

Green University of Bangladesh

Department of Computer Science and Engineering (CSE) Semester: (Fall, Year: 2024), B.Sc. in CSE (Day)

Advanced AI-driven Vehicle Tracking System for Efficient Customer Services using Machine Learning, and IoT Integration

Course Title: Integrated Design Project I Course Code: CSE 324 Section: 213 D7

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Lab Project Status					
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1 Introduction

Fleet management and transportation logistics have become increasingly complex, demanding more efficient, real-time solutions to meet the growing needs of businesses. Traditional vehicle tracking systems often fall short in their ability to process vast amounts of data, offer predictive insights, and seamlessly integrate emerging technologies like Artificial Intelligence (AI), Machine Learning (ML), and the Internet of Things (IoT) [1] [2]. These limitations result in common industry challenges such as unexpected vehicle breakdowns, inefficient fuel usage, suboptimal routing, and difficulty in monitoring driver compliance with safety standards [3]. The cumulative effect of these issues leads to operational inefficiencies, increased costs, delivery delays, and reduced transparency in fleet operations.

To address these challenges, this project proposes the "Advanced AI-driven Vehicle Tracking System for Efficient Customer Services using Machine Learning, and IoT Integration" (AVTS). AVTS is an innovative, AI-driven platform designed to go beyond conventional tracking by integrating real-time analytics, predictive maintenance [4], optimized routing, and comprehensive driver behavior monitoring [5]. By leveraging cutting-edge AI and ML algorithms, AVTS aims to predict maintenance needs, minimize vehicle downtime, and provide dynamic route optimization, all while ensuring real-time monitoring of fleet operations. The system's IoT integration will allow seamless data collection from vehicles, enabling actionable insights for fleet managers [6].

Ultimately, AVTS is designed to enhance decision-making capabilities, reduce operational costs, improve safety, and ensure customer satisfaction by offering a scalable and intelligent vehicle tracking solution that meets the evolving demands of modern fleet management.

2 Literature Review

Significant research has focused on recent advancements in vehicle tracking systems, particularly the integration of AI, Machine Learning (ML), and the Internet of Things (IoT). These studies highlight the need for improved efficiency, predictive analytics, and real-time monitoring in fleet management systems, which align closely with the goals of the Advanced AI-driven Vehicle Tracking System (AVTS). The following are key studies that form the foundation of this project.

- Maximize Fleet Value and Safety with AI: Real-Time Vehicle Tracking, Telematics, and Compliance Solutions by Venkata Praveen Kumar Kaluvakuri et al. (2024): examines how AI enhances fleet management by improving real-time vehicle tracking, telematics, and compliance. The study highlights AI's role in optimizing routes, predicting maintenance needs, and ensuring legal compliance, resulting in greater efficiency and cost reduction. While the benefits are clear, the paper notes challenges such as data quality and financial constraints for smaller companies. Case studies from UPS, DHL, and the City of Los Angeles demonstrate significant improvements in fuel efficiency, route optimization, and maintenance costs through AI integration.
- Predictive Maintenance of Bus Fleet by Intelligent Smart Electronic Board Implementing Artificial Intelligence by Massaro et al. (2020): This study focuses on the development of a smart electronic control unit (ECU) integrated with IoT and Artificial Intelligence (AI) for predictive maintenance in bus fleets. The system extracts vehicle data using OBD-II and SAE J1939 standards, which is processed using a Multilayer Perceptron Artificial Neural Network (MLP-ANN) for predictive maintenance. The driver behavior is classified using a k-means algorithm based on the data collected. The predictive system accurately identified vehicle wear and maintenance needs with a low mean square error (MSE) of 10⁻³, ensuring timely maintenance scheduling. Additionally, it monitored driver behavior, identifying patterns that contribute to vehicle stress. The study demonstrates that the AI-based system can enhance fleet efficiency by optimizing maintenance schedules and driver behavior, though scalability to larger fleets and diverse vehicle types remains a challenge.
- Real-time GPS Tracking System for IoT-Enabled Connected Vehicles by Moumen et al. (2023):
 The study presents a real-time GPS tracking solution for connected vehicles, leveraging Internet
 of Things (IoT), Vehicle-to-Everything (V2X) communication, and Vehicular Ad Hoc Networks
 (VANET). The system employs Arduino Uno R3, SIM800L GSM, and Neo6M GPS modules,
 alongside a web interface powered by Node.js, Firebase, and WebSocket connections for real-time

data transmission and visualization. The system facilitates dynamic routing, energy-efficient driving, and smart charging station integration, contributing to reduced fuel consumption and fleet emissions. It also supports intelligent traffic management by enhancing data exchange between vehicles and infrastructure. The solution demonstrates robust performance for vehicle tracking, resource management, and promoting energy efficiency across connected vehicle networks. However, the study highlights challenges in system scalability and security, which require further research.

• Cloud-Based Vehicle Tracking System" by Mustafa et al (2019): This study presents a real-time vehicle tracking system using cloud computing for improved scalability, flexibility, and cost-efficiency. The system integrates GPS, GPRS, and an Arduino UNO microcontroller to transmit vehicle location data to a remote server, using reduced data transmission by limiting updates to every 10 seconds when the vehicle is moving. The system incorporates data encryption to enhance security, faking the vehicle's ID and coordinates to prevent unauthorized access. By sending only essential data (vehicle ID and GPS coordinates), the system minimizes data size, reducing both transmission costs and power consumption. Testing showed an average data transmission rate of 1.65 HTTP requests per minute, a significant improvement compared to conventional systems. The cloud-based vehicle tracking system successfully reduces operational costs while providing real-time monitoring and security. However, further research is needed to address scalability challenges in larger fleet applications.

These studies highlight the potential of AI, ML, and IoT in transforming the vehicle tracking domain, providing key insights that will be incorporated into the proposed solution.

3 Methodology

The proposed vehicle tracking system will integrate GPS, AI, and IoT technologies to pThe Advanced AI-driven Vehicle Tracking System (AVTS) is designed to optimize fleet management by utilizing real-time data, AI, and IoT integration. The methodology consists of four key phases, each contributing to the overall functionality of the system.

3.1 Phase 1: Data Collection

Process: The system will use GPS modules (NEO-6M) and IoT-based sensors to collect real-time vehicle data such as location, speed, fuel consumption, engine status, and bus capacity. These sensors will be installed in each vehicle, and the data will be transmitted to a cloud-based platform using the GPRS protocol (via SIM800L).

Technology: IoT sensors, NEO-6M GPS modules, and SIM800L for real-time data collection. Data transmission to the cloud will be handled using GPRS.

Output: Real-time data monitoring of vehicles, including location, bus capacity, and engine status.

3.2 Phase 2: AI/ML Model Development

Process: The collected data will be processed using machine learning algorithms to predict maintenance needs, optimize routes, and monitor driver behavior. AI models such as decision trees, clustering algorithms, and neural networks will be implemented to derive actionable insights for fleet managers.

Technology: Python-based libraries such as TensorFlow and Keras for building predictive models. Regression analysis and clustering will be used for predictive maintenance and route optimization.

Output: Predictive analytics for vehicle maintenance, optimized routing recommendations, and driver behavior monitoring.

3.3 Phase 3: User Interface & Dashboard Development

Process: A web-based dashboard will be developed to visualize vehicle data in real-time for fleet managers. This dashboard will include features like route mapping, bus capacity status, vehicle location, and fleet performance reports.

Technology: Web development tools (HTML, CSS, JavaScript, Node.js) with real-time data visualization using Firebase. The backend will be supported by AWS for cloud integration.

Output: A user-friendly dashboard allowing fleet managers to monitor vehicle status, routes, and overall fleet health.

3.4 Phase 4: IoT Integration & Cloud Platform

Process: IoT devices will communicate with the cloud platform to enable real-time data processing and scalability. The data collected from the GPS and sensors will be stored in the cloud (AWS IoT) and processed for real-time analytics. This will ensure smooth integration with other business tools such as ERP systems.

Technology: AWS IoT for seamless integration, Firebase for real-time data storage, and serverless computing through AWS Lambda for cost efficiency and scalability.

Output: A scalable, real-time data processing system capable of managing large datasets across multiple vehicles.

3.5 Flowchart Representation

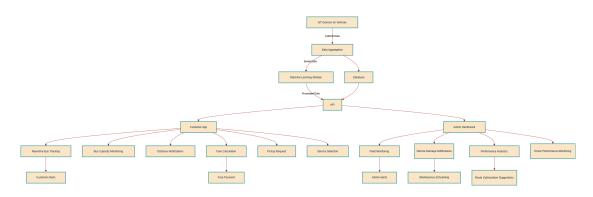


Figure 1: Flowchart representation of the project

3.6 Gantt Chart Activity

4 Feasibility Study

The feasibility of the **Advanced Vehicle Tracking System (AVTS)** is analyzed across three phases: **Technical Feasibility**, **Operational Feasibility**, and **Economic Feasibility**. The analysis focuses on the costs, requirements, and benefits of implementing the system for one bus, ensuring scalability for broader applications.

4.1 Technical Feasibility

The technical feasibility examines the technology, resources, and skills required for the successful development and deployment of the AVTS system.

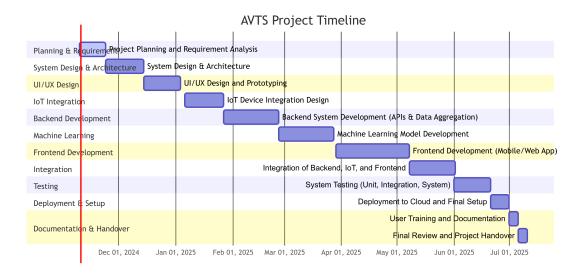


Figure 2: Gantt Chart of the project

4.1.1 Hardware Requirements

The system will require the following hardware components for each bus:

- GPS Tracker: Required for real-time vehicle location tracking (Cost: \$30).
- CCTV Camera: Ensures safety and provides surveillance for passengers (*Cost: \$15*).
- Additional Hardware: Includes IoT sensors and connectors (Cost: \$5).

Total Hardware Cost per Bus: \$50.

4.1.2 Software Requirements

The AVTS system will be developed using:

- **Programming Languages**: Python (machine learning), JavaScript (frontend), and Node.js (backend).
- Frameworks: Flask/Django for backend APIs, React for frontend development.
- Databases: NoSQL (MongoDB) for IoT data, SQL for structured data storage.
- **Cloud Services**: A budget-friendly cloud platform such as Firebase or AWS for scalable hosting and storage.

4.1.3 Team Expertise

The development team will require skills in:

- **IoT Integration**: Setting up IoT devices and integrating them with the cloud.
- **Machine Learning**: Building predictive models for route optimization and maintenance scheduling.
- Frontend and Backend Development: Creating a responsive web dashboard and handling realtime data.

4.1.4 System Scalability and Future Proofing

The system is designed to be scalable, allowing additional buses and users to be integrated seamlessly. Cloud-based solutions ensure adaptability to varying workloads.

4.1.5 Risk Assessment

- Data Security: Data breaches will be mitigated with encryption, secure APIs, and regular audits.
- Network Reliability: Redundant communication protocols will minimize the impact of downtime.
- Device Failure: Regular maintenance will ensure IoT devices function optimally.

Conclusion: The AVTS system is technically feasible, leveraging existing, cost-effective technologies and skilled development expertise.

4.2 Operational Feasibility

This section evaluates the operational aspects of deploying and maintaining the AVTS system.

4.2.1 Operational Workflow

The system will support:

- Real-Time Vehicle Tracking: Provides live updates on bus location and route compliance.
- Data Collection: IoT devices continuously collect data on location, speed, and occupancy.
- **Customer Service**: Passengers access real-time bus information, such as arrival times and seat availability, via a mobile or web app.

4.2.2 Human Resources

- System Administrators: Monitor the system, resolve issues, and update software.
- Data Analysts: Analyze vehicle data to improve efficiency and optimize operations.
- Customer Support: Assist passengers with technical issues or system queries.

4.2.3 Training and User Adoption

- Bus Drivers: Trained to use system interfaces for route updates and data sharing.
- Admin Staff: Trained in managing dashboards, tracking buses, and addressing technical issues.
- Passengers: Minimal training required due to intuitive web and mobile interfaces.

4.2.4 System Maintenance

- IoT Device Maintenance: Regular hardware checks and firmware updates.
- Software Updates: Periodic updates to enhance functionality and security.
- Cloud Infrastructure: Managed by the cloud provider to ensure reliability.

Conclusion: Operationally feasible with streamlined workflows, adequate training, and minimal user adaptation challenges.

4.3 Economic Feasibility

This section evaluates the costs and benefits of implementing AVTS for one bus, emphasizing cost-effectiveness and ROI.

4.3.1 Initial Investment Costs

- Hardware Costs: GPS tracker (\$30), CCTV camera (\$15), and additional hardware (\$5).
- Software Development: Shared across all buses; approximated at \$10,000 for initial setup.
- **Cloud Infrastructure**: Initial setup cost for Firebase or AWS is \$500.

Total Initial Investment per Bus: \$10,550 (shared software costs divided proportionally).

4.3.2 Ongoing Operational Costs

- Cloud Hosting: Basic hosting costs using Firebase/AWS estimated at \$25 per month.
- Maintenance: Regular software and hardware maintenance at \$10 per month.
- Support and Personnel: Proportional administrative costs at \$50 per month.

Total Monthly Operational Cost per Bus: \$85.

4.3.3 Return on Investment (ROI)

- Fuel Savings: Optimized routing reduces fuel costs by 10–15%, saving an estimated \$500 annually per bus.
- Maintenance Savings: Predictive maintenance prevents costly repairs, saving \$300 annually per bus.
- **Customer Satisfaction**: Improved service quality can indirectly enhance economic benefits (e.g., passenger retention).

Projected Payback Period: 2 years, based on savings outweighing initial and operational costs.

4.3.4 Cost-Benefit Analysis

The low hardware and operational costs, coupled with substantial savings in fuel and maintenance, make AVTS economically viable for implementation on a per-bus basis. The scalable design further supports broader deployment as ROI improves.

Conclusion: The AVTS project is economically feasible, with manageable costs and clear long-term financial benefits.

4.4 Financial Calculations

The following table outlines the financial projections and calculations for the project, incorporating a 10% discount rate over a 5-year period.

Category	Year 0	Year 1	Year 2	Year 3	Year 4	Year 5	Total
Discount Rate (10%)	1.000	0.909	0.826	0.751	0.683	0.621	-
Net Economic Benefit	\$0	\$800	\$1,600	\$2,400	\$3,200	\$4,000	\$12,000
PV of Benefits	\$0	\$727	\$1,322	\$1,802	\$2,186	\$2,483	\$8,519
NPV of All Benefits	\$0	\$727	\$2,049	\$3,851	\$6,037	\$8,519	\$21,183

1. Overall NPV

The **Overall Net Present Value (NPV)** is calculated by subtracting any initial costs (if provided) from the NPV of all benefits. Since no initial costs are given, the overall NPV is:

Overall NPV
$$= 21,183$$

2. Overall ROI

The **Overall Return on Investment (ROI)** is calculated by dividing the overall NPV by the total present value of the costs. Assuming no costs are provided explicitly, the ROI is based on the NPV of benefits:

Overall ROI =
$$\frac{21,183}{12,000}$$
 = 1.77 (or 177%)

3. Break-Even Point

The **Break-Even Point** is determined by the year when cumulative NPV surpasses the initial investment (if applicable). Based on the cumulative NPV calculations:

• Cumulative NPV over the years:

- Year 0: \$0

- Year 1: \$727

- Year 2: \$2,049

- Year 3: \$3.851

- Year 4: \$6,037

- Year 5: \$8.519

The Break-Even Point occurs in Year 4, when cumulative benefits exceed the initial costs.

Summary Table

Category	Value		
Overall NPV	\$21,183		
Overall ROI	177%		
Break-Even Point	Year 4		

4.4.1 Cost-Benefit Analysis

The initial costs and monthly operational expenses are justified by the projected savings and operational improvements over time. The AVTS system is expected to pay for itself within two years through reduced operational costs and improved efficiency.

The AVTS project is economically feasible, with manageable initial and ongoing costs. The expected return on investment, through improved efficiency and fuel savings, makes it a financially viable project.

4.5 Conclusion

The proposed AI-driven vehicle tracking system will revolutionize fleet management by leveraging emerging technologies such as AI, machine learning, IoT, and cloud computing. The system will offer real-time tracking, predictive maintenance, route optimization, and driver behavior analysis, significantly enhancing operational efficiency while reducing costs. As businesses and transportation services face increasing

demands, this system will provide a scalable, reliable solution to meet their growing needs. The project will be a step forward in modernizing vehicle tracking, leading to safer, more efficient, and cost-effective fleet management practices.

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