

# EBERHARD KARLS TÜBINGEN









**USER** 



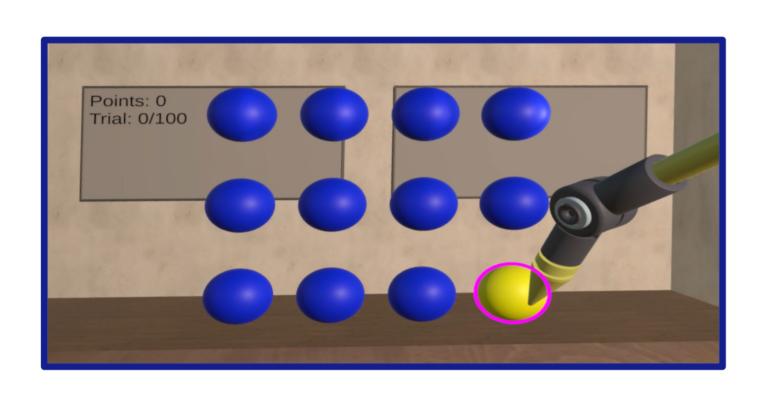
# Communication breakdown: Gaze-based prediction of system error for an Al-assisted robotic arm simulated in VR

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No Target Visualization Group

#### canceled 75% no error Goal: to gain points by not canceled properly detecting the target. canceled The target is selected 25% error through gaze. not canceled Can we

SYSTEM

#### Motivation

#### Accessability with Al

- Al has shown promise accassibility improvements, e.g., rehabilitation [1] and wheelchair navigation [2]
- Al has reached a stage where it can significantly enhance the functionality and effectiveness of these support systems

- Procedure
- Design of a VR simulation that emulates an interactive scenario involving a robotic arm supported by AI
- Evaluation of gaze features as predictors to determine whether users can predict errors

# Experiment

Select

- Look at the yellow ball
- Press button to select the target
- (Participant gets visual feedback from the system)

- Robotic arm starts to move to the selected ball
- Participant is able to cancel the trial by pressing stop button

Return

 After ending the trial, the robotic arm moves back to start position

## **Experiment:**

- Starting with calibration
- Explanation inside of the VR
- 2 training trials
- Starting the experiment

#### **Distribution of Errors:**

- 20 Trials without any Errors
- Trial 21 to 40 randomly 25% errors
- Trial 41 to 100 randomly 33% errors

### References

- [1] Min Hun Lee et al. (2022) Enabling AI and robotic coaches for physical rehabilitation therapy: iterative design and evaluation with therapists and post-stroke survivors. International Journal of Social Robotics (2022), 1-22.
- [2] Walid Zgallai et al. (2019) Deep learning Al application to an EEG driven BCI smart wheelchair. In 2019 Advance in Science and Engineering Technology International Conferences (ASET). IEEE, 1-5
- [3] Candace E. Peacock et al. (2022) Gaze as an Indicator of Input Recognition Errors. Proc. ACM Hum.-Comput. Interact. 6, ETRA, Article 142 (May 2022), 19 pages.

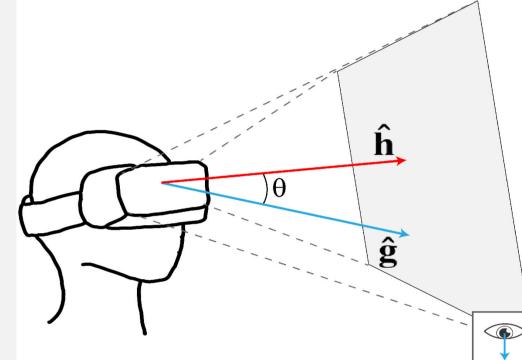
#### Feature

Determination of the first gaze vector after selecting the target

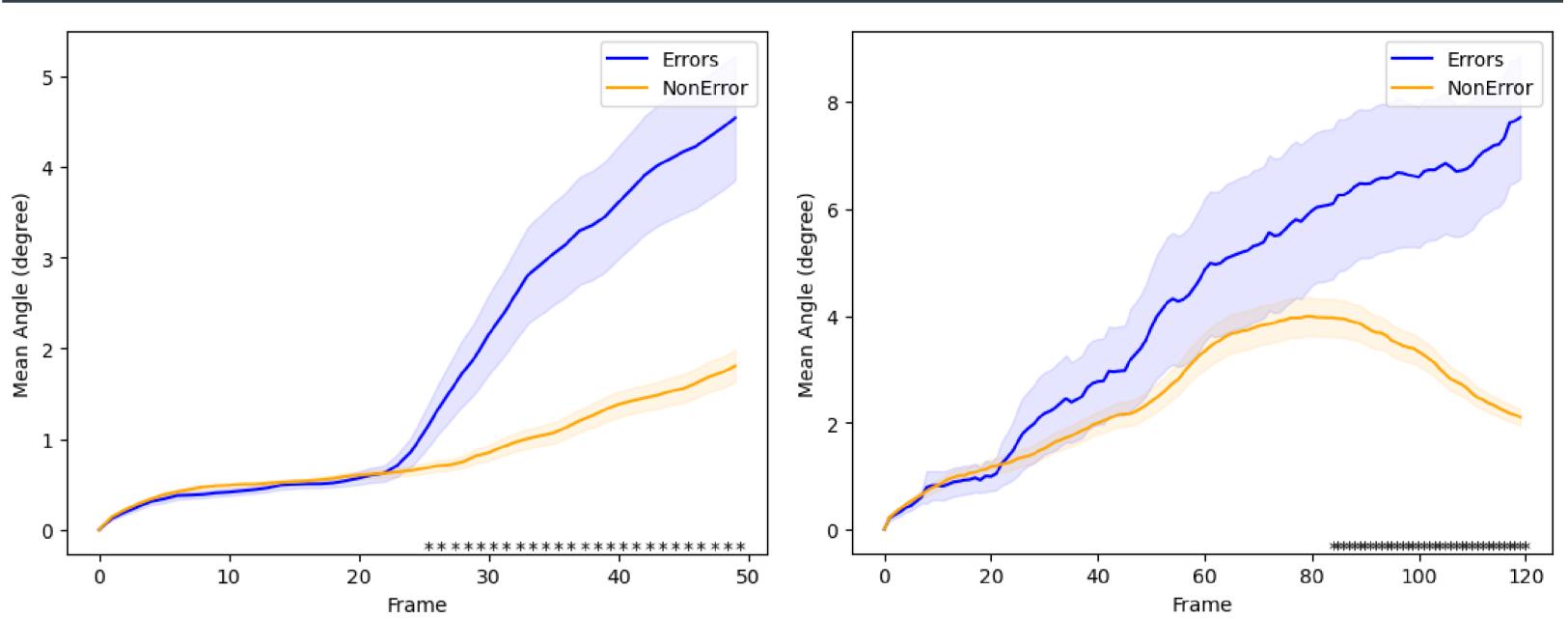
predict errors

with gaze?

- Calculate the angle between the first gaze vector and the following one
- This gives an impression of how far the gaze is from the first gaze vector after selection



#### Results



#### With visualization:

- Significant differences after 25 frames (≈300ms)

- Input first order difference of the gaze vector of the last 40
- Performance very subject dependent

#### Without visualization:

 Significant differences after 85 frames (≈900ms)

#### Classification with visualization:

- Using a TCN [3]
- frames

		Accuracy	Chance	Precision	Recal
	All	0.701	0.749	0.346	0.218
	Subj12	0.933	0.667	1.000	0.800
	Subj17	0.724	0.655	0.667	0.400
	Subj08	0.500	0.667	0.353	0.600
	Subj07	0.655	0.655	0.500	0.200

### Summary

- Design VR simulation that emulates an interactive scenario involving a robotic arm supported by Al
  - Choosing a target with gaze
  - Robotic arm will reach it
- Evaluation of reaction to errors based on the interface visualization system
  - Possibility of false target selection by the simulated Al
  - Gaze based features found with significant differences
- Using gaze features as predictors to determine system errors in an online setup
  - Gaze features are used for classification
  - Results are very subject dependent