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Research Paper

AI-Enabled Adaptive Online Proctoring with Neuro-Cognitive Monitoring

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Abstract— The AI-Enabled Adaptive Online Proctoring with Neuro-Cognitive Monitoring system is designed to improve the security and reliability of online examinations by using artificial intelligence and computer vision techniques. The proposed framework continuously monitors candidates through a webcam and analyzes their behaviour during the examination without requiring constant human supervision. It detects important events such as face presence, head movement, facial expressions, multiple person detection, and the presence of prohibited objects like mobile phones or books. These behavioural cues are combined to identify suspicious activities and help maintain examination integrity. A Convolutional Neural Network (CNN) is employed for emotion recognition, while computer vision models are used for object detection and head pose analysis. The system provides real-time monitoring and generates alerts whenever unusual behaviour is observed. Experimental evaluation shows that the proposed approach delivers accurate detection with minimal false predictions, making it suitable for practical online assessment environments. Overall, the framework offers an intelligent, efficient, and cost-effective solution for secure remote examinations by reducing manual monitoring effort and enhancing the fairness and credibility of online assessments.

Keywords— Artificial Intelligence (AI), Online Proctoring, Convolutional Neural Network (CNN), Neuro-Cognitive Monitoring, Emotion Recognition, Head Pose Estimation, Object Detection, Computer Vision, Behaviour Analysis, Examination Integrity.

I. INTRODUCTION

The rapid adoption of online education has transformed the way teaching, learning, and examinations are conducted across educational

institutions. As remote assessments become increasingly common, maintaining examination integrity has emerged as a significant challenge because candidates are no longer supervised in a controlled classroom environment [1]. The absence of direct invigilation creates opportunities for unfair practices such as using unauthorized resources, communicating with others, or switching between multiple applications during examinations [3]. These challenges have increased the demand for intelligent online proctoring systems that can automatically monitor candidate activities without continuous human intervention. Artificial intelligence and computer vision technologies provide effective solutions by analysing student behaviour through webcam-based monitoring and identifying suspicious actions in real time [20]. Such automated systems improve the reliability of online assessments while reducing the workload of human invigilators, making them suitable for modern digital learning environments.

Artificial intelligence has significantly improved the ability of online proctoring systems to monitor candidate behaviour with greater accuracy and efficiency. Modern computer vision techniques can analyse facial expressions, detect eye movements, estimate head pose, and recognise multiple individuals during an online examination [2]. Deep learning models such as Convolutional Neural Networks (CNNs) automatically learn visual features from images and video streams, allowing them to identify behavioural patterns without requiring manual feature extraction [4]. Facial emotion recognition further helps determine whether a candidate displays signs of distraction, stress, or suspicious behaviour while taking an examination [6]. By combining behaviour analysis with intelligent object detection, AI-based monitoring systems can generate reliable decisions while minimizing false alarms. These capabilities make automated proctoring more effective than conventional monitoring approaches and support secure online examinations.

Maintaining examination security requires more than simply observing the candidate through a webcam. Intelligent monitoring systems must also identify unauthorized objects, detect the presence of additional individuals, and monitor changes in head movement or gaze direction throughout the examination session [10]. Computer vision libraries and machine learning frameworks have made it possible to develop real-time monitoring systems that process video streams efficiently while maintaining high detection accuracy [8]. Advanced face detection and landmark localization techniques improve the precision of head pose estimation, allowing the system to identify unusual movements that may indicate examination misconduct [21]. The integration of these technologies enables online proctoring systems to continuously analyse multiple behavioural cues simultaneously, resulting in more dependable examination supervision and improved academic integrity.

Neuro-cognitive monitoring enhances traditional online proctoring by examining behavioural indicators that reflect a candidate's level of attention and interaction during an examination. Eye gaze tracking, facial emotions, and head pose estimation provide valuable information for identifying abnormal activities that may suggest cheating or distraction [18]. Behavioural analysis techniques have been successfully applied in e-learning environments to understand user attention and improve interaction between learners and intelligent systems [11]. Biometric authentication methods further strengthen examination security by continuously verifying candidate identity throughout the examination process instead of relying only on login credentials [12]. Combining these intelligent monitoring techniques enables the system to provide continuous and transparent supervision while maintaining fairness for genuine candidates and reducing dependence on manual invigilation.

The proposed AI-Enabled Adaptive Online Proctoring with Neuro-Cognitive Monitoring framework integrates artificial intelligence, deep learning, and computer vision into a unified system for secure online examination monitoring. The system analyses webcam video to detect facial emotions, head movements, multiple person presence, and prohibited objects such as mobile phones and books during examinations [1]. A CNN-based emotion recognition model is combined with object detection and behavioural analysis algorithms to evaluate candidate activities and identify suspicious events in real time [7]. The developed graphical interface allows administrators to load trained models, monitor examinations efficiently, and review detected events through an easy-to-use

application [3]. Overall, the proposed framework offers an intelligent, reliable, and cost-effective solution for enhancing examination integrity while supporting educational institutions in conducting secure and transparent online assessments.

II. RELATED WORK

Neelesh et al., (2021) [1] proposed an artificial intelligence-based online proctoring system to improve the security and reliability of remote examinations. Their work focused on replacing manual invigilation with automated monitoring using computer vision and machine learning techniques. The system continuously observed candidates through a webcam and analysed their activities to identify suspicious behaviour such as unusual head movements, multiple person detection, and the presence of unauthorized objects. The authors highlighted that automated monitoring reduces the workload of human invigilators while maintaining examination integrity. Experimental results demonstrated that the proposed approach effectively detected examination malpractice and improved monitoring efficiency. This study shows that artificial intelligence can provide a practical solution for secure online assessments and serves as a strong foundation for developing intelligent online proctoring systems.

Wang et al., (2020) [2] presented a facial emotion recognition system for analysing student behaviour during online education. The proposed approach used deep learning algorithms to identify different facial expressions from webcam images and evaluate student emotions during virtual learning sessions. The authors explained that facial emotion recognition can help understand student engagement and attention while participating in online activities. Their model was trained using facial image datasets and achieved reliable recognition performance under different conditions. The study also emphasized that emotion analysis can improve intelligent educational applications by providing valuable behavioural information. The research demonstrates that facial expression recognition can be integrated into AI-based online proctoring systems to monitor candidate behaviour more effectively and improve the overall quality of online examinations.

Prathish et al., (2016) [3] developed an intelligent online examination monitoring system that combined webcam video, audio recording, and active window tracking to detect examination malpractice. The proposed framework continuously monitored candidate behaviour and analysed multiple factors such as head movements, facial position, and application switching during the

examination. A rule-based inference system was used to determine whether suspicious activities had occurred. The authors reported that integrating multiple monitoring techniques improved the accuracy of examination supervision compared to traditional methods. Their experimental results demonstrated that the proposed system effectively reduced the need for constant human monitoring while maintaining examination security. This work highlights the importance of combining artificial intelligence with behavioural analysis for secure and efficient online examination environments.

Clarke et al., (2013) [5] introduced e-Invigilator, a biometric-based supervision system designed for secure online examinations. The proposed framework continuously verified the identity of candidates throughout the examination using biometric authentication instead of relying only on login verification. The authors explained that continuous identity verification helps prevent impersonation and strengthens examination security. A prototype implementation demonstrated that the system could monitor candidates without interrupting the examination process. The study concluded that biometric authentication provides a reliable solution for remote assessments while reducing dependence on manual invigilators. Their research established an important foundation for intelligent online proctoring systems by combining continuous authentication with automated monitoring to improve the fairness and credibility of online examinations.

Cheung and Peng, (2015) [18] proposed a webcam-based eye gaze tracking method for analysing user attention in desktop environments. The system estimated eye movement using computer vision techniques without requiring expensive hardware, making it suitable for practical applications. The authors demonstrated that eye gaze information could accurately determine where users focus their attention while interacting with a computer. Experimental results showed that the proposed approach achieved reliable gaze estimation under normal operating conditions. The study suggested that combining eye gaze tracking with other behavioural analysis techniques can improve user monitoring and activity recognition. This research provides valuable support for AI-based online proctoring systems by enabling continuous observation of candidate attention and helping identify suspicious behaviour during online examinations.

III. DATASET DETAILS

The dataset used in this project consists of facial images collected for emotion recognition along with

images containing people and common objects used during online examinations. It includes different facial expression categories such as angry, disgust, fear, happy, neutral, sad, and surprised, which are used to train the emotion recognition model. The dataset also contains images of objects such as mobile phones, books, and persons to support object detection during online proctoring. These images are organized into separate classes, allowing the deep learning model to learn meaningful visual patterns for accurate classification. Before training, the dataset is examined to remove duplicate or low-quality images and ensure proper class distribution. This well-structured dataset provides the foundation for developing an intelligent proctoring system capable of recognizing candidate emotions, detecting unauthorized objects, and monitoring examination activities accurately.

Before training the CNN model, several preprocessing steps are performed to improve dataset quality and model performance. All images are resized to a fixed resolution so that they can be processed consistently by the neural network. Pixel values are normalized to reduce variations in lighting conditions and improve learning stability. The dataset is then labelled according to its corresponding emotion or object category and divided into training and testing sets. Shuffling is applied before splitting the data to ensure that different classes are evenly distributed during model training. This preparation enables the CNN model to learn important visual features efficiently and perform reliable emotion recognition and object detection during real-time online examination monitoring.

IV. PROPOSED METHODOLOGY

The proposed AI-Enabled Adaptive Online Proctoring with Neuro-Cognitive Monitoring system follows a systematic approach to provide secure and intelligent online examination monitoring. Initially, facial image and object detection datasets are collected and prepared for model development. The preprocessing stage includes resizing images, removing unwanted noise, normalizing pixel values, and organizing images into different categories such as facial emotions, persons, mobile phones, and books. The processed dataset is then shuffled and divided into training and testing sets to ensure balanced learning and reliable performance evaluation. These steps enable the CNN model to learn meaningful visual features and improve its ability to recognize different examination-related activities accurately.

After preprocessing, a Convolutional Neural Network (CNN) is trained to recognize facial emotions, while computer vision techniques are

used to detect head movements, multiple persons, and unauthorized objects during online examinations. The trained model is evaluated using performance metrics such as accuracy, precision, recall, F1-score, and the confusion matrix to measure its effectiveness. Once the model achieves satisfactory performance, it is integrated with a webcam-based monitoring system for real-time examination supervision. The system continuously analyses candidate behaviour and generates alerts whenever suspicious activities are detected, providing an intelligent and reliable solution for maintaining examination integrity.

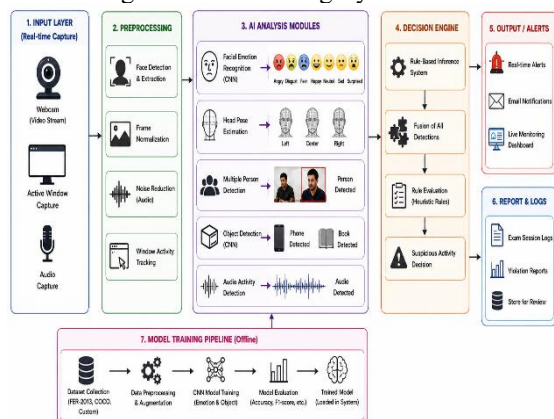


Figure [1]: AI-Enabled Online Proctoring System Architecture

Figure [1] illustrates the workflow of the proposed AI-enabled online proctoring system. Webcam video and examination data are captured and preprocessed before being analysed using CNN-based emotion recognition, head pose estimation, object detection, and behaviour monitoring techniques. The detected activities are evaluated by the decision engine to identify suspicious behaviour during online examinations. Finally, the system generates real-time alerts and monitoring reports to maintain examination integrity and ensure secure online assessments.

V. RESULT AND DISCUSSION

The experimental evaluation demonstrates that the proposed AI-Enabled Adaptive Online Proctoring with Neuro-Cognitive Monitoring system performs effectively in monitoring candidates during online examinations. Before training, the facial image dataset is preprocessed through image resizing, normalization, and organization to improve the quality of the input data. The processed dataset is then used to train the Convolutional Neural Network (CNN) for facial emotion recognition. Along with emotion detection, the system also identifies head movements, multiple person presence, and unauthorized objects using computer vision

techniques. The trained model is assessed using performance measures such as accuracy, precision, recall, F1-score, and the confusion matrix. The evaluation results indicate that the model correctly classifies most facial expressions while producing only a small number of incorrect predictions. After successful training, the model is integrated into a webcam-based proctoring application that continuously analyses candidate behaviour during examinations. The generated alerts and monitoring results demonstrate that the proposed system provides reliable and efficient supervision, helping educational institutions conduct secure and transparent online examinations.

CNN Accuracy : 96.98711250435387
CNN Precision : 96.07940891654252
CNN Recall : 97.03414005307192
CNN FScore : 96.51618633810959

Figure [2]: CNN Model Training Results

Figure [2] presents the CNN model after the training process is completed. The interface displays the evaluation metrics, including accuracy, precision, recall, and F1-score, which are used to assess the performance of the trained model. These results indicate that the CNN model has learned facial features effectively and is capable of recognizing candidate emotions accurately. The trained model is then loaded into the online proctoring system for real-time examination monitoring.

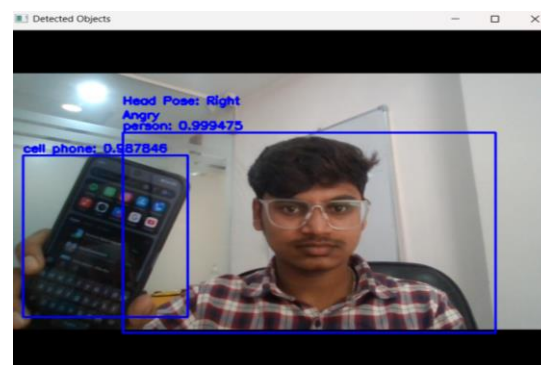


Figure [3]: Real-Time Online Proctoring and Object Detection

Figure [3] shows the real-time online proctoring process during an examination. The system continuously monitors the candidate through a webcam and detects facial emotions, head pose, and unauthorized objects such as mobile phones. It also displays confidence scores for each detected activity, enabling automatic identification of suspicious behaviour and helping maintain the integrity of online examinations.

DISCUSSION

The results of this project demonstrate that integrating artificial intelligence with computer vision provides a practical solution for intelligent online examination monitoring. Proper preprocessing of facial images enables the CNN model to learn meaningful visual features, improving the accuracy of emotion recognition and behavioural analysis. The system effectively identifies different facial expressions while simultaneously detecting suspicious activities such as head movement, multiple person presence, and prohibited objects. The confusion matrix and other evaluation metrics confirm that the trained model performs consistently across different categories with minimal prediction errors. Real-time webcam monitoring allows the application to analyse candidate behaviour continuously without requiring constant human supervision. This automated approach reduces the workload of invigilators while improving examination security and fairness. The user-friendly interface further enhances the usability of the system by providing instant monitoring results and alerts whenever suspicious activities are detected. Overall, the proposed framework offers an accurate, dependable, and cost-effective solution for maintaining academic integrity and ensuring the credibility of remote online examinations.

VI. CONCLUSION

The proposed AI-Enabled Adaptive Online Proctoring with Neuro-Cognitive Monitoring system provides an intelligent and reliable solution for improving the security of online examinations. By integrating artificial intelligence, computer vision, and deep learning techniques, the system continuously monitors candidates throughout the examination process without requiring constant human supervision. It effectively recognizes facial emotions, estimates head movements, and detects unauthorized objects such as mobile phones, books, and the presence of multiple persons. These behavioural observations help identify suspicious activities that may indicate examination malpractice. The developed framework also provides real-time alerts, enabling administrators to respond quickly whenever unusual behaviour is detected. Experimental results demonstrate that the trained CNN model performs consistently in recognizing facial expressions and supporting accurate behavioural analysis. Although the system significantly improves automated examination monitoring, certain complex situations may still require manual review to ensure fair decision-making. Overall, the proposed framework offers a practical, efficient, and cost-effective approach for maintaining examination integrity. It reduces the burden on human invigilators while promoting transparency, fairness, and credibility in remote online assessment environments.

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