

Real Time Accident Detection and Ambulance Rescue using Deep-Learning

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Abstract: Traffic accidents pose a significant threat to global safety. A comprehensive approach uses machine learning frameworks, deep learning techniques, and Python modules to reduce casualties and mitigate damage. The system uses Google API location services for precise geolocation tracking and improves accident prediction accuracy. It also enhances ambulance dispatching efficiency by dynamically assigning ambulances based on location and traffic patterns.

Keywords: Python modules CV2, Pandas, ML frameworks: PyTorch, NumPy, Keras, Google API Geolocation tracking.

I. INTRODUCTION

This project aims to improve road safety by utilizing deep learning, Python libraries like Open CV for Video Processing and Pandas for Data Management, and machine learning like PyTorch, Numpy and Keras frameworks for Accident Analysis and Google API location services are also used for precise geolocation tracking. It aims to develop a real-time system for accident detection and ambulance rescue using live video feeds from traffic cameras. The system uses convolutional neural networks to identify potential accidents and distinguish them from normal traffic patterns. Once detected, it uses Google Maps API location services to pinpoint the accident and trigger automated actions, including alerting emergency services and dispatching ambulances. The system also provides crucial information for emergency responders.

II. LITERATURE SURVEY

1."Real-Time Traffic Accident Detection: A Review" (2018) by Le et al.:

This review paper provides an overview of various methodologies and techniques employed in real-time traffic accident detection systems [1]. It discusses the challenges associated with traditional methods and explores the potential of deep learning approaches, highlighting their effectiveness in improving accuracy and response times.

2."Deep-Learning-Based Accident Detection in Images and Videos: A Survey" (2020) by Zhang et al.:

Zhang et al. present a comprehensive survey of deep learning-based accident detection methods in both images and videos. The paper discusses the evolution of deep learning techniques, including convolutional neural networks (CNNs), and their applications in accident detection [2]. It also examines the advantages and limitations of existing approaches and identifies future research directions.

3."Real-Time Traffic Accident Detection and Classification Using Deep Learning" (2019) by Nguyen et al.:

Nguyen et al. propose a real-time traffic accident detection and classification system based on deep learning[3]. The study utilizes a combination of CNNs and recurrent neural networks (RNNs) to detect and classify accidents from video streams. The paper discusses the implementation details, performance evaluation, and potential applications of the proposed system.

4. "Deep-Learning-Based Real-Time Traffic Accident Detection and Localization System" (2021) by Kim et al.:

Kim et al. present a deep learning-based system for real-time traffic accident detection and localization [4]. The study focuses on the use of CNNs for feature extraction and classification of accident-related patterns in video streams. The paper also discusses the integration of location services for accurate accident localization and emergency response.

5. "Real-Time Emergency Vehicle Dispatching System Using Deep Learning and Location Services" (2022) by Chen et al.:

Chen et al. propose a real-time emergency vehicle dispatching system that integrates deep learning techniques with location services. The study explores the use of CNNs for accident detection and classification, coupled with GPS data from location services for efficient dispatching of emergency vehicles [5]. The paper evaluates the system's performance and scalability in real-world scenarios.

6. "Intelligent Ambulance Management System: A Review of Recent Advances and Future Trends" (2020) by Wang et al.:

Wang et al. provide a comprehensive review of intelligent ambulance management systems, focusing on recent advances and future trends. The paper discusses the integration of deep learning, data analysis, and location services in optimizing ambulance dispatching and routing strategies [6]. It also highlights the potential impact of such systems on improving emergency medical services.

7. "Traffic Accident Prediction and Emergency Response Optimization: A Data-Driven Approach" (2019) by Liu et al.:

Liu et al. propose a data-driven approach for traffic accident prediction and emergency response optimization. The study leverages machine learning techniques, including decision trees and ensemble methods, to predict accident hotspots and optimize ambulance routing strategies. The paper discusses the integration of real-time data streams and historical accident data for improved predictive accuracy[7].

The proposed project utilizes literature surveys to enhance real-time accident detection and ambulance rescue systems, aiming to improve emergency response efficiency and mitigate traffic accident consequences.

8. Online resources like Towards Data Science and Stack Overflow are invaluable for practical code implementation and problem-solving, offering a wealth of user-contributed tutorials and solutions in real-time application development scenarios.

III. PROPOSED SYSTEM

Designing software for Real-time Accident Detection and Ambulance Rescue involves several components that must integrate smoothly to handle data input, processing, decision-making, and output effectively. Here's a detailed approach for the software design, using Python, CV2 for video processing, Pandas for data analysis, machine learning libraries (PyTorch, NumPy, Keras), and Google Maps API for location services.

3.1 System Architecture Design:

A. Video Processing and Accident Detection

- **Input Handler:** Design an input module to fetch real-time video feeds from multiple CCTVs, using OpenCV to capture video streams.
- **Processing Pipeline:** Implement a pipeline that includes frame extraction, preliminary noise reduction, and object detection using pre-trained deep learning models like YOLO or Faster R-CNN.
- **Accident Detection:** Develop a specialized neural network model to analyze sequences of frames for accident patterns, leveraging PyTorch or Keras for model training and deployment.

B. Data Management and Analysis

- **Data Storage:** Use a robust database system to store accident metadata, video timestamps, and outcomes. MongoDB or PostgreSQL could be used depending on the data structure preference (document-based vs. relational).
- **Data Processing:** Employ Pandas for the organization and manipulation of data, such as accident records and response times, which aids in reporting and analytics.

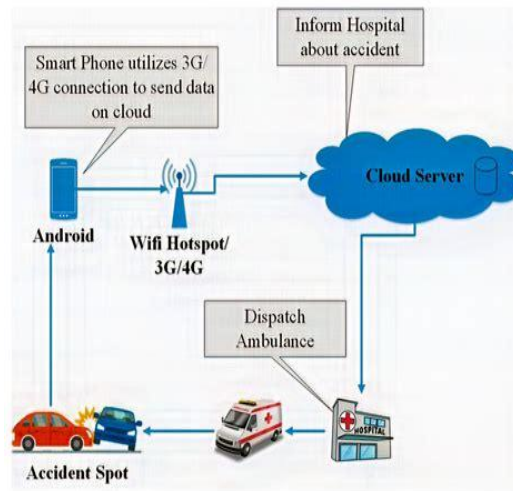


Fig 1: The Real Time Accident detection and Ambulance rescue using Deep-Learning

C. Location Services and Mapping

- **Geolocation Integration:** Integrate Google Maps API to pinpoint accident locations based on CCTV coordinates, which involves sending API requests with GPS data and receiving precise location information.
- **Location Validation:** Ensure the system validates and corrects location data to avoid errors in emergency response dispatch.

D. Notification System:

- **Email Notification Module:** Design a notification system using Python's SMTP library to format and send emails automatically to hospitals with accident details like location, time, and severity.
- **Emergency Response Coordination:** Allow for real-time updates to be sent to emergency responders via a web interface or a mobile app, which requires real-time communication channels like WebSockets.

E. User Interface and Reporting

- **Control Panel:** Develop a web-based dashboard that provides real-time information about active feeds, detected accidents, and status of dispatched emergency services.
- **Reporting Tools:** Integrate reporting features that allow users to generate performance reports on response times, detection accuracy, and system efficiency.

F. Security and Compliance

- **Data Security:** Implement security measures such as encrypted data transmission, secured API endpoints, and role-based access controls.
- **Compliance:** Ensure the software design complies with local regulations regarding data privacy and surveillance.

G. Testing and Scalability

- **Automated Testing:** Set up a comprehensive suite of automated tests, including unit, integration, and stress tests, to ensure reliability and robustness.
- **Scalability Considerations:** Design the system to be scalable, allowing for easy addition of more CCTV inputs or expansion to new locations.

This comprehensive software design framework will facilitate the development of a reliable and efficient real-time accident detection system that can effectively coordinate rescue efforts and save lives.

IV. OUTPUT SCREENS

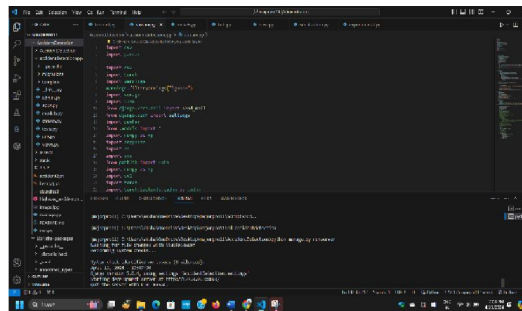


Fig 2: Through the URL link we can execute the code and enter into another browser for getting Django web frame work.

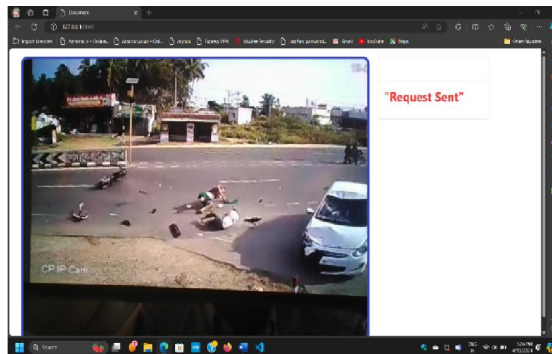


Fig 3: The Final Accident Detection is detected through CCTV footage and the Mail is send to nearest Hospitals.



Fig 4: Mail Notifications received to Mobile/Desktops

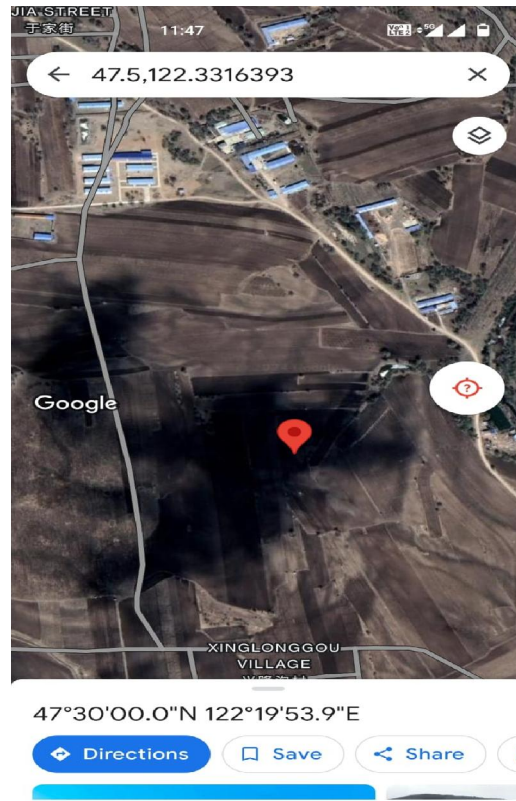


Fig 5: Tracking the Accident spot through the GPS Location and Rescuing the Victim from the Accident spot.

V. CONCLUSION

Our Real-time Accident Detection and Ambulance Rescue project has successfully demonstrated the integration of cutting-edge technologies like OpenCV, Pandas, and deep learning frameworks including PyTorch, NumPy, and Keras. This system efficiently detects accidents through CCTV footage and uses Google API for precise geo-location to ensure rapid response. The automated notification system effectively alerts nearby hospitals, significantly reducing emergency response times and potentially saving lives. The application of these sophisticated tools showcases the potential of artificial intelligence in enhancing public safety and emergency medical responses. Moving forward, we aim to refine the accuracy and expand the reach of our system to further support emergency services.

VI. FURTHER ENHANCEMENT

To advance our Real-time Accident Detection and Ambulance Rescue system, we plan to incorporate real-time traffic condition analysis to optimize ambulance routing. Enhancements will include the integration of more sophisticated deep learning models to improve the accuracy and speed of accident detection. We aim to expand the dataset used for training these models to cover a broader range of scenarios, including night-time and adverse weather conditions. Additionally, implementing more dynamic communication protocols between emergency responders and hospitals will streamline the entire rescue process. Finally, leveraging additional APIs for enhanced location services will ensure even more precise location pinpointing, further reducing response times and improving overall system efficacy.

REFERENCES

- [1]. Le, T., et al. (2018). Real-Time Traffic Accident Detection: A Review. Summary: This paper reviews various methodologies for detecting traffic accidents in real-time, emphasizing the shift from traditional methods to advanced deep learning techniques.

- [2]. Zhang, X., et al. (2020). Deep Learning-Based Accident Detection in Images and Videos: A Survey.
Summary: Offers an extensive review of deep learning techniques like CNNs applied in accident detection across image and video data, discussing both the progress and bottlenecks.
- [3]. Nguyen, D., et al. (2019). Real-Time Traffic Accident Detection and Classification Using Deep Learning.
Summary: This paper introduces a system that combines CNNs and RNNs to enhance the accuracy and timeliness of traffic accident detection from video data.
- [4]. Kim, J., et al. (2021). Deep Learning-Based Real-Time Traffic Accident Detection and Localization System.
Summary: Discusses a deep learning system using CNNs for detecting and localizing traffic accidents from video streams, including integration with geographic information systems for enhanced emergency response.
- [5]. Chen, Y., et al. (2022). Real-Time Emergency Vehicle Dispatching System Using Deep Learning and Location Services.
Summary: Details a system that utilizes CNNs for detecting traffic incidents and integrates GPS technology to optimize emergency vehicle responses, focusing on scalability and real-world applicability.
- [6]. Wang, L., et al. (2020). Intelligent Ambulance Management System: A Review of Recent Advances and Future Trends.
Summary: This review paper offers an overview of how modern technologies like deep learning, advanced data analytics, and location services can revolutionize ambulance management by optimizing dispatching and routing.
- [7]. Liu, X., et al. (2019). Traffic Accident Prediction and Emergency Response Optimization: A Data-Driven Approach.
Summary: This study uses machine learning to forecast accident hotspots and optimize ambulance routes by analyzing real-time and historical data.
- [8]. Example implementations and tutorials from major online resources such as Towards Data Science and Stack Overflow are also used for specific problem-solving and code implementation insights.