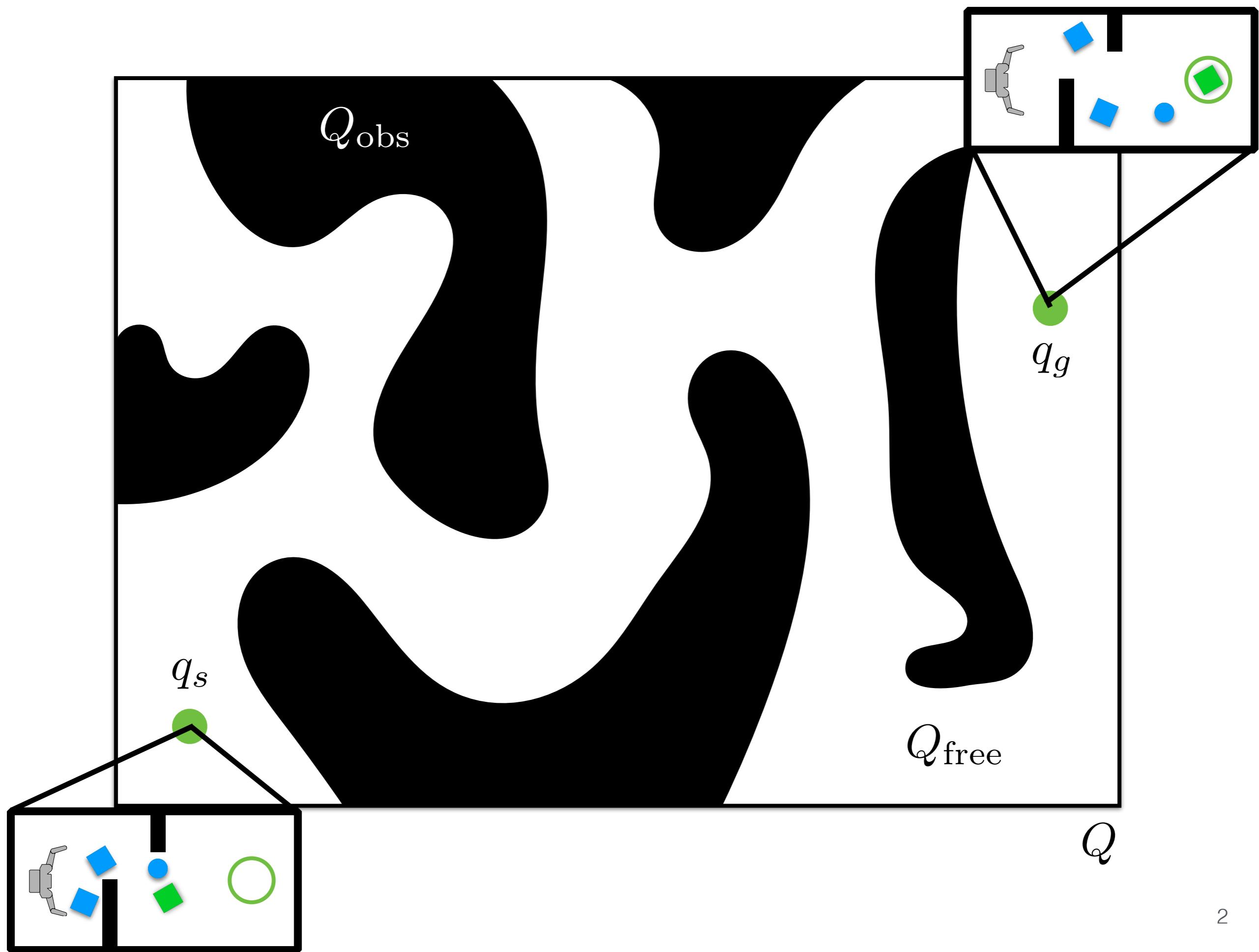


Physics-Based Manipulation under Uncertainty

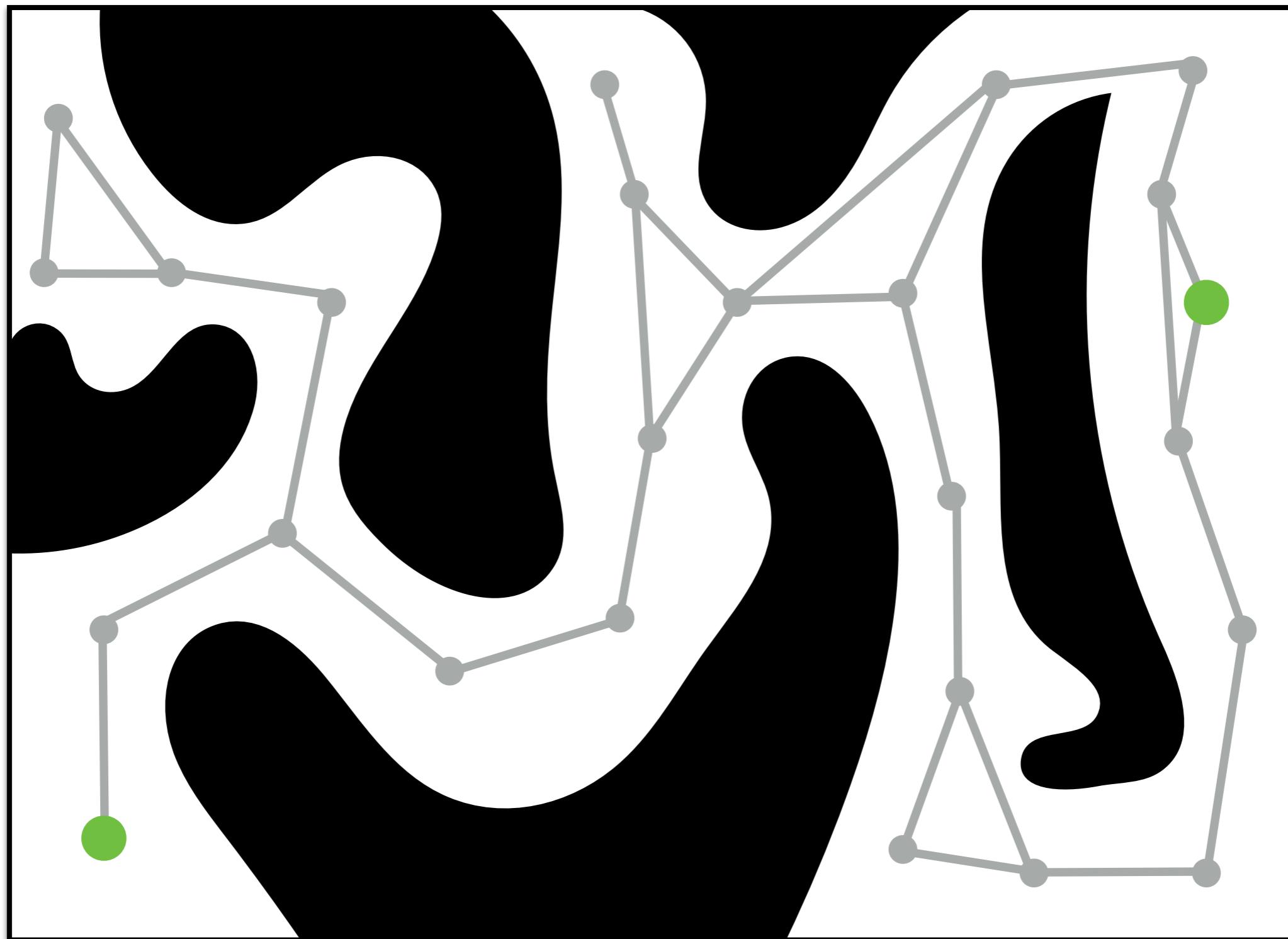


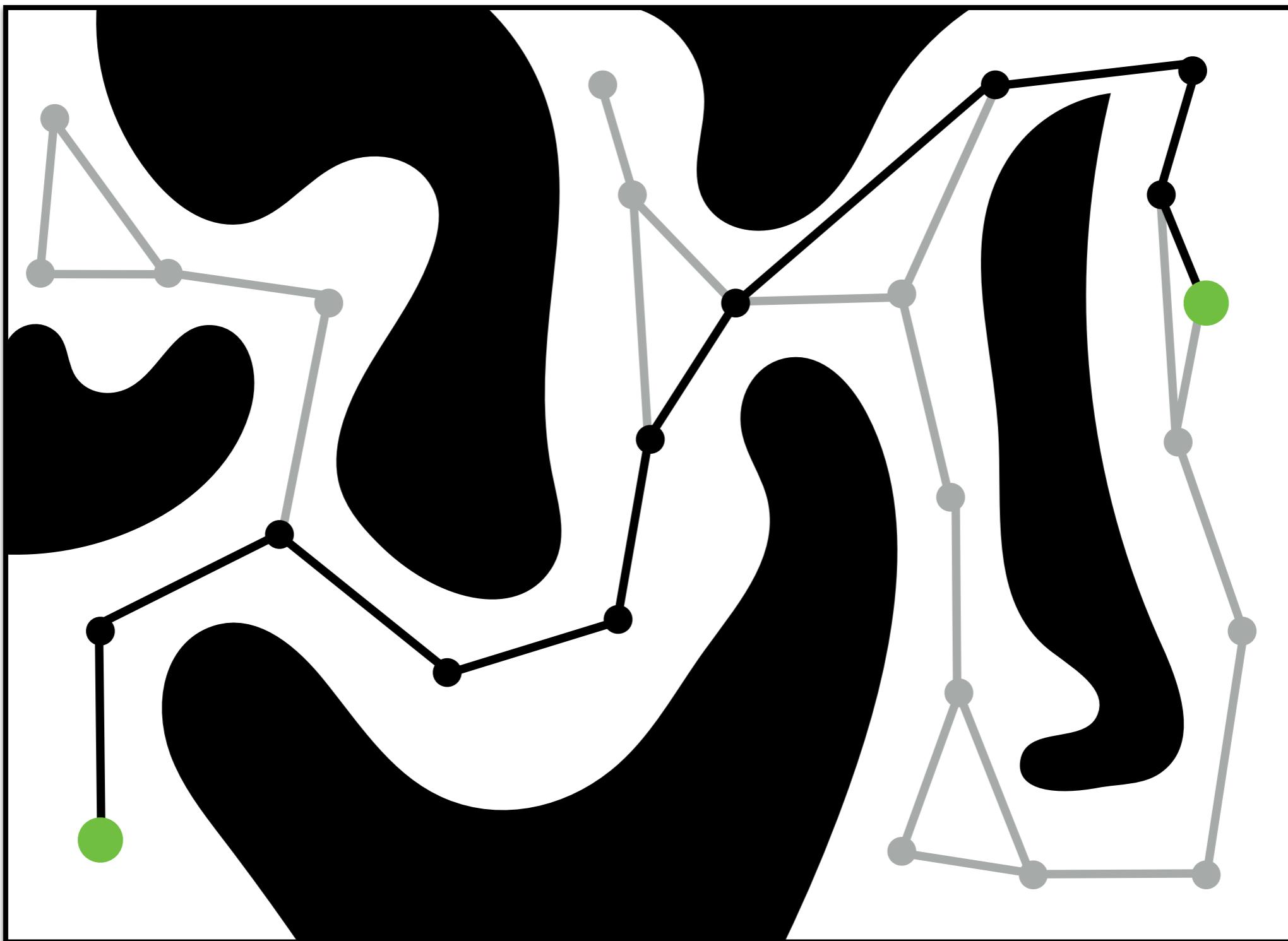
Michael Koval
mkoval@cs.cmu.edu

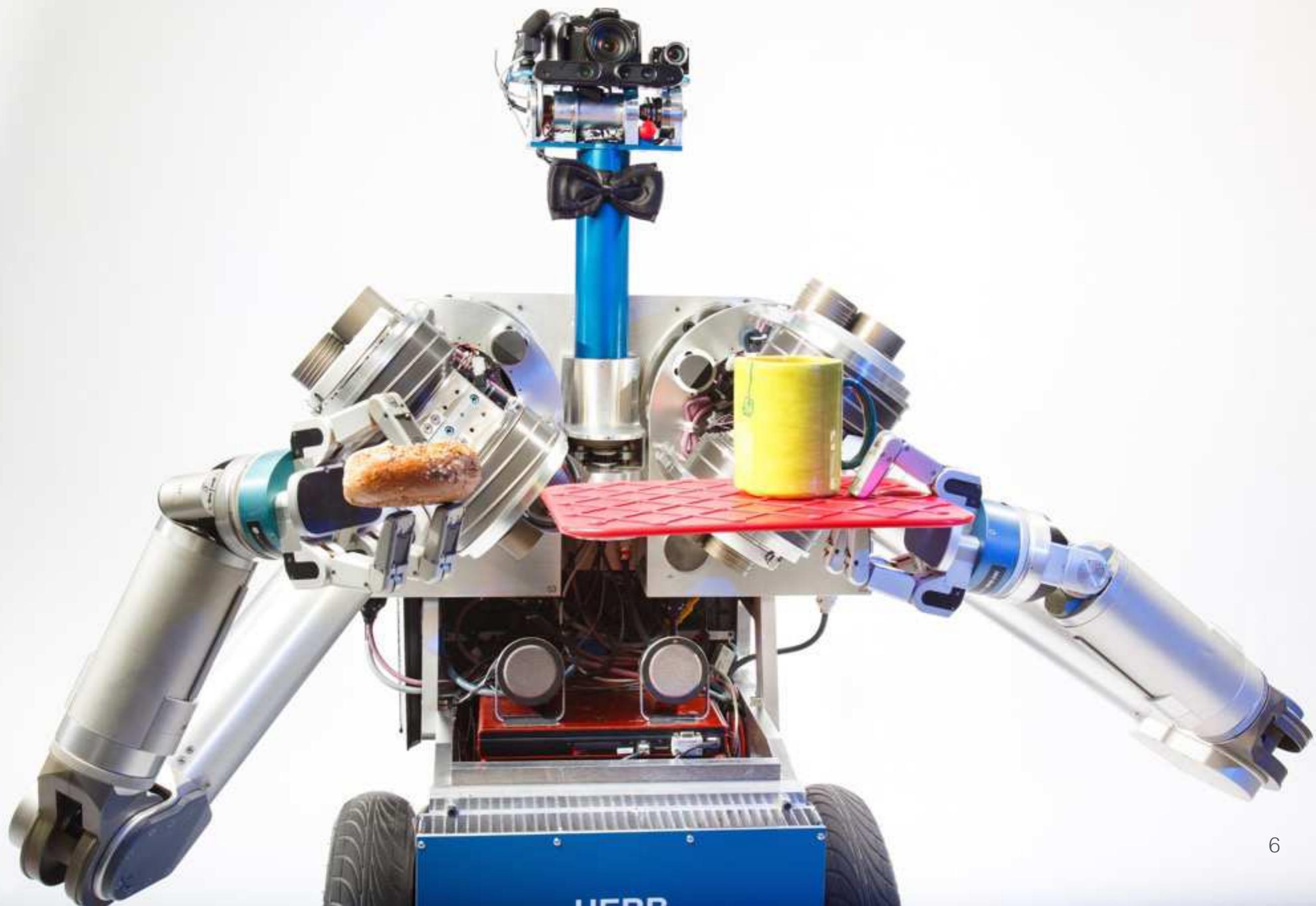
February 9, 2016
Hands: Design and Control for
Dexterous Manipulation



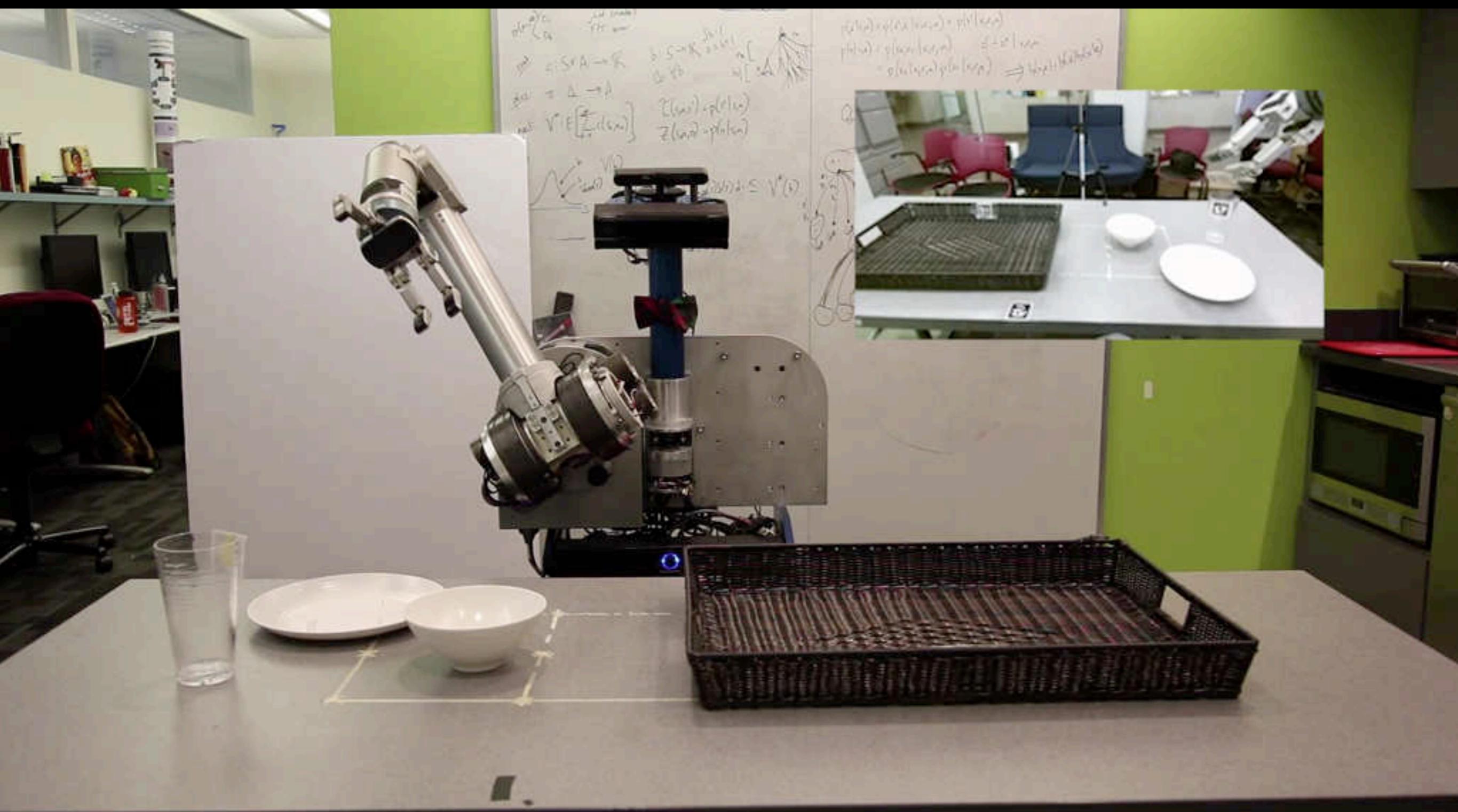


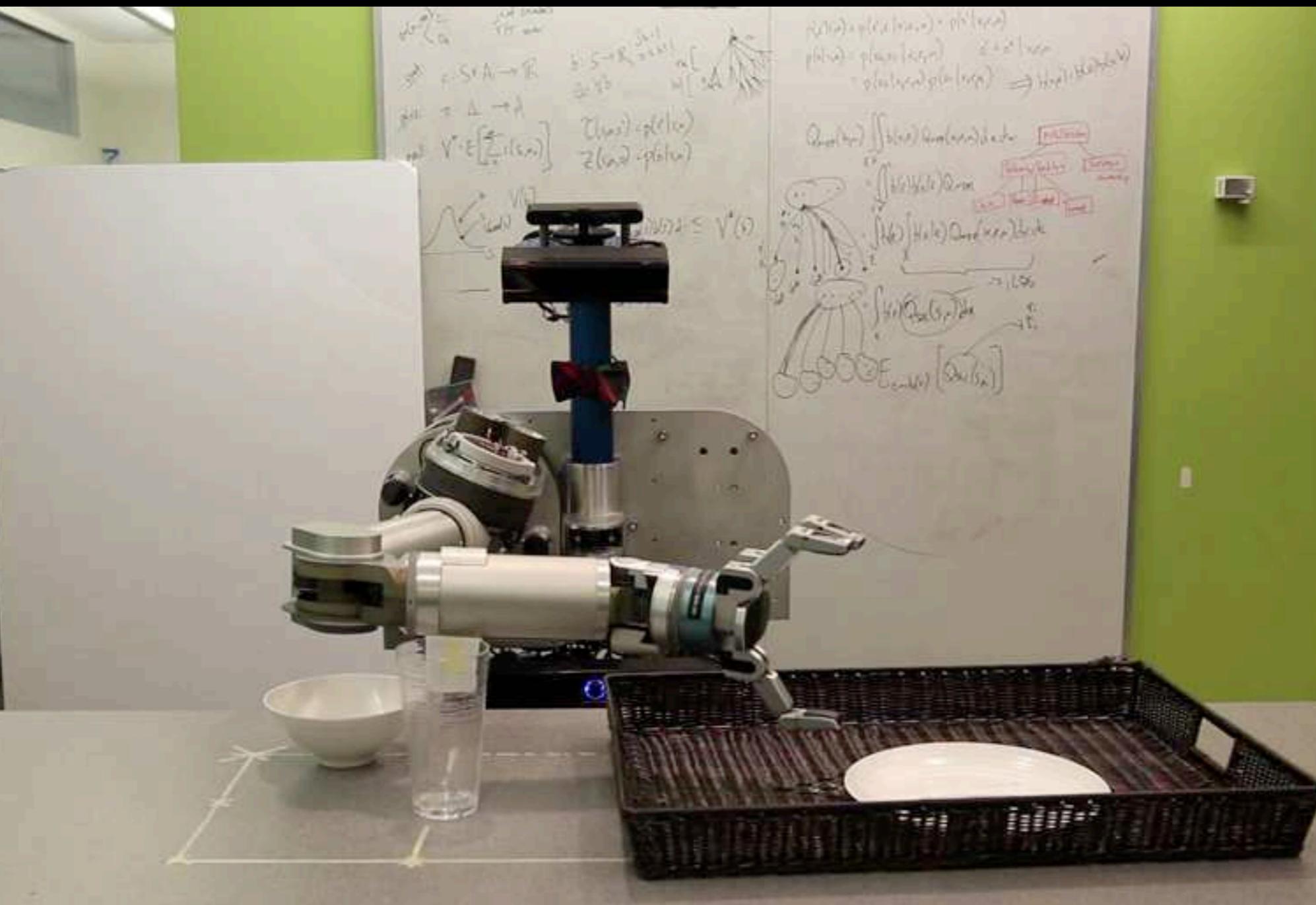






(4×)



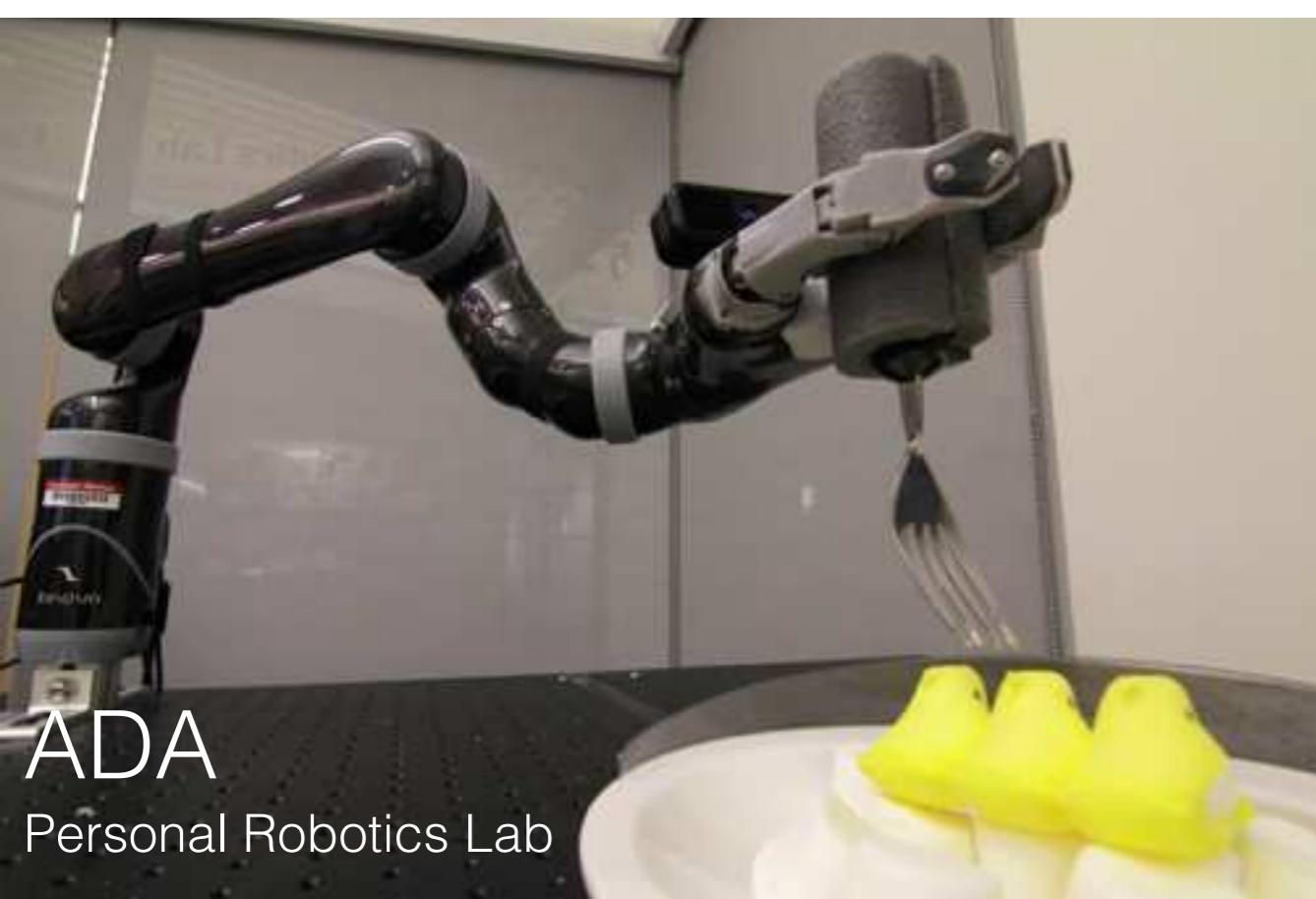






HERB

Personal Robotics Lab



ADA

Personal Robotics Lab



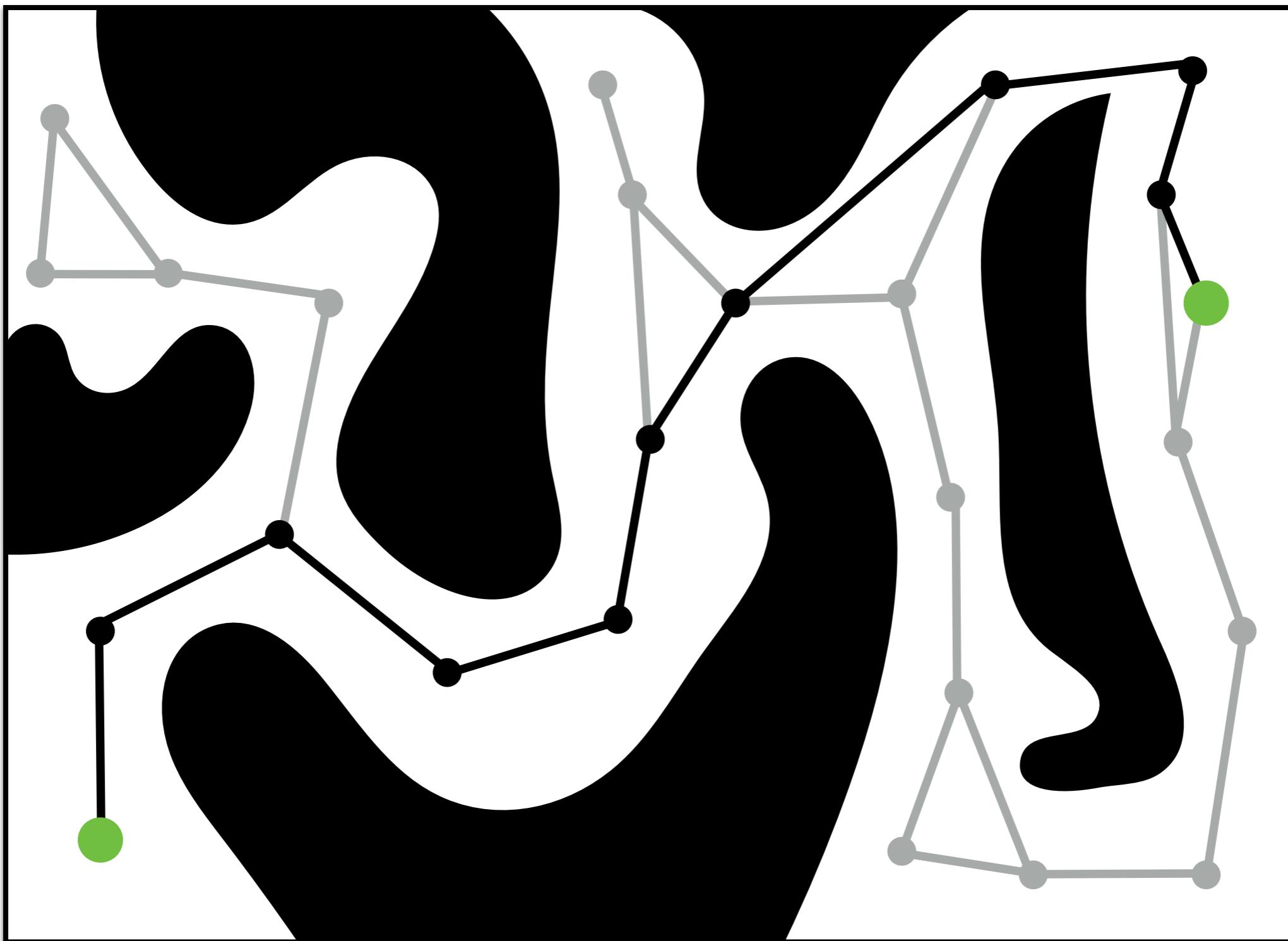
Robonaut 2

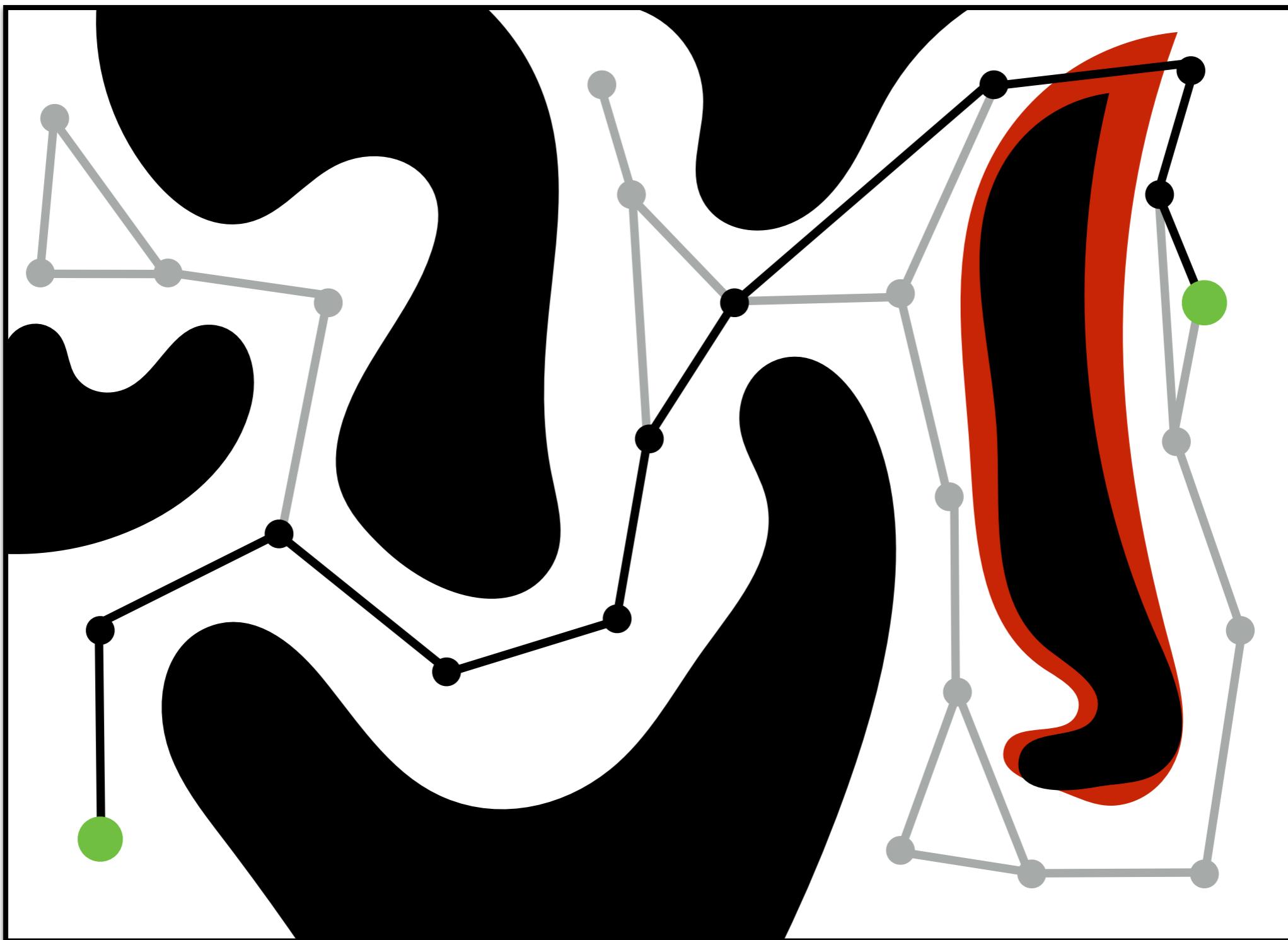
NASA/JSC

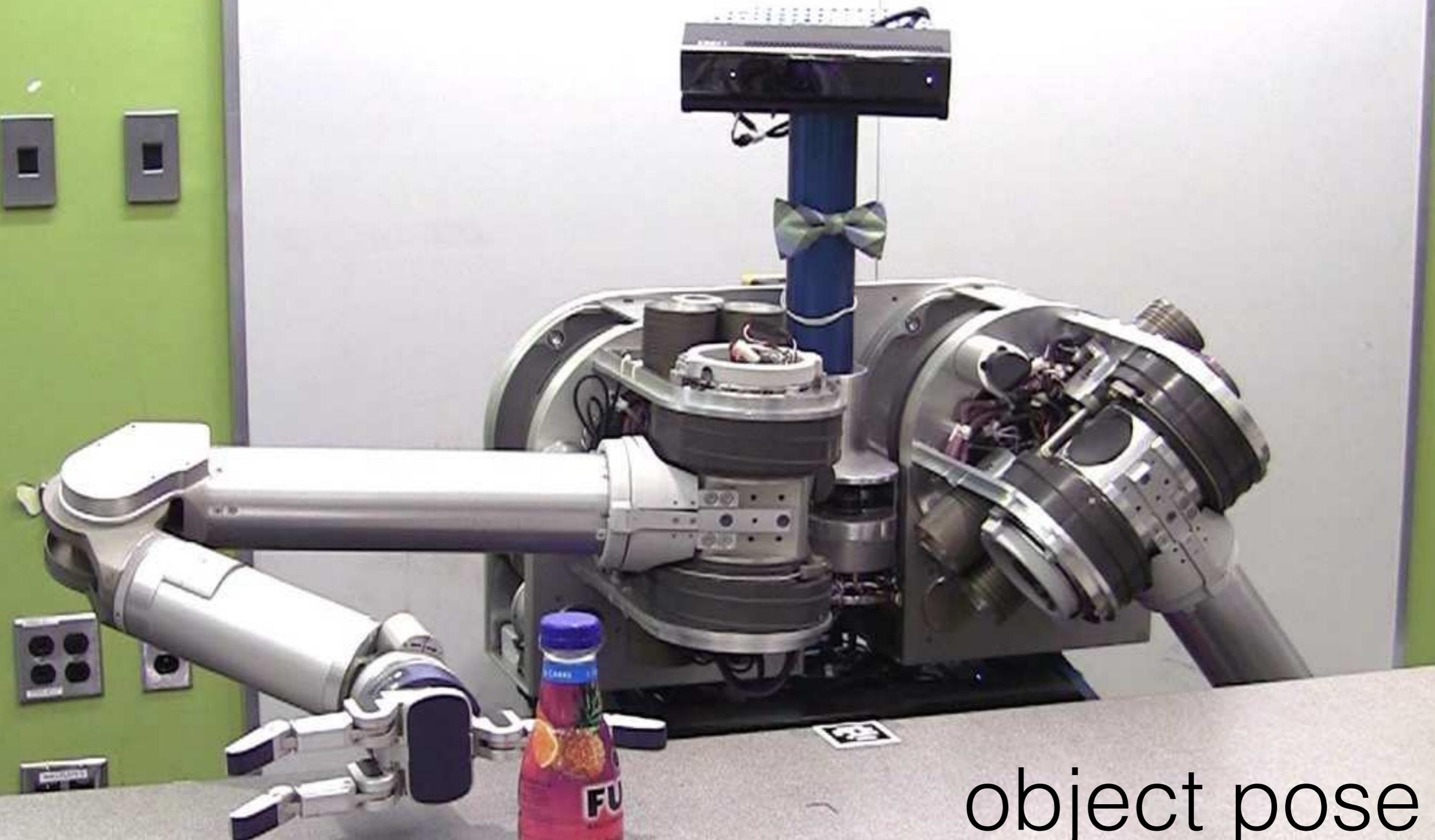
Andy

CMU/NREC

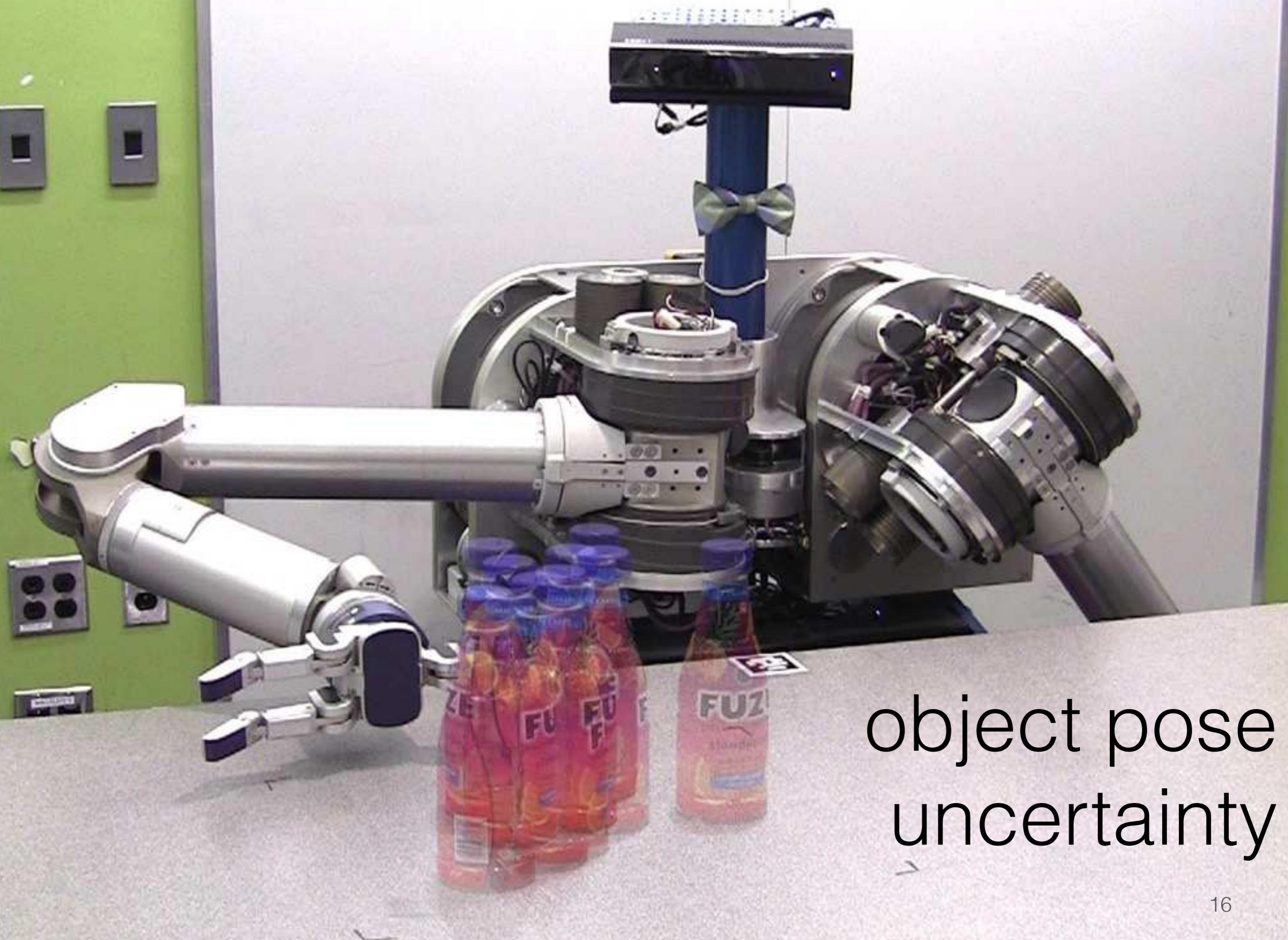
What caused these failures?



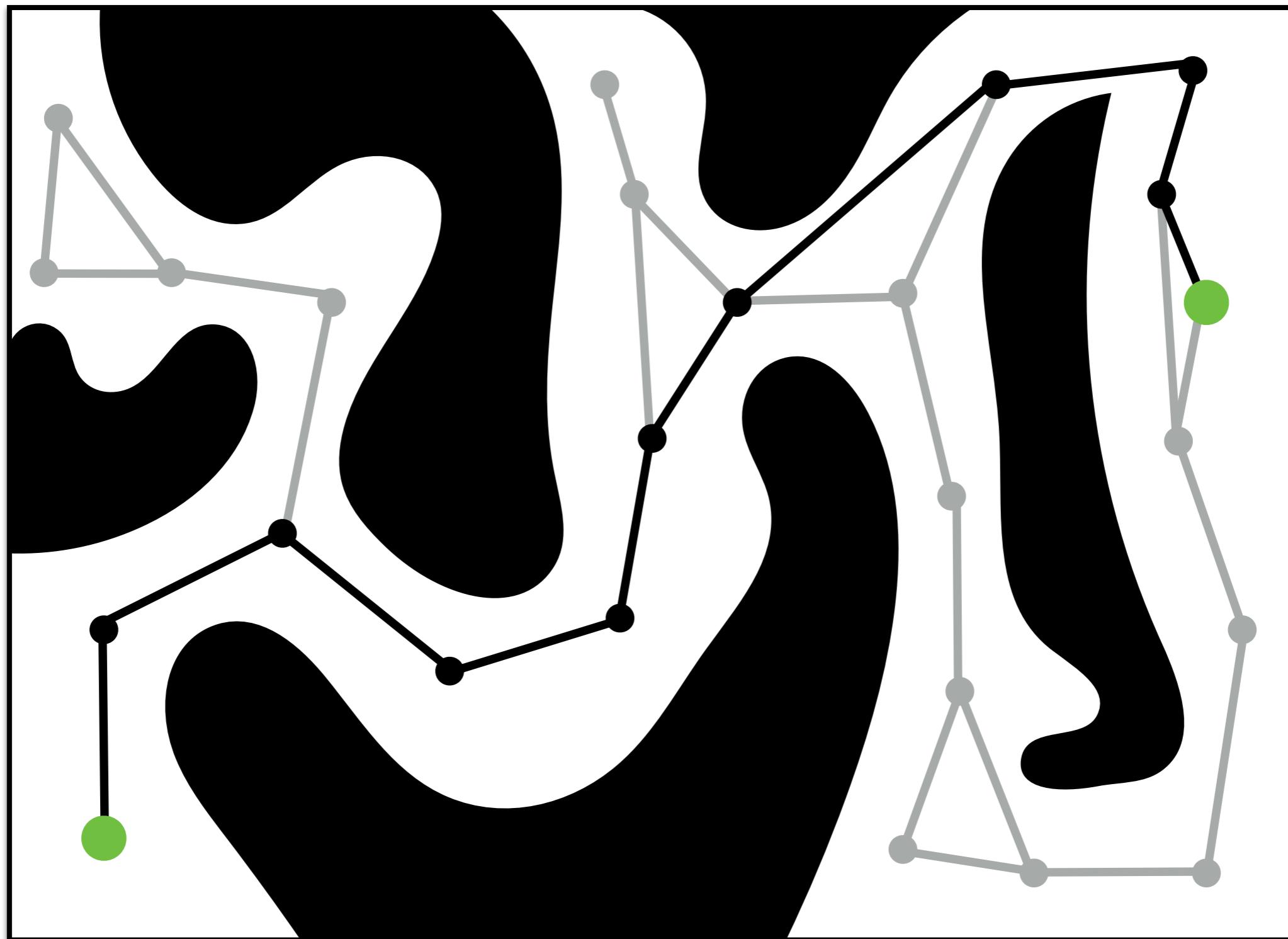


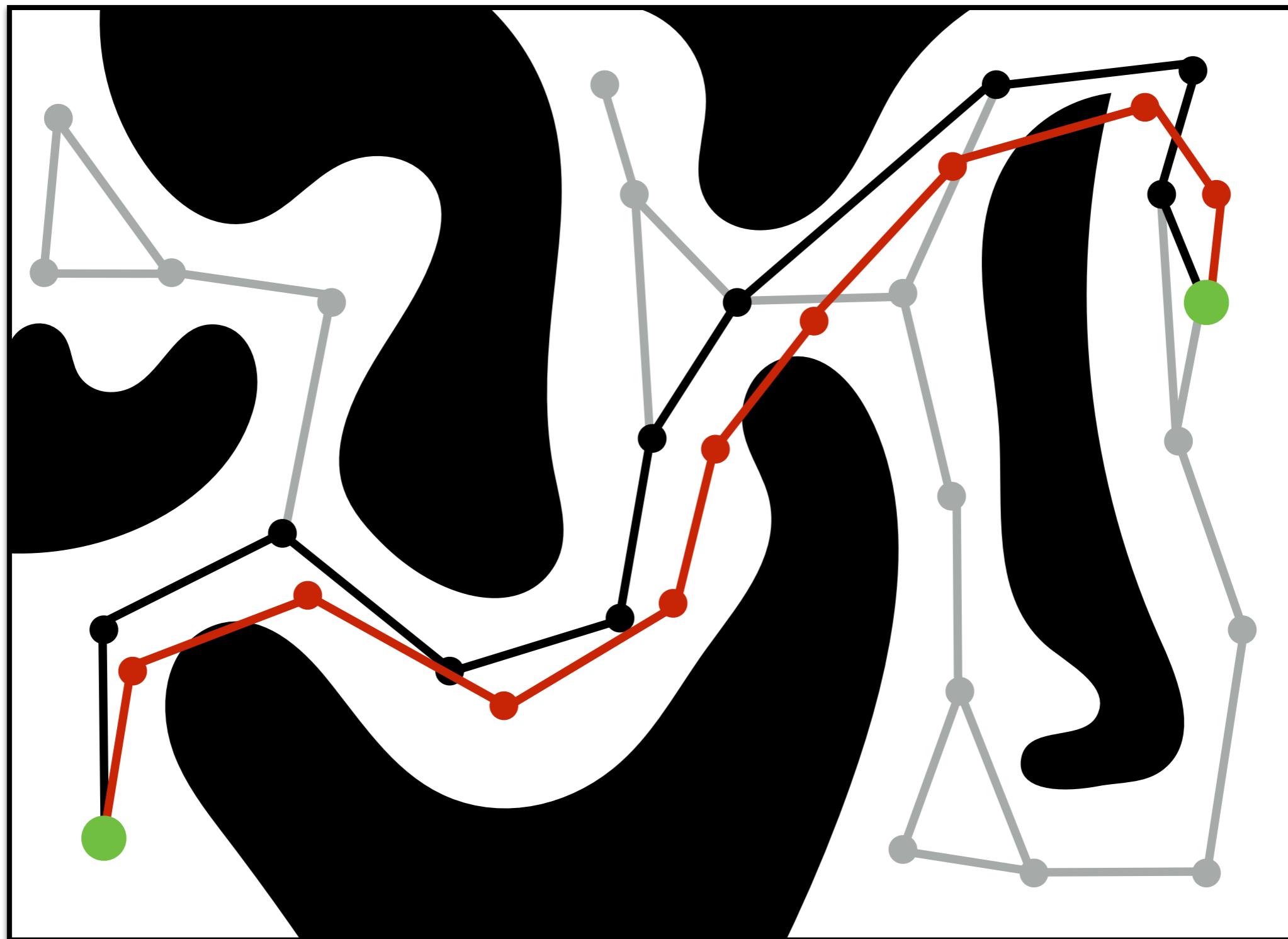


object pose
uncertainty



object pose
uncertainty







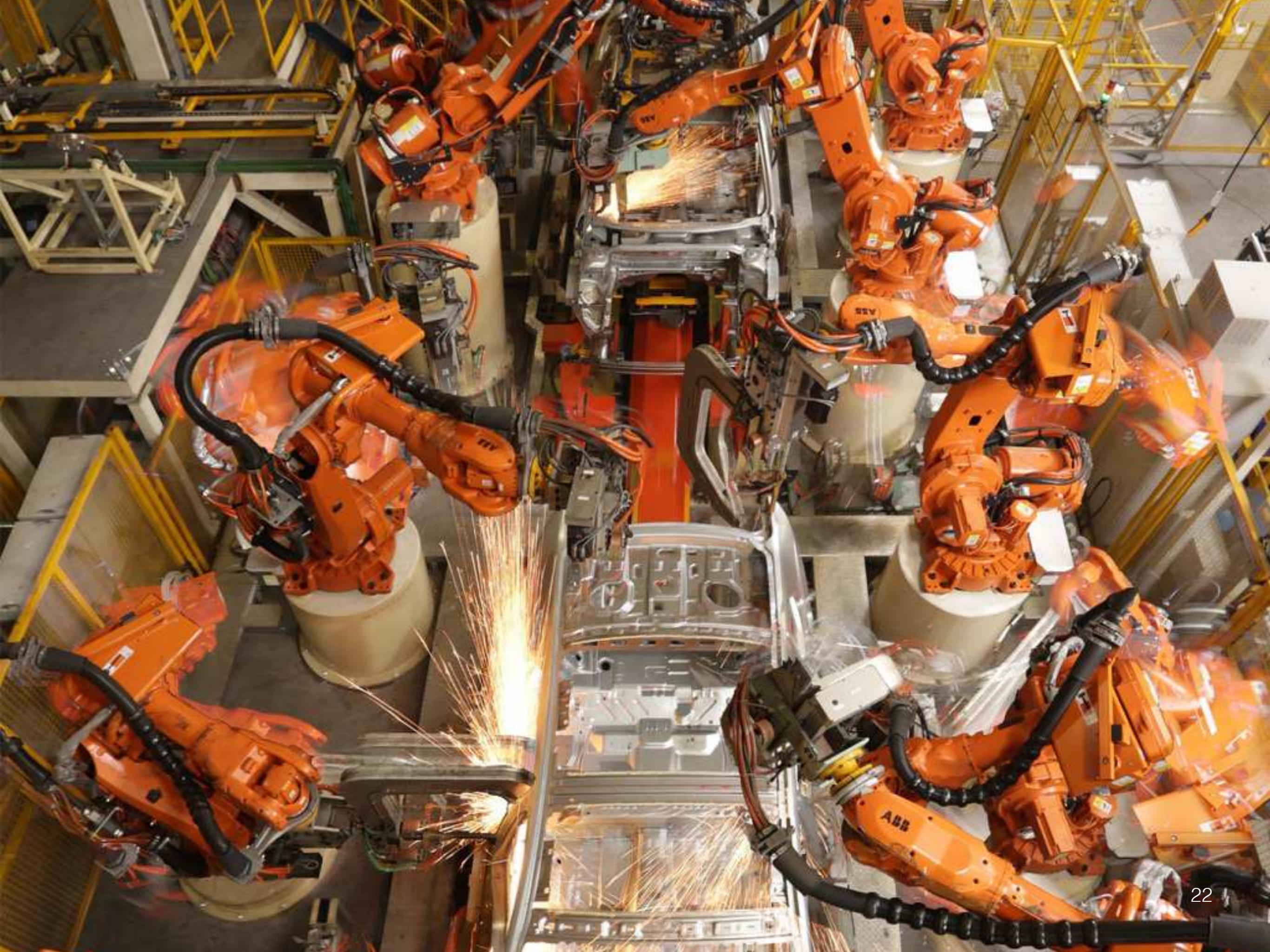
proprioceptive
uncertainty

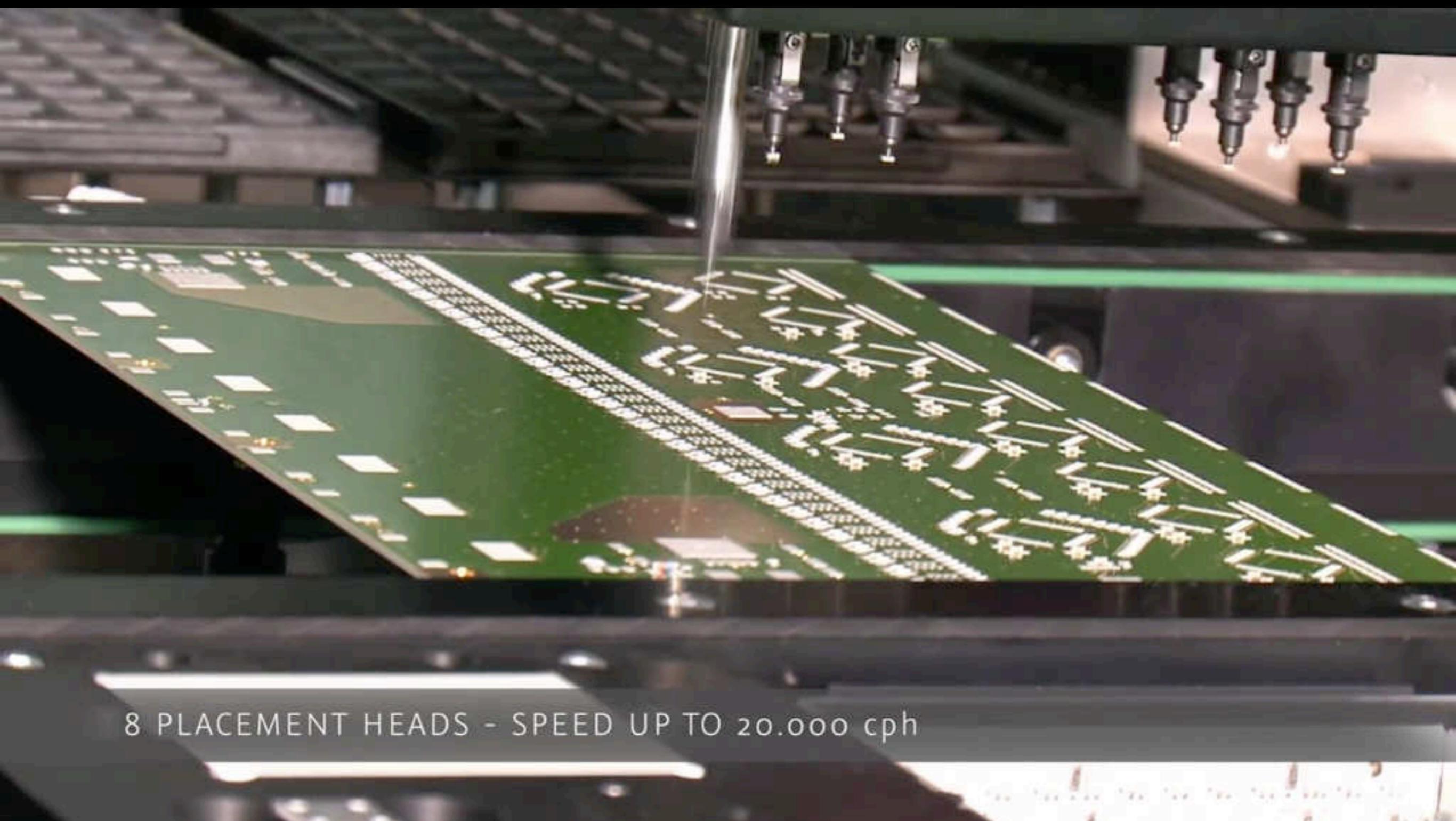
A close-up photograph of a robotic hand, likely a Dobot, positioned in the upper right corner. The hand is holding a small, translucent red cube. The background is a light-colored surface with a black and white checkered pattern in the top left corner.

proprioceptive
uncertainty

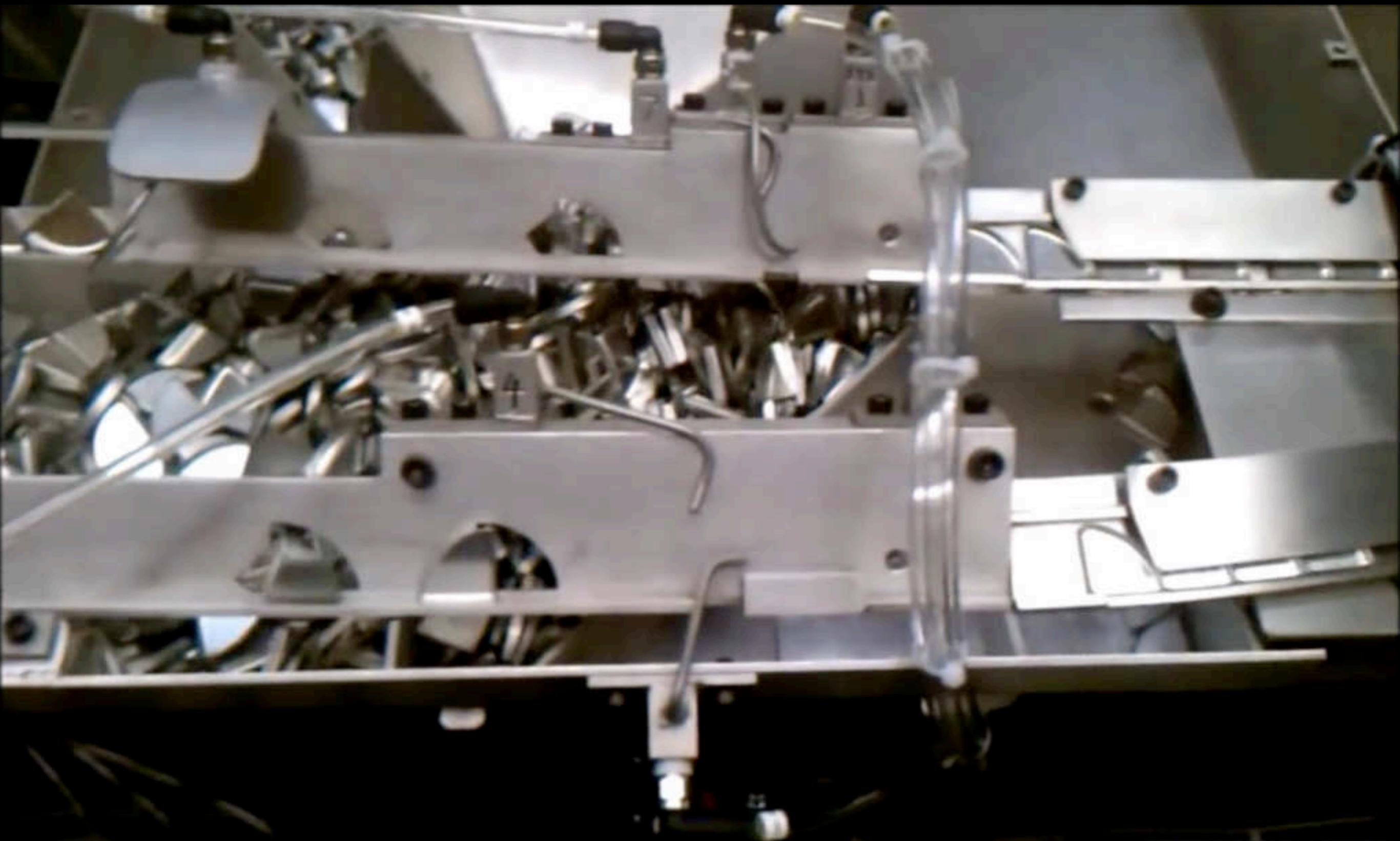


model
uncertainty



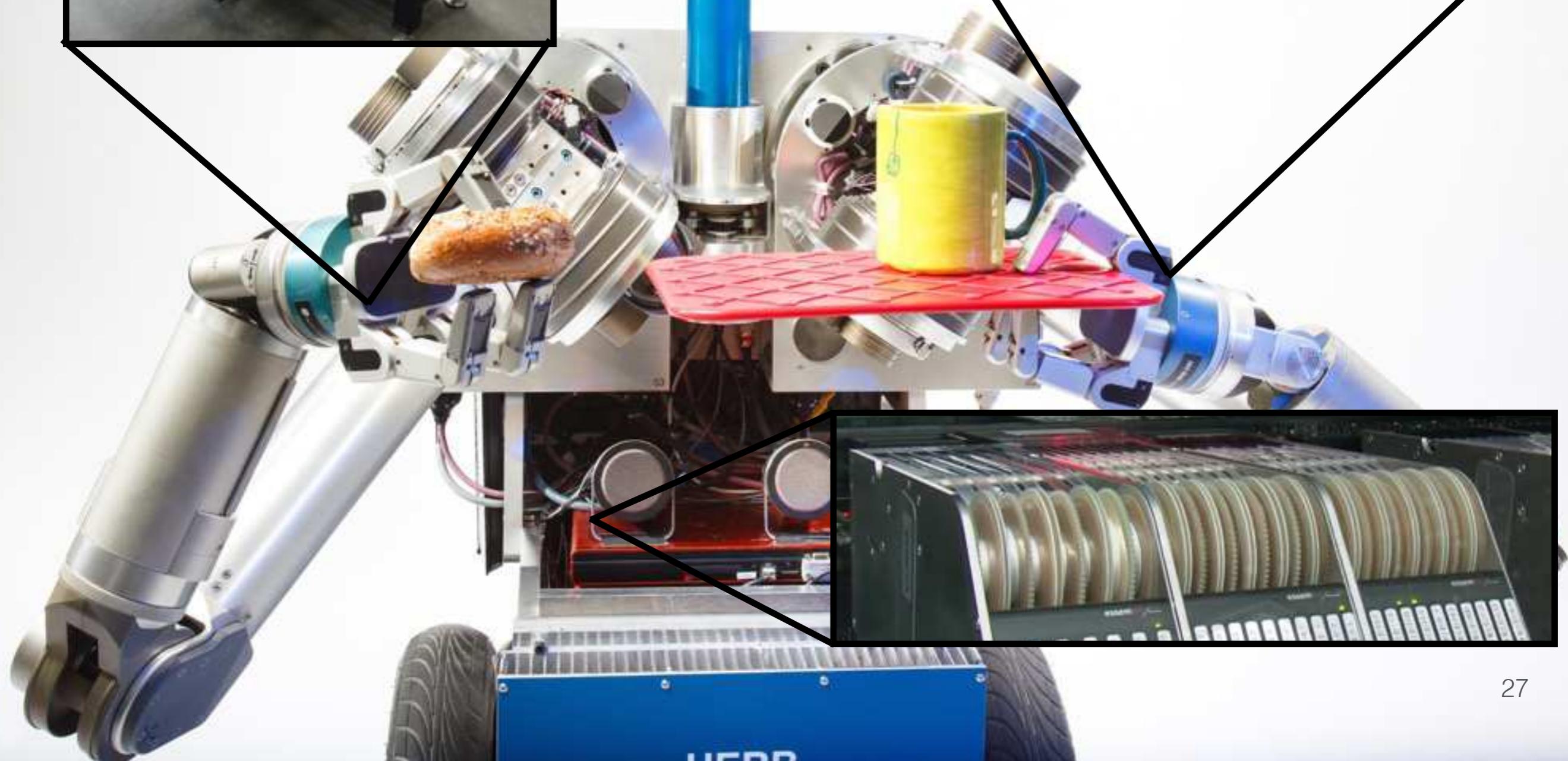


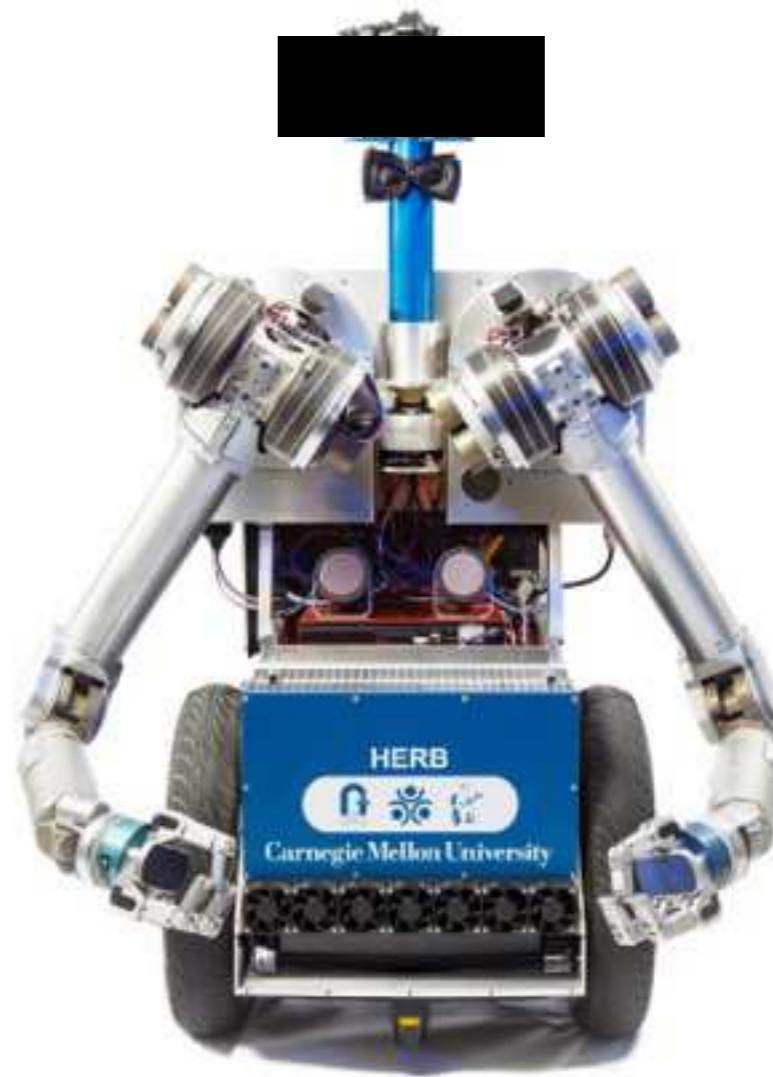
8 PLACEMENT HEADS - SPEED UP TO 20.000 cph





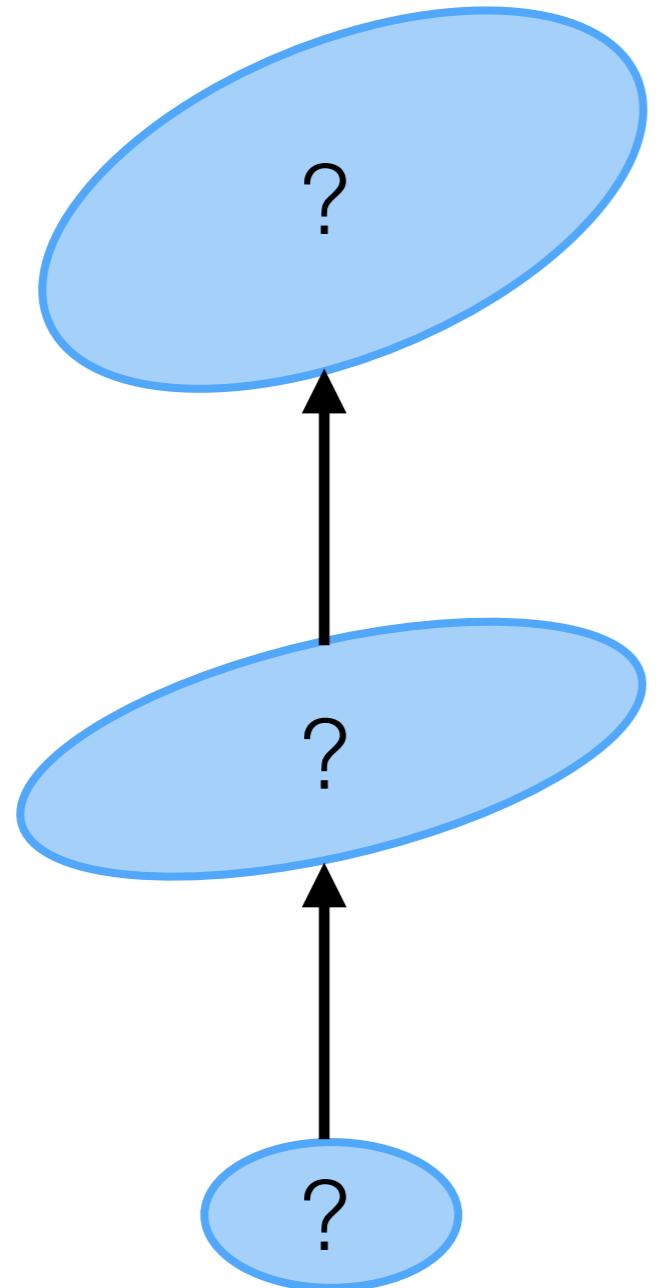
How can we manipulate under uncertainty?



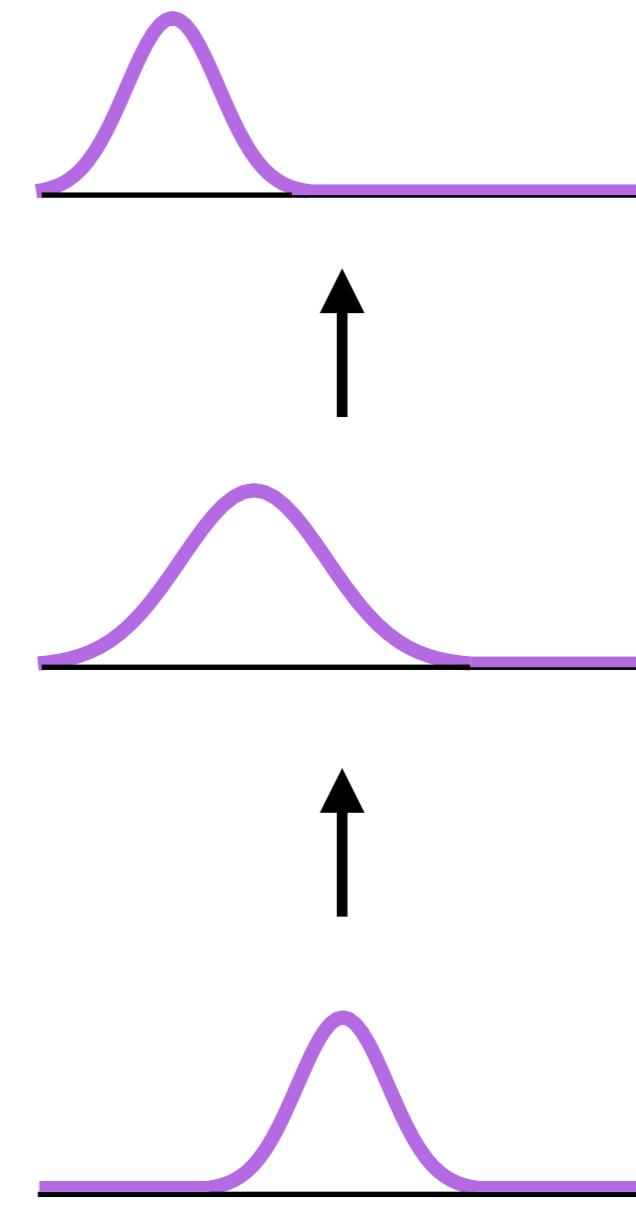


Closed-loop or open-loop?

“worst case”

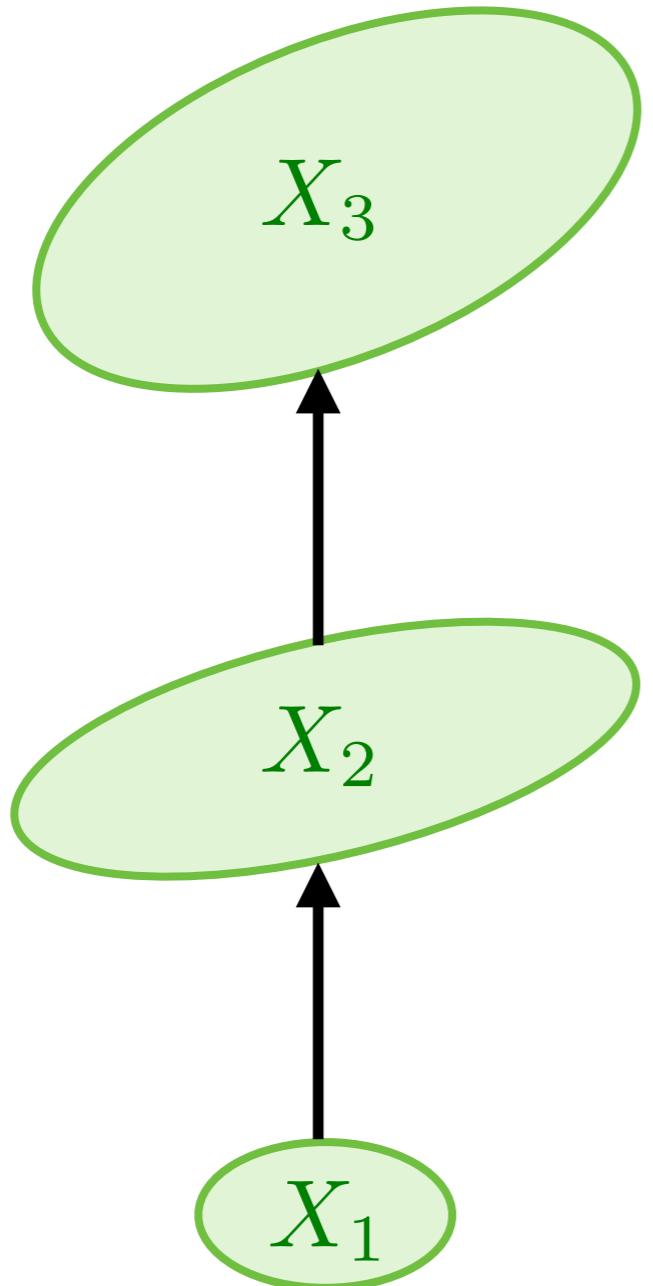


“average case”

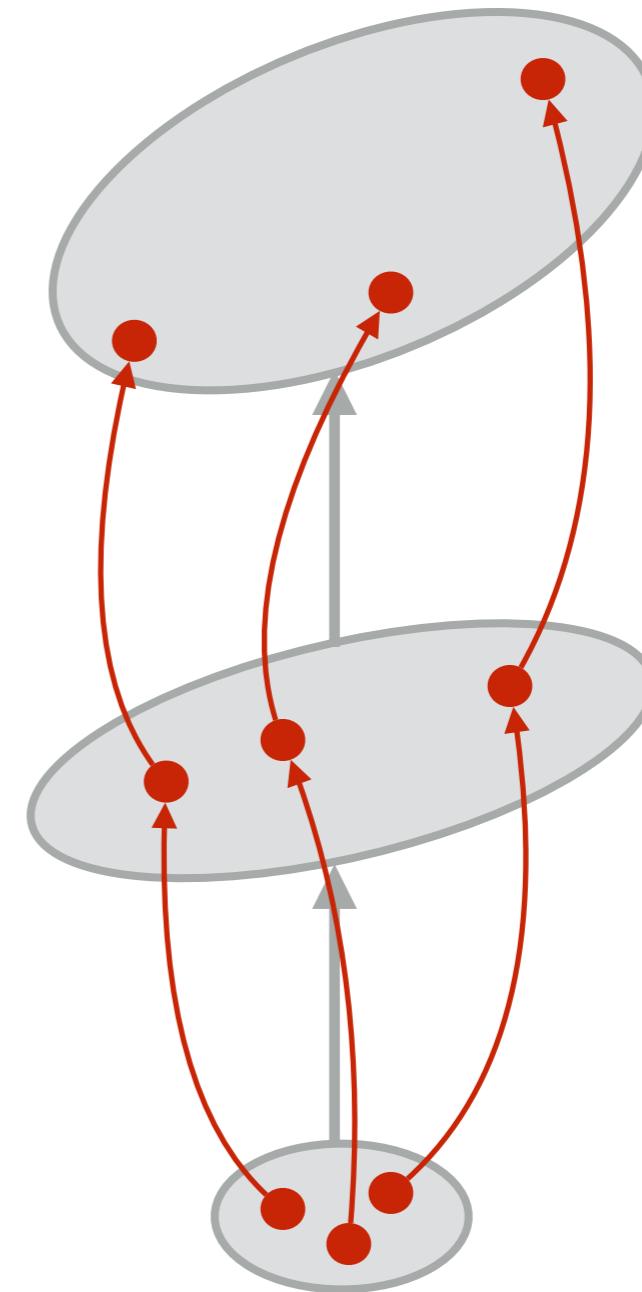


Non-deterministic or probabilistic uncertainty?

“Kalman filter”



“particle filter”



Closed-form or sample-based representation?

Estimate, react, or plan?



Visual Feedback



Tactile Feedback



No Feedback

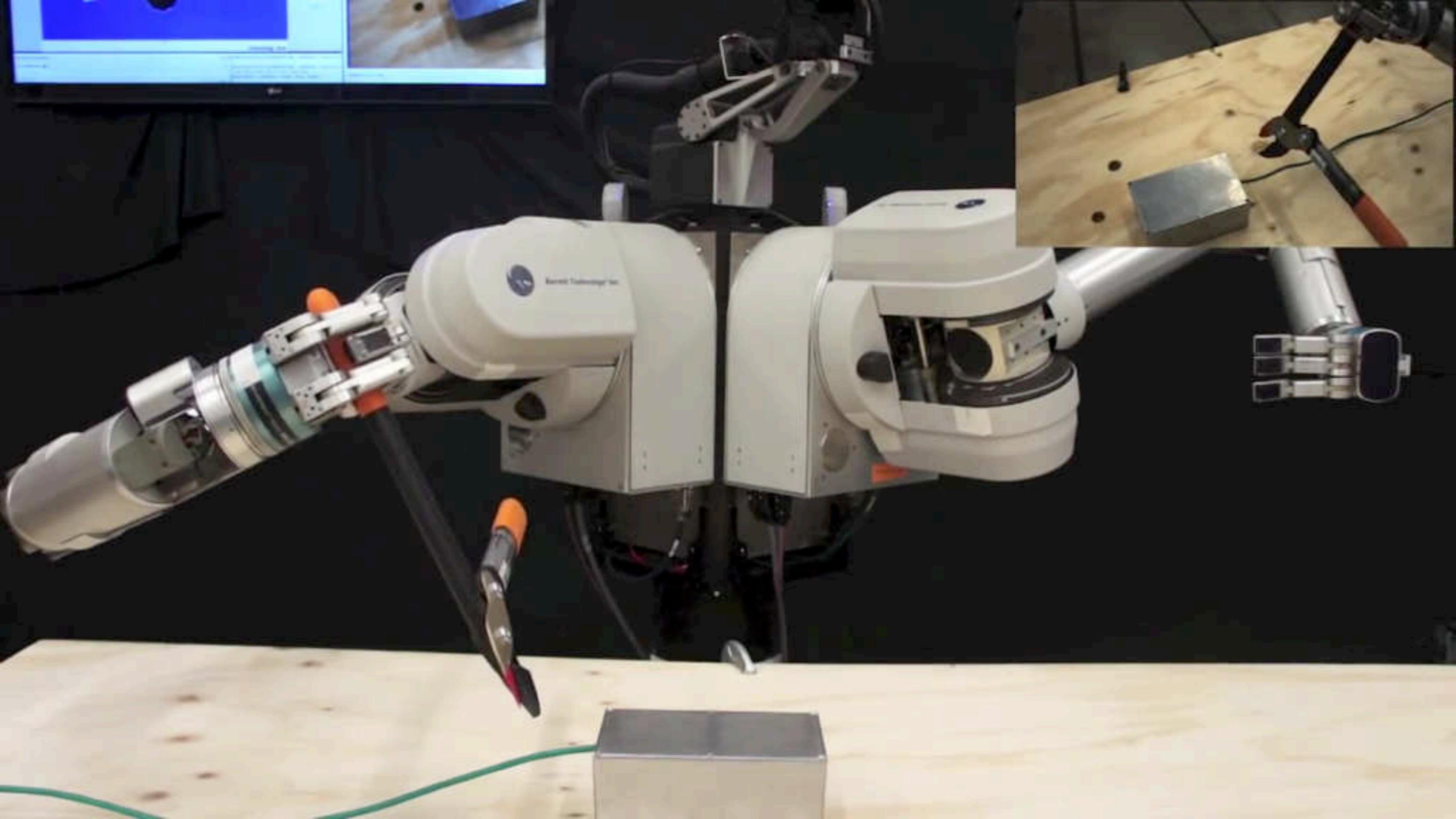
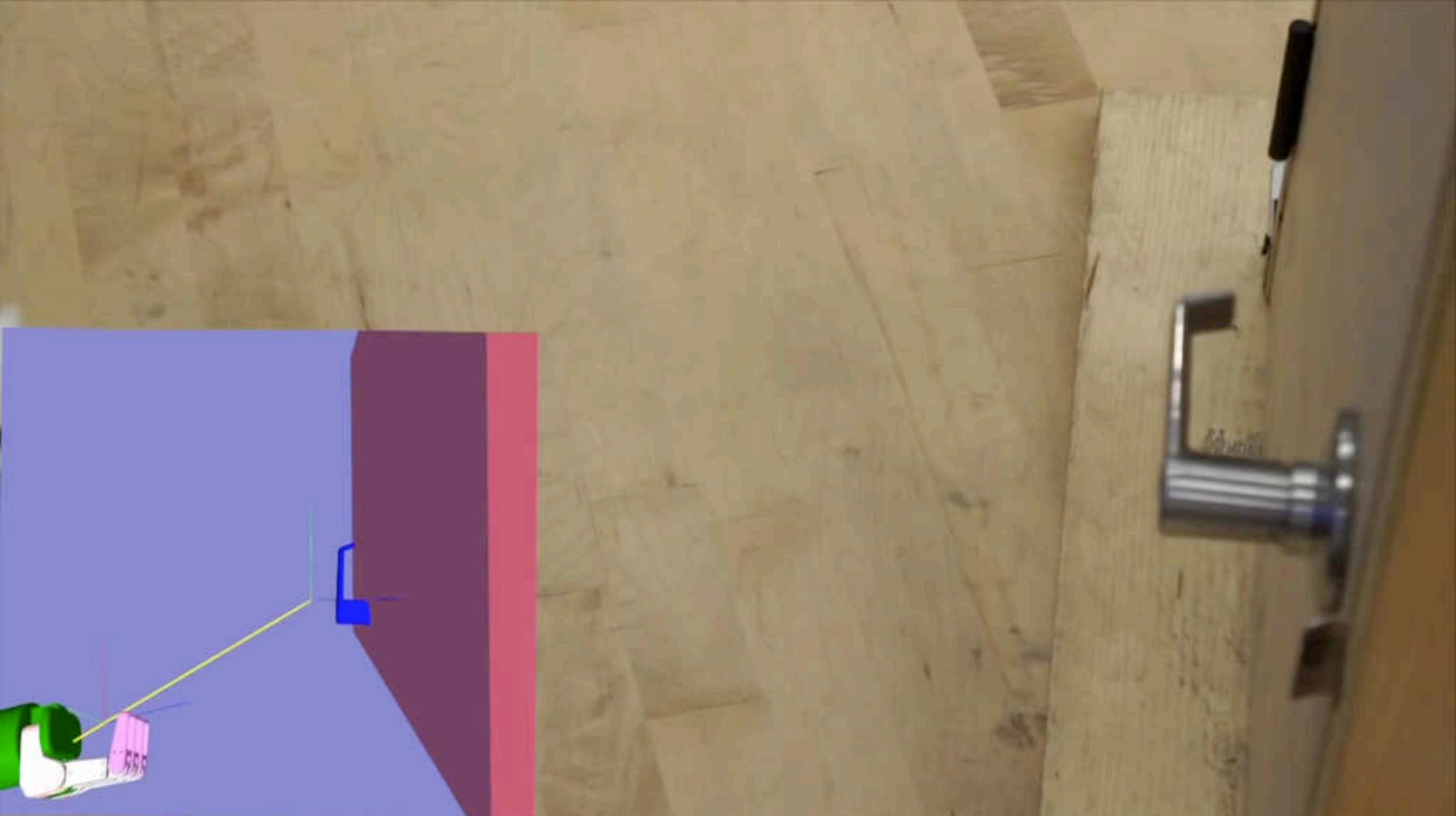
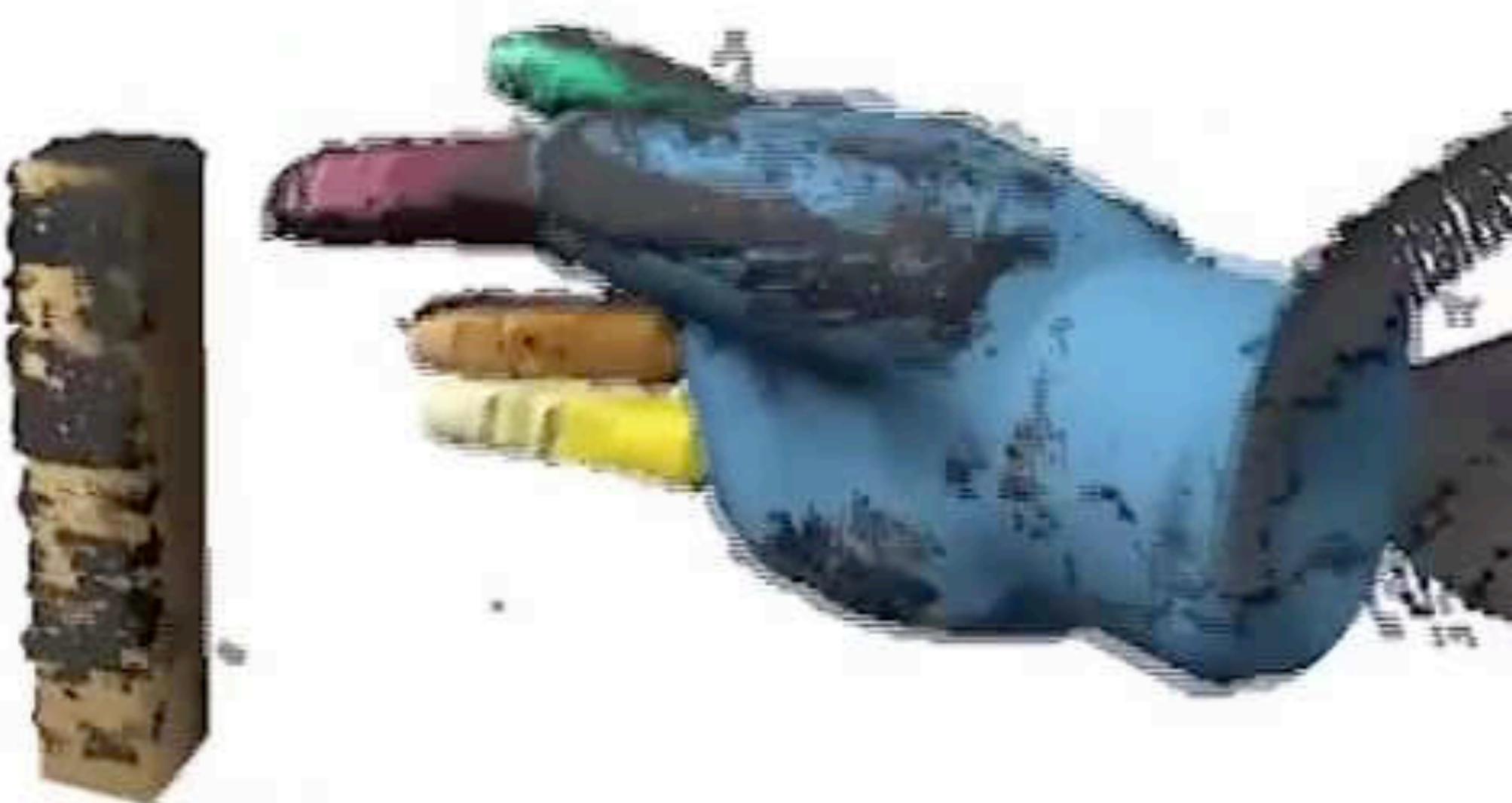


Image-space visual servoing



Markerless real-time articulated tracking

realtime tracking of articulated objects



dark grey: depth measurements
colored: tracked model



Visual Feedback



Tactile Feedback



No Feedback



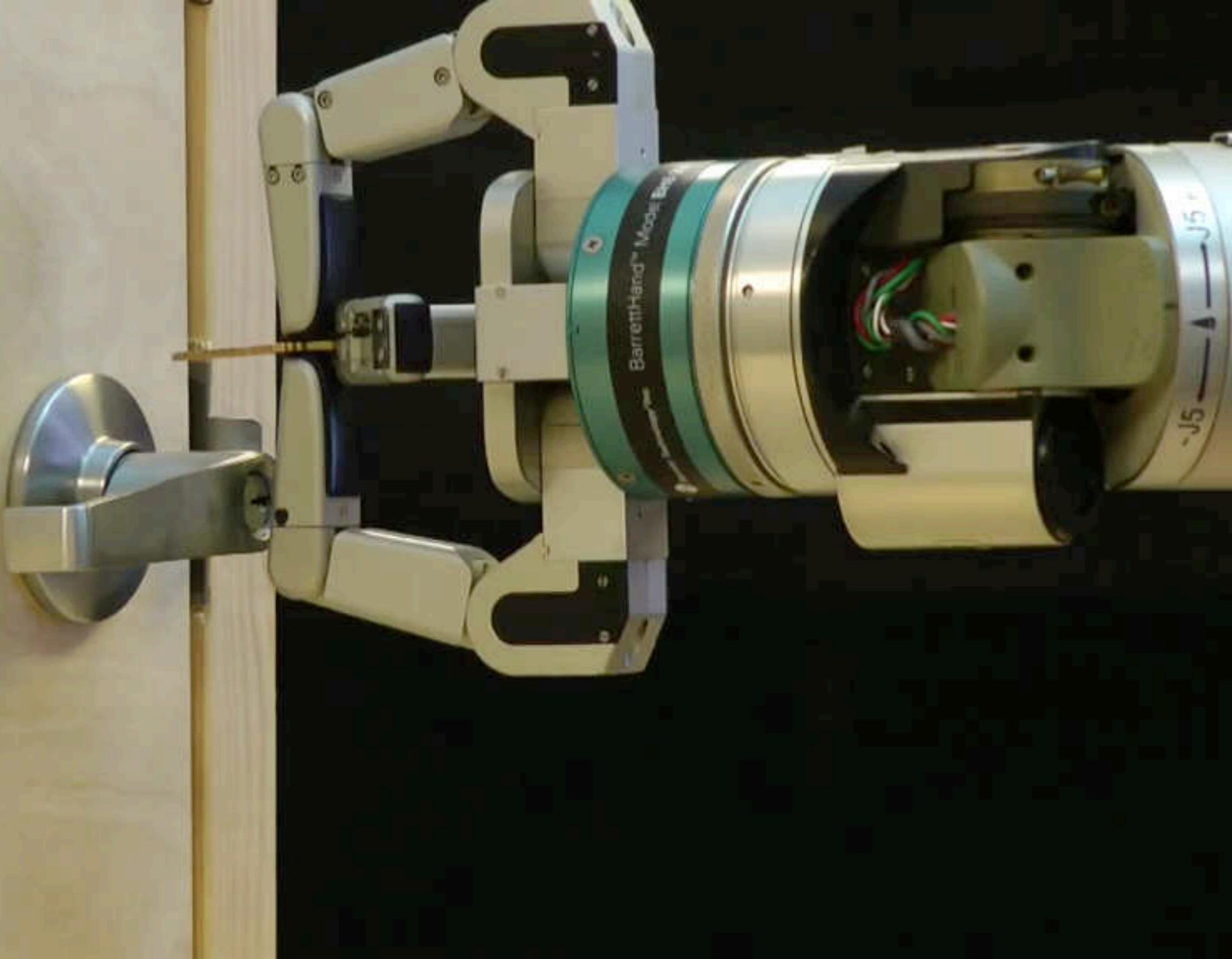
Visual Feedback



Tactile Feedback



No Feedback



2X Realtime

Use “guarded moves” to reduce uncertainty

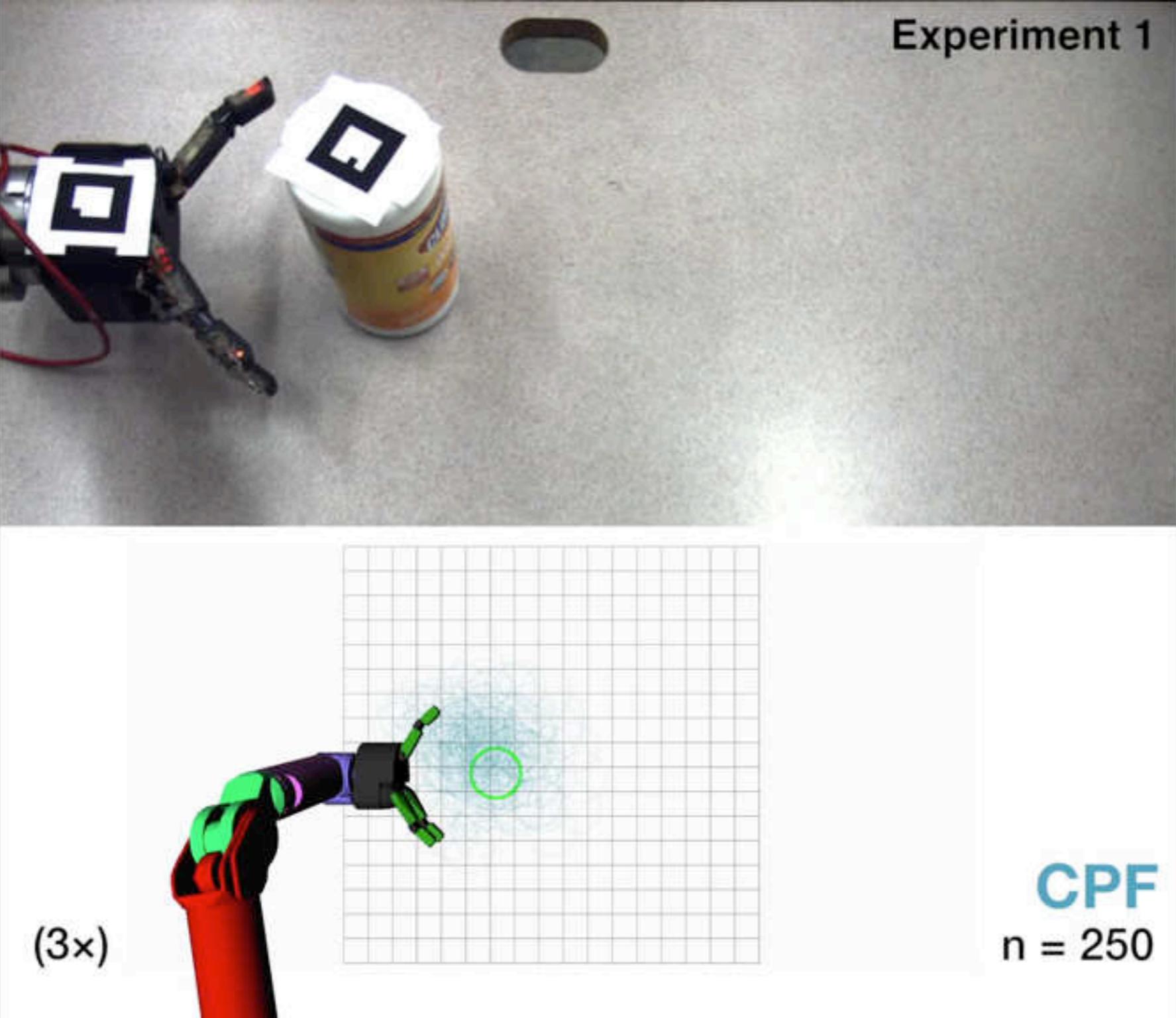


Selecting next action...

Ix

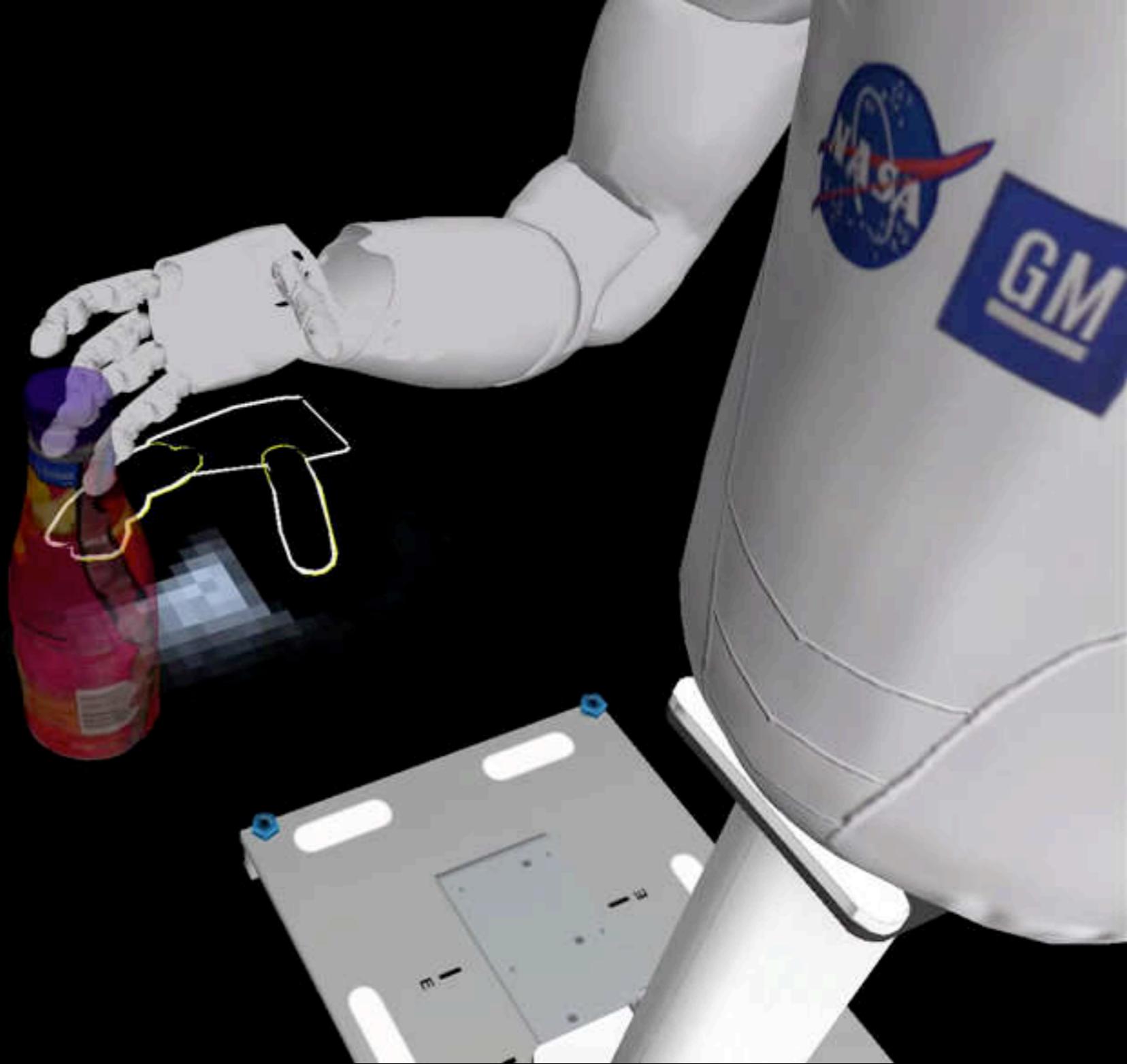
Plan a sequence that maximizes information gain

Experiment 1



Estimate the pose of the object using tactile sensing

M.C. Koval, M.R. Dogar, N.S. Pollard, and S.S. Srinivasa. "Pose estimation for contact manipulation using manifold particle filters." IROS, 2013.
M.C. Koval, N.S. Pollard, and S.S. Srinivasa. "Manifold representations for state estimation in contact manipulation." ISRR, 2013.
M.C. Koval, N.S. Pollard, and S.S. Srinivasa. "Pose estimation for planar contact manipulation with manifold particle filters." IJRR, 2015.



Closed-loop grasping using contact sensing



Goal of the learned movement Misplaced object



clmc.usc.edu

Learn feedback policies that use sensor feedback



Policy Input



**conv3 response map
feature 1 of 32**

A grayscale heatmap showing a response map for a convolutional layer. The map is highly localized, with a bright central region and darker surrounding areas, indicating which pixels in the input image triggered the first feature in the third convolutional layer.

Learn feedback policies that use sensor feedback



Visual Feedback



Tactile Feedback



No Feedback



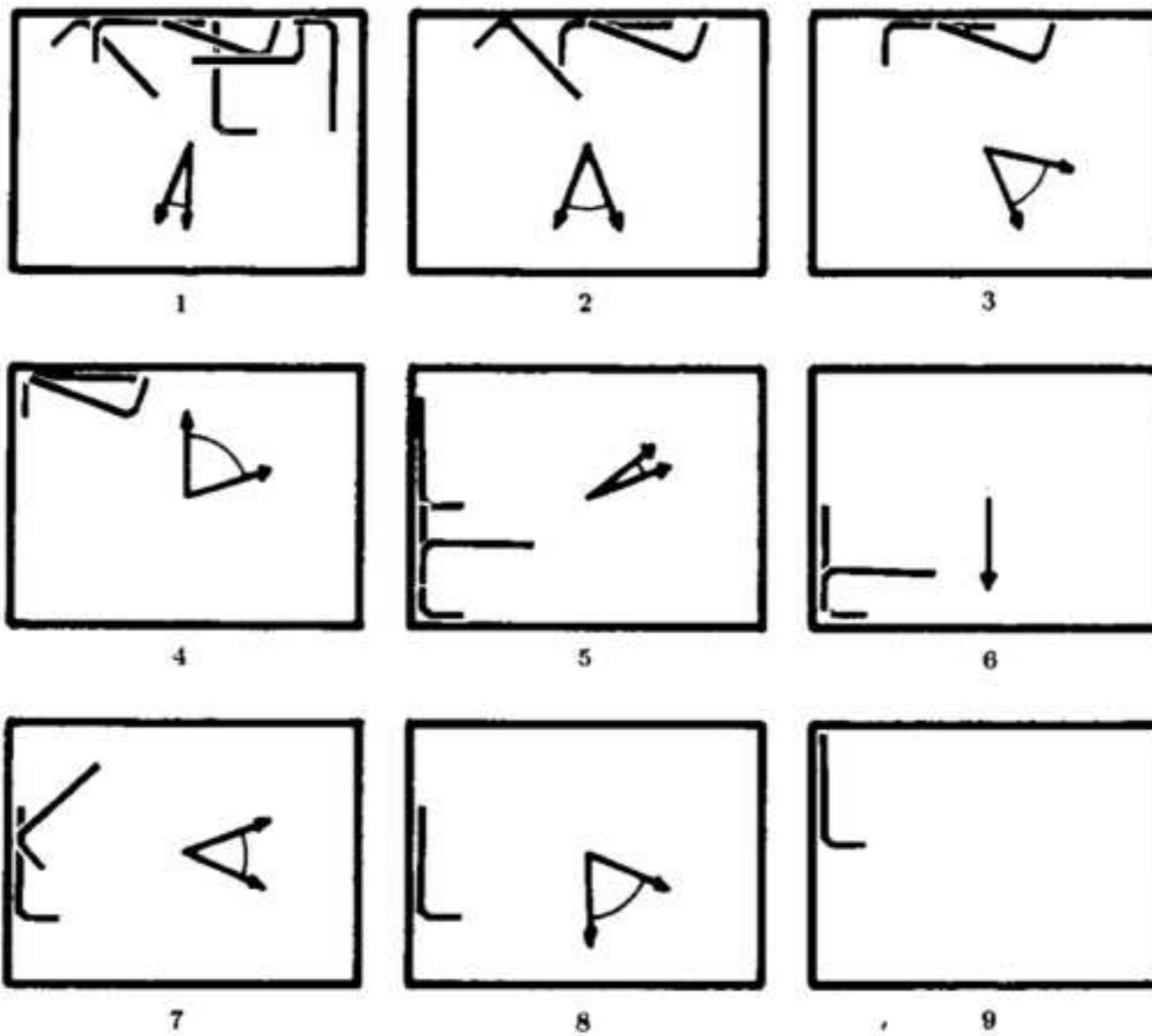
Visual Feedback



Tactile Feedback



No Feedback



Open-loop robotic part alignment



Push Grasping

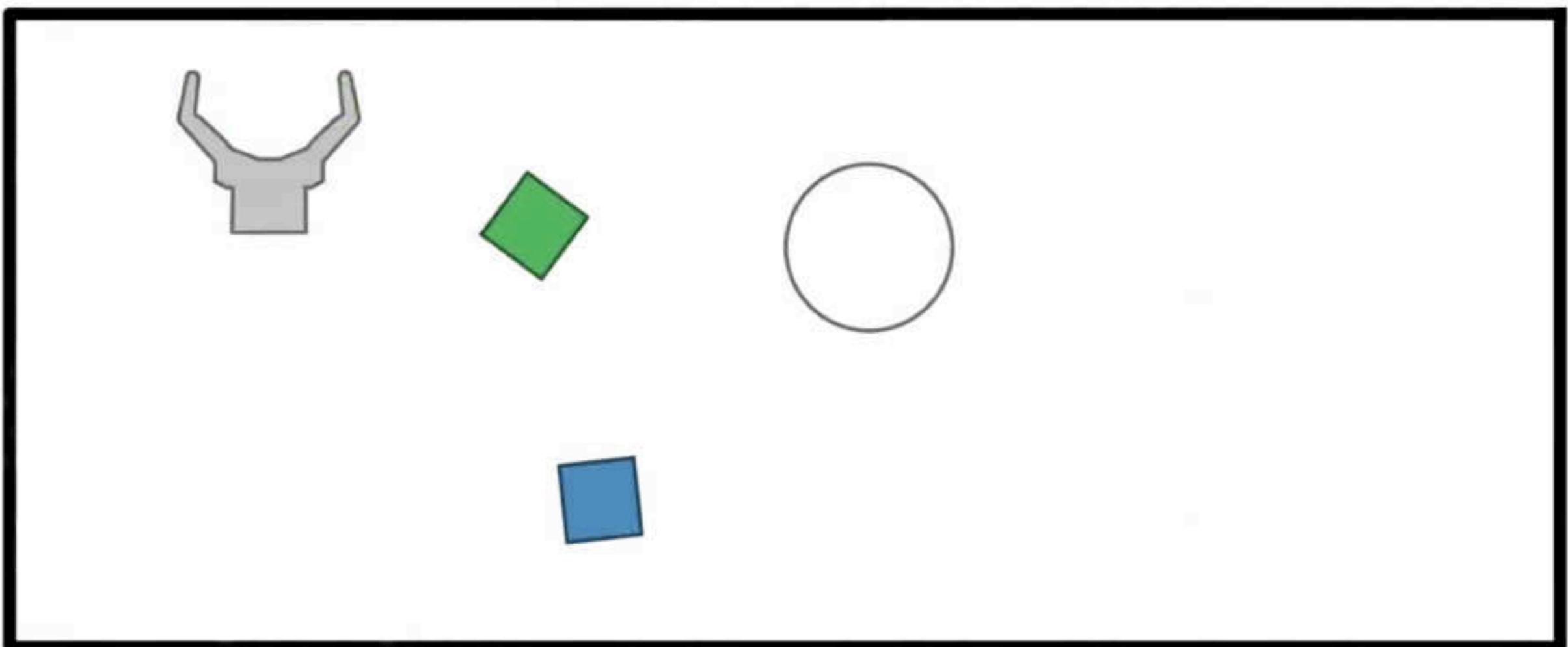
M. Dogar and S. Srinivasa. "Push-grasping with dexterous hands: Mechanics and a method." IROS, 2010.

M. Dogar and S. Srinivasa. "A framework for push-grasping in clutter." RSS, 2011.

M. Dogar, K. Hsiao, M. Ciocarlie, and S. Srinivasa "Physics-based grasp planning through clutter." RSS, 2012.



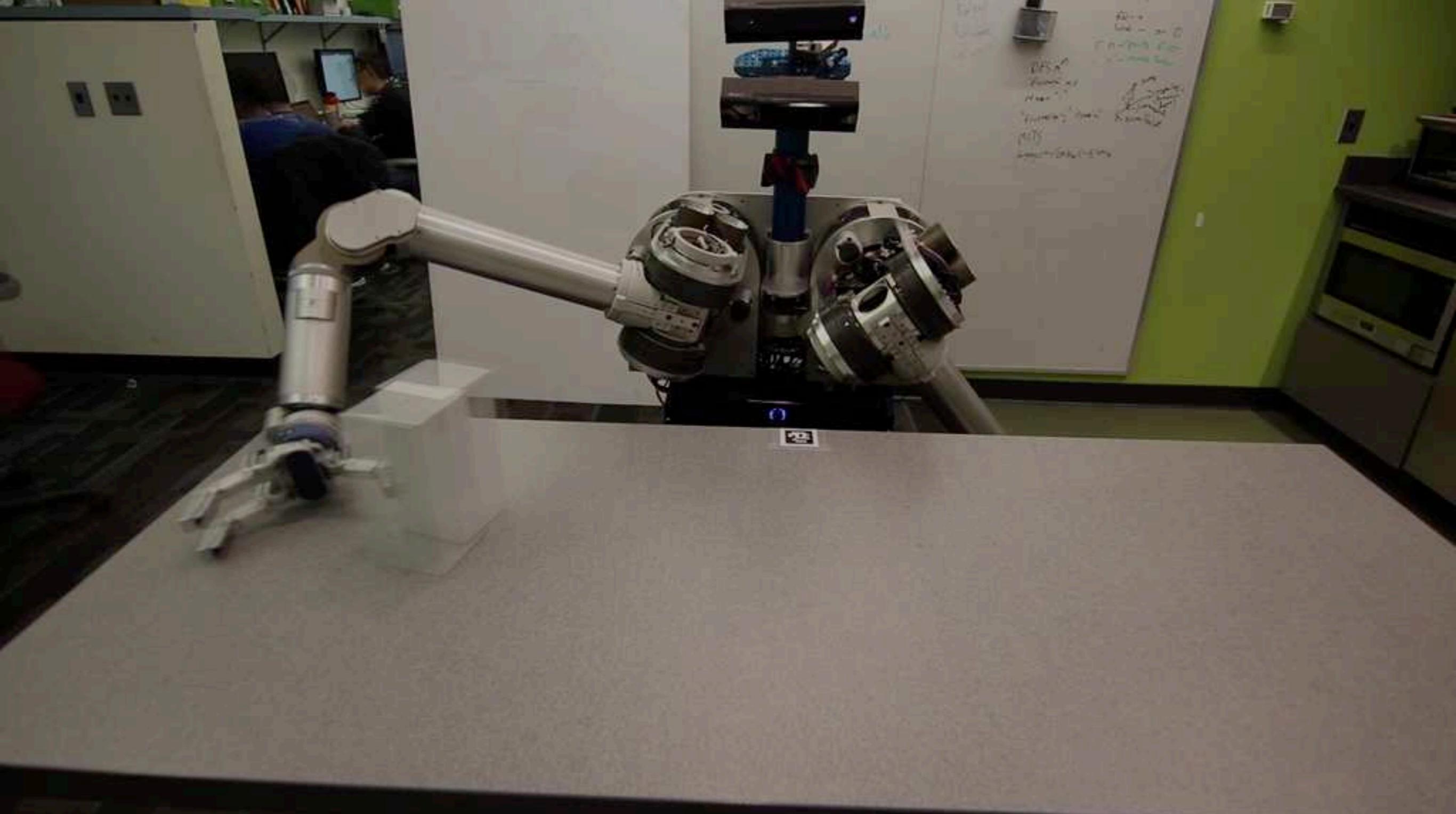
Rearrangement Planning



Plan 1

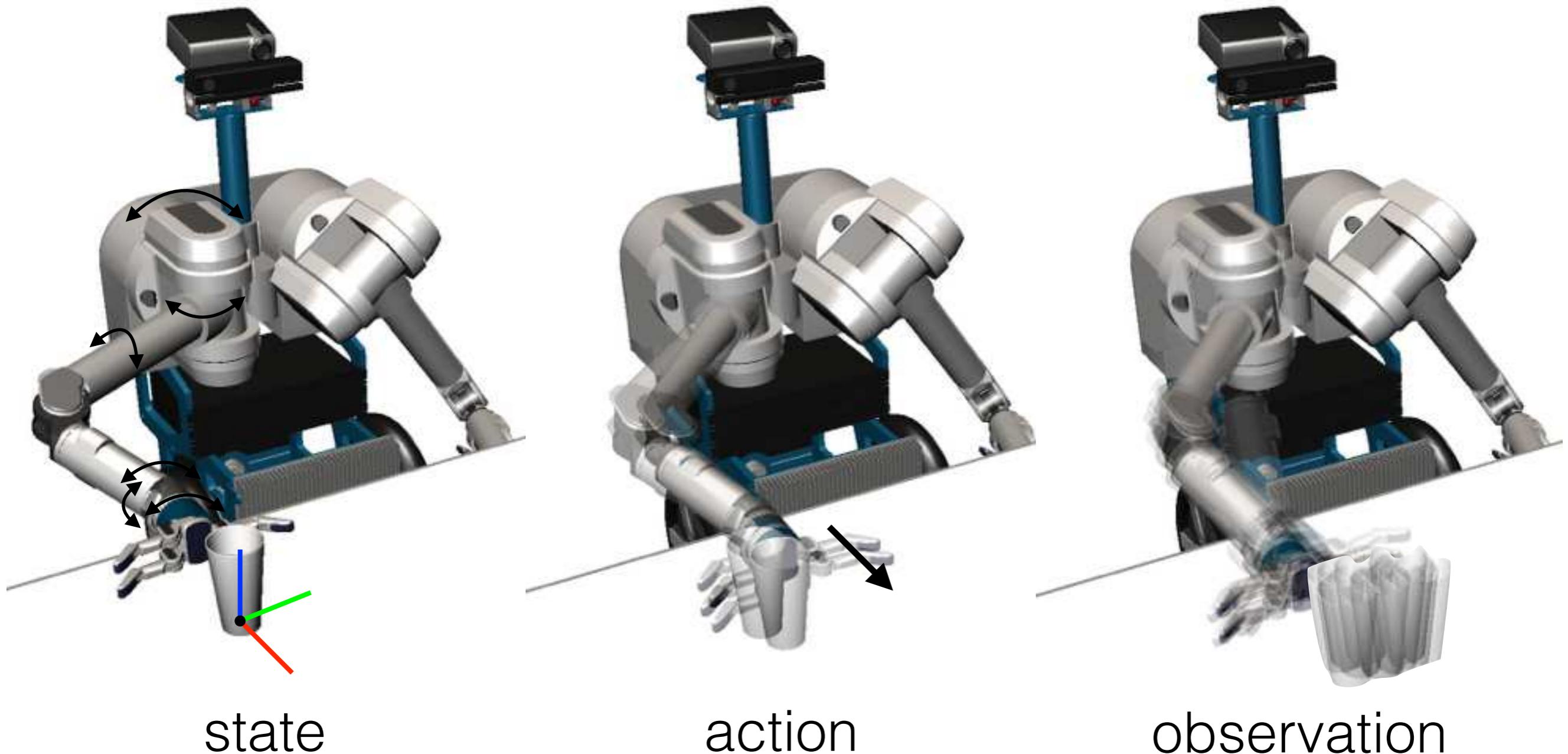
(two objects, nominal trajectory)

Robust Trajectory Selection



Convergent Planning

A brief introduction to POMDPs.



state

$$s = (q, x)$$

action

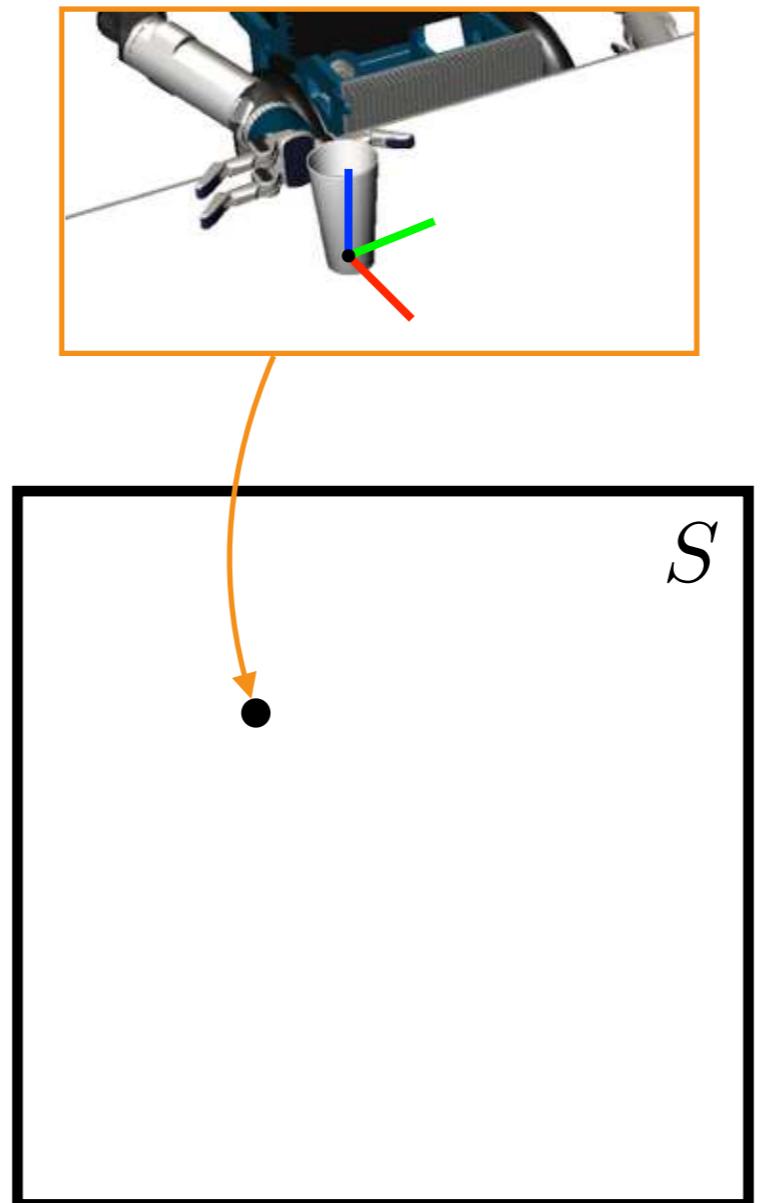
$$a = (\dot{q}, \Delta t)$$

$$T = p(s'|s, a)$$

observation

$$o = (o_q, o_c)$$

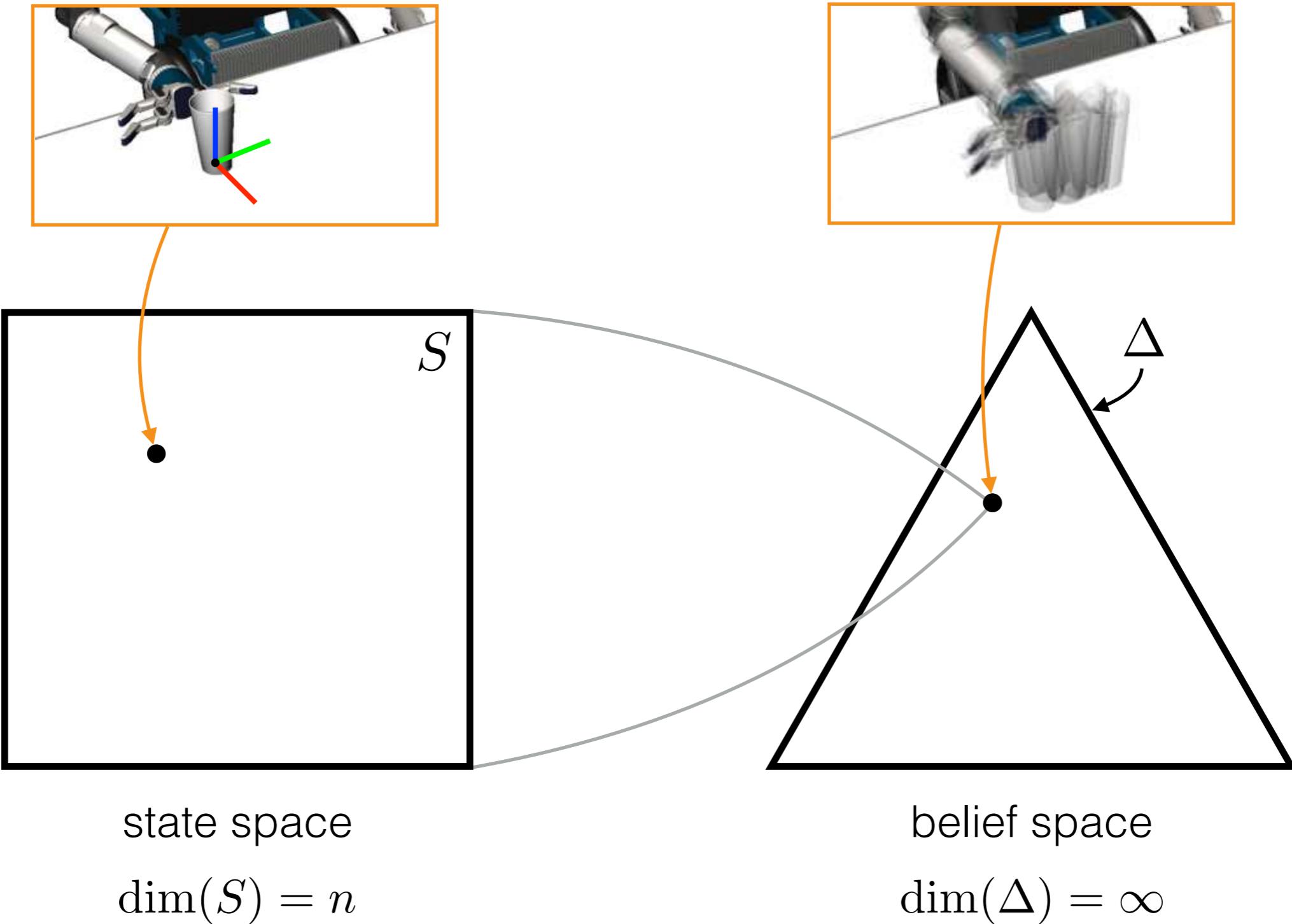
$$\Omega = p(o|s, a)$$



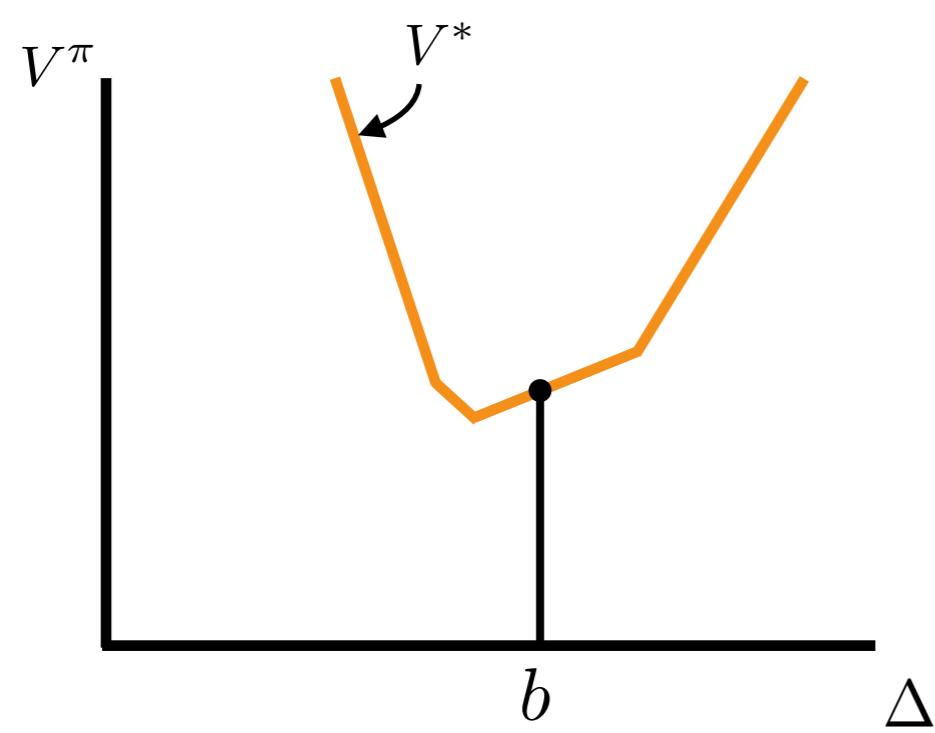
state space

$$\dim(S) = n$$

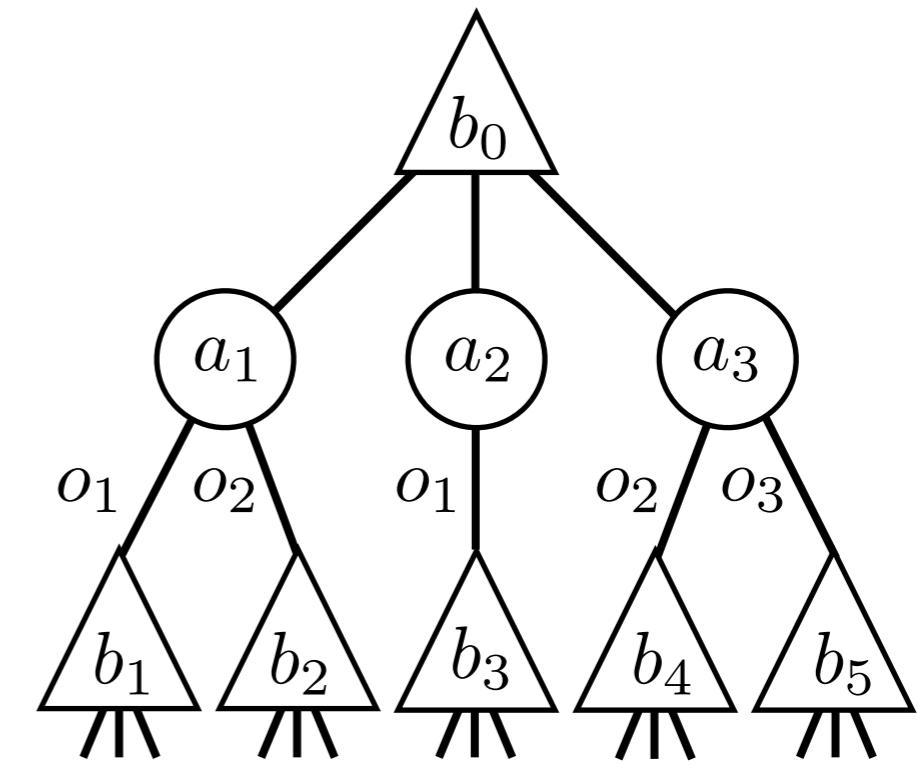
Planning in Belief Space



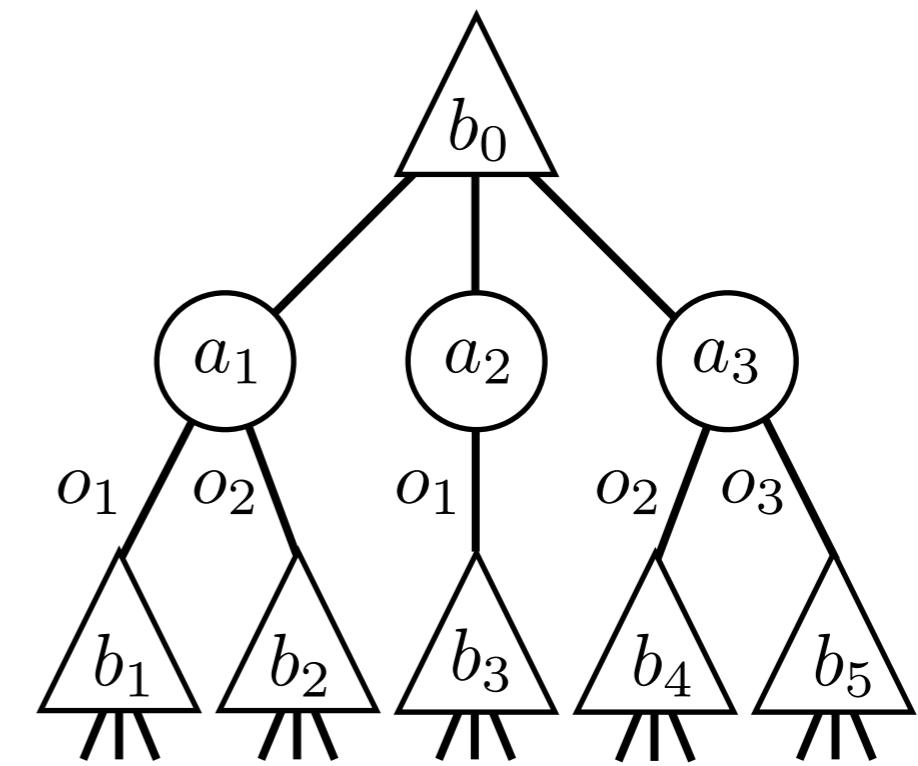
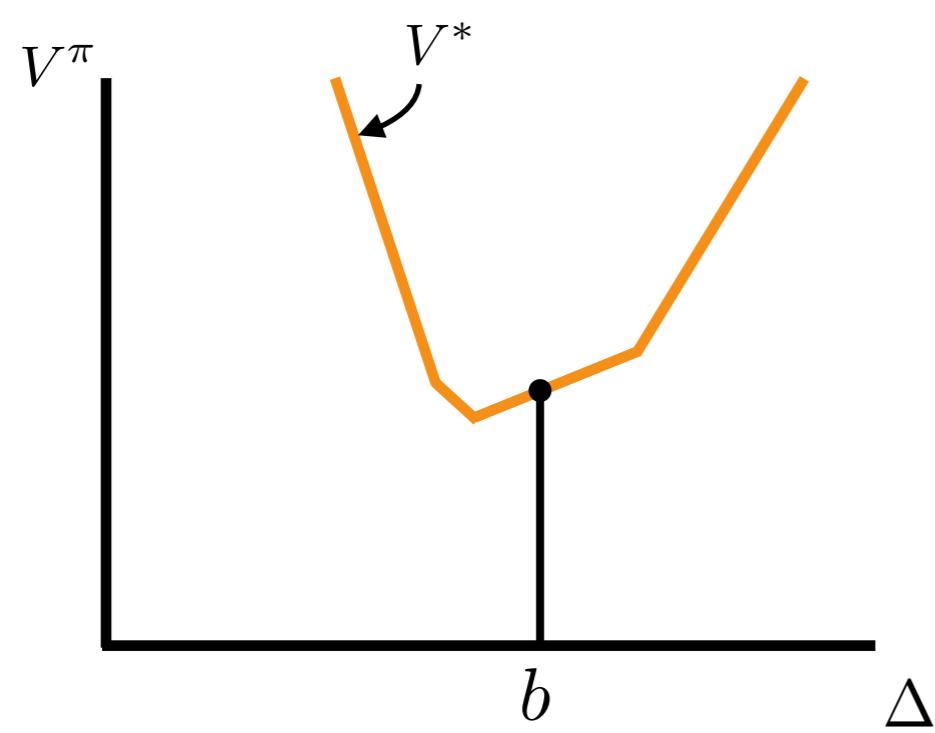
Planning in Belief Space



Offline Planning
Point-Based Methods



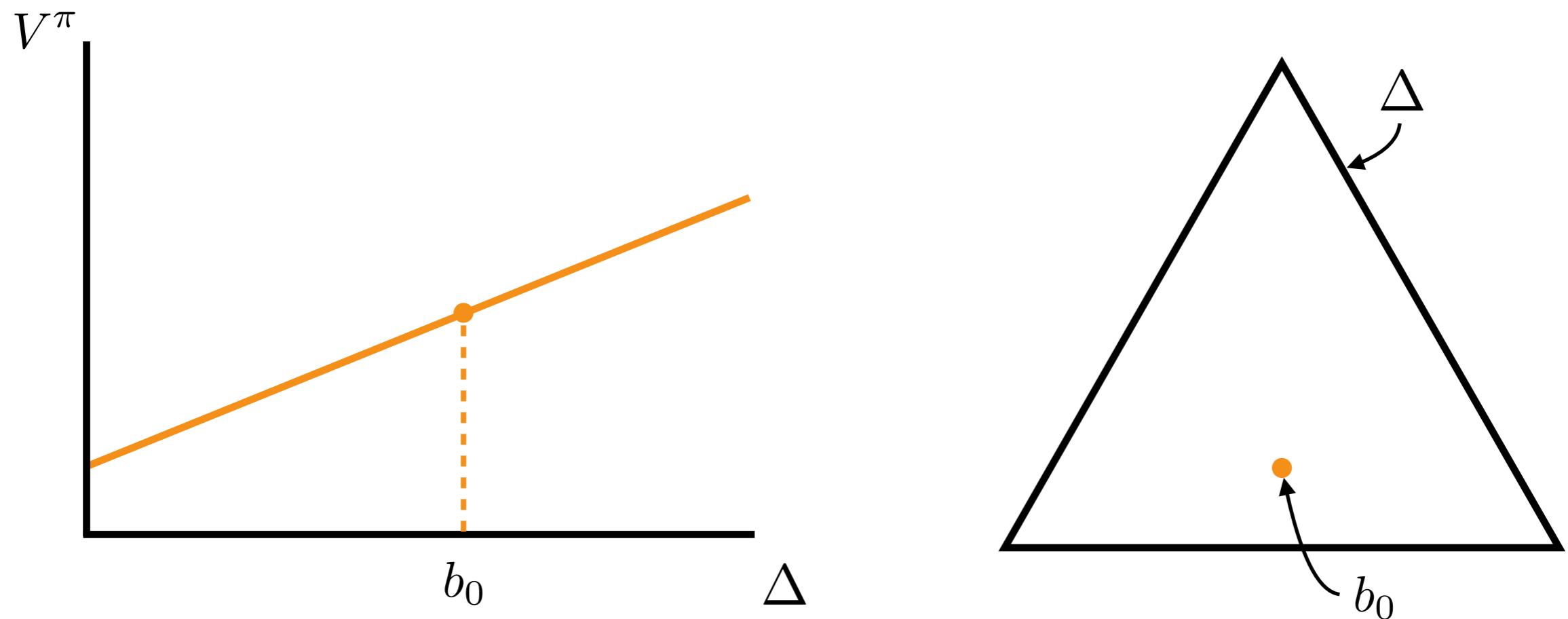
Online Planning



Offline Planning
Point-Based Methods

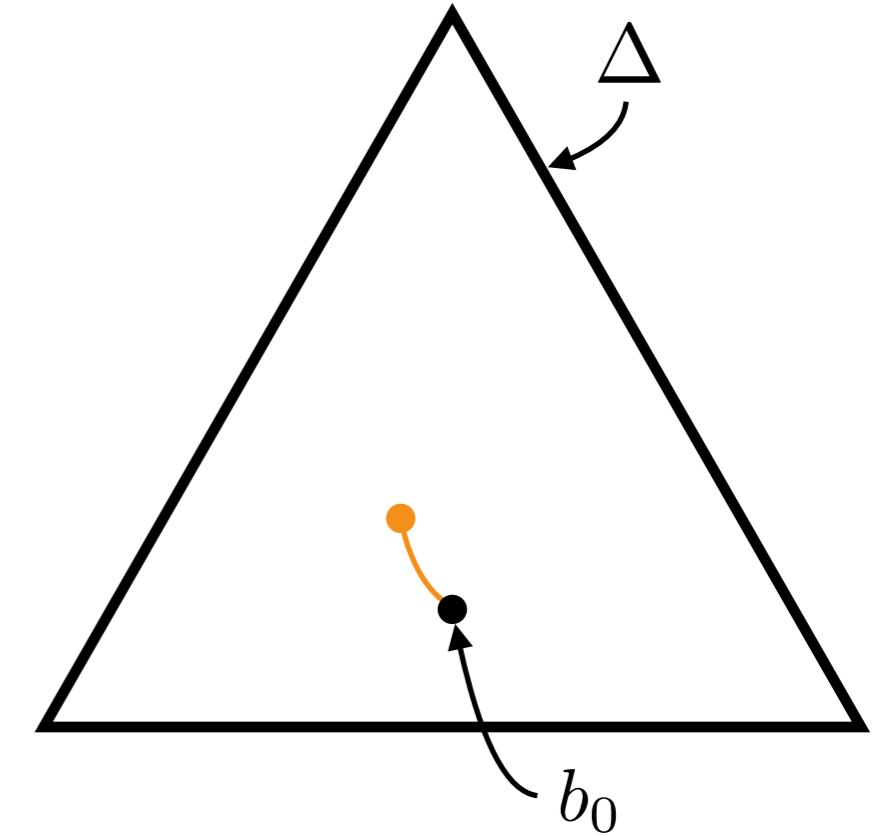
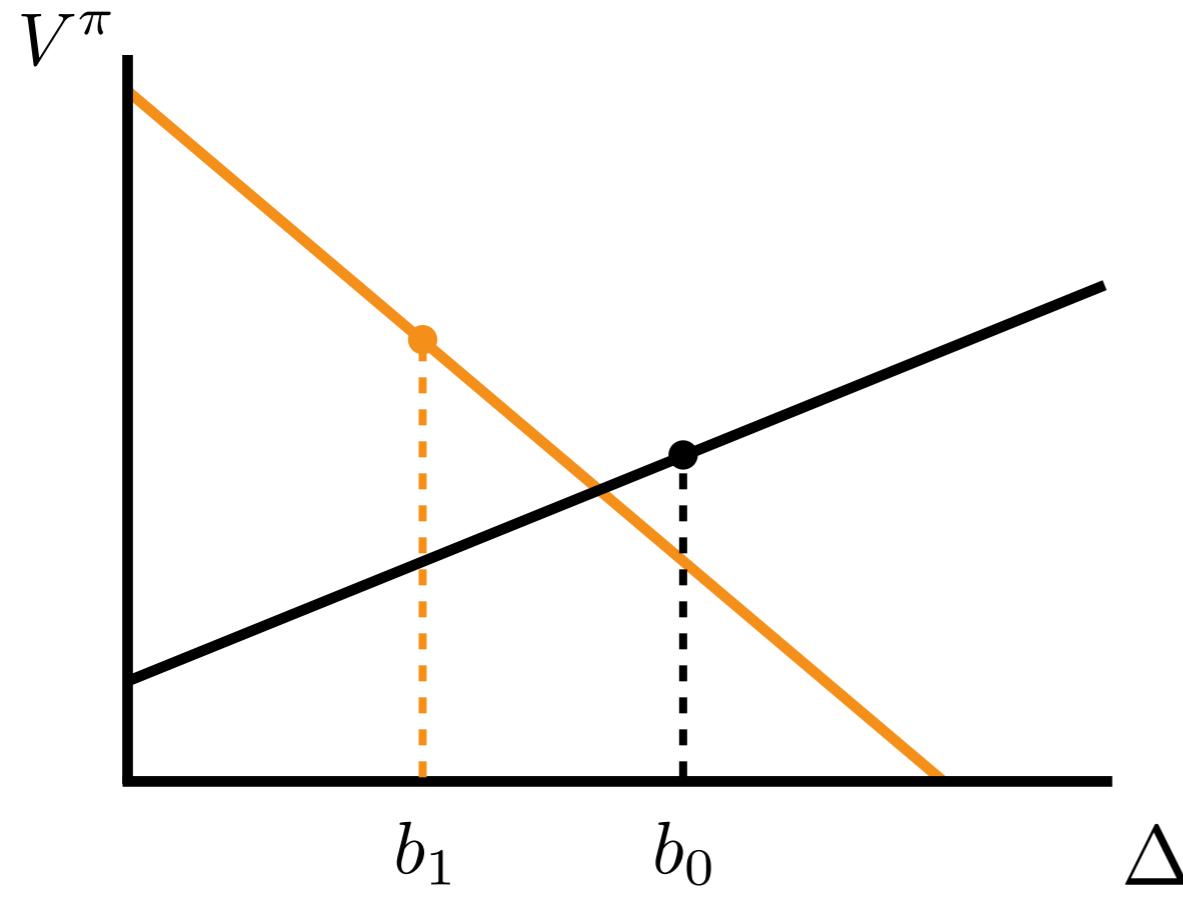
Online Planning

$$V^\pi = \sum_{t=1}^{\infty} \gamma^t R(s_t, a_t)$$



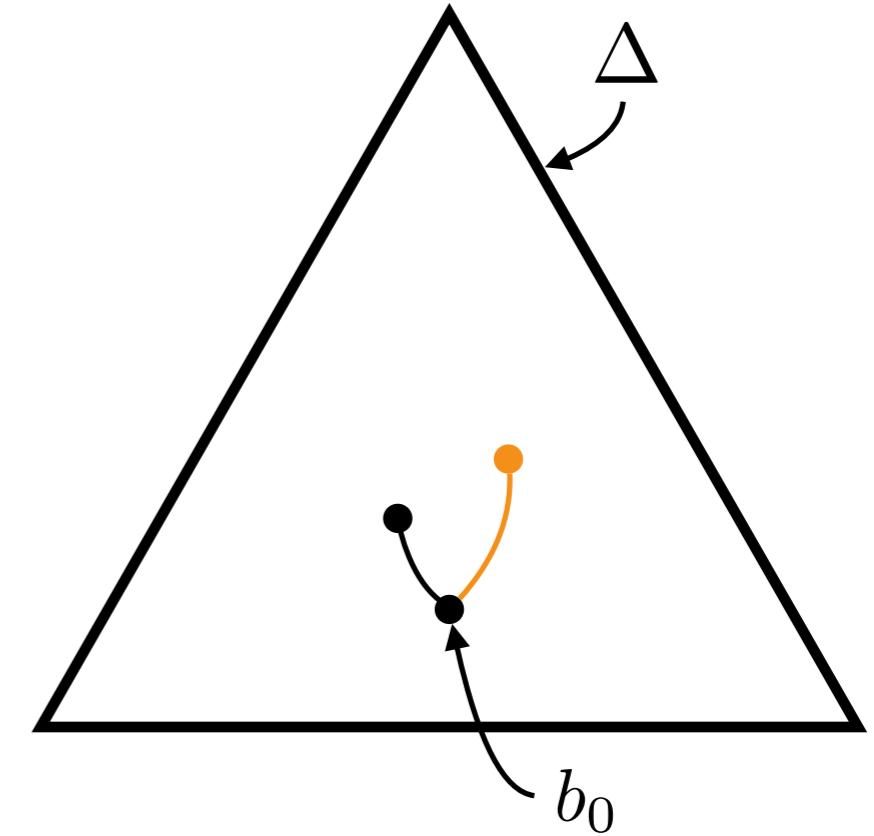
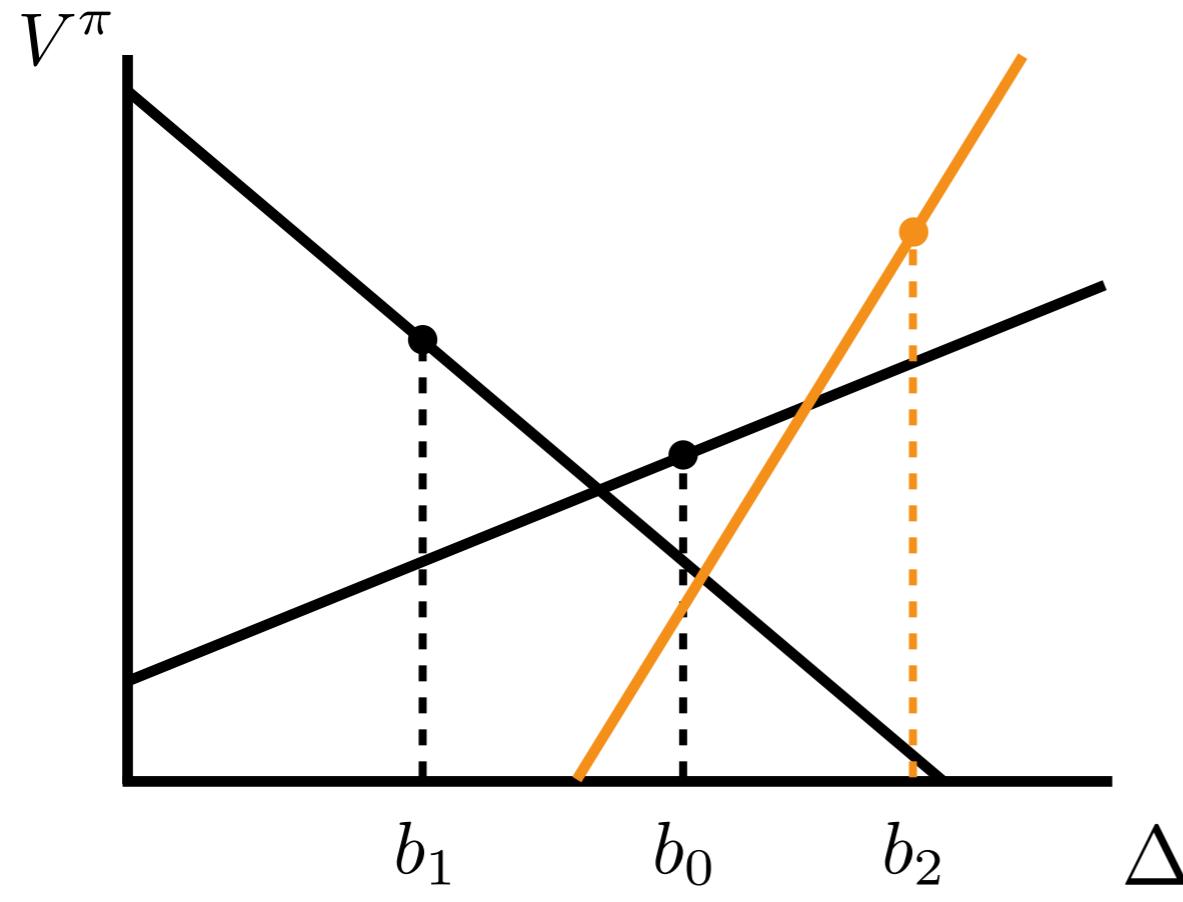
Point-based solvers

$$V^\pi = \sum_{t=1}^{\infty} \gamma^t R(s_t, a_t)$$



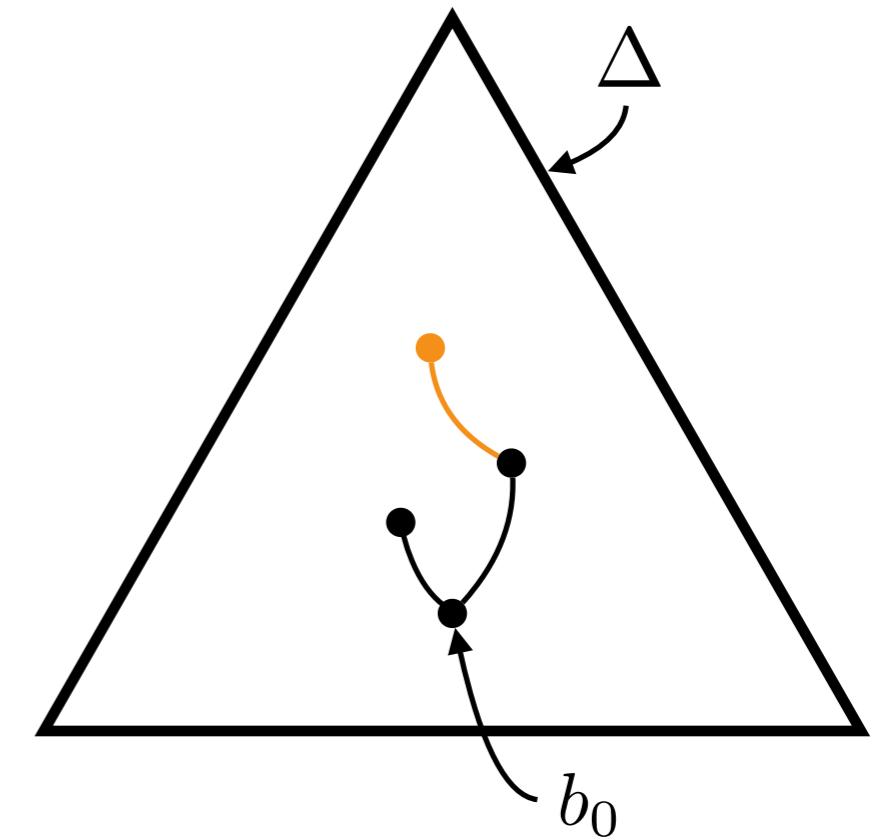
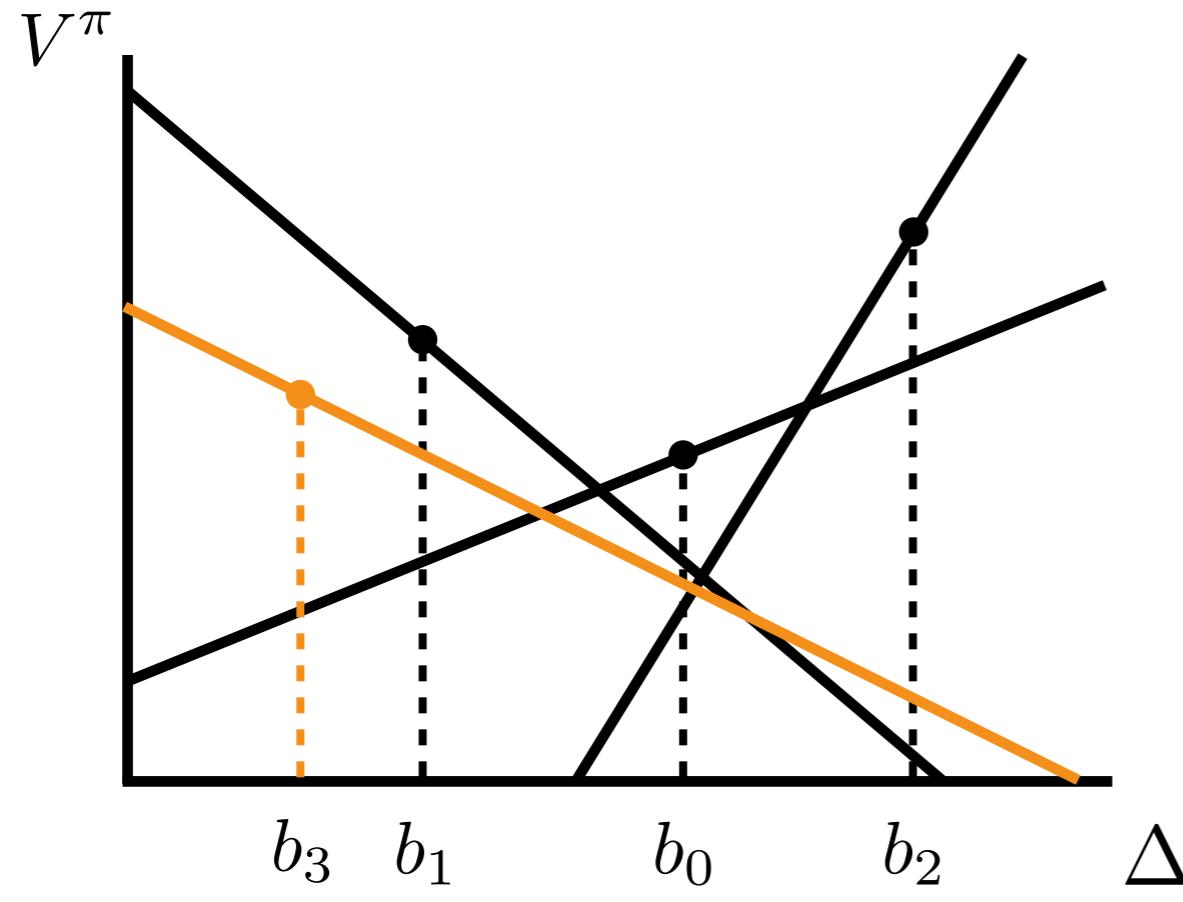
Point-based solvers

$$V^\pi = \sum_{t=1}^{\infty} \gamma^t R(s_t, a_t)$$



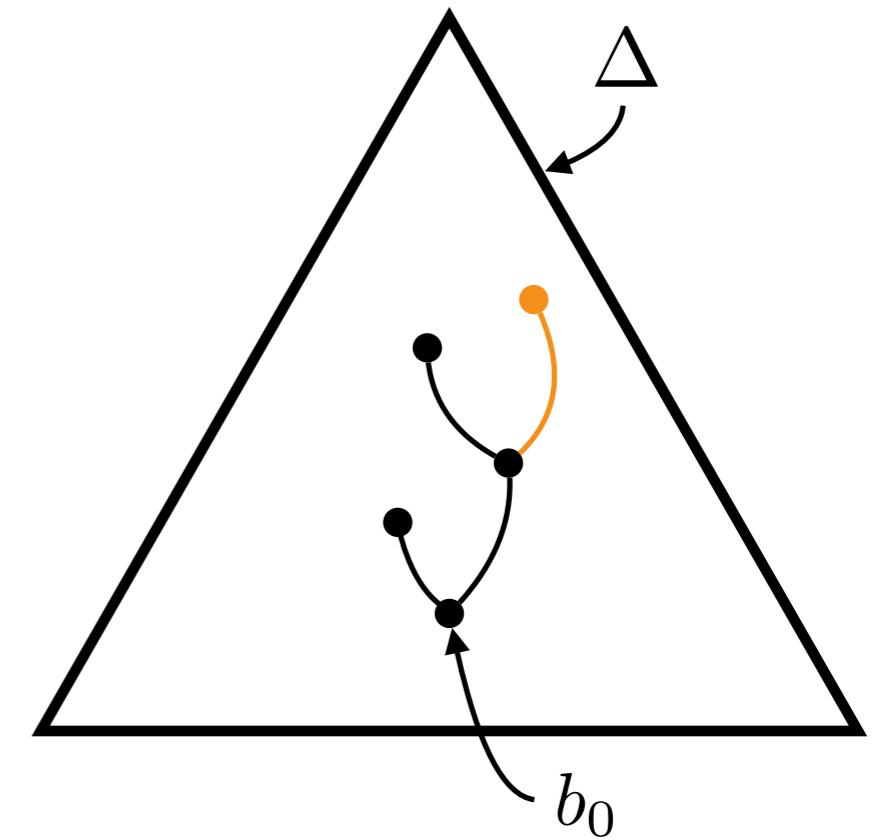
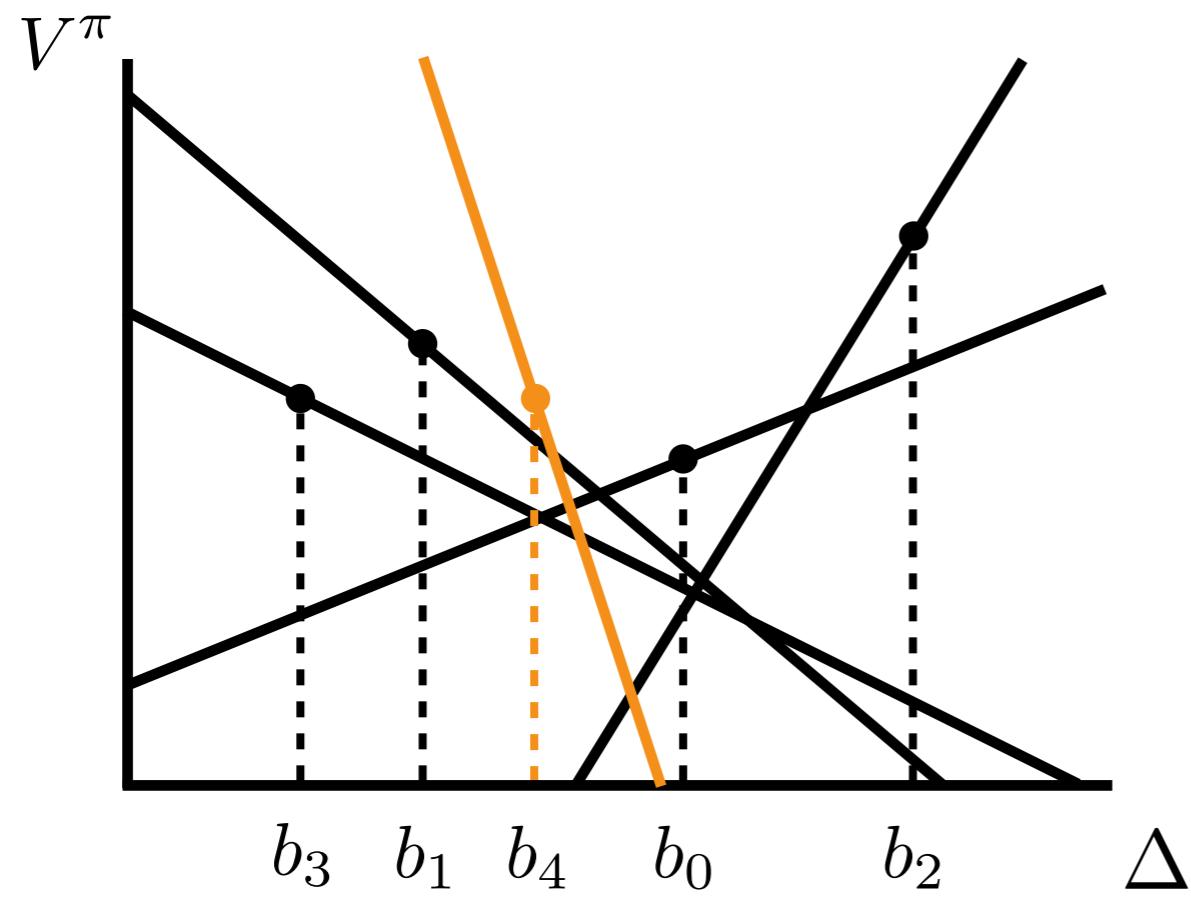
Point-based solvers

$$V^\pi = \sum_{t=1}^{\infty} \gamma^t R(s_t, a_t)$$



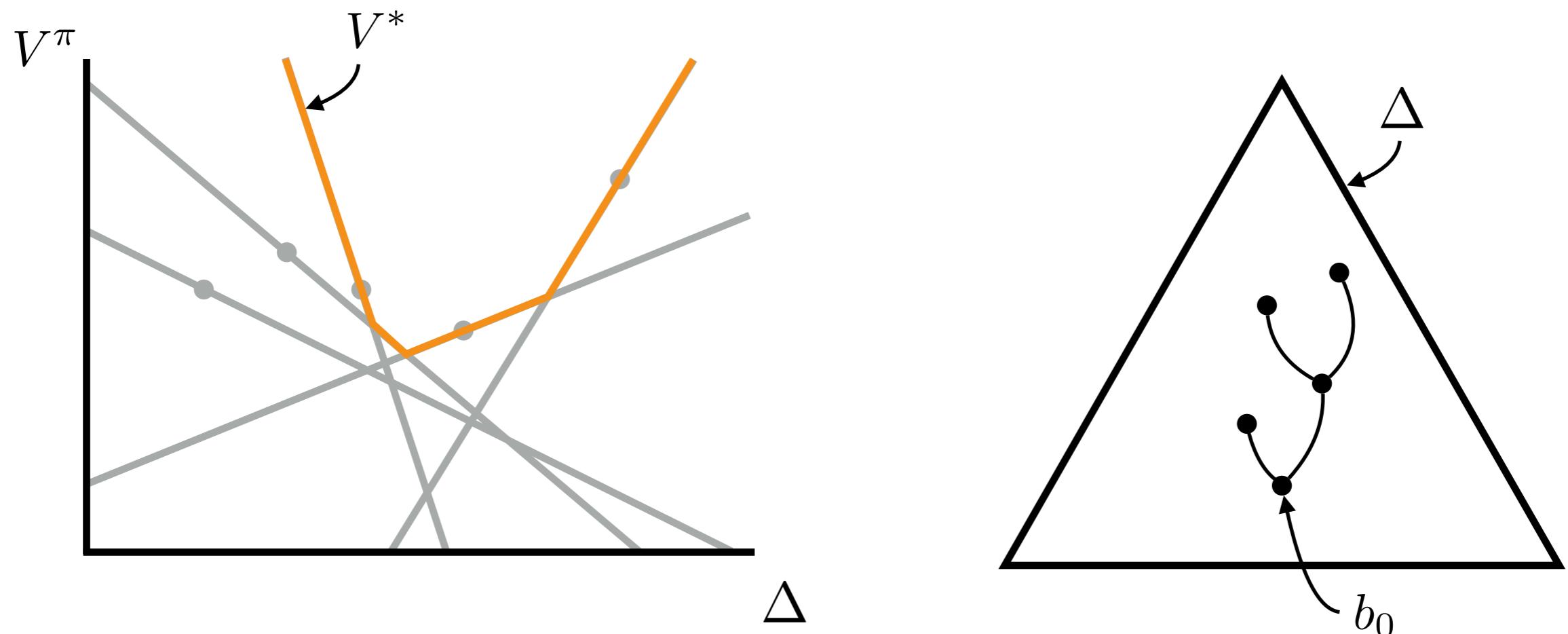
Point-based solvers

$$V^\pi = \sum_{t=1}^{\infty} \gamma^t R(s_t, a_t)$$

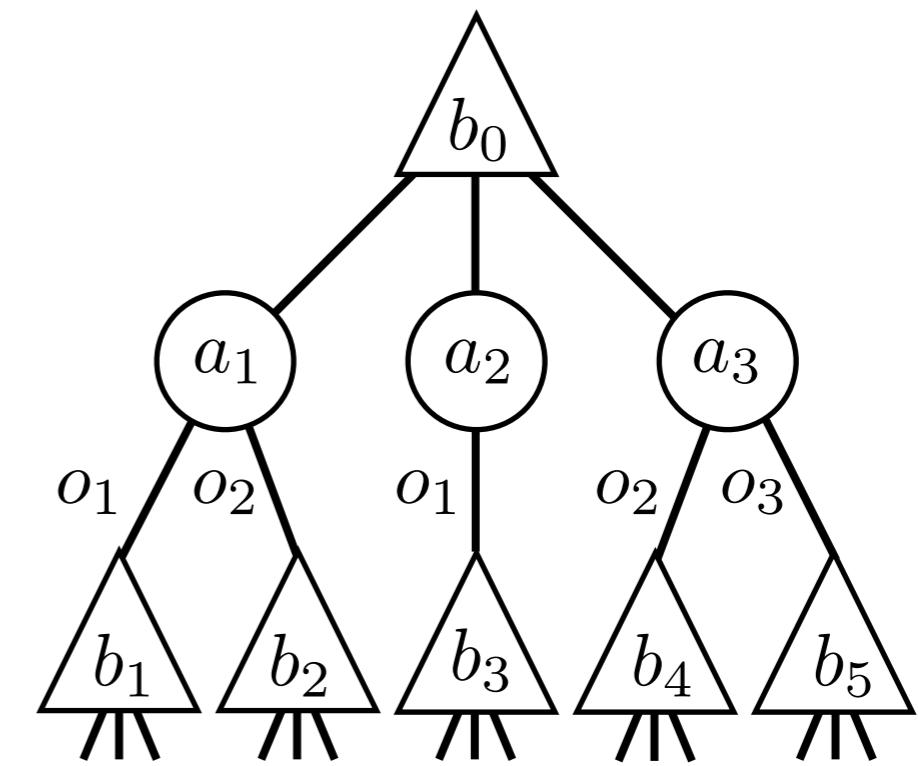
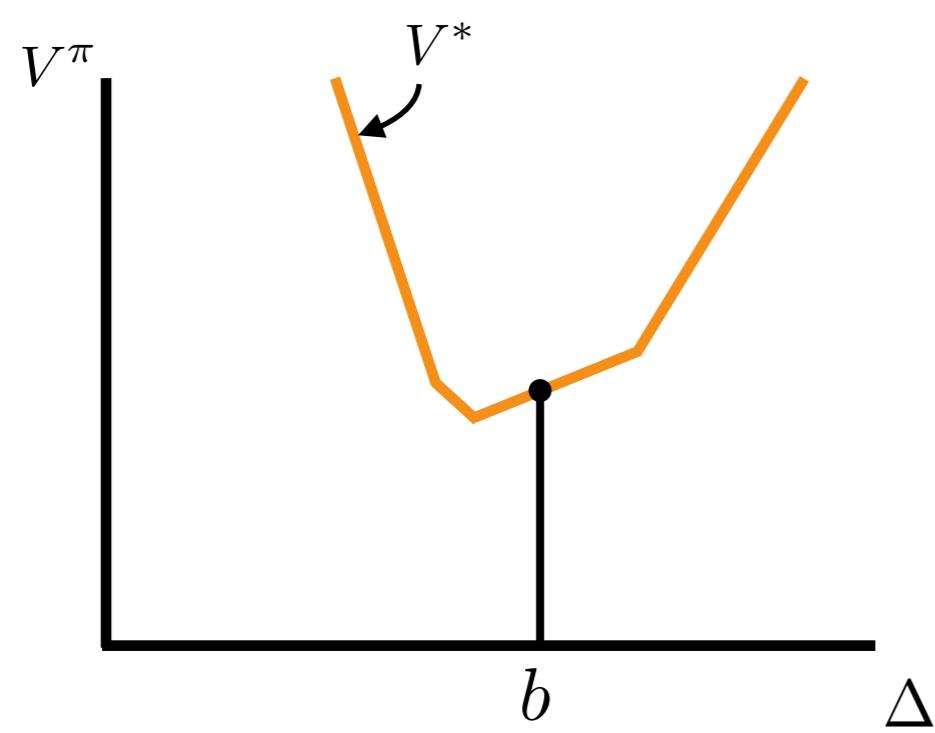


Point-based solvers

$$\pi^* = \arg \max_{\pi} V^{\pi} [b(s_0)]$$

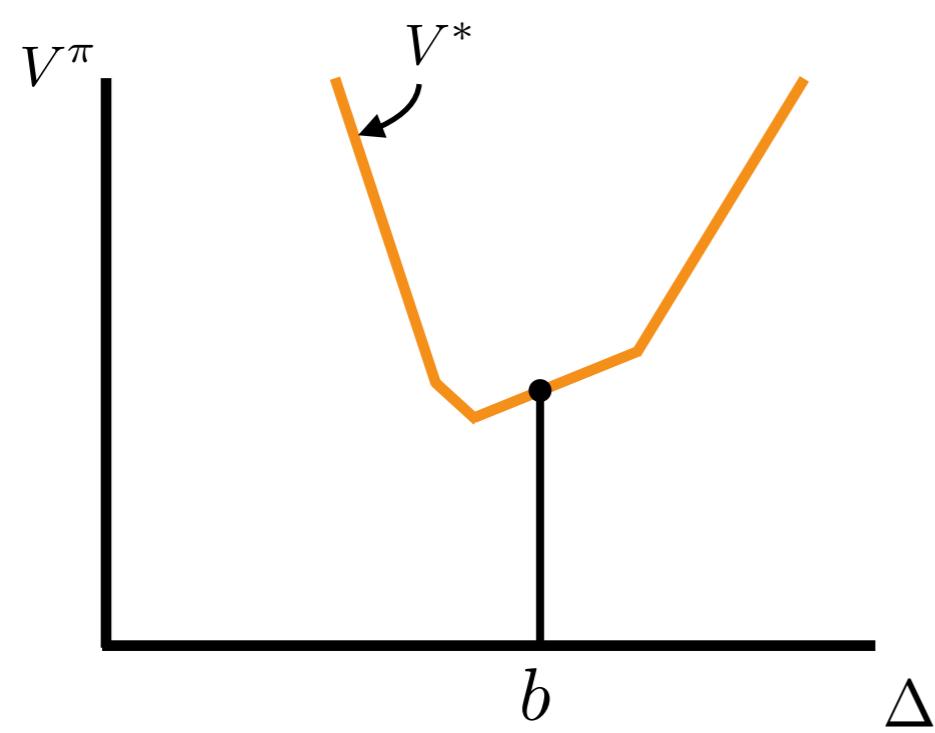


Point-based solvers

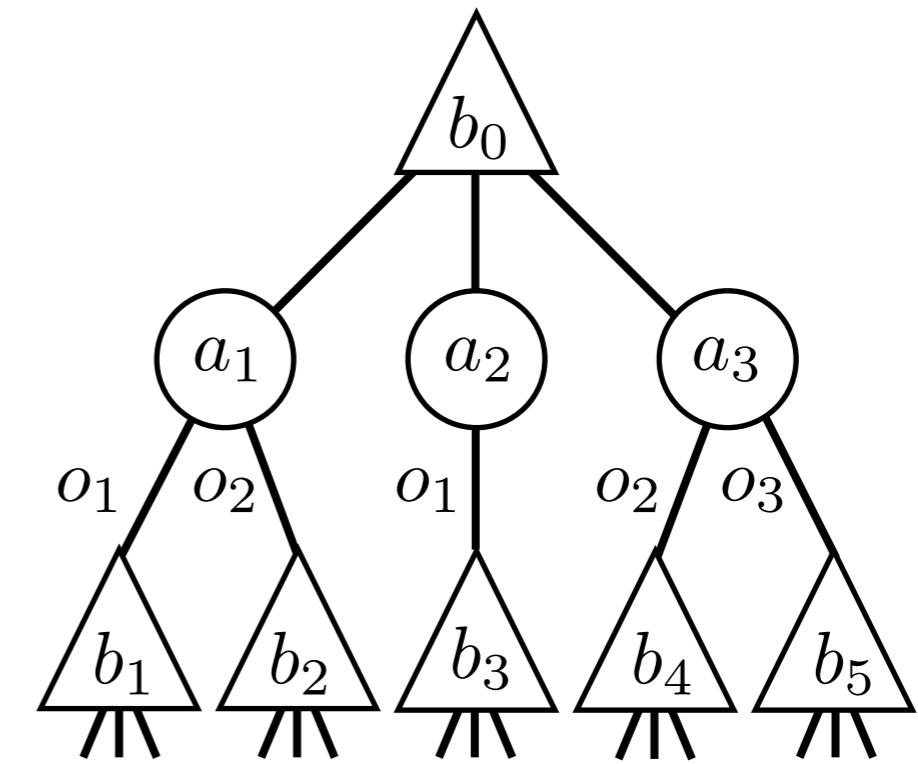


Offline Planning
Point-Based Methods

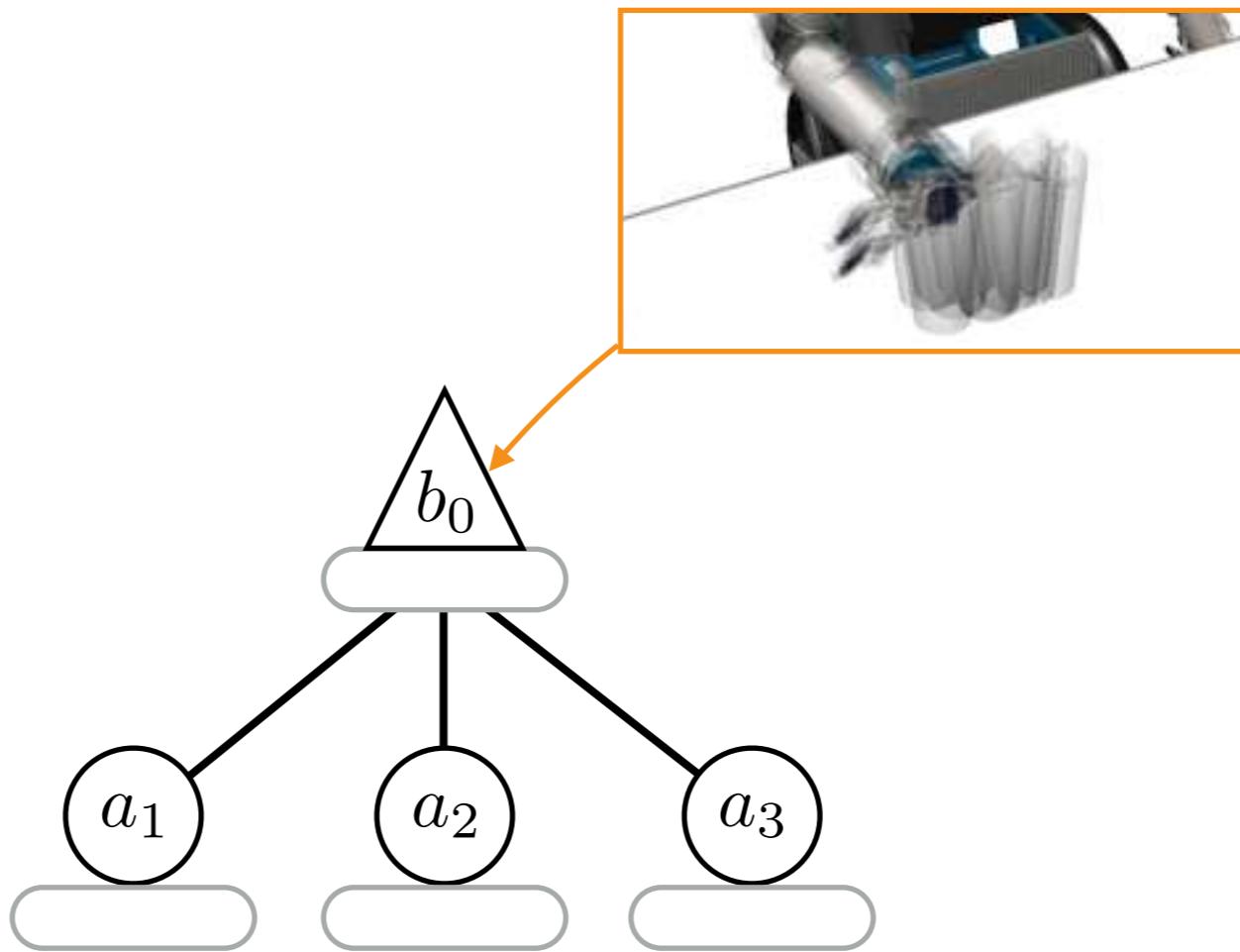
Online Planning

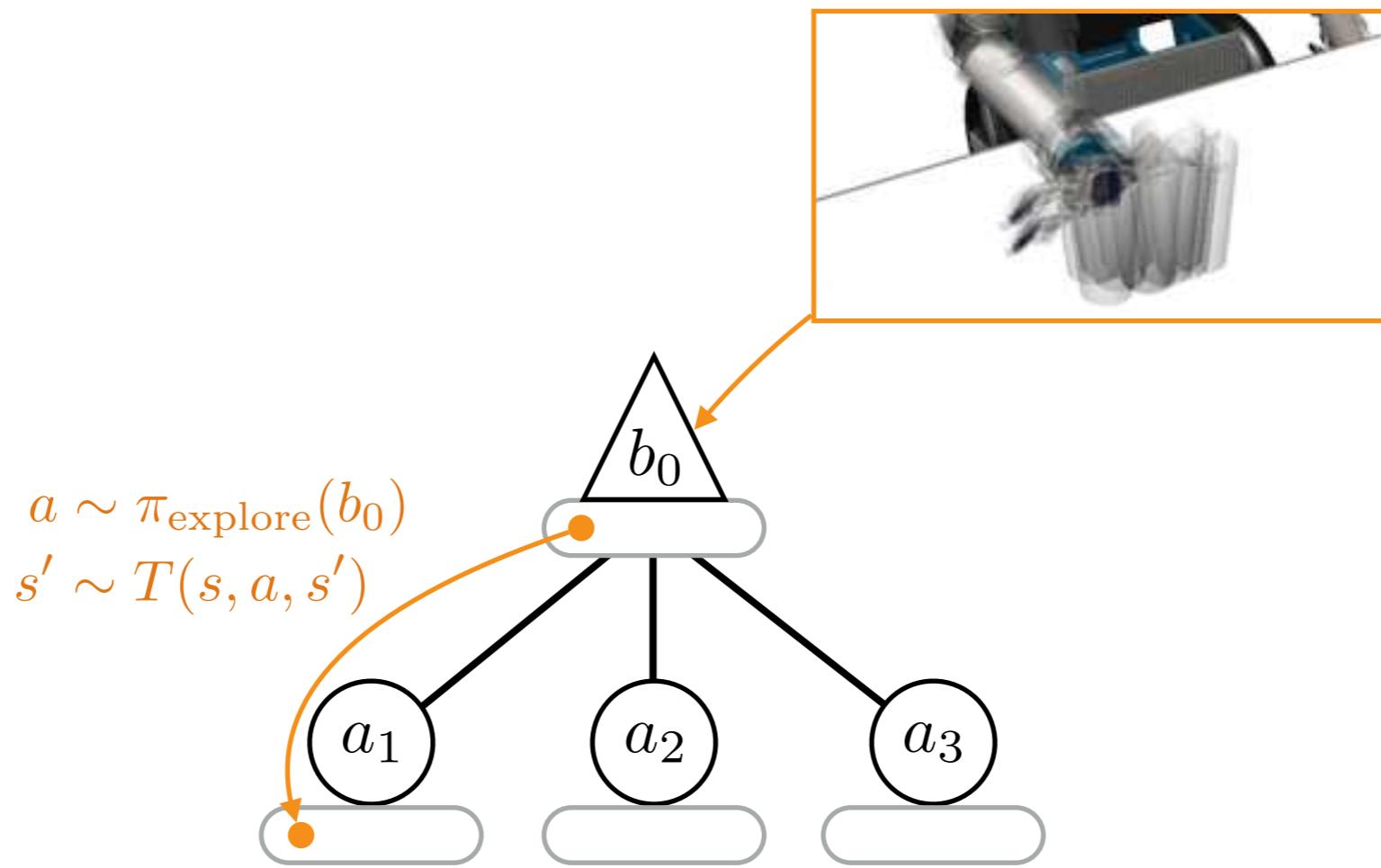


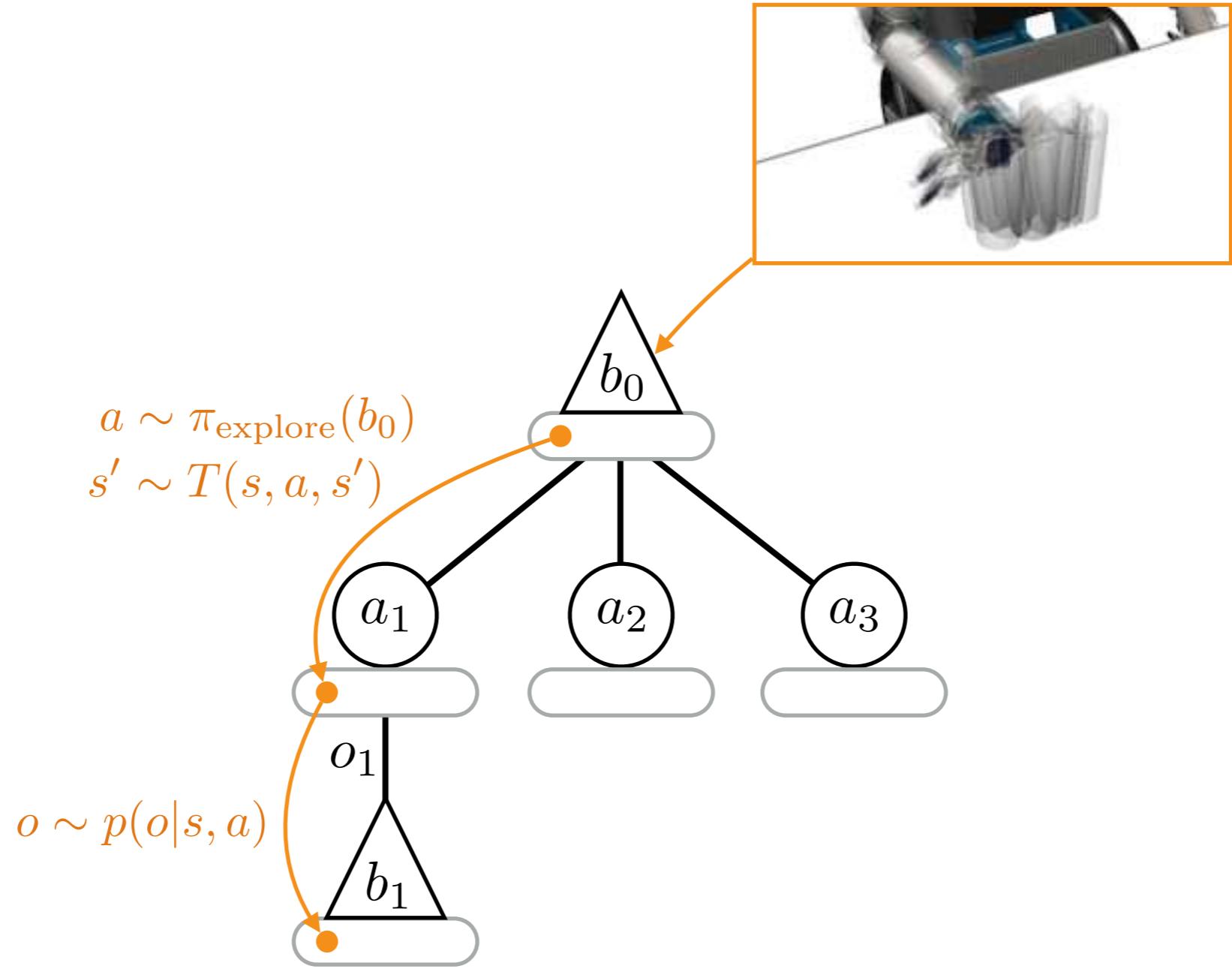
Offline Planning
Point-Based Methods

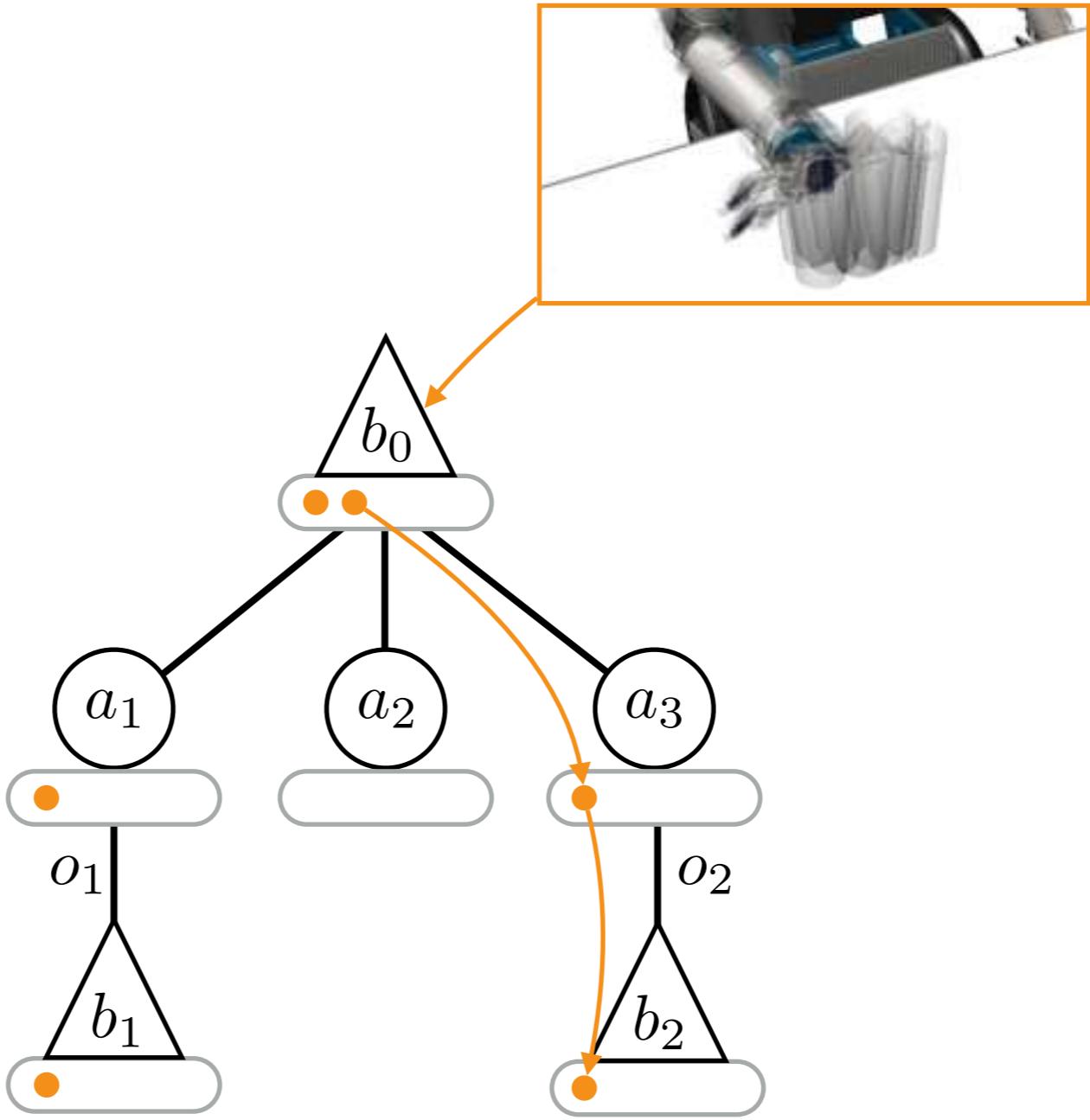


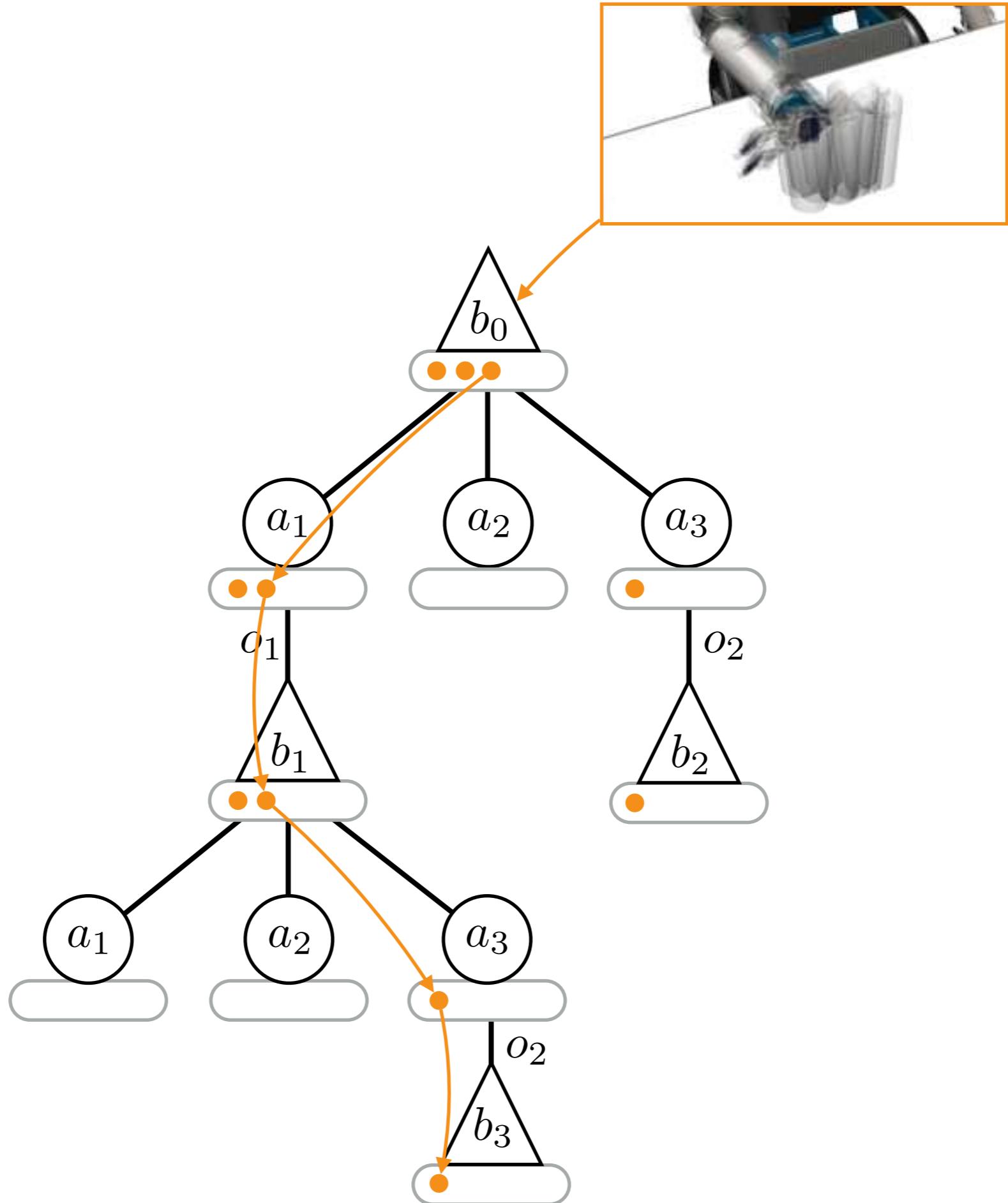
Online Planning

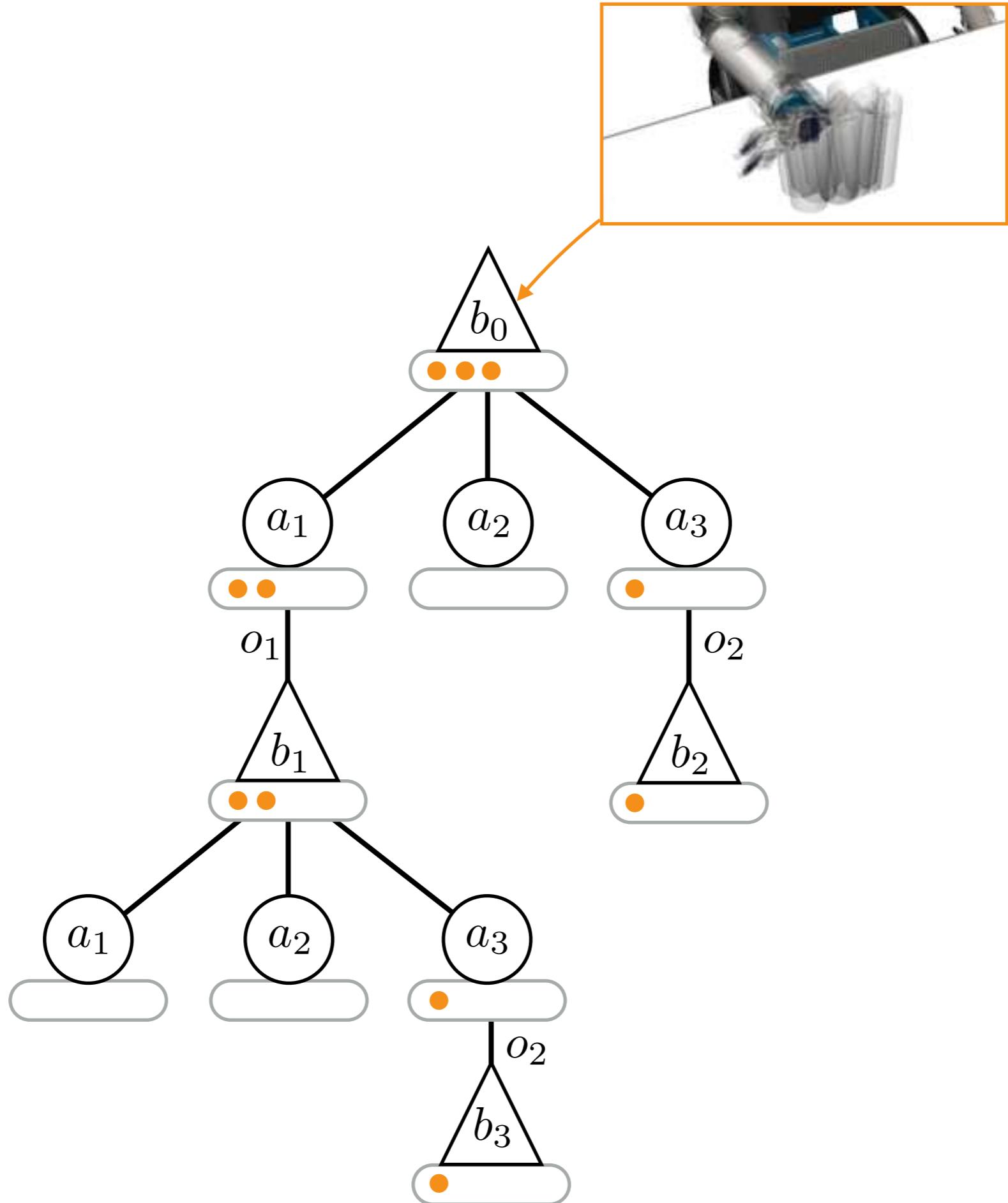




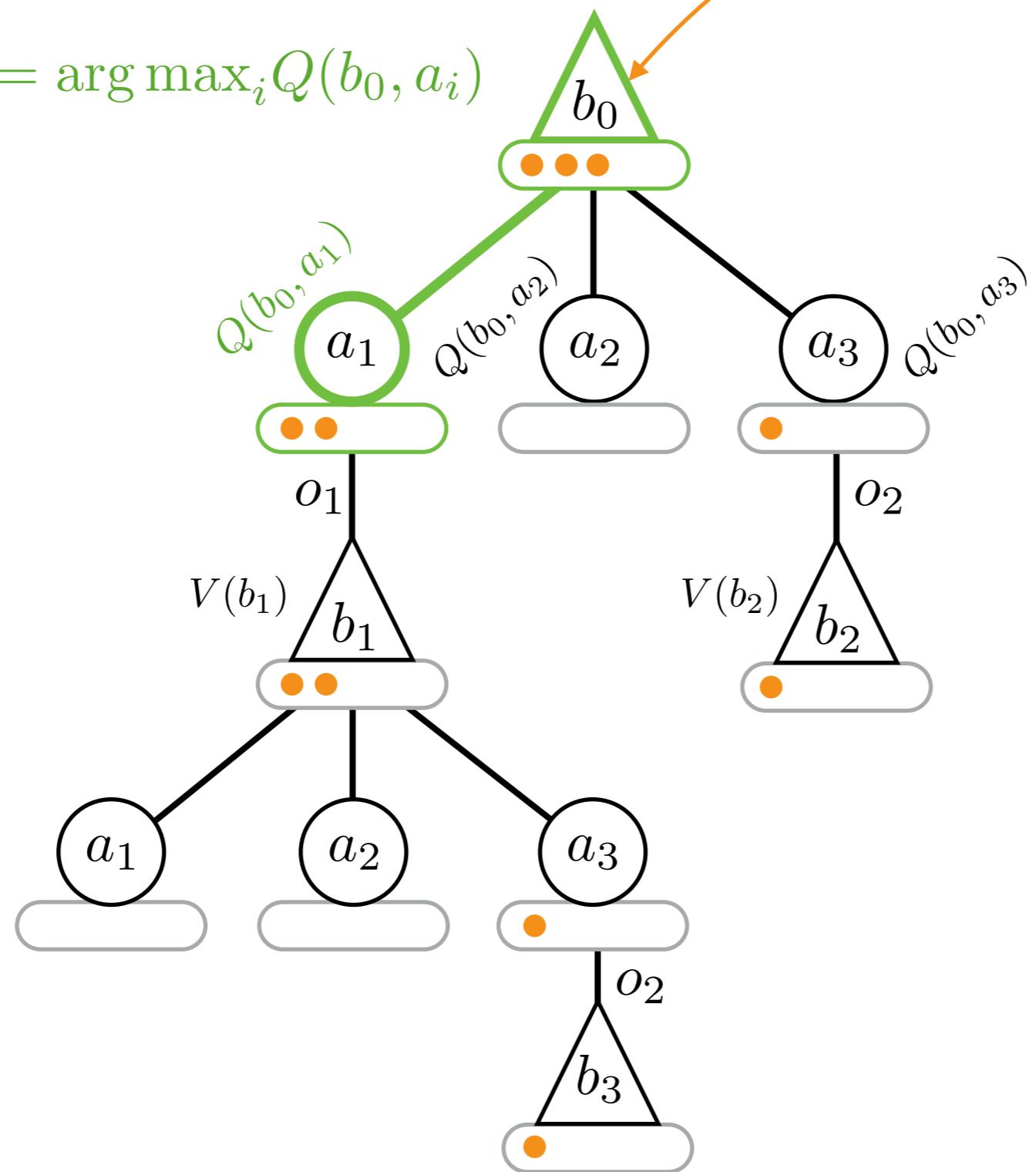


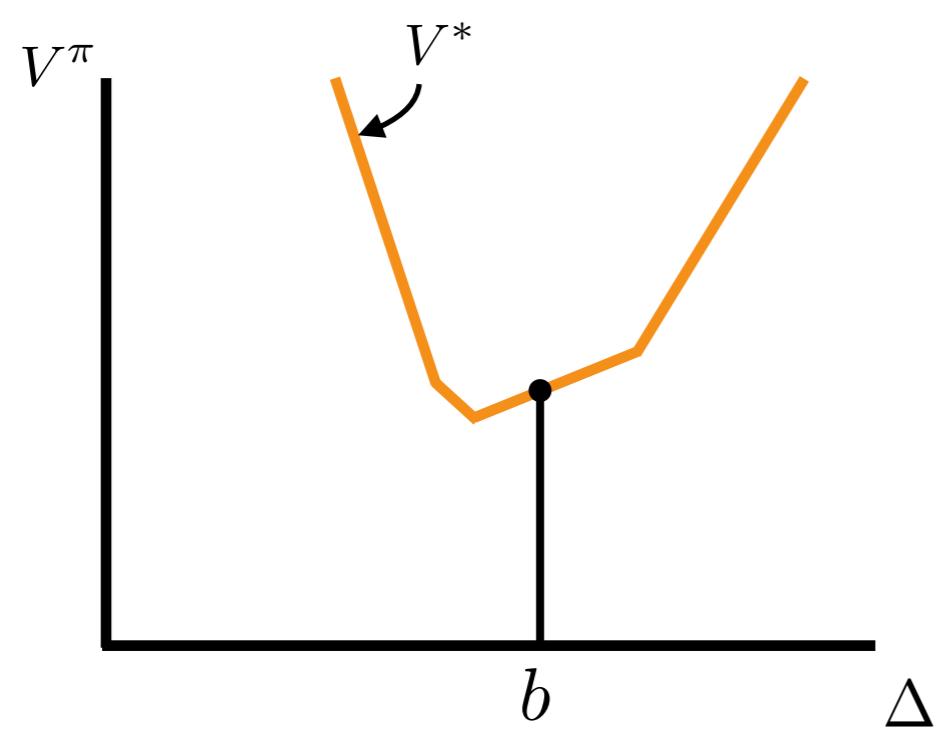




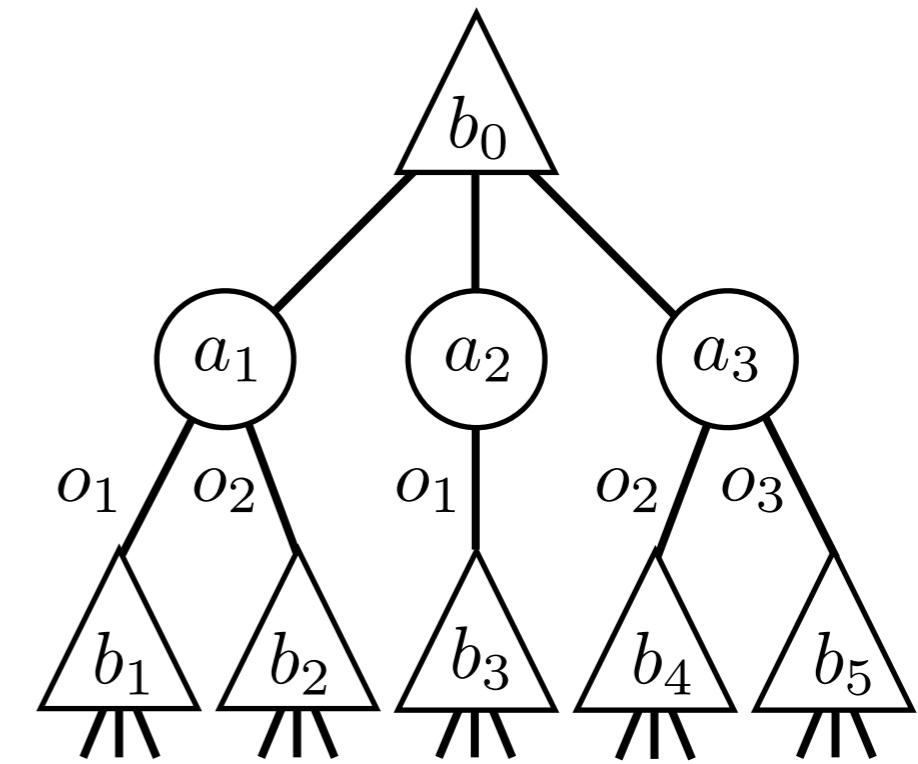


$$a^* = \arg \max_i Q(b_0, a_i)$$

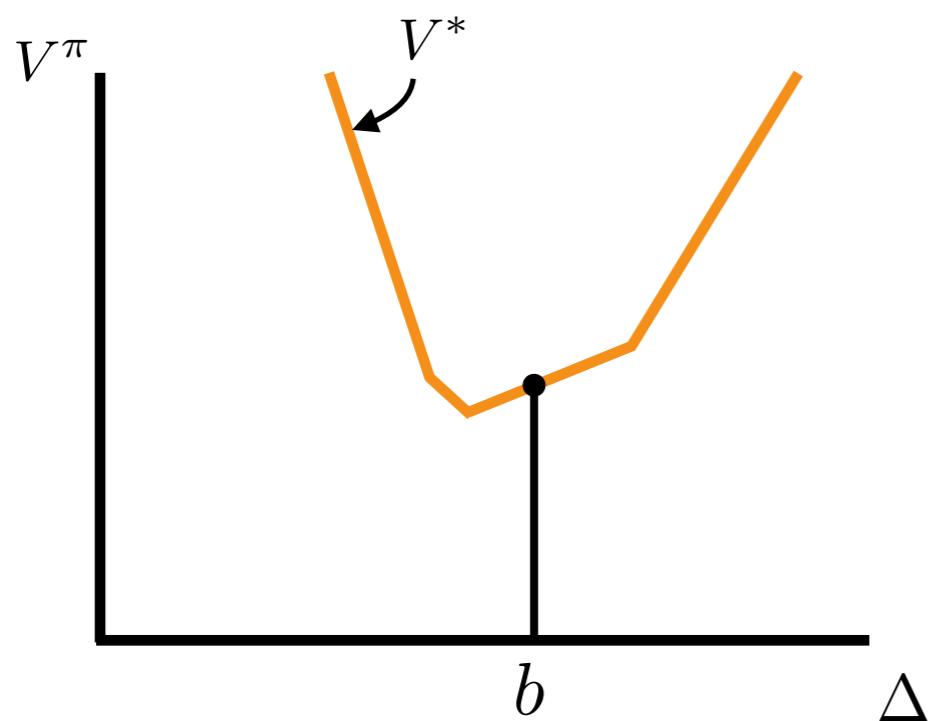




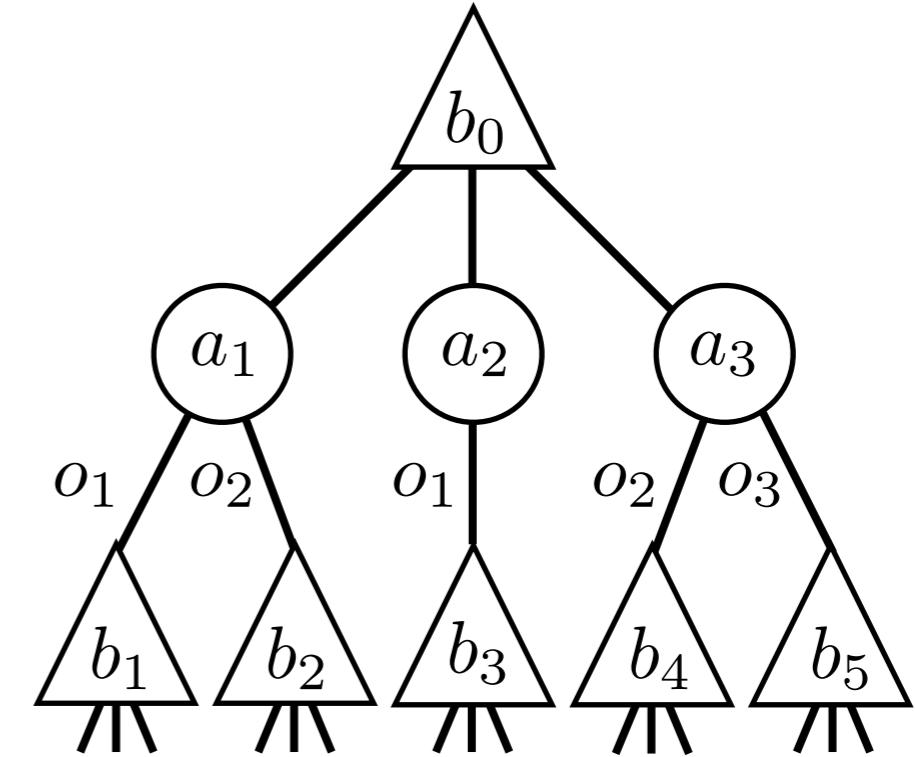
Offline Planning
Point-Based Methods



Online Planning



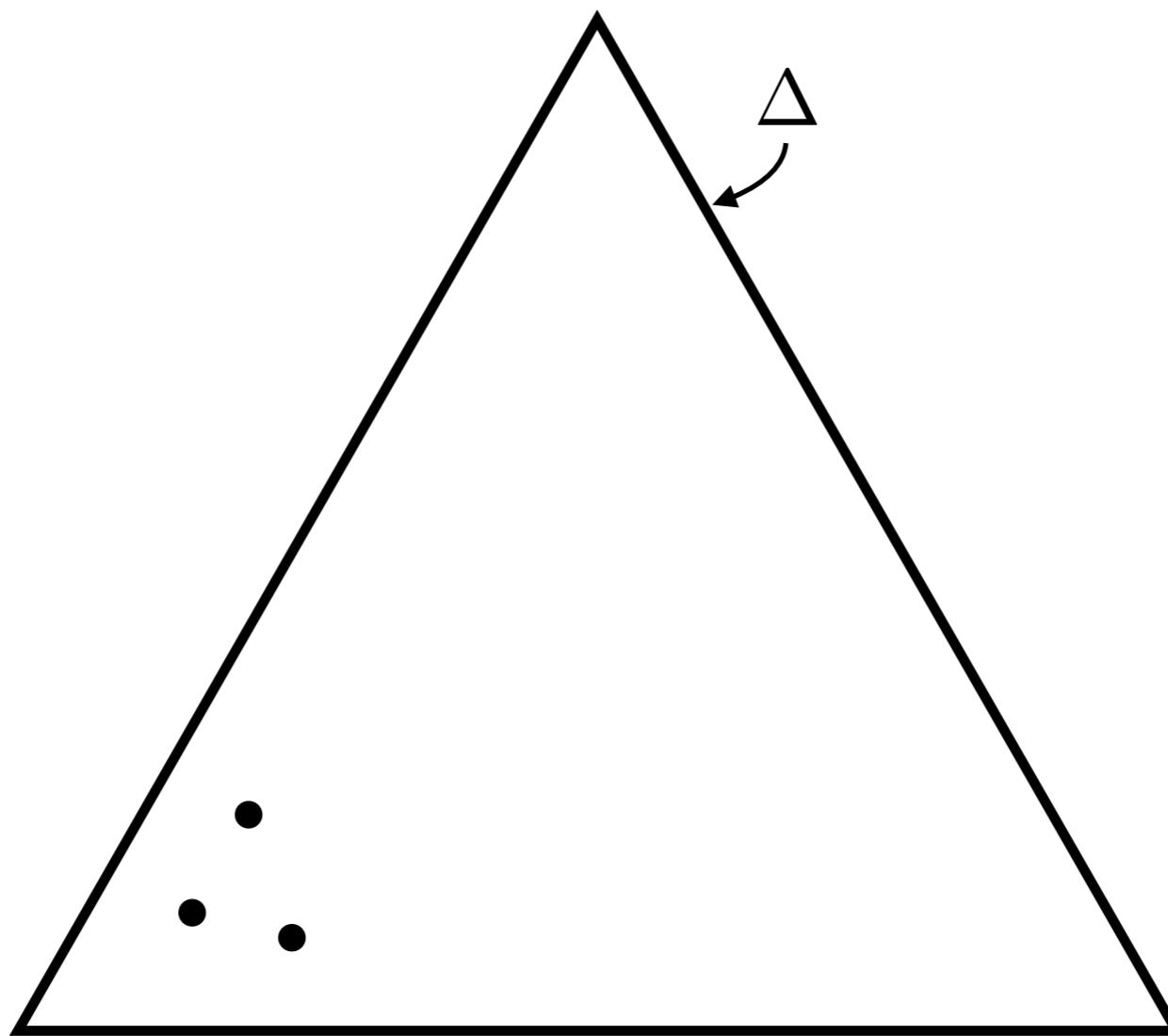
Offline Planning
Point-Based Methods



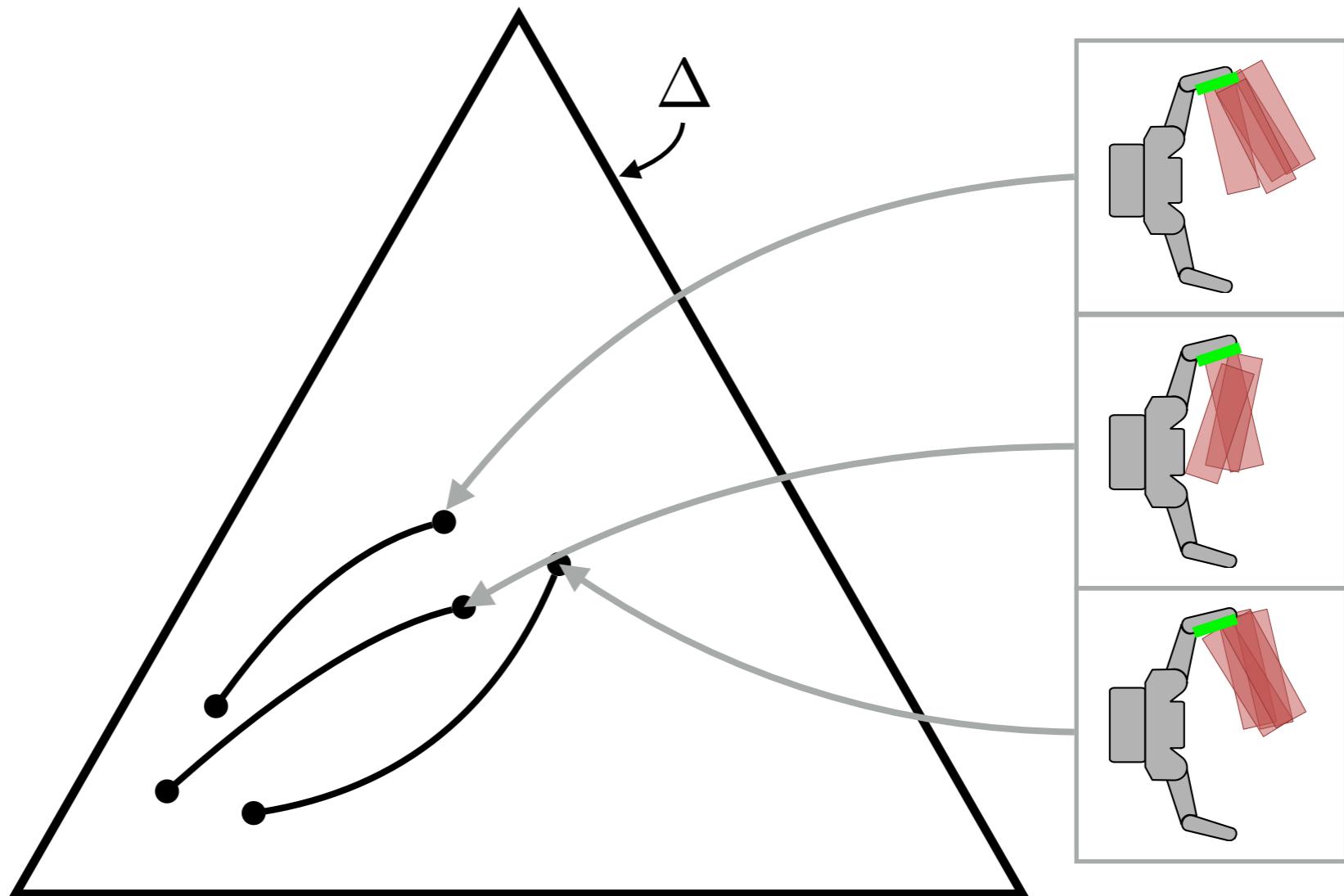
Online Planning

Heuristics / Bounds

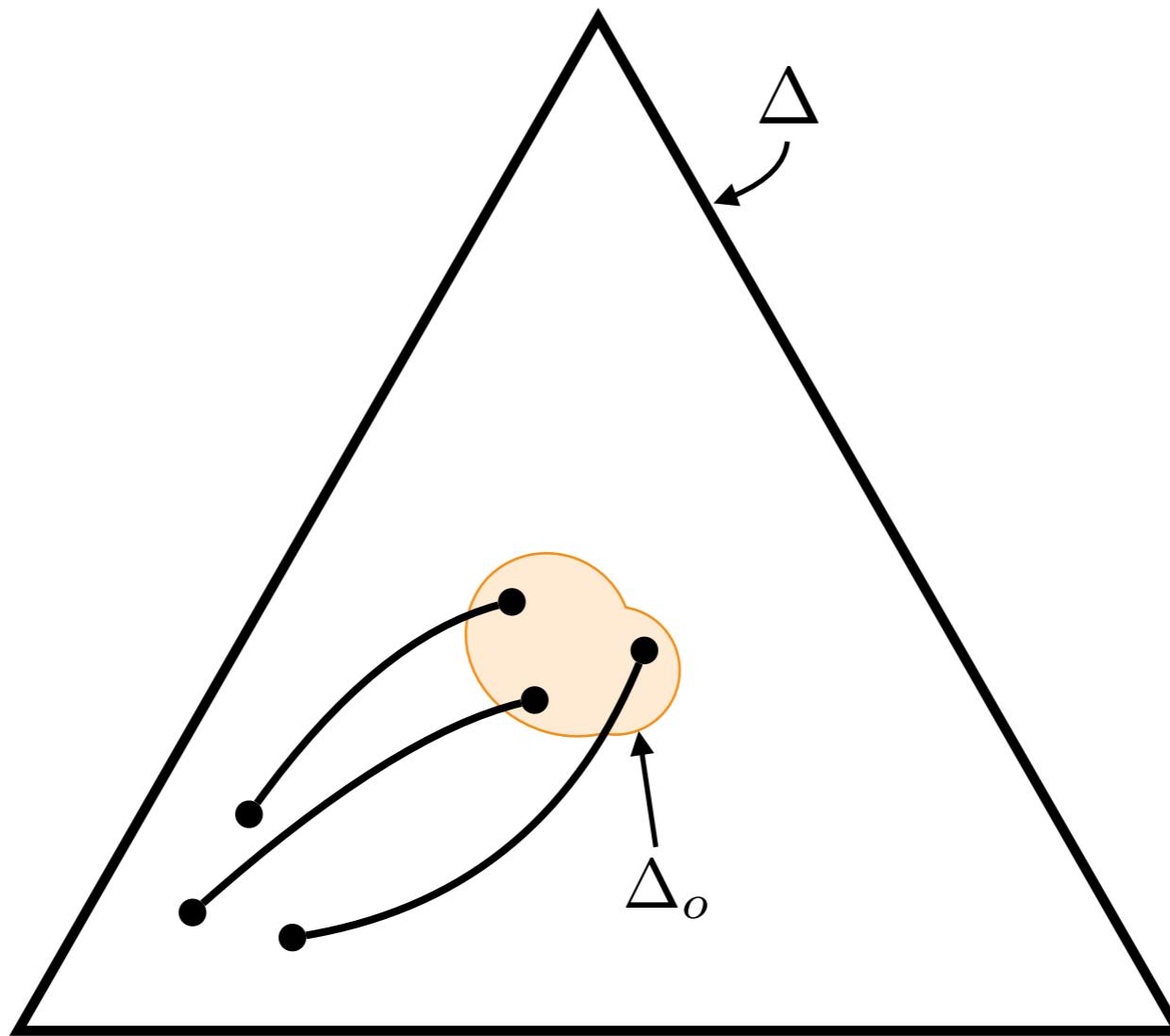
Combine online and offline planning.



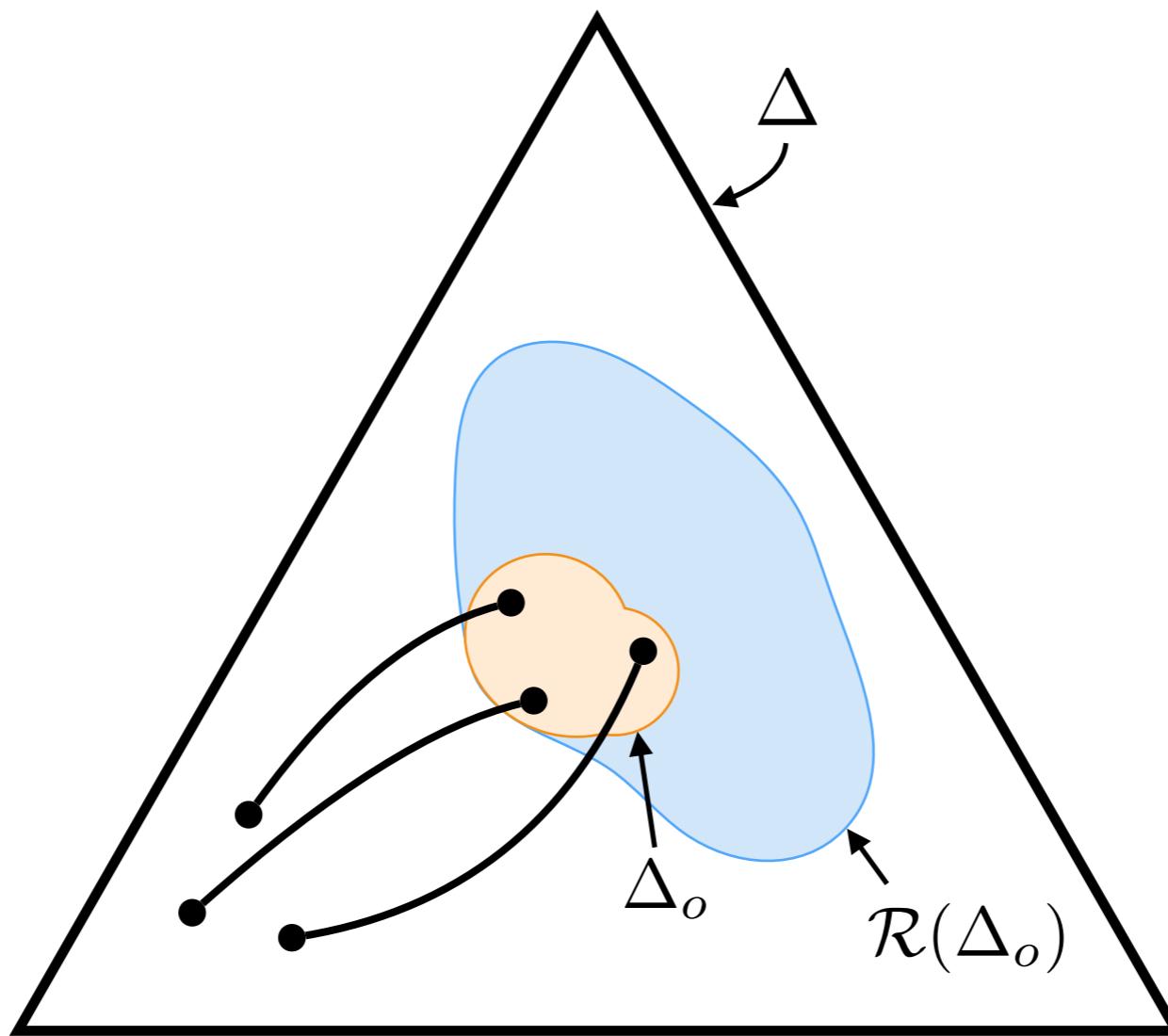
The post-contact belief space is small



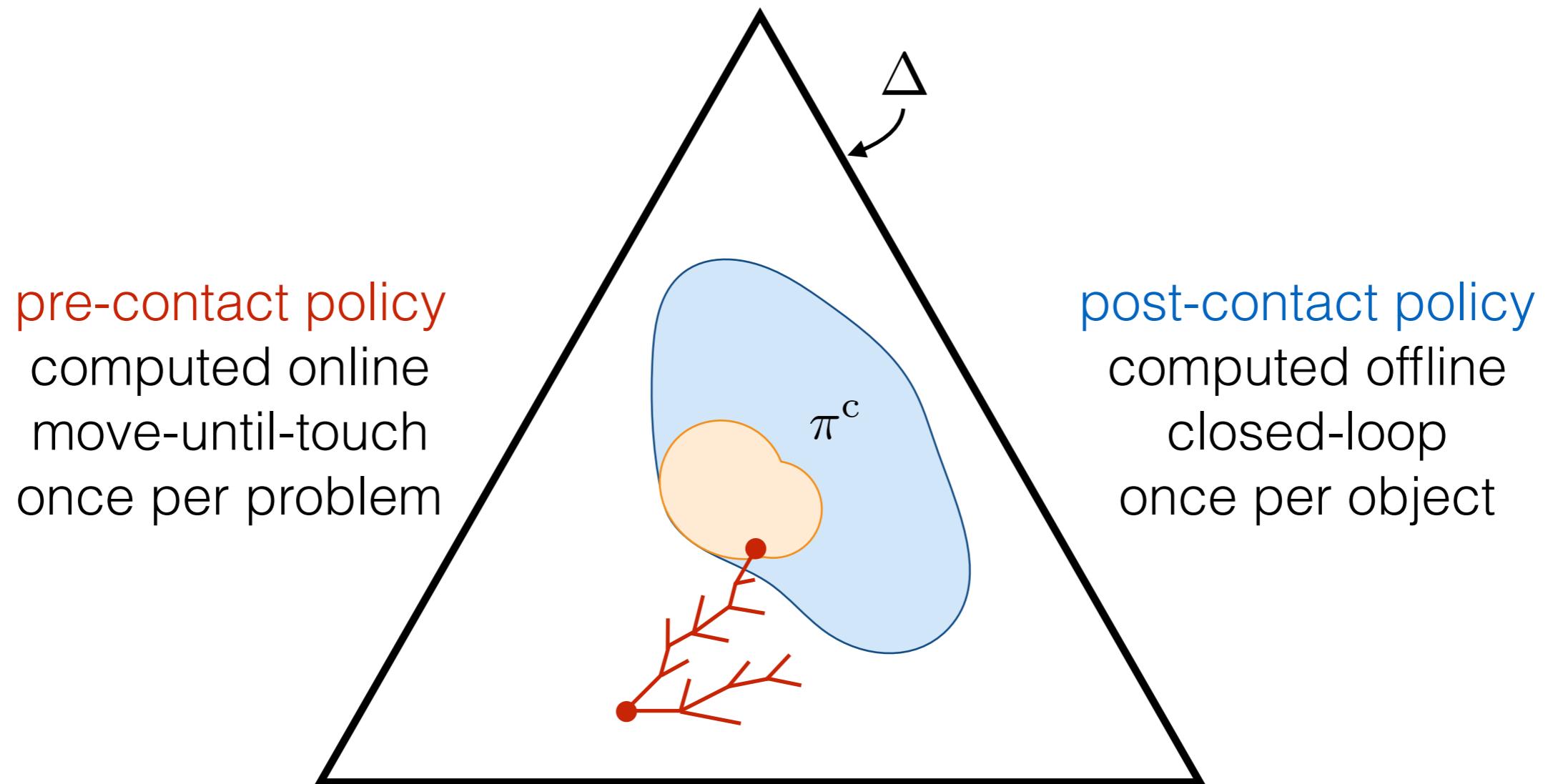
The post-contact belief space is small



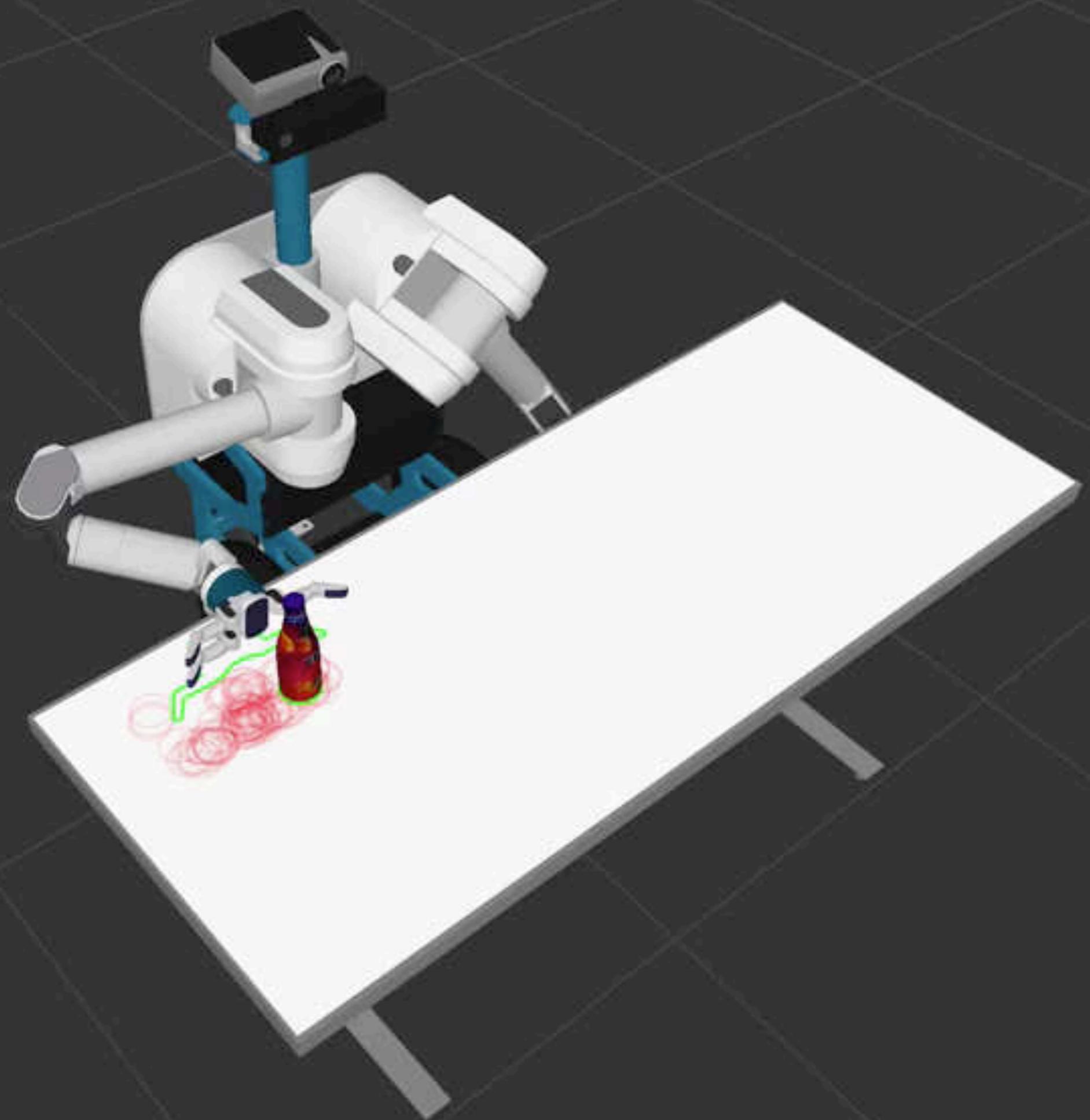
The post-contact belief space is small



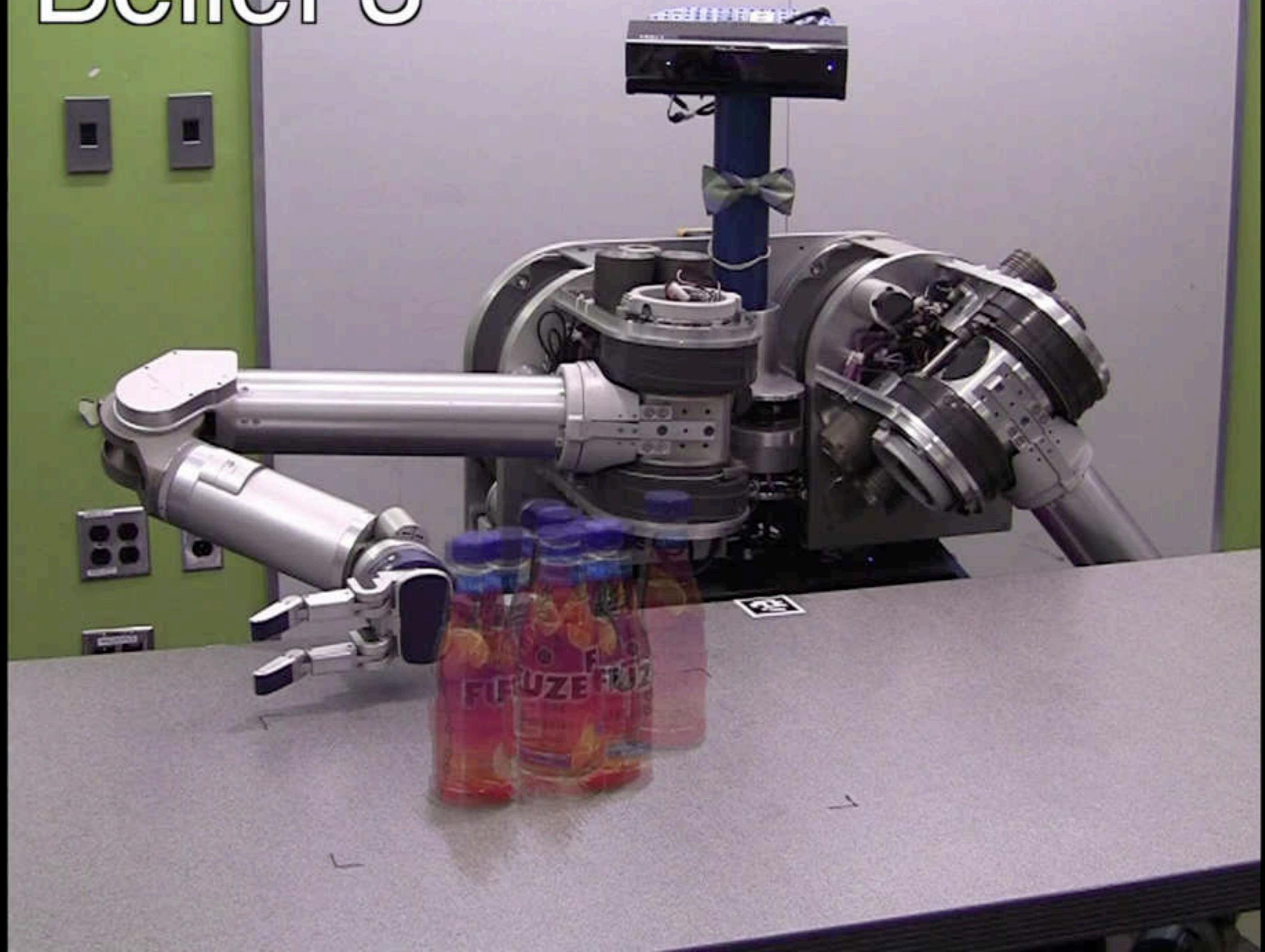
The post-contact belief space is small



Decompose into pre- and post-contact policies



Belief 3



Physics-Based Manipulation under Uncertainty



Michael Koval
mkoval@cs.cmu.edu

February 9, 2016
Hands: Design and Control for
Dexterous Manipulation