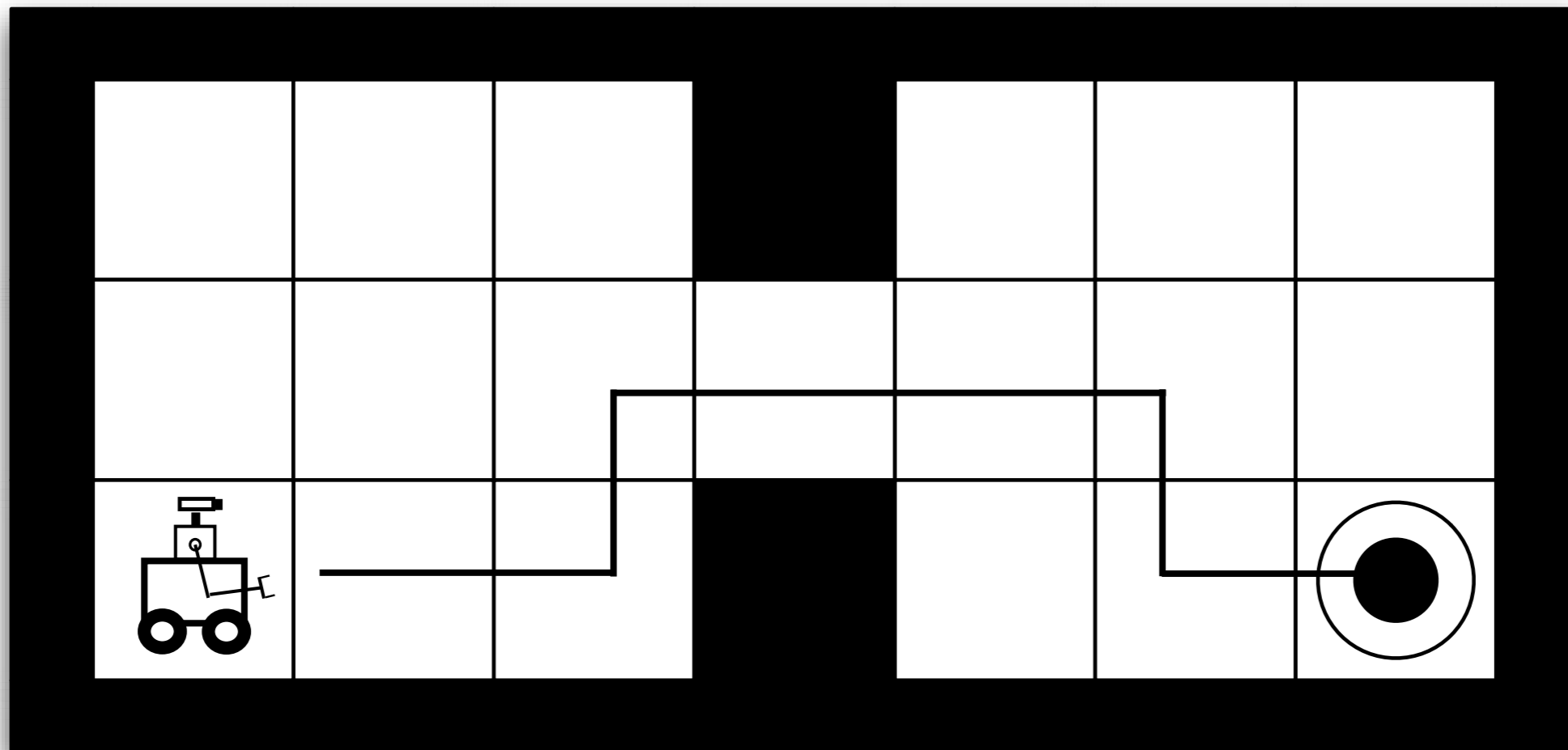# Robot Motion Planning

George Konidaris
gdk@cs.brown.edu

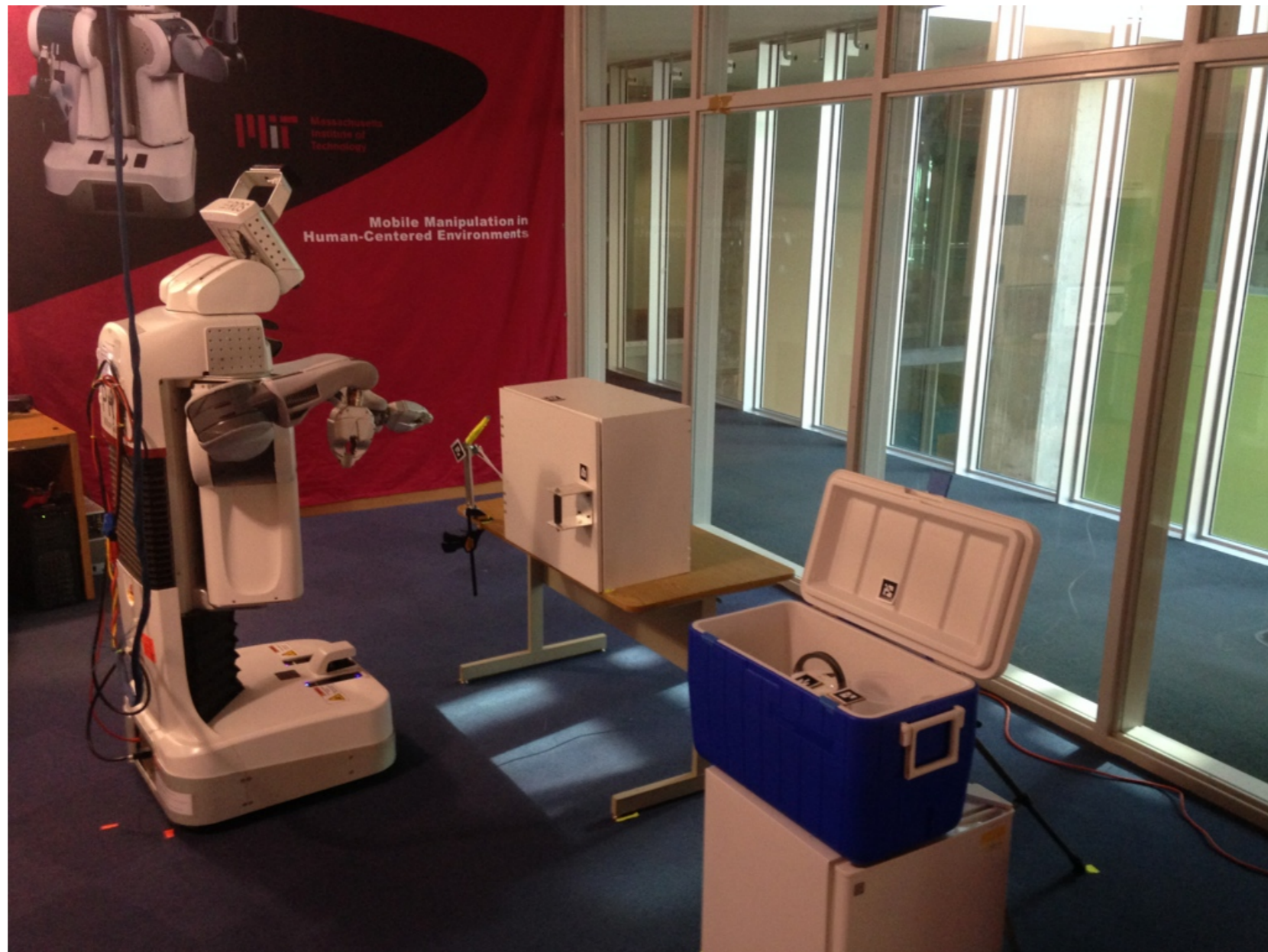**Fall 2021**

The Planning Problem

Finding a sequence of actions to achieve some goal.

Planning

Fundamental to AI:
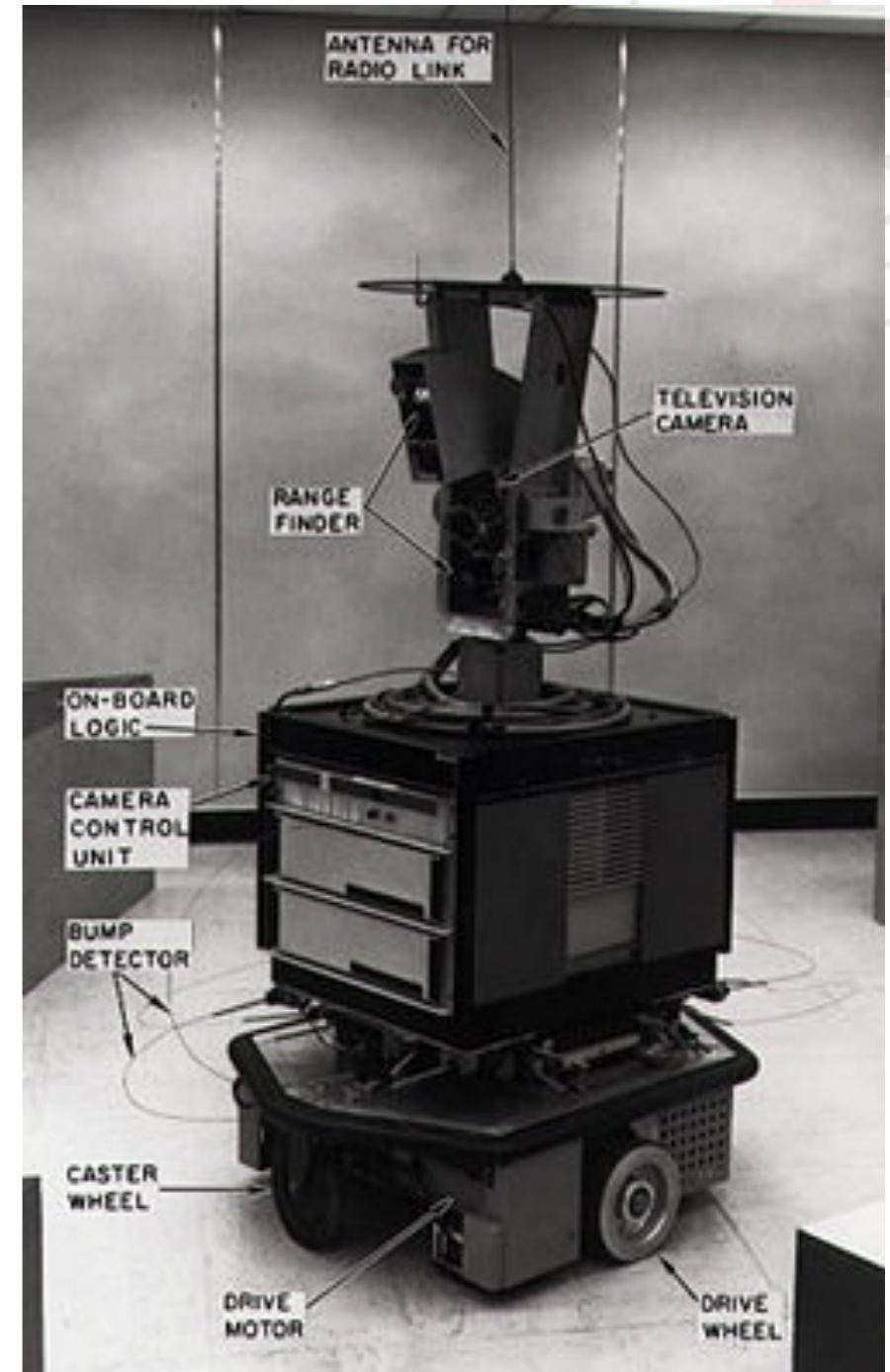
- Intelligence is about behavior.

Shakey the Robot

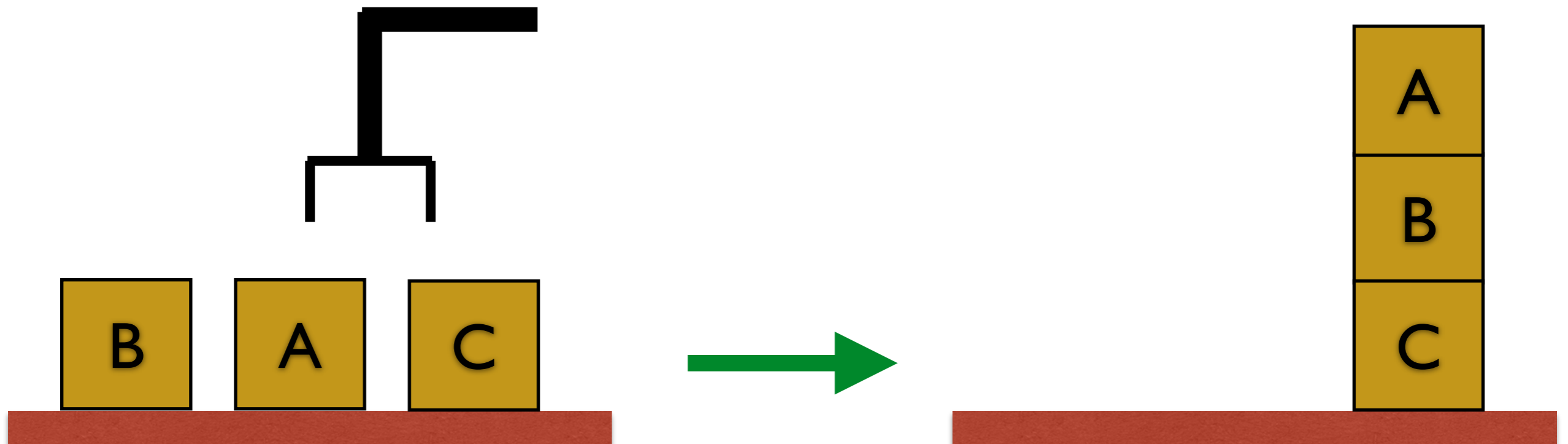Research project started in 1966.

Integrated:
- Computer vision.
- Planning.
- Control.
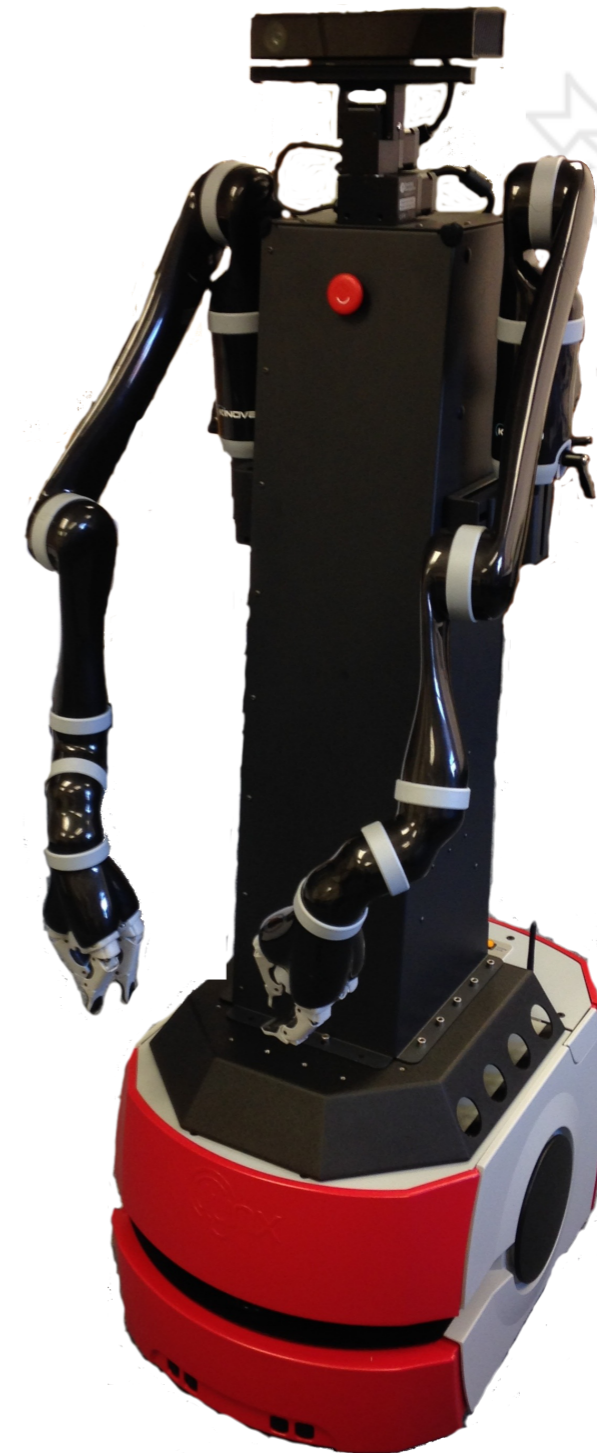- Decision-Making.
- KRR

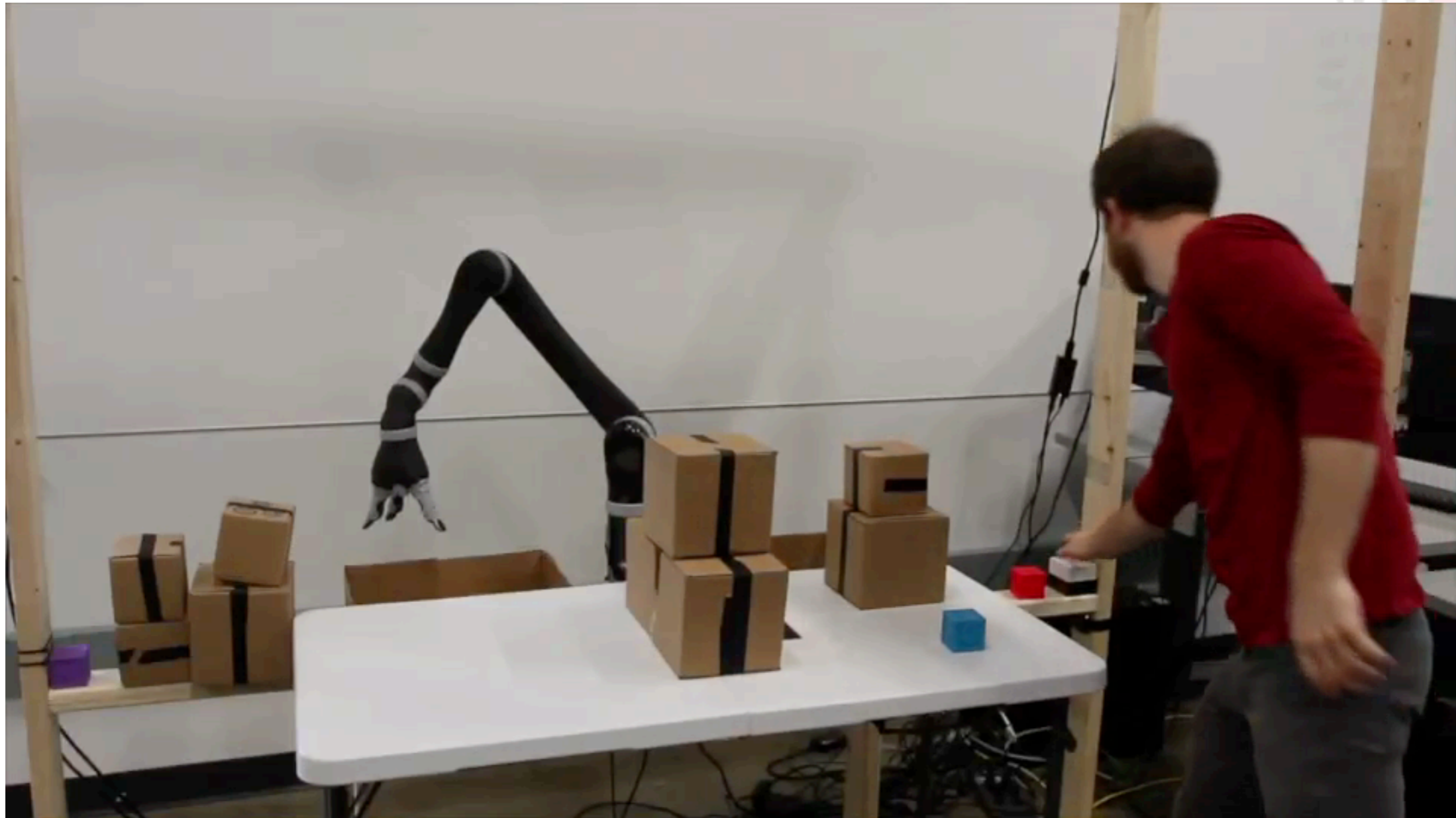# Classical Planning

```
(define (problem pb3)
   (:domain blocksworld)
   (:objects a b c)
   (:init (on-table a) (on-table b)   (on-table c)
         (clear a)  (clear b) (clear c) (arm-empty))
   (:goal (and (on a b) (on b c))))
```
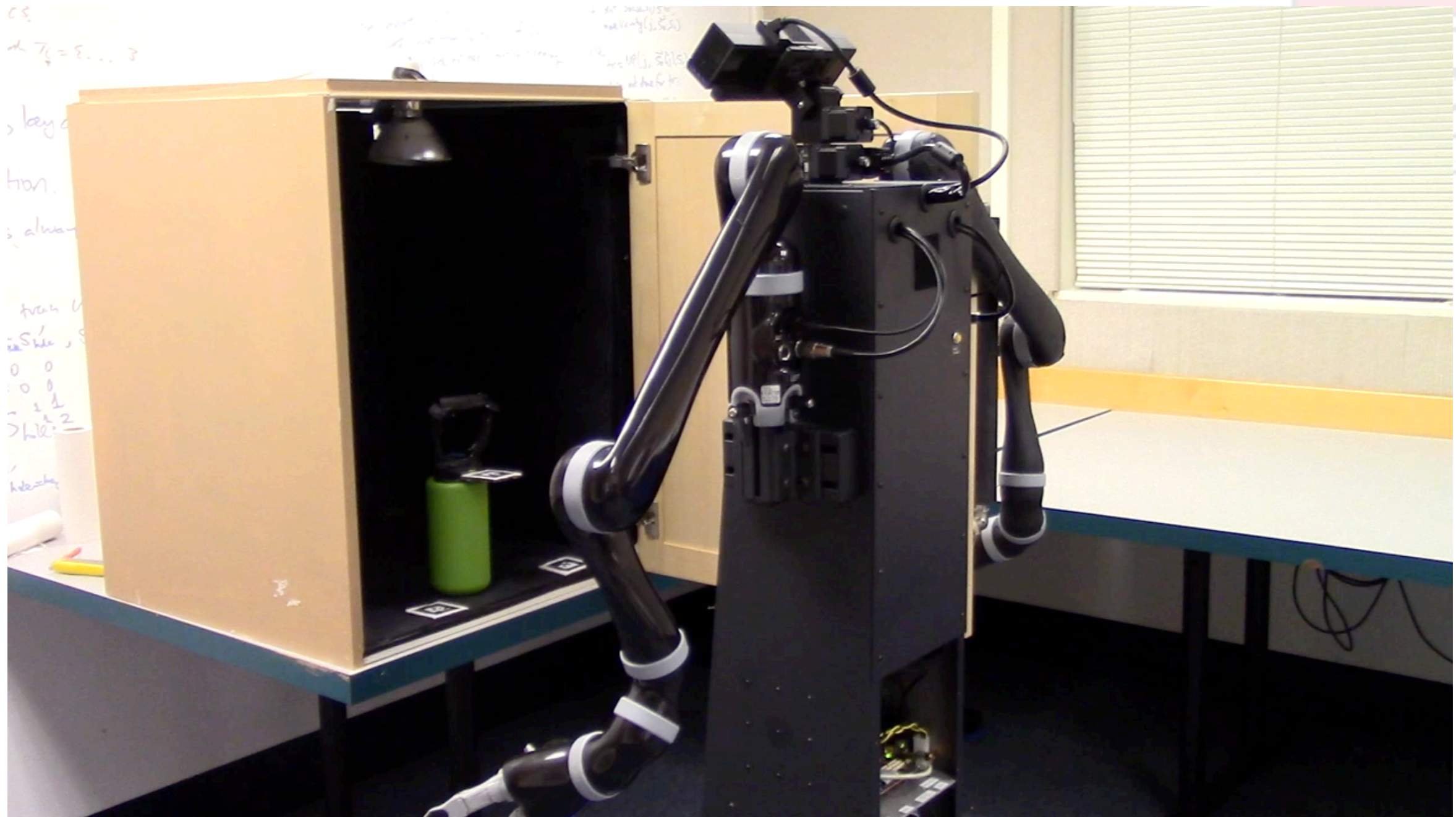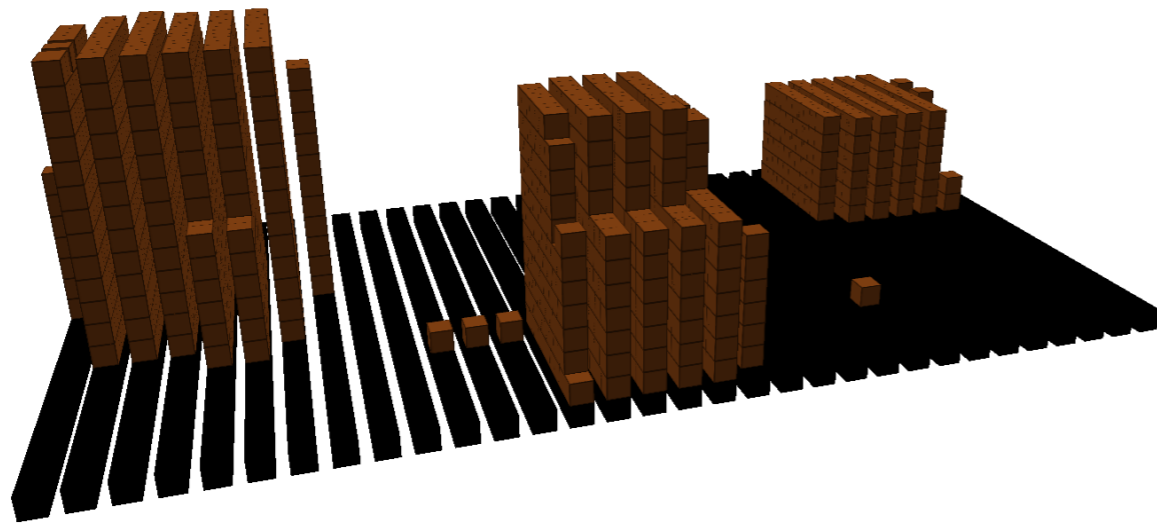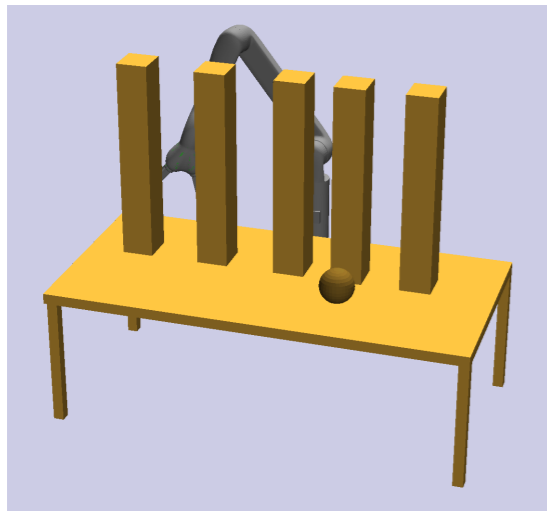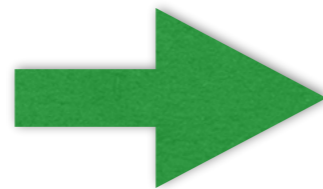
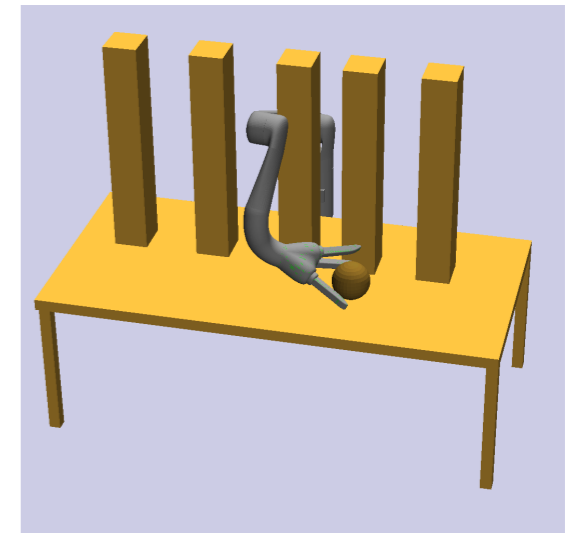# Robot Motion Planning

Motion Planning
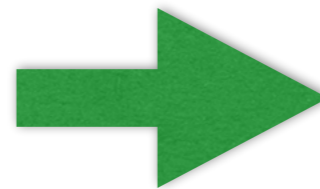
Motion Planning
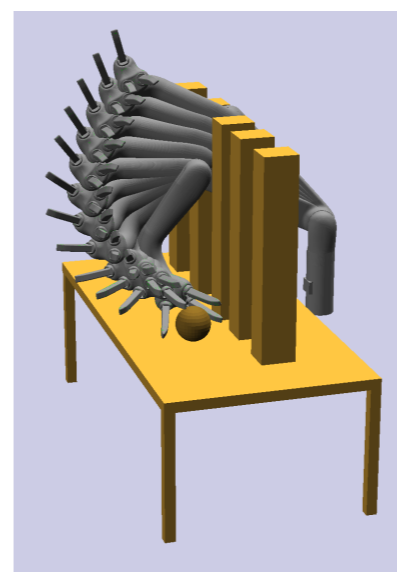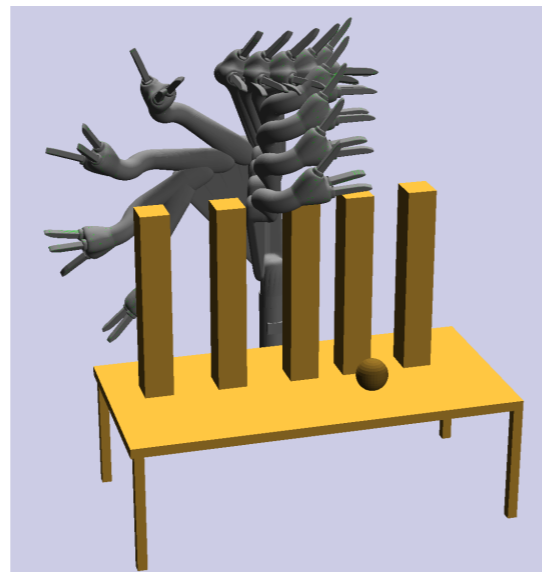
Motion Planning

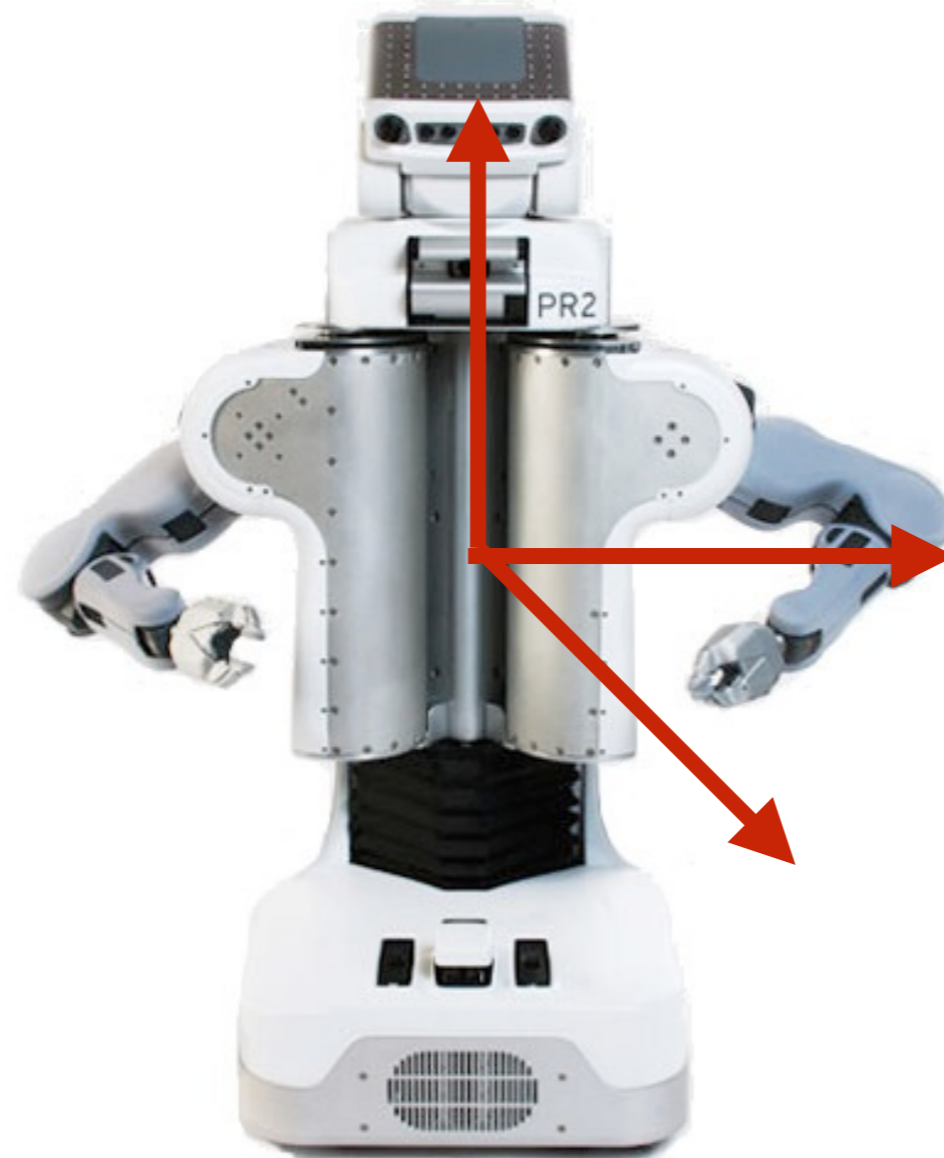# Motion Planning

start pose

??

goal

Configuration Space
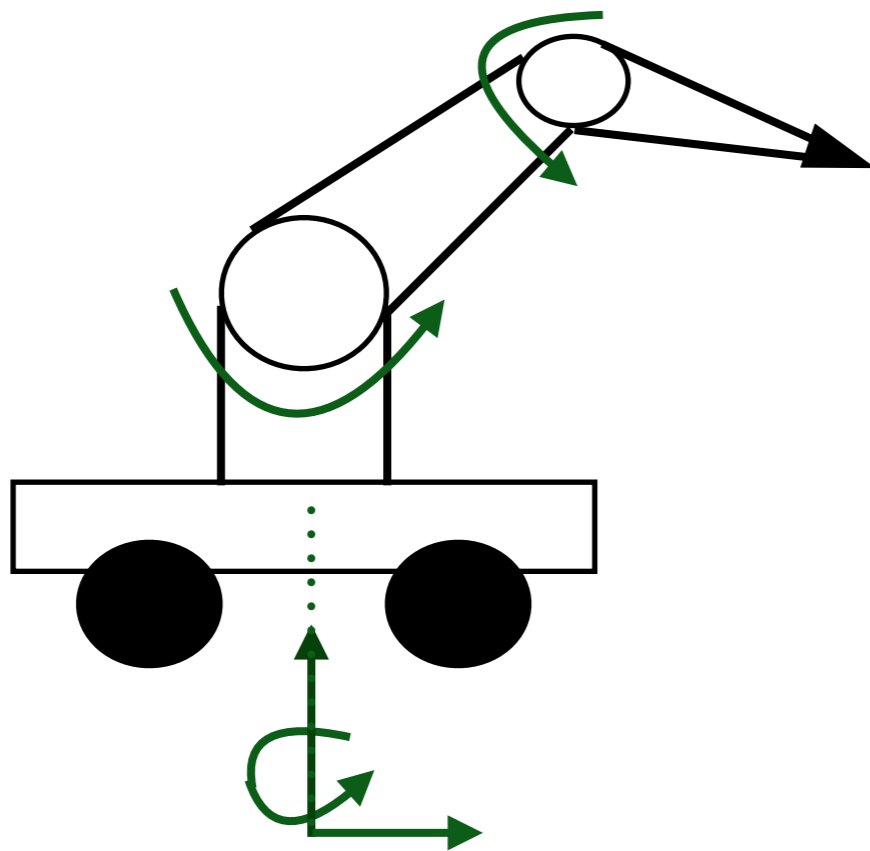
Robot has a **configuration space (C-space)**:
- Values for each joint
- Overall pose of reference frame

Configuration Spaces

Each joint is a **dimension** of the configuration space.

Let's say we have a robot with a movable base, and an arm with two revolute joints.

Configuration Spaces

Each joint is a **dimension** of the configuration space.

Let's say we have a robot with an arm with two revolute joints.
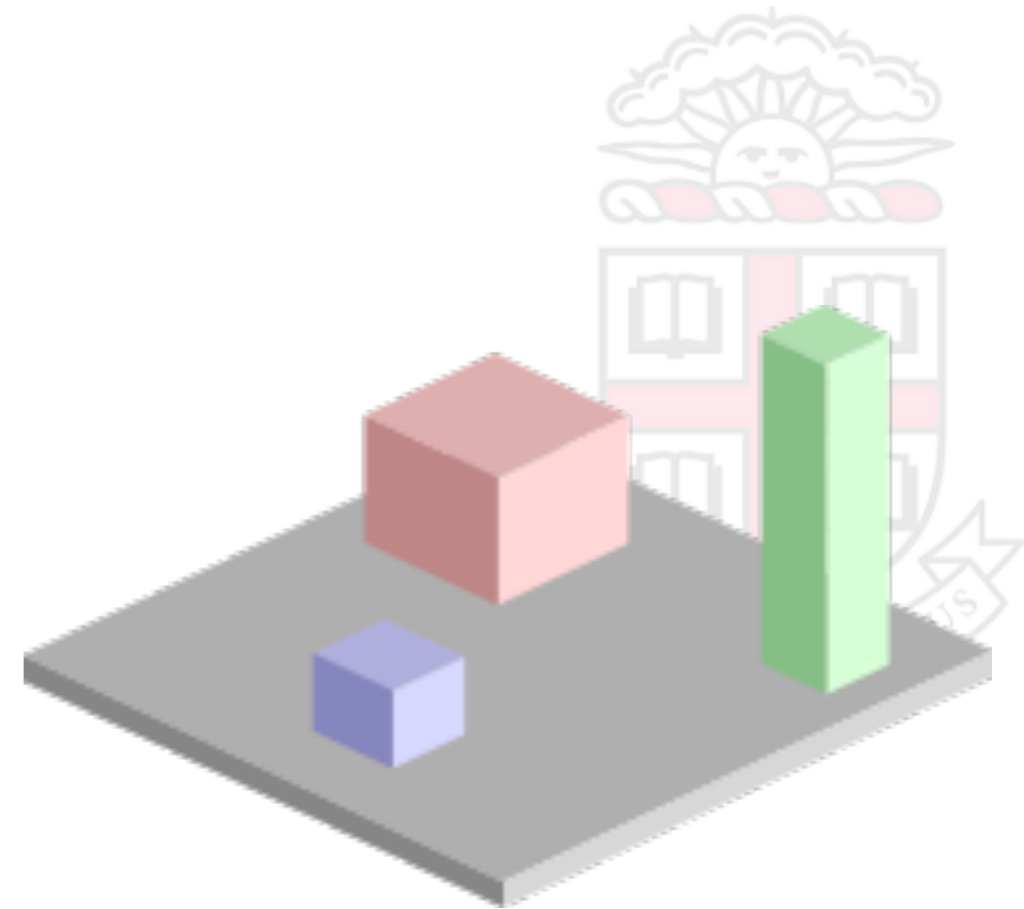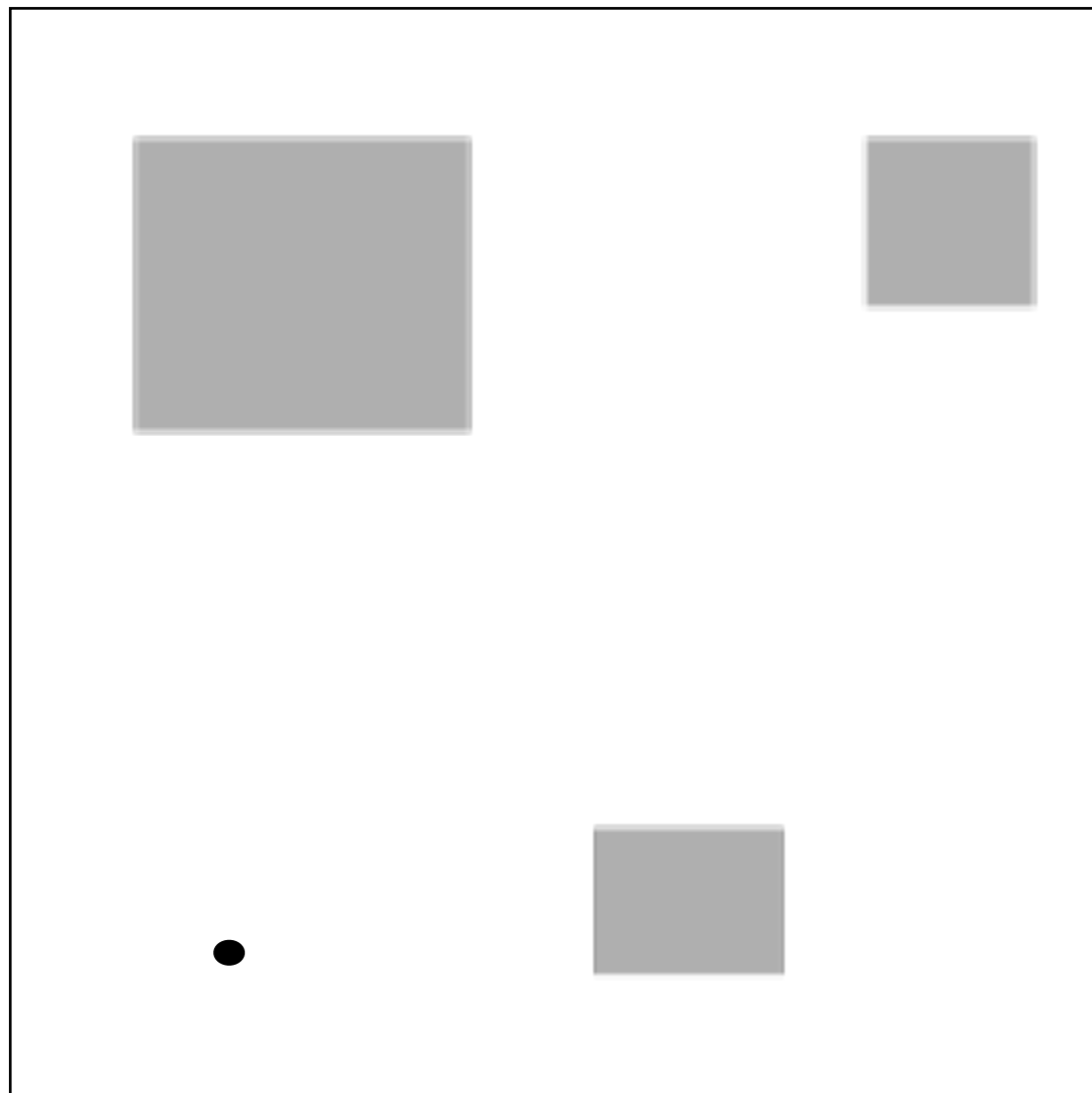
Configuration space:
- *x, y, theta* of *base frame*
- angle of first joint
- angle of second joint

A configuration is a *setting of values* to these 5 variables.
Configuration space is the *space of all such settings*.
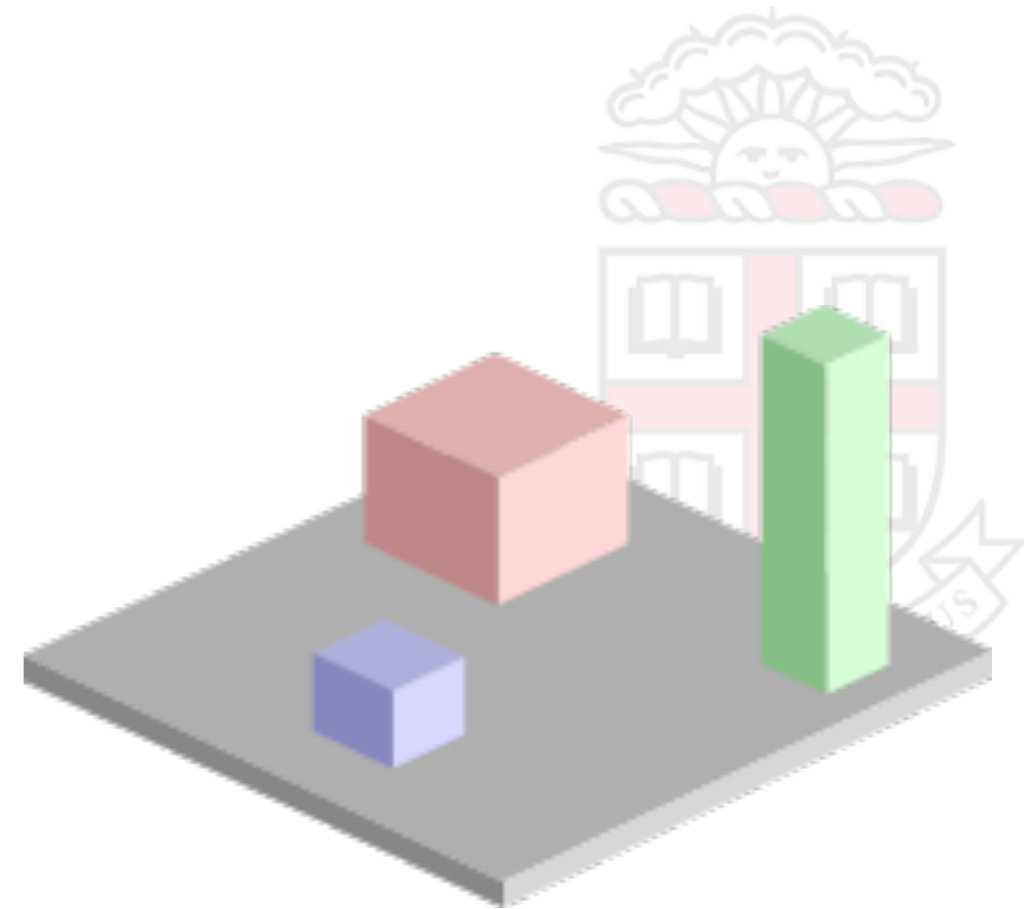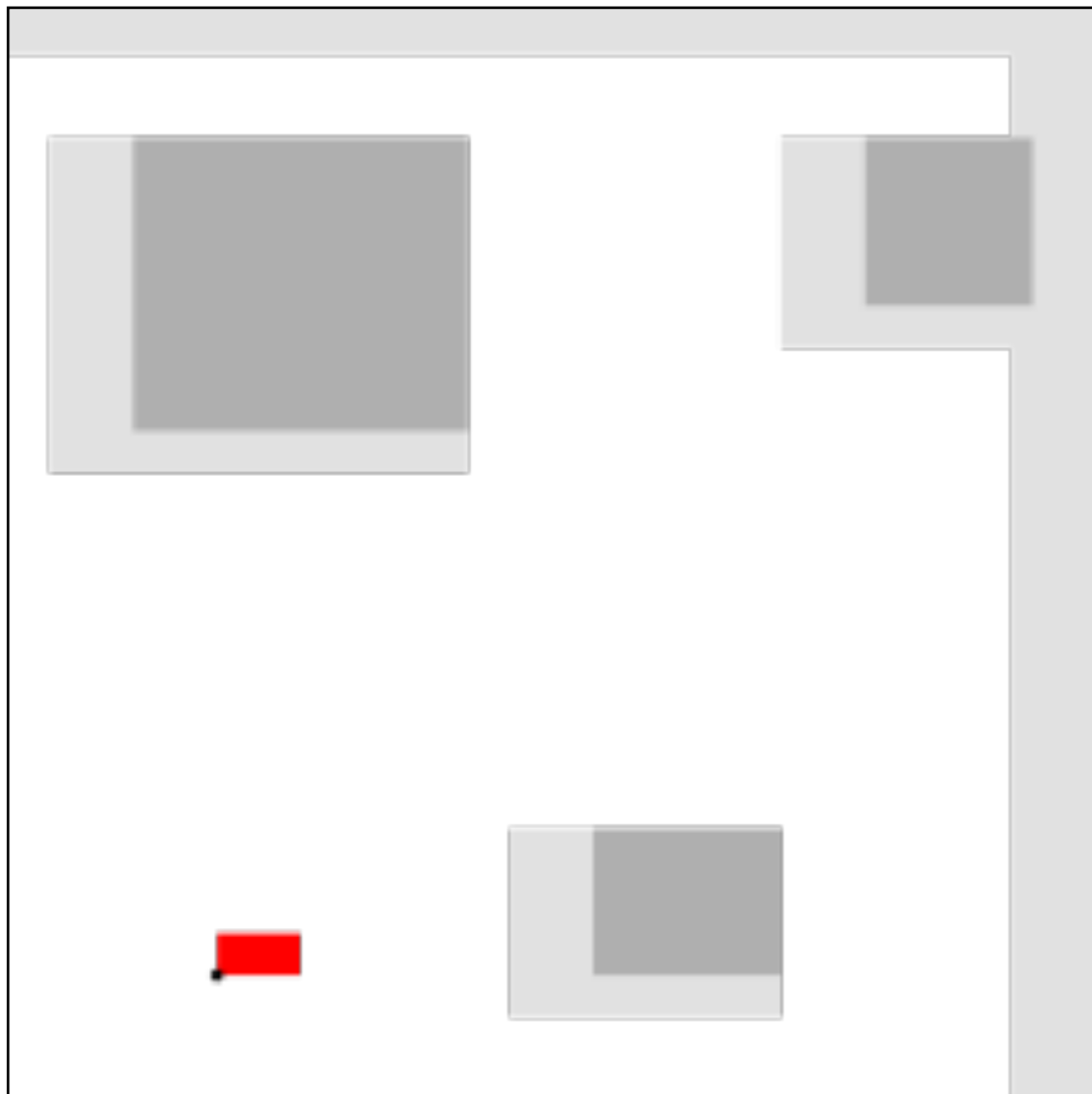
Configuration Space

Obstacles are no-go regions
of configuration space.
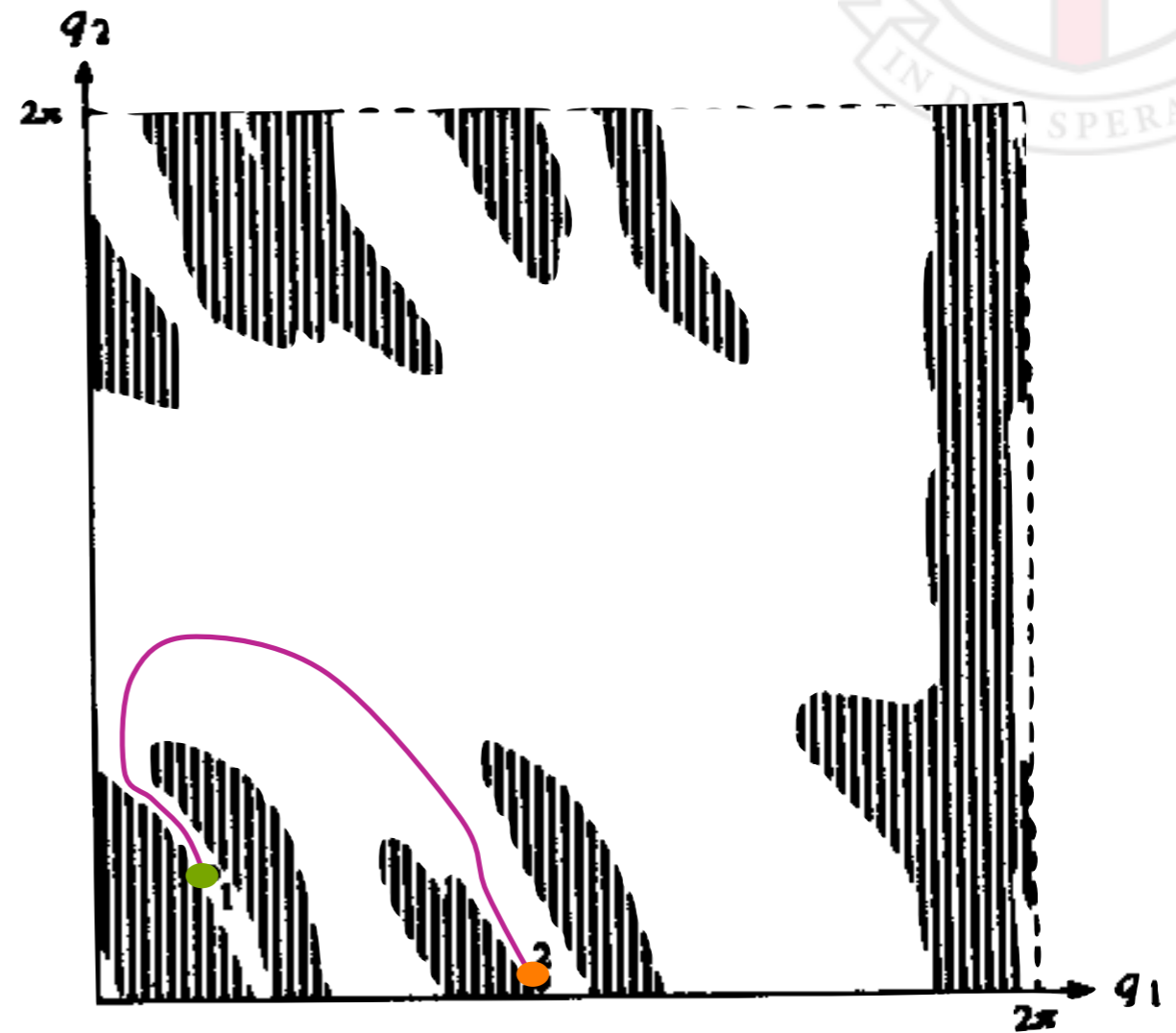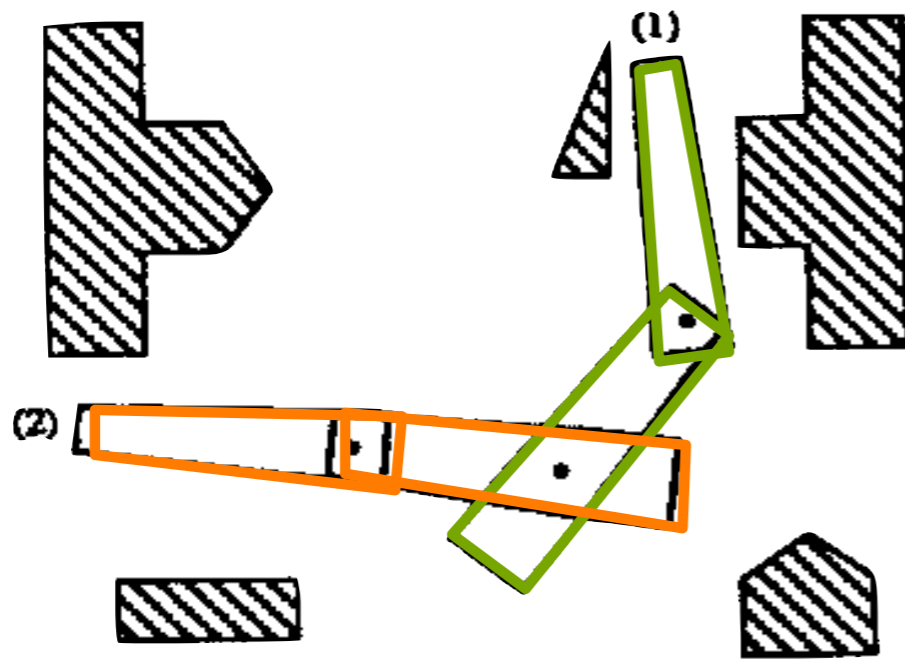
(images from Wikipedia)

Configuration Space

Obstacles are no-go regions
of configuration space.

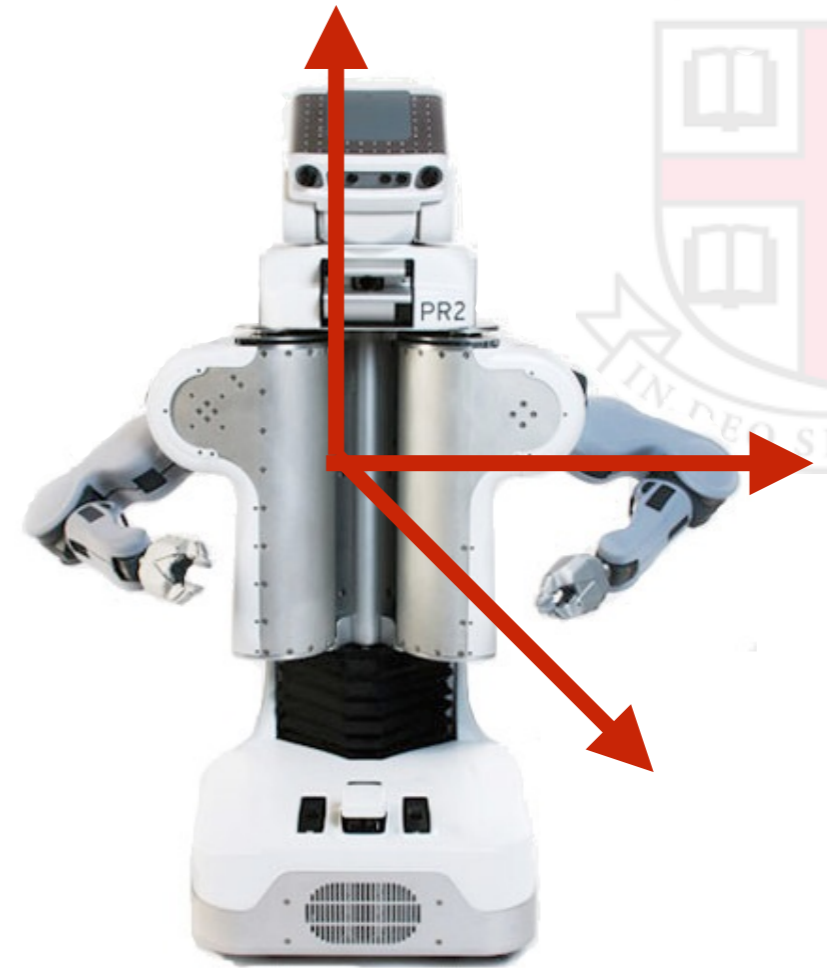(images from Wikipedia)

Configuration Space

[from Lozano-Perez 87]

Problem Definition

Given:

- Configuration space
- Start point in C-space
- Goal region in C-space
- Set of obstacles
  - Dense regions of 3D-space
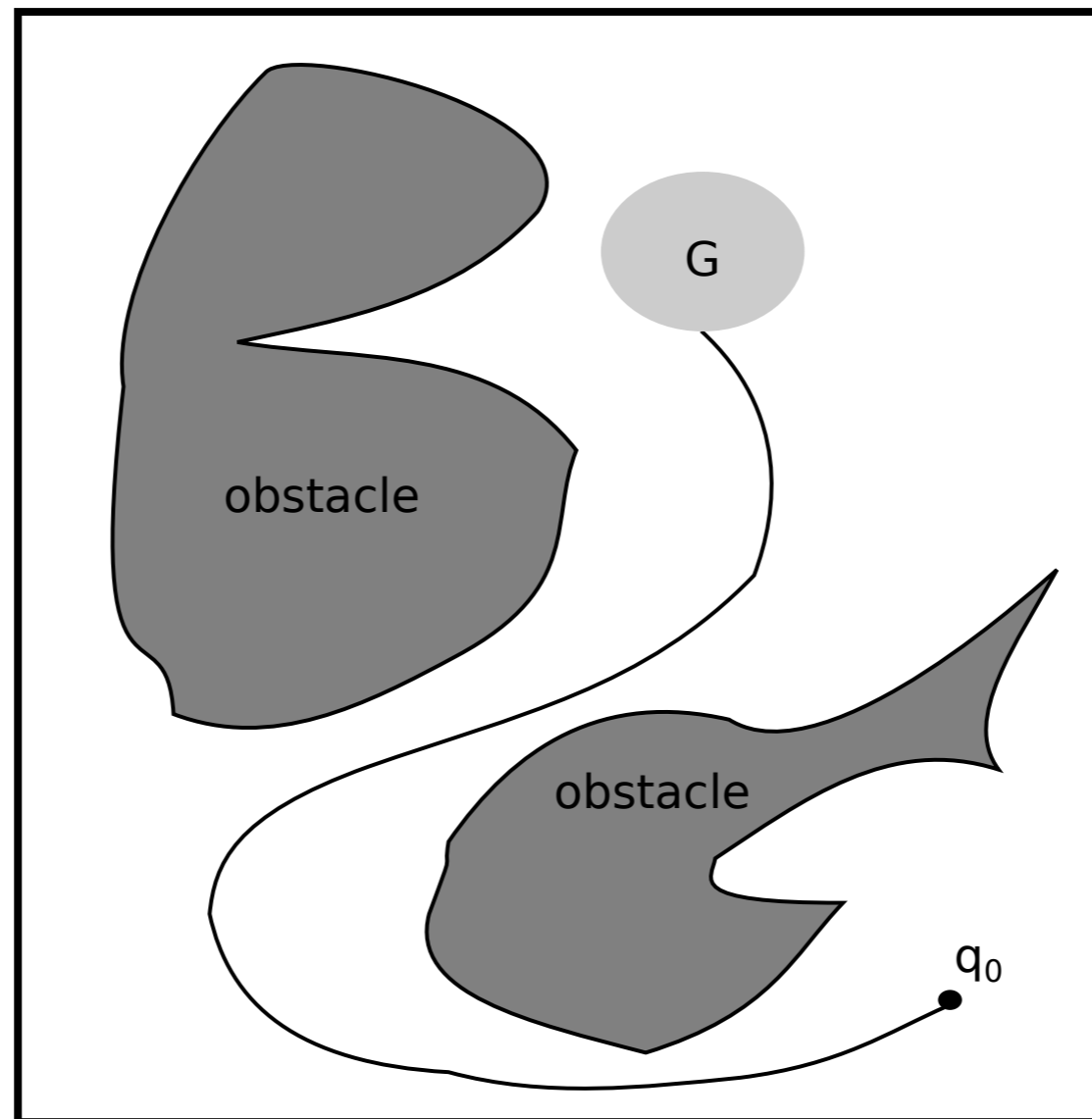 - (*Also* regions of C-space)

Find: *feasible*, *obstacle-free* (possibly cost-minimizing) path through C-space from start to a point in goal.

Planning

We wish to find a path through configuration space such that:
- Path feasible
- No collisions
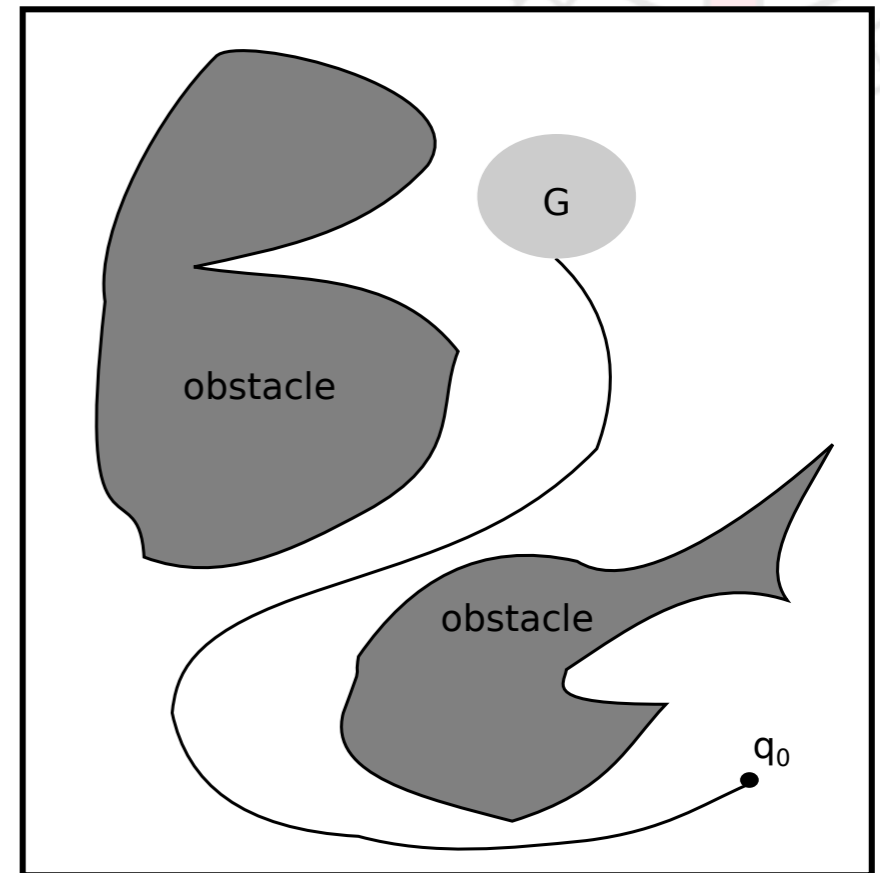- Minimize cost

a path in free space

Paths

Simple definition of a path:

- Sequence of points $p = \{p_1, \ldots, p_n\}$
- "Easy" to go between p_i and p_{i+1}.
- Additive cost $C(p_i, p_{i+1})$

Solution - path such that:

- p_1 = start
- p_n inside goal
- No collision between any p_i and p_{i+1}.
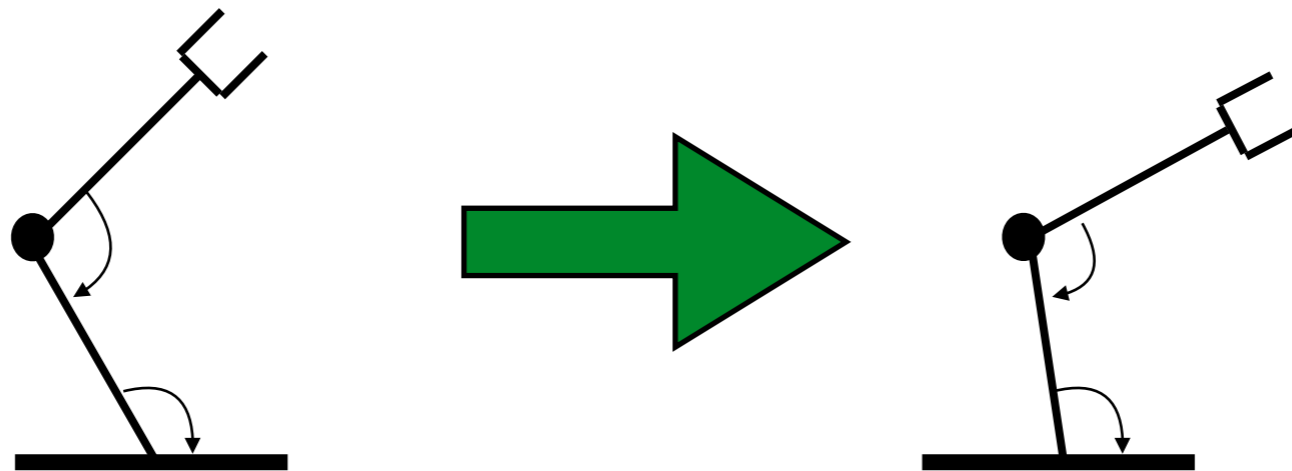
- $\min \sum_{i=1}^{n-1} C(p_i, p_{i+1})$

a path in free space

Local Controller

What does "easy to go between p_i and p_{i+1}" mean?

It means you can **control** the robot directly from point p_i to point p_{i+1}, without considering obstacles.

There may also be constraints on motions (e.g., maximum speed or jerk, maximum rate of angular acceleration).
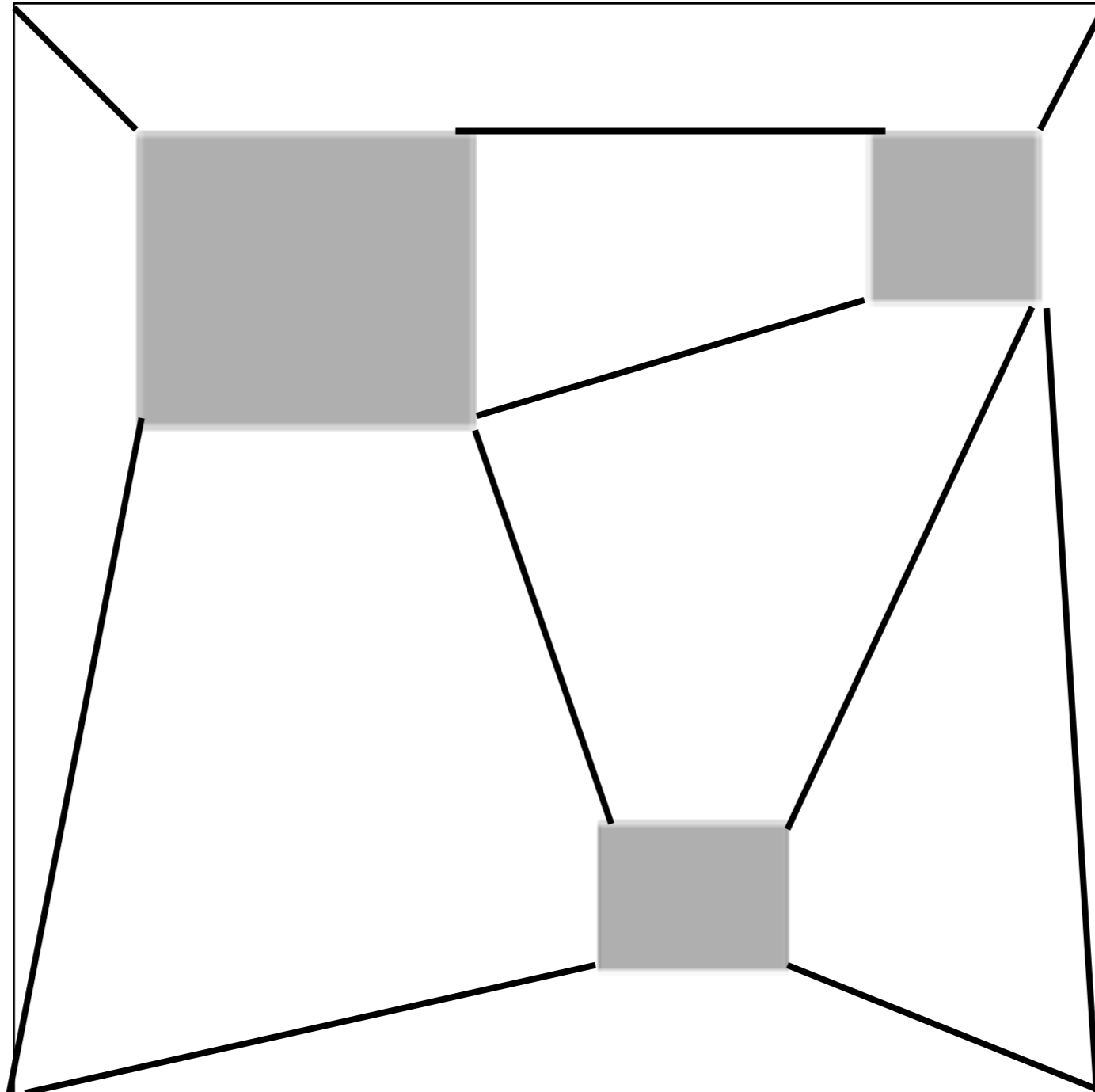
Collision Detection
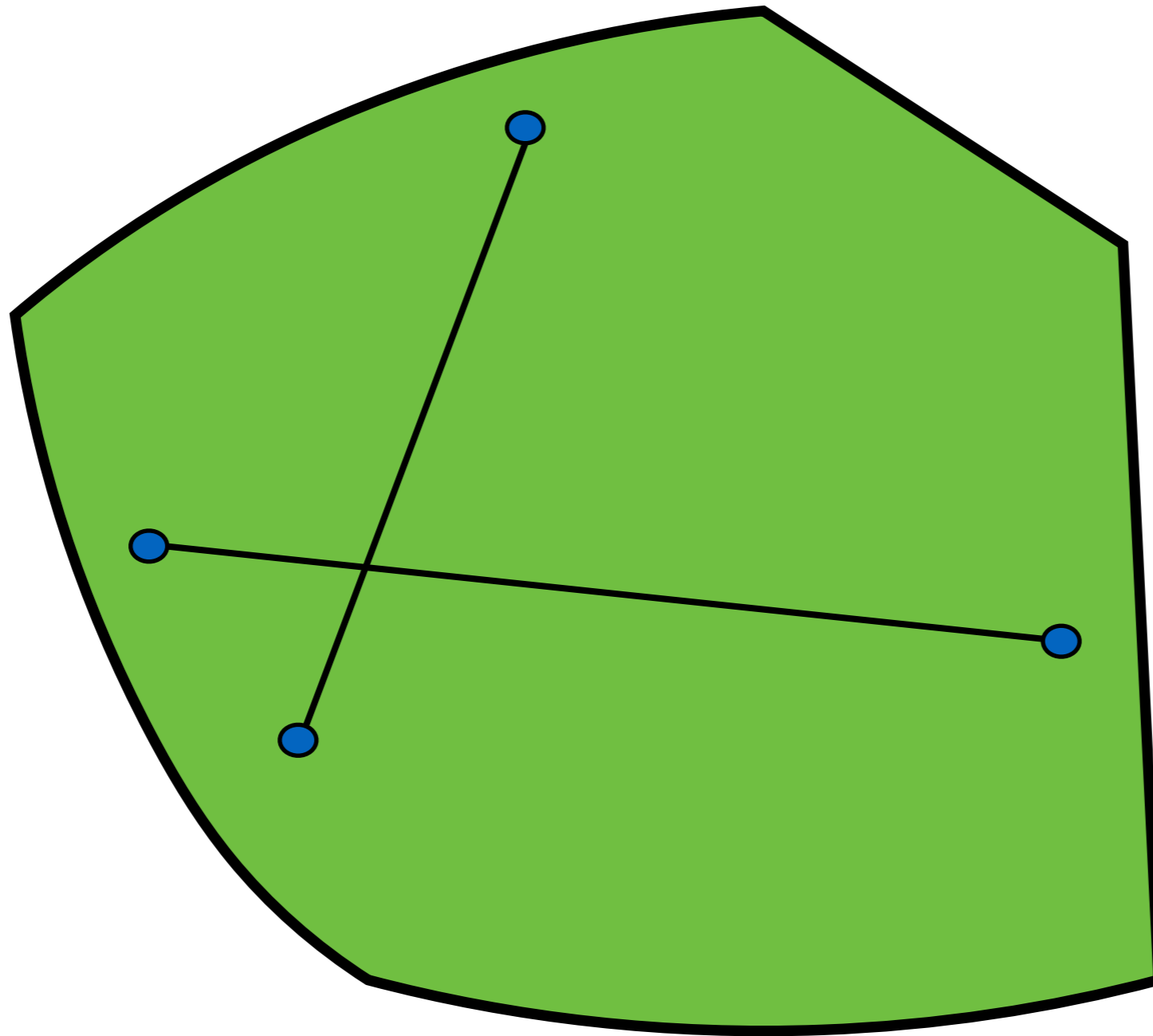
What does collision-free mean?

Must test: collision between *obstacle* and *swept volume*.
This can be done in 3-space.
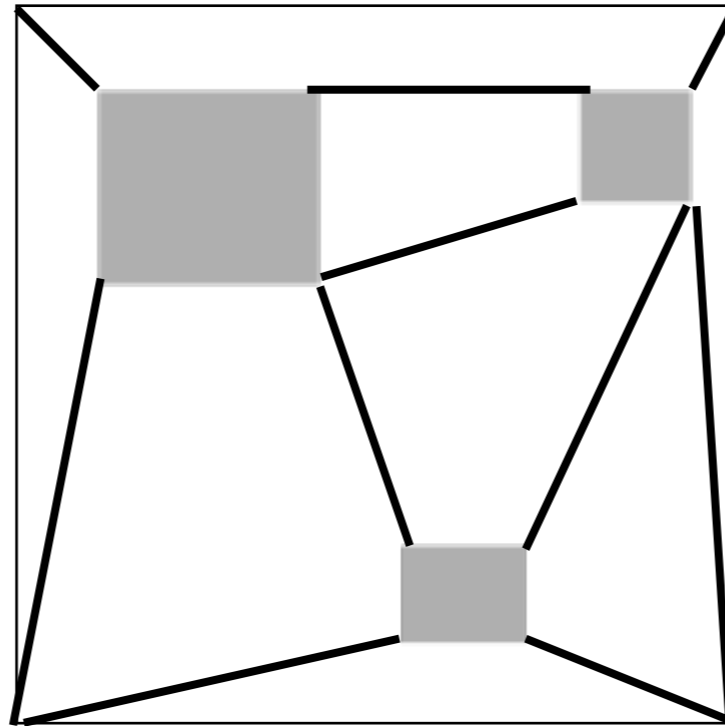
Visibility Graphs

Initial approaches: geometric.

Convex Regions

Convex region: the line connecting any two points inside the region lies itself wholly within the region.

Visibility Graphs
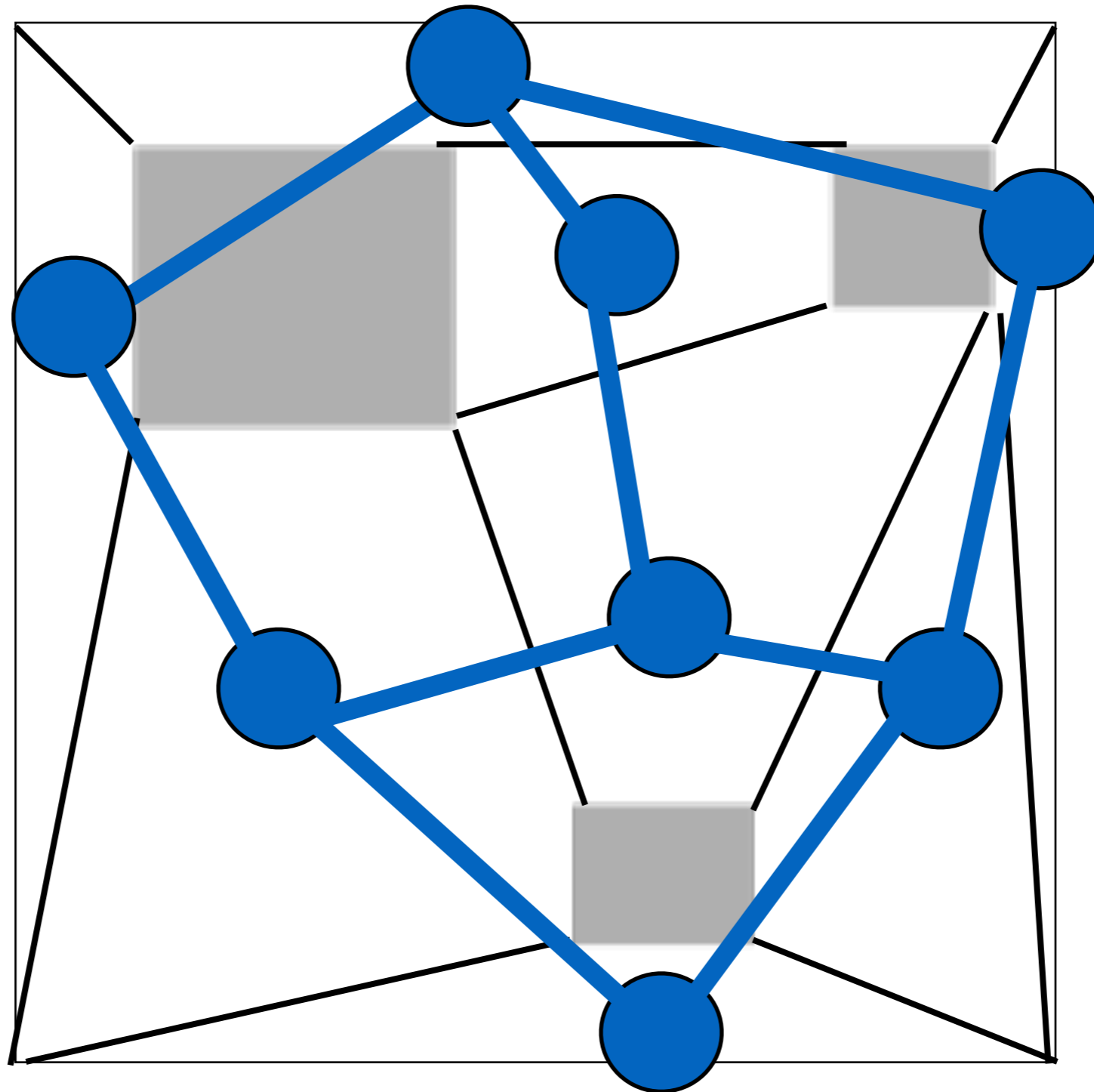
1. Break *C-space* up into convex regions.

2. Build a graph: each node convex region, edge when they share a face.
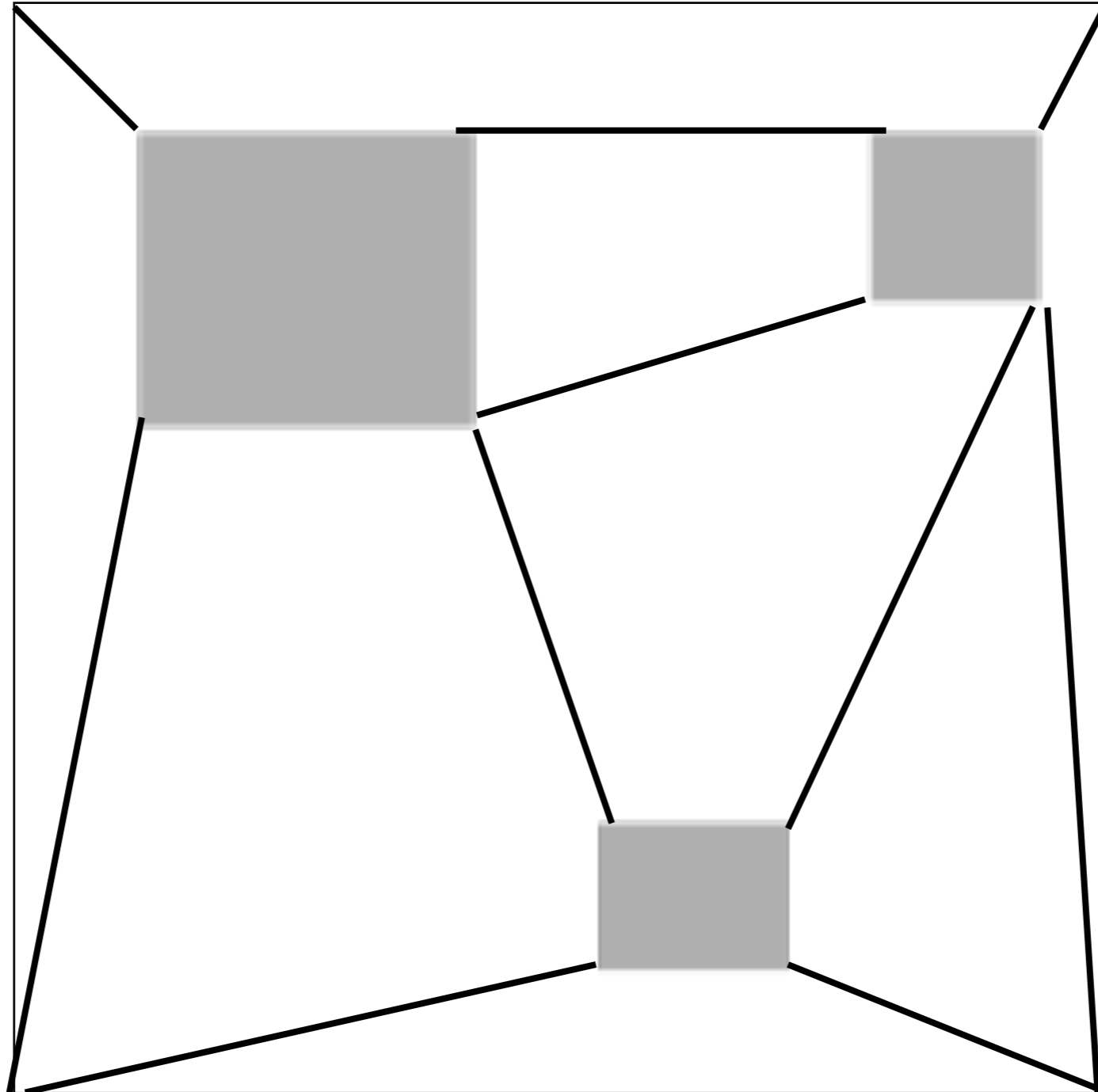
3. Do search on the graph.
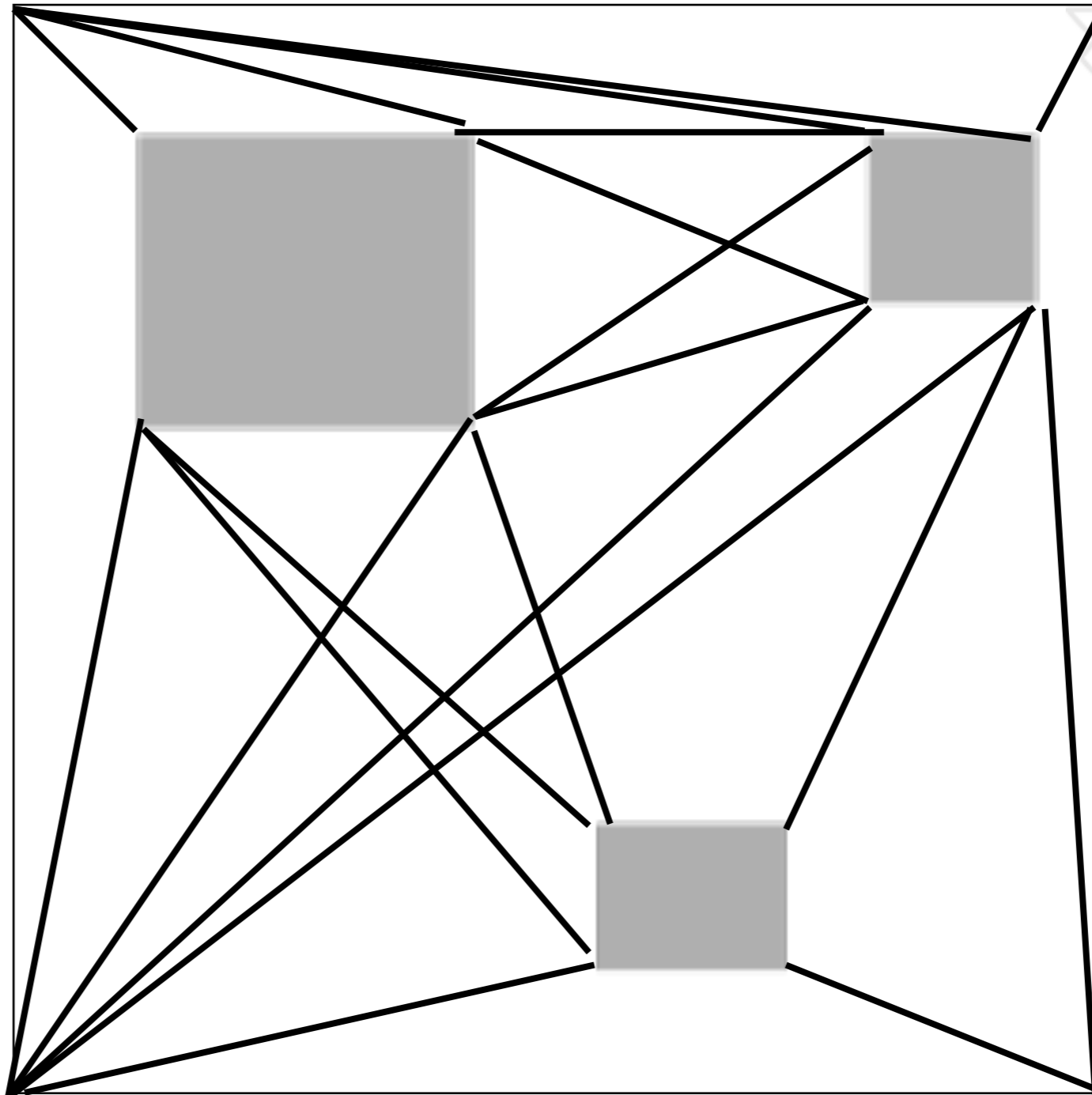
Visibility Graphs

Optimality
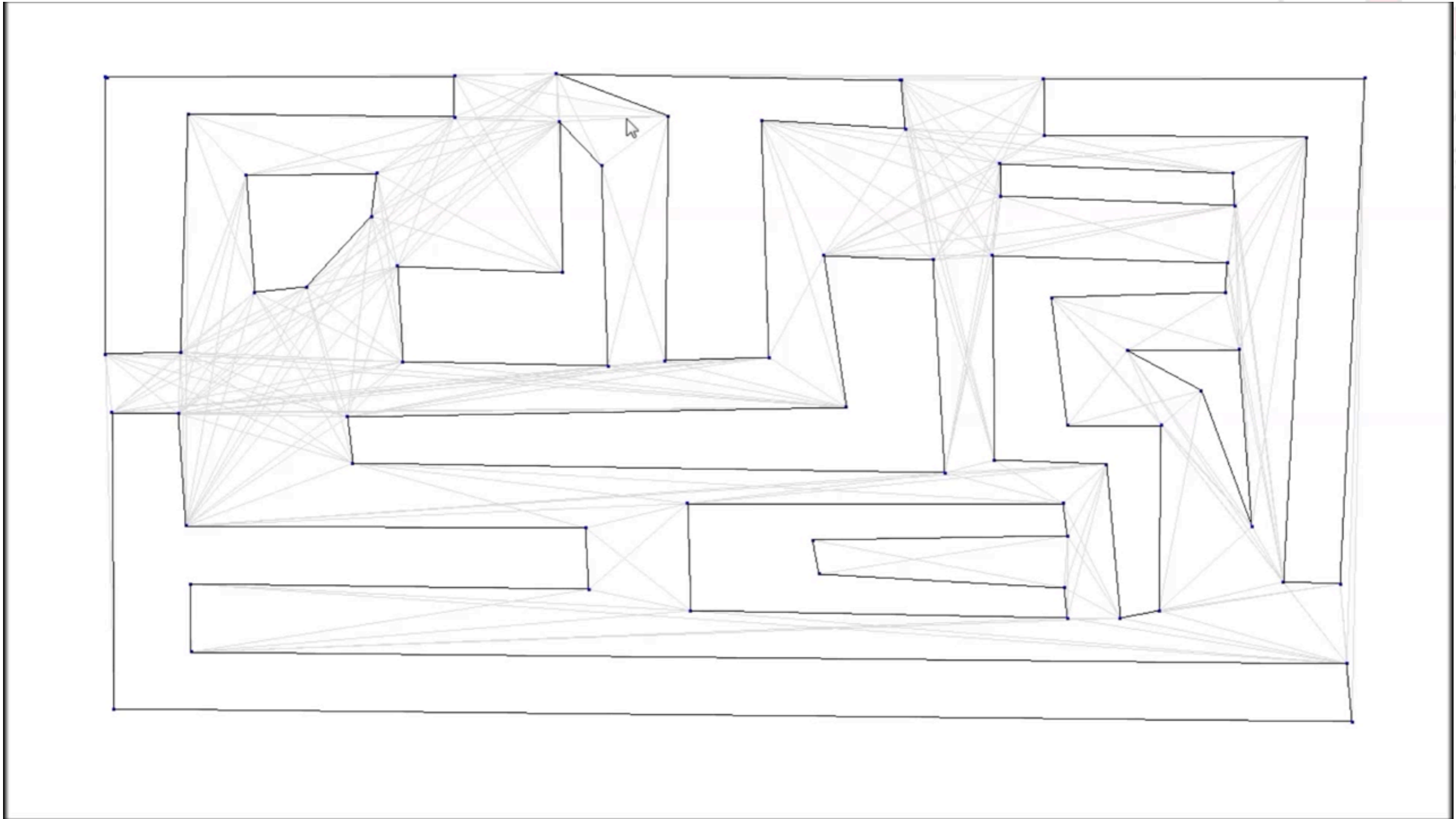
Issue: these paths may not be optimal. Why?

Optimality

Go a bit further: break into *triangles*, each vertex lies on an obstacle vertex.

Video

credit: Ulf Biallas

Issues

These are hard to use:

- Convex region numbers grow exponentially with dimension.
- Need analytical model of each obstacle *in C-space.*
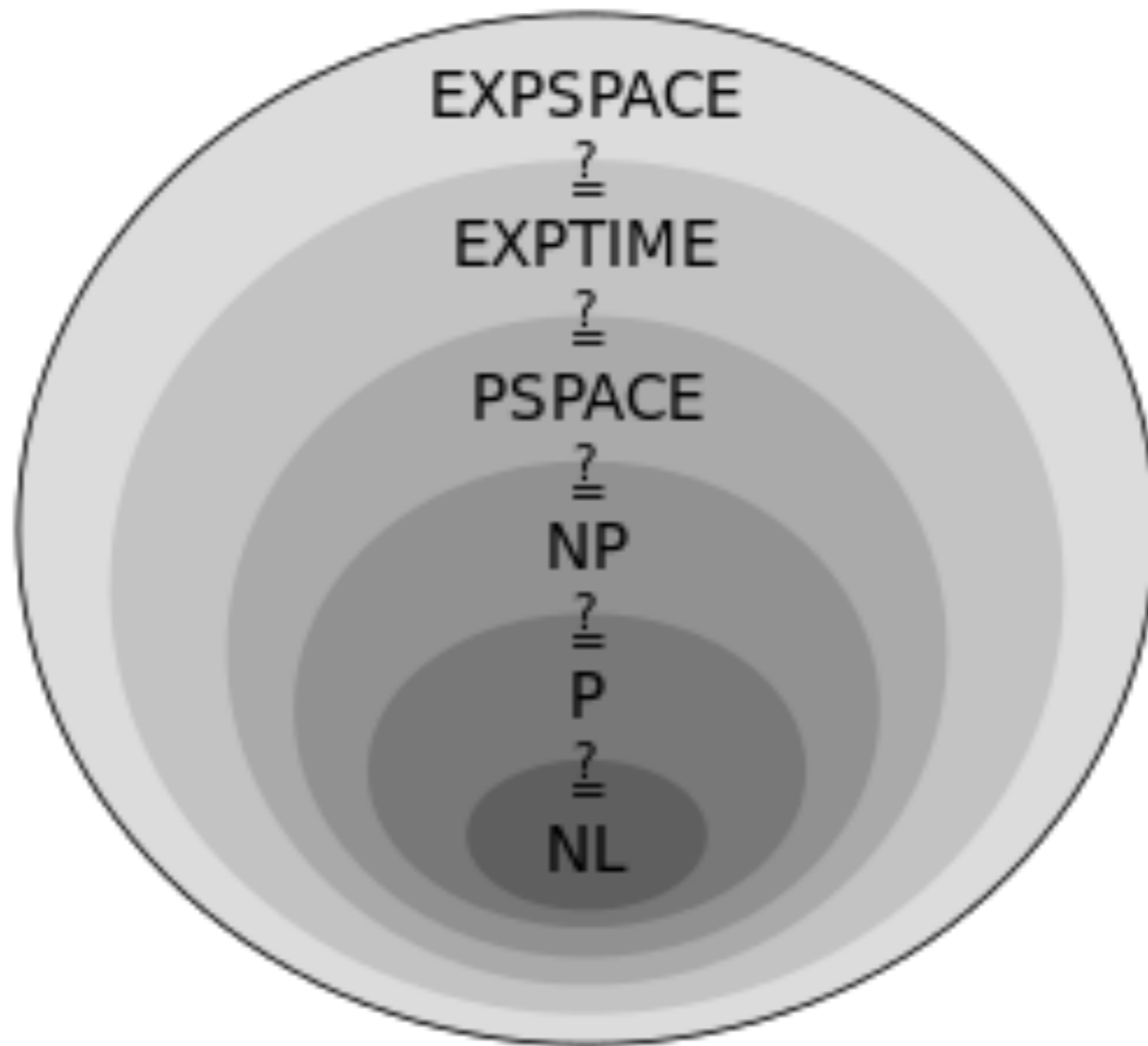- Need analytical model of C-space!

This is a lot of work.

Consequently, these methods only used for very low-*d* problems.

Complexity
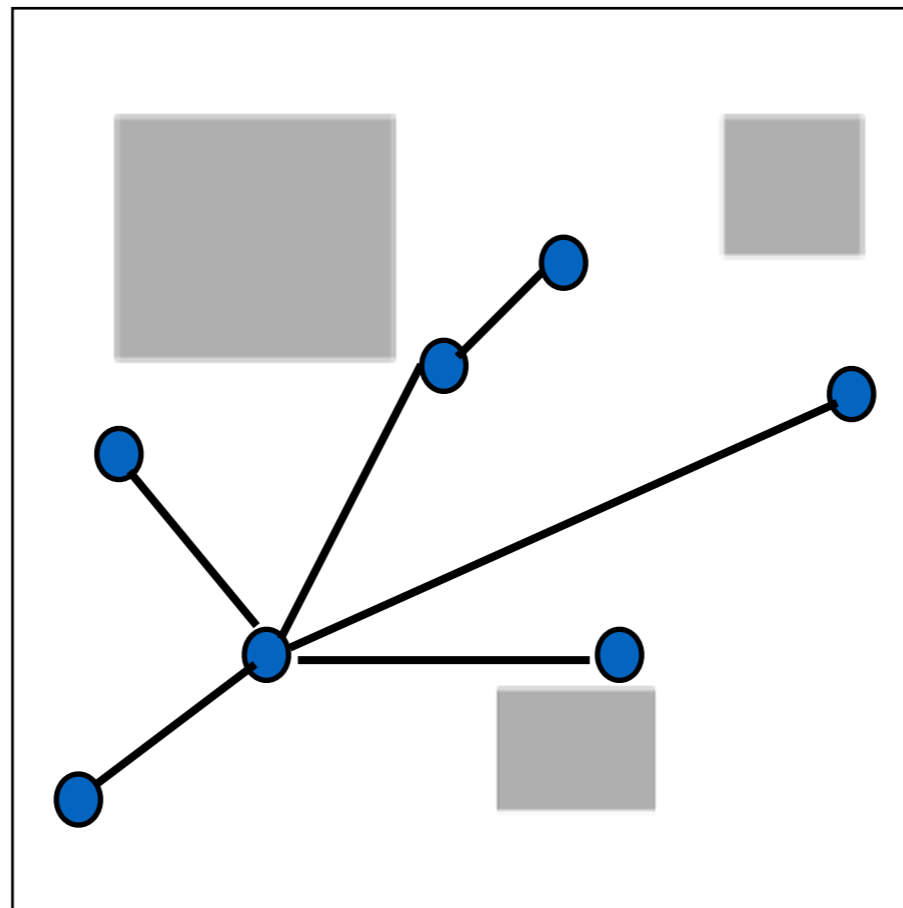
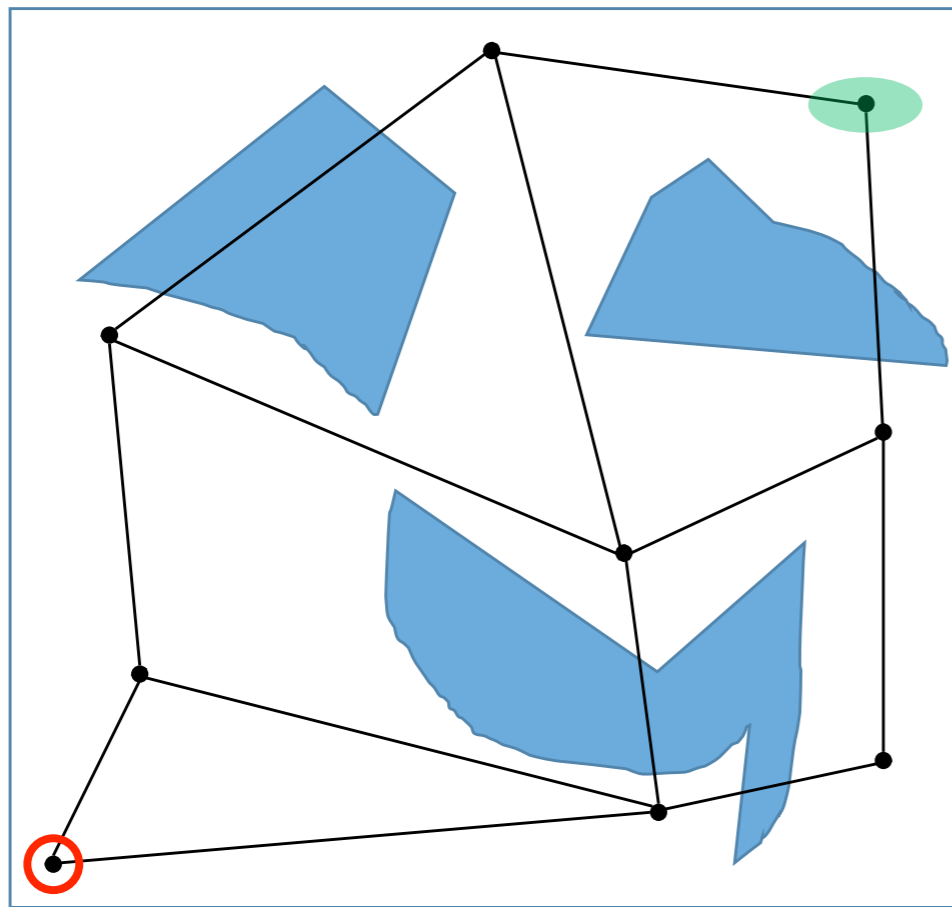Issue: motion planning is P-SPACE complete (Reif, 1979).

Randomization

Alternative solution:

- Rely on randomized algorithms.
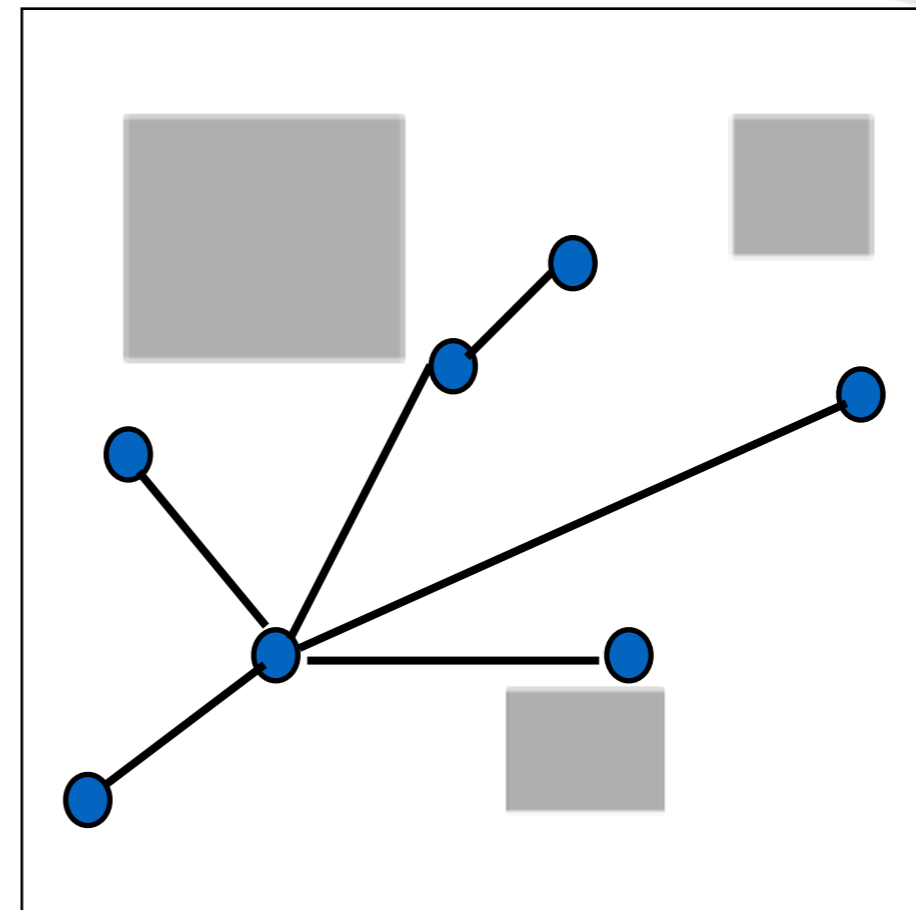- Expensive but probabilistic guarantees.
- Typically very simple to code.

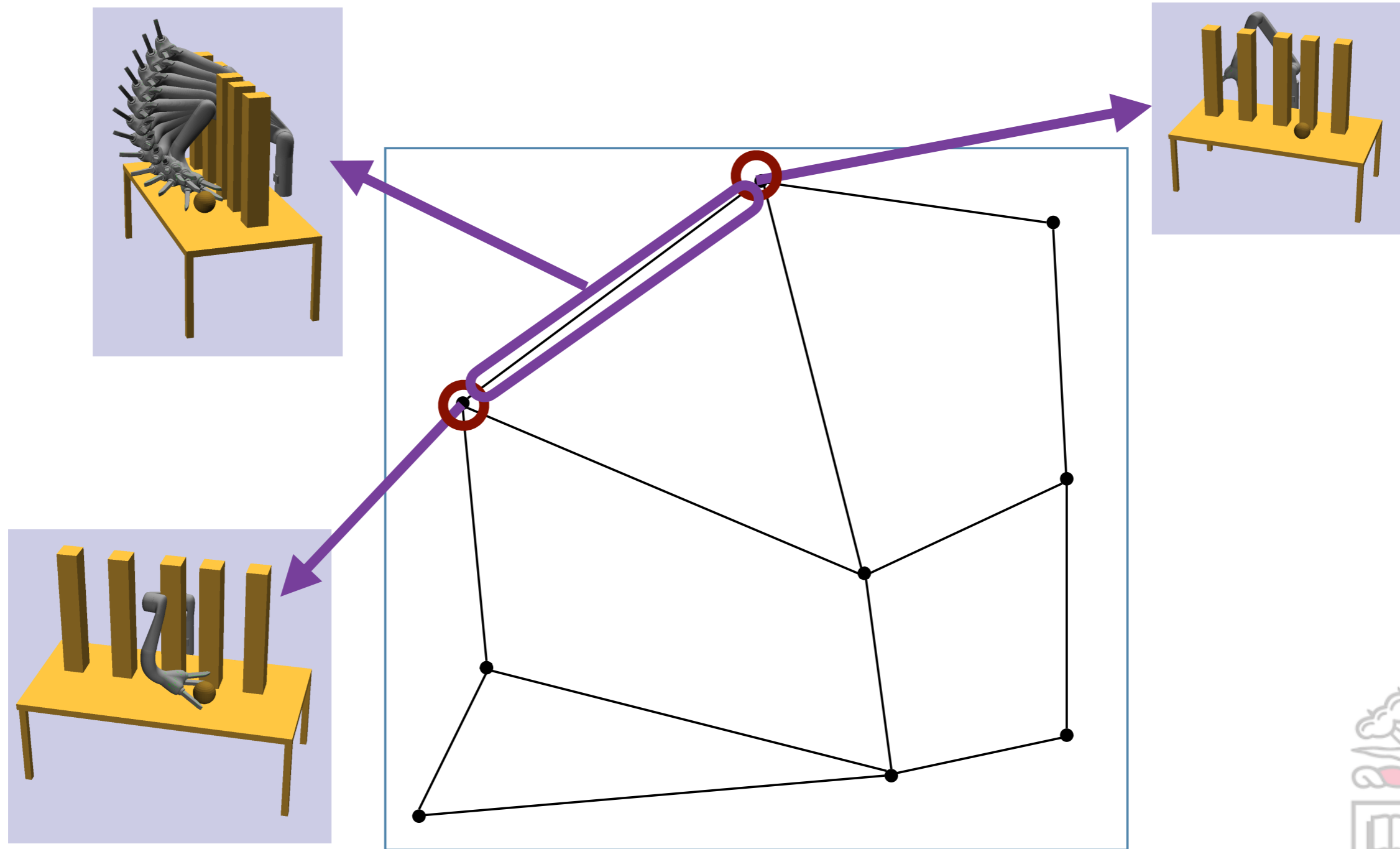# Randomized Algorithms

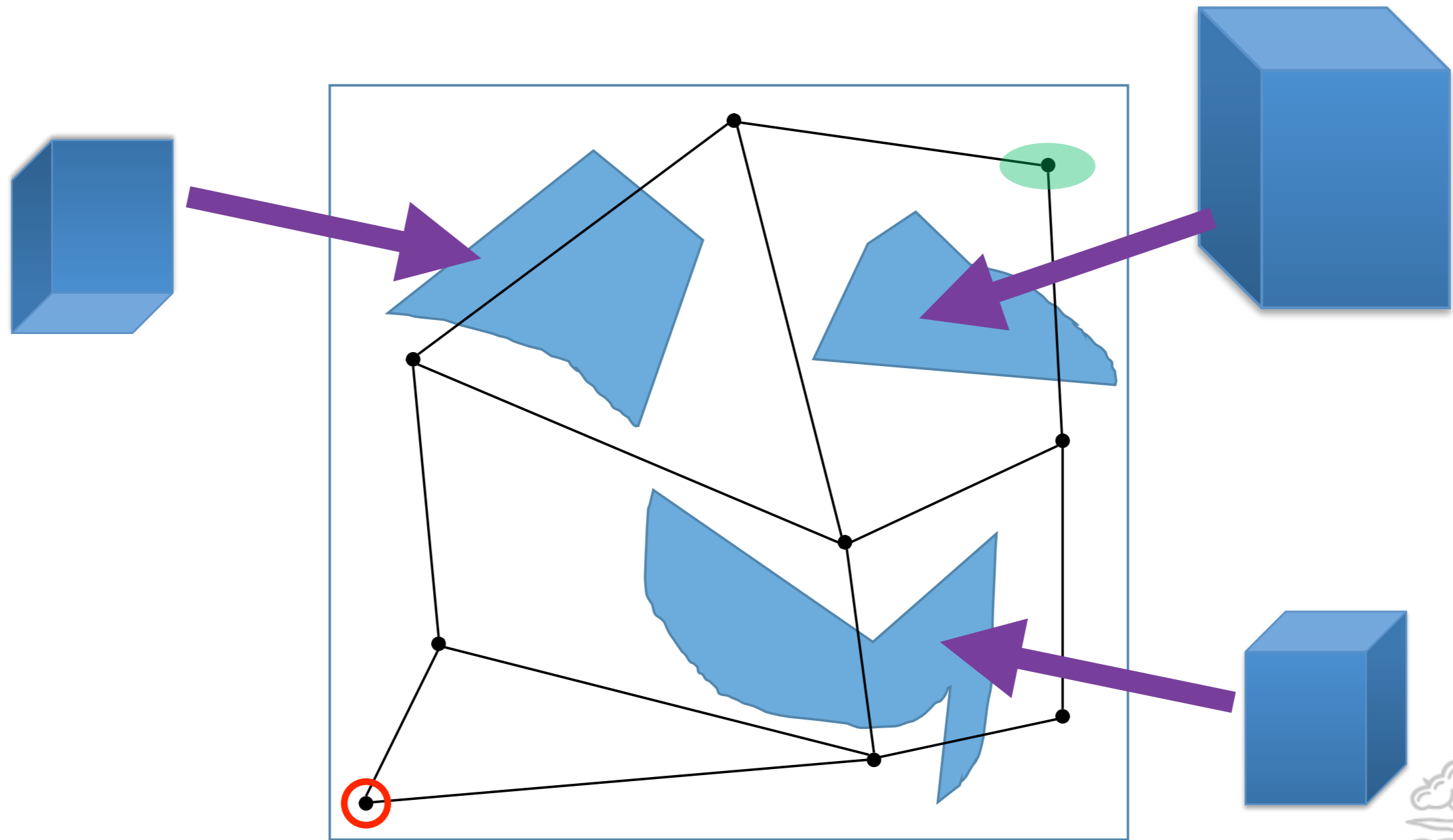Two major types:

Graphs
(multi-query)

Trees
(single-query)

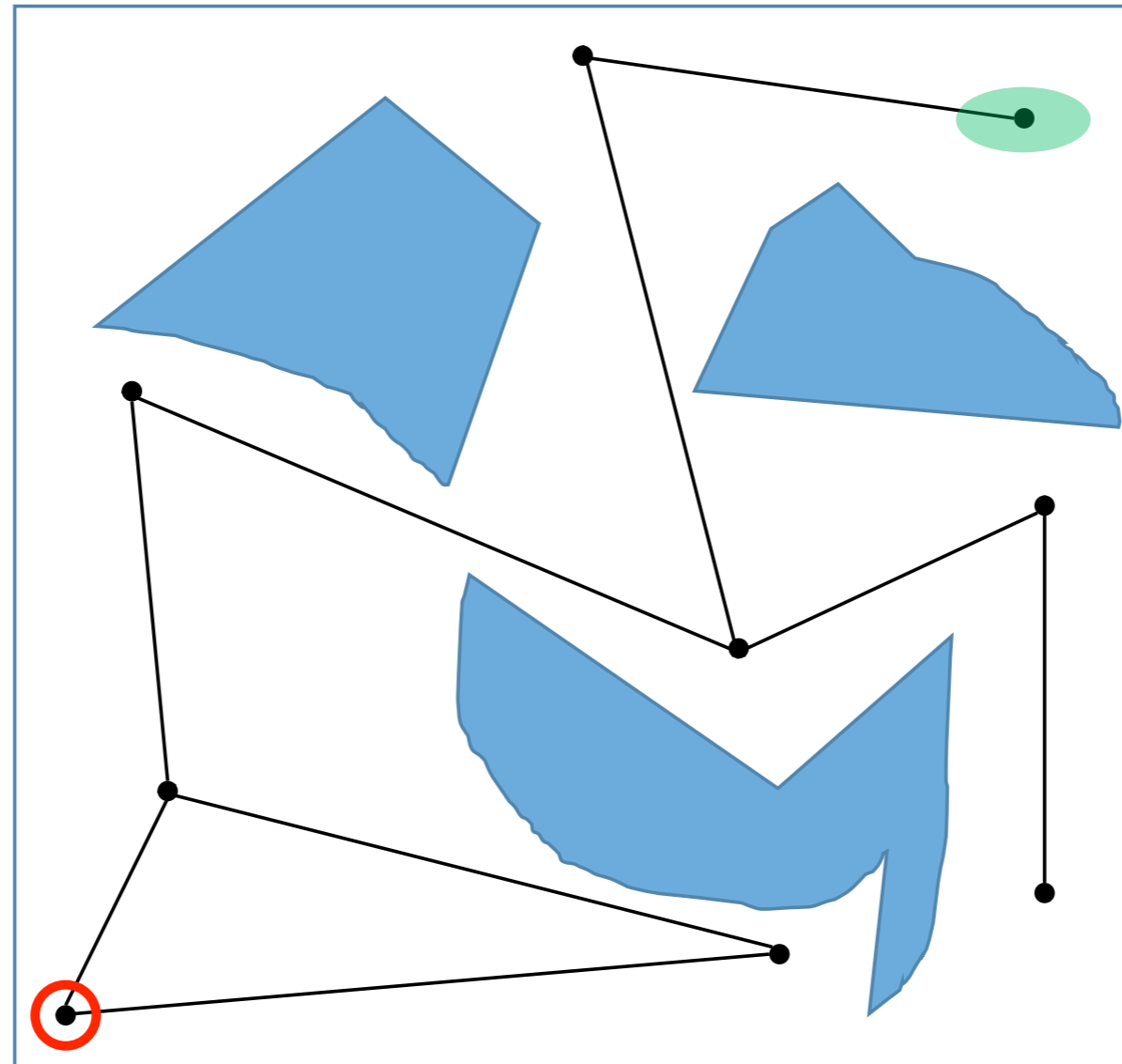# Probabilistic Roadmaps

[Leven and Hutchinson 2002]

PRMs

[Leven and Hutchinson 2002]

PRMs

[Leven and Hutchinson 2002]

Pros and Cons

Pros

- Initial computation of PRM can be slow
- Reused in many scenarios
- Very simple algorithm

Cons

- Must precompute PRM!
- Collision: 99% of compute time [Bialkowski et al. 2011]
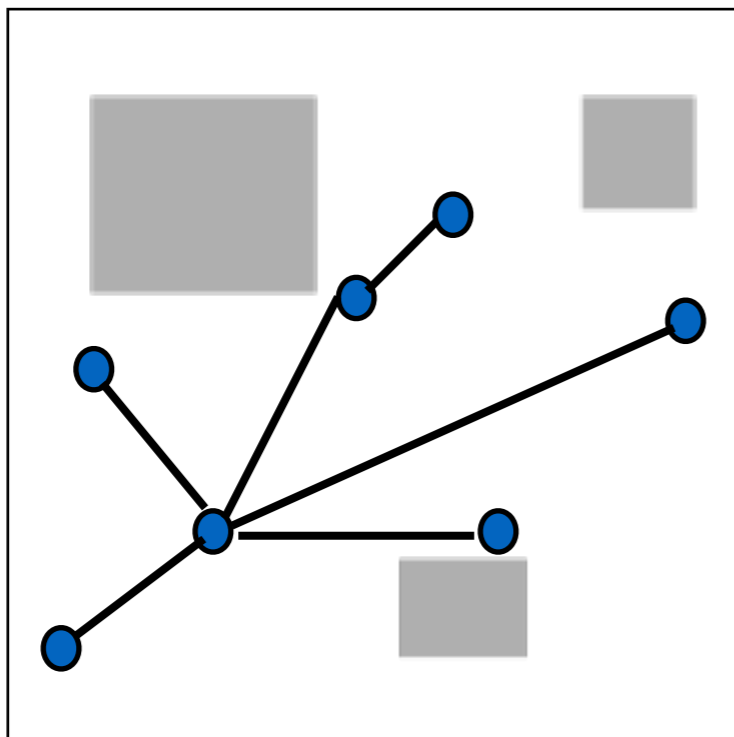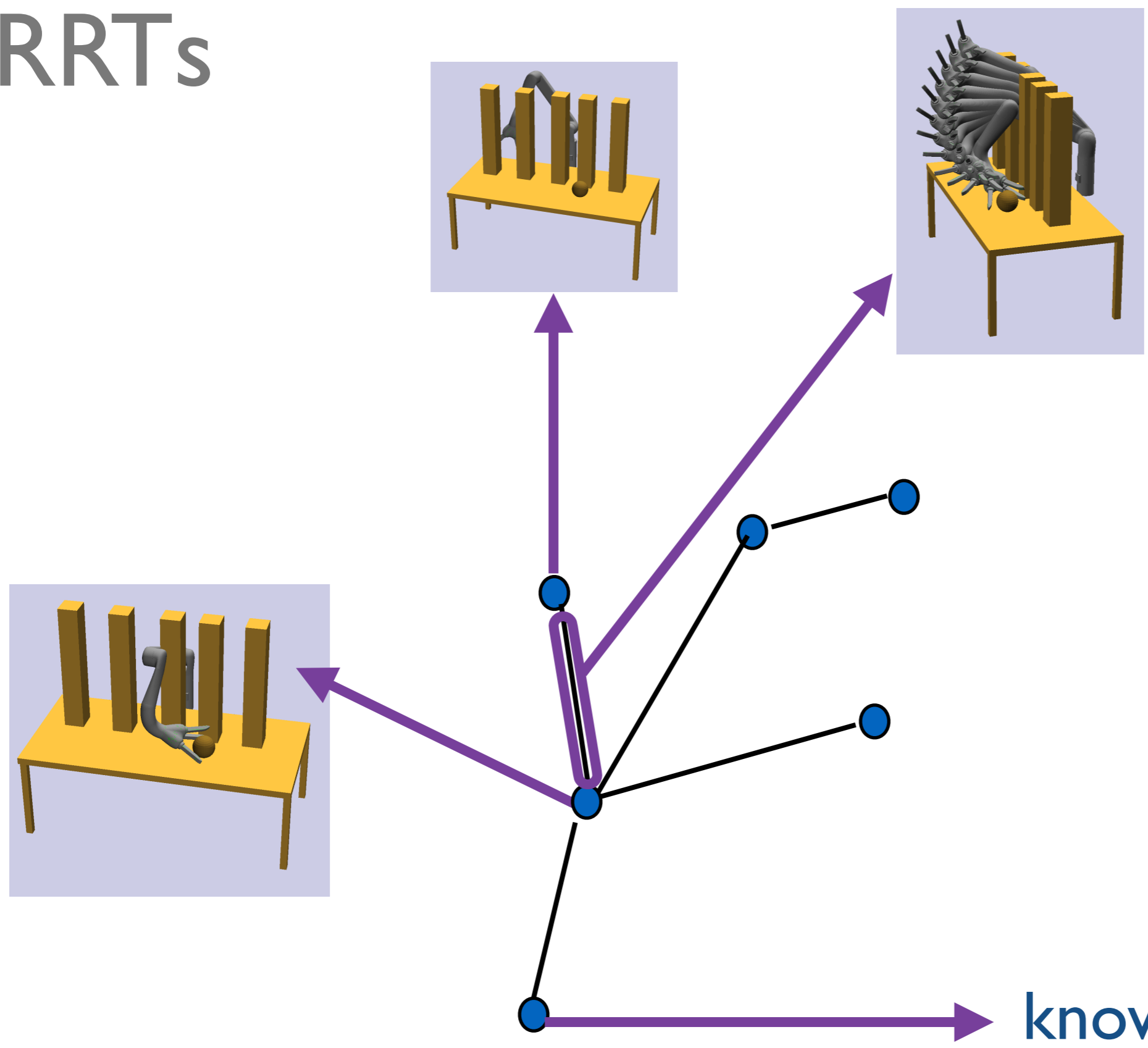- ***Just as fast (or faster) to recompute***

RRTs

Don't build a graph in advance - build a tree at query time!

Rapidly Exploring Random Trees
- Build a tree *starting from the start state.*
- Sample in C-space at random
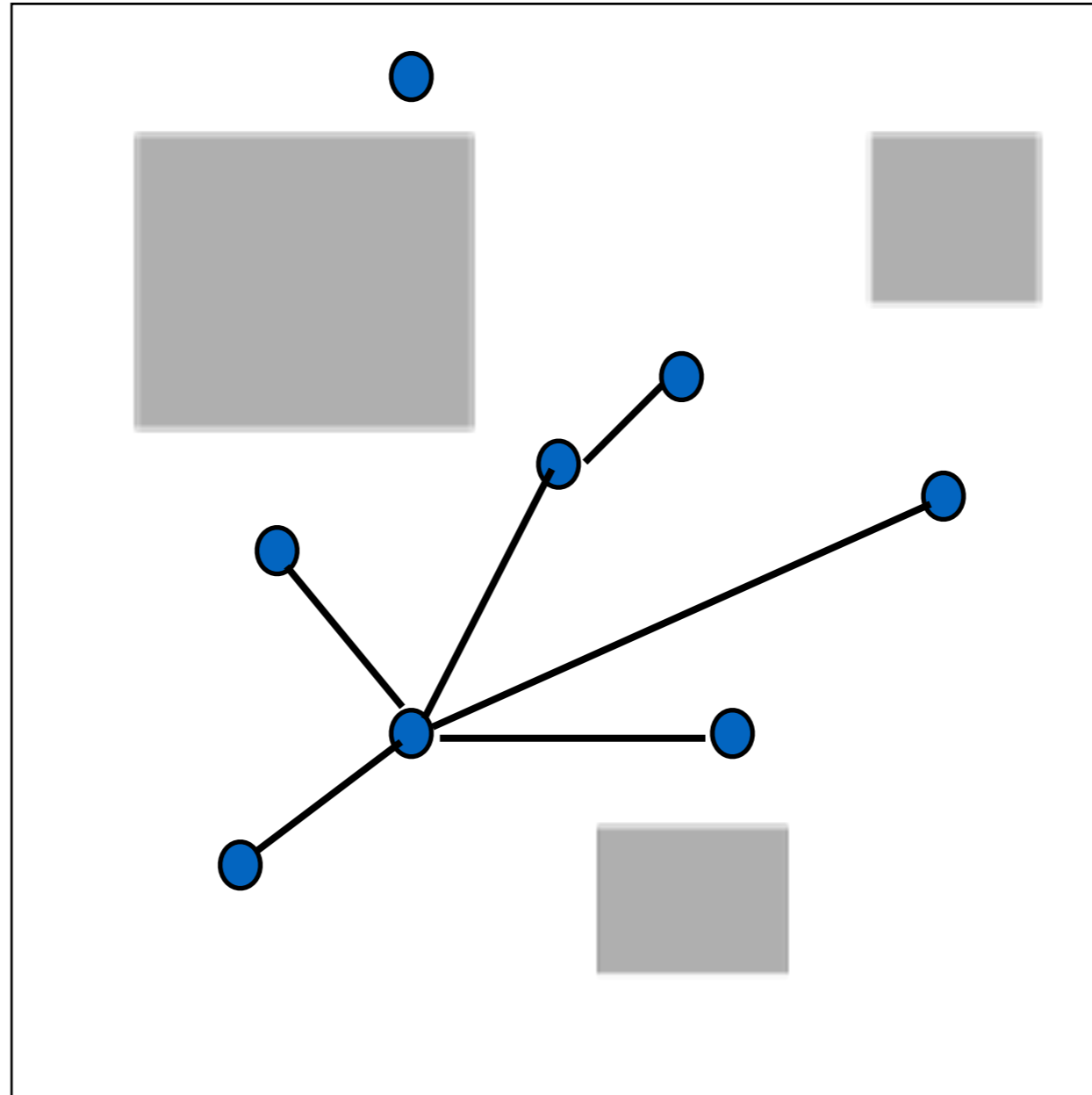- Try to connect sample to tree
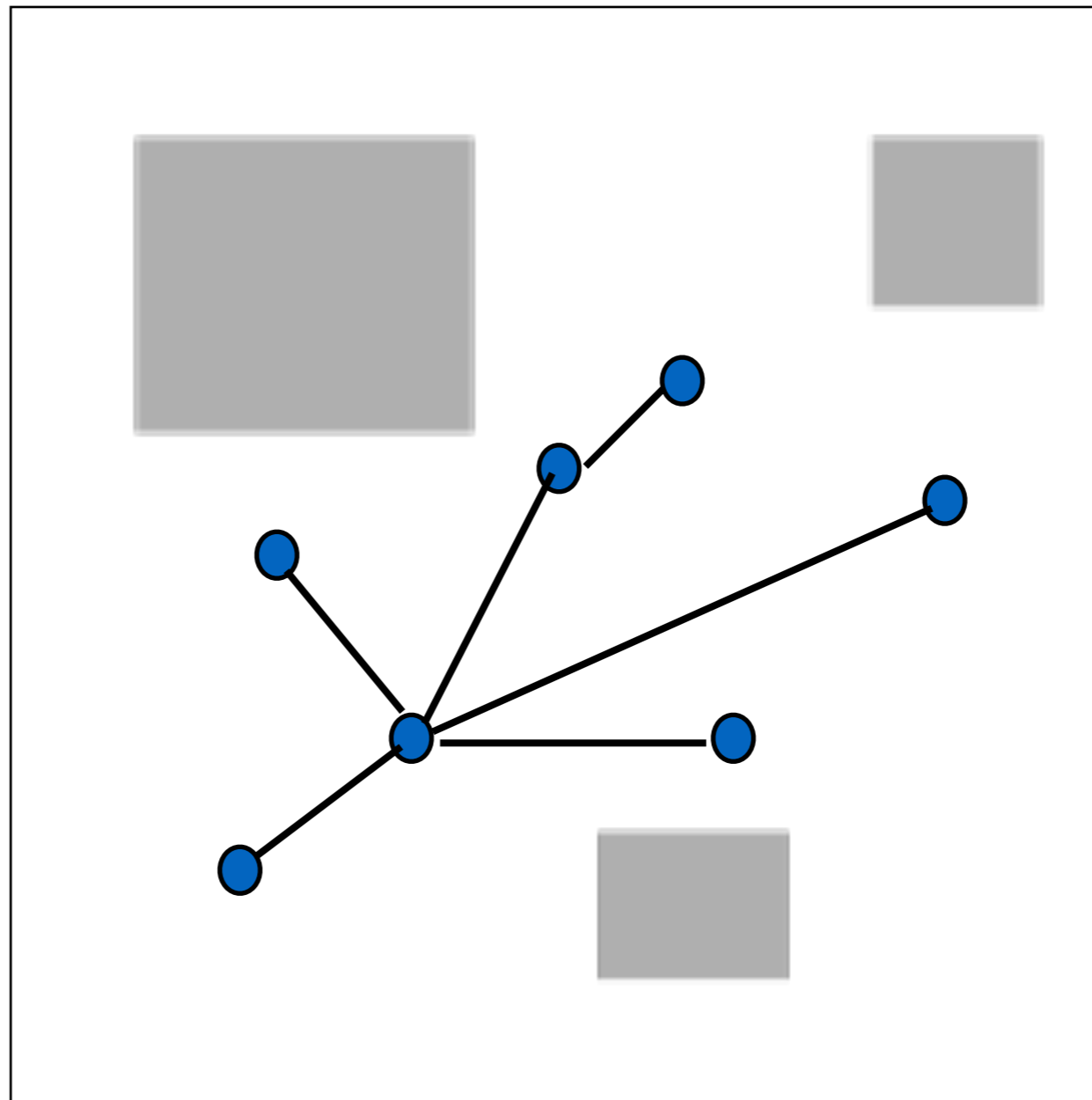- Stop when you hit the goal

RRTs

known start state

RRTs

RRT

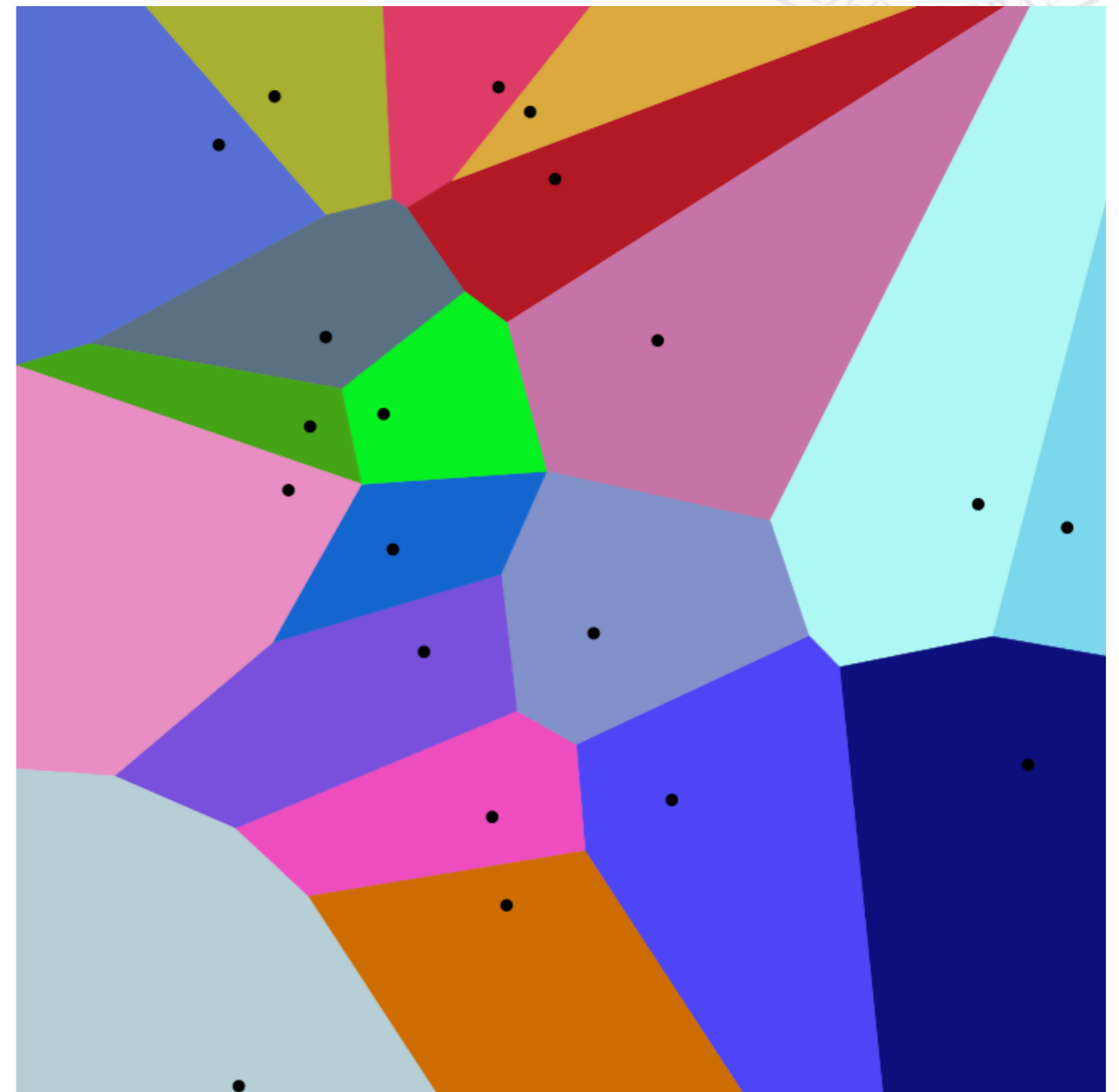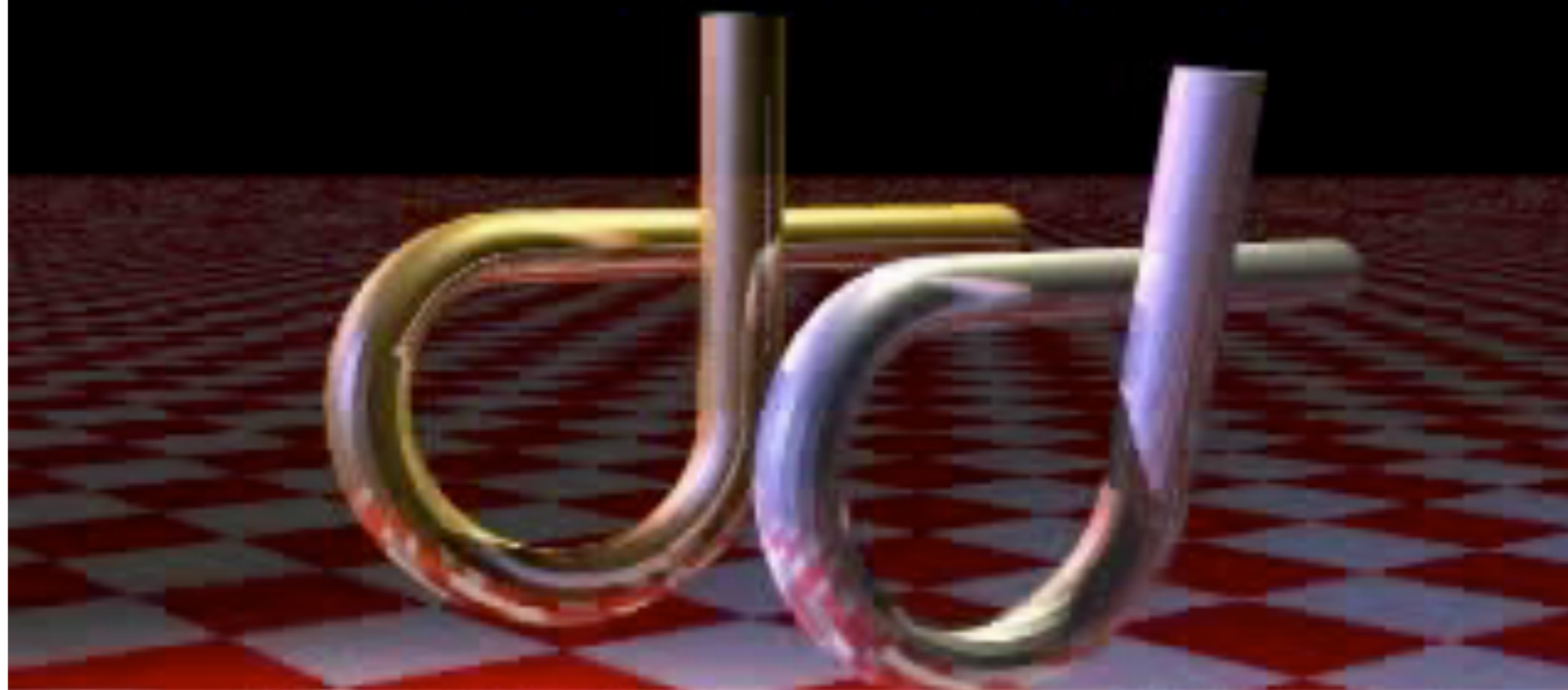Property: the tree *rapidly expands* to fill free space.

Why?

RRT: Voronoi Bias

Property: the tree *rapidly expands* to fill free space.

(via Steve LaValle)

More Videos

More Videos

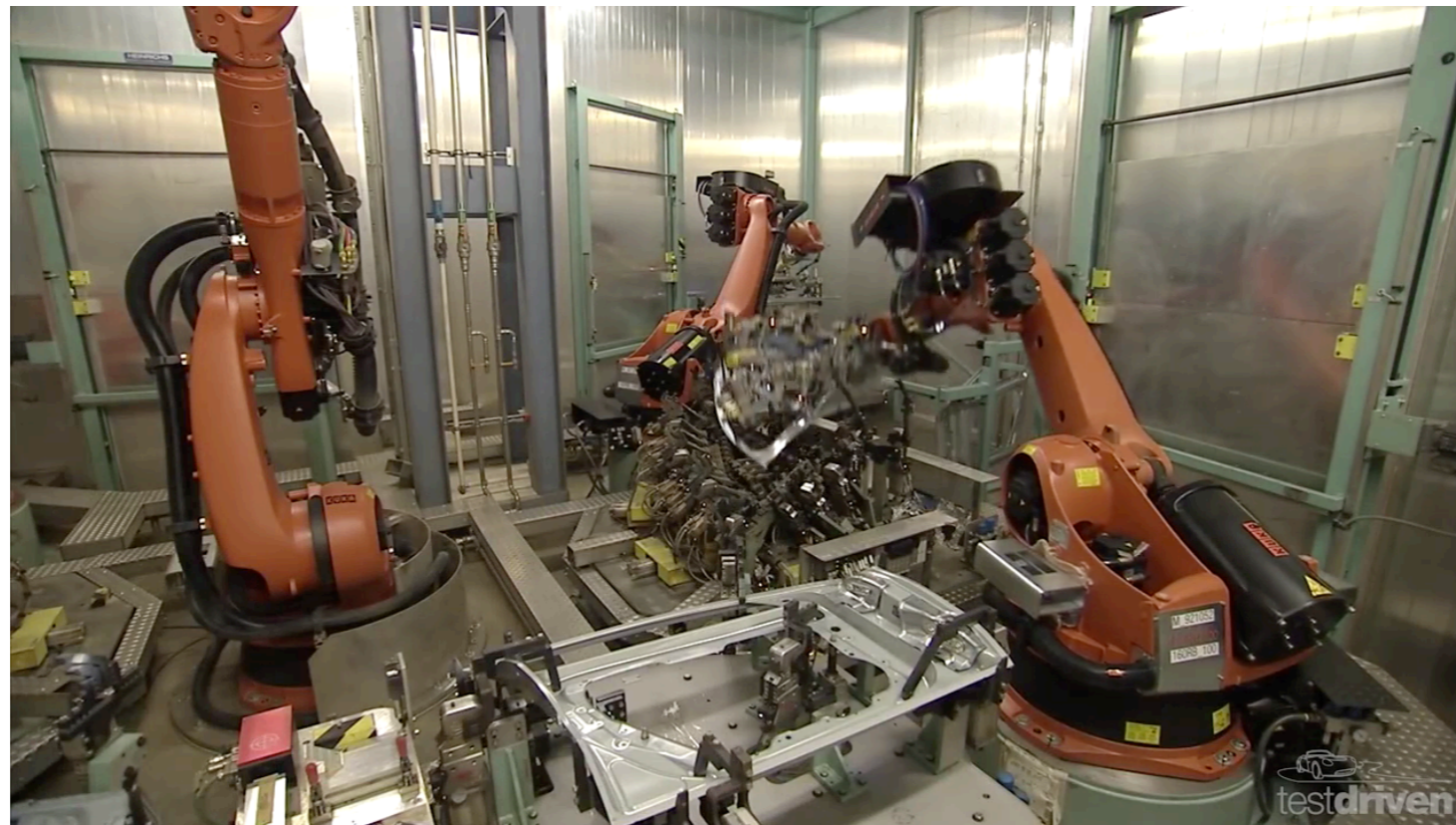# Robot Motion Planning

Critical for robots in **semi/un-structured** environments.

But:

watch this space
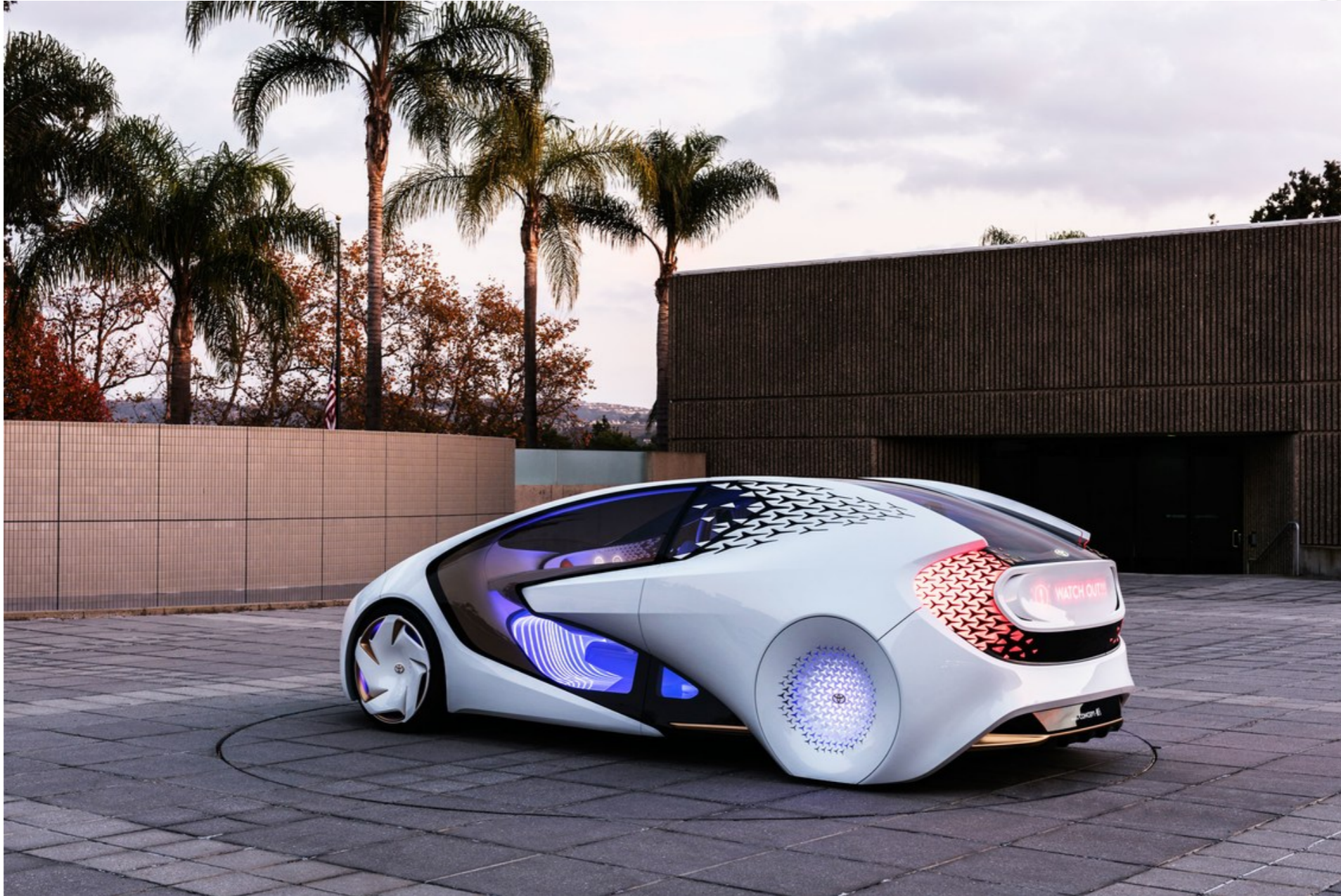
- Fundamentally hard.
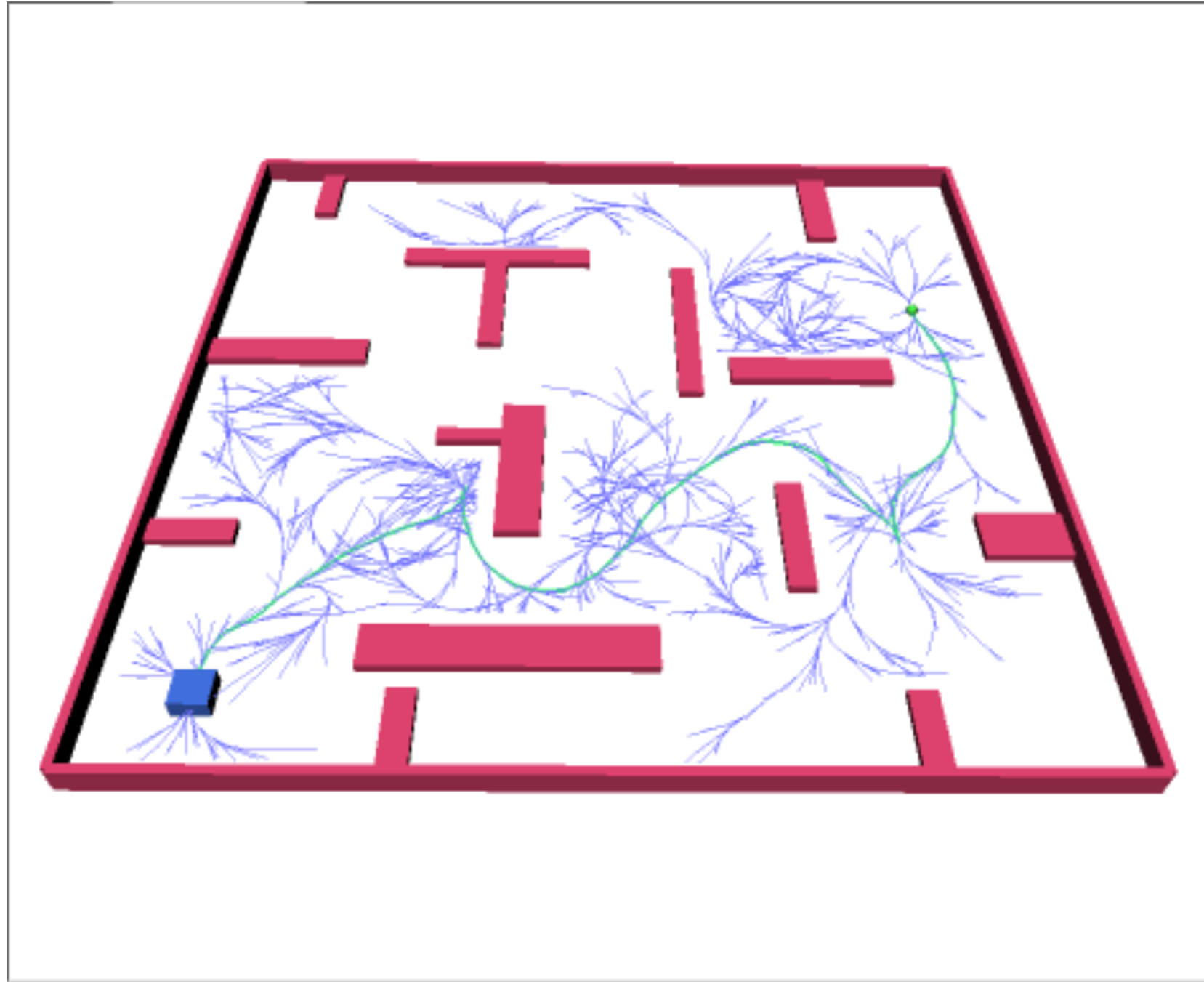- Very well studied (30 years)
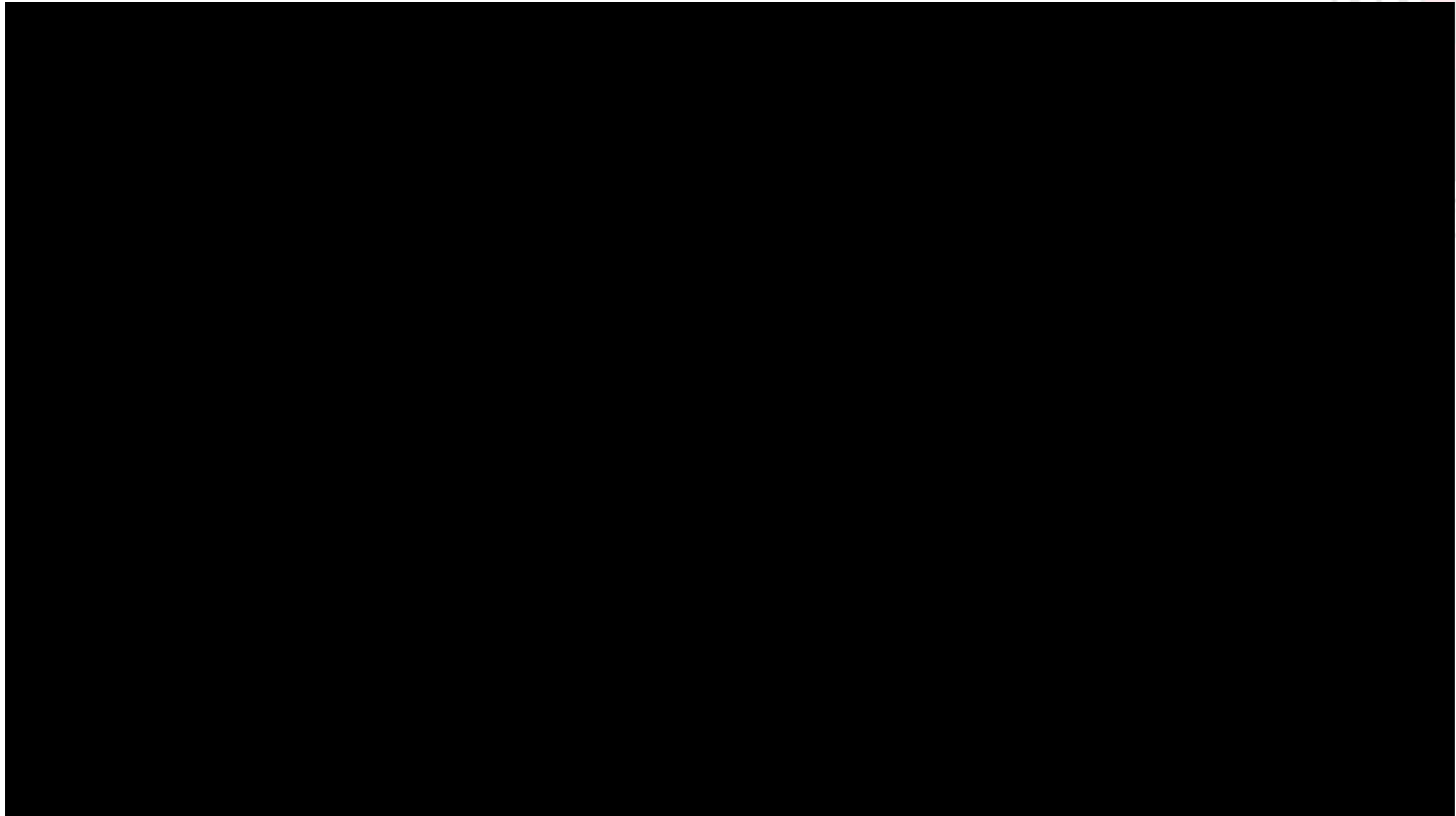- No real-time solutions.

Autonomous Cars

Autonomous Cars

(via Steve LaValle)

Video

Chen, Rickert, and Knoll IROS 2015