



AI VirtualPainter

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Abstract—The AI Virtual Painter project is an innovative application that merges computer vision, real-time hand tracking, and artistic expression for an immersive experience. Developed for artistic expression, interactivity, education, entertainment, and as a technology showcase, it utilizes OpenCV, MediaPipe, and machine learning to enhance hand tracking accuracy. With a user-friendly interface, it represents a significant advancement in Human-Computer Interaction, transforming digital creativity through the synergy of computer vision and artificial intelligence. This project holds the capacity to revolutionize our interaction with the digital world by offering a versatile tool for seamless and intuitive digital expression.

- **Gesture Recognition:** The project identifies and interprets a variety of hand gestures such as selecting colors and deletion of drawing.
- **Color Selection:** Users can select and choose different colors from a palette to create their artwork .
- **Eraser:** The project includes tools like an eraser for correcting mistakes and the ability to delete the entire artwork.
- **User-Friendly Interface:** An intuitive and user-friendly interface simplifies the interaction with the system, making it accessible to a wide range of users.
- **Artistic Filters and Effects:** Users might have access to artistic filters and effects to enhance their digital artwork.
- **Accessibility Features:** Accessibility features to accommodate users with varying levels of mobility or dexterity.
- **Performance Optimization:** The system is optimized for real-time performance to ensure smooth and responsive interaction between the user.

I. INTRODUCTION

The AI Virtual Painter project embodies a creative and immersive exploration at the intersection of art and technology, rooted in the continually advancing fields of computer vision and human-computer interaction. The central concept involves hand tracking, where a device monitors and traces a user's hand movements in real-time through sophisticated computer vision algorithms. Employing MediaPipe and OpenCV, the project generates immersive environments for an interactive art experience, facilitating the creation of simple digital artwork without the need for physical drawing tools. Delving deeper into AI Virtual Painter reveals potential applications in education, gaming, and accessibility, reshaping the methods of teaching and learning while empowering users to unleash their creativity. The project's features enable users to craft digital art through finger tracking facilitated by computer vision technology.

II. CHARACTERISTICS

Following are the characteristics of the AI Virtual Painter:

- **Real-Time Hand Tracking:** The system can track the user's hand or fingers in real-time, allowing them to interact with the digitally.

III. RELATED WORK

Several recent studies have explored diverse approaches in the field of hand gesture recognition for interactive applications.

[1] Arwoko et al. proposed a system based on Keypoint Vector for hand gesture recognition, as presented in the 2022 International Electronics Symposium (IES).

[2] Bakheet and Al-Hamadi, in their work published in the Eurasip Journal on Image and Video Processing, introduced a robust hand gesture recognition system employing multiple shape-oriented visual cues.

[3] Harish S. presented a computer vision-based hand gesture recognition system in a 2023 ResearchGate publication.

[4] Reddy, Kavya, and Sudheer developed a Virtual Paint and Volume Control system using hand gestures, as described in their work.

[5] Deval et al. explored hand gesture detection and recognition in the International Journal of Advanced Research in Science, Communication, and Technology.

[6] Akshat Rastogi, Aryan Aggarwal, Ayushi Vashista, Anmol Rastogi have worked on virtual AI painter using OpenCV and MediaPipe was presented in the European Chemical Bulletin.

[7] Shinde proposed a hand gesture recognition system using a camera in an IJERT publication.

[8] Patil et al. discussed virtual painting with OpenCV using Python.

[9] Niharika M., Neha J., Mamatha Rao, and Vidyashree K. P. developed a virtual paint application utilizing hand gestures, as detailed in their paper.

[10] To understand the broader context of human-computer interaction, the book "Human-Computer Interaction" by Dix et al. offers a comprehensive resource.

IV. METHODOLOGIES

The AI Virtual Painter project employs diverse datasets for machine learning model training, integrating OpenCV, MediaPipe, and custom algorithms. The user-friendly interface combines real-time hand tracking, artistic effects, and educational features. Rigorous testing ensures optimal performance, delivering an innovative and immersive digital art experience for artistic expression, education, and entertainment

- **Data Collection:** Collect diverse datasets of hand movements, artistic gestures, and color preferences to train machine learning models for precise hand tracking and gesture recognition.
- **Preprocessing:** Apply preprocessing techniques to enhanced datasets, including normalization, background subtraction, and noise reduction, ensuring optimal model training.
- **Technology Stack Selection:** Choose OpenCV and MediaPipe for computer vision and hand tracking functionalities. Integrate machine learning algorithms to enhance hand tracking accuracy.



Fig. 1. MediaPipe's Hand Landmark Model

[11]

- **Model Training:** Train machine learning models using the collected datasets to recognize hand movements, gestures, and color preferences.
- **System Integration:** Integrating machine learning models into a cohesive system and ensuring seamless communication for accurate hand tracking and gesture recognition.

- **User Interface Design:** Design an intuitive and user-friendly interface, incorporating artistic elements and interactive features. Consider accessibility for users with diverse abilities.
- **Enhancements:** Integrate features that enhance the value, such as dynamic artistic effects, interactive elements, and engaging user interactions.
- **Performance Optimization:** Optimize the system for real-time performance using parallel processing and efficient algorithms, ensuring a smooth and responsive user experience.
- **Testing and Validation:** Conduct rigorous testing to validate hand tracking accuracy, gesture recognition, and overall system functionality. Iterate on the design based on testing results and user feedback.

V. PROPOSED SYSTEM

AI Virtual Painter revolutionizes digital drawing by eliminating the need for traditional input devices. Leveraging computer vision, it utilizes the MediaPipe and OpenCV libraries for real-time hand tracking. Users can draw on computer screen and canvases effortlessly using their fingers, with the system translating finger movements into strokes. This innovation ensures a seamless and intuitive drawing experience, marking a departure from conventional input methods. The integration of advanced technologies promises to redefine the way users interact with digital canvases, making artistic expression more accessible and engaging.

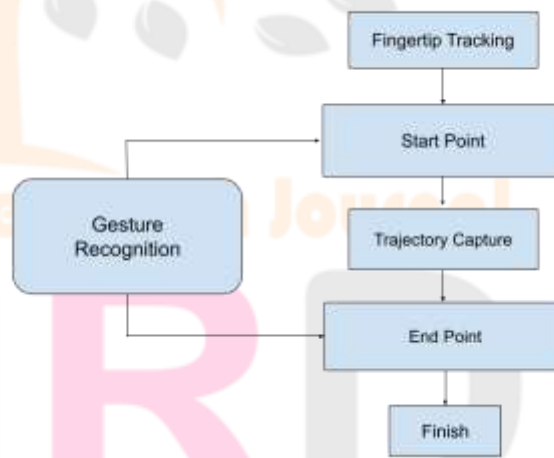


Fig. 2. Flowchart of Proposed System

VI. ALGORITHMS

The AI Virtual Painter project combines several computer vision and machine learning algorithms to achieve its immersive painting experience. Here's a breakdown of the key algorithms involved:

- **Background Subtraction:** It is a common technique used in computer vision for hand detection, particularly in scenarios where the hand is the moving object against

a relatively static background. The idea is to differentiate between the foreground (hand) and the background by identifying pixels that have changed significantly between consecutive frames.

Equations and Formulas: While the process involves image manipulation and thresholding, there aren't specific equations or formulas universally applied. The key steps involve simple arithmetic operations and logical thresholding:

Frame Difference:

$$\text{FrameDifference} = \text{CurrentFrame} - \text{BackgroundModel} \quad (1)$$

- **Kalman Filter:** The Kalman filter combines predictions from a mathematical model of the system with noisy measurements to provide a more accurate and reliable estimate of the system's state. The basic equations of the Kalman filter are as follows:

Prediction Step:

$$\hat{x}_{k|k-1} = F_k \hat{x}_{k-1|k-1} + B_k u_k \quad (2)$$

Covariance Prediction:

$$P_{k|k-1} = F_k P_{k-1|k-1} F_k^T + Q_k \quad (3)$$

Update Step:

$$K_k = P_{k|k-1} H_k^T (H_k P_{k|k-1} H_k^T + R_k)^{-1} \quad (4)$$

State Update:

$$\hat{x}_{k|k} = \hat{x}_{k|k-1} + K_k (z_k - H_k \hat{x}_{k|k-1}) \quad (5)$$

Error Covariance Update:

$$P_{k|k} = (I - K_k H_k) P_{k|k-1} \quad (6)$$

where:

$\hat{x}_{k|k-1}$ is the predicted state estimate at time k given measurements up to time $k-1$.

F_k is the state transition matrix.

B_k is the control-input matrix.

u_k is the control input.

$P_{k|k-1}$ is the predicted error covariance.

Q_k is the process noise covariance.

H_k is the measurement matrix.

z_k is the measurement vector at time k . R_k is

the noise covariance.

K_k is the Kalman gain.

- **Contour Analysis:** Computer Vision technique used to identify and analyze the contours or outlines of objects within an image. The contours represent the boundaries of objects, and contour analysis involves extracting and processing these contours for various purposes, such as object recognition, shape analysis, and feature extraction. A parametric representation of a 2D contour could be given by a set of equations:

$$x(t) = f_x(t) \text{ and } y(t) = f_y(t) \quad (7)$$

Here, $x(t)$ and $y(t)$ are the parametric equations representing the contour, and t is a parameter that varies along the contour.

- **Mean Shift Algorithm:** It is a non-parametric clustering algorithm commonly used for image segmentation and tracking. It operates by iteratively shifting a set of data points toward the mean of the points in their neighborhood until convergence. The algorithm is especially useful for handling irregularly shaped clusters.

For each seed point x_i :

$$m(x_i) = \frac{\sum_{x_j \in N(x_i)} K(x_j - x_i) \cdot x_j}{\sum_{x_j \in N(x_i)} K(x_j - x_i)} \quad (8)$$

where $N(x_i)$ is the neighborhood of x_i .

Repeat these steps until convergence.

- **Stroke Prediction and Smoothing:** Stroke prediction and smoothing are techniques commonly used in computer graphics and digital drawing applications to enhance the fluidity and aesthetics of user-drawn strokes. These techniques aim to predict the trajectory of the stroke and reduce the impact of jitter or irregularities introduced by the user's hand movements.

Stroke Prediction: Let P_i represent the predicted point at time i , and P_{i-1} represent the previously drawn point. A simple linear prediction equation might be:

$$P_i = P_{i-1} + \alpha \cdot (P_{i-1} - P_{i-2}) \quad (9)$$

where α is a prediction coefficient. *Stroke Smoothing:* Assuming a series of points P_i represent the raw stroke data:

$$\text{SmoothedPoint}_i = (1 - \beta) \cdot P_i + \beta \cdot \text{SmoothedPoint}_{i-1} \quad (10)$$

where β is a smoothing factor.

- **Convolutional Neural Networks:** It enhance user interaction by recognizing hand gestures, allowing intuitive control. Recurrent Neural Networks (RNNs) predict and smooth drawing strokes, improving fluidity and accuracy. CNNs apply artistic filters for visual enhancement, while hand tracking accuracy benefits from CNN-based refinement. Utilizing a CNN for artistic style transfer, allowing users to apply various filters and effects to their digital artwork. RNNs analyze user interactions for a personalized interface experience. This integrated use of CNNs and RNNs revolutionizes digital creativity, offering a seamless and intuitive tool for artistic expression in real-time.

VII. DESIGN DETAILS

The AI Virtual Painter design includes multiple components that work together to create a seamless user experience. Here's the detailed breakdown:

- **System Architecture:**

- Camera Interface: This module connects to camera hardware to capture video input. Access and edit camera feeds using libraries such as OpenCV.
 - MediaPipe Integration: This component integrates his MediaPipe library for hand tracking. Processes video images to identify and track the user's hands.
 - Algorithmic approach: Depending on the approach chosen, this module can identify individual fingers using techniques such as contour analysis, convex hulls, and machine learning models.
 - Mapping algorithm: This component maps the detected finger position from camera space to screen space. Accurately track hand movements and convert them to screen coordinates.
- **User Interface (UI):**
 - UI elements: Graphical User Interface (GUI) elements allow users to select drawing tools (such as brush) and set parameters such as size and color.
 - Color selection interface: Users can select colors from a palette that can be implemented as a graphical UI component.
 - Visual indicators: The interface provides visual feedback indicating the selected color, brush size, etc.
 - **Handling interactions:**
 - Gesture recognition: If gestures are integrated, you will need a module that recognizes gestures (pinch to zoom, two-finger swipe, etc.).



Fig.3.Flowchart of AI Virtual Painter

- **Integration with external libraries:**
 - MediaPipe integration: API Interface: This module manages integration with the MediaPipe library for hand tracking.
 - OpenCV integration: Camera feed processing: This component handles interaction with the OpenCV library for camera input processing.

VIII. EXPERIMENTAL SETUP

A. Details About Input to Systems

The AI Virtual Painter involves using computer vision techniques for hand tracking and gesture recognition to allow humans to draw on the screen with their fingers. The primary input to the application is the live video stream captured from a webcam or connected camera.

- **Camera input:** The project relies on a camera as its primary input source. This camera can be a webcam or another type of digital camera that can record video in real-time.
- **MediaPipe:** MediaPipe is a framework developed by Google that provides a variety of prebuilt machine learning models for various computer vision tasks such as hand tracking.
- **Hand Tracking:** The system uses MediaPipe's hand tracking model to detect and track the user's hands in real time. The aim is to identify key points on the hand, such as the fingertips, palm, and joints.



Fig.4. Hand Tracking using MediaPipe

- **OpenCV:** OpenCV (an open-source computer vision library) may be used to access and manipulate video images. OpenCV provides a wide range of computer vision tools and features essential for tasks such as image processing and feature recognition.
- **Gesture Recognition:** Implementing gesture recognition to interpret the movements of the user's hands and poses. For example, certain hand gestures may be associated with actions such as drawing, erasing, changing colors, and selecting tools.
- **User interaction:** Recognized hand movements and gestures are used to interact virtually on the screen. The user's movement is captured with their finger, the system interprets the movement and draws corresponding lines and shapes on the screen.

IX. IMPLEMENTATIONSTEP

Implementing the AI Virtual Painter project involves several steps, combining computer vision, hand tracking, and interactive graphics. Below is a simplified outline of the implementation process:

- **Set Up Development Environment:** Install necessary software and libraries, including Python, MediaPipe, OpenCV, and any other dependencies.
- **UnderstandMediaPipeandOpenCV:**Familiarizeyourself with the functionalities of MediaPipe and OpenCV, especially the modules related to hand tracking and computer vision.
- **Camera Setup:** Connect and configure the camera to capture video input. Ensure that the camera has a clear view of the user’s hand movements.
- **Initialize Hand Tracking:** Use MediaPipe to initialize handtrackingbydetectingandlocalizingtheuser’shands in the camera feed.

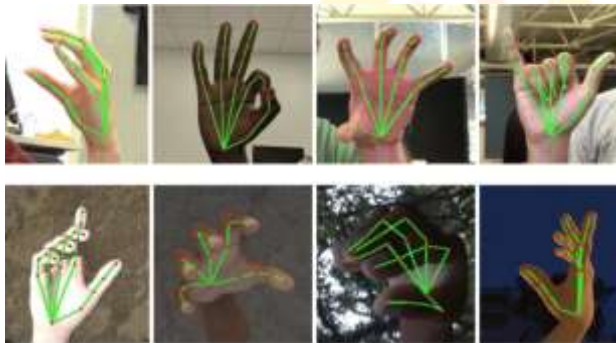


Fig.5.HandPositionTrackingusingCamera [12]

- **Extract Hand Landmarks:** Utilize the hand tracking results to extract the landmarks or key points of theuser’s hand. These landmarks will be used to track hand movements.
- **Define Canvas:** Set up a virtual canvas or interactive space where users can draw. This could be a 2D or 3D space, depending on the project’s goals.



Fig.6.Header

- **Optimization:**Optimizethecodeforefficiency,consideringfactors suchasframerate,responsiveness,andoverall performance.
- **Real-time Rendering:** Implement real-time rendering of thedigitalartworkontheinteractivecanvas.Thisrequires updating the display continuously as the user interacts with the system. [13]



Fig.7.RealTimeRendering

- **Testing:** Test the system extensively to ensure that hand tracking is accurate, drawing is smooth, and the overall user experience meets expectations.
- **Draw on the Canvas:** Based on the tracked hand landmarks, implement a drawing mechanism that allows users to create digital artwork. This could involve using gestures or movements to control the drawing tools.



Fig.8.Canvas

- **Iterate and Improve:** Gather feedback from users and iterate on the project, making improvements based on user experience and any identified issues.

X. PERFORMANCEMETRICS

The performance evaluation parameters for the project include accuracy in hand and gesture recognition, low latency for real-time interactions, high frame rate for smooth interactions, tracking robustness under various conditions, precision and recall in gesture recognition, quick response time, and compatibility across diverse devices and operating systems.

For the given application the video is recorded in 30fps for time 10s. Here are some potential performance metrics with associated equations:

- **Hand Tracking Accuracy:** Percentage of accurately tracked hand movements.

$$\begin{aligned}
 \text{Acc.} &= \frac{\text{No. of accurately tracked frames}}{\text{Total number of frames}} \times 100\% \\
 &= \frac{255}{300} \times 100\% \\
 &= 85\%
 \end{aligned} \tag{11}$$

- **Gesture Recognition Accuracy:** Accuracy in recognizing predefined gestures.

$$\begin{aligned}
 \text{Acc.} &= \frac{\text{No. of correctly recognized gestures}}{\text{Total number of gestures}} \times 100\% \\
 &= \frac{224}{300} \times 100\% \\
 &= 74.6\%
 \end{aligned} \tag{12}$$

- **Frame Rate:** Number of frames processed per second.

$$\begin{aligned}
 \text{Frame Rate} &= \frac{\text{No. of Frames}}{\text{Total Time (in seconds)}} \\
 &= \frac{300}{10} \\
 &= 30 \text{ fps}
 \end{aligned} \tag{13}$$

- **Drawing Accuracy:** Accuracy in translating hand movements to on-screen strokes.

$$\begin{aligned}
 \text{Accuracy} &= \frac{\text{No. of accurately drawn strokes}}{\text{Total number of strokes attempted}} \times 100\% \\
 &= \frac{282}{300} \times 100\% \\
 &= 94\%
 \end{aligned} \tag{14}$$

- **Responsiveness:** Time delay between a user's action and system response.

$$\begin{aligned}
 \text{Responsiveness} &= \frac{\text{No. of interactions}}{\text{Total processing time}} \\
 &= \frac{20}{30} \\
 &\approx 0.67 \text{ s}
 \end{aligned} \tag{15}$$

- **User Satisfaction (Survey-based):** User feedback on the overall satisfaction with the system.

$$\begin{aligned}
 \text{User Satisfaction} &= \frac{\text{Sum of user ratings}}{\text{No. of user responses}} \times 100\% \\
 &= \frac{20}{22} \times 100\% \\
 &= 90\%
 \end{aligned} \tag{16}$$

XI. CONCLUSION

The AI Virtual Painter project represents a groundbreaking fusion of computer vision, gesture recognition, and real-time interaction, transforming digital creativity. Through the integration of technologies like MediaPipe and OpenCV, the system allows users to draw on a screen using their fingers, eliminating the need for traditional input devices. The project

excels in accuracy, tracking robustness, and gesture recognition precision. It achieves low latency and high frame rates, ensuring a seamless and responsive drawing experience.

The precision and recall metrics validate the robustness of gesture recognition, while response time optimization contributes to smooth user interactions. The system's compatibility across devices and operating systems enhances its accessibility and usability. In conclusion, the AI Virtual Painter not only pioneers intuitive digital painting but also exemplifies the potential of computer vision and gesture-based interfaces in revolutionizing interactive and immersive applications.

Thus, the AI Virtual Painter project offers a glimpse into the future of how technology can enhance creativity and interaction. With continued development and innovation, it has the potential to become a valuable tool for artistic expression, education, and entertainment.

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