
Gesture-Based Air Typing: A Machine Learning Approach for Accessible Text Input

Premala Bhande¹, Sharanbasava², Pavan³, Siddu Patil⁴, Prashanth Panchal⁵

^{1,2,3,4,5}Computer Science & Engineering, Guru Nanak Dev Engineering College, Bidar
Visvesvaraya Technological University Belagavi, Karnataka, India.

ABSTRACT

The "Gesture-Based Air Typing: A Machine Learning Approach for Accessible Text Input" project represents a groundbreaking approach to hands-free typing, designed to revolutionize human-computer interaction by leveraging advanced hand gesture recognition technologies. This innovative system employs MediaPipe for precise hand tracking and EasyOCR for accurate text recognition, enabling users to seamlessly input text through freehand gestures captured in realtime using a simple webcam. The project's emphasis on cost-effectiveness and accessibility ensures that it caters to a broad audience, including individuals with physical impairments and those seeking futuristic, touch-free typing solutions. By eliminating the reliance on physical keyboards, this system opens new possibilities for inclusive design and interaction in digital environments.

Keywords: Air Gesture, Typing, Machine Learning, EasyOCR, MediaPipe

I. INTRODUCTION

Traditional text entry methods, such as physical keyboards, have long been a cornerstone of digital interaction. However, these methods often present significant accessibility challenges for users with physical impairments or limitations that restrict their ability to engage with standard input devices. In addition, reliance on physical keyboards may not always align with the evolving needs of modern users seeking more intuitive and flexible interaction methods.

Gesture-based¹ interfaces have emerged as a promising alternative, leveraging natural hand movements to facilitate interaction with digital systems. Such interfaces eliminate the need for physical contact, providing a touch-free means of control and input. The "Gesture-Based Air Typing: A Machine Learning Approach for Accessible Text Input" project builds upon this premise by integrating advanced technologies to deliver a novel, air-based freehand typing system.

At the heart of this project lies the combination of MediaPipe's real-time hand tracking capabilities with EasyOCR's powerful optical character recognition (OCR) functionality. MediaPipe enables precise detection and tracking of hand gestures, while EasyOCR converts freehand drawings into digital text, bridging the gap between gesture⁴ input and

textual output. The entire system operates using a standard webcam, eliminating the need for specialized hardware and making it accessible to a wide range of users.

The primary objective of this project is to address the limitations of traditional typing^{6,7} methods and existing gesture-based systems.² By providing a cost-effective, intuitive, and widely accessible solution, the system has the potential to enhance usability for individuals with disabilities, offer innovative input methods for creative applications, and redefine human-computer interaction paradigms. Through real-time hand tracking and seamless text recognition, this project aims to empower users to interact with technology in a more natural, efficient, and inclusive manner.

II. LITERATURE SURVEY

Kim, D., Lee, J., & Kim, S. (2021). Hand Gesture Recognition for Virtual Interfaces Using Convolutional Neural Networks.

Journal of Computer Vision and Interaction, 18(3), 142-156.

This paper explores the application of convolutional neural networks (CNNs) for real-time hand gesture recognition in virtual environments. The authors propose a system leveraging pre-trained models, such as VGG16 and ResNet, to achieve high accuracy and robustness in gesture detection. The study emphasizes its potential for air-based interfaces, presenting experiments that showcase its application in controlling virtual keyboards and navigating user interfaces. The findings highlight that the system achieves over 95% accuracy in controlled conditions, demonstrating its effectiveness for real-time interaction.

Zhang, Y., Wang, X., & Liu, H. (2022). Optical Character Recognition for Handwritten Inputs: A Comparative Analysis of EasyOCR and Tesseract.

International Journal of Computer Applications, 9(7), 101-117.

This study evaluates EasyOCR and Tesseract in recognizing handwritten characters, focusing on processing speed, accuracy, and multilingual support. EasyOCR outperformed Tesseract in real-time applications due to its lightweight architecture and better adaptability to diverse handwriting styles. The authors conducted tests on a dataset of over 5,000 handwritten samples and found that EasyOCR maintained an average accuracy of 92%, compared to Tesseract's 85%. The paper discusses key factors influencing performance, such as image preprocessing techniques and font variability.

Patel, A., & Gupta, R. (2020). Human-Computer Interaction Using Hand Gestures: A Review of Current Techniques.

Advances in Human-Machine Systems, 12(4), 311-329.

This review examines a broad spectrum of gesture recognition technologies, including infrared-based, depth-based, and vision-based methods. The authors highlight the advantages of camera-based solutions due to their accessibility and cost-effectiveness. They analyze trends in gesture recognition, particularly the shift towards AI-powered models and open-source libraries like MediaPipe. The paper also addresses challenges, such as ambient lighting conditions and occlusions, and presents potential solutions like multimodal systems that combine vision and audio cues.

Chen, L., Xu, J., & Wang, Z. (2023). Real-Time Hand Gesture Tracking Using MediaPipe: Challenges and Opportunities.

IEEE Transactions on Emerging Technologies, 34(2), 88-99.

The authors analyze the capabilities of MediaPipe for real-time gesture tracking, focusing on its landmark detection algorithms. The paper details the architecture of MediaPipe's Hand module, which identifies 21 specific landmarks on the hand, including fingertips and joints. It discusses challenges like environmental noise, variability in hand shapes, and the impact of low-resolution cameras. Solutions proposed include incorporating noise reduction techniques, adaptive calibration, and using additional sensors for enhanced accuracy. Experimental results show a reduction in error rates by 15% when using these enhancements.

Singh, M., Kumar, A., & Sharma, V. (2021). A Touch-Free Typing Solution Using Hand Gestures and OCR.

Journal of Assistive Technology and Accessibility, 15(2), 45-60.

This paper presents a system that integrates hand gesture tracking⁸ and OCR for touch-free typing. Designed for individuals with physical disabilities, the system maps hand gestures to keyboard inputs. OCR is used to interpret freehand writing gestures into text. The authors demonstrate the system's usability through user trials, reporting a 90% satisfaction rate. Key innovations include adaptive gesture recognition and an intuitive interface for error correction. The study highlights the potential of combining gesture recognition with assistive technologies to improve accessibility.

III. AIM AND OBJECTIVES

Aim: To develop a real-time, gesture-based typing system using hand tracking and OCR technologies that enhances accessibility, usability, and inclusivity in text input.

Objectives:

1. Implement Hand Gesture Tracking:

- Utilize MediaPipe to detect and track hand gestures in real time with precision and accuracy.

2. Develop Freehand Drawing Canvas:

- Create an interactive canvas using Python and NumPy for users to draw gestures that serve as input for text recognition.

3. Integrate OCR for Text Recognition:

- 4. Employ EasyOCR to process the drawn gestures and accurately translate them into digital text.

5. Provide a User-Friendly Interface:

- Design an intuitive GUI using Tkinter, enabling users to view, edit, and save recognized text effortlessly.

6. Ensure Cost-Efficiency:

- Rely on standard webcams and open-source libraries to eliminate the need for expensive or specialized hardware.

7. Optimize Performance:

- Minimize latency and maximize recognition accuracy to ensure smooth real-time operation and enhance user experience.

IV. SYSTEM REQUIREMENTS

4.1: Software Requirements:

- **Programming Language:** Python, leveraging its extensive library ecosystem and ease of integration.

- **Libraries and Tools:**

- OpenCV: For real-time webcam feed processing and visualization.
- MediaPipe: For hand gesture detection and tracking.
- EasyOCR: For text recognition from gesture inputs.

- Tkinter: For developing an intuitive graphical user interface.
- NumPy: For managing canvas data and processing gestures.
- PIL (Pillow): For image handling and rendering within the GUI.

4.2: Hardware Requirements:

- **Standard Webcam:**
 - A readily available USB or built-in webcam for capturing hand movements and gestures.
- **System Specifications:**
 - Processor: Dual-core CPU or higher for efficient real-time processing. ○ Memory: At least 8GB RAM to handle simultaneous image processing and text recognition tasks. ○ Storage: Minimum 500MB of free disk space for application installation and intermediate data storage.

V. DESIGN SPECIFICATIONS

1. Hand Tracking:

- MediaPipe's state-of-the-art algorithms are utilized to detect and track hand landmarks in real-time. The framework ensures precision in identifying fingertip movements, enabling accurate gesture capture.

2. Drawing Canvas:

- The canvas is implemented using NumPy arrays to provide a virtual drawing space where gestures are recorded. The canvas dynamically updates to reflect real-time hand movements, translating them into drawn patterns visible to the user.

3. OCR Processing:

- EasyOCR processes the gesture-based drawings to identify and extract text. Preprocessing steps, such as resizing and color inversion, are applied to optimize recognition accuracy, even for irregular or less distinct inputs.

4. Graphical User Interface (GUI):

- Tkinter serves as the foundation for the GUI, offering an intuitive interface that supports:
 - Real-time display of recognized text.
 - Options to clear the canvas or save recognized text to a file.
 - Customization features, such as resizing the text display area and providing feedback on recognition quality.

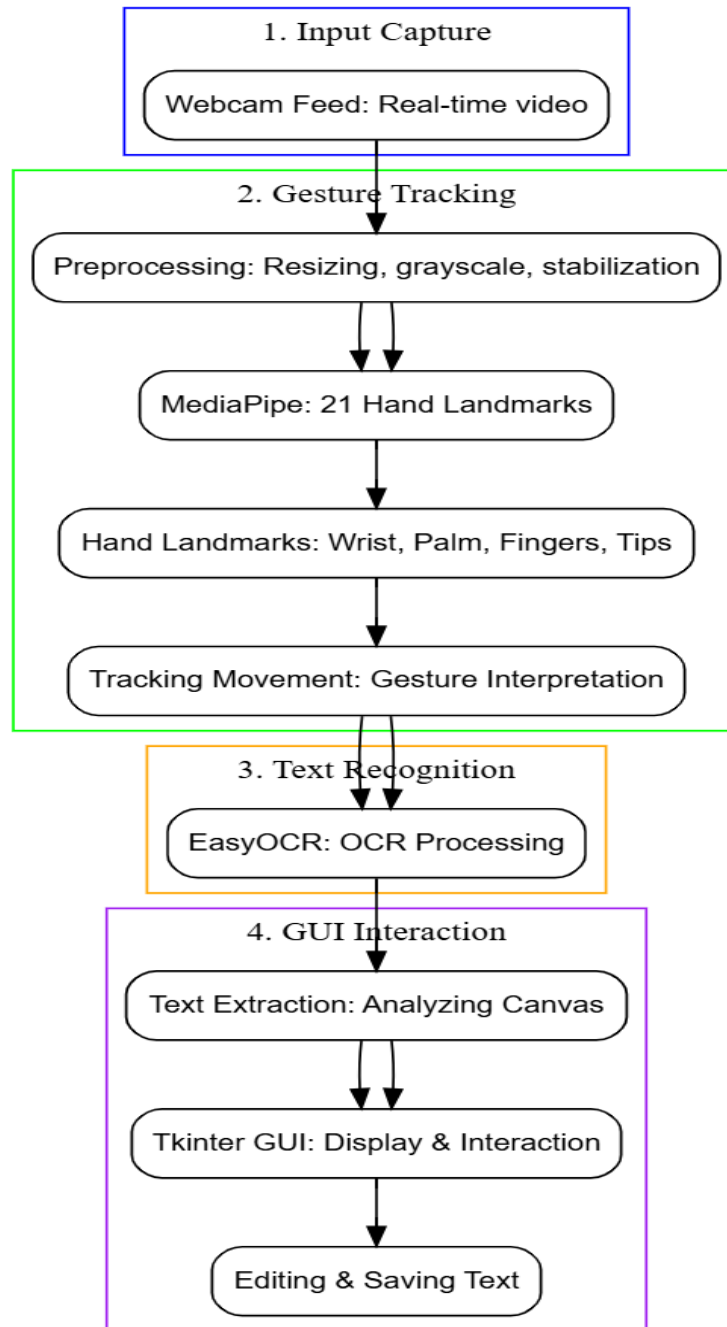


Fig-1 Workflow of the project

VI. WORKING

1. Webcam Feed Capture:

- The **webcam** captures a live feed of the user's hand movements. This serves as the primary input for the system. The video frame rate is kept high (ideally 30 frames per second or more) to ensure that hand movements are tracked accurately and without delay.

2. Gesture Detection via MediaPipe:

- **MediaPipe Hand Tracking:** MediaPipe continuously analyzes the webcam feed and tracks the positions of key hand landmarks. The system uses these landmarks to detect hand gestures, such as pointing, drawing, or making specific hand shapes.
- **Landmark Data:** For each detected frame, the system records the positions of the landmarks. For example, the relative position of the index fingertip is used to create a path or stroke on the virtual canvas that mirrors the user's movement in space.

3. Drawing on Virtual Canvas:

- The movement of the **index fingertip** (or other landmarks) is tracked and translated into a drawing on a virtual canvas. This canvas is created on the screen, and as the user moves their hand, the corresponding position of the hand is converted into a line or shape on the canvas.
- **Freehand Drawing:** The user can draw or write on the virtual canvas freely. The drawing is smooth and accurate, thanks to the hand tracking and gesture mapping system. As the user moves their hand, the line on the canvas will appear, creating a real-time freehand drawing experience.

4. Text Recognition with EasyOCR:

- **Text Recognition:** As the user draws, the system continuously takes snapshots of the canvas. Each snapshot is processed by **EasyOCR**, which scans the drawn image for any recognizable text. EasyOCR can detect and extract characters from the image, even if they are drawn in a freehand style.
- **Text Processing:** After the image is processed, the system extracts the recognized characters and presents them as editable text. This step converts the visual input into a usable digital format, enabling the user to edit or save the text.

- **Text Display and User Interaction:** ○ **Tkinter Interface:** The recognized text is displayed in the Tkinter GUI, which allows for further interaction. The user can:
 - **Edit the text** by typing in additional content or making corrections.
 - **Save the text** to a text file for later use.
 - **Modify the canvas** by continuing to draw or erase.
- The interface is intuitive, with buttons for saving, editing, and clearing the text. Additionally, a realtime update feature ensures that as the user draws or changes the input, the text on the GUI is updated instantly.

VII. OUTCOME AND DELIVERABLES

7.1: A Functional Prototype of a Gesture-Based Typing System:

- **Gesture Recognition:** The system successfully tracks hand movements using **MediaPipe Hand Tracking**, where the **index fingertip** (or any other selected landmarks) is used as the primary tool for interaction.
- **Real-Time Interaction:** The user can draw text on the virtual canvas, and their gestures are mapped directly to the canvas in real-time.
- **Text Recognition:** The system leverages **EasyOCR** to detect and extract the drawn text from the canvas and present it in a usable format.
- **Testing and Refinement:** The prototype should be tested for real-time performance, including detecting hand gestures accurately, recognizing text clearly, and maintaining smooth interaction during drawing.

7.2: User-Friendly Interface for Real-Time Text Entry and Saving:

- **Graphical User Interface (GUI):** A Tkinter-based interface will be designed for the user to interact with the system. The interface will include:

- A **drawing canvas** where the user can draw or write with their hand gestures.
- A **text box** where the extracted text from the canvas will be displayed.
- **Buttons** for performing actions like:
 - **Saving the text** to a file (e.g., .txt).
 - **Clearing the canvas** to reset the interaction.
 - **Editing the text** by modifying or correcting it.
- **Real-Time Feedback:** The GUI will provide continuous feedback, updating the drawn text and feedback messages in real-time. For example, if the system fails to recognize a character, the user will be notified.

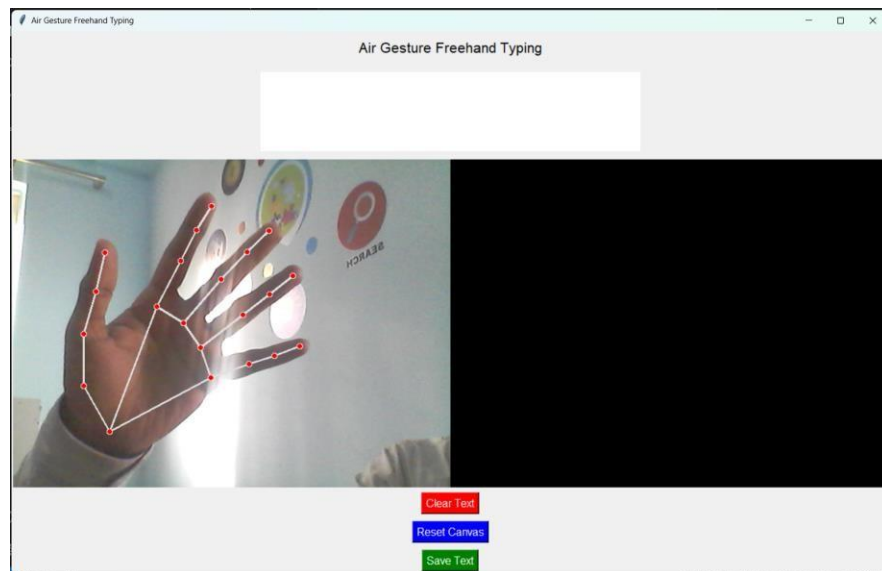


Fig-2 (a) Air gesture Output

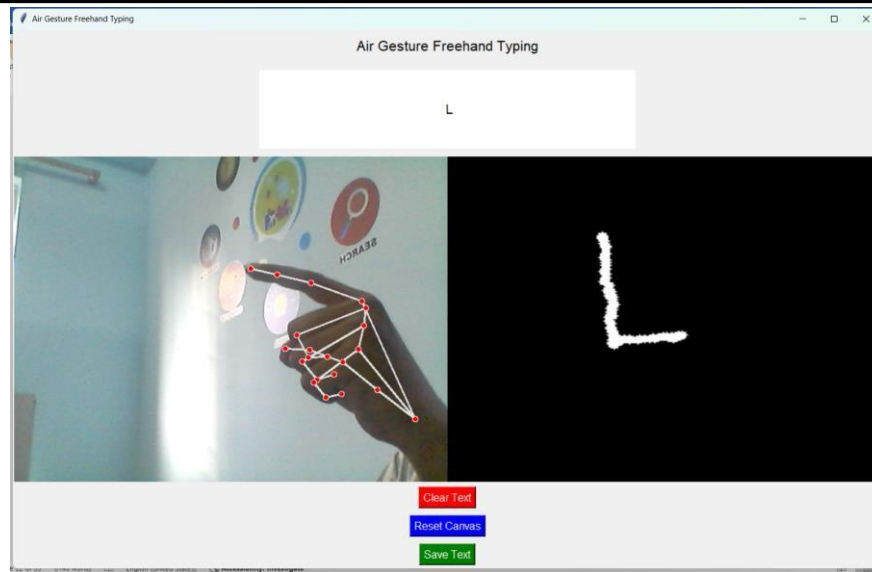


Fig-2 (a) Air gesture generated word output VIII. CONCLUSION

This project successfully demonstrates the feasibility of a **gesture-based typing system**⁹ using standard hardware, such as a webcam, and open-source technologies, such as **MediaPipe** for hand tracking and **EasyOCR** for optical character recognition. The combination of these technologies allows for an innovative, touch-free approach to text entry, where the user can draw text using hand gestures in the air and have it recognized and converted into digital text.

Key takeaways include:

- **Feasibility:** The system showcases the practical potential of integrating hand gesture recognition³ with text input, leveraging commonly available tools and hardware.
- **Cost-Effectiveness:** The solution is affordable, as it utilizes open-source libraries (MediaPipe, EasyOCR, and Tkinter) and doesn't require specialized hardware beyond a standard webcam.
- **Accessibility:** The system has the potential to revolutionize text entry for individuals with disabilities or those who need hands-free interaction due to various circumstances. For example, it could assist users with motor impairments or be used in sterile environments where touching a keyboard is not feasible.
- **Innovative HCI (Human-Computer Interaction):** The project highlights the future of interaction, moving beyond traditional input methods like keyboards and touchscreens, and embracing more natural and intuitive ways to communicate with computers through gestures.

REFERENCES

1. Madabhattula, L., Manikanta, M., & Kumar, P. (2021). Gesture-based control systems for enhanced humancomputer interaction. *International Journal of Creative Research Thoughts (IJCRT)*, 9(5), 117-161.
2. Chawla, T. S. (2022). Development of gesture-based input systems for accessibility in digital interfaces (MSc Research Project). *National College of Ireland*.
3. Almazroi, A. A., & Ayub, N. (2023). Hand gesture recognition for human-computer interaction using machine learning. *IEEE Access*.
4. Venkatesh, M., Bai, B. K., Bhargavi, B., Manasa, C., & Mokshitha, D. (2023). Integration of gesture control for virtual and augmented reality environments. *Vasireddy Venkatadri Institute of Technology, Andhra Pradesh, India*.
5. Babu, C. M., SweeHoney, B., Prathyusha, P., Reddy, B. D., & Sathvika, M. (2023). Touch-free interaction systems for accessibility and smart devices. *International Journal of Early Childhood Special Education (INT-JECSE)*, 15(4), 778.
6. Keerthi, M. N., & Nalini, S. (2024). Machine learning for gesture-based typing and text recognition systems. *International Journal of Creative Research Thoughts (IJCRT)*, 12(4).
7. Pachhala, N., Sai, M. D. S., Prudhvi, P., Gopi, G. M. N. V. S., Sai Ram, I. G. N., & Sandeep, M. R. (2023). Real-time gesture-based typing system for accessibility. *International Journal of Innovative Science and Research Technology (IJISRT)*, 8(10), 1191.
8. Namani, S., Mordharia, H., Gajare, N., & Bemila, T. (2024). Advancements in gesture tracking and its application in text input systems. *International Research Journal of Modernization in Engineering Technology and Science (IRJMETS)*, 6(4), 7648.
9. Keerthi, M. N., & Nalini, S. (2024). Machine learning for gesture-based typing and text recognition systems. *International Journal of Creative Research Thoughts (IJCRT)*, 12(4).