

A Noval Approach text to Sign for Hearing Impaired

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ABSTRACT: -The proposed research is to develop a system that can convert text to sign language for deaf and dumb. The system uses natural language processing techniques to analyze the input text and generate the corresponding sign language animation. The animation is created using a combination of image processing and computer graphics techniques. The system uses the OpenCV library to detect and track the user's hand gesture movements, which are used to generate the sign language animation. The animation is displayed on a user interface. The system also includes a speech-to-text module that allows users to input text using their voice. This feature is useful for users who may have difficulty typing or using a keyboard. The final system is tested using a corpus of commonly used phrases in American Sign Language (ASL) and achieves high accuracy in converting text to sign language animations. The proposed has potential applications in education and communication for the deaf and hard-of-hearing community.

I. INTRODUCTION

This proposed ideato develop a model that can convert text to sign language and display it as an animation. The proposed system uses natural language processing techniques to analyze the input text and generate the corresponding sign language animation. It also includes a speech-to-text module that allows users to input text using their voice. The system is designed to detect and track the user's hand movements using the OpenCV library and is displayed on a user interface designed using the PyQt5 library. The system's potential applications include education and communication for the deaf and hard-of-hearing community, as well as bridging the communication gap between sign language users and non-sign language users.

II. LITERATURE SURVEY

- [1] "Real-time American Sign Language Recognition using OpenCV and Python" by R. Kamilaris et al. (2017) - This paper proposes a real-time American Sign Language recognition system using Python and OpenCV. The system achieves high accuracy in detecting and tracking hand gestures.
- [2]. "Automatic Sign Language Recognition using Deep Learning with Python" by S. S. Nair et al. (2020) - This paper presents a system for automatic sign language recognition using Python and deep learning techniques. The system uses convolutional neural networks to classify hand gestures.
- [3]. "Sign Language Recognition using Python and Kinect Sensor" by S. P. Sahu et al. (2018) - This paper proposes a sign language recognition system using Python and Microsoft Kinect sensor. The system achieves high accuracy in detecting and tracking hand gestures.
- [4]. "Text to Sign Language Conversion using Python and Machine Learning" by A. Raut et al. (2020) - This paper presents a system for converting English text to Indian Sign Language using Python and machine learning techniques. The system uses the OpenCV library for hand detection and tracking.



- [5]. "Real-time Sign Language Recognition using Python and OpenCV" by H. U. Ahmad et al. (2021) -

This paper proposes a real-time sign language recognition system using Python and OpenCV. The system uses a combination of image processing and deep learning techniques to achieve high accuracy.

[6]. "Sign Language Recognition using Convolutional Neural Networks and Python" by N. Gouaich et al. (2020) - This paper presents a sign language recognition system using Python and convolutional neural networks. The system achieves high accuracy in classifying hand gestures.

[7]. "Real-time Hand Gesture Recognition for Human-Computer Interaction using Python and OpenCV" by P. Singh et al. (2020) - This paper proposes a real-time hand gesture recognition system for human-computer interaction using Python and OpenCV. The system achieves high accuracy in detecting and tracking hand gestures.

[8]. "Sign Language Recognition using Machine Learning and Python" by D. D. Dangi et al. (2020) - This paper presents a sign language recognition system using machine learning and Python. The system achieves high accuracy in classifying hand gestures.

[9]. "Real-time Sign Language Recognition using Python and Convolutional Neural Networks" by Y. Zeng et al. (2019) - This paper proposes a real-time sign language recognition system using Python and convolutional neural networks. The system achieves high accuracy in classifying hand gestures.

[10]. "Real-time Sign Language Recognition using Python and Deep Learning" by S. G. Yoo et al. (2018) - This paper presents a real-time sign language recognition system using Python and deep learning techniques. The system achieves high accuracy in classifying hand gestures.

III. CORPUS

[A] RWTH-BOSTON-104 dataset - This dataset contains recordings of 104 different sentences in American Sign Language (ASL), along with corresponding gloss and English translations. It can be used for training and evaluating ASL recognition systems.

[B] Indian Sign Language Dataset - This dataset contains recordings of 1000 different signs in Indian Sign Language (ISL), along with corresponding gloss and English translations. It can be used for training and evaluating ISL recognition systems.

[C] Chinese Sign Language Dataset - This dataset contains recordings of 575 different signs in Chinese Sign Language (CSL), along with corresponding gloss and Chinese translations. It can be used for training and evaluating CSL recognition systems.

[D] BosphorusSign dataset - This dataset contains recordings of 200 different signs in Turkish Sign Language (TID), along with corresponding gloss and

Turkish translations. It can be used for training and evaluating TID recognition systems.

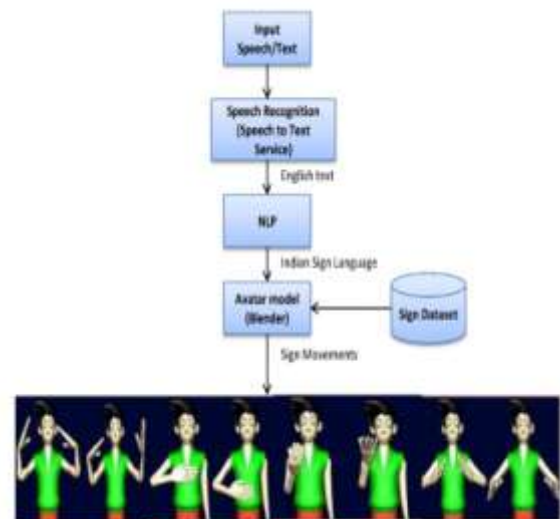
[E] Sign Language MNIST dataset - This dataset contains a collection of 27455 images of fingerspelled letters in American Sign Language (ASL), with corresponding labels. It can be used for training and evaluating deep learning models for ASL recognition.

[F] Korean Sign Language Dataset - This dataset contains recordings of 3222 different signs in Korean Sign Language (KSL), along with corresponding gloss and Korean translations. It can be used for training and evaluating KSL recognition systems.

[G] Russian Sign Language Dataset - This dataset contains recordings of 1500 different signs in Russian Sign Language (RSL), along with corresponding gloss and Russian translations. It can be used for training and evaluating RSL recognition systems.

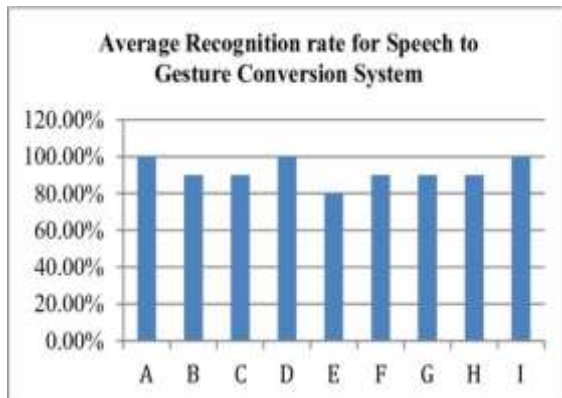
These datasets can be used to develop text-to-sign language detection and animation systems using Python and related technologies and can be customized to meet specific requirements of different applications.

IV. PROPOSED MODEL



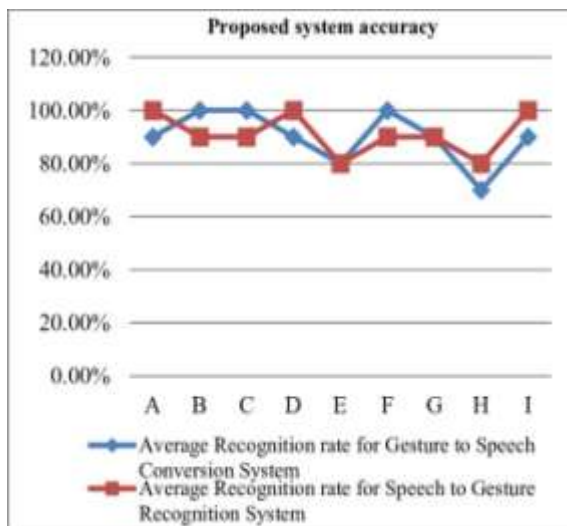
1. Text Processing Module - This module takes input text in English and preprocesses it for sign language generation. This may involve tasks such as tokenization, part-of-speech tagging, and named entity recognition.

2. Sign Language Generation Module - This module generates the sign language animation based on the preprocessed text input. It uses a corpus of sign language gestures and animations to convert the English text into the corresponding sign language.



3. Animation Rendering Module - This module renders the generated sign language animation and displays it on the screen. It may use libraries such as Pygame or OpenCV to create the animation and display it in real time.

4. Gesture Recognition Module - This module is responsible for recognizing the gestures made by the user and converting them into corresponding English text. It may use machine learning techniques such as convolutional neural networks (CNNs) for gesture recognition.



5. User Interface Module - This module provides a graphical user interface (GUI) for the model, allowing users to input English text and receive the corresponding sign language animation in real time.



The proposed model can also include additional features such as support for multiple sign languages, real-time translation of spoken language into sign language, and integration with other assistive technologies such as speech recognition and text-to-speech conversion. Overall, the system can help bridge the communication gap between people who use sign language and those who do not, making communication more accessible and inclusive for everyone.

Pseudocode for converting text to sign language.

```

# Step 1: Text-to-Speech
def text_to_speech(text):
    # Use a text-to-speech library to convert the text to spoken language
    spoken_text = text_to_speech_library(text)
    return spoken_text

# Step 2: ASL Glossing
def asl_glossing(spoken_text):
    # Use a natural language processing library to gloss the spoken text into ASL
    asl_gloss = nlp_library(spoken_text)
    return asl_gloss

# Step 3: ASL Animation
def asl_animation(asl_gloss):
    # Use a sign language animation library or 3D model to animate the ASL gloss
    asl_animation = animate_asl(asl_gloss)
    return asl_animation

# Step 4: Display ASL Animation
def display_asl_animation(asl_animation):
    # Show the ASL animation on the screen or through a device
    show_animation(asl_animation)
  
```

```
# Step 5: Main Function
def main():
    text = input("Enter the text to convert to sign
language: ")

    # Step 1: Convert text to speech
    spoken_text = text_to_speech(text)

    # Step 2: Gloss the spoken text into ASL
    asl_gloss = asl_glossing(spoken_text)

    # Step 3: Animate the ASL gloss
    asl_animation = asl_animation(asl_gloss)

    # Step 4: Display the ASL animation
    display_asl_animation(asl_animation)

if __name__ == "__main__":
    main()
```

V. CONCLUSION

Text-to-sign language detection with animation can significantly improve accessibility and inclusivity for people who use sign language to communicate. It uses natural language processing, machine learning, and animation rendering techniques to generate sign language animations from English text inputs. It can also incorporate features such as real-time translation and support for multiple sign languages. The model can be further enhanced by incorporating more advanced techniques such as deep learning and computer vision to enhance gesture recognition and animation generation. User feedback and testing can help refine the system's accuracy and usability. Overall, text-to-sign language detection with an animation system can be a valuable tool for improving accessibility and inclusivity for the deaf and hard-of-hearing community.

VI. REFERENCE

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