



Automated Digital Presentation Control using Hand Gesture Technique

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ABSTRACT

In the modern digital age, utilizing a slideshow for presentations is an efficient and appealing way for speakers to convey information and persuade their audience. Traditional methods for controlling slides involve devices such as mice, keyboards, or laser pointers, which require prior knowledge and experience to operate effectively. Recently, gesture recognition has gained prominence and is employed in various applications like media players, robotics, and gaming. Existing hand gesture recognition systems often rely on gloves or markers, which increase the system's cost. This proposal introduces an AI-based hand gesture detection system. Users can navigate through presentation slides, moving forward or backward, using simple hand gestures. This approach simplifies interaction, making it convenient and eliminating the need for additional devices. The proposed method aims to enhance presentations by enabling natural and efficient communication with the computer. It is particularly advantageous over laser pointers, as hand gestures are more noticeable and can better capture the audience's attention.

Keywords : OpenCV, Hand Gesture Recognition, Human Computer Interaction, Machine learning, Artificial Intelligence, Presentation Slides, Media Control, Virtual Pointer.

Introduction

Over recent decades, hand gesture recognition has emerged as a revolutionary technique in human-computer interaction (HCI), marking a significant departure from traditional input methods like keyboards and mice. This innovative approach is characterized by its automatic, intuitive, and user-friendly nature, creating a seamless communication channel between users and computing systems.

In the realm of presentation software, gesture recognition opens up exciting possibilities, allowing users to navigate slides, execute commands, and engage with content effortlessly through natural hand movements. This advancement enhances the user experience by providing a hands-free and responsive interaction method, aligning with the broader trend of using human gestures as a powerful input across various software applications. Integrating gesture recognition into presentation software exemplifies the technological evolution towards more natural and user-friendly interaction styles, making computing experiences more enjoyable and efficient.

Traditionally, presenters navigate PowerPoint slides using devices such as mice, keyboards, or remote controls. While these methods have been standard, they come with inherent inconveniences and discomforts. Using a mouse can tether presenters to a specific location, limiting their mobility during presentations. Keyboards can be cumbersome and may disrupt the flow of the presentation by diverting attention from the audience. Remote controls, although offering mobility, can be prone to technical glitches or misplacement.

To address these issues, there is a growing interest in simpler and less disruptive alternatives. One promising solution is gesture recognition technology. This innovative approach provides presenters with greater freedom of movement, smoother interaction with presentation software, and a better connection with the audience. The transition from cumbersome devices to intuitive gestures is paving the way for a more fluid and engaging presentation experience.

Literature Survey

The presented papers encompass a diverse range of hand gesture recognition applications, each with unique methodologies, techniques, and focuses. In [1] A comprehensive survey on hand gesture recognition is conducted, offering an extensive overview of various recognition techniques for both static and dynamic gestures. While it provides a broad understanding of the field and highlights potential applications, it lacks specific implementation details.

The next [2] focus on practical applications of hand gestures, particularly in the context of PowerPoint presentations. The introduces a novel approach for navigation using distance transform and skin color segmentation, offering a hands-free interaction during presentations. However it may be limited to specific hand positions or conditions.

The [3] concentrates on finger-pointing gestures for slide presentations, potentially enhancing engagement and interactivity, but it may lack diversity in the types of gestures considered.

The [4] proposes real-time hand gesture recognition for human-computer interaction, addressing the need for efficiency but lacks specific algorithmic details. Lastly, the [5] integrates real-time object detection with TensorFlow and OpenCV, showcasing the potential to enhance user interaction in real-time scenarios, albeit with performance variations influenced by the complexity of detected objects, contributing to the evolving field of gesture recognition. Overall, the papers collectively contribute to the evolving field of gesture recognition, with strengths in surveying, practical applications, and real-time interaction. Yet, certain limitations, such as the lack of detailed implementations or specificity in gesture types, should be considered for future research.

SL No.	Title	Authors	Accuracy
1	A survey on hand gesture recognition	Lingchen Chen, Feng Wang, Hui Deng, Kaifan Ji	92.6%
2	Navigation of PowerPoint Using Hand Gestures	Dnyanada R Jadhav, L. M. R. J Lobo	89.7%
3	Finger-Pointing Gesture Analysis for Slide Presentation	Maisevli Harika, Ary Setijadi P, Hilwadi Hindersah, Bong-Kee Sin	95.4%
4	Real Time Hand Gesture Recognition for Human Computer Interaction	Rishabh Agrawal, Nikita Gupta	90%
5	Detection of Real Time Objects Using TensorFlow and OpenCV	Ajay Talele, Aseem Patil, Bhushan Barse	97%

Fig 1 - Literature Survey

Methodology and Analysis:

The system was developed using the Python programming language, integrating key libraries such as OpenCV, MediaPipe Hands, and OpenCV was essential for capturing frames from the camera system and included additional functionalities to optimize MediaPipe's directional capture capabilities. MediaPipe Hands captured hand movements from the frames provided by OpenCV, identifying user-generated gestures to control slide transitions. The hand images were processed as a mesh of 21 points to facilitate gesture recognition. Once a gesture was recognized, PyWin32 executed the corresponding system tasks.

Unlike existing systems, this gesture recognition system takes a different approach. Instead of relying on machine learning algorithms, it employs mathematical logic derived from the FingerUp function within the Hand Tracking Module developed using OpenCV and MediaPipe. This method eliminates the need for extensive data collection and training, enabling gesture recognition without depending on machine learning techniques.

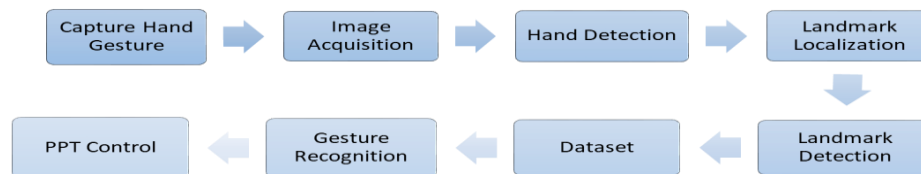


Fig. 2 – Methodology Flow

Computer Vision

OpenCV, an open-source computer vision library, was pivotal in the project, offering predefined algorithms for various computer vision tasks. This included capturing frames from the camera and supporting essential functions such as object detection, face detection, and biometric recognition.

B. MediaPipe

MediaPipe, Google's open-source framework for real-time multimedia processing, provided pre-built machine learning models and a flexible architecture for tasks like image processing, object detection, and facial recognition. Its modular design and real-time performance capabilities simplified the development of computer vision applications for mobile devices and embedded systems.

C. MediaPipe Hands

MediaPipe's Hand Tracking solution uses machine learning for real-time hand detection and tracking in video or image inputs. It accurately detects and tracks 21 points on the hand, enabling precise hand movement representation. This functionality supports applications like hand gesture recognition, virtual try-ons, and augmented reality interactions.

- Palm Detection Model
- MediaPipe introduced a specialized palm detection model that estimates bounding boxes for rigid objects such as palms and fists. This approach simplifies hand detection by focusing on palm detection and employs advanced feature extraction and loss optimization methods. The model achieves impressive results, with an average precision of 95.7% in palm detection.
- Palm Key Points Model
- The Palm Key Points Model predicts the coordinates of 21 key positions on the palm using regression methods for precise localization. Trained on a diverse dataset of real hand images and synthetic palm models, the model ensures accurate localization even in challenging situations like self-occlusions or partially visible hands.

D. PyWin32

PyWin32, a Python library, provides access to the PowerPoint COM object model, enabling interaction with presentation elements such as slides and shapes. The win32com.client module within PyWin32 allows for the creation of COM objects and facilitates actions such as opening presentations and manipulating slide content

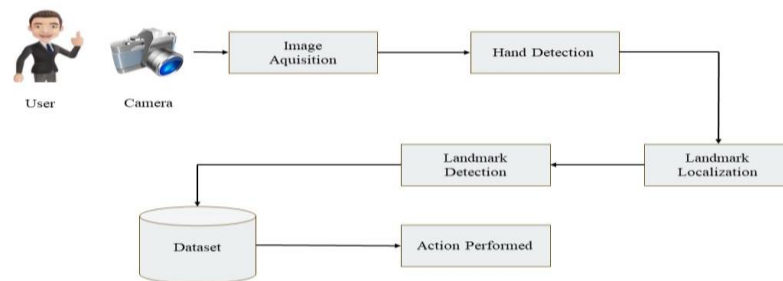


Fig. 3 – Architecture Diagram

The architecture diagram illustrates the process flow for a hand gesture recognition system designed for controlling presentation slides. The process begins with the user making gestures in front of a camera, which captures the images (Image Acquisition). These images are then processed to detect the presence of a hand (Hand Detection). Once the hand is detected, the system moves to the Landmark Localization phase, where specific points on the hand are identified. This information is used for Landmark Detection to accurately pinpoint and analyze key features of the hand's position and movement. The identified landmarks are then cross-referenced with a pre-existing dataset to recognize the gesture. Finally, based on the recognized gesture, the corresponding action is performed, such as navigating presentation slides. This system eliminates the need for physical devices, offering a more natural and intuitive interaction method.

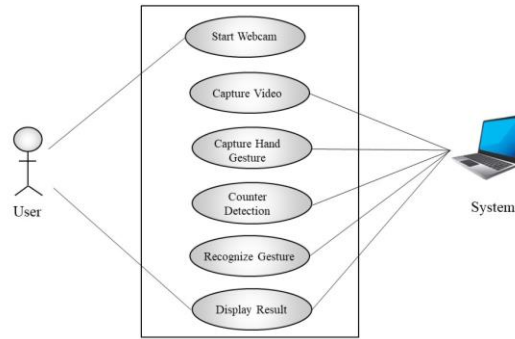


Fig. 4 – UseCase Diagram

The use case diagram depicts the interaction between the user and the system for hand gesture-based presentation control. The process begins with the user initiating the system by starting the webcam (Start Webcam). The webcam captures a continuous video feed (Capture Video), which is then processed to detect the user's hand gestures (Capture Hand Gesture). The system performs counter detection (Counter Detection) to identify the specific gesture being made. Once the gesture is recognized (Recognize Gesture), the corresponding action, such as navigating presentation slides, is executed. Finally, the result of the gesture recognition is displayed to the user (Display Result). This diagram highlights the steps involved in capturing and interpreting hand gestures to control presentation software, emphasizing a seamless interaction between the user and the system.

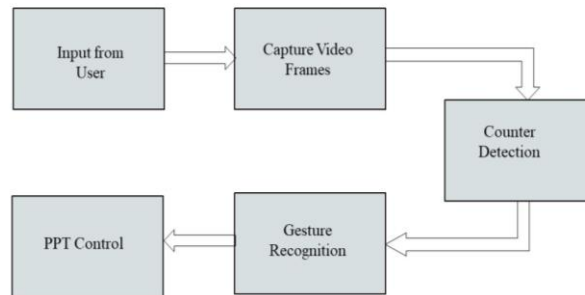


Fig. 5 – Block Diagram

The block diagram illustrates the workflow of a hand gesture recognition system designed for controlling PowerPoint presentations. It starts with the user performing hand gestures, which are captured by the camera in the Input from User block. These gestures are then processed in the Capture Video Frames block, where the system records the video frames containing the hand movements. The captured frames are analyzed in the Counter Detection block, where the system identifies and counts specific gestures. This information is sent to the Gesture Recognition block, which interprets the gestures based on predefined patterns or models. Finally, the recognized gestures are used to control the presentation in the PPT Control block, enabling actions such as moving slides forward or backward. This system offers an intuitive and hands-free way to interact with presentation software, enhancing the user experience and engagement.

Result

4.1 Gestures

1. Forward Gesture: By doing Forward gesture we can move to next slide.

2. Backward Gesture: By doing Backward Gesture we can move to previous slide.

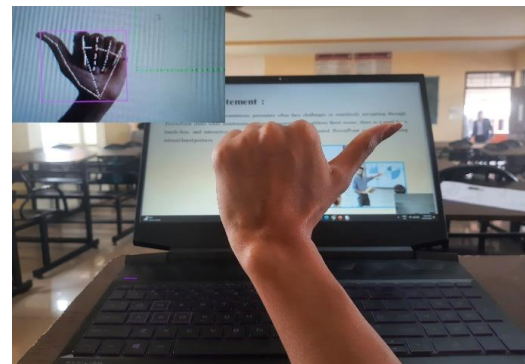
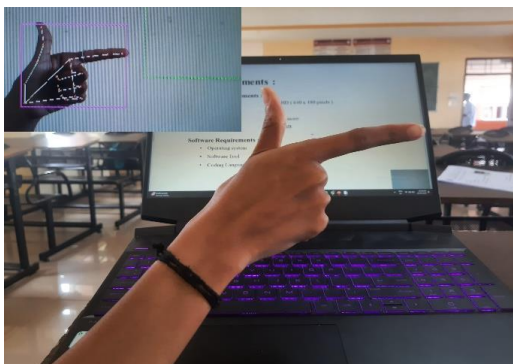


Fig 6 – Forward Gesture

Marker Gesture: By Doing this gesture we can initialize just by Marker on the screen using index finger.

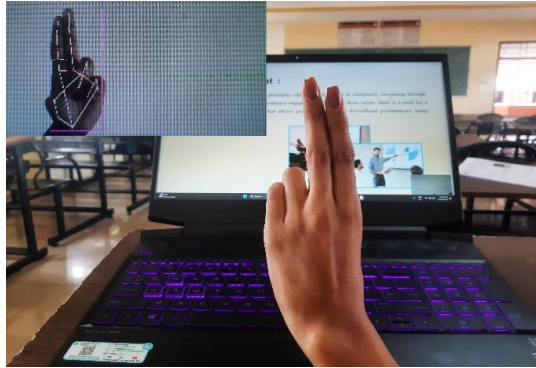


Fig 8 – Marker Gesture

5) Eraser Gesture: By doing this Gesture we can erase the the whole Marks done by the marker screen.

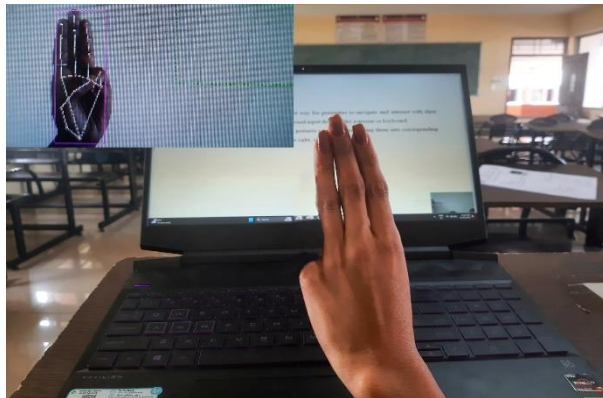


Fig10 – Eraser Ge

7) Zoom In Gesture : By doing this we can Zoom into the ppt in the ppt.

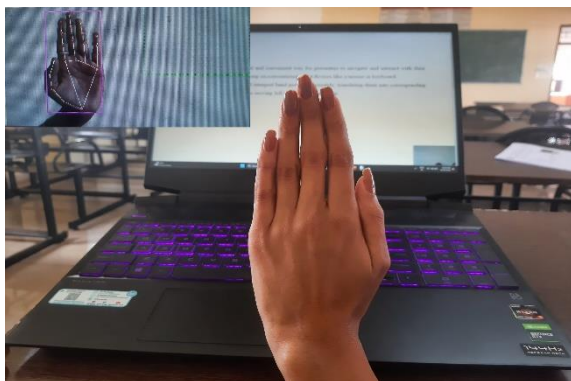


Fig 12 - Zoom In Gesture

Fig 7 – Backward Gesture

4) Pointer Gesture : By doing this gesture we can use marker



Fig 9 – Pointer Gesture

6) Clear Gesture: By doing this Gesture we can Clear

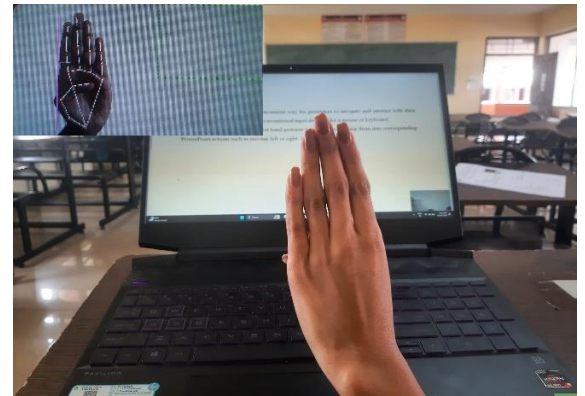


Fig 11 – Clear All Gesture

8) Zoom Out Gesture: By doing this we can Zoom out

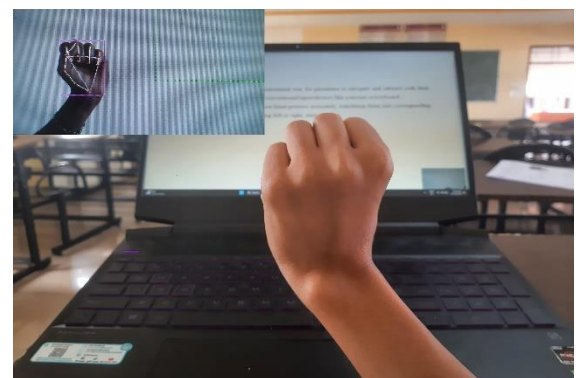


Fig 13 – Zoom Out Gesture

Discussion

Hand gestures provide a more natural means of interaction as they are an integral part of body language, unlike other devices. Utilizing hand gestures eliminates the need for additional devices and simplifies interaction. In this proposed system, an AI-based methodology for hand gesture detection is introduced. Implementing hand gestures will facilitate easier presentation delivery for speakers. The goal of this system is to develop software that enables presenters to control presentation slides using various hand gestures. With this software, there will be no need for devices such as a keyboard, mouse, or remote control to change slides.

Conclusion

In summary, the papers presented collectively contribute to the expanding domain of hand gesture recognition through diverse methodologies and applications. The extensive survey offers a comprehensive understanding of recognition techniques, while practical implementations in PowerPoint navigation and slide presentations introduce innovative approaches to improve user interaction. The focus on real-time gesture recognition and object detection reflects the increasing demand for responsive human-computer interfaces. However, limitations such as insufficient detailed implementations in certain cases and potential specificity in gesture types highlight the necessity for further research to refine and advance the efficacy of these approaches. Overall, these papers represent valuable milestones in the ongoing quest for more refined and inclusive hand gesture recognition systems.

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