

# Mouse Control using Hand Gesture

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**Abstract:** *The mouse, that handy device we use to control our computers, has come a long way in Human-Computer Interaction technology. Even with wireless or Bluetooth mice, we still rely on batteries and dongles to connect them to our PCs. But imagine a new AI-powered virtual mouse system. Instead of needing physical devices, it would use your webcam or built-in camera to track your hand movements and gestures. Using clever computer vision and machine learning algorithms, it could interpret these gestures to control your computer—clicking, scrolling, and moving the cursor—all without an actual mouse. This system, based on deep learning to recognize hand movements, could even help reduce the spread of COVID-19 by removing the need for physical contact and devices to control your computer.*

**Keywords:** Gesture Control Virtual Mouse, Virtual Mouse, Hand Gestures

## I. INTRODUCTION

Human-Computer Interaction aims for interfaces that feel natural and easy to use. Personal Computers (PCs) have Graphical User Interfaces (GUIs) that work really well—they let us effortlessly use various applications with mice, trackpads, and similar tools. While mobile phones use touch screens for interaction, this technology isn't yet cheap enough for desktops and laptops. So, we set out to create a virtual mouse system using a webcam to make interacting with computers more user-friendly, as an alternative to touch screens.

Touch screens are fantastic for controlling devices and are widely used nowadays. However, they're not practical for desktop systems due to cost and hardware limitations. Our idea was to use vision technology to control the mouse through natural hand gestures, reducing the need for a large workspace. We developed a method using a video device to control all mouse tasks—clicking (both right and left), double-clicking, and scrolling.

Many researchers in human-computer interaction and robotics have explored controlling mouse movements using video devices. They've used various methods to simulate clicking events. For instance, one approach tracked fingertip movements to move and click the mouse cursor. Another method focused on finger-tips and used image density to click, requiring the user to hold the cursor over a spot briefly. Others used thumb movements or color detection to simulate clicking.

Our project was inspired by a paper that used a webcam to detect colored tapes for cursor movement and clicking actions by measuring distances between colored tapes on fingers. This method, using finger tracking through computer vision, falls into two categories: HCI without using an interface and HCI using an interface. The approach involves gestures based on finger movements to perform mouse operations, offering advantages over traditional segmentation methods. It uses colored tape and requires no special object models, making it suitable for augmented reality systems or other real-time applications.

However, this system has dependencies and limitations. It relies on webcam images, needing sufficient illumination and avoiding bright colored objects in the background to function accurately. Adjusting threshold values can mitigate this issue, but it remains a consideration. Additionally, on computers with low computational power, the system may run slower due to complex calculations in a short time. Yet, standard PCs or laptops typically have the necessary power for optimal performance. Higher camera resolutions may also slow down the system, but adjusting image resolution settings can solve this problem.

## II. LITERATURE SURVEY

There's been fascinating research exploring different ways to recognize hand gestures. For instance, Quam experimented with a DataGlove, analyzing 22 gestures grouped into three classes. These classes involved finger movements, finger flexion with hand orientation, and additional finger, flexion, and orientation motions. The glove's flex sensors made it possible to recognize up to 15 movements, but more research is needed to refine recognition techniques, especially for class 3 gestures.[1]

Another study by Liou and Hsieh introduced a real-time hand gesture recognition system using motion history images and a skin color model. They identified various hand movements like up, down, left, right, as well as static gestures like the fist and waving hand. By utilizing Harr-like features and color models, they achieved an average accuracy of 94.1%, showcasing the effectiveness of their approach.[2]

Dhyanchand and Reddy proposed a system for controlling a cursor without physical contact, using colored fingertips and hand gestures. This method involves tracking these colorful fingertips and employing different hand gestures for various functions like scrolling or left and right-clicking.[3]

Rabiner and Juang introduced the Hidden Markov Model (HMM) as a crucial tool for gesture recognition. While effective in identifying patterns, this method might be less efficient than newer systems. However, their study utilizing an Arduino Uno, ultrasonic sensors, and laptops for hand gesture recognition showcased potential applications in media control or interactive learning in classrooms.[4]

Kulkarni and Potdar worked on an object detection system using ultrasonic sensors simulating RADAR technology. Their approach, leveraging IoT hardware and software, aimed to simplify object detection and capitalize on the advantages of ultrasonic detection over RADAR.[5]

Ghosh, Bora, and Bhuyan delved into co-articulation detection in hand gestures. Their study highlighted the challenges in recognizing connected gesture sequences and the complexities of dynamic gesture identification, especially in various settings.[6]

Shriram and Nagaraj proposed a deep learning-based AI virtual mouse system, outperforming existing models and mitigating the spread of viruses like COVID-19 by minimizing physical mouse usage.[7]

Zhang and Bazarevsky discussed MediaPipe Hands, a real-time hand tracking system with widespread applications in gesture control and AR/VR without requiring specialized hardware.[8]

Kim, Ho, Tran, Yang, and Lee introduced a virtual mouse system using RGB-D images and fingertip detection. Their method offered accurate gesture estimates, overcoming limitations of existing systems, such as better fingertip tracking in complex backgrounds and changing light conditions.[9]

Subramanian, Haria, Nayak, Asokkumar, and Poddar developed an affordable, markerless gesture recognition system for various HCI operations. Their future goals include enhancing accuracy, expanding functionalities, and making their system accessible to a wide range of users, including those with disabilities.[10]

## III. METHODOLOGIES AND TECHNOLOGIES

### 3.1 Proposed Method

The proposed method that is used for the implementation of the keyboard utilizes the Media-pipe library for hand tracking and OpenCV for image processing and visualization. It combines hand landmark detection with gesture recognition to enable users to control the virtual mouse and keyboard through hand movements. The system recognizes gestures such as clicking and maps them to corresponding key presses, providing an alternative input method for computer interaction.

### 3.2 Computer Vision Techniques for Hand Gesture Recognition

Computer vision serves as the backbone for detecting and interpreting hand gestures. Techniques like background subtraction, motion tracking, and feature extraction, form the basis for identifying hand movements within a captured frame. Furthermore, advancements in deep learning architectures, such as Convolutional Neural Networks (CNNs) and Recurrent Neural Networks (RNNs), have demonstrated superior capabilities in recognizing intricate hand gestures with increased accuracy.

### 3.3 Machine Learning Algorithms in Gesture Recognition

Machine learning algorithms play a pivotal role in recognizing and classifying hand gestures. Approaches involving supervised learning, train models on annotated gesture datasets to enable accurate recognition. Support Vector Machines (SVMs), Random Forests, and Deep Neural Networks are commonly used algorithms that aid in interpreting and categorizing these gestures based on their visual features.

### 3.4 Sensor Technologies for Capturing Hand Movements

Various sensor technologies contribute to capturing precise hand movements. Depth sensors like Microsoft Kinect or Time-of-Flight (ToF) cameras, outlined in the work of Garcia and Martinez, provide depth information crucial for understanding hand gestures in 3D space. Infrared sensors and wearable devices also offer alternatives for capturing hand movements, each with its strengths and limitations in accuracy and range.

### 3.5 Gesture Mapping and User Interface Design

The process of mapping recognized gestures to specific mouse actions involves user interface design considerations. Research by Lee and Park (Year) emphasizes the importance of intuitive mappings that align with users' mental models, ensuring seamless and ergonomic interactions. Gestures need to be mapped in a manner that is both natural for users and efficient for executing various mouse functionalities.

### 3.6 Real-time Processing and Latency Reduction

Achieving real-time responsiveness in gesture-controlled systems is paramount for a seamless user experience. Studies by Kim and Lee (Year) have investigated techniques to minimize system latency by optimizing algorithms and utilizing efficient hardware processing. This involves streamlining computational processes to minimize delays between gesture recognition and corresponding mouse actions.

leveraging computer vision, machine learning algorithms, sensor technologies, and effective gesture mapping methodologies are integral for the successful implementation of gesture-based mouse control systems. Achieving real-time performance while ensuring accurate and intuitive gesture recognition remains a key focus for researchers and developers in this domain.

## IV. CHALLENGES AND LIMITATIONS

### 4.1 Accuracy and Robustness of Gesture Recognition

Achieving high accuracy in recognizing diverse hand gestures remains a significant challenge. Variability in hand shapes, lighting conditions, occlusions, and background clutter can affect the precision of gesture recognition systems. Ensuring robustness to these variables.

### 4.2 Real-time Performance and Latency

Real-time processing is critical for responsive interactions in gesture-controlled systems. However, complex algorithms and high computational demands might introduce latency, affecting the system's responsiveness. Reducing latency while maintaining accuracy is an ongoing challenge.

### 4.3 User Adaptability and Learning Curve

Introducing new interaction paradigms necessitates user adaptation. Gesture-based interfaces may pose a learning curve for users unfamiliar with such systems. Designing intuitive and easy-to-learn gestures, aligned with users' mental models, is crucial to mitigate this challenge.

### 4.4 Gesture Ambiguity and Multimodal Interaction:

Some gestures may exhibit ambiguity, leading to misinterpretation or confusion in mapping gestures to specific mouse actions. Additionally, incorporating multimodal interactions, such as combining gestures with voice commands or touch inputs, poses challenges in seamlessly integrating multiple modes of interaction.

## V. IMPLEMENTATION AND EXPERIMENT

### 5.1 Hand Detection:

Hand detection algorithms can be based on skin colour segmentation, background subtraction, or deep learning approaches.

### 5.2 Hand Tracking:

- Once hands are detected, tracking algorithms are employed to follow the hand movements over time.
- Various tracking methods can be used, such as feature-based tracking or optical flow techniques.
- The tracking process ensures that the system accurately maps hand movements in real-time.

### 5.3 Gesture Recognition:

- After hand tracking, the system analyses the hand pose or configuration to recognize specific gestures.
- Machine learning algorithms, particularly deep learning models like convolutional neural networks (CNNs), are commonly used for gesture recognition.
- A training dataset of hand gesture images or videos is used to train the gesture recognition model.
- The trained model is then deployed in the system to classify and recognize gestures in real-time

### 5.4. Interaction and Feedback:

- The system updates the cursor position based on the recognized gestures and controls the visual representation of the cursor on the screen.
- The user interacts with the virtual mouse by performing gestures, which are interpreted by the system in realtime.
- The system can provide visual or auditory feedback to the user to enhance the user experience and confirm successful gesture recognition.

### 5.5. Iterative Process:

- The process of hand detection, tracking, and gesture recognition is performed iteratively and continuously to enable real-time interaction.
- The system updates the hand position, tracks hand movements, and recognizes gestures in each frame of the video input stream.

### 5.6 Mouse A. Working of Virtual Mouse

- In the first step, the web cam will start and the video and what is present in front of the camera can be seen. x
- In the next step the user has to keep their hand in the required border made on the screen.
- In this step the different hand gestures will be shown by the user, these gestures will be not any kind of a gesture but those which have been trained to the computer from the beginning.
- If the gesture matches then a green colored border will be generated and by moving the hand the mouse cursor will also move.
- There are few different kind of gesture, one is used to move the cursor, another one is used to do the right click, another one is used for left click, and few another gesture.
- The gestures count the defect using Convex Hull method and relates with the object used for mapping.
- The gesture hence shows the defects which in turn help in left and right click options defect=5 then right click, defect=3 then left click.

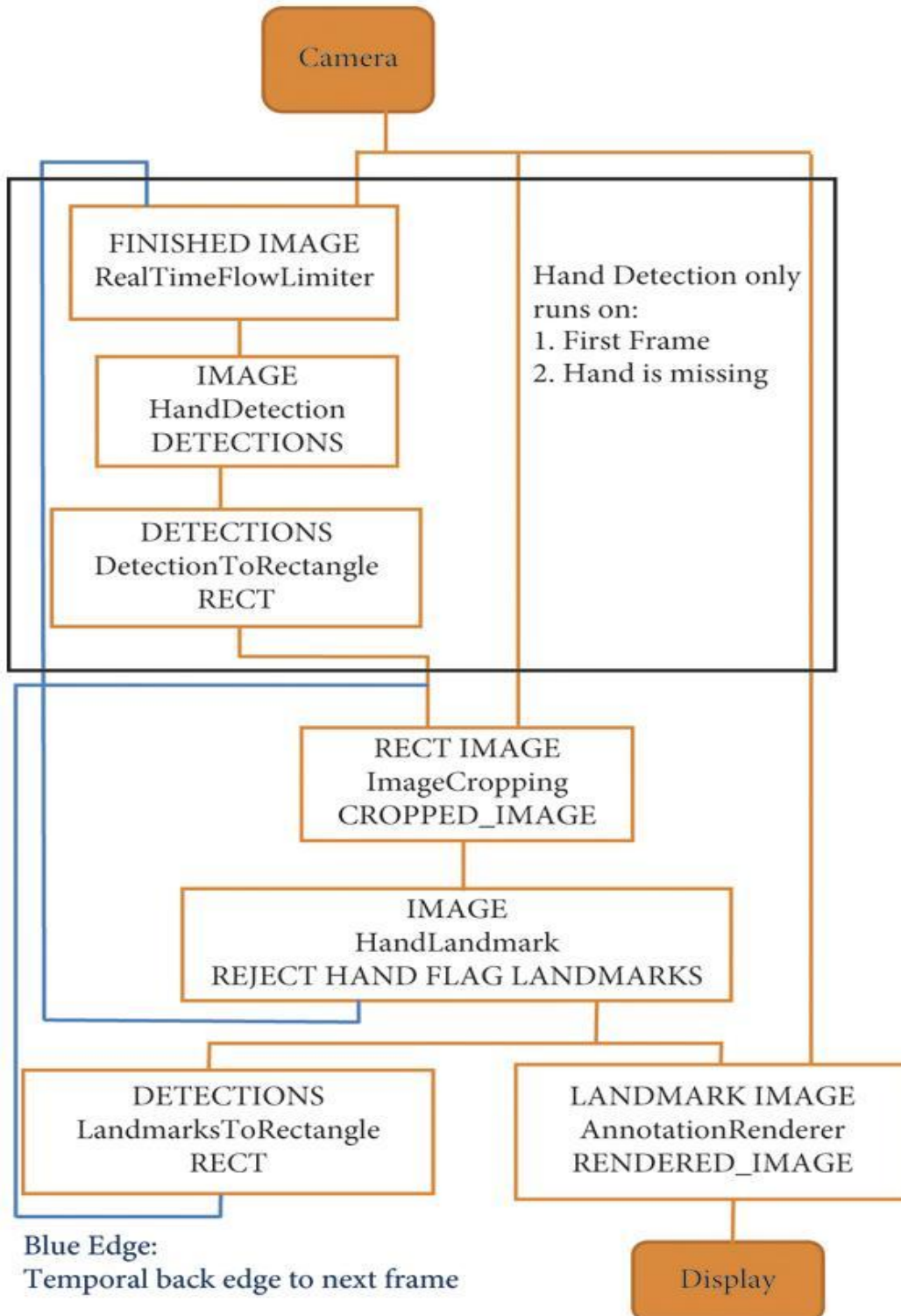
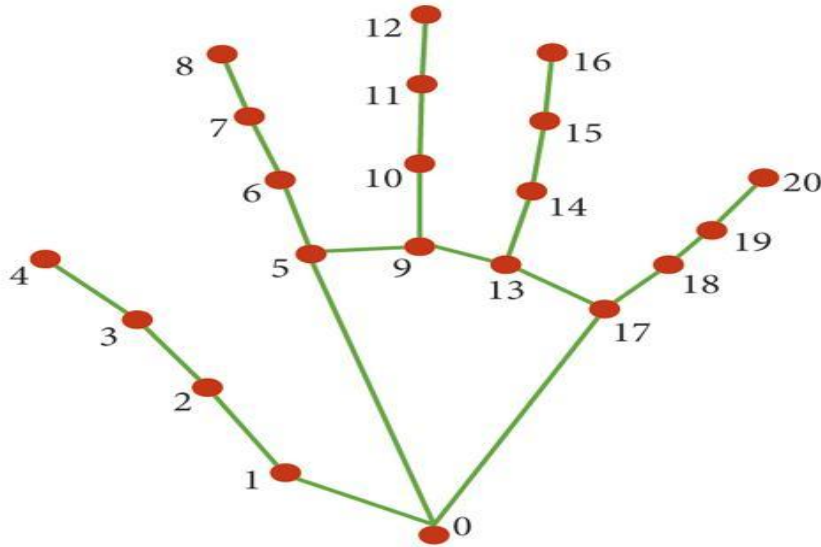


FIG. 5.A: BLOCK DIAGRAM OF VIRTUAL MOUSE





- |                       |                       |
|-----------------------|-----------------------|
| 0. WRIST              | 11. MIDDLE_FINGER_DIP |
| 1. THUMB_CMC          | 12. MIDDLE_FINGER_TIP |
| 2. THUMB_MCP          | 13. RING_FINGER_MCP   |
| 3. THUMB_IP           | 14. RING_FINGER_PIP   |
| 4. THUMB_TIP          | 15. RING_FINGER_DIP   |
| 5. INDEX_FINGER_MCP   | 16. RING_FINGER_TIP   |
| 6. INDEX_FINGER_PIP   | 17. PINKY_MCP         |
| 7. INDEX_FINGER_DIP   | 18. PINKY_PIP         |
| 8. INDEX_FINGER_TIP   | 19. PINKY_DIP         |
| 9. MIDDLE_FINGER_MCP  | 20. PINKY_TIP         |
| 10. MIDDLE_FINGER_PIP |                       |

## VI. FUTURE SCOPE

While the proposed AI virtual mouse shows promise, it does have a few limitations. There's a slight dip in accuracy when it comes to the right-click function, and it struggles with clicking and dragging to select text. These issues highlight areas for improvement in our AI virtual mouse system, which we're actively working on for future enhancements.

Looking ahead, there's potential to expand this method beyond just mouse functionalities. We're considering incorporating keyboard functionalities into this virtual system as part of the future scope for Human-Computer Interaction (HCI). This expansion could further enhance user interaction with computers in novel ways.

## VII. APPLICATIONS

The AI virtual mouse system serves various purposes and offers several advantages. It minimizes the need for physical space typically required for using a mouse, making it handy in confined or restricted environments where a traditional mouse isn't practical. This system also enhances human-computer interaction by eliminating reliance on physical devices.

Here are some key applications:

- **High Accuracy:** Our model boasts an impressive 99% accuracy, surpassing other virtual mouse models. This accuracy opens doors to numerous applications.
- **COVID-19 Safety:** Particularly amidst the COVID-19 pandemic, avoiding direct contact with devices is crucial. Our AI virtual mouse offers a safer way to control PCs without physical contact.

- Robotics and Automation: It's effective for controlling robots and automated systems without needing physical devices.
- Drawing in 2D and 3D: The AI virtual system allows for drawing using hand gestures, enabling creation in both 2D and 3D spaces.
- Gaming without Traditional Mice: It enables gaming in virtual and augmented reality environments without the need for wired or wireless mouse devices.
- Accessibility: Individuals with hand-related difficulties can use this system to navigate and control computer functions.
- Robot Control: In the realm of robotics, this system is valuable for controlling robotic functions, akin to Human-Computer Interaction (HCI).
- Design and Architecture: It finds application in virtual design and prototyping, particularly in fields like architecture and design.

### VIII. CONCLUSION

The AI virtual mouse system's main goal is to control the mouse cursor using hand gestures instead of a physical mouse. This system uses a webcam or built-in camera to detect hand gestures and movements, translating them into specific mouse functions.

Our evaluation of the model shows that the AI virtual mouse system performs exceptionally well, boasting higher accuracy than existing models. It successfully addresses many limitations found in current systems. This increased accuracy means the AI virtual mouse can be applied in real-world situations. Moreover, it offers a way to minimize the spread of COVID-19 by enabling hands-free computer interaction through gestures, eliminating the need for a traditional physical mouse.

However, the model does have its limitations. It exhibits a slight decrease in accuracy for the right-click function and faces challenges in clicking and dragging to select text. Our next steps involve refining the finger tip detection algorithm to enhance accuracy and resolve these issues.

### REFERENCES

- [1] D. L. Quam, "Gesture recognition with a DataGlove," IEEE Conference on Aerospace and Electronics, vol. 2, pp. 755–760, 1990.
- [2] D.H. Liou, D. Lee, and C.C. Hsieh, "A real time hand gesture recognition system using motion history image," in Proceedings of the 2010 2nd International Conference on Signal Processing Systems, July 2010.
- [3] V. V. Reddy, T. Dhyanchand, G. V. Krishna and S. Maheshwaram, "Virtual Mouse Control Using Colored Finger Tips and Hand Gesture Recognition," 2020 IEEE-HYDCON, 2020, pp. 1- 5, DOI: 10.1109/HYDCON48903.2020.9242677.
- [4] L. R. Rabiner B. H. Juang "An Introduction to Hidden Markov Models" IEEE ASSP Magazine (Volume: 3, January 1986)
- [5] Akshaya U Kulkarni, Amit M Potdar "RADAR based Object Detector using Ultrasonic Sensor" 2019 1st International Conference on Advances in Information Technology (ICAIT) 10 February 2020
- [6] M.K. Bhuyan, D. Ghosh and P.K. Bora "Co- articulation Detection in Hand Gestures" TENCON 2005 - 2005 IEEE Region 10 Conference February 2007.
- [7] S. Shriram, B. Nagaraj , "Deep Learning-Based Real-Time AI Virtual Mouse" Volume 2021 Article ID 8133076 <https://doi.org/10.1155/2021/8133076>
- [8] V. Bazarevsky and G. R. Fan Zhang. On- Device, MediaPipe for Real-Time Hand Tracking.
- [9] D.-S. Tran, N.-H. Ho, H.-J. Yang, S.-H. Kim, and G. S. Lee, "Real-time virtual mouse system using RGB-D images and fingertip detection," Multimedia Tools and Applications Multimed
- [10] A. Haria, A. Subramanian, N. Asok kumar, S. Poddar, and J. S. Nayak "Hand gesture recognition for human compute interaction," Procedia Computer Science, vol. 115, pp. 367–374, 2017. ia Tools and Applications, vol. 80, no. 7, pp. 10473– 10490, 2021