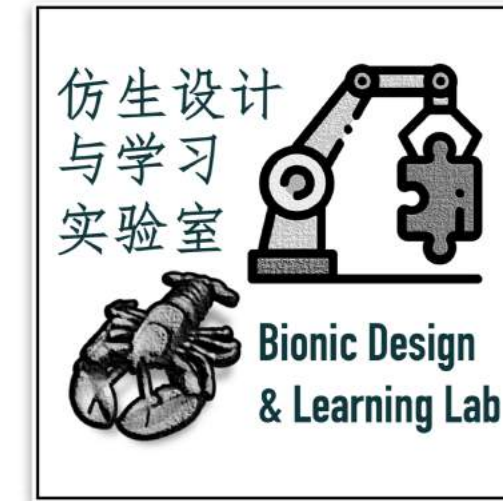


Welcome to ME336



ME336

Collaborative Robot Learning

Week 01 Lecture 1

Wednesday, 1400-1550, Room 235, New Engineering Building

Song Chaoyang | Asst. Prof. | Department of Mechanical & Energy Engineering | SUSTech | songcy@sustech.edu.cn

Agenda

Week 01, Wednesday January 13, 2021



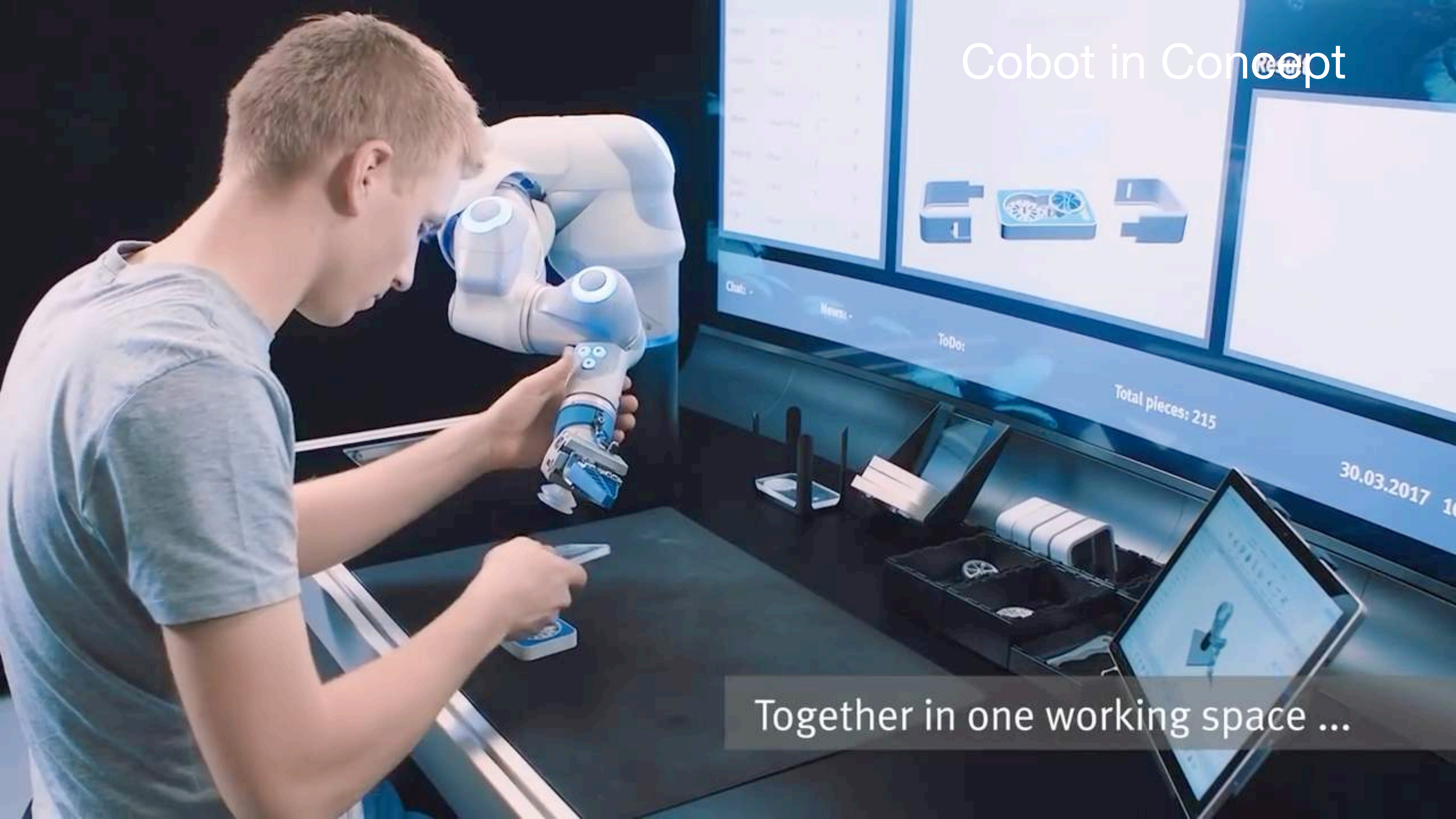
Collaborative Robots at Work

Robot Learning in Research

A Brief Review of Robot Manipulation Learning Problems

Course Overview

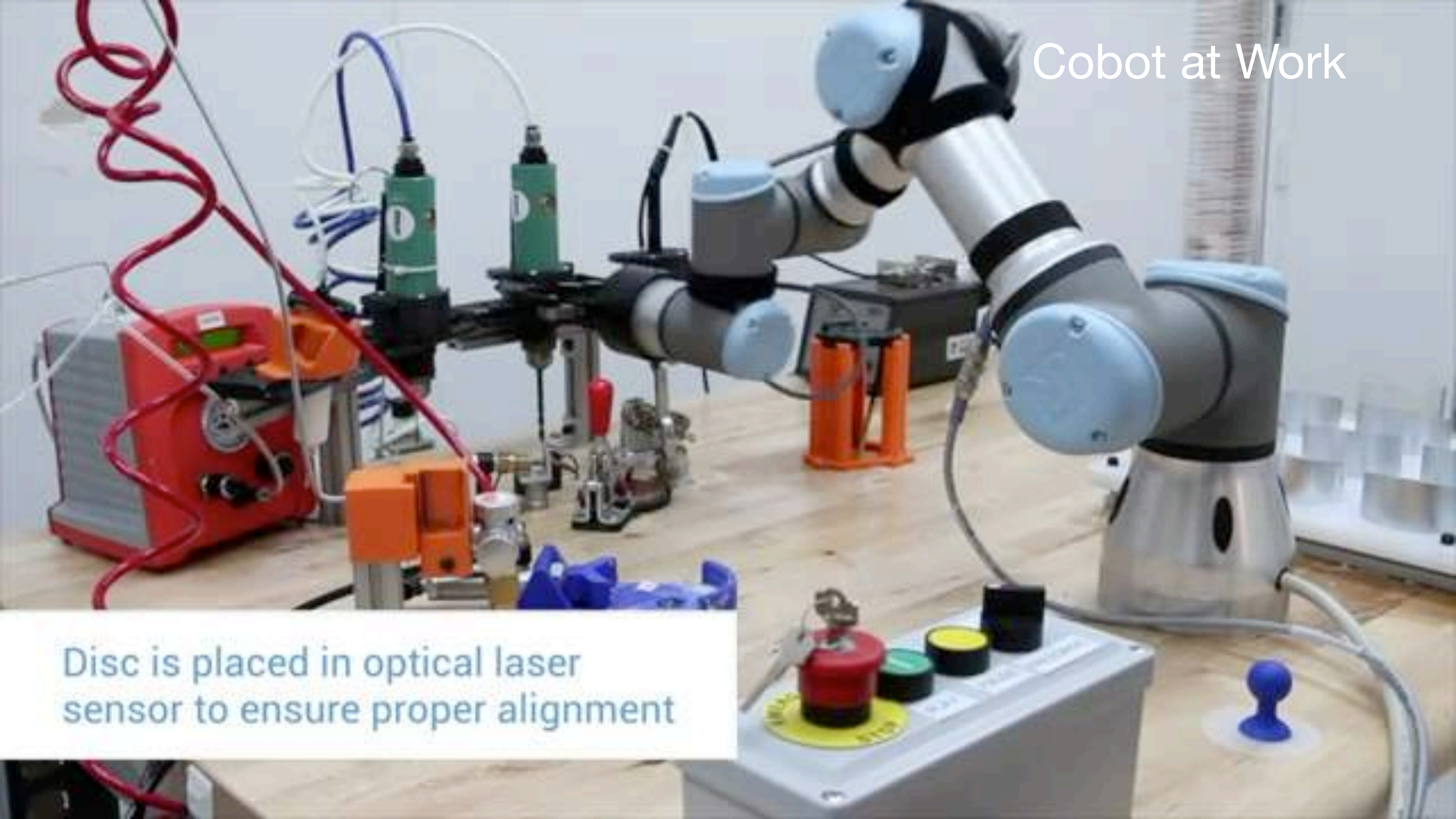
Cobot in Concept



Together in one working space ...

Cobot at Work

Disc is placed in optical laser sensor to ensure proper alignment



Common Applications of Cobot in Automation

Highly repetitive tasks that require different levels of dexterity

- **Object Relocation**

- Handling object from one location to another
- Pick & Place | Machine Tending | Packing and Palletizing

- **Material Releasing**

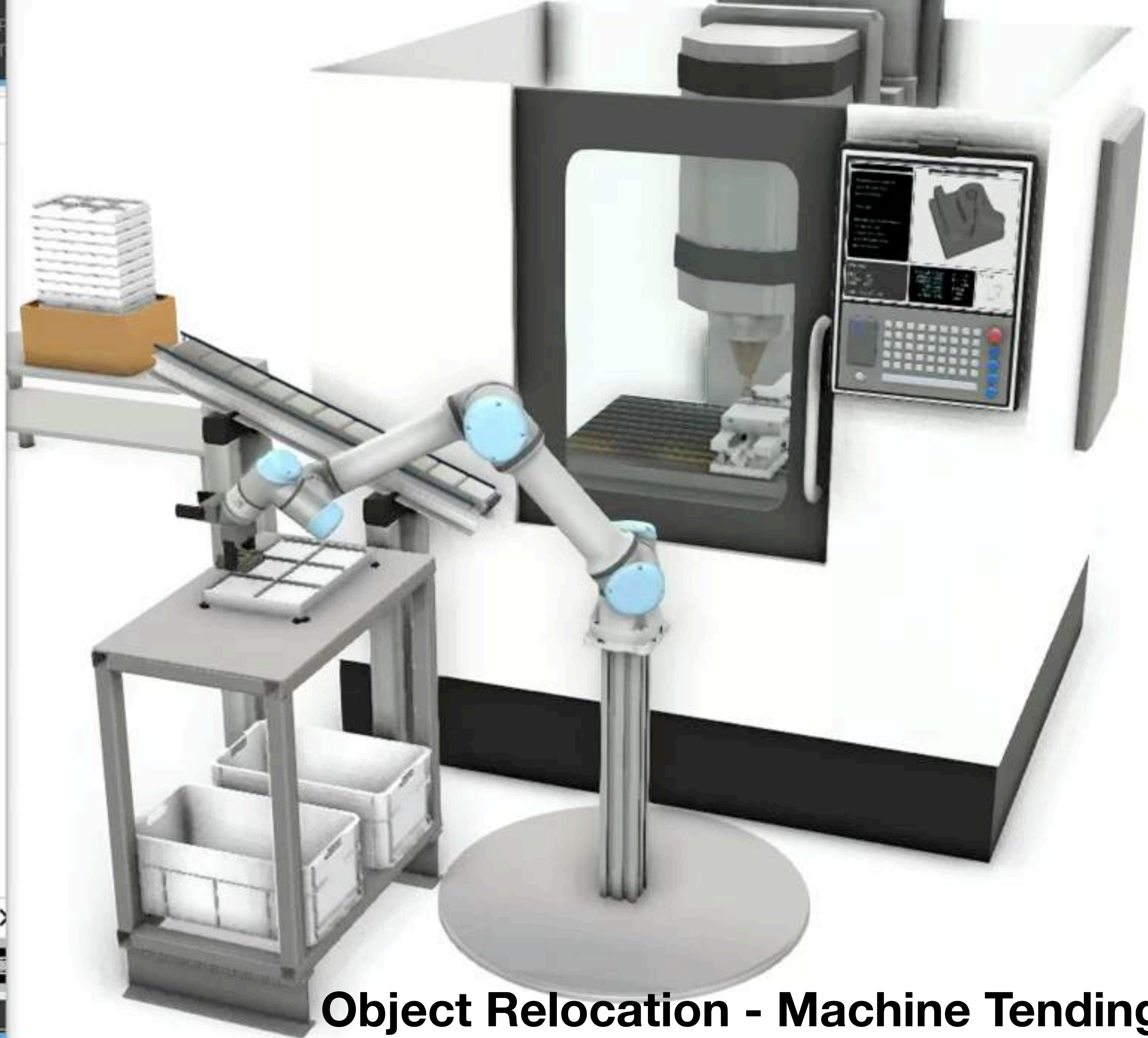
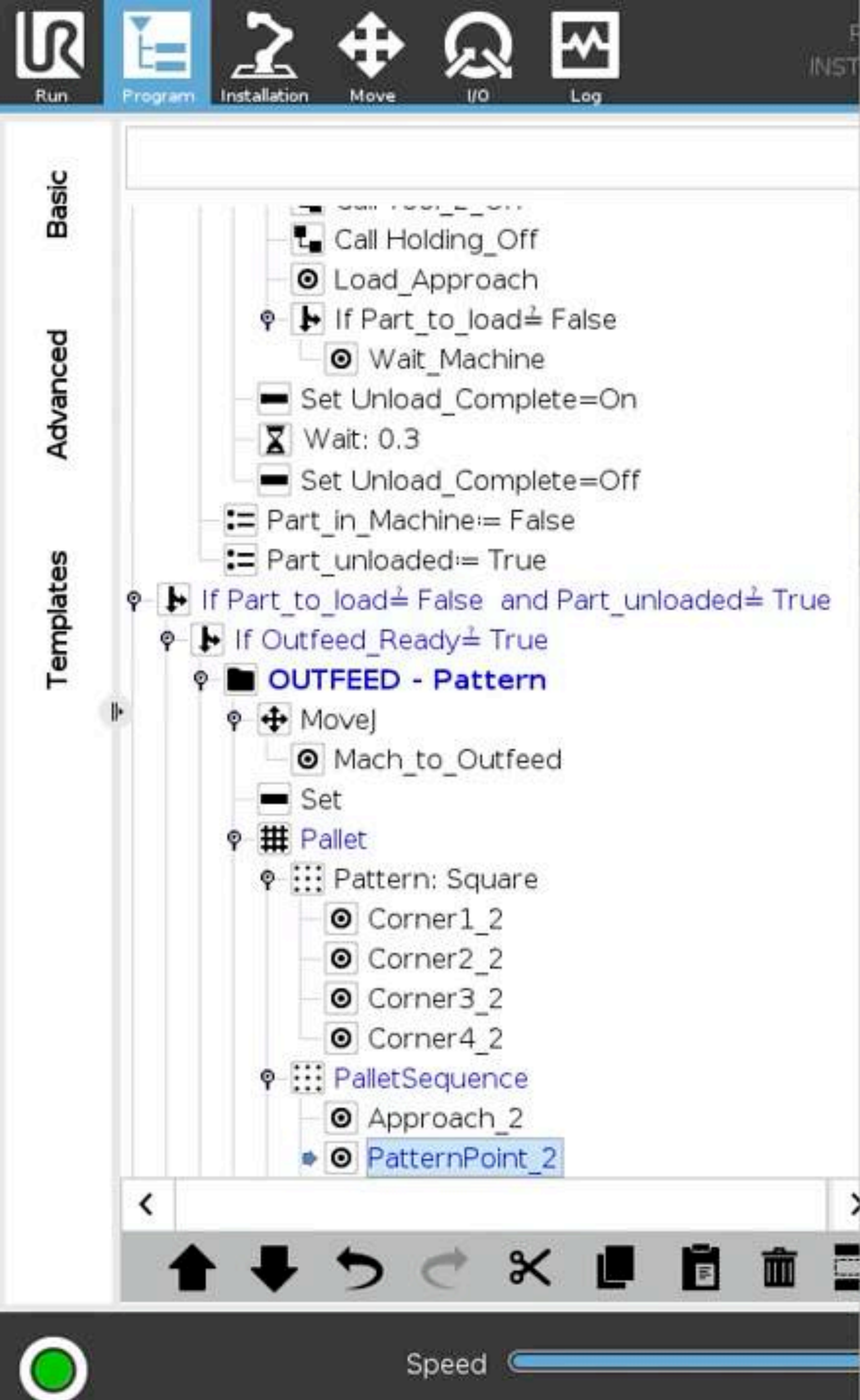
- Releasing material from the robot to the target location
- Gluing | Dispensing | Welding | Screwdriving

- **Material Removal**

- Removing material from the target object using the robot
- Polishing | Grinding | Deburring

- **Information Gathering**

- Collecting information using sensors attached to the robot
- Quality Inspection



Object Relocation - Machine Tending

Object Relocation - Pick & Place

Intel® RealSense™
Depth Camera D415

Universal Robots
e-Series cobots

RHR GripperV5

2nd Gen
RightPick.AI

Common Applications of Cobot in Automation

Highly repetitive tasks that require different levels of dexterity

- Object Relocation

- Handling object from one location to another
- Pick & Place | Machine Tending | Packing and Palletizing

- **Material Releasing**

- Releasing material from the robot to the target location
- Gluing | Dispensing | Welding | Screwdriving

- Material Removal

- Removing material from the target object using the robot
- Polishing | Grinding | Deburring

- Information Gathering

- Collecting information using sensors attached to the robot
- Quality Inspection

UR Run Program Installation Move I/O Log

Basic
Init Variables
Robot Program

Advanced
■ SCREW PRESENTER
MoveL
Presenter_Appro
Wait ScrPresenterRdy = True and joint_nu
Presenter_Screw
Set ScrDriv_Reverse=On
Wait: 0.25
Set ScrDriv_Reverse=Off
Presenter_Exit

Templates
■ SCREW CHECK
MoveL
Check_Screw
If ScrewSensor = True
screw_on_bit := True
Else
screw_on_bit := 'Screw not detected on bit.'

URCaps
■ SCREWDRIVING
If screw_on_bit = True
MoveL
Check_Exit
If joint_number = 1
Switch screw_number

Speed

Material Removal - Screwdriving



Cobots, or collaborative robots, are robots intended for direct human robot interaction within a shared space, or where humans and robots are in close proximity. Cobot applications contrast with traditional industrial robot applications in which robots are isolated from human contact.

Collaborative Robot Learning

Common Applications of Cobot in Automation

Highly repetitive tasks that require different levels of dexterity

- Object Relocation

- Handling object from one location to another
- Pick & Place | Machine Tending | Packing and Palletizing

- Material Releasing

- Releasing material from the robot to the target location
- Gluing | Dispensing | Welding | Screwdriving

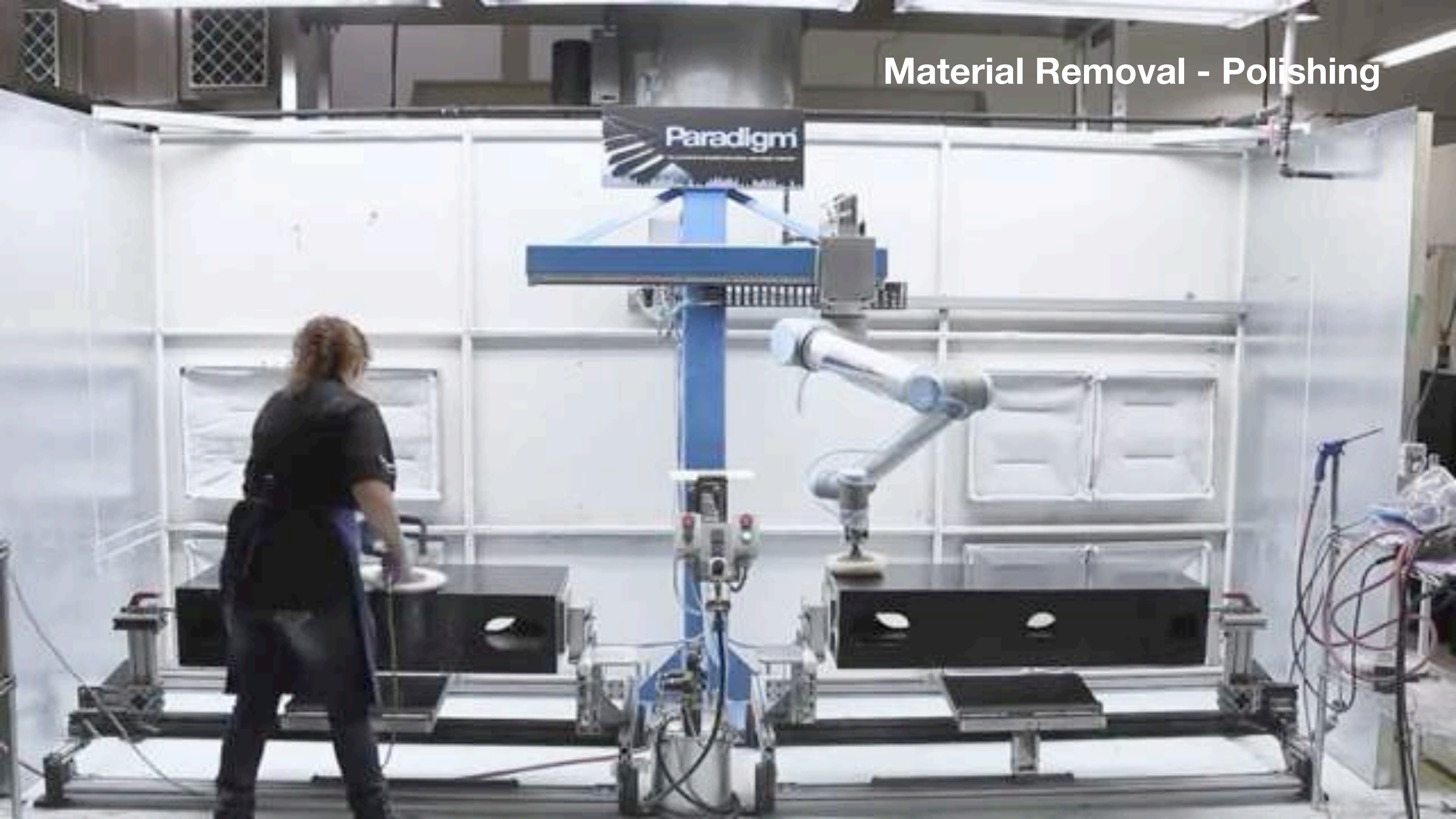
- Material Removal

- Removing material from the target object using the robot
- Polishing | Grinding | Deburring

- Information Gathering

- Collecting information using sensors attached to the robot
- Quality Inspection

Material Removal - Polishing



Common Applications of Cobot in Automation

Highly repetitive tasks that require different levels of dexterity

- Object Relocation

- Handling object from one location to another
- Pick & Place | Machine Tending | Packing and Palletizing

- Material Releasing

- Releasing material from the robot to the target location
- Gluing | Dispensing | Welding | Screwdriving

- Material Removal

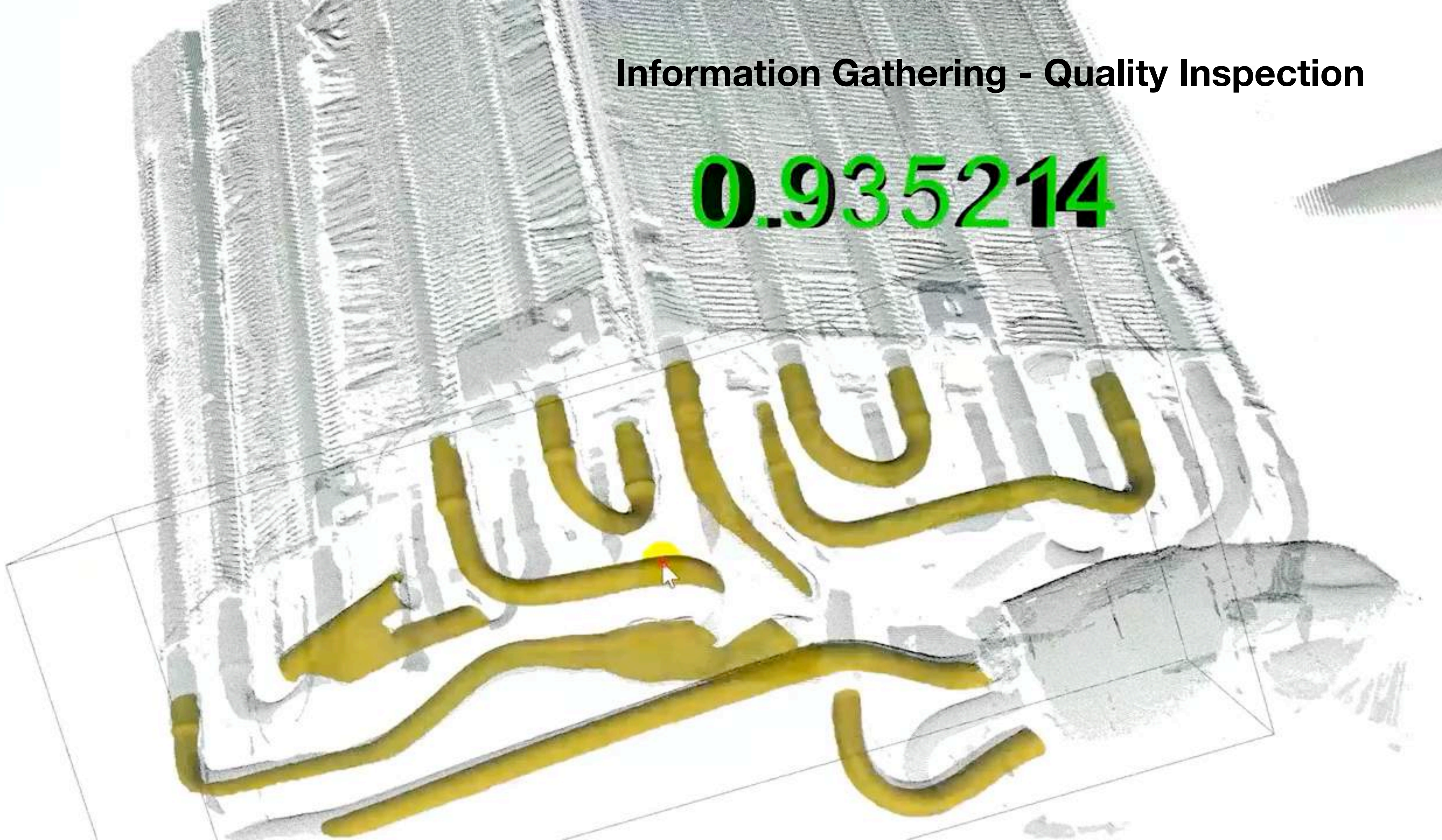
- Removing material from the target object using the robot
- Polishing | Grinding | Deburring

- Information Gathering

- Collecting information using sensors attached to the robot
- Quality Inspection

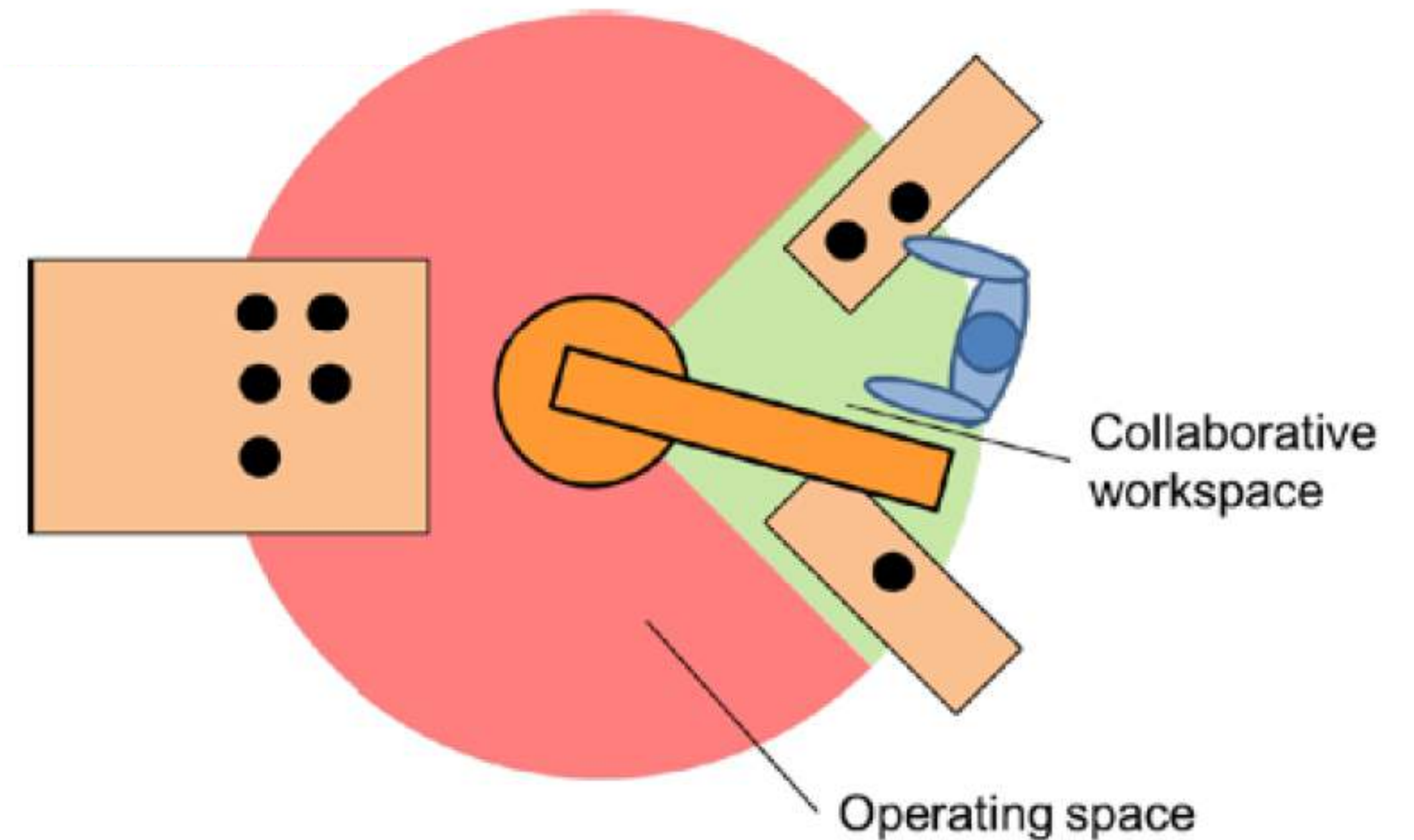
Information Gathering - Quality Inspection

0.935214



How to define Collaboration with Robots?

Collaborative Robot Technical Specification ISO/TS 15066



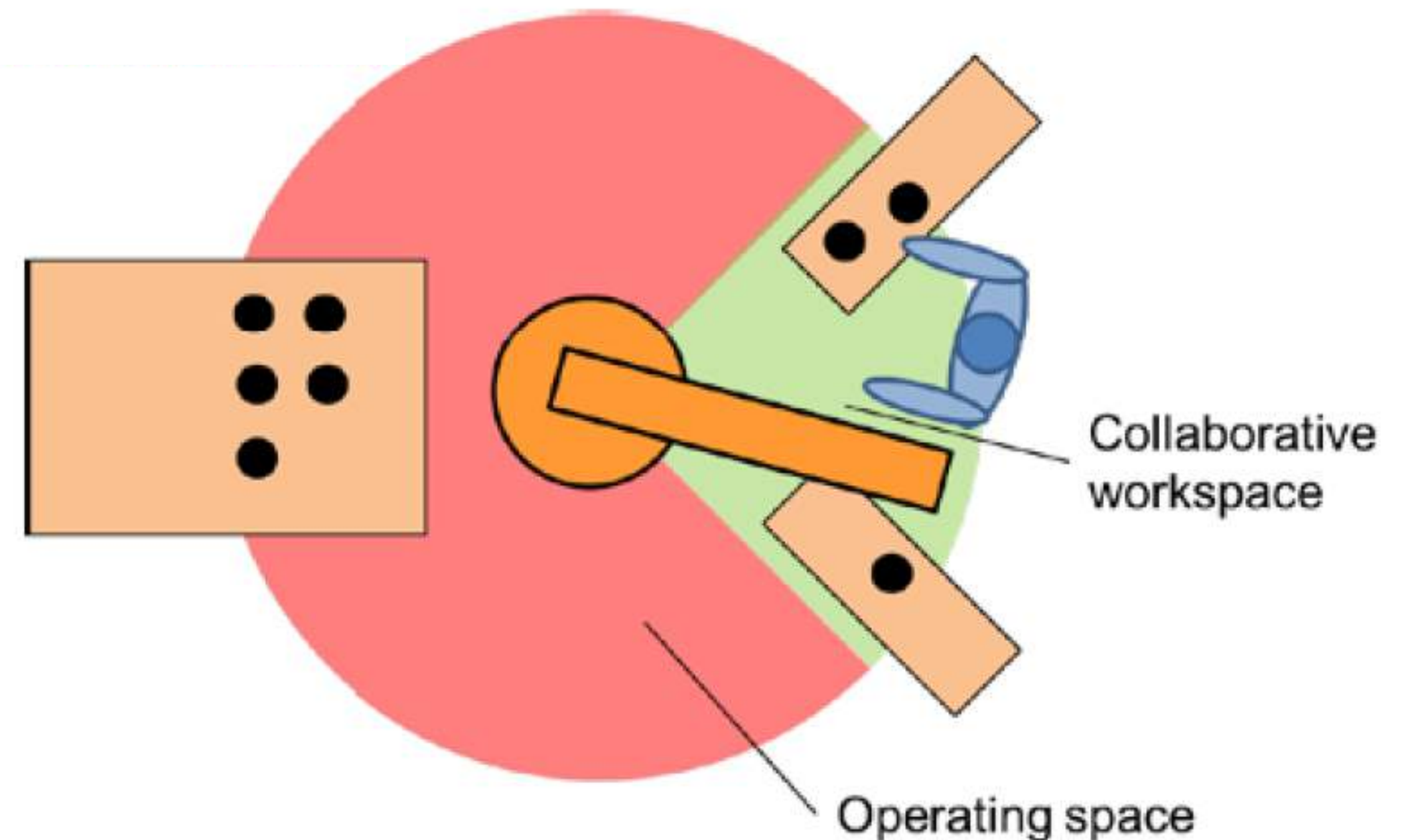
How to define Collaboration with Robots?

Collaborative Robot Technical Specification ISO/TS 15066

- A robot that **CAN** (capable) for use in a collaborative operation
 - purposely designed robot systems work in **direct cooperation with a human** within **a defined workspace**

Collaborative Workspace: space within the operating space where the robot system (including the workpiece) and a human can perform tasks concurrently during production operation.

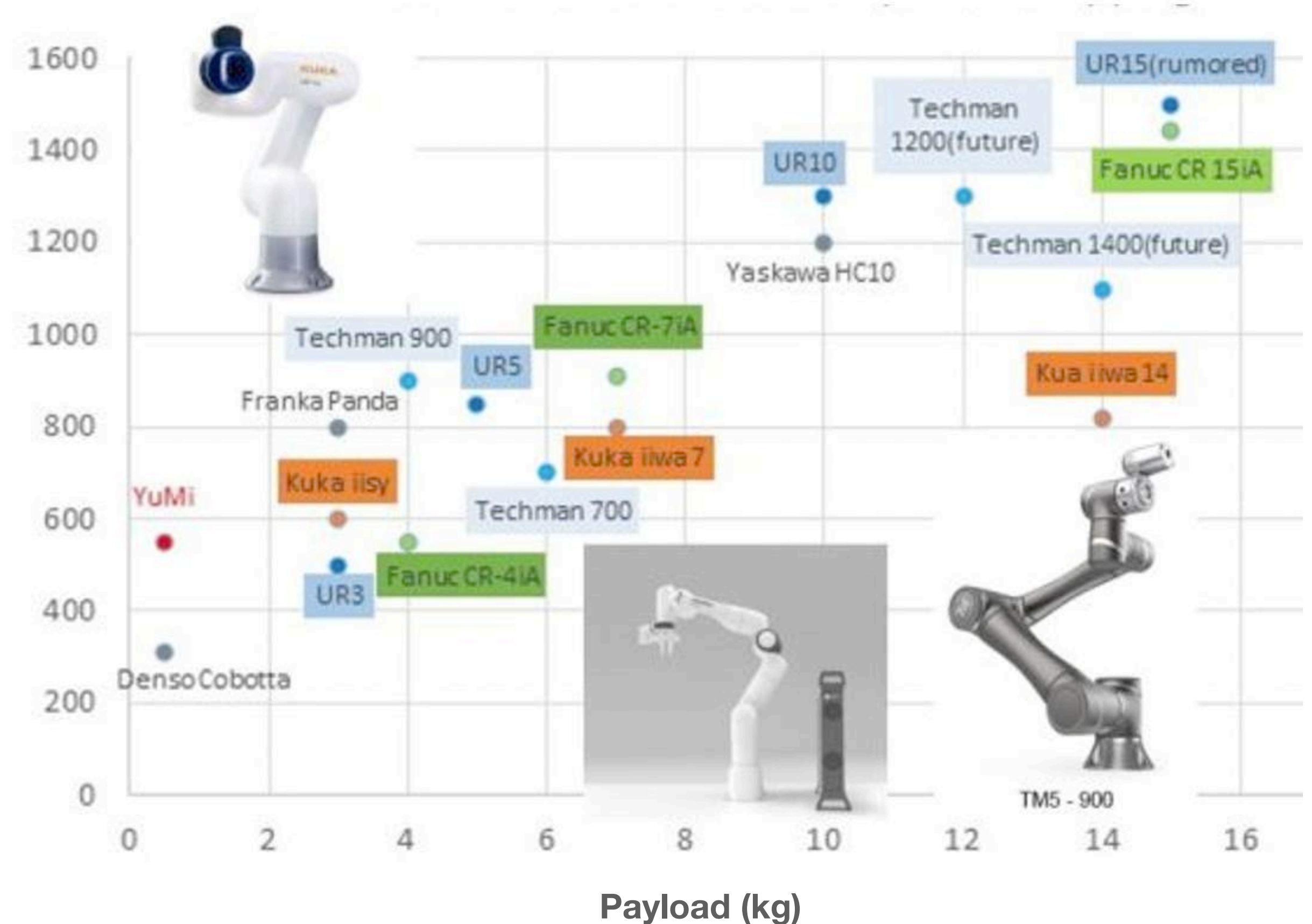
- Robot: Robot arm & robot control
- Robot System: Robot, end-effector & workpiece



Safety vs. Risk against Cost

The Design Need for Robotic Collaboration

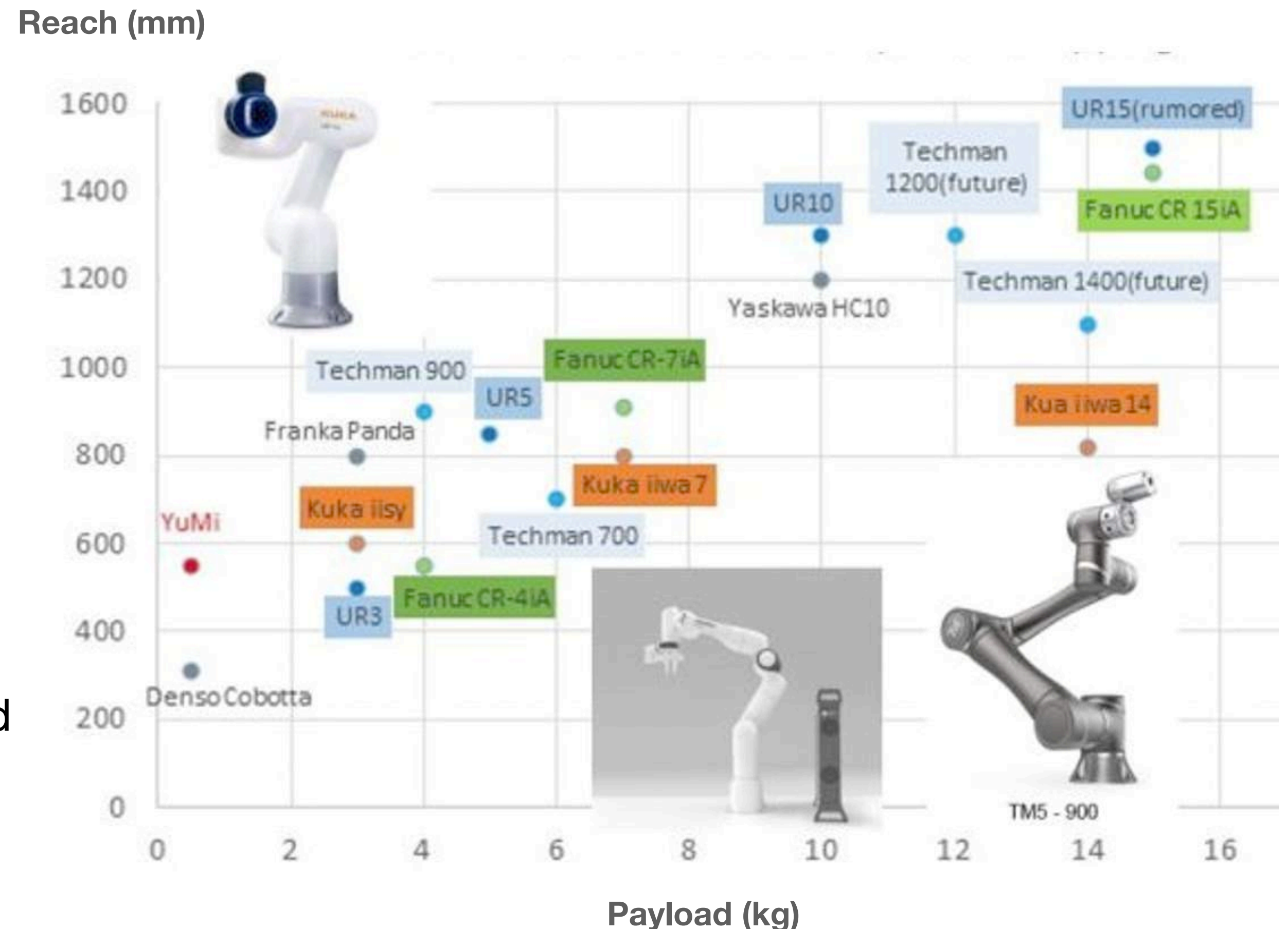
Reach (mm)



Safety vs. Risk against Cost

The Design Need for Robotic Collaboration

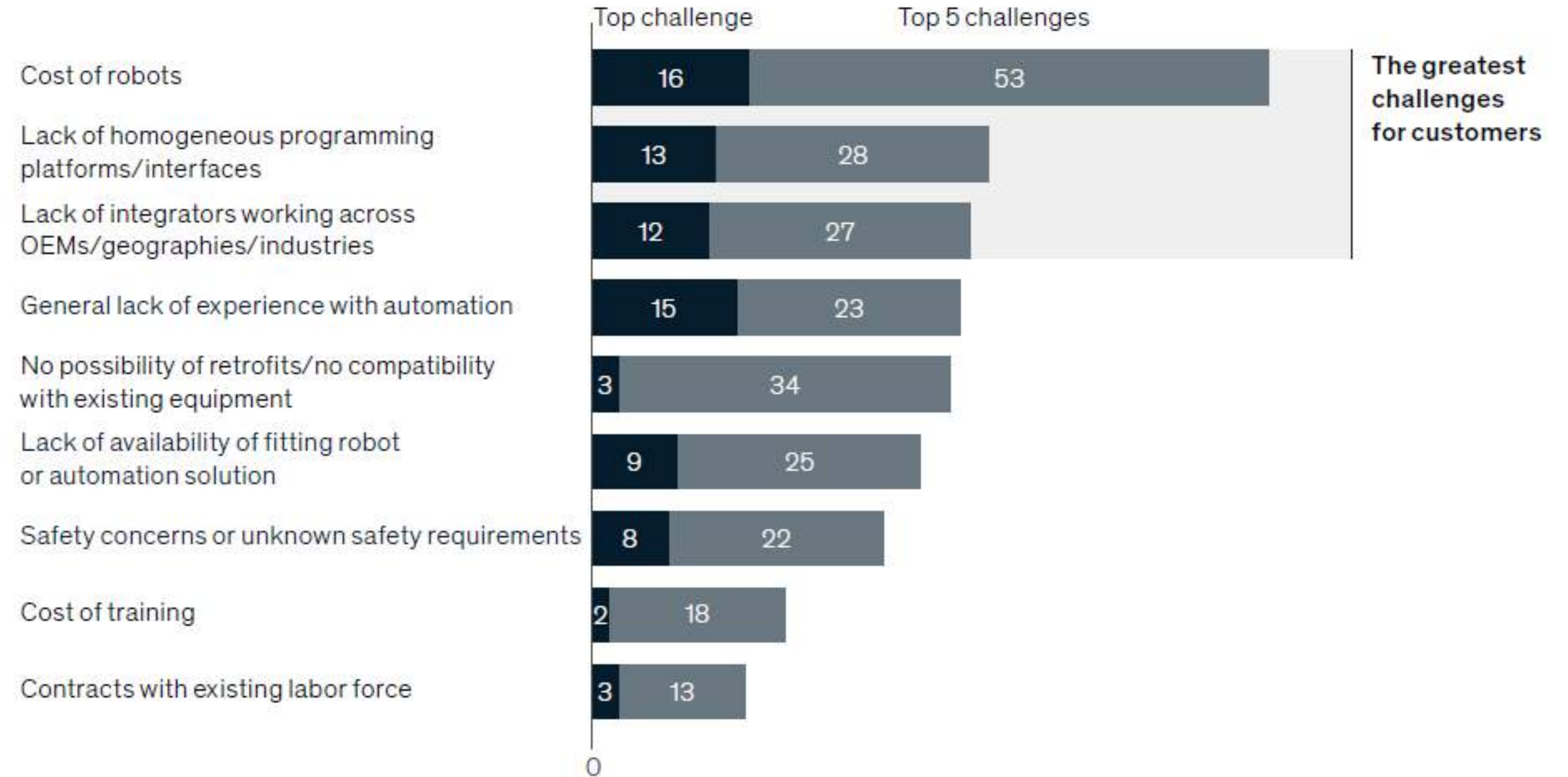
- **Small payload**
 - Force limiting for safe interaction
- **Small footprint**
 - Less disruption to the existing automation line
- **Highly repetitive**
 - Labor replacement for added value
- **Ease of integration**
 - Flexible implementation for the changing demand
- **Cost-Effectiveness**
 - Lower cost in purchase, use, and maintenance



Challenge to Adoption

Customers indicating top challenge and top 5 challenges

Percent

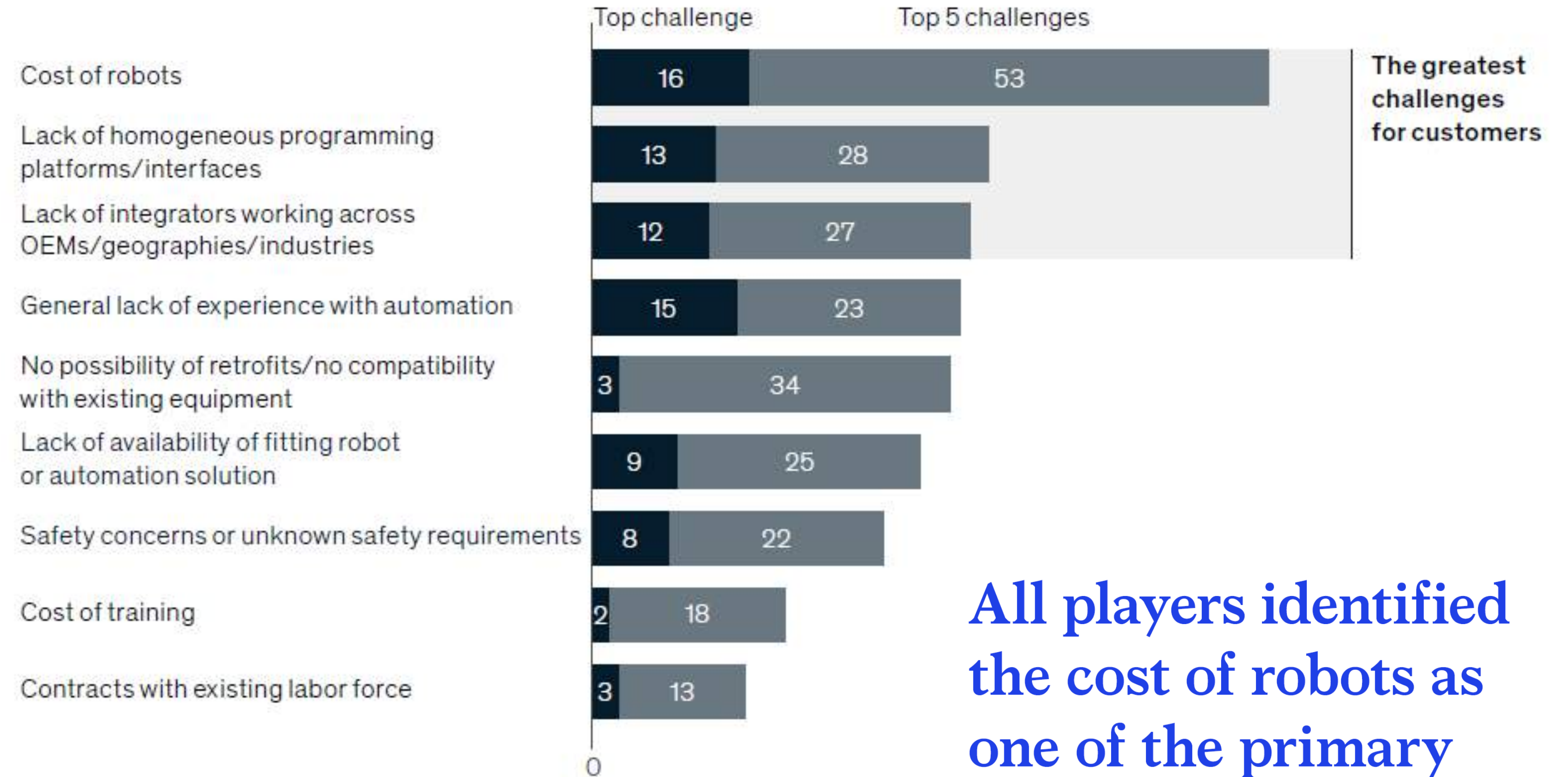


Source: McKinsey Global Robotics Survey 2018

Challenge to Adoption

Customers indicating top challenge and top 5 challenges

Percent



Source: McKinsey Global Robotics Survey 2018

All players identified the cost of robots as one of the primary challenges to adoption

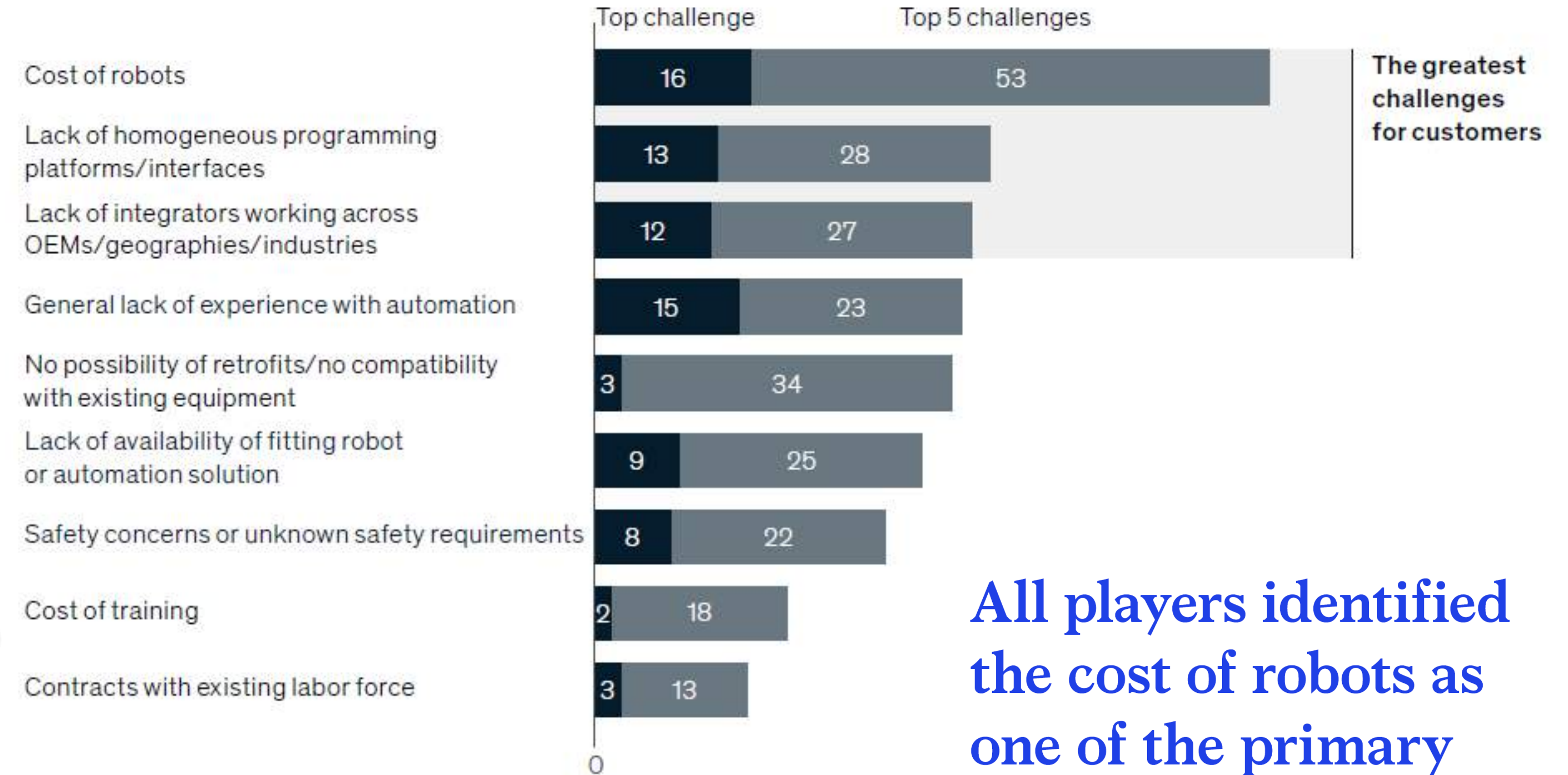
Challenge to Adoption

The cost of which is higher?

- human or
- robot?

Customers indicating top challenge and top 5 challenges

Percent



Source: McKinsey Global Robotics Survey 2018

All players identified the cost of robots as one of the primary challenges to adoption

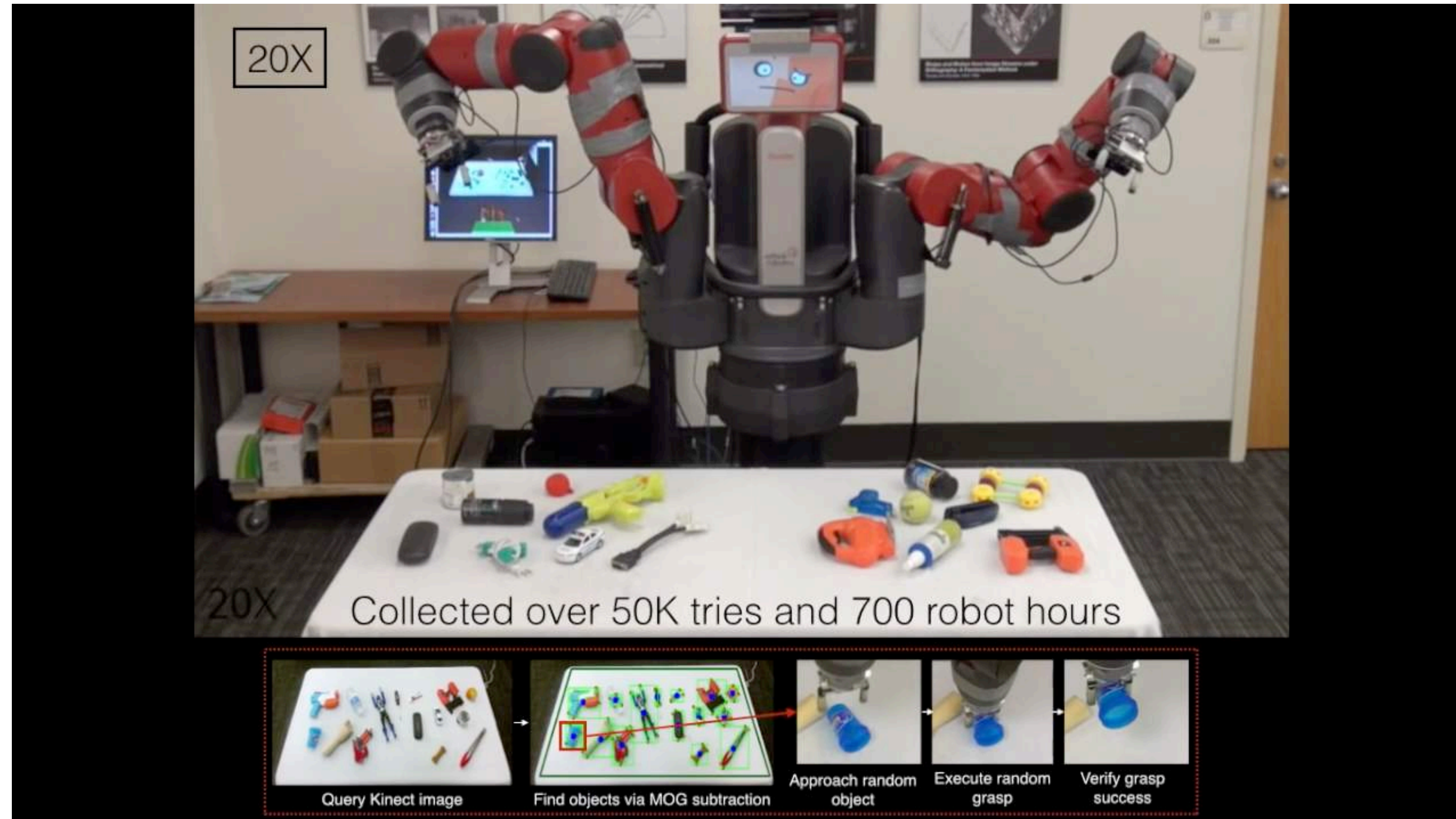
A research field at the intersection of machine learning and robotics that studies techniques allowing a robot to acquire novel skills or adapt to its environment through learning algorithms.

Collaborative Robot Learning



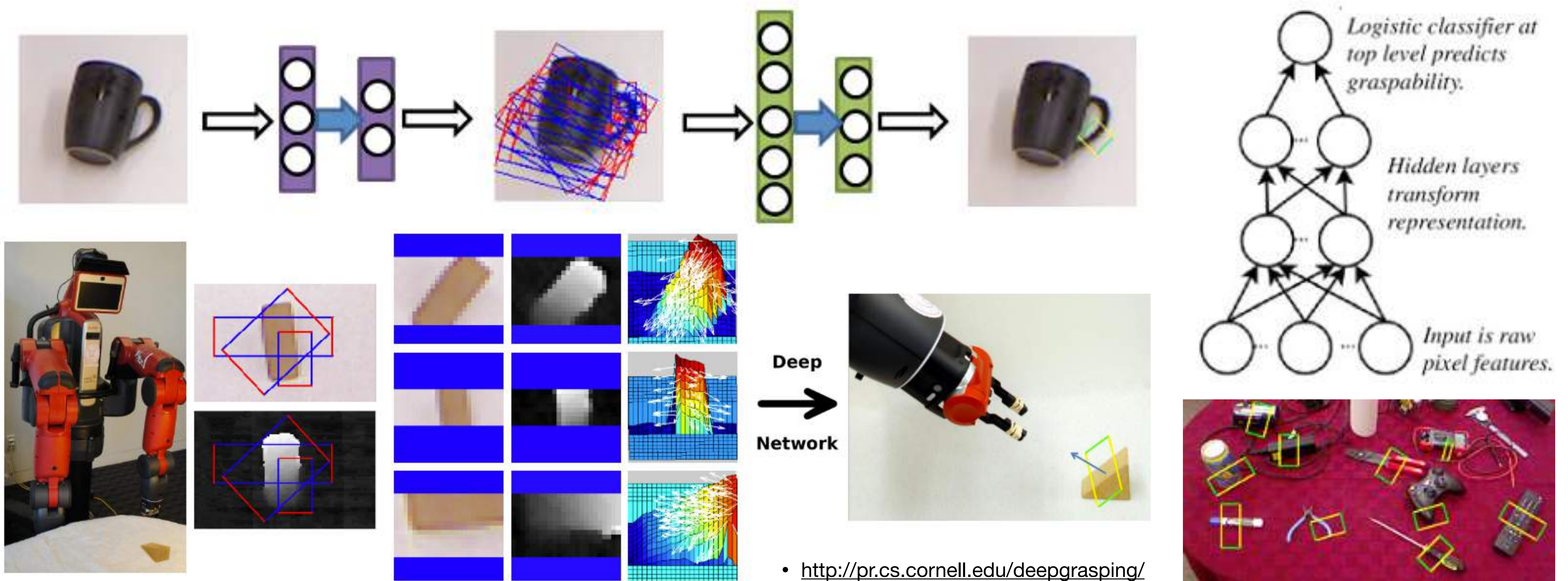
Supersizing self-supervision

By Pinto, Lerrel, and Abhinav Gupta @ CMU



Deep Learning for Detecting Robotic Grasps

By Ian Lenz, Honglak Lee, and Ashutosh Saxena @ Cornell



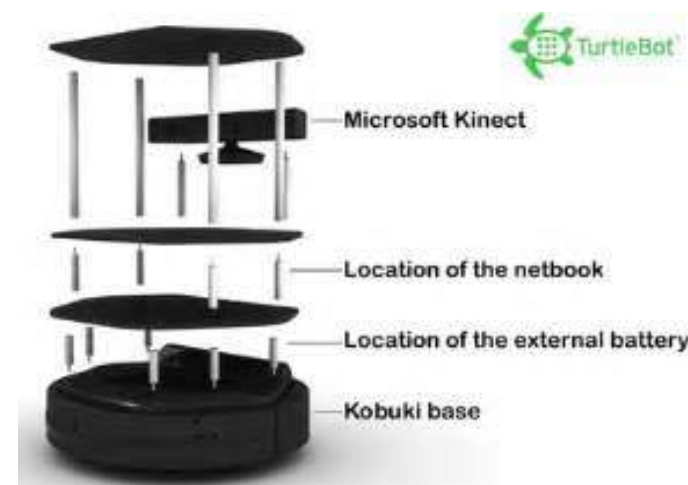
Why Making the Robots to Learn?

Autonomous Vehicles

Research
Challenge
(Science 1st)



Consumer
Electronics
(Cost 1st)



Service
Integration
(App 1st)



Why Making the Robots to Learn?

Autonomous Vehicles

Autonomous Drones

Research
Challenge
(Science 1st)



Consumer
Electronics
(Cost 1st)



Service
Integration
(App 1st)



Why Making the Robots to Learn?

Autonomous Vehicles

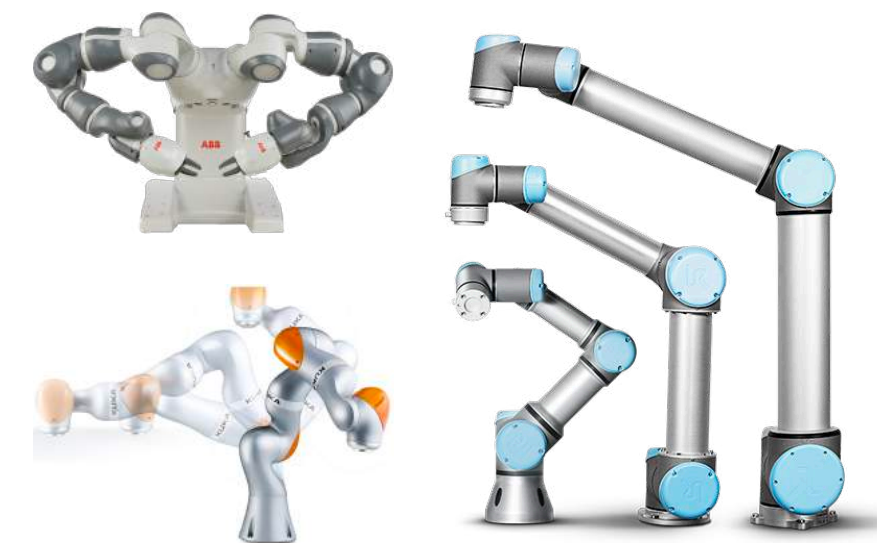
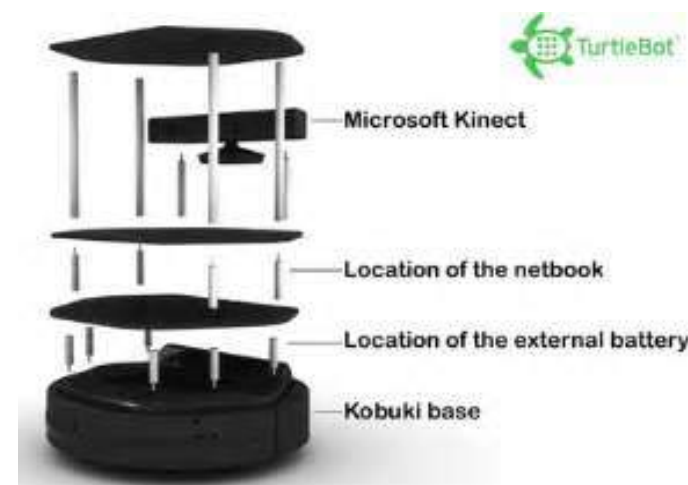
Autonomous Drones

Arm-Type Robots

Research
Challenge
(Science 1st)



Consumer
Electronics
(Cost 1st)



Service
Integration
(App 1st)



Still **Expensive** to buy/integrate,
Difficult to use/learn,
Un-safe to work with, ...

Why Making the Robots to Learn?

Autonomous Vehicles

Autonomous Drones

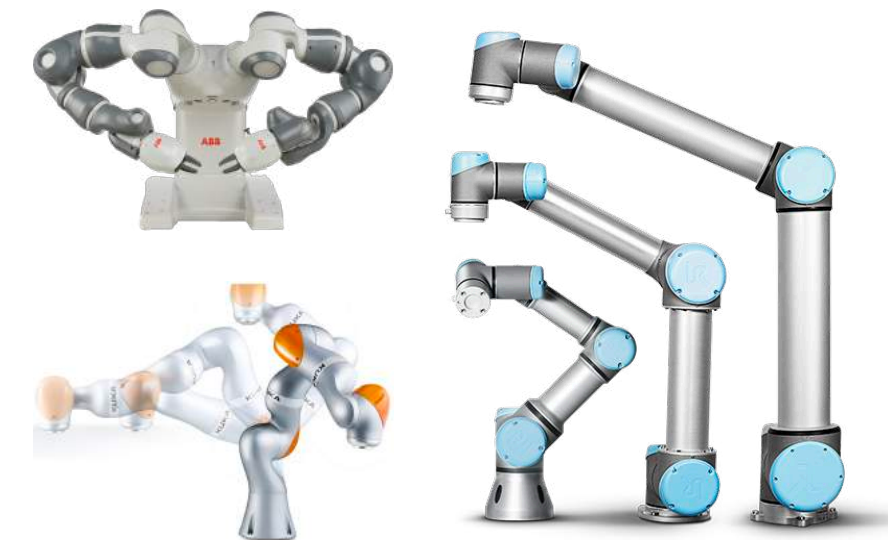
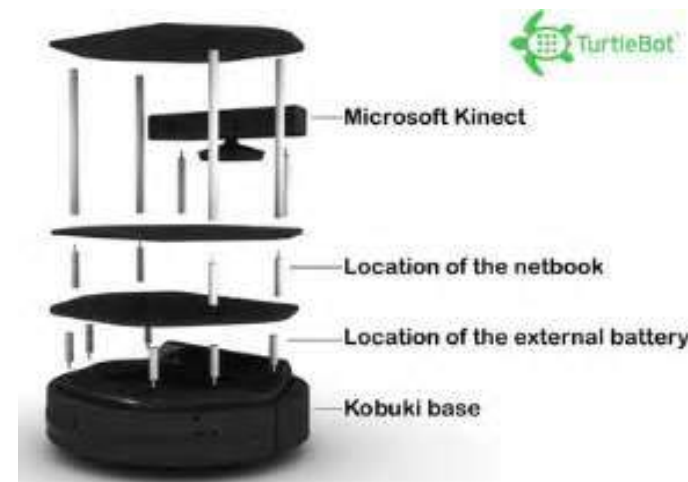
Arm-Type Robots

Research
Challenge
(Science 1st)



Industry
Need

Consumer
Electronics
(Cost 1st)



?

Service
Integration
(App 1st)



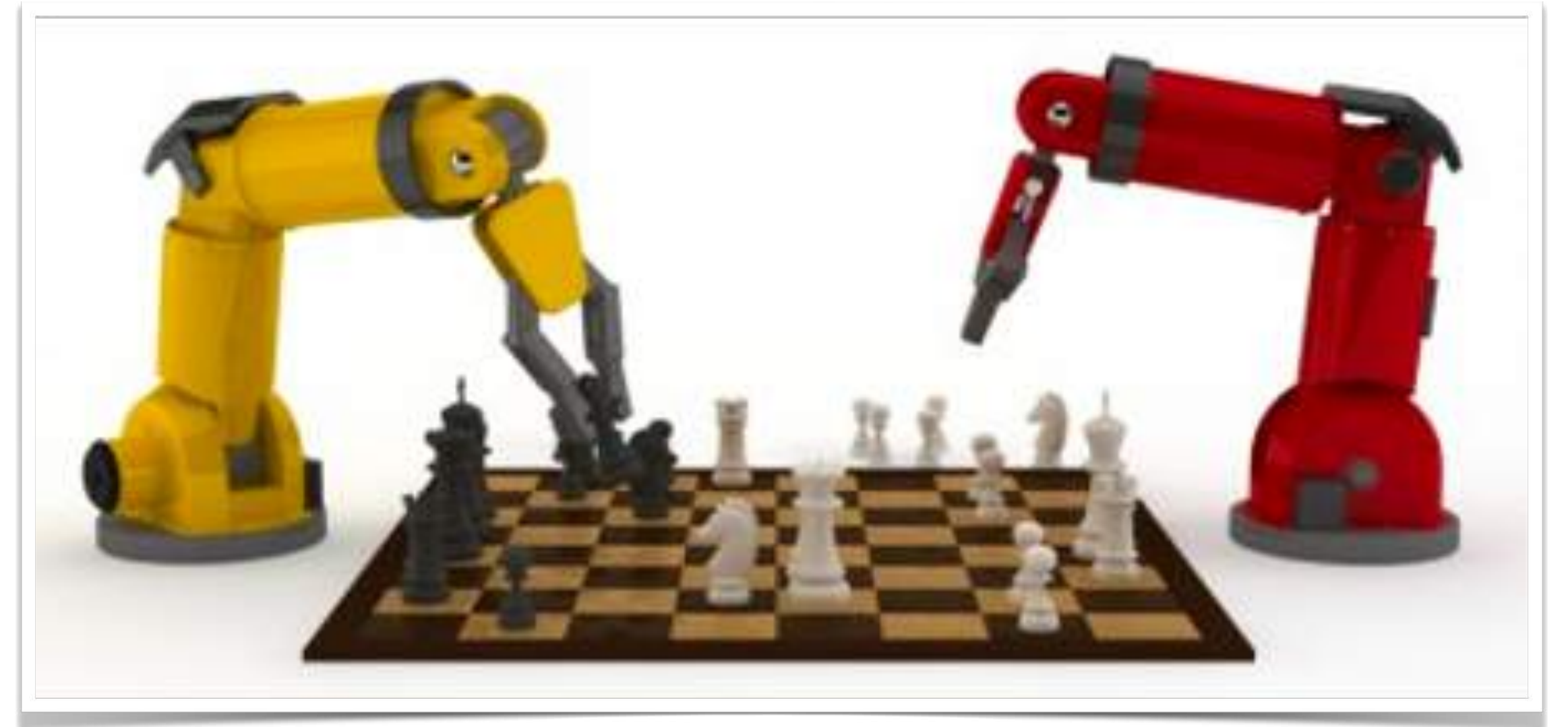
Service
Penetration

Still **Expensive** to buy/integrate,
Difficult to use/learn,
Un-safe to work with, ...

Why Making the Robots to Learn?

Translating Success in Machine Learning for Robotics

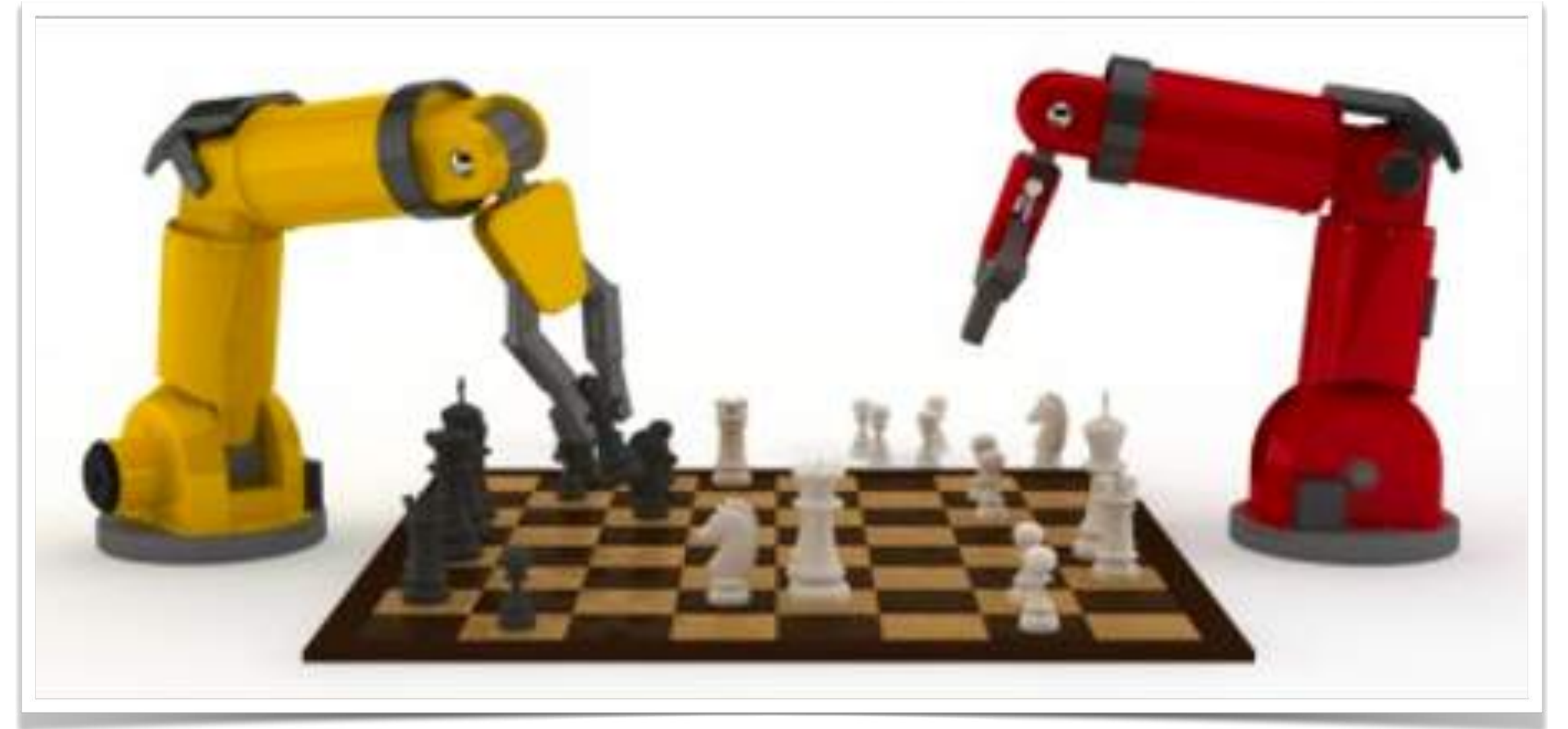
- Computing Unit
- Advanced Algorithms
- Big Data



Why Making the Robots to Learn?

Translating Success in Machine Learning for Robotics

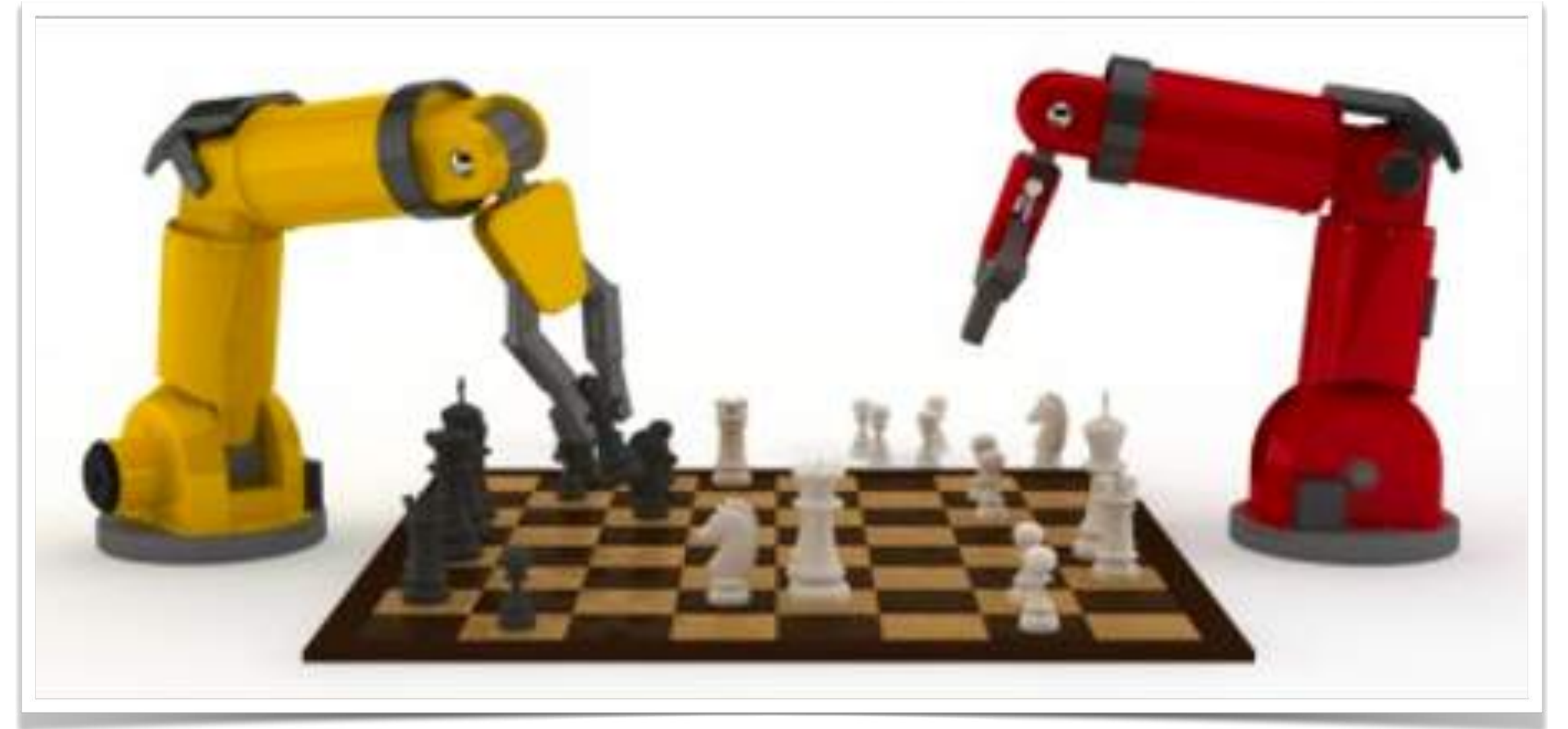
- Computing Unit ✓
- Advanced Algorithms ✓..
- Big Data ✗



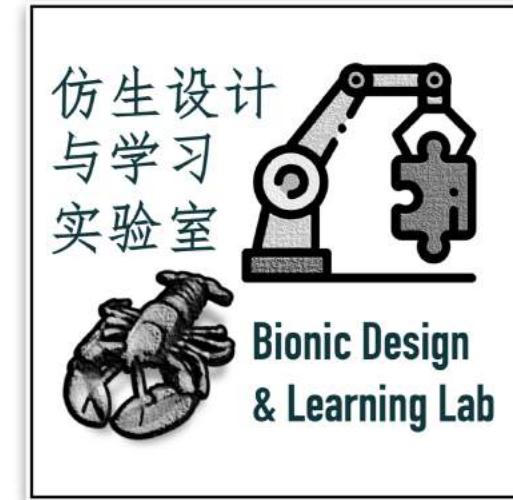
Why Making the Robots to Learn?

Translating Success in Machine Learning for Robotics

- Computing Unit ✓
- Advanced Algorithms ✓..
- Big Data ✗



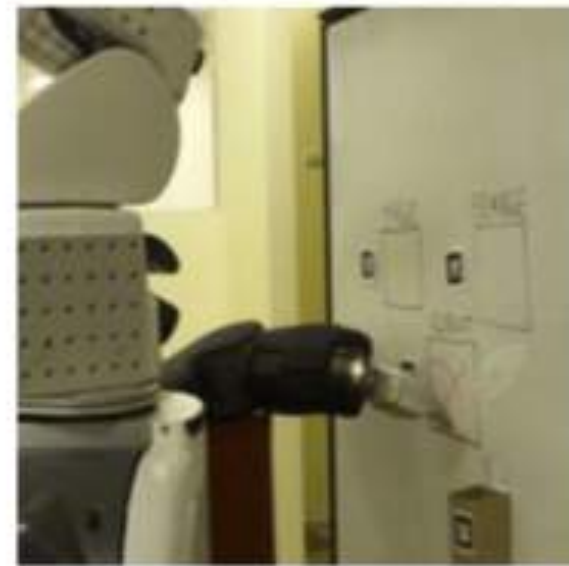
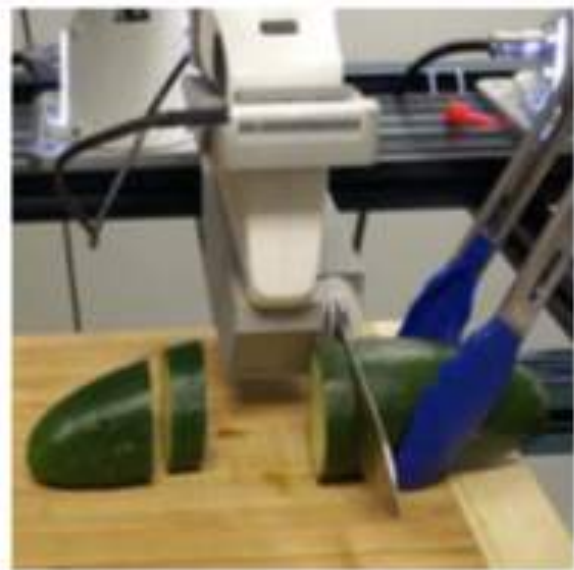
- *How to cost-effectively acquire large, quality, robotic data for learning?*



A Brief Review of Robot Manipulation Learning

Robot Manipulation

How a robot should learn to manipulate the world around it



- Example manipulation skills:
 - inserting,
 - stacking,
 - opening,
 - pushing,
 - cutting,
 - screwing,
 - pouring, and
 - writing ...

Common Concepts in Learning for Manipulation

Internal structure of a manipulation task

- Manipulations as Physical Systems
 - Laws of physics and the structure they impose provide strong prior knowledge
 - Exploit such concepts using learning algorithms and making learning skill tractable

Common Concepts in Learning for Manipulation

Internal structure of a manipulation task

- Underactuation
 - The DOFs of the physical environment can be easily
 - larger than those of the robotic system

Common Concepts in Learning for Manipulation

Internal structure of a manipulation task

- Nonholonomic Constraints
 - a system whose state depends on the path taken in order to achieve it.
 - (Controllable DOFs \neq Total DOFs caused by non-integrable constraints)

Common Concepts in Learning for Manipulation

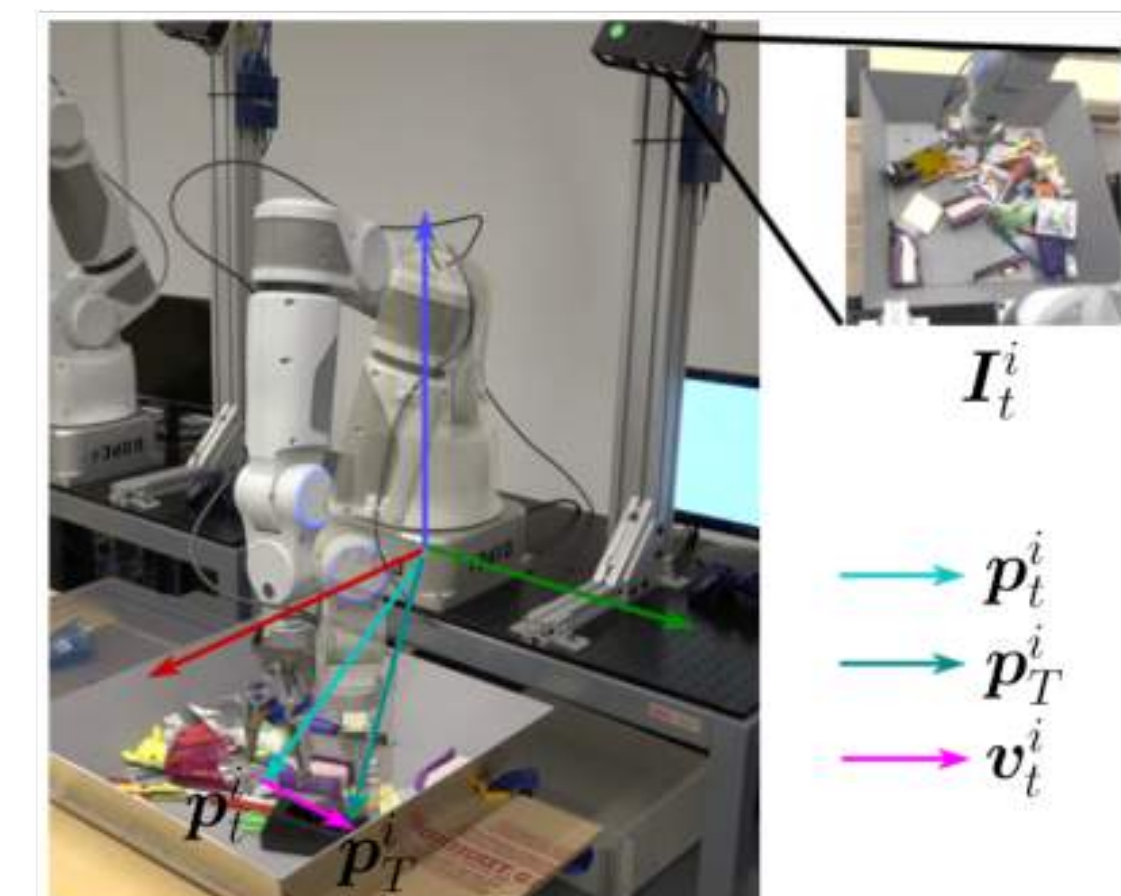
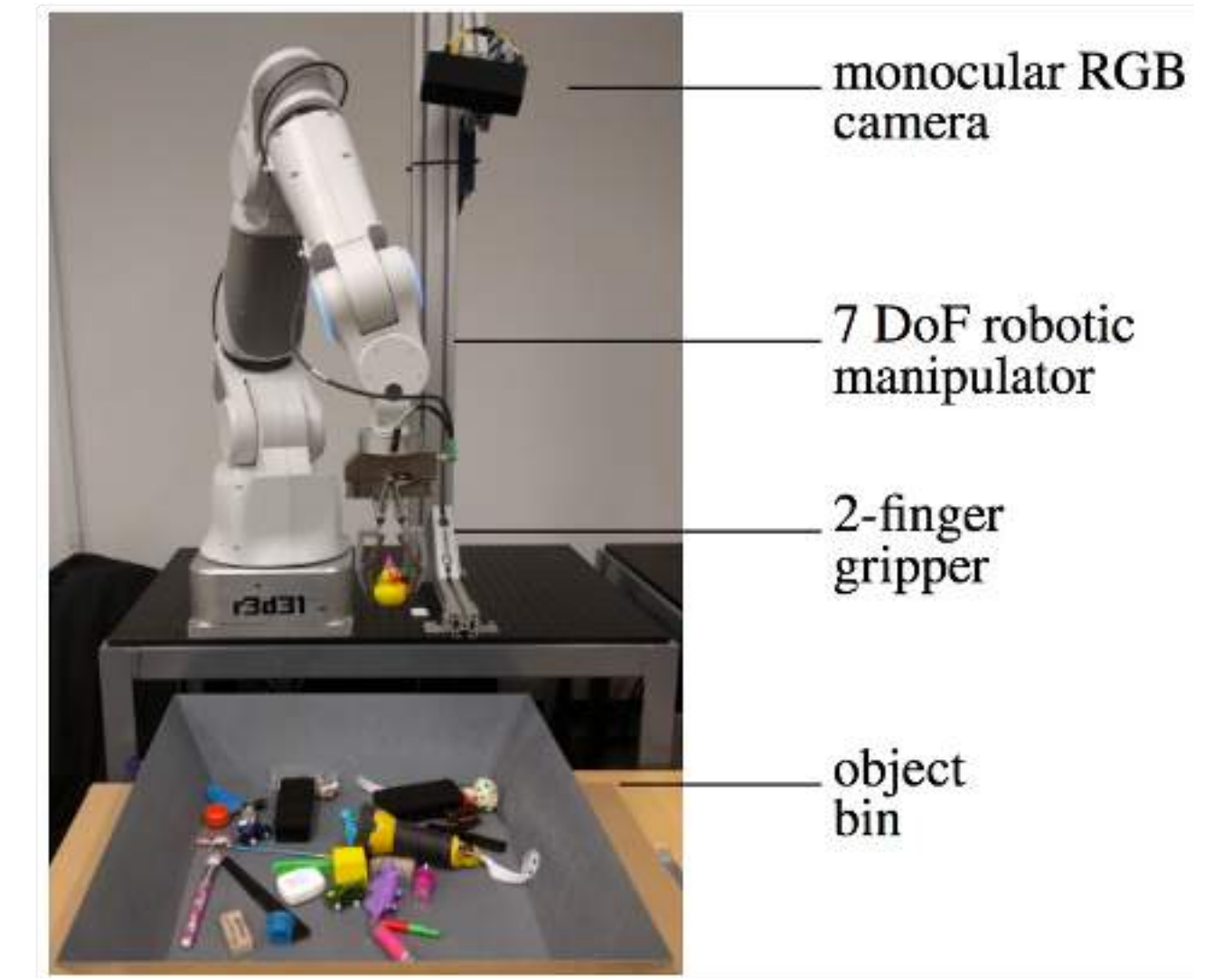
Internal structure of a manipulation task

- Modes in Manipulations
 - Breaking or making of contacts, i.e. collision with obstacles
 - A modular structure for convenient implementation, but will make the manipulation tasks inherently discontinuous
 - The robot must reach a suitable mode before it can perform a desired manipulation

Interactive Perception and Verification

Paving the pathway for supervised learning

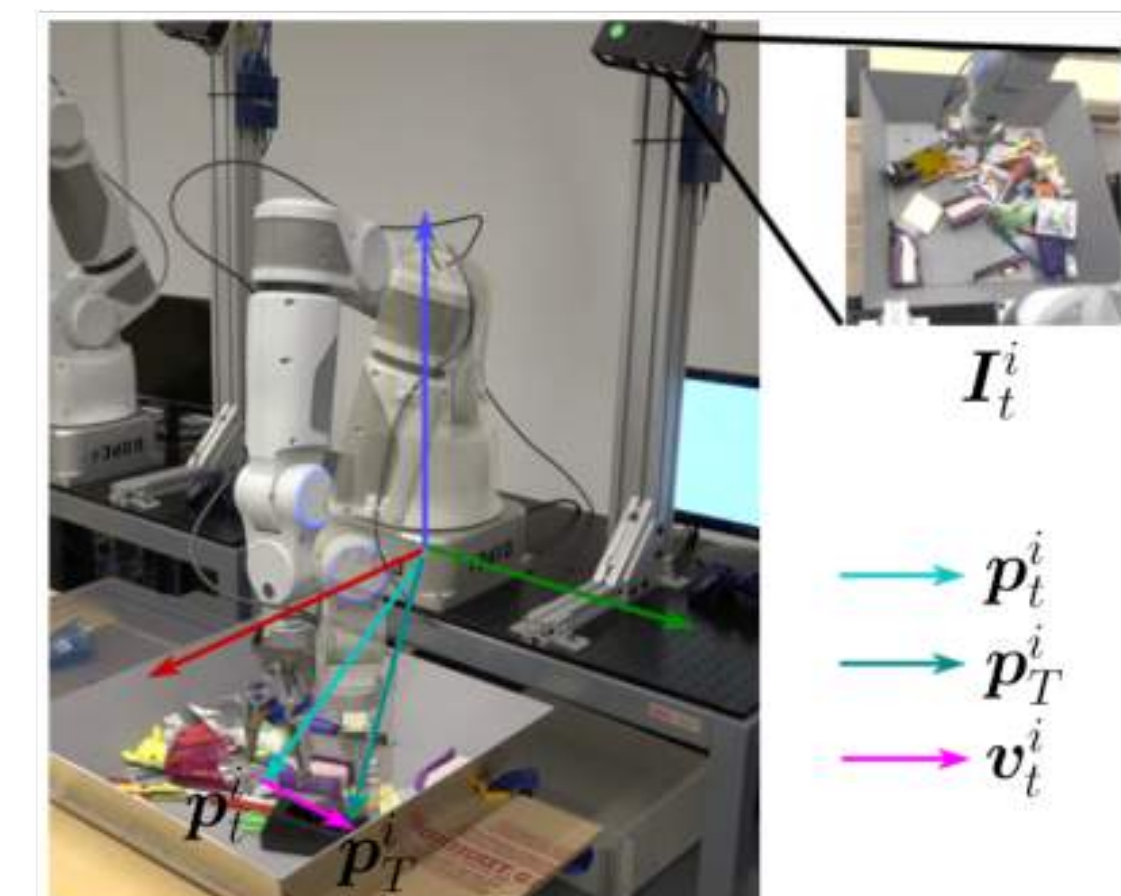
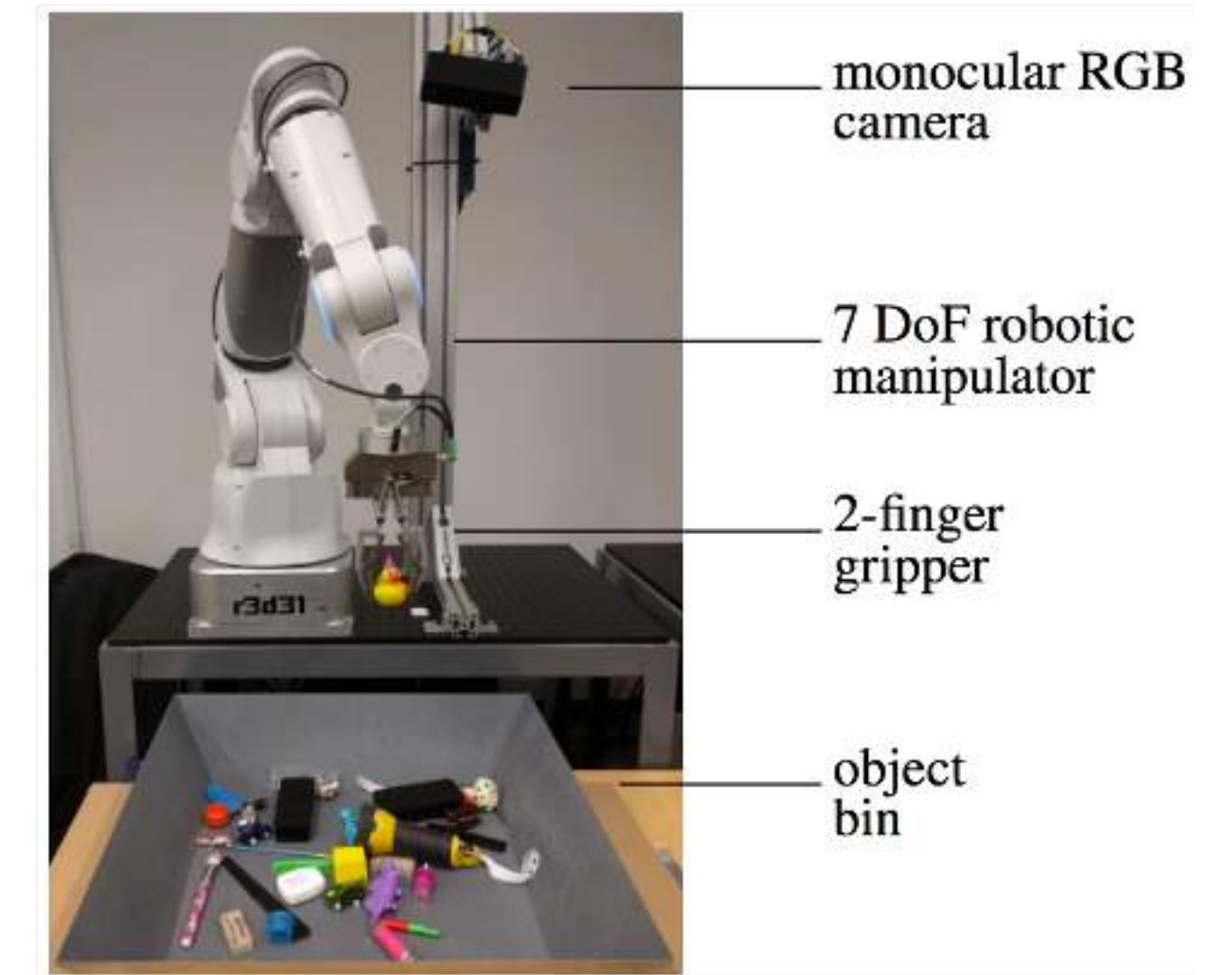
- **Interactive Perception** enables the robot to perceive latent object properties by observing the outcomes of different manipulation actions
- **Verification** is usually done through interactive perception to get the ground truth value for supervised learning with passive perception
- **Active learning** is the process of actively selecting samples to label to maximize learning performance, often used together with interactive perception



Interactive Perception and Verification

Paving the pathway for supervised learning

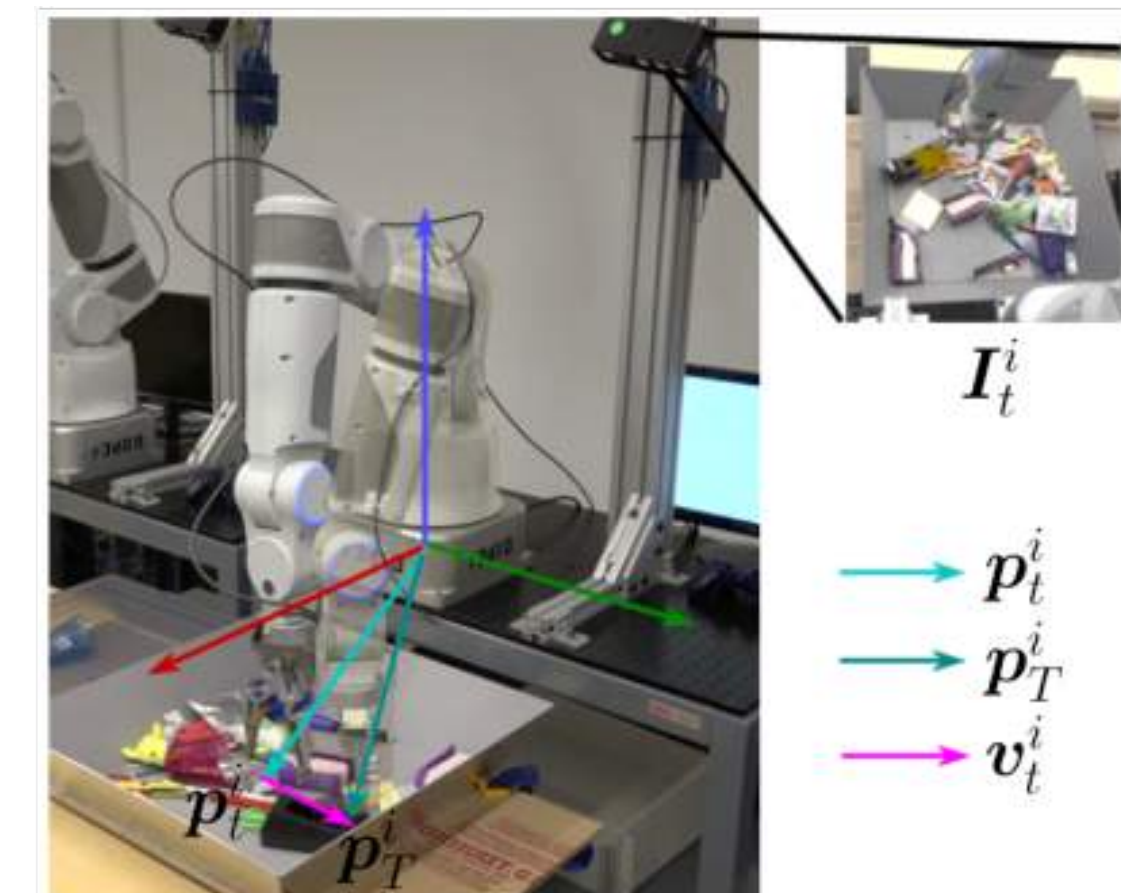
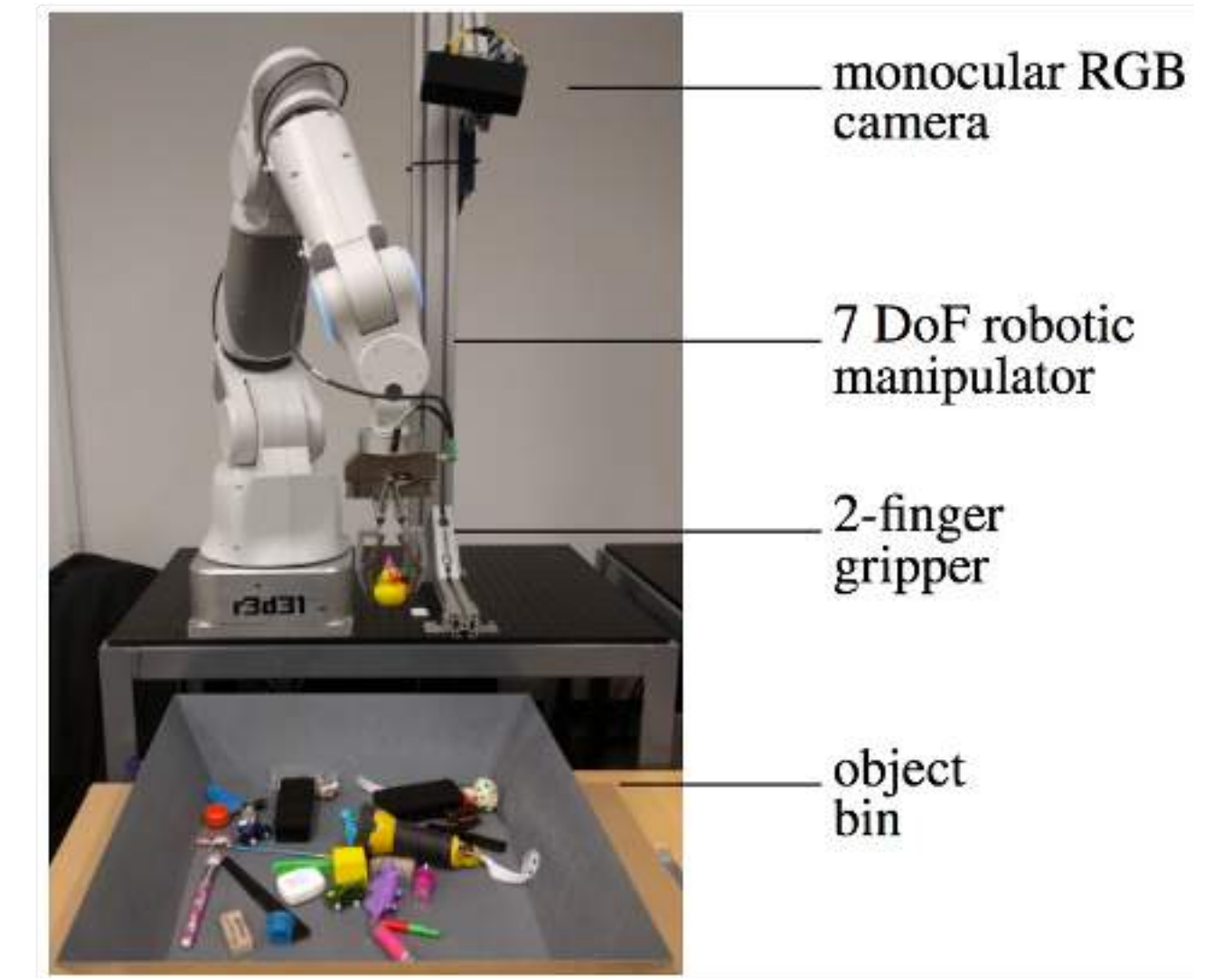
- **Interactive Perception** enables the robot to perceive latent object properties by observing the outcomes of different manipulation actions
- **Verification** is usually done through interactive perception to get the ground truth value for supervised learning with passive perception
- **Active learning** is the process of actively selecting samples to label to maximize learning performance, often used together with interactive perception



Interactive Perception and Verification

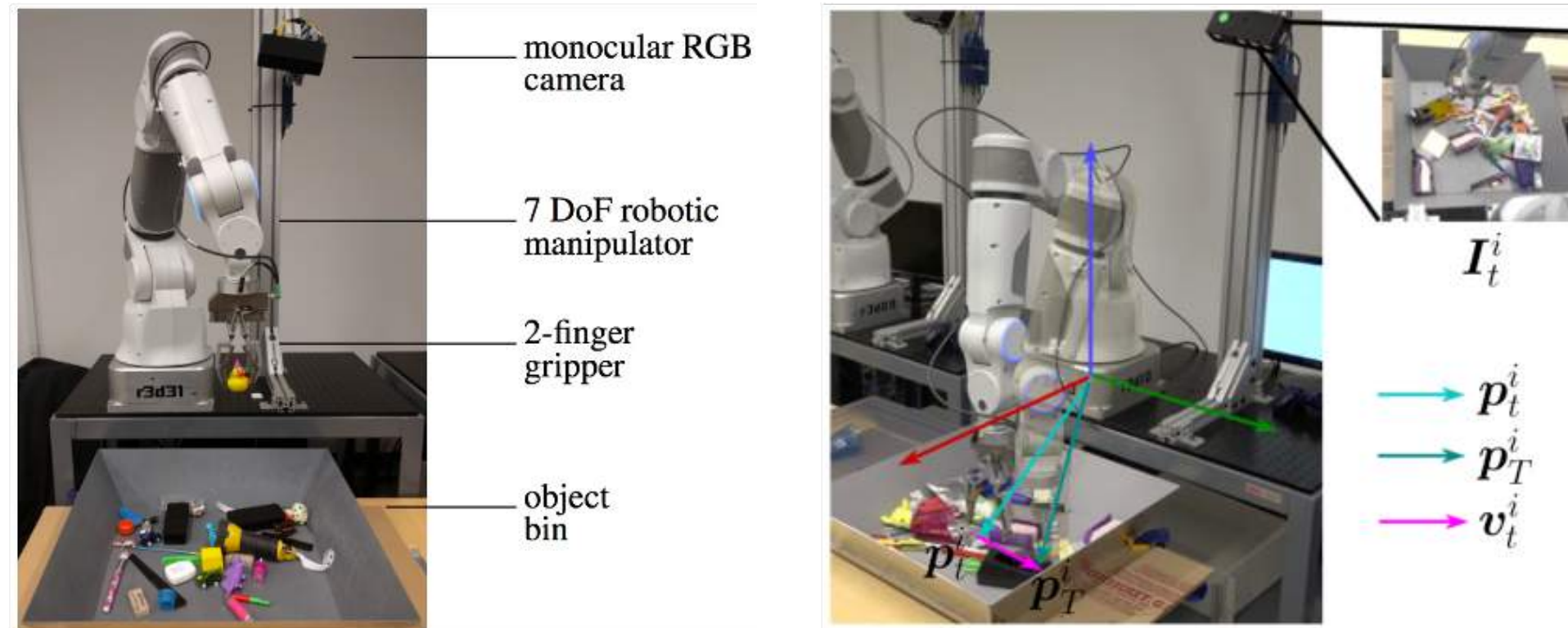
Paving the pathway for supervised learning

- **Interactive Perception** enables the robot to perceive latent object properties by observing the outcomes of different manipulation actions
- **Verification** is usually done through interactive perception to get the ground truth value for supervised learning with passive perception
- **Active learning** is the process of actively selecting samples to label to maximize learning performance, often used together with interactive perception



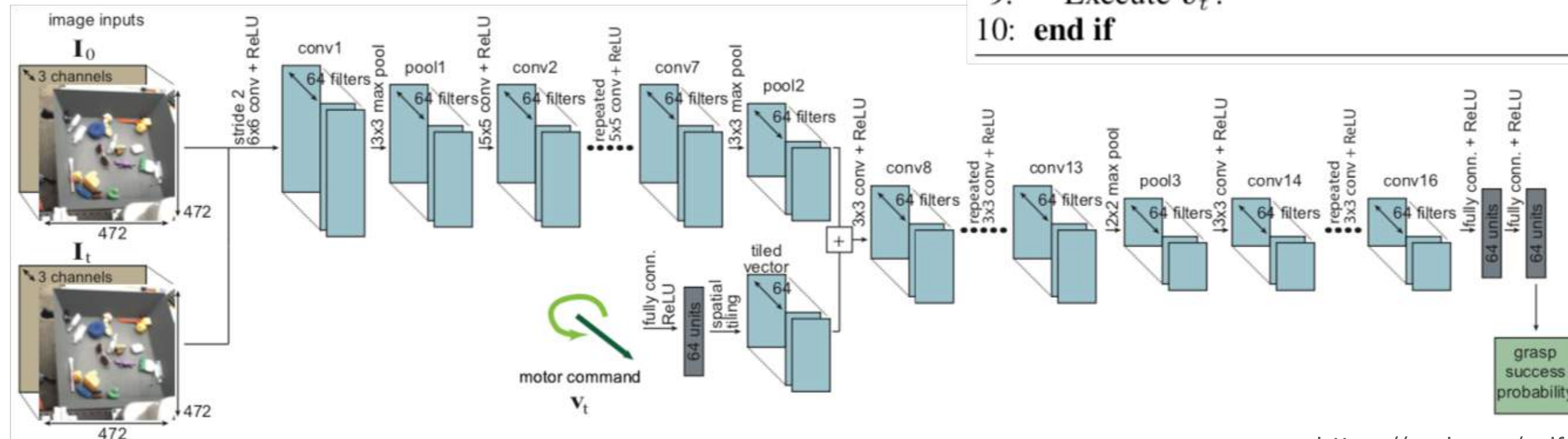
Interactive Perception and Verification

Paving the pathway for supervised learning



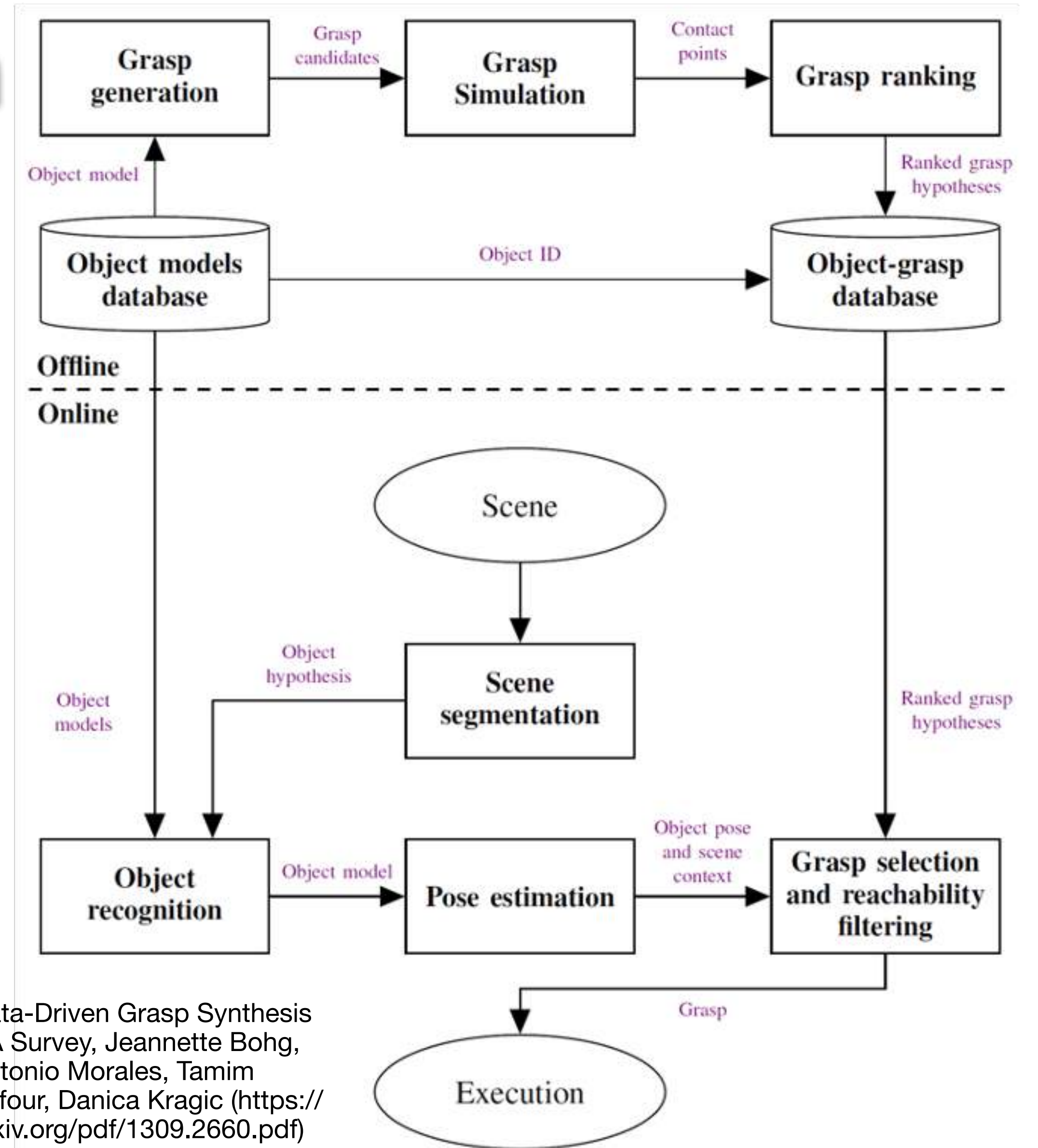
Algorithm 1 Servoing mechanism $f(I_t)$

- 1: Given current image I_t and network g .
- 2: Infer v_t^* using g and CEM.
- 3: Evaluate $p = g(I_t, \emptyset) / g(I_t, v_t^*)$.
- 4: **if** $p = 0.9$ **then**
- 5: Output \emptyset , close gripper.
- 6: **else if** $p \leq 0.5$ **then**
- 7: Modify v_t^* to raise gripper height and execute v_t^* .
- 8: **else**
- 9: Execute v_t^* .
- 10: **end if**



<https://arxiv.org/pdf/1603.02199v1.pdf>

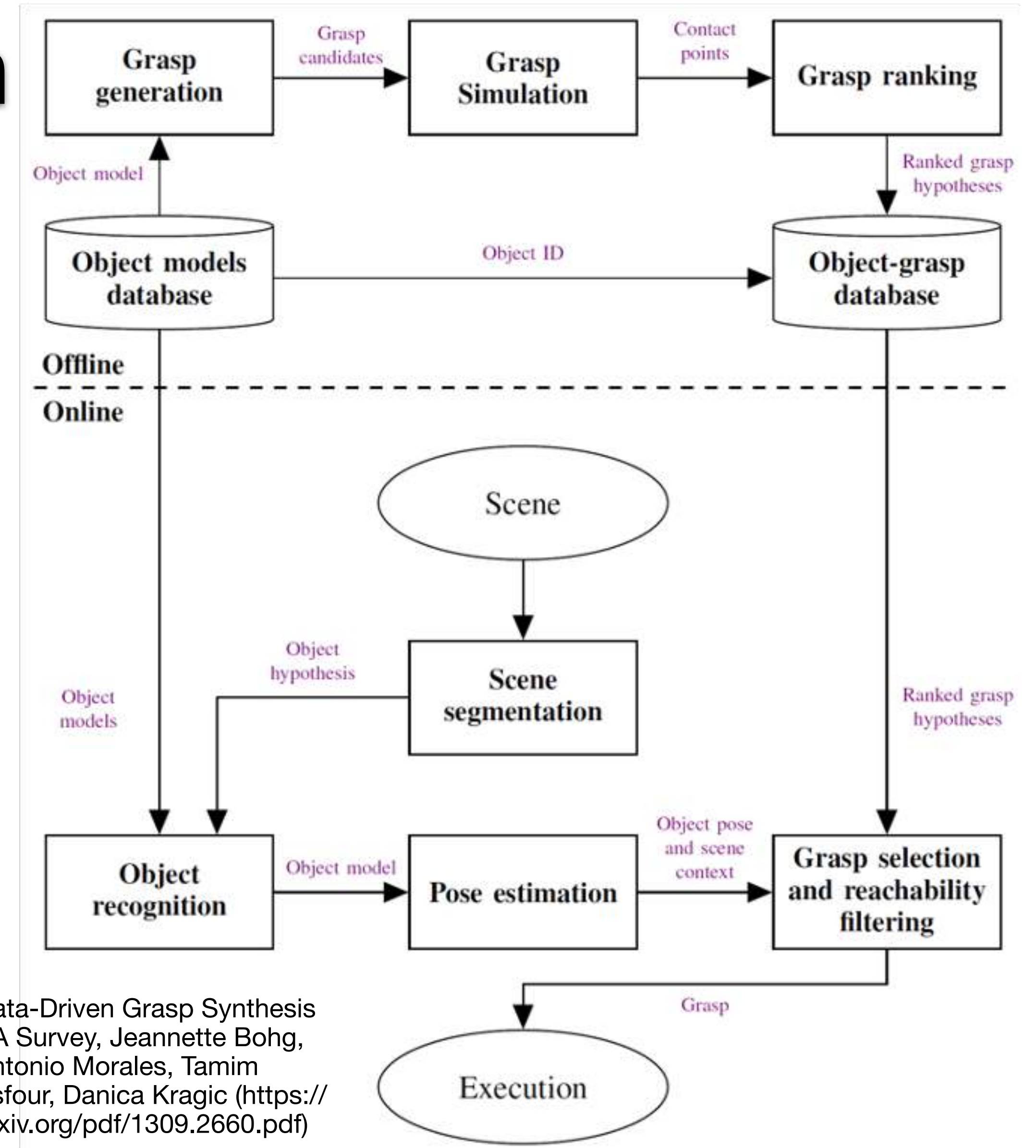
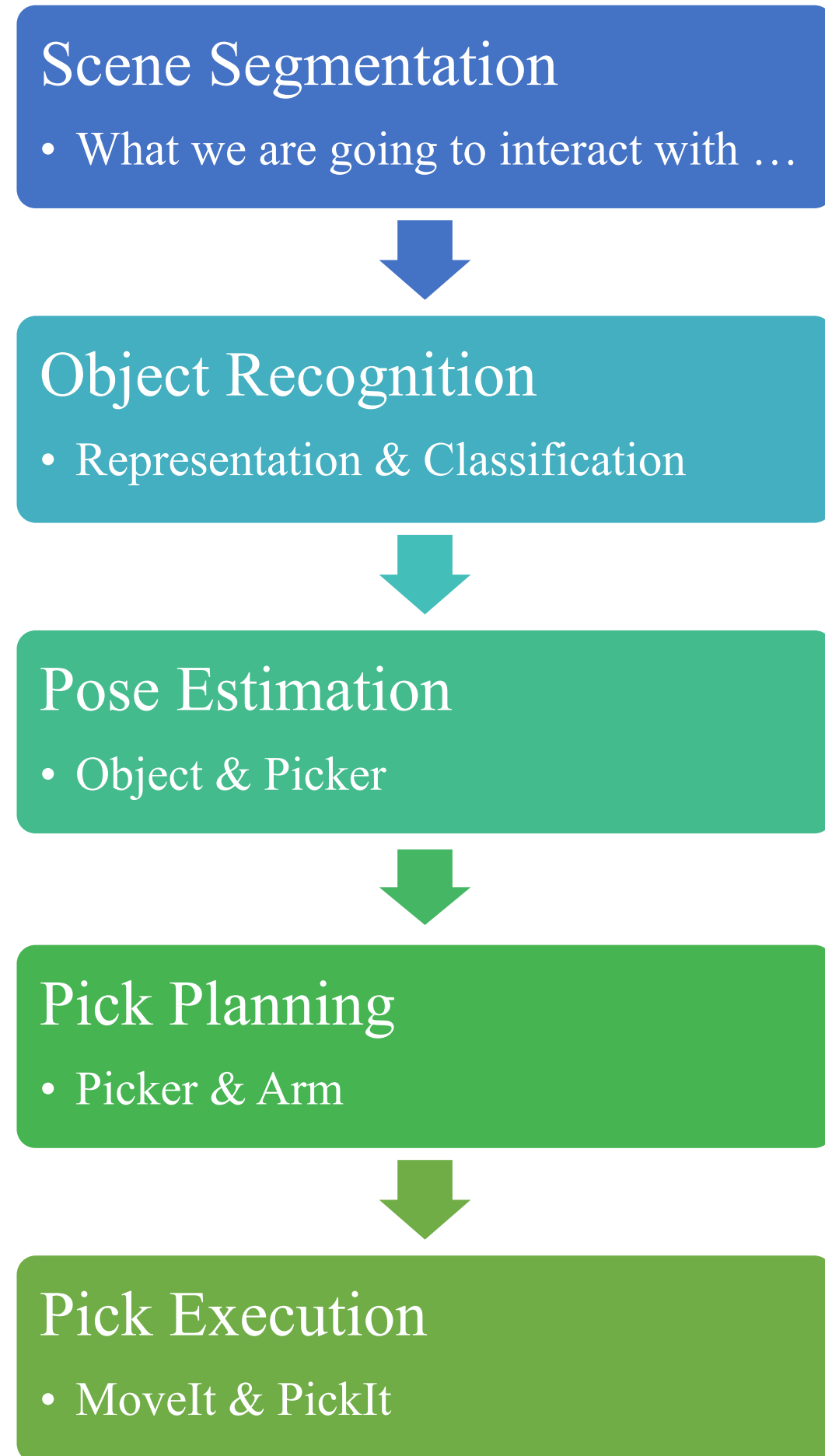
Structured Decomposition of Vision-based Picking



- Data-Driven Grasp Synthesis - A Survey, Jeannette Bohg, Antonio Morales, Tamim Asfour, Danica Kragic (<https://arxiv.org/pdf/1309.2660.pdf>)

Structured Decomposition of Vision-based Picking

Hierarchical Task Decompositions and Skill Reusability



- Data-Driven Grasp Synthesis - A Survey, Jeannette Bohg, Antonio Morales, Tamim Asfour, Danica Kragic (<https://arxiv.org/pdf/1309.2660.pdf>)

System Integration of Object-centered Robotic Manipulation

Object-Centric Generalization



- **Generalization via objects**
 - across different objects, between similar (or identical) objects in different task instances
- Usually enough to generalize across task instances
 - Generalizing across different objects will require both motor skills and object models that adapt to variations in object shape, properties, and appearance.

System Integration of Object-centered Robotic Manipulation

Object-Centric Generalization

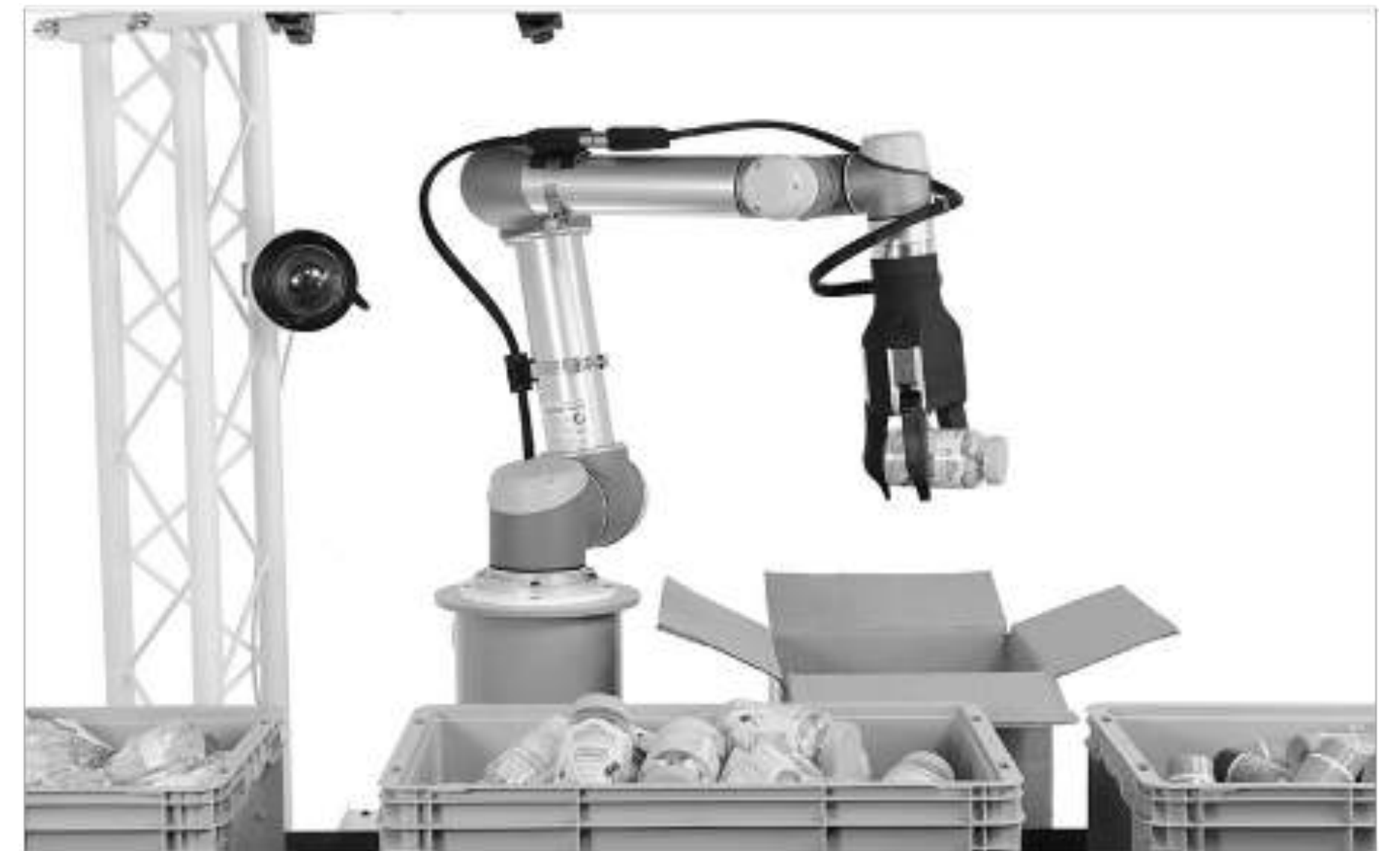


- Generalization via objects
 - across different objects, between similar (or identical) objects in different task instances
- **Usually enough to generalize across task instances**
 - Generalizing across different objects will require both motor skills and object models that adapt to variations in object shape, properties, and appearance.

System Integration of Object-centered Robotic Manipulation

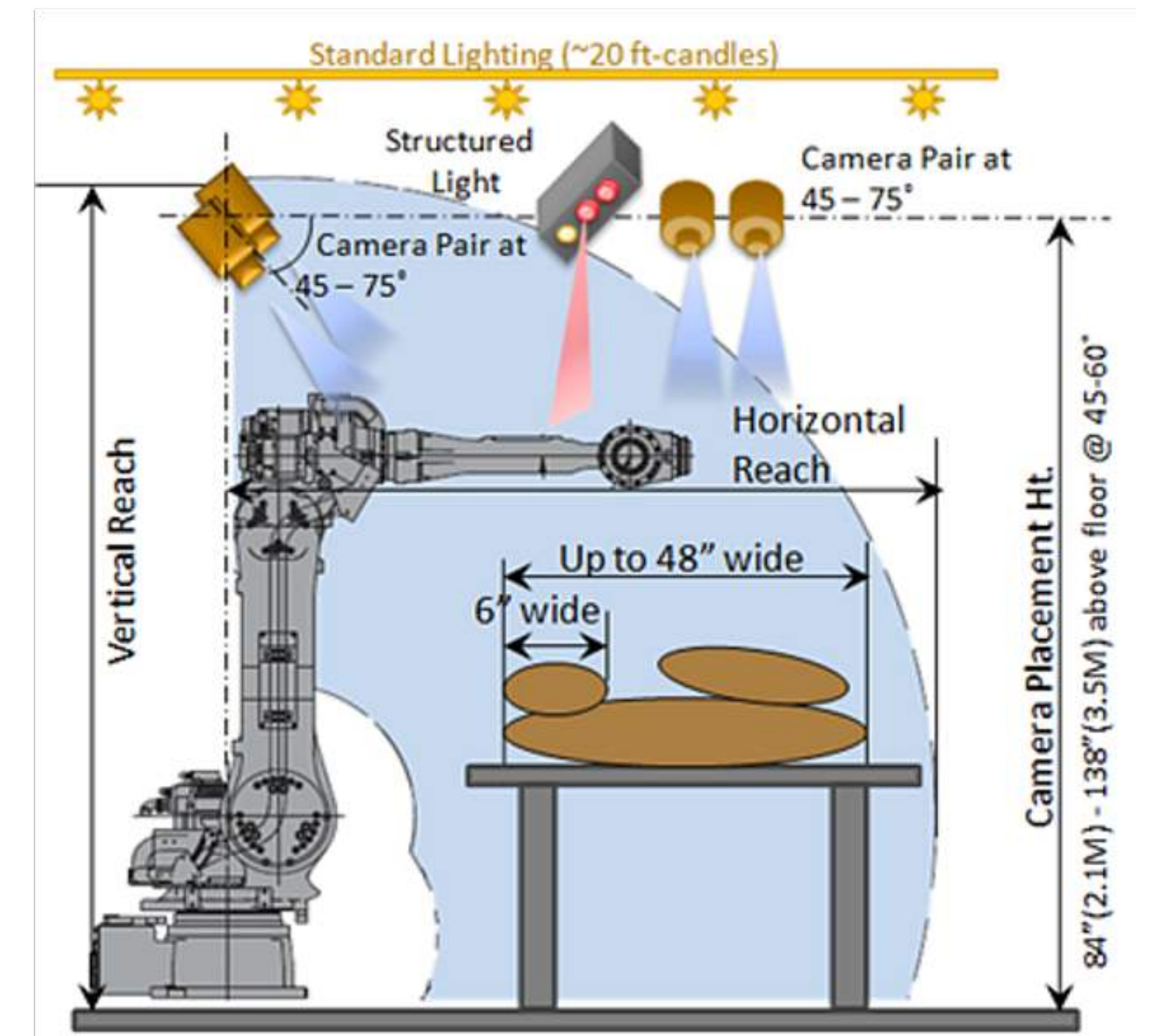
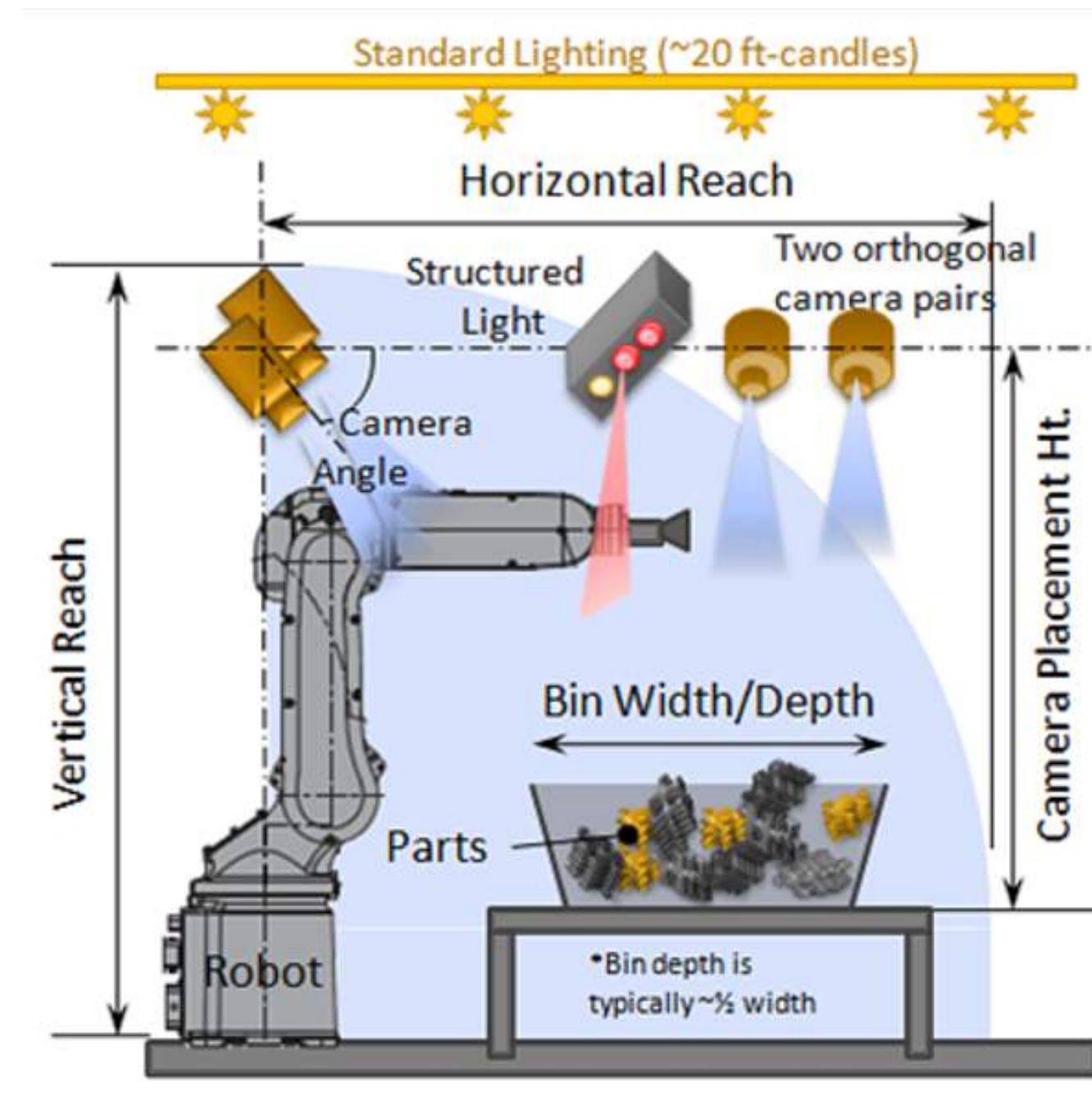
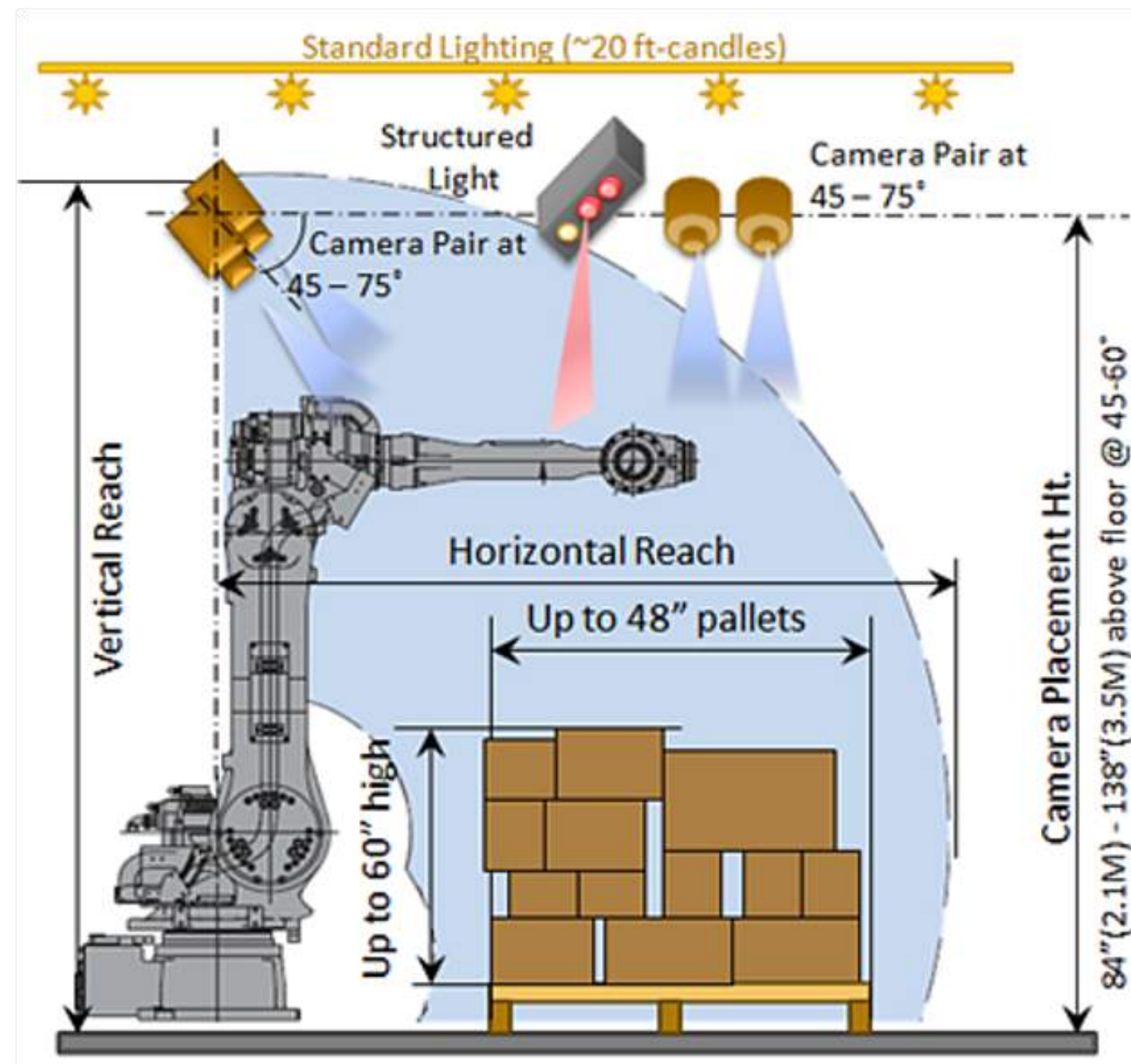
Object-Centric Generalization

- **Abstract representations**
 - Find a representation under which we can consider a family of objects to be equivalent or identical, even though they vary substantially at the pixel or feature level.



System Integration of Object-centered Robotic Manipulation

Object-Centric Generalization



Discovering Novel Concepts and Structures

A structured hierarchy for learning structural skills

Can We Generalize Learning through Physical Interaction?

The diagram illustrates the interface of physical interaction between a robot and objects, categorized into Rigid and Soft, and Finger and Object.

- Rigid:** OnRobot RG6 robot arm.
- Soft:** Omni Adaptive Soft Fingers.
- Finger:** Selected YCB Objects (e.g., Windex, Maxwell House, banana, cup).
- Object:** Stuffed Toys (e.g., teddy bear, banana, horse).

The interaction is shown in the **DeepClaw Benchmarking Station**, where the robot arm is interacting with a tray of objects.

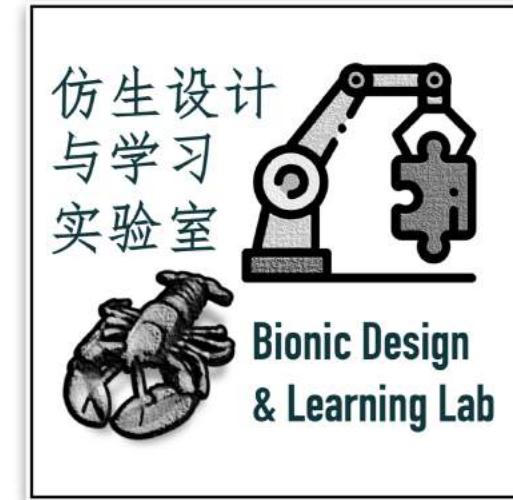
AncoraSIR.com

Rigid-Soft Interactive Learning for Robust Grasping

Bionic Design & Learning Lab @ SUSTech

SUSTech

2



Course Overview

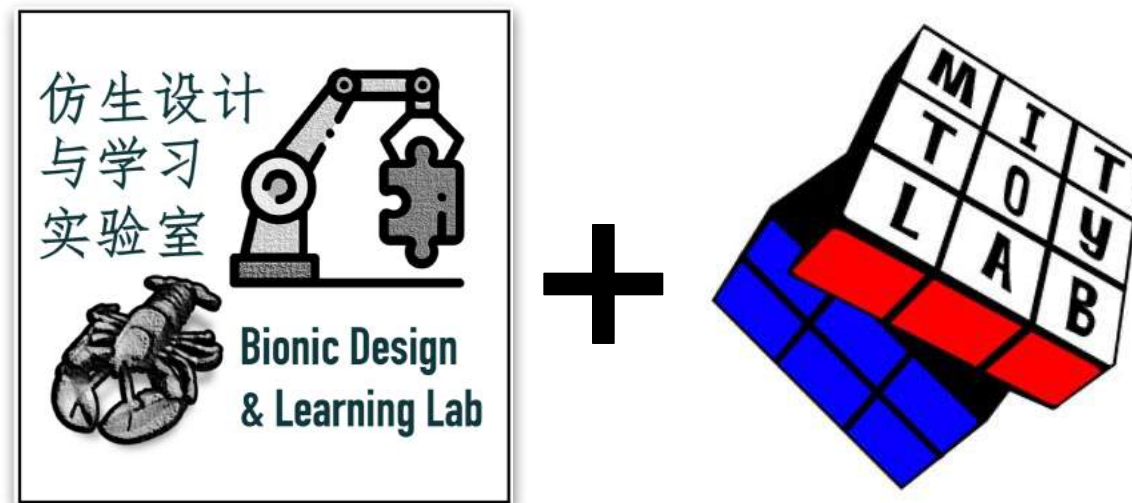
Welcome to [ME336]

Machine Intelligence Design & Learning Lab

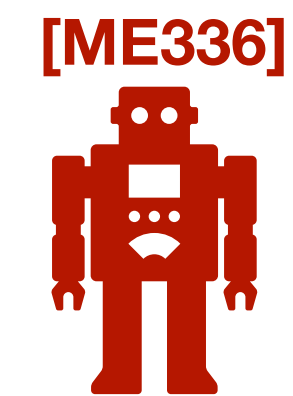
Course
Instructor



Course
Advisor



[ME303]
Machine
Design
(in Autumn)



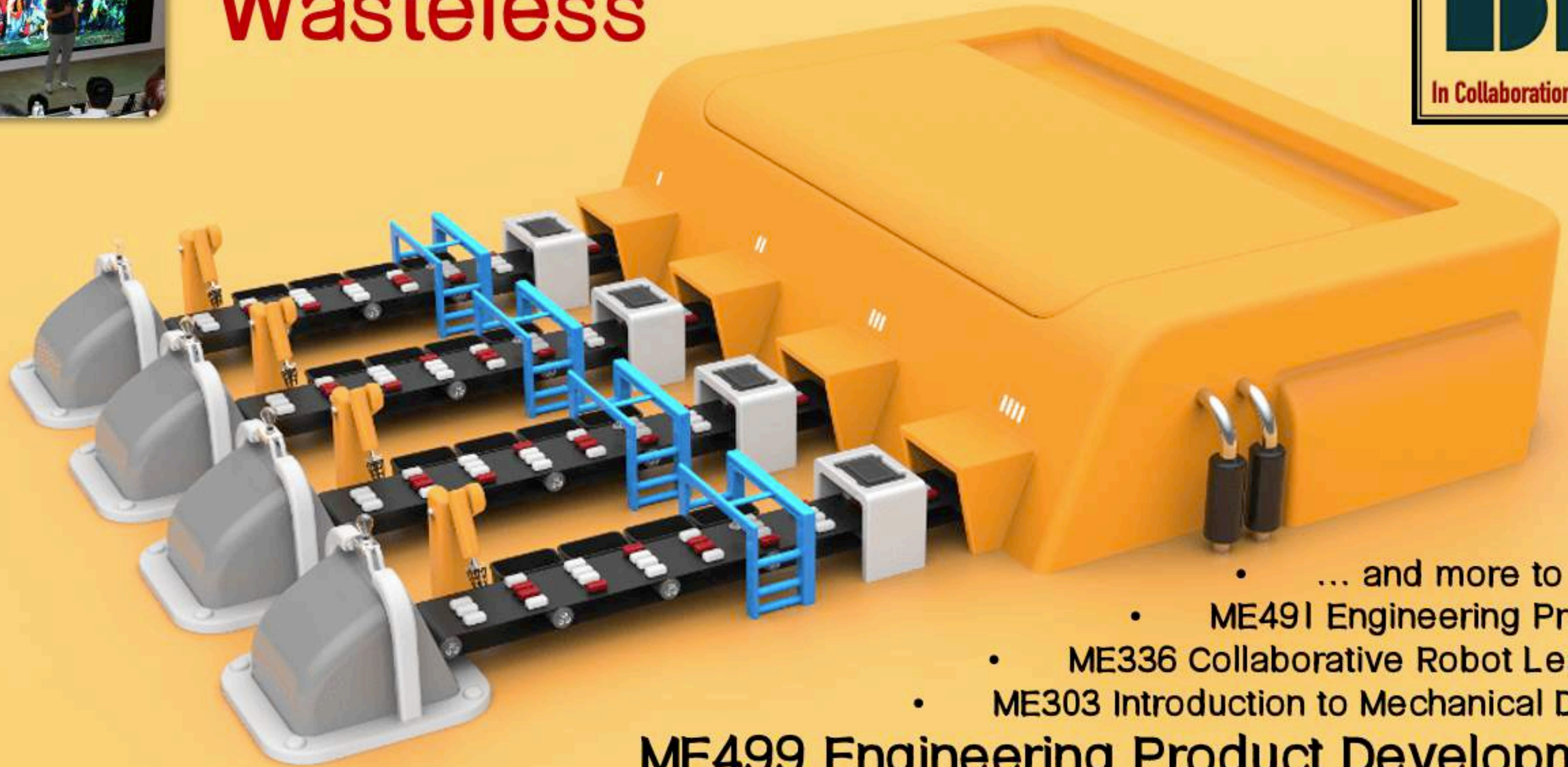
[ME336]
Robot
Learning
(in Spring)



[ME491]
Industrial
Practice
(in Summer)



Design Theme of the Year
Wasteless



- ... and more to come
- ME491 Engineering Practice
- ME336 Collaborative Robot Learning
- ME303 Introduction to Mechanical Design

ME499 Engineering Product Development



Song Chaoyang



Liu Xiaobo



Wang Zhenhong



Guo Ning



Yang Linhan



Wan Fang



Sun Haoran



Yu Chengming



Guo Yuqin



Feng Shihao



Chen Mindong



Ge Sheng



He Jin



He Haibin



Fu Tian



Wang Teng



Zhao Dan

Location:

- Room 235, New Engineering Building

Time:

- Lecture on Wednesday: 1400~1550

- Lab on Friday: 0800-0950

Course Instructors & Teaching Support

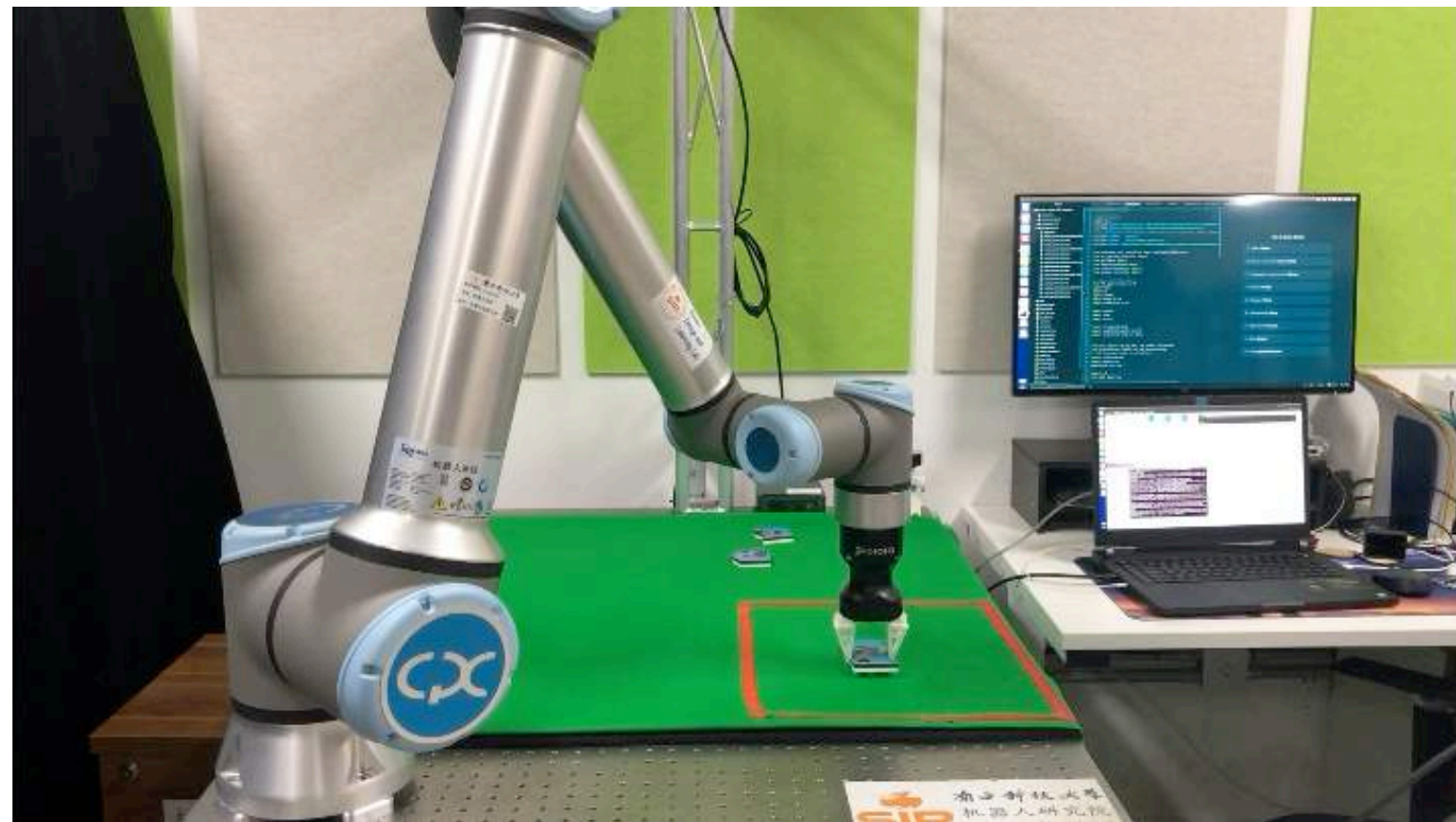
3 Mini Courses + 2 Lab Assignments + 1 Themed Project

Room 235, New Engineering Building

Subject to Change

• Human-Centered Robot Games (5 weeks)

- Introduction to the human-centered design of learning systems for competitive gaming with a robot player.

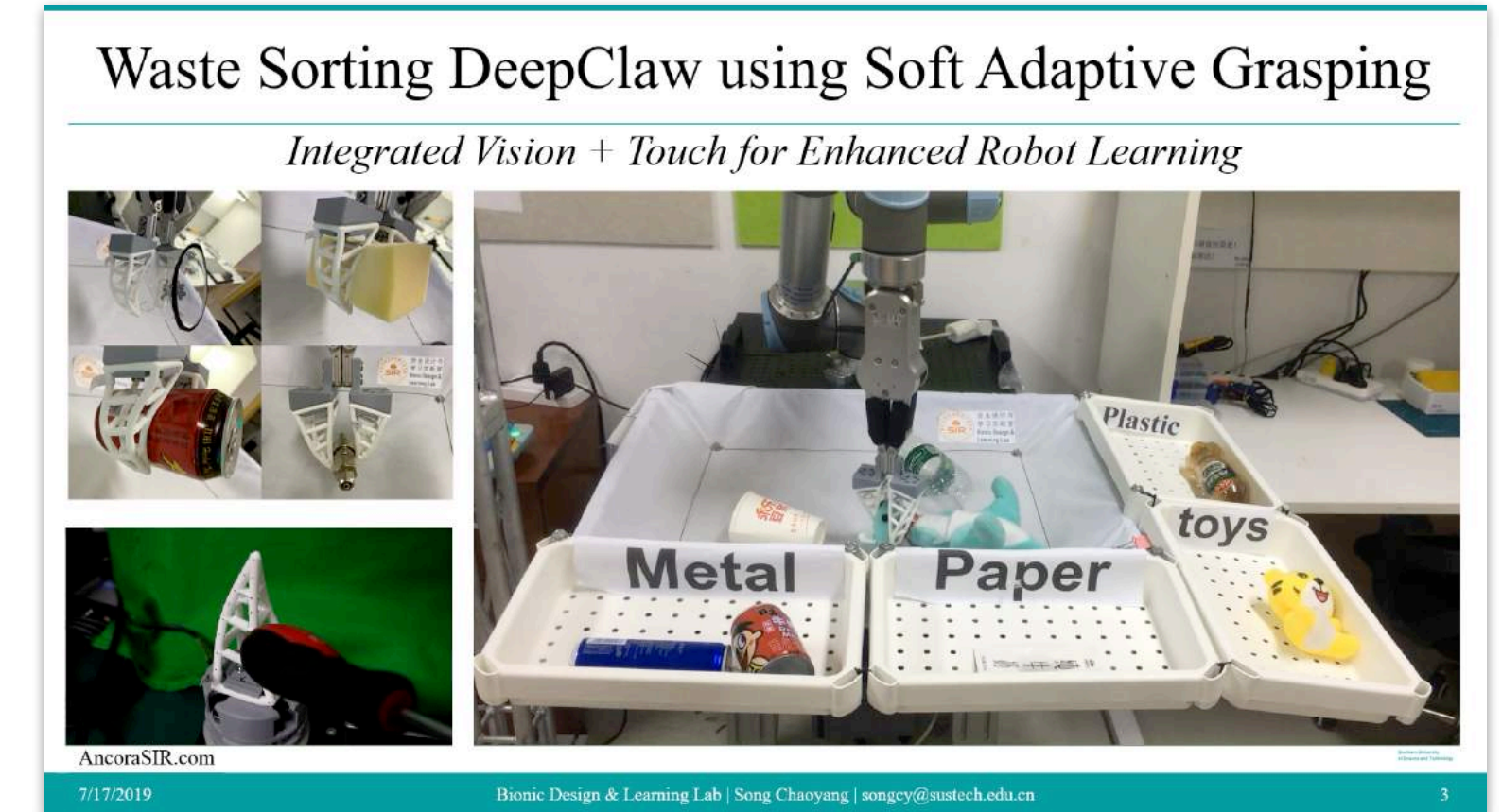


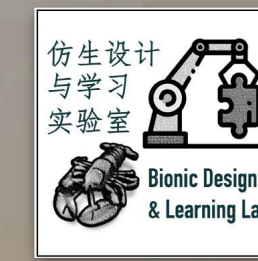
• Dive Deep with Robot Learning (5 weeks)

- Advanced topics focusing on the system integration and algorithm implementation of robot learning.

• Wasteless Robot Learning Challenge (5 weeks)

- A themed challenge among student teams to design and develop a learning system for a competitive picking for waste sorting



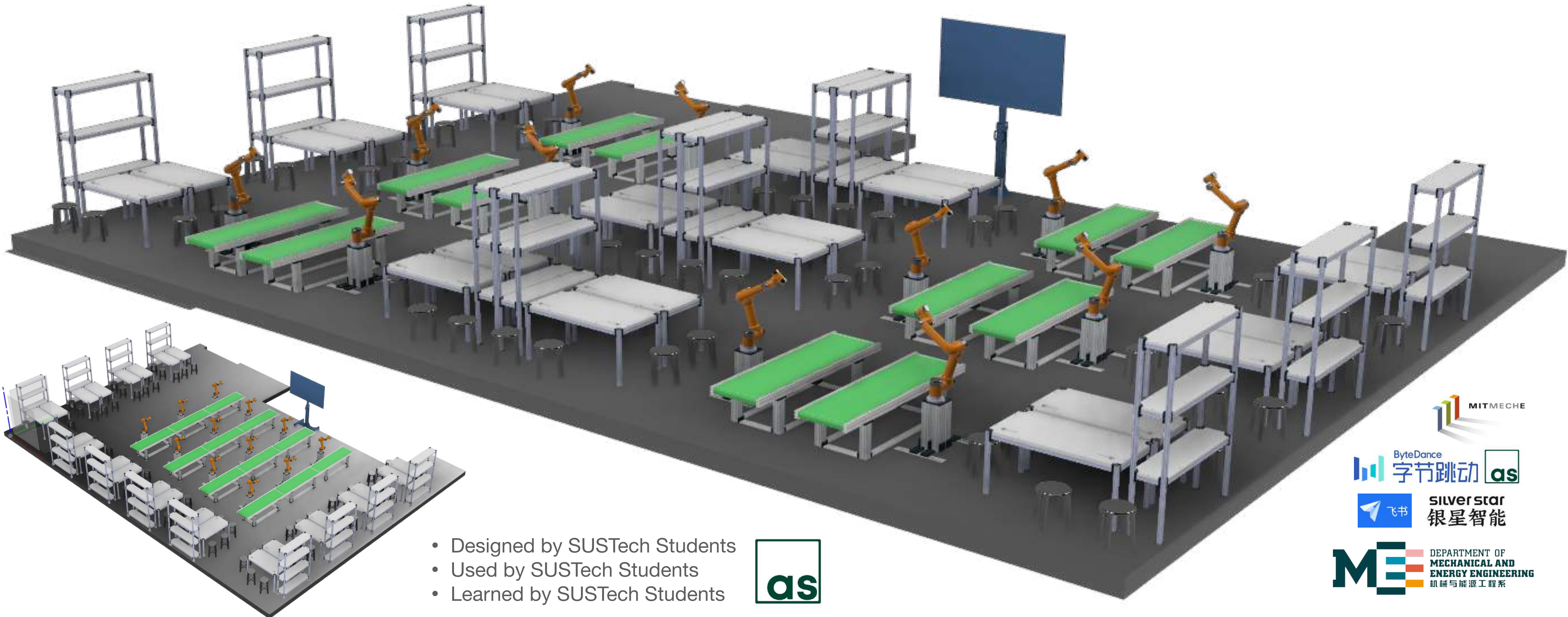


ME303 INTRO TO MECH DESIGN



Machine Intelligence Design & Learning Lab

“Wasteless” Themed Lab Project, Space, and Equipment



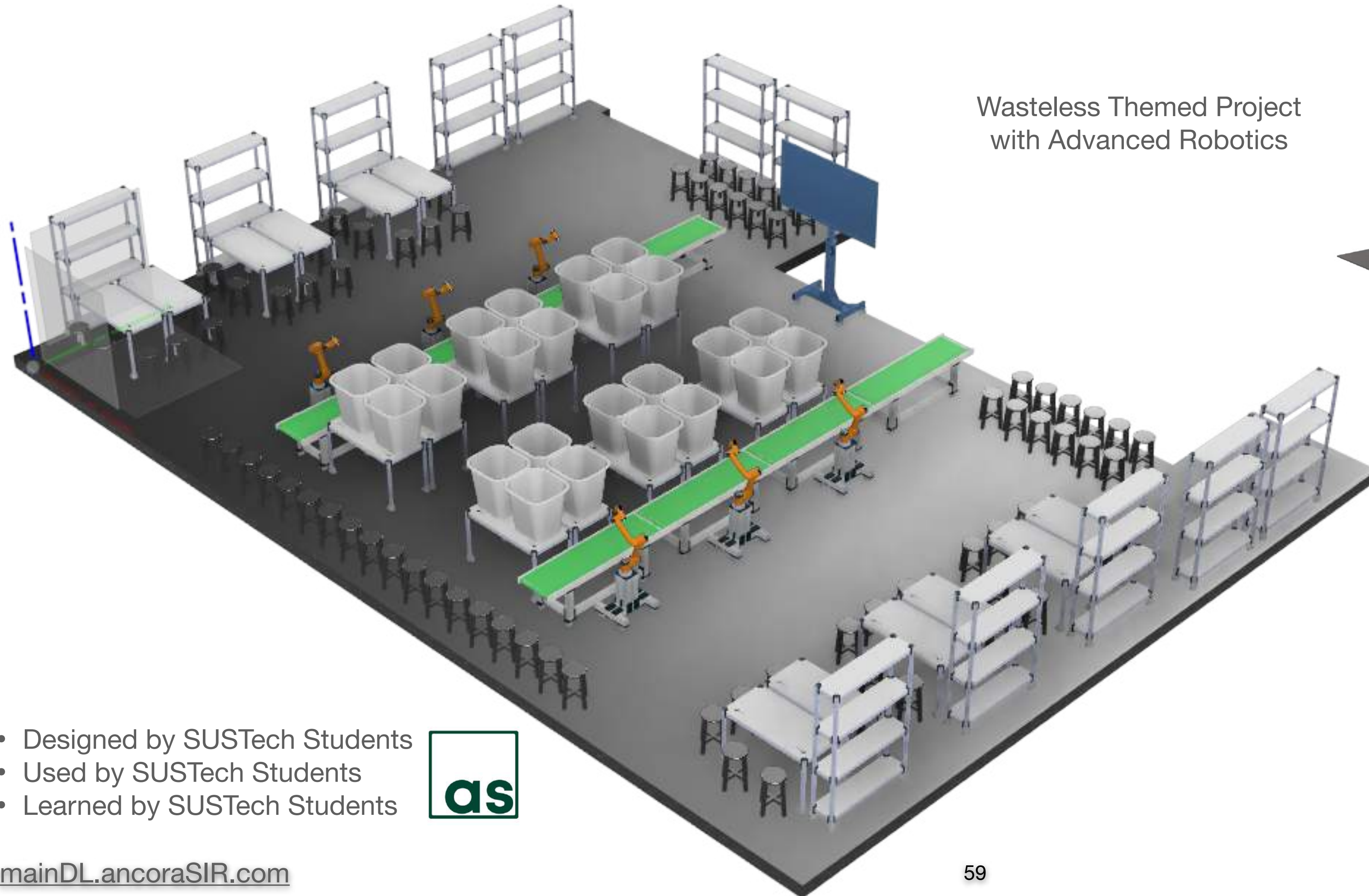
- Designed by SUSTech Students
- Used by SUSTech Students
- Learned by SUSTech Students



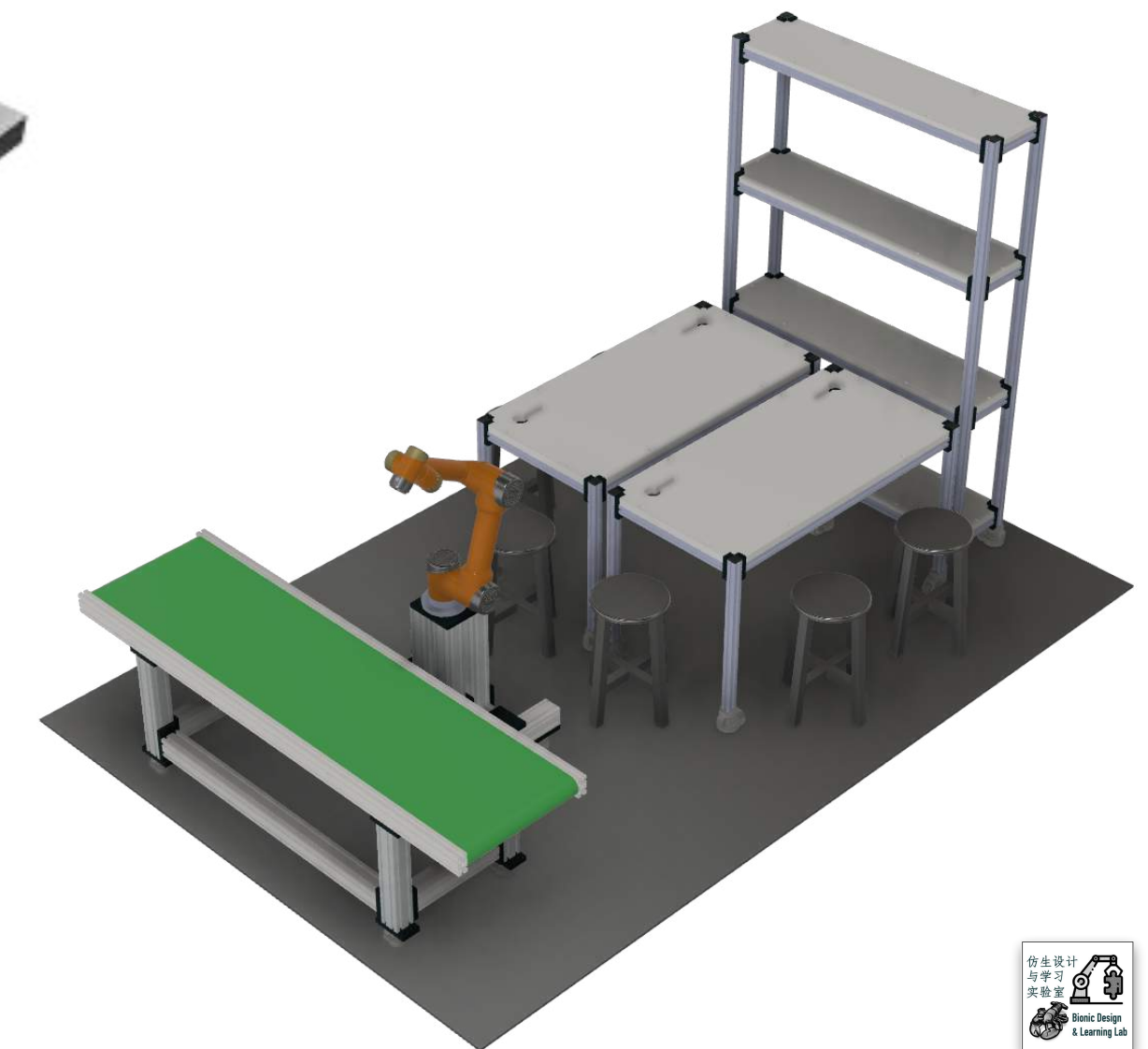
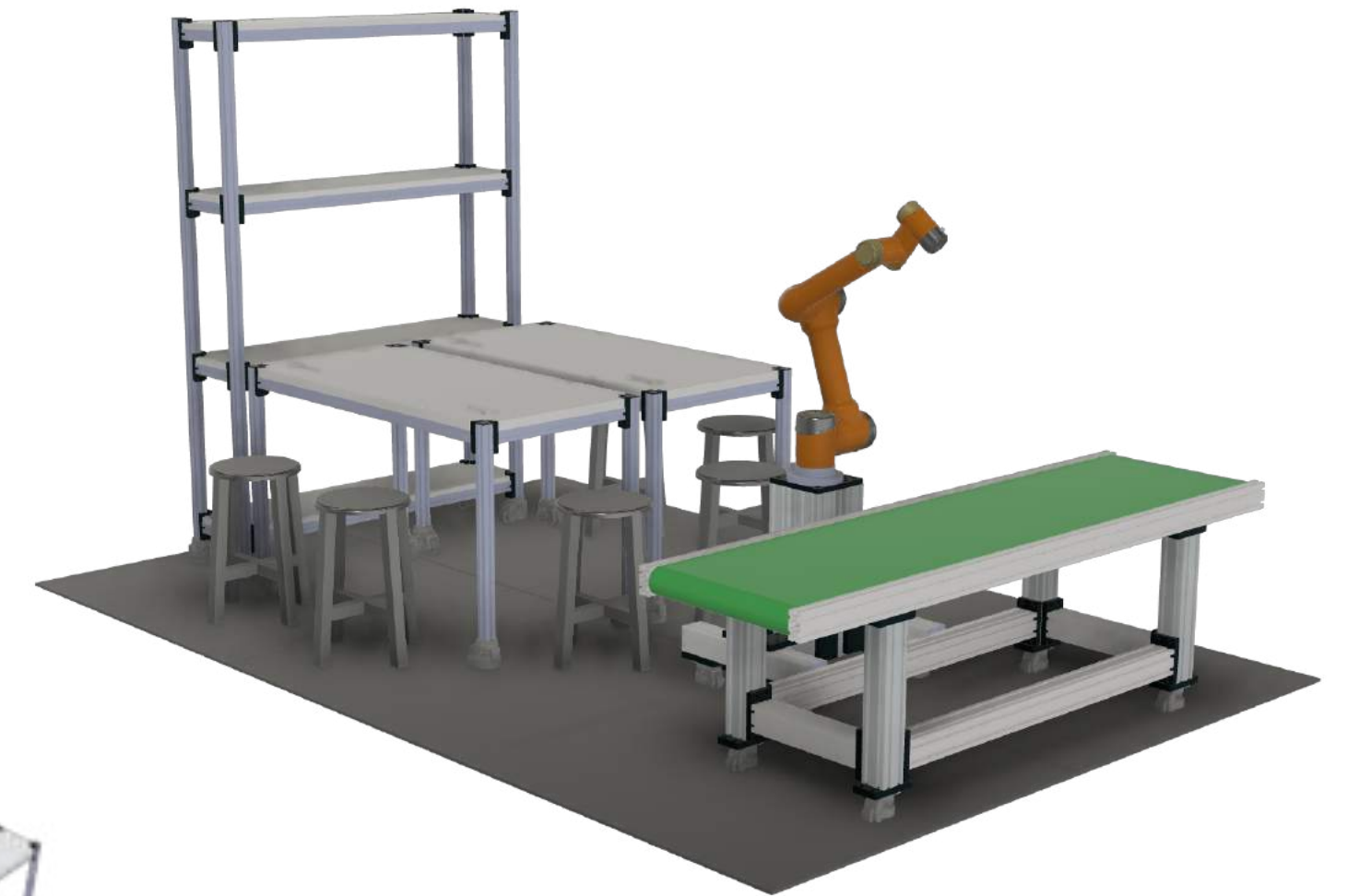


For ME336, From ME303

Reconfigurable Automation SYSTEM for Higher Education



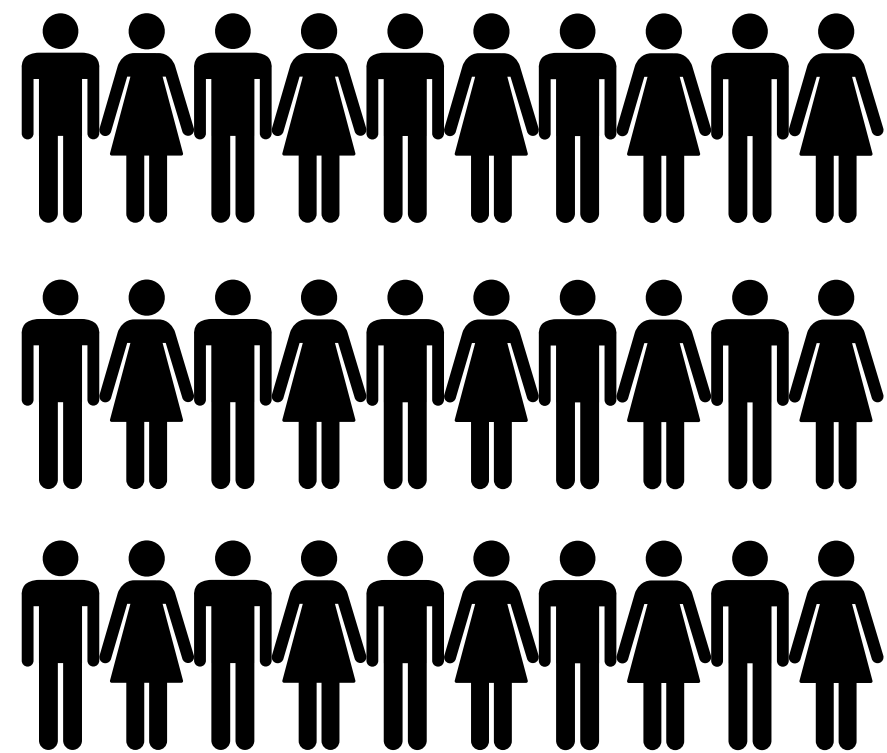
Wasteless Themed Project
with Advanced Robotics



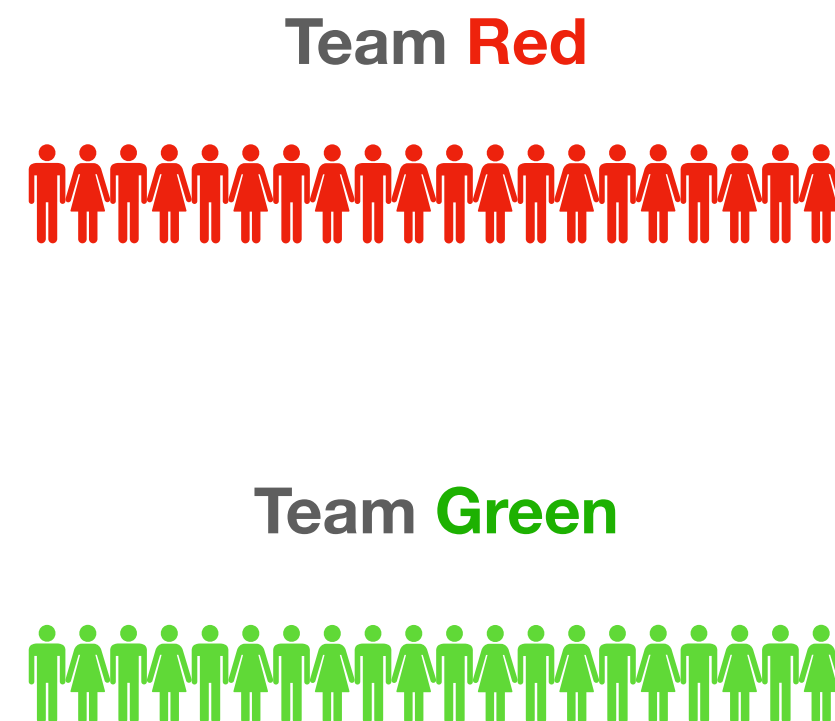
- Designed by SUSTech Students
- Used by SUSTech Students
- Learned by SUSTech Students



All Students of the Class
(~30 per class)



2 Teams of Student Designer
(~15 per team)



3 Task Forces of Student Designers
(7~8 per task force)



8 Team Roles
per team

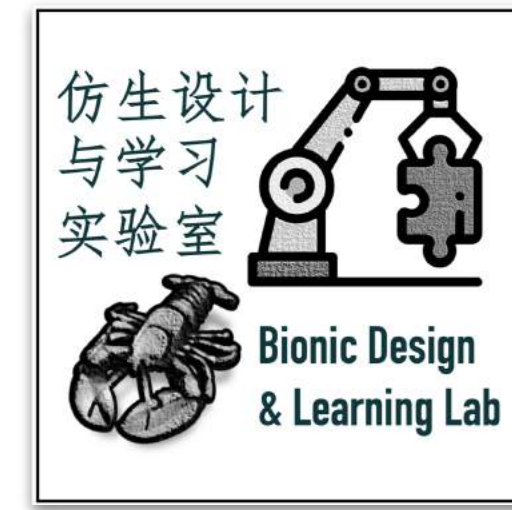
- System Integrator
- Financial Officer
- Tool Officer
- Information Officer
- Team Site Master
- Safety Officer
- Yoda Officer
- Video Log Officer

5 Task Roles
per task force

- Design Engineer
- Algorithm Engineer
- System Engineer
- Software Engineer
- Data Engineer

ME336

Teams & Task Forces

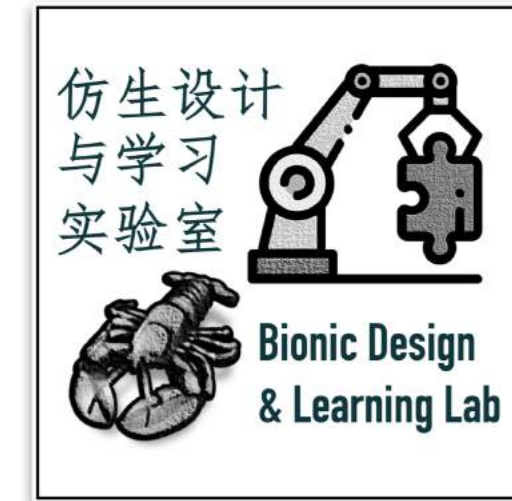


ME336

Collaborative Robot Learning

For more information, please visit mainDL.ancoraSIR.com

Song Chaoyang | Asst. Prof. | Department of Mechanical & Energy Engineering | SUSTech | songcy@sustech.edu.cn



Thank you

Friday, 0800-0950, Room 235, New Engineering Building

Song Chaoyang | Asst. Prof. | Department of Mechanical & Energy Engineering | SUSTech | songcy@sustech.edu.cn