

A Review Paper on Sign Language Recognition System For Deaf And Dumb People using Image Processing

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Abstract—Communications between deaf-mute and a normal person have always been a challenging task. This paper reviews a different methods adopted to reduce barrier of communication by developing an assistive device for deaf-mute persons. The advancement in embedded systems, provides a space to design and develop a sign language translator system to assist the dumb people, there exist a number of assistant tools. The main objective is to develop a real time embedded device for physically challenged to aid their communication in effective means.

Keywords—Sign language identification, Hidden Markov Model, Artificial Neural Network, Data glove, Leap motion controller, Kinectic Sensor.

I. INTRODUCTION

Sign language system is a way of communication between deaf and dumb people. While communicating with dumb and deaf peoples, those who have knowledge of sign language, can talk and hear properly. But untrained people cannot communicate with dumb and deaf people, because the person can communicate to dumb people by training sign language. Sign language to text system will be more useful for such a impaired people for communicate with normal people more fluently.

Sign language is a physical action by using hands and eye with which we can communicate with dumb and deaf people. They can express their feeling with different hand shapes and movement. The task is to convert that shape or their sign language into text or speech.

Due to advancement in the field of image processing, an automatic sign language converter system is developed. Few researchers have developed tools to help to convert sign language into text or speech. Researchers in the field of sign language are broadly categorized in two ways, Data glove & Image processing. In data glove system, user needs to wear glove. Glove consists of flex sensor, accelerometer and motion tracker. Sensor output signals are sending to the

computer for processing and analyze the gesture and convert into text or speech.

In image processing, image is captured through web camera. Rest of this paper has organized as follows: the section 2 describes the technique for sign language into text conversion. a review on sign language converter system describes in section 2, section 3 describes the methods of sign language converter system. Finally conclusion of paper describes in section 4.

II. LITETATURE SURVEY

2.1 Sign Acquiring Methods:

A. Leap Motion

Leap Motion controller (figure 1) is a sensor which detects the hand movement and converts that signal into computer commands. It consists of two IR cameras and three infrared LED's. LED generates IR light signal and camera generates 300 frames per second of reflected data. These signals are sending to the computer through USB cable for further processing.



Figure 1: Leap motion controller with USB

P. Karthick et al. [1] used model that transform Indian sign language into text using leap model. The Leap device

detects the data like point, wave, reach, grab which is generated by a leap motion controller. Combination of DTW and IS algorithm is used for conversion of hand gesture into text. Neuron network was used for training the data.

Leigh Ellen Potter et al. [2] used leap motion controller for recognition of Australian sign language. Leap motion controller used to sense the hand movement and convert that hand movement into computer commands. Artificial neuron network is used for training symbols. The disadvantage of that system was low accuracy and fidelity.

B. Kinect Sensor

Kinect is Microsoft motion sensor with Xbox 360 gaming console shown in figure 2. It consists of RGB camera, depth sensor and multi-array microphone. It recognizes facial movement and speech.

Cao dong et al. [3] used Microsoft Kinect to recognize American sign language. Depth camera is Kinect sensor used to detect ASL alphabet. Distance adaptive scheme was used for feature extraction. Support vector machine and RF classifier algorithm used for classification purpose. Training of data was done using ANN network. The accuracy of the system was 90%. Yanyao et al. [4] used Kinect sensor for recognition of hand gesture. Firstly it detects hand movement and then matched with counter model. Second task was to locate multi colour glove and detect different colour regions. Gaussian colour model used for training data and per pixel classifier used for classification. This system has one drawback that is limited accuracy.



Figure 2: Kinect for Xbox 360

C. Data Glove

This method uses different sensors to detect hand gesture signal. Hand gesture signal is in the form of analog. ADC is used to convert analog signal into digital form. It consists of flex sensor and accelerometer. Flex sensor is used to detect bend signal [5]. Figure 3 shows the data glove.



Figure 3: Data glove with flex sensor

Anarabasi Rajamohan et al. [6] used data glove based method for recognition of American Sign Language. The

system consists of flex sensor, accelerometer and tactile sensor. This sensor is used to detect hand gesture and converted into code. Accuracy of that system was 90%.

D. Vision Based

In this method, a web camera is used to capture images. After that, image segmentation is done. Features like palm, finger extracted from input image. Different hand motions that are half closed, fully closed, semi closed were detected. Data is saved in a vector and that vector is used for recognition of alphabets [7].

Paulo Trigueiros et al. [8] used vision-based technique for recognition of Portuguese language. For their implementation, hand gesture was captured in real time. SVM algorithm is used for classification purpose. In this system, vowels were recognized with 99.4% accuracy and consonants with 99.6% accuracy.



Figure 4: Sample of vision based technique

Generally, while capturing the image for experiments, head movement is also mixed with hand images. To solve this overlap between hand and head movement, the camera is mounted above the signers [9]. But due to this, face and body gestures are lost. Nilsen et al. [10] used less hand gesture for fast recognition process.

III. METHODS FOR SIGN IDENTIFICATION SYSTEM

A. Artificial Neural Network

An artificial neuron is a computational model inspired by natural neurons. The advantage of ANN is its accuracy and generality. It has the ability to learn relationships from modeled data and at the same time to recognize the constraints [11]. In [12], Arabic sign language is converted into static hand gesture. To recognize that language, two recurrent neural networks are used, i.e., Partial recurrent network and fully recurrent network.

In this, input image was captured through a digital camera. Colored gloves were worn on hands. HIS model was used for segmentation process. After that, training and testing of images was done. The result of the fully recurrent network was better than the partially recurrent network.

A real-time 2D tracking system [13] is used for recognition of Myanmar alphabetic language. Tin Hanin implemented this system to recognize the hand gesture for MAL. Input image is digitized photographs and applied to Adobe Photoshop for recognizing edges of images. Histogram is used for feature extraction. For further processing, a neural network is used.

To recognize hand gesture for Japanese sign language, MLP neural network was used. Here, input was taken from data glove interface and fed to MLP neural network. Then data was trained and tested. Major drawback in this system is that data glove was unable to measure gesture direction. Shiga used this system for JSL [14].

Gonzalo et al. [15] implemented continuous time recurrent neural network real time hand gesture recognition system. Wireless mouse and tri axial accelerometer was used for capturing hand gesture. Genetic algorithm was used.

B. Hidden Markov Model

Liang et al. [16] implemented two HMM models for a continuous system for the Taiwanese sign language using a data glove. It consists of grammar and semantics for matching sentences. The main aim of this model is to provide estimates of probability of a sequence of movements. Due to that it increases the recognition rate.

British sign language recognition by using markov chain in combination with independent component analysis [17], data was captured through image technique. Feature extraction was used to extract motion and shape of hands.

Tani bata et al. [18] proposed HMM for isolated for JSL recognition system. Baum–Welch algorithm used to model parallel left and right hand data. Viterbi algorithm was used for verification purpose.

The multilayer architecture in sign language recognition for the signer independent CSL recognition, in which combination of DTW and HMM are used. To solved the confusion set in vocabulary space DTW/ISODATA algorithms are used [19]. The recognition accuracy was greater than the HMM based recognition system.

Volger [20] Proposed system for recognition of American sign language using parallel hidden markov model. In this system only phonemes were used for continuous recognition system. Two channels are used from which one channel is for left hand and other for right hand. Word is divided into fundamental phonemes as same word used in speech recognition. The accuracy of that model was high.

IV. CONCLUSION

In this review paper, different techniques of sign language recognition are reviewed on the basis of sign acquiring methods and sign identification methods. For sign acquiring methods, vision based methods and for sign identification methods, artificial neuron network proves a strong candidature.

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