Path Planning for Point Robots

Sung-Eui Yoon (윤성의)

Course URL: http://sglab.kaist.ac.kr/~sungeui/MPA

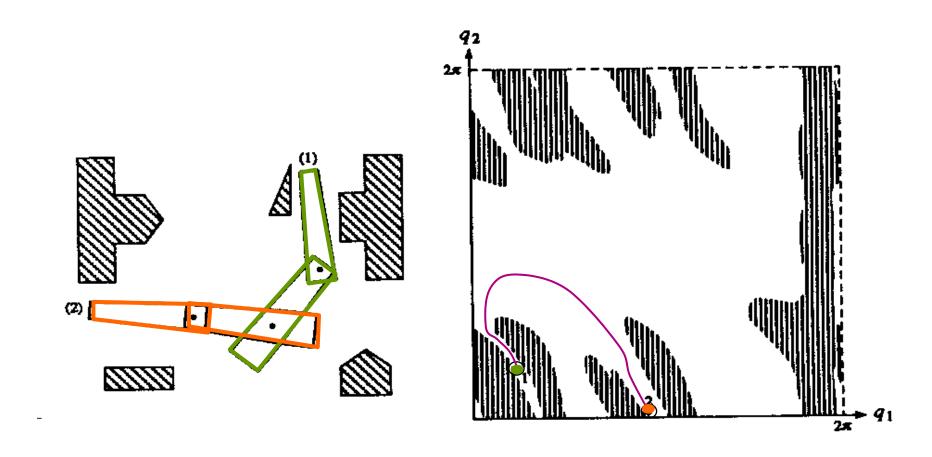


Class Objectives

- Motion planning framework
- Classic motion planning approaches



Configuration Space: Tool to Map a Robot to a Point



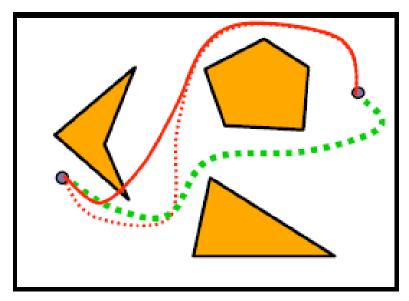


Problem

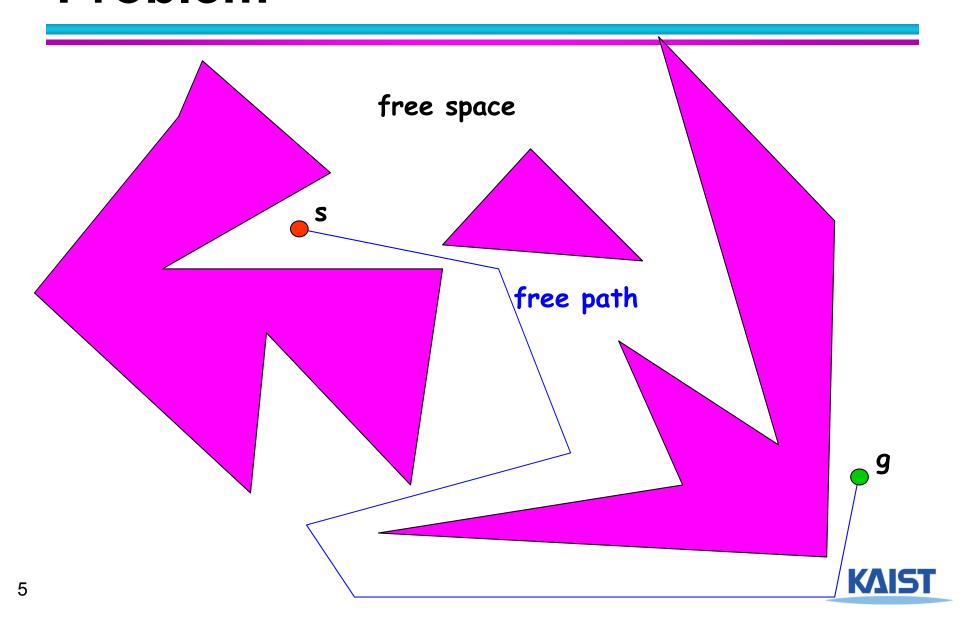
Input

- Robot represented as a point in the plane
- Obstacles represented as polygons
- Initial and goal positions
- Output
 A collision-free path between the initial and goal positions

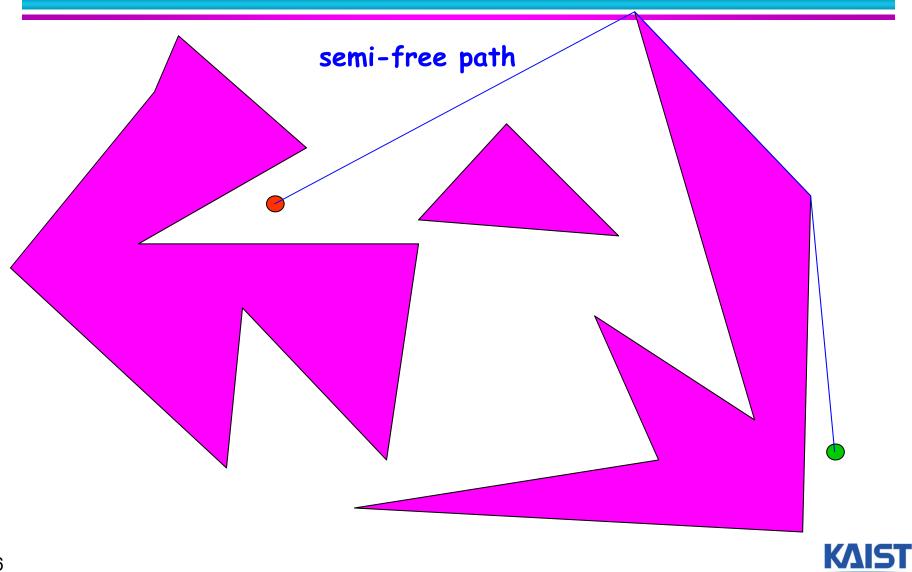




Problem



Problem

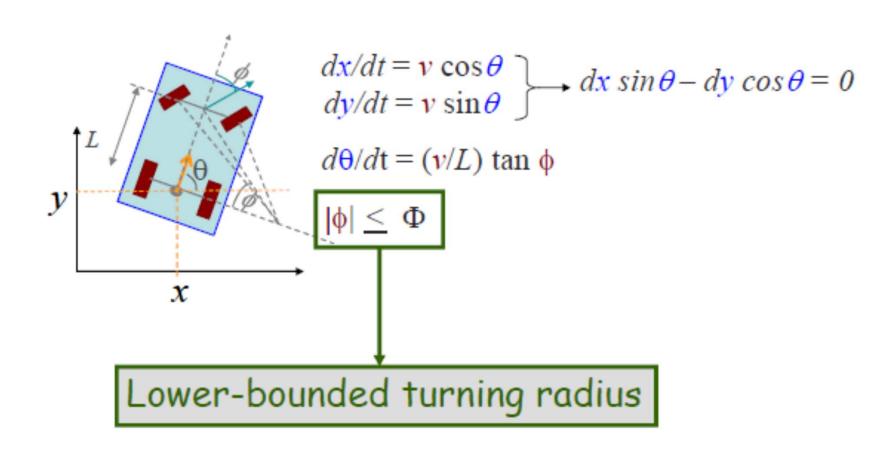


Types of Path Constraints

- Local constraints: lie in free space
- Differential constraints:
 have bounded curvature
- Global constraints: have minimal length



Example: Car-Like Robot



Motion-Planning Framework

Continuous representation

(configuration space formulation)

Discretization

(random sampling, processing critical geometric events)

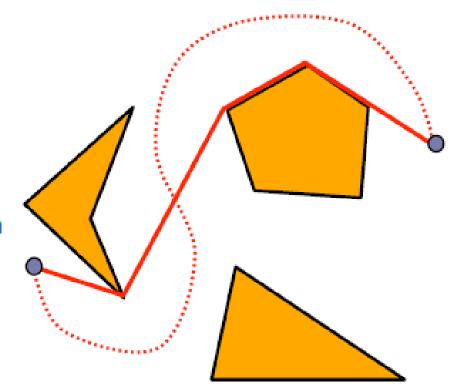
Graph searching

(blind, best-first, A*)



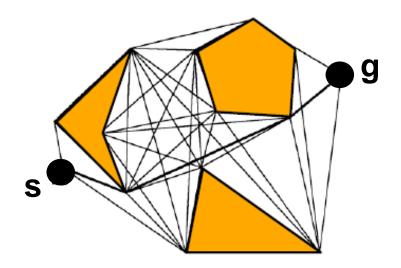
Visibility graph method

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



 A polygonal path is a piecewise linear curve.

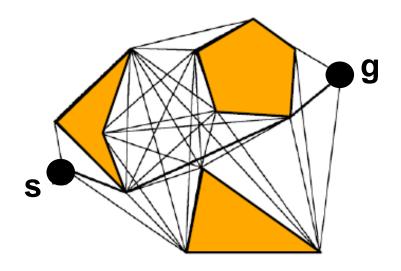
Visibility Graph



- A visibility graph is a graph such that
 - Nodes: s, g, or obstacle vertices
 - Edges: An edge exists between nodes u and v if the line segment between u and v is an obstacle edges or it does not intersect the obstacles



Visibility Graph



- A visibility graph
 - Introduced in the late 60s
 - Can produce shortest paths in 2-D configuration spaces



Simple Algorithm

- Input: s, q, polygonal obstacles
- Output: visibility graph G

```
    for every pair of nodes u, v
    if segment (u, v) is an obstacle edge then
    insert edge (u, v) into G;
    else
    for every obstacle edge e
    if segment (u, v) intersects e
    go to (1);
    insert edge (u, v) into G;
    Search a path with G using A*
```



Computation Efficiency

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

- Simple algorithm: O(n³) time
- More efficient algorithms
 - Rotational sweep O(n² log n) time, etc.
- O(n²) space



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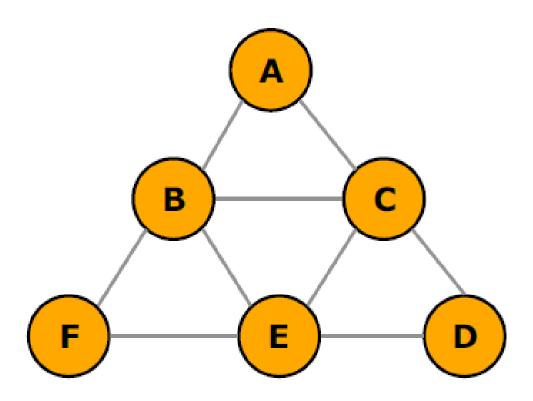
(blind, best-first, A*)

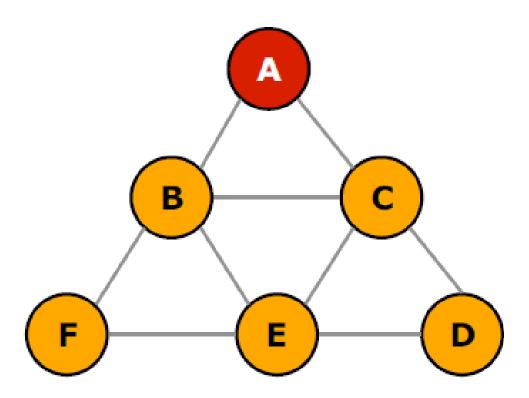


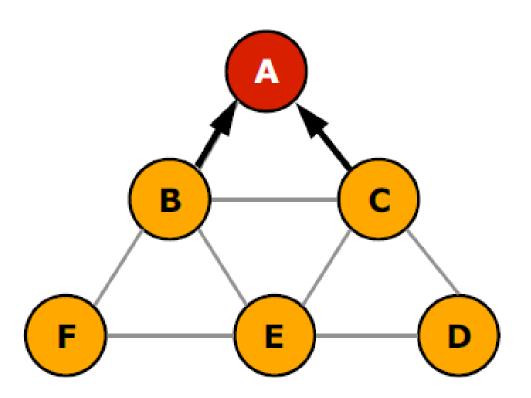
Graph Search Algorithms

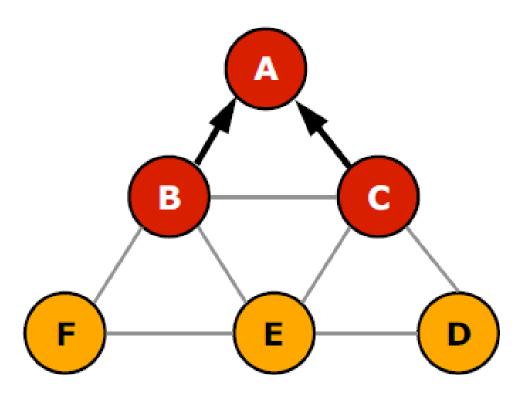
- Breadth, depth-first, best-first
- Dijkstra's algorithm
- A*











Dijkstra's Shortest Path Algorithm

- Given a (non-negative) weighted graph, two vertices, s and g:
 - Find a path of minimum total weight between them
 - Also, find minimum paths to other vertices
 - Has O (|V| Ig|V| + |E|)



Dijkstra's Shortest Path Algorithm

A set S

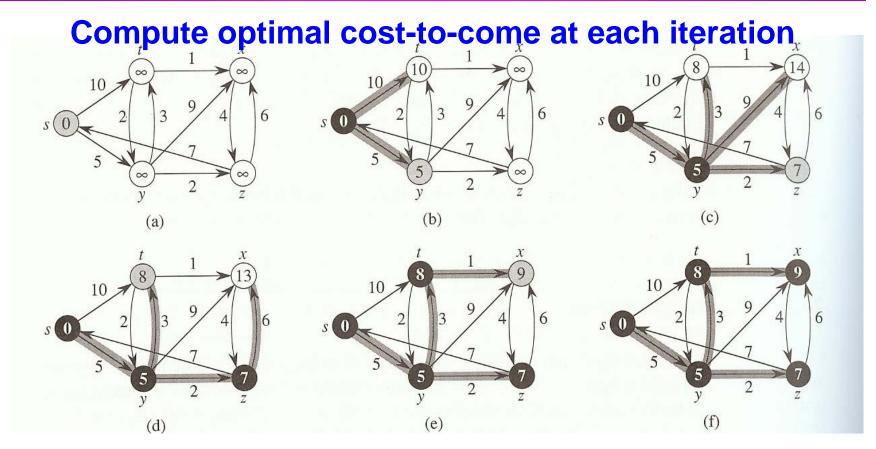
 Contains vertices whose final shortest-path cost has been determined

DIJKSTRA (G, s)

- 1. Initialize-Single-Source (G, s)
- 2. $S \leftarrow empty$
- 3. Queue ← Vertices of G
- 4. While Queue is not empty
- 5. **Do** u ← min-cost from Queue
- 6. S \leftarrow union of S and $\{u\}$
- for each vertex v in Adj [u]
- 8. **do** RELAX (u, v)



Dijkstra's Shortest Path Algorithm



Black vertices are in the set.
White vertices are in the queue.
Shaded one is chosen for relaxation.



A* Search Algorithm

- An extension of Dijkstra's algorithm based on a heuristic estimate
 - Conservatively estimate the cost-to-go from a vertex to the goal
 - The estimate should not be greater than the optimal cost-to-go

Sort vertices based on "cost-to-come + the estimated cost-to-go"

free space

Can find optimal solutions with fewer steps

Best-First Search

- Pick a next node based on an estimate of the optimal cost-to-go cost
 - Greedily finds solutions that look good
 - Solutions may not be optimal
 - Can find solutions quite fast, but can be also very slow



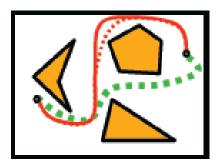
Framework

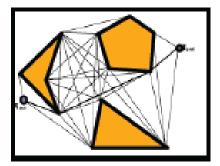
continuous representation

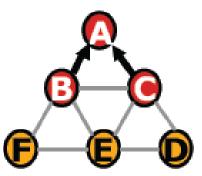


construct visibility graph

↓ graph searching







Computational Efficiency

- Running time O(n³)
 - Compute the visibility graph
 - Search the graph
- Space O(n²)

• Can we do better?



Classic Path Planning Approaches

Roadmap

 Represent the connectivity of the free space by a network of 1-D curves

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 function over the free space that has a global minimum at the goal configuration and follow its steepest descent



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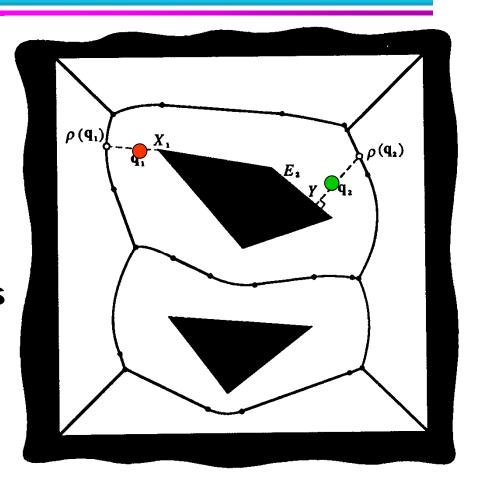
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Roadmap Methods

- Visibility Graph
 - Shakey project, SRI [Nilsson 69]
- Voronoi diagram
 - Introduced by computational geometry researchers
 - Generate paths that maximize clearance
 - O(n log n) time and O(n) space





Other Roadmap Methods

- Visibility graph
- Voronoi diagram
- Silhouette
 - First complete general method that applies to spaces of any dimension and is singly exponential in # of dimensions [Canny, 87]
- Probabilistic roadmaps



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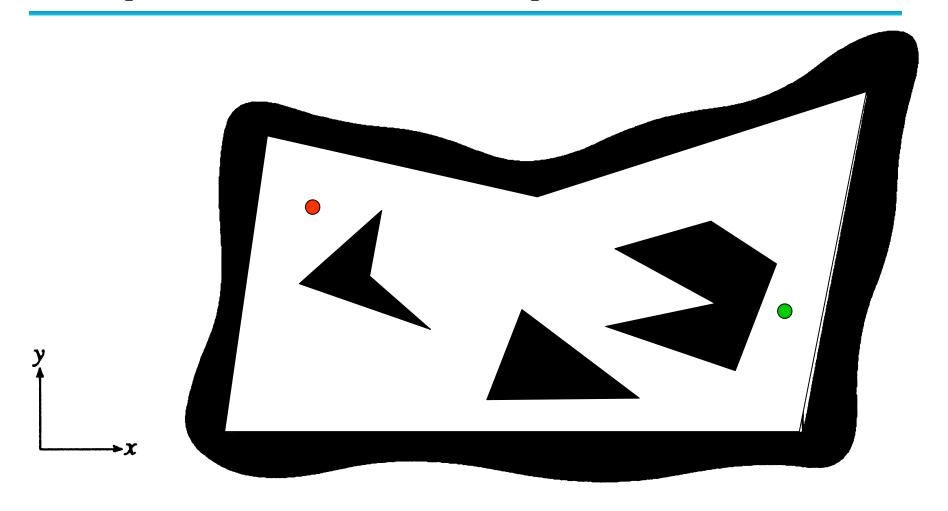
Cell-Decomposition Methods

- Two classes of methods:
 - Exact and approximate cell decompositions

- Exact cell decomposition
 - The free space F is represented by a collection of non-overlapping cells whose union is exactly F
 - Example: trapezoidal decomposition

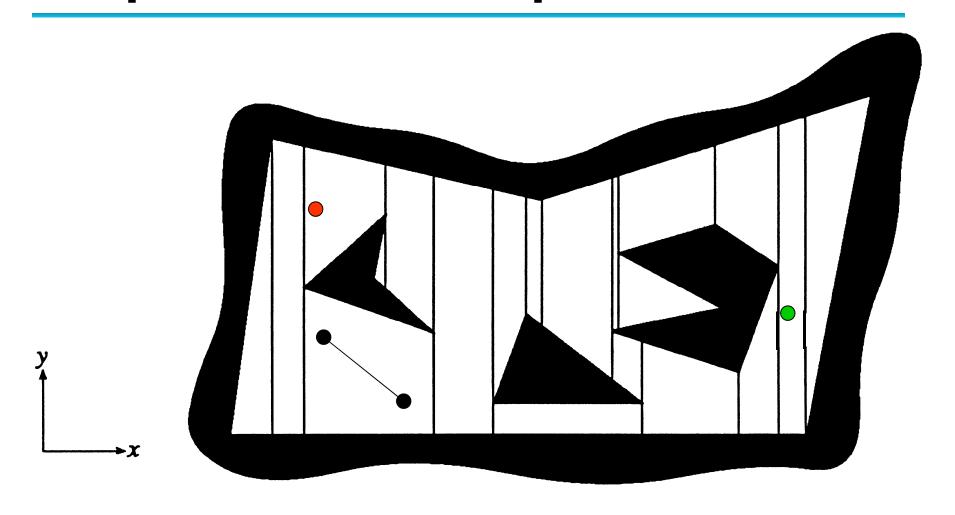


Trapezoidal Decomposition



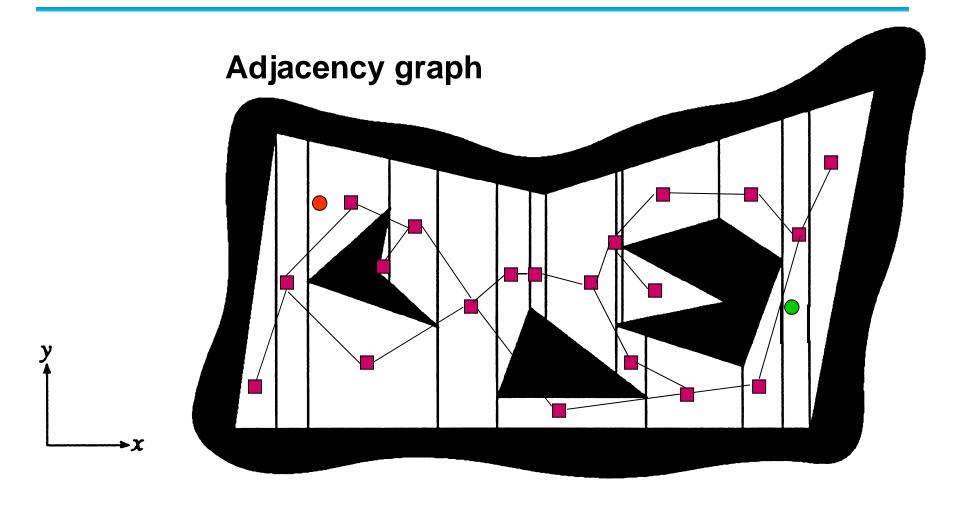


Trapezoidal Decomposition



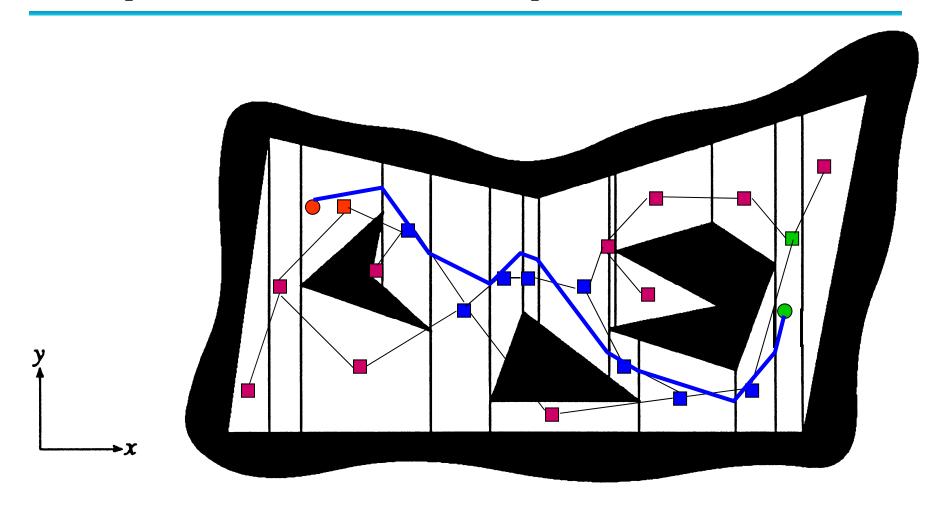


Trapezoidal Decomposition



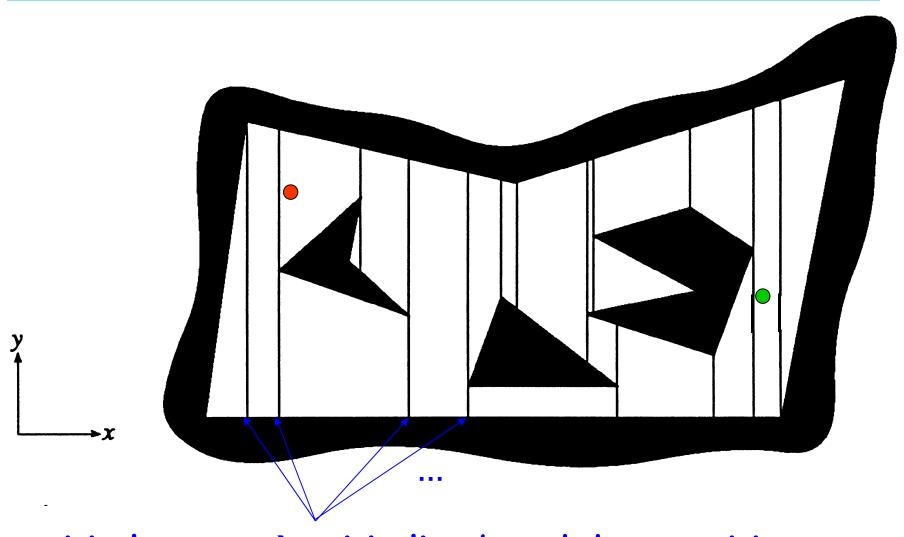


Trapezoidal Decomposition



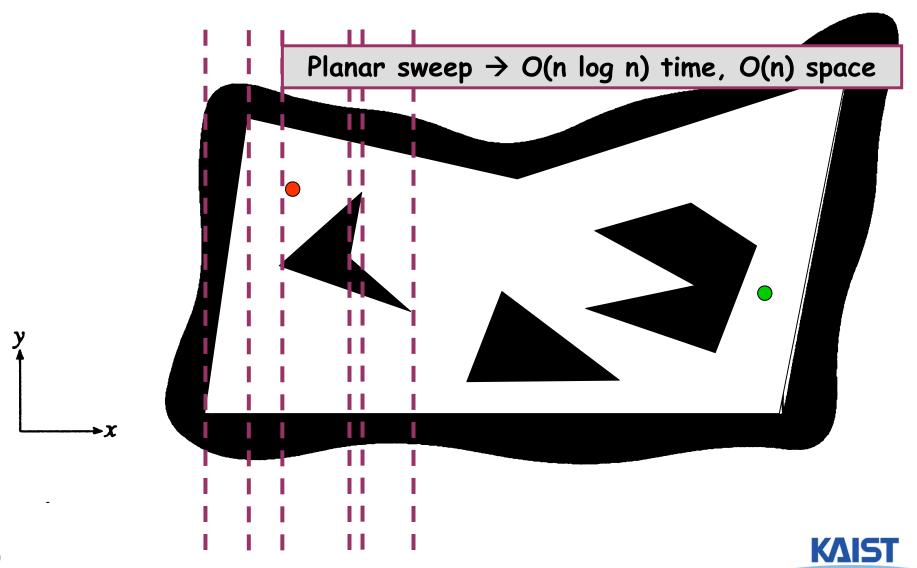


Trapezoidal Decomposition



critical events -> criticality-based decomposition KAIST

Trapezoidal Decomposition



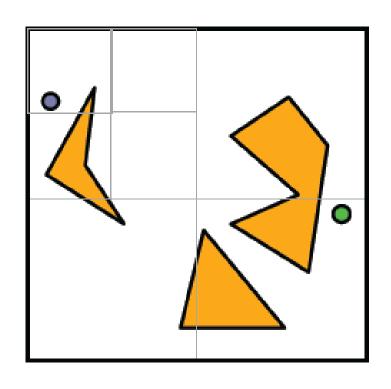
Cell-Decomposition Methods

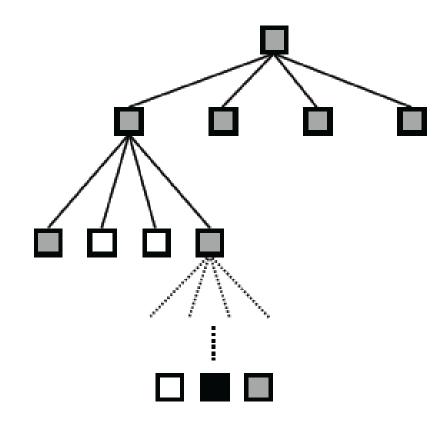
- Two classes of methods:
 - Exact and approximate cell decompositions

- 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 and squares)
 - Facilitates hierarchical space decomposition



Quadtree decomposition



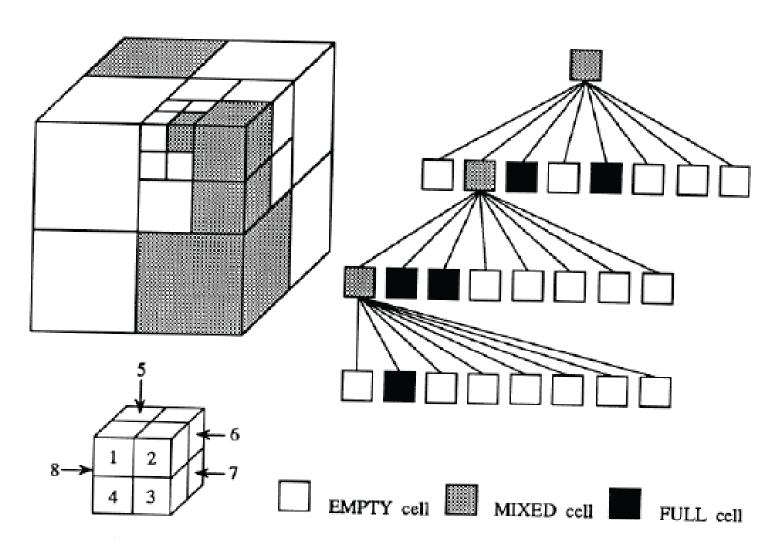


empty

mixed

full

Octree decomposition



Sketch of Algorithm

- 1. Decompose the free space F into cells
- 2. Search for a sequence of mixed or free cells that connect that 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

Roadmap

 Represent the connectivity of the free space by a network of 1-D curves

Cell decomposition

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Potential field

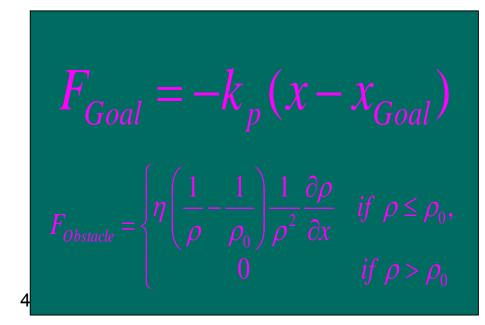
 Define a function over the free space that has a global minimum at the goal configuration and follow its steepest descent

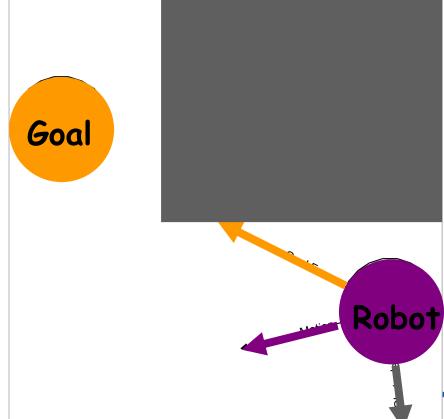


Potential Field Methods

 Initially proposed for real-time collision avoidance [Khatib, 86]

Hundreds of papers published on it



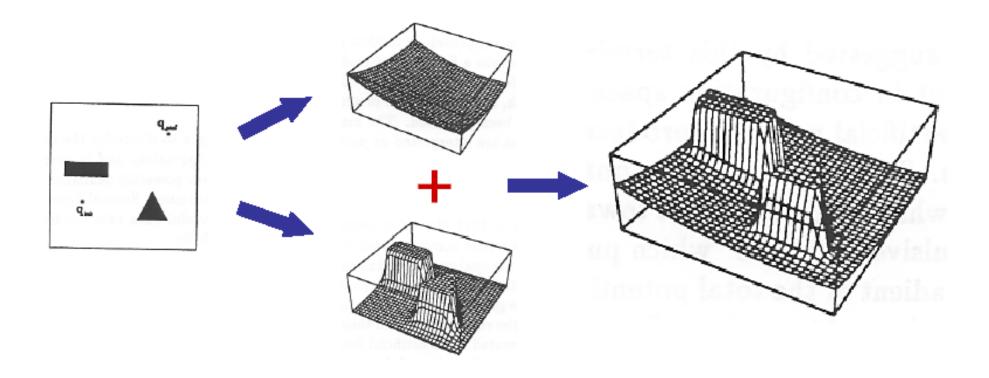


Potential Field

- A scalar function over the free space
- To navigate the robot applies a force proportional to the negated gradient of the potential field
- A navigation function is an ideal potential field that
 - Has global minimum at the goal
 - Has no local minima
 - Grows to infinity near obstacles
 - Is smooth

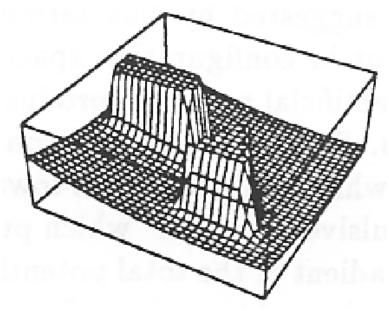


Attractive and Repulsive fields





Local Minima



- What can we do?
 - Escape from local minima by taking random walks
 - Build an ideal potential field that does not have local minima



Sketch of Algorithm

- Place a regular grid G over the configuration space
- Compute the potential field over G
- Search G using a best-first algorithm with potential field as the heuristic function



Question

 Can such an ideal potential field be constructed efficiently 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 algorithm complete? Yes
 - How about the exact cell decomposition algorithm and the potential field algorithm?



Class Objectives were:

- Motion planning framework
- Classic motion planning approaches



Homework for Every Class

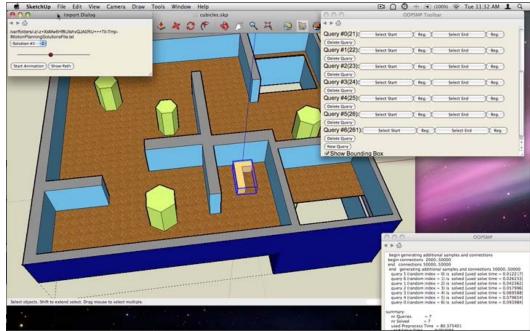
- Go over the next lecture slides
- Come up with one question on what we have discussed today and submit at the end of the class
 - 1 for typical questions
 - 2 for questions with thoughts or that surprised me
- Write a question at least 10 times
 - Do that out of 2 classes



Homework

- Install <u>Open Motion Planning Library</u> (<u>OMPL</u>)
- Create a scene and a robot

Find a collision-free path and visualize the path





Homework

- Deadline: 11:59pm, Sep.-30
- Delivery: send an email to TA (<u>limg00n@kaist.ac.kr</u>) that contains:
 - An image that shows a scene with a robot with a computed path
- Our TA: 임장관, x7851, N1, 924호



Conf. Deadline

- ICRA
 - Sep., 2013
- IROS
 - Mar., 2014





Next Time....

Configuration spaces

