

2018

Year 2 of Interactive Grounded Language Understanding

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Summary

What was done during the second year of the IGLU project:

- ▶ Beginning of the integration of developed algorithms on multiple robotic platforms (e.g. IRL-1, Baxter).
- ► Study of the effects of grounding in multimodal neural machine translation.
- ► Speech-related processing in terms of source separation and visual detection of the active speaker.
- Generative modeling for language learning using probabilistic frameworks.
- Goal-oriented visual and dialogue tasks for language learning and grounding.
- Evaluation framework for grounded language understanding in cognitive agents.

Evaluation of Language Learning Agents

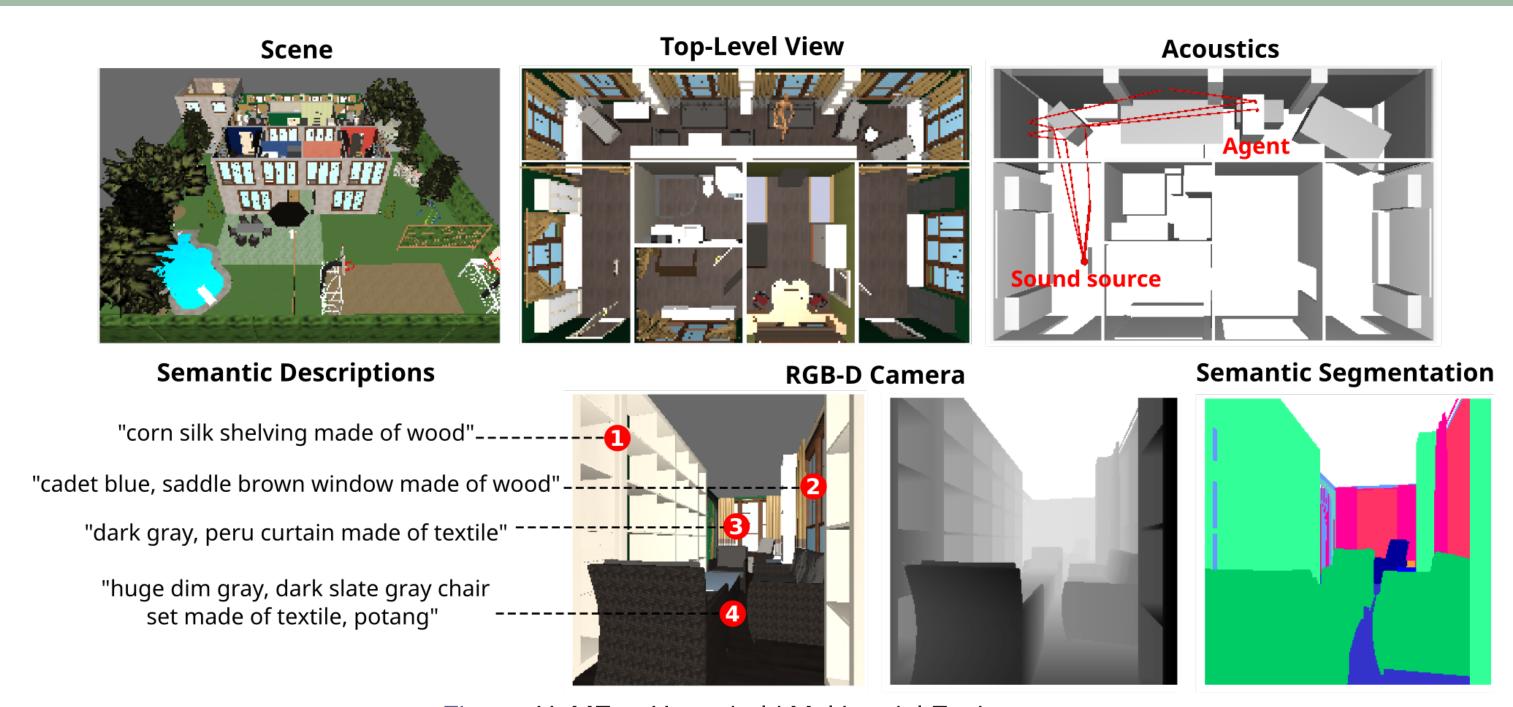


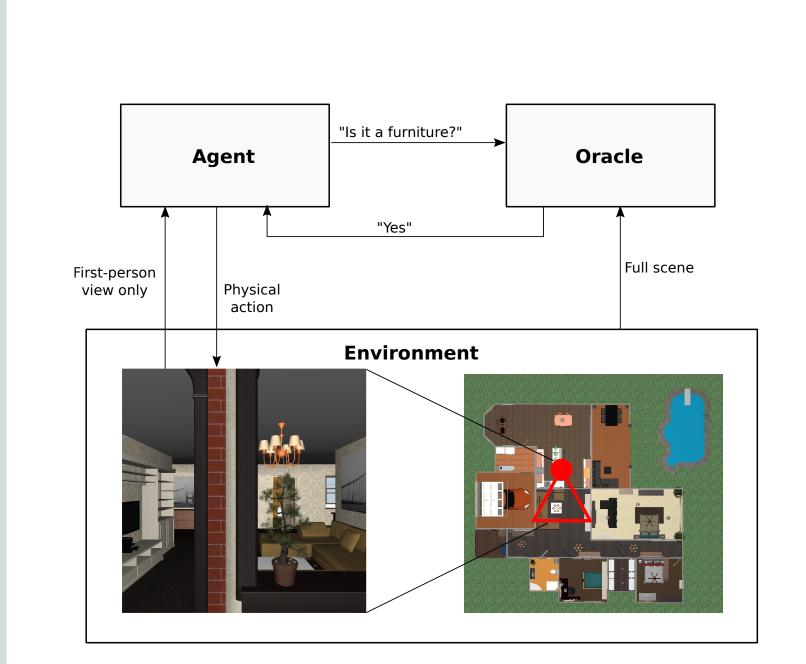
Figure: HoME: a Household Multimodal Environment

there -

Research aims:

- Objective evaluation of the grounding abilities of artificial agents.
- Use goal-oriented dialogue games to learn language and ground it in multimodal perception.
- ► Realistic and complex environments, yet controllable and reproducible research.

Language Learning with Goal-oriented Visual Tasks



more → GRU yellow → GRU ResBlock 2 Conv things → GRU ReLU Linear

Answer: Yes

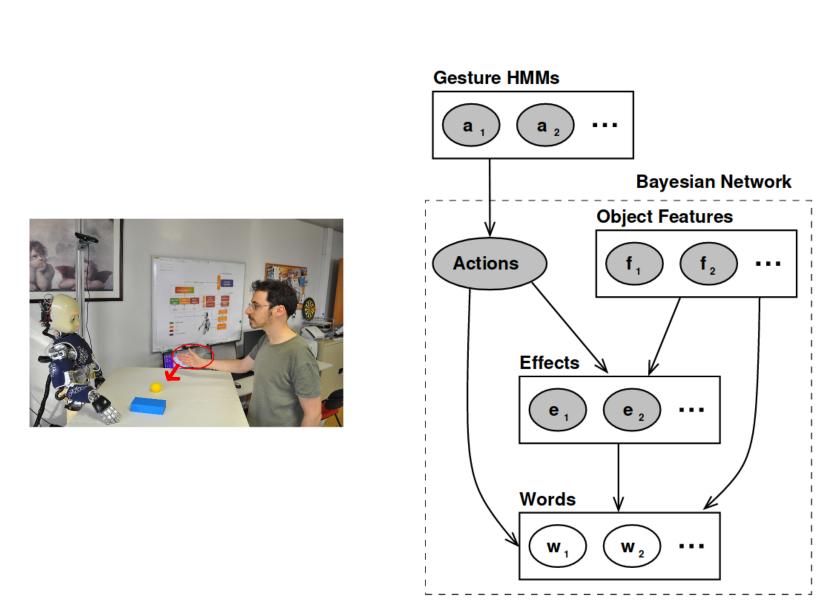
Figure: Situated dialogue on HoME

Figure: Visual question answering on CLEVR

Research aims:

- Situated dialogue and visual question answering that require effective language and vision integration.
- ► General reasoning over visual scenes with attention mechanisms (CBN and FiLM).

Generative Modeling for Language Learning



Input modalities Past events compression Time [sec] representation Predicted modality

Figure: Learning of gestures, language and affordances

Figure: Event-based generative modeling

Multimodal

Density estimation

Research aims:

- Probabilistic frameworks for cognitive agents, with reasoning by building internal model of the environment.
- ▶ Joint learning of robot affordances and word descriptions with statistical recognition of human gestures.
- Multimodal event-based representations and probabilistic generative modeling of robot sensory data.

Integration on Robotic Platforms

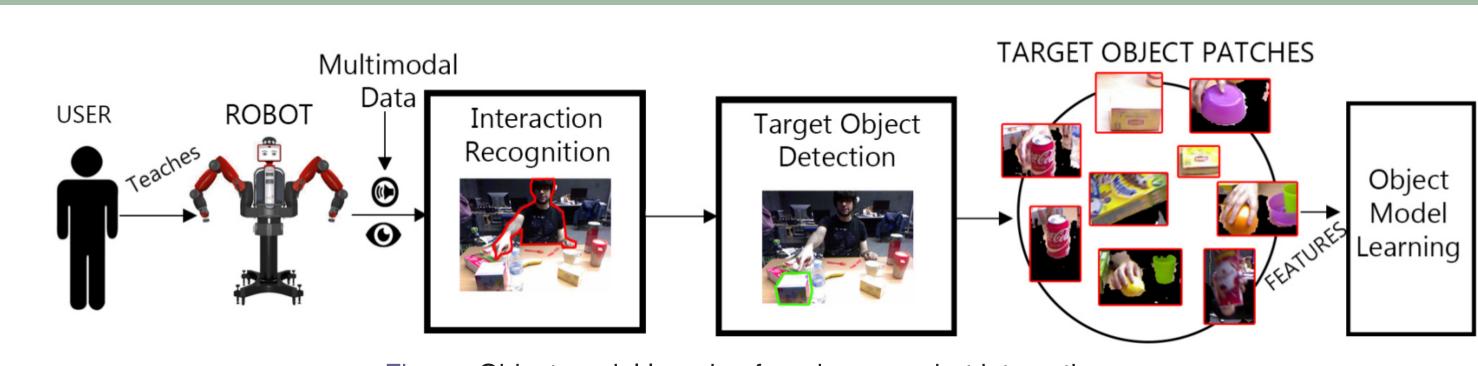


Figure: Object model learning from human-robot interactions

Research aims:

- ► Develop an API for robot incremental learning that supports multiple robotic platforms.
- ► Build a partially annotated dataset for object modeling from natural human-robot interactions.
- ► Acquire object models from interaction data (point, show and speak) using an end-to-end pipeline.
- ► Perform knowledge transfer from simulation to real robots using generative models.

Multimodal Neural Machine Translation



Figure: Translation with attentional mechanism for sentence "Einkleiner schwarzer Hund springt über Hindernisse".

Research aims:

- Solve ambiguities in machine translation by providing the visual context.
- Develop better attentional mechanisms (CBP and CBN) to find objects in the image.
- Improve visual and word representations (DenseCap and Glove).



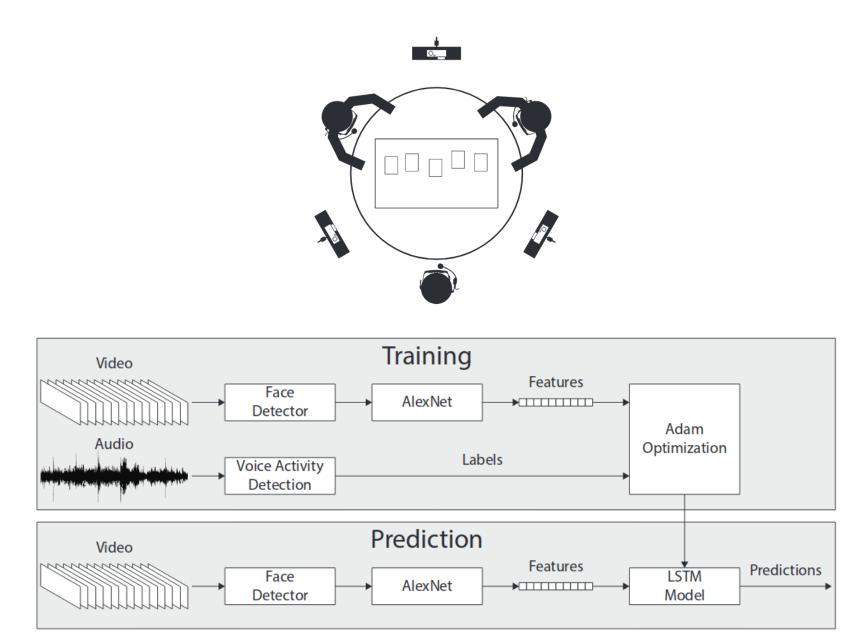
A person in a red jacket with black pants holding rainbow ribbons. Eine Person in einer roten Jacke und schwarzen

Hosen hält Regenbogenbänder. Eine Person in roter Jacke mit schwarzen Hosen hält eine Znde in der Hand

Eine Person in einer roten Jacke und schwarzen Hosen hält einen Zebnde.

Figure: Multimodal translation task

Speech Enhancement and Processing



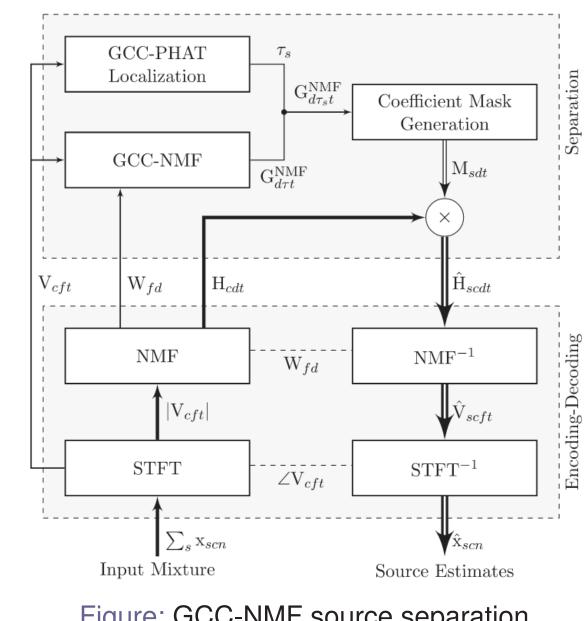


Figure: Active speaker detection

Figure: GCC-NMF source separation

Research aims:

- Automatic visual detection of the active speaker in multiparty interactions.
- ► Real-time source separation system with GCC-NMF, running on embedded systems.

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