



Source: <https://blenderartists.org/forum/showthread.php?316399-Industrial-Robots>

Robotics for Future Industrial Applications

Introduction to Industrial Robots & Challenges

Philipp Ennen, M.Sc.

- I. Organizational
- II. Introduction
 - I. Rise of Robotics and AI
 - II. Smart Robots for the Manufacturing Industry
- III. Artificial Intelligence in Robotics
 - I. Definition
 - II. Approaches for AI in Robotics

Your Lecturers



Philipp Ennen, M.Sc.
PhD Candidate
„Robot Learning for Contact-Rich Assembly Tasks“


Coffee
Jazz
Wine



Haoming Zhang, M.Sc.
PhD Candidate
„Optimal Control, Machine Learning“

Big Bang Theory
Microcontroller

Organizational

Time	Sunday 13 th August	Monday 14 th August	Tuesday 15 th August	Wednesday 16 th August	Thursday 17 th August	Friday 18 th August	Saturday 19 th August	
9:00-9:30	Free Time for Excursions, Sight-Seeing and Self-Study	Introduction to Industrial Robots & Challenges	Fundamentals of Robot Learning and Control Theory		Robot Learning with iterative Linear- Quadratic Regulator	Practical Unit: Robot Learning of an Assembly Task	Free Time for Excursions, Sight-Seeing and Self-Study	
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20:00-20:30				OPTIONAL International Tuesday with the INCAS Student Organisation 		„Karaoke Night“ with the Summer School Team and your Buddies 		
20:30-21:00								
21:00-21:30								
21:30-22:00								

Website and Wifi

- Website for this week: <https://goo.gl/X5twgi>

Home

Summerschool 2017: Intelligent Motion Planning

We (Haoming and Philipp) are glad to be organizing a seminar on intelligent motion planning for industrial robots. This seminar will be held in week two of the summer school "Robotics for Future Industrial Applications" at the Cybernetics Lab of the RWTH Aachen University (14. august - 18. august). This page will provide you with lecture materials and introductions for the practicals. If you have any further questions, feel free to contact us via email.

Contact details of Haoming and Philipp



- Wifi-Access:
 - SSID: ariz
 - Password: ariz_ab12345!

Day 1: Introduction

Lecture: Introduction to Industrial Robots and Challenges

I. Organizational

II. Introduction

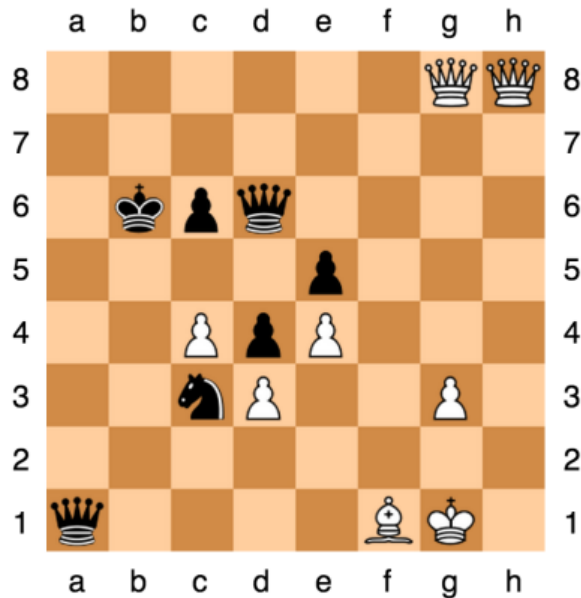
- I. Rise of Robotics and AI
- II. Smart Robots for the Manufacturing Industry

III. Artificial Intelligence in Robotics

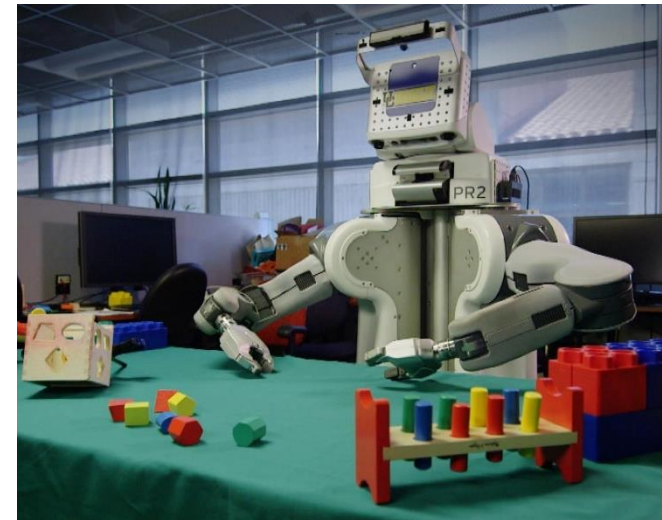
- I. Definition
- II. Approaches for AI in Robotics

„In robotics, the easy problems are hard and the hard problems are easy“

S. Pinker. The Language Instinct. New York: Harper Perennial Modern Classics, 1994



Building a computer that defeats
Chess World Champion Garri
Kasparow: **Easy**



Build a robot with "healthy human
understanding" (i.e. motor skills):
Difficult

The rise of Robotics and AI

Fueled by advances in computing power and connectivity, the fields of robotics and artificial intelligence have grown rapidly

1921
The term **robot** is first used by Czech writer Karel Capek



1939
Elektro, a humanoid robot, debuts at the World's Fair, smoking cigarettes and blowing up balloons



1948
William Grey Walter creates the first autonomous robot with complex behavior



1950
Alan Turing publishes paper about the possibility of machines that think, develops idea known as the **Turing's Test**.
It tests a machine's ability to "think" by answering a series of questions. In essence, the tester must think the machine's answers are coming from a human

1941
Isaac Asimov formulates the **Three Laws of Robotics**:



A robot may not injure a human being or, through inaction, allow a human being to be harmed

A robot must obey orders given it by human beings except where such orders would conflict with the First Law

A robot must protect its own existence as long as such protection does not conflict with the First or Second Law

1954
George Devol invents the first digitally operated and programmable robot

1951
Marvin Minsky builds the first neurocomputer, **SNARC**

1956
IBM 305, the first hard disk drive **5MB**

1970
IBM 1330 **100MB** per pack

1985
IBM 0665, a 5.25" disk with **20-40MB**

Minimize and maximize
Shrinking disk sizes and exponentially growing capacity help fuel robotics and AI

1956
Field of AI research founded at a conference at Dartmouth

1960
Frank Rosenblatt constructs **Mark I Perceptron**, a computer that learned new skills by trial and error

1968
Mobile robot **"Shakey"** is introduced. It's controlled by a computer the size of a room



1972
Stanford researcher develops **PARRY**, designed to simulate a paranoid schizophrenic.

1961
GM installs **Unimate** robot to lift and stack hot pieces of metal

1974
Intel produces its second-generation 8080 general-purpose chips

1979
SCARA, an articulated robot arm, is developed for assembly lines



1984
Doug Lenat and his team start **Cyc**, to codify millions of pieces of knowledge that compose human common sense

virtual reality



1984
The **RB5X**, developed by General Robotics Corp., includes software enabling it to learn from its environment



1985
Jaron Lanier's VPL Research, Inc., sells first VR glasses and gloves; Lanier coins the phrase

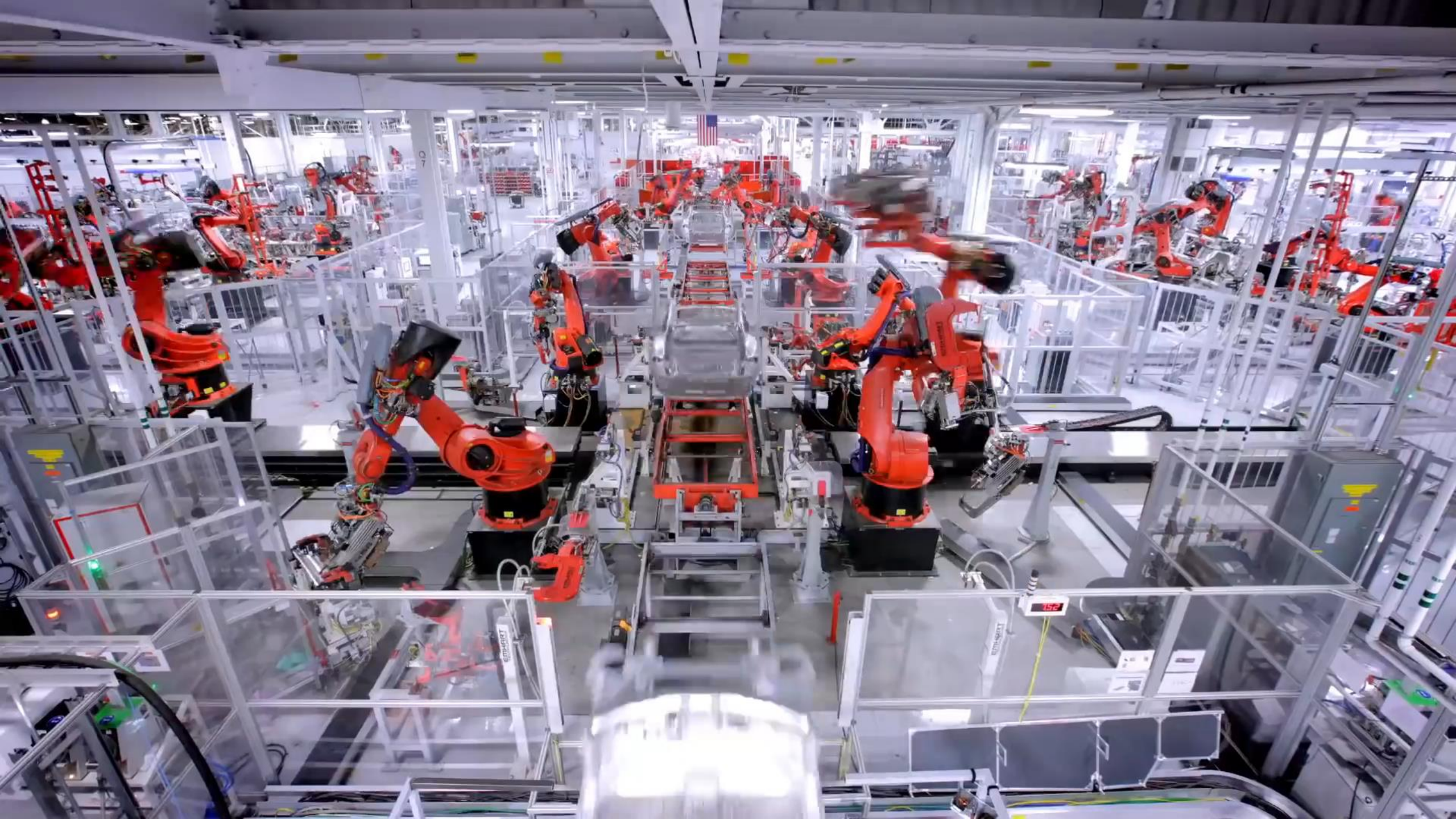
1986
Honda creates the E0, the first of a series of humanoid robots that walk on two feet

1988
Researchers launch **Jabberwacky**, an AI chatbot designed to learn through conversation



1988
The first **HelpMate service robot** begins work at Danbury Hospital

Nope, I'm human.

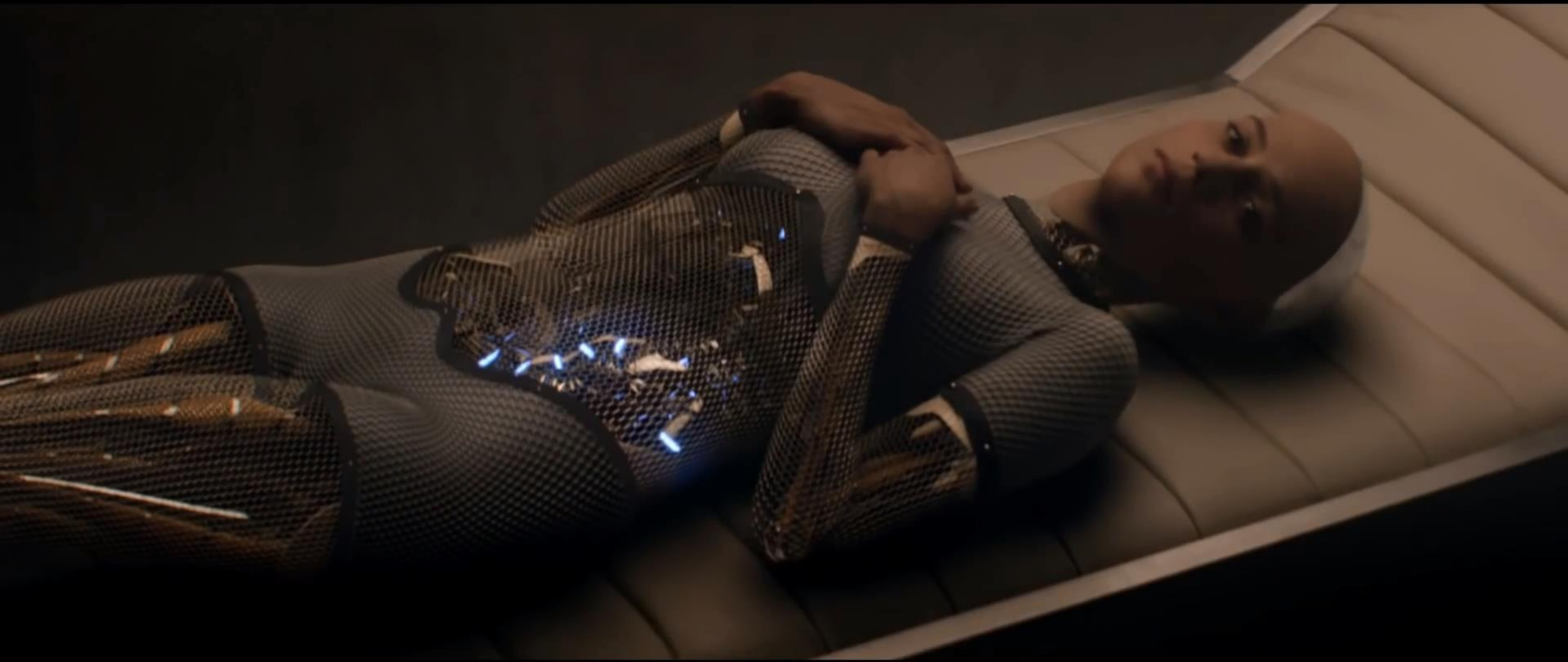


Traditional Industrial Robots

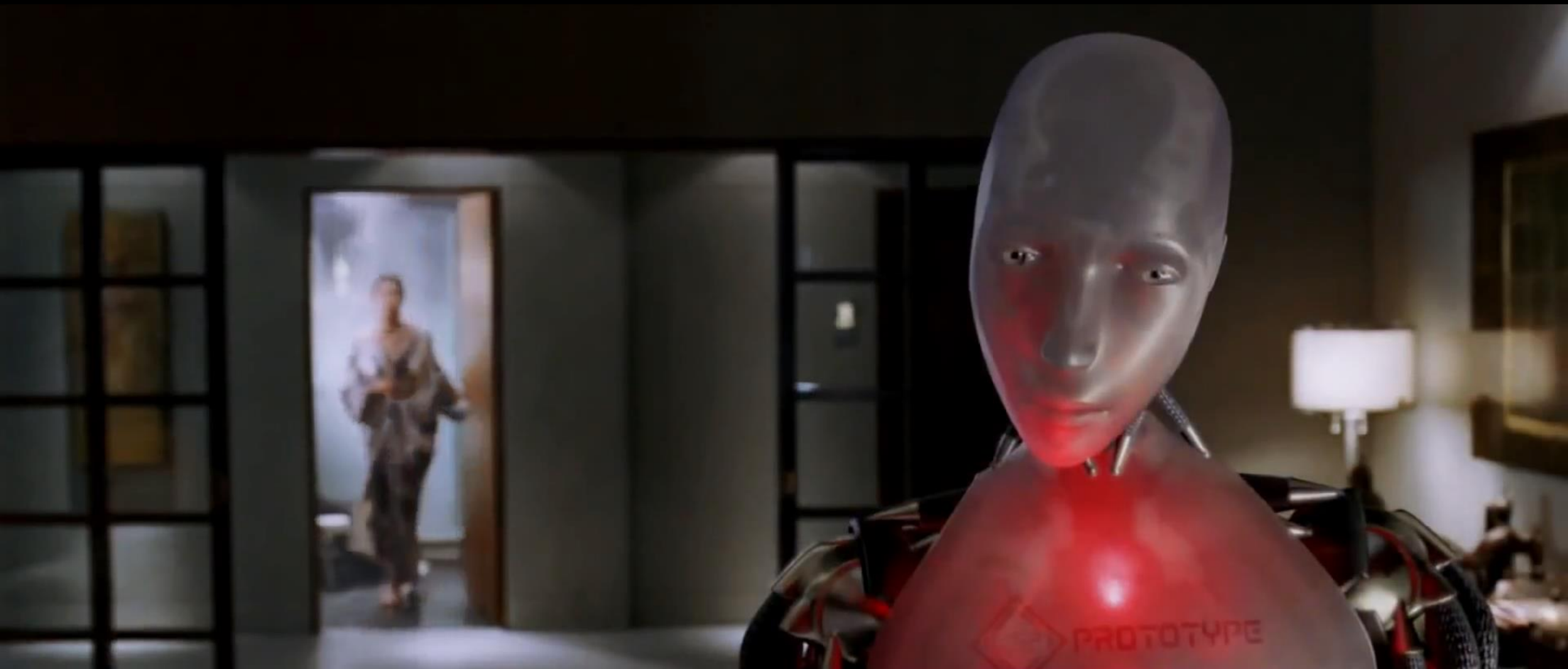
- ... high precision and velocity
- ... high programming effort (through experts)
- ... inflexible
- ... stationary
- ... high driven potential, protection fences necessary
- ... high integration/commissioning effort
- ... only isolated sensor integration, no environment detection



What do we really want?



What do we really want?



Robots getting smart!



- Adaptability
- Motion Ability

Robots getting smart!



- **Adaptability**
- Motion Ability

Adaptability for Industrial Robots

Robot reacts to changes in the operating environment

- Auto-Configuration strategies
- Self-Learning strategies

Robots getting smart!



- Adaptability
- **Motion Ability**

Motion Ability for Industrial Robots

Kinematics/Dynamics for advanced reconfigurable work cells

- Positioning
- Navigation

Robots getting smart!



- Adaptability
- Motion Ability
- **Interaction Ability**

Interaction Ability for Industrial Robots

Interaction with Operators, Robots and other Systems

- Interaction must be safe
- Interaction must be intuitive
- Interaction must appropriate

Robots getting smart!



- Adaptability
- Motion Ability
- Interaction Ability
- **Manipulation Ability**

Manipulation Ability for Industrial Robots

Handle material objects and tools

- Adaptability
- Robustness
- Accuracy
- Repeatability

Robots getting smart!



- Adaptability
- Motion Ability
- Interaction Ability
- Manipulation Ability
- **Perception Ability**
- Cognitive Abilities
- Decisional Autonomy

Perception Ability for Industrial Robots

Environment Sensing

- Choice of Sensing Modality
- Efficient Signal and Data Analysis
- Generating Maximum Information Output from the Data
- Guaranteed Safe Perception

Robots getting smart!



- Adaptability
- Motion Ability
- Interaction Ability
- Manipulation Ability
- Perception Ability
- **Cognitive Abilities**
- Decisional Autonomy

Cognitive Abilities for Industrial Robots

“In the context of manufacturing, the **greatest potential** is for functions that contribute to a **reduction of programming and configuration requirements** in deployed systems. There are clear benefits for small lot size systems in reducing the time and skill needed to reconfigure an adapt systems to new processes.”

Robots getting smart!



- Adaptability
- Motion Ability
- Interaction Ability
- Manipulation Ability
- Perception Ability
- Cognitive Abilities
- **Decisional Autonomy**

Decisional Autonomy for Industrial Robots

Increase level of responsibility in the control process

- Reducing energy consumption
- Increasing throughput
- Providing Context Aware Task Control

Robots getting smart!



- Adaptability
- Motion Ability
- Interaction Ability
- Manipulation Ability
- Perception Ability
- Cognitive Abilities
- Decisional Autonomy

What will be the impact of smart robots in the manufacturing industry?

Industrial Robots of the Future

- ... work together with humans
- ... do not pose a risk to humans
- ... interact intuitively with humans (multimodal)
- ... are easy to program
- ... possess a wide range of sensory abilities
- ... are flexible to use
- ... have a high degree of autonomy



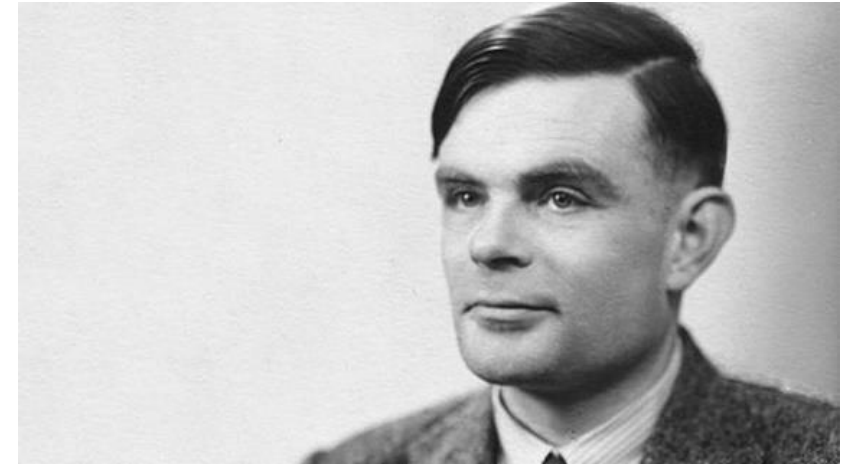
This week topic

How can we implement
intelligence in industrials
robots?

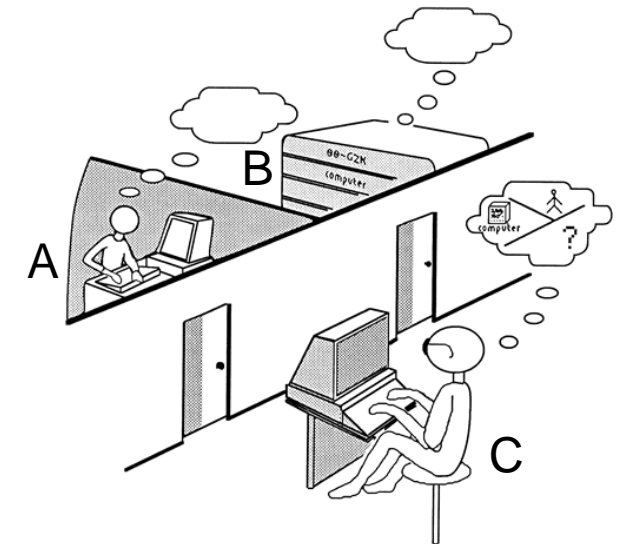
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What is Artificial Intelligence?

“Instead of trying to produce a programme to simulate the adult mind, why not rather try to produce one which simulates the child's? If this were then subjected to an appropriate course of education one would obtain the adult brain.”, Alan Turing



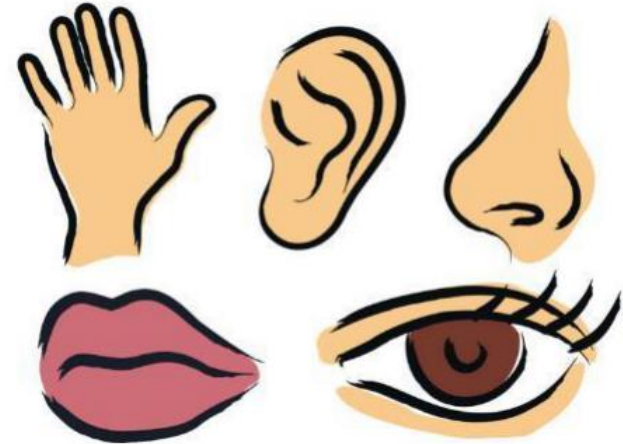
- The (simplified) **Turing Test** proves the existence of an AI
 - Human C talks with A and B
 - A and B try to convince that both are intelligent humans
 - If human C is not able to find out if A or B is a human, the Turing-Test is passed



What do we need for an intelligent machine?

- Need for interpreting heterogeneous sensor values
 - (What das interpreting mean?)

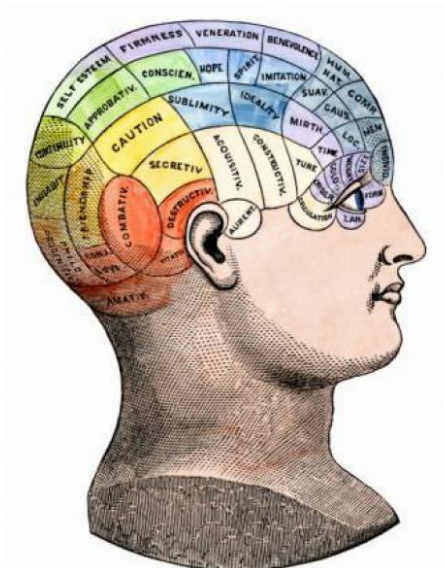
- Need for choosing and acting complex actions



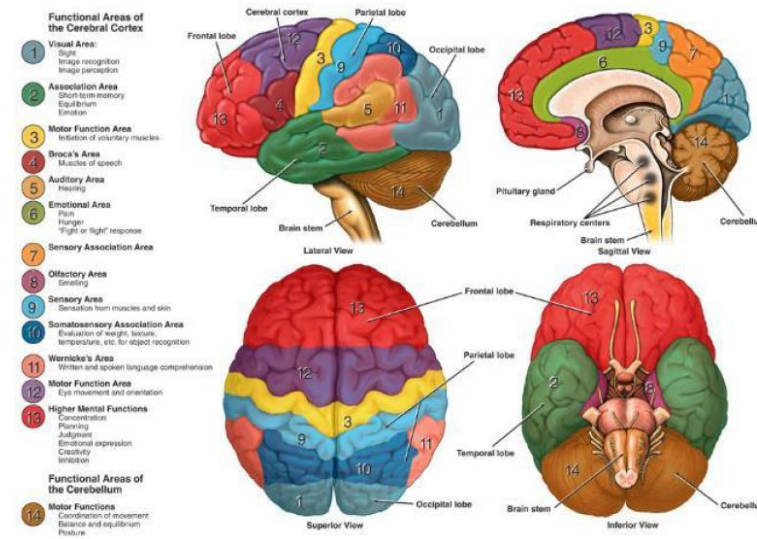
Introduction

How would you start to build an intelligent machine?

“We have a brain for one reason and one reason only: to produce adaptable and complex movements”, Daniel Wolpert TED 2011 (Neuro-Scientist)

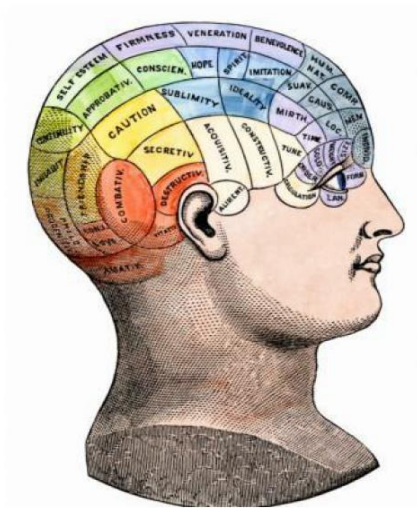


Anatomy and Functional Areas of the Brain



How would you start to build an intelligent machine?

An algorithm for each component in the brain?



Standard technology

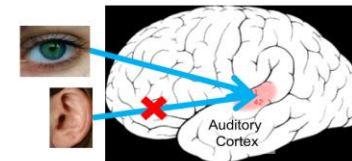
A single flexible algorithm?



Seeing with your tongue



Human echolocation (sonar)



[BrainPort; Martinez et al; Roe et al.]
adapted from A. Ng



End-to-End Learning
(z.B. Deep Learning)

Standard Cooking vs End-to-End Cooking



Standard

The central questions of robotics

What is it all about?

MOTION



Robots as an Example for Intelligent Machines

The central questions of robotics

Three main question:

Where am I?



- Easy in Industrial Robotics
- Hard in Mobile Robotics

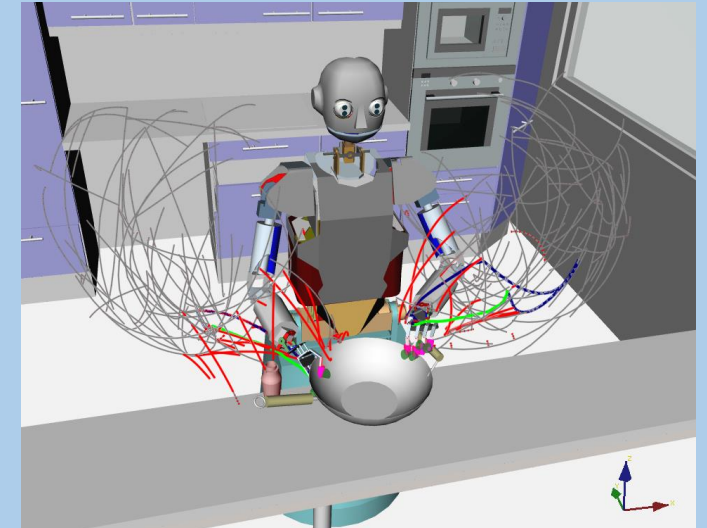
Where should I go?



- Often hand-engineered
- Or a Result of Task Planning

This weeks focus!

How do I go there?

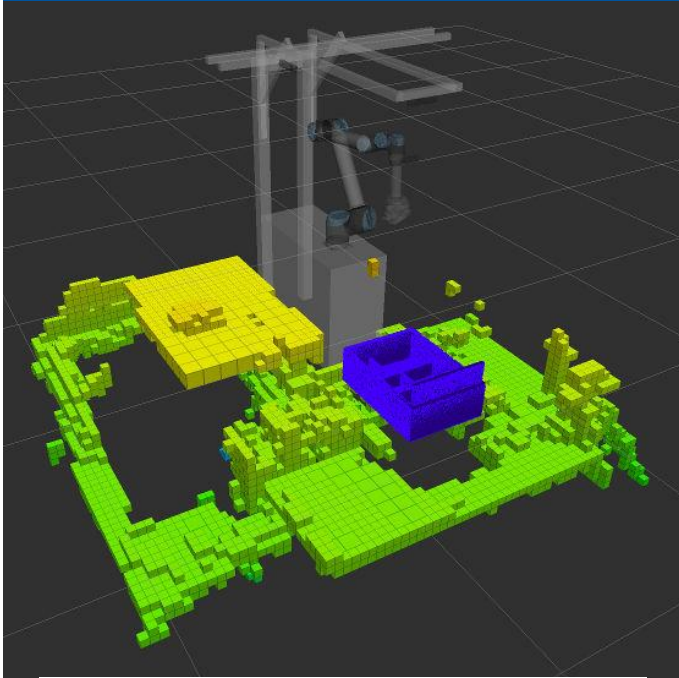


- Comparable Easy in Mobile Robotics
- Hard in Industrial Robotics

Robots as an Example for Intelligent Machines

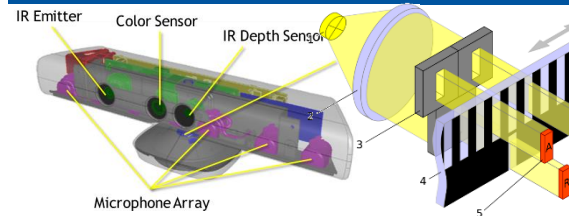
How do I (the robot) go there?

Model of the Robot and Environment



$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\phi} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \frac{-(I + ml^2)b}{I(M+m) + Mml^2} & \frac{m^2 gl^2}{I(M+m) + Mml^2} \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ \phi \end{bmatrix} + \begin{bmatrix} 0 \\ \frac{I + ml^2}{I(M+m) + Mml^2} \\ \frac{mgl(M+m)}{I(M+m) + Mml^2} \end{bmatrix} u$$

Sensing



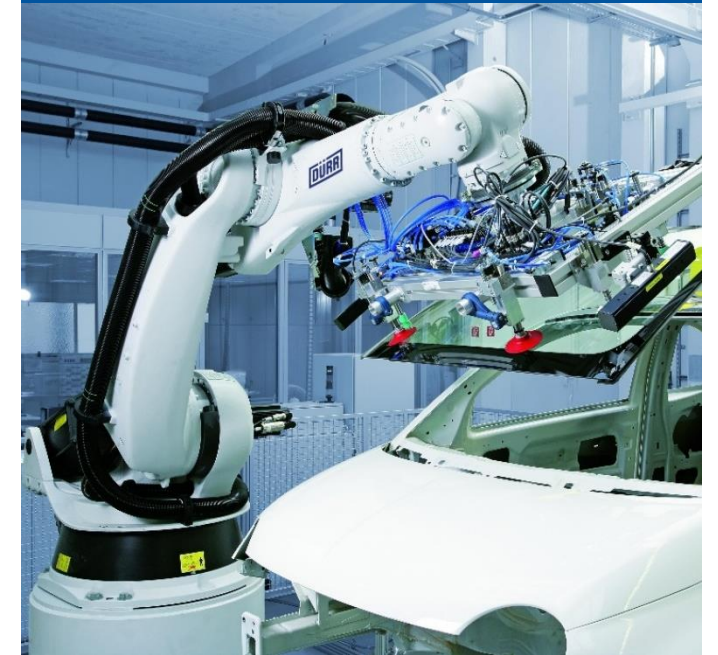
This weeks topic!

Motion Planning

Requires Goalstate:

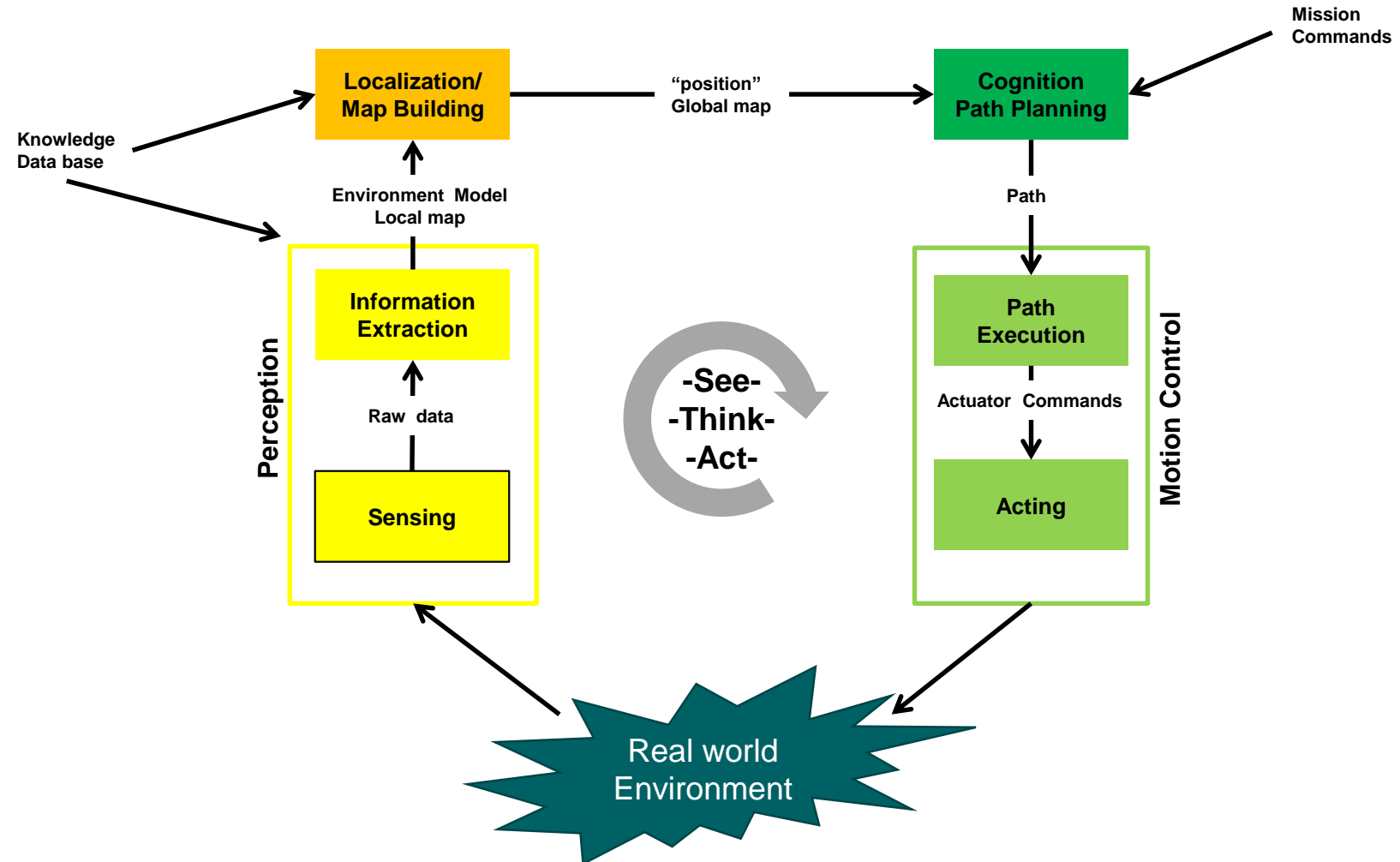
- i.e. hand-engineered
- i.e. via a cost function

Motion Control and Execution



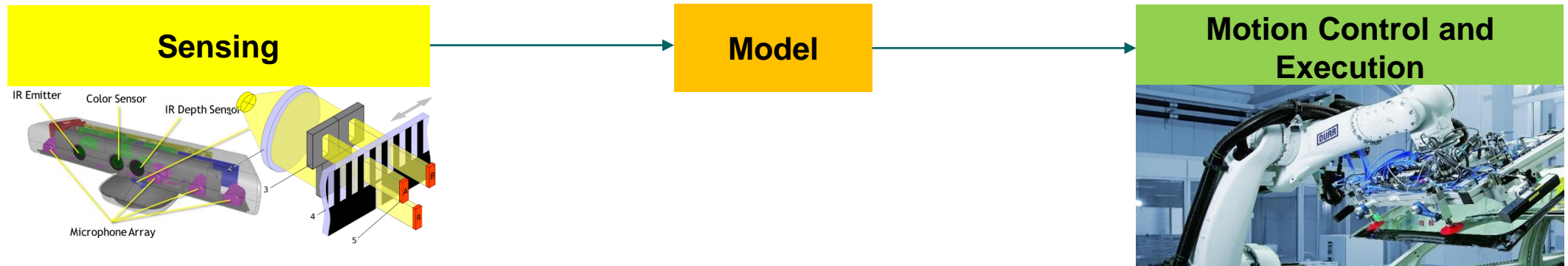
Robot-Architecture: See-Think-Act Cycle

- Popular in Mobile Robotics
 - See: Perceive the environment
 - Think: Path Planning
 - Act: Execute Path

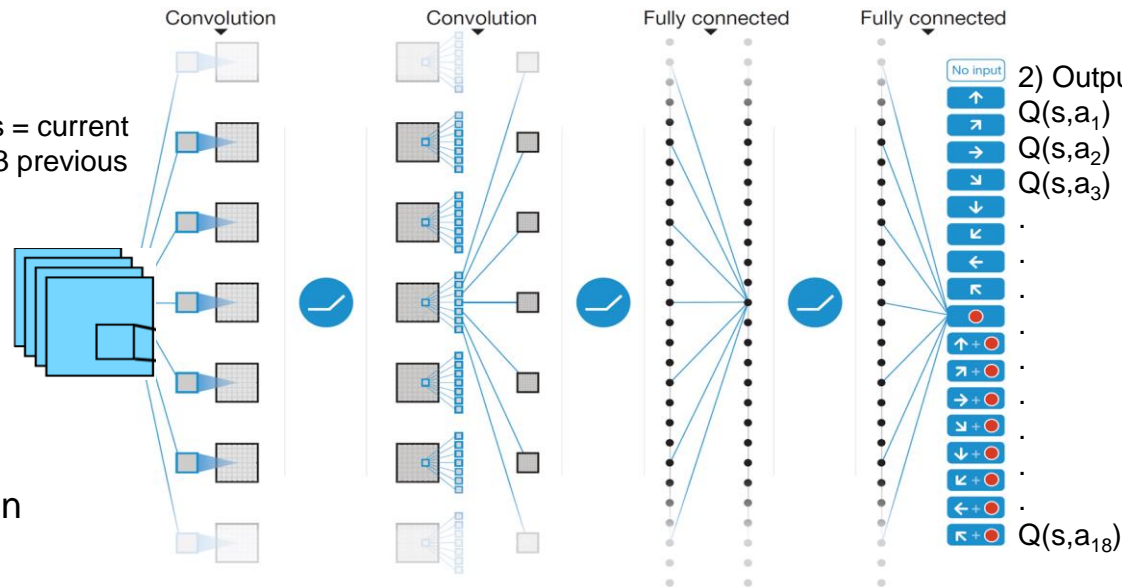


Robots as an Example for Intelligent Machines

Robot-Architecture: End-To-End Control



1) Input:
4 images = current
frame + 3 previous



2) Output: $Q(s, a_i)$

$Q(s, a_1)$

$Q(s, a_2)$

$Q(s, a_3)$

\vdots

\vdots

\vdots

\vdots

\vdots

\vdots

\vdots

\vdots

\vdots

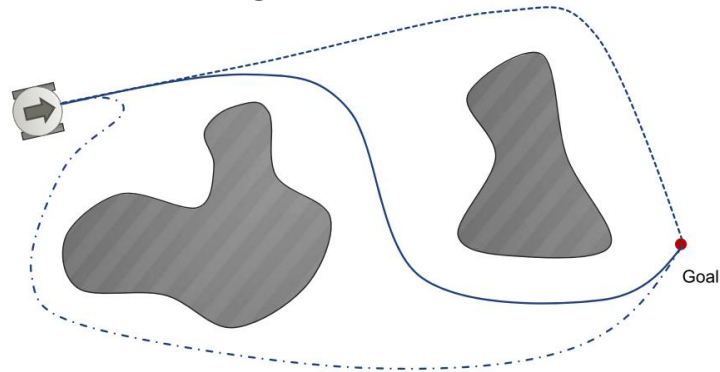
$Q(s, a_{18})$

- Popular in Industrial Robotics

- Sensing: Perceive the state
- Model: Calculates Action
- Motion C. and E.: Executes Action

Motion Planning

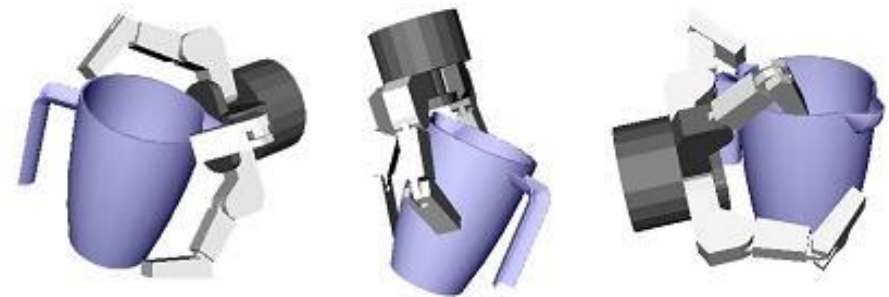
Path-Planning



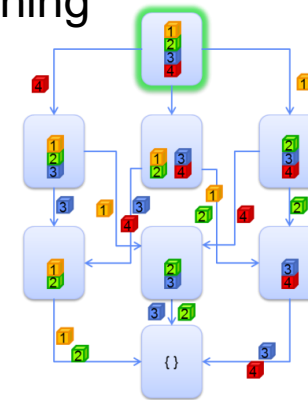
Task-Planning



Grasp-Planning



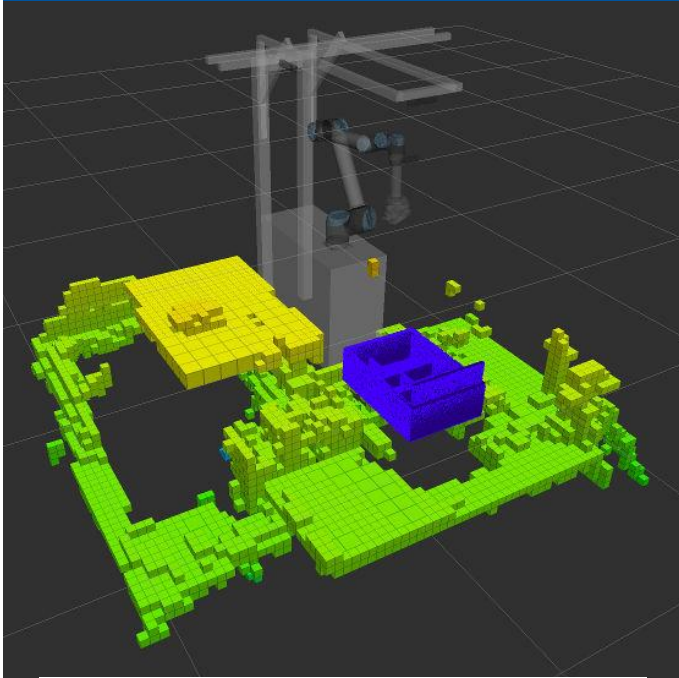
Assembly-Planning



Robots as an Example for Intelligent Machines

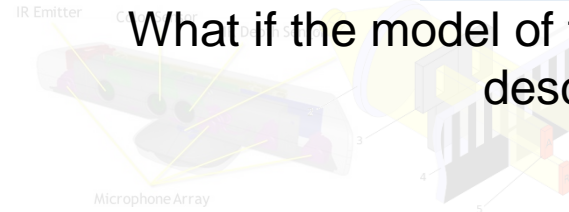
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Model of the Robot and Environment



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Sensing



What if the model of the robot and environment is hard to describe (or unknown)?

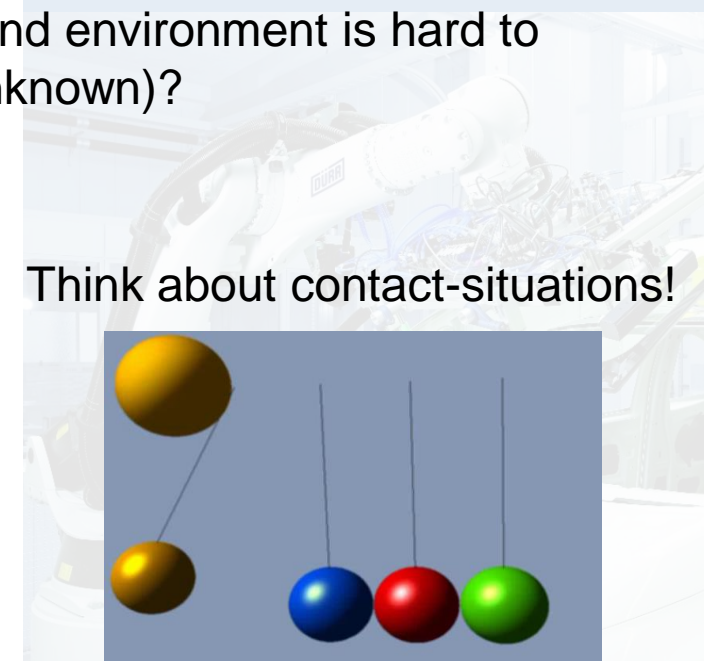
Think about flexible objects!



Requires Goalstate:

- i.e. hand-engineered
- i.e. via a cost function

Motion Control and Execution

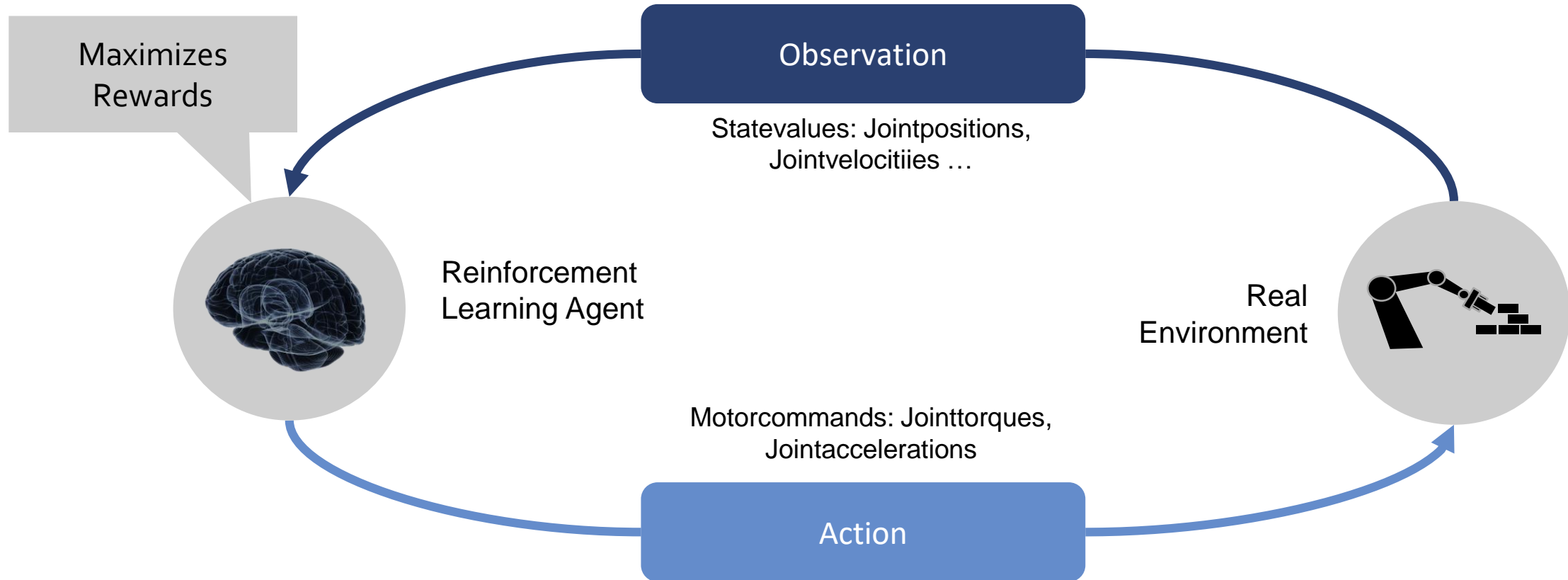


Think about contact-situations!

Robots as an Example for Intelligent Machines

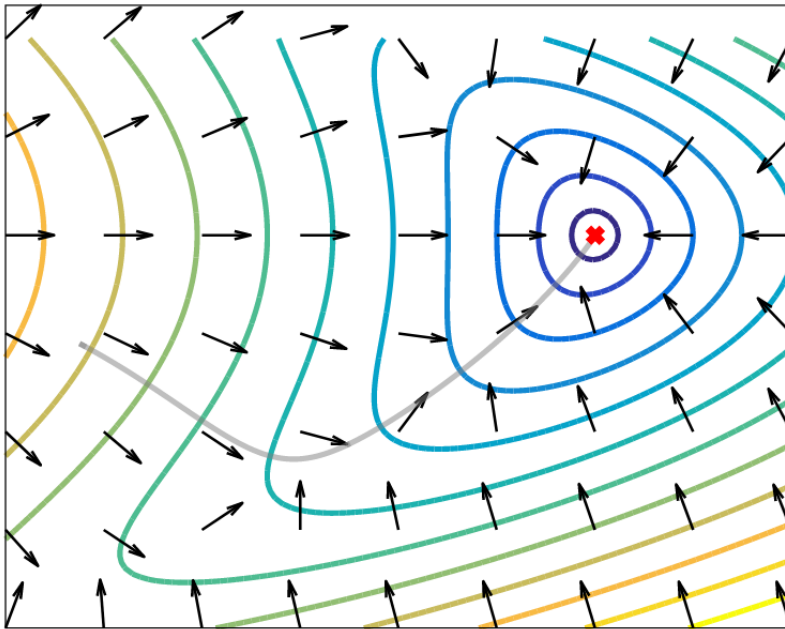
What if the model of the robot and environment is hard to describe (or unknown)?

- Use an Reinforcement Learning Agent!



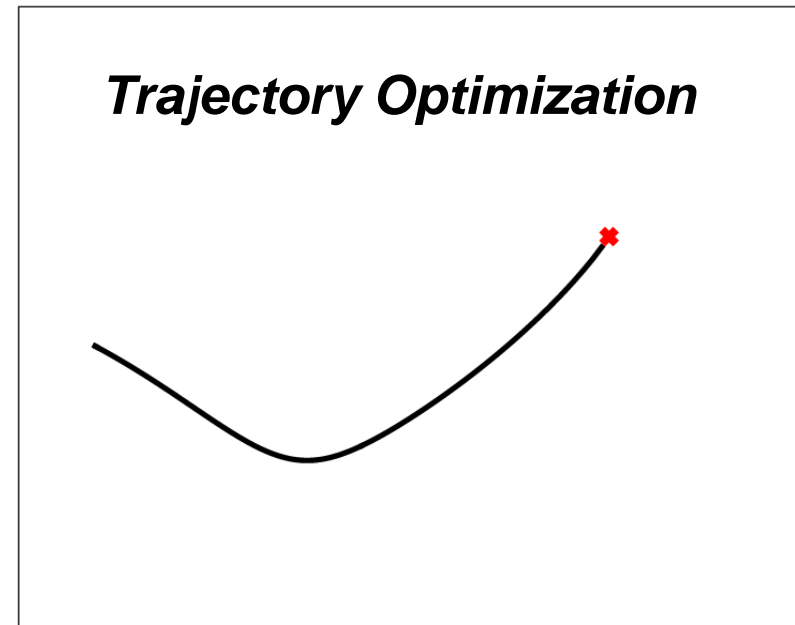
What if the model of the robot and environment is hard to describe (or unknown)?

Closed-Loop
 $u = u(x)$



- Global motion strategy (a “**policy**”)
- Robot chooses an appropriate action in each state

Open-Loop
 $u(t)$



- Local motion strategy (a “**trajectory**”)
- Only valid in a specific region

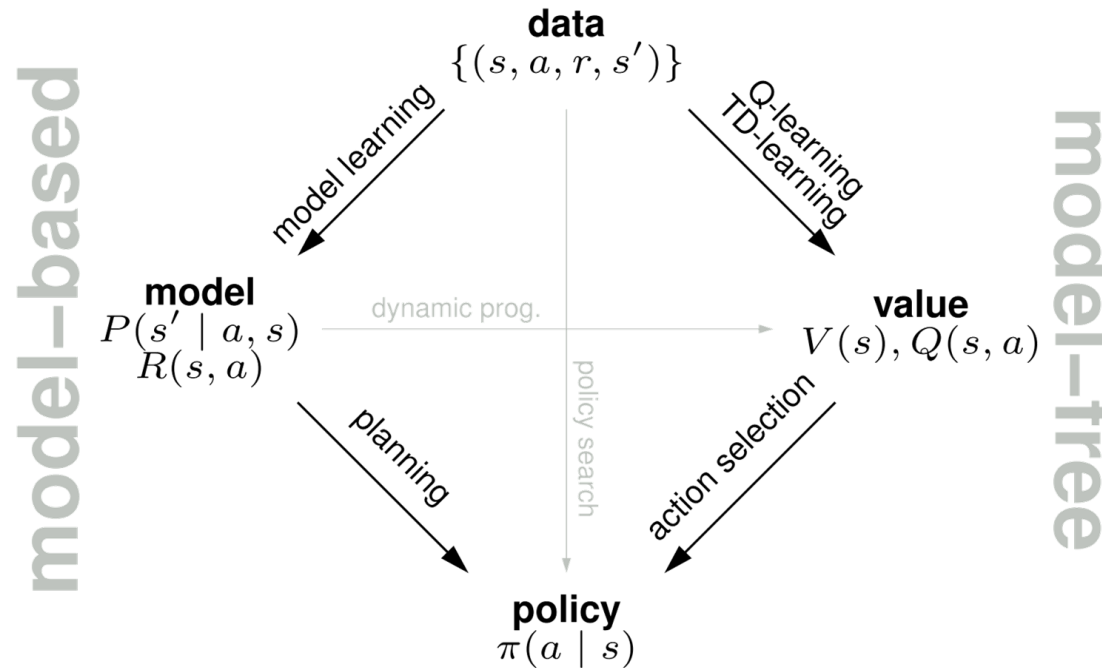
Robots as an Example for Intelligent Machines

What if the model of the robot and environment is hard to describe (or unknown)?

This weeks topic!

Model-based RL:

- Learn to predict next state: $P(s'|s, a)$
- Learn to predict immediate reward $P(r'|s, a)$






Model-free RL:

- Learn to predict value: $V(s)$ or $Q(s, a)$

Reinforcement Learning with End-to-End Technology



Organizational

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