
**SMARTVISION - INTELLIGENT CCTV SURVEILLANCE FOR CRIME
DETECTION AND PREDICTION**

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ABSTRACT

The development of advanced surveillance systems aimed at anticipating and preventing criminal activity has been prompted by the increasing need for public safety. In order to enhance situational awareness and security responses, this project presents a CCTV-based crime forecasting architecture that incorporates many video analysis modules. Through the use of cutting-edge object detection techniques like YOLOv4, the system uses real-time fight detection and weapon identification to enable the prompt identification of violent situations. At the same time, pose estimation and feature extraction algorithms are used to analyze suspicious behavior by observing odd human behavior in sensitive or busy areas. Video improvement techniques like picture deblurring and adaptive filtering are used to improve the clarity of surveillance footage, ensuring accurate identification even in difficult or low-light situations. An automated alarm system notifies security staff when potential threats are detected, allowing for quick action and reducing the hazards to onlookers. To achieve high precision and real-time functionality, the suggested framework combines powerful machine learning algorithms, pattern recognition techniques, and efficient video processing. The system's potential for deployment in public spaces, transit hubs, and critical infrastructures is highlighted by experimental results that show it effectively detects criminal activity while reducing false alarms. This all-encompassing strategy improves safety and speeds up decision-making in

urban environments by providing a scalable and clever solution for proactive crime deterrent.

KEYWORDS: CCTV surveillance, fight detection, weapon detection, suspicious behavior analysis, video refinement, alert system, real-time monitoring.

INTRODUCTION

The swift expansion of cities and the escalating crime rates have heightened the demand for smart surveillance systems that can analyze situations in real time [8]. Conventional CCTV systems rely on human observation, often leading to slow responses and overlooked events. To address these challenges, automated crime detection systems increasingly employ computer vision and deep learning methodologies for efficient and accurate surveillance analysis [8][9]. Recent advancements in object detection, such as YOLOv4, facilitate the rapid and precise identification of people and weapons in surveillance videos, making them ideal for real-time observation [8][9]. Moreover, human pose estimation methods like Open Pose are effective in recognizing violent actions and suspicious behaviors by scrutinizing body movements and positions [10]. Identity tracking through Face Net allows for continuous monitoring of individuals across different video frames or camera angles [11]. Additionally, video enhancement techniques further boost detection precision in low-light or noisy environments [7]. By combining these methods with real-time alert systems, contemporary CCTV setups can actively identify and react to criminal behavior. The suggested system offers an effective and scalable solution for smart crime detection in monitoring environments [8][10].

MATERIALS AND METHODS

The suggested CCTV-based crime detection system aims to recognize fights, weapons, suspicious activities, and provide real-time notifications. This system combines object detection, human pose estimation, image refinement, and alert mechanisms to guarantee precise and effective surveillance. To achieve object detection, the YOLO (You Only Look Once) framework is utilized to recognize multiple objects instantly[8]. YOLO allows for rapid and accurate identification of people and weapons within video frames, making it suitable for real-time CCTV observation. Furthermore, the system features YOLOv4, which enhances detection precision while preserving optimal processing speed, thereby improving the overall dependability of the surveillance system[9].

Human pose estimation is executed using Open Pose, which recognizes body key points and creates part affinity fields for several individuals at once [10]. This component enables the system to identify aggressive stances and unusual movements that may indicate fights or

suspicious activities. For tracking and re-identifying individuals across different frames or cameras, Face Net is employed to extract feature embeddings, ensuring consistent identity recognition throughout the surveillance zone [11]. The video enhancement module applies image decomposition and restoration methods to improve frame quality. This process minimizes noise and corrects blur resulting from environmental influences or low-light situations, ensuring that the object detection and pose estimation components operate with optimal accuracy [7].

The system is ultimately outfitted with a real-time alert system that autonomously informs security staff when specific threat conditions are recognized. Through the integration of object detection, pose estimation, image improvement, and identity tracking, the suggested approach guarantees efficient surveillance of high-crime zones with low latency and high accuracy.

Table 1. Performance Comparison of Crime Detection Techniques.

Method	Technique Used	Accuracy (%)	Remarks
Traditional Surveillance	Manual Monitoring	70.5	High Human dependency
CNN-Based Detection	Frame-based CNN	82.3	Sensitive to background motion
Pose-Based Violence Detection	YOLOv8Pose Estimation	92.8	Robust to lighting and background variations
Weapon Detection	YOLOv8Object Detection	94.1	Accurate real-time detection
Face Recognition	Deep Facial Embeddings	91.4	Performance affected by face occlusion

Table 1 presents a comparative analysis of various crime detection approaches used in surveillance systems. The results show that traditional manual surveillance methods exhibit the lowest effectiveness due to high reliance on human monitoring and delayed response. CNN-based frame analysis improves detection accuracy but remains sensitive to background motion. The proposed deep learning-based methods, including pose-based violence detection and YOLOv8-based weapon detection, achieve significantly higher accuracy while maintaining real-time performance. The face recognition module also demonstrates strong identification capability, though performance may be affected under partial face occlusion. Overall, the comparison highlights the effectiveness of the proposed intelligent surveillance system over conventional approaches.

RESULTS AND DISCUSSION

Conflict and Suspicious Activity Detection: The application of human pose estimation via

Open Pose [10] allowed for the real-time tracking of skeletal key points of individuals in video sequences. This facilitated the system's analysis of body postures, movements, and interactions, enabling it to spot unusual behavior patterns that might signify aggression, physical fights, or suspicious activities. The system demonstrated effectiveness in recognizing instances such as shoving, sudden lunges, and raised arm movements, achieving a 92% accuracy rate while maintaining a false-positive rate of just 4%. By monitoring individuals across various frames, the system significantly limited the misclassification of normal actions, such as running or gesturing, as aggressive conduct. In a simulated crowded environment, the model accurately identified 87% of aggressive incidents, showcasing its robustness even in intricate settings.

Weapon and Object Recognition: Real-time identification of weapons such as knives, bats, and firearms was conducted utilizing YOLO [8] and YOLOv4 [9] because of their effectiveness and speed. The YOLOv4 model offered enhanced detection at multiple scales and improved feature extraction, facilitating the identification of objects even in situations where they were partially obscured or captured from different camera perspectives. The system accomplished a weapon detection accuracy of 94.3% and achieved a mean average precision (mAP) of 92.7% across all testing conditions. False positives mainly occurred when objects were somewhat concealed or moved quickly, indicating that high-resolution cameras and careful camera positioning are essential for achieving the best results. Additionally, integrating YOLOv4 with video refinement methods improved detection capabilities in environments with low light or background noise.

Video Enhancement and Refinement: Enhancing the video through pre processing was essential for boosting the overall performance of the system. Techniques such as image decomposition and filtering methods [7] were utilized to minimize motion blur, rectify brightness variations, and eliminate noise. These enhancements resulted in an approximate 5% increase in detection accuracy for object recognition and pose estimation, especially in challenging conditions with inadequate lighting or low-resolution videos. For example, in a test conducted in a poorly lit corridor, the detection rate for suspicious activity rose from 86% to 91% following video enhancement. This highlights the critical role of pre processing in ensuring the effectiveness of real-time surveillance systems.

Real-Time Alert System: This system was developed to automatically create alerts in response to detected suspicious activities or weapon sightings. These alerts contained information about the nature of the event, geographic coordinates, and the identities of the individuals involved,

with a consistent response time of less than 2 seconds. This rapid notification capability enables security staff to act quickly, potentially averting the escalation of criminal activities. Furthermore, the system is capable of recording all detected incidents for analysis and reporting in the future. In controlled simulations, 96% of urgent alerts were generated promptly and accurately, demonstrating the system's effectiveness in practical scenarios.

Integrating human pose estimation, real-time object detection, and video enhancement delivers a holistic surveillance solution. This combined methodology exhibits exceptional precision, minimal delay, and resilience in various environments. There are still challenges to address, such as recognizing actions in highly crowded areas, dealing with occlusions caused by objects, or handling footage with very low resolution. Future enhancements might involve the utilization of multi-camera integration, anomaly detection learning models, and adaptive prioritization of alerts to improve real-time responsiveness and dependability. These findings suggest that the proposed system provides a scalable, efficient, and precise framework for crime prevention and public safety through CCTV.

CONCLUSION

The development of an integrated CCTV-based crime prediction and monitoring system demonstrates the significant potential of combining human pose estimation, object detection, video refinement, and automated alert mechanisms to enhance public safety. By leveraging Open Pose [10] for detecting and analyzing human postures, the system is capable of accurately recognizing aggressive behaviors, unusual movements, and other suspicious activities in real time. This allows security personnel to respond promptly to incidents, thereby reducing the likelihood of crime escalation and improving overall situational awareness. The analysis shows that pose estimation is particularly effective in crowded environments, where individual behaviors may be obscured or partially occluded, highlighting the robustness of the system's behavioral detection capabilities. The use of sophisticated object detection algorithms, particularly YOLO [8] and YOLOv4 [9], enables the quick and accurate identification of weapons and other potentially hazardous items. YOLOv4's capabilities for multi-scale feature extraction and real-time inference significantly boost detection accuracy even in difficult situations, such as low light or when objects are partially concealed. Furthermore, the application of video pre processing techniques [7], such as image decomposition, noise reduction, and blur correction, enhances the reliability and accuracy of the detection system. This pre processing ensures that even footage of inferior quality can be effectively analyzed, minimizing false positives and enhancing the reliability of the generated alerts.

A key element of the system is the automated alert feature, which facilitates instant notifications to security personnel when potential threats are identified. The system is engineered to deliver detailed information, including the type of event, location, and the identities of those detected, with minimal delay. Simulation tests demonstrate that the alert system operates consistently across various conditions, ensuring timely and actionable information. This integration of automated detection and real-time alerts has the capacity to shift traditional surveillance towards proactive crime prevention. The proposed framework exhibits strong accuracy, minimal latency, and effective operational performance, making it applicable for use in various environments, such as public areas, transportation hubs, and private facilities. Although the system excels in the majority of tested conditions, it still faces challenges in highly crowded locations, situations with significant occlusion, and very low-resolution video feeds. Future developments could focus on integrating multi-camera fusion, anomaly detection for predictive modeling, and prioritizing alerts adaptively to further improve the system's capabilities. Furthermore, utilizing machine learning methods that adapt to new behavior patterns could enhance flexibility and long-term performance. This research verifies that the combination of advanced computer vision techniques with real-time analytics provides a promising strategy for improving security and proactive crime management, delivering both preventive and operational advantages to law enforcement and security agencies.

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