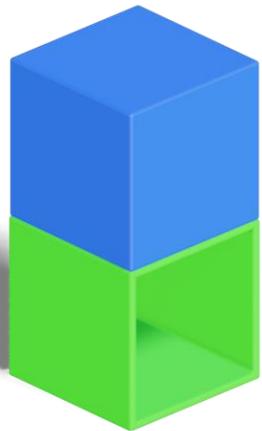


AVHRC 2020

Combining Reinforcement Learning with Supervised Deep Learning for Neural Active Scene Understanding

Dano Roost, Ralph Meier, Giovanni Toffetti Carughi,
and Thilo Stadelmann

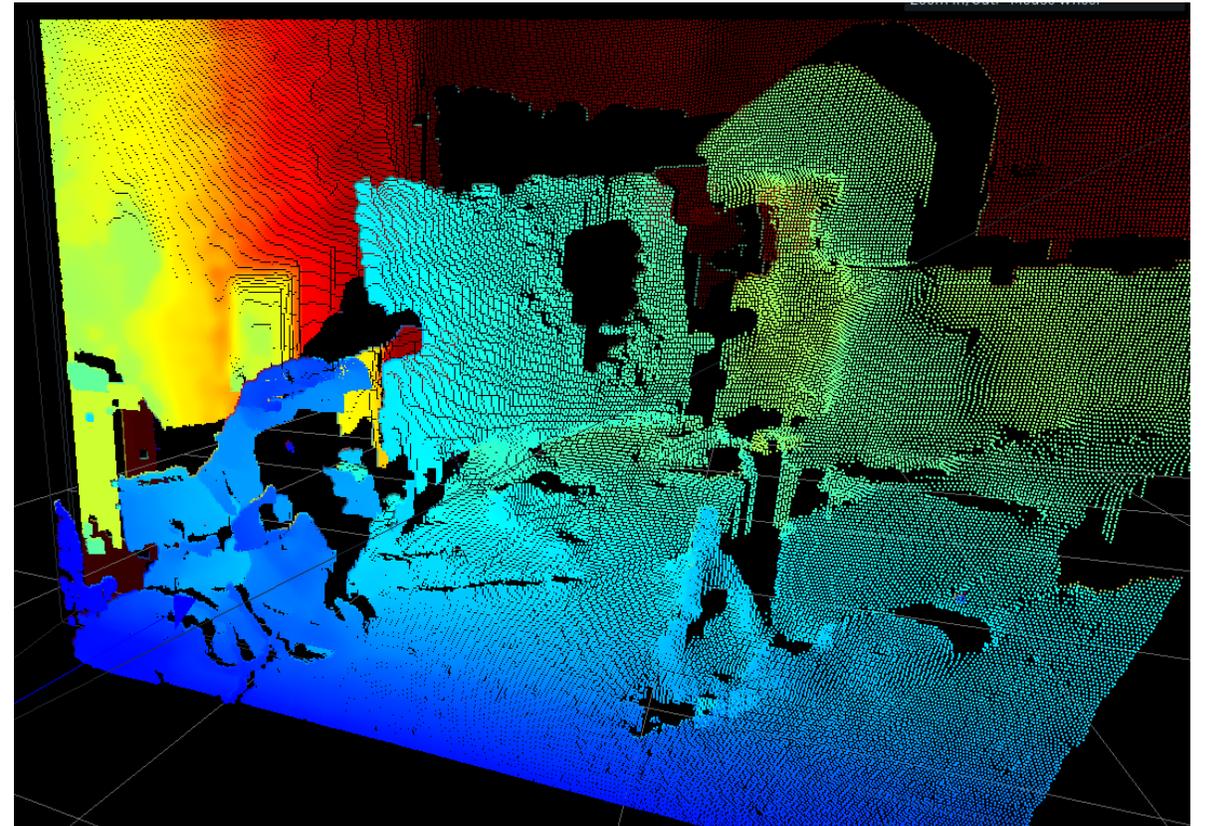
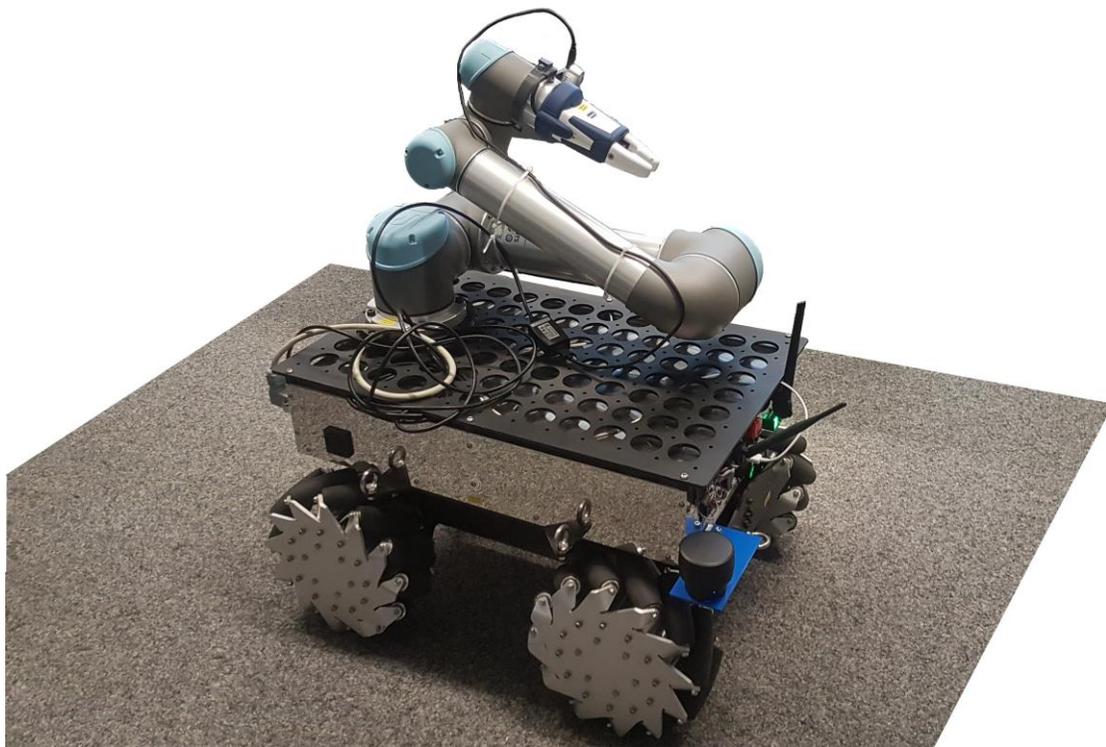


Zurich University
of Applied Sciences

**zh
aw** School of
Engineering

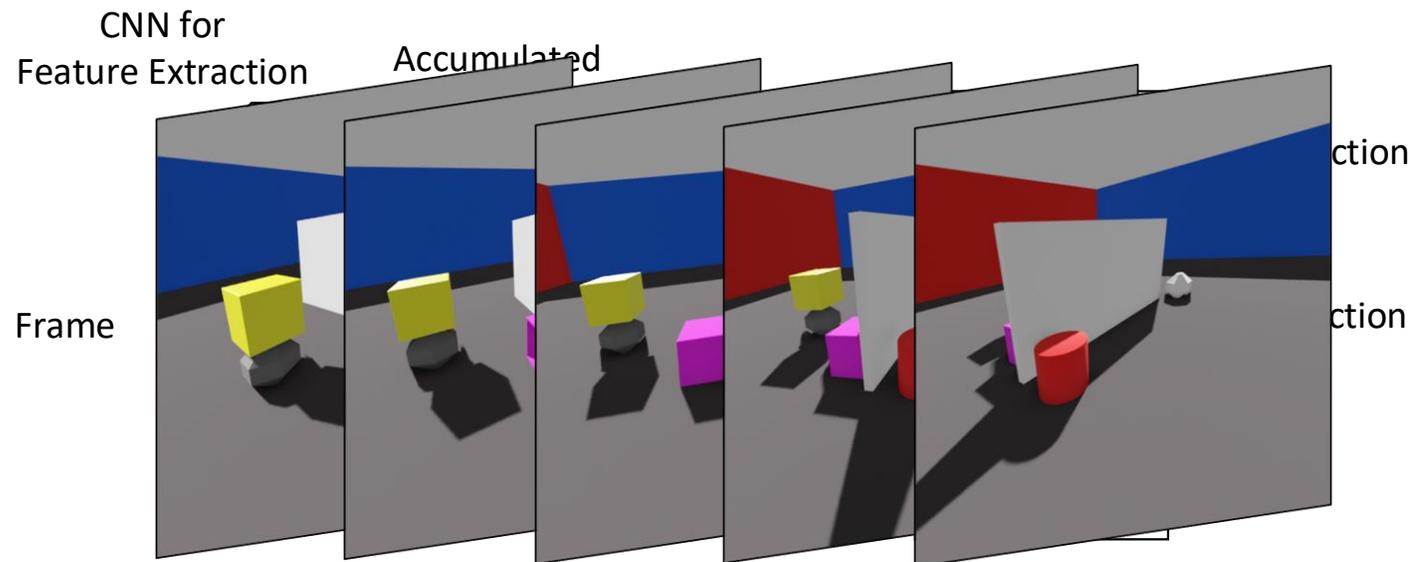
Traditional mobile robotics: 3D Mapping

→ Handcrafted pipeline to identify, locate & interact



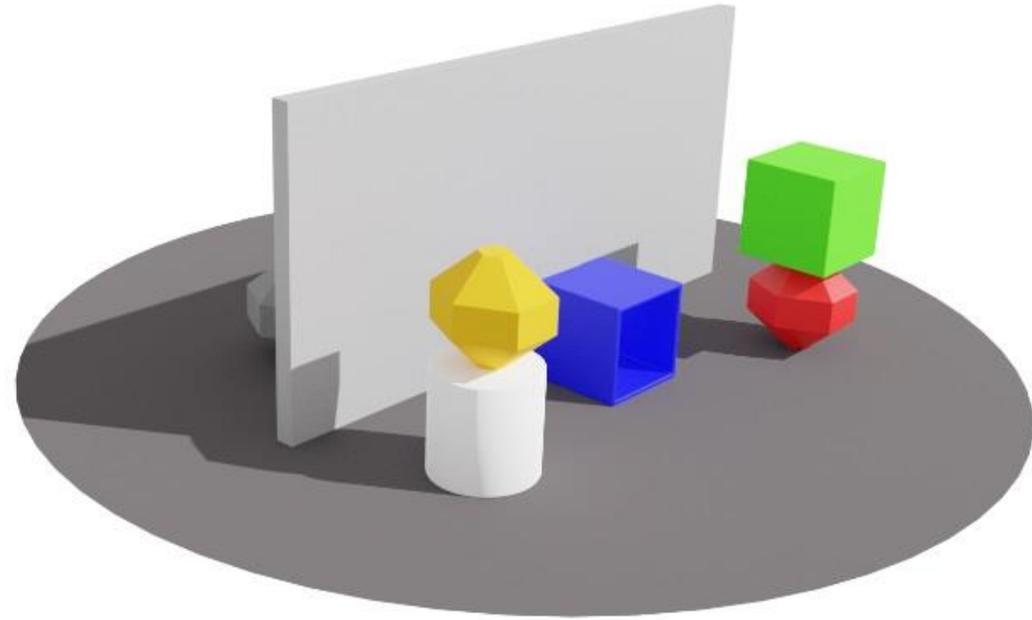
Idea: From Handcrafted to Learned Capabilities

- Use sequences of RGB-D frames
- Accumulate information in **neural network** instead of **point cloud**
- Different outputs → Force network to capture a lot of information



Possible Questions for the System

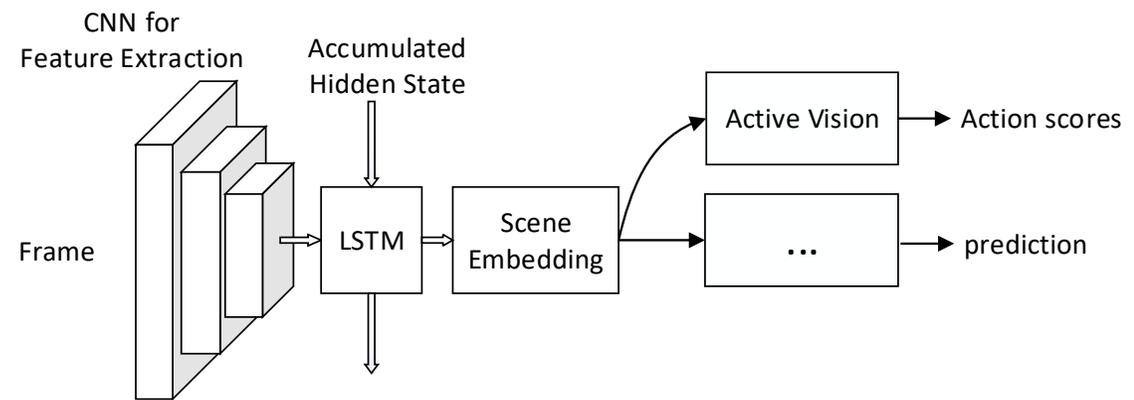
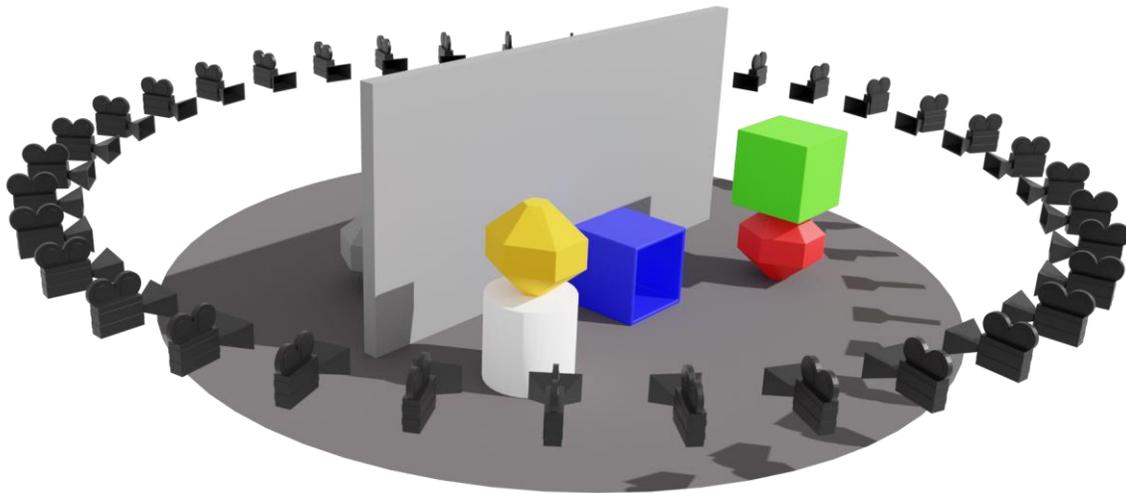
- Color of object at given position?
- Relationship between two objects?
- List all present objects in the scene with another LSTM-layer



- Each output has own loss
- Idea: Use **total loss reduction** as **reward** signal for reinforcement learning based active vision

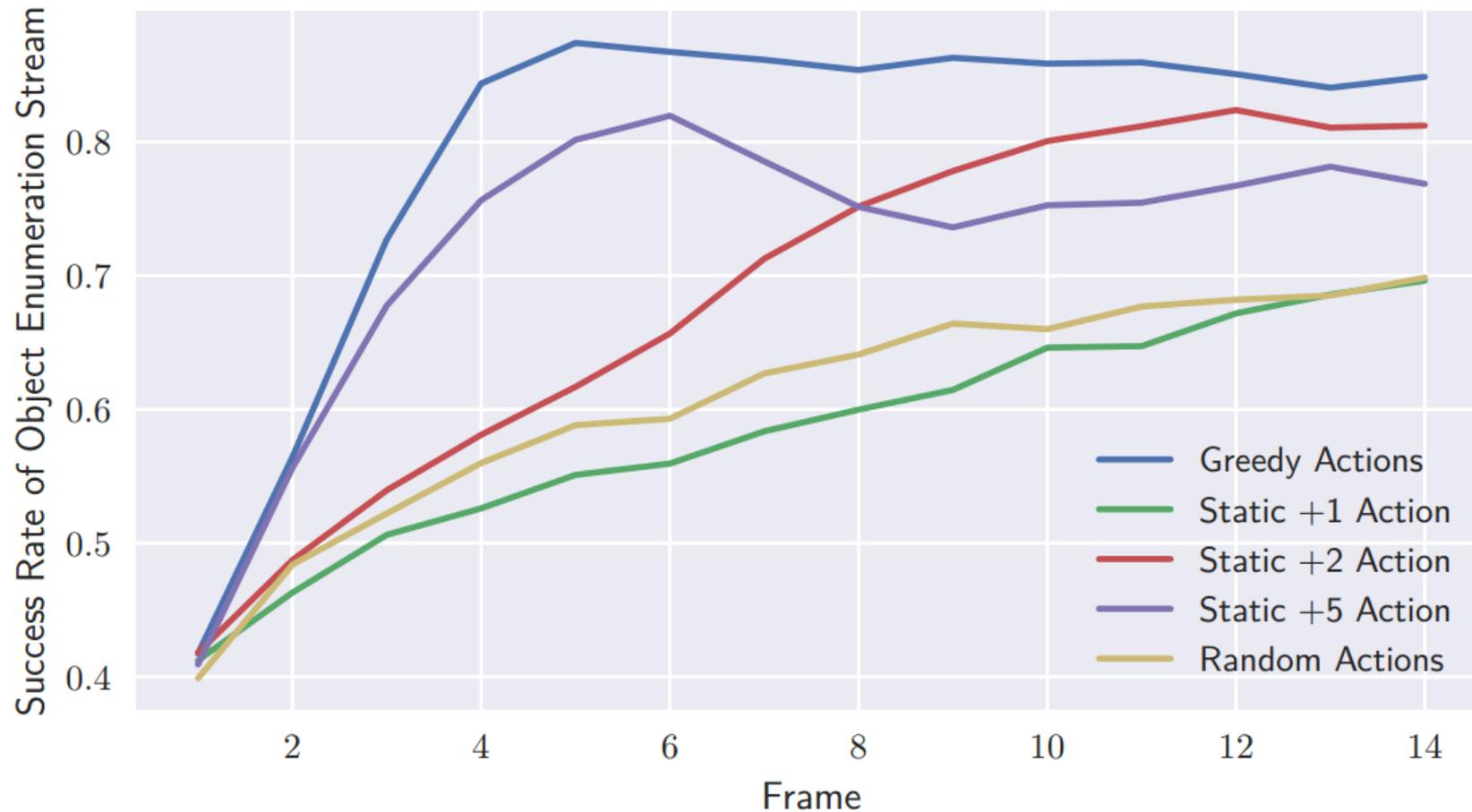
Control Camera with RL

- **Discrete camera positions**
- Actions: 5 left, 2 left, 1 left, stay here, 1 right, 2 right, 5 right
- Use **Q-Learning** to learn score for each action, given the accumulated information



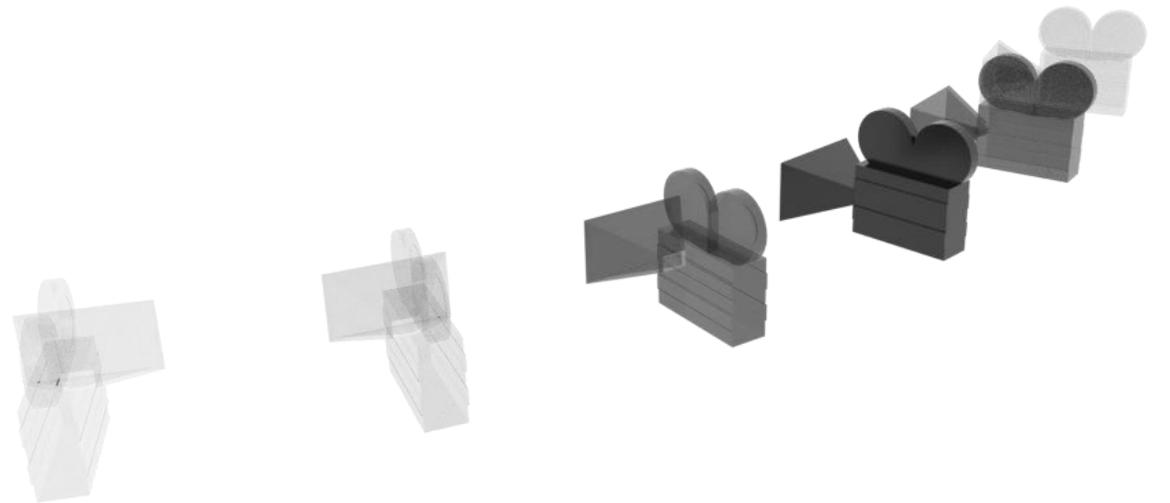
Impact of active vision

For benchmarking: We evaluate stream that should output all objects



Summary

- 40,000 simple synthetic scenes for training
- Easier to train additional outputs – only 200 scenes needed
- System is capable of remembering relative object positions, **even when camera moves and objects are occluded**



Next steps/further research

- Most important: Check, whether this works with real world data
- Maybe: use Transformer instead of LSTM
- Switch to Continuous Camera Control



Thanks for joining!

Check out our work on Github:

<https://github.com/Danoishere/ba-brain-net>