

Recognizing highly variable American Sign Language in virtual reality

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Introduction

- Over 5% (430 million) of the world's population has some form of hearing loss, which is projected to increase to 2.5 billion by 2050¹.
- American Sign Language (ASL) recognition in 2D/3D is a relatively mature research area than virtual reality (VR).
- As immersive technology grows, ASL interactions in VR are more relevant and timely.



Objectives

- The objectives of this research are to teach ASL in VR and make the learning process fun and entertaining ^{2,3}.
- State-of-the-art ASL recognition research mostly focused on 2D or RGB-D-based cameras where users cannot feel the real-world 3D experience. Our focus is to develop more interactive learning environment.
- Provide real-time feedback to the users.

2. Quandt, L. C., Lamberton, J., Leannah, C., Willis, A., & Malzkuhn, M. (2022). Signing avatars in a new dimension: Challenges and opportunities in virtual reality. In *Proceedings of the 7th International Workshop on Sign Language Translation and Avatar Technology (SLTAT)*

3. Quandt, L. C., Lamberton, J., Willis, A. S., Wang, J., Weeks, K., Kubicek, E., & Malzkuhn, M. (2020). Teaching ASL signs using signing avatars and immersive learning in virtual reality. In *The 22nd International ACM SIGACCESS Conference on Computers and Accessibility (ASSETS '20), October 26–28, Virtual Event, Greece.*



Concept

- Interactive learning system in VR
- Student
 - User of the system (ASL learner).
 - Will get feedback from the system about sign accuracy.
- Teacher
 - Animated avatar who teaches ASL.
 - Mocap data from a native signer.
 - "Decides" whether the student's sign is correct or not.





Signing avatars in a 3D environment

- We designed a 3D interactive coffee shop environment
- The avatar acts as the Teacher and shows ASL signs
- A motion capture system (Vicon) was used to animate the 3D avatar
- We have nine different coffee shop sign in our sign vocabulary





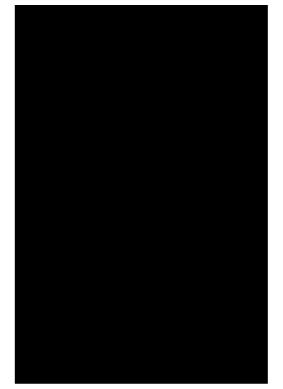
Data Collection

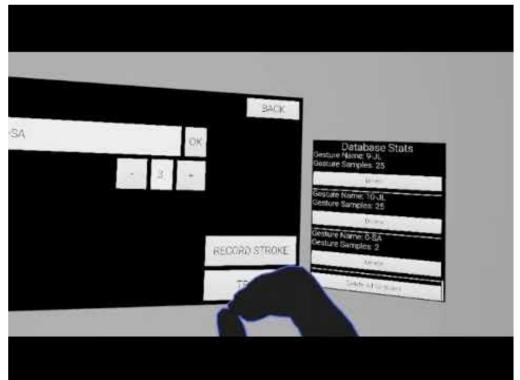
- Data plays an important role in any machine learning and deep learning research.
- For the training process, we gathered ASL sign data from native signers.
- A user interface (UI) was designed where users could interact with some basic buttons and record ASL signs without much external intervention.
- User can see the existing and new sign in the right window (see next slide).

				BACK	
Gesture Name					Database Stats Number of gestures: 0
Gesture Duration	-	3 -	+		Doloto Ali Cestures
				RECORD STROKE	
				TRAIN	



Data Collection - Video

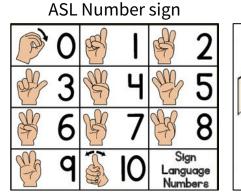




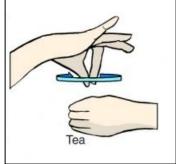


Datasets

- We trained two different models with different dataset-
 - ASL numbers 0-10
 - $\circ \quad \mathsf{TEA}\,\mathsf{sign}$



ASL TEA sign



Number Signs 0-10

- Total signs: 2500
- Participants: 10
 - Men: 4
 - Women: 6
 - Age range: 22-46
- Hearing status: D, H, HH

TEA Sign

- Total signs: 500
- Participants: 10
 - Men: 4
 - Women: 6
 - Age range: 22-46
- Hearing status: D, H, HH



Experimental Setup

- VR Device: Oculus Quest 2 (software version 44.0.0.169.455).
- Environment: Unreal Engine v4.27
- Al Plugin: MiVRy v2.5
- PC:
 - Windows 11 pro 64 bit
 - Memory: 32GB
 - Processor: Core i9 3.50Ghz



Methods

- Sign recognition is performed by an AI model.
- MiVRy Unreal Engine plugin is used for sign recognition.
- The AI model is generated based on training data.
- The model provides similarity values in real time and this is the backbone of our feedback system.





Results

- Each number was signed 10 times.
- Average accuracy was 46%. •
- Best and worst result found for numb 10 and 6, respectively. The signer signer signer the signer signer the signer sign
- *The recognition accuracy of the sign TEA is around 55%.

			The model guesses (recognizes) this:											
			Ø	G	Ð	ØL	W	m	\mathbb{R}	R	R	B		
			0	1	2	3	4	5	6	7	8	9	10	
	Ø	0	4	1	0	0	0	0	0	0	0	0	3	
	¢	1	1	7	0	0	0	0	0	0	0	0	2	
	B	2	0	3	3	0	0	0	0	0	0	0	4	
	M	3	0	0	4	3	0	0	0	0	0	0	3	
	W	4	0	0	0	0	4	3	0	0	0	0	3	
	M	5	0	0	0	0	3	6	0	0	0	0	1	
	\mathbb{H}	6	0	0	0	7	0	0	2	0	0	0	1	
	R	7	1	0	5	0	0	0	0	3	0	0	1	
	R	8	0	4	0	0	0	0	0	0	3	0	2	
	m	9	4	1	0	0	0	1	0	0	0	3	1	
	i d'illi	10	2	0	0	0	0	0	0	0	0	0	8	



Conclusion

The ten signers in this initial study were *diverse* in age, sex, ASL proficiency, and hearing status, with most being deaf lifelong ASL users.

Next steps:

- Improve our recognition accuracy by modifying the AI model
- Add more content / signs
- Continue building the 3D environments
- Continue adding rich facial expressions to the avatars



Thank you! Questions?





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