

## SMART EYE

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### ABSTRACT:

In today's security-conscious society, the demand for effective surveillance solutions is ever-increasing, especially in areas like hospitals, public spaces, and institutions, where safeguarding assets is crucial. With urban population growth contributing to rising crime rates, video surveillance has become a necessity. While closed-circuit television (CCTV) remains popular, its high cost, power consumption, and storage requirements pose challenges. To address this, a sophisticated intelligent video surveillance system will be proposed, tailored for settings with sporadic human presence. This cutting-edge system employs Machine Learning through a software library to operate without constant camera

surveillance, conserving both electricity and storage space. By capturing video and processing it frame by frame, the system activates recording only upon detecting human presence, minimizing unnecessary footage. If movement is identified, the security system activates, collecting and storing data in a local database. This recorded video aids in identifying intruders, facilitating subsequent tracking and investigation. Such innovation is valuable not only in public spaces but also in private locations with irregular human traffic, such as homes and bank vaults. The system's selective recording feature optimizes resources while enhancing security, making it a promising advancement in surveillance technology.

**Keywords:** Machine learning, surveillance technology.

## I INTRODUCTION

In today's world, everyone wants their valuables to be safe and secure. The development of the urban population has coincided with an increase in crime. Using video surveillance to monitor a specific region, such as hospitals, institutions, public parks, and buildings, has become necessary. Citizens have been greatly impacted by video monitoring. Video surveillance is required for preventing thefts, monitoring day-to-day activities, protecting property, employee safety, public safety, event video surveillance, traffic monitoring, and so on. Daytime and nighttime video cameras are available in color, monochrome (with or without IR lighting), LLL intensified, thermal IR, analogue and digital, simple and full featured. Cameras with built-in VMD can inform security personnel, improving their ability to detect and locate people as well as be notified to activity on the scene. All of these types of video cameras continuously monitor and record footage in accordance with the specifications, which consumes more storage space and energy. The cost of keeping these systems up to date is considerably higher. There is no need for

continual monitoring in some places where individuals are irregular, like as homes and bank vaults. Automatic video surveillance systems leveraging machine learning represent a cuttingedge application of artificial intelligence (AI) in bolstering security and safety across diverse environments. These systems harness advanced algorithms and computer vision techniques to analyze video streams in real-time, enabling proactive threat detection, behavior analysis, and anomaly detection. This comprehensive exploration delves into the intricacies of automatic video surveillance systems in machine learning, elucidating their underlying principles, technological advancements, practical implementations, and ethical considerations. At the heart of automatic video surveillance systems lies machine learning, a subfield of AI that empowers computers to learn from data patterns and make informed decisions without explicit programming. Through the convergence of machine learning and computer vision, surveillance systems can efficiently process vast amounts of video data, extract 2 meaningful insights, and respond to potential security threats in real-time. The foundation of these systems lies in training algorithms on labeled datasets, enabling them to recognize objects, detect

motion, and classify events with high accuracy. One of the key components of automatic video surveillance systems is object detection and tracking, which involves identifying and monitoring specific entities within a video feed. Traditional methods relied on handcrafted features and heuristics to detect objects, but the advent of deep learning revolutionized this process. Convolutional Neural Networks (CNNs) and other deep learning architectures excel at learning hierarchical representations of visual data, enabling robust object detection and tracking even in complex scenes. Behavior analysis is another critical aspect of automatic video surveillance systems, where machine learning algorithms discern normal patterns of behavior from anomalous activities. By analyzing motion trajectories, spatial relationships, and temporal dynamics, these systems can identify suspicious behaviors such as loitering, trespassing, or erratic movement. Anomaly detection algorithms, often based on unsupervised learning techniques, flag deviations from established norms, triggering alerts for further investigation.

## II. LITERATURE SURVEY

### 1. TheSfinX Video Surveillance System

**Authors:** RajuRangaswami,  
ZoranDimitrijević, Kyle Kakligian,

**Edward Chang, Yuan Fang**  
**EduUniversity of California, Santa**  
**Barbara**

It emphasizes the increasing adoption of video surveillance as a means to enhance public security, enabled by the widespread availability of inexpensive cameras and high-speed broadband wired/wireless networks. This accessibility has made it economically and technically feasible to deploy a larger number of cameras for security surveillance. the limitations of current state-of-the- art commercial video surveillance equipment, predominantly comprising analog cameras and tape- based VCRs. These systems face functional limitations such as lack of simultaneous recording and reviewing capabilities, the need to convert analog data to digital format for analysis, and manual retrieval of archived videos, which is time-consuming. Cameras are mounted at the edges of the sensor network to collect signals. When activities are detected, the signals are compressed and transmitted to a server. The server combines data from multiple sensors and creates spatiotemporal descriptors to display the captured activities. The server indexes and stores the video signals with their metadata on the RAID storage. Users of the system (top left in the image) are

alerted to unusual events and can make online inquiries to retrieve and view video clips of interest. Video signals are captured by the video capture module. At the same time, tracking algorithms are used to track objects in the captured video streams, and the video stream is encoded and sent to be stored in Xstream a real-time streaming storage system. To help effectively track blocked objects and gain consensus on object location in ambiguous situations, the multi-tracker module combines tracking information from different cameras that cover a common physical area and feeds back global information to individual camera tracking modules.

**2.A System for Video Surveillance and Monitoring Authors: The Robotics Institute, Carnegie Mellon University, Pittsburgh PA | The Sarnoff Corporation, Princeton, NJ**

The project was conducted under a Defense Advanced Research Projects Agency (DARPA) contract to develop cooperative multi-sensor surveillance technology to increase battlefield awareness and improve situational awareness for military and federal law enforcement agencies. The primary goal was to create an automated video understanding technology that would allow a single human operator to monitor

activities in a complex area using a distributed network of active video sensors. The developed technology aims to revolutionize battlefield intelligence by automatically collecting and disseminating real-time information to enhance commanders' and staff's situational awareness. This technology finds versatile applications, ranging from bolstering perimeter security for soldiers to tracking peace deals or refugee movements using unmanned aerial vehicles (UAVs). It also encompasses strengthening security at embassies or airports and monitoring drug or terrorist hideouts. Central to achieving these objectives is the Video Surveillance and Analysis Module (VSAM), which plays a pivotal role in analyzing people and vehicles from raw video footage, determining their relocation, and integrating them into a dynamic visualization of the scene.

**3.Video Surveillance System Authors: LubosOvensik, Jan Turan Department of Electronics and Multimedia Communications, Faculty of Electrical Engineering and Informatics, Technical University of Kosice**

The focus of much of the current research in this domain revolves around developing algorithms capable of analyzing video and other media from multiple sources to

automatically detect significant events. Example applications cited include intrusion detection, activity monitoring, and pedestrian counting. The fundamental problem addressed by these vision systems lies in extracting moving objects from a video sequence. For systems employing static cameras, the method typically used for segmenting moving regions in image sequences is background subtraction. This method involves comparing each frame to a model of the scene background. The system comprises the function of object detection, tracking, recognition and classification. The problem of object detection has been tackled using statistical models of the background image frame differences techniques or a combination of both. Several techniques have also been used for object tracking in video sequences in order to cope with multiple interacting targets. Object recognition and classification in video surveillance systems are vital for detecting and identifying relevant entities within a monitored scene. These tasks are often accomplished through a combination of statistical pattern recognition and neural network techniques. Statistical pattern recognition methods analyze various features extracted from video frames, including geometric characteristics such as

bounding box aspect ratio, motion patterns, and color histograms. Neural networks, particularly convolutional neural networks (CNNs), are adept at learning complex patterns and structures in visual data, enabling accurate classification of objects.

### III SYSTEM ANALYSIS

#### EXISTING SYSTEM

The existing surveillance system continuously monitors the region and records all the footage.

The existing surveillance system, while providing a means to monitor regions consistently, suffers from several drawbacks that compromise its efficiency and practicality. One of the primary shortcomings lies in its approach of continuous monitoring, which entails recording all footage indiscriminately. This indiscriminate recording not only consumes significant storage space but also poses challenges in managing and analyzing vast amounts of data effectively.

#### Disadvantages

- The Existing system continuously monitors the area.

- It takes more storage for all the recorded footage
- It uses more electricity
- Maintenance costs are also prohibitively expensive.

## PROPOSED SYSTEM

The proposed surveillance system represents a significant advancement in monitoring technology, harnessing the power of the OpenCV open-source computer vision library to enhance surveillance capabilities. By utilizing the Haarcascade classifier, the system effectively identifies faces with remarkable ease, making it particularly suitable for environments where human presence is sporadic but requires attention. This approach offers numerous advantages over traditional surveillance methods, including reduced storage usage, energy consumption, and maintenance costs, ultimately providing a cost-effective and sustainable solution for monitoring targeted areas. At the core of the proposed system is the utilization of the OpenCV open-source computer vision library, which serves as a versatile and powerful tool for executing surveillance setups. This library provides a

comprehensive suite of functionalities for image and video processing, enabling the implementation of advanced surveillance algorithms with ease. By leveraging the capabilities of OpenCV, the proposed system can efficiently analyze surveillance footage in real-time, detecting faces using the Haarcascade classifier with remarkable accuracy and speed.

One of the key advantages of the proposed system is its ability to run efficiently on large databases, enabling seamless integration with existing surveillance infrastructure. Unlike traditional surveillance methods that rely on continuous monitoring, the proposed system optimizes camera usage by selectively activating surveillance measures when human presence is detected. This targeted approach not only conserves energy but also reduces the strain on storage resources, minimizing the amount of footage that needs to be stored and managed. By reducing the amount of time that cameras are used to continuously monitor the region, the proposed system significantly lowers energy consumption, contributing to cost savings and environmental sustainability. Moreover, the selective activation of surveillance measures helps to mitigate storage utilization,

reducing the need for extensive storage infrastructure and lowering associated costs. This not only translates into financial savings but also streamlines data management processes, making it easier to store and retrieve relevant surveillance footage.

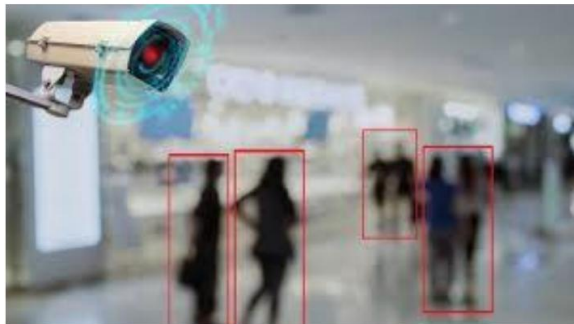


Fig: 1. Reference image to proposed system

Furthermore, the proposed system offers substantial cost savings on maintenance, as it requires fewer resources to operate and maintain compared to traditional surveillance systems. With its reliance on OpenCV and the Haarcascade classifier, the system benefits from a robust and well-supported framework that is easy to implement and maintain. This ease of use and accessibility ensures that the proposed system can be deployed by a wide range of users, from small businesses to large enterprises, without requiring extensive technical expertise.

**Advantages**

- It runs efficiently on large databases.
- To reduce the amount of time that cameras are used to continuously monitors the region.
- In order to save energy.
- To reduce storage utilization.
- To save money on maintenance.

**IV IMPLEMENTATION**

**Architecture:**

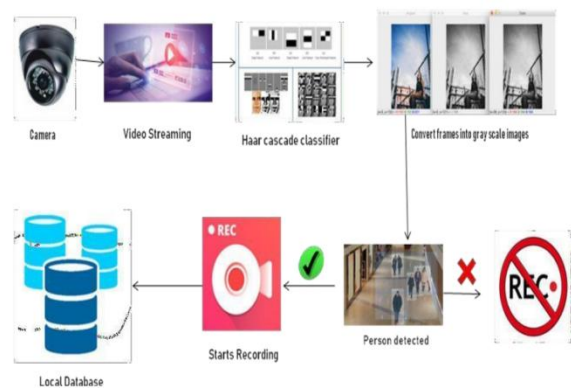


Fig: 2. Architectures of the system model

In the context of a video surveillance system, the process begins with activating the camera and initiating video streaming, allowing for the capture of continuous footage. This stream of video frames serves as the raw input data for subsequent analysis and processing. One pivotal component in this system is the Haar cascade classifier, a machine learning based algorithm specifically designed for object detection



tasks, particularly human faces in this case. By loading and utilizing this classifier, the system gains the ability to identify and locate human faces within the video stream accurately. Once the Haar cascade classifier is integrated into the system, the next step involves converting the individual frames of the video stream into grayscale images. This conversion is essential for simplifying the subsequent face detection process. Grayscale images, being one-dimensional representations of the original frames, strip away color information while preserving crucial details related to luminance and intensity variations. As a result, the identification of human faces becomes more efficient and reliable, as the algorithm can focus solely on the structural aspects of facial features without being influenced by color variations. Upon converting the frames to grayscale, the system proceeds with the face detection process. The Haar cascade classifier is applied to each frame, scanning for patterns resembling human facial features.

When a human face is detected within a frame, the system triggers a series of actions to capture and record the corresponding footage. However, if no human face is detected within the frame, indicating the

absence of any significant human presence in the monitored area, the system automatically deactivates the camera to conserve resources and minimize unnecessary recording. In the event of a successful face detection, signaling the presence of a human within the monitored space, the camera initiates recording, capturing the ongoing activity for further analysis and archival purposes. The recorded footage is then saved into a local database, allowing for convenient storage, retrieval, and review of past events. This database serves as a repository of recorded incidents, facilitating post-event analysis.

In summary, the described process outlines the operational workflow of a video surveillance system equipped with face detection capabilities. By integrating a Haar cascade classifier and leveraging grayscale image conversion, the system can accurately identify human faces within a video stream, enabling proactive monitoring and recording of relevant activities. Through this approach, the system enhances security measures while minimizing resource consumption and false alarms, ultimately contributing to the effective surveillance and management of the monitored area.

## MODULES



## 1. Video Acquisition

High-resolution cameras capture video footage from designated areas under surveillance, which serves as input data for subsequent analysis.

## 2. Pre-processing

Raw video data undergoes pre-processing steps such as noise reduction, stabilization, and frame alignment to enhance the quality and consistency of the input data. Feature

## 3.Extraction

Computer vision algorithms extract relevant features from video frames, including objects, shapes, motion patterns, and spatial-temporal relationships.

## 4. Saves Image

All recordings with human presence are saved.

## 5. Opencv

OpenCV (Open Source Computer Vision Library) is a library of programming functions mainly aimed at real-time computer vision. Originally developed by Intel, it was later supported by Willow Garage then Itsee's (which was later acquired by Intel). The library is cross-platform and

free for use under the open-source BSD license. OpenCV's application areas include:

- 2D and 3D feature toolkits
- Segmentation and recognition Stereopsis stereo vision: depth perception from 2 cameras
- Structure from motion (SFM)

## V RESULT AND DISCUSSION

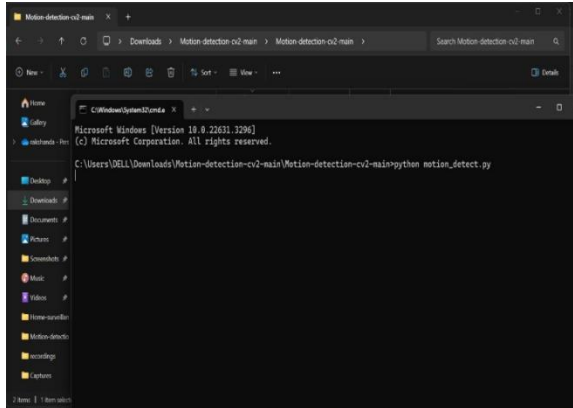
Code

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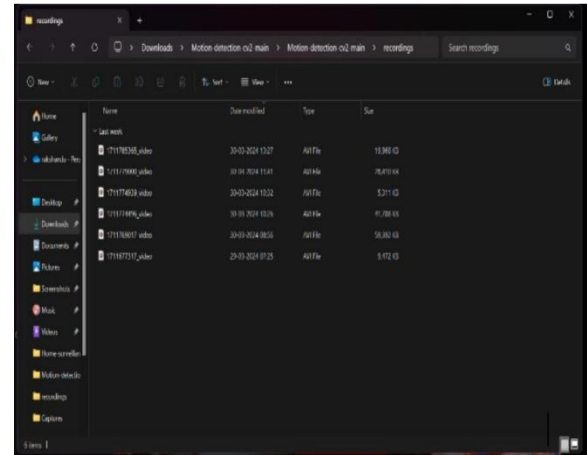
1 # Video Acquisition
2 source = cv2.VideoCapture(0)
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4 # Video Feature Extraction
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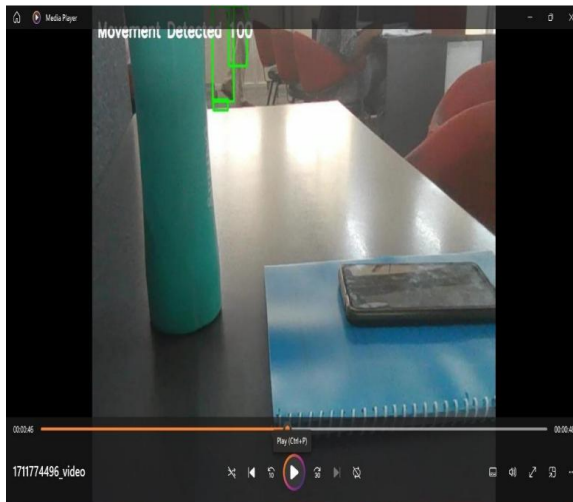
Code Execution



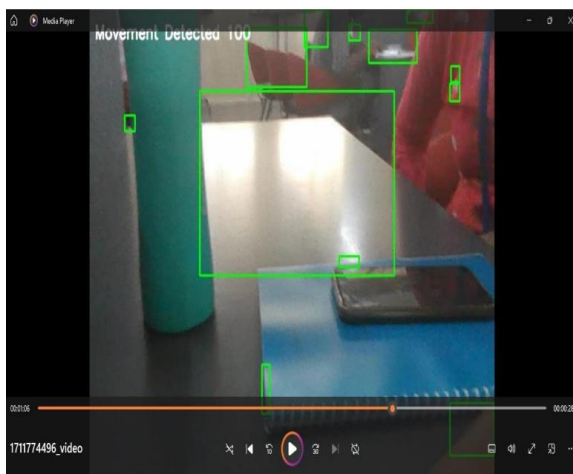
Videos saved to local database



Movement Detection captured 1



Movement Detection Captured 2



## VI CONCLUSION

Automatic video surveillance systems, with their ability to efficiently monitor and analyze footage, represent a significant advancement in security technology that not only enhances safety but also contributes to resource and energy conservation, ultimately reducing costs. By automating the monitoring process, these systems reduce the need for continuous human oversight, thus saving labor costs and allowing personnel to be allocated to other critical tasks. Moreover, automatic video surveillance systems enable more targeted and effective use of resources by focusing attention on areas of concern or potential threats. This targeted approach ensures that resources are allocated where they are most needed, optimizing efficiency and

minimizing waste. The integration of automatic video surveillance systems not only enhances security measures but also delivers tangible benefits in terms of cost savings, resource optimization, and energy conservation. By leveraging these technologies responsibly and implementing best practices, organizations can achieve a balance between security requirements and operational efficiency, ultimately leading to safer, more sustainable environments.

#### **FUTURE ENHANCEMENT**

Eye tracking glasses and webcam eye tracking, there are plenty of other applications of smart eye tracking technology. One of the main benefits of such technology is it can be used to record and analyses visual behavior objectively and accurately.

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