

# Knowledge-based environment models

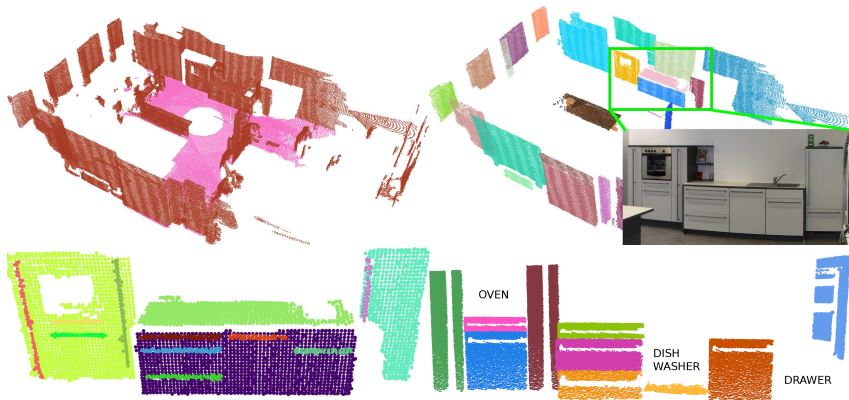
Moritz Tenorth and other members of the  
Intelligent Autonomous Systems Group  
Technische Universität München



# A kitchen...

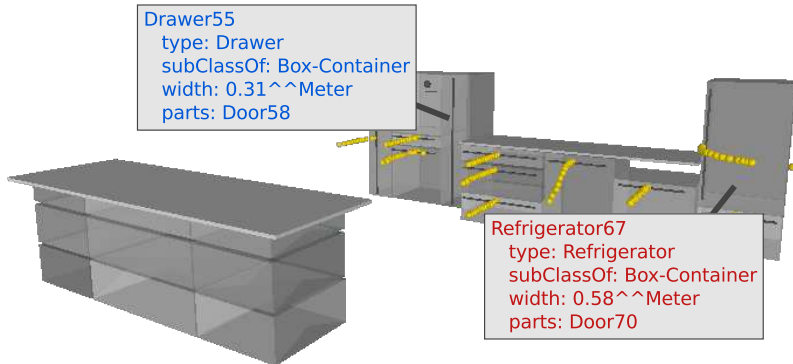


# ...as segmented point cloud...



- Work by Dejan Pangercic, Nico Blodow et al.

# ...and its semantic environment map

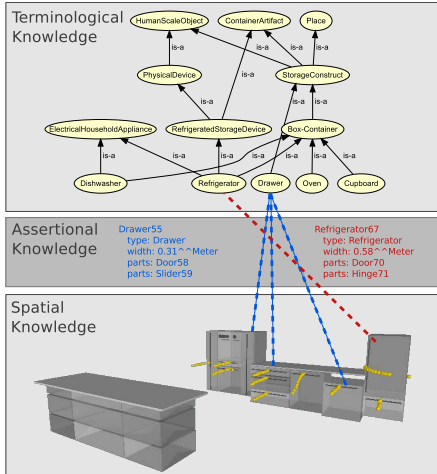


- OWL-based representation in the robot's knowledge base

# Objectives

- ▶ Turn environment maps into a knowledge resource for robots
- ▶ Uniform interface for low-level information (poses, trajectories) and high-level semantics
- ▶ Facilitate integration of map information with other knowledge sources (common sense knowledge, knowledge from WWW, ...)

# Semantic map representation



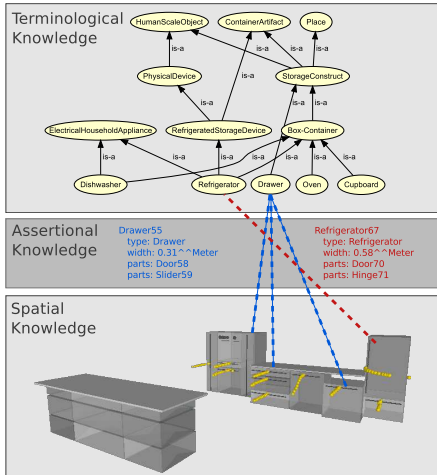
Abstract knowledge  
about object classes

Object instances and  
component hierarchy

Poses in the environment  
and their changes over time

Related: TBOX/SBOX, Galindo et al (RAS 2008)

# Semantic map representation

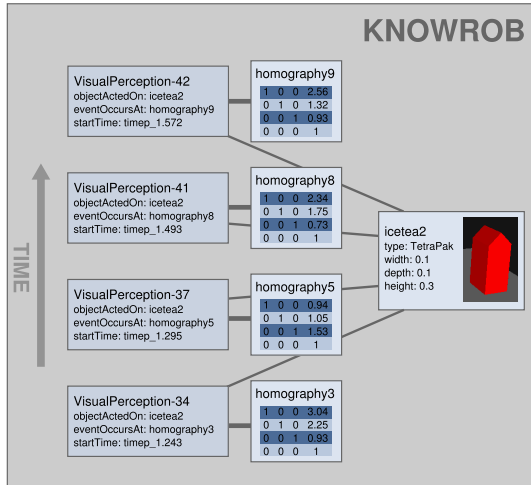


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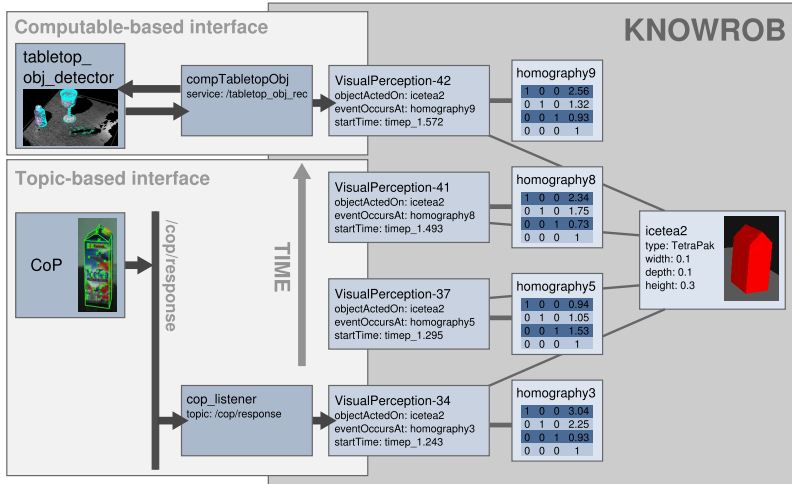
Poses in the environment  
and their changes over time

# Spatio-temporal object representation

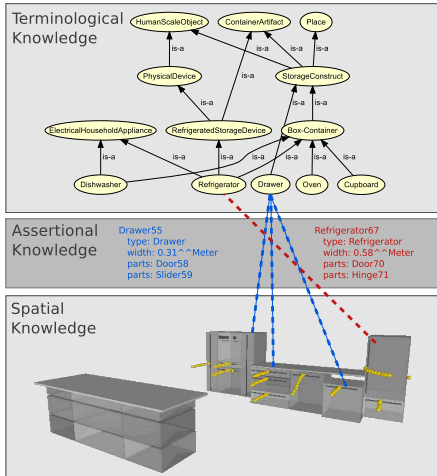




# Integrating perception



# Semantic map representation

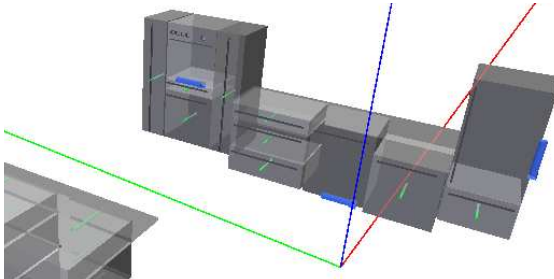


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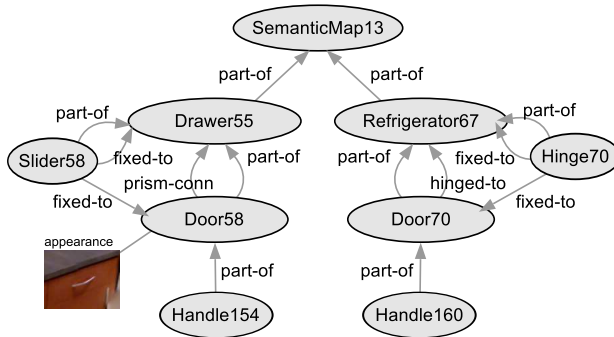
Poses in the environment  
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# Representing articulated objects



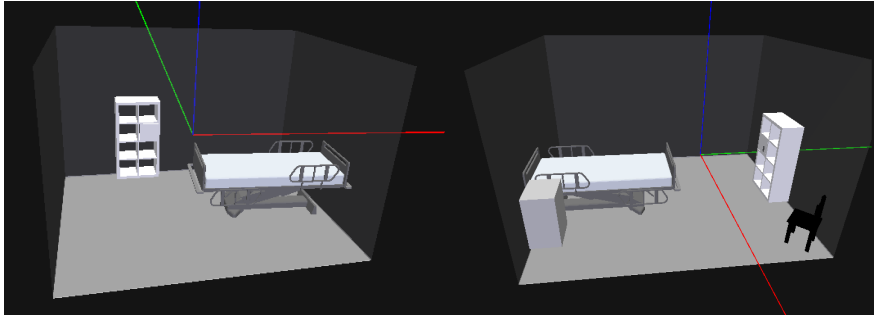
- ▶ Poses and joint limits of hinges and prismatic joints
- ▶ Enable robot to open doors and drawers

# Object part composition



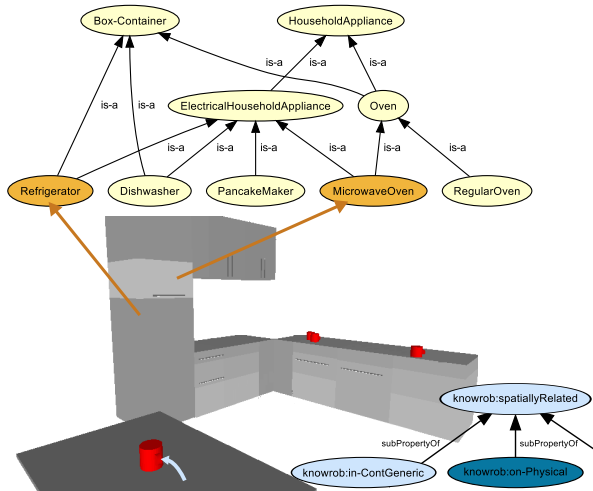
- Part-of hierarchy: composition of objects from parts
- Hinged-to/fixed-to: kinematic properties

# Integrating CAD models

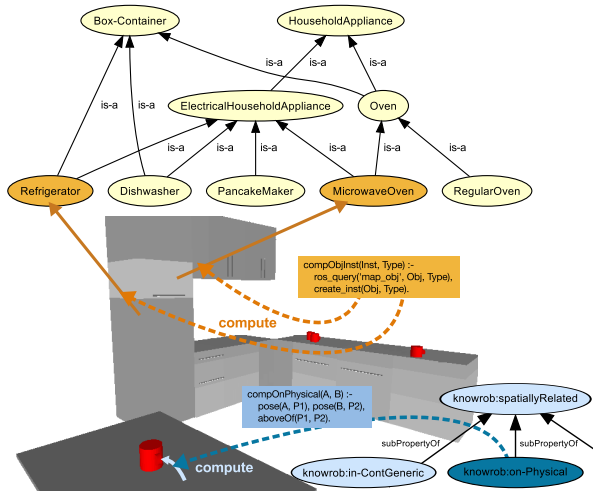


- ▶ Visualization and computation of qualitative spatial relations
- ▶ Soon: segmentation and interpretation of object parts

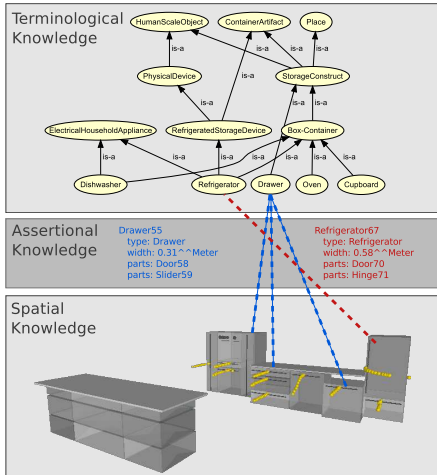
# Computing qualitative spatial relations



# Computing qualitative spatial relations



# Semantic map representation



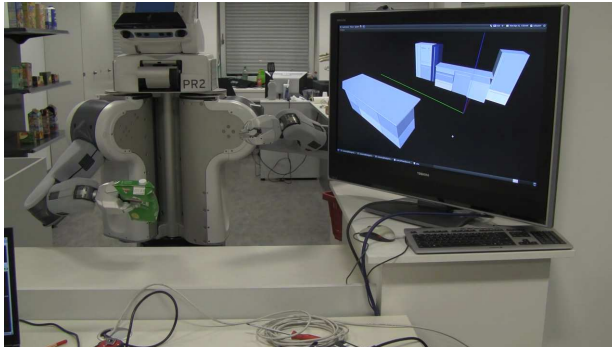
**Abstract knowledge  
about object classes**

**Object instances and  
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**Poses in the environment  
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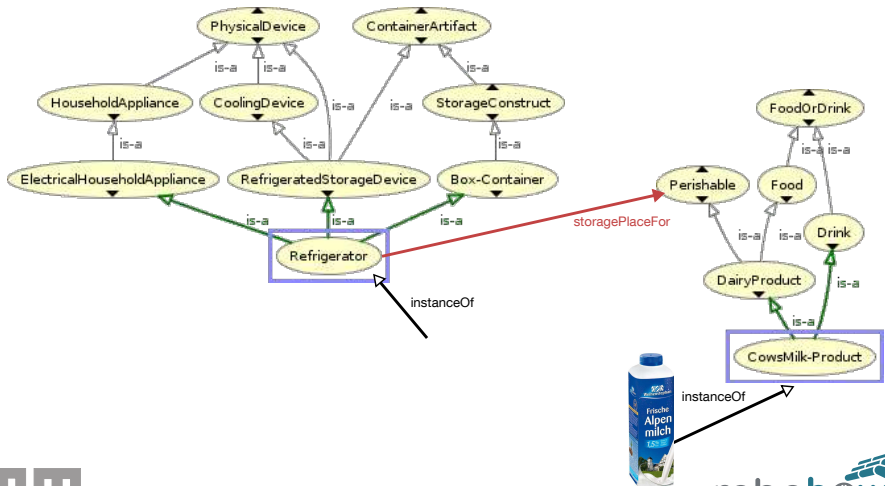


# Video: Unpacking a shopping basket



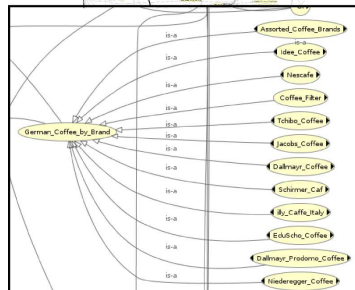
Thanks to Dejan Pangercic, Máthé Koppány, Zoltan-Csaba Marton, Lucian Goron, Monica Opris and Thomas Rühr for making this live demonstration possible.

# Infer storage location based on generic class knowledge

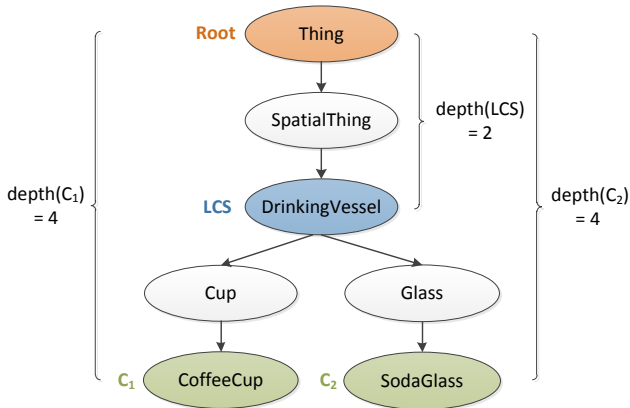


# Object ontology

- ▶ Automatically created ontology of >7500 objects from the online shop [germandeli.com](http://germandeli.com)
- ▶ Class hierarchy from categories + perishability, weight, price, origin, ...
- ▶ SIFT recognition models from product pictures (work by Dejan Pangercic)

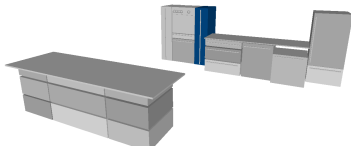


# Infer storage locations based on semantic object similarity



# Infer storage locations based on semantic object similarity

```
?- highlight_best_location_dtree(  
orgprinciples:'CoffeeFilter1', Canvas).
```



Best location: knowrob:Drawer7

Objects at location knowrob:Drawer7:

WUP similarity: object (class)

0.87500: orgprinciples:CoffeGround1

(germandeli:Dallmayr\_Classic\_Ground\_Coffee\_250g)

0.75000: orgprinciples:EspressoBeans1

(germandeli:illy\_Espresso\_Whole\_Beans\_88\_oz)

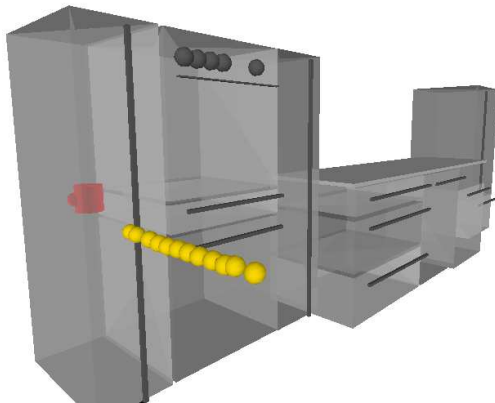
0.70588: orgprinciples:Sugar1

(germandeli:Nordzucker\_Brauner\_Teezucker\_500g)

0.66667: orgprinciples:Tea2

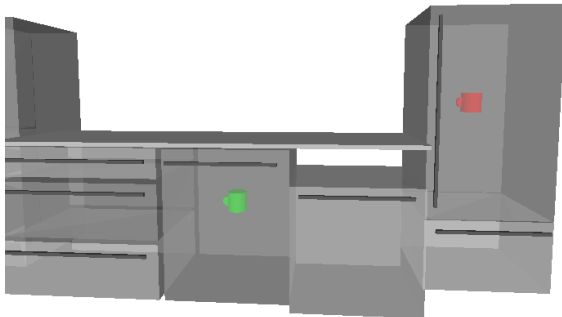
(germandeli:Teekanne\_Rotbusch\_Tee\_Vanille\_20\_Bags)

# Open the drawer where cups are stored



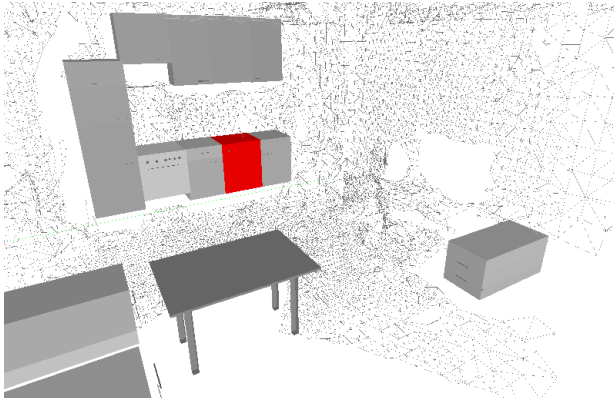
- ▶ Infer most likely storage location
- ▶ Read articulation model from semantic map

# Check if objects are placed correctly



- ▶ Infer most likely storage location
- ▶ Compare with actual locations of objects

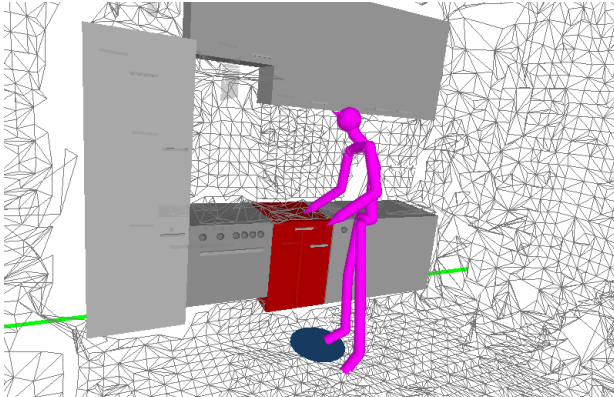
# Select objects based on their purpose



- ▶ Combine abstract object knowledge with map information
- ▶ Here: “appliance that can be used for washing dishes”

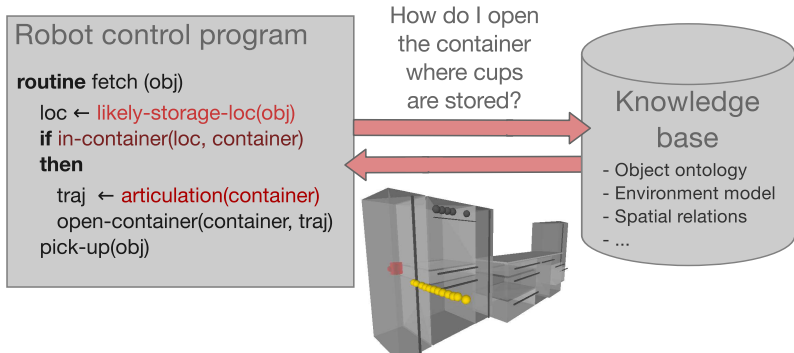


# Integrate with human tracking data



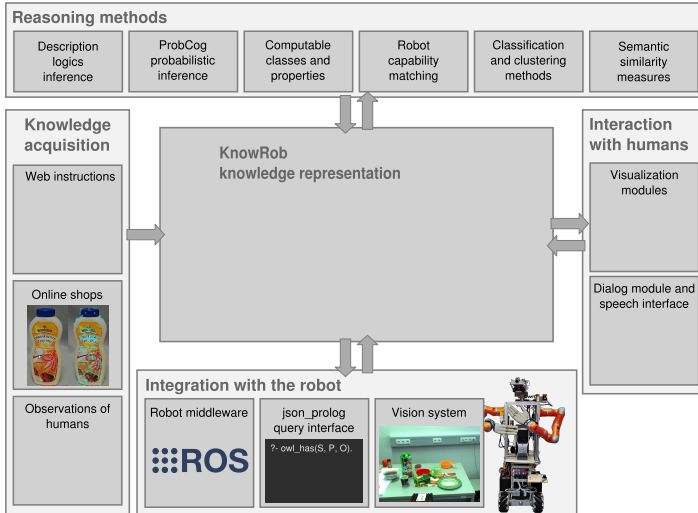
- Observations of humans are represented in the KB
- Reason about their interaction with objects in the map

# Making decisions using map knowledge

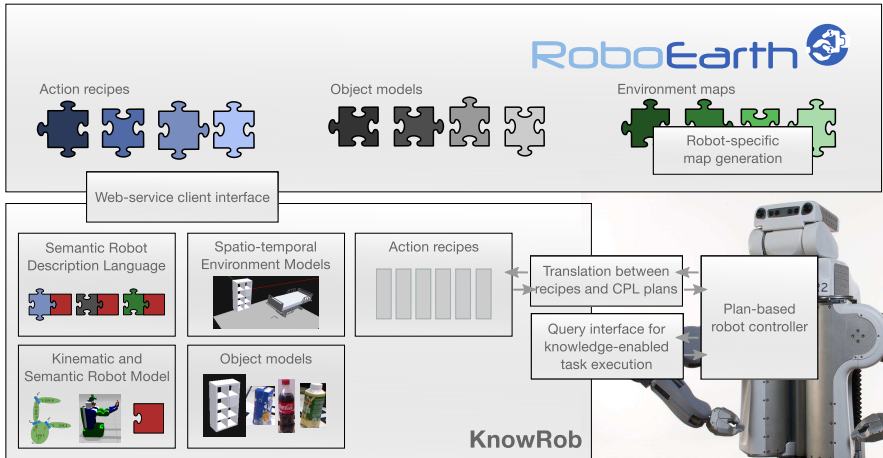


- Formulate control decisions as inference tasks
- Separate control flow from execution context
- Increase re-usability of robot plans

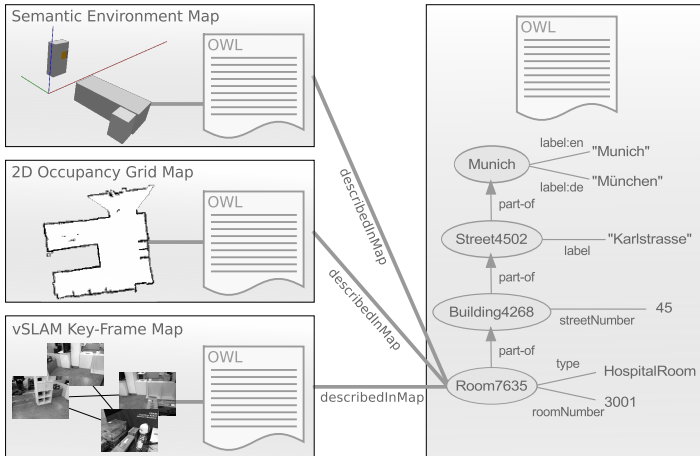
# The KnowRob robot knowledge base



# Exchanging maps via RoboEarth



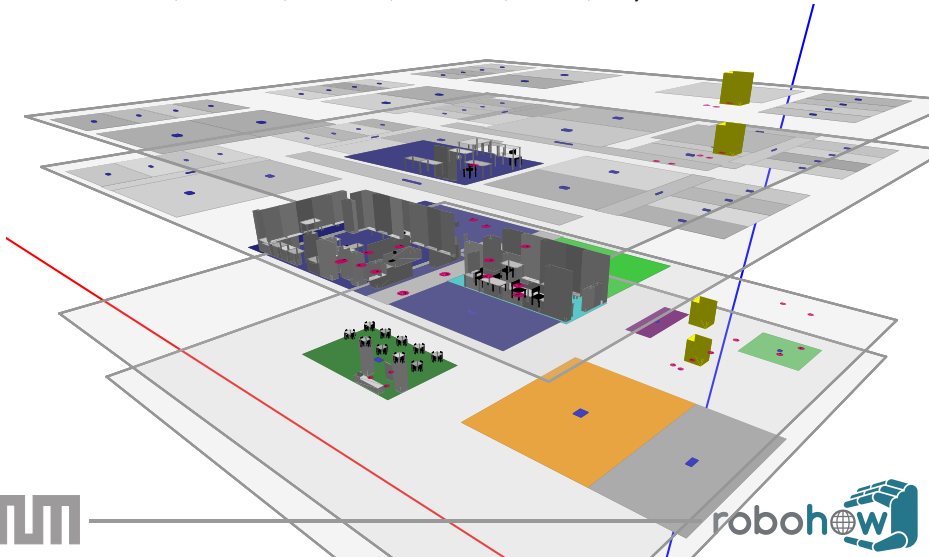
# Finding maps for an environment



# Scaling up to large environments

Searching Objects in Large-Scale Indoor Environments: A Decision-Theoretic Approach.

Lars Kunze, Michael Beetz, Manabu Saito, Haseru Azuma, Kei Okada, Masayuki Inaba. ICRA 2012



# Conclusions

- ▶ Semantic environment maps as robot knowledge bases
- ▶ Spatio-temporal object representation
- ▶ Object composition and articulation properties
- ▶ Integration with abstract knowledge about objects
- ▶ Available as open-source ROS components incl. tutorials

<http://www.ros.org/wiki/knowrob>



**Thank you for your attention!**

Acknowledgments:

