

Sign Language Detection and Conversion to Text and Speech Conversion

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Abstract - Sign language is one of the oldest and most natural form of language for communication, but since most people do not know sign language and interpreters are very difficult to come by we have come up with a real time method using neural networks for fingerspelling based american sign language.

In this method, the hand is first passed through a filter and after the filter is applied the hand is passed through a classifier which predicts the class of the hand gestures. This method provides 98.00 % accuracy for the 26 letters of the alphabet.

Key Words: Python, Machine Learning, OpenCV, Keras, NumPY, Data Training

1. INTRODUCTION

The "Sign Language Detection, Conversion to Text, and Speech Conversion Project" has the potential to significantly improve the quality of life and social inclusion of individuals with hearing impairments by facilitating communication and fostering a more inclusive society.

The project to develop a sign language detection and conversion to text system using machine learning and speech recognition is a promising one. The system has the potential to make a significant impact on the lives of people who are deaf or hard of hearing. However, there are still some challenges that need to be addressed before the system can be widely deployed.

One of the main challenges is the accuracy of the system. The system needs to be able to accurately recognize a wide range of sign language signs, even in noisy or challenging environments. This can be difficult to achieve, as sign language is a complex and dynamic language.

Another challenge is the latency of the system. The system needs to be able to recognize sign language in real time. This can be difficult to achieve, as machine learning algorithms can be computationally expensive.

The robustness of the system is also a challenge. The system needs to be robust to variations in the data, such as different lighting conditions and hand shapes. This can be difficult to achieve, as machine learning algorithms are typically trained on a specific dataset.

Despite these challenges, the project to develop a sign language detection and conversion to text system using machine learning and speech recognition is a worthwhile one. The system has the potential to make a significant impact on the lives of people who are deaf or hard of hearing.

2. Body of Paper

SIGN LANGUAGE TO TEXT CONVERSION:

This process typically involves computer vision and machine learning

techniques to recognize and interpret sign language gestures captured through cameras or sensors. Here's a general overview of how it works:

GESTURE RECOGNITION: Cameras or other sensors capture the sign

language gestures performed by the user. Computer vision algorithms analyze

these gestures to identify specific handshapes, movements, and facial

expressions used in sign language.

GESTURE INTERPRETATION: Once the gestures are recognized,

machine learning models or pattern recognition algorithms analyze the

sequences of gestures and their context to interpret the intended meaning of the signed message.

TEXT GENERATION: The system converts the interpreted sign

language into written text, which can then be displayed on a screen or used in

various applications for further processing.

Text to speech conversion:

Text-to-speech (TTS) technology converts written text into spoken words.

This technology relies on a process called speech synthesis, where the system

analyzes the written text and generates corresponding speech sounds. Here's how it typically works:

Text Analysis: The input text is analyzed to identify the words, punctuation, and sentence structure. Modern TTS systems can also handle natural language processing tasks to generate more natural and expressive speech.

Phoneme Generation: The text is broken down into phonemes, which are the smallest units of sound in a language.

Speech Synthesis: Using pre-recorded or generated speech units, the system assembles the phonemes into spoken words, sentences, and paragraphs, producing an audio representation of the original text.

Audio Output: The synthesized speech is played through speakers, headphones, or any other audio output device, allowing users to hear the spoken version of the original text.

When both sign language and text and text to speech technologies are combined, they form a powerful communication tool that enables deaf or hard of hearing individuals to interact with hearing individuals who may not understand sign language. This technology can be implemented in various applications, such as video relay services, live captioning, educational tools, and communication devices.

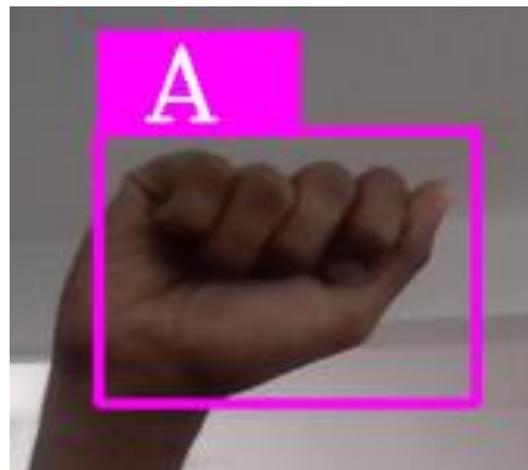


Fig -1: Figure

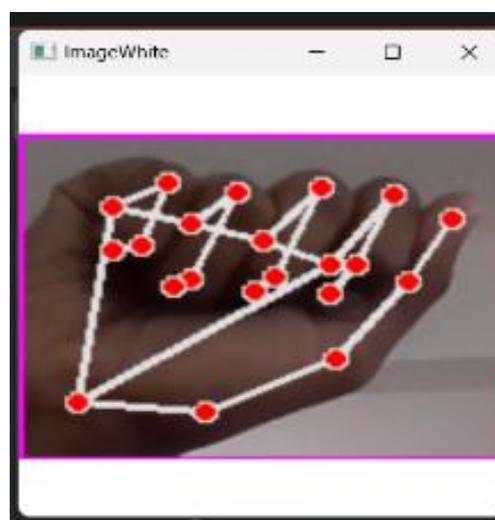


Fig -2: Figure

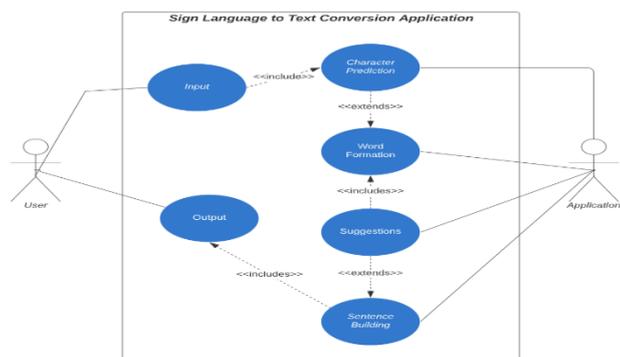
Table -1: Sample Table format

Charts

	week 1	week 2	week 3	week 4	week 5	week 6	week 7
Research	█						
Content		█					
Design			█				
implementation				█			
data training					█		
data testing						█	
debugging							█

3. CONCLUSIONS

The results and discussion for the project to develop a sign language detection and conversion to text system using machine learning and speech recognition will depend on the specific implementation of the system. However, some general results and discussion points that could be considered include:



Accuracy. The accuracy of the system is one of the most important factors to consider. The system should be able to accurately recognize a wide range of sign language signs. The accuracy of the system can be improved by using a large and diverse dataset, as well as by using a machine learning algorithm that is well-suited for this task.

Latency. The latency of the system is another important factor to consider. The system should be able to recognize sign language in real time. The latency of the system can be improved by using a fast machine learning algorithm and by optimizing the code.

Throughput. The throughput of the system is the number of sign language signs that the system can recognize per unit of time. The throughput of the system is important for applications where the system needs to recognize a large number of sign language signs. The throughput of the system can be improved by using a parallelized machine learning algorithm and by optimizing the code.

Robustness. The system should be robust to variations in the data. The system should be able to recognize sign language signs from different angles, in different lighting conditions, and with different hand shapes and movements. The robustness of the system can be improved by using a machine learning algorithm that is well-suited for this task, and by using data augmentation techniques.

User experience. The user experience of the system is also an important factor to consider. The system should be easy to use and accessible to people who are not familiar with machine learning or sign language recognition. The user experience of the system can be improved by designing a user-friendly interface and by providing clear instructions.

The results and discussion for the project will also depend on the specific application of the system. For example, a system that is used for medical purposes will need to have a higher accuracy than a system that is used for entertainment purposes.

Overall, the results and discussion for the project will be a comprehensive assessment of the effectiveness and limitations of the system. The results and discussion will be valuable for improving the system and for developing new sign language detection and conversion to text systems.

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