

Convolutional Neural Network Based Bidirectional Sign Language Translation System

Anil Sonawane¹, Vinit Shahane², Onkar Devkar³, Prof. Ghadage R. A⁴

Students, Department of Computer Engineering^{1, 2, 3}

Assistant Professor, Department of Computer Engineering⁴

Vishwabharati Academy's College of Engineering, Ahmednagar, India

Abstract: *This paper focuses on experimenting with different segmentation approaches and unsupervised learning algorithms to create an accurate sign language recognition model. To more easily approach the problem and obtain reasonable results, we experimented with just up to 10 different classes/letters in our self-made dataset instead of all 26 possible letters. We collected 12000 RGB images and their corresponding depth data using a Microsoft Kinect. Up to half of the data was fed into the auto encoder to extract features while the other half was used for testing. We achieved a classification accuracy of 98% on a selected set of test data using our trained model. In addition to the work, we did on static images, we also created a live demo version of the project which can be run at a little less than 2 seconds per frame to classify signed hand gestures from any person. The problem we are investigating is sign language recognition through unsupervised feature learning. Being able to recognize sign language is an interesting computer vision problem while simultaneously being extremely useful for deaf people to interact with people who don't know how to understand American Sign Language (ASL). Hand gesture is one of the methods used in sign language for non-verbal communication. Various sign language systems have been developed by many makers around the world but they are neither flexible nor cost-effective for the end users.*

Keywords: Image processing, Noise removal, Feature extraction and matching, Static and dynamic gesture

REFERENCES

- [1]. Lance Fernandes, Prathamesh Dalvi, Manisha Bansode. Convolutional Neural Network based Bidirectional Sign Language Translation System. Proceedings of the Third International Conference on Smart Systems and Inventive Technology (ICSSIT 2020) IEEE Xplore Part Number: CFP20P17-ART; ISBN: 978-1-7281-5821-1. IEEE, 2020.
- [2]. P. Y. Simard. Best practices for convolutional neural networks applied to visual document analysis, August 2003. Seventh International Conference on Document Analysis and Recognition.
- [3]. X. Teng. A hand gesture recognition system based on local linear embedding, April 2005. Journal of Visual Languages and Computing.
- [4]. M. Ranzato. Efficient learning of sparse representations with an energy-based model, 2006. Courant Institute of Mathematical Sciences.
- [5]. D. Metaxas. Sign language and human activity recognition, June 2011. CVPR Workshop on Gesture Recognition.
- [6]. S. Sarkar. Segmentation-robust representations, matching, and modeling for sign language recognition, June 2011. CVPR Workshop on Gesture Recognition, Co-authors: Barbara Loeding, Ruiduo Yang, Sunita Nayak, Ayush Parashar.
- [7]. P Vijayalakshmi and M Aarthi. Sign language to speech conversion. In 2016 International Conference on Recent Trends in Information Technology (ICRTIT), pages 1–6. IEEE, 2016.
- [8]. Teeperson. Sign language mnist. [Online]. Available: <https://www.kaggle.com/datamunge/sign-language-mnist>, 2017.
- [9]. Shaheer Bin Rizwan, Muhammad Saad Zahid Khan, and Muhammad Imran. American sign language translation via smart wearable glove technology. In 2019 International Symposium on Recent Advances in



IJARSCT

Impact Factor: **6.252**

IJARSCT

ISSN (Online) 2581-9429

International Journal of Advanced Research in Science, Communication and Technology (IJARSCT)

Volume 2, Issue 2, July 2022

Electrical Engineering (RAEE), volume 4, pages 1–6. IEEE, 2019.