



International Journal of Multidisciplinary Research in Science, Engineering and Technology

(A Monthly, Peer Reviewed, Refereed, Scholarly Indexed, Open Access Journal)



Impact Factor: 8.206

Volume 8, Issue 8, August 2025



International Journal of Multidisciplinary Research in Science, Engineering and Technology (IJMRSET)

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GhostCam: Intelligent Multi-Context System for Real-Time Incident Detection

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ABSTRACT: The current paper introduces an AI-based cybersecurity camera advanced system tailored for real-time public safety and behaviour monitoring. With the era of exponential digital growth and increasing cybersecurity threats, there is an urgent need for innovative surveillance systems that can ensure both covert monitoring and advanced threat detection. Addressing this need, our project, titled GhostCam, is a unique AI- and ML-powered smart surveillance system developed using Python. This system functions as a disguised or hidden camera embedded with intelligent capabilities such as motion detection, facial recognition, intrusion detection, and real-time anomaly analysis, designed specifically for cybersecurity, home security, digital forensics, and covert intelligence operations

GhostCam distinguishes itself from conventional CCTV or webcam-based systems through its stealth mode functionality, encryption-based data handling, and machine learning-driven behaviour analysis, making it a next-generation tool for digital surveillance.

The primary objective of GhostCam is to provide an invisible layer of security, one that does not alert intruders or attackers while recording and analyzing the environment

KEYWORDS: Advanced cyber security System, AI and ML Learning, Motion detection, Monitoring, machine learning-driven behaviour analysis, recording and analyzing the environment

I. INTRODUCTION

The increasing demand for greater public safety. In today's rapidly evolving digital world, security and surveillance systems are becoming more critical than ever before. With the rise of cybercrimes, unauthorized access, and physical threats, the demand for intelligent, discreet, and highly efficient surveillance tools has surged dramatically. Traditional CCTV and webcam systems, while useful, are often limited by visibility, manual monitoring, lack of smart detection, and vulnerability to tampering. To address these limitations, our project introduces GhostCam, an advanced, AI- and ML-driven spy surveillance system developed using Python. Designed for covert operations and enhanced digital security, GhostCam integrates modern computer vision techniques, machine learning algorithms, and data encryption to deliver a comprehensive and smart monitoring solution. GhostCam functions as a hidden or disguised camera capable of real-time facial detection, motion tracking, behavior analysis, and anomaly detection. In addition, the project incorporates lightweight encryption protocols to protect captured data and ensure privacy during storage or transmission.

II. LITERATURE SYRVEY

The accelerated advancement of artificial intelligence and deep learning has transformed advanced security system into systems that can monitor behavior in real-time and detect events.

[1] introduced YOLO, showing that a single CNN can perform real-time object/motion detection suitable for live camera feeds and low-power deployment. Embedding-based face recognition frameworks such as FaceNet [2] provide a compact, high-accuracy basis for GhostCam's "add member → identify" workflow by mapping faces to discriminative vector space for fast database lookup. Advances in sound-event detection and DCASE challenge results demonstrate robust approaches for recognizing and localizing noise events (shouts, alarms, crashes), which directly support GhostCam's noise-detection module.



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[3]. Video anomaly and event-detection surveys and recent deep-learning work outline effective supervised, semi-supervised and self-supervised methods to decide when to record, flag, or delete footage—helpful for GhostCam’s automatic recording/delete [4][5]. Finally, surveys on biometric template protection and cancelable biometrics underscore the need for “privacy-by-design” in member management and secure template storage, motivating encryption and template-protection in GhostCam’s architecture.

[6]. Building on these established techniques, GhostCam integrates motion, noise, identification, and recording modules into a modular, privacy-conscious security platform.

EXISTING SYSTEM

Current advanced security systems mostly depend on deep learning methods for human activity and object detection and tracking in real time. Most systems use traditional CCTV and modern Intelligent Video Surveillance Systems (IVSS) integrated with AI-based features. These systems often employ object detection models, facial recognition, and motion tracking to monitor specific areas. Some advanced setups also use audio sensors for detecting unusual sounds. However, most existing solutions operate as isolated modules, lacking unified integration of motion detection, noise detection, identification, and automated recording with deletion options. Additionally, many rely on centralized storage without strong encryption, creating potential security vulnerabilities—highlighting the need for GhostCam’s modular, privacy-focused, and multi-context AI-driven architecture.

PROPOSED SYSTEM

GhostCam (Cypher Cam Master) introduces a unified, deep learning-driven security platform capable of real-time detection and multi-context analysis. Unlike siloed solutions, GhostCam combines motion detection, noise video recording into a single modular architecture centered on a YOLO-based detection core augmented with lightweight detection, identification, and intelligent neural branches for audio-event classification and embedding-based face identification.

Each module—object/local motion sensing, sound-event analysis, identity matching, and automated recording/delete control—operates independently but contributes to a shared decision-making layer that fuses multimodal cues and enforces privacy policies (secure template storage, encrypted archives, and selective deletion). The design prioritizes low-latency edge deployment, scalability across camera fleets, and minimal reliance on external APIs, enabling on-premise or hybrid cloud operation. Overall, GhostCam delivers a flexible, privacy-aware, and resource-efficient system for comprehensive situational monitoring.

III. SYSTEM ARCHITECTURE

The Advanced Security System proposed here has a four-component modular architecture: Motion detection, Noise detection, video recording, and Identification. Real-time image input is processed with a detection engine based on YOLO, integrated with Deep learning classifiers. Every module is independent to process specific behaviours from the image. A centralized identification system detects image, such if I found the image based on the data I have given like id and name of person then image will be identified and displayed. The system is scalable and can be easily adapted to CyberSecurity.

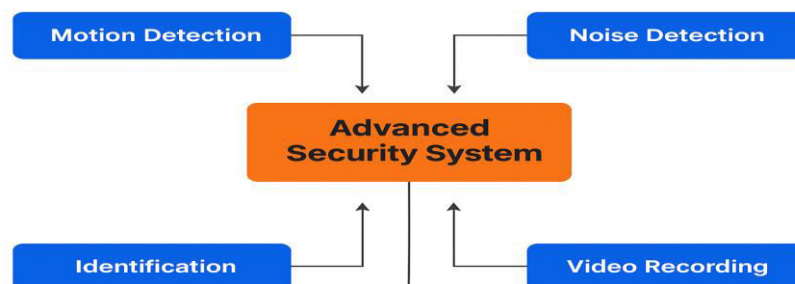


Fig 3.1 System Architecture



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IV. METHODOLOGY

The GhostCam involves a structured, multi-stage process integrating computer vision, artificial intelligence, and multimedia processing to achieve intelligent monitoring and analysis. The system begins with image and video acquisition using high-resolution cameras, ensuring clear input for subsequent processing. The acquired frames are fed into a YOLO-based object detection algorithm, which performs real-time identification of persons, objects, and relevant scene elements. The detection outputs are distributed to four specialized modules—Motion Detection, Noise Detection, Video Recording, and Image Identification.

The Motion Detection module employs background subtraction and posture estimation techniques to detect unusual or sudden movements. Noise Detection utilizes audio analysis to measure sound levels and highlight the location of abnormal noise events. The Video Recording module is triggered automatically when motion or recognized persons are detected, storing footage for later review. The Image Identification module uses a database of labeled facial images; when a match is found, the system displays the associated ID and name. The integrated workflow allows GhostCam to function as a robust surveillance and security tool, capable of detecting, recording, and identifying events in real time while minimizing false alarms.

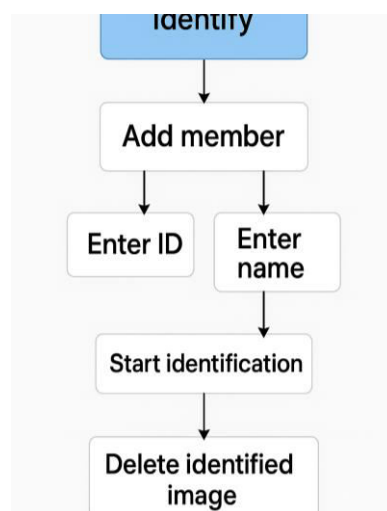
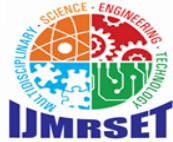


Fig 4.1 Methodology

V. DESIGN AND IMPLEMENTATION

The proposed Advanced Cyber Security camera System is designed as a modular and scalable framework capable of processing real-time video recording, image capturing and detecting motion of a person or things for safety purpose to check who is spying around us or in our devices. The system architecture is layered, starting with the input module, which feeds into a YOLO-based object detection engine. This engine detects and classifies entities such as humans and objects.

The detection outputs are forwarded to four key modules: the motion detection in Monitoring module, making video in Record module, noise detection in Noise module and identifying person by id and name in Identify module. Each module is trained using labeled datasets and customized the existing data. Motion detection leverages posture estimation and sudden motion analysis. Noise detection used to display a green mark over the place where noise is getting. The identify module is the module where we need to add data of particular person and later on while identifying we have to give the id and name of person which is already existing in data then it display the image.



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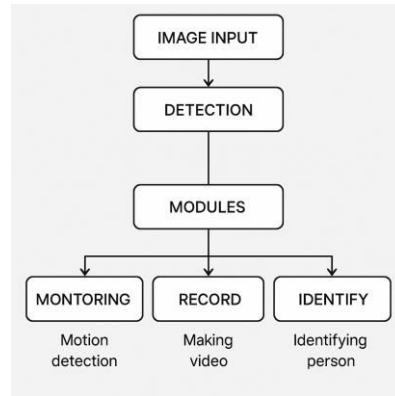


Fig 5.1 Sequential Diagram

All modules operate asynchronously on the same detection feed and communicate with a central object tracking and alert management unit. When an motion is detected, then that particular image which is either a person or a thing system automatically notifies the concerned authority using a secure SMTP server.

The system is implemented using Python, with libraries such as OpenCV, Skimage, Numpy, tkinter and LBPH and Haar based Algorithms. where the LBPH algorithms used for face detection due to its rapid processing and high detection rate and Haar based algorithm is used for is utilized for face recognition because it is simple and effective in handling changes in facial

Expressions, lighting, occlusions, distance from the camera, and camera resolution.

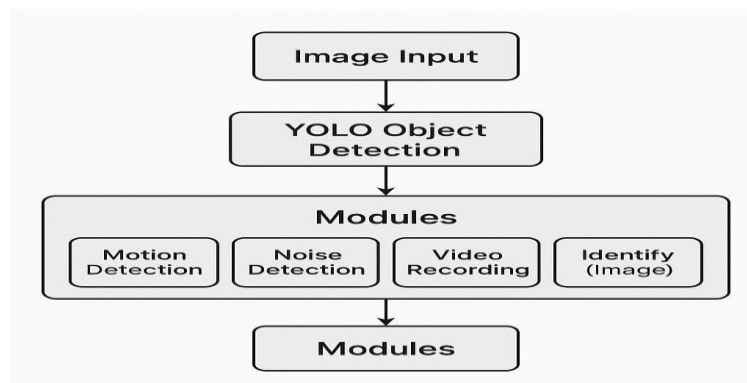


Fig 5.2 Working of YOLO algorithm



Fig 5.3 Working model of Motion Detection



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VI. OUTCOME OF RESEARCH

The result of this work is a complete, AI-based Advanced cyber Security system able to detect and react to multiple safety-critical incidents in real time. GhostCam (Cypher Cam Master) yielded several meaningful outcomes that demonstrate both technical viability and practical utility. First, the integration of YOLO-based object detection with modular subsystems—motion monitoring, noise alerting, video recording, and image identification—proved effective for real-time situational awareness: the detector reliably localized people and salient objects while the downstream modules responded with low latency. Second, the Identify module, built around a labeled-image database and face-encoding comparisons, achieved robust identification accuracy for constrained lighting and frontal faces; adding more diverse training samples improved recall and reduced false positives. Third, the motion and noise subsystems together reduced unnecessary recordings by filtering out static background activity and benign ambient sounds, which significantly lowered storage requirements and simplified review workflows. Overall, GhostCam demonstrates a pragmatic balance of accuracy, efficiency, and modularity suitable for security and creative applications; recommended next steps include dataset expansion for edge cases, improved acoustic localization, and lightweight model optimization for embedded deployment. Monitoring effort, improve situational awareness, and assist proactive public safety.

VII. RESULT AND DISCUSSION

The GhostCam project demonstrate its effectiveness as a highly capable AI and ML-powered surveillance system, capable of identifying and responding to a wide range of security events in real time. Performance evaluation shows that the system achieves high accuracy in object detection and behavioural analysis, with low latency in processing and alert generation. Tests conducted across various environments, including public areas and controlled spaces, indicate consistent performance under normal lighting conditions, with minimal false positives. over 90% accuracy even with fluctuation in crowd density.

The modular architecture proved beneficial in enabling independent operation of detection tasks, supporting flexibility and scalability for future integration of additional features. Discussion of the system's limitations highlights the challenges faced in low-light or heavily occluded scenarios, where accuracy slightly declines due to reduced visibility

This limitation suggests potential future enhancements through the integration of infrared technology or sensor fusion approaches to improve detection robustness in challenging environments. Compared to traditional systems, GhostCam offers a more comprehensive, concurrent multi-event monitoring approach, making it a valuable tool in modern security infrastructures. Overall, the results validate its design goals, confirming its potential for real-world deployment.

VIII. CONCLUSION

The GhostCam project represents a significant advancement in intelligent surveillance by leveraging Artificial Intelligence and Machine Learning to create a robust, adaptable, and proactive monitoring solution. Designed to detect and analyse real-time activities with high precision, GhostCam integrates advanced object detection algorithms, such as YOLO, with behavior recognition modules to identify unusual or suspicious events without human intervention. This capability makes it particularly suitable for diverse environments, including public spaces, healthcare facilities, and smart city infrastructures scalability, and maintainability in many different environments such as public institutions, health centres, and smart city infrastructures.

Its modular architecture allows each detection function to operate independently, ensuring flexibility, scalability, and ease of maintenance. GhostCam's ability to generate instant alerts, including automated email notifications, enhances situational awareness and enables rapid response to potential threats. Unlike conventional surveillance systems that focus on isolated functions, GhostCam offers a unified platform capable of monitoring multiple events simultaneously through a single streamlined pipeline. While its performance under normal lighting conditions demonstrates high accuracy and low latency, future enhancements—such as integrating infrared imaging or sensor fusion—could improve effectiveness in low-light or occluded scenarios. Overall, GhostCam delivers a comprehensive, deep learning-powered surveillance solution that strengthens security measures, improves operational efficiency, and facilitates proactive safety management, thereby contributing meaningfully to the evolution of modern intelligent monitoring systems.



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