

# Probabilistic Scene Understanding using Virtual Reality and Markov Logic Networks

Dominik Dieckmann

Institute for Artificial Intelligence  
University Bremen

18.12.2018

# Autonomous Robots in Household Environments

- ▶ perception component
  - ▶ detect objects
  - ▶ analyse objects
- ▶ reasoning component
  - ▶ identify/classify objects based on their visual cues
  - ▶ needs to be trained

# Creation of Training Data

time and resource intensive:

- ▶ manually creating scenarios and images
- ▶ no groundtruth

→ synthetic images from a game engine

# System Setup

- ▶ list of objects, classes and scenarios
- ▶ UNREAL ENGINE to create *Unreal Images*
- ▶ ROBOSHERLOCK analysis the images
- ▶ learn a *Markov Logic Network*
- ▶ classify the objects in the images

# Unreal Engine

- ▶ photorealism
- ▶ rendering in realtime
- ▶ open source

# Unreal Images

- ▶ Assets
  - ▶ scanned 3D-models of the objects
  - ▶ kitchen environment
- ▶ URoboVision plugin
  - ▶ create RGBD image from a scene
  - ▶ create *ObjectImage* and *ObjectMap*
  - ▶ send them to ROBOSHERLOCK
- ▶ RSpawnBox class
  - ▶ visual representation of scene space
  - ▶ rotates camera around the scene

# Unreal Images

- ▶ 114 scenes
- ▶ 2 - 5 objects per scene
- ▶ 5 viewpoints per scene
- ▶ only objects of one scenario per scene
- ▶ total of 570 images

- ▶ based on *UIMA*
- ▶ segments images and creates object hypotheses
- ▶ annotates attributes of the hypotheses



# Perceptionpipeline

Annotates the following attributes:

- ▶ color
- ▶ size
- ▶ shape
- ▶ goggles\_{Logo, Text, Product}
- ▶ instance
- ▶ object

Annotates the groundtruth by:

- ▶ counting the color of pixels in the ObjectImage
- ▶ looking up the corresponding asset
- ▶ looking up the groundtruth for that asset
- ▶ setting the groundtruth

# Markov Logic Networks

- ▶ consists of formulas and real-valued weights
- ▶ describes a *Markov Network* with a joint probability distribution
- ▶ is able to:
  - ▶ model relations between objects
  - ▶ compensate inconsistencies
  - ▶ reason about every aspect represented in the model
- ▶ used Toolbox: PRACMLN

## PR2 Looking at Things

### Experiments:

- ▶ ROBOSHERLOCK + MLNs
- ▶ 50 images
- ▶ 21 object categories
- ▶ 4 scenarios

### Results:

- ▶ Annotators are complementary
- ▶ classification rate of 69%

# Experiments

MLN model:

# Experiments

some results

# Conclusion

it just works