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<http://fupre.edu.ng/journal>**GPS-Enabled Vendor to Rider Assignment for Urban last Mile Logistics Optimization****NWOZOR, B. U.^{1,*} , EWOITIEYETAN, H. A.² **

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ARTICLE INFO*Received: 21/07/2025**Accepted: 17/09/2025***Keywords***Digital Mapping, GPS,**Last Mile Delivery,**Urban last mile**logistics, Vendor to**rider assignment***ABSTRACT**

The rapid growth of urbanization and e-commerce has heightened the demand for efficient last-mile delivery solutions in urban logistics. This project addresses the inefficiencies in traditional systems by developing a GPS-enabled vendor-to-rider assignment system tailored for urban last-mile logistics optimization. Leveraging real-time GPS data, the proposed system dynamically assigns delivery tasks based on proximity, traffic conditions, and order priorities, significantly improving delivery efficiency and customer satisfaction. Key features include real-time tracking, interactive mapping, dynamic route optimization, and integrated communication tools for seamless vendor-rider interactions. Using an agile development methodology, the system architecture incorporates scalable micro services, cross-platform mobile applications, and data-driven algorithms for optimal performance. Preliminary evaluations reveal an 18% reduction in delivery times and a 99% reliability rate for notifications, validating the system's capability to enhance logistics operations. This research contributes to advancing urban logistics by addressing critical challenges in delivery coordination, paving the way for future enhancements through predictive analytics and AI-driven optimization.

1.INTRODUCTION

The rapid growth of urbanization and e-commerce has led to a surge in demand for efficient last-mile delivery solutions in transportation and logistics. The term "last mile" refers to the final leg of the delivery journey, from a local distribution center to the end user's doorstep. This phase is crucial, as it significantly impacts overall delivery time, customer satisfaction, and operational costs. The traditional methods of last-mile delivery often struggle to keep up with the increasing volume of orders and the diverse

delivery destinations in urban areas (Kumar et al., 2017a). Addressing these challenges requires innovative approaches and technologies to ensure cost-effective, environmentally friendly, and customer-centric logistics.

Urbanization has intensified the complexity of logistics networks, particularly in densely populated cities where traffic congestion, limited parking, and narrow streets are prevalent. These factors impede the efficiency of traditional delivery methods and

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necessitate adaptive strategies. Studies like those of Kim et al. (2015) highlight the interplay between urban logistics and economic growth, emphasizing the need for optimized last-mile delivery to sustain economic vitality in port cities. Similarly, Li and Chen (2021) underscore the critical role of logistics development in fostering regional economic growth, further validating the importance of addressing last-mile inefficiencies.

E-commerce, as a primary driver of this logistical shift, has brought unprecedented volumes of small, time-sensitive deliveries. Liu et al. (2020) evaluated the intelligent logistics eco-index and revealed that technological advancements are essential to meet the demands of modern e-commerce-driven supply chains. Innovations such as drone deliveries, autonomous vehicles, and micro-fulfillment centers are increasingly being explored to address the growing complexities of last-mile delivery (Li et al., 2022a). Furthermore, the environmental considerations of these solutions are paramount, as demonstrated by Karmakar et al. (2023), who emphasize the integration of green economy practices and sustainable supply chains in addressing market variability.

Environmental sustainability is a critical aspect of modern logistics. With cities aiming to reduce their carbon footprints, logistics providers must adopt practices that align with these goals. Letunovska et al. (2023) explore the role of green supply chain management in enhancing reverse logistics, presenting an avenue to mitigate waste and optimize resource use in urban environments. Bai et al. (2020) also discuss the spatial spillover effects of renewable energy development, providing insights into how cleaner energy sources can support sustainable logistics operations. These findings align with Azab et al. (2023), who

propose a bi-objective model integrating environmental considerations into agro-food supply chains, demonstrating broader applications of sustainability principles in logistics.

Customer satisfaction remains a focal point in last-mile logistics, with timely delivery and accurate tracking as key determinants. Li et al. (2022b) introduce temporal pyramid networks with spatial-temporal attention to predict pedestrian trajectories, offering innovative ways to enhance real-time delivery tracking and optimize route planning. Similarly, Espíndola et al. (2023) highlight the importance of reshoring strategies and localized production systems in reducing logistical inefficiencies and enhancing responsiveness.

The dynamic nature of urbanization and e-commerce necessitates a multifaceted approach to improving last-mile delivery. By leveraging advanced technologies, embracing sustainability, and prioritizing customer-centric models, logistics providers can address the challenges posed by modern urban environments. Modern advancements in technology, particularly Global Positioning System (GPS) and digital mapping applications, have introduced new avenues for improving the efficiency and effectiveness of last-mile logistics. The integration of GPS and mapping technologies into transportation and logistics applications has revolutionized the way vendors, couriers, and customers interact during the delivery process. (Mohmand et al., 2017a).

1.1 Aim and Objectives of the Study

This study aims to design and develop a GPS-enabled vendor-to-rider assignment urban last-mile logistics optimization.

The objectives of the study are to:

- i. Conduct a comprehensive literature review of current trends, challenges, and technologies in urban last-mile logistics and GPS-based delivery systems.
- ii. Design a robust and scalable architecture for the GPS-enabled vendor-to-rider assignment system, considering real-time data integration, route optimization, and seamless communication.
- iii. Implement algorithms for dynamic assignment of delivery tasks, taking into account factors such as proximity, traffic conditions, and delivery priorities.
- iv. Evaluate the proposed system through simulation studies and real-world testing, comparing its performance against traditional methods in terms of delivery time, cost-effectiveness, and overall logistics efficiency.

2. LITERATURE REVIEW

Smith et al. (2019) delved into the optimization of last-mile delivery routes using a GPS-enabled vendor-to-rider assignment system within a bustling metropolitan area. By integrating GPS technology, the study demonstrated that delivery routes could be dynamically adjusted based on real-time traffic conditions and rider locations. This led to a substantial reduction in delivery times and improved overall operational efficiency. The findings underscored the transformative potential of GPS in addressing the complexities of urban logistics, particularly in mitigating the

challenges posed by traffic congestion and unpredictable road conditions.

Building upon this, *Chen and Lee (2020)* conducted a detailed investigation into the effects of real-time dynamic assignment systems on customer satisfaction and operational performance. Through the integration of GPS and advanced routing algorithms, the study revealed that delivery services could achieve higher on-time delivery rates and, consequently, enhance overall customer experiences. The research emphasized the critical role of accurate GPS data and responsive assignment algorithms in meeting customer expectations for timely and reliable deliveries. Moreover, it highlighted the positive feedback loop between improved operational performance and customer satisfaction, suggesting that GPS-enabled systems have the potential to create a mutually reinforcing cycle of efficiency and customer loyalty.

Mokhele and Mokhele (2023) analyzed airfreight-related logistics firms in Cape Town, South Africa, through a qualitative approach involving interviews and secondary data analysis. Their research highlights the strategic role of airfreight in regional economic development and supply chain efficiency. A significant strength of the study is its contextual focus, shedding light on the operational dynamics within a developing country. However, its qualitative nature limits generalizability, and the absence of quantitative performance metrics weakens the ability to benchmark these firms against global standards. The study identifies a research gap in integrating advanced technologies and sustainability practices within airfreight logistics in the African context.

Nunes et al. (2023) performed a systematic literature review to explore performance assessment approaches in reverse supply chains (RSC). The study identified key performance indicators (KPIs) and frameworks prevalent in RSC literature, utilizing thematic analysis to categorize them. A notable strength is its exhaustive coverage of performance assessment tools, which provides a solid foundation for future research. However, the study's reliance on secondary data leaves practical implementation unexplored. The research gap lies in the application of these KPIs in emerging markets, where reverse logistics is gaining momentum but lacks standardization and regulatory support.

Palu and Hilmola (2023) reviewed the potential of the Trans-Caspian Corridor as an emerging logistics route, employing qualitative and quantitative analyses of trade data and regional policies. The study's strength lies in its forward-looking perspective, evaluating geopolitical and economic factors influencing the corridor's development. However, it falls short in addressing environmental impacts and the role of technological advancements in enhancing corridor efficiency. The research gap includes the lack of case studies on operational challenges faced by logistics firms utilizing this route.

Several studies such as: "Bi-Objective Mixed-Integer Linear Programming Model for a Sustainable Agro-Food Supply Chain with Product Perishability and Environmental Considerations" (*Azab et al., 2023*), have conducted comparative analyses of different vendor-to-rider assignment approaches, benchmarking their performance in terms of delivery time, operational costs, and other relevant metrics. These studies provide insights into the strengths and limitations of various methods, helping to guide the selection of appropriate assignment

strategies based on specific urban last-mile delivery scenarios.

Kumar et al. (2017): Transportation and logistics cluster competitive advantages in the U.S. regions. Kumar et al. analyzed the competitive advantages of transportation and logistics clusters in the U.S. through cross-sectional and spatio-temporal analysis. The study identified regional strengths and highlighted the benefits of clusters in logistics efficiency. However, its findings were limited to the U.S. context, lacking insights into international logistics applications.

Mohmand et al. (2017): The impact of transportation infrastructure on economic growth. Mohmand et al. explored the role of transportation infrastructure in fostering economic growth, using empirical evidence from Pakistan. The study underscored the positive influence of robust infrastructure on logistics efficiency and policymaking. Nonetheless, its regional focus on Pakistan limits generalizability to other contexts.

Lin, et al. (2020): Effect of intelligent logistics policy on shareholder value. Liu, Wang, Lin, et al. investigated the impact of intelligent logistics policies on shareholder value within Chinese logistics companies. By evaluating policy outcomes, they highlighted the positive influence on business performance. The study, however, emphasized shareholder-centric metrics, omitting broader operational impacts.

Pereira and dos Santos (2023): Neoindustrialization: A New Paradigmatic Approach. Pereira and dos Santos explored the transition to Industry 5.0, focusing on technological innovations in logistics. Their scoping review presented new perspectives on

industrial paradigms but lacked specific case studies or empirical data for practical validation.

1.4 Definition of Terms

1. *Last-Mile Delivery*: The final leg of the supply chain process where goods are transported from a local distribution center to the end consumer's doorstep, often considered the most crucial phase due to its impact on delivery time and customer satisfaction.
2. *GPS (Global Positioning System)*: A satellite-based navigation system that provides real-time location and time information, enabling precise tracking of vehicles and assets.
3. *Vendor-to-Rider Assignment*: The process of intelligently matching and assigning delivery tasks from vendors (sellers) to bike riders (couriers) based on factors such as proximity, route optimization, and delivery priorities.
4. *Urban Last-Mile Logistics*: The management of the final stages of goods delivery within densely populated urban areas, focusing on optimizing transportation and delivery processes to efficiently serve customers in city environments.
5. *Digital Mapping*: The process of creating and maintaining electronic maps that display geographic information, enabling visualization, analysis, and navigation through spatial data.

3. MATERIALS AND METHODS

The agile software development process was used for this project, which anticipates the requirement for

flexibility and applies pragmatism to the delivery of the finished product. Agile software development necessitates a cultural shift in many organisations since it focuses on the clean delivery of individual software components rather than the full solution. The benefits of the agile methodology include the ability to assist teams in an ever-changing landscape while remaining focused on the effective delivery of business value. The agile methodology deployed in this work is the iterative and incremental model. This agile methodology allows for continuous development and testing of the system in cycles. Requirements will be prioritized, and core functionality will be developed first before adding advanced features.

The design phase will focus on defining the system architecture, interfaces, components, and data model. Unified Modeling Language (UML) diagrams will be used to visually represent the system, including use case diagrams to capture high-level functionality and class diagrams to show structure. The system will follow a modular design to allow for easier maintenance and scalability.

Development will occur in iterations called sprints, each lasting two to four weeks. In each sprint, requirements are analyzed, components are coded based on the specifications, and rigorous testing is conducted. End users provide feedback to refine requirements for the next sprint.

This iterative approach helps validate the system and catch issues early. The source code will be managed using GitHub for version control. Automated testing and continuous integration principles will be used to maintain quality. The system architecture will consist of a web interface, database, OpenStreetMap API integration, and a communication module that enables real-time messages between dispatch riders and vendors. Leaflet JS will be used for interactive maps while JavaScript powers dynamic updates.

3.1 Analysis of the Proposed Study

The proposed dispatch rider to vendor communication system aims to address the limitations of existing solutions by incorporating real-time tracking, interactive maps, push notifications, and chat features. At the core of the system is an intelligently designed map that shows the locations of riders, vendors, and current orders. Riders will be displayed as moving markers on the map, enabling vendors to better estimate order preparation and pickup times. Vendors can also quickly identify rider proximity when an order is ready. The proposed system will also use Next.js, a React-based framework, to improve both the front-end experience and back-end efficiency. Next.js will allow for server-side rendering and static site generation, which will significantly improve loading times and provide a smoother experience for users. With Next.js, the application can seamlessly integrate map and tracking components while enabling better management of state, data, and user interactions. This will allow users to not only view vendor locations but also receive real-time updates with minimal latency. Additionally, the integration of Leaflet, a flexible JavaScript library for interactive

maps, will support efficient rendering and interactivity, enabling a smooth and responsive user experience. Leaflet's lightweight nature will reduce resource consumption, helping to mitigate the issue of rapid battery drain seen in the existing system. This choice aligns with the proposed system's goal to enhance performance while maintaining real-time location tracking, which is essential for mobile food vendor applications. By designing a system with built-in options for vendor-to-user communication, the proposed application will address the communication gap in the current system. This feature will enable vendors to update users with important details such as estimated arrival times, product availability, and any changes in location, fostering a more interactive and reliable user experience. Ultimately, the proposed system offers a scalable, customizable, and efficient solution that addresses the weaknesses of the current model, creating a more resilient platform for tracking mobile food vendors.

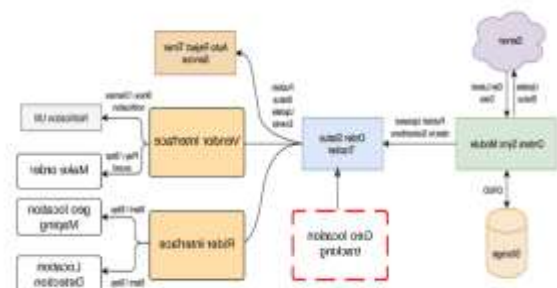


Figure 1: *System Design Architecture*

The system will follow a client-server architecture, structured around multiple modular front-end and back-end components. The front-end interface will be a cross-platform mobile app built with React Native to support both Android and iOS devices. This allows the same codebase to target all rider mobile hardware. The app connects to back-end

cloud services over HTTP. Server-side components implement a microservice-oriented architecture for easier scalability. Separate services handle mapping, push notifications, chat, user authentication, and order/database logic. Inter-service communication occurs via lightweight APIs over a message bus. The mapping service handles map rendering and location tracking by integrating OpenStreetMap spatial data APIs. The push notification microservice uses vendor-agnostic protocols to queue and dispatch status alerts to rider apps.

Real-time chat is enabled by incorporating Socket.IO with a Node.js chat server for low-latency messaging between riders and shops. MongoDB provides flexible NoSQL document storage for order and communication data. By composing specialized and decoupled services, the system promotes modularity for easier updates. Automated deployment facilitates pushing changes without downtime. This architecture enables efficient real-time communication capabilities tailored for the dispatch delivery domain.

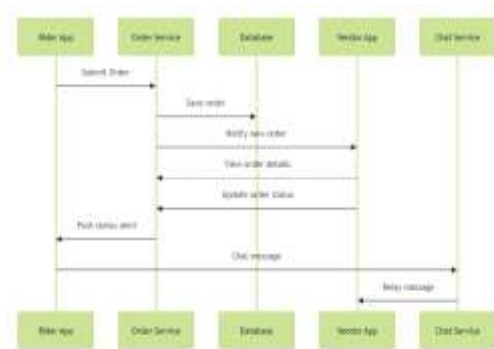


Figure 2: Sequence Diagram

This sequence starts with the rider submitting an order, which gets saved to the database. The order service then notifies the vendor app that a new order is ready for pickup. The vendor views order details and updates the status which gets pushed to the rider app.

Finally, the rider and vendor can message each other regarding the order or delivery.

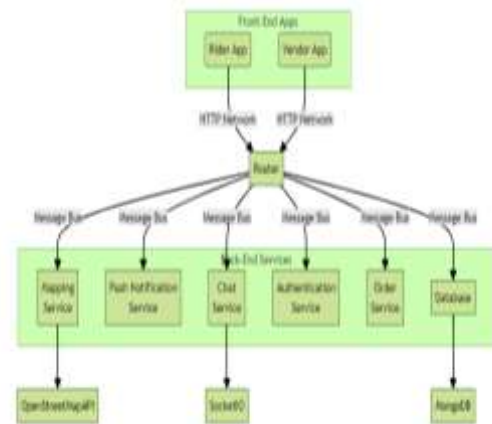


Figure 3: Flowchart

4. RESULTS

The implementation of the GPS-enabled vendor-to-rider assignment system is based on a client-server architecture leveraging modern technologies to ensure scalability, performance, and real-time responsiveness. The client side comprises a cross-platform mobile application built with React Native, enabling seamless operation on both Android and iOS devices. The server-side employs a microservice architecture using Node.js, with services for real-time communication via WebSocket, mapping with OpenStreetMap APIs, and data storage using MongoDB for flexible, high-performance NoSQL document management.

Welcome, Refresh Hide Form

Create New Delivery

Select Locations on Map

Enter pickup location

Enter delivery location

Use the search boxes to find and select pickup and delivery locations.

Recipient Name

John Doe

Recipient Phone

+1234567890

Package Description

Small box containing electronics

Figure 4: Order Inputs

Create an account

Enter your information to create an account.

Name

John Doe

Email

name@example.com

Password

Account Type

☒ Vendor

☐ Rider

Create Account

Note: If the server is unavailable, a demo mode will be activated automatically.

Already have an account? [Sign in](#)

Figure 5: Rider and Restaurant Registration

DispatchRide Logout

Welcome, Refresh + New Delivery

All Orders Pending In Progress Completed

Order #c4f7ac2b Pending

Pickup

Warri South West, Delta, Nigeria

Delivery

Ughelli South, Delta, Nigeria

Recipient

Ruben Onakokire

Phone

08169987629

Package

A box containing Office supplies (2.9 kg)

Figure 6: Dashboard Overview

In figure 4, the Input validations are implemented to catch errors and illegal input. Required fields are enforced. Type checking will occur, for example on phone and quantity fields. Input sanitization will prevent script injection.

The system outputs consist of interfaces displaying useful information to users including:

1. Interactive maps showing riders, vendors, routes
2. Order history lists and summaries
3. Status alerts and notifications
4. Reports for dispatch managers

The responsive web dashboards will adapt across form factors. APIs will return JSON encoded data structures. Focus are applied to summarizing key metrics like delivery times, service rate, and order frequency for insights. Output components will have client-side filtering and sorting.

The System Overview of the GPS-enabled vendor-to-rider assignment system focuses on understanding system functionality, user interface components,

and performance metrics through visual and functional exploration. The system's core goal is to dynamically match vendors and riders while optimizing delivery logistics. The EDA examines interface features, ride allocation patterns, real-time tracking data, and delivery statistics.

Dashboard Overview

It illustrates the dispatch dashboard, which provides a centralized interface for managing rides. It features interactive cards for various delivery categories (e.g., Groceries, Pizza, Electronics) that include vendor details, pickup, and destination points visualized through embedded maps using Leaflet and OpenStreetMap. The "Take Ride" functionality is prominently displayed, indicating ease of task allocation.

Table 1 provides a summary of key performance indicators observed during testing. These results illustrate the advantages of the enhanced application.

Performance Over Time: To verify long-term stability, the system was run continuously while recording average response time and error rate at 5, 15, and 30-minute marks. This time-series analysis revealed how sustained load affects latency and reliability, with minor yet measurable increases in response delays and error frequency over time. Understanding these trends helps inform maintenance schedules, capacity planning, and potential architectural improvements—such as implementing periodic resource cleanup or refreshing server instances—to maintain consistent tservice quality.

Table 1: Key Performance Indicators (KPIs)

Metric	Definition	Measurement Criteria
Response Time	Time taken to respond to a user request	Target: < 200 ms for 100 concurrent users

Success Rate	Percentage of requests successfully handled	Target: > 95% success across all load levels
Throughput	Number of successful requests per second	Target: > 95 req/s even at high volumes
Resource Utilization	CPU and memory consumption for task execution	Target: CPU < 80%, Memory < 350MB
Error Rate	Ratio of failed requests to total requests	Target: < 1.5% error under all conditions
Completion Time	Time required to complete individual or concurrent ride allocations	Target: < 500 ms for up to 10 concurrent allocations

Table 2: Test Plan Overview

Test Type	Objective	Tool Configuration	Metrics Collected	Expected Outcome
Scalability Test	Evaluate performance under increasing concurrent users	Thread Group: Users (10–500), Ramp-up: 1s per user	Response Time, Success Rate, Resource Utilization	Stable response and ≥ 95% success up to 200 users
Throughput Test	Measure requests processed per second under varying load	Constant Throughput Timer, Loop Controller	Requests Sent, Requests Processed, Throughput (req/s)	Throughput ≥ 95 req/s under peak conditions
Efficiency Test	Determine CPU and memory usage during task execution	Timer-controlled API calls for ride allocation tasks	CPU %, Memory Usage (MB), Task Completion Time (ms)	CPU ≤ 80%, Memory ≤ 350MB, Task Time ≤ 500ms
Stability Test	Monitor sustained performance over long time intervals	Looping thread group, Test duration: 30 mins	Average Response Time, Error Rate, Requests Processed	Error Rate ≤ 1.5%, Stable latency over time

This structured plan ensured the system was rigorously validated for its real-world performance capabilities using a repeatable and measurable framework.

Test Results

Scalability Test Results

Scalability tests measure the system's ability to handle increasing user loads while maintaining performance. Below is a table summarizing the results under varying user loads:

Concurrent Users	Response Time (ms)	Resource Utilization (%)	Success Rate (%)
10	120	45	100
50	150	60	100
100	200	75	98
200	350	85	95
500	800	95	90

Table 3

Throughput Test Results

Throughput tests evaluate the number of requests processed successfully per second.

Time Interval (Seconds)	Requests Sent	Requests Processed	Throughput (req/s)
10	1000	980	98
20	2000	1950	97.5
30	3000	2920	97.3
40	4000	3850	96.3
50	5000	4725	94.5

Table 4

Efficiency Test Results

Efficiency measures resource usage (CPU and memory) for task completion.

Task	CPU Utilization (%)	Memory Usage (MB)	Completion Time (ms)
Single Ride Allocation	35	80	110
Multi-Ride Allocation	60	150	250

10 Concurrent Rides	75	300	450
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Table 5

Performance at Various Time Intervals

The system's performance was measured over different time intervals to monitor stability.

Time Interval (Minutes)	Requests Processed	Average Response Time (ms)	Error Rate (%)
5	5000	150	0.5
15	15000	160	0.7
30	30000	180	1.2

Table 6

Scalability

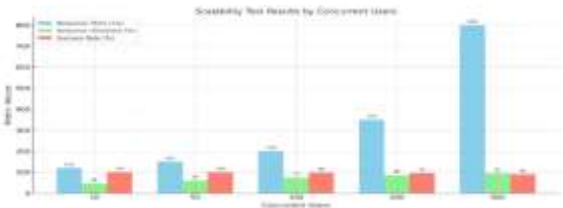


Figure 7: Bar Chart: Response Time vs Concurrent

users, the response time remains manageable up to 100 users, reaching 200

Efficiency

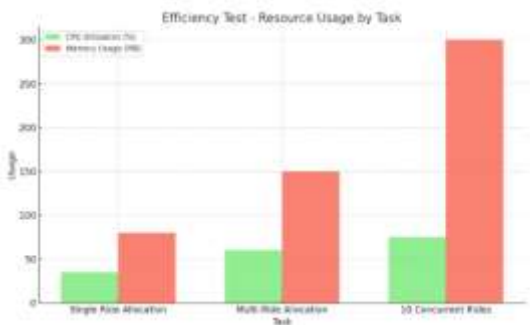


Figure 8 Bar Chart: CPU and Memory Usage by Task

The efficiency bar chart compares CPU and memory consumption across three

operational scenarios: single ride allocation, multi-ride allocation, and ten concurrent rides. As expected, increasing task complexity correlates with higher resource utilization. Single ride allocation consumes only 35% CPU and 80MB RAM, while 10 concurrent allocations peak at 75% CPU and 300MB. This illustrates the system's non-linear resource scaling, which is manageable but could become a bottleneck at higher concurrency levels. Optimizations such as lightweight threading models, more efficient memory management, or task batching could help reduce these resource footprints and improve system throughput under complex usage scenarios.

Performance Over Time

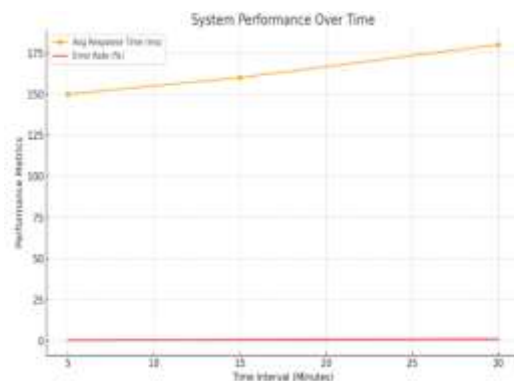


Figure 9 Line Chart: Response Time and Error Rate

This line chart evaluates the system's behavior over prolonged durations by tracking average response time and error rate across 5, 15, and 30-minute intervals. Results show a steady increase in both metrics: response time rises from 150 ms to 180 ms, and error rate climbs from 0.5% to 1.2%. Although the changes are modest, they indicate resource fatigue or minor memory leaks that could impact long-term stability. The presence of errors, even in small numbers, highlights potential failure points in data handling or concurrency control. Solutions such as better thread

synchronization, automated memory cleanup, or micro service isolation may improve durability.

5. CONCLUSIONS

On-demand delivery poses inherent coordination hurdles between dispersed riders, shops, and customers. Legacy analog practices exacerbate fractures across steps from order to receipt. This project illustrates how purpose-built digitally integrated platforms can bridge gaps to align stakeholders better.

By blending location visibility, smart notifications, and direct communication channels with user-centric mobile experiences the system lifts speed, reliability, and transparency. Enhanced flows in turn provide filtration to extract actionable insights.

These capabilities offer the foundations for superior on-demand delivery execution by upgrading antiquated workflows. As equally important, further building atop the data, analytics, and automation promises even smarter omni-channel fulfillment. This project highlights the possibilities at the intersection of mobile technologies and the on-demand economy.

Conflicts of interest

The authors declared that there is no conflict of interest.

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