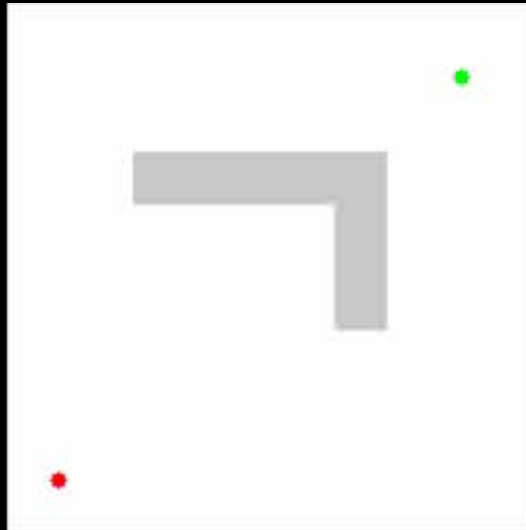


Motion Planning I – Point Robots

Last time...

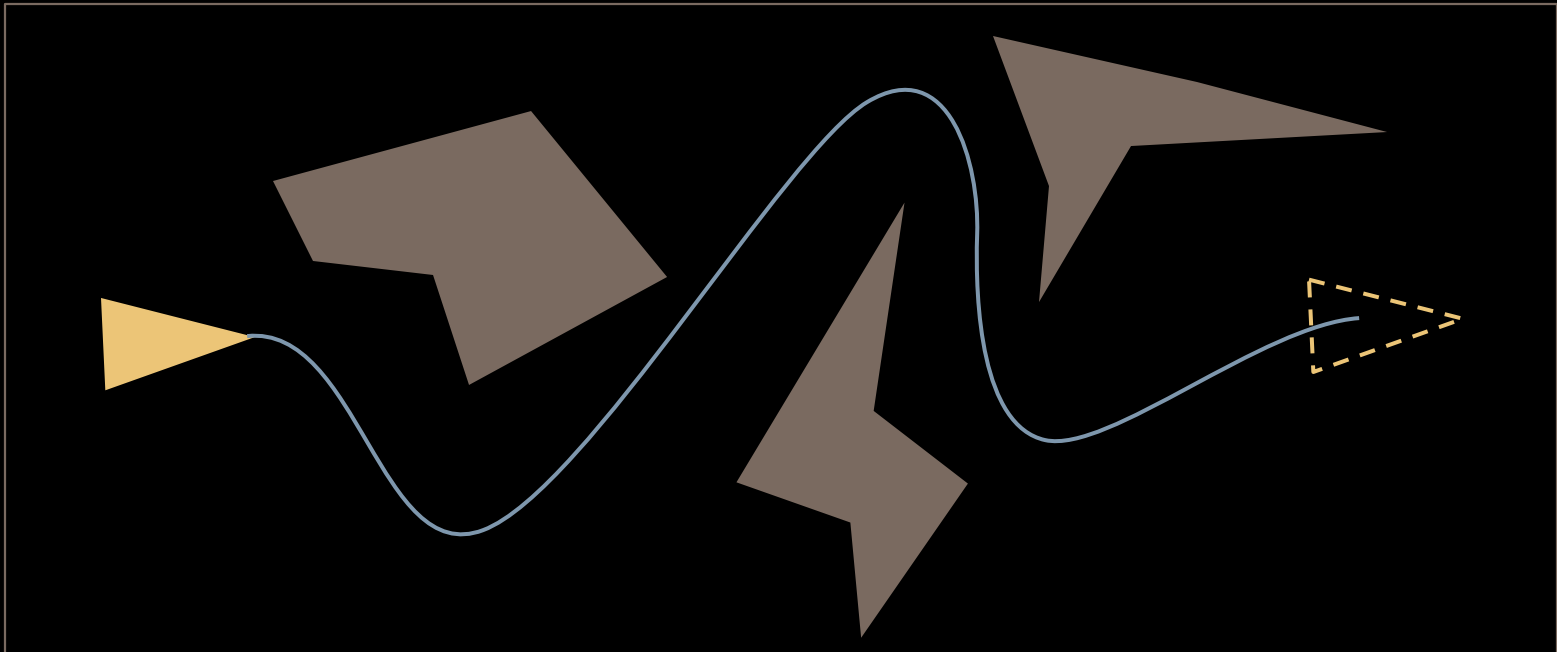
- We saw how to search for a path in a graph



- Today we will frame the problem of searching for a path for a robot
- We'll use some graph search methods to solve it

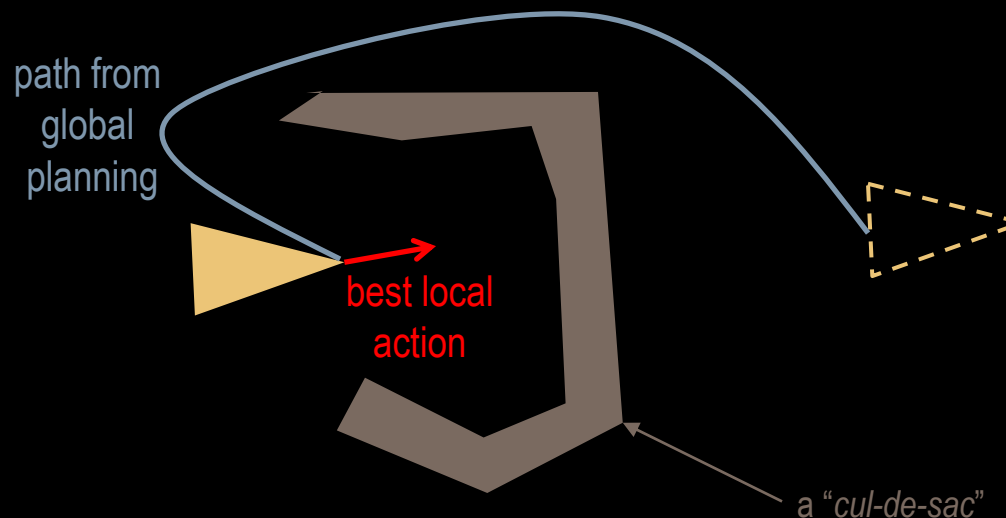
What is motion planning?

- The automatic generation of motion

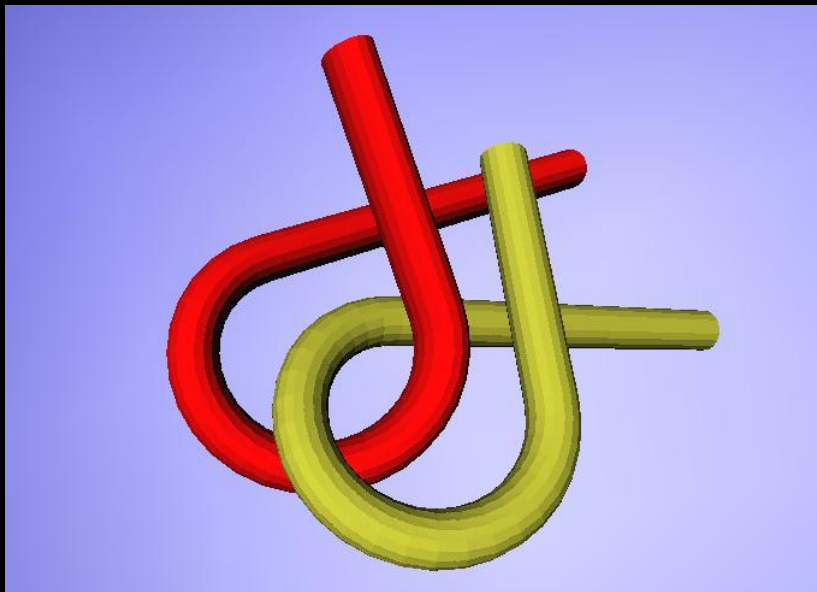


Why Motion Planning Instead of Obstacle Avoidance?

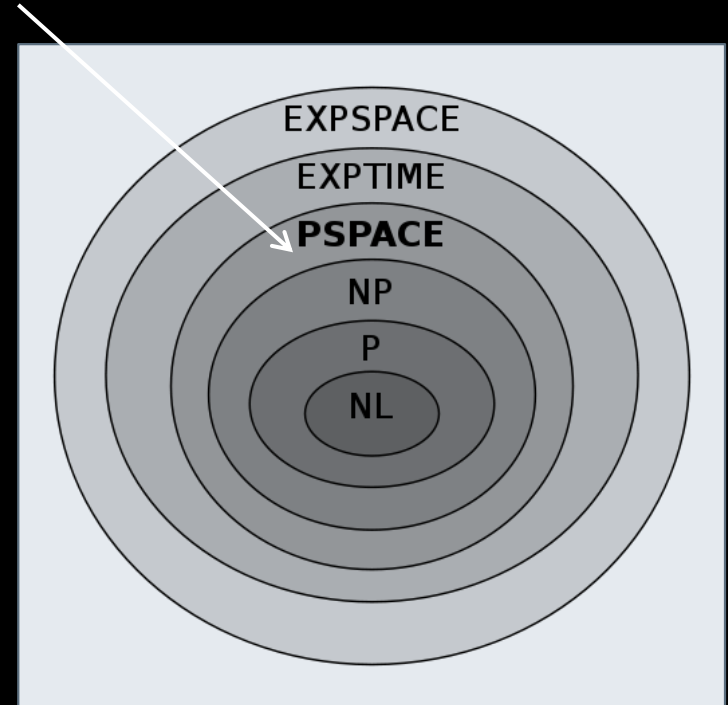
- Path planning
 - low-frequency, time-intensive search method for global finding of a (optimal) path to a goal
- Obstacle avoidance (aka “local navigation”)
 - fast, reactive method with local time and space horizon
- Distinction: Global vs. local reasoning



Is motion planning hard?



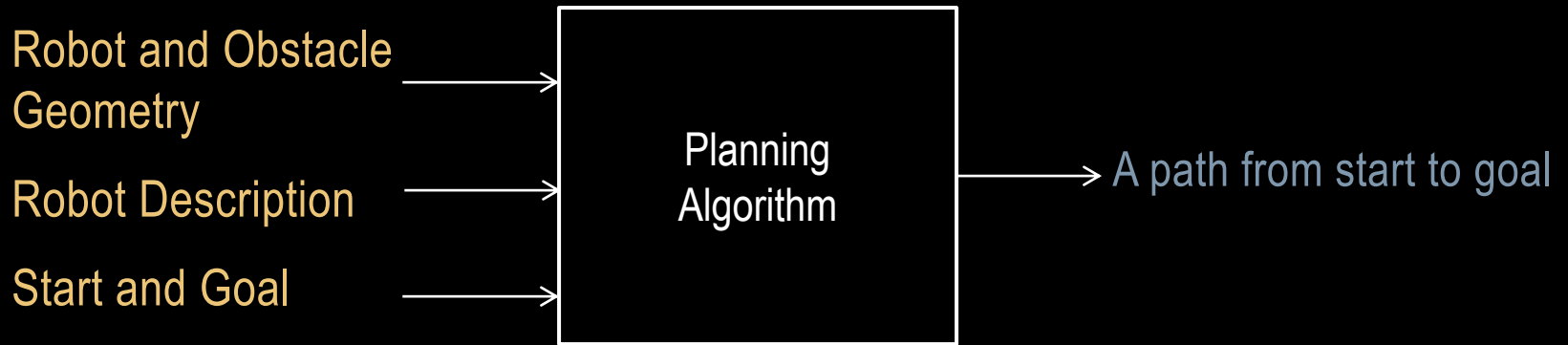
Basic Motion
Planning Problems



Complexity

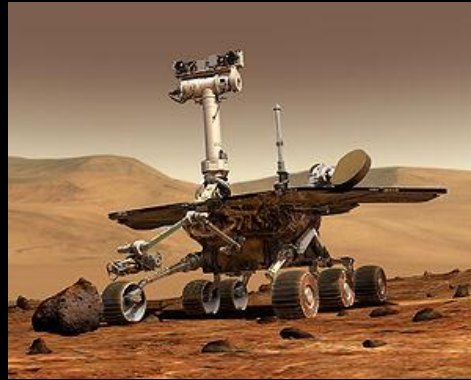
Basic Problem Statement

- *Automatically compute a path for an object/robot that does not collide with obstacles.*

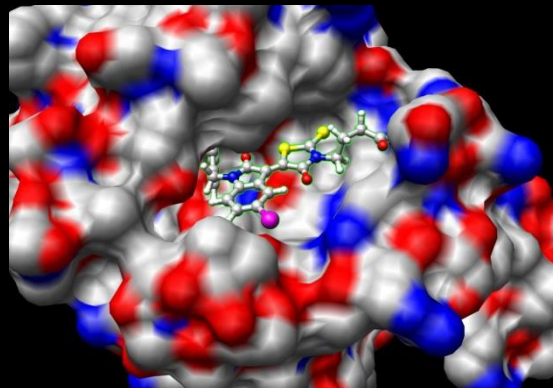
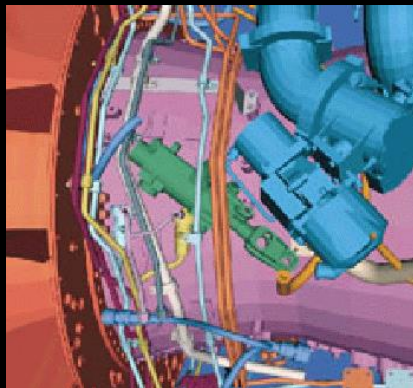


What can motion planning do?

- Automatically generate motion



- Automatically validate



Applications: Mobile Robots



Roomba iCreate



Mars Rovers



DARPA Urban Challenge



Google Self-Driving Car

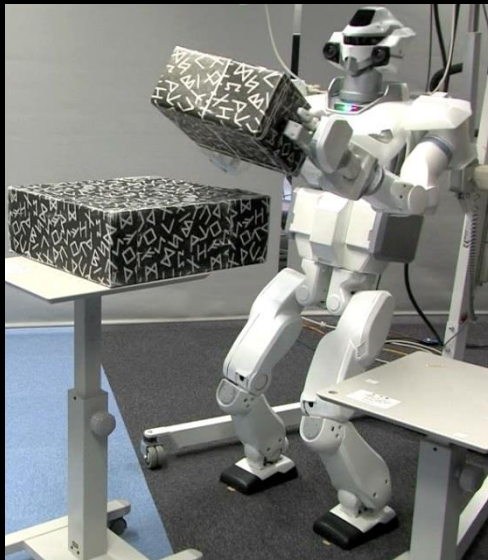
Applications: Robotic Manipulation



Factory Automation



Personal Robots



Humanoid Robots



Personal Robots

Applications: Computer Games/Graphics



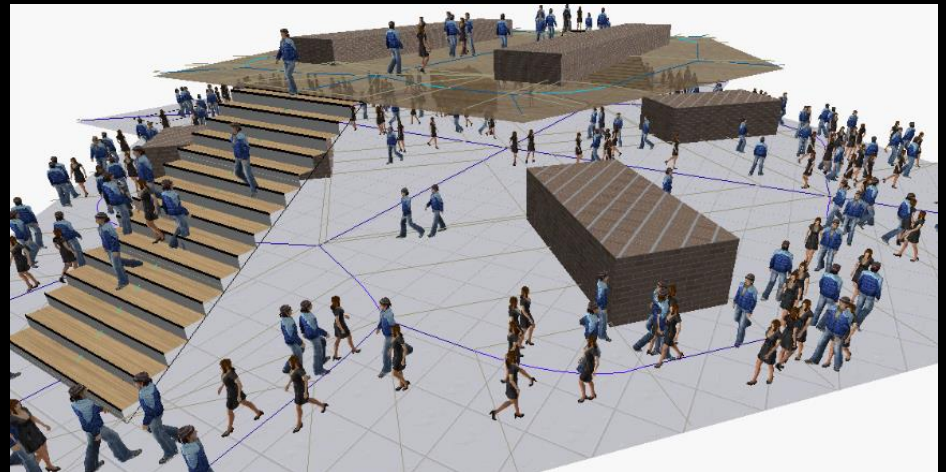
Path Finding in Games



Character Animation

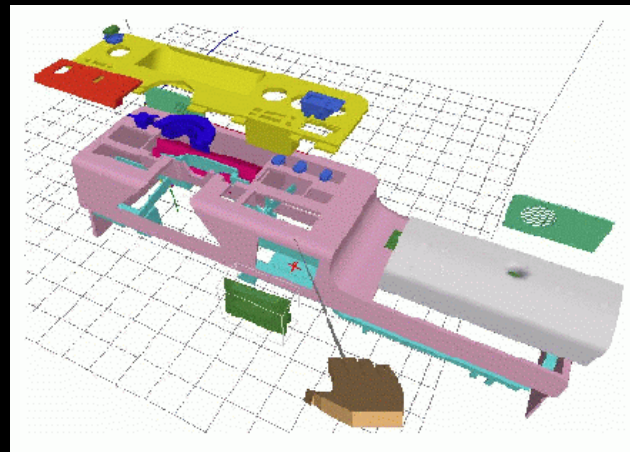
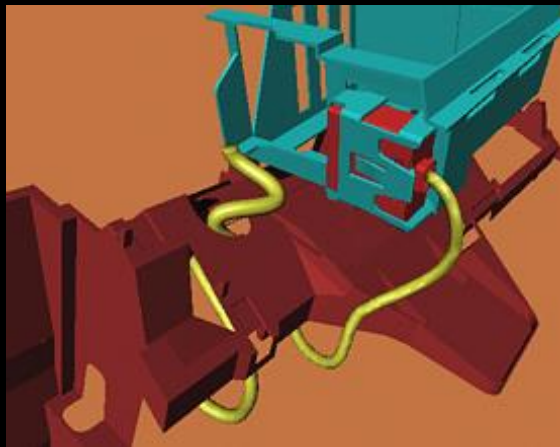
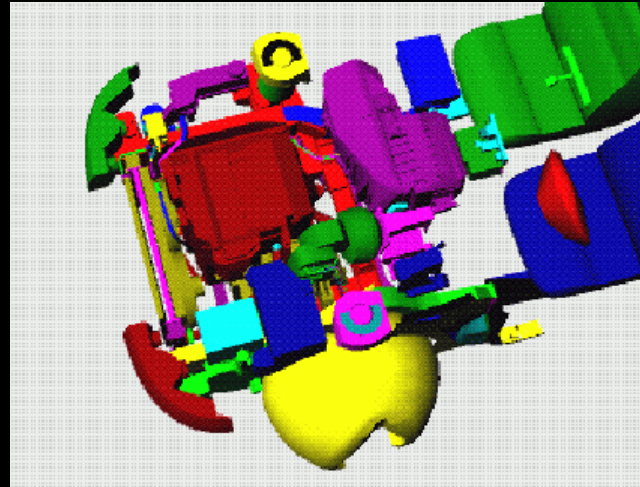
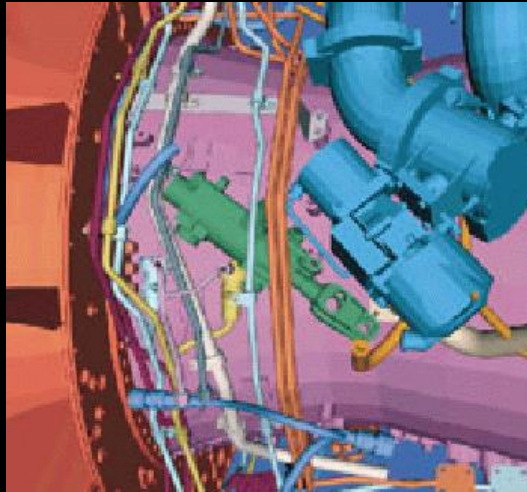


Retargeting Motion Capture

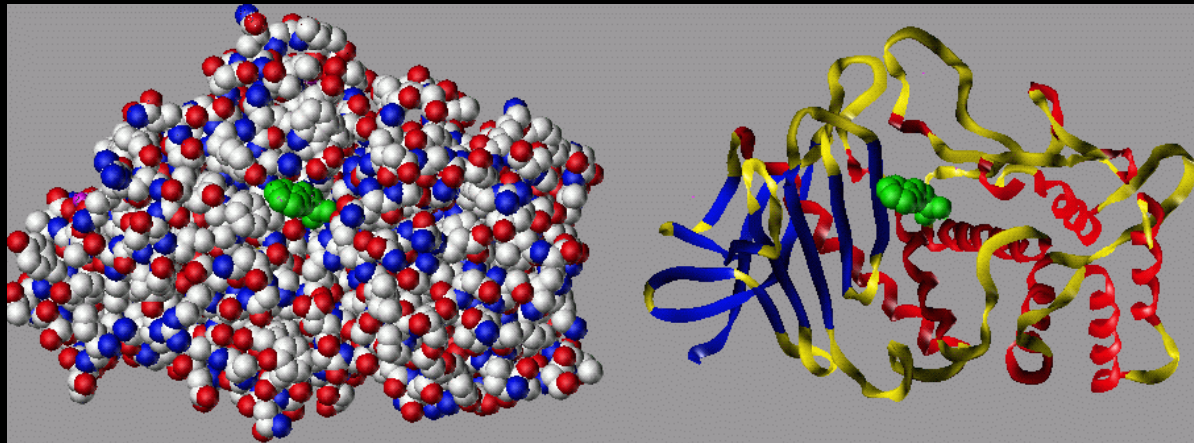
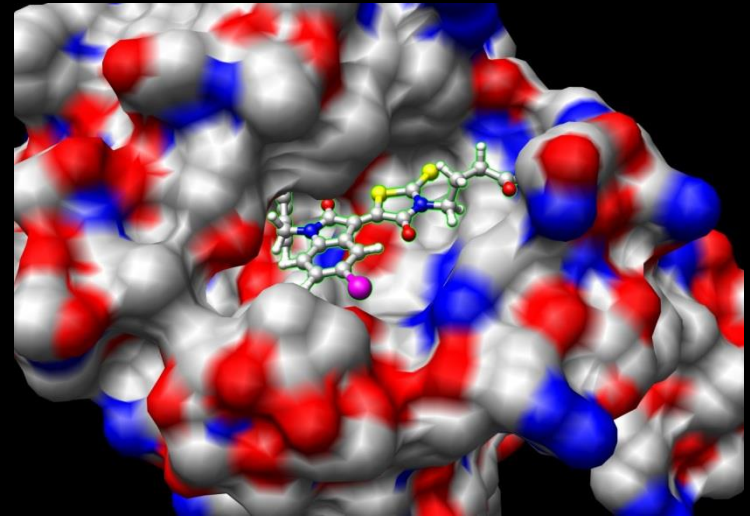
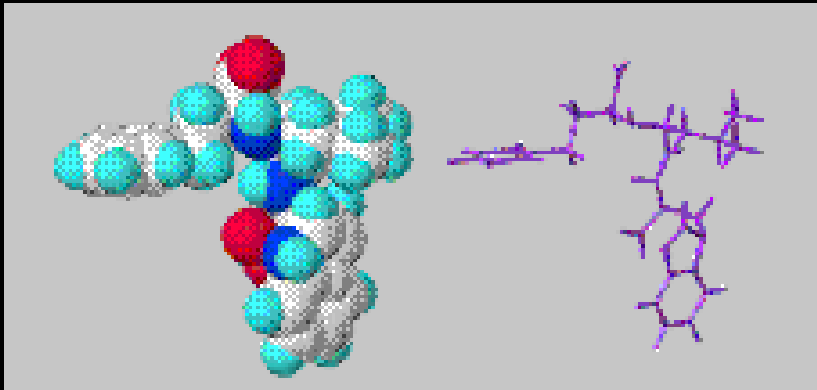


Animation of Crowds

Applications: Assembly Planning

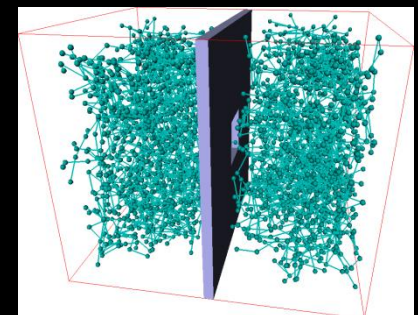
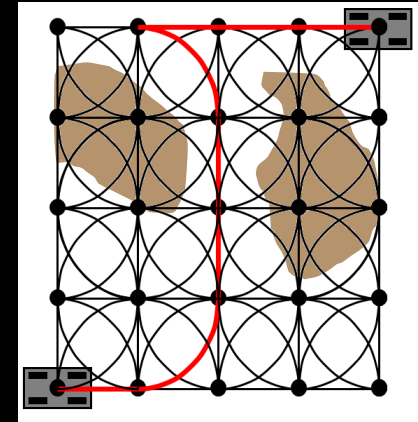
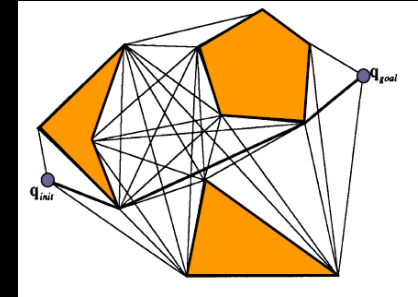


Applications: Computational Biology



Approaches

- Exact algorithms
 - Either find a solution or prove none exists
 - Very computationally expensive
 - Unsuitable for high-dimensional spaces
- Discrete Search
 - Divide space into a grid, use A* to search
 - Good for vehicle planning
 - Unsuitable for high-dimensional spaces
- Sampling-based Planning
 - Sample the C-space, construct path from samples
 - Good for high-dimensional spaces
 - Weak completeness and optimality guarantees



What matters?

- Motion planning algorithms are judged on
 - Completeness
 - Optimality
 - Speed (AKA efficiency)
 - Generality
- These vary in importance depending on the application

What matters: Completeness

- Will the algorithm solve all solvable problems?
 - Will the algorithm return no solution for unsolvable problems?
 - What if the algorithm is probabilistic?
-
- For what application(s) is completeness very important?
 - For what application(s) is completeness not important?

What matters: Optimality

- Will the algorithm generate the shortest path?
 - Will the algorithm generate the least-cost path (for an arbitrary cost function)?
 - Do we need optimality or is feasibility enough?
-
- For what application(s) is optimality very important?
 - For what application(s) is optimality not important?

What matters: Speed (AKA Efficiency)

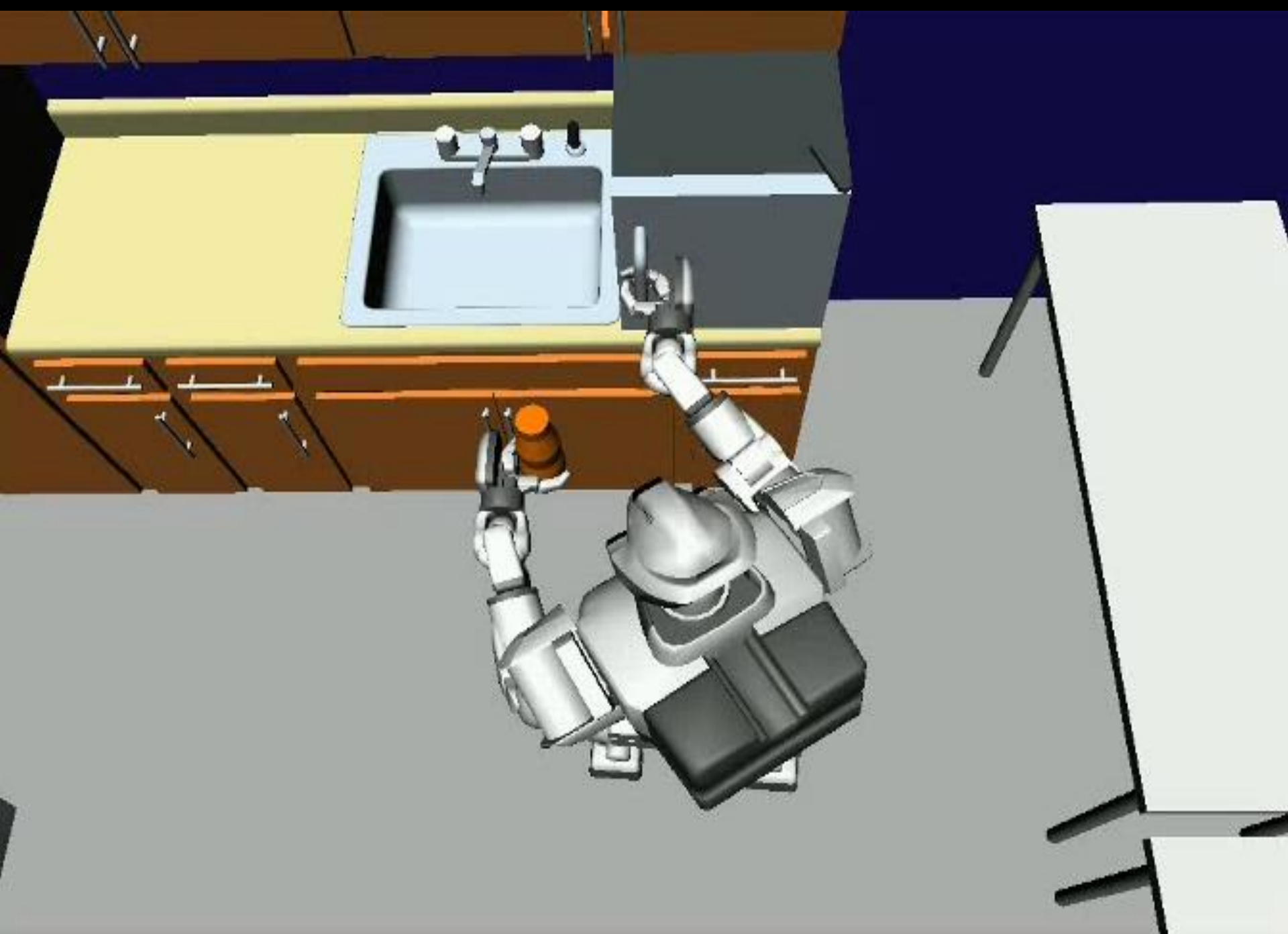
- How long does it take to generate a path for **real-world** problems?
- How does the run-time scale with dimensionality of the problem and complexity of models?
- Is there a quality vs. computation time tradeoff?
- For what application(s) is speed very important?
- For what application(s) is speed not important?

What matters: Generality

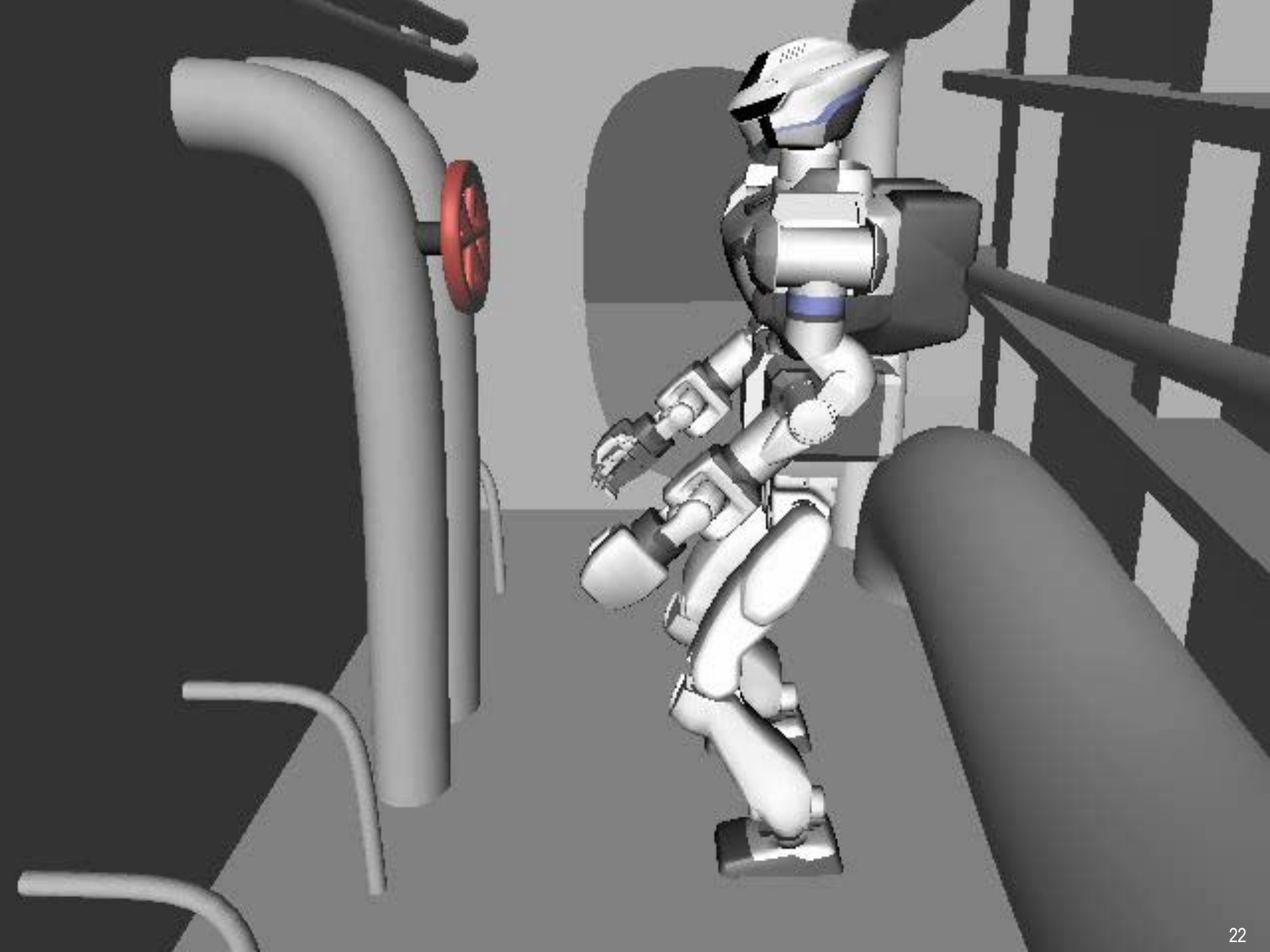
- Generality is the vaguest criterion, but often the most important
 - What types of problems can it solve?
 - What types of problems can't it solve?
-
- For what application(s) is generality very important?
 - For what application(s) is generality not important?

$w = 4.98\text{kg}$





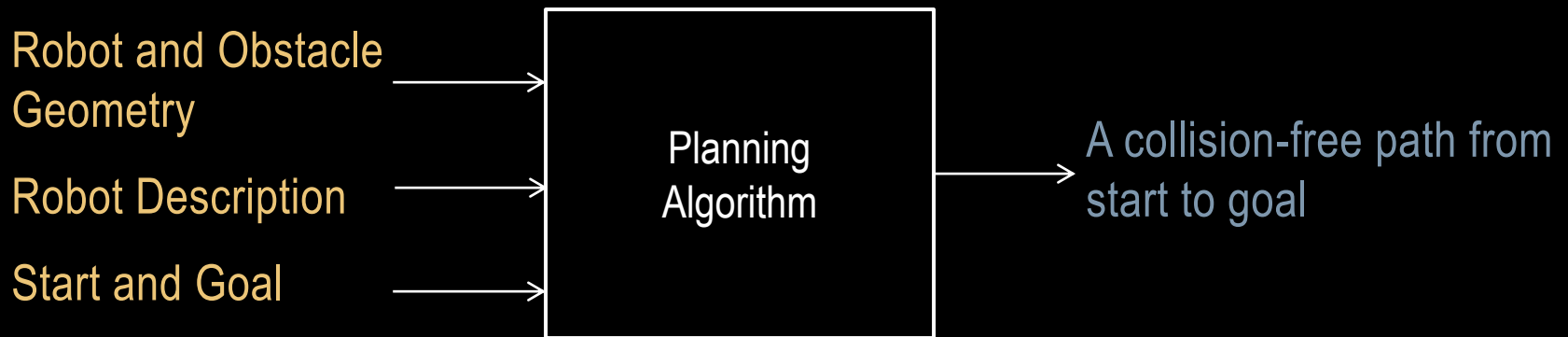




Path Planning for Point Robots

Basic Problem Statement

- *Automatically compute a path for an object/robot that does not collide with obstacles.*



- Start simple:
 - The robot is a point that can move freely
 - The environment is 2D with polygonal obstacles

Methods

- Visibility graph
- Cell decomposition
- Potential fields

Framework

continuous representation

(configuration space formulation)



discretization

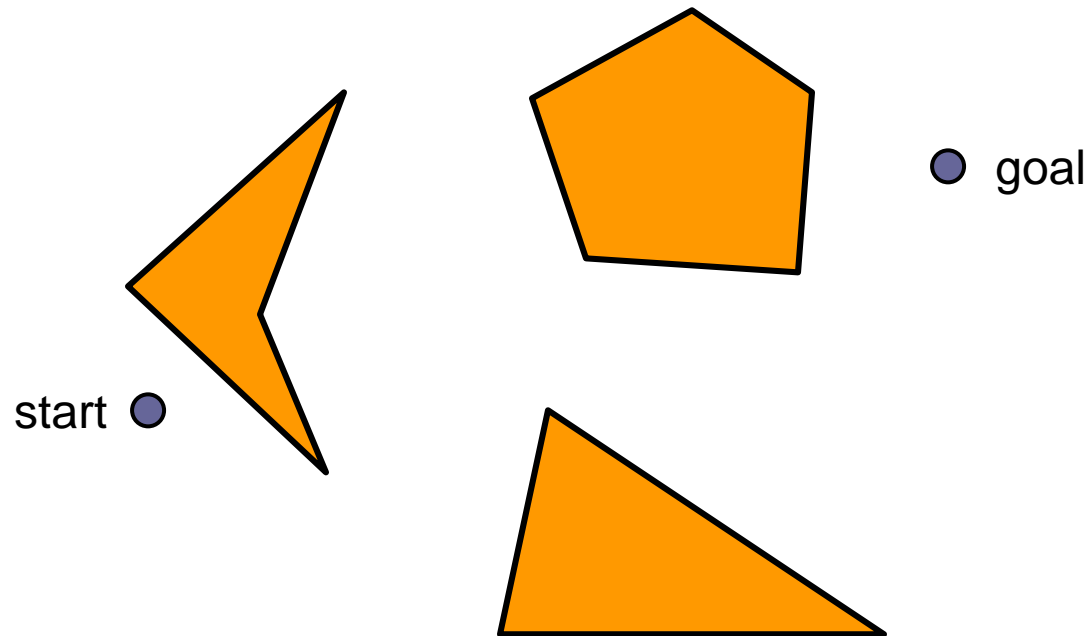
(random sampling, processing critical geometric events)



graph searching

(breadth-first, best-first, A*)

Continuous Representation



Framework

continuous representation



discretization

(random sampling, processing critical geometric events)

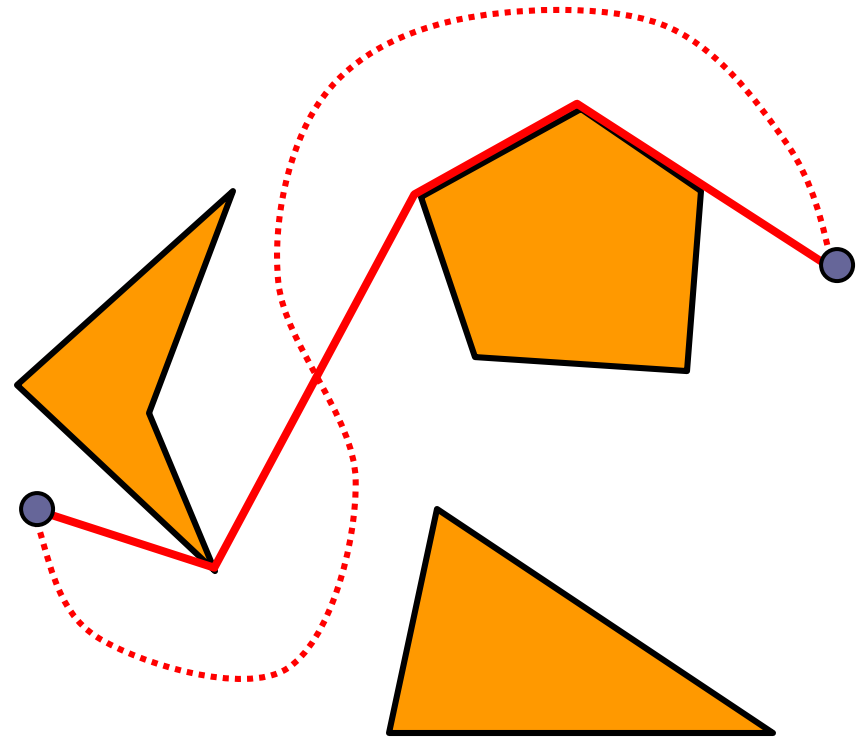


graph searching

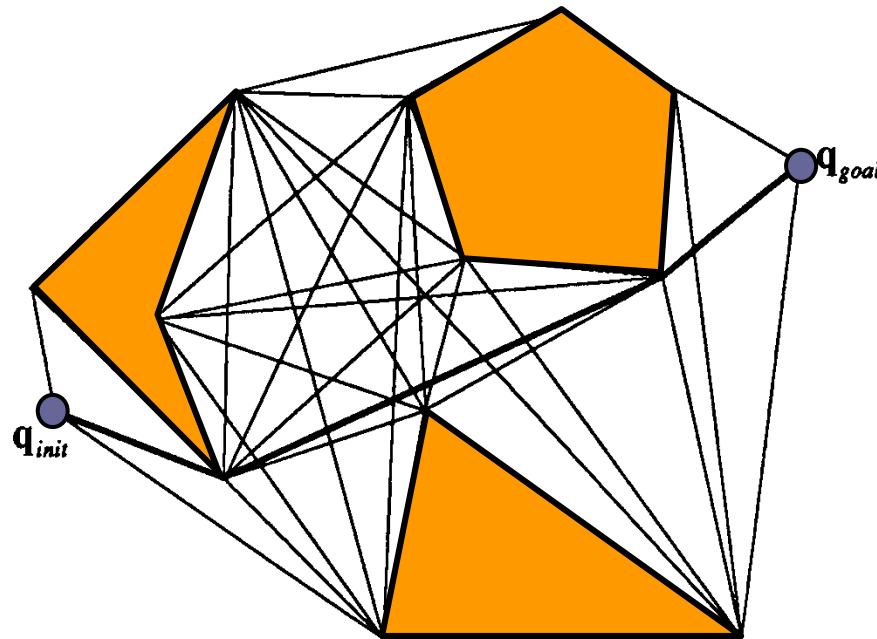
(breadth-first, best-first, A*)

Visibility graph method

- **Observation:** If there is a collision-free path between two points, then there is a piece-wise linear path that bends only at the obstacles vertices.
- **Why?**
Any collision-free path can be transformed into a piece-wise linear path that bends only at the obstacle vertices.



What is a visibility graph?



- A **visibility graph** is a graph such that
 - Nodes: q_{init} , q_{goal} , or an obstacle vertex.
 - Edges: An edge exists between nodes u and v if the line segment between u and v is an obstacle edge or it does not intersect the obstacles.

A simple algorithm for building visibility graphs

Input: q_{init} , q_{goal} , polygonal obstacles

Output: visibility graph G

```
1: for every pair of nodes  $u, v$ 
2:   if segment( $u, v$ ) is an obstacle edge then
3:     insert edge( $u, v$ ) into  $G$ ;
4:   else
5:     for every obstacle edge  $e$ 
6:       if segment( $u, v$ ) intersects  $e$ 
7:         go to (1);
8:     insert edge( $u, v$ ) into  $G$ .
```

Computational efficiency

```

1: for every pair of nodes  $u, v$   $O(n^2)$ 
2:   if segment( $u, v$ ) is an obstacle edge then  $O(n)$ 
3:     insert edge( $u, v$ ) into  $G$ ;
4:   else
5:     for every obstacle edge  $e$   $O(n)$ 
6:       if segment( $u, v$ ) intersects  $e$ 
7:         go to (1);
8:     insert edge( $u, v$ ) into  $G$ .
```

- Simple algorithm $O(n^3)$ time
- More efficient algorithms
 - Rotational sweep $O(n^2 \log n)$ time
 - Optimal algorithm $O(n^2)$ time
- $O(n^2)$ space

Framework

continuous representation

(configuration space formulation)



discretization

(random sampling, processing critical geometric events)



graph searching

(breadth-first, best-first, A*)

Which method
from last lecture
should we use?

Framework

continuous representation



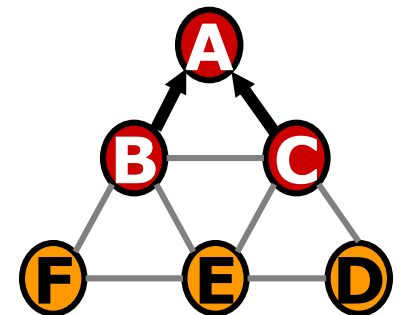
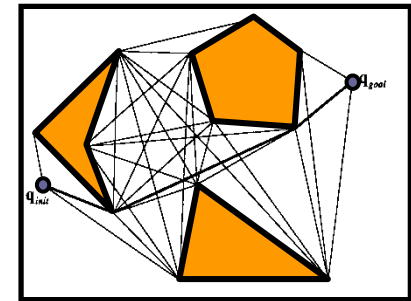
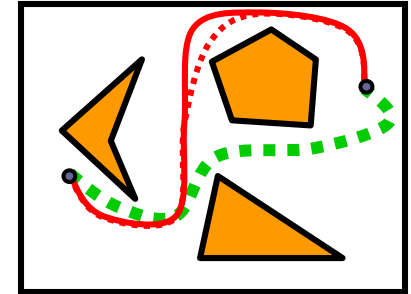
discretization

construct visibility graph



graph searching

A*



Computational efficiency

- Running time $O(n^3)$
 - Compute the visibility graph
 - Search the graph
 - An optimal $O(n^2)$ time algorithm exists.
- Space $O(n^2)$
- **Can we do better?**

Break

Classic path planning approaches

□ **Cell decomposition**

Decompose the free space into simple cells and represent the connectivity of the free space by the adjacency graph of these cells

□ **Potential field**

Define a potential function over the free space that has a global minimum at the goal and follow the steepest descent of the potential function

Classic path planning approaches

□ Cell decomposition

Decompose the free space into **simple** cells and represent the connectivity of the free space by the adjacency graph of these cells

□ Potential field

Define a potential function over the free space that has a global minimum at the goal and follow the steepest descent of the potential function

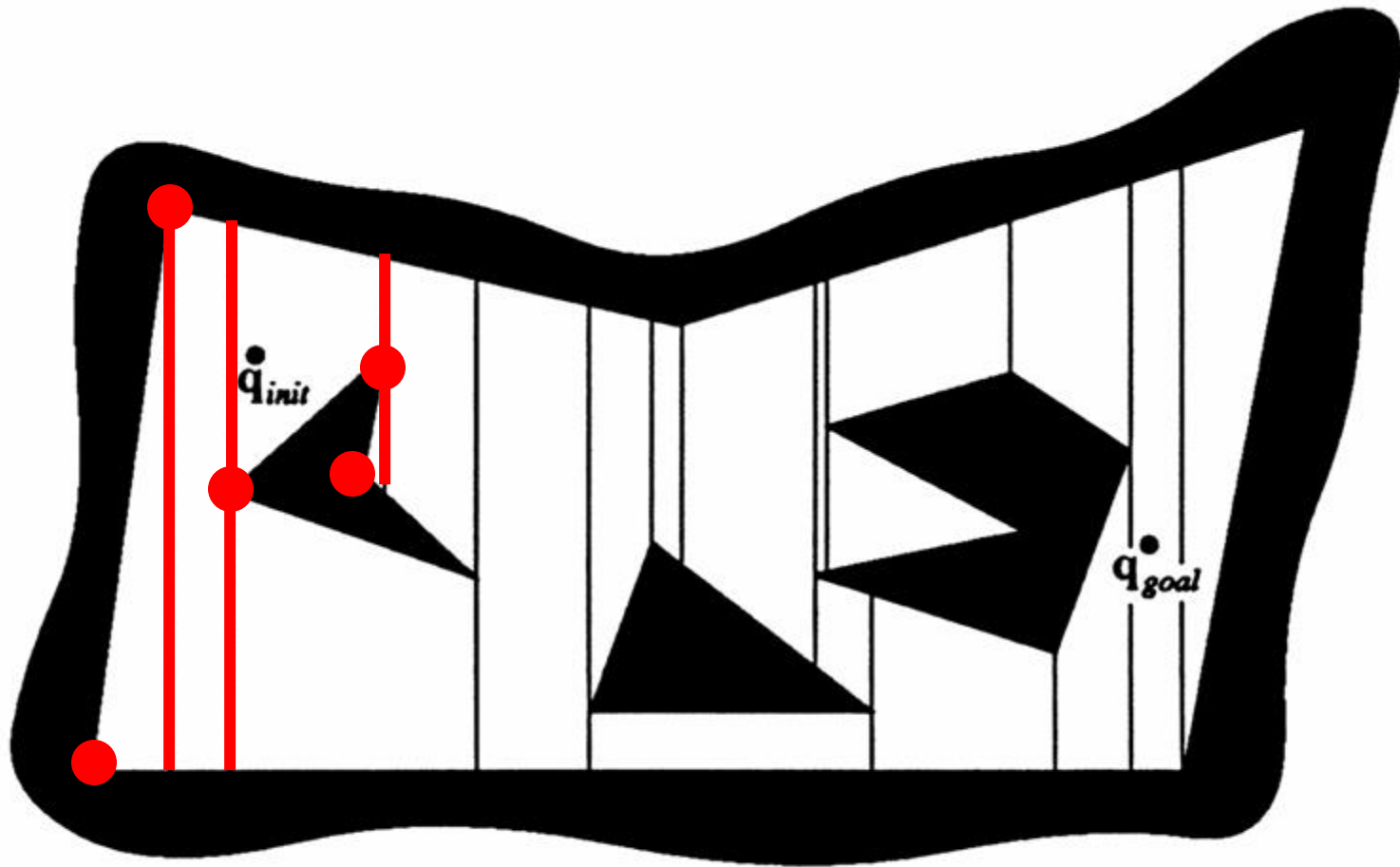
Cell-decomposition methods

□ Exact cell decomposition

The free space F is represented by a collection of non-overlapping simple cells whose union **is exactly** F .

- Examples of cells: trapezoids, triangles

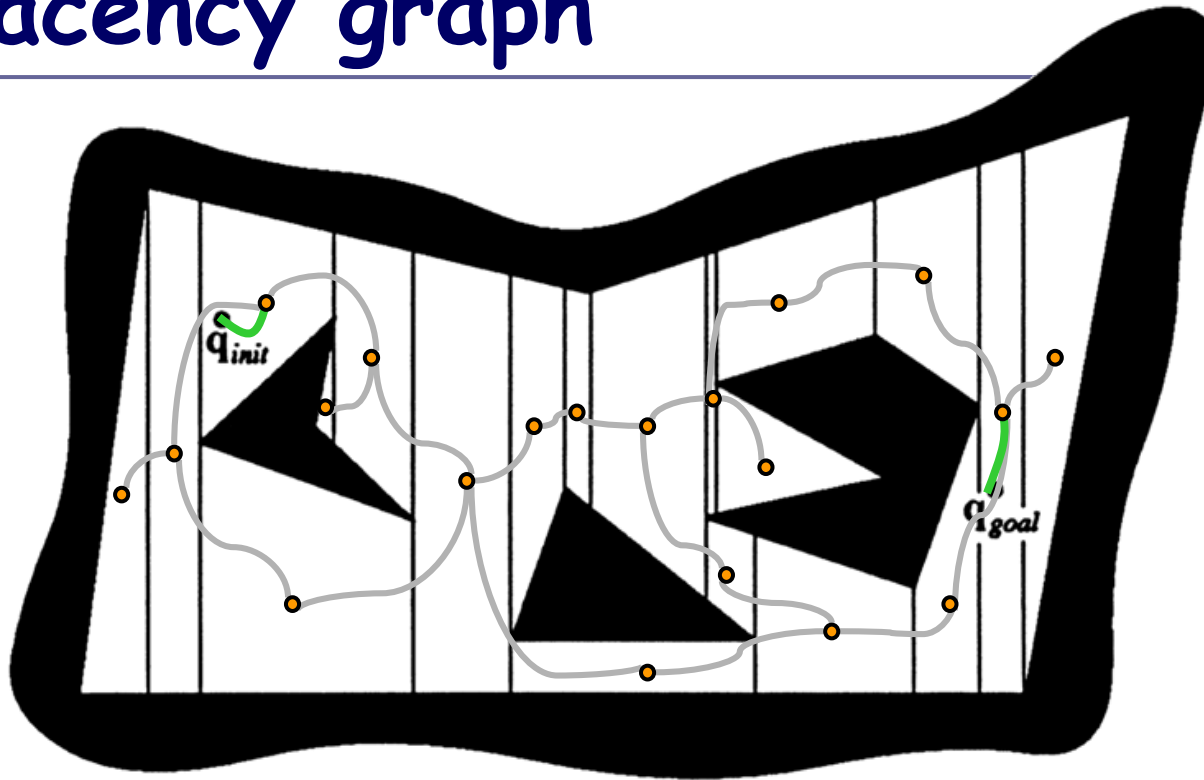
Trapezoidal decomposition



Computational efficiency

- Running time $O(n \log n)$ by planar sweep
- Space $O(n)$
- Mostly for 2-D environments

Adjacency graph



- **Nodes:** cells
- **Edges:** There is an edge between every pair of nodes whose corresponding cells are adjacent.
- A sequence of edges can be converted into a continuous path
 - This is easy to do when cells are convex. Why?

Framework

continuous representation



discretization

construct an adjacency graph of the cells



graph searching

search the adjacency graph

Cell-decomposition methods

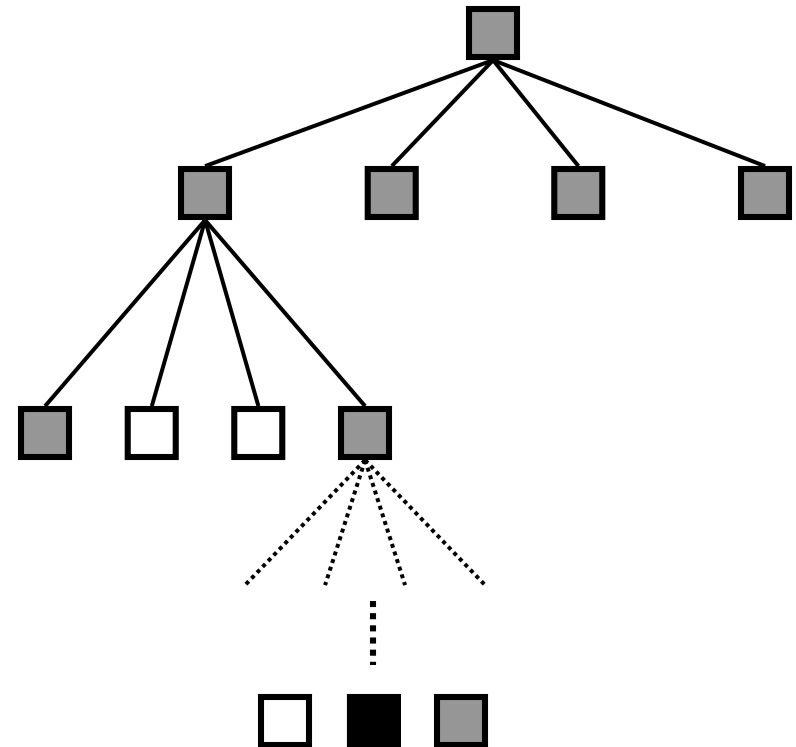
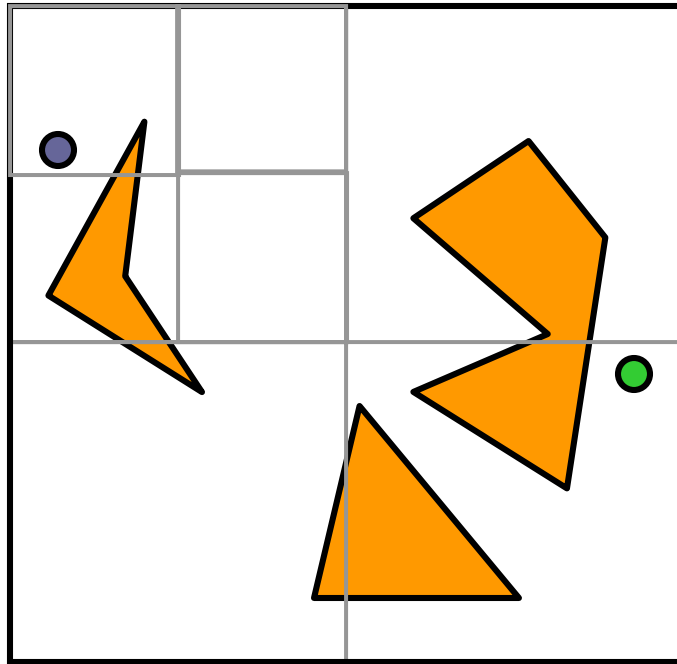
- Exact cell decomposition

- **Approximate cell decomposition**

The free space F is represented by a collection of non-overlapping cells whose union is **contained** in F .

- Cells usually have simple, regular shapes, *e.g.*, rectangles, squares.
- Facilitate hierarchical space decomposition

Quadtree decomposition

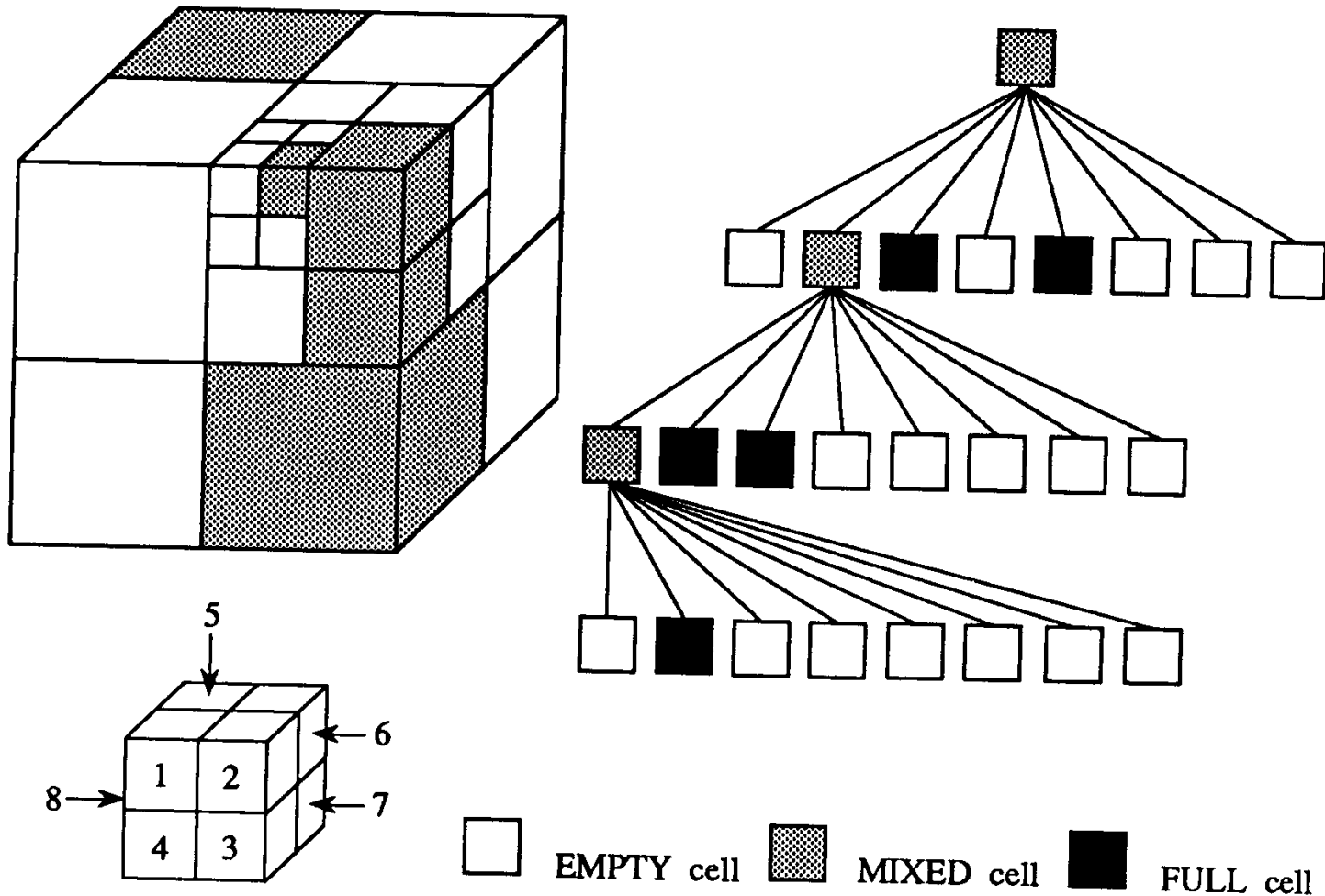


 empty

 mixed

 full

Octree decomposition



Sketch of the algorithm

1. Decompose the free space F into cells.
2. Search for a sequence of **mixed** or **free** cells that connect the initial and goal positions.
3. Further decompose the mixed.
4. Repeat (2) and (3) until a sequence of **free** cells is found.

Classic path planning approaches

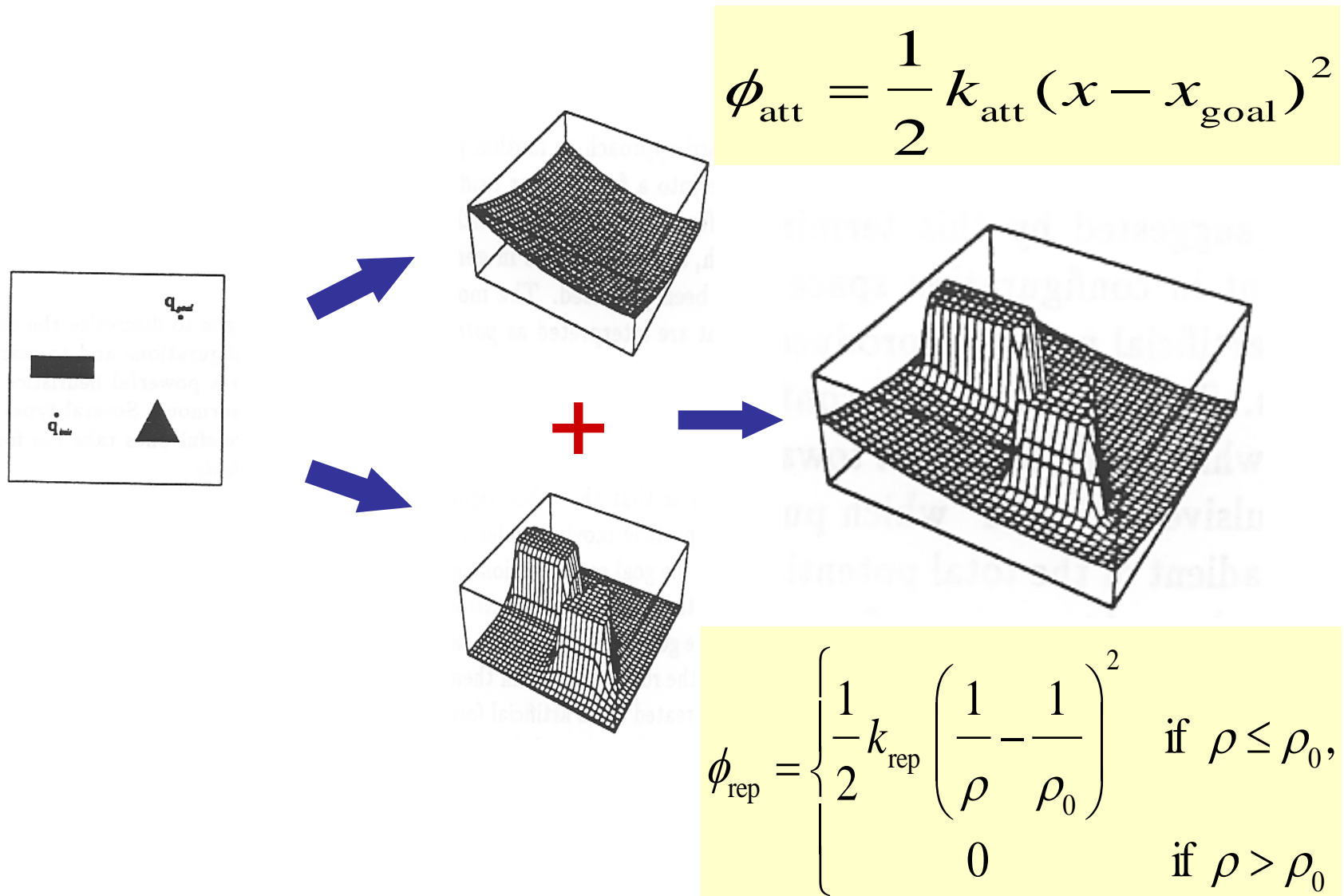
□ Cell decomposition

Decompose the free space into simple cells and represent the connectivity of the free space by the adjacency graph of these cells

□ Potential field

Define a potential function over the free space that has a global minimum at the goal and follow the steepest descent of the potential function

Algorithm in pictures



Attractive & repulsive fields

$$F_{\text{att}} = -\nabla \phi_{\text{att}} = -k_{\text{att}} (x - x_{\text{goal}})$$

$$F_{\text{rep}} = -\nabla \phi_{\text{rep}} = \begin{cases} k_{\text{rep}} \left(\frac{1}{\rho} - \frac{1}{\rho_0} \right) \frac{1}{\rho^2} \frac{\partial \rho}{\partial x} & \text{if } \rho \leq \rho_0, \\ 0 & \text{if } \rho > \rho_0 \end{cases}$$

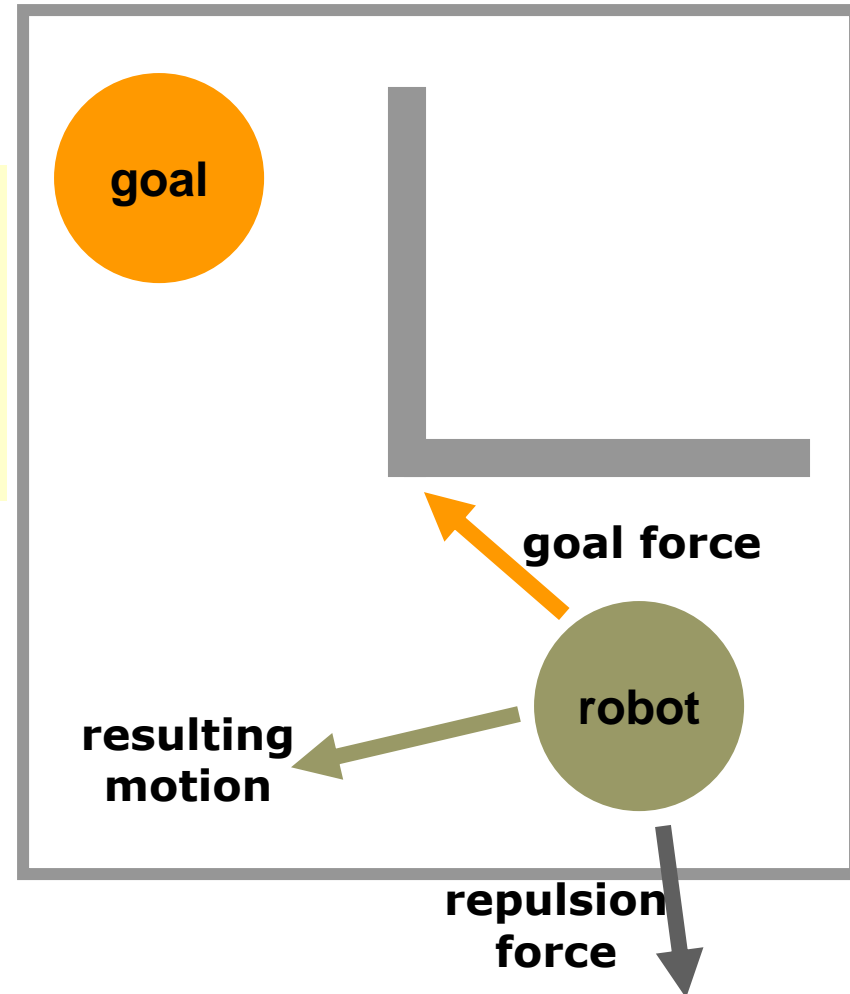
$k_{\text{att}}, k_{\text{rep}}$: positive scaling factors

x : position of the robot

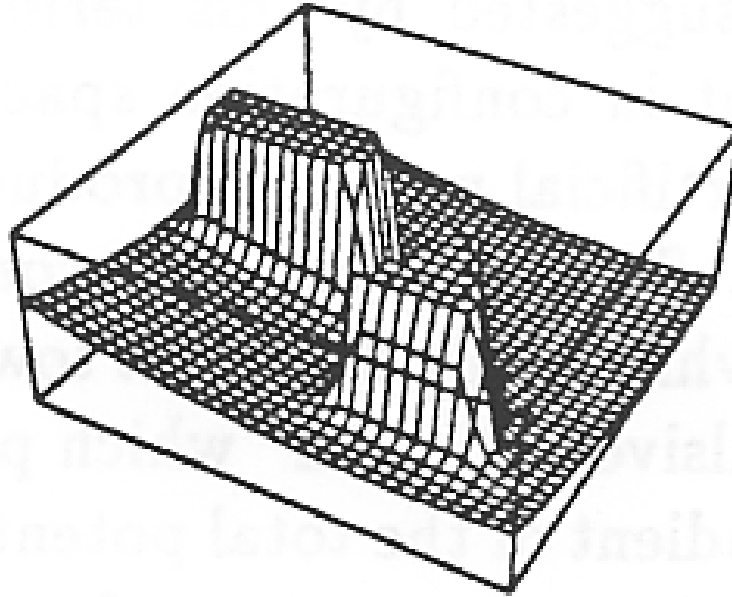
ρ : distance to the obstacle

ρ_0 : distance of influence

[Khatib, 1986]



Local minima



- What can we do?
 - Escape from local minima by taking random walks
 - Build an ideal potential field – navigation function – that does not have local minima
 - Computationally expensive in general

Completeness

- A **complete motion planner** always returns a solution when one exists and indicates that no such solution exists otherwise.
 - Is the visibility graph algorithm complete?
 - Is the exact cell decomposition algorithm complete?
 - Is the potential field algorithm complete?

Homework

- Read LaValle Ch. 4.0-4.3