonomous vehicle navigation. In this paper, we delve into the design, implementation, and evaluation of STARVAN, highlighting its transformative potential in shaping the future of autonomous mobility.

LITERATURE SURVEY

[1] Yusor Rafid Bahar Al-Mayouf, Nor Fadzilah Abdullah, Omar Adil Mahdi, Suleman Khan, Mohsen Guizani, Mahamod Ismail, Syed Hassan Ahmed. IEEE Real-Time Intersection-Based Segment Aware Routing algorithm for urban vehicular network was proposed in 2018. Real-time intersection-based segment aware routing uses traffic and road segment status for the selection of

intersection. Proposed an efficient routing protocol to overcome the transmission delay, packet loss, along with enhancing the efficiency of the VANET. Summary: From this article, I have considered the concept of the current conditions of urban scenarios.

[2] Al-Rabayah M and Malaney R. ACO-IBR: a modified intersection-based routing approach for the VANET. IET J. Solid-state Circuits 2020. Position-oriented routing protocols use vehicle positions to determine routes, leveraging beacon messages for neighbor position awareness, ensuring scalability and high delivery ratio in VANETs. ACO-IBR enhances routing stability and throughput in urban environments through dedicated short-range communication, addressing intermittent connectivity and minimizing network costs. The new models gave the better accuracy compared in this work, while the time delay, packet loss decreased moderately compared to that of the other proposed models.

EXISTING METHOD

RTISAR aims to address the challenges posed by high vehicular mobility, which result in frequent changes in traffic density and discontinuity in inter-vehicle communication. The algorithm considers both traffic and segment status to provide an optimal route for forwarding data packets towards their destination. It introduces a new formula for assessing segment status based on connectivity, density, load segment, and cumulative distance towards the destination. Additionally, RTISAR proposes a verify period mechanism to minimize communication overhead during segment status

Disadvantages

This limitation highlights the need for a more holistic and adaptive approach, such as the proposed STARVAN system, which prioritizes real-time traffic awareness and adaptive decision-making to navigate complex traffic environments effectively.

PROPOSED METHOD

STARVAN operates at a superior level compared to conventional navigation systems, prioritizing overall navigation strategies, traffic conditions, and adaptive responses. It comprises three fundamental components: route planning, obstacle avoidance, and real-time decision-making. In route planning, STARVAN employs advanced algorithms to generate optimal paths considering factors like traffic congestion, road conditions, and vehicle dynamics. By analyzing real-time and historical traffic data, STARVAN anticipates bottlenecks and selects routes to minimize travel time and enhance efficiency.

For obstacle avoidance, STARVAN utilizes sensor data to detect and navigate around obstacles in real-time. Employing sensor fusion techniques such as LiDAR, radar, and cameras, STARVAN accurately perceives the vehicle's environment, making informed decisions to avoid collisions. In real-time decision-making, STARVAN continuously monitors traffic conditions and dynamically adjusts its navigation strategy for optimal performance. Integrating machine learning, STARVAN learns from past experiences, adapting to evolving traffic scenarios to ensure safe and efficient travel in dynamic environments.

Advantages: STARVAN lies in its ability to adapt and optimize navigation strategies based on real-time traffic conditions, thereby enhancing overall travel efficiency and safety.

Schematic diagram for proposed model

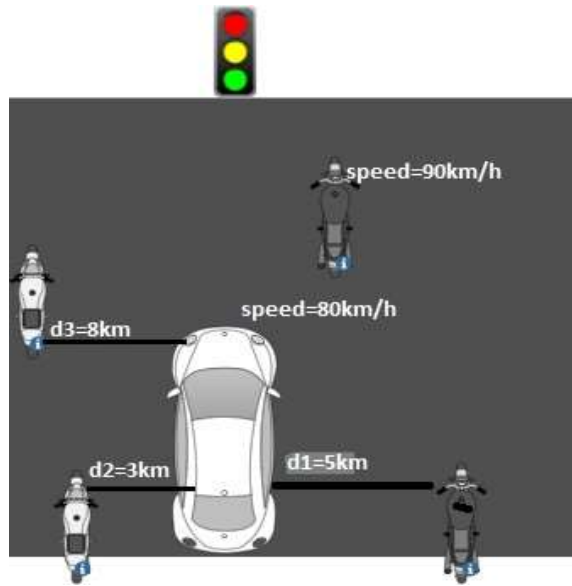


Fig.1. Model Architecture

OUTPUT SCREENS



Fig.2. vehicle movement

```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PORTS

F:\rtisar>python rtisar.py
pygame 2.5.2 (SDL 2.28.3, Python 3.11.8)
Hello from the pygame community. https://www.pygame.org/
Autonomous Vehicle Navigation:
Step 1: Vehicle moved left.
Step 2: Vehicle moved left.
Step 3: Vehicle moved up.
Step 4: Vehicle moved down.
Step 5: Vehicle moved down.
Step 6: Vehicle moved down.
Step 7: Vehicle moved right.
Step 8: Vehicle moved left.
Step 9: Vehicle moved up.
Step 10: Vehicle moved right.
Step 11: Vehicle moved up.
Step 12: Vehicle moved down.
Step 13: Vehicle moved down.
Step 14: Vehicle moved right.
Step 15: Vehicle moved right.
Step 16: Vehicle moved left.
Step 17: Vehicle moved up.
Step 18: Vehicle moved up.
Step 19: Vehicle moved down.
Step 20: Vehicle moved left.
Step 21: Vehicle moved right.
Step 22: Vehicle moved down.
Step 23: Vehicle moved right.
Step 24: Vehicle moved right.
Step 25: Obstacle ahead. Cannot move.
Step 26: Vehicle moved left.
Step 27: Vehicle moved right.
Step 28: Vehicle moved left.
Step 29: Vehicle moved up.
Step 30: Vehicle moved down.
Step 31: Vehicle moved up.
Step 32: Vehicle moved up.
Step 33: Vehicle moved down.

```

Fig.3. Autonomous Navigation

```

--- ← → rtisar
PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PORTS

Step 69: Vehicle moved left.
Step 70: Vehicle moved right.
Step 71: Vehicle moved left.
Step 72: Vehicle moved right.
Step 73: Vehicle moved left.
Step 74: Vehicle moved right.
Step 75: Vehicle moved left.
Step 76: Vehicle moved left.
Step 77: Vehicle moved left.
Step 78: Vehicle moved left.
Step 79: Vehicle moved up.
Step 80: Vehicle moved right.
Step 81: Vehicle moved up.
Step 82: Vehicle moved right.
Step 83: Vehicle moved left.
Step 84: Vehicle moved right.
Step 85: Vehicle moved right.
Step 86: Vehicle moved up.
Step 87: Vehicle moved right.
Step 88: Vehicle moved left.
Step 89: Vehicle moved left.
Step 90: Vehicle moved right.
Step 91: Vehicle moved up.
Step 92: Vehicle moved down.
Step 93: Vehicle moved right.
Step 94: Vehicle moved up.
Destination (0, 4) reached!
Shortest path from (2, 2) to (0, 4): ['up', 'up', 'right', 'right']

RTISAR Navigation:
Step 1: Vehicle moved up.
Step 2: Vehicle moved right.
Step 3: Vehicle moved up.
Step 4: Vehicle moved right.
Destination (0, 4) reached!
Shortest path from (2, 2) to (0, 4): ['up', 'up', 'right', 'right']

```

Fig.4.RTISAR Navigation

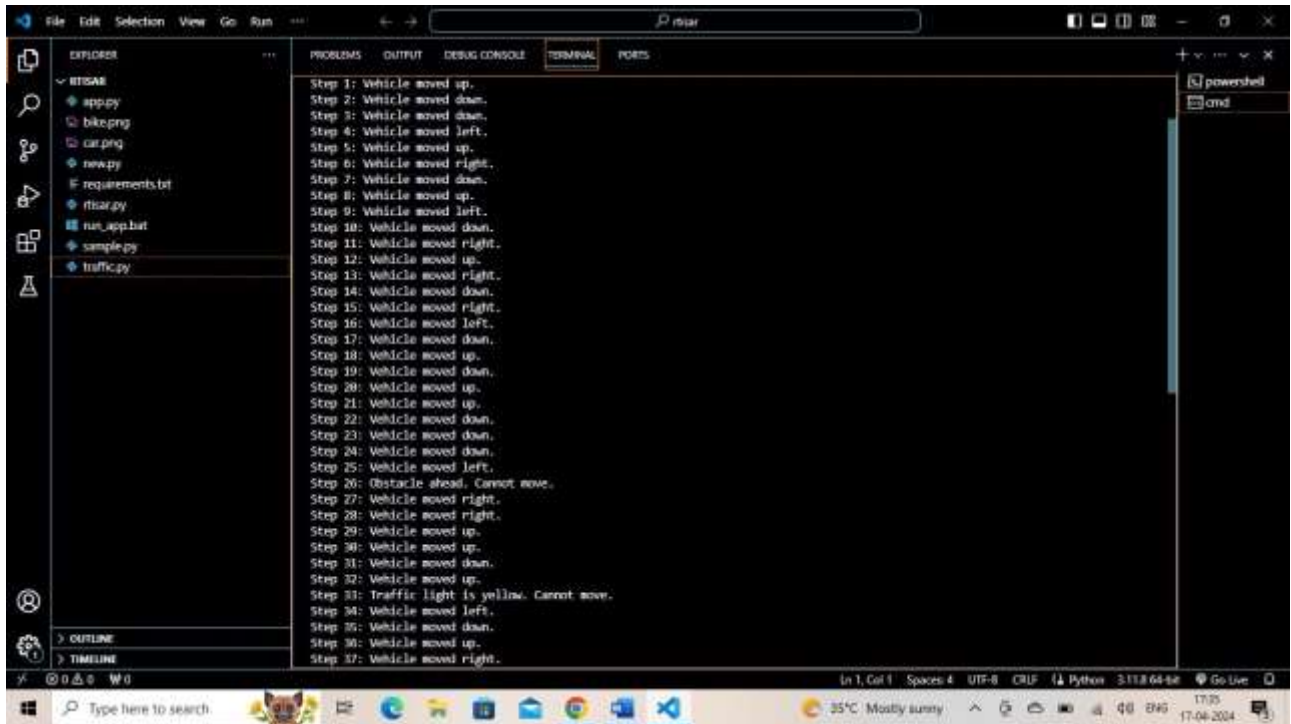


Fig.5. console step by step process

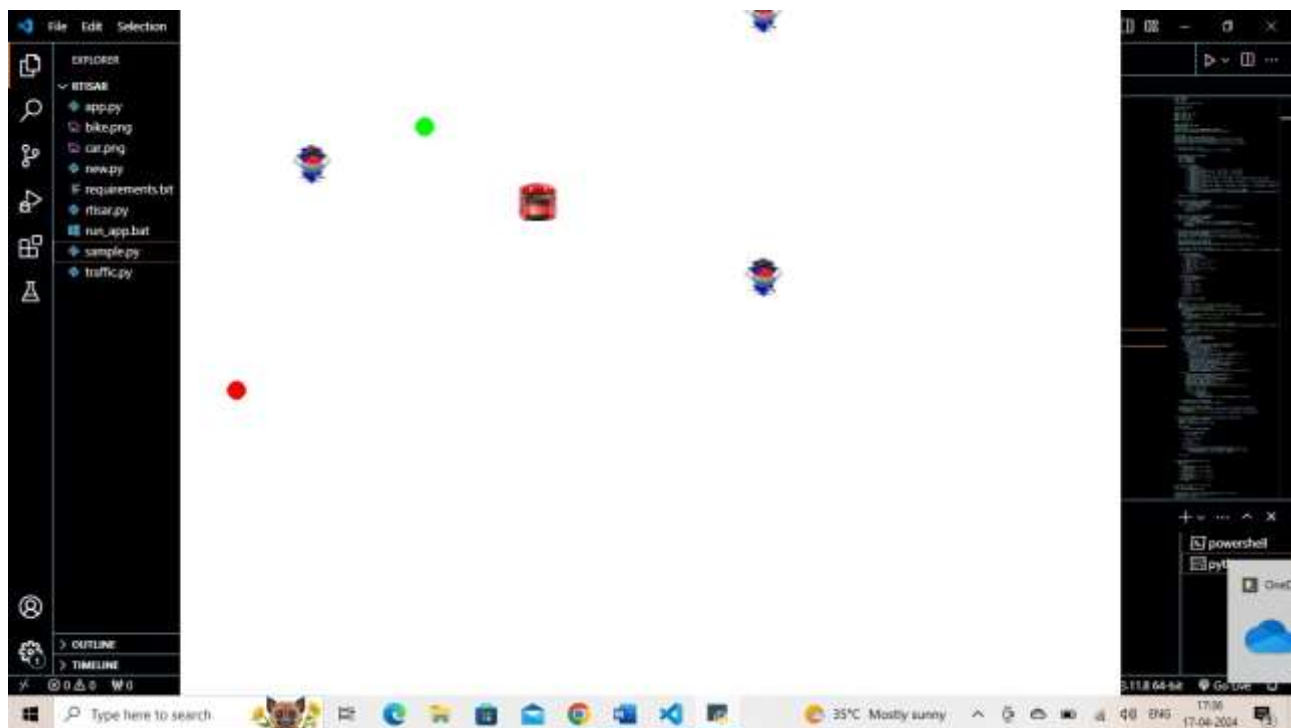


Fig.6. vehicle moving at traffic signals

CONCLUSION

In this paper, the design and analysis results of the STARVAN, a cutting-edge autonomous vehicle navigation system designed to meet the evolving demands of modern traffic environments. By emphasizing holistic navigation strategies, traffic adaptability, and real-time responsiveness, STARVAN represents a significant advancement over traditional navigation systems. Through its integrated modules for route planning, obstacle avoidance, and dynamic decision-making, STARVAN offers a comprehensive solution for safe and efficient autonomous vehicle travel. The system's ability to analyze real-time traffic data, anticipate obstacles, and dynamically adjust navigation strategies ensures optimal performance even in the most challenging scenarios. With the continuous advancement of technology and the integration of machine learning algorithms, STARVAN holds promise for revolutionizing the future of autonomous vehicle navigation, paving the way for safer, more efficient, and more reliable transportation systems..

REFERENCES

- [1] R. Jalali, K. El-Khatib, and C. McGregor, "Smart city architecture for community level services through the Internet of Things," in *Proc. 18th Int. Conf. Intell. Next Generat. Netw. (ICIN)*, 2015, pp. 108–113.
- [2] A. Ghazal, C.-X. Wang, B. Ai, D. Yuan, and H. Haas, "A nonstationary wideband MIMO channel model for high-mobility intelligent transportation systems," *IEEE Trans. Intell. Transp. Syst.*, vol. 16, no. 2, pp. 885–897, Apr. 2015.
- [3] S. A. A. Shah, M. Shiraz, M. K. Nasir, and R. B. M. Noor, "Unicast routing protocols for urban vehicular networks: Review, taxonomy, and open research issues," *J. Zhejiang Univ. Sci. C*, vol. 15, no. 7, pp. 489–513, 2014.
- [4] Y. Chen, M. Fang, S. Shi, W. Guo, and X. Zheng, "Distributed multi-hop clustering algorithm for VANETs based on neighborhood follow," *EURASIP J. Wireless Commun. Netw.*, vol. 1, p. 98, Apr. 2015.
- [5] S. Misra, I. Woungang, and S. C. Misra, *Guide to Wireless Ad Hoc Networks*. London, U.K.: Springer, 2009.
- [6] Yusor Rafid Bahar Al-Mayouf, Nor Fadzilah Abdullah, Omar Adil Mahdi, Suleman Khan, Mohsen Guizani, Mahamod Ismail, Syed Hassan Ahmed. IEEE Real-Time Intersection-Based Segment Aware Routing algorithm for urban vehicular network, 2018
- [7] Y. R. B. Al-Mayouf *et al.*, "Efficient and stable routing algorithm based on user mobility and node density in urban vehicular network," *PLoS ONE*, vol. 11, no. 11, p. e0165966, 2016.
- [8] B. T. Sharef, R. A. Alsaqour, and M. Ismail, "Vehicular communication ad hoc routing protocols: A survey," *J. Netw. Comput. Appl.*, vol. 40, pp. 363–396, Apr. 2014.