

# Semantische Wahrnehmung für den Griff in die Kiste

**Sven Behnke**

**Institut für Informatik VI  
Autonome Intelligente Systeme**

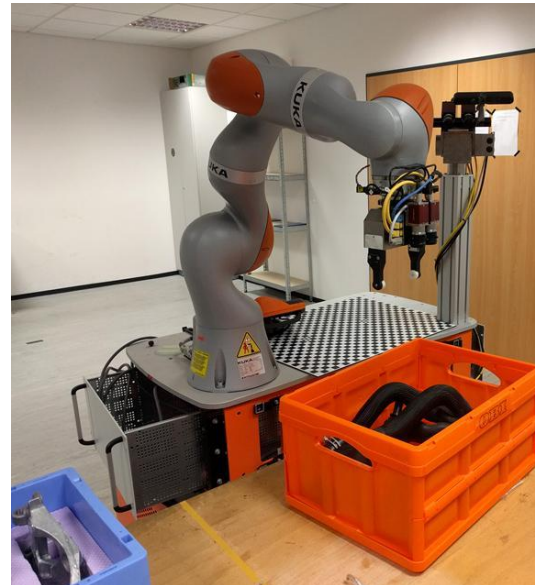


# Roboter für den Griff in die Kiste

ActReMa



EuRoC  
C1



STAMINA

Amazon  
Picking  
Challenge



EuRoC C2

# Serviceroboter



Dynamaid



Cosero

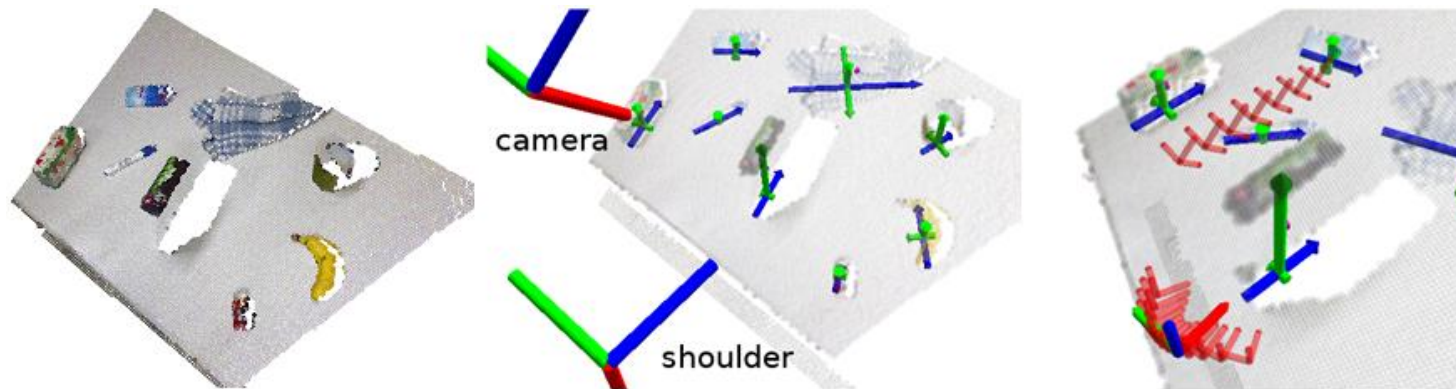
[Stückler et al.: Frontiers in AI and Robotics 2016]

# RoboCup 2013 Eindhoven

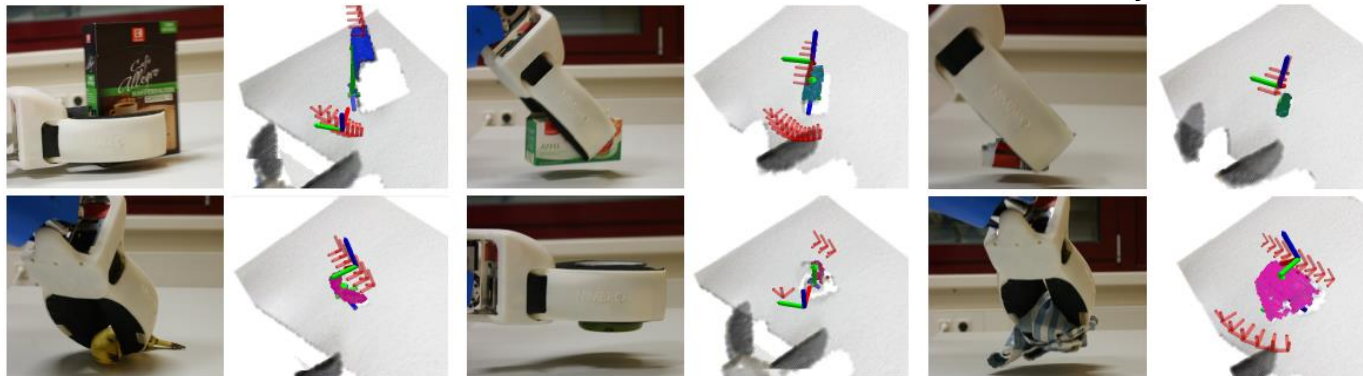


# Greifen unbekannter Objekte

- Detektion von Punktclustern über horizontaler Fläche
- Zwei Griffarten: Von oben und von der Seite



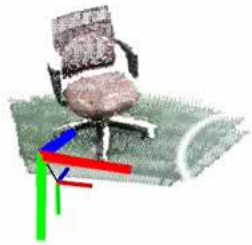
- Flexibles Greifen vieler unbekannter Objekte



# Lernen und Verfolgen von Objektmodellen

- Modeling of objects by RGB-D SLAM

[Stückler, Behnke:  
Journal of Visual Communication  
and Image Representation 2013]

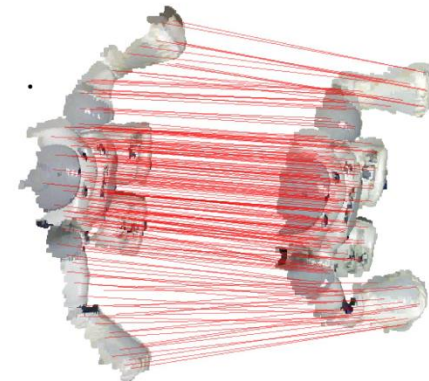
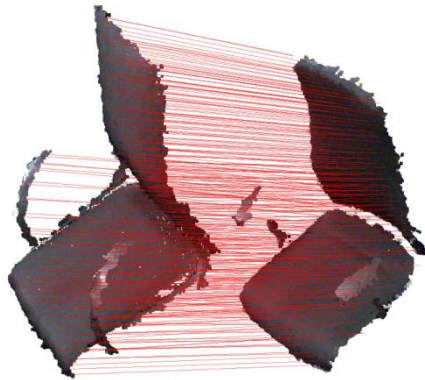
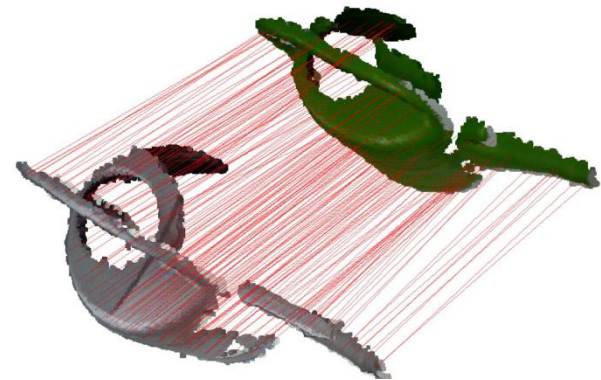
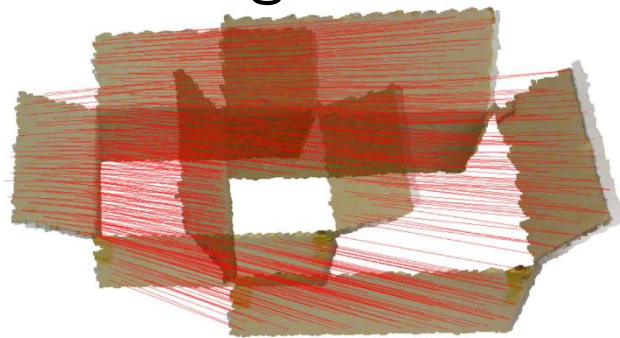


- Real-time registration with current RGB-D image



# Deformierbare RGB-D-Registrierung

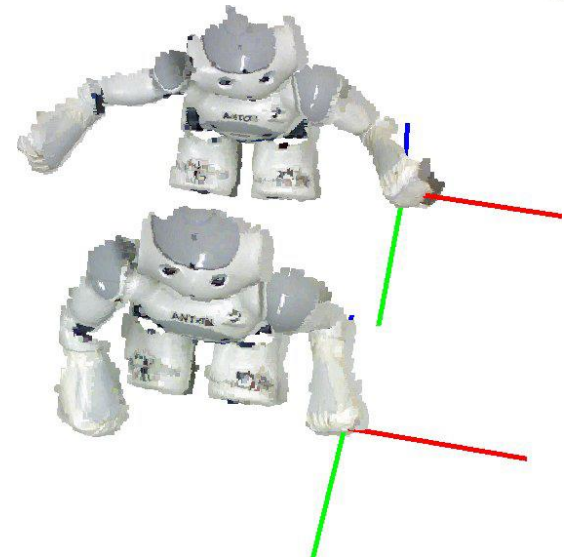
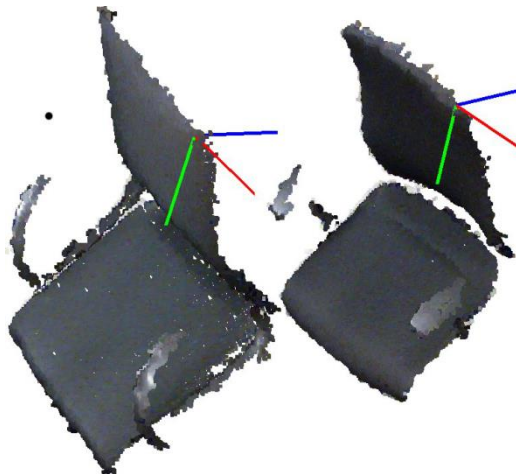
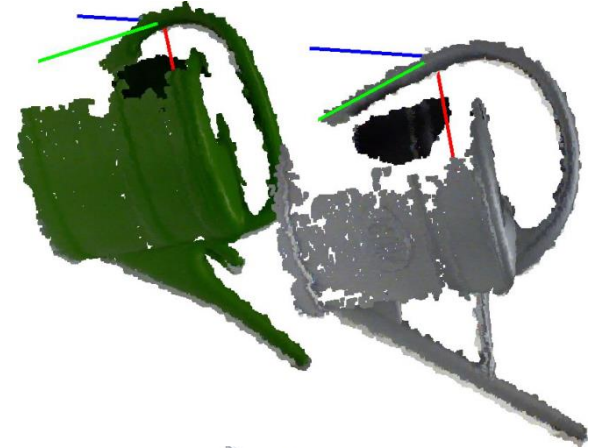
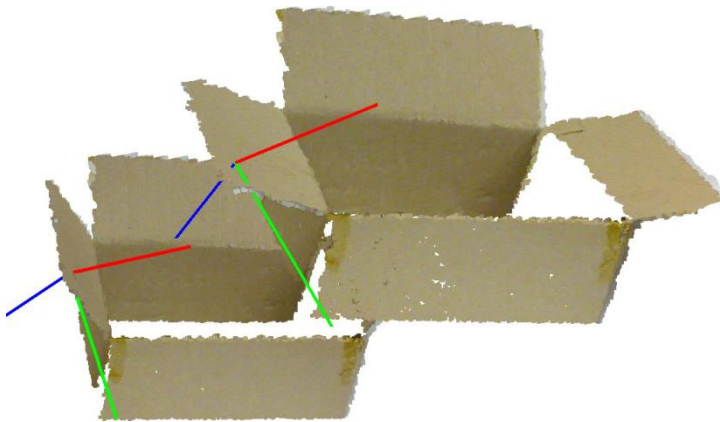
- Basierend auf Coherent Point Drift-Methode  
[Myronenko & Song, PAMI 2010]
- Multiresolutions-Surfel-Map erlaubt Echtzeit-Registrierung



[Stückler, Behnke, ICRA2014]

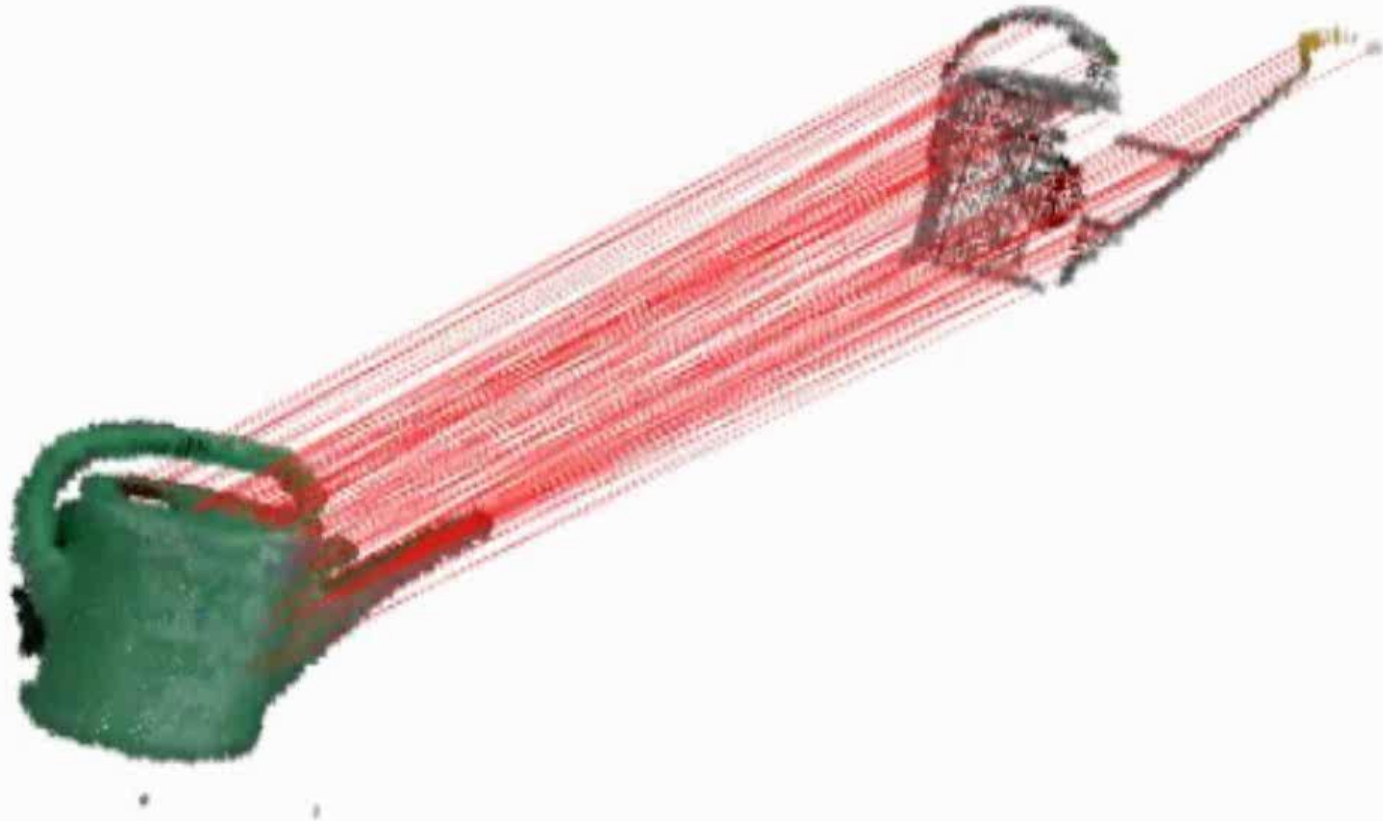
# Transformation von Schlüsselposen

- Abgeleitet aus Deformationsfeld





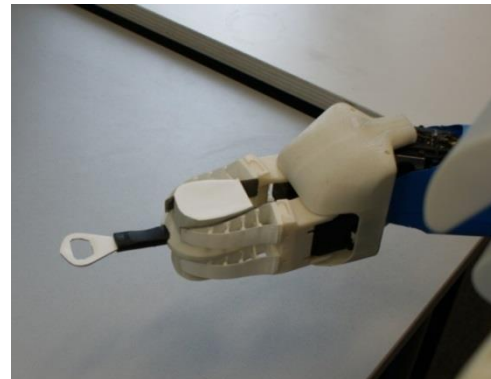
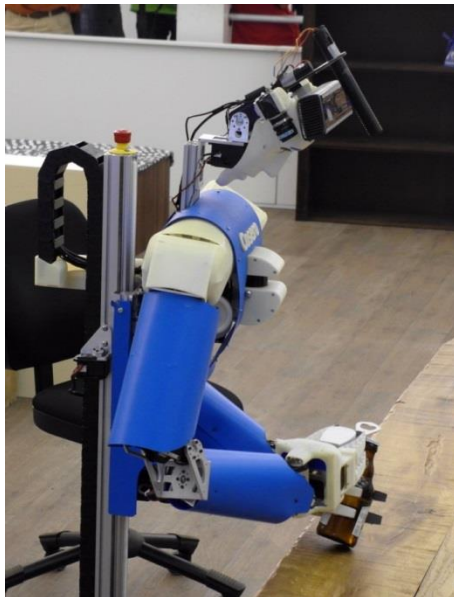
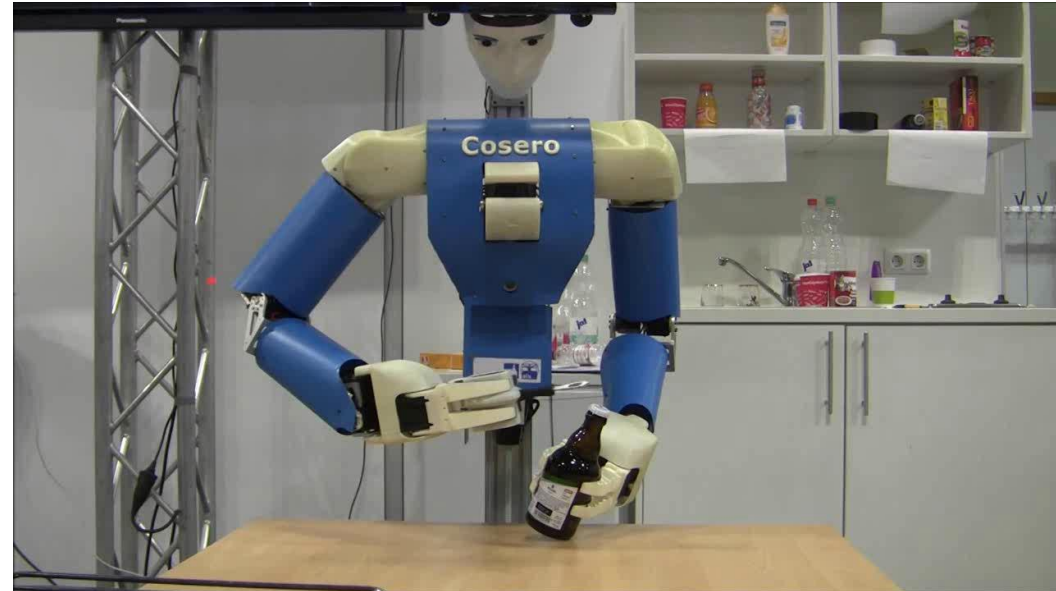
# Transfer von Handhabungswissen



- Demonstration beim RoboCup 2013

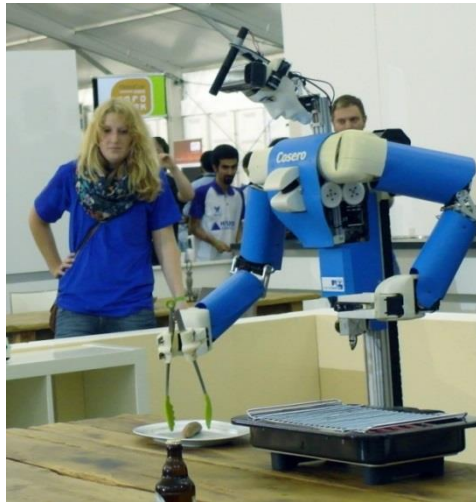
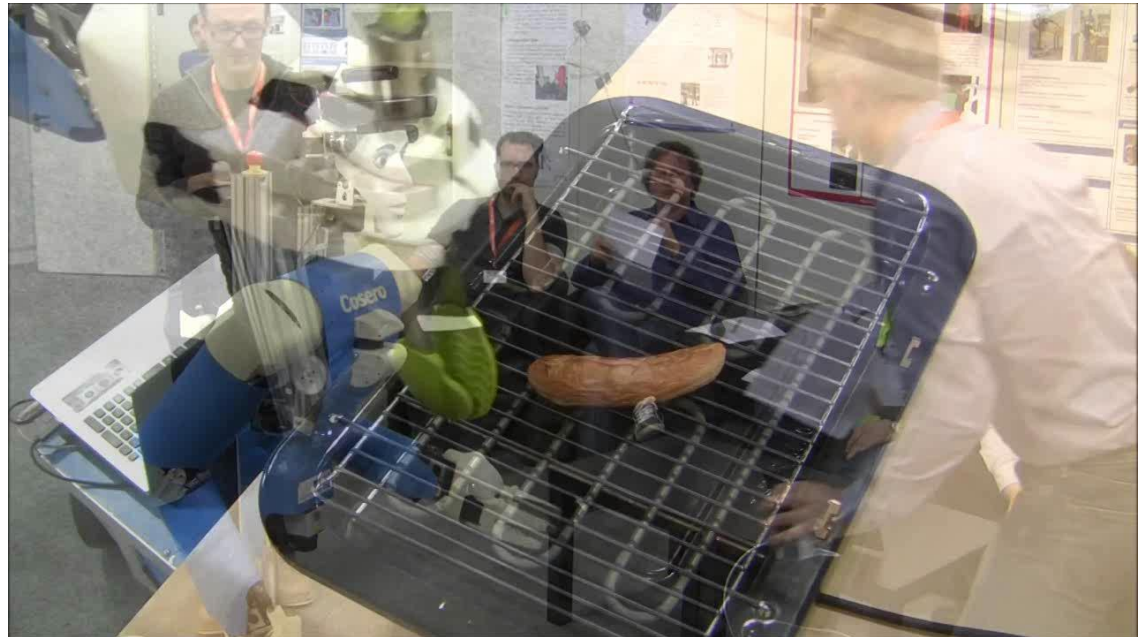
# Werkzeuggebrauch: Flaschenöffner

- Wahrnehmung von Werkzeugspitze und Kronkorken
- Erweiterung der Armkinematik
- Bewegungsanpassung



# Würstchenzange, Tablett

- Wahrnehmung von Zangenspitze und Würstchen
- Anpassung an Hauptachse des Würstchens



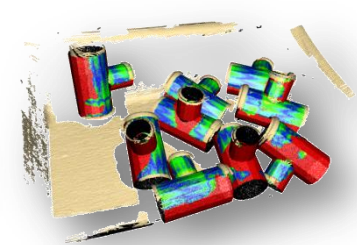
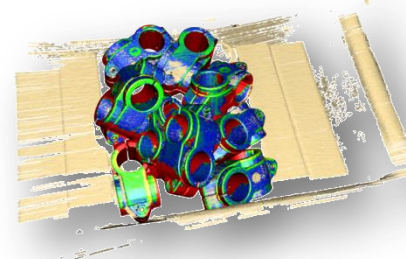
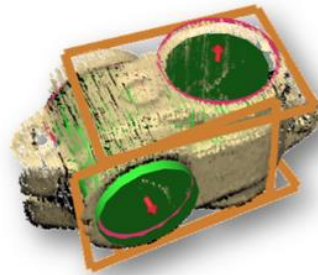
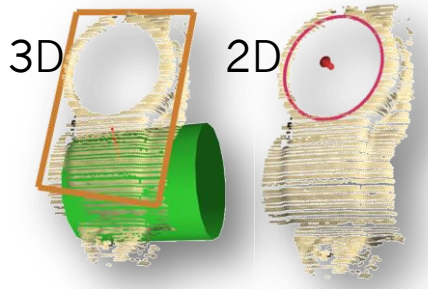
- Unser Team NimbRo hat drei Mal hintereinander die internationalen RoboCup@Home-Wettbewerbe gewonnen

# Griff in die Kiste

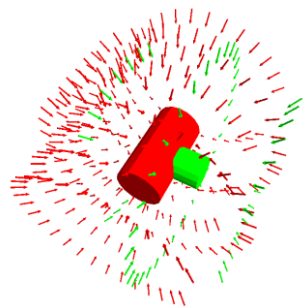
- Bekannte Objekte  
Kiste



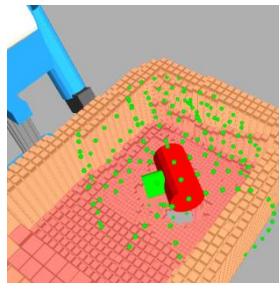
- Registrierung von 2D und 3D Formprimitiv-Graphen



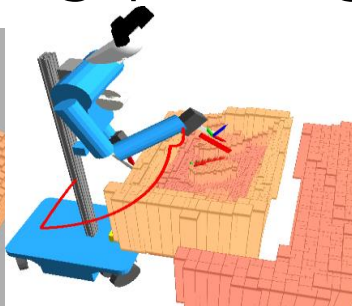
- Greif- und Bewegungsplanung



Offline

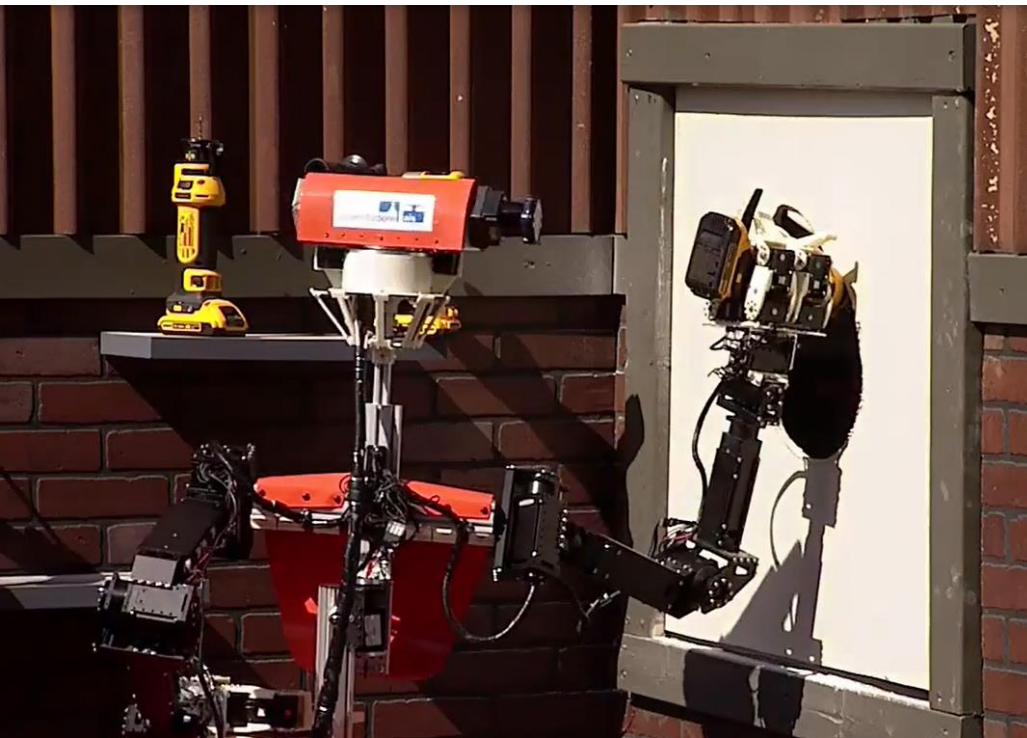


Online



# Rettungs- und Explorationsroboter

Mobiler Manipulationsroboter Momaro



[Schwarz et al.: Frontiers in Robotics and AI 2016, JFR 2017]

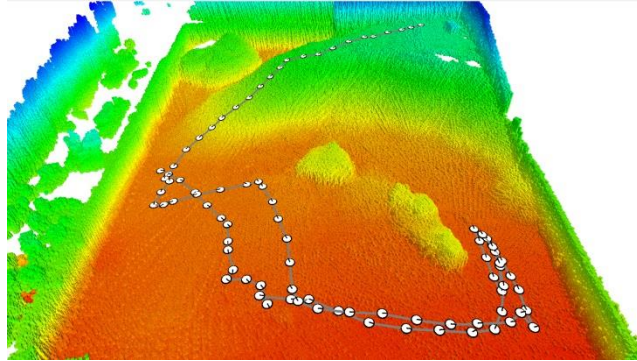
# DLR SpaceBot Camp 2015



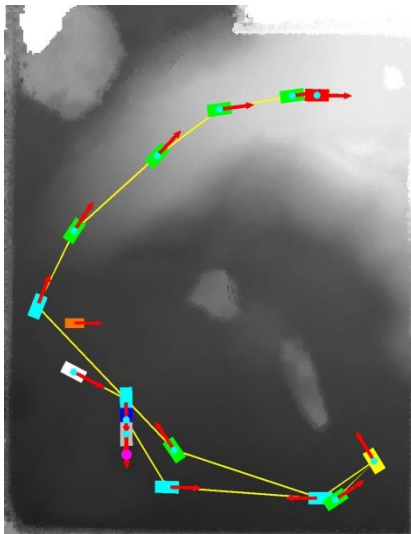
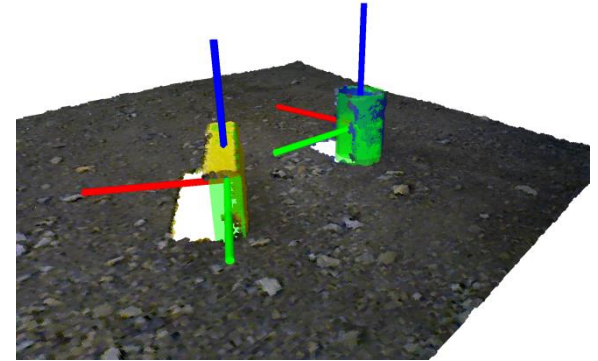
8X

# Autonome Missionsausführung

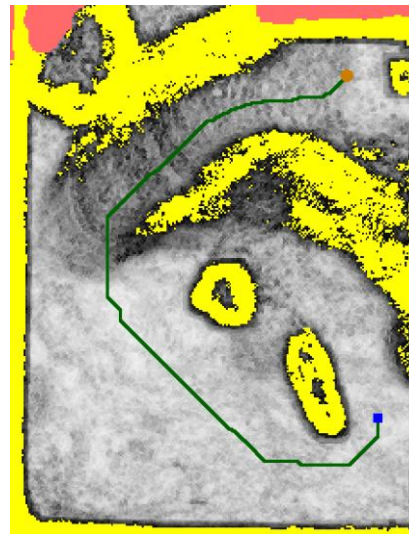
3D Mapping & Localization



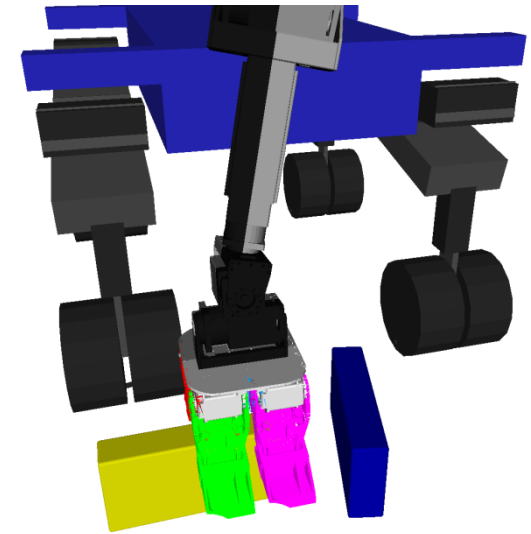
Object perception



Mission plan



Navigation plan

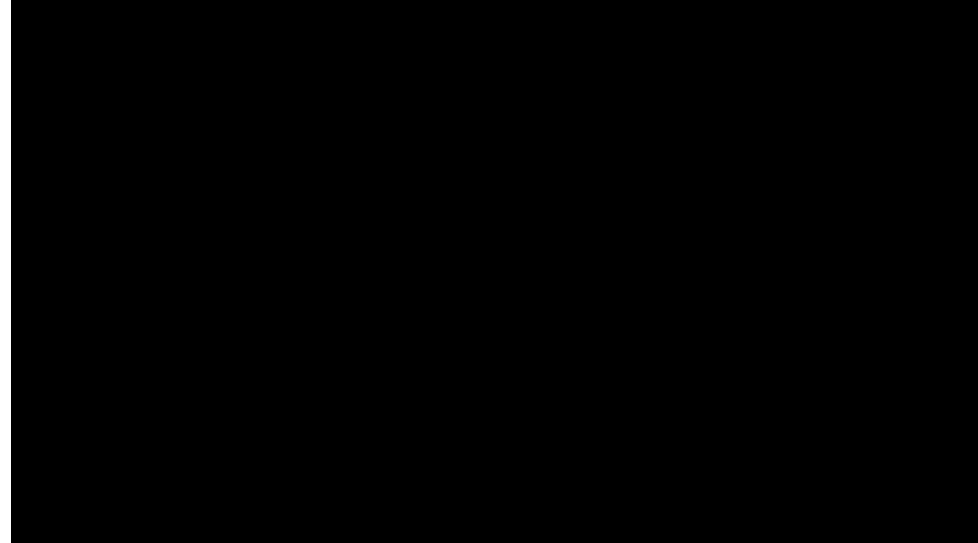


Grasping

[Schwarz et al., Frontiers in Robotics and AI 2016]

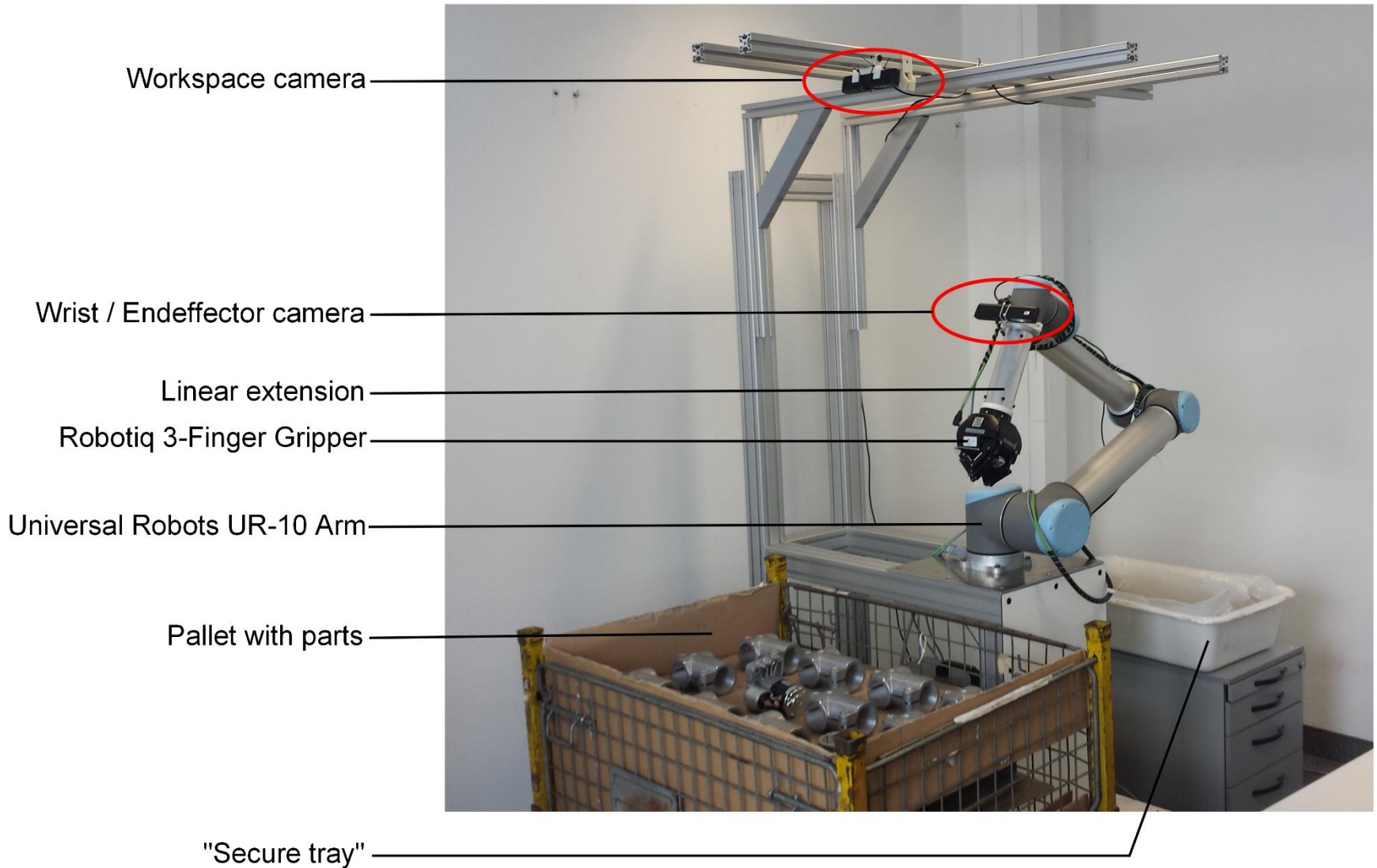
# Kitting in Automobillogistik

- Many car variants
- Collect the parts needed for the assembly of a particular car in a kit
- Parts in different variants are available in a supermarket
- Robot needs to
  - navigate to the transport boxes,
  - grasp the parts, and
  - place them in the kit



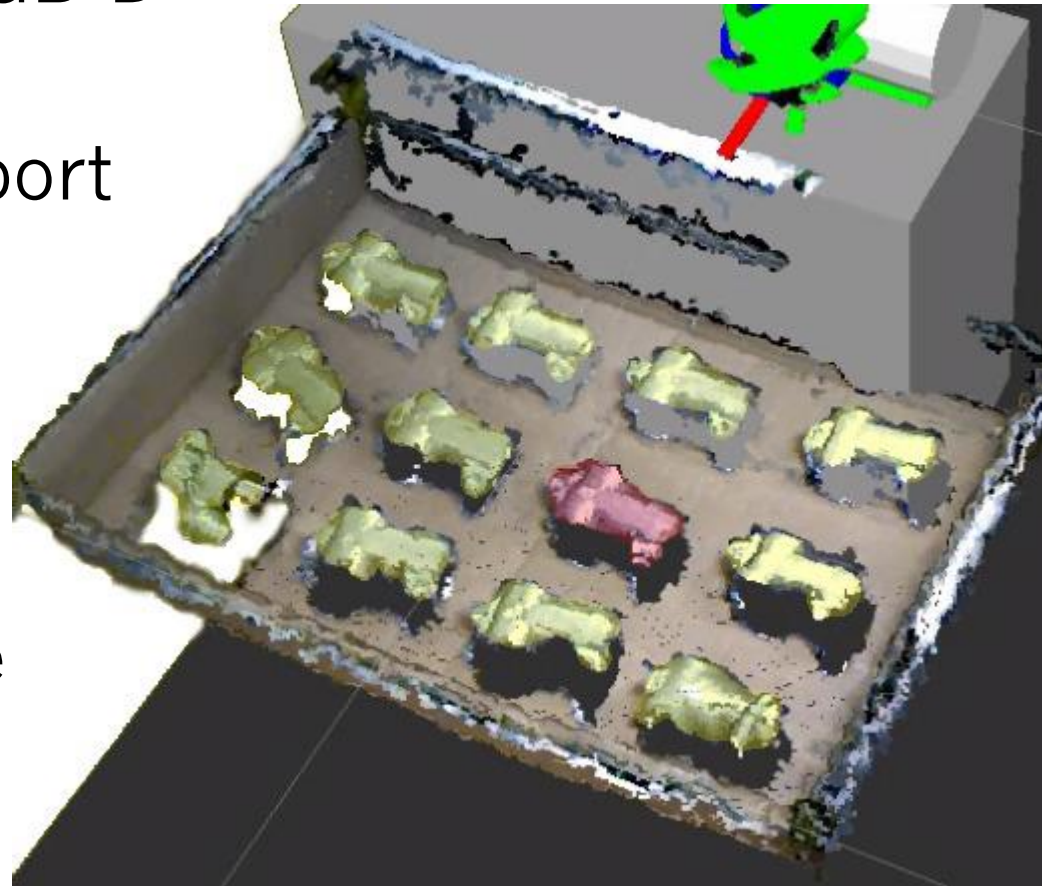


# Labor-Demonstrator



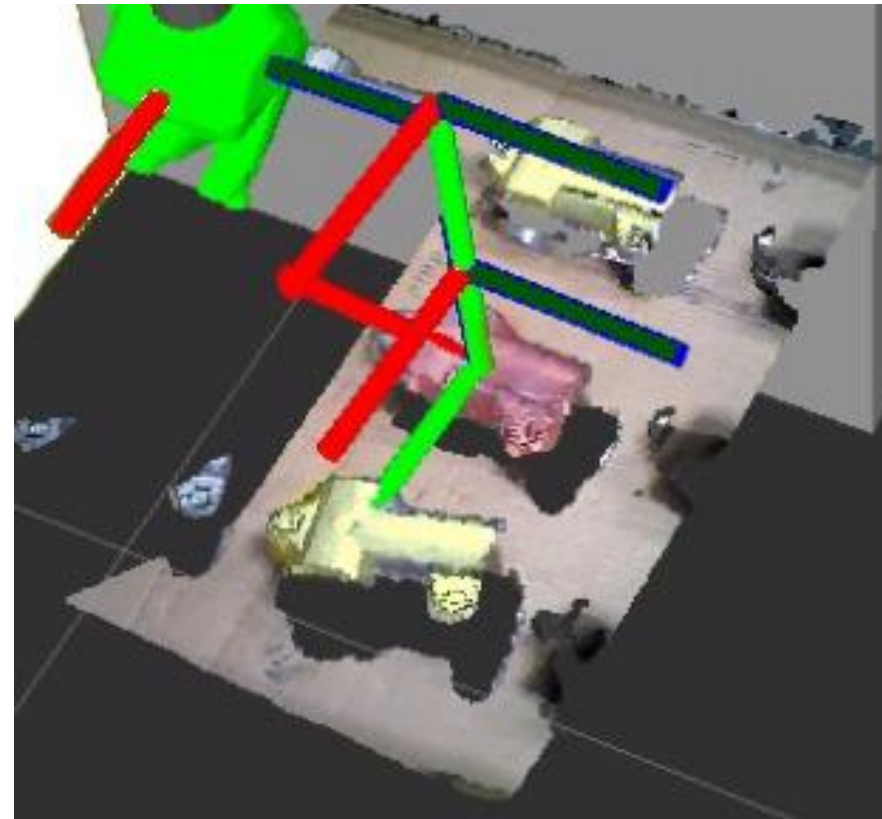
# Detektion von Teilen

- Using work space RGB-D camera
- Initial pose of transport box roughly known
- Detect dominant horizontal plane above ground
- Cluster points above support plane
- Estimate main axes



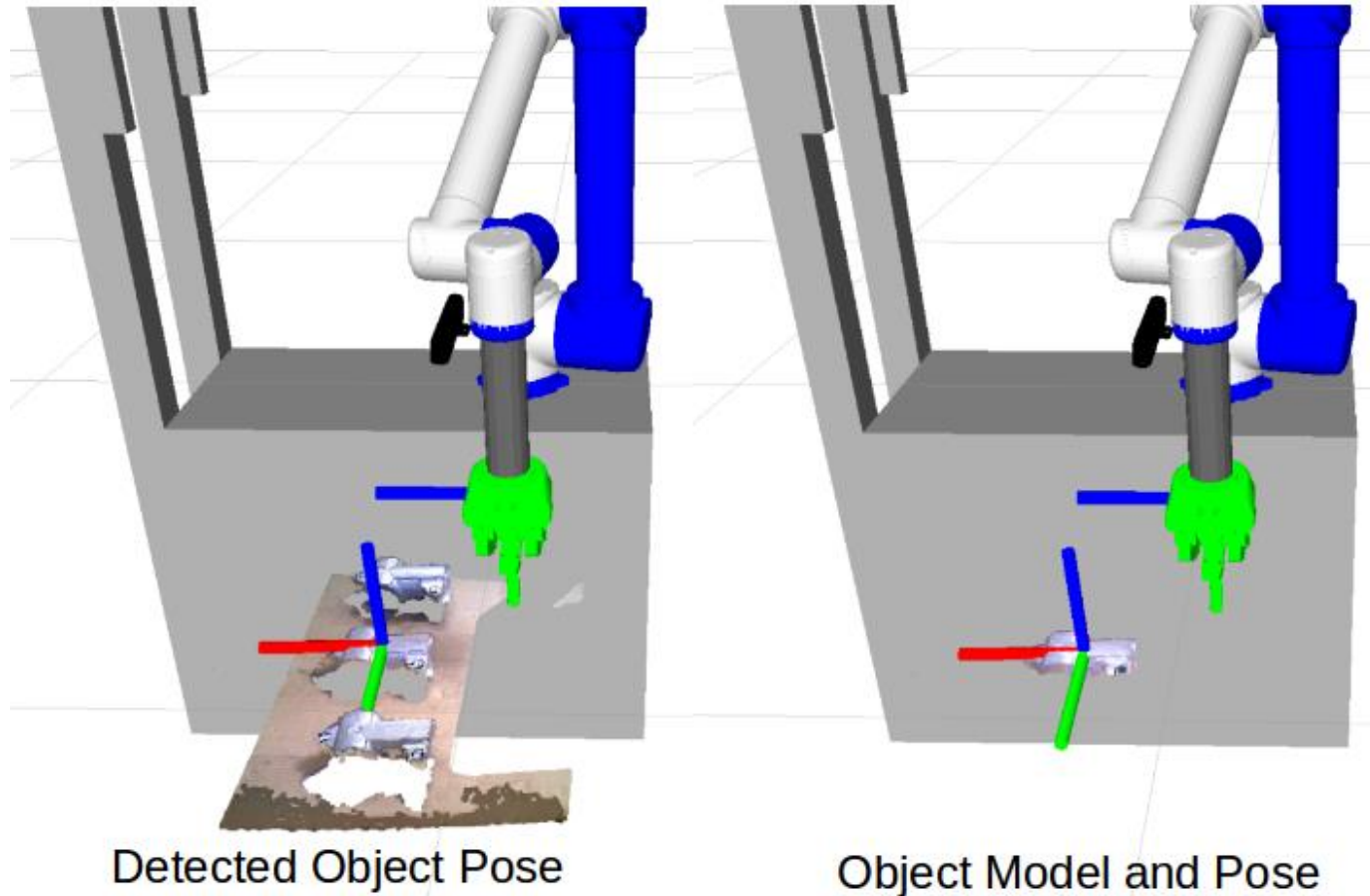
# Schätzung der Objektpose

- Wrist RGB-D camera moved above innermost object candidate
- Object views are represented as Multiresolution Surfel Map
- Registration of object view with current measurements using soft assignments
- Verification based on registration quality



# Registriertes Objektmodell

- Registration yields the object pose



# Definition von Griffen

- GUI for object model acquisition and grasp definition, relative to object model

The image shows a screenshot of the ObjectGraspEditor GUI on the left and a 3D visualization of a robot arm on the right. The GUI includes a table with the following data:

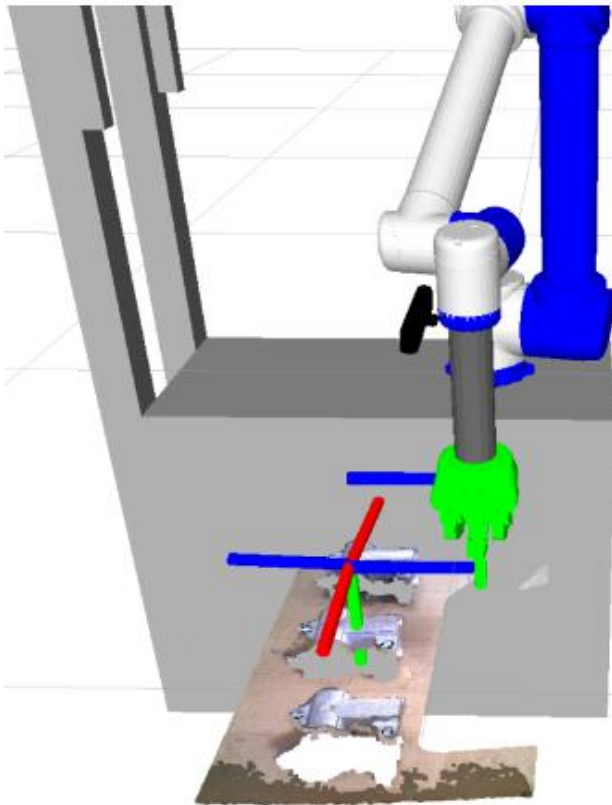
X	Y	Z	Roll	Yaw	Pitch	Grasp Type	
1	-0.009	0.003	0.128	-1.588	-0.009	1.557	BASIC
2	-0.008	-0.007	0.125	-1.550	0.006	-1.643	BASIC

Annotations in the image include:

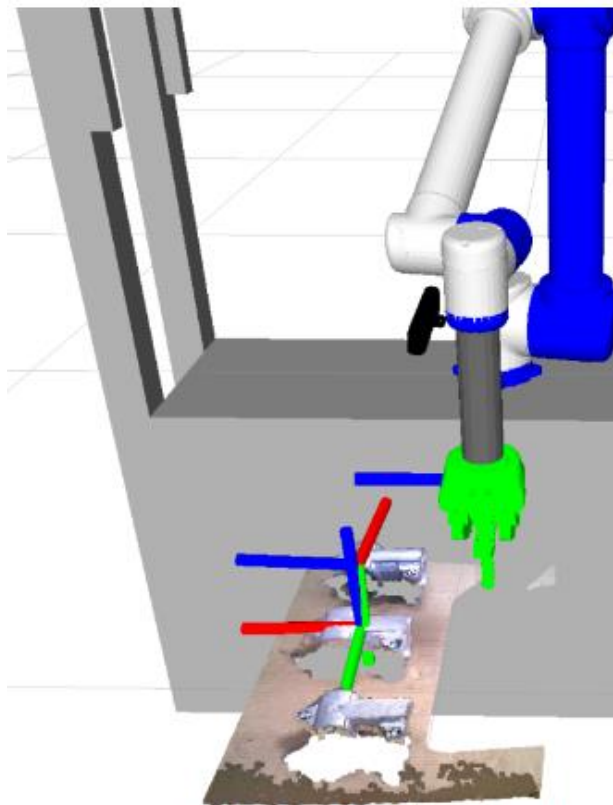
- Red arrows pointing to the 'Add Current Grasp' button and the second row of the table, with the text: **Save current robot pose as a new grasp pose**
- Red arrows pointing to the 'Load Object Axis' button and the second row of the table, with the text: **Find object axis**
- A red arrow pointing to the 'Take Snapshot' button, with the text: **Store an object snapshot**
- Black lines pointing to the green hand and the purple object in the 3D view, with the text: **Grasp pose** and **Detected object pose**

# Griff-Auswahl

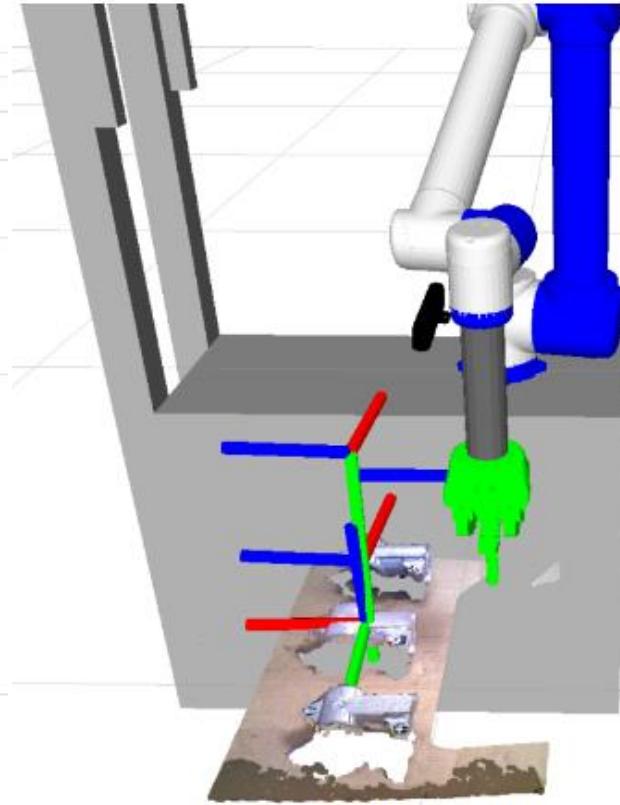
- Grasps are selected according to object pose



Grasp Candidates



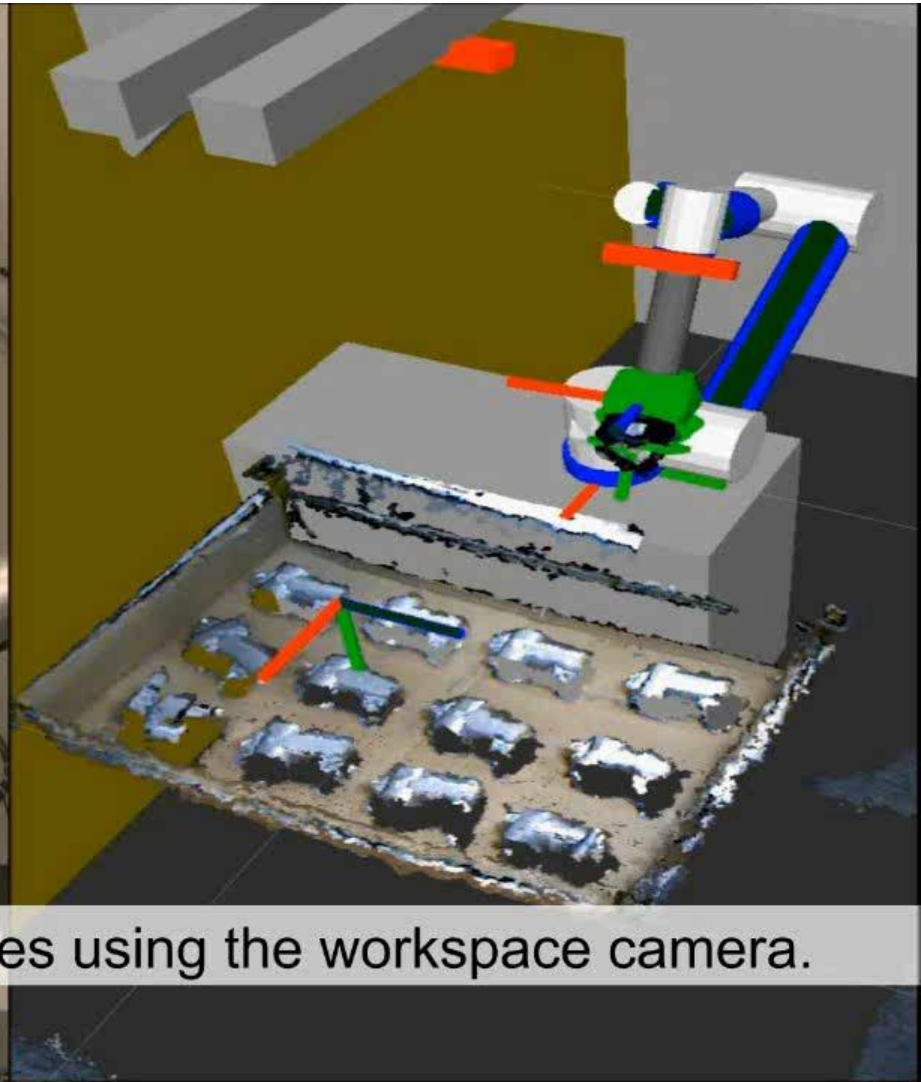
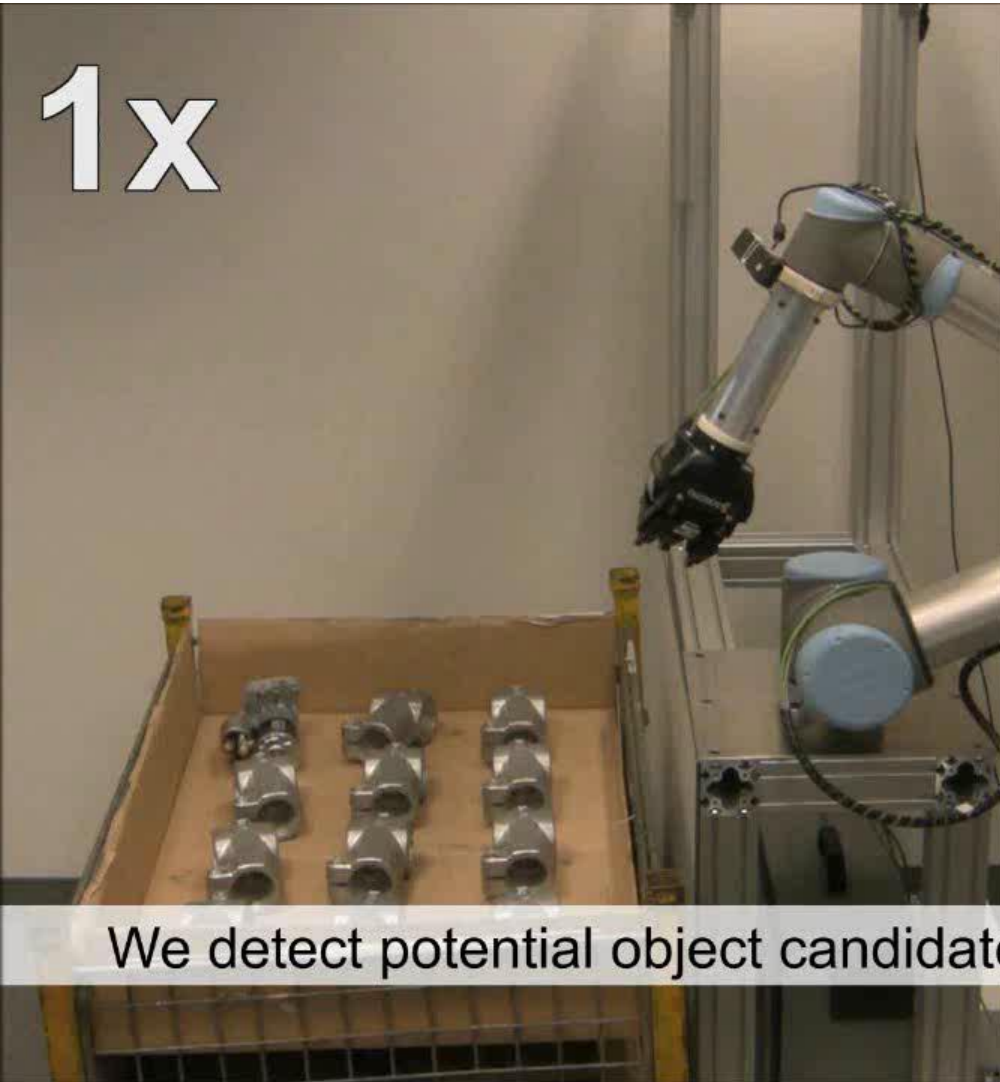
Selected Grasp and Object Pose



Pre-Grasp, Grasp and Object Pose

# Depalettierung von Teilen

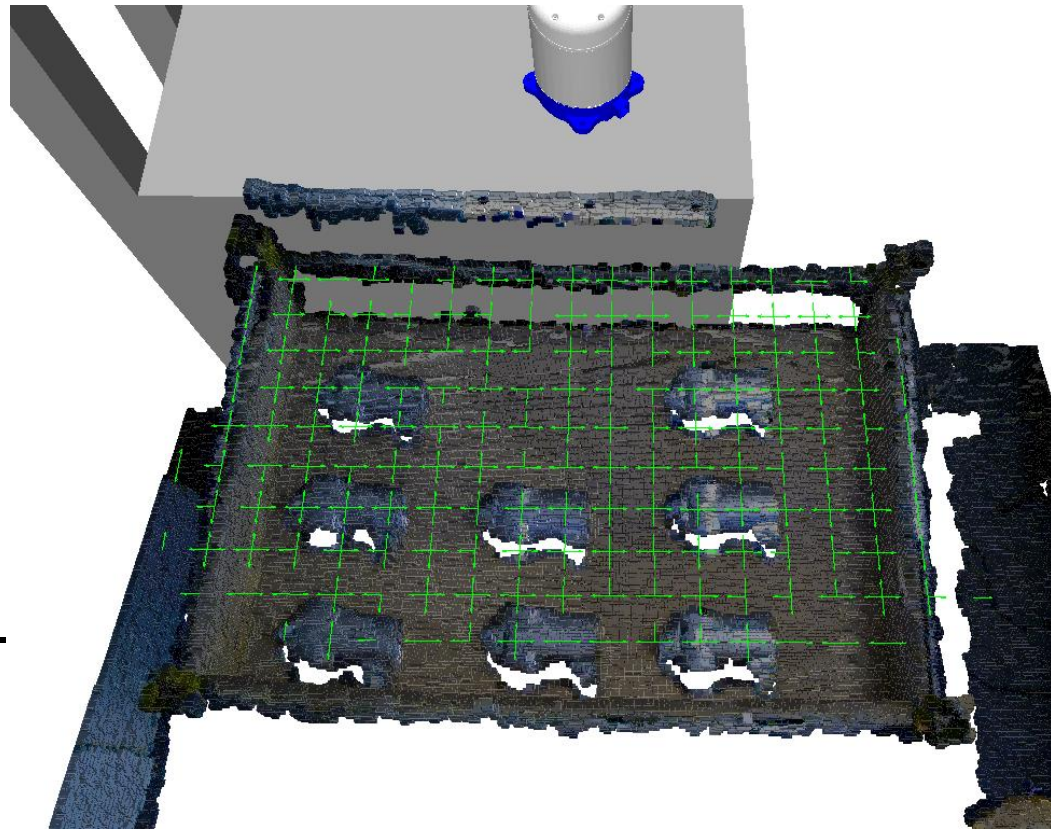
1x



We detect potential object candidates using the workspace camera.

# Bewegungsplanung

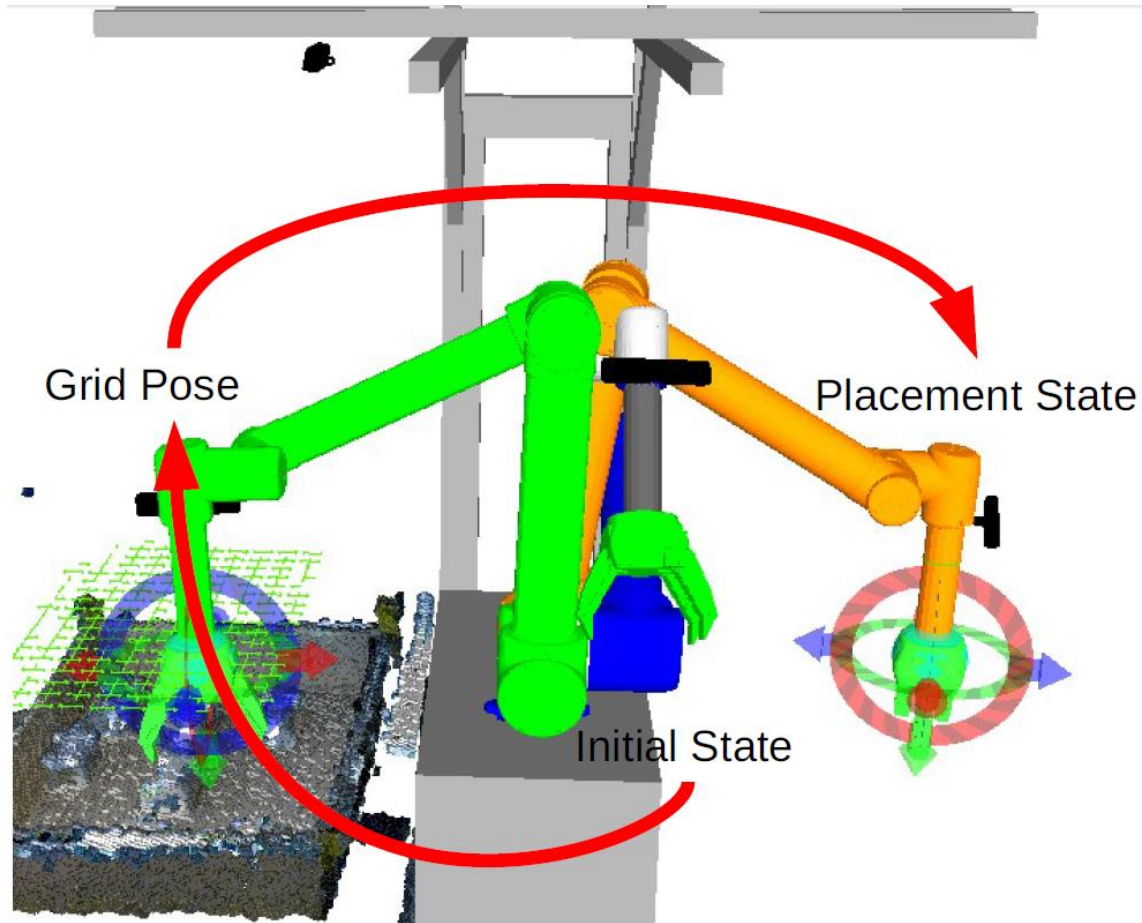
- Use ROS MoveIt for motion planning and execution
- Predefined poses (initial, placement) and grid of poses above the objects
- Preplanned paths
- Only short trajectories must be planned online





# Verbindung von Bewegungssegmenten

- Interpolation for smooth segment transitions



# Depallettierung von Anlassern



# Mobiler Manipulationsroboter

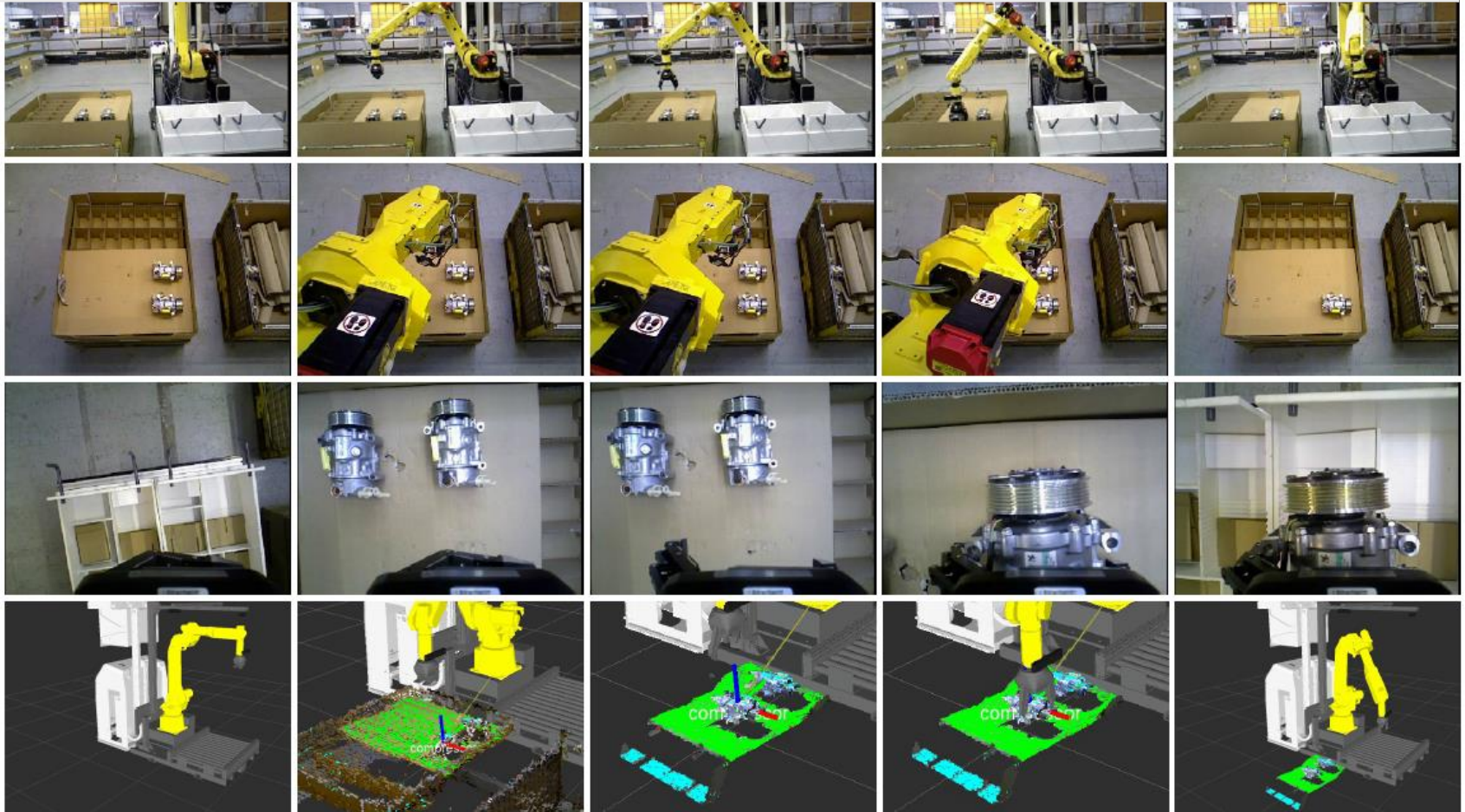
Initialization

Part  
detection

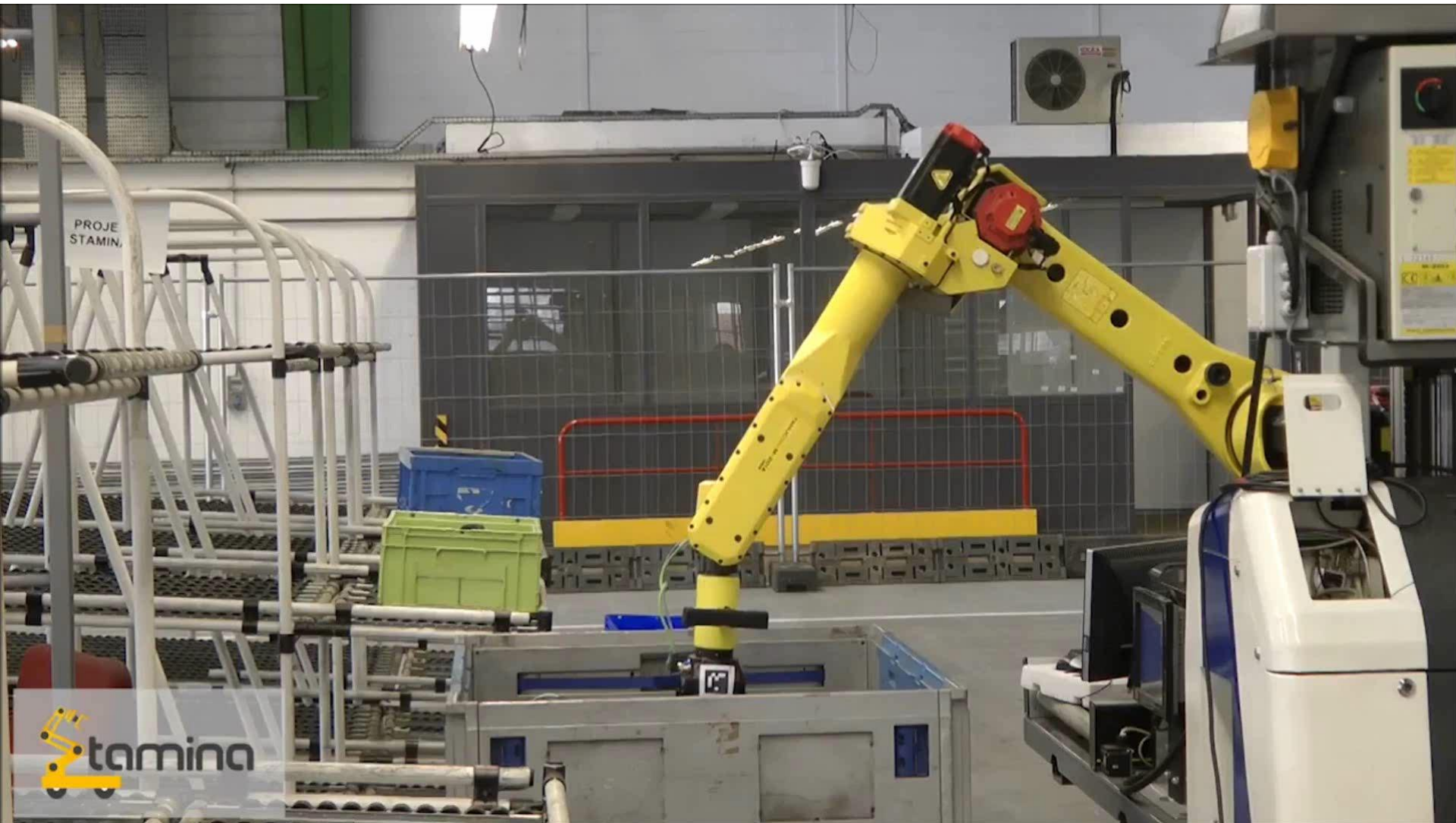
Approach

Grasping

Placing

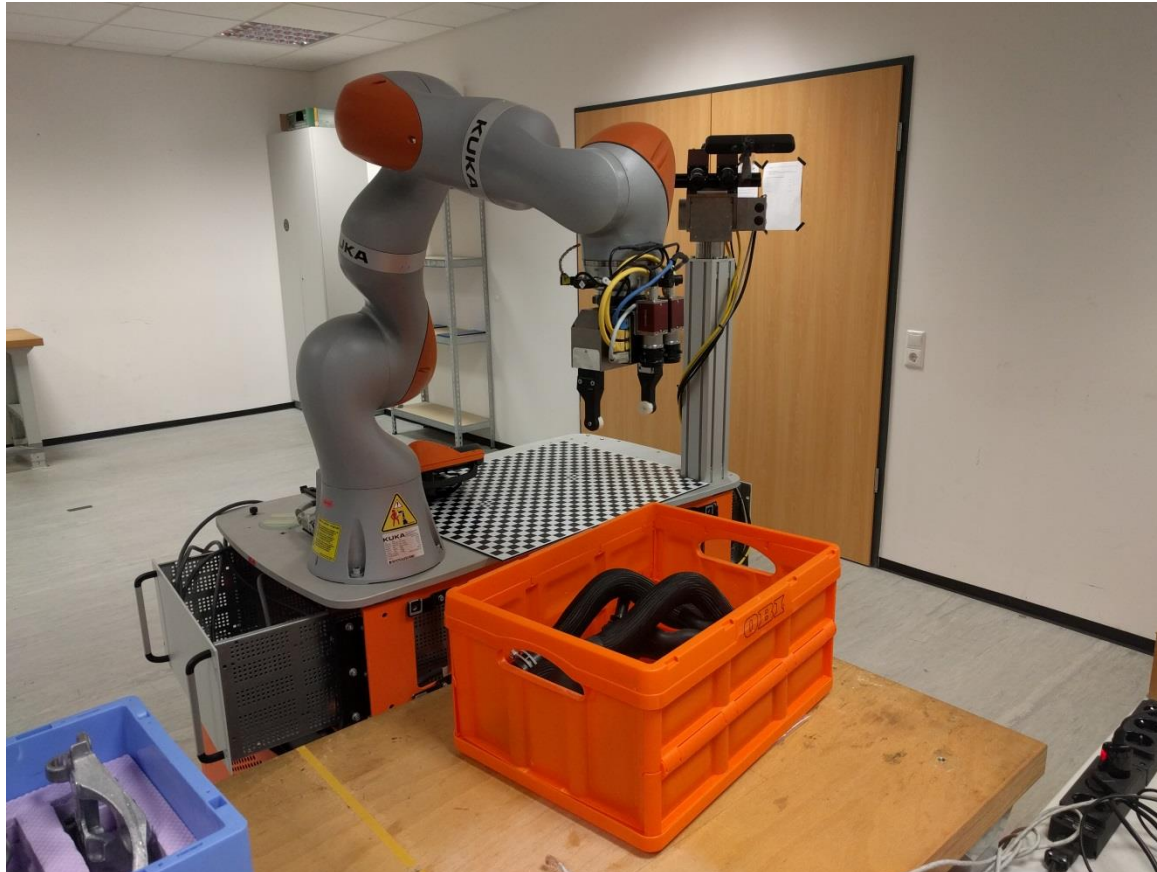


# STAMINA-Abschlussdemonstration



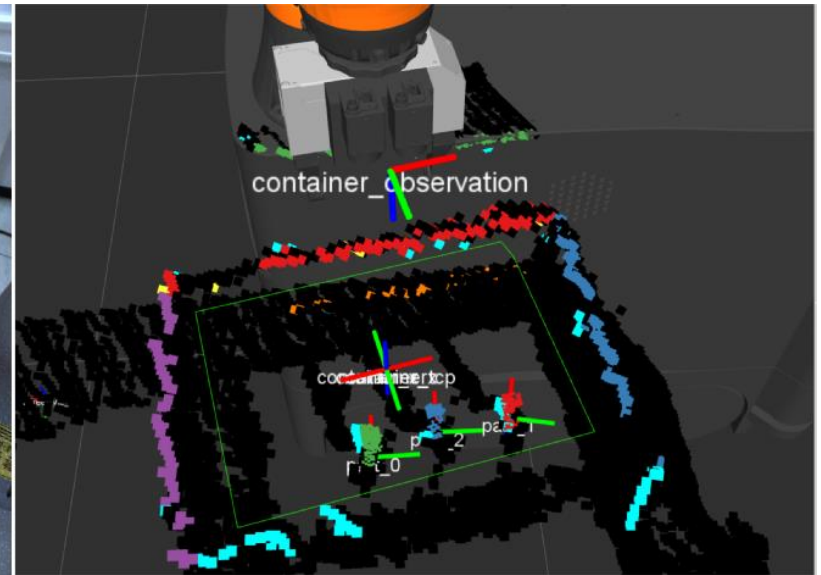
# EuRoC Challenge 2: KittingBot

- Kleinerer Roboter: Kuka miiwa
- Zusammenarbeit mit Menschen



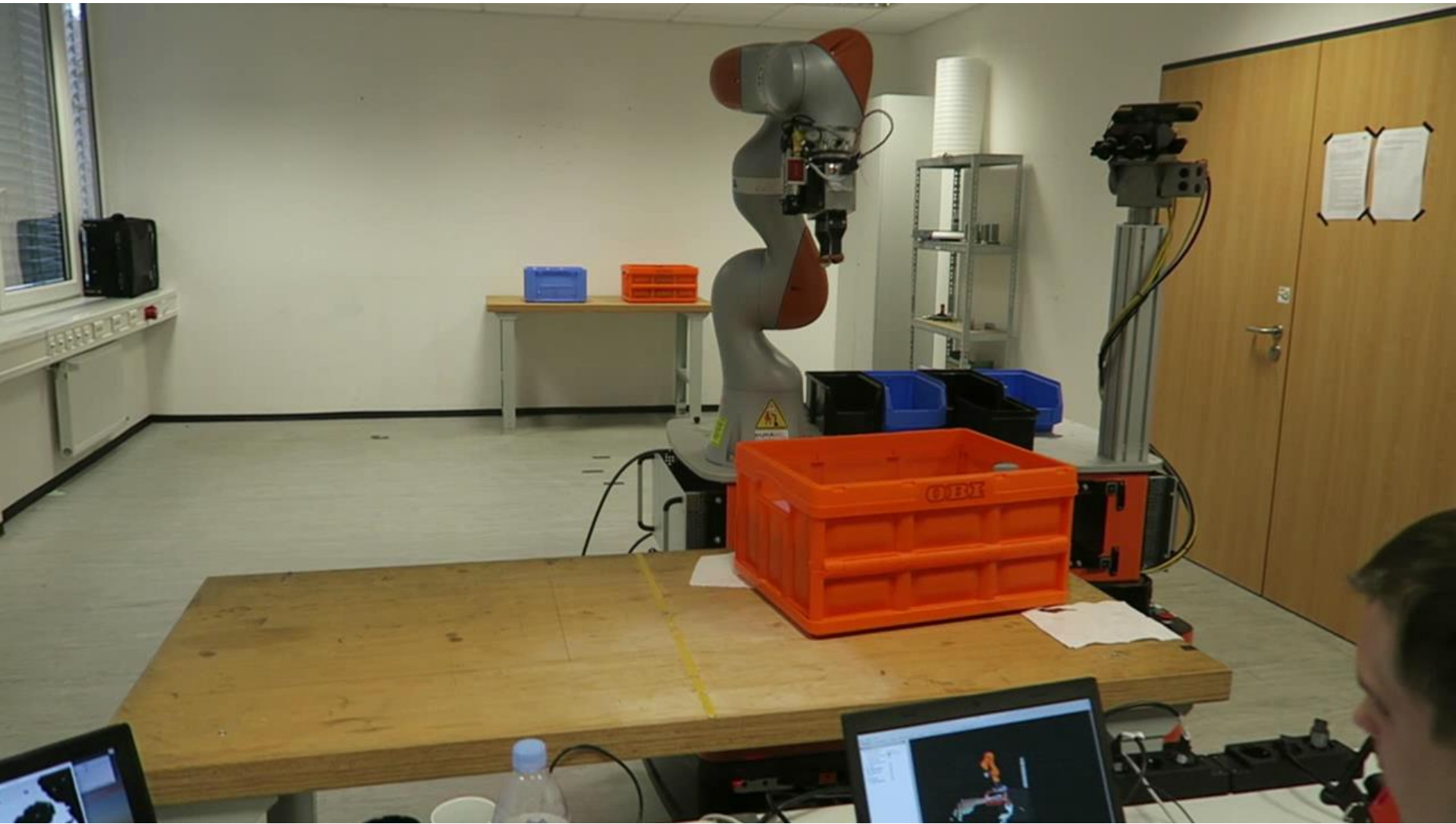
# Detektion der Kiste und Posenschätzung

- Detektion von Kanten im Tiefenbild
- Gruppierung zu Linien
- Posenschätzung für oberen Kistenrand



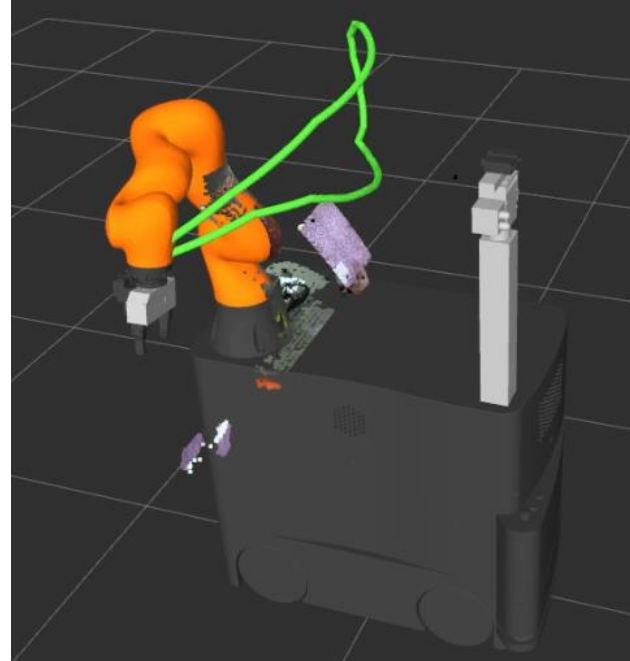
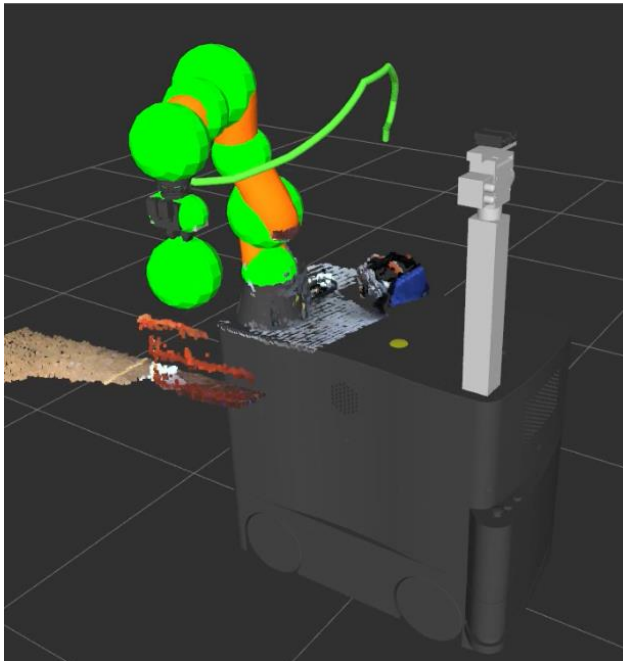
[Holz und Behnke: ISR 2016]

# KittingBot Showcase-Demonstration



# Online-Trajektorienoptimierung

- Kombination mehrerer Ziele
  - Hindernisvermeidung
  - Ausführungsdauer
  - Drehmomente



[Pavlichenko and Behnke: IROS 2017]



# KittingBot: Hindernisvermeidung

Factory workers might be operating in the  
vicinity of the arm

-> we need to detect collisions and re-plan to avoid them

# KittingBot Showcase-Demonstration

## Kitting Demonstrator

Engine Support 2  
9808515080



Engine Support 1  
9672950980



Engine pipe

# KittingBot Showcase-Demonstration

## Kitting Demonstrator

Engine Support 2  
9808515080

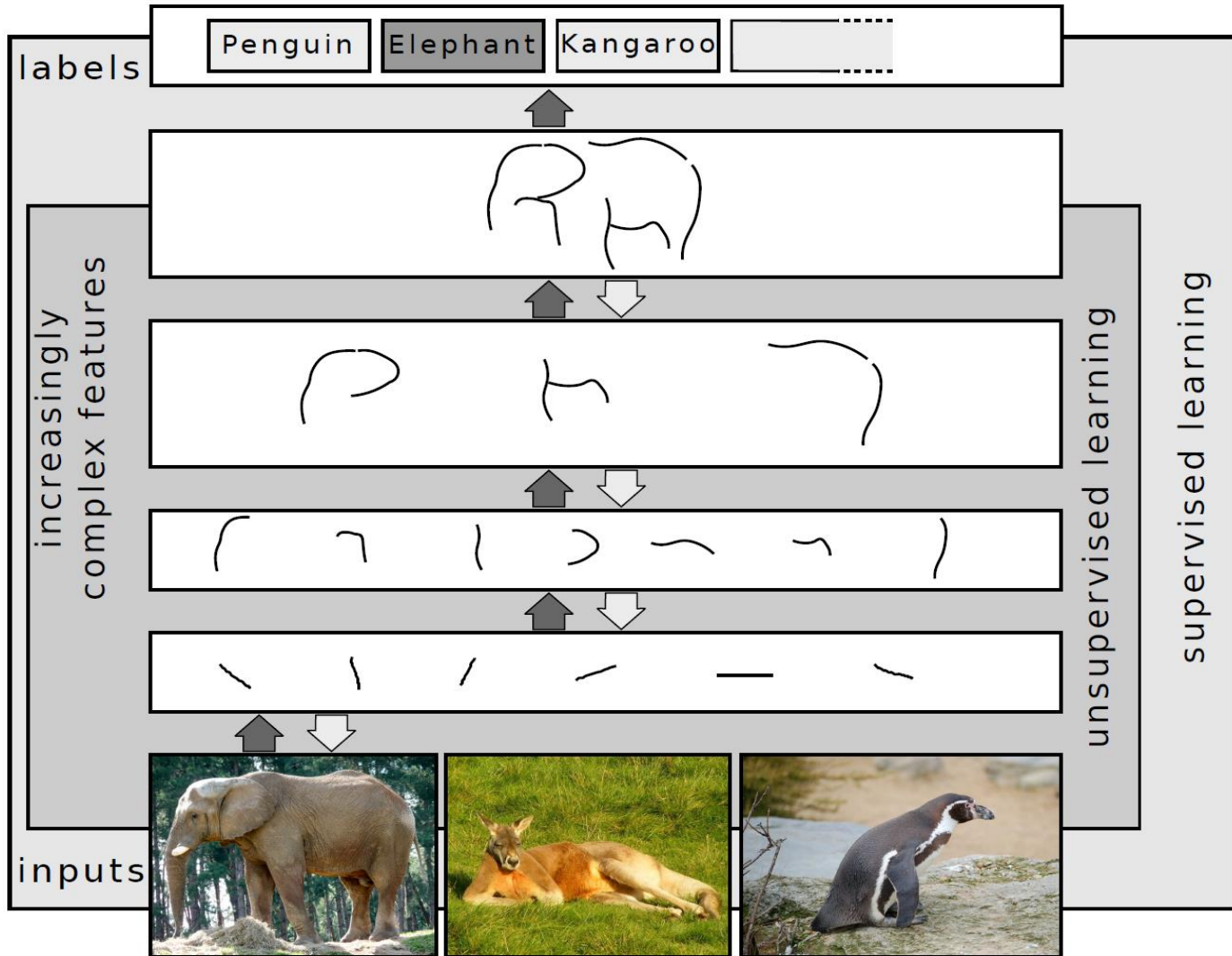


Engine Support 1  
9672950980



Engine pipe

# Deep Learning



[Schulz and Behnke, KI 2012]

# Objektkategorisierung: Besser als Menschen



GT: horse cart  
1: horse cart  
 2: minibus  
 3: oxcart  
 4: stretcher  
 5: half track



GT: birdhouse  
1: birdhouse  
 2: sliding door  
 3: window screen  
 4: mailbox  
 5: pot



GT: forklift  
1: forklift  
 2: garbage truck  
 3: tow truck  
 4: trailer truck  
 5: go-kart



GT: letter opener  
 1: drumstick  
 2: candle  
 3: wooden spoon  
 4: spatula  
 5: ladle



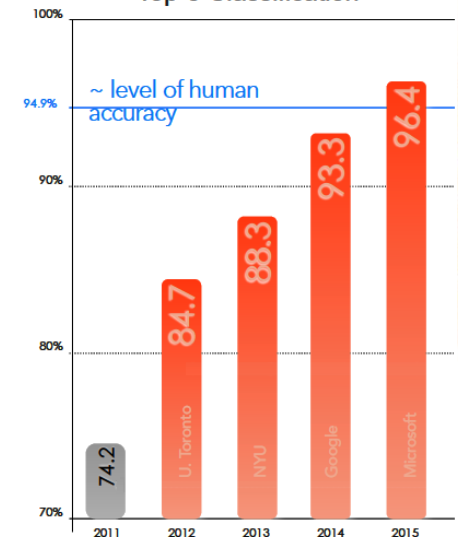
GT: coucal  
1: coucal  
 2: indigo bunting  
 3: lorikeet  
 4: walking stick  
 5: custard apple



GT: komondor  
1: komondor  
 2: patio  
 3: llama  
 4: mobile home  
 5: Old English sheepdog



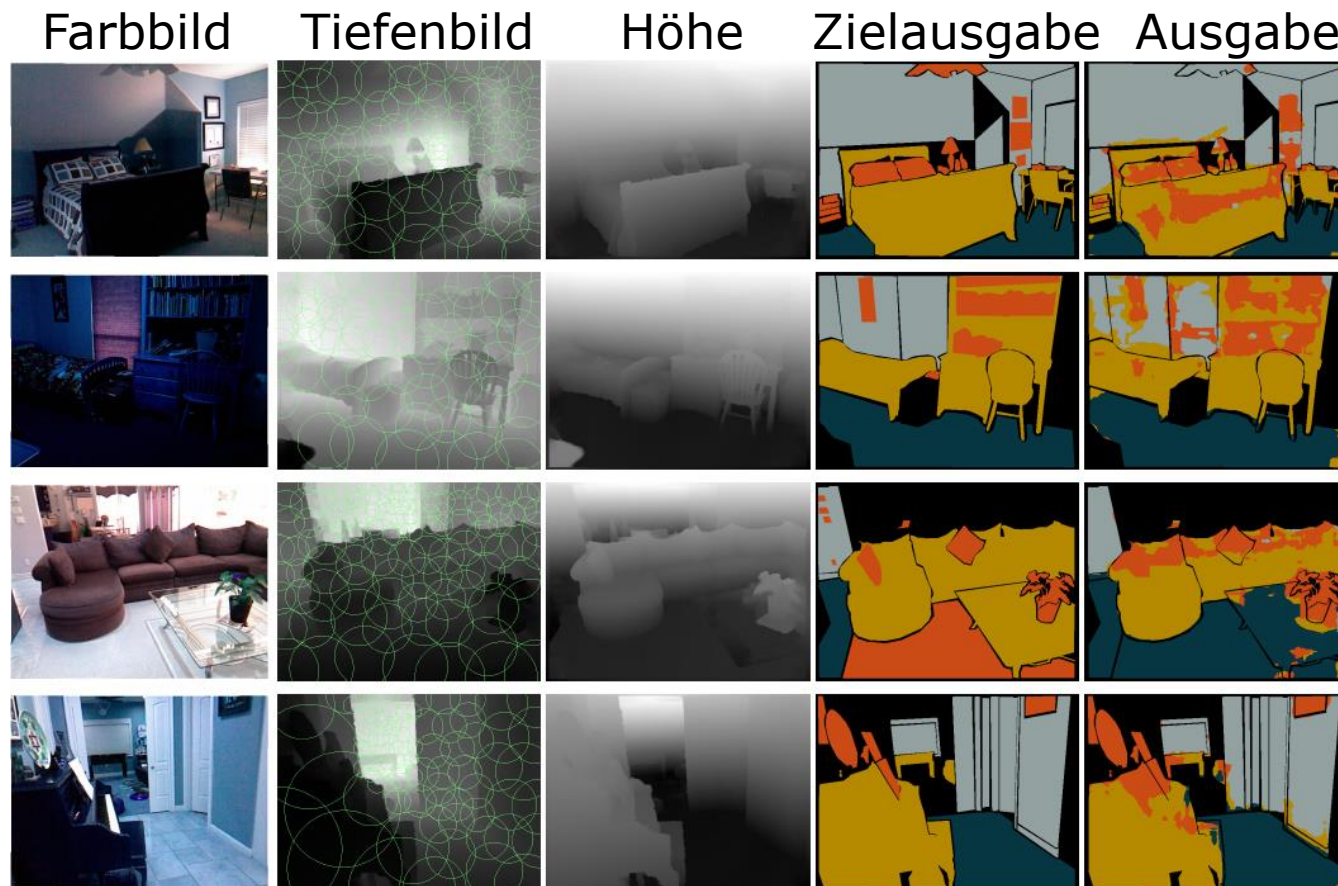
GT: yellow lady's slipper  
1: yellow lady's slipper  
 2: slug  
 3: hen-of-the-woods  
 4: stinkhorn  
 5: coral fungus



[He et al. 2015]

# Kategorisierung von Oberflächen

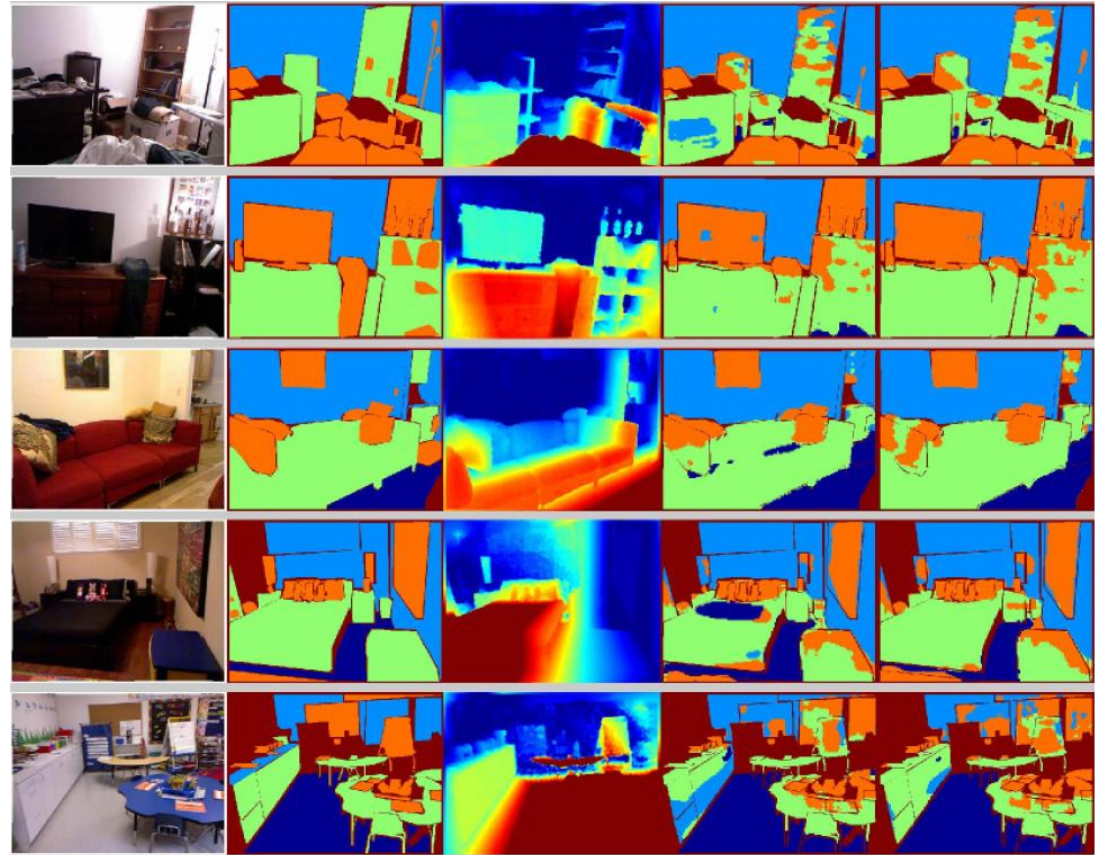
- RGB-D-Eingaben, Höhe über Boden geschätzt
- Skalierung der Eingabe mit der Tiefe



# Geometrische und Semantische Merkmale für RGB-D Objektklassensegmentierung

- Neues **geometrisches** Merkmal: Wandabstand
- **Semantische** Merkmale vortrainiert aus ImageNet
- Beide helfen signifikant

[Husain et al. RA-L 2016]



RGB

Truth

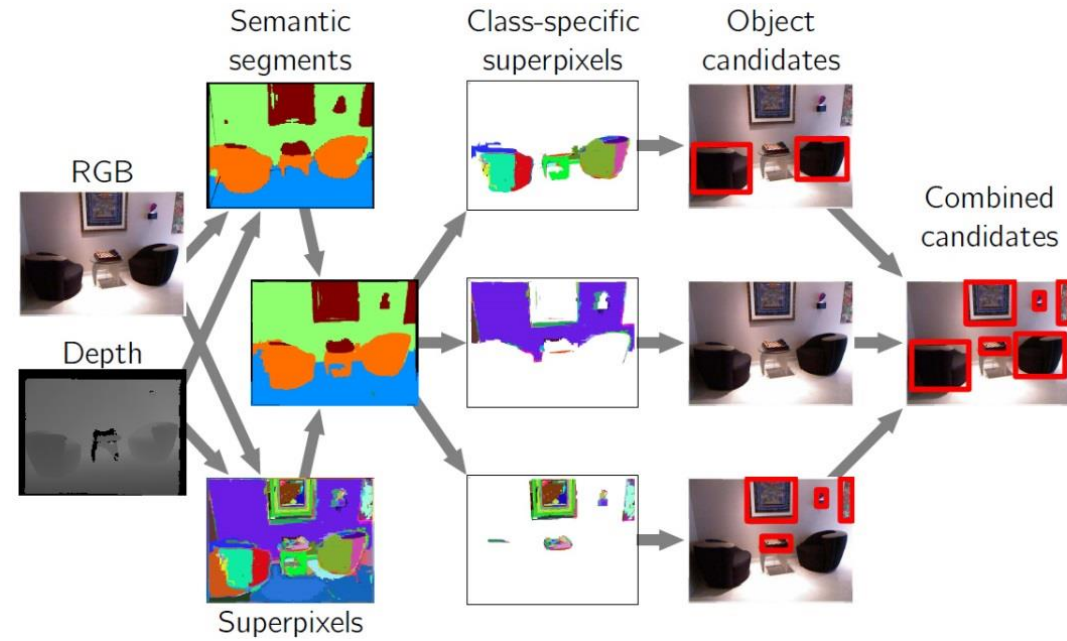
DistWall

OutWO

OutWithDist

# Semantische Segmentierung als Prior für Objektdetektion

- Combine bottom-up object discovery and semantic priors
- Semantic segmentation used to classify color and depth superpixels
- Higher recall, more precise object borders

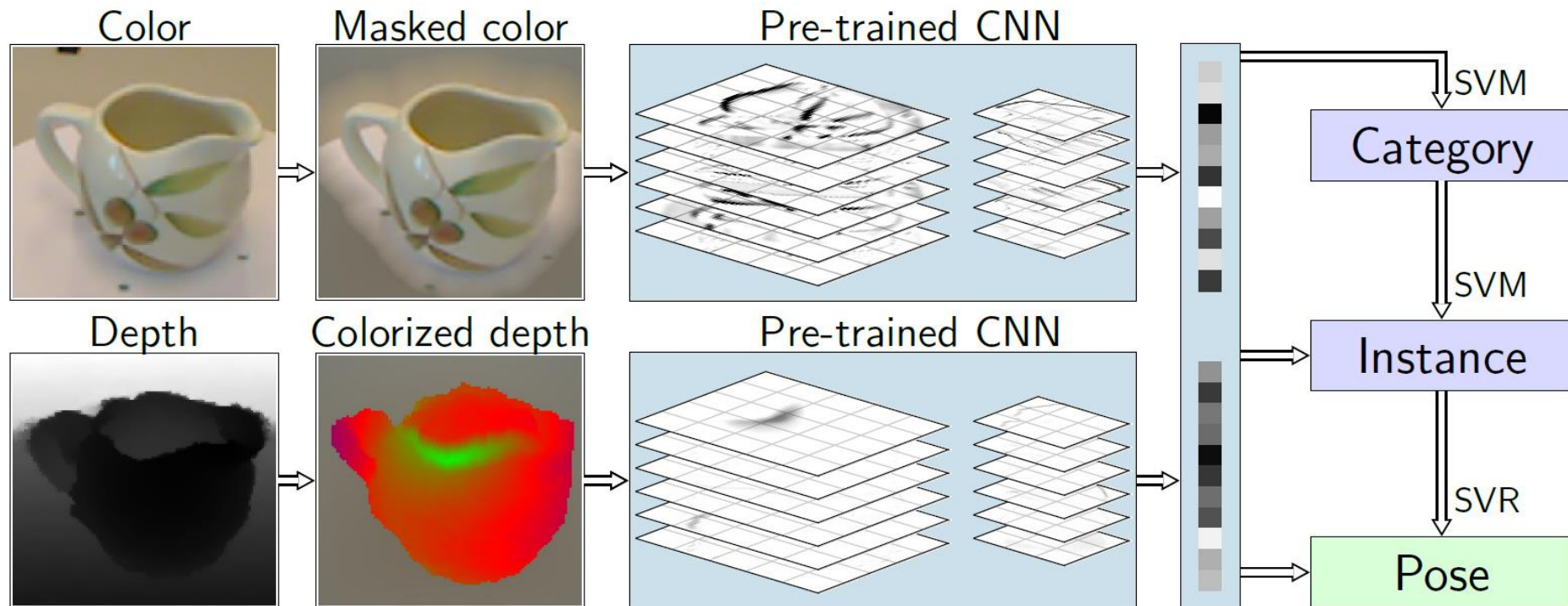


[Garcia et al. ICPR 2016]



# RGB-D Objekterkennung und Posenschätzung

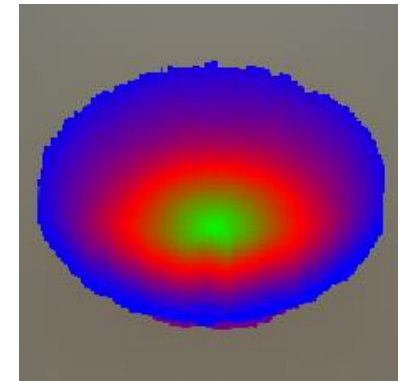
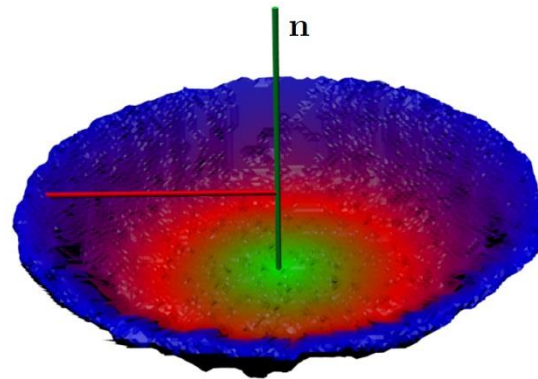
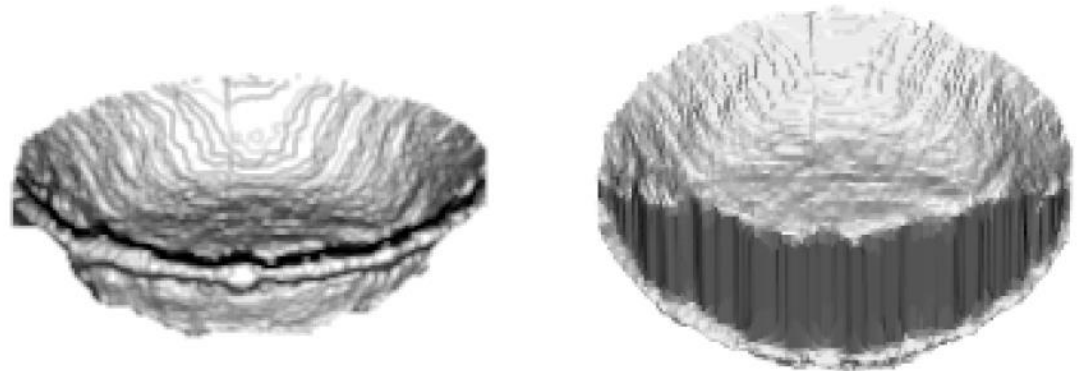
- Vortrainierte Merkmale von ImageNet



[Schwarz, Schulz, Behnke, ICRA2015]

# Kanonische Ansicht, Einfärbung

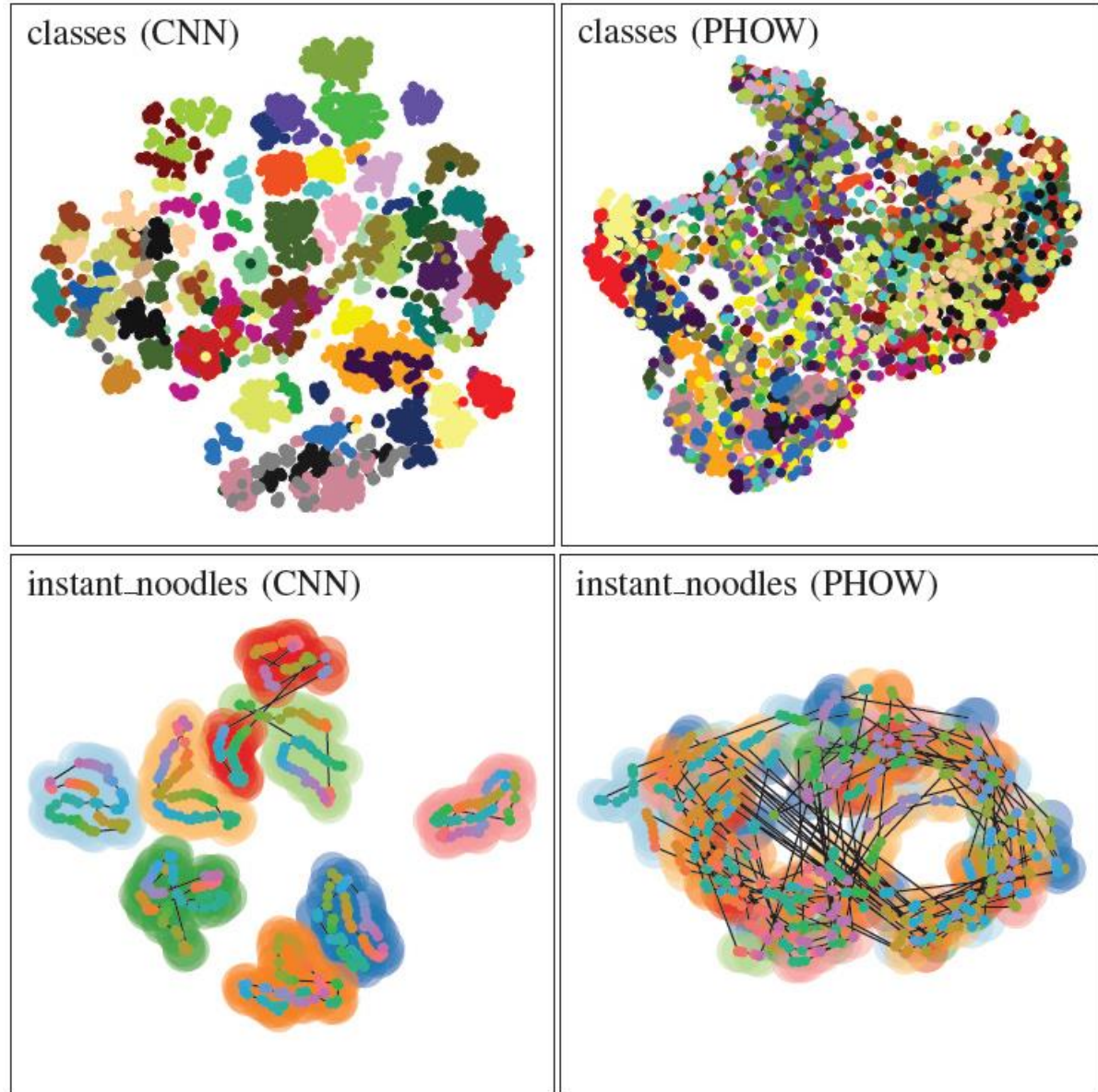
- Objektansichten aus verschiedenen Höhen
- Rendern einer kanonischen Ansicht
- Einfärbung anhand Distanz von Mittelachse



[Schwarz, Schulz, Behnke, ICRA2015]

# Merkmale Entwirren die Daten

- t-SNE Einbettung



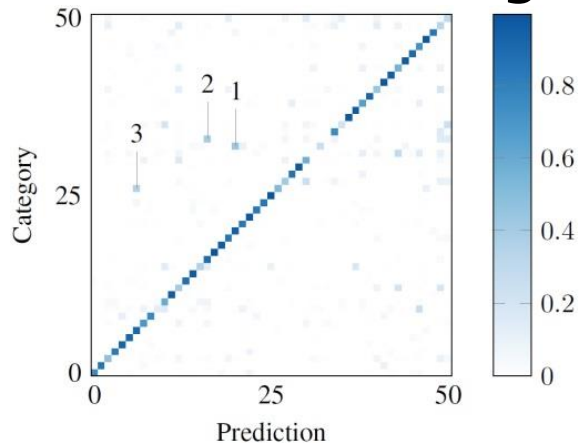
[Schwarz, Schulz,  
Behnke, ICRA2015]

# Erkennungsrate

## ■ Verbesserte Kategorie- und Instanzerkennung

Method	Category Accuracy (%)		Instance Accuracy (%)	
	RGB	RGB-D	RGB	RGB-D
Lai <i>et al.</i> [1]	74.3 ± 3.3	81.9 ± 2.8	59.3	73.9
Bo <i>et al.</i> [2]	82.4 ± 3.1	87.5 ± 2.9	<b>92.1</b>	92.8
PHOW[3]	80.2 ± 1.8	—	62.8	—
<b>Ours</b>	<b>83.1 ± 2.0</b>	88.3 ± 1.5	92.0	<b>94.1</b>
<b>Ours</b>	<b>83.1 ± 2.0</b>	<b>89.4 ± 1.3</b>	92.0	<b>94.1</b>

## ■ Verwechslungen



1: Kännchen / Kaffetasse

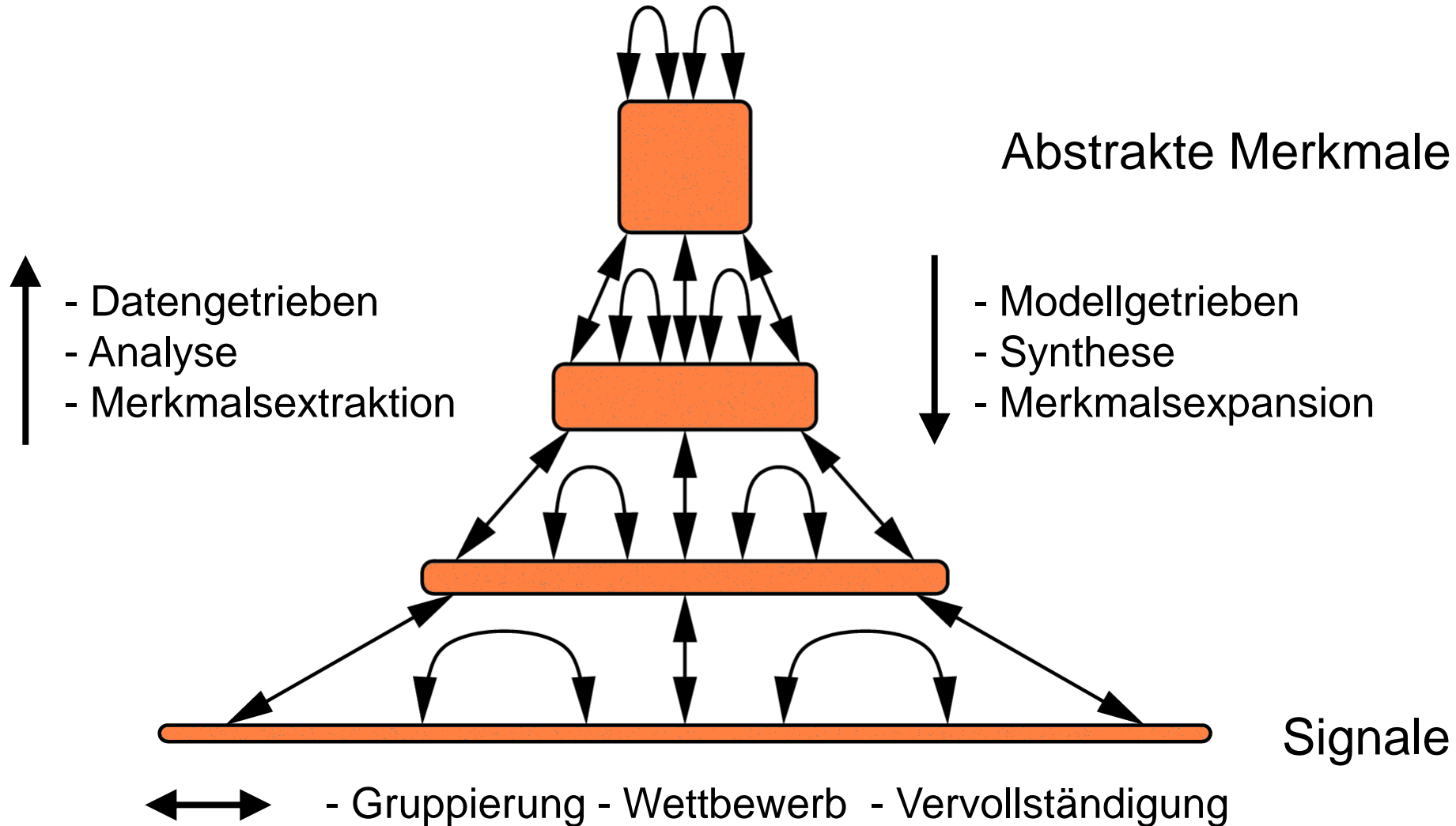


2: Pfirsich / Schwamm



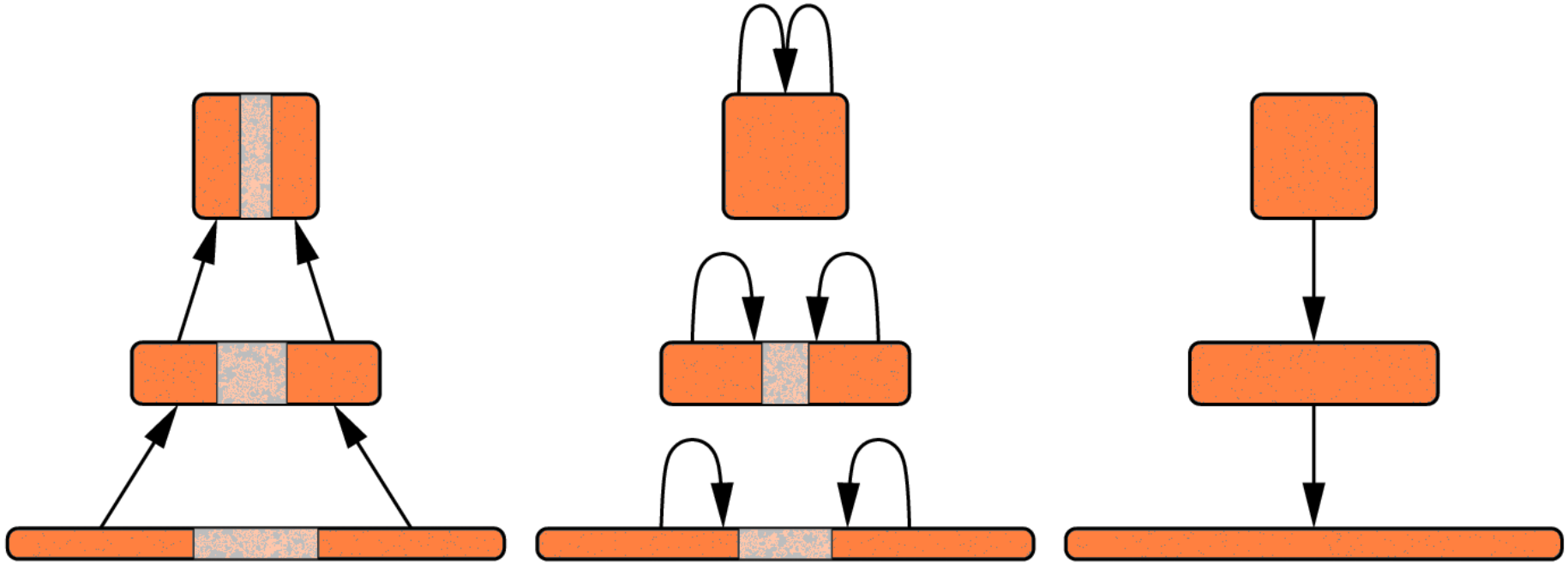
[Schwarz, Schulz, Behnke, ICRA2015]

# Neurale Abstraktionspyramide



# Iterative Interpretation

- Interpretiere die einfachsten Stellen zuerst



- Nutze Teil-Interpretationen als Kontext zur Auflösung von Mehrdeutigkeiten

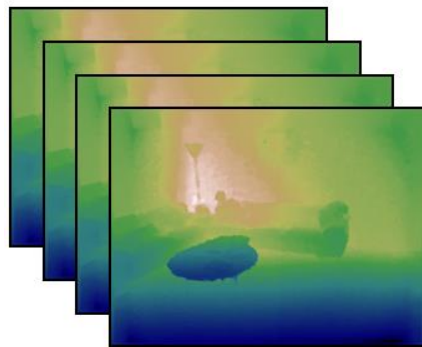
# Objektklassensegmentierung von RGB-D-Video

- Eingabe: RGB-D-Video (NYU Depth V2)

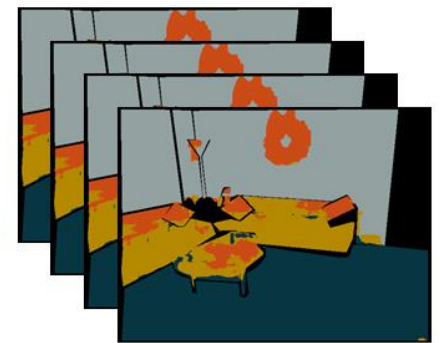
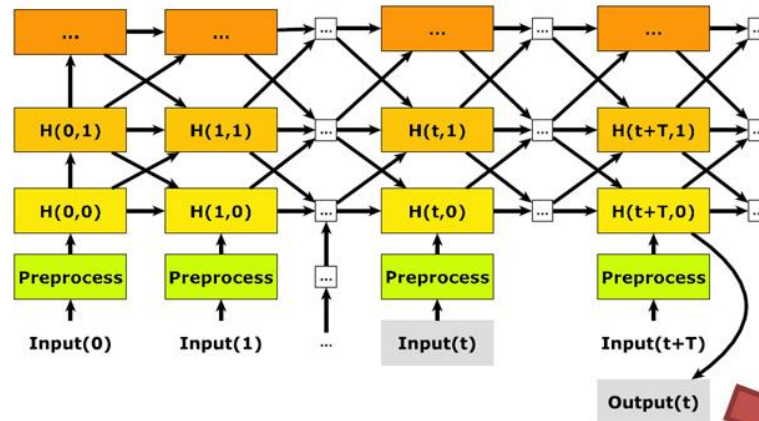
RGB



Neuronale Abstraktionspyramide



Tiefe



Ausgabe

- Rekursive Berechnung ermöglicht effiziente zeitliche Integration

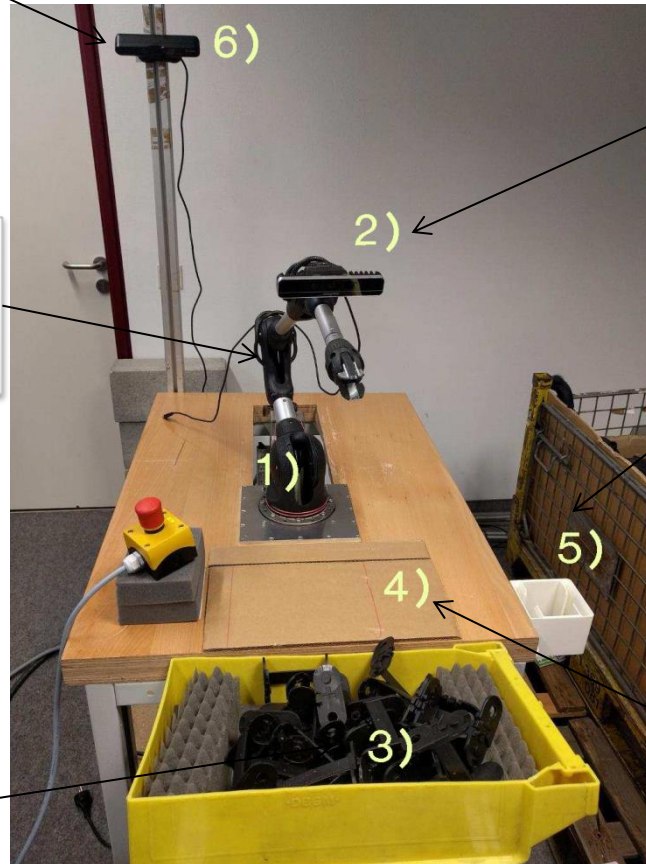
[Pavel, Schulz, Behnke, Neural Networks 2017]

# EuRoC Challenge 1: Robolink Feeder

ASUS Xtion RGB-D workspace camera

Cable-driven 6DOF igus-robolink® manipulator

Pile of the chain parts



SR300 RGB-D wrist camera

Energy chain part feeder

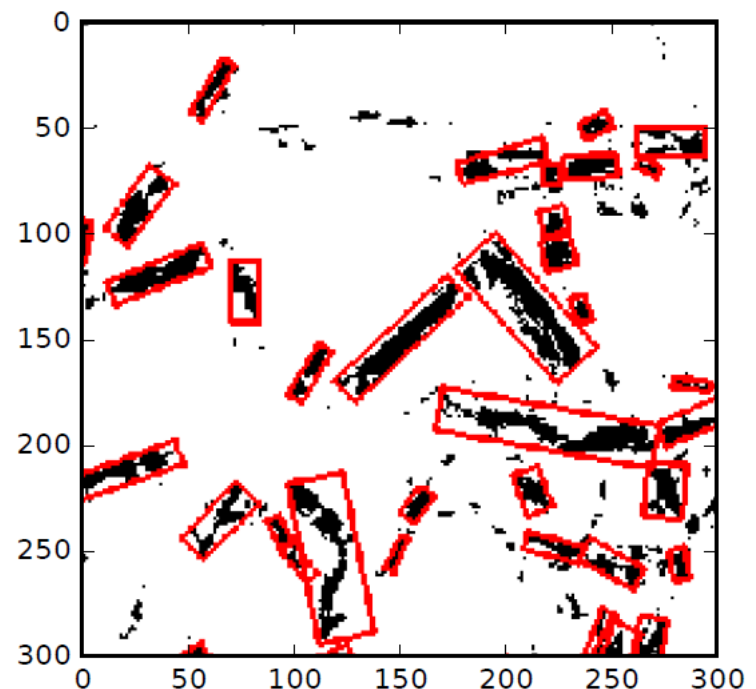
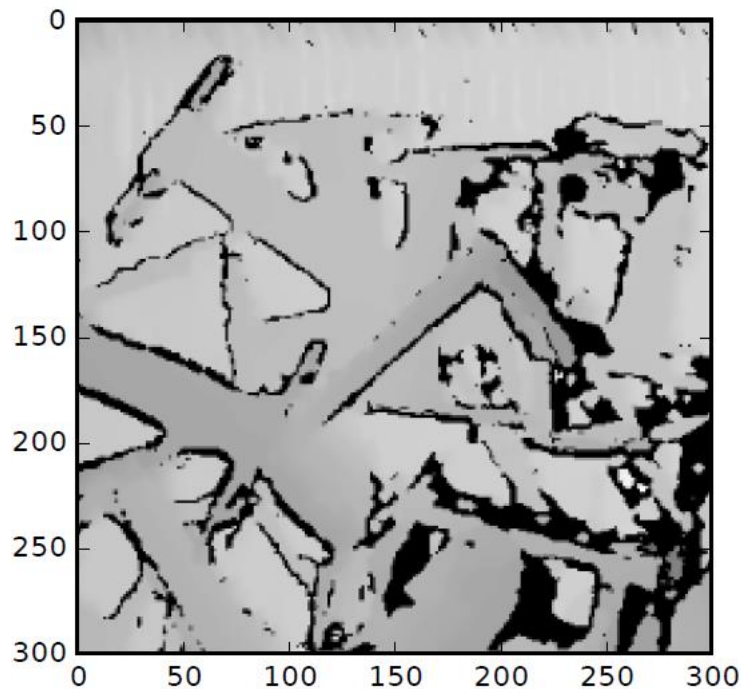
Place for regrasping

[Koo et al. CASE 2017]



# Detektion von Greifgelegenheiten

- Analyse des Tiefenbilds
- Detektion schmaler hervorstehende Bereiche



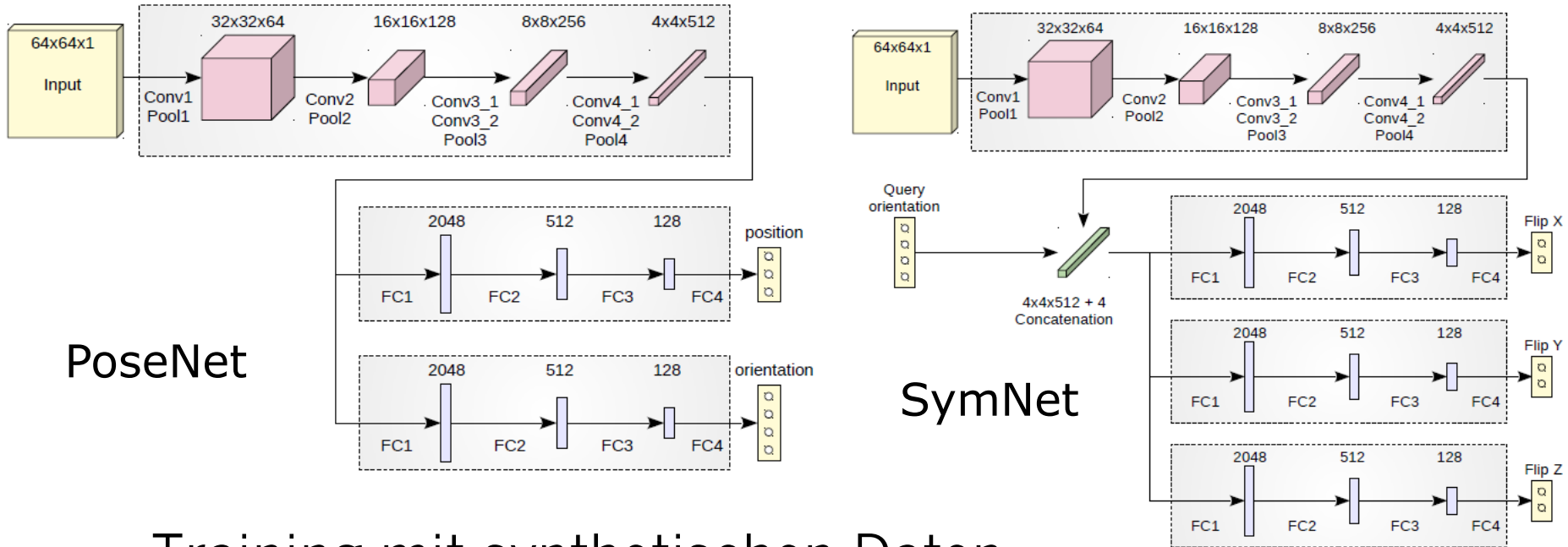
# Robolink Feeder: Griff in die Kiste

Part detection

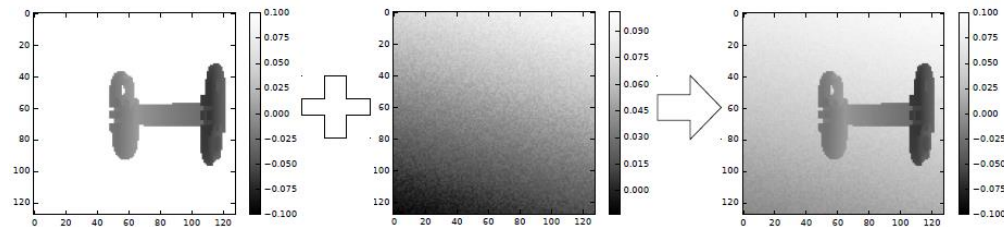


# Posenschätzung für Teile

- Zwei konvolutionale neuronale Netze



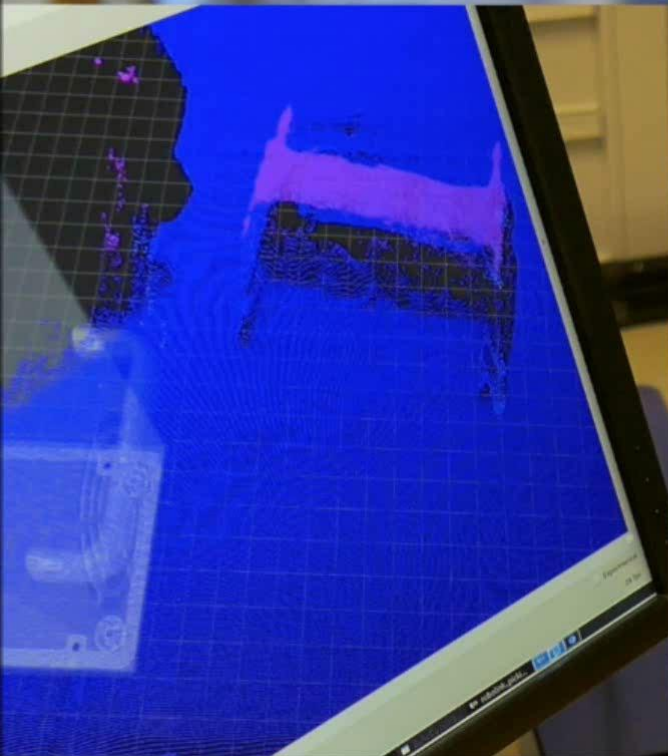
- Training mit synthetischen Daten



[Koo et al. CASE 2017]

# Robolink Feeder: Umgreifen und Platzierung

Pose estimation



[Koo et al. CASE 2017]



# System

Air velocity sensor

UR 10 Arm (6 DOF)

2x Intel RealSense SR300  
+ LED light

Bendable  
suction finger

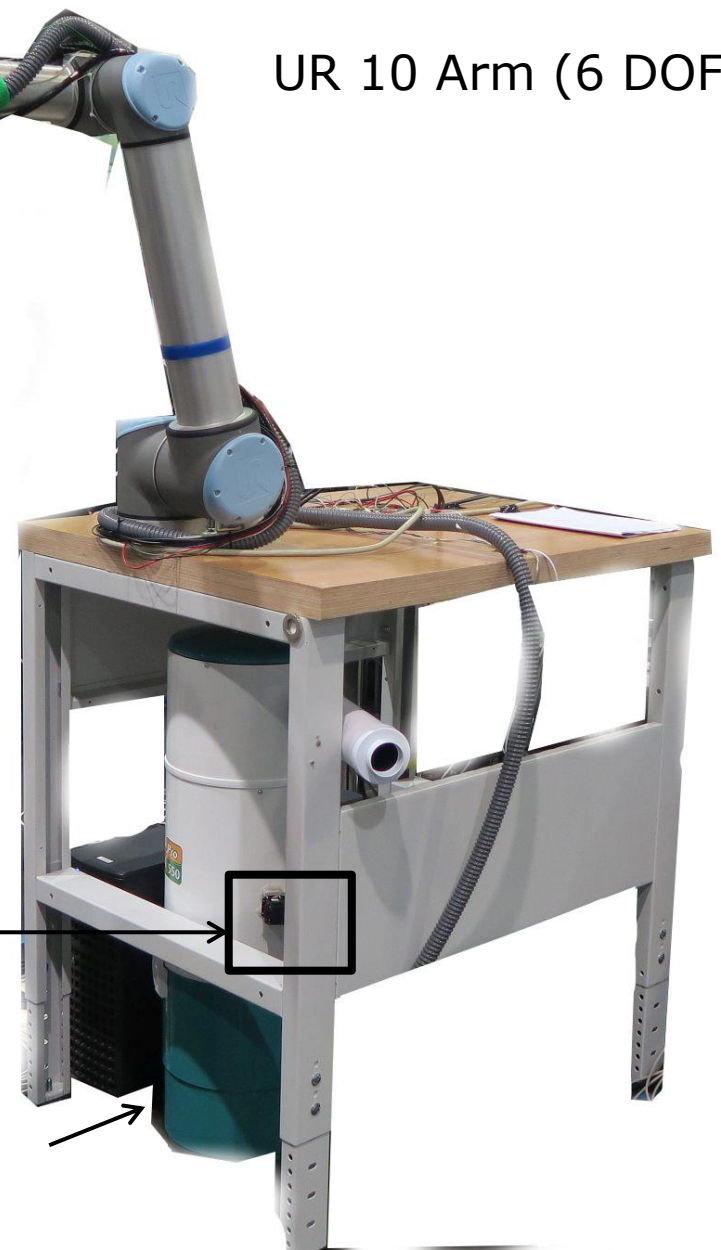
Linear actuator

Total:  
6+2 DOF

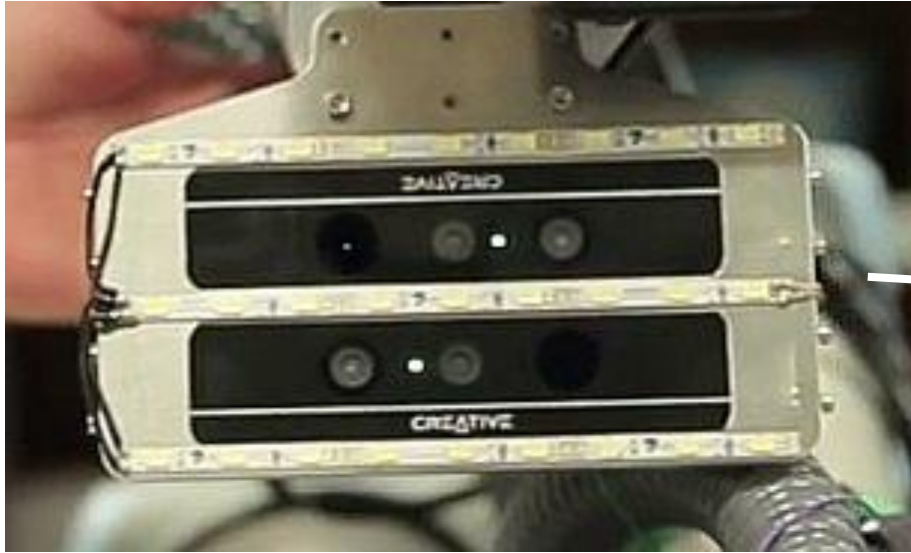
Suction strength control

Strong vacuum  
cleaner (3100 W)

[Schwarz et al. ICRA 2017]

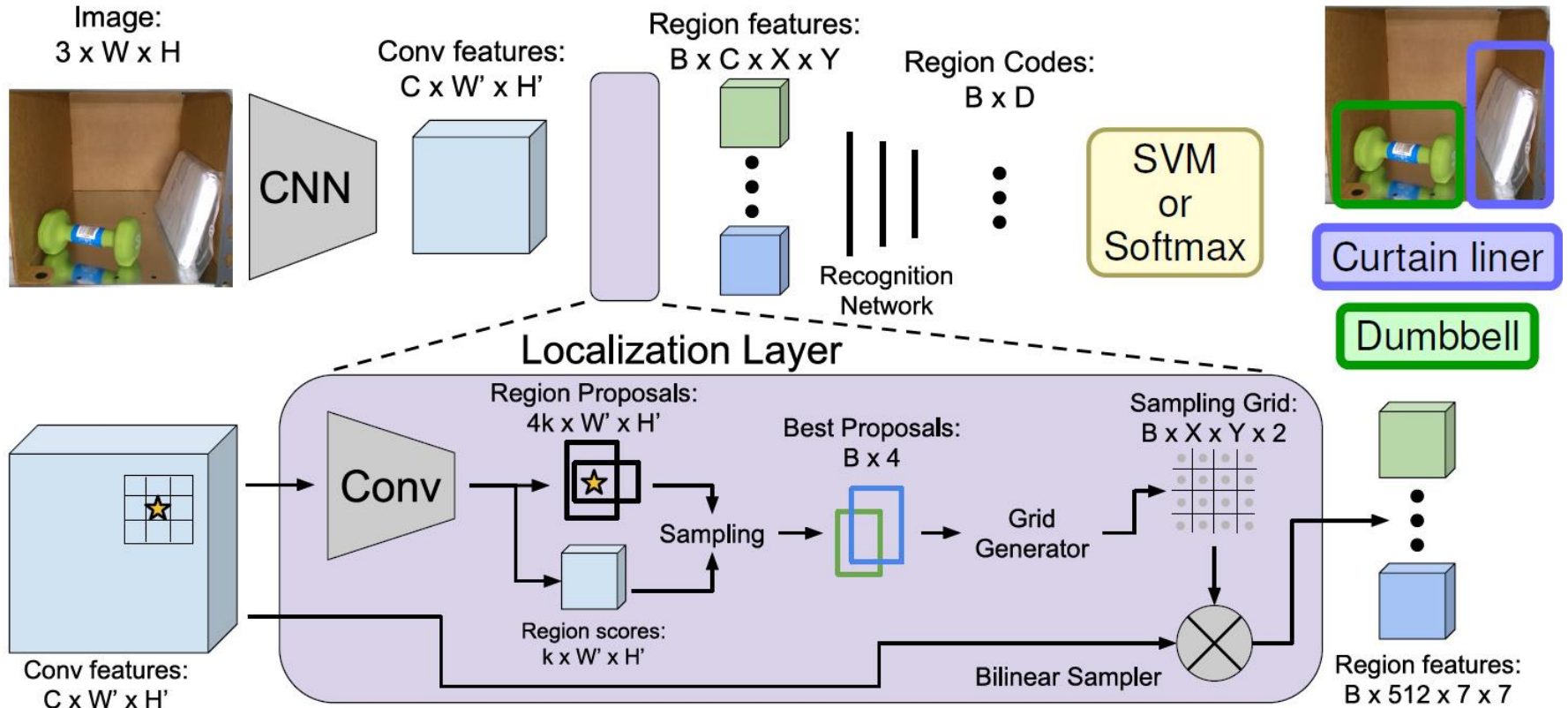


# RGB-D-Kameras



- 2x Intel RealSense SR300
- Fusion of three depth estimates per pixel (including RGB stereo)

# Objektdetektion



[Adapted from Johnson et al. CVPR 2016]

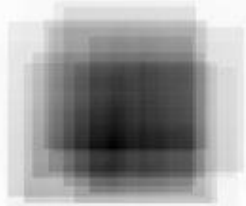
[Schwarz et al. ICRA 2017]



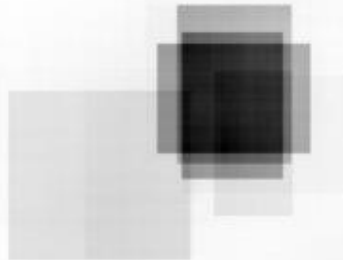
# Beispiel-Detektionen



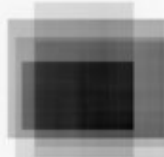
Gloves



Glue sticks



Sippy cup

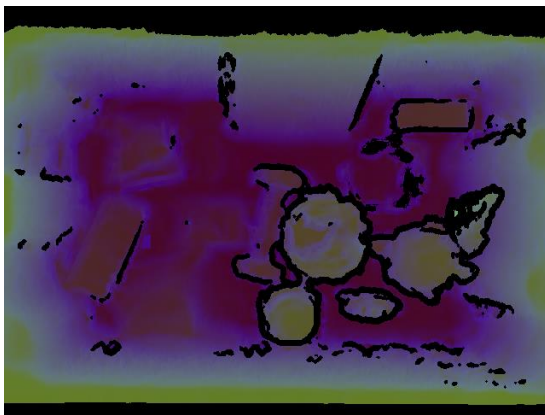


[Schwarz et al. ICRA 2017]

# Semantische Segmentierung

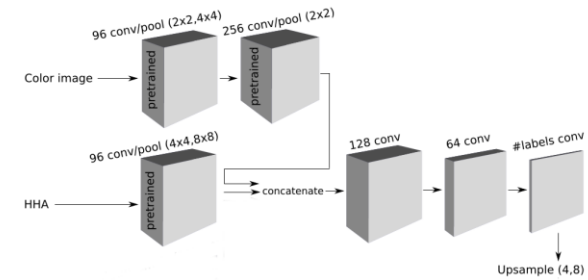
## ■ Tiefes Konvolutionales Netzwerk

RGB



HHA

Ausgabe



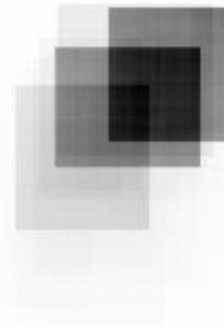
[Husain et al. RA-L 2016]

# Kombination von Detektion und Segmentierung

- Pixelweise Multiplikation



Detektion

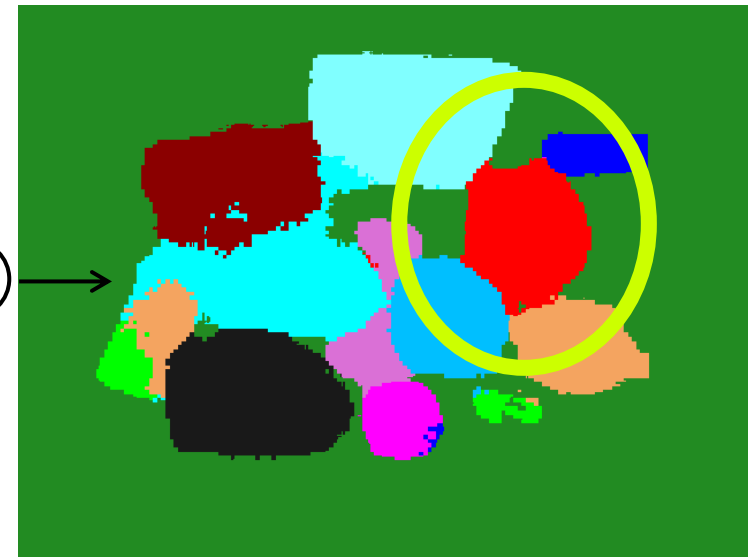


Segmentierung



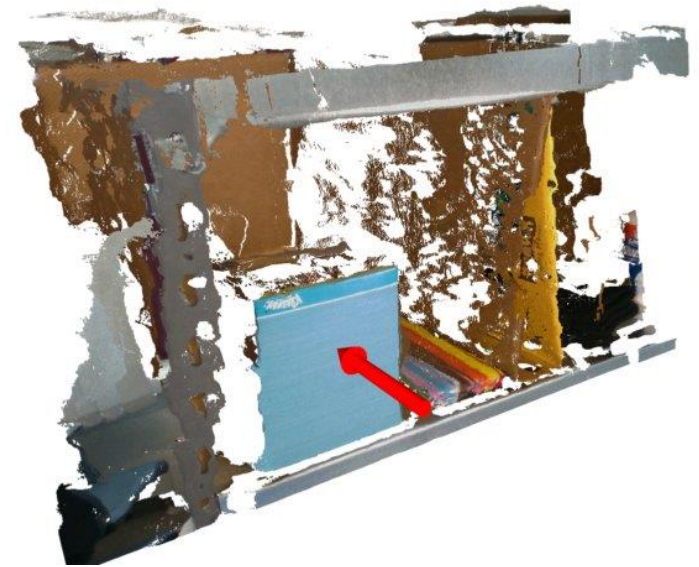
⊗

Ausgabe



# Auswahl der Saugpose

- Center grasp for “standing” objects:
  - Find support area for suction close to bounding box center
- Top grasp for “lying” objects:
  - Find support area for suction close to horizontal bounding box center



# Stowing



2x

# Picking



4x

# 6D-Posenschätzung

- Capture item on turn table
- Build 3D model
- Generate proposals
- Register to test image



# Schwierige Produkte

- Heavy / Large

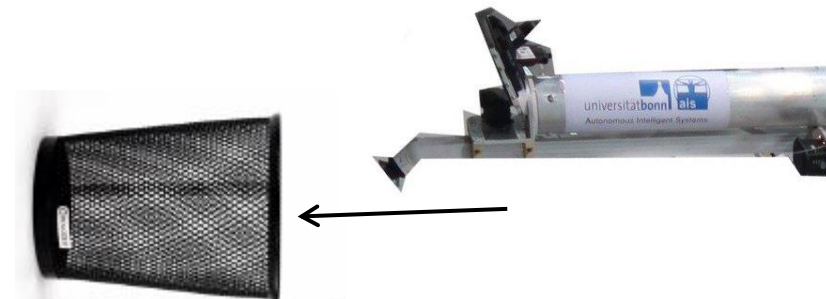
- Many holes / Meshes



Grasp one ball



Ensure that grasp is on **center of mass!**



Knock over and suck on bottom



# Greifen des Stiftbechers



# Team NimbRo Picking



- 2<sup>nd</sup> place stowing, 3<sup>rd</sup> place picking

# MBZIRC Challenge 2



2x

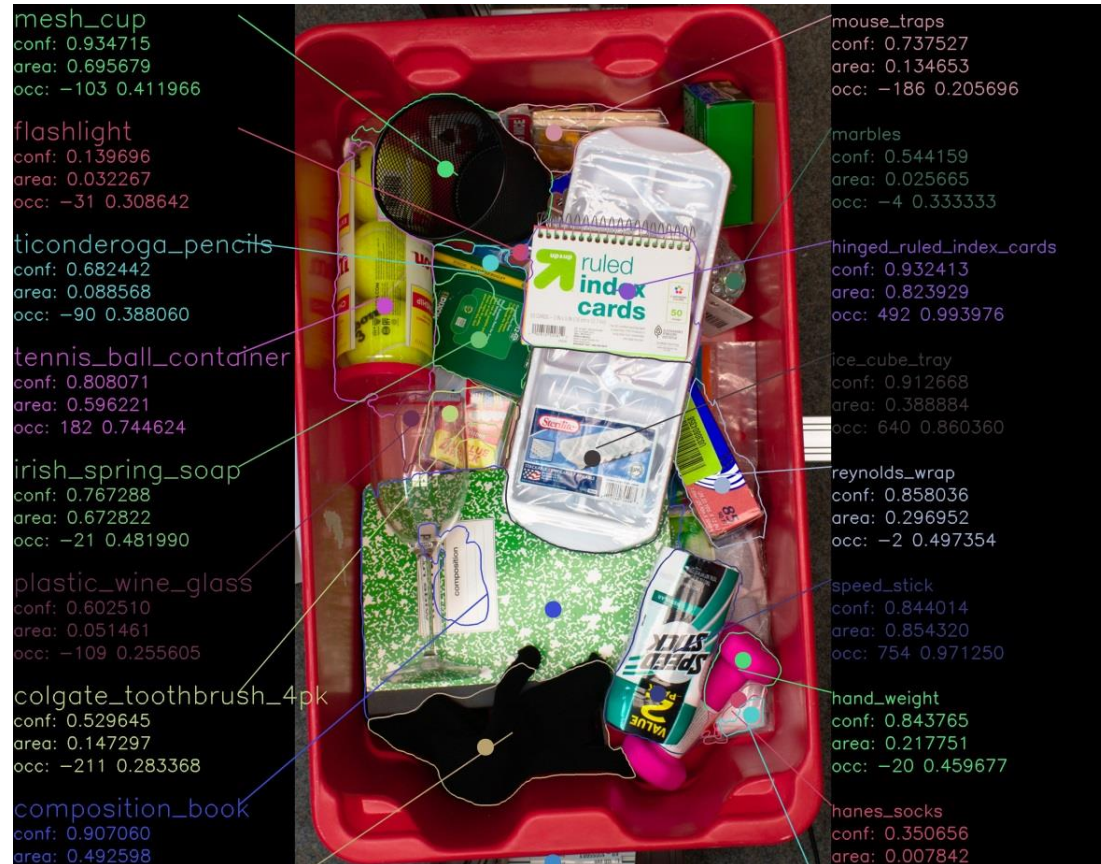
# Wahrnehmung der Schraubenschlüssel

- Detektion von Maul und Ring
- Assoziation der Teile, Auswahl anhand Länge



# Amazon Robotics Challenge 2017

- Höherer Schwierigkeitsgrad
- Training mit gerenderten Szenen



# Zusammenfassung

- Methoden zur Manipulation von Objekten und Werkzeuggebrauch entwickelt
  - Haushaltsroboter
  - Industrieautomatisierung
  - Suche und Rettung
  - Raumfahrtszenarien
- Herausforderungen
  - Variabilität der Objekte
  - Räumliche Beschränkungen
  - Aufgaben-Nebenbedingungen
- Forschungsbedarf
  - Flexible Greifer
  - Wahrnehmung
  - Manipulationsplanung
  - Lernen



**Vielen Dank für Ihre  
Aufmerksamkeit!**

**Fragen?**