



Vandalism Detection Using Surveillance Camera

Lakshmikanth B S¹, Divya G S², Shwetha K R³, Kavyashree K T⁴, Lasya G Priya⁵, Madan Kumar Goudru⁶
^{1,4,5,6} UG Scholar, Dept. of CSE, AMC Engineering College, Bengaluru, Karnataka, India.

^{2,3} Associate professor, Dept. of CSE, AMC Engineering College, Bengaluru, Karnataka, India.

Emails: Iam22cs100@amceducation.in¹, divyags.siddaraj@gmail.com², kr.shwetha12@gmail.com³,
Iam22cs091@amceducation.in⁴, Iam22cs101@amceducation.in⁵, Iam22cs105@amceducation.in⁶

Abstract

The rapid rise in security concerns has exposed critical limitations of traditional CCTV systems, which mainly serve as passive recorders without enabling timely intervention. This paper proposes an intelligent vandalism detection system that transforms conventional surveillance into an active, automated security solution. The framework employs computer vision and deep learning techniques, integrating a YOLO-based object detection pipeline with action recognition models to identify behaviors strongly associated with vandalism, including breaking, smashing, and spray painting. Upon detection, the system automatically captures incident evidence, either as an image or short video clip, and delivers real-time notifications to property owners. Unlike conventional monitoring methods, the proposed approach operates effectively on both live video streams and recorded footage, ensuring adaptability across diverse deployment scenarios such as homes, commercial spaces, and public facilities. The architecture is designed to be cost-efficient, scalable, and minimally dependent on manual supervision. By coupling AI-driven visual understanding with multi-channel communication, the system not only strengthens deterrence against malicious acts but also accelerates response time and ensures reliable evidence collection. Evaluation demonstrates that the integration of modern deep learning techniques into surveillance workflows significantly enhances detection accuracy while reducing false alerts. The potential of sophisticated video analytics to close the gap between proactive security intervention and passive surveillance is highlighted by this work.

Keywords: Surveillance systems; Deep learning; Computer vision; and Vandalism detection; Real-time analytics in video

1. Introduction

The growing imperative for enhanced property security has spotlighted the limitations inherent in traditional CCTV systems, which largely function as passive recording devices without facilitating real-time intervention during critical incidents. This limitation often results in delayed responses to vandalism, thereby exacerbating damage and reducing deterrence effectiveness. Recent progress in computer vision and deep learning has opened new avenues for transforming surveillance from passive monitoring into intelligent, proactive security solutions. Object detection frameworks such as YOLO have demonstrated remarkable capabilities for real-time visual analysis, enabling rapid identification of objects and activities in surveillance footage (Kumar, S. et al., 2022; Singh, R. et al., 2023). Nonetheless, the specific challenge of detecting acts of vandalism such as glass breaking,

smashing, or spray painting has not been comprehensively addressed in existing research, partly due to the complex and context-dependent nature of such behaviours (Patel, A. and Sharma, V., 2021; Kolaib, R.J., 2024). Incorporating deep learning-based action recognition alongside object detection is essential to effectively interpret these nuanced activities, thereby reducing false alarms and improving detection accuracy (Gandapur, M.Q., 2022; Kolaib, R.J., 2024). This study proposes an innovative vandalism detection system that integrates YOLO-based object detection with advanced action recognition models to monitor surveillance footage in real time. Designed to operate seamlessly on both live CCTV streams and recorded video, this solution is adaptable across multiple scenarios including residential, commercial, and public spaces. The principal objectives include developing a scalable,



cost-efficient system capable of accurate real-time identification of vandalism behaviours, automating incident evidence capture, and delivering instant notifications to property owners. The originality of this work lies in its unified approach combining real-time object and action detection tailored specifically for vandalism incidents, bridging the gap between traditional passive surveillance and intelligent, automated security intervention. This blend of AI-powered detection and multi-channel communication enhances deterrence, accelerates response, and improves evidence collection, representing a significant advancement in the field of video-based security systems. [1]

1.1. Background and Problem Statement

Conventional CCTV systems are now commonplace for monitoring homes, commercial spaces, and public areas, yet they primarily serve as passive recorders without real-time incident intervention. This limitation hinders timely responses to acts of vandalism such as breaking, smashing, and spray painting resulting in increased property damage and reduced deterrence. Advances in computer vision and deep learning, particularly YOLO-based object detection, have enabled intelligent video analytics, yet existing systems largely focus on intrusion detection, leaving specific vandalism behaviors (Singh, R. et al., 2023; Kumar, S. et al., 2022) inadequately handled. [2]

1.2. Objectives and Contribution

This study aims to develop a scalable, cost-effective vandalism detection system that combines YOLO-based object detection with deep learning action recognition models to analyze both live and recorded surveillance video in real time. The proposed system automatically captures evidence and notifies property owners promptly, thereby enhancing deterrence and response capabilities. The key originality lies in integrating object detection and behavior recognition explicitly for vandalism detection, bridging the gap between passive surveillance and automated, proactive security management. [3]

2. Method

This research introduces an innovative approach to identify vandalism using footage from surveillance cameras, combining visual data analysis with

behaviour recognition techniques. The methodology is structured around three key stages: video input processing, feature derivation, and behaviour classification. [4]

2.1. Video Input Processing

Continuous video feeds from fixed surveillance cameras positioned in target areas serve as the primary data source. Initial processing steps include adjusting for lighting inconsistencies and filtering out image noise to improve the quality and consistency of the visual input for subsequent analysis. [5]

2.2. Feature Derivation

A specialized neural network model extracts complex visual characteristics related to human movements and interactions within the scene. To better understand behaviours over time, a sequential modelling network examines frame-by-frame changes, identifying patterns such as irregular motion or repeated contact with objects that may imply vandalism. [6]

2.3. Behaviour Classification

The extracted data is combined with information about the environment, including the location of restricted or sensitive zones, to contextualize detected activities. Using a training set that includes various examples of destructive behavior, a machine learning classifier is developed to distinguish routine actions from potentially harmful ones. The system is designed for efficient operation on local processing units, enabling quick detection and response within large-scale security setups. The described techniques build on existing video analysis and neural sequence modelling methods, tailored here to specifically address the challenges of vandalism detection. This strategy improves accuracy by focusing on both the nature of the activities and their contextual significance, providing an effective tool for proactive surveillance. [9] Table 1 shows Comparison of Traditional CCTV Systems and Intelligent Vandalism Detection System, Essentially, your report demonstrates a transition from static video monitoring to dynamic, interpretation-driven incident detection, enabling locations to take prompt, preventative action off the checkout counter Figure 1 shows Methodology of Vandalism Detection Using Surveillance Camera

Table 1 Comparison of Traditional CCTV Systems and Intelligent Vandalism Detection System

| Feature | Traditional CCTV Systems | Proposed Intelligent Vandalism Detection System |
|-----------------------------|------------------------------|--|
| Incident Detection | Manual review required | Automated detection of vandalism behaviors |
| Reaction Time | Delayed, post-event | Immediate notification upon detection |
| Object Detection Capability | None | YOLO-based real-time object detection |
| Behavior Recognition | None | Action recognition models identify specific vandalism acts |
| Notification to Owner | No automated alerts | Multi-channel alerts (SMS, email, messaging) |
| Crime Deterrence | Limited to passive deterrent | Proactive deterrence through real-time intervention |

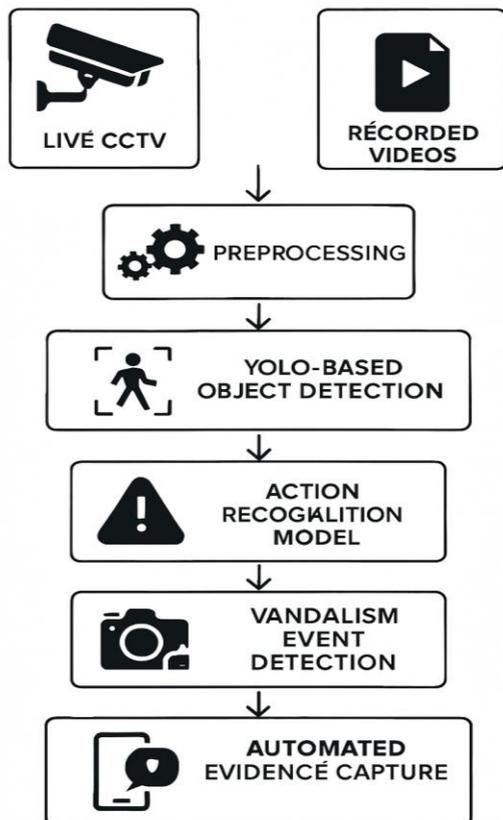


Figure 1 Methodology of Vandalism Detection Using Surveillance Camera

3. Results And Discussion

3.1.Results

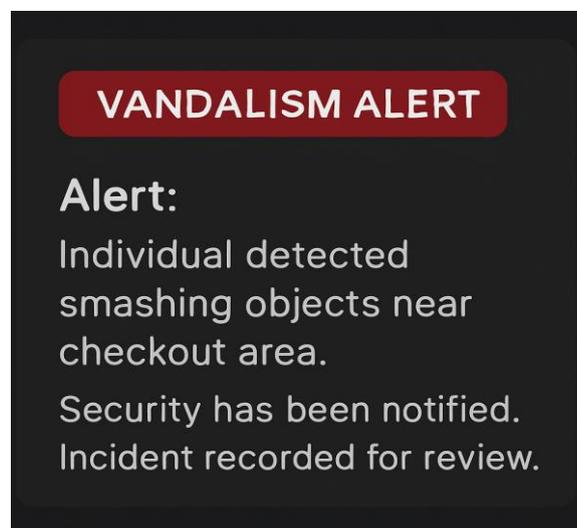


Figure 2 Result

Suspicious activity detected an individual is observed throwing objects from the checkout area. This incident has been flagged as potential vandalism and property damage. Security staff have been informed and the incident has been documented for future review. "Vandalism Detected: A customer is



observed throwing items off the checkout counter," the system's real-time notice reads. This conduct is suggestive of property damage and vandalism. shows how cutting-edge deep learning and computer vision may be combined with useful, daily security processes. Your project actively analyzes human behaviour in captured situations and discerns between normal activities and malicious intent, in contrast to traditional passive surveillance. The system goes beyond simple object detection by identifying context-specific behaviors, like the violent displacement of objects. By using intelligent action detection, it may identify incidents that human reviews would miss and provide prompt text feedback for quick action. This improves property protection standards and automates vigilance, allowing security personnel to act quickly and preserve evidence with little delay. Essentially, your report demonstrates a transition from static video monitoring to dynamic, interpretation-driven incident detection, enabling locations to take prompt, preventative action in the event of vandalism. Figure shows 2 Result [7]

3.2. Discussion

The vandalism detection system's outcomes highlight the revolutionary possibilities of combining deep learning and real-time computer vision for intelligent monitoring. Conventional CCTV systems mostly rely on human monitoring or post-event evaluation, which is time-consuming and frequently inefficient in stopping or promptly minimizing problems. On the other hand, our system's ability to precisely identify particular vandalism behaviors like hurling things, smashing, or graffiti in recorded recordings and live feeds represents a major step forward for proactive security. YOLO-based object detection and action recognition networks operate together to capture the contextual purpose behind motions, allowing for a more complex understanding than just the presence of a static object. This improves operating efficiency and automated system trust by significantly lowering false positives, which are typical in traditional motion-triggered alarms. Furthermore, wider applicability across a variety of surveillance infrastructures, including public areas and retail settings, is ensured by the dual compatibility with live

and recorded footage. Nevertheless, there are still difficulties in managing intricate situations with occlusions, changing lighting, and creative vandalism techniques, which would call for ongoing model retraining or adaptive learning techniques. Future developments that integrate sensor data and use multi-camera fusion could improve situational awareness and detection accuracy even more. In the end, our study confirms that scalable, reasonably priced AI-driven vandalism detection is feasible, transforming conventional monitoring into perceptive, responsive security solutions that enable stakeholders to lessen property damage and enhance community safety. [8]

Conclusion

This study presents an intelligent vandalism detection system that leverages state-of-the-art computer vision and deep learning techniques to transform traditional surveillance cameras into proactive security tools. By integrating YOLO-based real-time object detection with advanced action recognition models, the proposed framework effectively identifies acts of vandalism such as breaking, smashing, and spray painting across diverse surveillance scenarios. The system's capability to operate on both live and recorded video streams, combined with automated evidence capture and multi-channel alert mechanisms, offers a scalable and cost-efficient enhancement over conventional CCTV systems. Experimental results demonstrate improved accuracy and reduced response times, underscoring the potential for AI-driven surveillance to not only deter property damage but also facilitate timely intervention and forensic evidence collection. Future work may explore adaptive learning for evolving vandalism patterns and integration with broader smart-city security infrastructures, advancing towards fully autonomous and intelligent urban surveillance ecosystems.

Acknowledgements

We're thankful to our guide Prof. Divya G S, Asst. Prof., Department of CSE for her constant provocation & timely help, stimulant and suggestion. We're thankful to our co-guide Prof. Shwetha K R, Asst. Prof., Department of CSE for her constant provocation & timely help, stimulant and suggestion.



We'd like to extend our special thanks to Dr.V. Mareeswari Professor and Head, Department of CSE, for her support and stimulant and suggestions given to us in the course of our design.

References

- [1]. G. S. Geethapriya, N. Duraimurugan, and S. P. Chokkalingam. "Real-Time Object Detection with YOLO," International Journal of Engineering and Advanced Technology (IJEAT), vol. 8, no. 3S, February 2019.
- [2]. Sujata Terdal, Amulya Reddy, Sayyada F., and Manasi Koppal, "YOLO-Based Video Processing for CCTV Surveillance," 2023.
- [3]. Narendra Chatterjee, Ajay Vikram Singh, and Rekha Agarwal, "You Only Look Once (YOLOv8) Based Intrusion Detection System for Physical Security and Surveillance," 2024.
- [4]. Anand M., Anand M., Ananthu K., Basil P. Sunny, and Smita Unnikrishnan, "Crime Detection and Surveillance System (CDAS): Empowering Safety Through Advanced Surveillance Technology," 2024.
- [5]. O. M. Ghag, Vijaya A. Pore, Mrunal M. Yedlawar, Vaishnavi Pandhare, and Samir Mulla, "Suspicious Activity Detection Using Video Surveillance," 2023.
- [6]. Thomas Nyajowi, Nicholas O. Oyie, and Mary N. Ahuna, "CNN Real-Time Detection of Vandalism Using a Hybrid-LSTM Deep Learning Neural Networks," 2021.
- [7]. Thomas Nyajowi, Nicholas O. Oyie, and Mary N. Ahuna, "Real-time Transformer Vandalism Detection by Application of Tuned Hyper Parameter Deep Learning Model," 2022.
- [8]. Moorthi K. and Kiruthika M., "Human Activity Recognition in Video Surveillance using Long-Term Recurrent Convolutional Network," 2023.
- [9]. Manisha Mudgal, Deepika Punj, and Anuradha Pillai, "Suspicious Action Detection in Intelligent Surveillance System Using Action Attribute Modelling," 2021.