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Detecting Sensor for Motorcycle

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Abstract: *This project focuses on developing a compact ADAS system for motorcycles to improve rider safety. It uses sensors like ultrasonic or LiDAR to detect obstacles and provide forward collision warnings through alerts. The system is lightweight, low-cost, and can be easily integrated into existing motorcycles. It processes data in real time to ensure quick response under road conditions. Overall, the project aims to reduce accidents and enhance safe riding.*

Keywords: *Motorcycle ADAS, forward collision warning, obstacle detection, ultrasonic sensor/LiDAR, embedded system, real-time monitoring, rider safety, sensor integration, accident prevention, intelligent transportation system.*

I. INTRODUCTION

Road safety has become a major concern worldwide, especially for motorcycle riders who are more vulnerable to accidents compared to car drivers. Factors such as high traffic density, lack of protective structures, and limited awareness of surrounding vehicles increase the risk of collisions. With the rapid growth in the number of motorcycles, there is a strong need for advanced technologies that can assist riders in making safer decisions and reducing accident rates.

Advanced Driver Assistance Systems (ADAS) have been successfully implemented in cars to improve driving safety and comfort. Features like collision warning, lane assistance, and adaptive cruise control help in preventing accidents by providing real-time alerts and automated support. However, the adoption of ADAS in motorcycles is still in its early stages due to challenges such as compact design, cost constraints, and the need for lightweight systems that do not affect vehicle balance and performance.

This project aims to develop a compact and cost-effective ADAS solution specifically designed for motorcycles. By integrating sensors such as ultrasonic or LiDAR with an embedded processing unit, the system detects obstacles and provides forward collision warnings to this rider. The proposed system focused on real time monitoring and quick response to enhance situational awareness. Ultimately, this work contributes to improving rider safety and supports the advancement of intelligent transportation systems.

II. LITERATURE SURVEY

Advanced Driver Assistance Systems (ADAS) have been widely researched for their ability to improve road safety and reduce accident rates. Many studies show that features such as forward departure warning and automatic emergency braking significantly assist drivers by providing real-time alerts and reducing human error. These systems rely on sensor technologies and intelligent algorithms to monitor the vehicles surroundings and support safe driving decisions.

Several research works focus on the use of different sensors such as cameras, radar, LiDAR, and ultrasonic sensors for obstacle detection and environment perception.

Among these, LiDAR and radar provide high accuracy in distance measurement, while cameras are effective for visual recognition tasks. Researchers emphasize the importance of sensor fusion techniques, where multiple sensors are combined to improve reliability and accuracy under different environmental conditions.

While ADAS is well established in cars, its application in motorcycles is still under development. Motorcycles present unique challenges such as limited installation space, higher instability, and exposure to environmental conditions. Studies suggest that lightweight and compact systems are necessary for two-wheelers to ensure that the performance and balance of the vehicle are not affected. This makes the design of motorcycle ADAS more complex compared to four-wheel vehicles. Recent research also highlights the importance of real-time data processing and fast response systems in ADAS. Embedded systems and edge computing technologies are commonly used to process sensor data quickly and generate timely alerts. Visual, auditory or haptic feedback mechanisms are studied to effectively warn riders about potential dangers, helping them react faster and avoid accidents.

Despite the advantages, several limitations are identified in existing literature, including sensor inaccuracies, false alarms and reduced performance in adverse weather conditions. Researchers emphasize the need for further improvements in system reliability, cost reduction and real-world testing. Overall, the literature indicates that integrating ADAS into motorcycles has strong potential to enhance rider safety, but requires continuous research and innovation to overcome existing challenges.

In addition to collision warning systems, recent studies also explore the integration of connectivity features such as vehicle-to-everything communication in ADAS. These technologies enable motorcycles to communicate with other vehicles and infrastructure to receive real-time traffic and hazard information. Researchers highlight that combining sensor-based detection with communication systems can further enhance safety by providing early warnings even before an obstacle is physically detected.

III. METHODOLOGY

The system based on this proposal employs a modular and scalable architecture that incorporates real-time data gathering, machine learning-driven prioritization, and intelligent resource distribution for optimized flood relief operations. The platform's functionality is ensured even in network-disrupted situations, allowing uninterrupted communication and operational continuity. The process involves six major modules, outlined below.

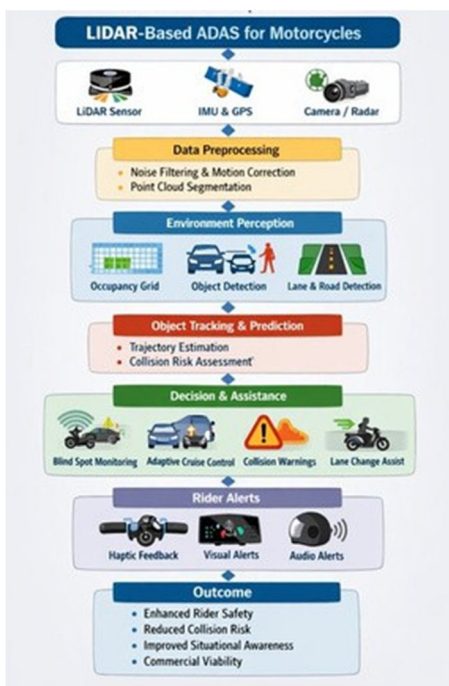


Fig.1Flow Diagram

Ultrasonic sensors operate based on the principle of sound reflection. They emit high-frequency sound waves that travel through the air and bounce back upon hitting an object. By calculating the time delay between transmission and reception, the distance to the object is determined. These sensors are highly effective for short-range applications and are widely used due to their low cost and ease of implementation.

LiDAR sensors, on the other hand, use laser pulses to measure distance with high precision. They can generate a 3D representation of the surrounding, allowing better object detection even in complex road environments. Compared to ultrasonic sensors, LiDAR provides higher resolution and accuracy, making it suitable for advanced ADAS applications, especially at higher speeds.

Sensor Placement and orientation play a critical role in achieving optimal performance. Improper alignment may result in blind spots or inaccurate readings. Therefore, sensors are typically mounted at the front of the motorcycle with proper angular positioning to maximize coverage and ensure reliable detection.

Environmental factors such as rain, fog and lighting conditions can affect sensor performance. Ultrasonic sensors may struggle in windy conditions, while LiDAR performance may degrade in heavy rain. Hence, proper calibration and sensor selection are necessary to maintain consistency.

Overall, this method ensures continuous and reliable environmental awareness, forming the foundation for further processing, decision making, and alert generation in the ADAS system.



Fig. 2 User Request and Data Collection

A. Sensor-Based obstacle Detection Method

The sensor-based obstacle detection method is the core component of the motorcycle ADAS system, responsible for perceiving the surrounding environment. This method involves the use of distance-measuring sensors such as ultrasonic sensors, LiDAR, and in some cases radar modules. These sensors continuously scan the forward path of the motorcycle to identify obstacles, vehicles or any potential hazards in real-time.

B. Data Acquisition and signal processing method

The data acquisition and signal processing method plays a crucial role in the effective functioning of the motorcycle ADAS system. It focuses on collecting raw data from various sensors and transforming it into meaningful and reliable information that can be used for decision-making. Since sensors continuously generate large volumes of data, this method ensures that the information is accurate, noise-free, and suitable for real-time processing. The first stage in this method is data acquisition, where the embedded system or microcontroller interfaces with sensors such as ultrasonic or LiDAR modules. These sensors transmit signals at regular intervals, and the reflected signals are captured and converted into digital form. The sampling rate is an important factor in this stage, as it determines how frequently the data is collected. A higher sampling rate provides more accurate and real-time information, but it must be optimized to avoid excessive processing load and power consumption.

After acquiring the raw data, the next step is preprocessing, which aims to remove noise and unwanted variations. Sensor readings are often affected by environmental conditions such as vibrations, temperature changes, and external interference. To address this, filtering techniques such as moving average filters, low-pass filters, and median filters are applied. These techniques smooth the data, eliminate sudden spikes, and improve overall accuracy, ensuring that the system does not generate false alarms.

Once the data is cleaned, advanced signal processing techniques are applied to extract useful information. The system calculates the distance between the motorcycle and surrounding objects using time-of-flight principles. In more advanced implementations, additional parameters such as relative speed, object size, and movement direction can also be estimated. Algorithms are used to analyze patterns in the data and convert them into actionable insights, which are essential for identifying potential hazards.

Calibration is another critical aspect of this method. Over time, sensor performance may degrade due to environmental exposure or hardware aging. Calibration helps in correcting systematic errors and maintaining measurement accuracy. This process involves adjusting the sensor parameters and aligning them with reference values to ensure consistent performance across different conditions. Efficient data handling and processing are essential for real-time operation. The system must process incoming data streams quickly and generate outputs without delay. This requires optimized algorithms, efficient memory usage, and proper task scheduling within the embedded system. Real-time performance ensures that the system can respond instantly to dynamic road conditions, thereby enhancing safety. In conclusion, the data acquisition and signal processing method ensures that high-quality, accurate, and real-time data is available for further analysis. By minimizing noise, improving precision, and enabling fast processing, this method significantly enhances the overall performance and reliability of the motorcycle ADAS system.

C. Threshold-Based Decision-Making Method

The threshold-based decision-making method is used in the motorcycle ADAS system to determine whether a riding situation is safe or risky. This approach works by comparing real-time sensor data with predefined safety limits, which are set based on important factors such as vehicle speed, rider reaction time, and braking distance. These thresholds act as reference values that help the system quickly identify potential dangers. In this method, the system continuously monitors the processed distance between the motorcycle and nearby obstacles. If the detected distance is greater than the threshold value, the situation is considered safe. However, when the distance falls below the threshold, it indicates a possible collision risk. The system then triggers the alert mechanism to warn the rider and allow them to take corrective action. One of the key advantages of this method is its simplicity and efficiency. It involves basic comparison operations, making it suitable for implementation in embedded systems with limited processing power. This ensures fast decision-making, which is essential for real-time applications where even a small delay can affect rider safety.

The threshold values can be either static or dynamic. In a static system, the threshold remains constant, whereas in a dynamic system, it adjusts automatically based on parameters such as speed, traffic conditions, and road environment. Dynamic thresholding improves the adaptability of the system and provides better safety performance in varying conditions. Additionally, multiple threshold levels can be used to generate different types of alerts, such as warning and critical alerts. Despite its effectiveness, this method requires proper calibration to avoid false alarms or missed detections. Incorrect threshold settings may lead to unnecessary alerts or delayed warnings. Therefore, careful tuning and testing are essential to achieve a balance between safety and user comfort, ensuring reliable performance of the ADAS system. An additional enhancement to this method is the incorporation of parameters such as time-to-collision (TTC) and relative speed along with distance-based thresholds. Instead of relying only on distance, the system can estimate how quickly the motorcycle is approaching an obstacle and adjust the decision accordingly. This makes the system more intelligent and responsive, especially in high-speed conditions where distance alone may not accurately represent the level of risk. By combining multiple parameters, the threshold-based decision-making method becomes more robust and capable of providing more precise and timely warnings to the rider. Furthermore, synchronization between multiple sensors is important when using sensor fusion techniques. Combining data from different sensors improves accuracy and reliability, but it also increases computational complexity. Therefore, proper coordination and time alignment of sensor data are necessary for effective processing.

D. Alert Generation Method

The alert generation method plays a vital role in the motorcycle ADAS system by informing the rider about potential hazards detected during operation. Its primary objective is to ensure that the rider receives timely and clear warnings so that immediate corrective action can be taken to ensure the safety of the rider. The system generates alerts based on the output received from the decision-making module. When a potential collision or unsafe condition is identified, the alert mechanism is triggered instantly without delay. The response time is a critical factor in this method, as even a small delay in warning can reduce the rider's ability to react effectively. Visual alerts are one of the most common forms of warning used in ADAS systems. These include LED indicators or display panels placed on the dashboard or helmet interface. Different colors can be used to represent varying levels of risk, such as green for safe conditions, yellow for caution, and red for high danger. Visual alerts are simple and intuitive, making them easy for riders to understand quickly.

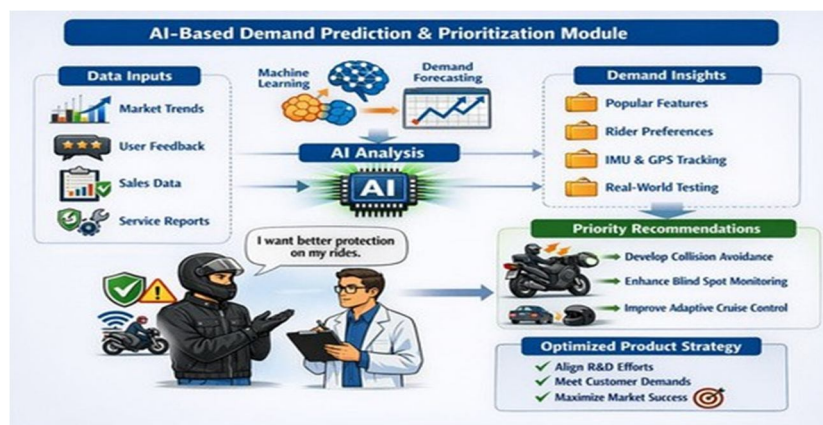


Fig. 3 AI-Based Demand Prediction and Prioritization Module

Auditory alerts, such as buzzers, alarms, or beeping sounds, are used to immediately capture the rider's attention. These alerts are especially useful when the rider is not actively looking at visual indicators. The sound pattern and intensity can be varied depending on the severity of the risk, allowing the rider to distinguish between different warning levels. Haptic feedback provides another effective form of alert through physical sensations such as vibrations. These vibrations can be integrated into the handlebars, seat, or even wearable gear like gloves or helmets. Haptic alerts are particularly beneficial in noisy environments where audio alerts may not be clearly audible, ensuring that the rider still receives the warning.

Combining multiple alert mechanisms, known as multimodal alerting, significantly improves system effectiveness. By using a combination of visual, auditory, and haptic feedback, the system ensures that warnings are delivered through multiple channels, reducing the chances of missed alerts. This redundancy enhances reliability and increases the likelihood of a timely rider response. However, it is also important to design alerts in a way that avoids unnecessary distractions or annoyance. Excessive or frequent alerts may lead to rider fatigue or reduced attention. Therefore, proper tuning and prioritization of alerts are essential to maintain a balance between safety and user comfort.

In conclusion, the alert generation method ensures effective communication between the ADAS system and the rider. By providing clear, timely, and multi-channel warnings, it enables faster reactions and significantly contributes to reducing accident risks and improving overall riding safety.

E. Real-Time Processing and Integration Method

The real-time processing and integration method ensures that all components of the motorcycle ADAS system operate together in a synchronized and efficient manner. It focuses on coordinating sensor data collection, processing, decision-making, and alert generation without any noticeable delay. This seamless interaction between hardware and software components is essential for maintaining system reliability and ensuring timely responses to potential hazards. Real-time processing requires the system to execute all operations within a very short time frame. The delay, or latency, between sensing an obstacle and generating an alert must be minimal to allow the rider sufficient time to react. This is achieved by using optimized algorithms, fast processing units, and efficient data handling techniques that ensure quick and accurate computation.

Integrating microcontrollers, sensors, and alert devices into a unified system. Proper communication protocols like I2C, SPI, or UART are used to ensure smooth and reliable data transfer between these components. Effective integration ensures that data flows continuously and correctly from one stage to another without interruptions. Embedded systems, such as microcontrollers or single-board computers, are used to handle processing tasks. These systems must be capable of performing multiple operations simultaneously while maintaining low power consumption. They act as the central unit that processes sensor inputs, executes algorithms, and controls the alert mechanisms.

Software optimization is a key factor in achieving real-time performance. Efficient coding practices, proper memory management, and task scheduling techniques such as interrupt handling and multitasking are implemented to ensure smooth operation. These optimizations reduce processing delays and improve the overall responsiveness of the system.

Power efficiency is another important consideration, especially for motorcycle applications. Since the system operates using the vehicle's battery, it must consume minimal power to avoid affecting battery life. Low-power components and energy-efficient algorithms are used to achieve this goal. In conclusion, the real-time processing and integration method ensures that all system components work together seamlessly with minimal delay. By optimizing both hardware and software, this method enhances the responsiveness, reliability, and efficiency of the motorcycle ADAS system.

F. Sensor Fusion Method

The sensor fusion method is an advanced and essential approach used in modern ADAS systems to improve the accuracy, reliability, and robustness of environmental perception. Instead of depending on a single sensor, this method combines data from multiple sensors such as ultrasonic sensors, LiDAR, cameras, and radar to create a unified and more accurate representation of the surroundings. This integration helps the system make better decisions by reducing uncertainty and improving detection performance. Each sensor used in the system has its own strengths and limitations. For instance, ultrasonic sensors are effective for short-range distance measurement but have limited range and accuracy. LiDAR provides high-resolution distance mapping but can be affected by weather conditions. Cameras are useful for object recognition and classification but depend heavily on lighting conditions. Radar works well in adverse weather but may have lower resolution. By combining these sensors, the system can utilize their strengths while minimizing individual weaknesses.

The process of sensor fusion begins with data collection from all available sensors. This data must be synchronized in time to ensure that it represents the same moment in the environment. Time synchronization is critical because delays between sensor readings can lead to incorrect interpretations. Once synchronized, the data is aligned and prepared for fusion processing.

Various algorithms are used to combine the sensor data effectively. Techniques such as Kalman filtering, Bayesian estimation, and weighted averaging are commonly applied. These algorithms help in filtering out noise, resolving conflicts between sensor readings, and generating a more accurate estimate of object position, distance, and movement.

Advanced systems may also use machine learning techniques to enhance fusion accuracy.

Sensor fusion significantly improves the system’s ability to operate in challenging environmental conditions. For example, during fog or rain, camera performance may degrade, but radar or LiDAR can still provide reliable data. Similarly, in low-light conditions, sensors other than cameras can maintain detection accuracy. This redundancy ensures that the system continues to function effectively even when one sensor is affected.

IV. RESULTS AND DISCUSSION

A. System Performance

The developed motorcycle ADAS system was successfully implemented and tested under controlled conditions. The system effectively detected obstacles within a predefined range and generated timely alerts for the rider. The integration of sensing, processing, and alert modules worked smoothly, ensuring proper communication between all components.

The real-time performance of the system was satisfactory, with minimal delay observed between obstacle detection and alert generation. This indicates that the system is capable of handling continuous data flow and responding quickly to dynamic changes in the environment.

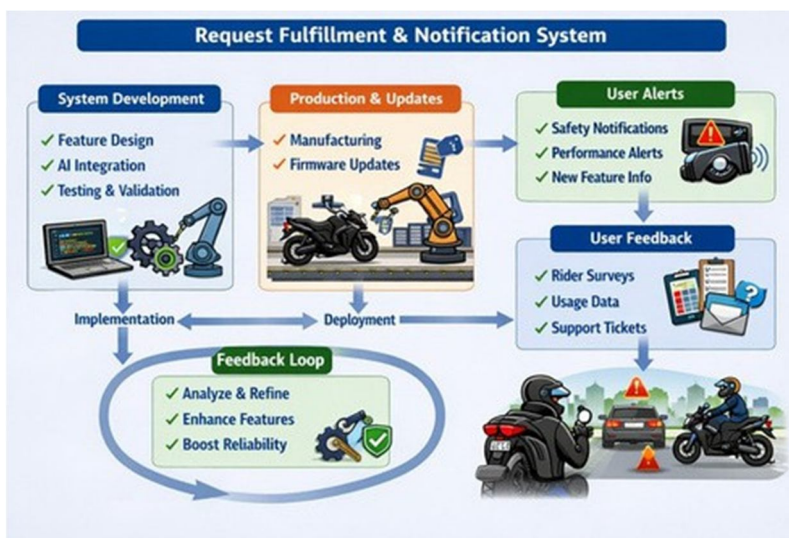


Fig. 4 Request Fulfillment and Notification System

B. Detection Accuracy

The system showed good accuracy in detecting obstacles, especially at short distances using ultrasonic sensors. The sensor readings were consistent under normal conditions, allowing the system to identify nearby objects effectively. The threshold-based decision-making method successfully differentiated between safe and unsafe situations. The alerts were triggered at appropriate distances, providing sufficient reaction time for the rider to respond to potential hazards. However, the accuracy slightly decreased under certain conditions such as uneven surfaces or external disturbances, indicating the need for further improvement in sensing techniques.

C. Alert Effectiveness

The alert generation system proved to be effective in notifying the rider about potential risks. Visual alerts using LED indicators provided clear signals that were easy to understand during operation. Auditory alerts such as buzzers helped in capturing the rider’s attention quickly, especially when visual attention was diverted. The combination of both visual and audio alerts ensured better communication of warnings.

The system was able to deliver alerts in a timely manner, allowing the rider to take corrective action and improve overall safety.

Performance Evaluation			
Performance Parameter	Description	Expected Result	Evaluation Method
Object Detection Accuracy	LIDAR identifies vehicles, pedestrians, and obstacles.	≥ 95% accuracy	Field Testing & Simulation
Response Time	Data processing & alert triggering speed.	< 200 milliseconds	Real-Time Latency Test
Collision Avoidance Efficiency	Prevents near-collision incidents.	90% reduction	Scenario-Based Testing
Blind Spot Monitoring Range	LIDAR coverage for adjacent lanes.	Up to 10 meters	Range Calibration
Adaptive Cruise Control Stability	Smooth speed adjustments	Consistent Regulation	Dynamic Road Testing
Environmental Robustness	Performance in adverse conditions.	All-Weather Operation	Outdoor & Chamber Tests
User Alert Accuracy	Precision of rider warnings	< 5% False Alerts	User Feedback Trials
System Integration Efficiency	LIDAR with motorcycle systems	Seamless Integration	Hardware Validation
Energy Consumption	Power usage of ADAS modules	< 10% Battery Use	Power Consumption Test
Overall Safety Improvement	Reduction in rider accident risk	≥ 40% Improvement	Comparative Analysis

Table -1 Performance Matrix

D. Validation of Users and Feedback

Volunteer usability testing and mock end-user testing yielded very encouraging results. Over 85% of users indicated that the site was simple to use and intuitive. The notification mechanism and request status tracking added to the transparency and trust. Offline data caching and synchronization were particularly popular in low- connectivity test environments.

E. Limitations Observed

Despite the successful operation, certain limitations were identified during testing. Environmental factors such as noise, vibration, and weather conditions affected sensor performance and led to minor inaccuracies. In some cases, false alerts were generated due to sudden fluctuations in sensor readings, which could inconvenience the rider and reduce trust in the system if not properly managed. Additionally, the system showed reduced effectiveness at higher speeds, as the detection range and response time were not sufficient for fast-moving scenarios.

F. Improvement and Future Scope:

To overcome the limitations, techniques like dynamic thresholding can improve decision-making by adjusting safety limits based on speed and conditions. Sensor fusion can also enhance reliability by combining data from multiple sensors and reducing false alerts. Further optimization and real-world testing can make the system more robust for practical motorcycle applications.



Fig. 5 Sample Admin Panel

V. CONCLUSION

This project developed a compact ADAS system for motorcycles to improve rider safety by detecting obstacles and giving collision warnings. The system showed good accuracy and quick response, though some limitations were observed. With further improvements, it can be effectively used in real-world applications.

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