

Robots that Learn to Communicate through Natural Human Dialog

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Our Goal

Develop a human-robot dialog system that communicates through natural language, and learns to become more capable from repeated user interactions



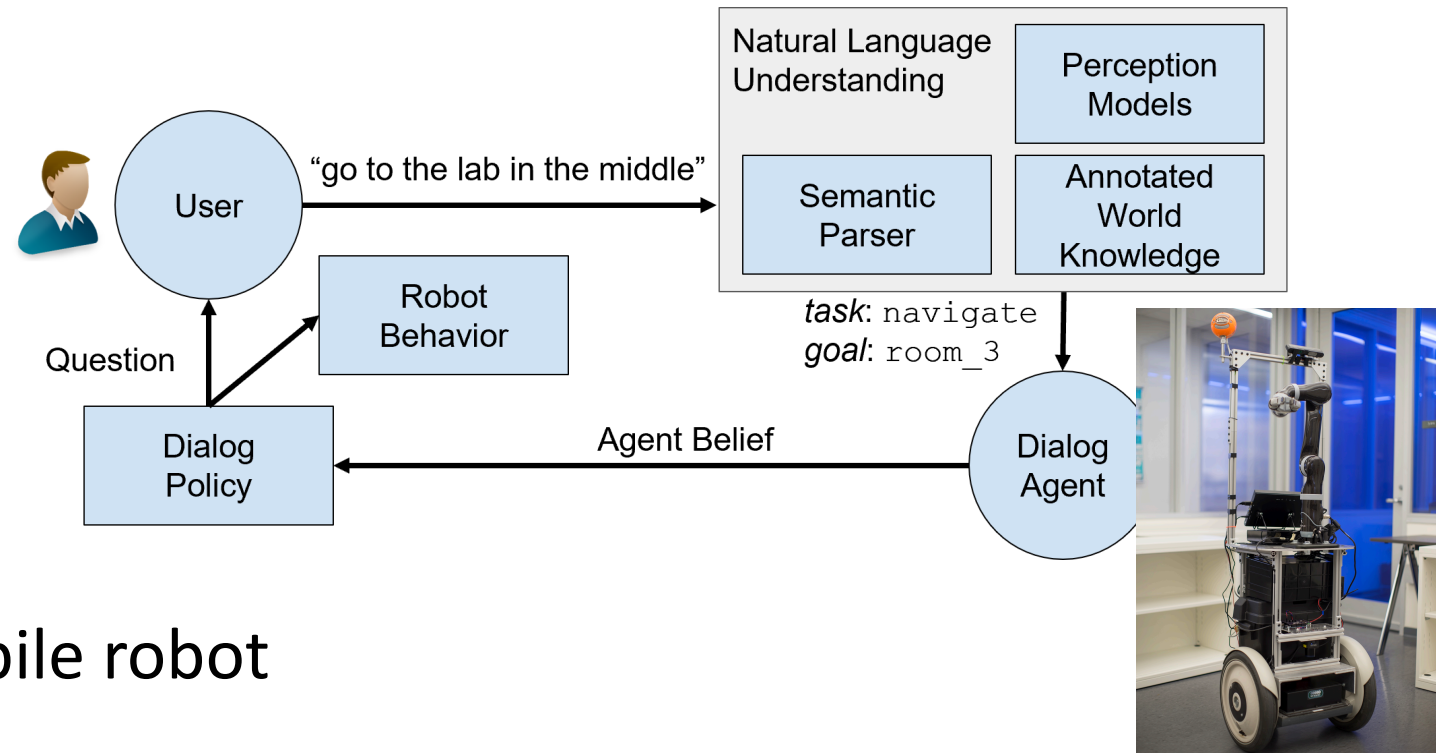
Robot, fetch me the green empty bottle



Method and Novelty

We integrate and adapt learning techniques from:

- Perceptual language grounding
- Semantic parsing
- Dialog management
- Active learning



And implement them on a mobile robot

Perceptual language grounding

Associating words and phrases in language to objects, properties and relations in the world as perceived by the robot's sensors

Vision - RGB(255,0,0)

*“Get the **red** heavy mug!”*

Haptic - joint torque

Exploring objects for all possible sensorimotor contexts takes time



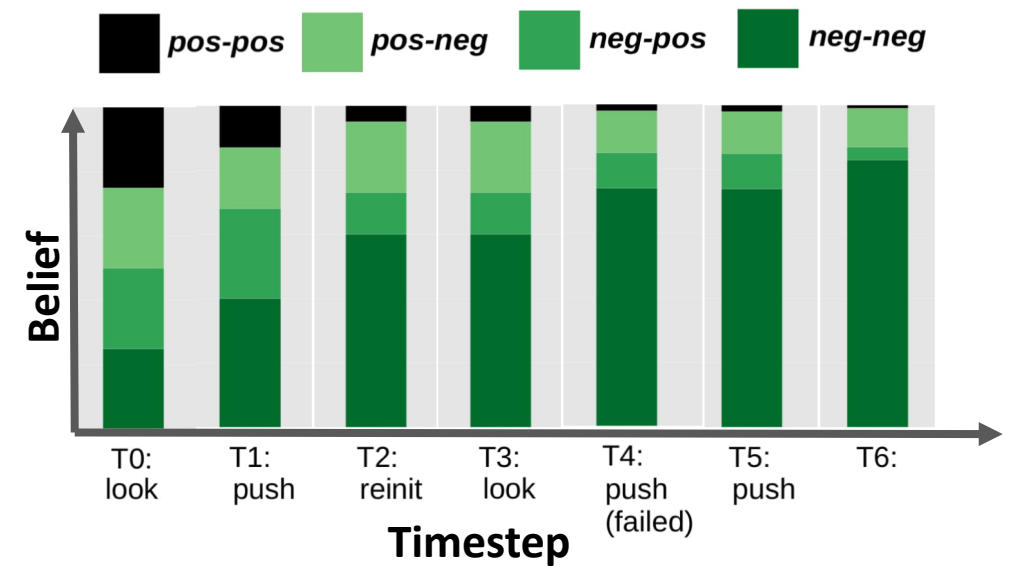
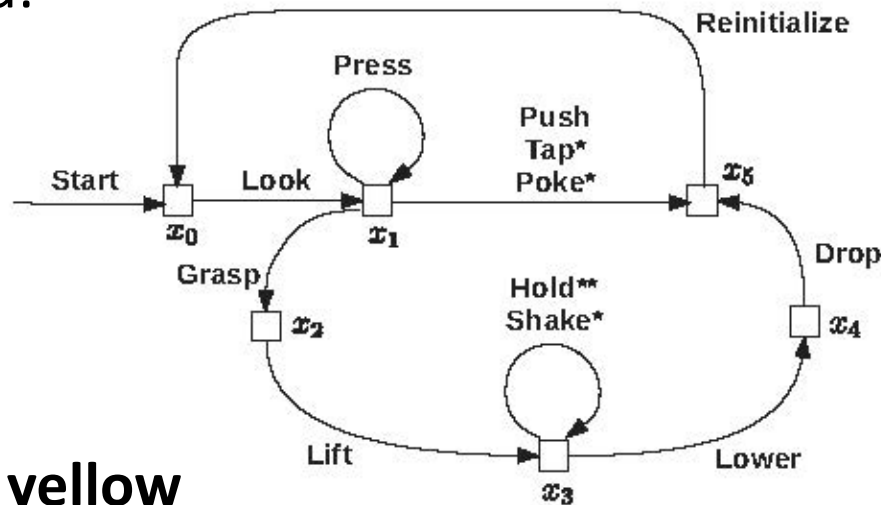
Multi-modal Predicate Identification using Dynamically Learned Robot Controllers , Amiri et al. (IJCAI 2018)

Efficient exploratory behavior using MOMDP

Object selected:

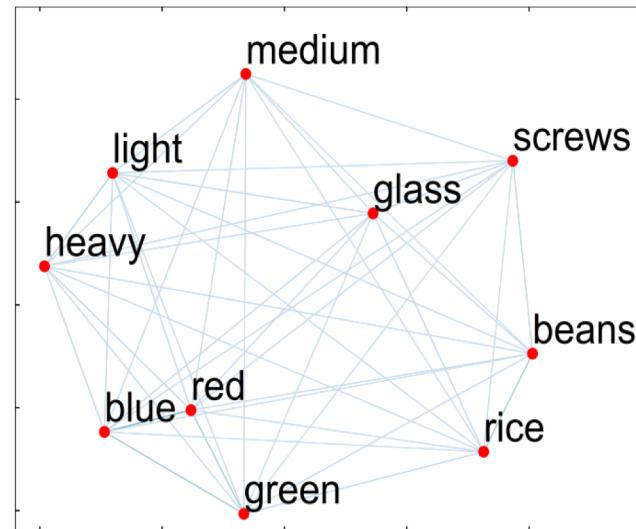
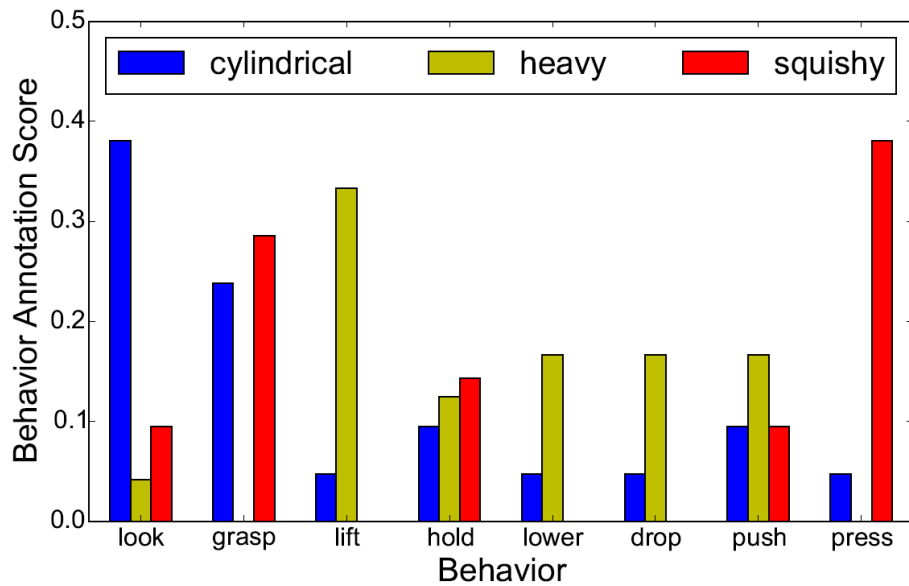


Query: Is this a **yellow metallic** object?

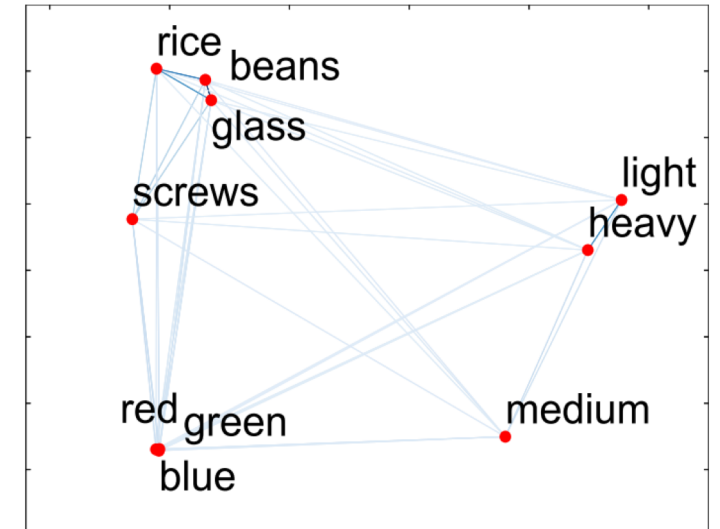


Guiding exploratory behaviors for multi-modal grounding of linguistic descriptions. Thomason et al. (AAAI 2018)

Efficient exploratory behavior using linguistic closeness



Word embeddings



Sensorimotor embeddings

Opportunistic active learning for grounding natural language descriptions. Thomason et al. (CoRL 2017)

Asking off topic questions to improve **future** interactions in object retrieval tasks



Fetch me the **green** bottle

On Topic

Would you describe this object as **green**?

Off Topic

Would you describe this object as **red**?







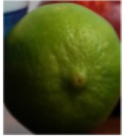







The inquisitive agent achieved **better accuracy** and was rated as **more fun** to interact with

Learning a Policy for Opportunistic Active Learning

Padmakumar et al. (EMNLP 2018)

Using **reinforcement learning** to learn a policy for opportunistic active learning using the Visual Genome dataset

Active Training Set			Dialog		Active Test Set	
Train_1 	Train_4 		Robot	Describe the object I should find.	Test_1 	Test_2 
Train_2 			Human	A white umbrella		
Train_3 	Train_5 		Robot	Is there something in Train_6 that can be described as yellow?	Test_3 	
			Human	No		
Train_6 	Train_7 	Train_8 	Robot	Can you show me an image with something that can be described as white ?		
			Human	Train_1	Test_4 	
			Robot	My guess is Test_4		
			Human	Correct		

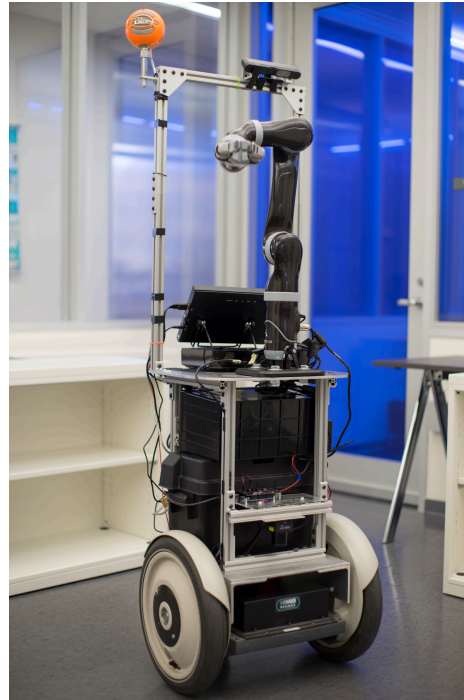
The agent needs to identify which object is best described by a given phrase

Policy	Success rate	Average Dialog Length
Learned	0.44	12.95
Static	0.29	16

The learned policy proved **more accurate** than the static policy with a **shorter dialog** length

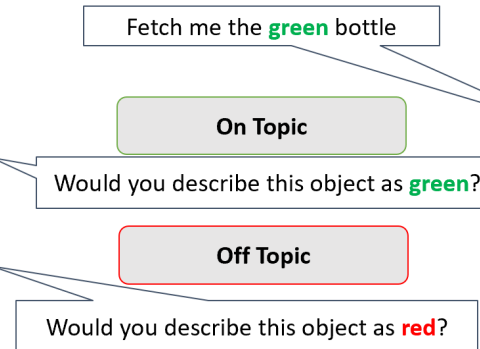
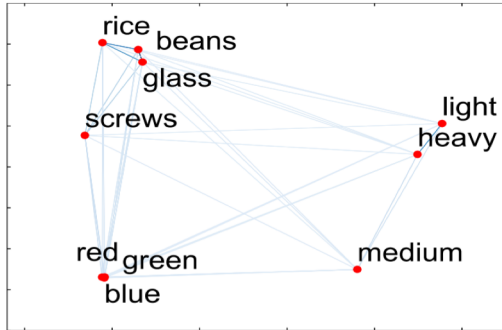
Jointly improving parsing and perception for natural language commands through human-robot dialog. **(RSS 2018)**

Integration on a mobile robot + Demo



Contributions

- A human robot dialog system that **learns to ground new predicates**
- Object exploration as an **MDP** and by modeling **linguistic closeness**
- RL-based **opportunistic active learning** to improve parsing and perception
- Fully **integrated** on a **mobile robot**



Acknowledgements



Ray Mooney



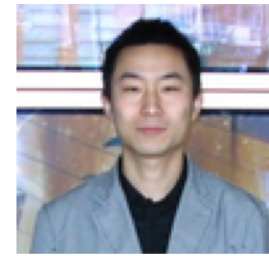
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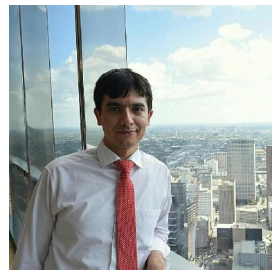
Jivko
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