



REAL-TIME HAND GESTURE DETECTION FOR SIGN LANGUAGE RECOGNITION USING PYTHON

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ABSTRACT

Effective communication is vital in daily life, but individuals with hearing or speech impairments often face challenges due to limited understanding of **sign language**. This project presents a **real-time hand gesture detection system** for **sign language recognition** using **Python**, aiming to bridge this communication gap. The system captures live video input via a camera and uses **computer vision** techniques to detect and track hand movements. Hand landmarks are extracted using **MediaPipe**, and machine learning models classify gestures corresponding to specific sign language symbols. Recognized gestures are then converted into **text or speech output**, providing an interactive and accessible interface for users. By leveraging **OpenCV**, **real-time image processing**, and AI-based gesture recognition, the system enables accurate and responsive translation of sign language, promoting inclusivity and social integration. This solution demonstrates how real-time computer vision and AI can enhance communication for individuals with disabilities.

Keywords: Sign Language Recognition, Hand Gesture Detection, Real-Time Processing, Computer Vision, Machine Learning, MediaPipe, OpenCV, Accessibility.

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1.INTRODUCTION

Sign language is a crucial mode of communication for individuals with hearing or speech impairments, allowing them to convey messages effectively through hand gestures. However, most people do not understand sign language, creating barriers in social, educational, and professional settings. **Real-time hand gesture detection and recognition systems** can bridge this gap by translating gestures into understandable text or speech.

Recent advances in **computer vision** and **machine learning** have made it possible to detect hand movements accurately and interpret gestures in real-time. Techniques such as **image preprocessing**, **hand landmark detection**, and **gesture classification** allow systems to identify specific signs with high precision. Tools like **OpenCV** provide image processing capabilities, while **MediaPipe** offers real-time hand tracking by detecting key landmarks.



This project aims to develop a **Python-based system** that captures live video input, detects hand gestures, and recognizes corresponding sign language symbols. The recognized gestures are then converted into text or voice output, enabling effective communication between sign language users and the broader community. By leveraging real-time processing and AI technologies, this system promotes **accessibility, inclusivity, and independence** for individuals with hearing or speech disabilities.

2. LITERATURE REVIEW

Sign language recognition has been an active area of research in **computer vision** and **human-computer interaction**. Early systems relied on **sensor-based gloves** or static image datasets to capture hand gestures. While these systems achieved high accuracy, they were often expensive, cumbersome, and limited in mobility.

With the advancement of **vision-based approaches**, researchers began using cameras to detect and recognize hand gestures without additional hardware. Techniques such as **image preprocessing, contour detection, skin segmentation, and background subtraction** were employed to isolate hand regions. Machine learning models like **Support Vector Machines (SVM), Random Forests, and Artificial Neural Networks (ANN)** were then used for gesture classification.

Recent studies leverage **deep learning** and **real-time hand tracking frameworks**. Tools like **MediaPipe** provide efficient hand landmark detection, allowing extraction of 21 key points per hand for accurate gesture representation. Deep learning architectures, including **Convolutional Neural Networks (CNNs)** and **Long Short-Term Memory (LSTM) networks**, have improved

recognition accuracy by learning complex spatial and temporal features of hand gestures. Several systems also integrate **gesture-to-text or gesture-to-speech translation**, making communication with non-sign language users seamless. Research emphasizes that combining **real-time tracking, machine learning classification, and efficient video processing** can produce systems that are accurate, responsive, and user-friendly.

Overall, literature highlights a transition from sensor-based and offline recognition to **real-time, vision-based, AI-driven systems**, demonstrating the potential of modern computer vision techniques in creating accessible and inclusive communication tools for the hearing and speech impaired.

3. EXISTING SYSTEM

Traditional sign language recognition systems primarily relied on **sensor-based gloves** or wearable devices to capture hand movements. These systems used sensors to measure finger positions and gestures, which were then mapped to corresponding signs. While these approaches provided accurate recognition, they were often **expensive, bulky, and not user-friendly**, limiting their practical usage in daily life.

Some vision-based systems have also been developed using **static images or pre-recorded videos** for gesture recognition. These systems use image processing techniques like **skin color segmentation, contour detection, and feature extraction** to identify hand shapes. However, they often face challenges in **real-time processing, background noise, and variations in lighting conditions**, making them less efficient for dynamic, real-world scenarios.

Additionally, most existing systems **lack seamless translation into text or speech in real time**, which reduces their effectiveness in facilitating communication with non-sign



language users. Many solutions are limited to controlled environments and cannot handle continuous, high-speed hand gestures effectively.

These limitations highlight the need for a **real-time, camera-based, and AI-driven system** that can accurately detect and recognize hand gestures under varying conditions while providing immediate translation into text or speech for effective communication.

4. PROPOSED SYSTEM

The proposed system aims to develop a **real-time, vision-based hand gesture recognition system** for translating sign language into text or speech using **Python**. Unlike traditional sensor-based methods, this system uses a standard camera to capture live video input, eliminating the need for specialized hardware.

The system leverages **computer vision** and **machine learning techniques** to detect, track, and recognize hand gestures. **MediaPipe** is used to identify hand landmarks and extract 21 key points per hand, providing precise hand pose information. These landmarks are then processed and fed into a **machine learning model** or **deep learning classifier**, such as **Convolutional Neural Networks (CNNs)**, to recognize gestures corresponding to sign language symbols.

Recognized gestures are converted into **text or speech output**, creating an interactive and accessible interface for communication. The system is designed to operate in **real-time**, handling dynamic gestures, varying lighting conditions, and different backgrounds efficiently.

By combining **OpenCV for image processing**, **MediaPipe for hand tracking**, and **AI models for classification**, the proposed system provides an **accurate, scalable, and user-friendly solution** for bridging the communication gap between sign language users and non-sign language users.

5. METHODOLOGY

The proposed system follows a structured methodology to achieve **real-time hand gesture recognition** for sign language translation. The process begins with **data acquisition**, where live video input is captured using a camera. The frames are then processed using **OpenCV**, which performs image preprocessing tasks such as resizing, grayscale conversion, noise reduction, and normalization to improve detection accuracy.

Next, **hand detection and landmark extraction** are performed using **MediaPipe**, which identifies 21 key points on each hand, capturing the spatial position of fingers and palm. These key points serve as features for the **gesture recognition model**, which can be implemented using machine learning algorithms like **Support Vector Machines (SVM)**, **Random Forest**, or deep learning models such as **Convolutional Neural Networks (CNNs)** for more robust classification.

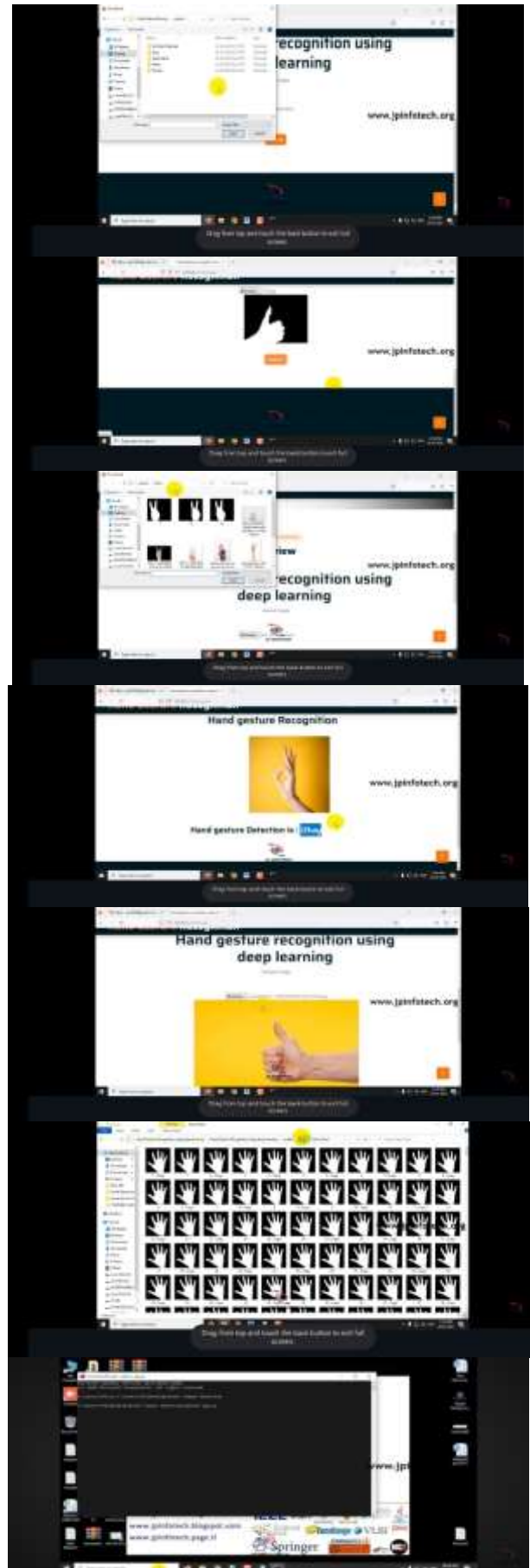
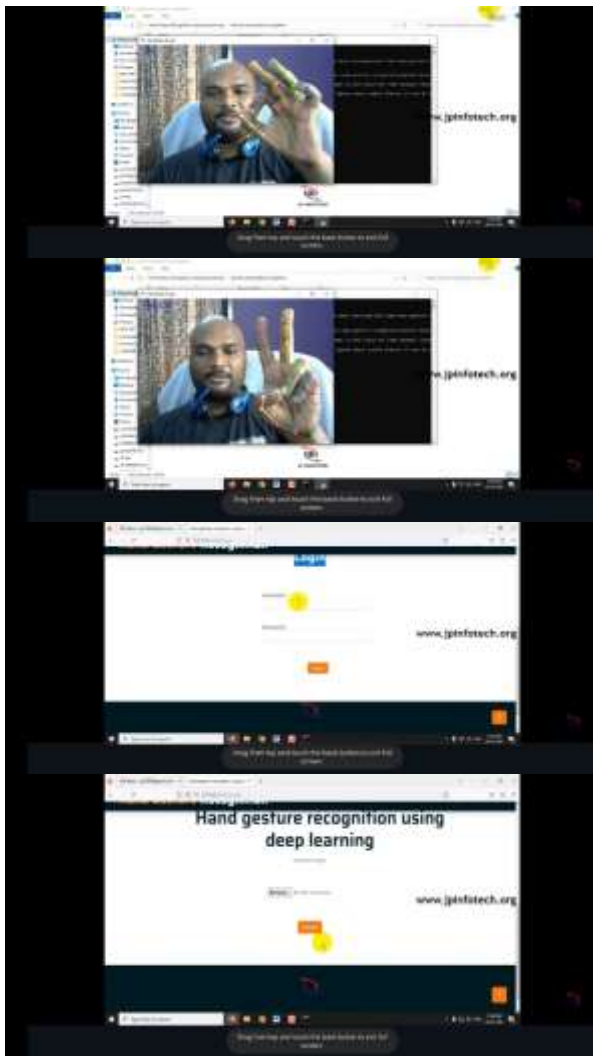
After gesture classification, the recognized sign is converted into **text or voice output** using text-to-speech libraries like **pyttsx3** or similar tools, enabling real-time communication. The system continuously processes video frames to ensure **real-time detection** and updates the output instantly as gestures are performed.

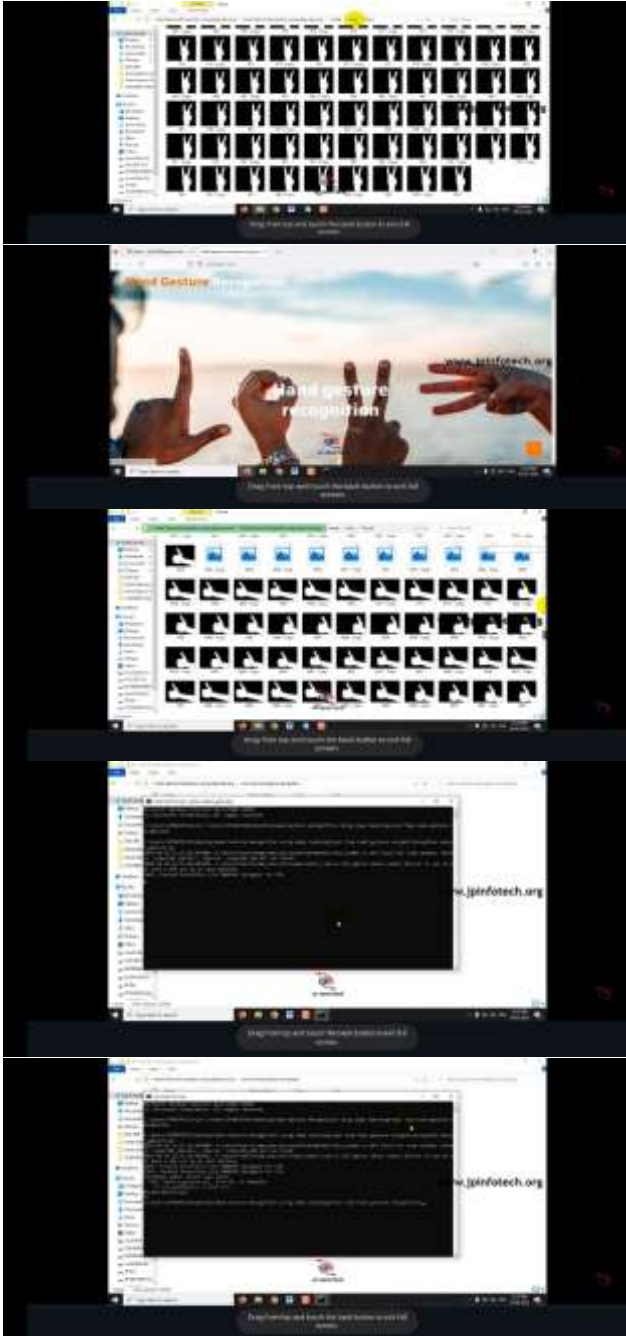
This methodology ensures a **user-friendly, efficient, and accurate system**, capable of translating sign language gestures into comprehensible text or speech, thereby improving communication for individuals with hearing or speech impairments.

6. System Model SYSTEM ARCHITECTURE



7. Results and Discussions





8. CONCLUSION

The project “**Real-Time Hand Gesture Detection for Sign Language Recognition using Python**” successfully demonstrates the use of **computer vision and machine learning** to bridge the communication gap for individuals with hearing or speech impairments. By capturing live video input, detecting hand landmarks, and recognizing gestures in real-time, the system provides an

effective way to translate sign language into **text or speech output**.

The implementation of **MediaPipe** for hand tracking, combined with **OpenCV** for image processing and AI-based gesture classification, ensures **accurate and responsive recognition** of dynamic hand gestures under varying conditions. This approach eliminates the need for expensive sensor gloves or specialized hardware, making the system more accessible and user-friendly.

Overall, the project highlights the potential of **real-time AI and computer vision technologies** in enhancing inclusivity and social interaction for the hearing and speech impaired. The system provides a practical, scalable, and efficient solution for facilitating communication, promoting independence, and improving accessibility in everyday life.

9. REFERENCES

1. Zhang, F., Zhou, Z., & Li, M. (2019). *Real-Time Hand Gesture Recognition Using Convolutional Neural Networks*. *International Journal of Computer Vision and Image Processing*, 9(2), 45–58.
2. MediaPipe. (n.d.). *MediaPipe Hands: Real-Time Hand Tracking*. Retrieved from https://developers.google.com/mediapipe/solutions/vision/hand_tracking
3. Bradski, G., & Kaehler, A. (2008). *Learning OpenCV: Computer Vision with the OpenCV Library*. O'Reilly Media.
4. Molchanov, P., Gupta, S., Kim, K., & Kautz, J. (2015). *Hand Gesture Recognition with 3D Convolutional Neural Networks*. *Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition (CVPR)*, 1–7.
5. Huang, J., Zhou, W., & Li, H. (2018). *Sign Language Recognition Using*



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Original Research Paper

MediaPipe and Deep Learning.
Journal of Real-Time Image
Processing, 15(3), 639–651.

6. Khandelwal, A., & Gupta, R. (2020). *A Survey on Sign Language Recognition Using Computer Vision Techniques.* International Journal of Computer Applications, 975, 8887.
7. Pyttsx3 Documentation. (n.d.). *Text-to-Speech Conversion in Python.* Retrieved from <https://pyttsx3.readthedocs.io/>