

An Introduction to Motion Planning

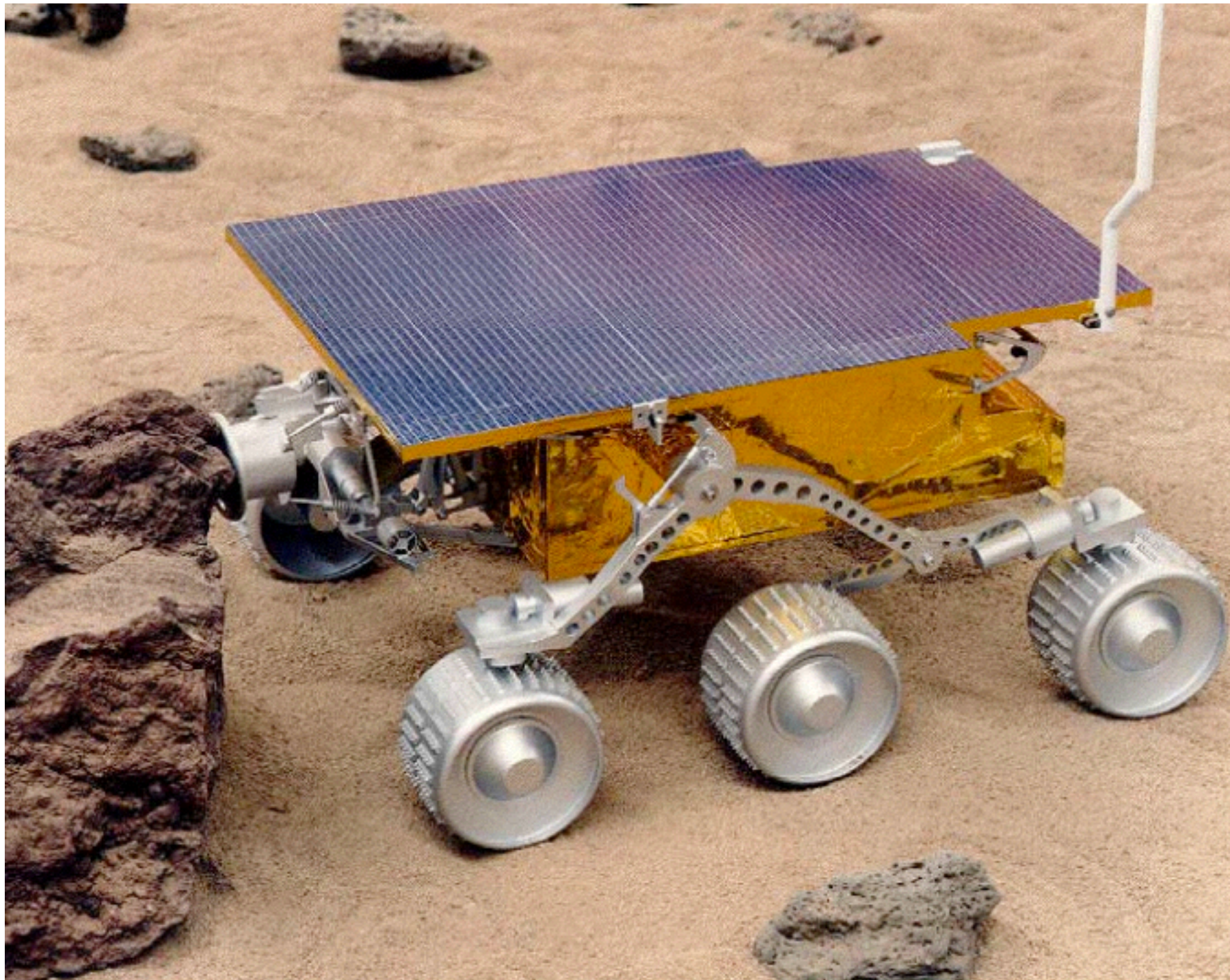
Alan Kuntz

2/4/15

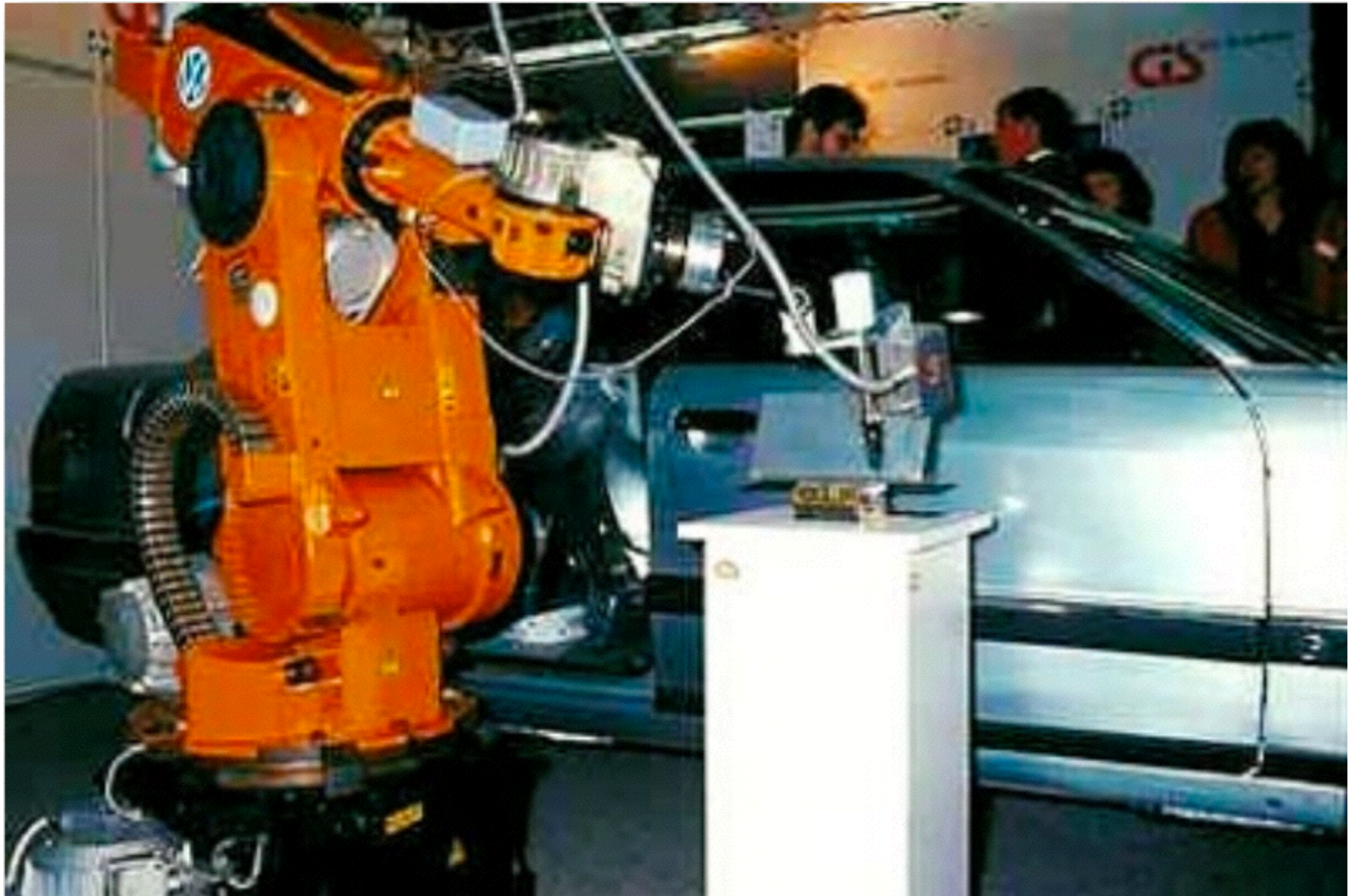
<http://cs.unc.edu/~adkuntz/MotionPlanning.pdf>



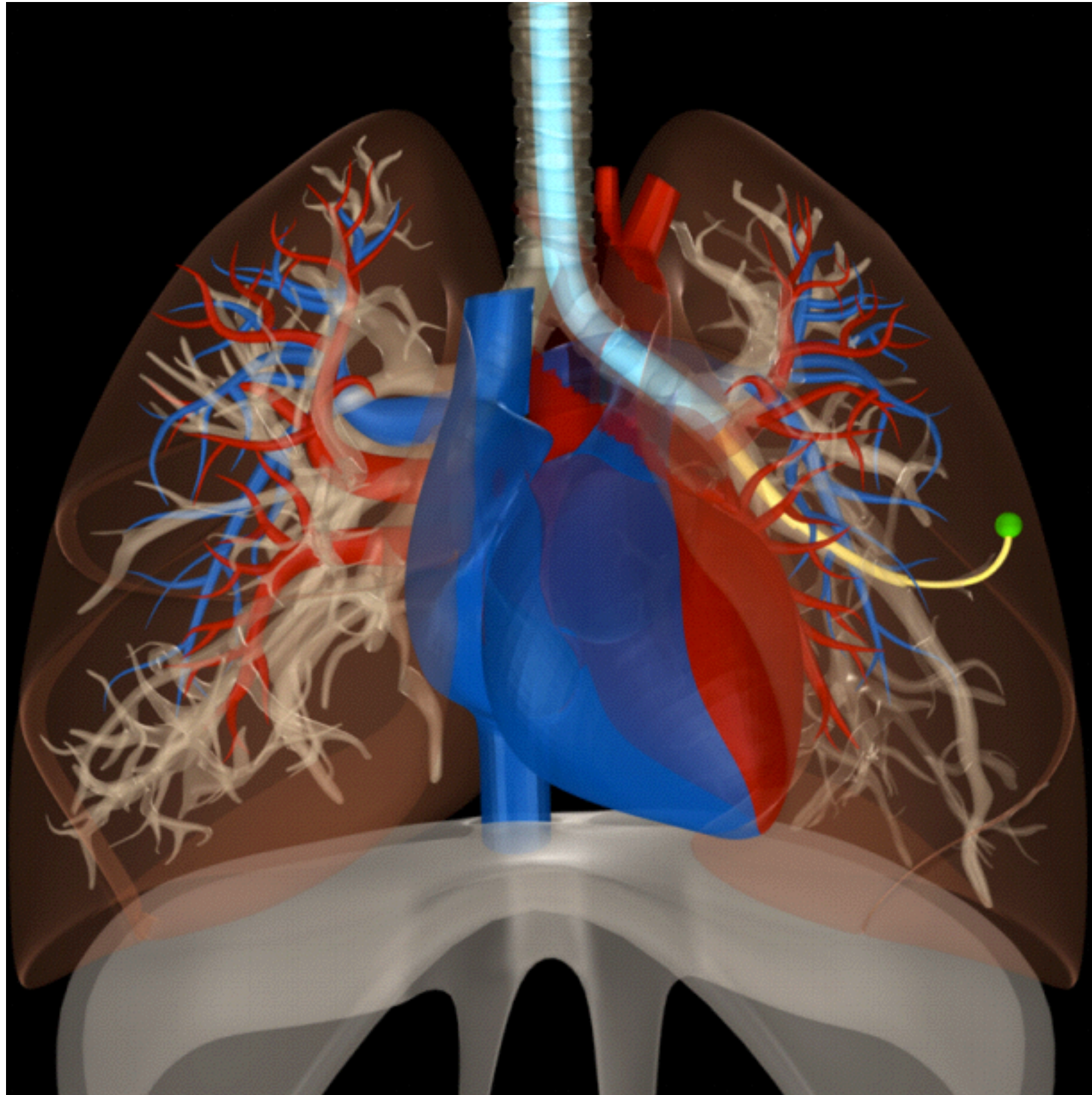
Applications



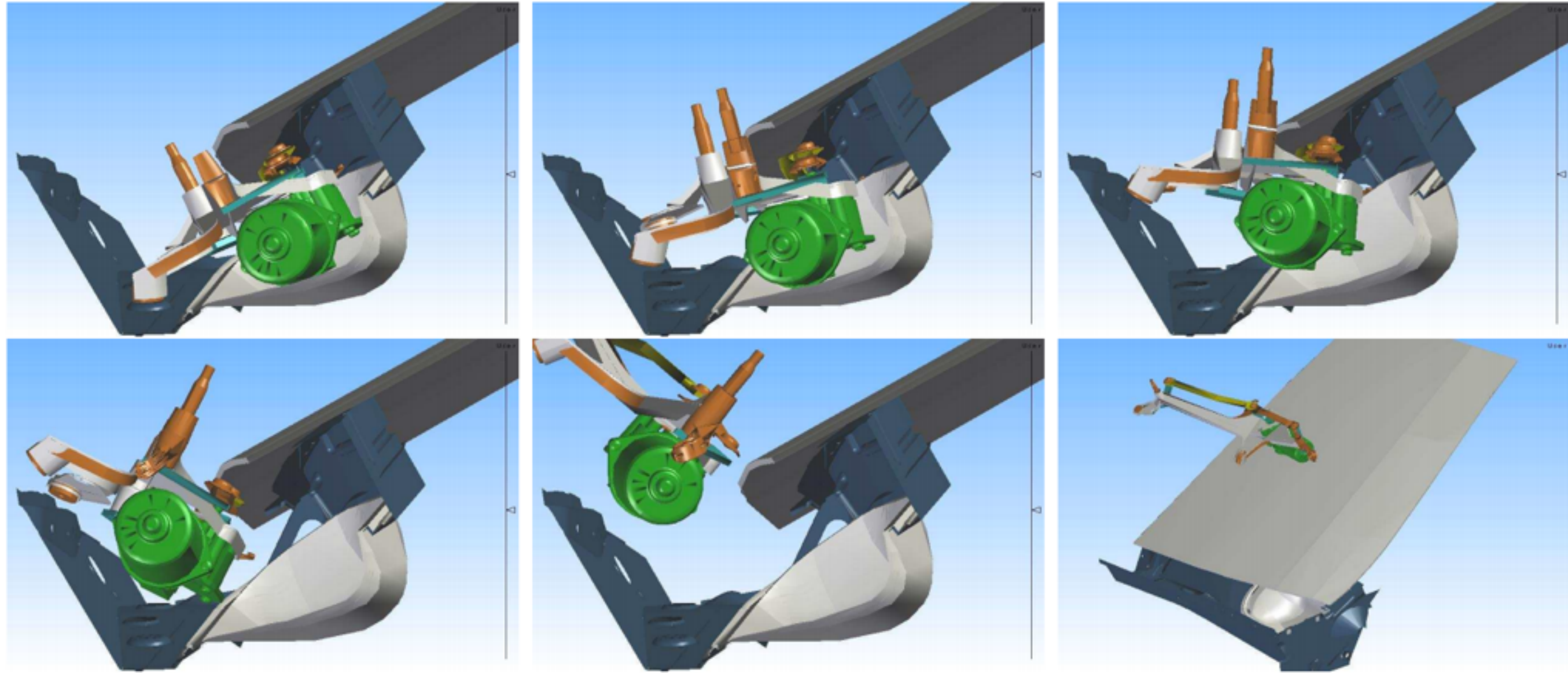
Applications



Applications



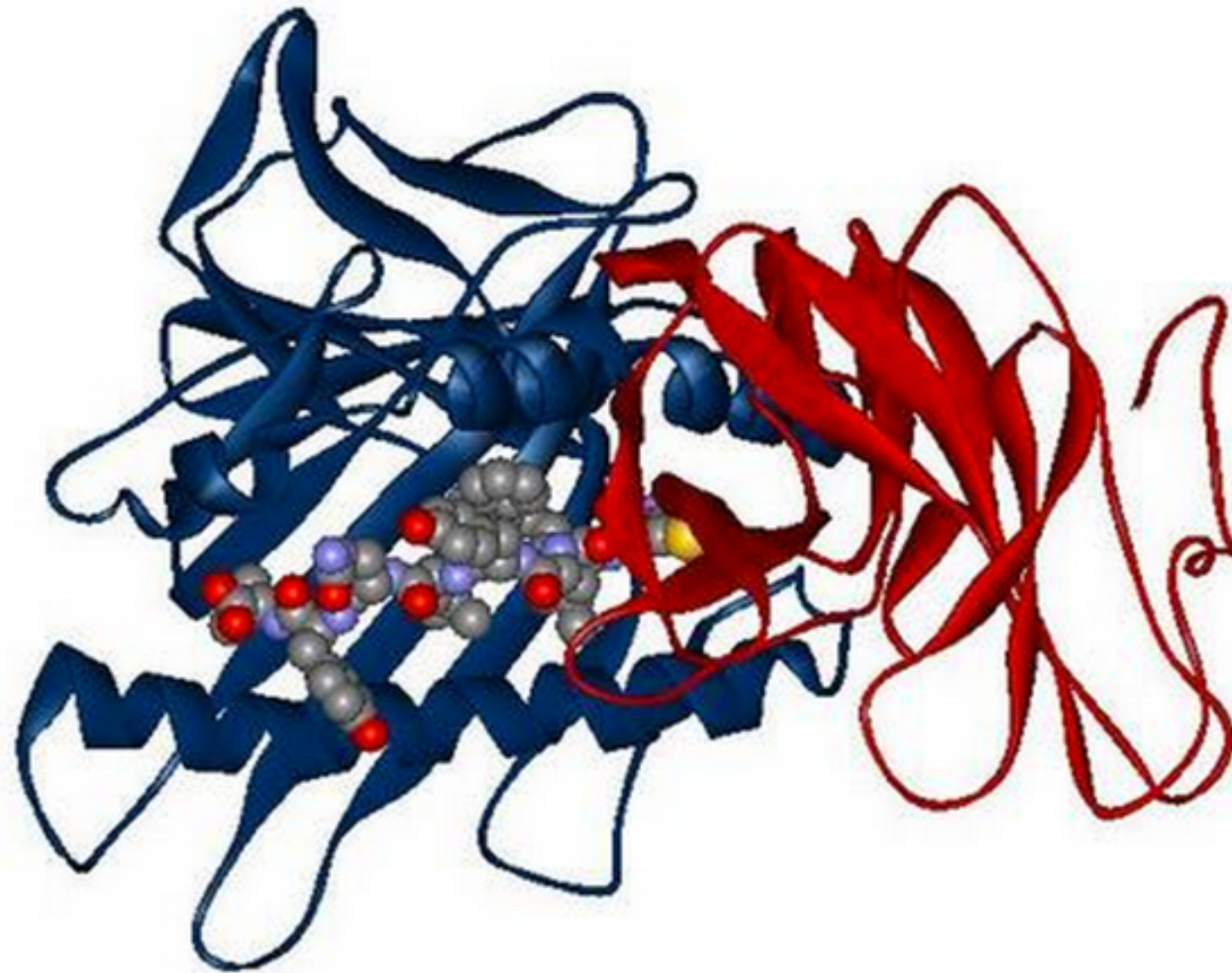
Applications



planning.cs.uiuc.edu



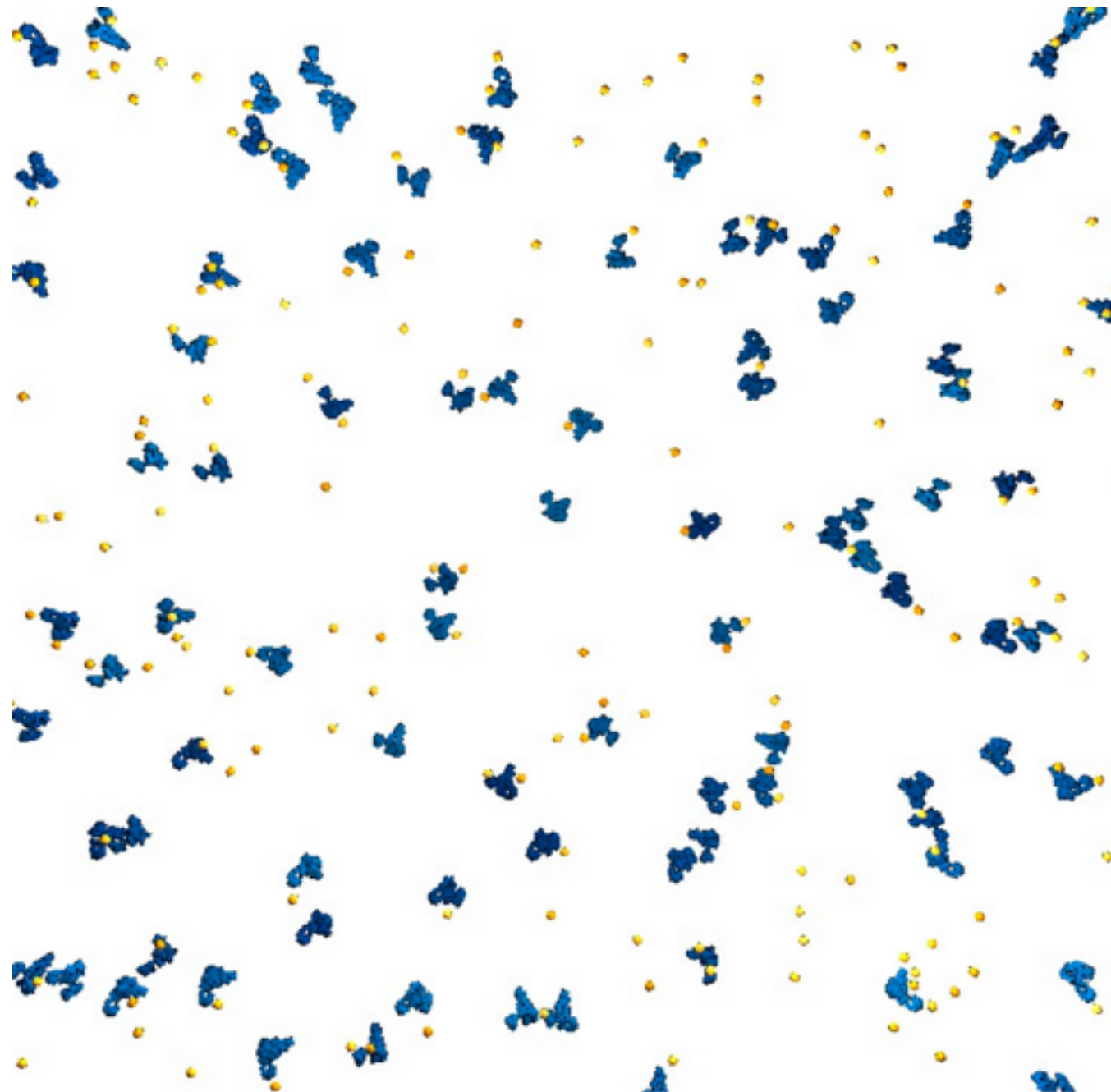
Applications



cs.unm.edu/amprg



Applications



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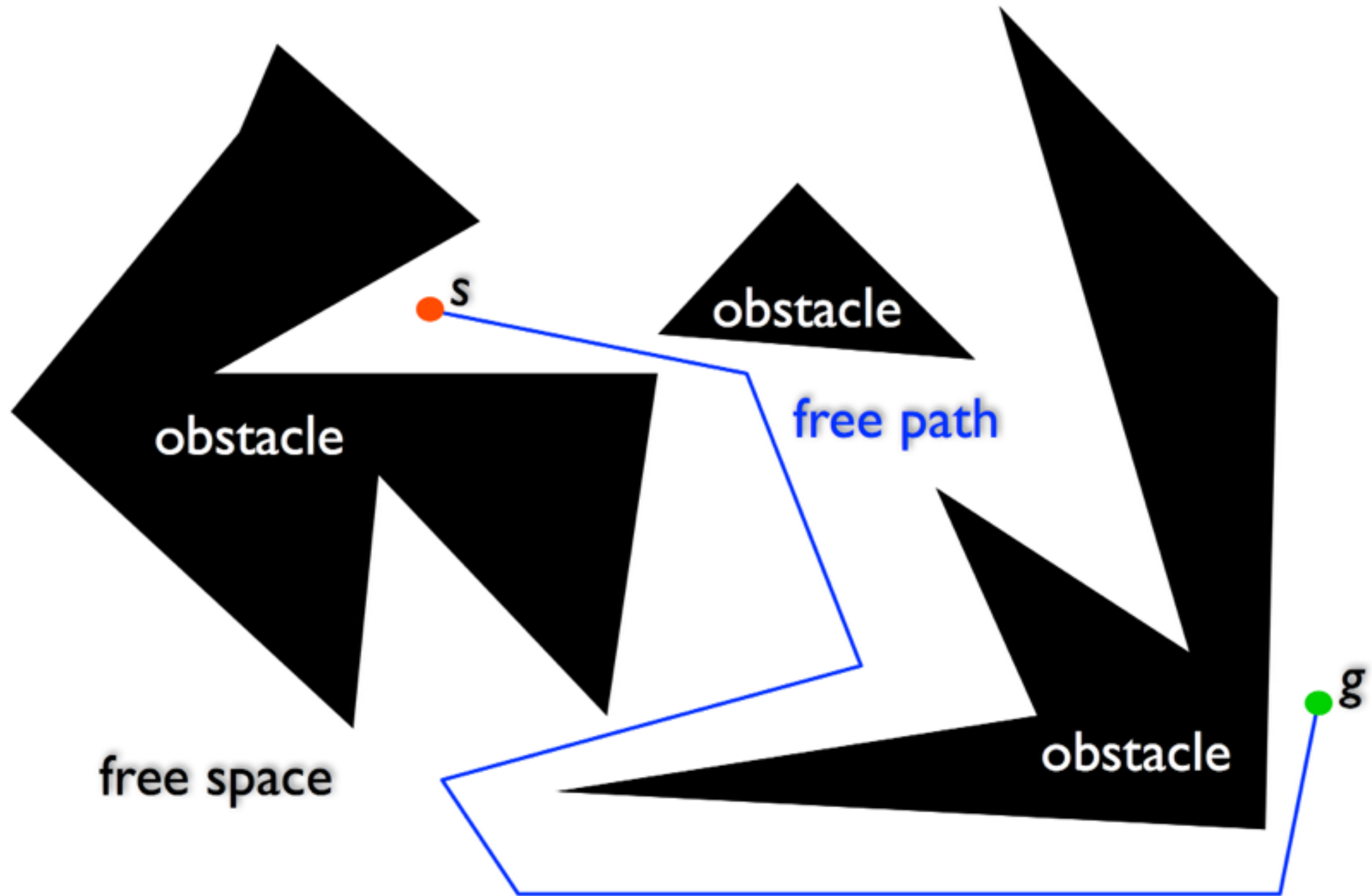


Problem Definition

- Compute a collision-free path for the robot/agent from a start configuration to a goal configuration
- Inputs
 - Geometry of robot/agent
 - Geometry of environment
 - Start and goal configurations
- Outputs
 - Continuous sequence of robot/agent configurations connecting the start and goal configurations



Problem Definition



Problem Definition

- Complete - Always return a solution plan if one exists, otherwise indicate there isn't one
- Optimal - Always return the best solution plan under some value metric

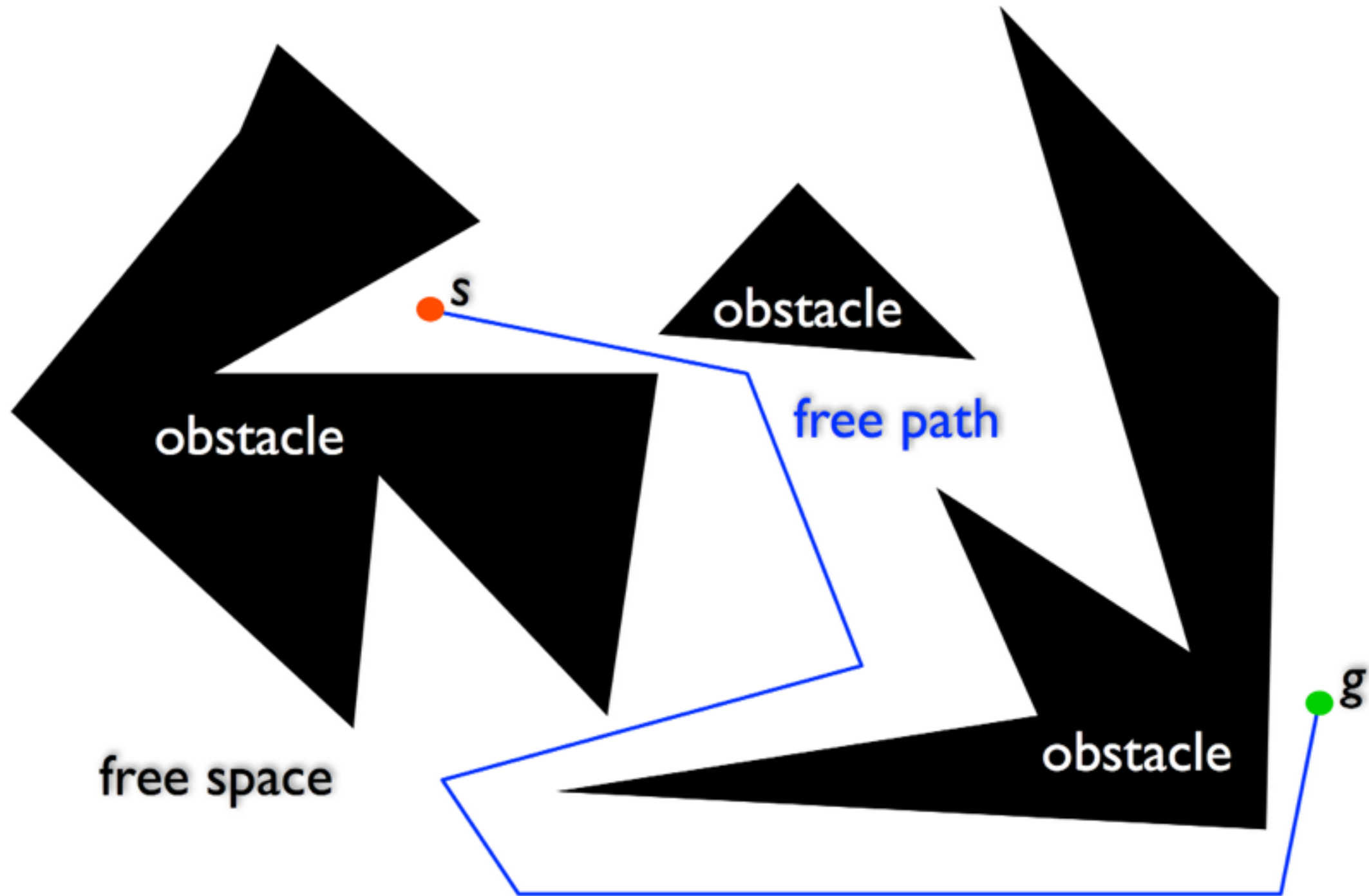


Problem Definition

- Completeness - In more than 2D, PSPACE-hard
- Exponential in DOFs, number of obstacles, etc.
- May require computation of entire C-space.
- Doable in simple cases, like 2D with point robot. Easy because C-space is workspace



Problem Definition



Problem Definition

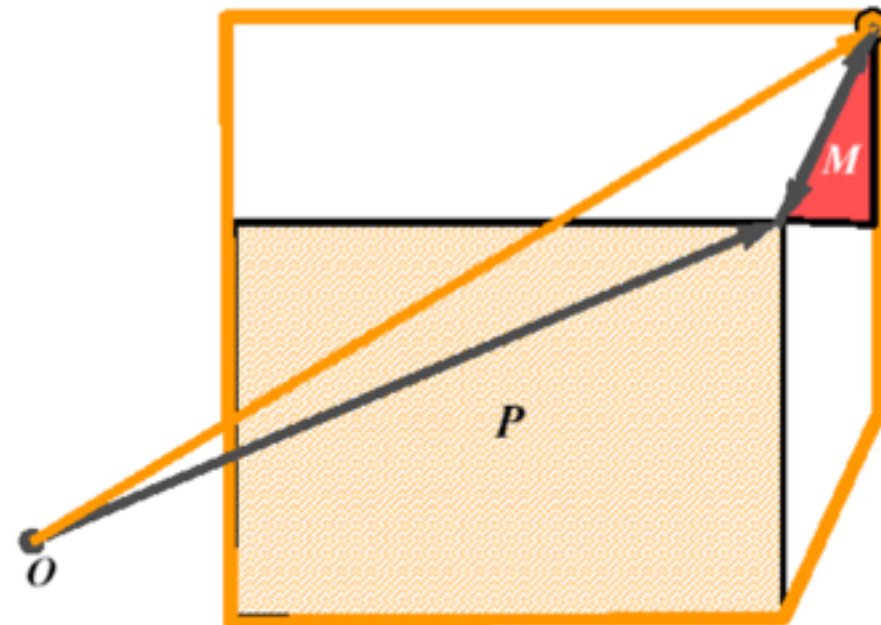
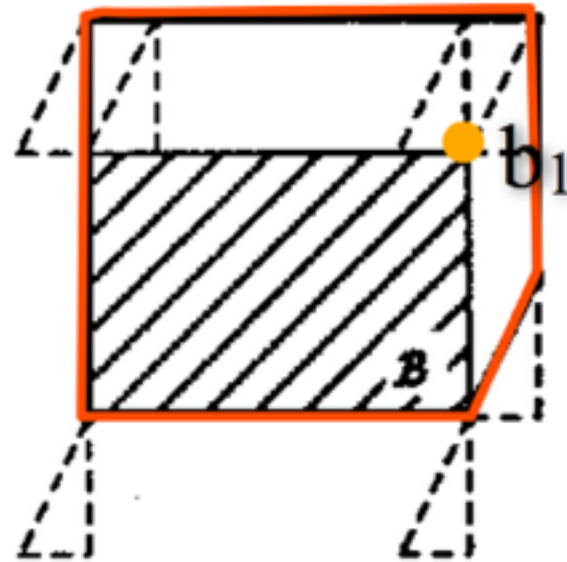
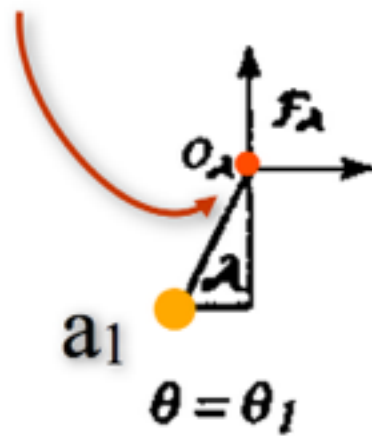
- What about for something more complex than a point?
- Next most complex - Polygonal robot that translates but does not rotate.
- Can also be done relatively easy in 2D space through Minkowski Sums/Differences



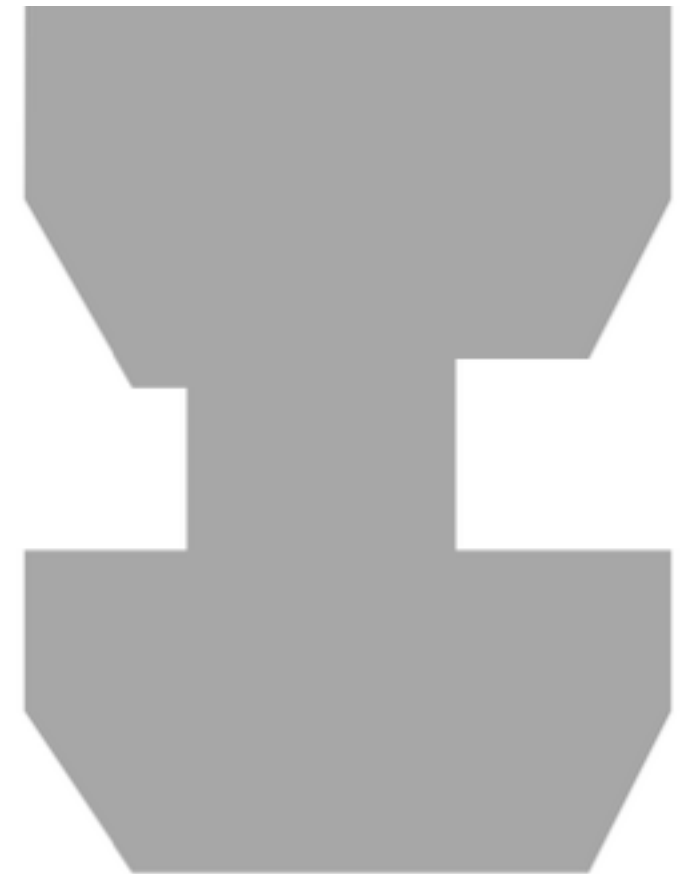
Minkowski Difference

$$CB = B \ominus A = \{b-a \mid a \in A, b \in B\}$$

reference
point (0,0)



Minkowski Difference



Obstacle

\mathcal{P}

Robot

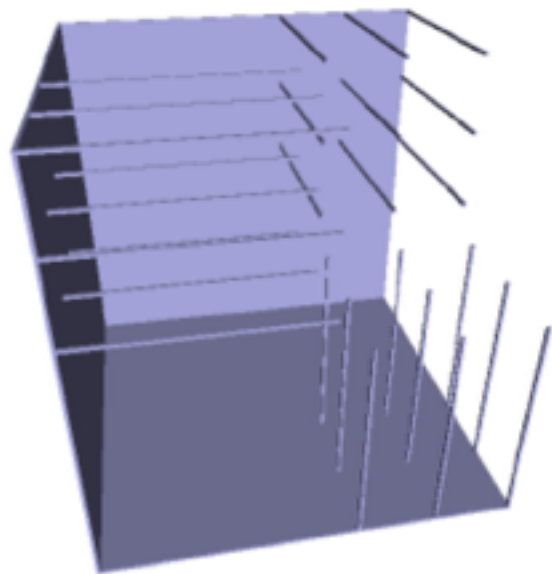
\mathcal{M}

C-obstacle

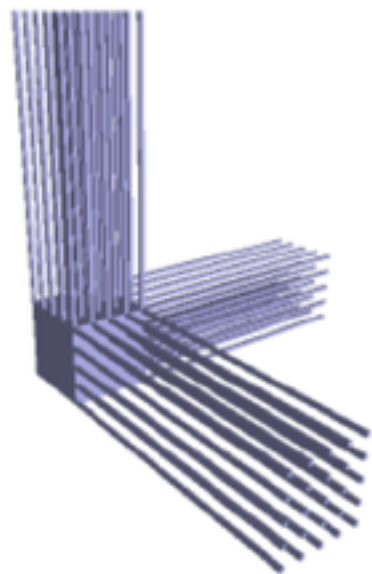
$\mathcal{P} \oplus -\mathcal{M}$

Classic result by Lozano-Perez and Wesley 1979

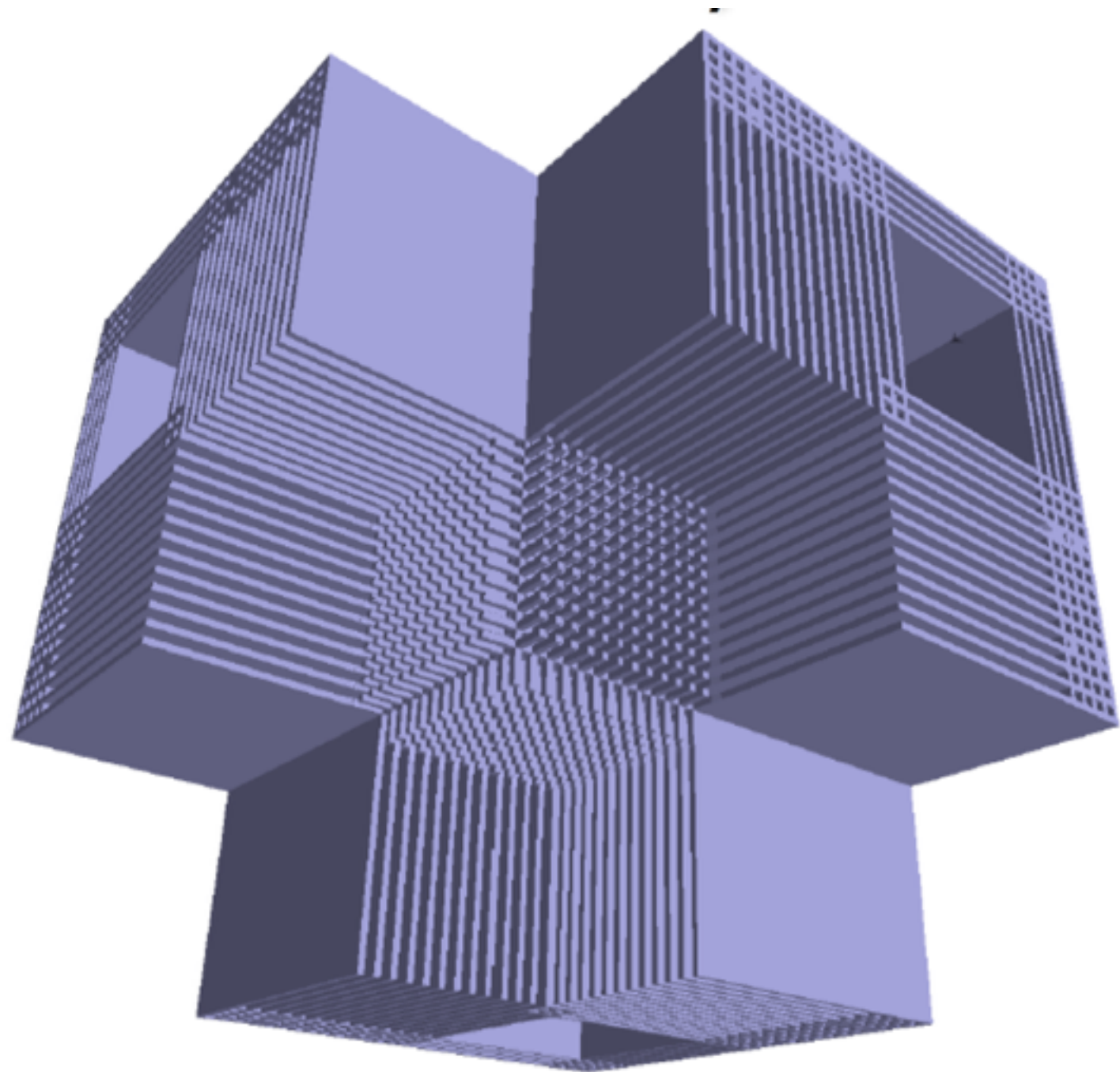
Minkowski Difference



Grate 1 (444 tris)



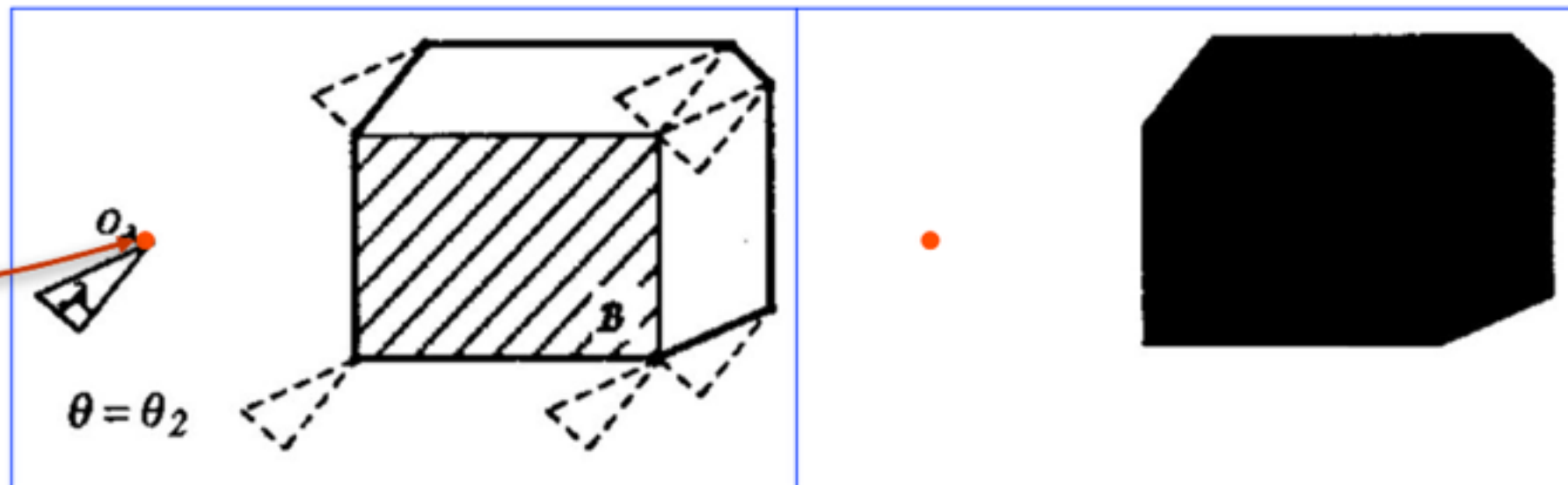
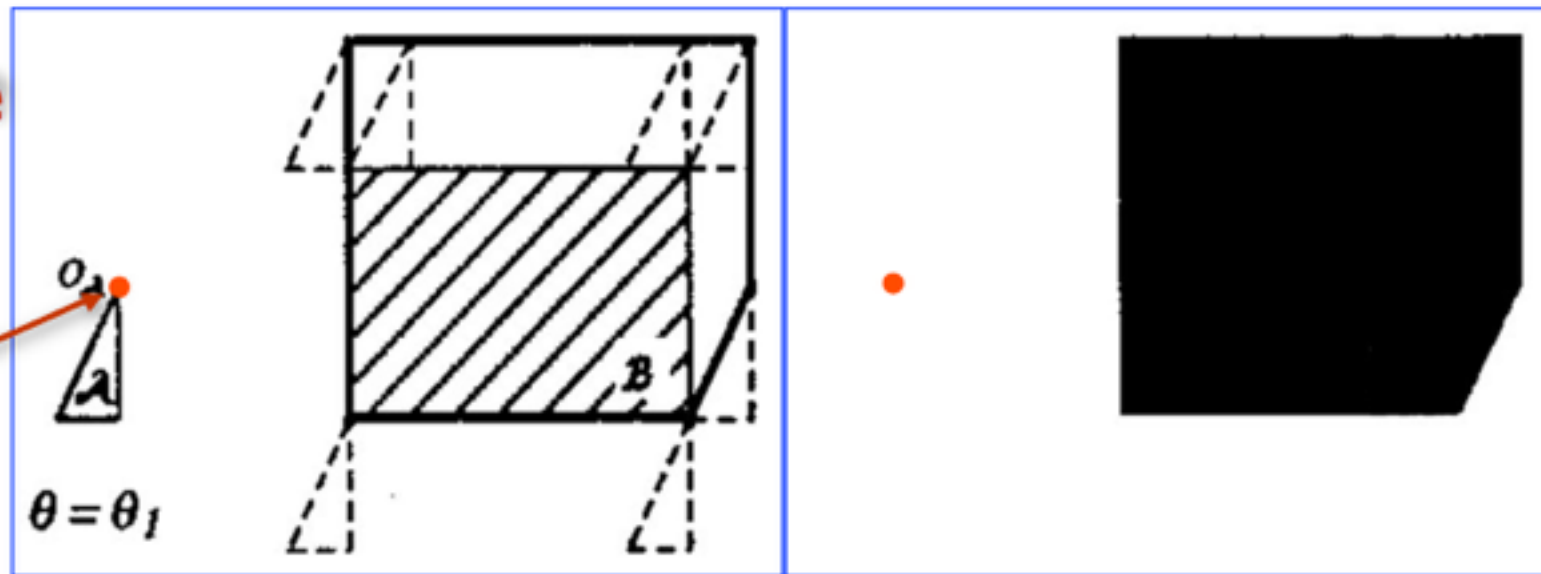
Grate 2 (1, 134 tris)



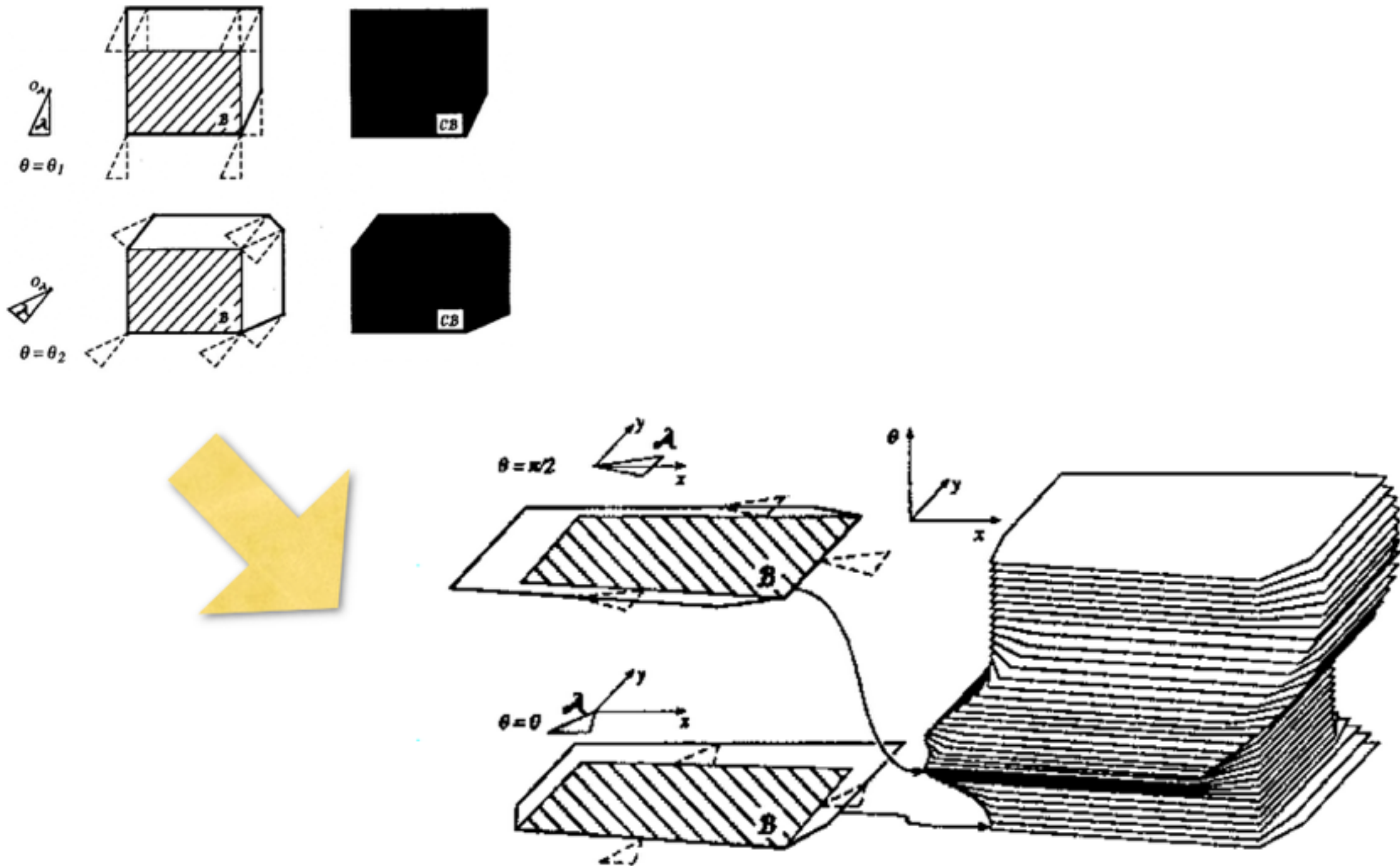
Grate 1 \oplus Grate 2 (Union of 66, 667 prims, 358K tris)

Minkowski Difference

reference
point



Minkowski Difference



Minkowski Difference

- That's an obstacle in C-space for a mobile robot.
- The problem has now become to navigate a point through this higher dimensional space.

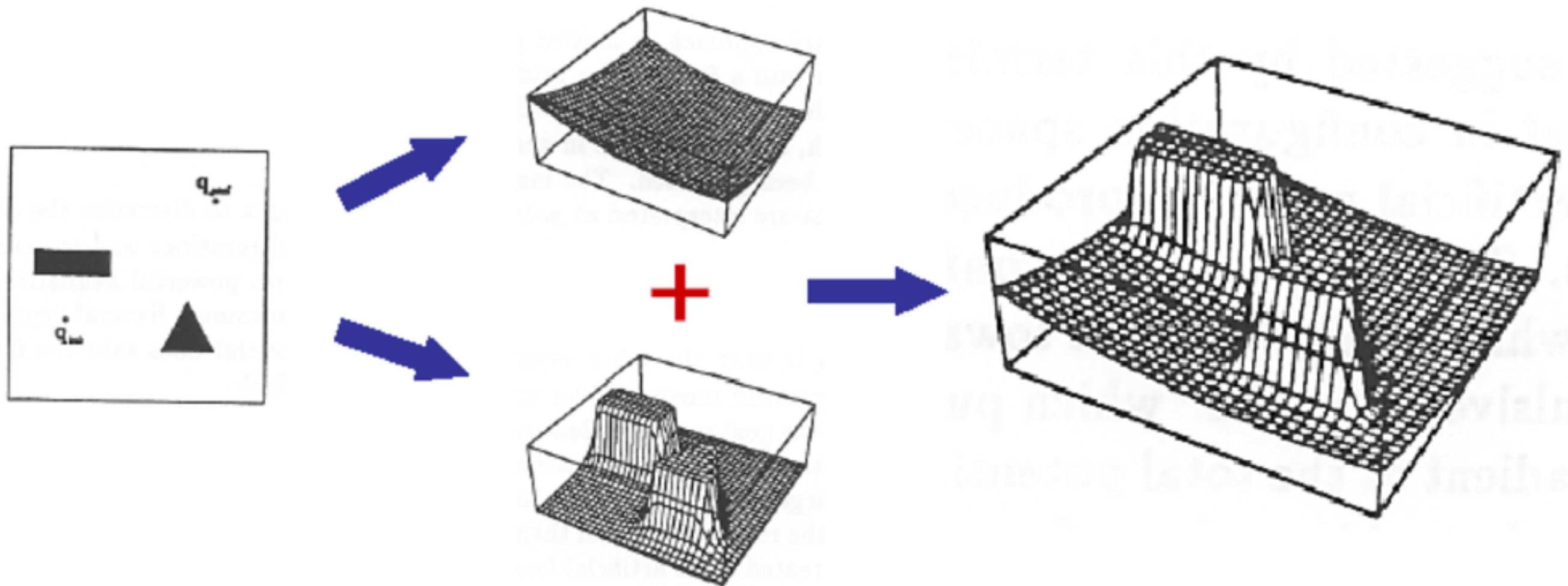


The Point

- A huge amount of motion planning concerns itself with navigating a point through some n -dimensional space.
- Why a point?
 - Points are easy.
 - Lines, Vectors, Graphs etc.
- Is this even useful?
 - Abstraction
 - Approximation
- How?



Potential Fields

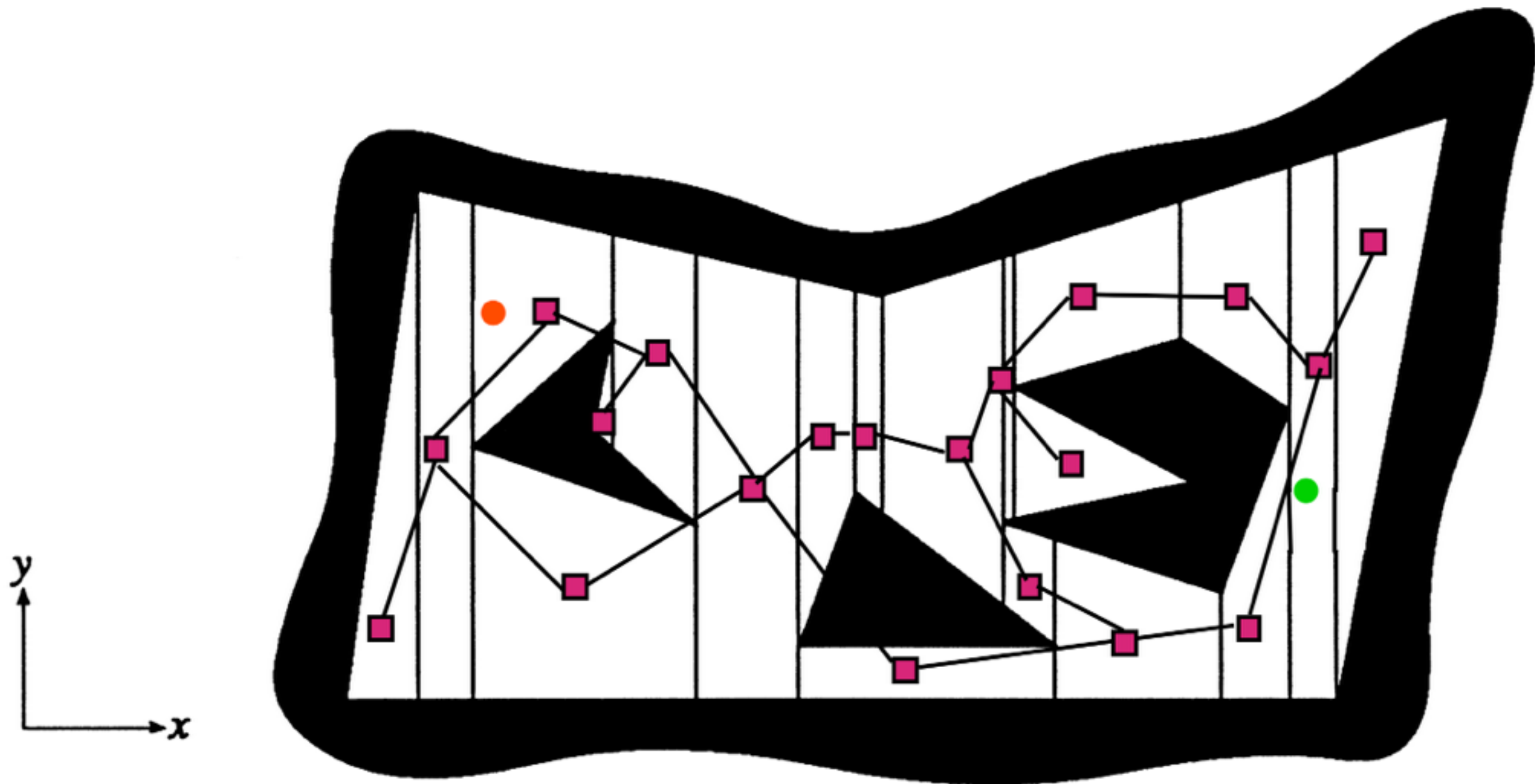


Back to low dimensions

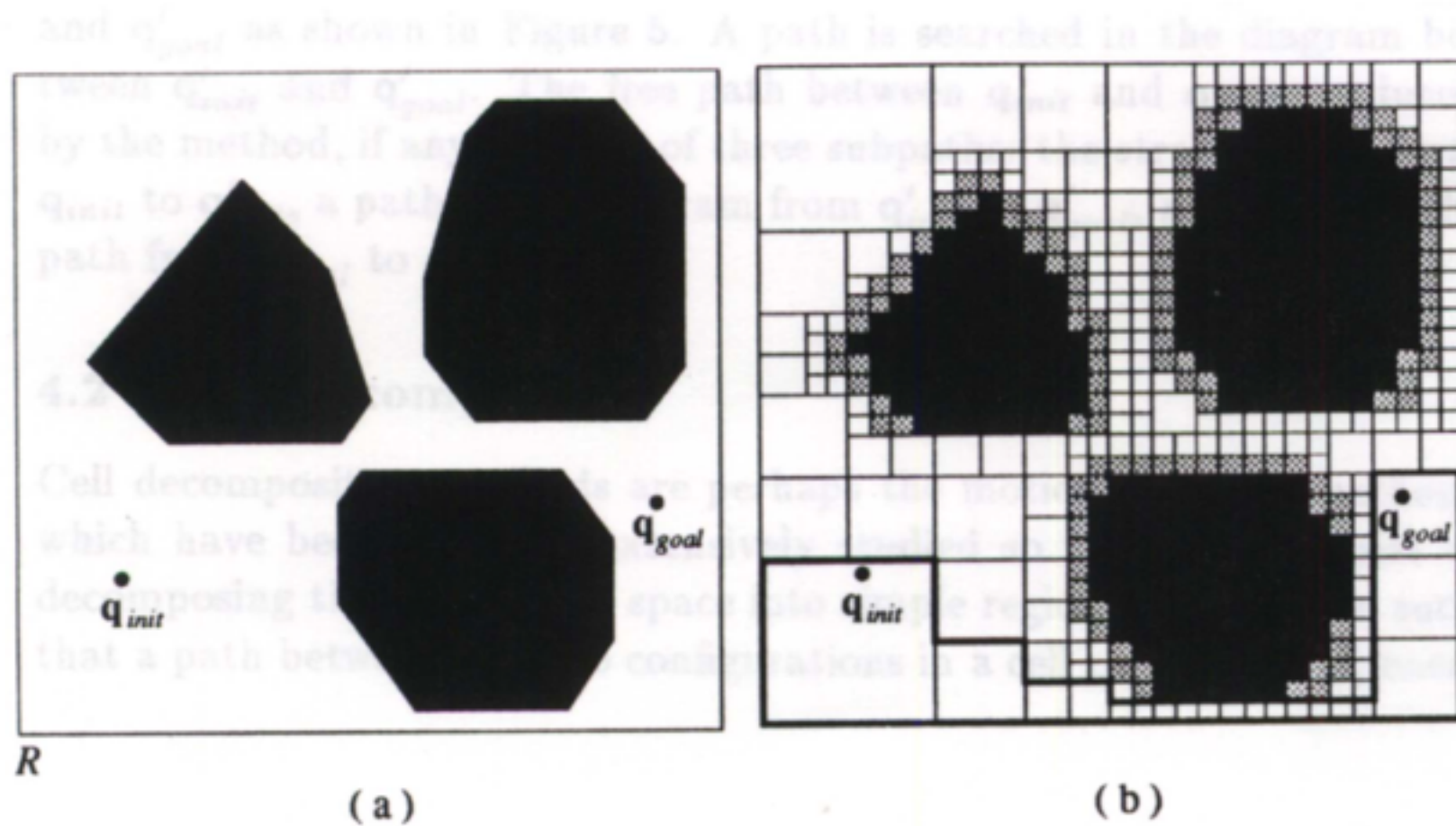
- How to plan the motion of a point?
- Discretize the space, construct a graph, search the graph.



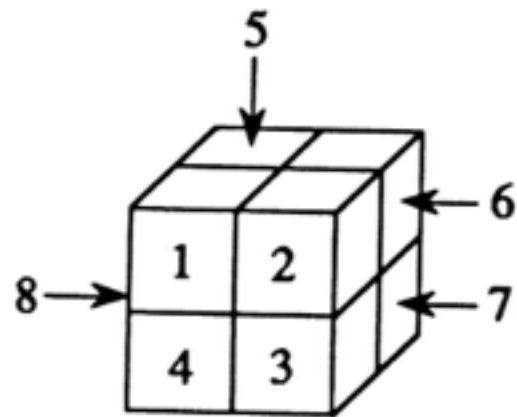
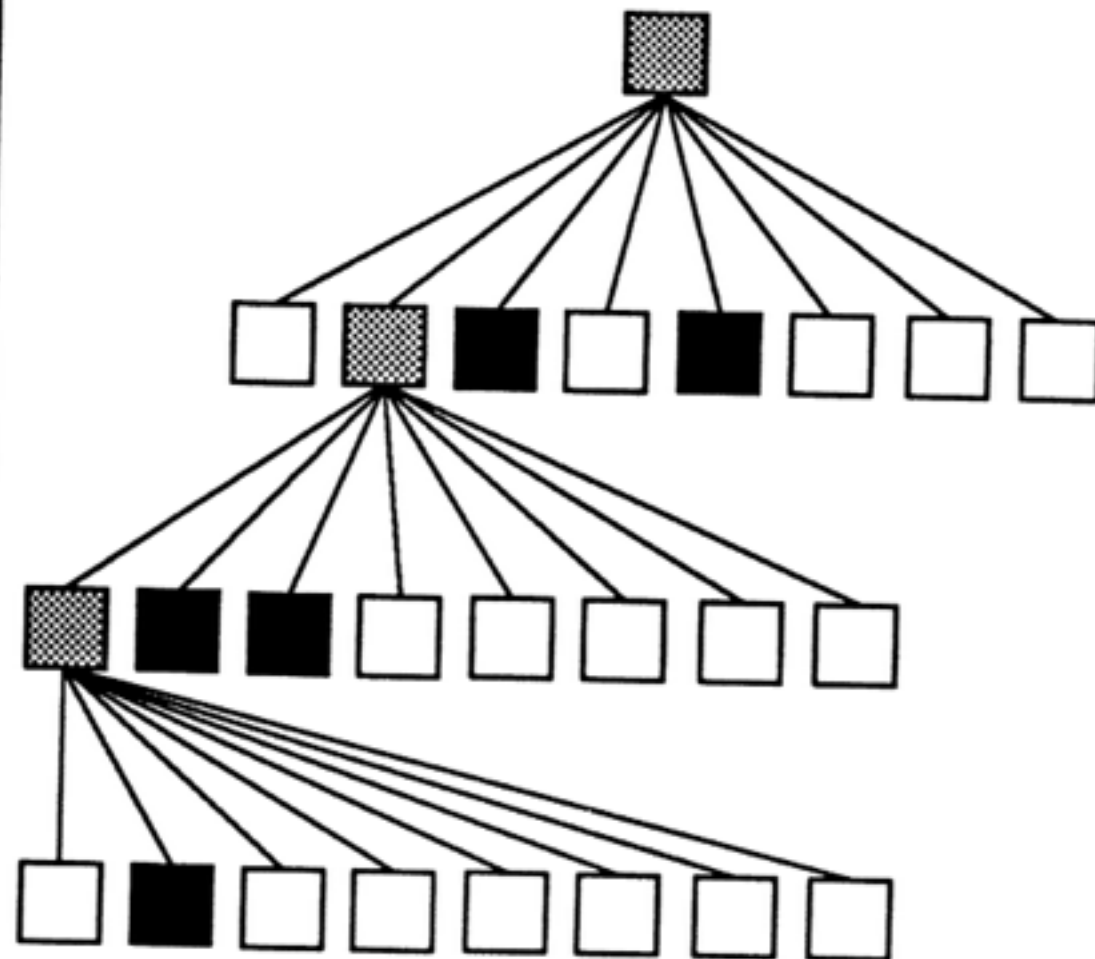
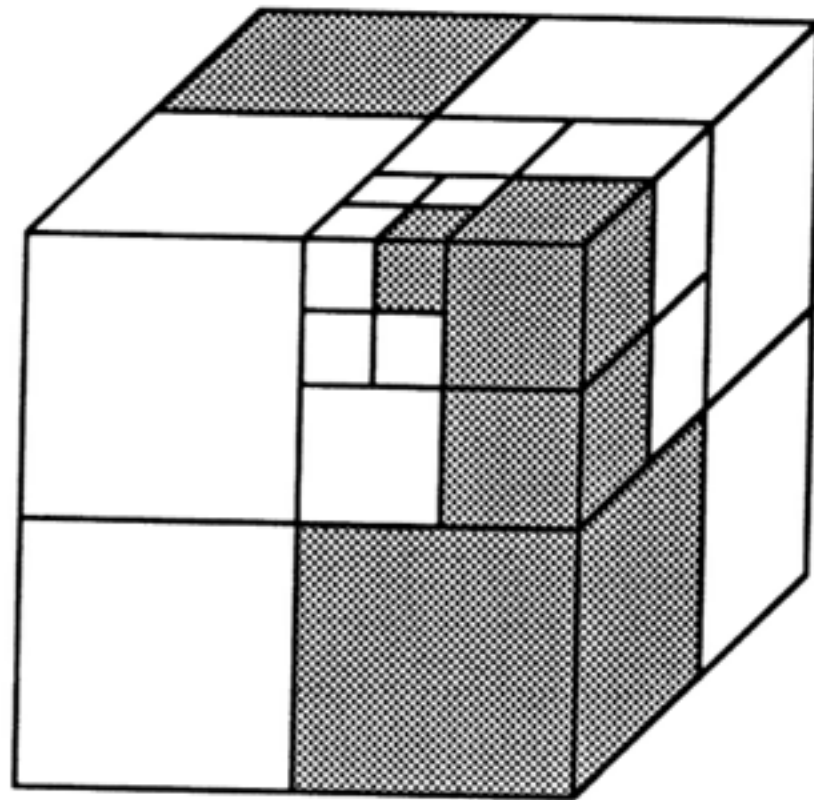
Trapezoidal Decomposition



Quadtree Decomposition



Octree Decomposition



 EMPTY cell  MIXED cell  FULL cell

The Problem

- Methods like these require a model of C-space
- These spaces becomes difficult/infeasible beyond three dimensions.
- How do we get around this?



The Point

- Describing the space is hard, but describing the state of a single point may not be.



Roadmaps

- Lets build a “roadmap” of the space, which requires much less evaluation.



Probabilistic Road Maps - PRM

- Learning Phase
 - Sample free points
 - Link samples to learn connectivity
 - Precomputed
- Query Phase
 - Add start and goal to roadmap
 - Connect to nearest neighbor
 - Compute path from start to goal
 - Multiple queries per road map

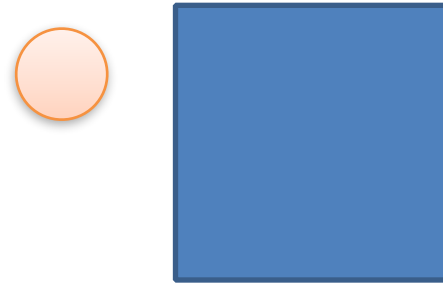


Probabilistic Road Maps - PRM

- Interactive Demo: <http://robotics.cs.unc.edu/interactive/prm.html>



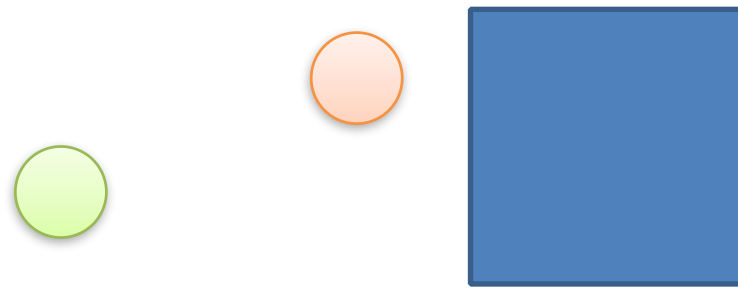
Rapidly Exploring Random Trees - RRT



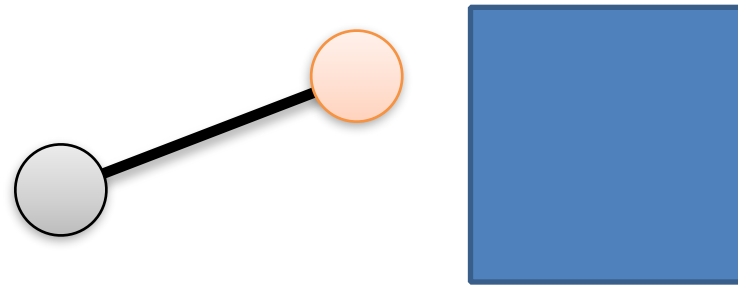
Rapidly Exploring Random Trees - RRT



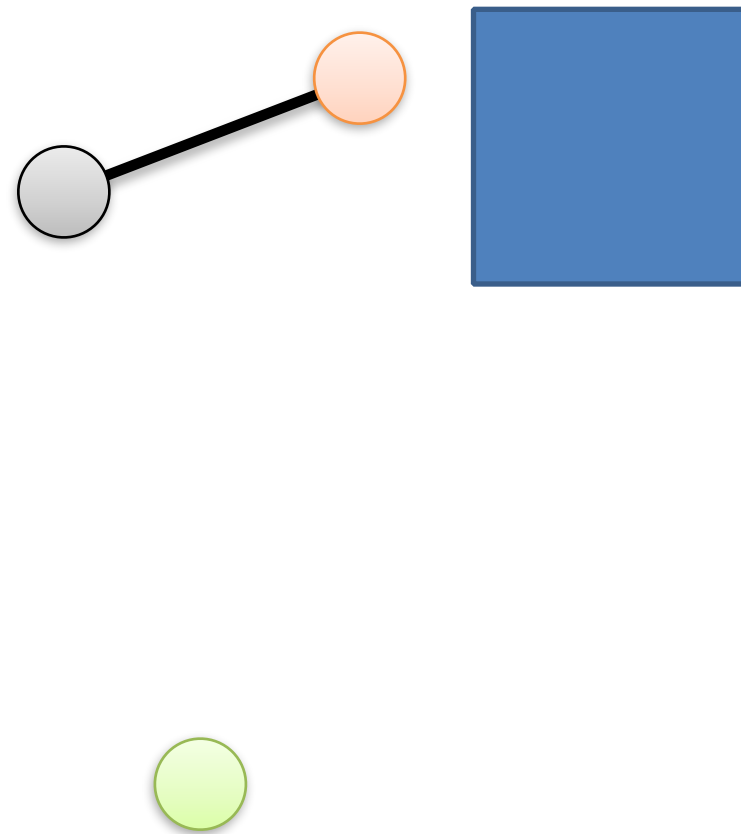
Rapidly Exploring Random Trees - RRT



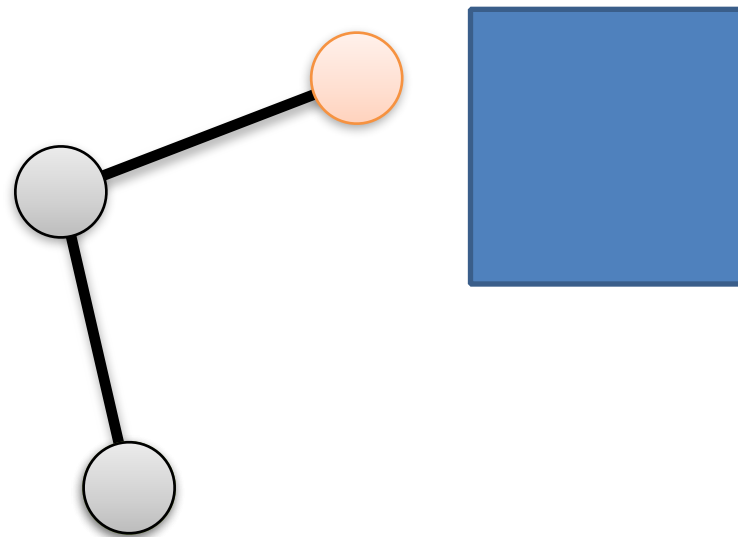
Rapidly Exploring Random Trees - RRT



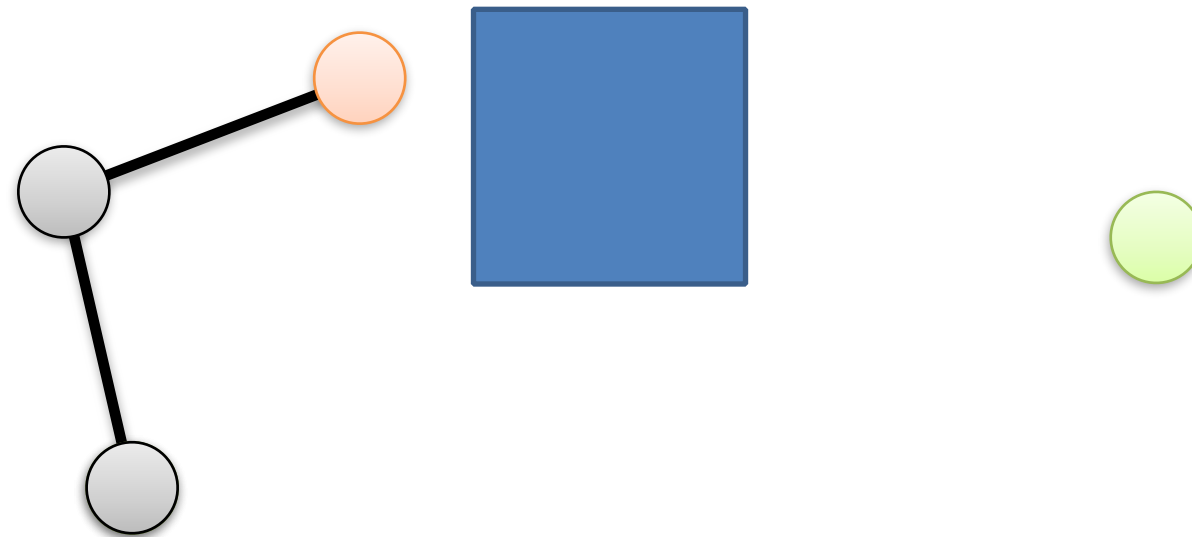
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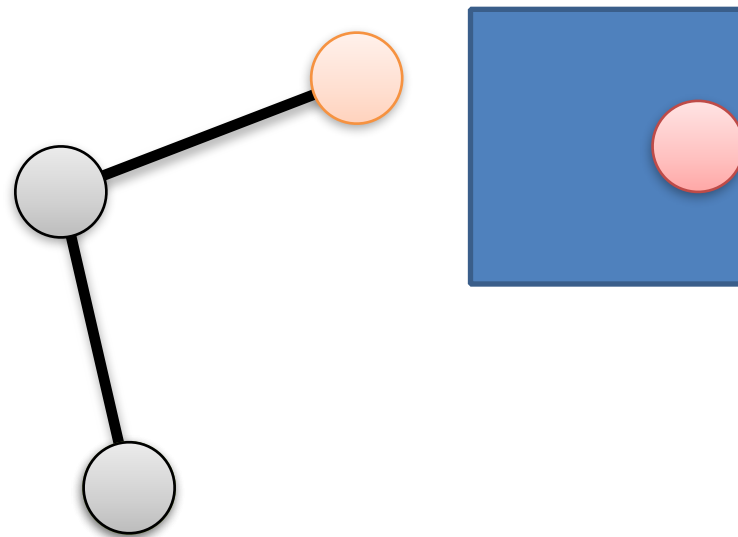
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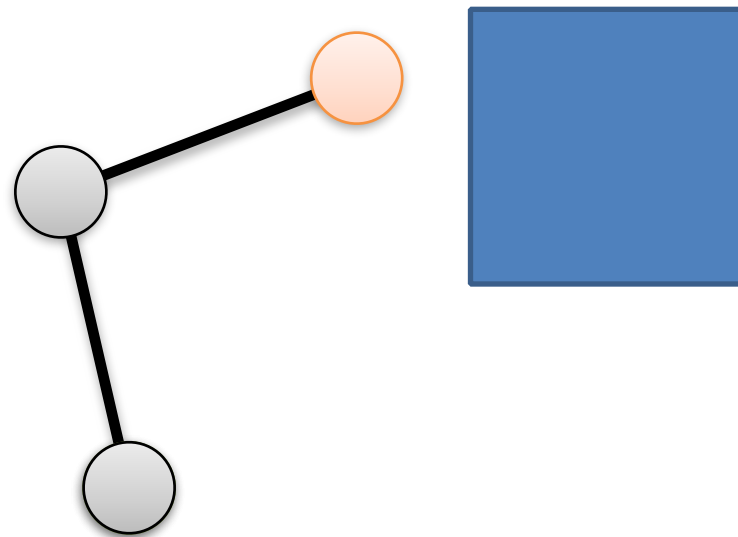
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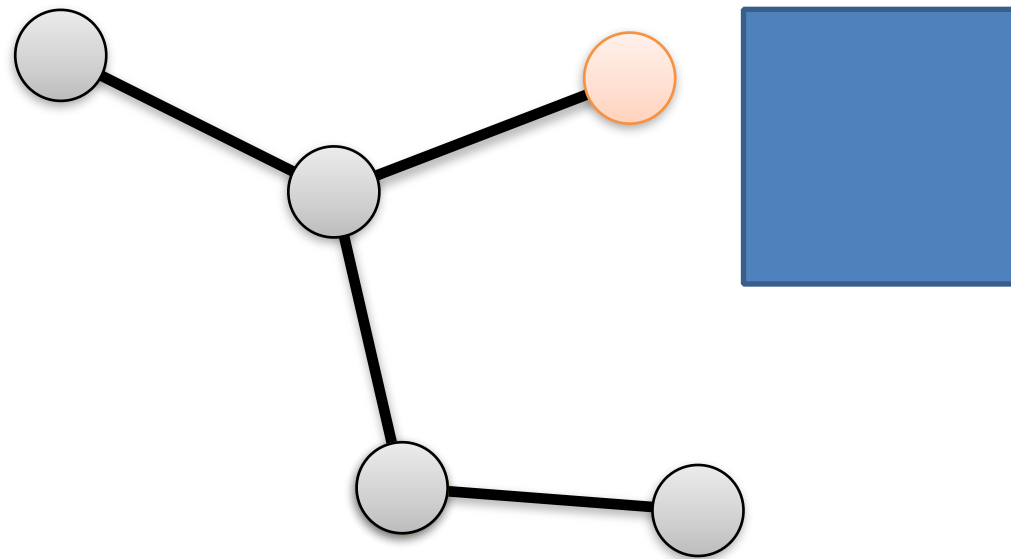
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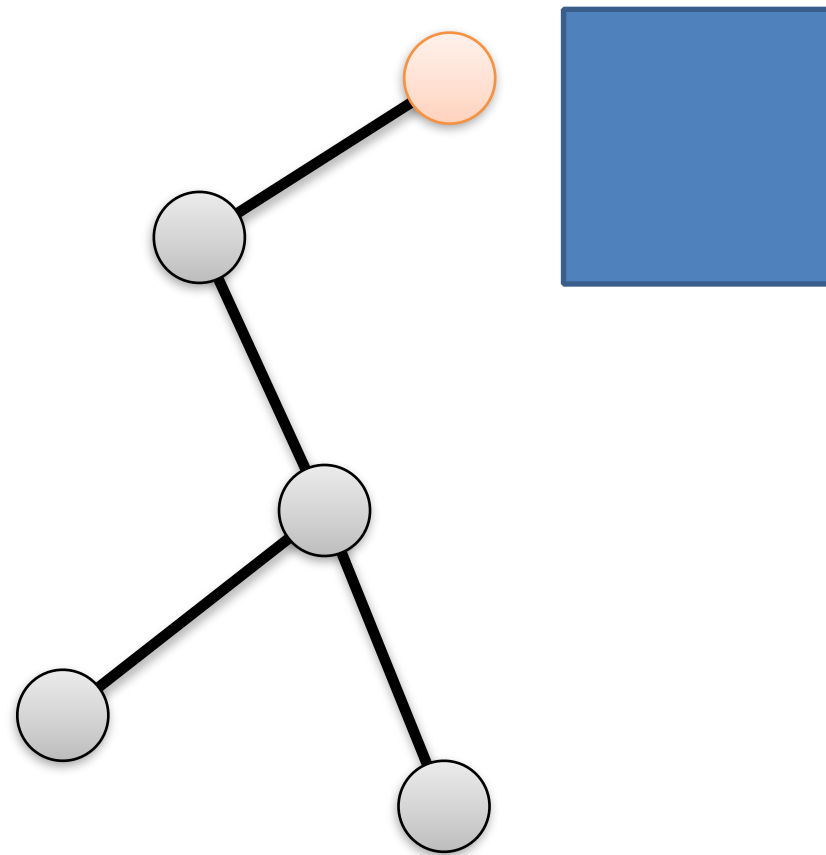
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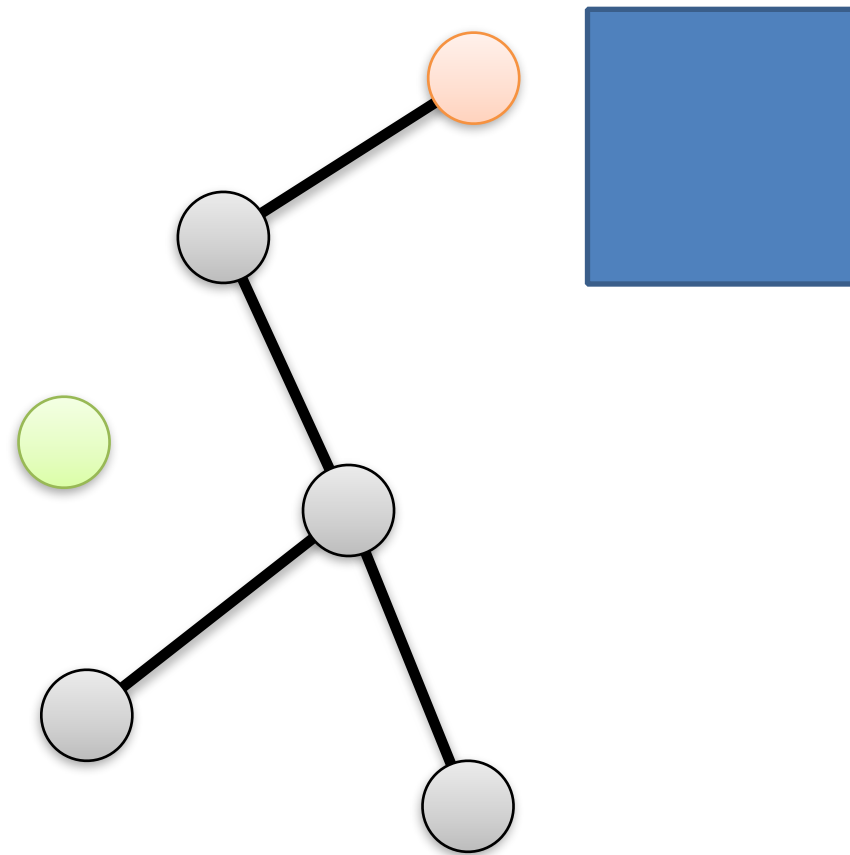
Rapidly Exploring Random Trees - RRT



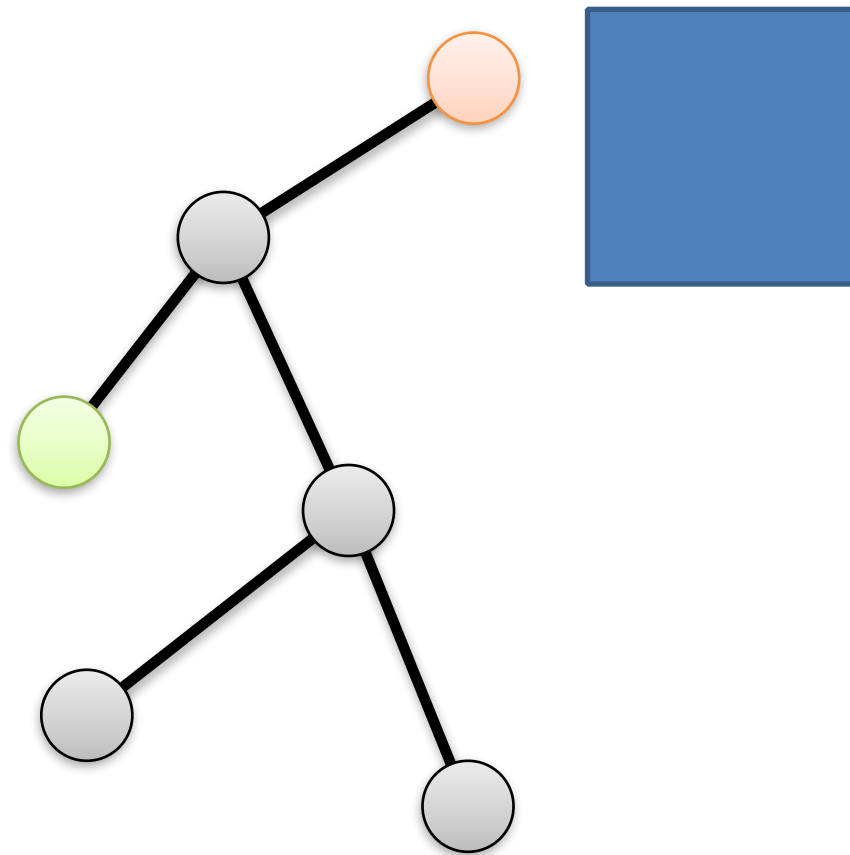
RRT*



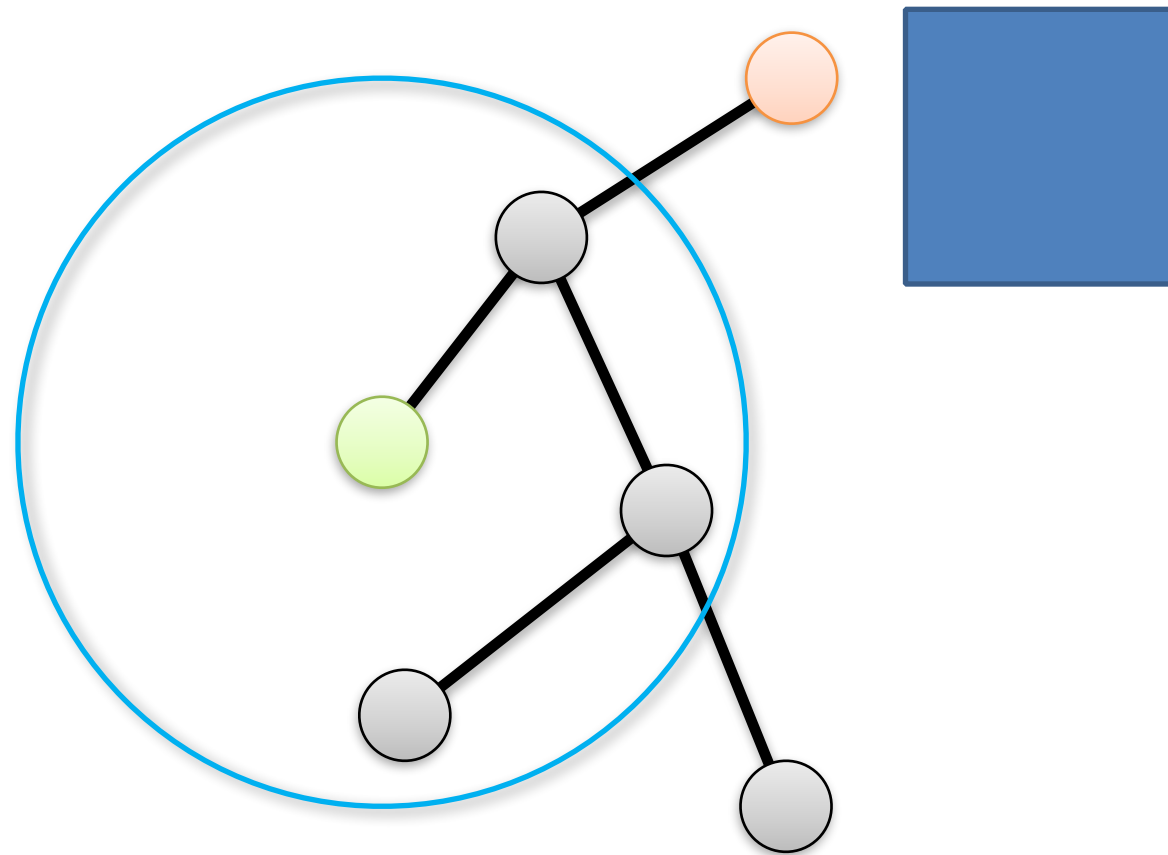
RRT*



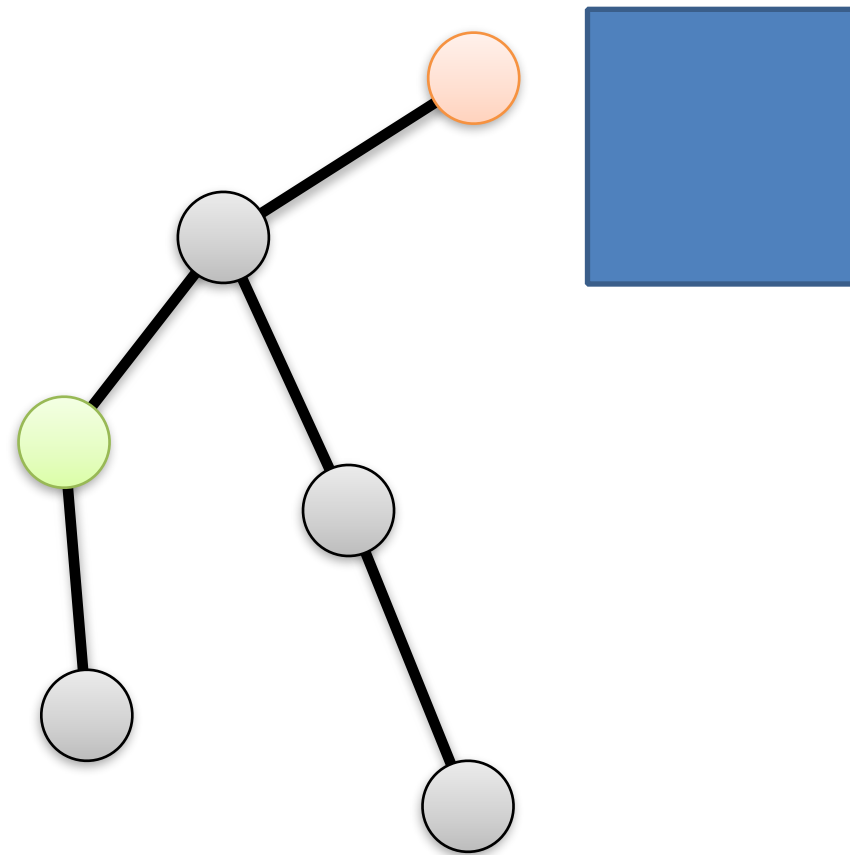
RRT*



RRT*



RRT*



Intuition

- Describe your system in terms of some high dimensional space
 - C-space
 - State space
 - Workspace
 - Trajectory space
 - A combination
- Plan a path through that space under some constraints



Space Choices

- Choice is frequently problem dependent
- Frequently require some approximation, so what model resolution is sufficient?
- May be influenced by the capabilities of your controller
 - One end of the spectrum, control propagation
 - Other end, maps.



Additional Considerations

- What space will allow for easy and effective implementation or adaptation of pre-existing algorithms?
- Space construction will affect topology, connectivity, obstacle definitions etc.



Increased Complexity

- Dynamic Environments
 - Ideas?
- Noisy Sensing/Actuation
 - Other Ideas?
- Nonholonomy
 - Even More Ideas?



Conclusion

- Many different classes of motion planning algorithms
- It is very difficult to generalize them
- The intuition gained from thinking about the abstractions will help you to understand the approaches as you encounter them.



Questions?

- Many images courtesy of Dr. Alterovitz's Robotics course.

