# ISSN (Online) 2278-1021



## International Journal of Advanced Research in Computer and Communication Engineering

Vol. 9, Issue 3, March 2020

# Sign Language Converter & Recognition

## Dhnashree Bhandarkar<sup>1</sup>, Amit Kale<sup>2</sup>, Prajakta Mukhekar<sup>3</sup>, Prof.Rohini Nere<sup>4</sup>

Student, Department of Computer Engineering, Trinity College of Engineering and Research, Pune, India<sup>1, 2, 3</sup>

Assistant Professor, Department of Computer Engineering, Trinity College of Engineering and Research, Pune, India<sup>4</sup>

Abstract: Sign language is the only way the speech and hearing impaired (i.e dumb and deaf) people can communicate. The main difficulty of this way of communication is normal people who cannot understand sign language can't communicate with these people or vice versa. Our project aims to bridge the gap between the speech and hearing-impaired people and the normal people. An individual's ability to communicate using speech and hearing is affected by Speech impairment disability. Sign language is other media of communication for people who are affected. The essential idea of this project is to make a system using which dumb people can importantly interact with all other people using their normal gestures. The system does not need the background to be perfectly black. It works on any background. The project requires image processing system to identify, mainly English alphabetic sign language used by the deaf people to communicate and converts them into text so that normal people can understand efficiently. To build a vision-based application which offers sign language translation to text thus aiding communication between signers and non- signers is the main objective of project. The proposed model takes video sequences of signers and extracts temporal and spatial features from them. We then use a CNN (Convolutional Neural Network) for recognizing spatial features.

**Keywords**: Machine learning, Computer vision, Sign language, Convolutional Neural Networks (CNN), Sign recognition.

#### I. INTRODUCTION

American Sign Language (ASL) substantially facilitates communication in the deaf community. However, there are only ~250,000-500,000 speakers which significantly limits the number of people that they can easily communicate with the alternative of written communication is cumbersome, impersonal and even impractical when an emergency occurs. In ASL recognition system that uses Convolution Neural Networks (CNN) in real time to translate a video of a user's ASL signs into text which reduce the obstacle of communication and permit dynamic communication. Inability to speak is measured to be true disability. People with this disability use diverse modes to communicate with others. There are number of methods presented for their communication one such common method of communication is sign language. Sign language application is useful for deaf people, as they will be able to communicate easily with even those who don't understand sign language

#### II. OBJECTIVE

- To interface camera to system for image capture
- To do CNN (Machine learning) based image classification to recognize sign in image
- To convert meaning text of detected sign into speech

To form communication simplex between deaf and dumb people by introducing Computer in communication path so that sign language can be automatically captured, acknowledged, translated to text.

Our software is designed especially for people who are mute but all other users can use it. The main purpose is to allow the user to use sign language in front of its Intel camera and then the application is recognizing the gesture. Every move acknowledged is translate and send to the different user Secondly, for people who are blind or partially-sighted, once they received the translation of the movement, the matter is also read. Then we need to allow more than two users to speak jointly. Our endmost purpose is to allow the self-learning on our application of new signs.

The camera is used to record the hand gestures and then the videos are preprocessed to frame sequences which are fed individually to the CNN for two possible outputs. The global pool layer gives us a 2048 sized vector, which possibly allows for more features to be analyzed by the CNN.

# **IJARCCE**



#### International Journal of Advanced Research in Computer and Communication Engineering

Vol. 9, Issue 3, March 2020

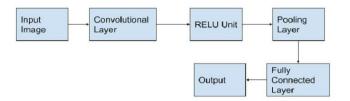


Fig.1 CNN Architecture

Specialized, deep learning CNN model to training and testing, every input picture will pass it through a order of convolution layers with filters (Kernals), Pooling, fully connected layers (FC) and apply Softmax function to sort an object with probabilistic values between 0 and 1. The above figure is a complete flow of CNN to process an input sign image and classifies the objects based on values.

#### CNN Algorithm:

- 1. Give an input image into convolution layer.
- 2. select parameters, apply filters with strides, padding if required. Do convolution on the image and apply RELU activation to the matrix.
- 3. Carry out pooling to reduce dimensionality size
- 4. Add as many convolutional layers until satisfied
- 5. Flatten the result and feed into a fully connected layer
- 6. Result of the class using an activation function (Logistic Regression with cost functions) and classifies images.

There are 3 major constraints for the project. They are as follows:

1) Image input: This module is used to take image from external and provide that image to image processing unit. The image processing unit consist basic segmentation and morphological operation to improve image features.



Fig.2 Sample image of 'signs'

- 2) Machine learning(CNN)image classification: This module is used to detect and classify image of sign.CNN machine learning algorithm is used to detection and classification of image.
- 3) Text to speech: This module is used to convert classified sign into text .The resultant text will convert into speech. Sign language is a type of language that uses manual communication to convey meaningful messages to other people. American Sign Language is one of the popular sign language used by most of deaf and dumb people to communicate with each other. American Sign Language is also referred to as ASL. A real-time sign language translator is required for facilitating communication between the deaf community and the general public.

We propose a system called Dynamic tool for American Sign Language (ASL) finger spelling interpreter which can consistently classify the letters a-z. The dataset consists of a set of American Sign Language videos. Our approach first converts the videos into frames and then preprocesses the frames to convert them into grayscale images. Then the Convolution Neural Network (CNN) classifier is used for building the classification model which classifies the frames

## **IJARCCE**



## International Journal of Advanced Research in Computer and Communication Engineering

Vol. 9, Issue 3, March 2020

into 26 different classes representing 26 English alphabets. Finally, the evaluation of the classification model is carried out with test data providing the output in the form of text or voice.

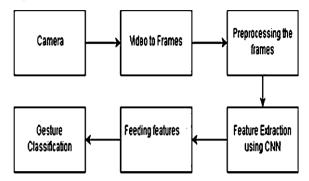


Fig.3 Architecture Diagram

The CNN model extracted temporal features from the frames which were used further to predict gestures based on sequence of frames. The CNN model used was Inception [6], which was a model developed by Google for image recognition as is widely regarded as the most image recognition neural network which exists right now.

We use two different approaches to classification:

- A. Using the predictions from the Softax layer and
- B. Using the output of the global pool layer.

In neural network, Convolutional neural network (ConvNets or CNNs) is one of the main categories to do images identification, images categorize. Objects discovery, recognition faces etc., are some of the areas where CNNs are widely used. CNN image classifications take an input image, process it and sort out it under certain category (Eg. Hand). Computer sees an input image as array of pixels and it depends on the picture resolution. Based on the picture resolution, it will see  $h \times w \times d(h = Height, w = Width, d = Dimension)$ . Eg., An picture of  $6 \times 6 \times 3$  array of matrix of RGB (3 refers to RGB values) and an picture of  $4 \times 4 \times 1$  array of matrix of grayscale picture.

## III. CONCLUSION

This project deals with the application of Convolution Neural Network for recognizing the hand gestures. One of the essential applications of hand gesture recognition is to identify the sign language which is a lively tool of communication for physically impaired, deaf and dumb people. This application will help to bridge the gap between normal and deaf/dumb people. From the outcome obtained above we can conclude that Convolution Neural Network provides a remarkable accuracy in identifying the sign language characters including alphabets and numerals. This piece of work can be further extended to building a real time application which can determine the sign language and including words, sentences to recognize instead of just characters or single word.

#### REFERENCES

- [1]. A new approach of sign language recognition system for bilingual users S. M. Kamrul Hasan , Mohiuddin Ahmad
- [2]. Bangla Sign Language Detection Using SIFT and CNN Shirin Sultana Shanta; Saif Taifur Anwar; Md. Rayhanul Kabir.
- [3]. Deep Convolutional Neural Networks for Sign Language RecognitionG. Anantha Rao1, K. Syamala, P.V. V. Kishore, A.S. C.S. Sastry.
- [4]. Two-handed hand gesture recognition for Bangla sign language using LDA and ANN, Rahat Yasir; Riasat Azim Khan, Rahat Yasir; Riasat Azim Khan
- [5]. A Novel Feature Extraction for American Sign Language RecognitionUsing Web- cam Ariya Thongtawee,Onamon Pinsanoh,YuttanaKitjaidure
- [6]. Sign language learning based on Android for deaf and speech impairedpeople.Setiawardhana,Rizky Yuniar Hakkun,Achmad Baharuddin
- [7]. The Comparison of Some Hidden Markov Models for Sign LanguageRecognition Suharjito; Narada Thiracitta; Herman Gunawan; GunawanWitjaksono
- [8]. Real time sign language recognition using the leap motion controller. Deepali Naglot; Milind Kulkarni
- [9]. An Efficient Sign Language Translator Device Using Convolutional Neural Net-work and Customized ROI Segmentation. Saleh Ahmad Khan, Amit Debnath Joy; S. M. Asaduzzaman; MorsalinHossain
- [10]. Real Time Bangladeshi Sign Language Detection using Faster R-CNN, Oishee Bintey Hoque; Mohammad Imrul Jubair; Md. Saiful Islam; Al-Farabi Akash; Alvin Sachie Paulso
- [11]. Kshitij Bantupalli, Ying Xie "American Sign Language recognition using deep learning and computer version" IEEE 2018.
- [12]. Umang Patel, Aarti G Ambekar "Moment Based Sign Language Recognition for Indian Languages" IEEE 2018.
- [13]. Mohamed Hassan, Khaled Assalhen "User- Dependent Sign Language Recognition Using Motion Detection" IEEE 2017.
- [14]. An Effective Sign Language Learning with Object Detection BasedROI Segmen-tation. Sunmok Kim; Yangho Ji; Ki-BaekLee.