

Semantische Wahrnehmung für den Griff in die Kiste

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**Institut für Informatik VI
Autonome Intelligente Systeme**



Roboter für den Griff in die Kiste

ActReMa

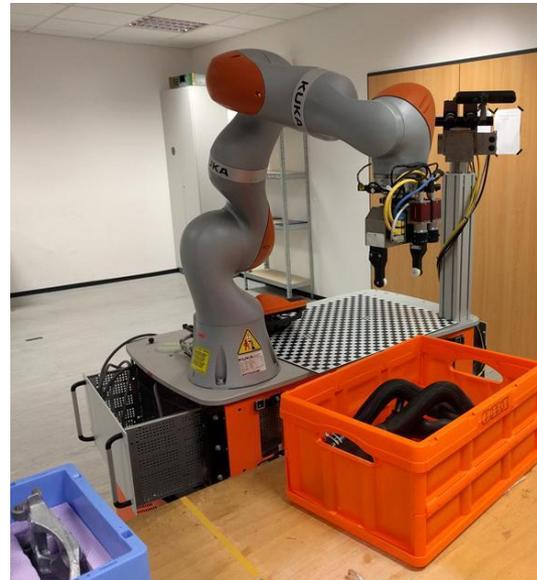


STAMINA

Amazon
Picking
Challenge



EuRoC
C1



EuRoC C2

Serviceroboter



Dynamaid



Cosero

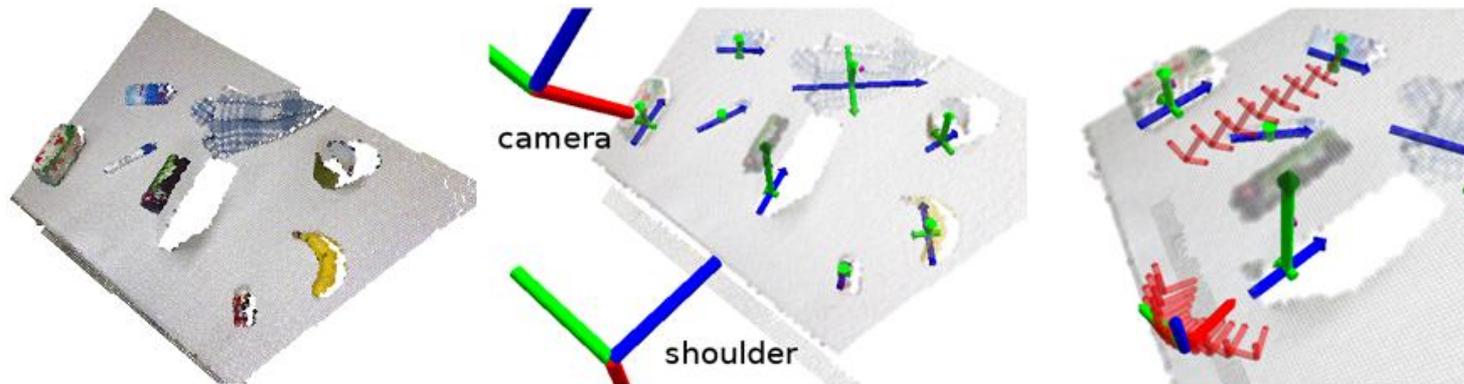
[Stückler et al.: Frontiers in AI and Robotics 2016]

RoboCup 2013 Eindhoven

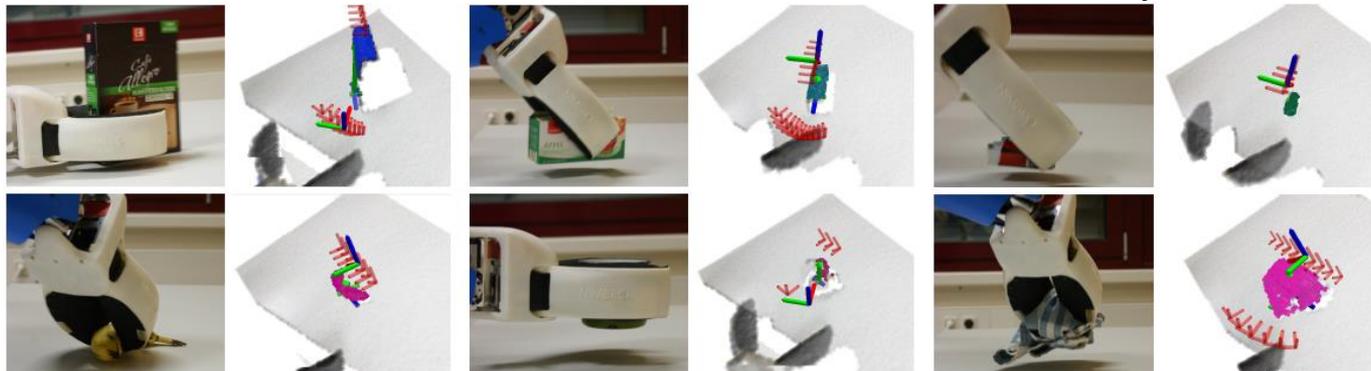


Greifen unbekannter Objekte

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



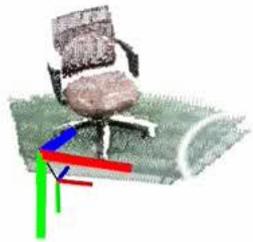
- Flexibles Greifen vieler unbekannter Objekte



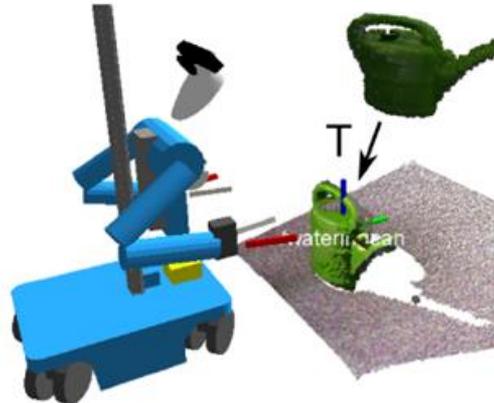
Lernen und Verfolgen von Objektmodellen

- Modeling of objects by RGB-D SLAM

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

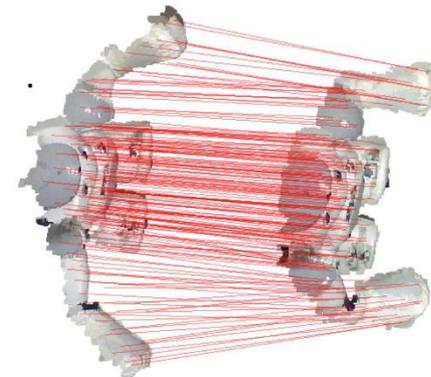
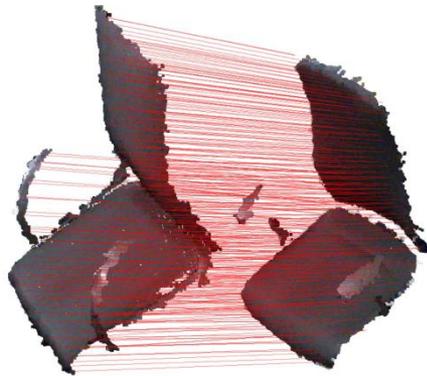
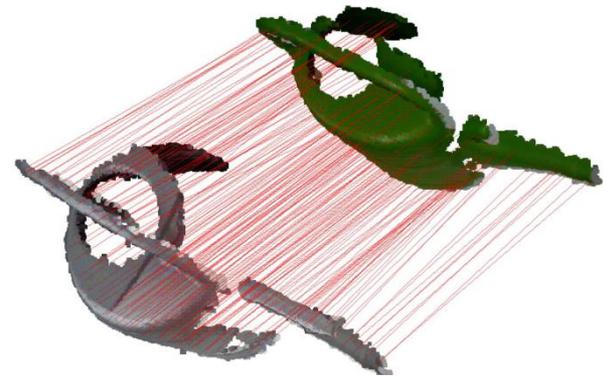
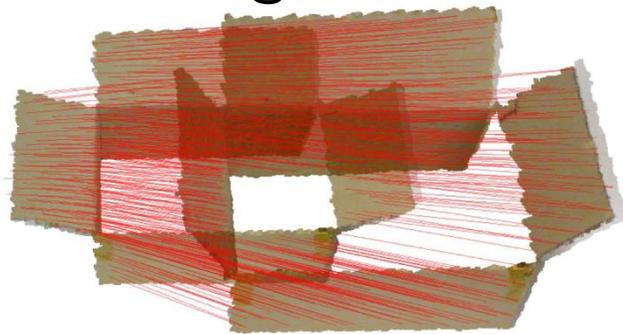


- Real-time registration with current RGB-D image



Deformierbare RGB-D-Registrierung

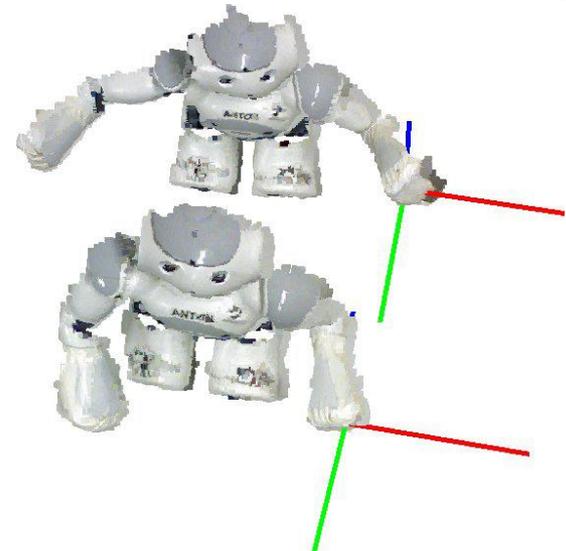
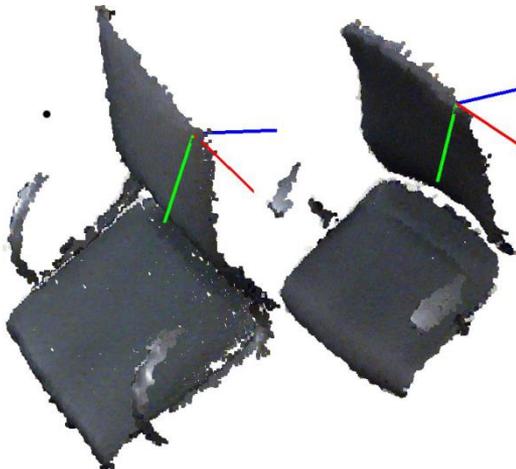
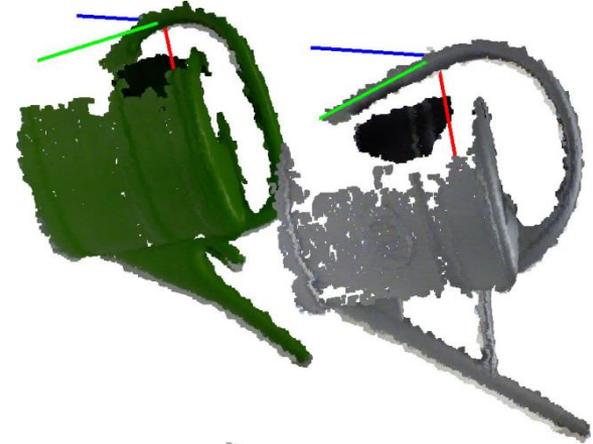
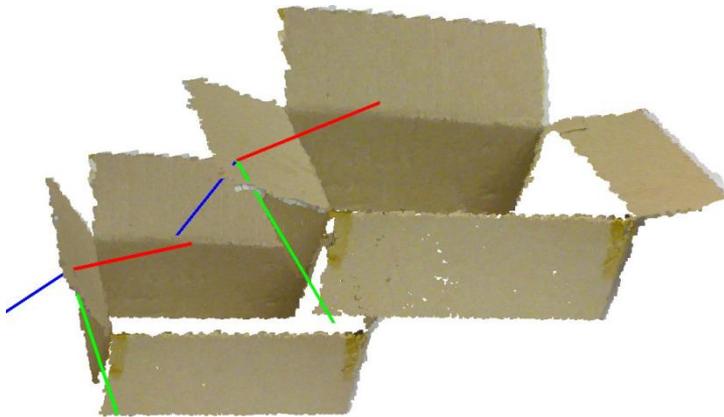
- Basierend auf Coherent Point Drift-Methode
[Myronenko & Song, PAMI 2010]
- Multiresolutions-Surfel-Map erlaubt Echtzeit-Registrierung



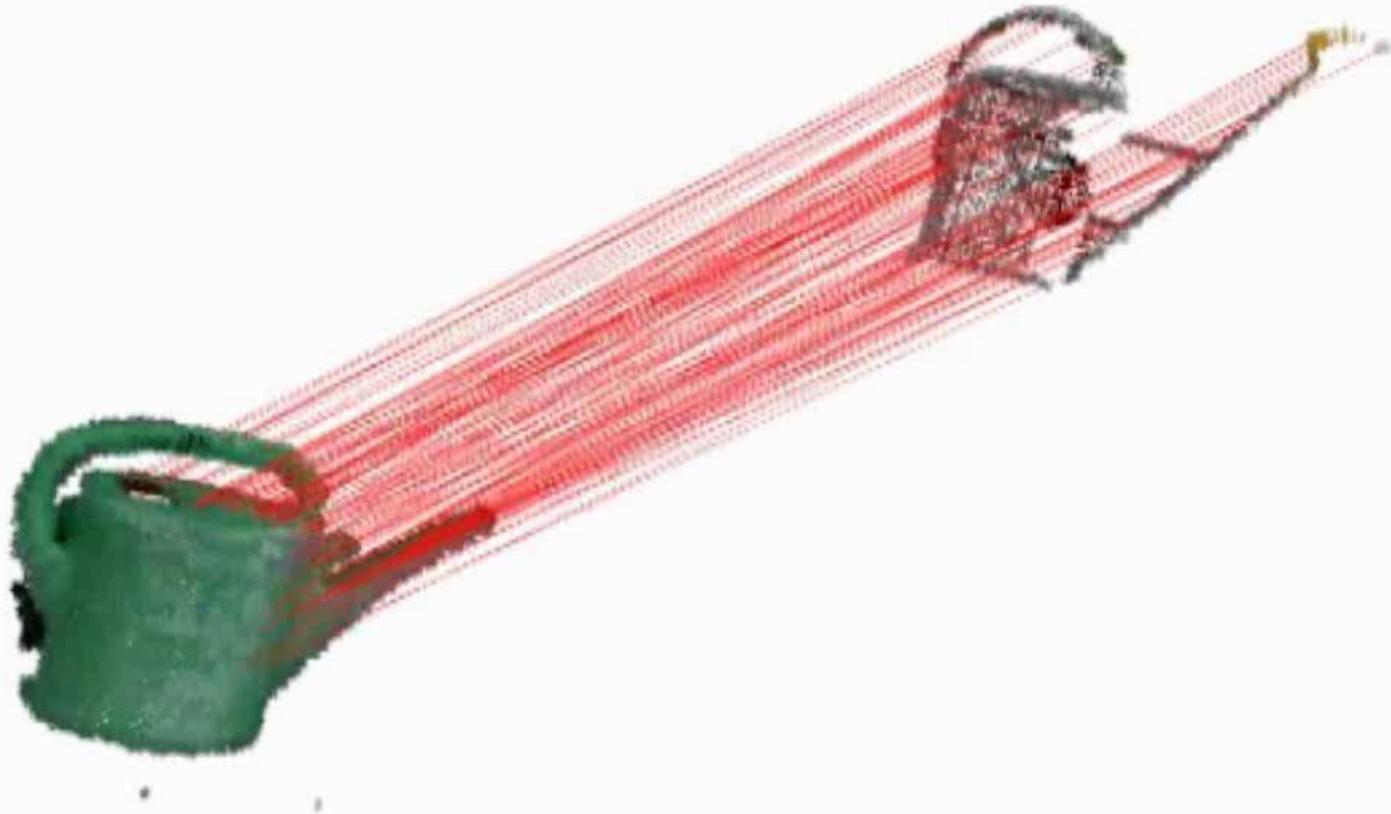
[Stückler, Behnke, ICRA2014]

Transformation von Schlüsselposen

- Abgeleitet aus Deformationsfeld



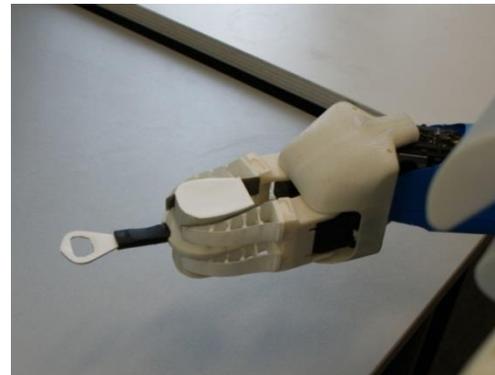
Transfer von Handhabungswissen



- 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



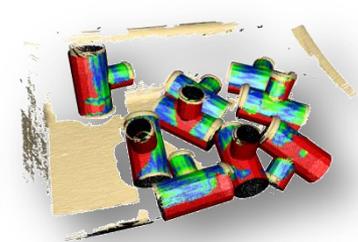
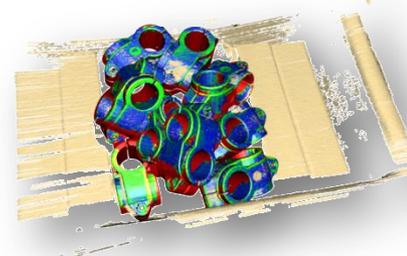
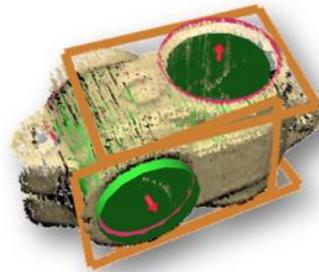
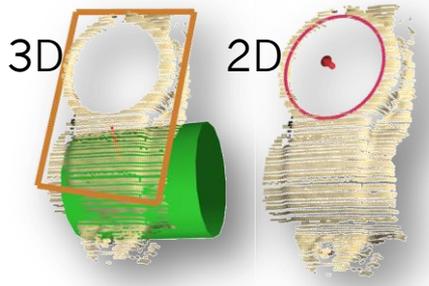
- Unser Team NimbRo hat drei Mal hintereinander die internationalen RoboCup@Home-Wettbewerbe gewonnen

Griff in die Kiste

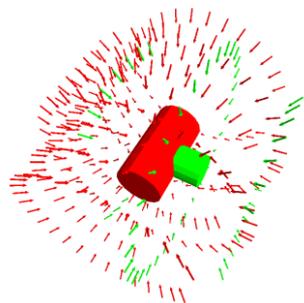
- Bekannte Objekte
Kiste



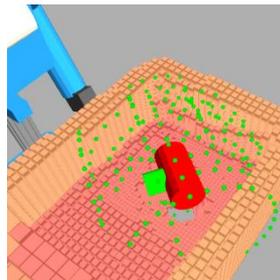
- Registrierung von 2D und 3D Formprimitiv-Graphen



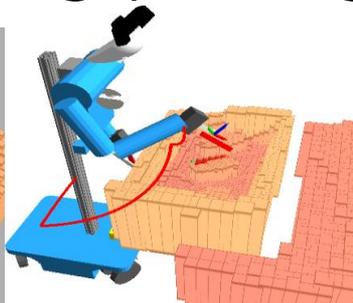
- Greif- und Bewegungsplanung



Offline



Online



Rettungs- und Explorationsroboter

Mobiler Manipulationsroboter Momaro



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

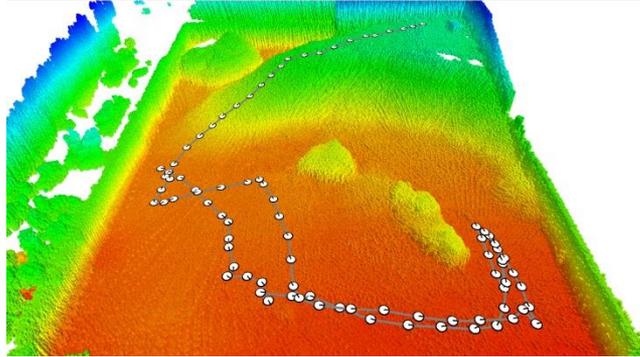
DLR SpaceBot Camp 2015



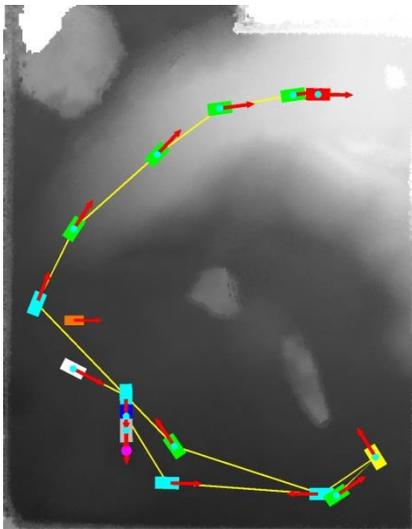
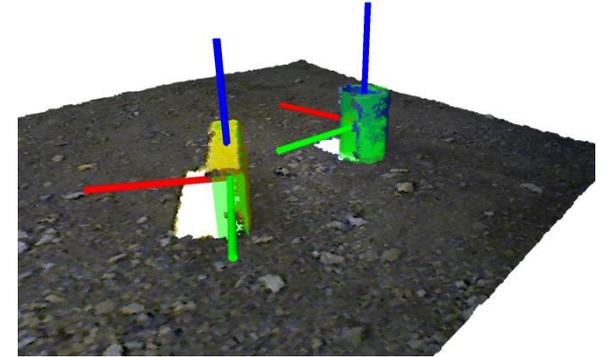
8X

Autonome Missionsausführung

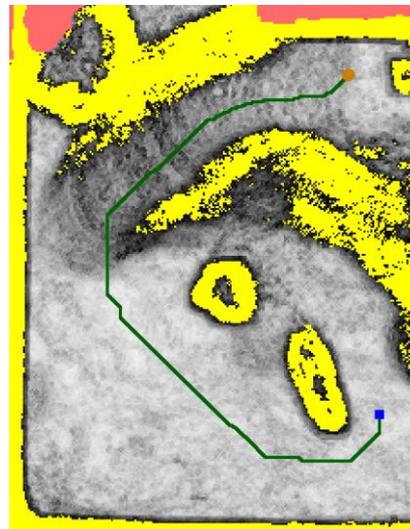
3D Mapping & Localization



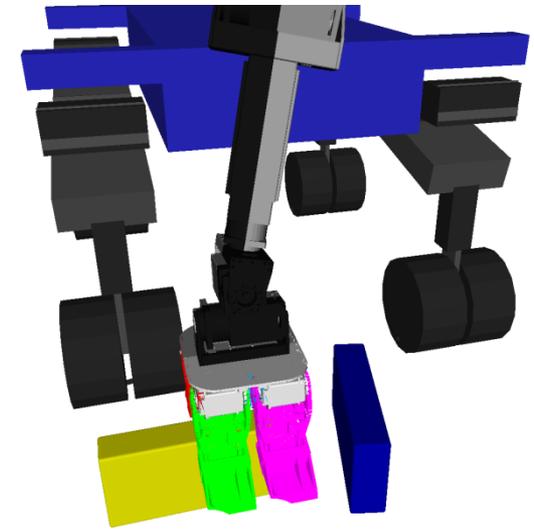
Object perception



Mission plan



Navigation plan



Grasping

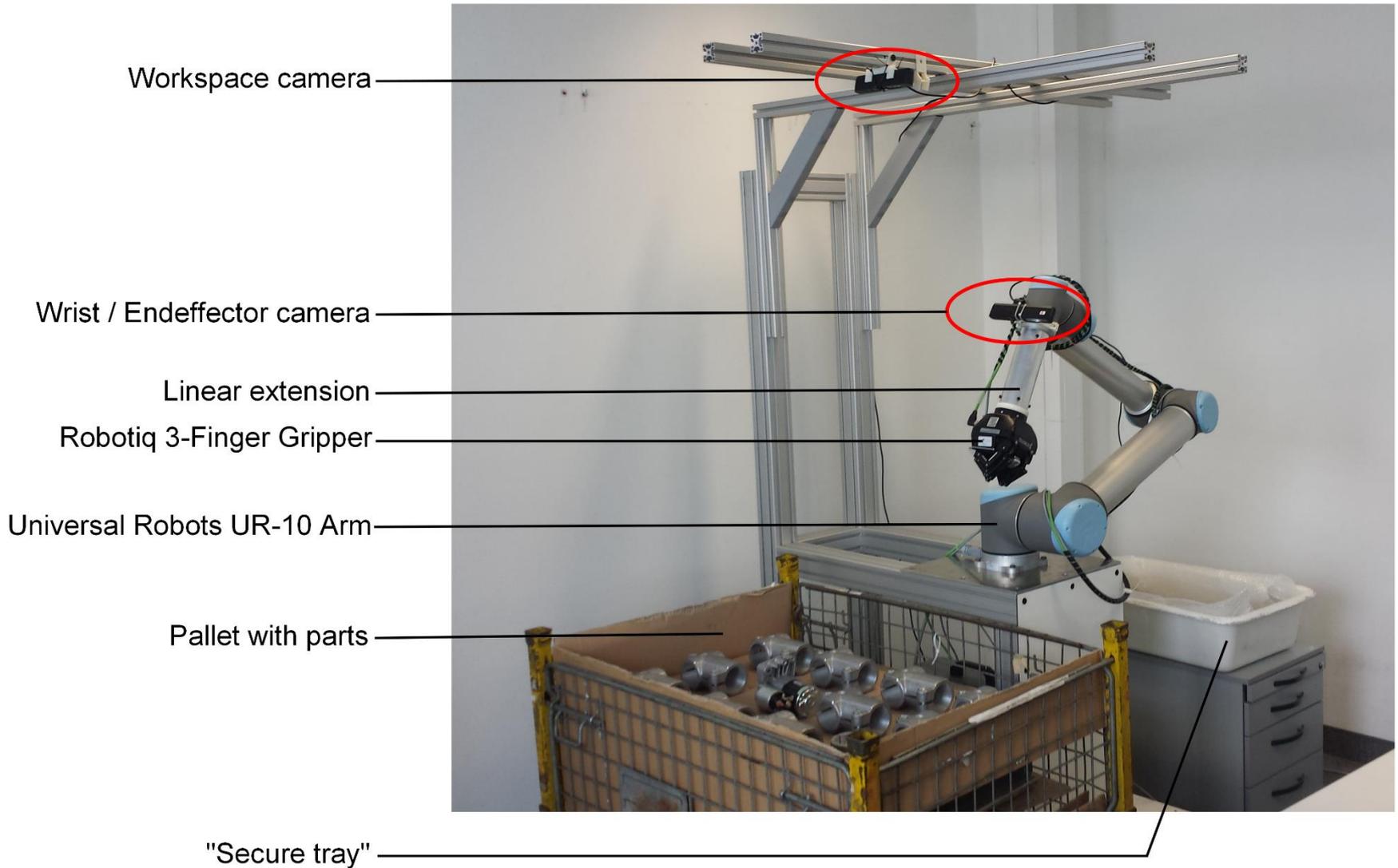
[Schwarz et al., Frontiers in Robotics and AI 2016]

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

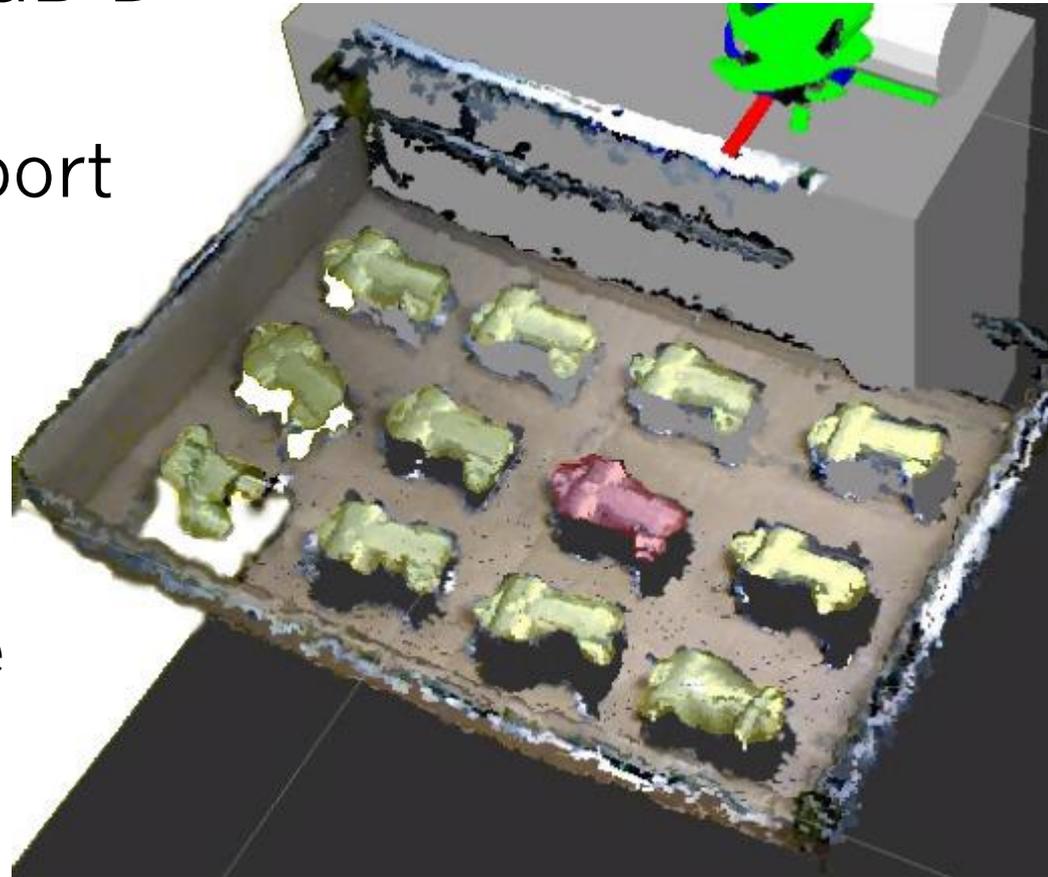


Labor-Demonstrator



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



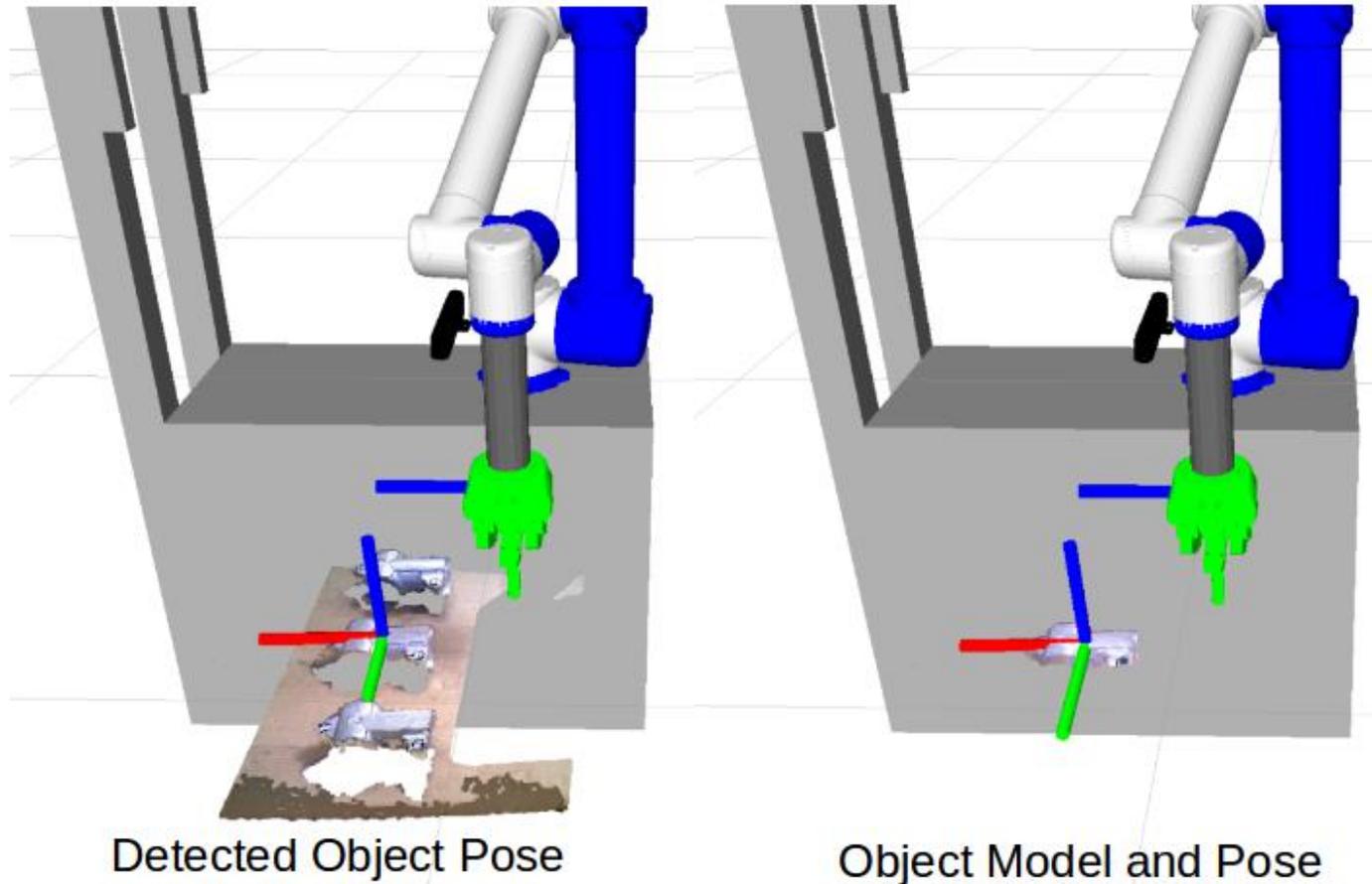
Schätzung der Objektpose

- 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:

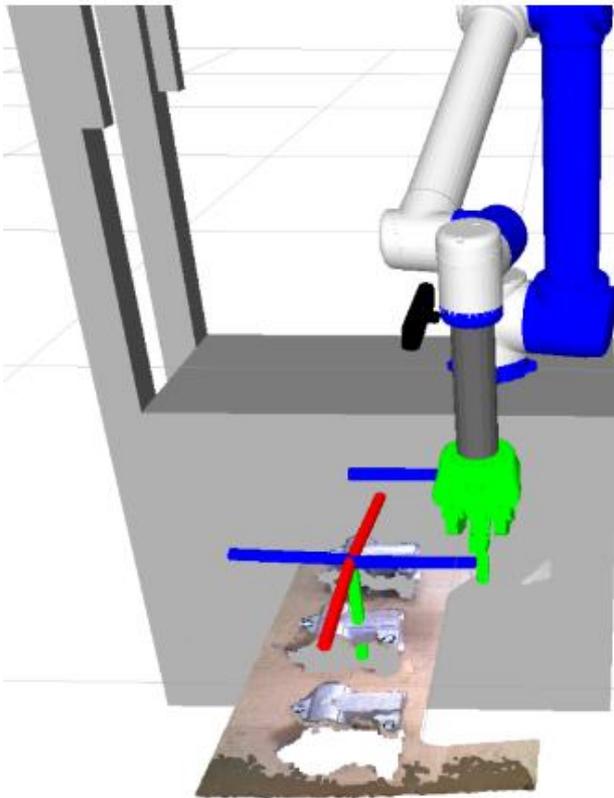
X	Y	Z	Roll	Yaw	Pitch	Grasp Type	
1	-0.009	0.003	0.128	-1.588	-0.009	1.557	BASIC
2	-0.008	-0.007	0.125	-1.550	0.006	-1.643	BASIC

Annotations in the image include:

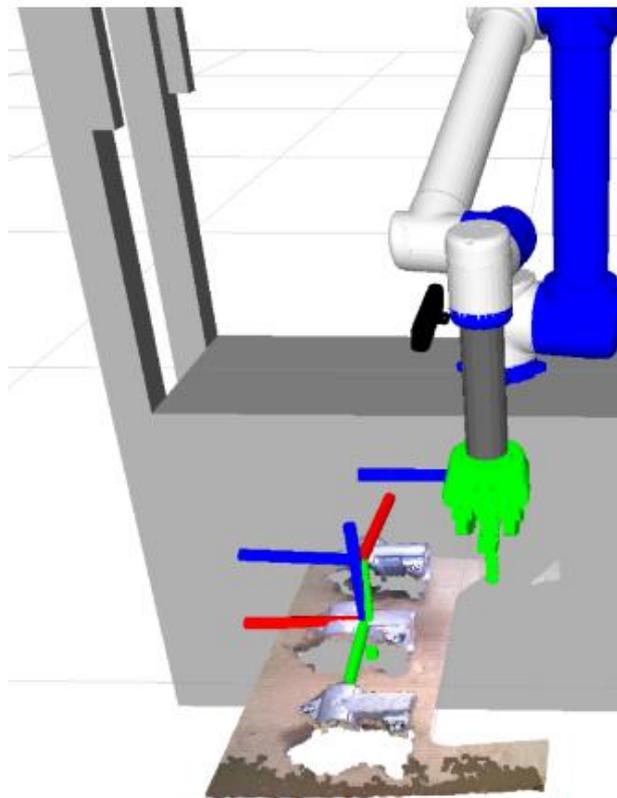
- Red arrows pointing to the 'Duplicate Grasp' and 'Add Current Grasp' buttons, with the text 'Save current robot pose as a new grasp pose'.
- Red arrows pointing to the 'X' and 'Z' columns of the table, with the text 'Find object axis'.
- A red arrow pointing to the 'Take Snapshot' button, with the 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

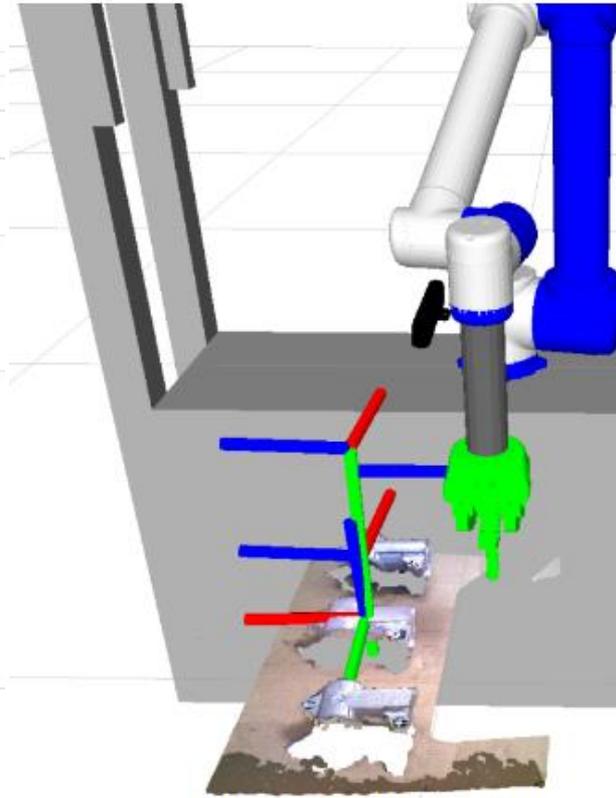
- Grasps are selected according to object pose



Grasp Candidates



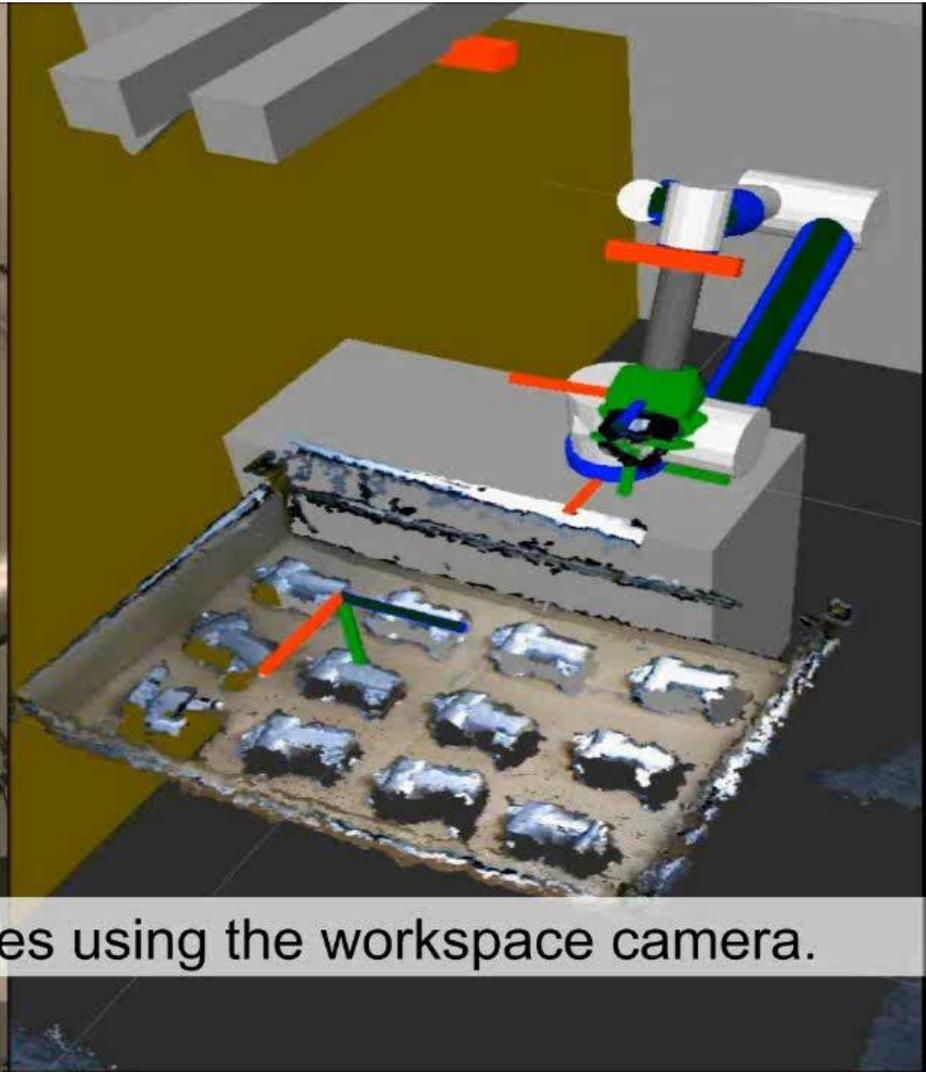
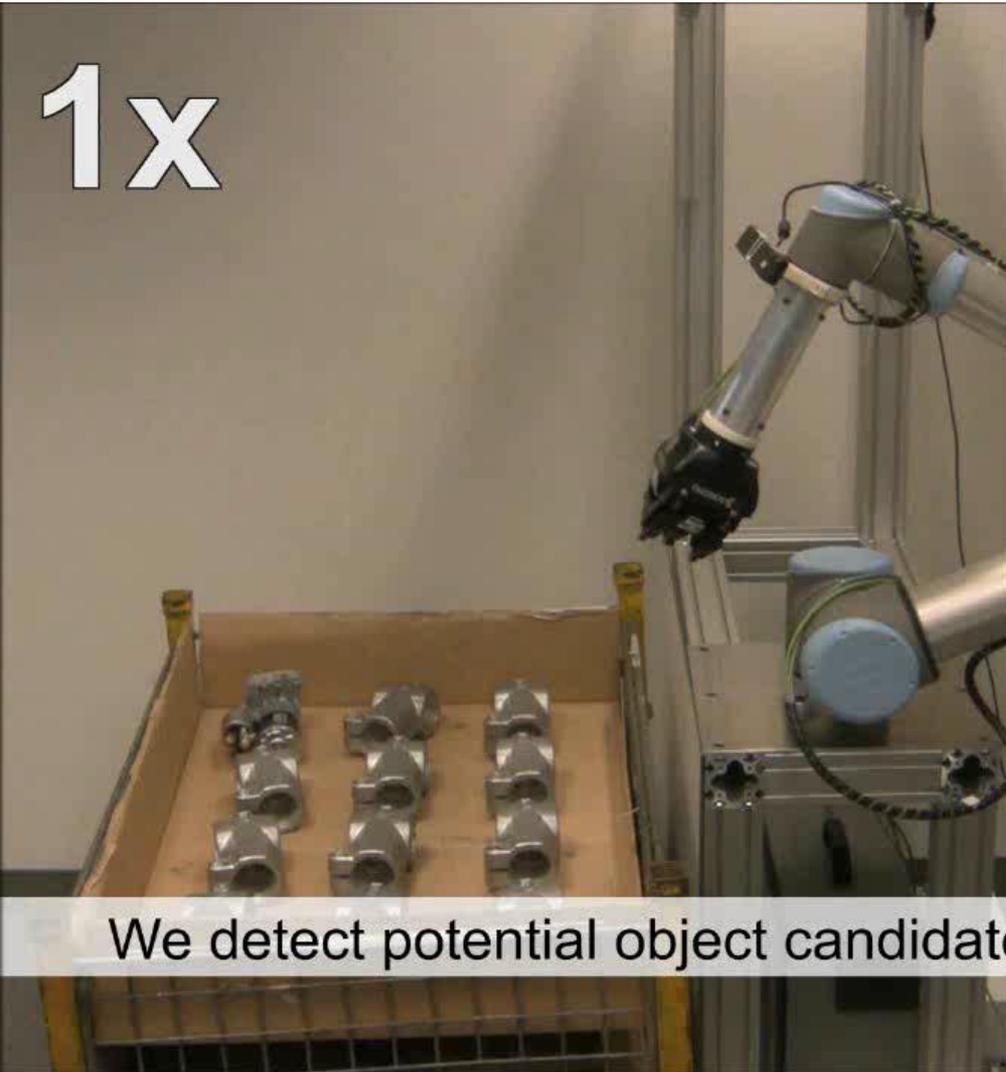
Selected Grasp and Object Pose



Pre-Grasp, Grasp and Object Pose

Depalettierung von Teilen

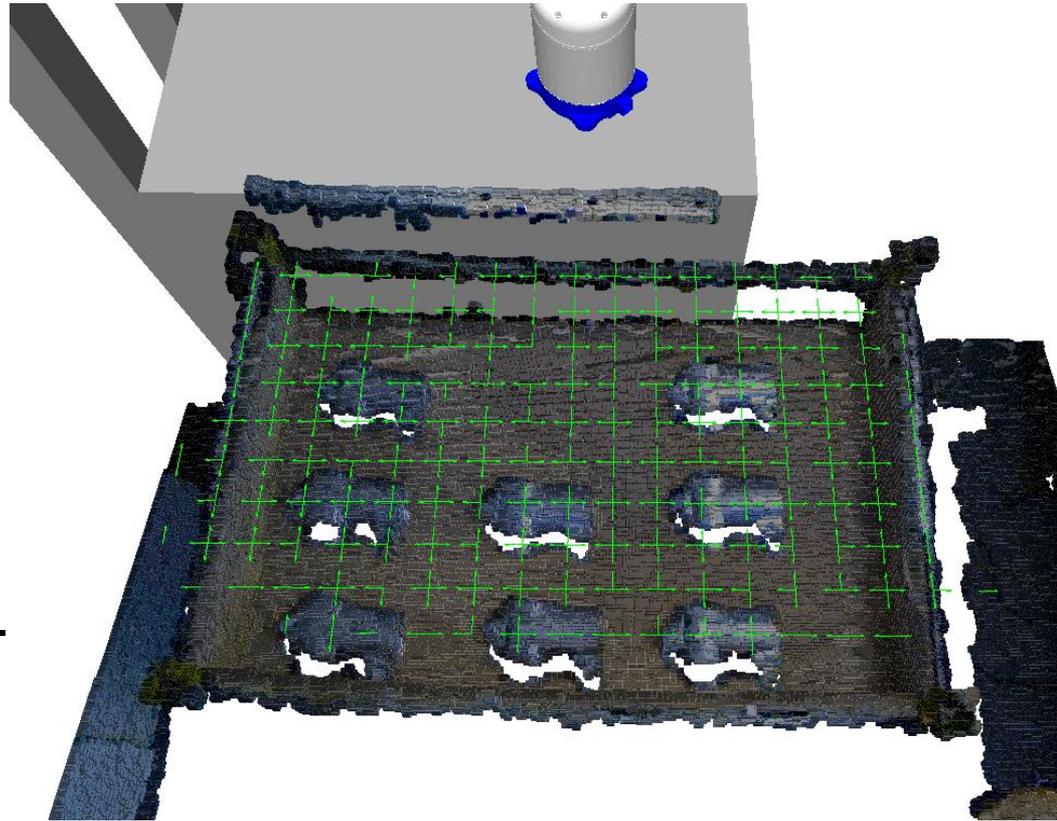
1x



We detect potential object candidates using the workspace camera.

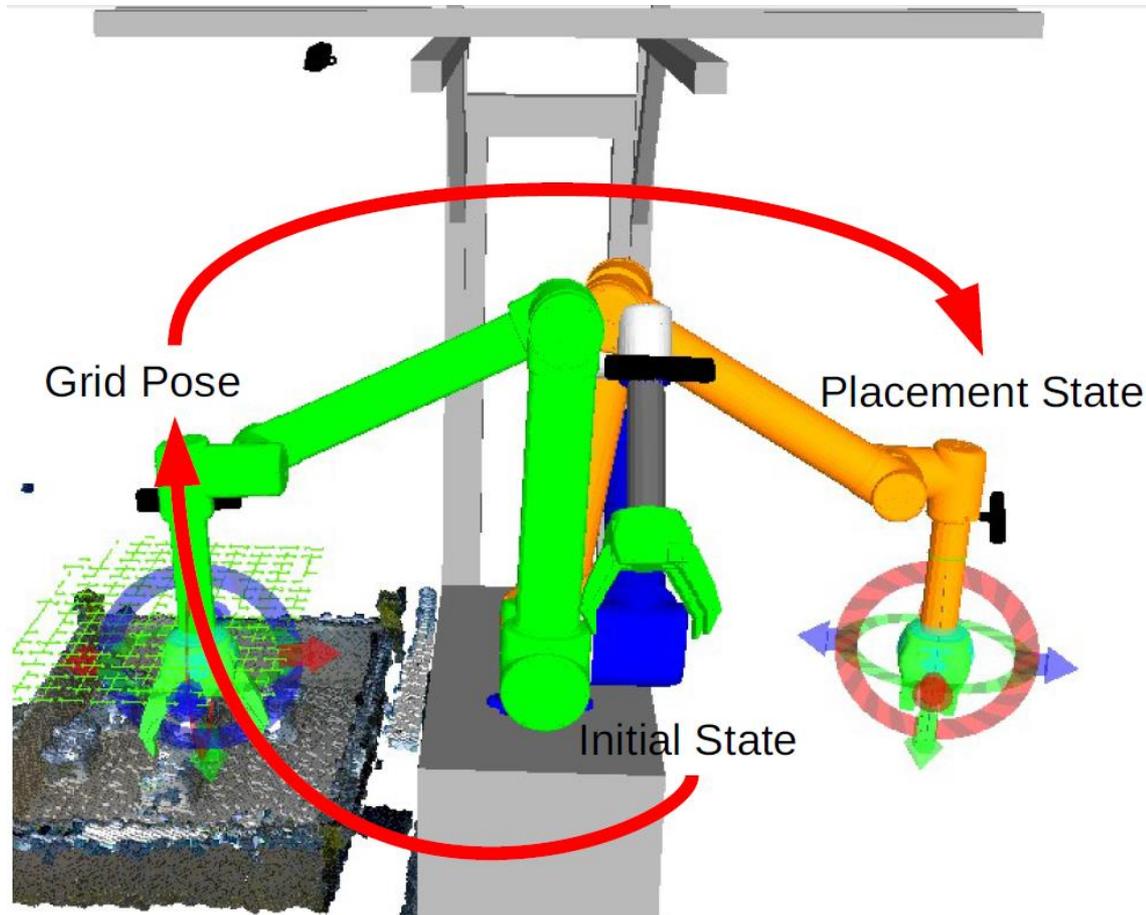
Bewegungsplanung

- 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

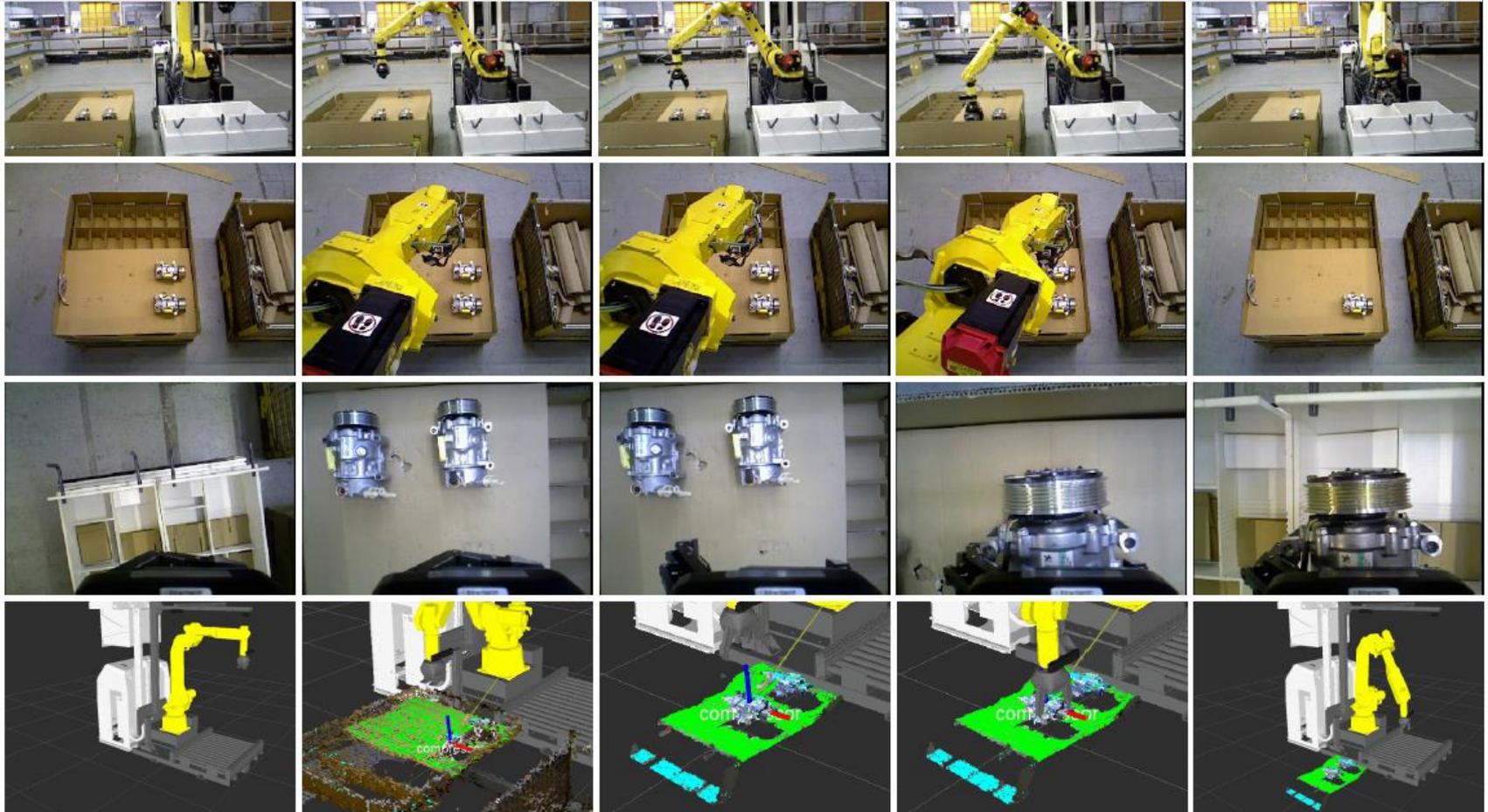
Initialization

Part
detection

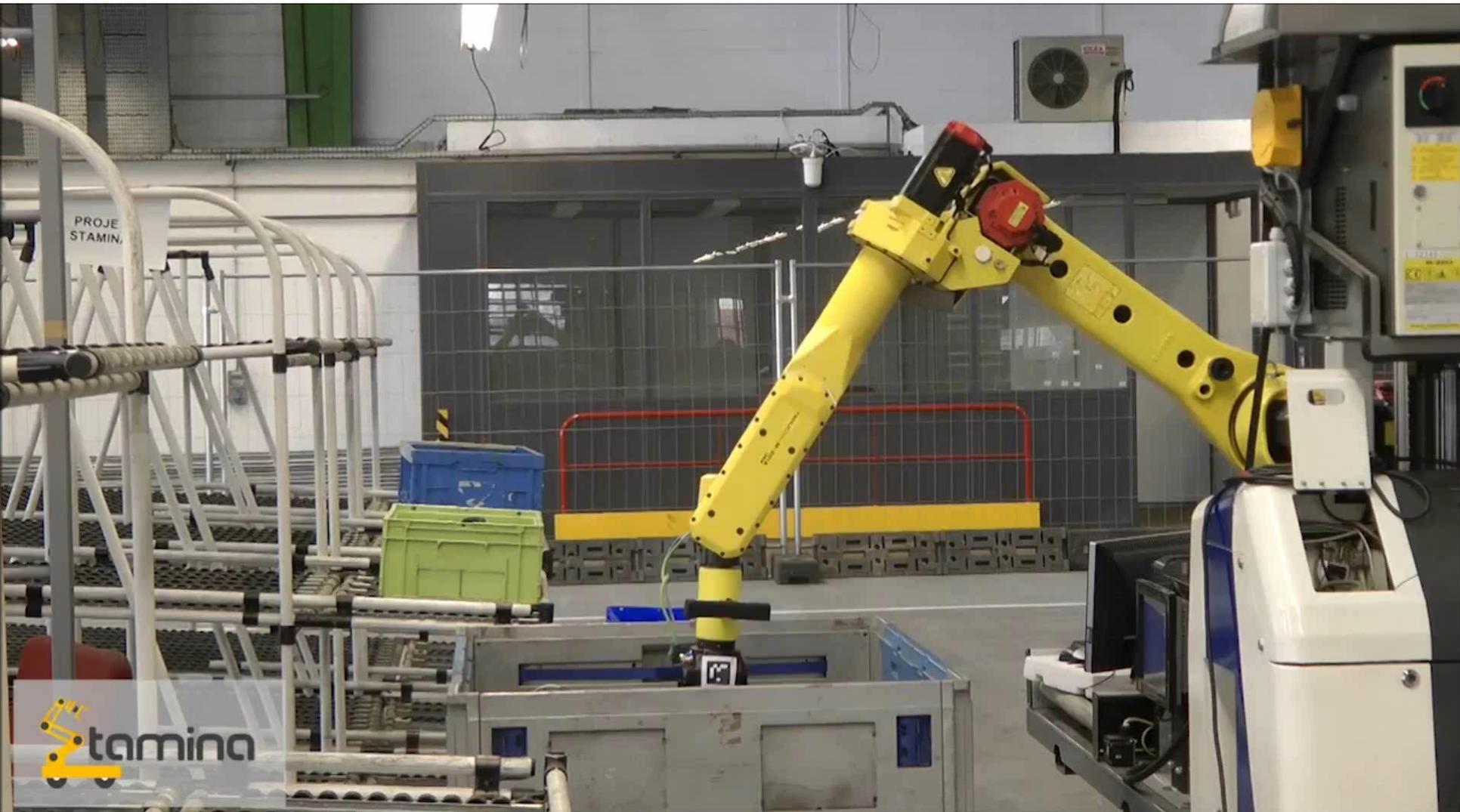
Approach

Grasping

Placing

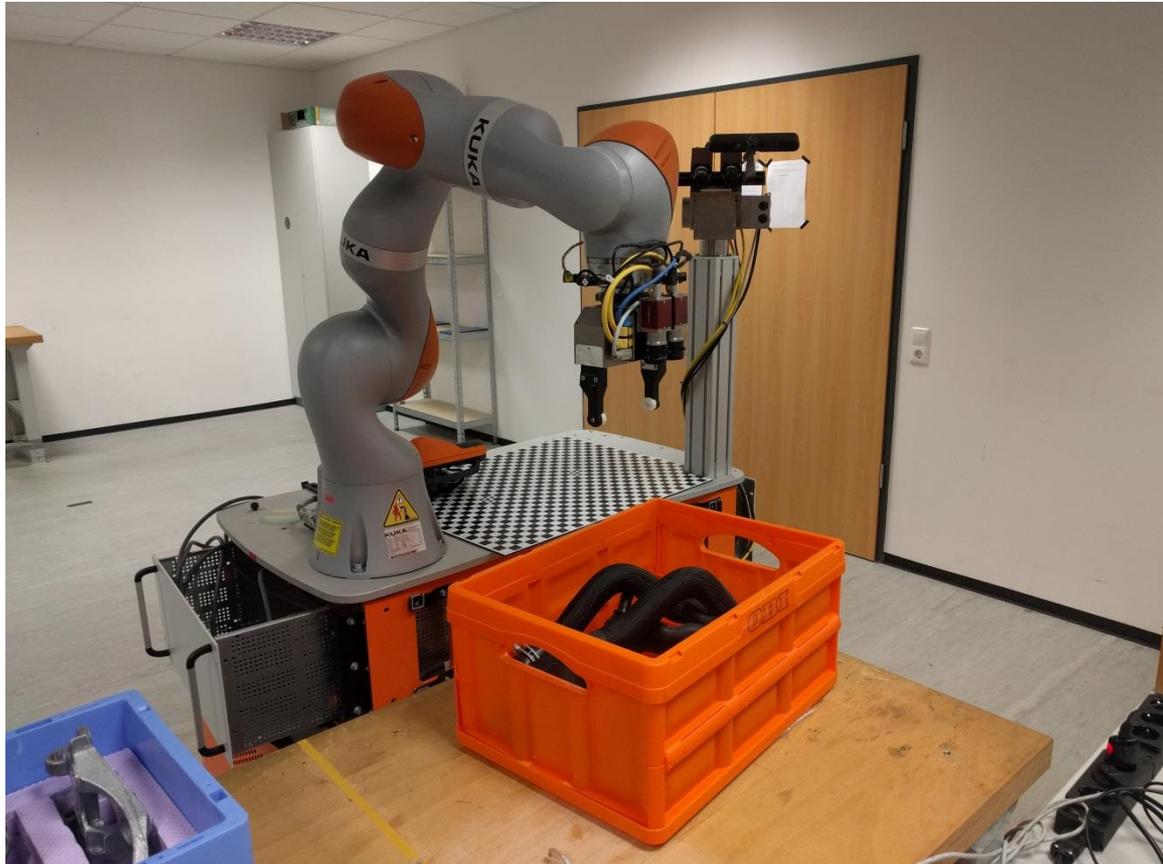


STAMINA-Abschlussdemonstration



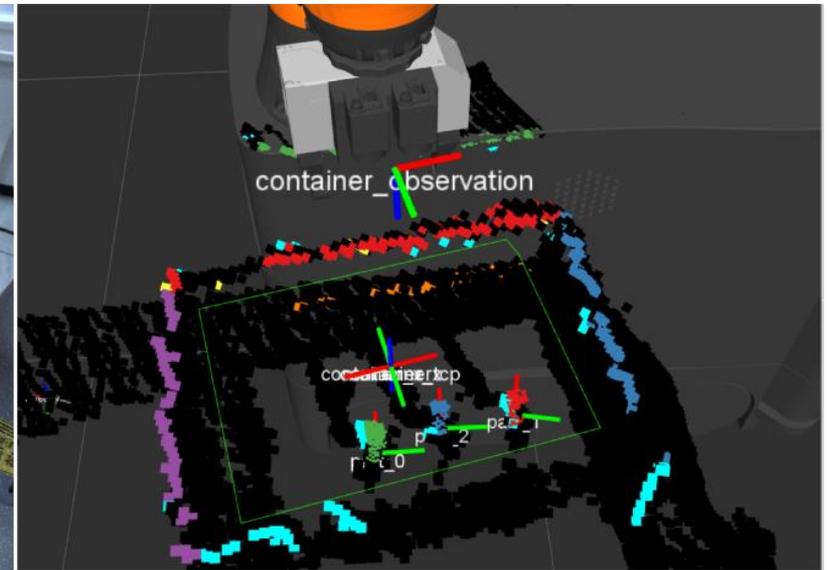
EuRoC Challenge 2: KittingBot

- Kleinerer Roboter: Kuka miiwa
- Zusammenarbeit mit Menschen



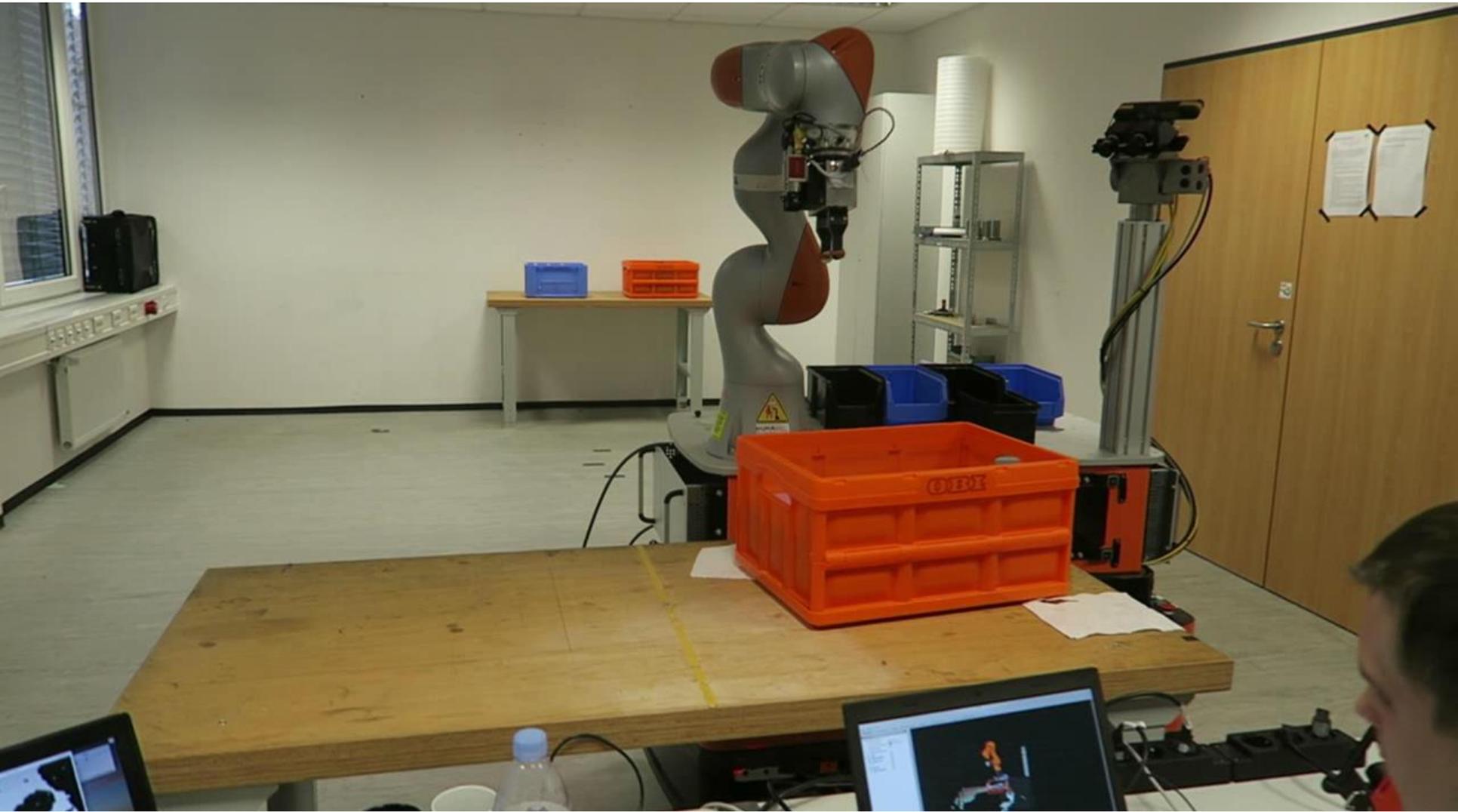
Detektion der Kiste und Posenschätzung

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



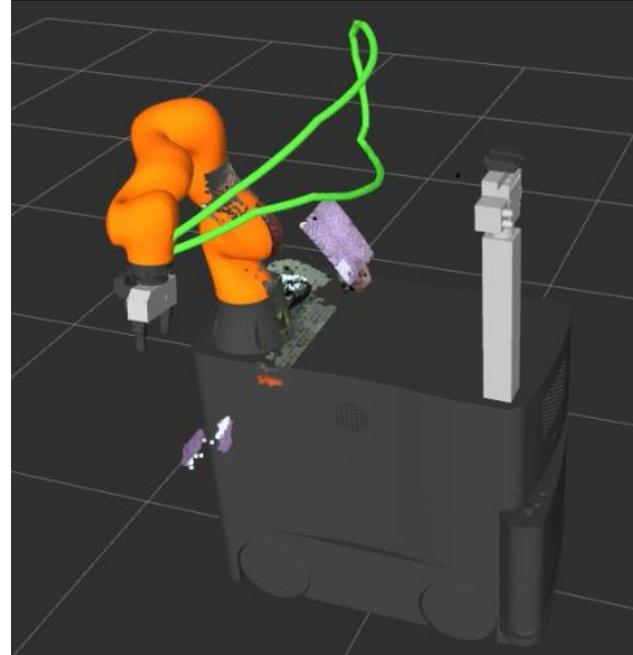
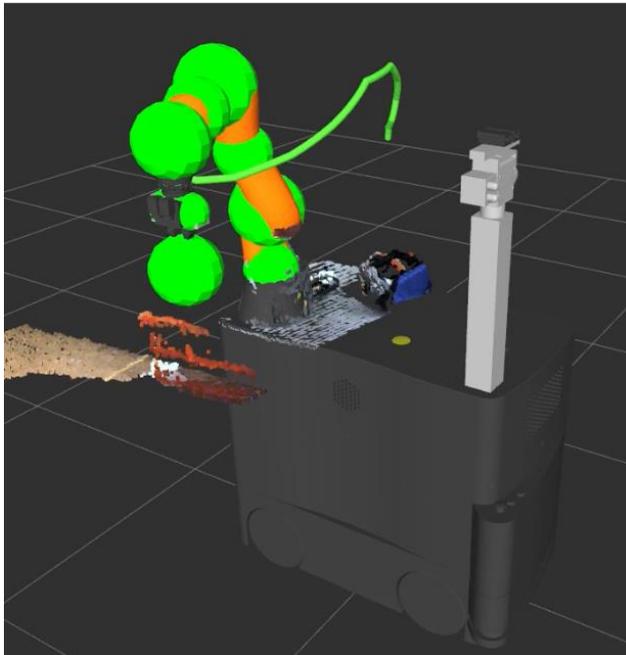
[Holz und Behnke: ISR 2016]

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



Germán M. García - Grzegorz Ficht - Dmytro Pavlichenko
Seongyong Koo - Sven Behnke

KittingBot Showcase-Demonstration

Kitting Demonstrator

Engine Support 2
9808515080

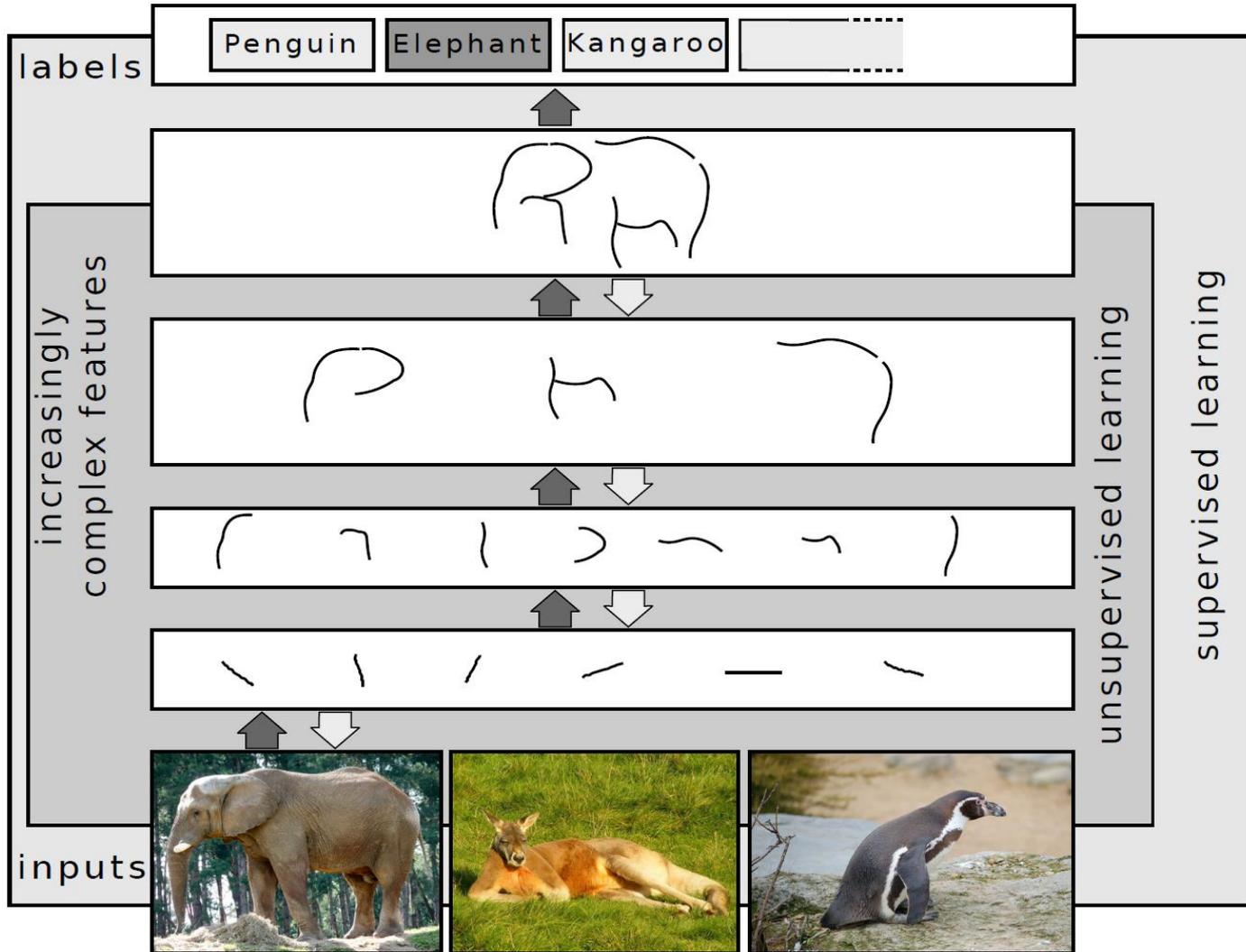


Engine Support 1
9672950980



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



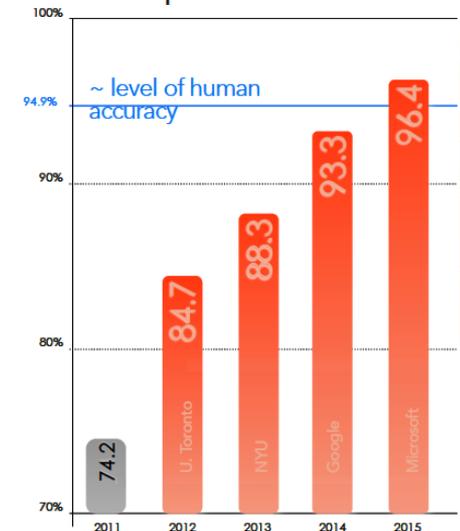
GT: coucal
1: coucal
 2: indigo bunting
 3: lorikeet
 4: walking stick
 5: custard apple



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



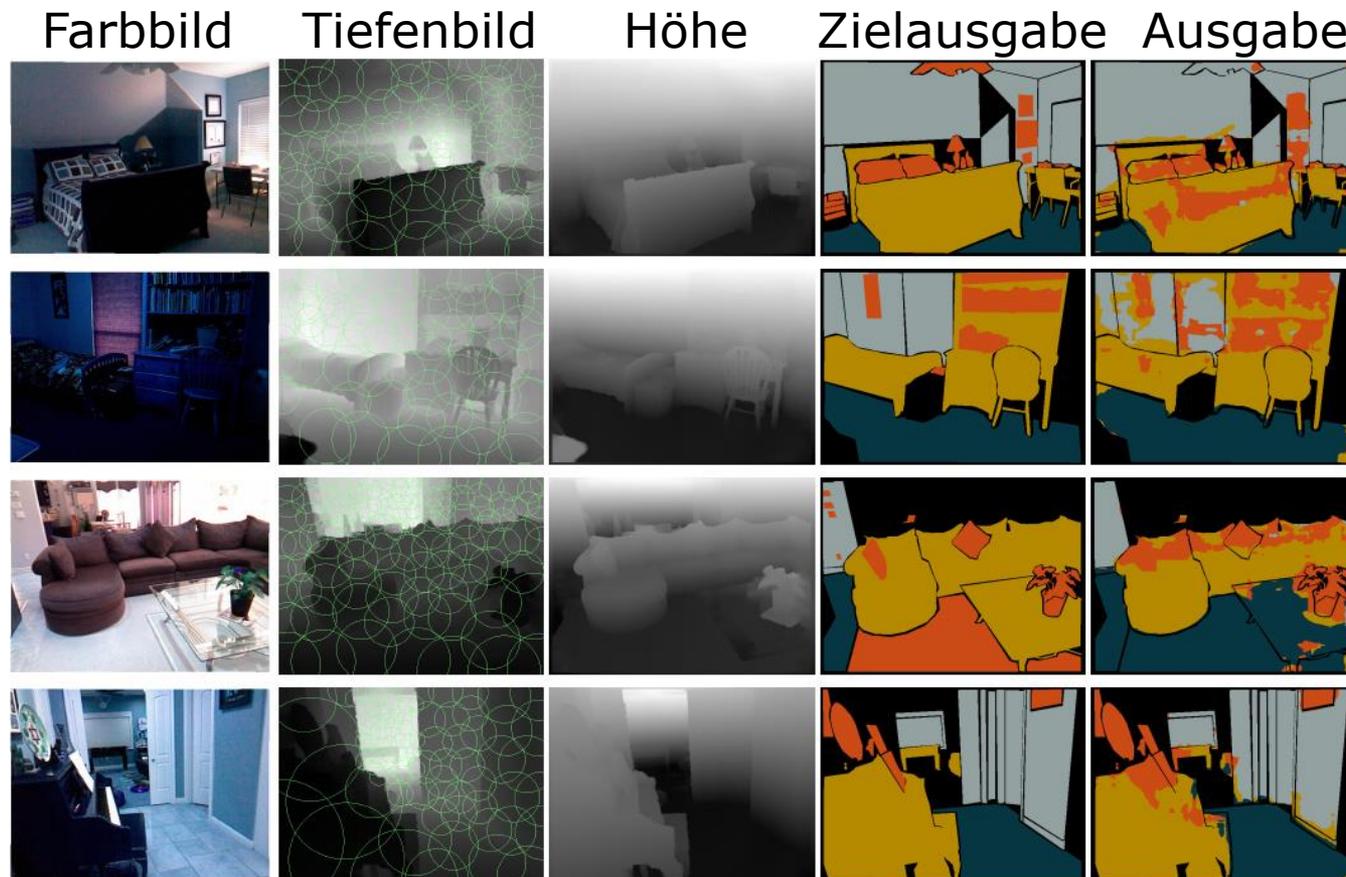
GT: yellow lady's slipper
1: yellow lady's slipper
 2: slug
 3: hen-of-the-woods
 4: stinkhorn
 5: coral fungus



[He et al. 2015]

Kategorisierung von Oberflächen

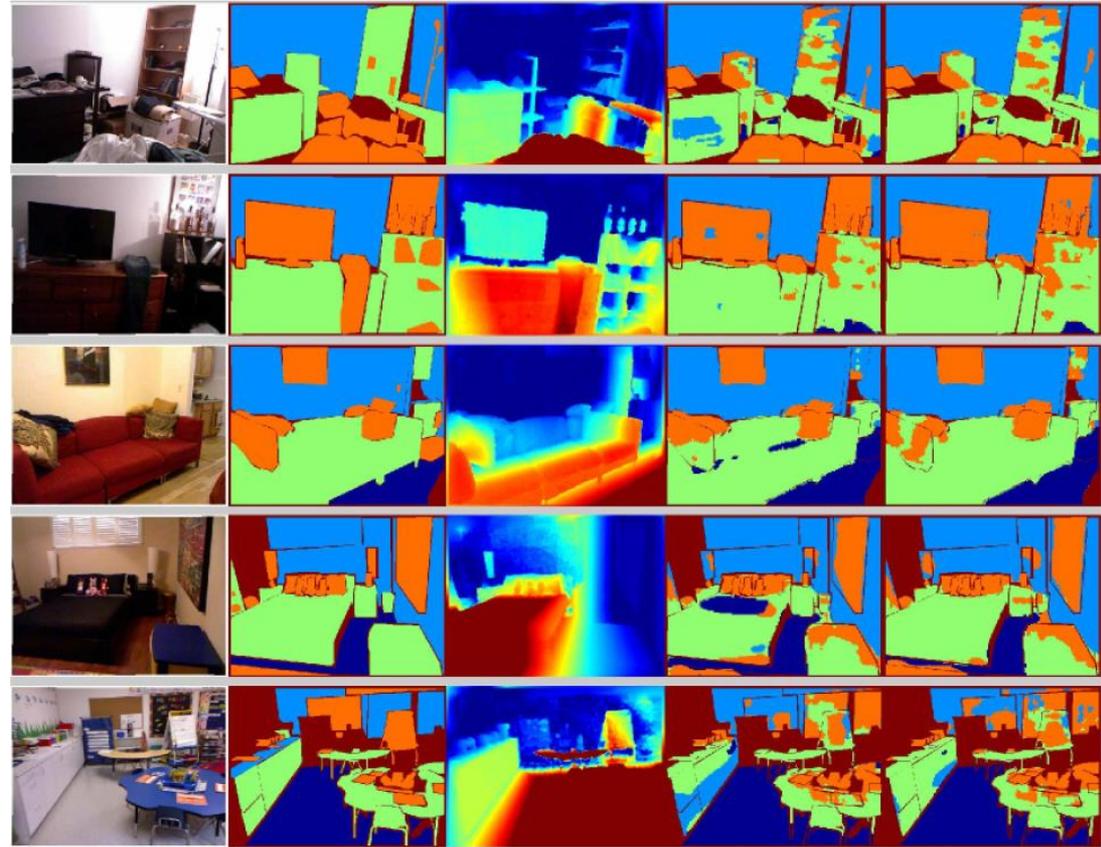
- RGB-D-Eingaben, Höhe über Boden geschätzt
- Skalierung der Eingabe mit der Tiefe



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

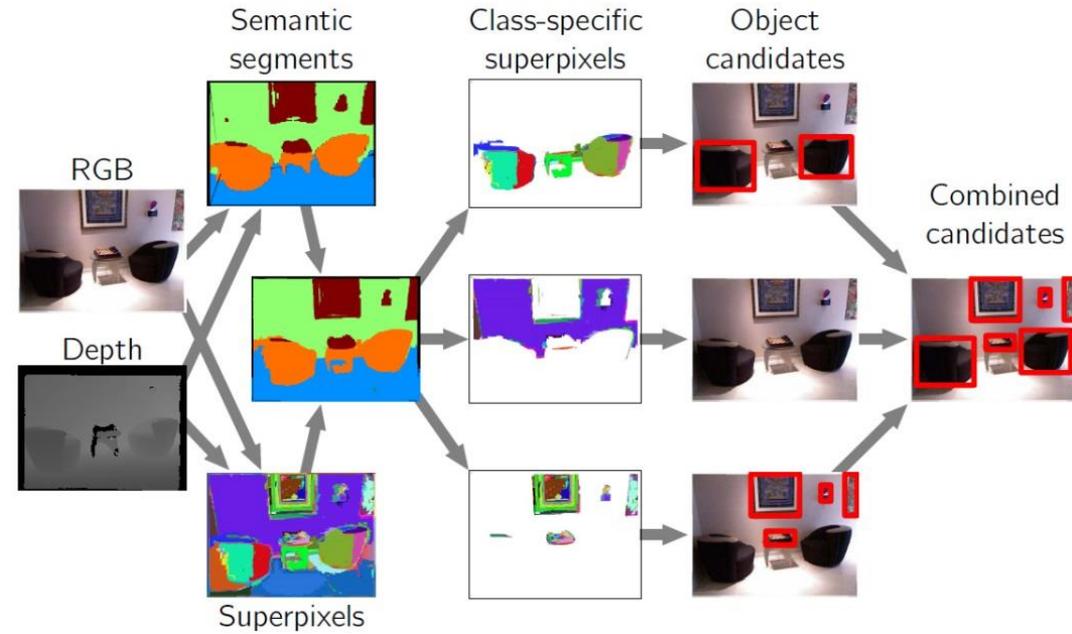
DistWall

OutWO

OutWithDist

Semantische Segmentierung als Prior für Objektdetektion

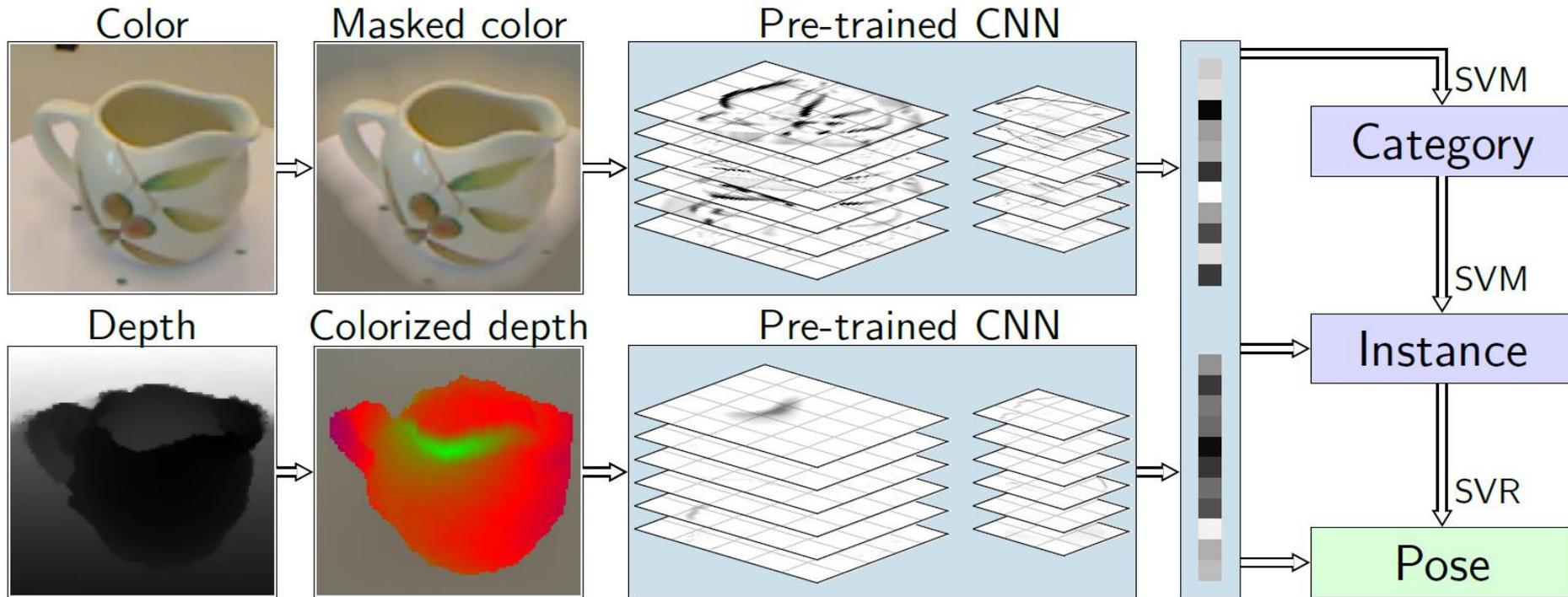
- Combine bottom-up object discovery and semantic priors
- Semantic segmentation used to classify color and depth superpixels
- Higher recall, more precise object borders



[Garcia et al. ICPR 2016]

RGB-D Objekterkennung und Posenschätzung

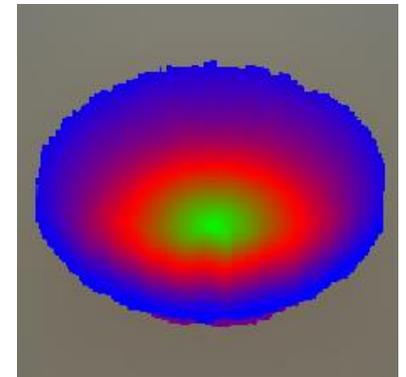
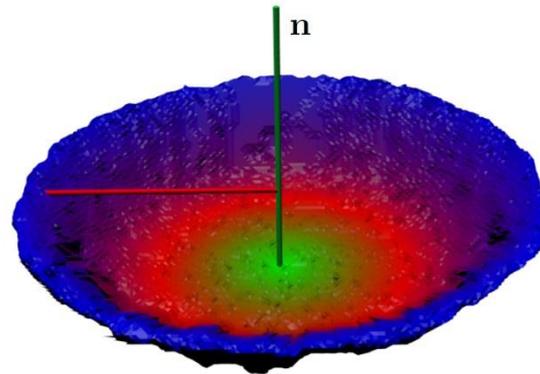
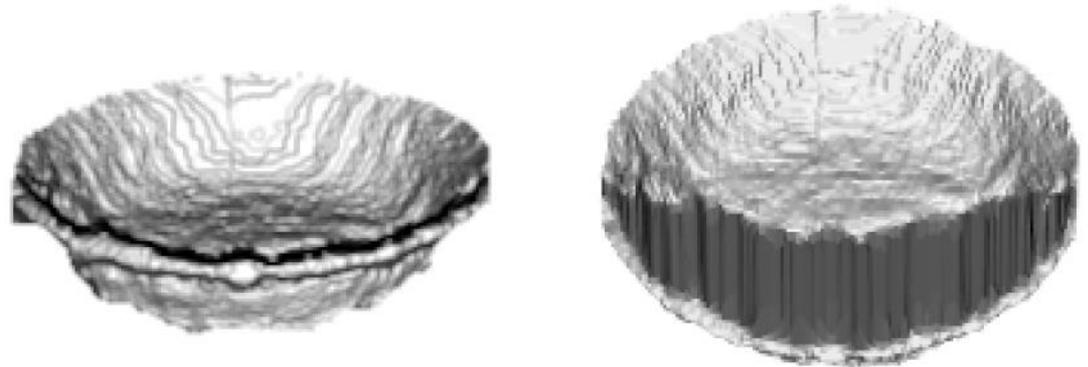
- Vortrainierte Merkmale von ImageNet



[Schwarz, Schulz, Behnke, ICRA2015]

Kanonische Ansicht, Einfärbung

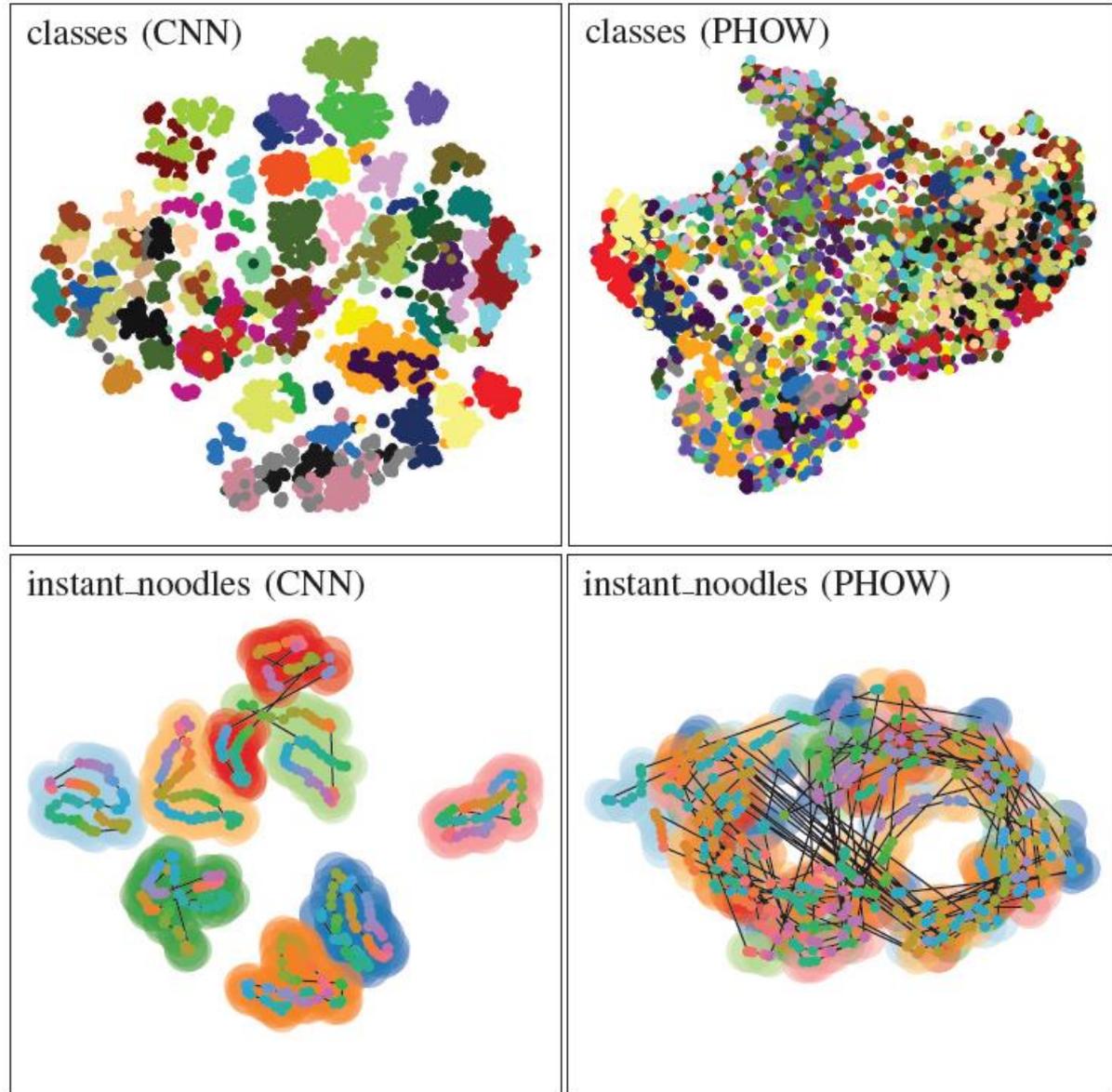
- Objektansichten aus verschiedenen Höhen
- Rendern einer kanonischen Ansicht
- Einfärbung anhand Distanz von Mittelachse



[Schwarz, Schulz, Behnke, ICRA2015]

Merkmale Entwirren die Daten

- t-SNE Einbettung



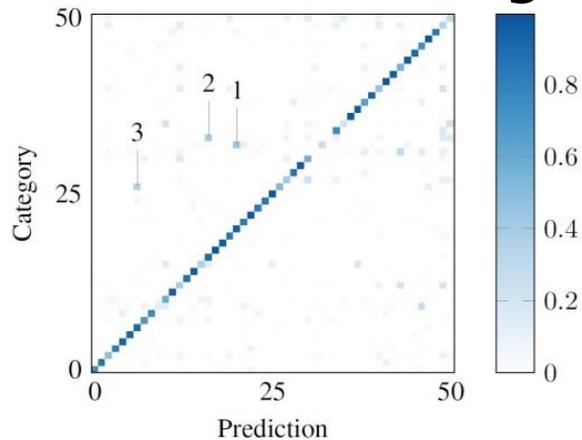
[Schwarz, Schulz,
Behnke, ICRA2015]

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	92.1	92.8
PHOW[3]	80.2 ± 1.8	—	62.8	—
Ours	83.1 ± 2.0	88.3 ± 1.5	92.0	94.1
Ours	83.1 ± 2.0	89.4 ± 1.3	92.0	94.1

■ Verwechslungen



1: Kännchen / Kaffetasse

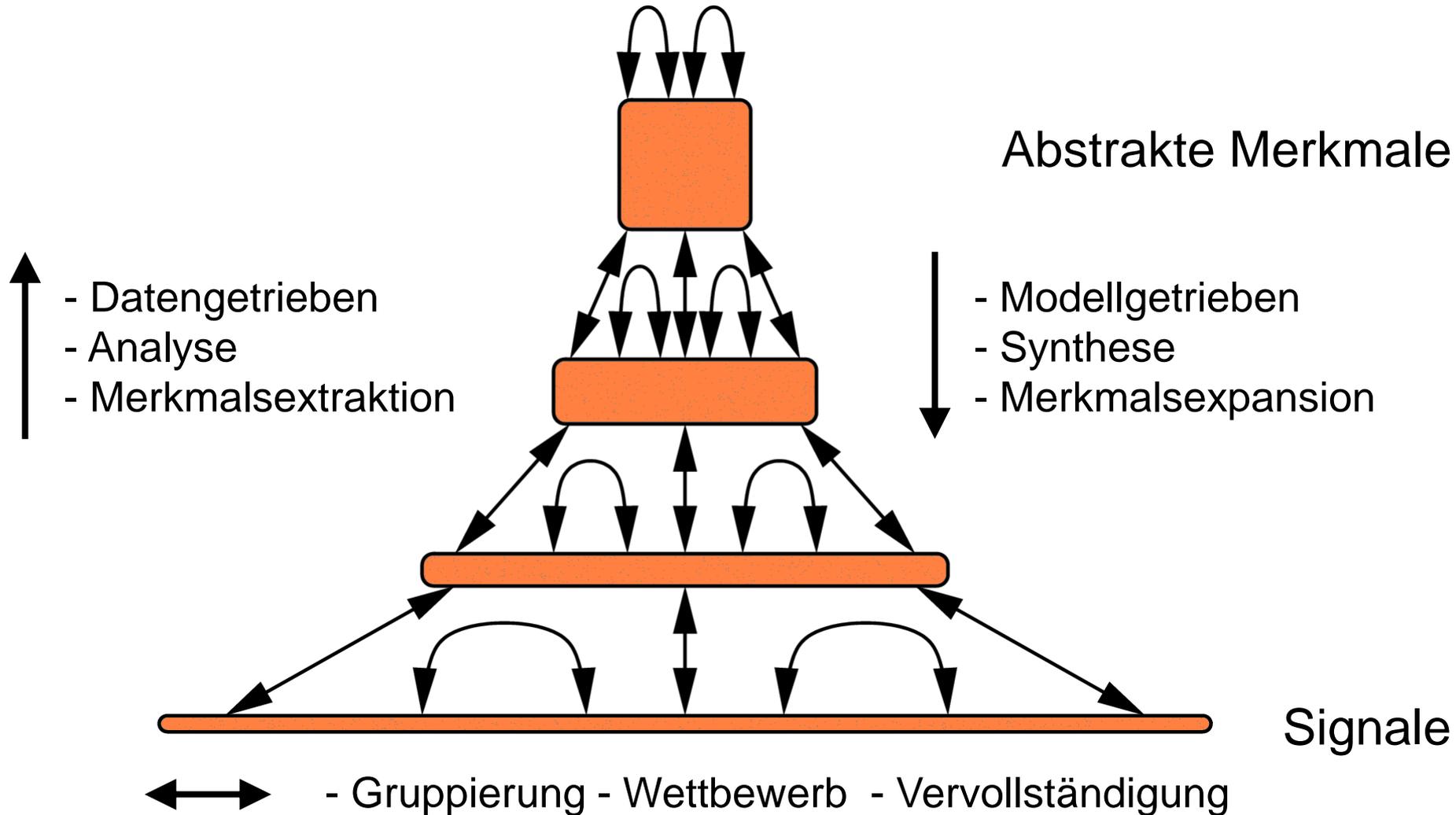


2: Pfirsich / Schwamm



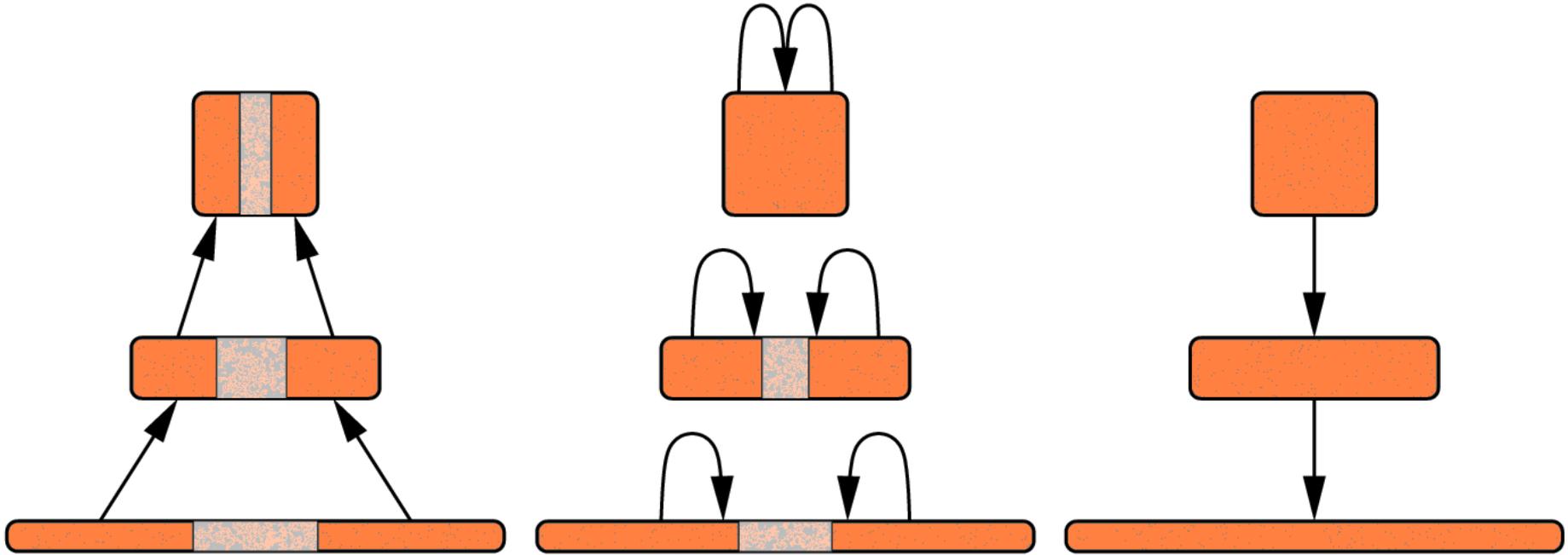
[Schwarz, Schulz, Behnke, ICRA2015]

Neurale Abstraktionspyramide



Iterative Interpretation

- Interpretiere die einfachsten Stellen zuerst



- Nutze Teil-Interpretationen als Kontext zur Auflösung von Mehrdeutigkeiten

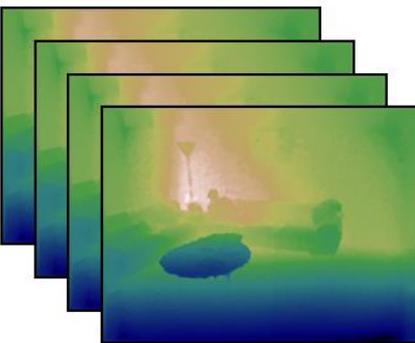
Objektklassensegmentierung von RGB-D-Video

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

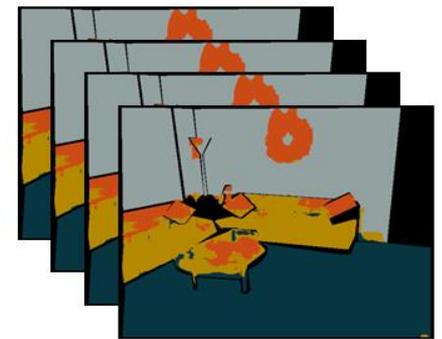
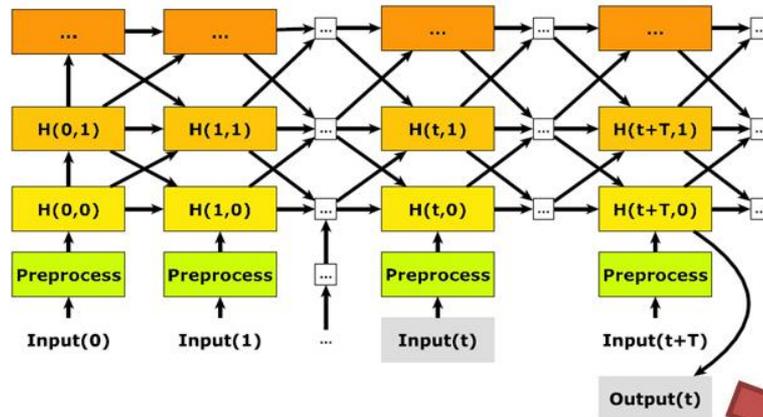
RGB



Neuronale Abstraktionspyramide



Tiefe



Ausgabe

- Rekursive Berechnung ermöglicht effiziente zeitliche Integration

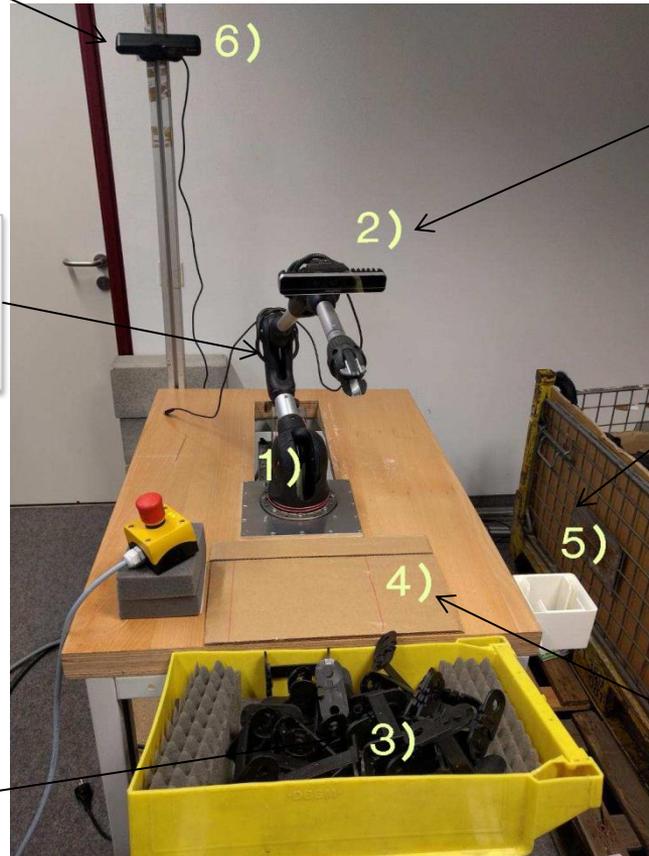
[Pavel, Schulz, Behnke, Neural Networks 2017]

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

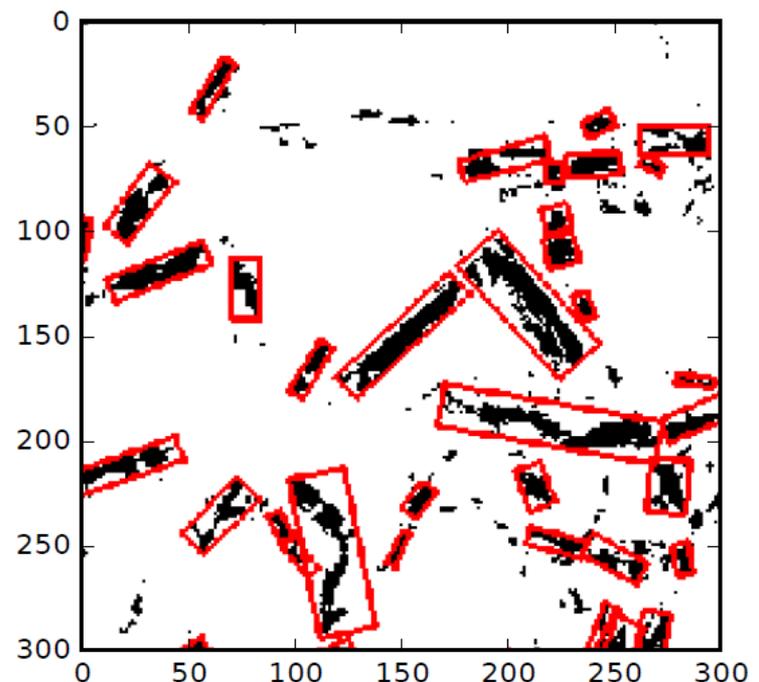
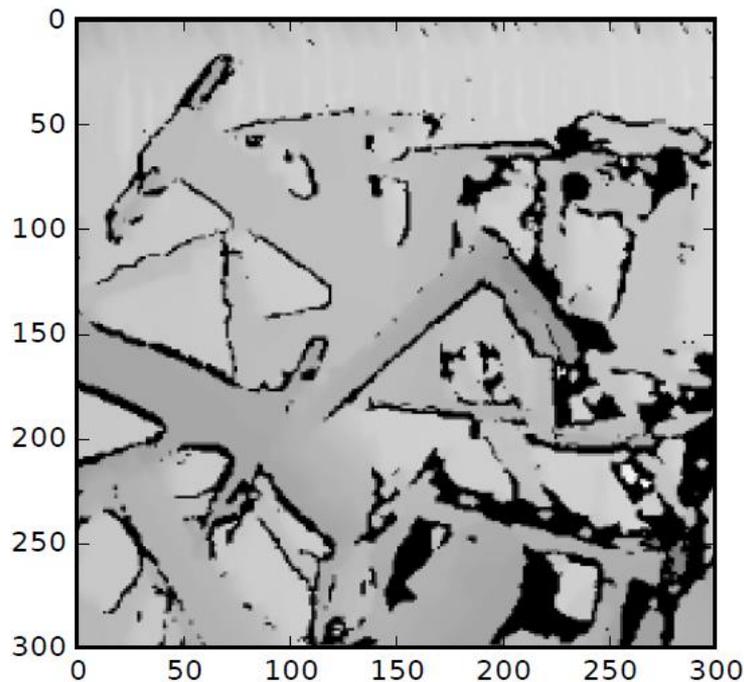
Energy chain part feeder

Place for regrasping

[Koo et al. CASE 2017]

Detektion von Greifgelegenheiten

- Analyse des Tiefenbilds
- Detektion schmaler hervorstechende Bereiche



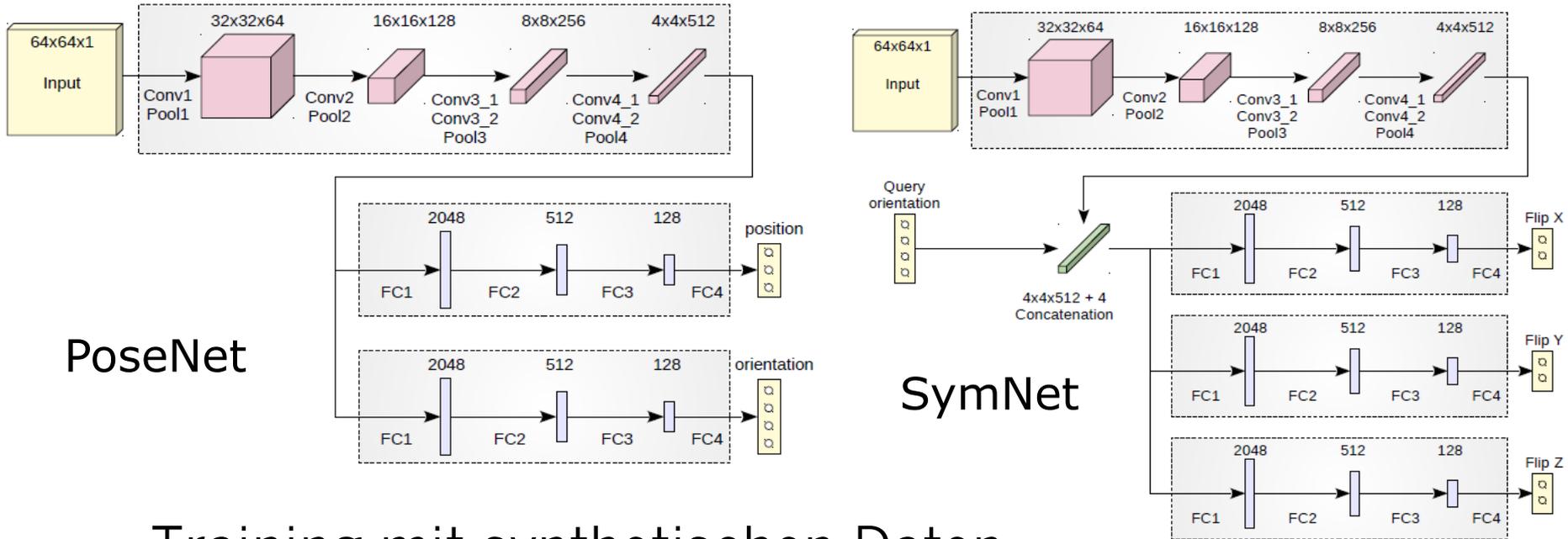
Robolink Feeder: Griff in die Kiste

Part detection

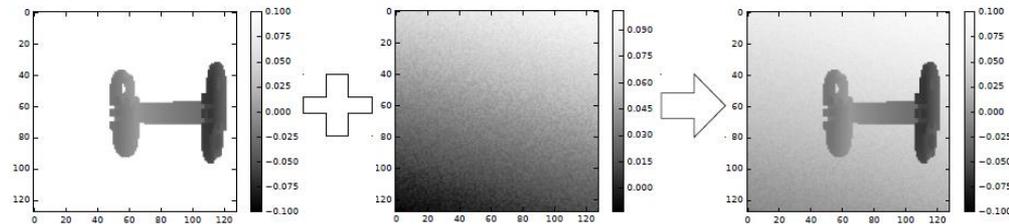


Posenschätzung für Teile

- Zwei konvolutionale neuronale Netze



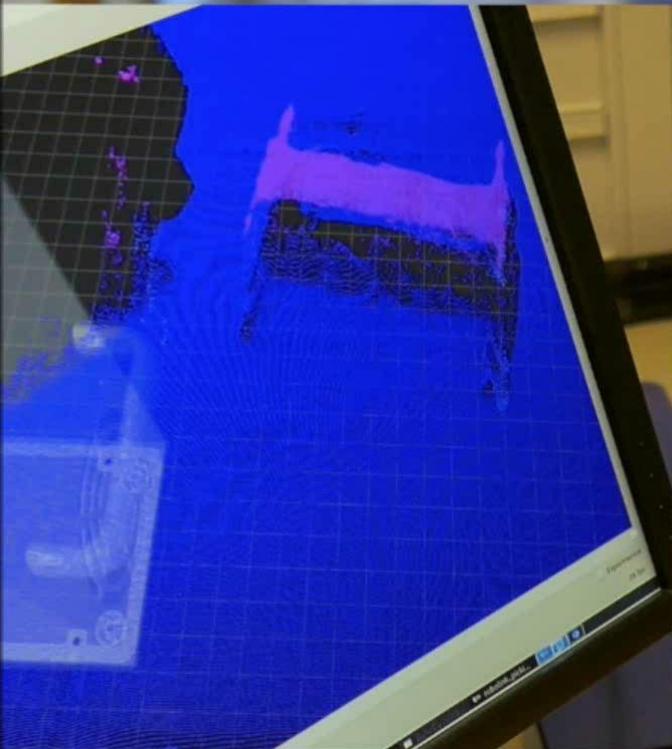
- Training mit synthetischen Daten



[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

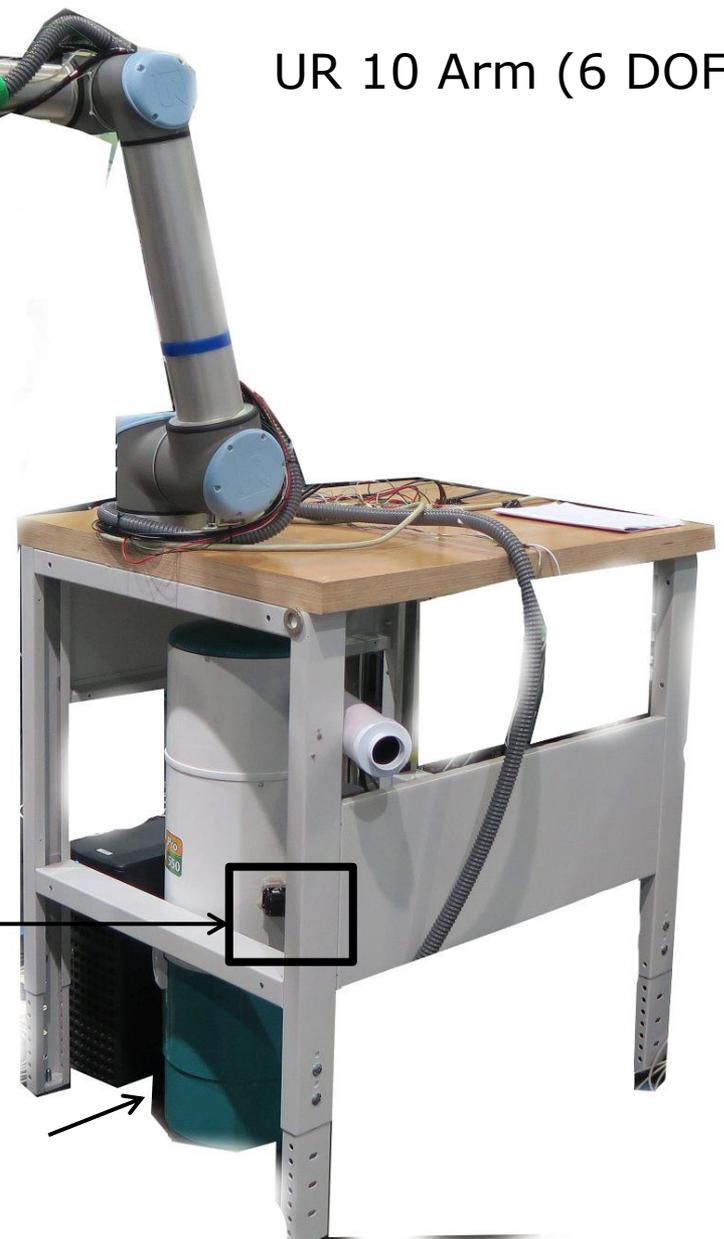
Linear actuator

Total:
6+2 DOF

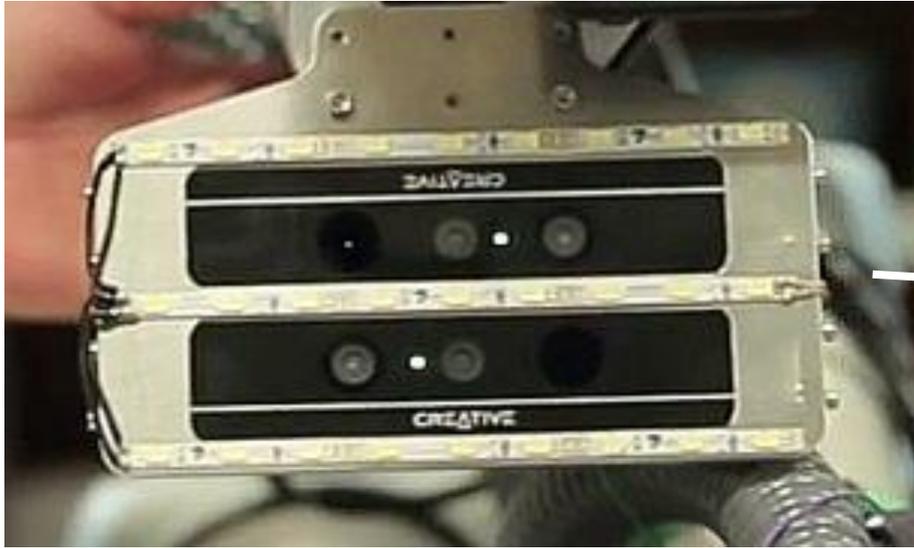
Suction strength control

Strong vacuum
cleaner (3100 W)

[Schwarz et al. ICRA 2017]

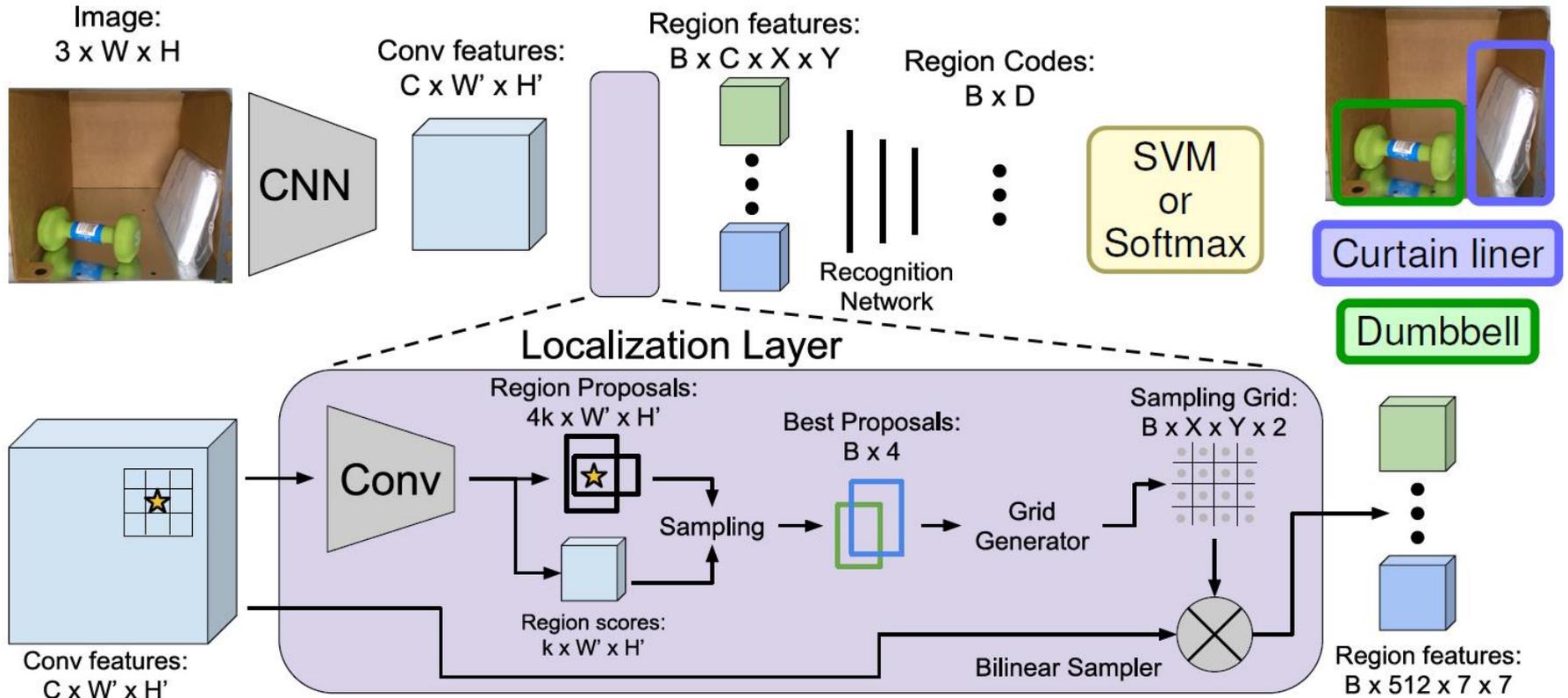


RGB-D-Kameras



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

Objektdetektion



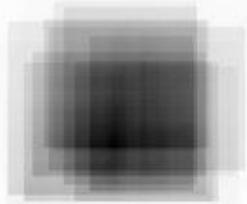
[Adapted from Johnson et al. CVPR 2016]

[Schwarz et al. ICRA 2017]

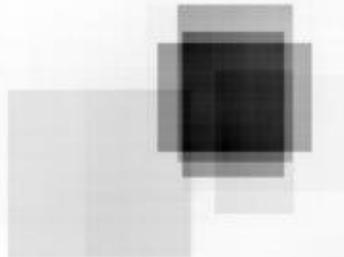
Beispiel-Detektionen



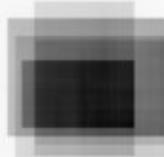
Gloves



Glue sticks



Sippy cup

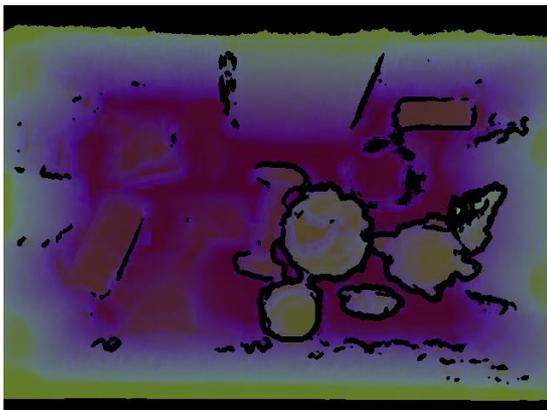


[Schwarz et al. ICRA 2017]

Semantische Segmentierung

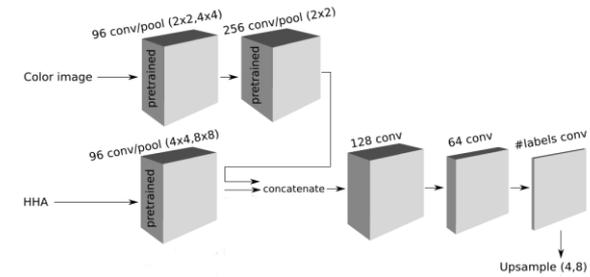
■ Tiefes Konvolutionales Netzwerk

RGB



HHA

Ausgabe



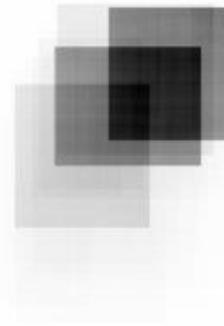
[Husain et al. RA-L 2016]

Kombination von Detektion und Segmentierung

- Pixelweise Multiplikation



Detektion

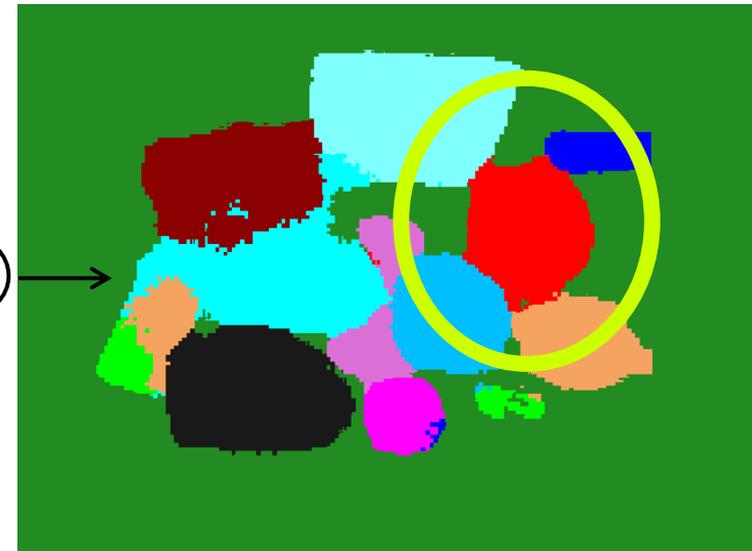


Segmentierung



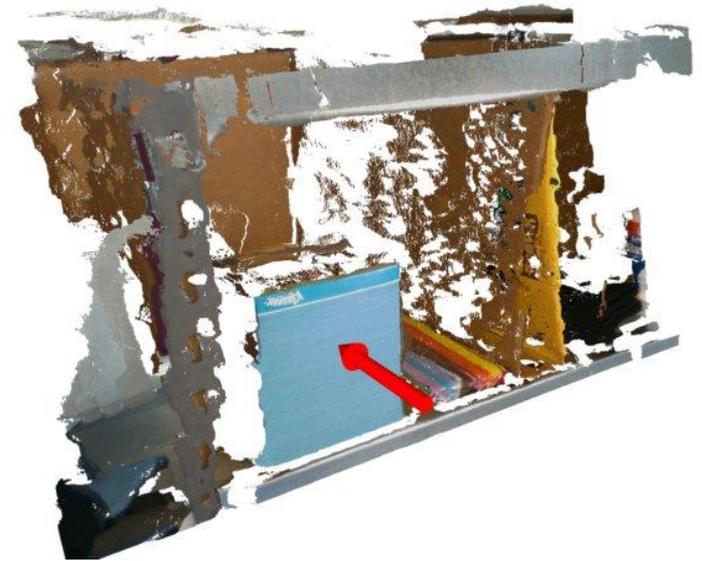
⊗

Ausgabe

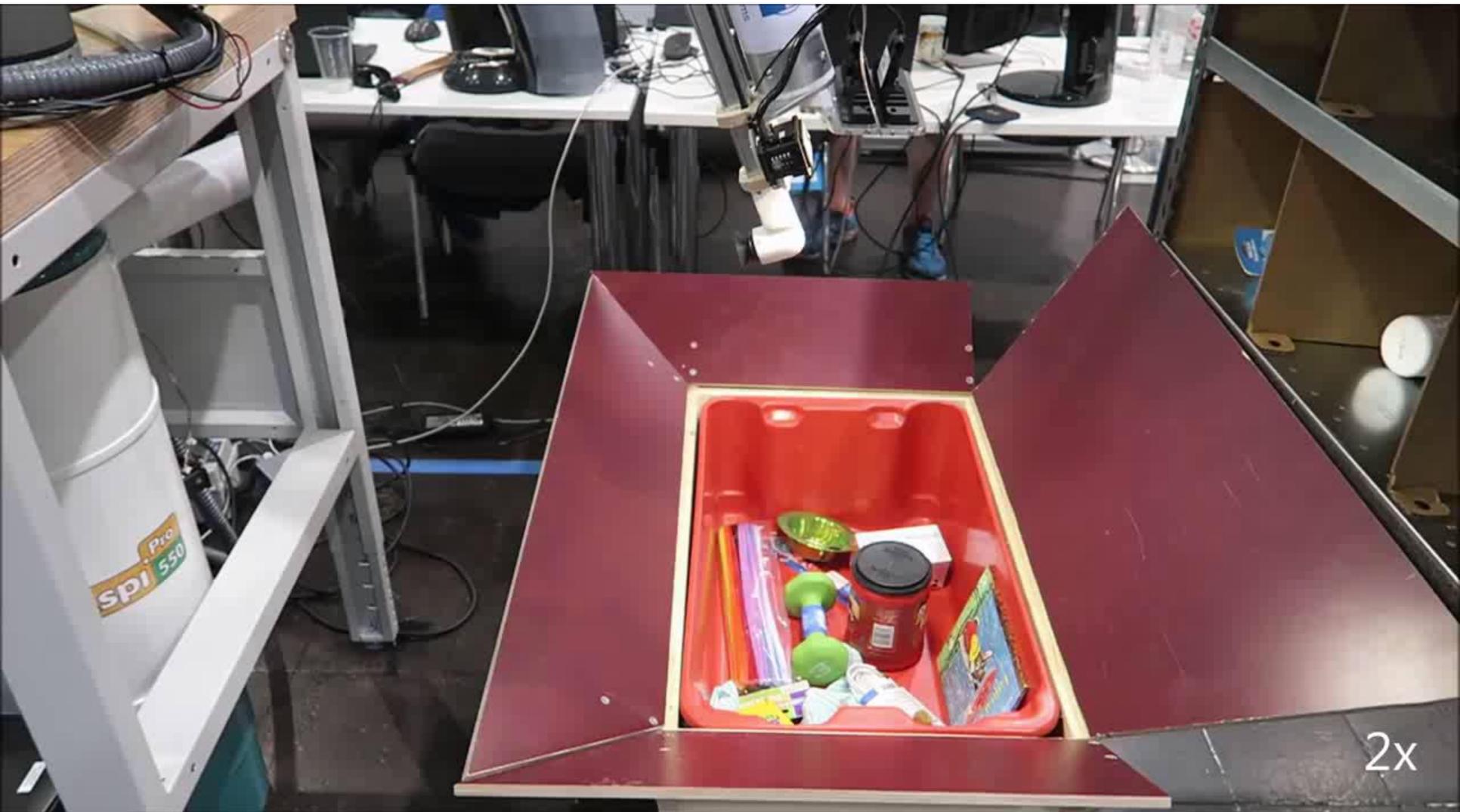


Auswahl der Saugpose

- 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



6D-Posenschätzung

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



Schwierige Produkte

- Heavy / Large

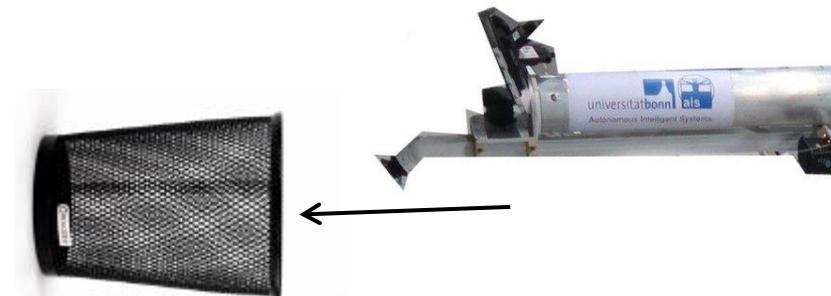
- Many holes / Meshes



Grasp one ball



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



Knock over and suck on bottom

Greifen des Stiftbechers

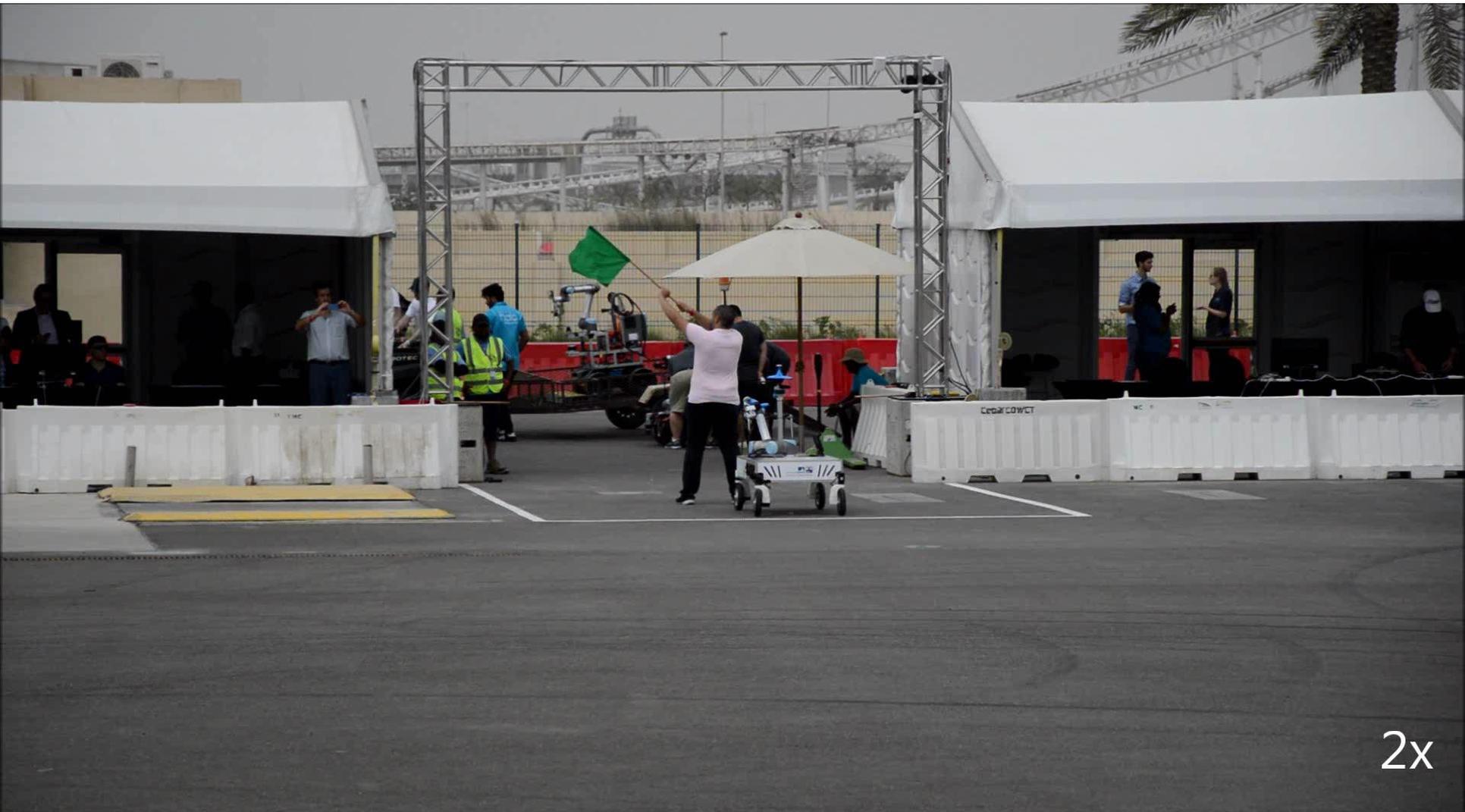


Team NimbRo Picking



- 2nd place stowing, 3rd place picking

MBZIRC Challenge 2



2x

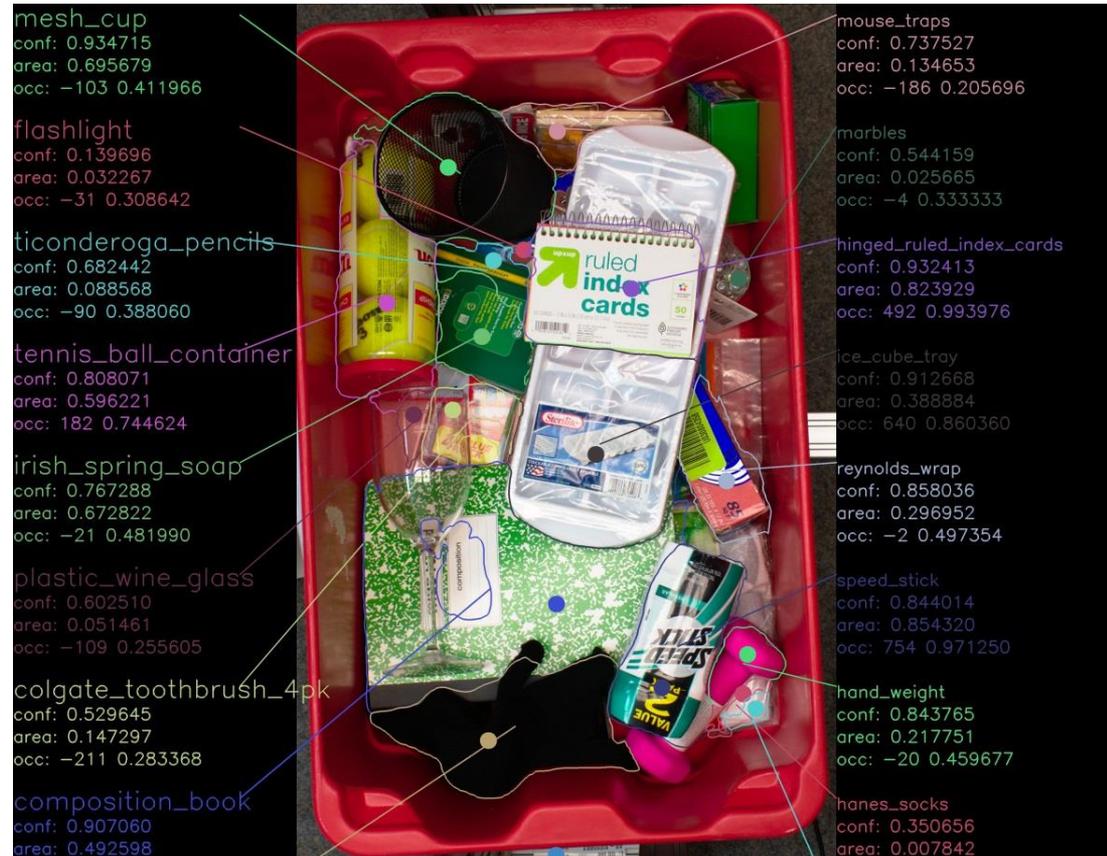
Wahrnehmung der Schraubenschlüssel

- 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?