Automatic Speech to American Sign Language Video Generation

Team 4

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Introduction & Motivation

- Sign language is essential for communicating with the hard-of-hearing or hearing-impaired, who make up 5% of the global population.
- While significant strides have been made in language recognition (SLR), sign language production (SLP) lags behind.
- **Bridge communication gap**: A system that translates spoken language to American sign language.
- Proposed System: An approach to SLP using Transformer and GAN models to produce realistic sign language videos from speech input.



Source: http://bu.edu/av/asllrp/dai-asllvd.html

System Requirements

Software Requirements:

- Language: Python, Javascript, Html
- Libraries: PyTorch, Tensorflow, DWPose, ngrok
- Framework : Flask
- Google Speech to Text API
- Google Colab
- AWS Sagemaker, Lambda Functions,
 API Gateway
- Storage tools: AWS S3 Bucket, Google drive

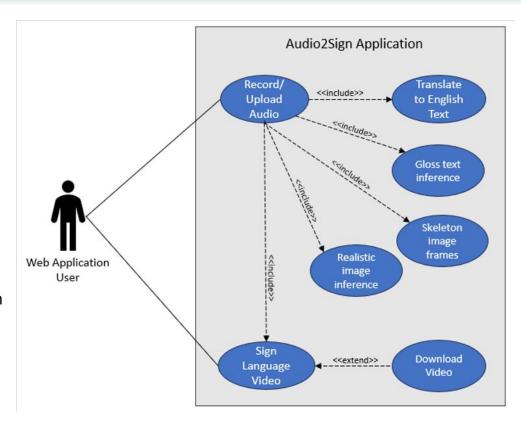
Hardware Requirements:

- Graphics Processing Unit (GPU)
- Memory (RAM)
- Microphone

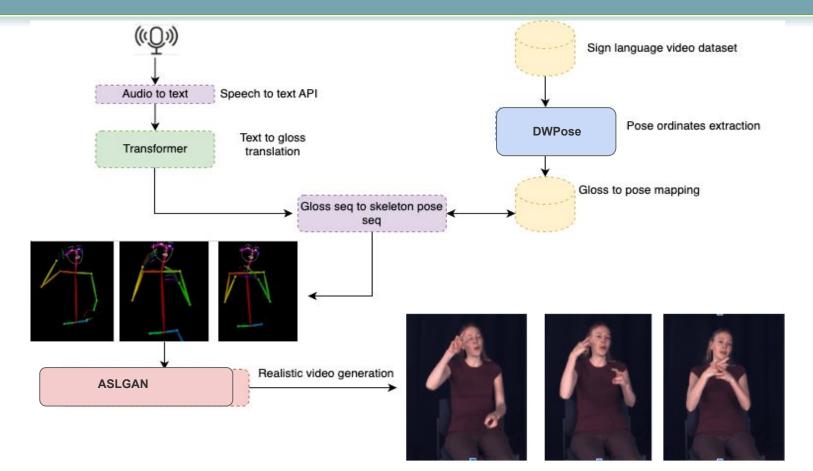
System Requirements

Functional Requirements:

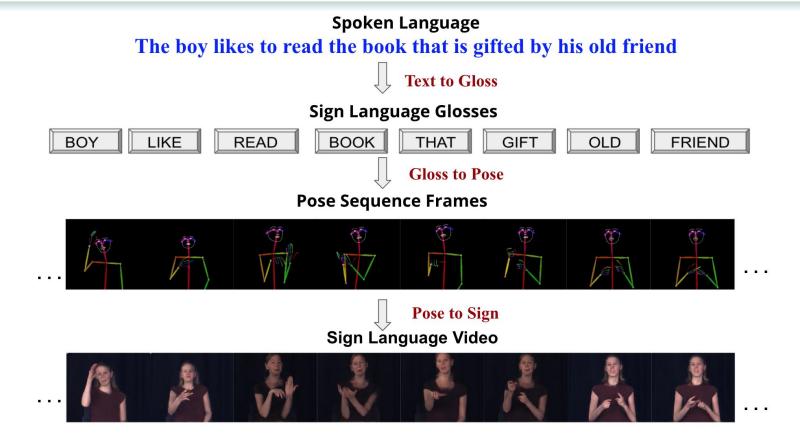
- Microphone Accessibility
- Upload File Accessibility
- Generate recorded Text
- Submit query
- Generate Sign lang Video
- Compatibility with operating systems such as Windows, macOS, or Linux



Pipeline Architecture



End to End Data Conversion



Datasets

ASLG-PC12 - text to gloss dataset

- A large parallel corpus of English written texts and American Sign Language glosses
- More than 80,000 pairs of sentences
- Open access for research
- Latest release in 2012

ASLLVD - gloss to sign dataset

- American Sign Language Lexicon Video Dataset
- Created by capturing videos of six signers with four synchronized cameras
- Videos represented for 9800 gloss tokens in more than 3300 video clips
- Video metadata file to map from gloss to video clip using scene id and sessionid along with start and end frame number of the video clip
- Videos accessible in 3 camera angles each of the resolution 640 x 480
 and 60 fps
- Open to access for research

Data Preprocessing - Text2Gloss

ASLG-PC12

Text-to-Gloss Dataset

ASLG-PC12 Dataset

Text sentence and gloss sentence mapping

Data preprocessing steps

- · Removing Email and URL Addresses
- Lowercasing
- Removing Punctuation and Unique Symbols except (a-z, A-Z, ".", "?", "!", ",")
- Tokenization
- Lemmatization
- Adding a start and an end token to the sentence

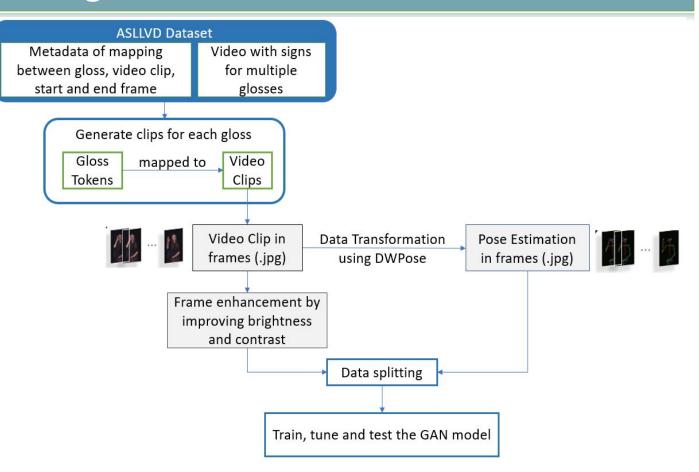
Data splitting

Train, tune, and test the Transformer model

Data Preprocessing - Gloss2Pose

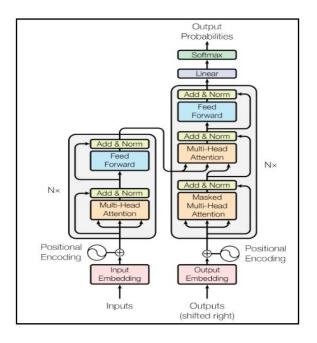
<u>ASLLVD</u>

Gloss-to-Pose Dataset

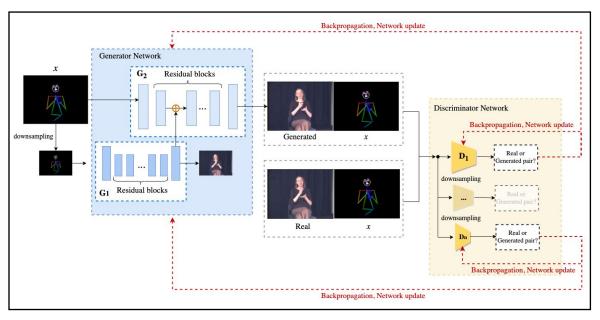


Models

Transformer



ASLGAN



https://www.tensorflw.org/text/tutorials/transforer

Model Evaluation

Model	Dataset	Evaluation
Transformer	ASLG_PC12	61.23 BLEU
ASLGAN + 5 Discriminators	ASLLVD	0.931 SSIM





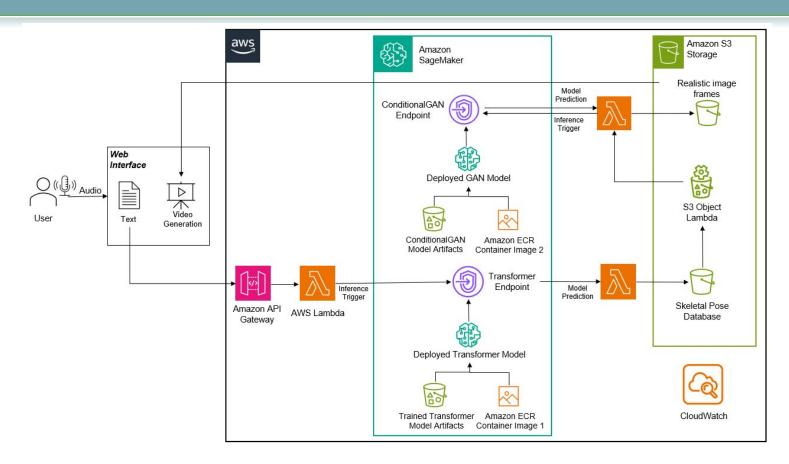
Real Image

Generated Image

System Design - Web UI Design



System Deployment Design



Achievements

Model Achievements:

- Developed a Transformer model from scratch that achieved state-of-the-art results translation results.
- Successfully generated high-quality videos using our proposed system.
- Pioneered the usage of CycleGAN for image generation from poses, reducing the need for extensive paired datasets.
- Integration of DWPose for pose extraction improved skeleton pose quality and inference speed with ASLGAN.

Sign Language Production Achievements:

- Project contributes an easy-to-understand yet foundational approach to sign language video production domain.
- Enhanced the accuracy and efficiency of the deep learning models that drive support to the hard of hearing community.

Constraints

Model Development Constraints

- Model's training was limited to a single signer.
- Hand keypoints need further enhancement.

Dataset Constraints

- Limited vocabulary in available datasets.
- Lack of high-resolution videos.
- Alternative datasets not feasible due to lack of gloss annotations.

System Constraints

- Deep Learning models require high GPU power for training and inference.
- Not possible to deploy SignSync on a single Amazon SageMaker endpoint due to differences in model frameworks (TensorFlow vs PyTorch).

Lessons Learned

Academic Research

- Systematic progress is the only solution to a complex problem.
- Keeping up with a progress made in GenAI (or any fast evolving domain) is essential to publish a promising paper.

Development and Deployment

- Preplanned continuous training and deployment on cloud service.
- Learn to estimate the resource allocation when dealing with high computational deep learning models.

Potential System and Model Applications

On-the-Go Communication

- Practical application for individuals with hearing difficulties in daily life.
- Supports communication needs in various settings, both indoors and outdoors.

Educational Integration

- Seamless integration into learning settings for real-time sign language interpretation.
- Enhances accessibility in online tutorials and courses.

Public Areas and Accessibility

- Integration into public spaces like government buildings and transit hubs.
- Enables real-time sign language interpretation via public address systems.

Contributions and Impacts on Society

- Addresses crucial issues in **communication accessibility** for individuals with hearing impairments.
- **Empowers users** to participate autonomously in various activities.
- Acts as a valuable tool for **instructional purposes**, facilitating easier learning of sign language.
- **Real-time translation** capabilities enhance the immediacy and effectiveness of sign language communication.
- Demonstrates the potential of responsible AI in assistive technology development.
- Can potentially curb the malpractices with suitable long-term enhancements.

Future Work

- Utilize **Computer Vision** and center the human figure on the GAN generated image and position the human in the center to avoid positional variations among different frames.
- Use of a **Stable Diffusion model** for generating the human-like images which leverages the choice of selecting the signer gender and other specificities as a prompt input.
- Evaluate the possibility of training GAN model on Multiple Signers.
- Include a Transition Mechanism while combining two different frames to generate a final video to have a smoother transition between poses generated.

Demo

Time to Demonstrate!

