

## e-ISSN:2582-7219



# INTERNATIONAL JOURNAL OF MULTIDISCIPLINARY RESEARCH IN SCIENCE, ENGINEERING AND TECHNOLOGY

### Volume 7, Issue 4, April 2024



6381 907 438

INTERNATIONAL STANDARD SERIAL NUMBER INDIA

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Impact Factor: 7.521

6381 907 438 ijmrset@gmail.com

| ISSN: 2582-7219 | www.ijmrset.com | Impact Factor: 7.521 | Monthly Peer Reviewed & Referred Journal |



Volume 7, Issue 4, April 2024

| DOI:10.15680/IJMRSET.2024.0704180 |

## **Sign Language Interpreter for Deaf People**

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**ABSTRACT:** Sign language stands as a timeless and instinctive mode of communication, yet its accessibility is often hindered by the scarcity of proficient interpreters. To bridge this gap, a pioneering real-time solution leveraging neural networks for fingerspelling has emerged. This groundbreaking method undergoes a two-step process: initially, the hand undergoes filtration, followed by classification to predict gesture classes. Remarkably, this innovation achieves an outstanding 95.7% accuracy across the 26 alphabet letters. For individuals unable to speak or hear, sign language serves as their sole means of communication, a vital lifeline enabling the expression of thoughts and emotions. Recognizing its profound significance, a novel approach to sign language recognition has been introduced. This innovative scheme integrates computer vision and neural networks to identify alphabets and gestures, facilitating seamless translation into text output. Embracing cutting-edge technologies such as Convolutional Neural Networks (CNNs), Image Processing, Edge Detection, and Hand Gesture Recognition, this solution marks a significant advancement in accessibility and inclusivity for the physically challenged community. By harnessing the power of AI and computer vision, barriers to effective communication in sign language are dismantled, empowering individuals to express themselves freely and connect with the world around them.

**KEYWORDS**: Sign LanguageRecognition1, Convolution Neural Network2, Image Processing3, Edge Detection4, Hand Gesture Recogniton5.

#### I. INTRODUCTION

In our daily lives, we encounter numerous individuals grappling with various health challenges, including conditions like deafness, muteness, and blindness. Communicating becomes a daunting task for them, often exacerbating feelings of isolation and frustration. While existing solutions predominantly rely on sensor-based technologies, they frequently fall short in providing a comprehensive remedy.

Enter a groundbreaking approach outlined in this paper, offering a novel method for virtual communication devoid of sensors. Leveraging the power of Convolutional Neural Networks (CNNs), this system revolutionizes the landscape by sidestepping the need for physical sensors. Instead, it harnesses the ubiquitous web camera to capture diverse gestures, serving as input data for the system. Upon capturing the gestures, the system, implemented in MATLAB, swiftly processes the images, employing sophisticated recognition algorithms to decipher the intricate hand movements. Subsequently, the identified gestures are seamlessly translated into text output, paving the way for effective communication. What sets this system apart is its bidirectional communication capabilities, facilitating seamless interaction between individuals with hearing and speech impairments and those without. Notably, the system adeptly converts sign language into both textual and vocal forms, ensuring inclusivity and accessibility for all parties involved. The voice output, generated through a voice replay kit, serves as a lifeline, bridging the communication chasm between diverse individuals. By enabling real-time interpretation of sign language gestures into comprehensible text and voice, this innovative solution heralds a new era of inclusive communication, empowering individuals of all abilities to connect, share, and engage without barriers.

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#### **II. LITERATURE SURVEY**

SN	<b>Topic Name</b>	Paper	Observations
1	Real time sign language recognition	Cristian Amaya and Victor Murray [1]	<ul> <li>Advantages:</li> <li>The algorithm shows correct predictions in more than 82% of evaluated images.</li> <li>The system was optimized for working with the five vowels showing results of a testing accuracy above 80% and an execution time of 59 milliseconds per frame.</li> <li>Disadvantages:</li> <li>It is applicable to only some letters</li> <li>The accuracy achieved is not up to 90%,</li> </ul>
2	Sign Language word translator using neural networks for the aurally impaired as a tool for communication	Balbin et al. [2]	<ul> <li>Advantages:</li> <li>The results show that the system can achieve 97.6% of recognition rate for 5 persons.</li> <li>Disadvantages:</li> <li>The system only recognized five Filipino words and used colored gloves for hand position recognition.</li> </ul>
3	RealtimeAmericansignlanguagerecognitionrecognitionusingwrist-wornmotionandsurfaceEMGsensors	Jian Wu [3]	<ul> <li>Advantages:</li> <li>Results show that after feature selection and conditioning, our system achieves 95.94% recognition rate.</li> <li>Disadvantages:</li> <li>The images are captured using electromyography which is costly, as it requires large-size datasets with diverse sign motion.</li> </ul>
4	Combining RGB and ToF cameras for real time 3D hand gesture interaction	Micha el Van den Bergh ETH Zurich et al. [4]	<ul> <li>Advantages:</li> <li>An improved hand detection algorithm is introduced based on adaptive skin color detection and depth. This approach significantly improves the interaction system, as the hands can overlap with the face, and other persons can be in the background.</li> <li>Disadvantages:</li> <li>Although this system gives good results, it only considers six classes of gestures.</li> </ul>

#### **III. OBJECTIVES**

The primary aim is to empower individuals with hearing impairments, providing them with the means to express their ideas and thoughts effectively. By leveraging cutting-edge technologies and tools, our goal is to develop a system that enables users to overcome the limitations imposed by their physical disability. Through this initiative, we aspire to enhance users' motivation and confidence, fostering a positive mindset and enabling them to overcome challenges with resilience. By addressing this global-level issue, we aim to promote inclusivity and create opportunities for individuals with hearing impairments to participate fully in society.

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#### **IV. METHODOLOGY**

CNNs, or Convolutional Neural Networks, represent a pinnacle in image and video recognition applications, serving as indispensable tools in tasks such as image recognition, object detection, and segmentation. The architecture of CNNs is characterized by its distinctive layers, each playing a vital role in the network's functionality:

- Convolutional Layer: Departing from the traditional neural network structure, CNNs exhibit localized connectivity, where only a subset of input neurons connects to neurons in the subsequent hidden layer. This unique arrangement facilitates efficient feature extraction and pattern recognition, enhancing the network's ability to discern intricate details within images.
- Pooling Layer: Integral to the CNN architecture, the pooling layer operates to reduce the dimensionality of feature maps generated by convolutional layers. Within the hidden layers of CNNs, multiple activation and pooling layers work in tandem to condense information while preserving essential features, contributing to hierarchical feature representation.
- Fully-Connected Layer: Situated towards the end of the network, fully connected layers receive input from the final convolutional or pooling layer. Here, the flattened feature vector undergoes complex nonlinear transformations, enabling the network to extract high-level features and make accurate predictions. The fully connected layer serves as a pivotal component in the classification process, facilitating the mapping of extracted features to corresponding output classes.

#### V. MATHEMATICAL MODEL

Let S represent the entire system, comprising: S = IP, Pro, OP

Where:

A. IP is the input to the system.

B. Pro is the procedure applied to the system to process the given input.

C. OP is the output of the system.

A. Input: IP =  $\{u, F\}$ 

Where:

- 1. u represents the user.
- 2. F denotes the set of files used for transmission.
- B. Procedure:
- 1. Process:
  - 1. Capture the image of the sign.
  - 2. Compare the captured image with the dataset.
  - 3. If a match is found:
  - Generate a voice alert message for the user.

#### C. Output:

- Upon sign detection, the system delivers a voice message alert to the user.

This mathematical model outlines the components and processes involved in the sign language interpretation system. It starts with user input, processes it through image comparison, and generates output in the form of voice alerts.

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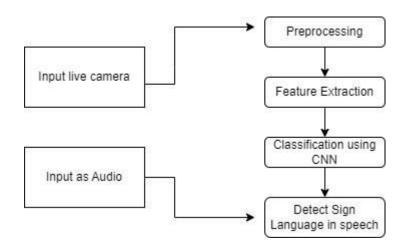


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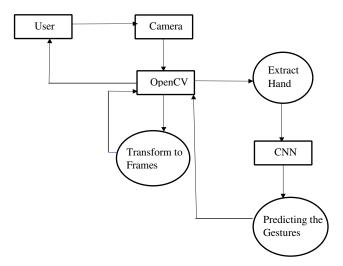
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VI. MODELING AND ANALYSIS

#### A] SYSTEM ARCHITECTURE



**B] DATAFLOW DIAGRAM** 



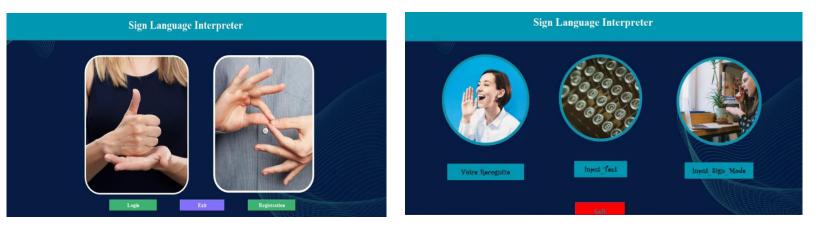
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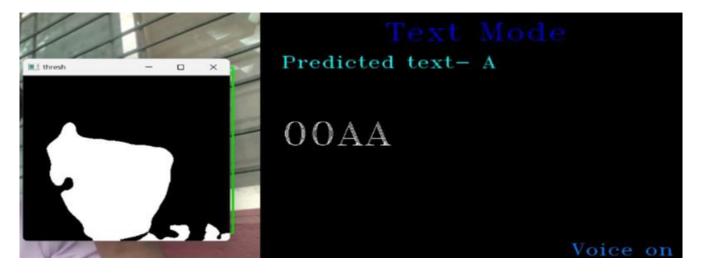
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VII. RESULT





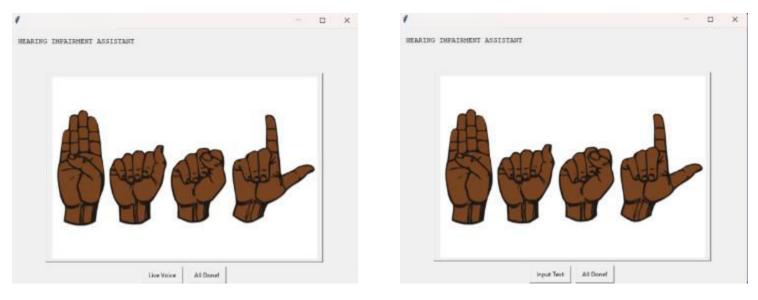


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#### VIII. CONCLUSION

The Sign Language Interpreter project is a groundbreaking endeavor with a mission to develop an intuitive, precise, and real-time system capable of recognizing sign language gestures and seamlessly translating them into either text or synthesized speech. This innovative system not only champions accessibility and inclusivity but also upholds ethical standards in technology utilization.

As technological advancements continue to shape our world, initiatives like the Sign Language Interpreter project showcase the immense potential of machine learning and Python programming to tackle real-world challenges. By enhancing the quality of life for individuals with hearing impairments, such projects contribute to creating a more inclusive and interconnected society.

In the ongoing journey of developing the Sign Language Interpreter, continuous collaboration, solicitation of user feedback, and a steadfast dedication to accessibility and ethical principles will be pivotal. These efforts will ensure that the project not only meets its objectives but also delivers a meaningful and positive impact on the lives of the deaf and hard of hearing community.

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