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Result Paper on Air Writing Recognition Using Machine Learning Algorithms

Prof. Y. L. Tonape, Haral Abijeet, Parkale Sudarshan, Pawar Rohit, Pawar Sandesh

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Abstract

Air writing is a novel gesture-based input technique that enables users to write in the air using their hand movements instead of relying on physical surfaces. This approach is particularly useful in accessibility solutions, human-computer interaction (HCI), and augmented/virtual reality (AR/VR) applications[1]. Traditional input methods like keyboards and touchscreens have inherent limitations, especially for individuals with motor disabilities or in scenarios requiring hands-free interaction.

In this study, we propose a Convolutional Neural Network (CNN)-based air-writing recognition system that processes real-time hand gestures captured via a standard webcam. The system employs OpenCV for hand tracking, extracts key movement features, and classifies them into characters using a deep learning model. Our method achieves high accuracy and real-time performance, making it feasible for applications in education, assistive technology, smart home interfaces, and digital signatures.

Introduction

Background and Motivation:

With the rapid advancement of human-computer interaction (HCI) technologies, there is a growing need for contactless, intuitive input methods. While traditional text input methods such as keyboards, styluses, and touchscreens are effective, they pose limitations in accessibility and flexibility. Many real-world scenarios require hands-free input, such as:

- 1. **Individuals with disabilities** Users who face difficulties in operating conventional input devices.
- 2. **Smart home and IoT interactions** Where voice or gesture commands provide a seamless experience.
- 3. **Industrial environments** Where physical input devices may be contaminated, damaged, or inconvenient.

Why Air Writing?

Air writing allows users to write characters in mid-air using simple hand gestures. These

movements can be captured using computer vision techniques and interpreted through machine learning models. This approach:

- Removes dependency on physical surfaces
- Supports hands-free and contactless interactions
- Improves accessibility for individuals with disabilities
- Offers a futuristic approach to text input

Objectives

This research aims to:

- 1. Develop a real-time air-writing recognition system using machine learning.
- 2. Enhance recognition accuracy across various handwriting styles.
- 3. Ensure real-time performance with minimal computational resources.
- 4. Enable scalability to support word-level and sentence-level recognition.

Problem Statement

The problem of air handwriting arises from the need to create a more accessible, efficient, and intuitive method of inputting text and commands into electronic devices. This approach is especially beneficial for individuals with physical disabilities, such as those with limited mobility or dexterity, who may find traditional input methods like keyboards or touchscreens difficult or impossible to use. Air handwriting enables handsfree interaction, offering a more inclusive solution that empowers individuals to communicate and interact with technology more effectively in a wide range of scenarios.

Limitations of Existing Systems Variability in Handwriting Styles:

- Different users have unique writing patterns that impact recognition accuracy.
- Speed, stroke shape, and slant variations increase complexity.

Real-Time Processing Limitations:

- Many existing handwriting recognition systems have high latency due to complex computations.
- The need for fast inference models is crucial.

Interference:

 Lighting conditions, background noise, and motion blur affect system performance.

High Hardware Dependency:

 Sensor-based air-writing recognition systems require specialized cameras or sensors.

Solution:

To address the limitations of existing air-writing recognition systems, our proposed system integrates enhanced deep learning algorithms, adaptive learning techniques, and an extended capability for advanced applications. The system

Image-Based Systems: Use cameras to capture hand movements but are affected by lighting and background noise, reducing accuracy.

Sensor-Based Systems: Use wearables with accelerometers and gyroscopes but struggle with variations in writing styles and speeds.[1]

Marker-Based Systems – Require wearable sensors, making them inconvenient.

Marker-Free Systems – Struggle with accurate hand tracking, especially in cluttered backgrounds.

Push-to-Write Mechanisms – Use predefined areas for writing, limiting natural movement. [2] **Sensor-Based Limitations**:

 Leap Motion and glove-based systems restrict user mobility and can be cumbersome. leverages computer vision and deep learning models to enable real-time and accurate airwriting recognition.

Key Enhancements in the Proposed System

1. Enhanced Gesture Recognition Algorithms

- We utilize OpenCV for precise hand tracking and segmentation.
- The system applies spatial and temporal filtering to remove background noise and improve accuracy.
- A Convolutional Neural Network (CNN) is employed to recognize character patterns with higher accuracy than traditional methods[2].

2. Adaptive Learning for Varied User Inputs

- The system incorporates self-learning mechanisms that adjust to different handwriting styles over time.
- It employs data augmentation techniques to generalize across various writing speeds and angles.
- An adaptive thresholding approach improves recognition for both fast and slow writers.

3. Extended Capability for Advanced Handwriting Recognition

- Beyond character recognition, the system can identify words and phrases through sequence modelling.
- It is capable of mathematical symbol recognition, allowing users to write equations and solve problems in realtime
- The architecture is designed to support multilingual handwriting recognition, enabling applications across diverse user bases.

Literature Review

 IMU-based wearable devices often require a surface, reducing the natural feel of airwriting.

Detection Challenges:

- Lack of clear pen-up/pen-down motion makes segmentation difficult.
- Existing models struggle with variations in writing speed and style.

Computational Complexity:

 Some recognition methods have high processing demands, making real-time recognition challenging. [3]

Hardware Dependency – Many systems require specialized sensors (e.g., IMUs, radar, Leap Motion), making them less practical for general use.

Writing Variability – Differences in individual writing styles and motion speeds make recognition challenging.

Limited Word Coverage – Some systems struggle with a restricted vocabulary, affecting real-world usability. [4]

Sensor-Based Systems – Methods using accelerometers and gyroscopes required users to

The proposed Air Writing Recognition System follows a structured pipeline that captures hand gestures in the air, processes them through image processing and deep learning techniques, and outputs recognized characters in real-time. The architecture is designed to be efficient, hardware-independent, and scalable for multiple applications, including assistive technology, smart environments, and augmented reality (AR)/virtual reality (VR) interfaces[3].

The system workflow consists of multiple stages, starting from user interaction to final character recognition and output display. The flow of data and processes is as follows:

- **1. User Input:** The user performs air-writing using their index finger or hand movement, creating invisible strokes in the air.
 - Live Camera Feed (real-time video processing).
 - Static Images (pre-recorded handwriting samples).
- **2. Input Capture:** The system captures input through either:

3. Processing Pipeline:

Algorithm

After feature extraction, the processed data is fed into Convolutional Neural Network (CNN), which classifies the air-written characters.

CNN Model Architecture

- 1. Convolutional Layers: Extracts highlevel features such as edges, curves, and handwriting patterns.
- 2. Pooling Layers: Reduces dimensionality while retaining key features.
- 3. Fully Connected Layers: Processes extracted features and maps them to specific characters.
- 4. Softmax Output Layer: Predicts the probability of each character and selects the highest confidence score.

Training and Optimization

 The model is trained on a large dataset of air-written characters to generalize across users. wear extra hardware, making them less user-friendly.

Leap Motion-Based Systems – While providing precise finger tracking, these systems had limited accuracy, with models like BLSTM achieving around 86.88%.

Wi-Fi Signal-Based Methods – These systems offered good accuracy but were sensitive to environmental interference, affecting reliability. [5]

- Hand Detection & Tracking (Using OpenCV).
- Image Preprocessing: Conversion to grayscale, noise removal, edge enhancement.
- Feature Extraction: Identification of key hand movement features using deep learning techniques.
- Character Recognition: Classification using a Convolutional Neural Network (CNN).
 - 4. Output Display: The recognized character/text is displayed in real-time.

System Architecture



Fig1. System Architecture

- Data augmentation techniques (rotation, scaling) improve recognition robustness.
- The model is optimized using the Adam optimizer and cross-entropy loss function.

Application

1. Assistive Technology for Disabled Individuals

Use Case:

- People with motor disabilities or speech impairments often struggle with traditional input methods.
- Air writing provides a touchless and accessible way to interact with digital devices.

How It Helps:

• Users can write in the air using hand gestures, which is converted into text.

- It can be integrated with text-tospeech (TTS) systems for verbal communication.
- Supports custom gestures for individuals with unique movement patterns.

Example: A person with cerebral palsy can use air writing to communicate via a computer screen instead of using a keyboard or touch interface.

2. Smart Classrooms & E-Learning Platforms

Use Case:

- Teachers and students can write in the air during online classes, making digital learning more interactive and engaging.
- The system allows students to practice handwriting and mathematical equations without pen and paper.

How It Helps:

- Converts air-written characters into digital text for easy note-taking.
- Supports mathematical symbols and formulas, which are difficult to type on a keyboard.
- Enables real-time evaluation of handwriting skills in digital education platforms.

Example: A math teacher writes an equation in the air during a virtual class, and it gets displayed as digital text on students' screens.

3. Human-Computer Interaction (HCI) & Smart Devices

Use Case:

- Air writing serves as a natural interface for smart devices, replacing physical keyboards and touchscreens.
- It enhances gesture-based communication with computers and IoT devices.

How It Helps:

- Enables touch-free text input on smart TVs, tablets, and AR/VR devices
- Works as an alternative to voice commands in noisy environments.
- Can be integrated with smartwatches and wearables for quick note-taking.

Example: A user wearing a VR headset can type a message by writing in the air instead of using a virtual keyboard.

4. Digital Signature Authentication & Secure Transactions

Use Case:

- Handwritten signatures are widely used for verification in banking and security applications.
- Air writing provides a contactless way to sign documents in digital transactions.

How It Helps:

- Users can air-write their signature, which is verified using biometric authentication.
- Reduces the risk of forgery and fraud through AI-based signature verification.
- Can be used for identity authentication in banking and corporate sectors.

Example: A business executive signs an important contract by writing in the air, and the system verifies the authenticity of the signature.

5. Healthcare & Rehabilitation

Use Case:

- Patients recovering from stroke or hand injuries need rehabilitation for motor control.
- Air writing serves as a therapeutic tool to help them regain hand coordination.

How It Helps:

- Tracks hand movement patterns and measures improvement over time.
- Provides real-time feedback to therapists monitoring rehabilitation progress.
- Can be used in neurological studies to analyze motor control disorders.

Example: A stroke patient practices writing alphabets in the air, and the system tracks their improvement in handwriting over several weeks.

Results

System Performance & Accuracy

 The CNN model was trained on handwritten gestures, achieving an average accuracy of 92.5% on the test set. Performance is high in well-lit environments and moderate writing speeds but slightly decreases in low light or fast hand movements.

Real-Time Recognition Results

 The system successfully detects airwritten text in real-time, converting strokes into digital characters

Screenshots & Description

1. Landing Page - Providing small summary of the project.



Registration / Login Page: Here user can create account and login.



Final Page: From here you can do air writing.



Conclusion

The Air Writing Recognition System successfully converts hand gestures into digital text using CNN-based classification. It provides an efficient, contactless, and intuitive method for text input, making it particularly useful in assistive technology, smart devices, education, and healthcare applications. The system performs well under optimal conditions, ensuring smooth and accurate recognition, but faces minor

challenges in low-light environments and rapid hand movements. By integrating real-time tracking and feature extraction, it effectively detects characters, words, and mathematical symbols. Further improvements in gesture refinement, deep learning models, and adaptive learning techniques can enhance its performance. Additionally, multi-language support and integration with IoT devices could expand its usability. Overall, this system

demonstrates great potential for revolutionizing human-computer interaction with a touch-free and accessible writing solution.

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