

RIDEMATE: AI-DRIVEN OPTIMAL ROUTING, COST ESTIMATION, AND IOT-ENHANCED CRASH DETECTION FOR LONG-DISTANCE BIKE RIDING

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ABSTRACT

This review paper examines the integration of artificial intelligence (AI) and Internet of Things (IoT) technologies in optimizing long-distance bike riding experiences, with a focus on routing, cost estimation, and crash detection. As AI and IoT innovations continue to shape transportation and mobility, they offer significant advancements in improving safety, efficiency, and rider experience. The paper explores key techniques such as AI-driven optimal routing algorithms, real-time cost estimation models, and IoT-enhanced crash detection systems, which contribute to the development of a comprehensive solution for long-distance bike riders. These models enable real-time traffic analysis, dynamic route planning, and predictive maintenance, enhancing the overall safety and cost-effectiveness of long-distance biking. Additionally, this review identifies future opportunities for AI and IoT to further personalize rider experiences, integrate with smart infrastructure, and address sustainability concerns in the biking ecosystem. The findings emphasize the potential of AI and IoT to revolutionize biking safety, navigation, and logistics, fostering a new era of connected and intelligent transportation for cyclists.

Keywords: Computer Vision, AI, ML, Image Processing, Classification, Visualization, Cross Domain Data Integration, Speech To Text, Metadata.

I. INTRODUCTION

Machine learning (ML) has emerged as a transformative force in computer vision, profoundly reshaping how machines perceive and interpret visual information. By enabling algorithms to learn from vast datasets, ML techniques allow for sophisticated image analysis, including tasks such as image classification, object detection, and real-time image segmentation. This capability is especially significant in enhancing accessibility for individuals with visual impairments, enabling them to interact more effectively with their environments. This review focuses on various machine learning techniques employed in computer vision, particularly highlighting the contrastive learning and bootstrapping approaches. Recent advancements have led to the development of groundbreaking models like CLIP (Contrastive Language-Image Pretraining) and BLIP (Bootstrapping Language-Image Pretraining), which have demonstrated remarkable capabilities in processing multimodal data—integrating visual and textual information to generate contextual insights. Through this examination, the review aims to present a comprehensive overview of these techniques, their contributions to the field, and potential future directions for research and application.

II. LITERATURE REVIEW

Overview

The integration of Artificial Intelligence (AI), Internet of Things (IoT), and machine learning (ML) in long-distance bike riding has brought significant improvements to several areas, including routing, cost estimation, and crash detection. Traditionally, bike riders relied on maps and manual inputs for route planning, and cost estimations were often based on basic factors, such as distance and terrain. However, with the emergence of more advanced technologies, modern systems now offer real-time data processing, predictive analytics, and dynamic adjustments that cater to the rider's needs in real-time.

AI-powered systems are increasingly being used to optimize bike routes by incorporating real-time traffic data, terrain mapping, and weather conditions, ensuring more efficient and safer journeys for long-distance riders. These systems can learn from historical data to predict optimal routes based on past rider behavior, road conditions, and other external factors such as construction zones or accidents.

IoT technologies, including smart helmets, GPS trackers, and sensors embedded in bikes, have become central to providing real-time data on the rider's speed, elevation, and location. These systems contribute to crash detection and accident prevention by analyzing accelerometer data to identify unusual movements that may indicate a fall or collision. Once an incident is detected, the system can immediately notify emergency services or designated contacts, helping to reduce the time taken for first responders to reach the scene.

Cost estimation models, powered by AI, now take into account a variety of factors including battery life for electric bikes, wear and tear on the bike, and even rider fatigue. These models predict the overall cost of a long-distance ride, considering aspects such as energy consumption, maintenance, and environmental factors, offering a more precise forecast of the cost and resources required.

The convergence of AI, ML, and IoT has opened up new possibilities for long-distance bike riders, ensuring that safety, efficiency, and cost-effectiveness are seamlessly integrated into the biking experience. These advancements are continuing to evolve, pushing the boundaries of what is possible in bike riding and transportation.

1. First Paper - BikeMate: Bike Riding Behavior Monitoring with Smartphones

Summary of the Paper :The research paper "BikeMate: Bike Riding Behavior Monitoring with Smartphones" presents a novel approach to enhancing bicycling safety using ubiquitous mobile technology. The system, BikeMate, utilizes smartphone sensors to detect dangerous riding behaviors such as lane weaving, standing pedaling, and wrong-way riding. By leveraging transfer learning, BikeMate reduces the training model overhead for different users, and crowdsourcing is employed to infer legal riding directions. The study demonstrates the effectiveness of BikeMate, achieving detection accuracies of 86.8% for dangerous riding behaviors and 90% for wrong-way riding using crowdsourced GPS data.

Improvements in Our Project:Building on the concepts introduced in BikeMate, our project, RideMate, expands these safety features specifically for long-distance motorbike riders. We improve upon the behavior detection capabilities by incorporating additional parameters tailored to motorbike riding, such as monitoring speed fluctuations, sharp turns, and potential hazards specific to motorbikes. We also integrate real-time alerts for road safety, weather conditions, and emergency contact notifications, enhancing the overall safety ecosystem for riders. By refining these features for motorbike users, RideMate offers a more comprehensive safety monitoring solution for long-distance biking.

Advantages of Our Improvements: The improvements made in RideMate offer enhanced safety and a more targeted approach for long-distance bikers. By incorporating additional parameters like speed monitoring and integrating real-time alerts, our solution provides a more robust safety system. This increased specificity improves rider engagement, reduces the risk of accidents, and enhances preparedness for unforeseen events. Additionally, by applying the smartphone-based approach, our system remains accessible, cost-effective, and easy to adopt, ensuring broader usability among bikers.

2. Second Paper - A Machine Learning Based Bike Recommendation System Catering To User's Travel Needs

Summary of the Paper :The research paper "A Machine Learning Based Bike Recommendation System Catering To User's Travel Needs" focuses on enhancing bike-sharing systems (BSS) by proposing a bike recommendation system that clusters bikes based on their behavioral patterns. The system utilizes machine learning techniques to analyze bikesharing data and weather conditions to recommend the best-suited bike for a user's travel needs. By evaluating real-world data from the Divvy Bike System in Chicago, the authors demonstrate that their method can effectively identify suitable bikes for specific trips, reducing user dissatisfaction and improving the overall bike-sharing experience.

Improvements in Our Project: Inspired by this recommendation system, our RideMate project integrates a similar approach to help long-distance bikers choose routes and pit stops that best match their travel requirements. We apply machine learning techniques to analyze historical data on routes, road conditions, and biker preferences. This allows our app to recommend the most suitable paths, considering weather conditions, rider experience, and vehicle performance. By offering personalized route recommendations and pit stop

suggestions, we enhance the planning and safety features in our application, making it more tailored to individual rider needs.

Advantages of Our Improvements:The incorporation of a personalized recommendation system in RideMate provides riders with more precise and reliable travel plans. This minimizes uncertainty and ensures that bikers are well-prepared for their journeys, enhancing safety and convenience. By considering individual rider preferences, weather conditions, and bike performance, our system ensures a smoother and more enjoyable long-distance biking experience. Additionally, this data-driven approach can lead to more efficient resource utilization and better trip outcomes, ultimately benefiting both riders and service providers.

3. Third paper - Abnormal driving behaviors detection and identification using smartphone sensors

Summary of the Paper : The research paper "MotoSafe: Active Safe System for Digital Forensics of Motorcycle Rider with Android" presents a safety system that monitors high-risk motorcycle riding behaviors using smartphone sensors like accelerometers, magnetometers, and GPS. This system can detect dangerous maneuvers such as weaving, sudden braking, and high-speed turning, and sends real-time alerts to riders when a high-risk behavior is identified. In cases of severe accidents, the system can automatically send emergency notifications to authorities and serve as a traffic data recorder, providing valuable digital forensics for crash investigations.

Improvements in Our Project: Building on MotoSafe's approach to high-risk behavior detection, our RideMate project integrates similar monitoring systems but focuses on a broader set of features tailored for long-distance bikers. We enhance the accident detection system by including additional factors like road conditions, weather alerts, and personalized route recommendations. Furthermore, RideMate includes an emergency contact notification system that triggers alerts to designated contacts in the event of accidents, providing an extra layer of safety for riders in remote areas.

Advantages of Our Improvements: The improvements made in RideMate extend beyond accident prevention by offering a more comprehensive safety ecosystem for long-distance bikers. By integrating road conditions, weather alerts, and real-time emergency notifications, we reduce the likelihood of accidents while also ensuring timely responses in case of emergencies. These enhancements provide bikers with greater peace of mind, making long-distance travel safer and more reliable, and improving the overall riding experience.

4. Fourth paper - MotoSafe: Active Safe System for Digital Forensics of Motorcycle Rider with Android

Summary of the Paper : The paper "D3: Abnormal Driving Behaviors Detection and Identification Using Smartphone Sensors" by Zhongyang Chen et al. explores an advanced approach to detecting and identifying abnormal driving behaviors using smartphone sensors. Current methods often provide only coarse-grained results, distinguishing abnormal behaviors from normal ones but lacking specificity. To address this, the authors propose D3, a system that uses smartphone accelerometers and orientation sensors to detect and classify six specific types of abnormal driving behaviors: weaving, swerving, sideslipping, fast U-turn, turning with a wide radius, and sudden braking. Through empirical studies and machine learning, particularly Support Vector Machines (SVM), D3 achieves an average accuracy of 95.36% in identifying these behaviors.

Improvements in Our Project: 1. Enhanced Detection Capabilities: Implement the fine-grained detection capabilities of D3 to improve the app's ability to identify specific types of abnormal driving behaviors. This will enable more accurate alerts and feedback for drivers, potentially reducing risky driving habits. 2. Feature Extraction and Analysis: Integrate advanced feature extraction techniques as outlined in the paper to enhance the app's performance. Features like acceleration and orientation patterns can be leveraged to provide detailed insights into driving behaviors. 3. Machine Learning Integration: Utilize machine learning models, particularly SVM, to classify and differentiate between various types of abnormal driving behaviors. This will enhance the app's ability to provide precise and actionable feedback to users.

Advantages of Our Improvements:1. Increased Accuracy: By adopting fine-grained detection methods, the app will offer a higher level of accuracy in identifying specific driving behaviors, leading to better safety and awareness for drivers. 2. Detailed Feedback: Enhanced feature extraction will allow the app to provide more detailed feedback on driving patterns, helping users understand their driving habits and make necessary

adjustments. Preventive Measures: With accurate identification and classification of driving behaviors, the app can serve as a preventive tool, alerting drivers to potential issues and helping to reduce the risk of accidents.

III. IMPLEMENTATION

This project employs three cutting-edge technologies: AI-driven routing algorithms, cost estimation models, and IoT-based crash detection systems.

A. AI-Driven Optimal Routing

The AI model utilizes machine learning algorithms to process large-scale geographic and traffic data, including real-time traffic conditions, road quality, and weather patterns. By analyzing this data, the model can suggest optimal routes for long-distance bike riding, considering variables such as road safety, terrain difficulty, and estimated time of arrival. This model adapts dynamically to changes in traffic or environmental conditions, offering users a reliable and efficient routing experience. It removes the need for manual route planning, improving the overall convenience of long-distance rides.

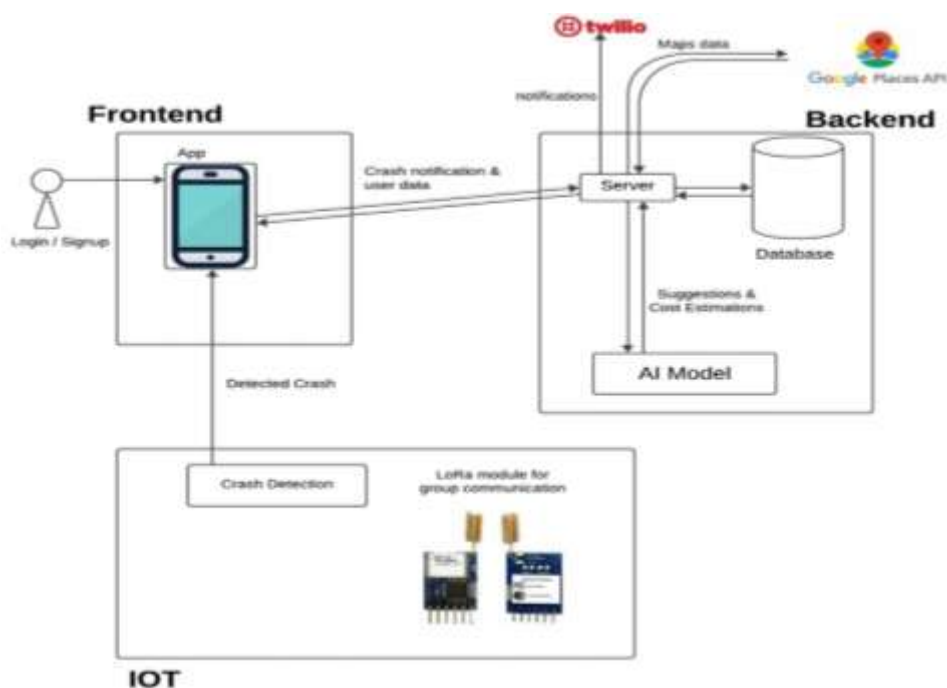
B. Cost Estimation for Long-Distance Bike Riding

Using machine learning techniques, this model predicts the total cost of a long-distance bike ride, including energy expenditure, time, and potential wear-and-tear on equipment. By integrating various data sources such as energy consumption, distance, and terrain complexity, it can offer riders detailed cost breakdowns before and during their journey. Additionally, the system uses real-time data to adjust cost predictions based on rider performance and changing conditions, offering a dynamic and accurate estimate that enhances planning and budgeting.

C. IoT-Enhanced Crash Detection

The IoT-based crash detection system integrates sensors embedded in the bike and rider's gear to monitor various indicators such as acceleration, impact, and rider's health metrics. If a crash or sudden fall is detected, the system sends real-time alerts to emergency contacts or authorities, ensuring quick response and minimizing potential injury severity. By combining data from various IoT sensors, including GPS and motion sensors, the system provides detailed crash data that can be used for safety analysis and further improvement of crash detection algorithms. This integration enhances the safety and reliability of long-distance rides, especially in remote or high-risk areas. The integration of these technologies within a modular architecture enables seamless application across various long-distance bike riding scenarios, offering optimal route planning, real-time cost estimation, and enhanced safety features for riders.

IV. SYSTEM ARCHITECTURE



V. FUTURE DIRECTIONS

Future research should prioritize improving the efficiency of AI models used in RideMate to ensure optimal routing, cost estimation, and crash detection capabilities, even in resource-constrained environments. This could involve exploring model compression techniques, lightweight architectures, and energy-efficient designs to reduce computational overhead without sacrificing performance. Furthermore, integrating AI-driven models with other sensory modalities like GPS, IoT devices, and real-time environmental data could open up new possibilities for more robust user interaction and situational awareness. Key areas for exploration include:

- **Ethical AI:** Ensuring that AI models used for bike routing and crash detection are free from biases, providing equitable and safe experiences for all riders regardless of location or demographic.
- **Augmented Reality (AR):** Integrating AI-based routing models with AR could enable real-time, context-aware navigation, offering riders turn-by-turn guidance, hazard warnings, and route highlights directly within their view, improving overall safety and experience.
- **Cross-Modal Learning:** Developing models that integrate data from multiple sources such as sensors, traffic data, and weather conditions, enabling the AI to dynamically adjust routes, optimize costs, and provide timely crash alerts, improving the overall adaptability and intelligence of the system.

VI. CONCLUSION

In summary, this review explores the transformative potential of machine learning techniques applied to long-distance bike riding, focusing on AI-driven optimal routing, cost estimation, and IoT-enhanced crash detection. The discussed systems leverage advanced AI models to optimize routes based on real-time data, predict and estimate travel costs accurately, and detect potential crashes using IoT devices. These innovations significantly improve the safety, efficiency, and user experience for bike riders, particularly on long-distance journeys. As the technology continues to evolve, models like these will play an essential role in enhancing the interaction between riders and their environments, providing smarter solutions that streamline travel and increase safety. The results emphasize the crucial role of AI in shaping a more efficient and secure future for bike riders, with applications that extend beyond conventional transportation systems, improving the overall quality of travel and safety in everyday riding experiences.

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