

# AI-Driven Vehicle Assistance Platform with Geolocation Services

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**Abstract-** It often has brought inconveniences of unsafe situations and discomfort to its customers owing to vehicular breakdown. Typical roadside assistant applications that come out face problems such as high response times, small cover-up areas, and lack of real-time diagnostic capabilities among other problems. This research proposal intends to establish an innovative, web-based platform called Repair that has AI and LBS technologies integrated to provide real-time assistance for vehicles. The core feature of Repair is an AI-powered chatbot that can troubleshoot the most common vehicle issues independently. Advanced NLP techniques are applied to guide users through the diagnostic steps and provide solutions to problems such as flat tires, dead batteries, or other engine issues. When the problem exceeds the capabilities of the chatbot, the system uses Geolocation API technology to pinpoint the user's exact location and dispatch the nearest available towing service. This seamless integration of AI and geospatial technology ensures faster response times, reducing user waiting periods and improving service efficiency.

**Index Terms-** Vehicle Breakdown, AI Chatbot, Geolocation API, Towing Services, Dynamic Pricing, Web Development, Emergency Assistance

## I. INTRODUCTION

Vehicle breakdowns are an unpredictable yet common problem for drivers, often resulting in logistical inefficiencies, delayed responses, and user frustration. Traditional roadside assistance systems have limited reach and responsiveness and cannot keep up with the demands of modern mobility. This paper introduces Repair, a novel platform that integrates AI and location-based services to address these challenges effectively. Repair reduces response times and provides more efficient service by using AI-powered chatbots for vehicle diagnostics and location-based services to dispatch towing assistance. The platform also offers dynamic pricing based on the weight of the vehicle, making it an equitable and transparent cost structure. This solution is technically feasible, supported by the adoption of geospatial technology across the board and high-bandwidth networks, and fills a critical gap in the current roadside assistance models. The total number of roadside assistance requests in a year will vary widely based on factors such as location, population size, weather conditions, and the number of vehicles on the road. However, we can make some rough estimates based on general figures. Here are some general estimates:

**Major Cities:** In big cities, with dense traffic and a high percentage of car ownership, the large roadside assistance companies can serve tens of thousands of requests every year.

For instance, a big company might get between 50,000 and 100,000+ requests in a big city every year.

**Smaller Cities or Towns:** The number may be significantly lower in less densely populated areas. A smaller town may see anywhere between 5,000 to 20,000 requests annually.

**National Scale:** On a national scale, if we consider larger countries, roadside assistance providers in the U.S., for example, could process millions of requests yearly across all regions. For example, AAA, one of the largest providers in the U.S., typically responds to over 30 million requests annually.

### Problem Statement

Traditional roadside assistance services often fail to provide quick and easily accessible support in case of a vehicle breakdown. The absence of real-time troubleshooting and a long time for service cause users to be dissatisfied with these services. This paper suggests Repair, a technology-enabled solution that integrates AI and geolocation services to enhance the user experience and speed up the towing and troubleshooting process.

## II. LITERATURE REVIEW

It can be noticed that the vehicle assistance system development over the years has been propelled significantly

with Artificial Intelligence, Location-Based Services, and data analytics. This literature review summarizes important research on the integration of AI in vehicle assistance systems, the role of LBS in the response time improvement, and dynamic pricing models to show how innovations directly inform the design and functionality of the Repair platform.

### 1. Roadside Assistance Systems with the Help of Artificial Intelligence

The inclusion of Artificial Intelligence has modified the interaction of assistance vehicle systems with users completely.

The AI-powered chatbot diagnosing and troubleshooting of car issues is a part and parcel of modern car assistance platforms. According to Garikapati et al. (2024), AI is in the emergency management phase of self-driven vehicles, where it generates real-time diagnostics and solving capabilities. These NLP-driven AI systems can comprehend a variety of inputs from users, which is the reason why these have found useful applications in roadside assistance services for diagnosing mechanical failure issues, such as flat tires, dead batteries, or even engine malfunction. Smith C Taylor (2023) discussed the increasing role of AI-powered chatbots in vehicle safety, noting their ability to manage common vehicle issues by using intelligent conversational agents that engage with users, help in troubleshooting, and escalate any complicated issues to human agents. Their work was able to show potential where AI can improve vehicle diagnostics both in terms of efficiency and accuracy, thus benefiting customers by reducing response times and instances of human intervention. Patel C Verma (2024) further discussed how AI, especially NLP and machine learning algorithms, can enhance the precision of vehicle diagnostics. Analyzing historical data from vehicles, AI systems can predict potential breakdowns and offer preventive advice, which could be integrated into roadside assistance platforms like Repair. This is preventive in nature and aligns with the goals of Repair, which is not only to address current breakdowns but also to prevent future incidents through AI-driven insights.

### 2. Location-Based Services and Real-Time Dispatching

Location-Based Services (LBS) is a critical component of the effectiveness of contemporary roadside assistance.

Chen C Zhao (2024) emphasized the significance of LBS in automotive platforms for emergency management, especially focusing on the potential of geospatial technology to enhance service response times.

Vehicle assistance platforms can track a user's location precisely with the help of GPS and geolocation APIs and dispatch towing services from nearby areas. This does away with the need for manning navigation, thereby reducing delays while increasing the speed of service. Wang C Liu (2024) also

discussed how LBS, in conjunction with IoT and AI, improves the responsiveness of roadside assistance systems. The system can dynamically allocate resources based on proximity because it incorporates real-time location data, ensuring that the nearest service provider is dispatched. This integration, apart from improving the efficiency of roadside assistance, also enables the system to optimize routing, considering traffic conditions and other factors. The use of LBS in the Repair platform follows these principles, enabling real-time tracking of user location and the ability to dispatch the closest towing service to minimize wait times and improve overall service efficiency.

### 3. Dynamic Pricing Models in Roadside Assistance

The conventional roadside assistance services often work on fixed pricing models that do not consider vehicle weight, distance, or service demand.

Banerjee C Gupta (2024) suggested dynamic pricing for towing services, which would involve using AI to determine charges based on factors like the weight of the vehicle, location, and demand at the time of service.

This model ensures a fair and transparent pricing structure while helping service providers optimize their revenue based on resource utilization.

Zhao C Huang (2023) further investigated the benefits of dynamic pricing in transportation services, including roadside assistance. They argued that such pricing models increase transparency for consumers and improve service accessibility by offering a cost structure that reflects real-time conditions. For example, during high-demand periods, the system can adjust pricing to encourage service providers to fulfill urgent requests while ensuring that users are aware of the pricing logic. The dynamic pricing model used in Repair is comparable to these studies, where towing fees are charged by weight of the vehicle. This method not only guarantees that users are charged appropriately according to their vehicle's requirements but also improves the efficiency of service providers by correlating costs with actual service demands.

### 4. The Future of Roadside Assistance: Predictive Analytics and Automation

Going forward, the future of integration with vehicle assistance platforms and predictive analytics holds a lot of promise. Kumar C Singh (2023) reviewed the role of predictive maintenance in automotive systems by focusing on how AI and machine learning algorithms predict vehicle failure before it happens. Such systems analyze data from vehicle sensors and historical performance records to identify early warning signs of mechanical issues so that drivers can take preventive action before the breakdown occurs. By adding predictive analytics to Repair, its abilities would be further enhanced. That is not just providing aid with the

current breakdowns but also preventing future incidents. It would alert users to issues with the vehicle health if continuously monitored by AI, which in turn, would provide timely maintenance recommendations and reduce the chances of major breakdowns.

### III. SYSTEM DESIGN

The Repair platform was developed on the principles of simplicity and efficiency; hence, it is user-friendly. Some of its major characteristics are as follows:

#### 1. User Interface Design

The interface is straightforward and simple, using the peaceful blue and white color scheme. This consists of:

- **Chatbot Section:** An interactive page where users can describe issues with their vehicles and seek troubleshooting.
- **Location Sharing:** Geolocation feature to share the location of the user, which activates the nearest towing services.
- **Service Buttons:** Simple call-to-action buttons like "Request Towing Service" and "Send Issue" for prompt action.

#### 2. Chatbot Functionality

The core of the service is an AI chatbot powered through Natural Language Processing (NLP). The chatbot identifies major vehicle problems (e.g., flat tire, dead battery) and gives solutions thereto. For more complex troubles, it asks users to call for the towing service. Over a period of time, a machine learning technique improves how the chatbot responds and tries to understand.

#### 3. Inclusion of Location-Based Services

Using the Geolocation API, the application captures the location data (latitude and longitude) of the users to identify the nearest towing services. This geospatial functionality ensures that help is dispatched to the user's exact location, reducing delays and the need for manual navigation.

### IV. METHODOLOGY

#### 1. Front-End Development

The website is built with HTML5, CSS3, and JavaScript to make it responsive. Bootstrap is used to optimize the layout and grid management to ensure the platform is mobile-friendly.

#### 2. Chatbot Development

The chatbot was made using Python and JavaScript, with its responses using rule-based responses triggered by user inputs. For even more complex functionality, TensorFlow and Dialog flow were used in order to increase the capability of the bot to

converse as widely as possible and to respond to even more kinds of inputs more effectively.

#### Chatbot Implementation

##### Requirement Analysis

The primary functions the chatbot was supposed to meet and fulfill included: Vehicle Diagnostics: Analyze user inputs to identify potential vehicle issues. Repair Guidance: Provide step-by-step instructions for minor issues.

**Service Coordination:** Assist users in requesting towing services or connecting with nearby mechanics.

Key user inputs include vehicle symptoms, error codes, and towing requests.

##### Tool Selection

The following tools and technologies were chosen for the chatbot development: Languages: Python for the backend AI logic and JavaScript for the front-end interface. Natural Language Processing (NLP):

Dialog flow for understanding user intent and managing conversational flows. TensorFlow for advanced AI models, enhancing the capability of the chatbot to process a wide range of inputs.

**Database:** Firebase for real-time user data storage and retrieval.

**APIs:** Google Maps API for location tracking and nearby service recommendations.

#### 3. Chatbot Development

##### Rule-Based Responses

Initially, rule-based logic was integrated using Python to yield responses for basic queries. For example, if someone complains that, "My car won't start," it recommends that the user checks the battery or ignition system.

##### Training NLP Model

The chatbot learns through the use of dialog flow trained on the intents like - Vehicle Issue Diagnosis

##### Request Tow

Nearby Mechanic Search Some training phrases were introduced in order for the system to recognize well and have proper responses.

##### Integration of TensorFlow

In-house AI models have been designed in order to gain maximum possible diagnostics capabilities. These AI models are trained on labelled datasets regarding common vehicle breakdowns.

#### 4. Back-end Development

The backend, developed on Python, helps support NLP Processing with sending user input to dialog flow for further processing of responses. Additionally, for Service Coordination, it will fetch and then present near-by services for towing or repair considering the location of the user.

**Action Triggers:** It can initiate actions like sending emergency alerts or connecting users to service providers.

#### 5. Front-End Development

A user-friendly interface was developed using JavaScript:

- **Chat Window:** It enables users to interact with the chatbot through text.
- **Real-Time Communication:** AJAX and WebSocket technologies allow smooth and instantaneous interactions with the backend.

#### 6. Testing and Refinement

Extensive testing was done to ensure the chatbot handles a variety of inputs effectively. Edge cases and anomalies were caught and rectified through refining the rule-based logic and retraining AI models.

#### 7. Deployment

The backend of the chatbot was deployed on a cloud platform (for example, AWS or Google Cloud) to scale up and be reliable.

The front-end chatbot interface was integrated with the mobile and web applications.

#### 8. User Feedback and Iteration

Post-deployment, user feedback was gathered to identify areas for improvement. Updates in the knowledge base of the chatbot were done frequently, along with updating new AI models retrained with new data for higher performance.

#### Integration of Towing Service into Location Sharing

The location-sharing feature of the platform depends on the HTML5 Geolocation API. After capturing the user's location, the system shows nearby towing services. The dynamic pricing model based on the weight of the vehicle ensures a transparent and fair price structure for the users. Google map API could also be used to display map on the user screens and they could share their current location.

#### Revenue Model

The revenue of the Repair platform will be derived from multiple sources:

**Chatbot-Assisted Services:** The users can access premium troubleshooting advice for free. Service Centers : Service Centers will be charged for showcasing their towing service and Repairing services on the website.

**Towing Service Fees:** The towing charges are calculated on the basis of the vehicle weight:

Small vehicles (<1000 kg): ₹500

Medium vehicles (1000–3000 kg): ₹1000 Heavy vehicles (>3000 kg): ₹2000

#### Results and Discussion

##### User Experience and Accessibility

The design of the platform has been based on user experience and accessibility as its guiding principles, ensuring that it has a seamless and intuitive interface for all users. Preliminary testing results are overwhelmingly positive, with praise from users on the ease of navigation and overall efficiency of the platform.

The clean user interface is very intuitive to operate in reporting vehicle issues on this platform. Using this application or chatbot, whether via text or voice commands, has never been easier. There are guided prompts to guide the users in their description of their problems to the chatbot so it provides relevant and actionable solutions.

**Chatbot Accuracy:** The chatbot has been very accurate in diagnosing the vehicle's problem during the test. With NLP and a comprehensive database of vehicle troubleshooting data, the chatbot is able to quickly identify the common causes of breakdowns and provide recommendations for the next course of action. This precision eliminates the guesswork usually involved in roadside emergencies.

**Efficient Location Sharing:** One of the standout features is efficient location sharing, powered by GPS integration. Users can easily share their exact location with service providers, thus facilitating faster response times. This feature is particularly beneficial in unfamiliar or remote areas where precise location details may be difficult to describe.

**Improved Response Times:** Compared to traditional roadside assistance services, the platform significantly reduces the time required to address user requests. Automated processes, such as instant towing dispatch and nearby service provider suggestions, ensure that help is on the way within minutes. Early testers reported feeling reassured by the platform's promptness and efficiency.

## V. CONCLUSION

The Repair platform is one of the biggest advancements in vehicle breakdown assistance, bringing together the power of AI and location-based services for efficient, user-centric solutions.

Its dynamic pricing model ensures fairness and accessibility while providing an intuitive interface that increases the user

experience and makes roadside assistance more reliable and convenient.

With promises to further develop to add the following: services; predictability diagnostics, scalability. The platform stands at par, and it promises to shift the industry with new measures of innovative adaptation that fits the changing requirements of users.

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