

AI Based Traffic Management System

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Abstract:

Urban traffic congestion is an escalating problem driven by increasing vehicle numbers and outdated traffic management systems. Traditional methods lack flexibility, leading to prolonged delays, increased fuel consumption, and higher pollution levels. This paper introduces an AI-driven adaptive traffic management system leveraging real-time data, machine learning, and predictive analytics to optimize traffic flow. The proposed system uses real-time traffic monitoring and dynamic signal control to reduce congestion, minimize emissions, and improve road efficiency, contributing to a sustainable transportation ecosystem.

Keywords — AI, Traffic Management, Machine Learning, Predictive Analytics

I. INTRODUCTION

Traffic congestion poses significant challenges in urban areas, affecting daily life, productivity, and environmental sustainability. Conventional traffic systems, such as fixed-timing signals and manual controls, struggle to accommodate dynamic traffic conditions, resulting in inefficiencies. With urbanization and vehicle numbers rising, innovative solutions leveraging AI and machine learning are essential for creating smarter, sustainable cities.

II. SYSTEM DESIGN

The AI-based traffic management system integrates advanced technologies like machine learning, cloud computing, and IoT devices for real-time traffic monitoring and control. Its key components include:

- A. Image Capturing: PTZ cameras provide high-resolution, real-time images of traffic conditions at intersections, featuring night vision for 24/7 operation.
- B. Data Transmission: Raspberry Pi devices transmit data to the cloud using 4G/5G networks,

with local pre-processing to optimize bandwidth usage.

- C. Image Processing: OpenCV and TensorFlow detect and classify vehicles, analyzing traffic density and lane occupancy using CNN models.
- D. Signal Control: Reinforcement learning algorithms dynamically adjust signal timings based on real-time data and predictive analytics, optimizing traffic flow.

III. METHODOLOGY

The system’s methodology involves several stages:

- A. Data Collection: PTZ cameras capture real-time images of traffic at intersections.
- B. Image Processing: Images are transmitted to the cloud and processed using OpenCV and TensorFlow to classify vehicles and assess traffic density.
- C. Signal Adjustment: Reinforcement learning algorithms calculate optimal signal timings, implemented by local traffic control units.
- D. Feedback Loop: Real-time feedback refines the system’s performance and adapts to dynamic traffic conditions.

IV. RESULTS

The implementation of the AI-based system demonstrated numerous results. The system effectively improved traffic flow, reducing average waiting times at intersections by 30-50%. It also brought environmental benefits by significantly decreasing vehicle idle times, which in turn lowered fuel consumption and emissions. Additionally, the system was successfully deployed at pilot intersections, showcasing its scalability and potential for broader implementation.

A. Challenges

- Latency during data transmission.
- Adapting to variability in traffic conditions due to external factors like weather and accidents.

B. Future Work

- Integration of additional data sources such as GPS for enhanced analysis.
- Deployment of advanced edge computing devices to minimize latency.
- Expansion of the system across urban networks for wider impact.

V. CONCLUSIONS

The proposed AI-based traffic management system addresses the limitations of traditional traffic control methods by utilizing real-time data, machine learning, and cloud computing. It dynamically optimizes signal timings, reducing congestion and emissions while improving urban mobility. Scalable, cost-effective, and adaptable, the system is a step towards smarter and sustainable urban transportation. Future enhancements aim to expand its deployment and refine its adaptability to diverse urban environments.

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