

SDU Robotics



Research

Robot Control

- Kinodynamic modelling and planning
- Force-based control
- Process Optimization

Computer Vision

- Object recognition and Pose estimation
- Optical issues in Computer Vision
- Inspection (NEW)

Model Based Robot Systems Engineering

- Digital Twins
- Visual Programming
- Co-simulation with MATLAB, ROS, ...

Selection and configuration hardware.



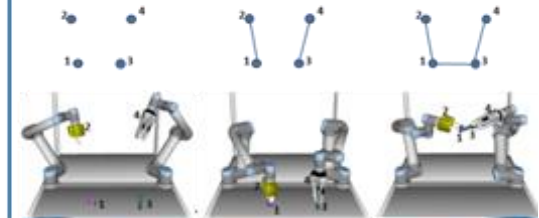
Modeling of hardware for simulation.



Sequence of events in the assembly process.



Configurations described by semantic event plans.



Configurations parameterized as e.g. PMF.



Simulation of trial configurations.



Testing of trial configurations





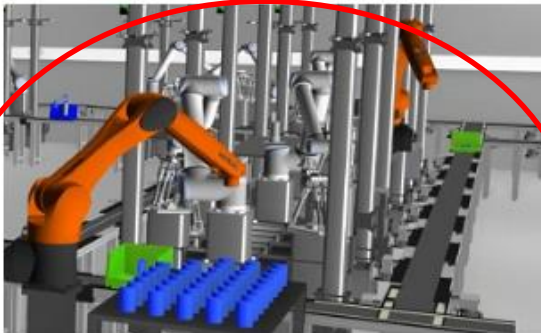
Robotic Solutions for Industrial Applications



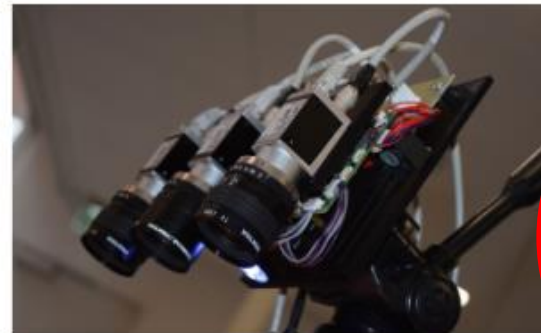
Welfare Robotics



Surgical Robotics



Mathematical Modelling and Simulation of Robotic Systems and Processes



Computer Vision based Object Recognition and Pose Estimation



Model and Learning-based Control of Robotic Systems

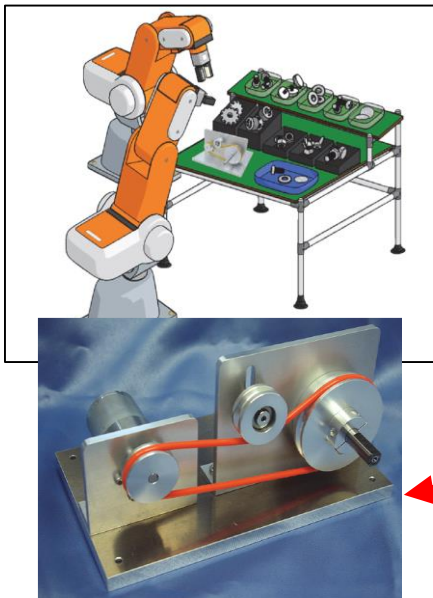


Industrial Robotics

World Robotic Challenge

SDU Robotics wins the “unofficial World Cup” in industrial robotics

- 4 day competition in Tokyo in October 2018
- 16 teams selected from 250 applicants
- Relevant industrial assembly challenges



History

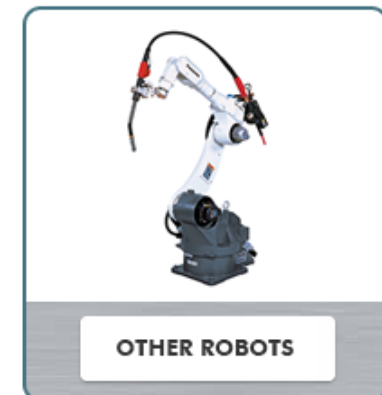
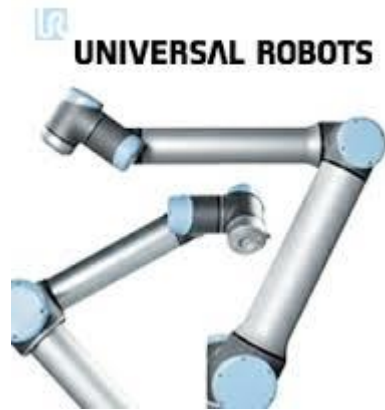
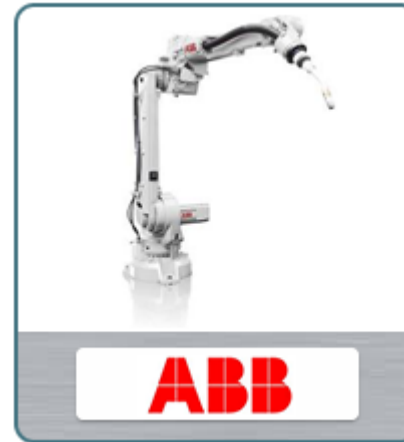
- Unimate 1956-70
 - Hydraulic



- 1970's
 - Electric
 - Up to 6 degrees of motion



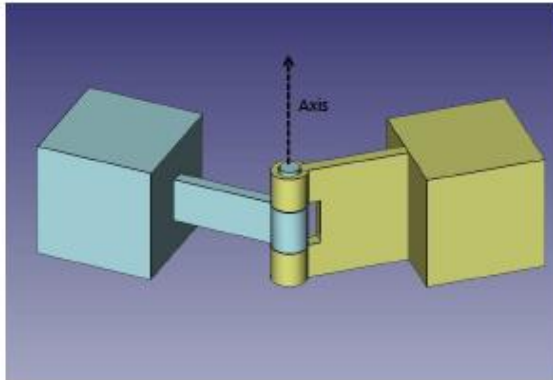
Pictures of different robots



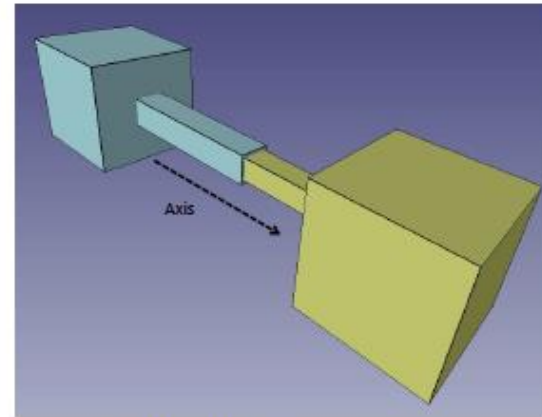
The next 23 slides will illustrate why an industrial robot manipulator arm is just a stupid machine that is easy to use as any other tool...



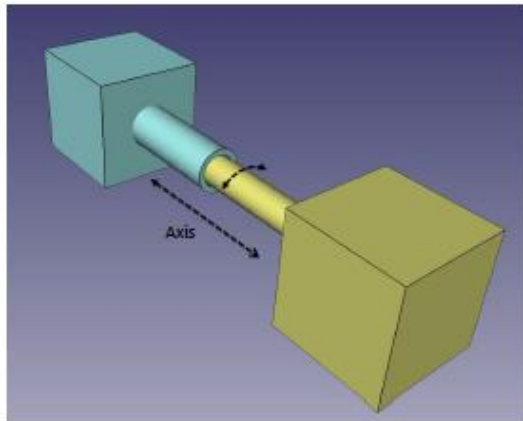
Robot joints



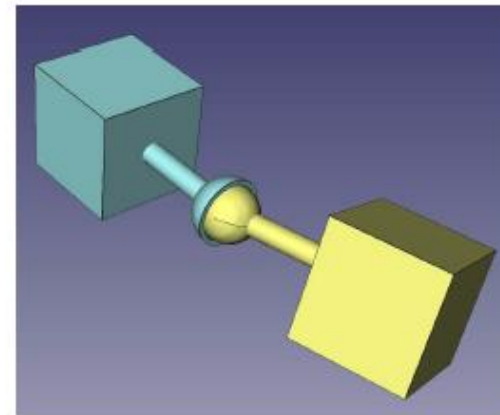
(a) 1 DOF Revolute joint.



(b) 1 DOF Prismatic joint.



(c) 2 DOF Cylindrical joint



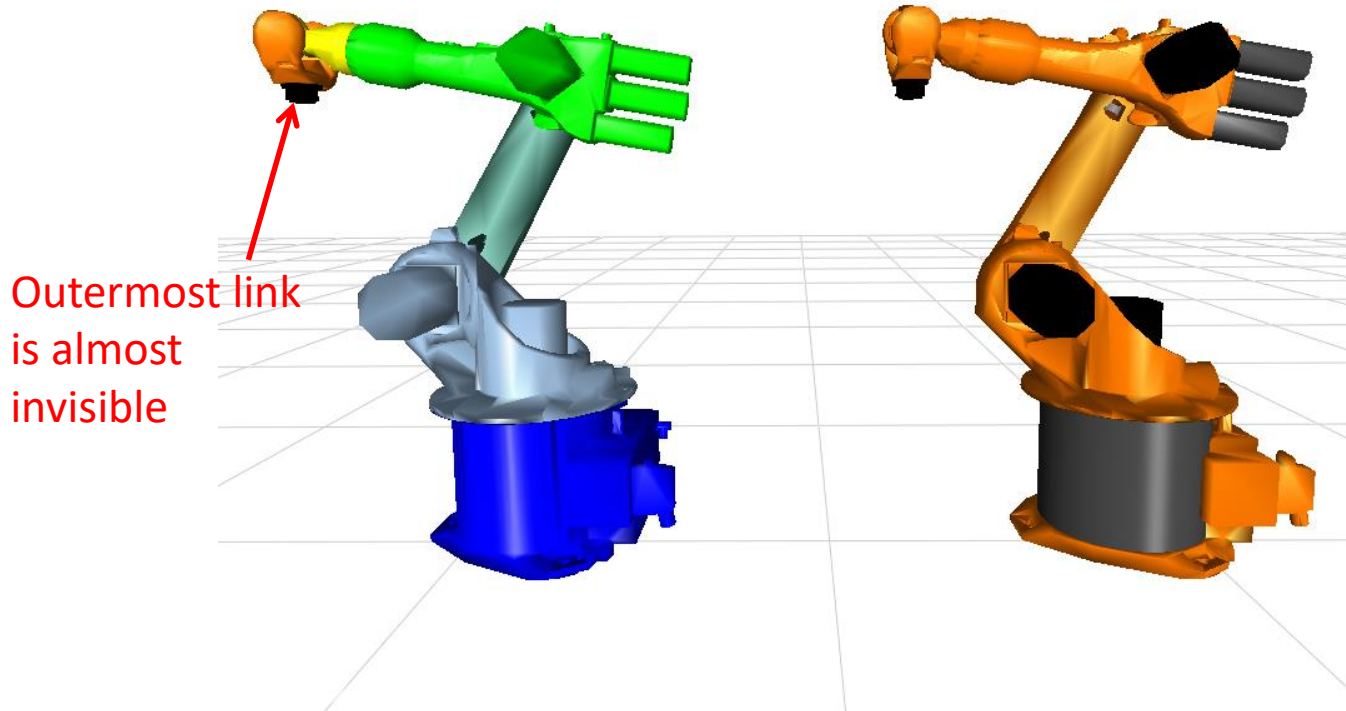
(d) 3 DOF Spherical joint

Figure 2.1: Joint types.



Robot links

Links marked with different colours

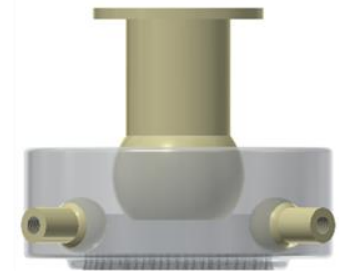
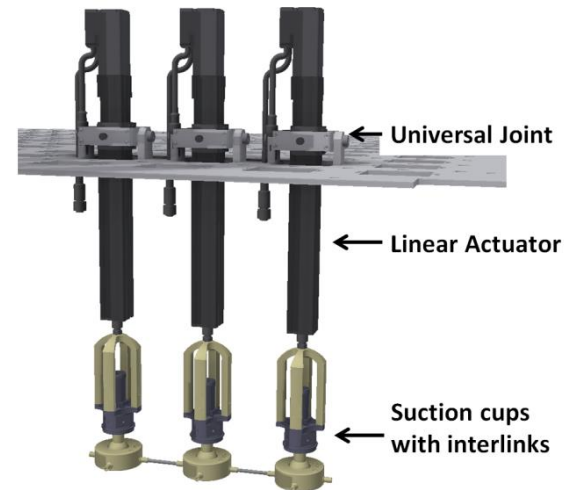


Constraints on Actuation and Topology

- The robot consists of a set of links and joints connected into a single structure
- Any two links are connected by at most one joint.
- Some, but not necessarily all, joints are motorized
- Each motorized joint can move independent of the other motorized joints.
- If all motorized joints are fixed at a given position, the robot will be kinematically stiff (typical case). Otherwise the robot is *underactuated*.

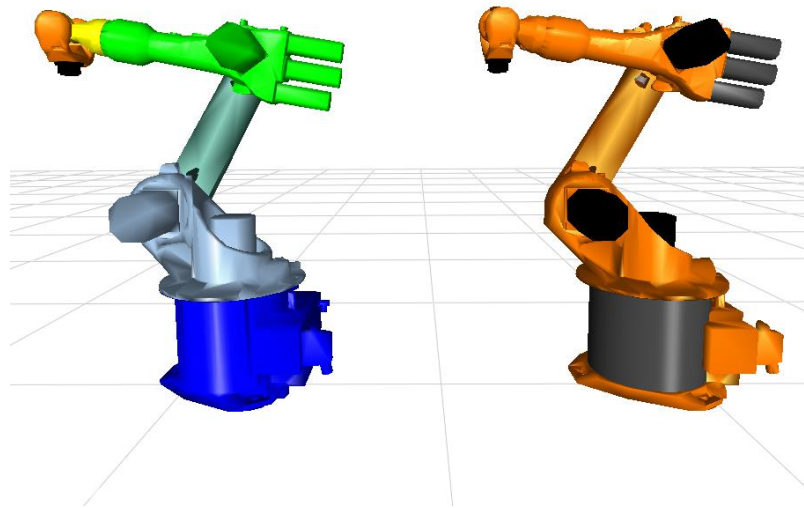


Underactuated Robot

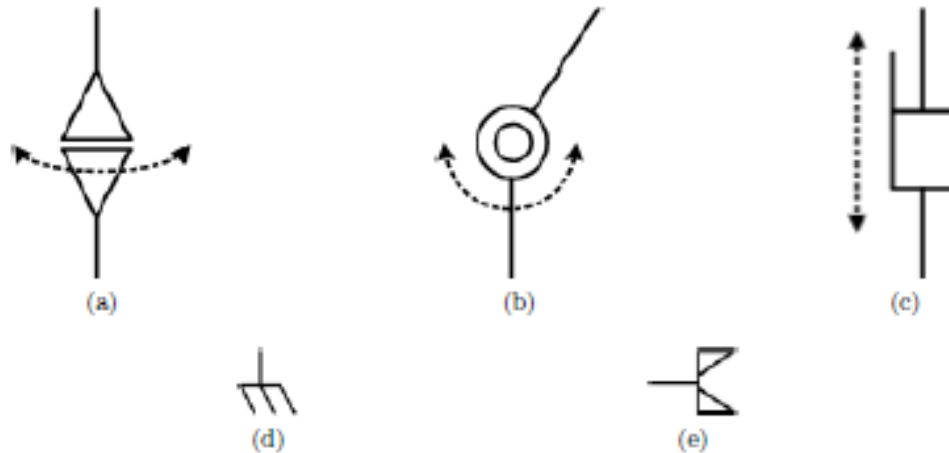


Kuka KR16

How do we describe the kinematics ?



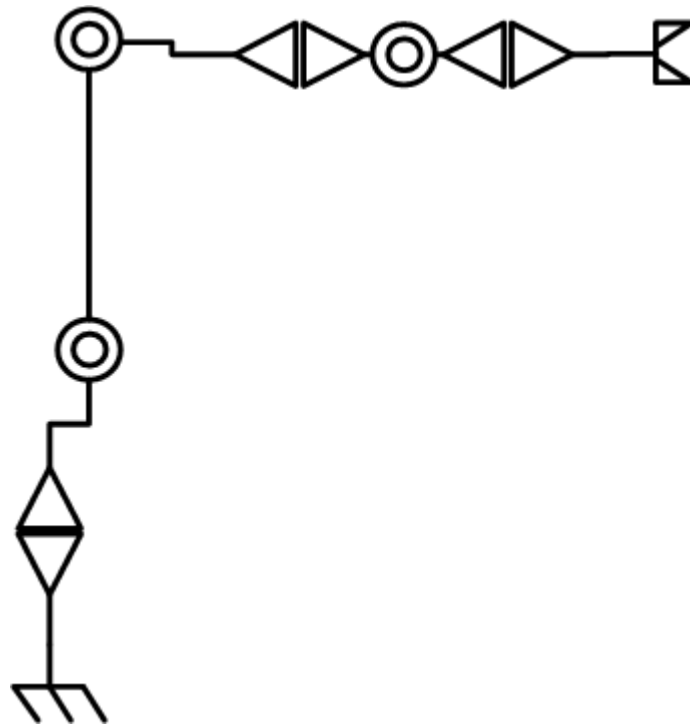
Joints, base and tool



- a): Revolute joint (vertical rotation axis)
- b): Revolute joint (rotation axis perpendicular to drawing plane)
- c): Prismatic (translational) joint
- d) Robot base
- e) Robot tool

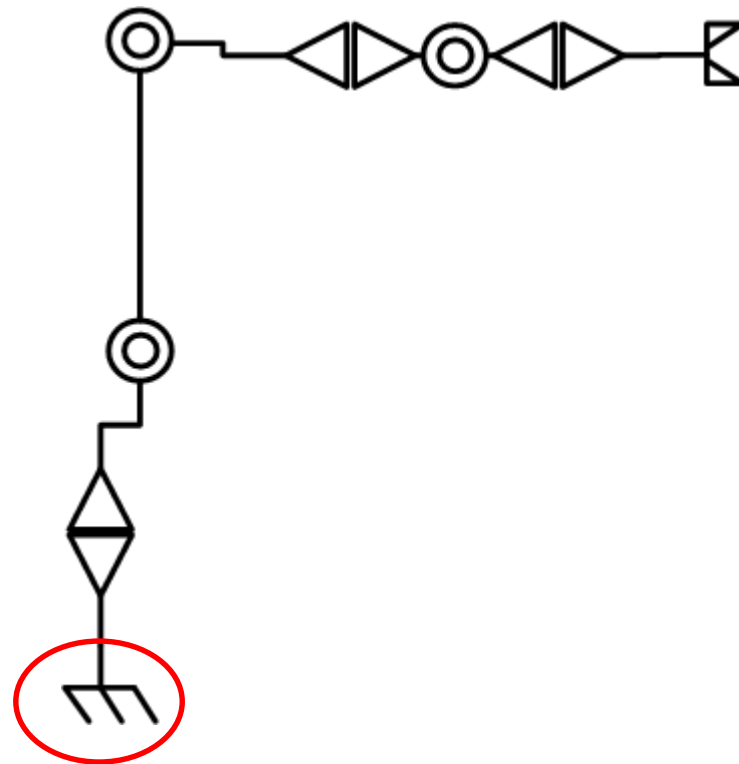


Schematic drawing of the Kuka KR16



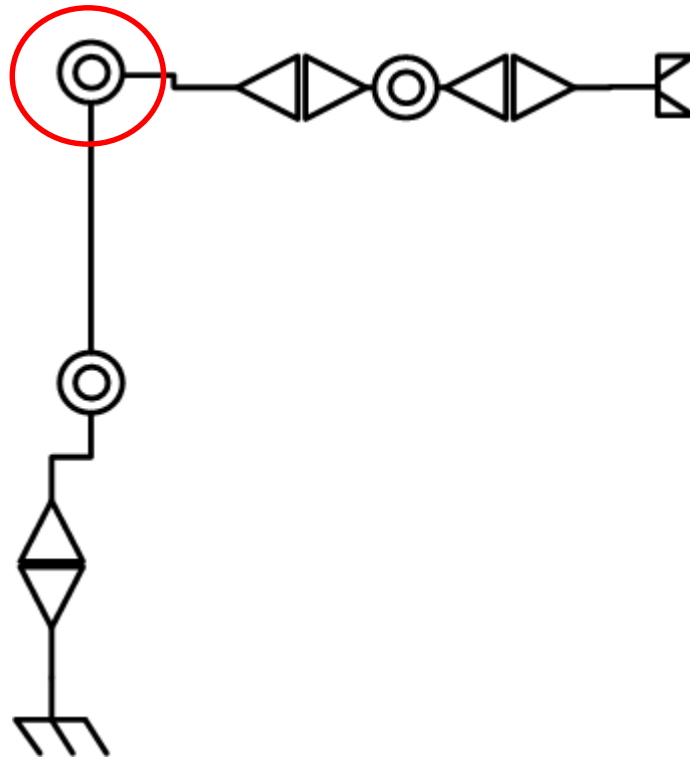
Schematic drawing of the Kuka KR16

(Robot) Base



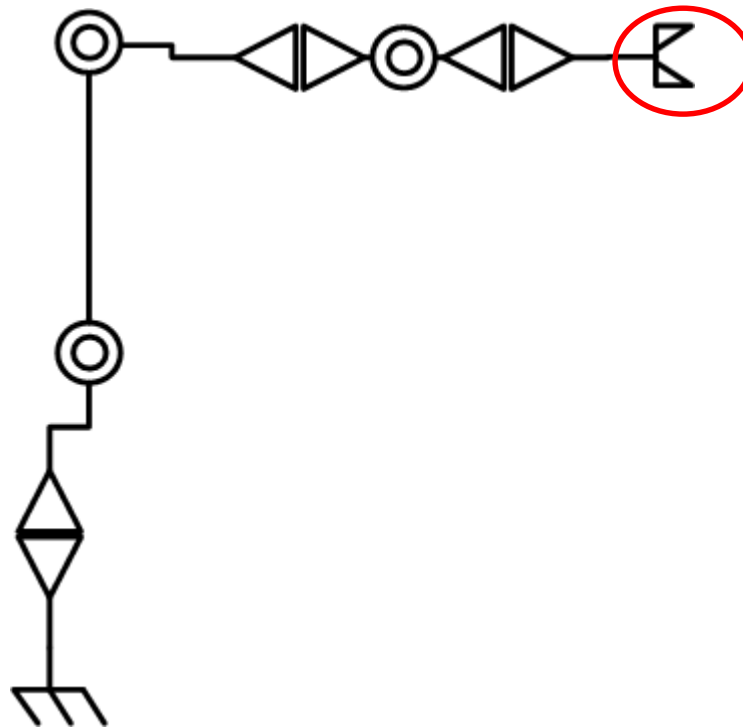
Schematic drawing of the Kuka KR16

Elbow joint (1)

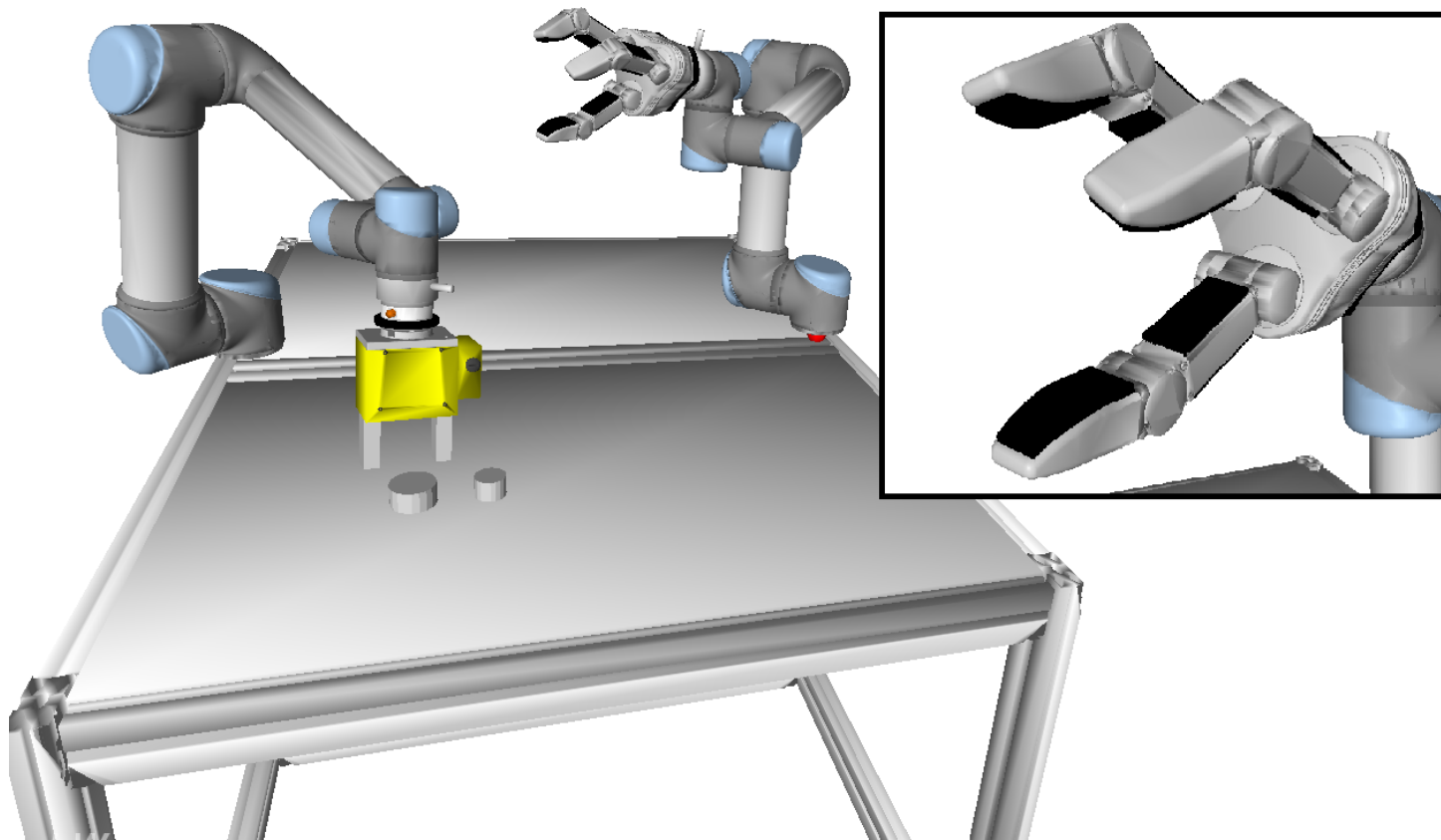


Schematic drawing of the Kuka KR16

Tool

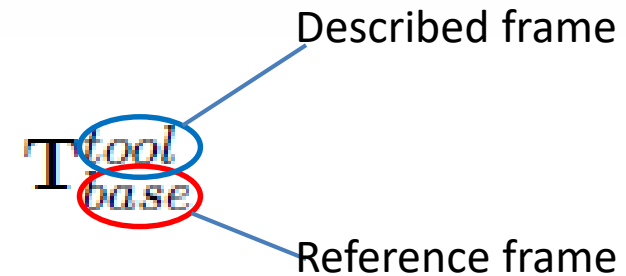
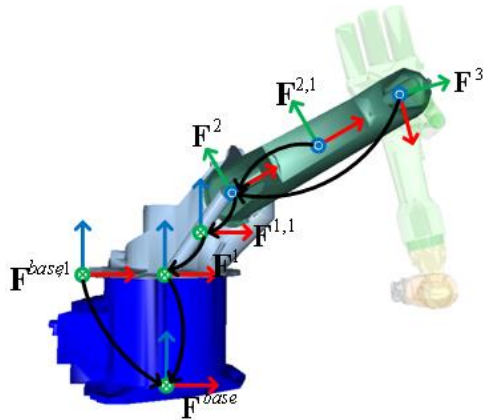


A two robot work cell



Homogeneous Transformation

$$\mathbf{T} = \begin{bmatrix} \mathbf{R} & \mathbf{p} \\ \mathbf{0}_{1 \times 3} & 1 \end{bmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} & x \\ r_{21} & r_{22} & r_{23} & y \\ r_{31} & r_{32} & r_{33} & z \\ 0 & 0 & 0 & 1 \end{bmatrix}$$



Joint vector q

- The *joint vector* is defined as the set of all joint variables of all robotic devices in a scene
- To simplify, we will from now on only consider a single serial robot in a scene.
- The joint vector uniquely determines the position and orientation of all robot links



Inverse Kinematics

Desired transformation $[\mathbf{T}_{base}^{tool}]_{desired}$

Find **a joint vector** \mathbf{q} so that $\mathbf{T}_{base}^{tool}(\mathbf{q}) = [\mathbf{T}_{base}^{tool}]_{desired}$

Nonlinear root finding problem.

Numerical approach – for any serial robot. Good initial guess required

Find **all joint vectors** \mathbf{q} so that $\mathbf{T}_{base}^{tool}(\mathbf{q}) = [\mathbf{T}_{base}^{tool}]_{desired}$

Analytical approach – for serial robots with special (idealized) kinematics.



The previous 23 slides has illustrated why an industrial robot manipulator arm is just a stupid machine that is easy to use as any other tool...

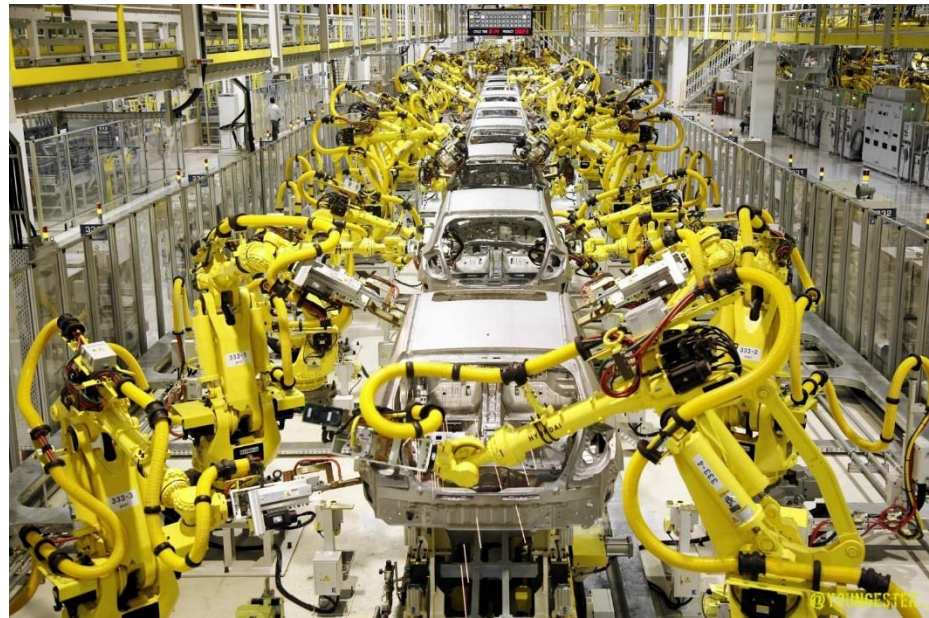
Challenges in industrial robotics are related to describing and optimizing the process

- Modelling the process
- Use the model to derive a desired robot tool path
- Complete the solution by inverse kinematics



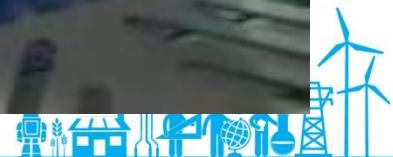
Traditional robot automation

Complex, but repetitive



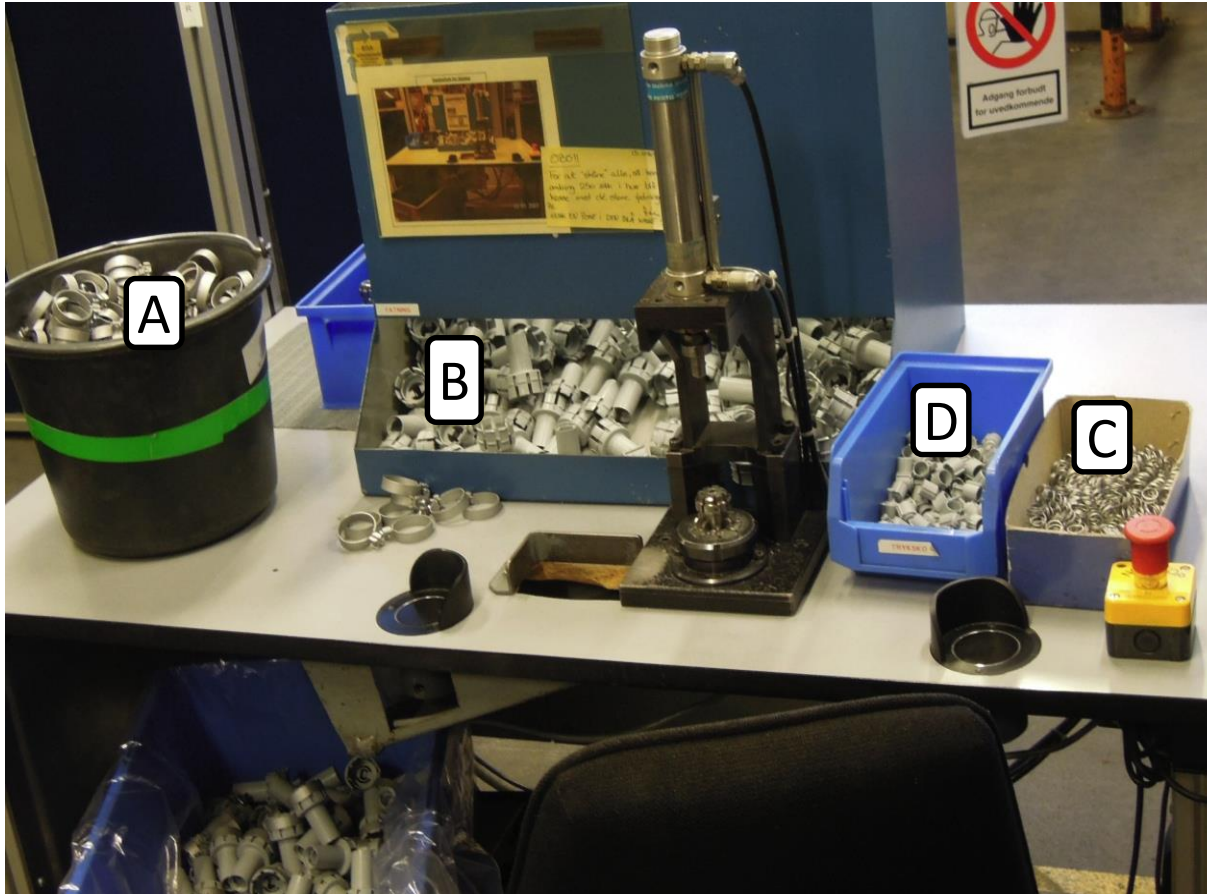
Traditional robot automation

Non-repetitive, but simple



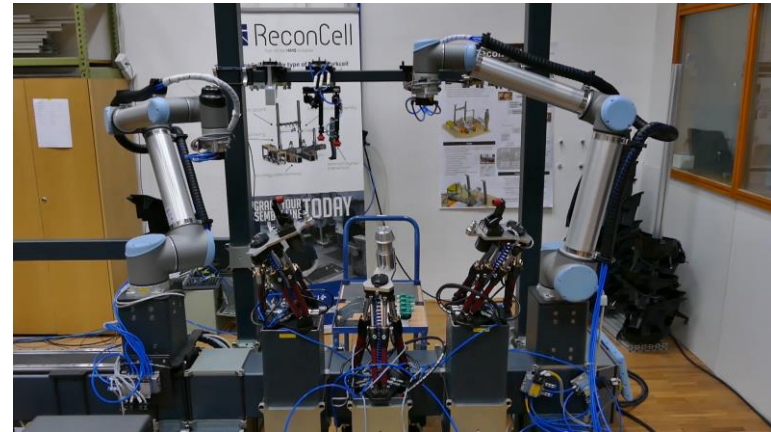
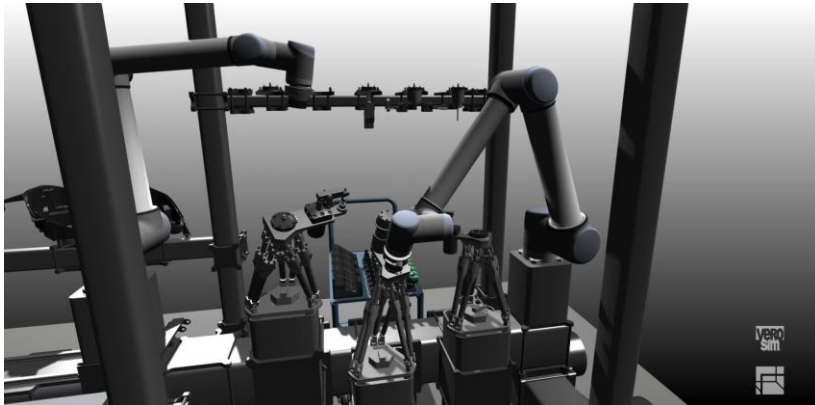
Trend in robotic automation

Non-repetitive and (somewhat) complex



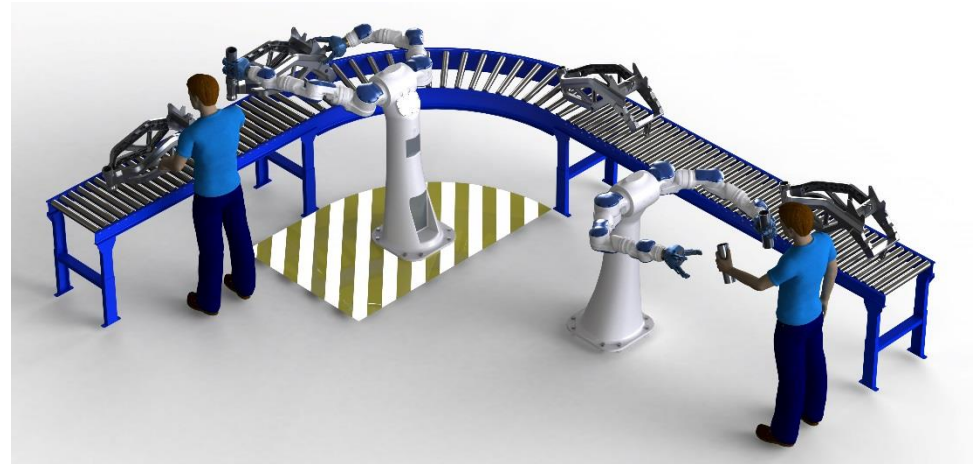
Programming and testing in virtual environments

Example from one of our European projects:
Headlight assembly for automotive

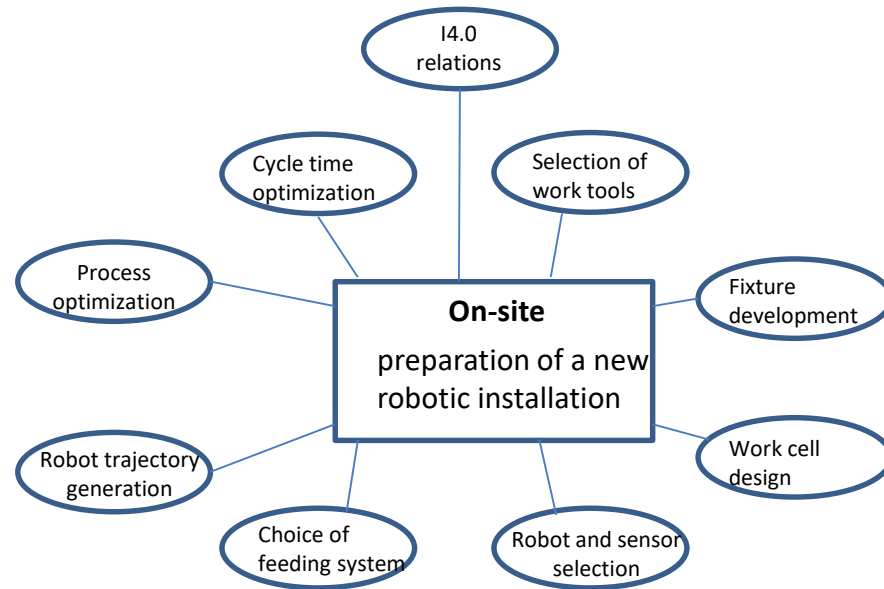


Trend in robotic automation

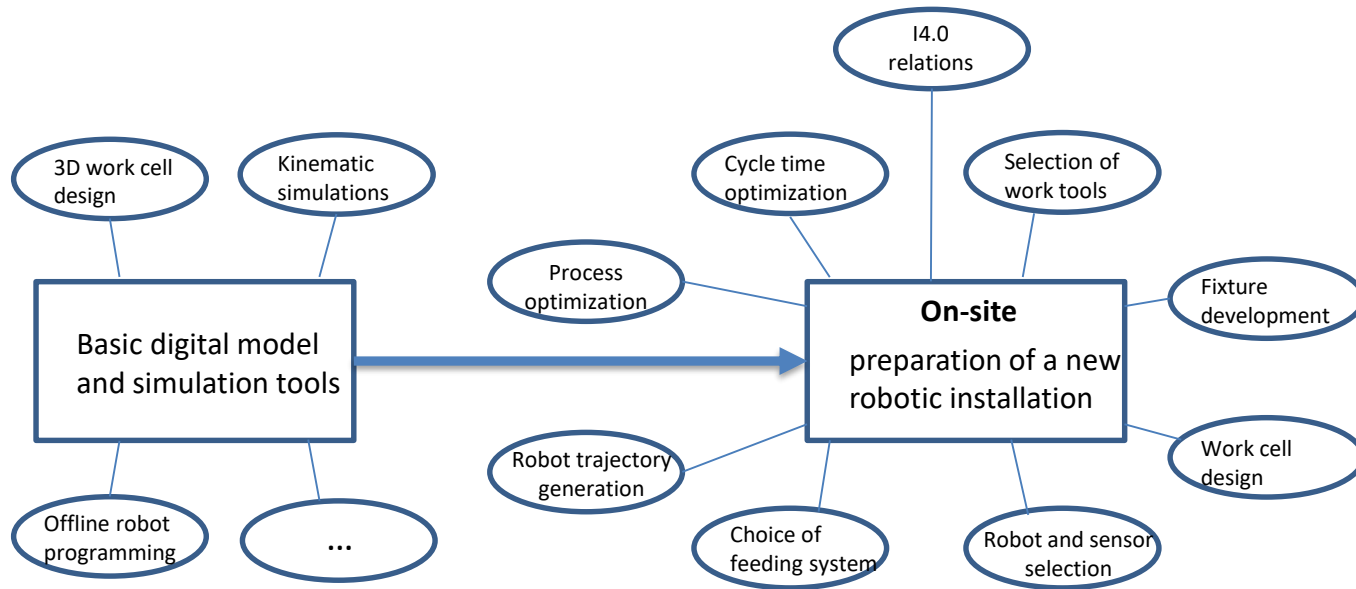
Collaborative robots



Traditional programming of robot systems



Using just simple digital tools



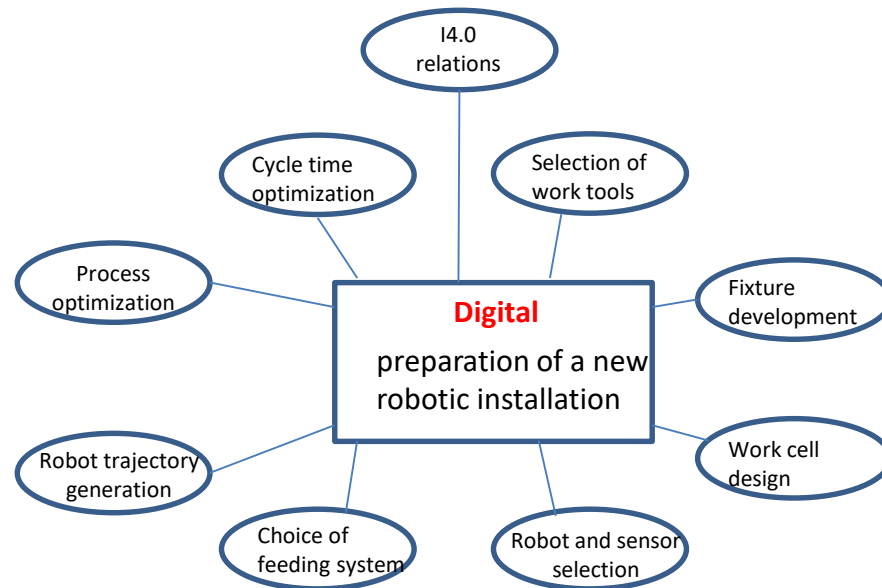
Problem: The robot systems integrator is often unable to offer even these basic digital tools

Problem: 3D CAD models of all relevant parts in the production should be available at the enduser

Problem: Unawareness of the advantages



Where we should be !!!



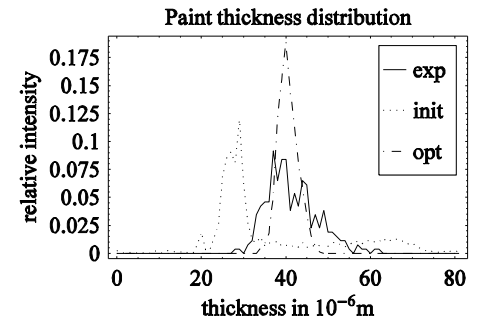
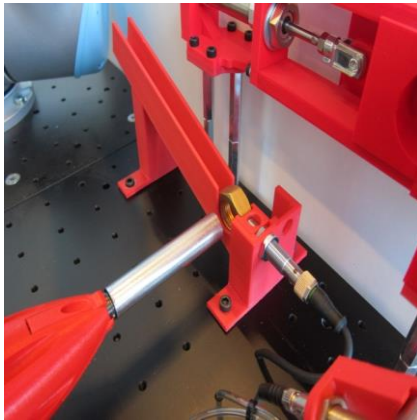
Three examples of digital modeling and simulation:

Choice of feeding system



Robot trajectory generation

Process optimization



Example 1: Bowl feeder design

Problem:

Part specific bowl feeders are implemented based on a trial and error procedure

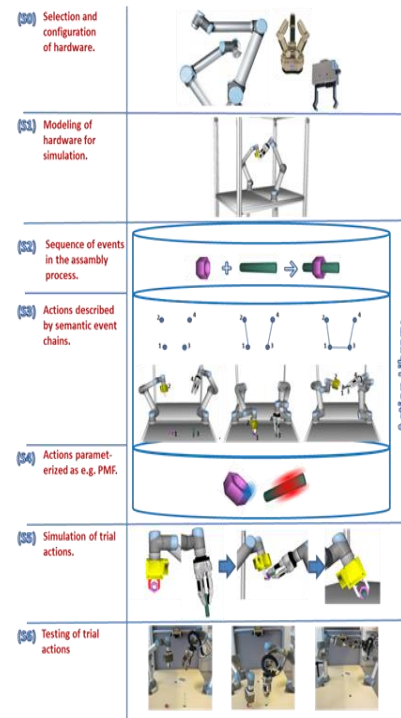
Requires very high expertise

Expensive bottleneck

Solution:

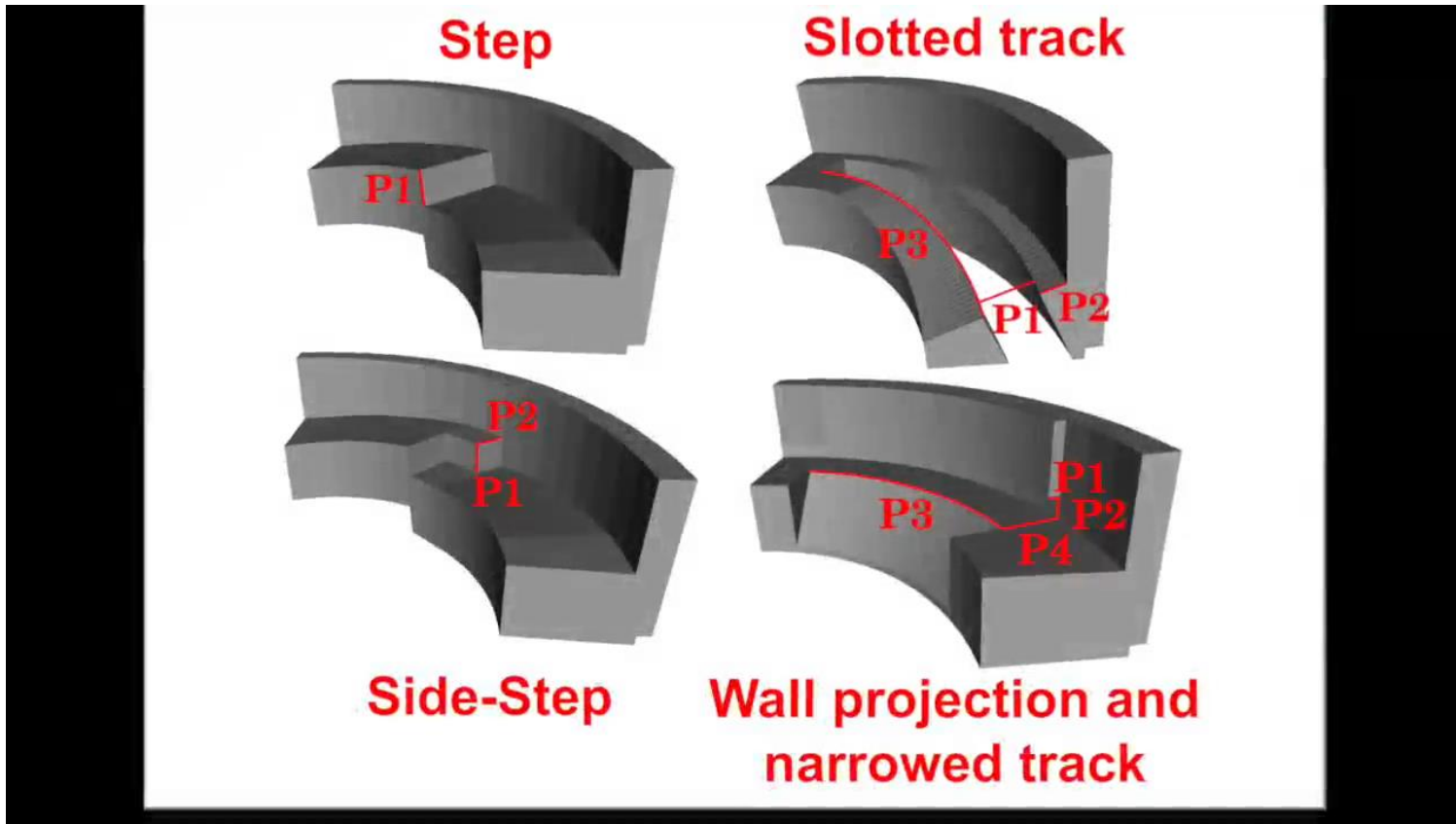
Use computer simulations as a replacement for the trial and error procedure

Reduces price and paves the way for the option of enduser designing these themselves



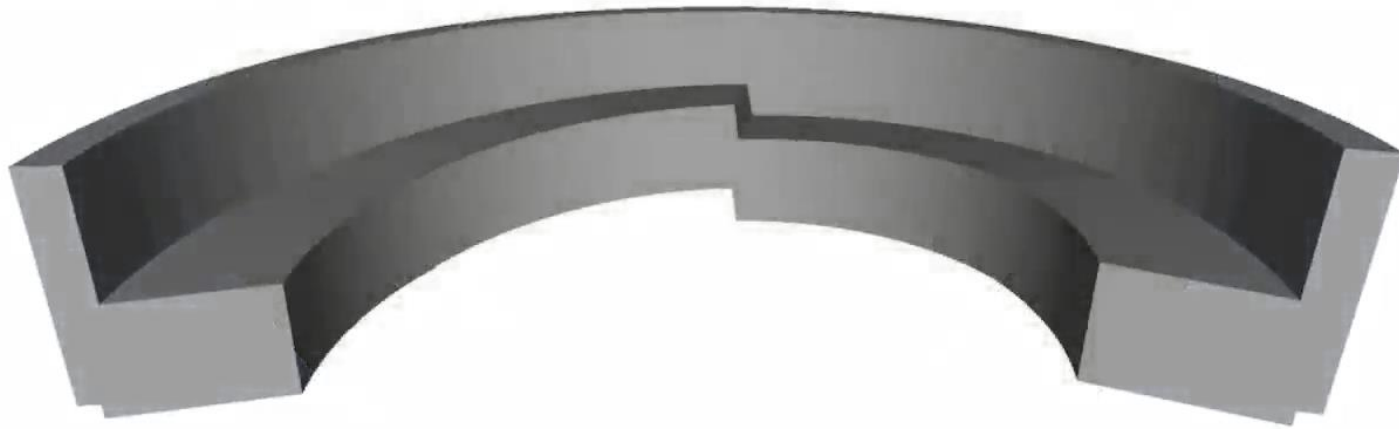
Choice of feeding system

Precise dynamic simulations of vibrational feeders for virtual design and optimization of traps and trap sequences



Choice of
feeding system

Precise dynamic simulations of vibrational feeders for virtual design
and optimization of traps and trap sequences



Choice of
feeding system

Precise dynamic simulations of vibrational feeders for virtual design
and optimization of traps and trap sequences



Example 2: Robot Spray Painting

Problem:

Double-curved surface to be painted (one of a kind)

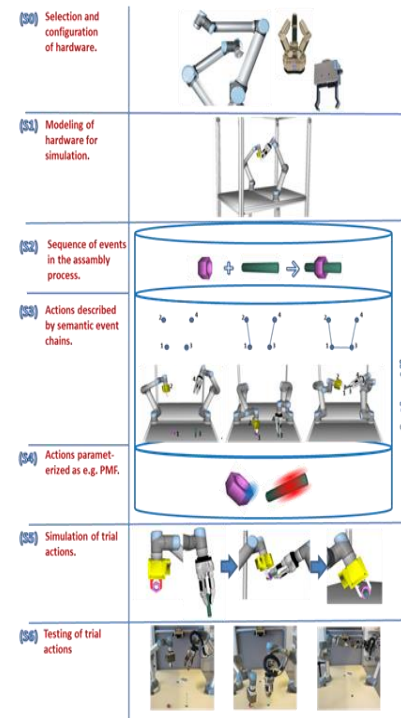
Desired thickness given

Automatical computation of spray nozzle trajectory

Solution:

Calibrate paint flux in spray nozzle

Compute nozzle trajectory by simulation based optimization



Spray cone flux model

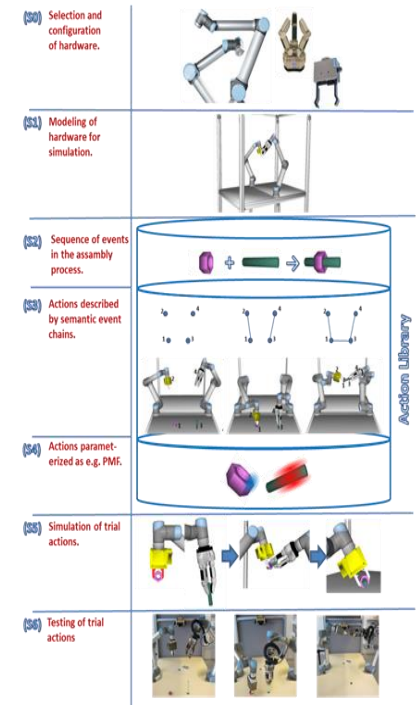
The streamlines radiate from the tool center, which ensures that the paint flux is contained within a cone whose silhouettes are straight lines,

If the cone is intersected by a plane perpendicular to the cone axis, the curves of constant flux in this plane are ellipses,

The divergence of \mathbf{j} must be zero, which ensures that the total flux across any surface completely intersecting the cone is a constant, expressing conservation of paint.

$$\mathbf{j}(\mathbf{r}, \mathbf{u}_1 \cdots \mathbf{u}_3) = \alpha Q ((\mathbf{r} \cdot \mathbf{u}_3) / (\mathbf{r}^T \mathbf{A} \mathbf{r})^{1/2}) \mathbf{r} / (\mathbf{r}^T \mathbf{A} \mathbf{r})^{3/2}$$

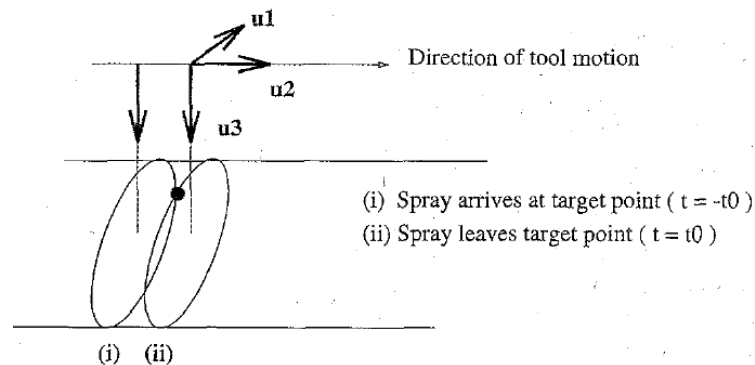
$$\mathbf{A} = \mathbf{u}_1(t) \mathbf{u}_1(t)^T + \alpha^2 \mathbf{u}_2(t) \mathbf{u}_2(t)^T + \mathbf{u}_3(t) \mathbf{u}_3(t)^T$$



Finding the flux parameters of the nozzle by calibration

- Paint flat surface
- Constant distance and speed
- Spray gun perpendicular to surface

$$c(u, v) dA = - \int_{-\infty}^{\infty} \mathbf{j}(\mathbf{r}(u, v) - \mathbf{r}_{TC}(t), \mathbf{u}_1(t), \mathbf{u}_2(t), \mathbf{u}_3(t)) \cdot (\mathbf{r}_u(u, v) \times \mathbf{r}_v(u, v)) du dv dt$$



After several shifts of variables and some tedious algebra...

$$C(w) = \int_0^w \frac{q(u)}{\sqrt{w-u}} du.$$

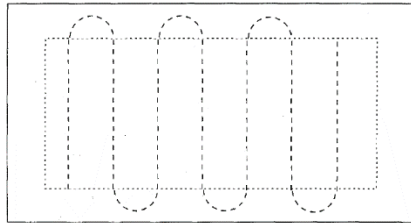
Analytical solution apparently first found by N.H. Abel in 1823:

$$q(u) = \frac{1}{\pi} \int_0^u \frac{C'(\eta) d\eta}{\sqrt{u-\eta}}.$$

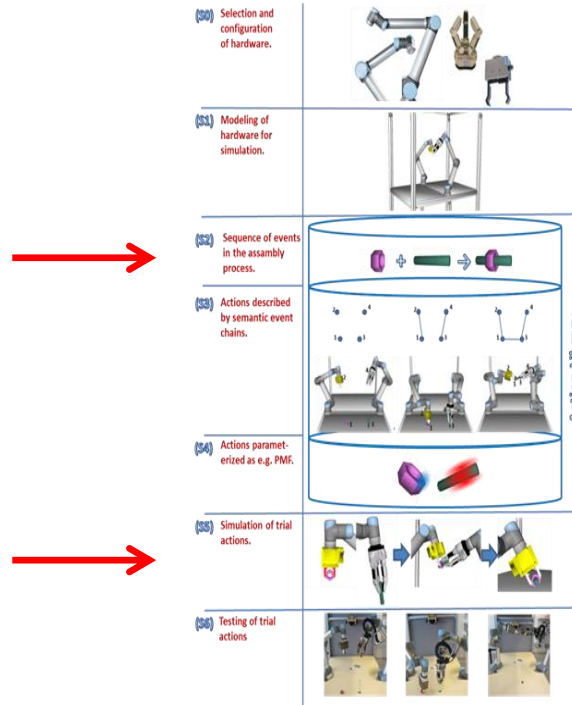


Optimizing the nozzle trajectory

Initial guess



Optimization by simulation (guess trajectory and compute coverage thickness)

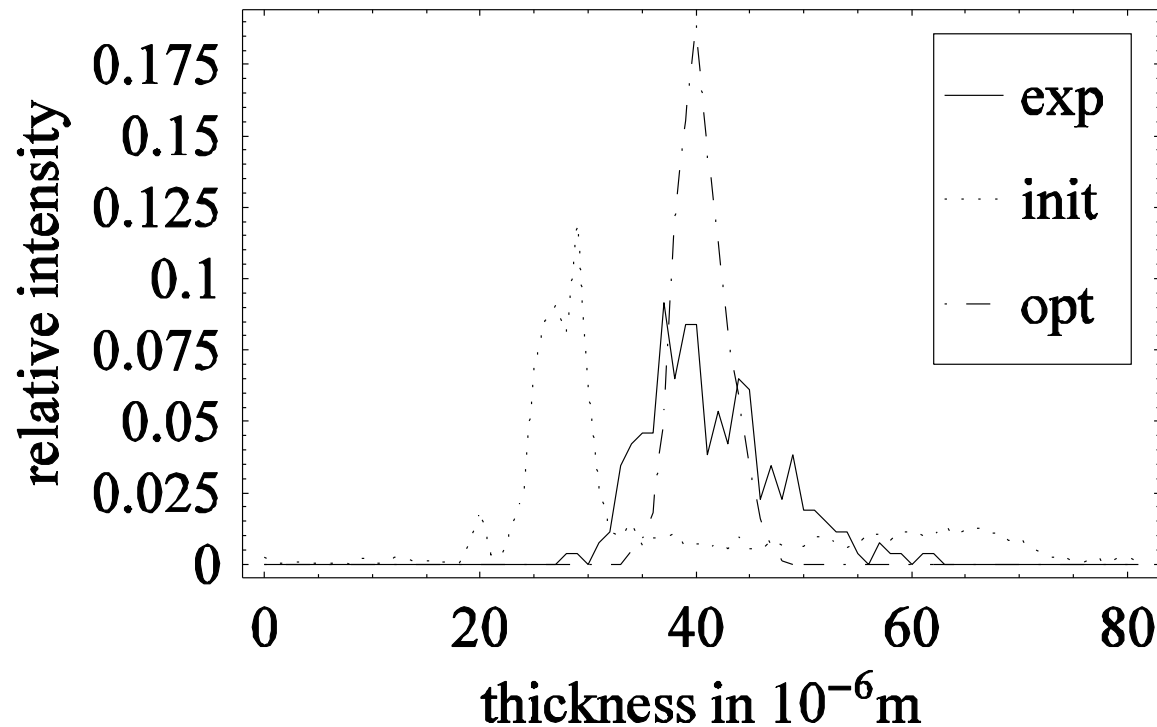


Painting a wheelbarrow



Simulation and experimental results

Paint thickness distribution



Example 3: Dynamic simulation models for robust robot assembly



Example 3: Dynamic simulation models for robust robot assembly

Problem:

Optimizing assembly operations on site can be difficult because errors are rare.

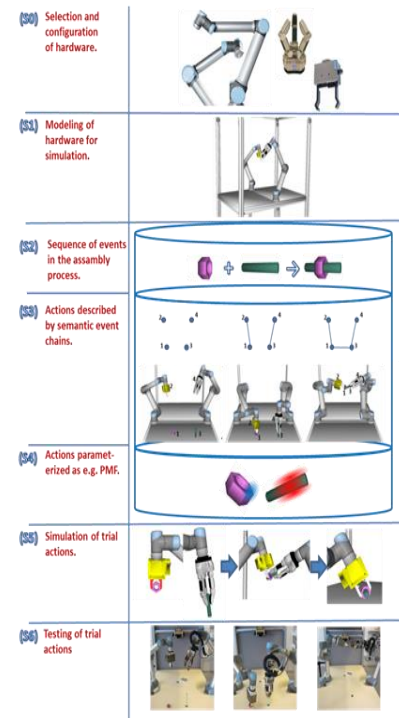
Hence the run-in time is long

Process is poorly understood

Solution:

Use computer simulations as a supplement for the trial and error procedure

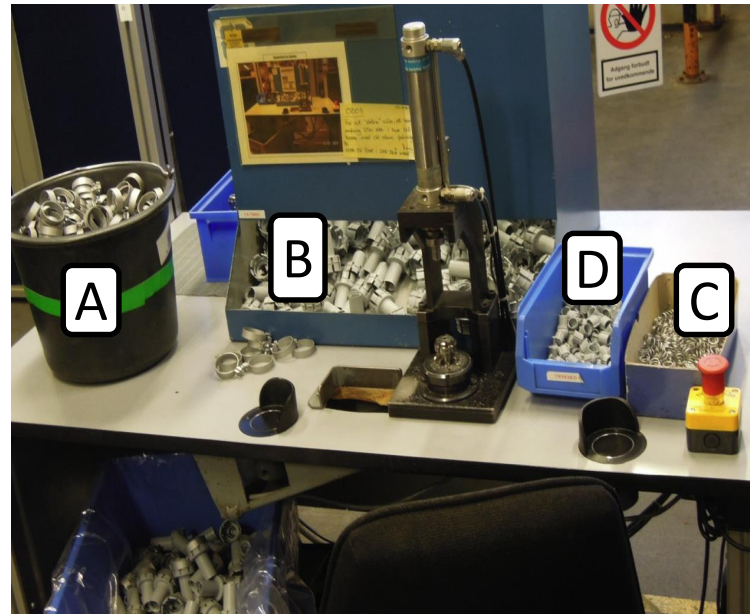
Requires a process model and hence forces process understanding



Usage: Simple movable, reconfigurable, adaptable platforms



Movable and reconfigurable



Adaptable to randomly located parts

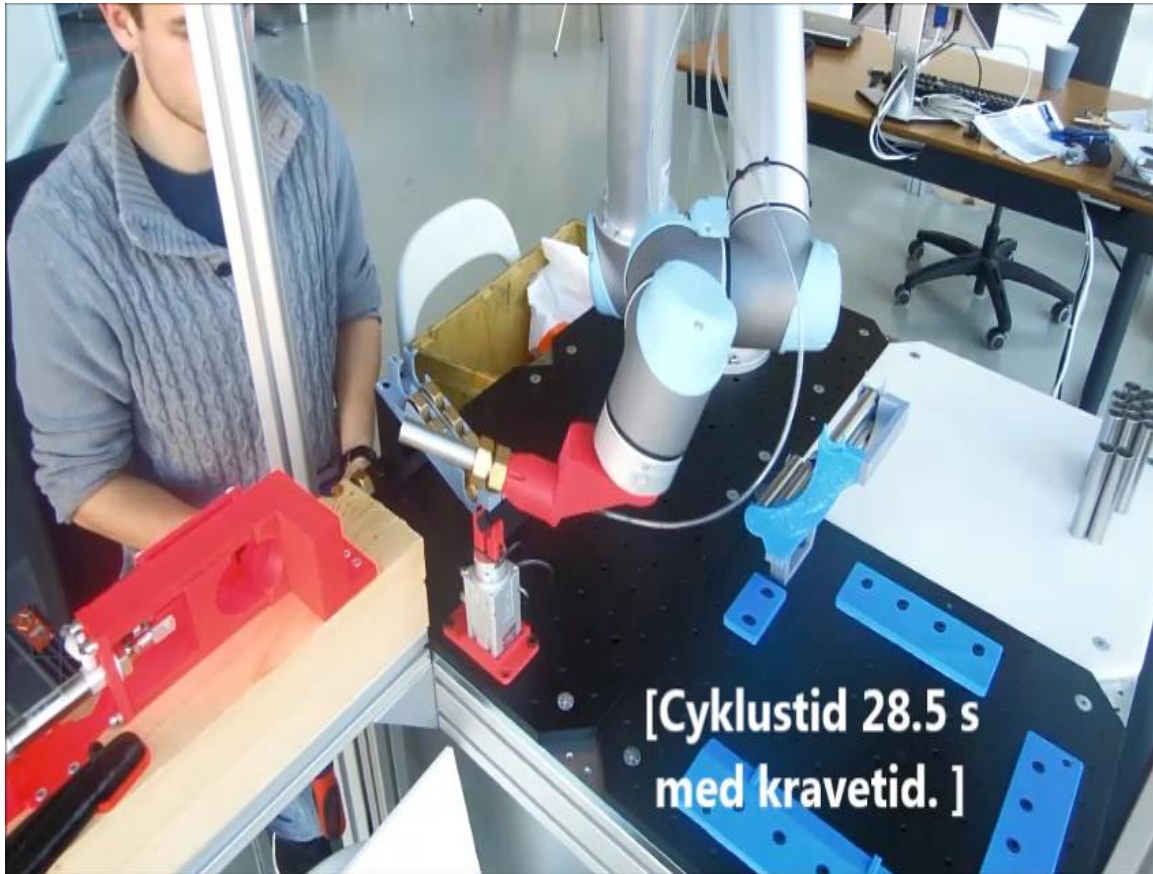


An example of a simple task

- Posed by the company KVM-Conheat
- Task: Place the two union nuts on the pipe as shown
- Several steps in the task, but main here is on getting the nuts onto the pipe
- Pipe and nut feeders used



Conventional teach-in programming of the task



Professor's negation field:

GET TO KNOW:
**YOUR ADVISOR'S
NEGATION FIELD**

A Professor's Negation Field is the unexplained phenomenon whereby mere spatial proximity to an experimental set-up causes all working demonstrations to fail, despite the apparent laws of Physics or how many times it worked right before he/she walked into the room.

BEWARE
ALSO:

The Sphere of Death.
Allowing your experiment
within arm's reach of your
Advisor risks the possibility
of immediate destruction.



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Professor's negation field:

Lack of **uncertainty handling**

GET TO KNOW: YOUR ADVISOR'S NEGATION FIELD

A Professor's Negation Field is the unexplained phenomenon whereby mere spatial proximity to an experimental set-up causes all working demonstrations to fail, despite the apparent laws of Physics or how many times it worked right before he/she walked into the room.

BEWARE ALSO:

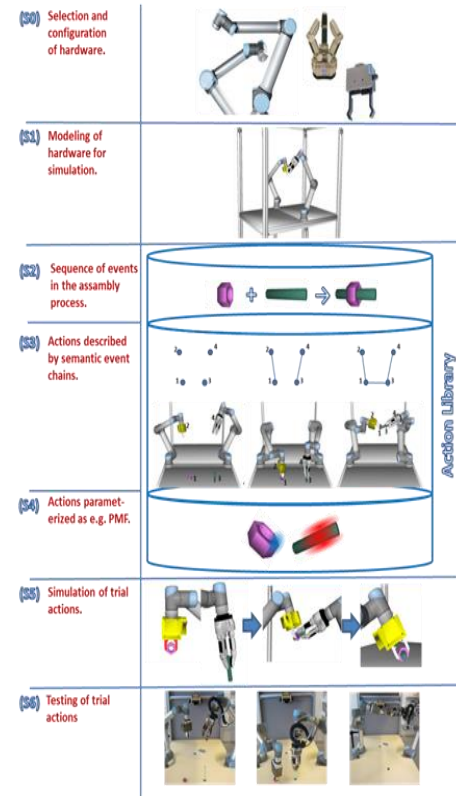
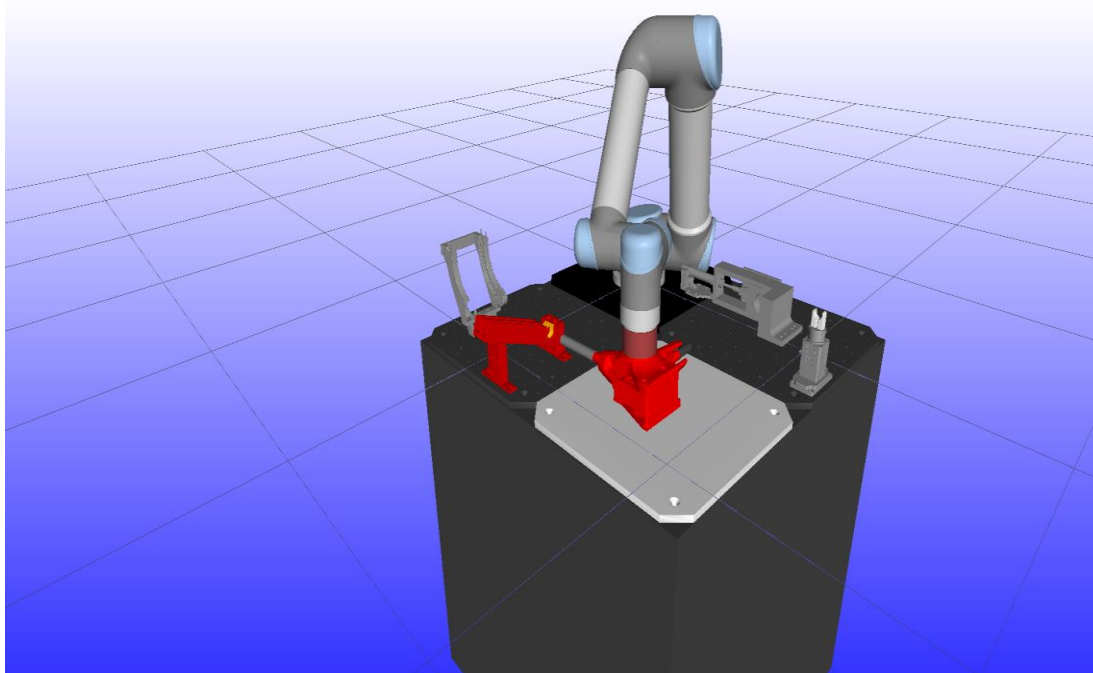
The Sphere of Death. Allowing your experiment within arm's reach of your Advisor risks the possibility of immediate destruction.



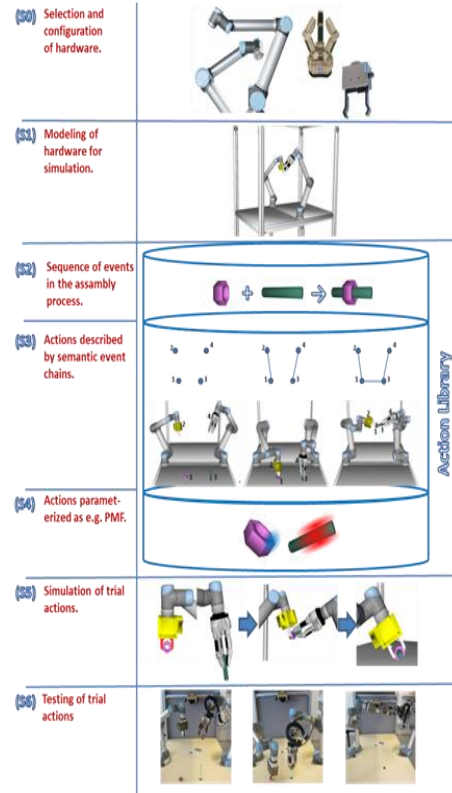
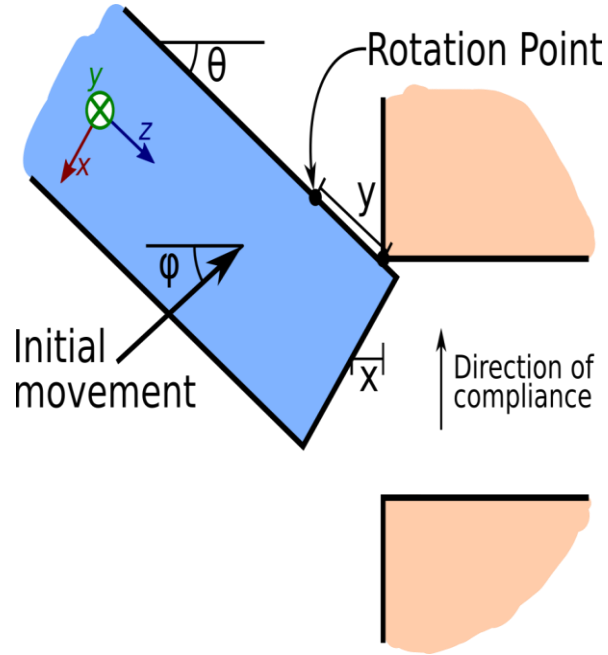
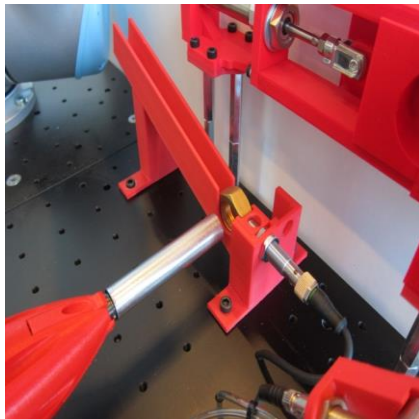
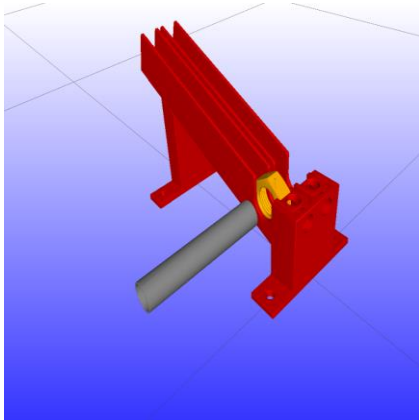
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Modeling of hardware for simulation



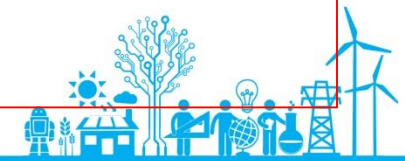
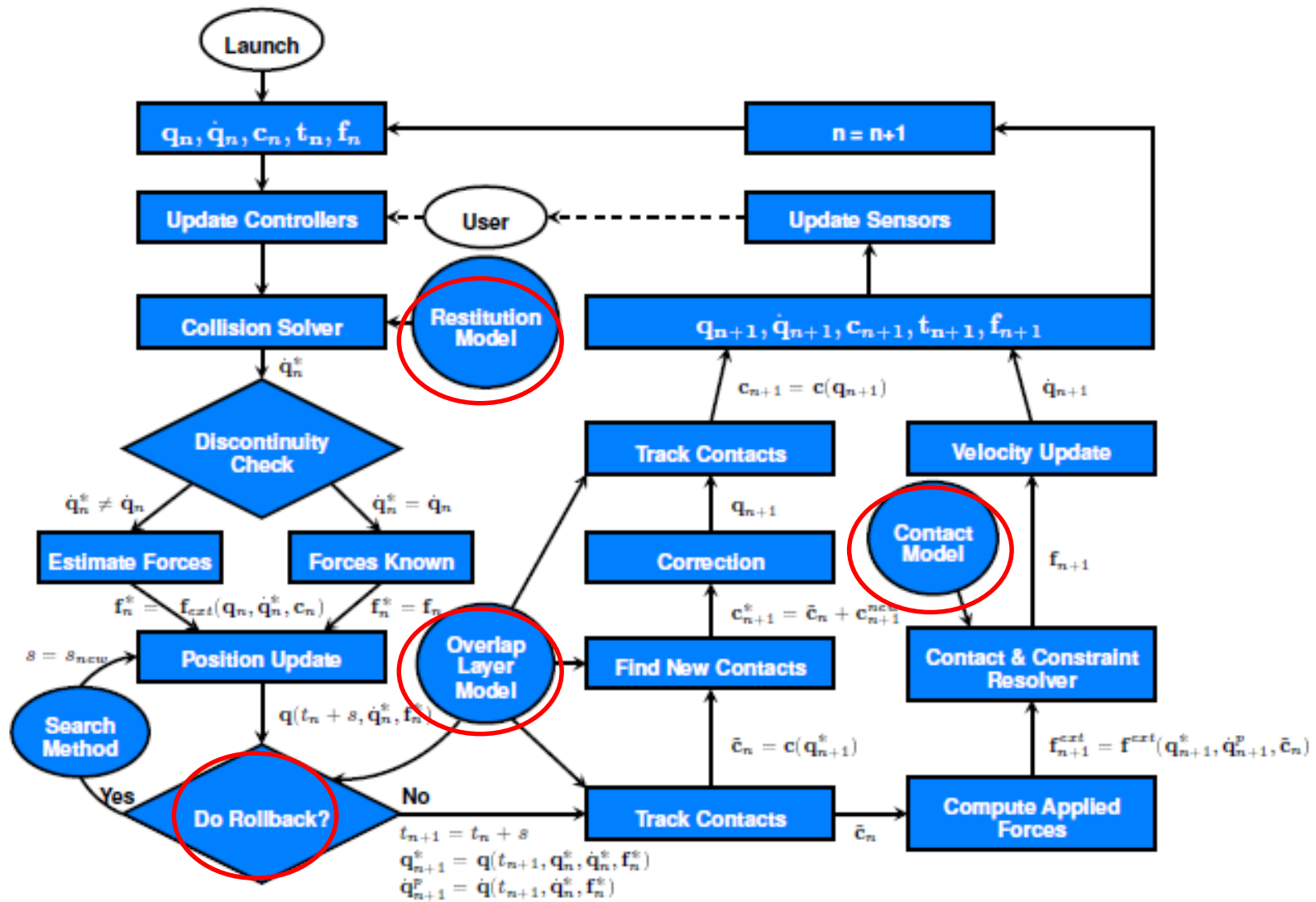
Action parametrization



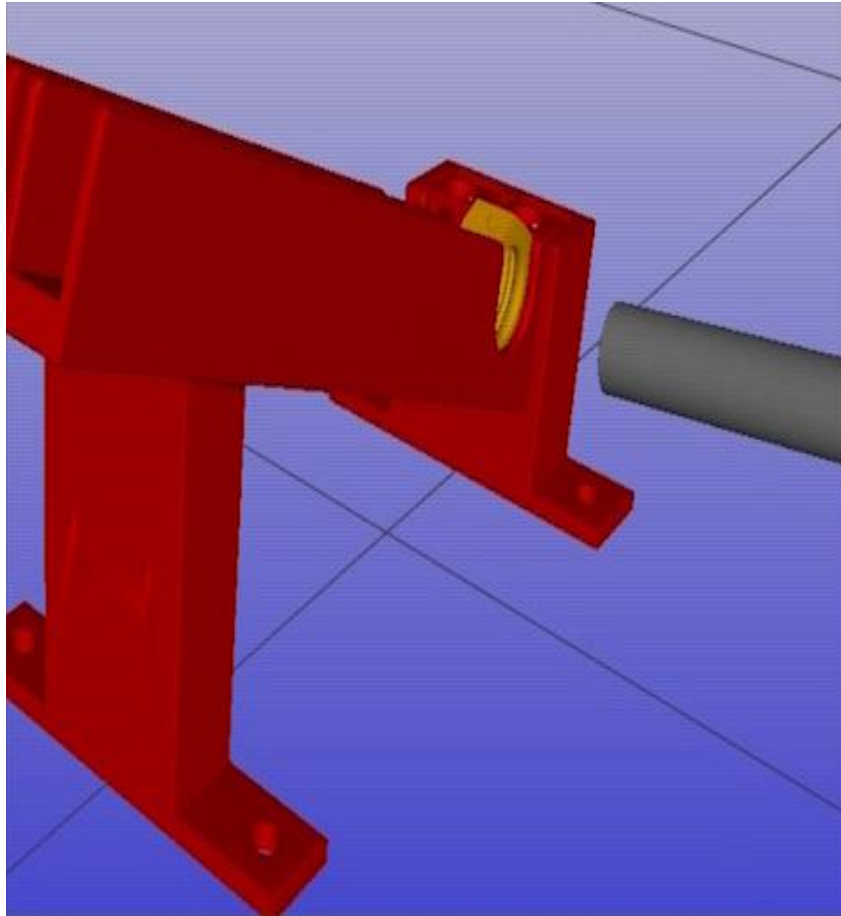
Dynamic simulations

- Newton/Euler rigid body equations of motion
- Bodies modeled as rigid
- Contact models (friction, restitution)
- Existing engines too inaccurate (aimed for computer games)

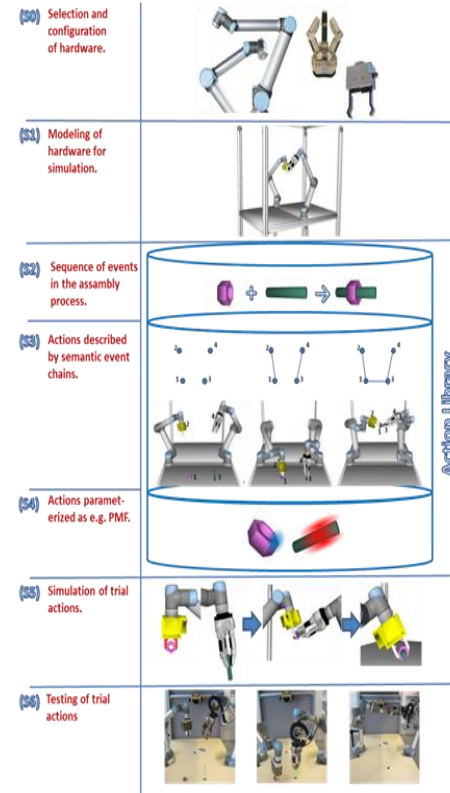




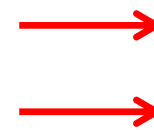
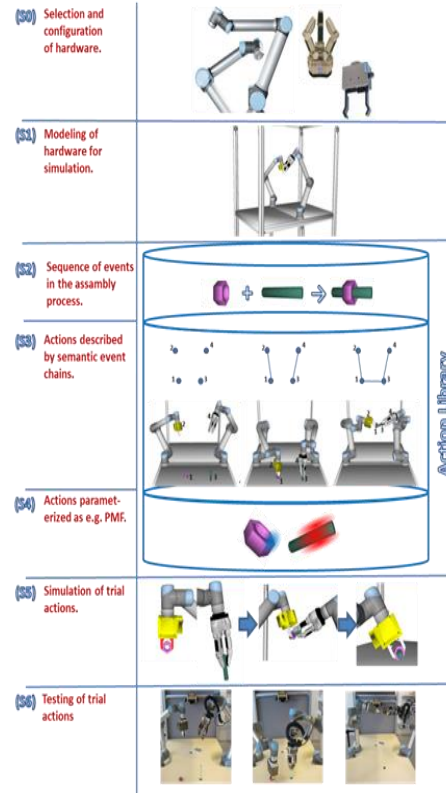
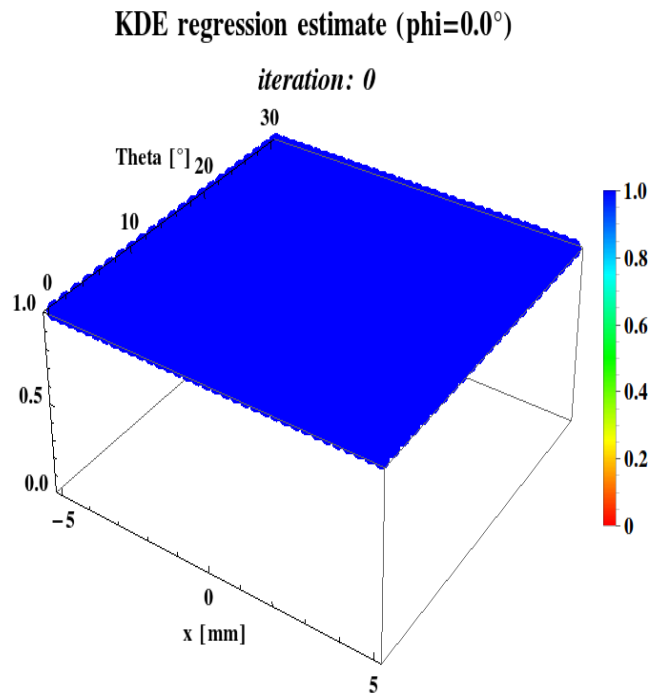
Simulation of trial actions:



A chosen set of action parameters
A randomly chosen pose perturbation



Learning promising action parameters:



$$\hat{\mu}_H(x) = \frac{\hat{p}_O(x, s)}{\hat{p}_X(x)} = \frac{n^{-1} \sum_{i=1}^n K_{H, x_i}(x) O_i}{n^{-1} \sum_{j=1}^n K_{H, x_j}(x)}$$

$$CI_{kde} = z \cdot \sqrt{\frac{\|K\|_2^2 \hat{\sigma}_H^2(x)}{n |H| \hat{f}_H(x)}}$$



Learned solution

