

# Motion Planning

**Jyh-Ming Lien**

Department of Computer Science  
George Mason University

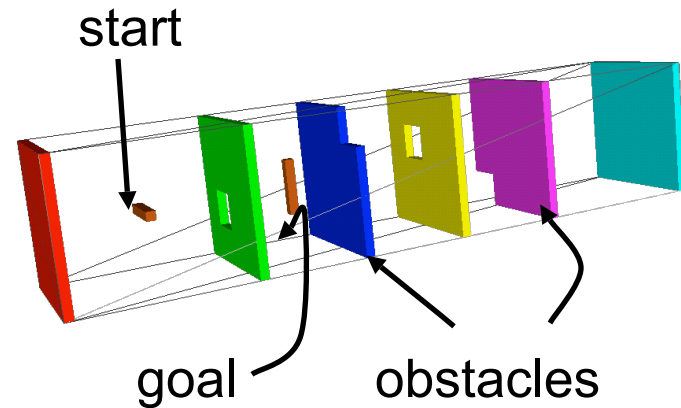
**Based on many people's lecture notes**

Seth Hutchinson at the University of Illinois at Urbana-Champaign, Leo Joskowicz at Hebrew University, Jean-Claude Latombe at Stanford University, Nancy Amato at Texas A&M University, Burchan Bayazit at Washington University in St. Louis

# Motion Planning in continuous spaces

## ***(Basic) Motion Planning*** (in a nutshell):

Given a *movable object*, find a *sequence of valid configurations* that moves the object from the start to the goal.



# Main Steps In Motion Planning

*Workspace*



*Configuration space*



*Discretization*



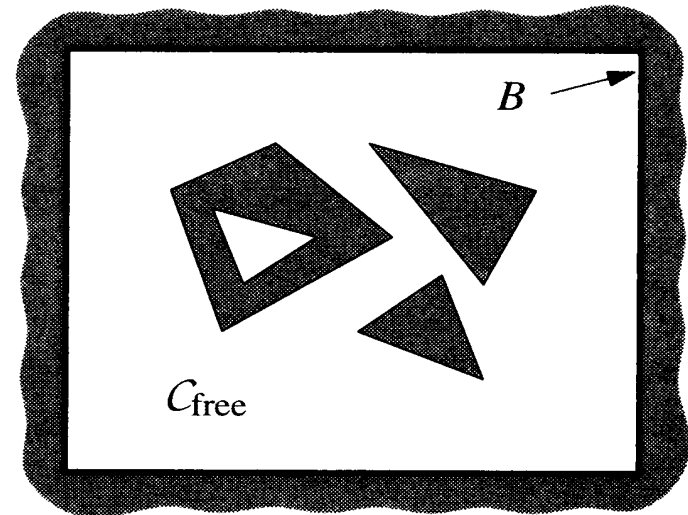
*Search*



*Path or no solution*

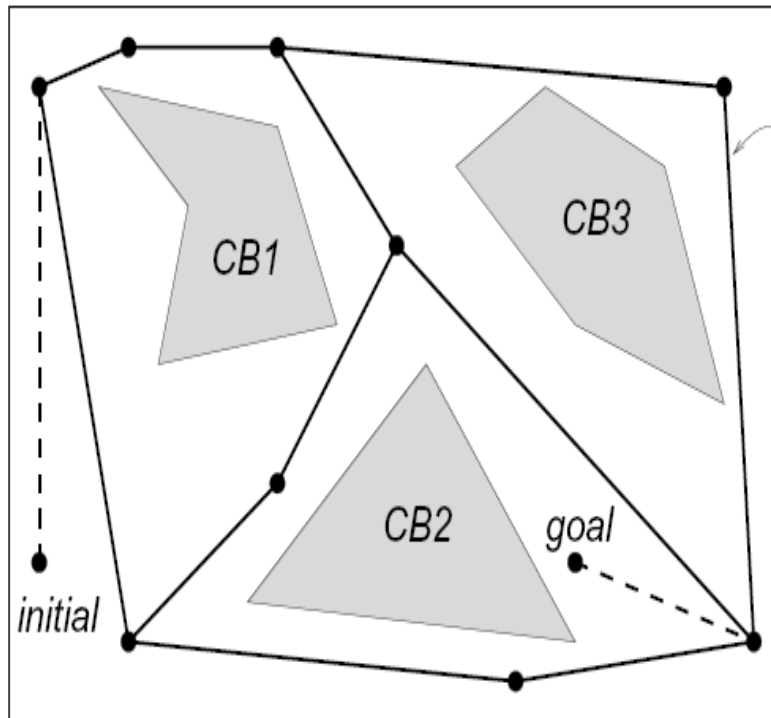
# Classical Motion Planning

- Given a **point robot** and a workspace described by polygons
- **Roadmap methods**
  - Visibility graph
  - Cell decomposition
  - Retraction

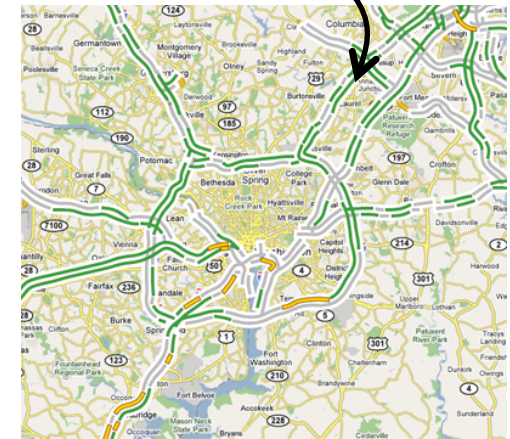


# Roadmap Methods

Capture the connectivity of  $C_{free}$  with a roadmap (graph or network) of one-dimensional curves



roadmap

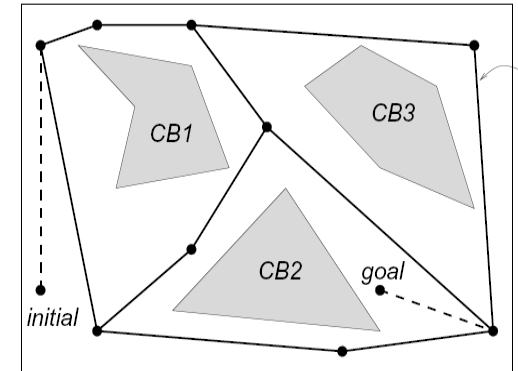


# Roadmap Methods

## Path Planning with a Roadmap

**Input:** configurations  $q_{init}$  and  $q_{goal}$ , and  $B$

**Output:** a path in  $C_{free}$  connecting  $q_{init}$  and  $q_{goal}$



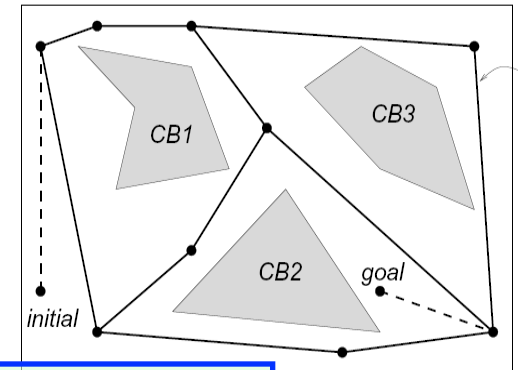
1. Build a roadmap in  $C_{free}$  (preprocessing)
  - roadmap nodes are free configurations (or semi-free)
  - two nodes connected by edge if can (easily) move between them
2. Connect  $q_{init}$  and  $q_{goal}$  to roadmap nodes  $v_{init}$  and  $v_{goal}$
3. Find a path in the roadmap between  $v_{init}$  and  $v_{goal}$ 
  - directly gives a path in  $C_{free}$

# Roadmap Methods

## Path Planning with a Roadmap

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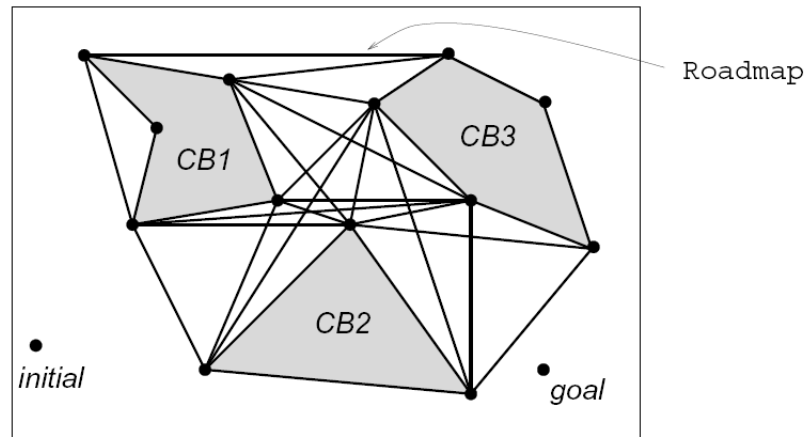
### 3. Find a path in the roadmap between $v_{init}$ and $v_{goal}$

- directly gives a path in  $C_{free}$

↓  
difficult  
part

# Visibility Graph

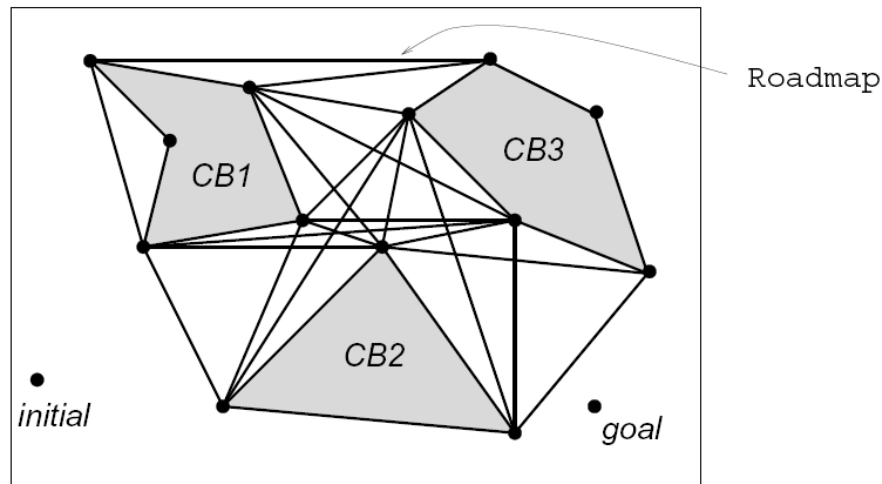
- A visibility graph of C-space for a given C-obstacle is an undirected graph  $G$  where
  - nodes in  $G$  correspond to vertices of C-obstacle
  - nodes connected by edge in  $G$  if
    - they are connected by an edge in C-obstacle, or
    - the straight line segment connecting them lies entirely in  $C_{free}$
  - (could add  $q_{init}$  and  $q_{goal}$  as roadmap nodes)





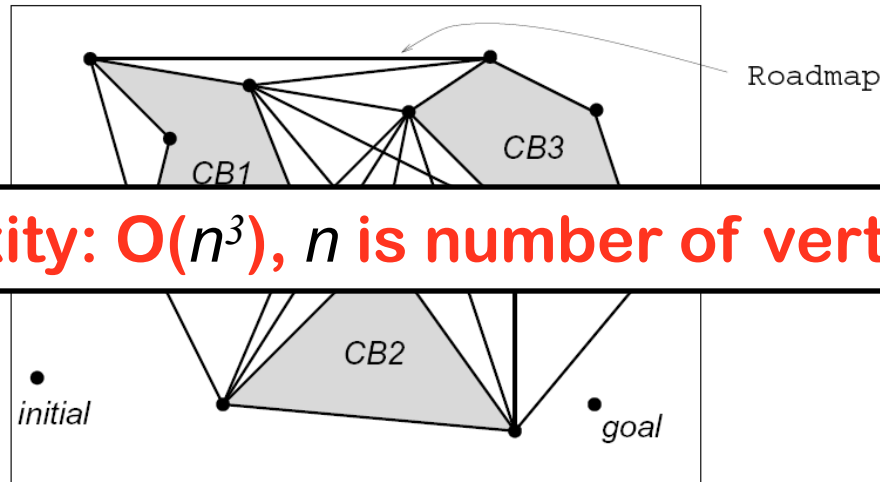
# Visibility Graph

- Brute Force Algorithm
  - add all edges in C-obstacle to  $G$
  - for each pair of vertices  $(x, y)$  of C-obstacle, add the edge  $(x, y)$  to  $G$  if the straight line segment connecting them lies entirely in  $cl(C\text{-free})$ 
    - test  $(x; y)$  for intersection with all  $O(n)$  edges of C-obstacle
    - $O(n^2)$  pairs to test, each test takes  $O(n)$  time



# Visibility Graph

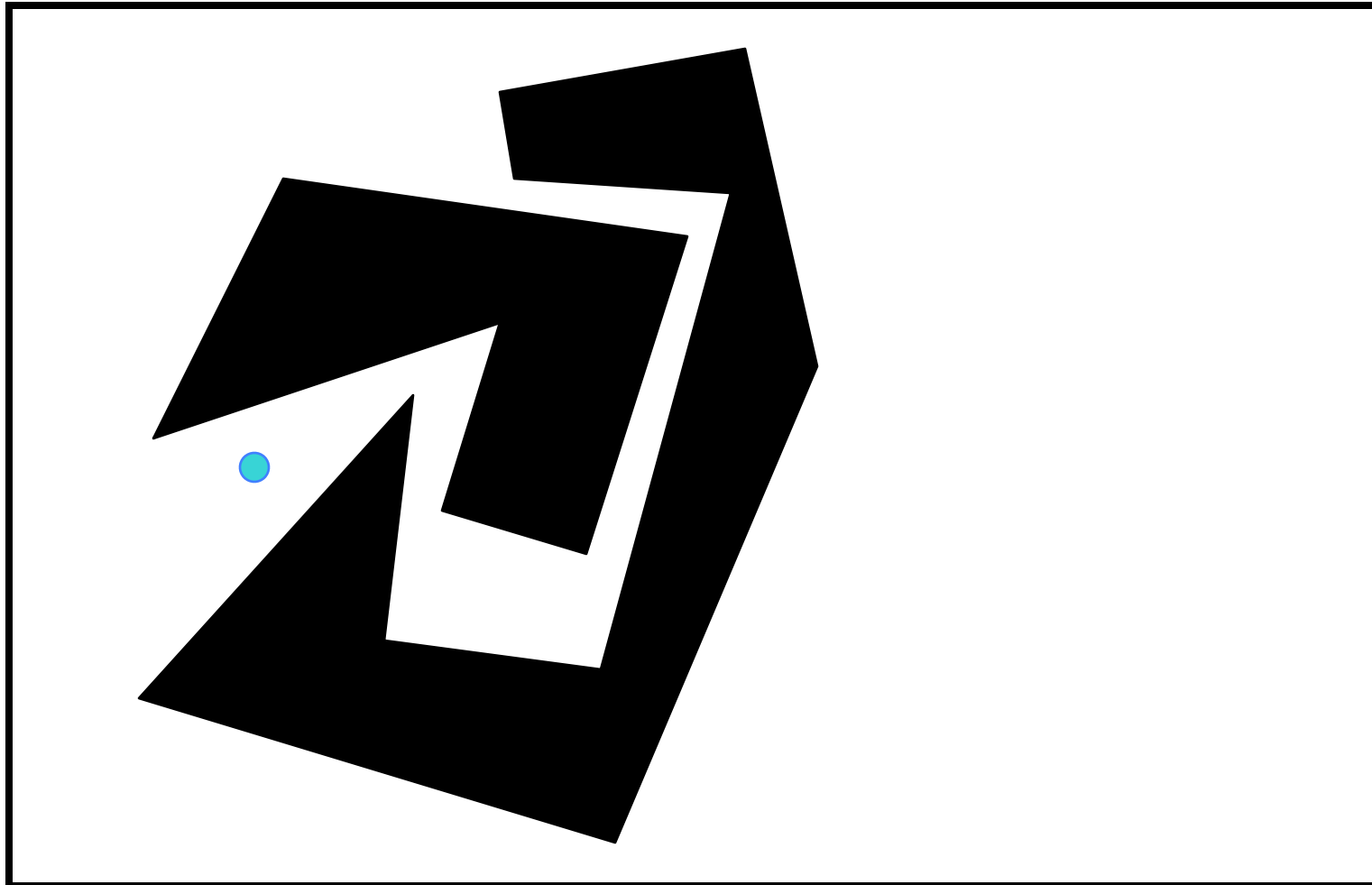
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**Complexity:  $O(n^3)$ ,  $n$  is number of vertices in C-obstacle**

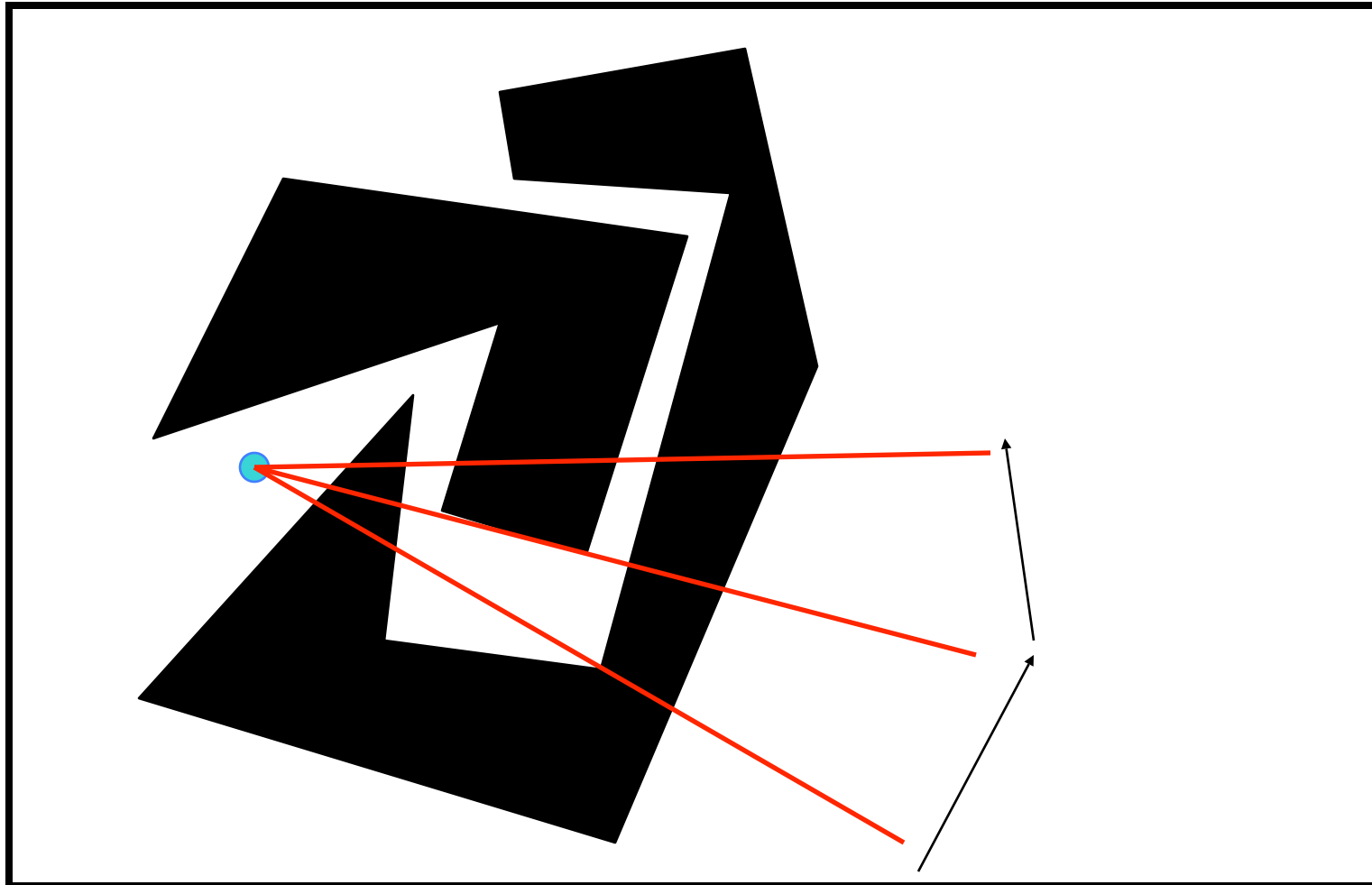
# Visibility Graph

- A better algorithm?



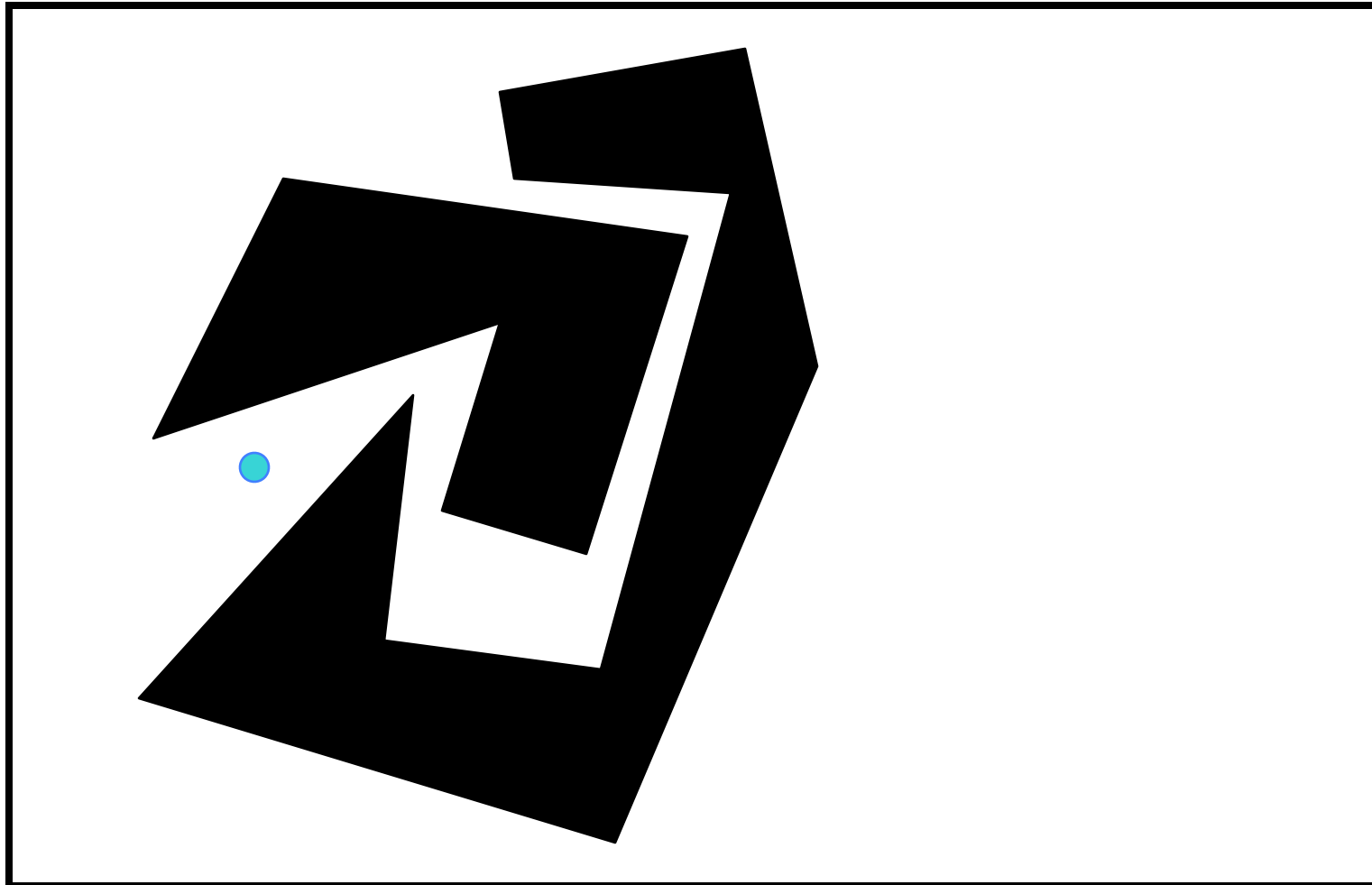
# Visibility Graph

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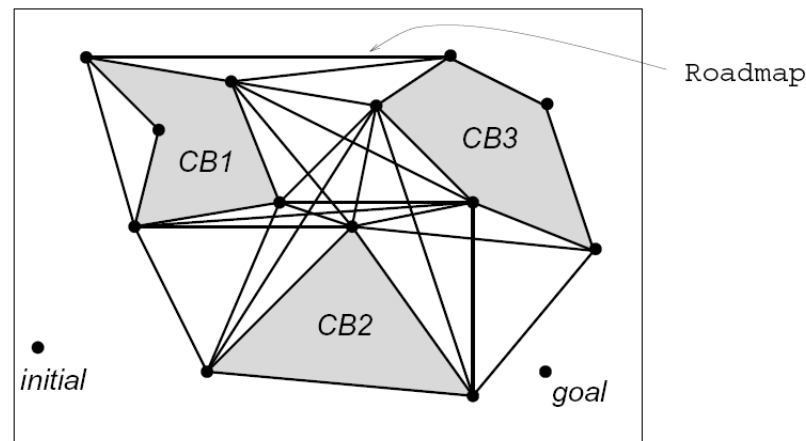
# Visibility Graph

- An even better algorithm?



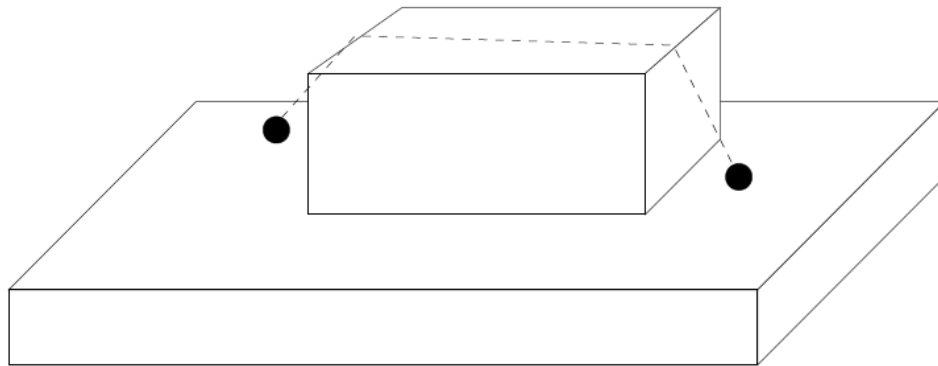
# Visibility Graph

- **Visibility graphs (Good news)**
  - are conceptually simple
  - shortest paths (if the query cannot see each other)
  - we have efficient algorithms if WS is polygonal
    - $O(n^2)$ , where  $n$  is number of vertices of C-obstacle
    - $O(k + n \log n)$ , where  $k$  is number of edges in  $G$
  - we can make a 'reduced' visibility graph (don't need all edges)



# Visibility Graph in 3-D

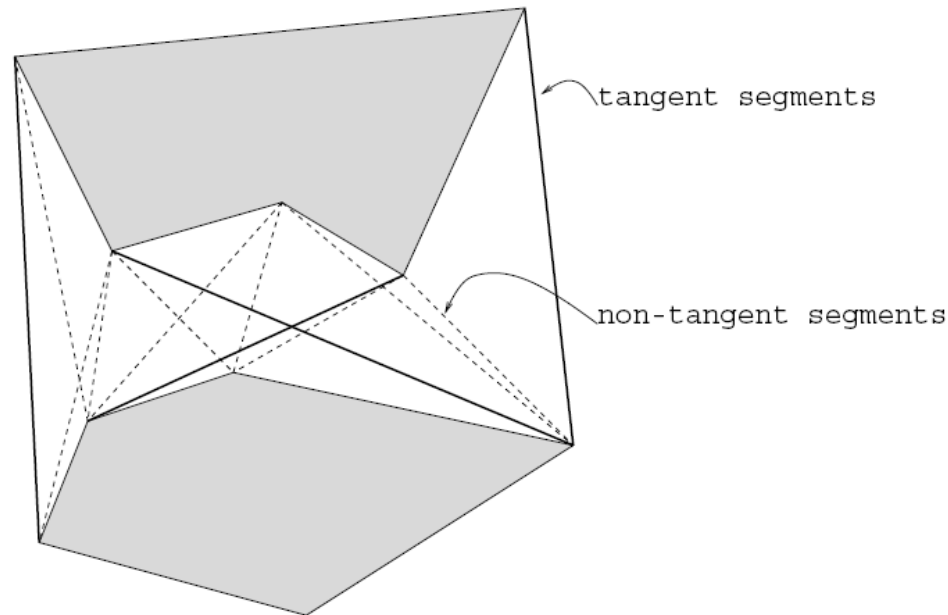
- Visibility graphs don't necessarily contain shortest paths in  $R^3$ 
  - in fact finding shortest paths in  $R^3$  is NP-hard [Canny 1988]
  - $(1 + \epsilon^2)$  approximation algorithm [Papadimitriou 1985]



**Bad news: really only suitable for two-dimensional C**

# Reduced Visibility Graph

- we don't really need all the edges in the visibility graph (even if we want shortest paths)
- **Definition:** Let  $L$  be the line passing through an edge  $(x; y)$  in the visibility graph  $G$ . The segment  $(x; y)$  is a tangent segment *iff*  $L$  is tangent to  $C$ -obstacle at both  $x$  and  $y$ .

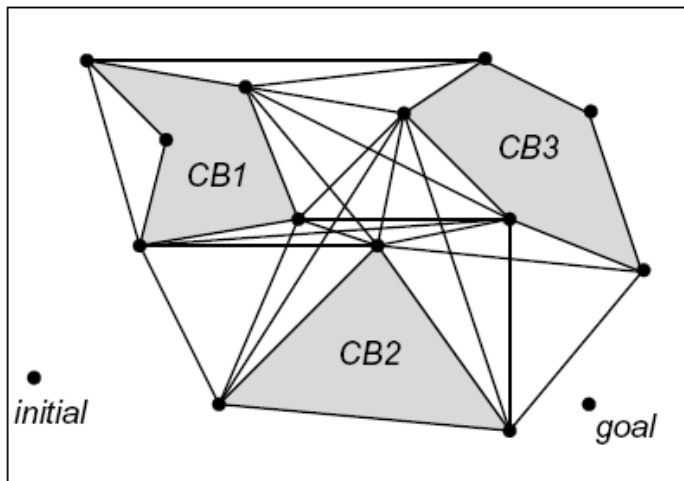




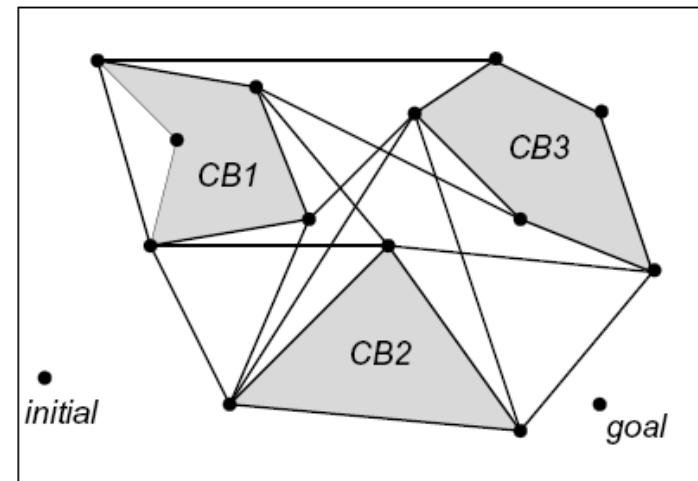
# Reduced Visibility Graph

- It turns out we need only keep
  - convex vertices of C-obstacle
  - non-CB edges that are tangent segments

Visibility Graph

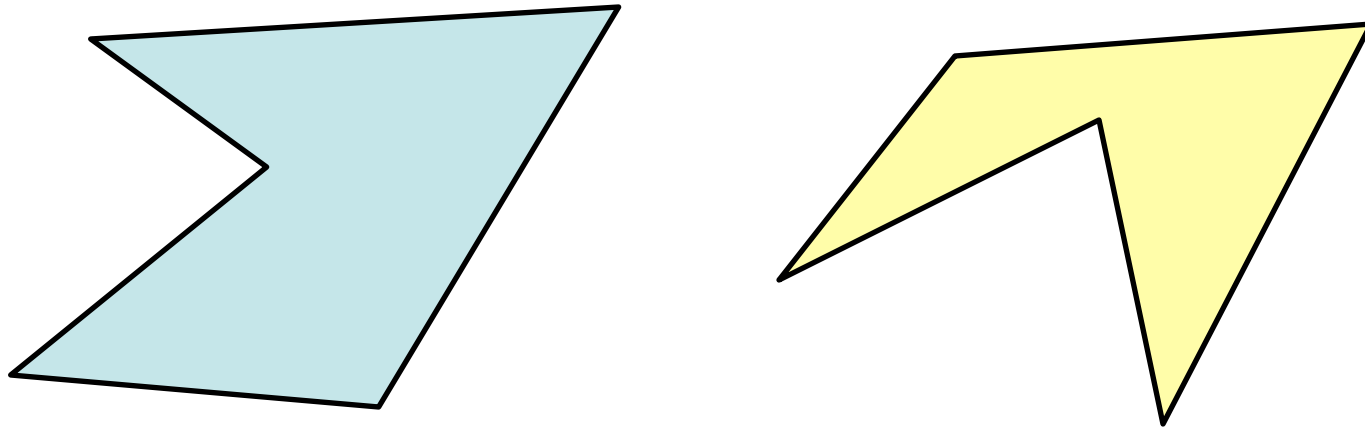


Reduced Visibility Graph



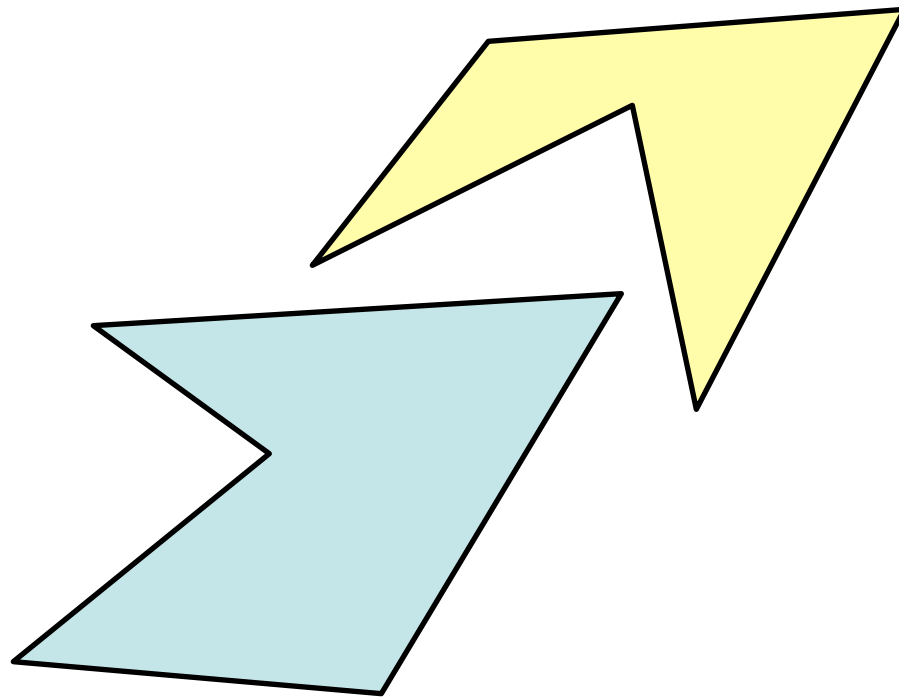
# Reduced Visibility Graph

- Reduced visibility graphs are easier to build
  - construct convex hull of each C-obstacle piece eliminate non-convex vertices
  - construct pairwise tangents between each convex C-obstacle piece
- easy to construct tangents between two convex polygons
  - How?



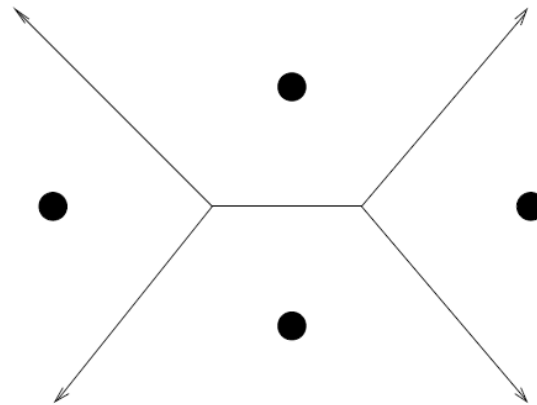
# Reduced Visibility Graph

- Reduced visibility graph does not work if the convex hulls of two obstacles intersect



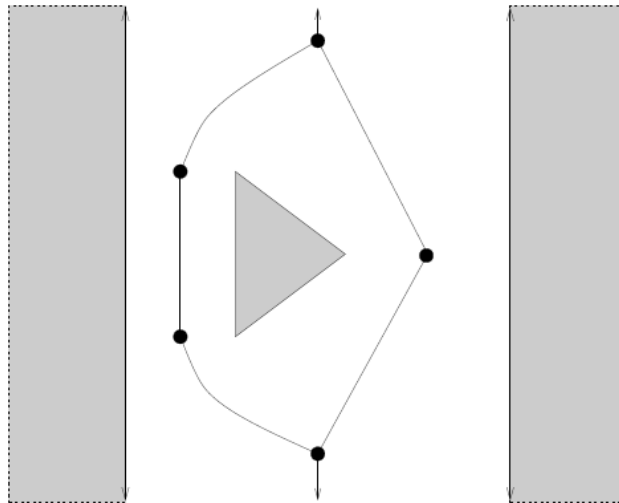
# Voronoi Diagram for Point Sets

- Voronoi diagram of point set  $X$  consists of **straight line segments**, constructed by
  - computing lines bisecting each pair of points and their intersections
  - computing intersections of these lines
  - keeping segments with more than one nearest neighbor
- segments of  $\text{Vor}(X)$  have **largest clearance** from  $X$  and regions identify closest point of  $X$



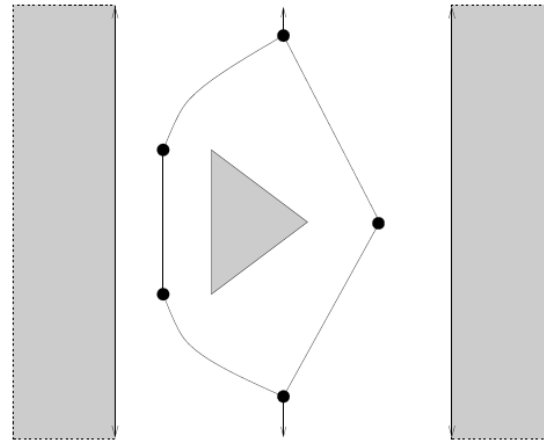
# Voronoi Diagram for Point Sets

- When  $C = \mathbb{R}^2$  and polygonal  $C$ -obstacle,  $\text{Vor}(C_{\text{free}})$  consists of a finite collection of **straight line segments** and **parabolic curve segments** (called arcs)
  - straight arcs are defined by **two vertices** or **two edges** of  $C$ -obstacle, i.e., the set of points equally close to two points (or two line segments) is a line



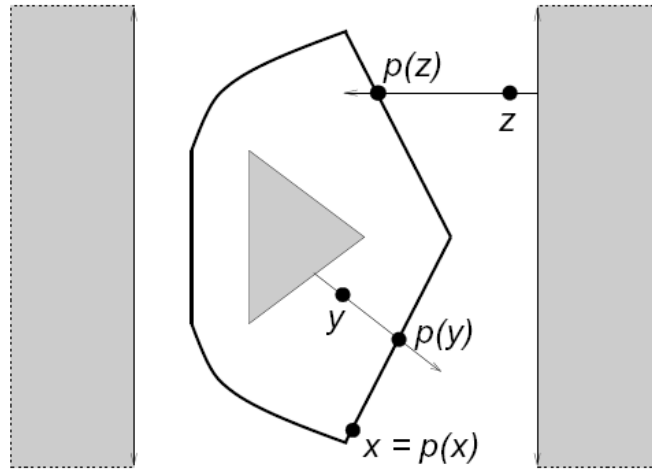
# Voronoi Diagram for Point Sets

- Naive Method of Constructing  $V$  or  $(C_{free})$ 
  - compute all arcs (for each vertex-vertex, edge-edge, and vertex-edge pair)
  - compute all intersection points (dividing arcs into segments)
  - keep segments which are closest only to the vertices/edges that



# Retraction

- Retraction  $\rho : C_{free} \rightarrow \text{Vor}(C_{free})$



## To find a path:

1. compute  $\text{Vor}(C_{free})$
2. find paths from  $q_{init}$  and  $q_{goal}$  to  $\rho(q_{init})$  and  $\rho(q_{goal})$ , respectively
3. search  $\text{Vor}(C_{free})$  for a set of arcs connecting  $\rho(q_{init})$  and  $\rho(q_{goal})$

# Cell Decomposition

- **Idea:** decompose  $C_{free}$  into a collection  $K$  of non-overlapping cells such that the union of all the cells exactly equals the free  $C$ -space
- Cell Characteristics:
  - geometry of cells should be **simple** so that it is easy to compute a path between any two configurations in a cell
  - it should be pretty **easy to test the adjacency** of two cells, i.e., whether they share a boundary
  - it should be pretty easy to find a path crossing the portion of the boundary shared by two adjacent cells
- Thus, cell boundaries correspond to 'criticalities' in  $C$ , i.e., something changes when a cell boundary is crossed. No such criticalities in  $C$  occur within a cell.



# Cell Decomposition

# Cell Decomposition



**Difficult**

# Cell Decomposition

- **Preprocessing:**

- represent  $C_{free}$  as a collection of cells (connected regions of  $C_{free}$ )
  - planning between configurations in the same cell should be 'easy'
- build connectivity graph representing adjacency relations between cells
  - cells adjacent if can move directly between them

- **Query:**

- locate cells  $k_{init}$  and  $k_{goal}$  containing start and goal configurations
- search the connectivity graph for a 'channel' or sequence of adjacent cells connecting  $k_{init}$  and  $k_{goal}$
- find a path that is contained in the channel of cells

- Two major variants of methods:

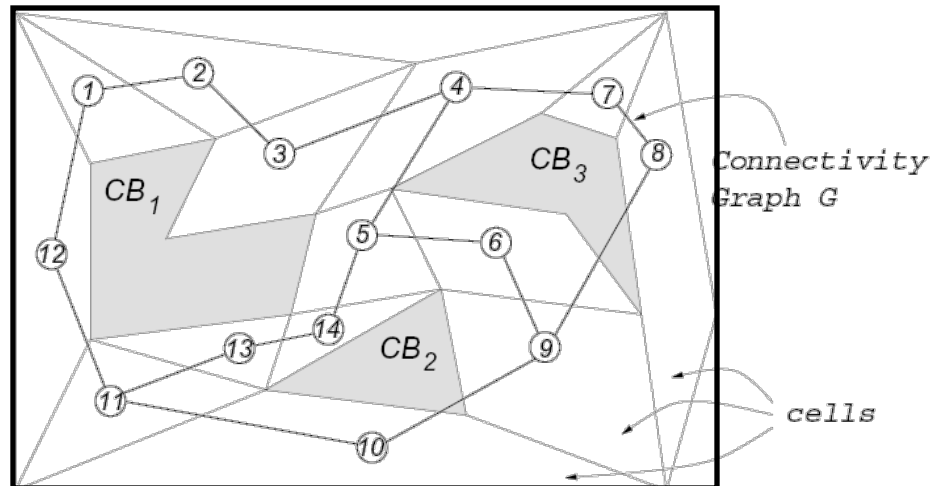
- exact cell decomposition:
  - set of cells exactly covers  $C_{free}$
  - complicated cells with irregular boundaries (contact constraints)
  - harder to compute
- approximate cell decomposition:
  - set of cells approximately covers  $C_{free}$
  - simpler cells with more regular boundaries



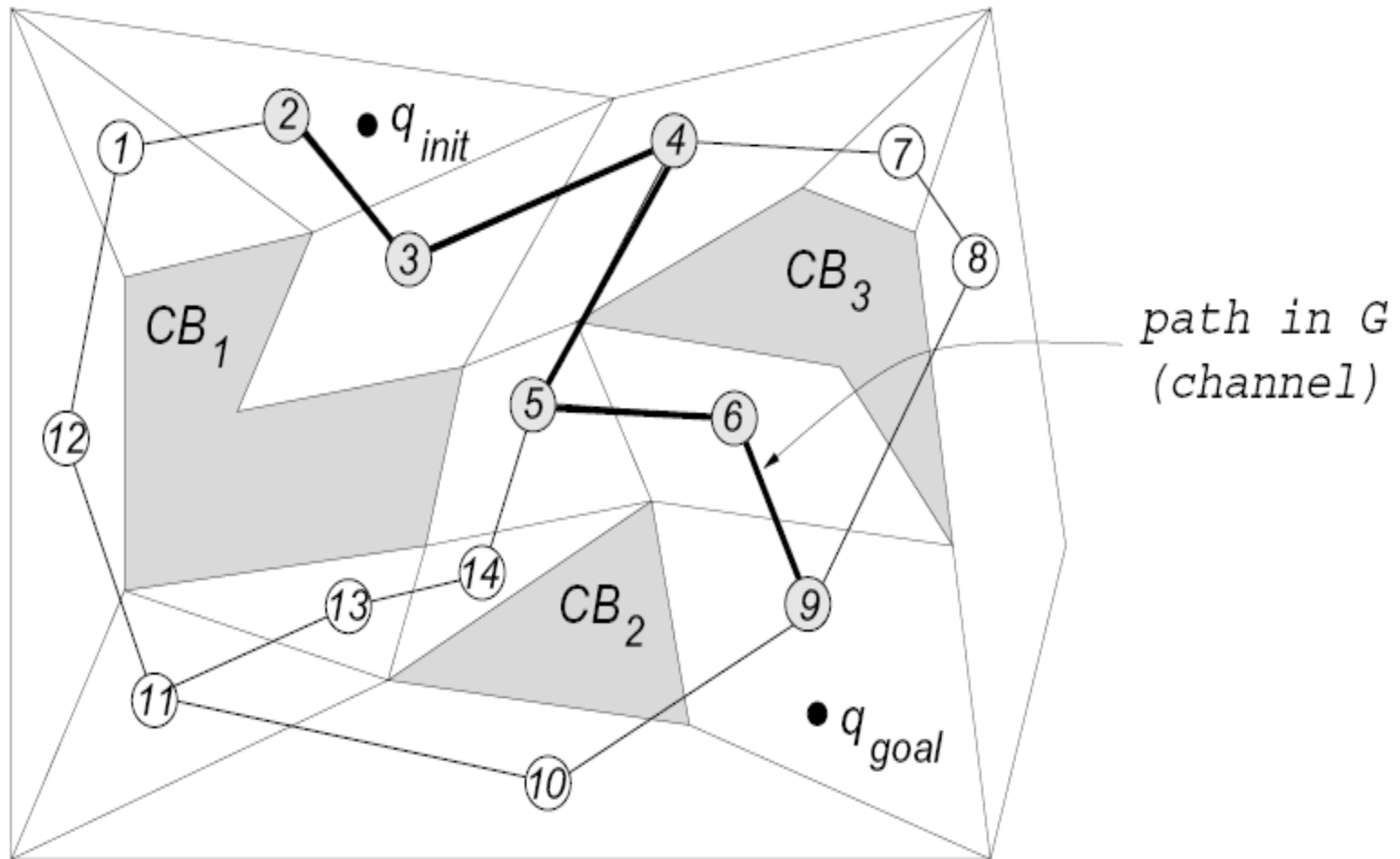
**Difficult**

# Convex Decomposition

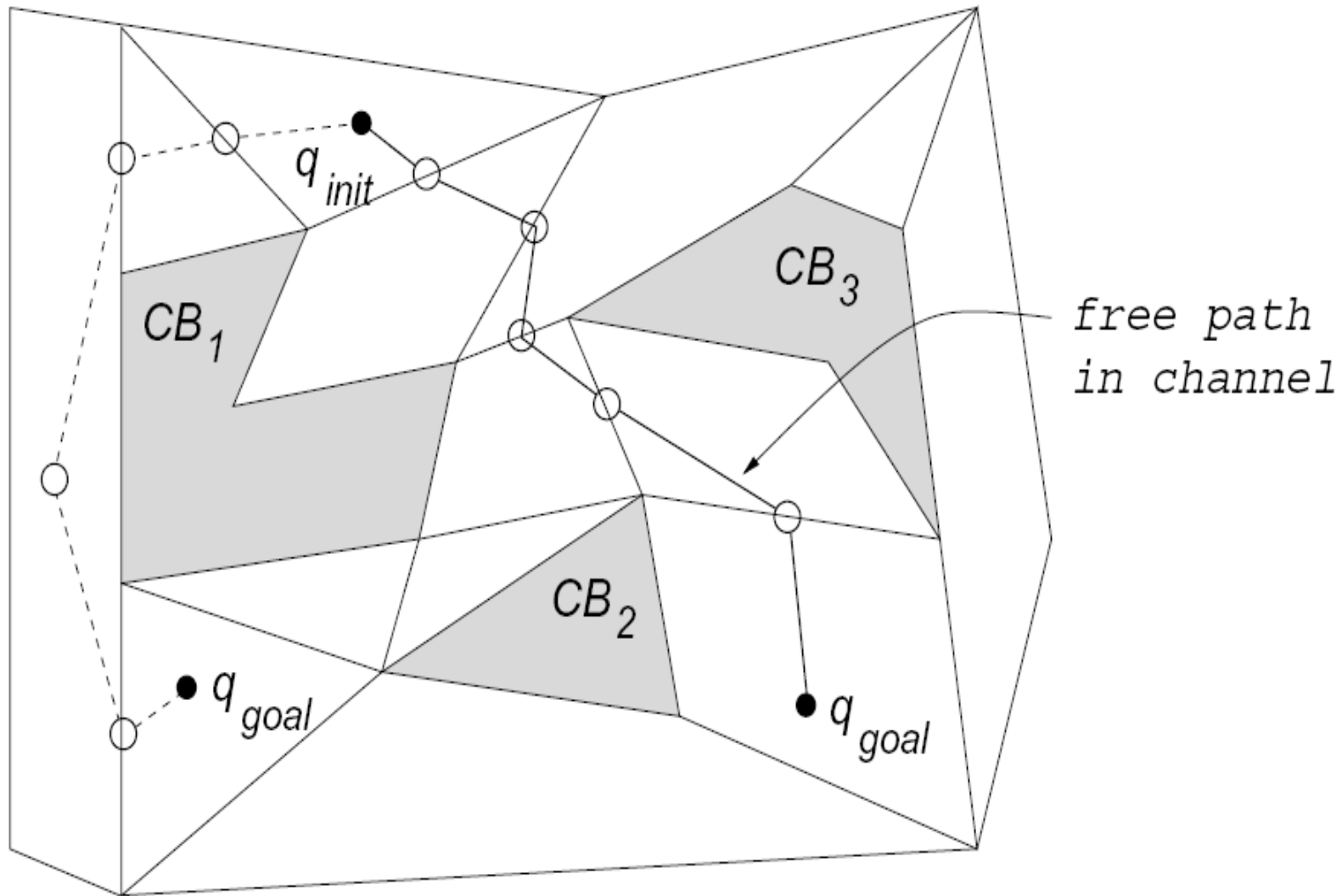
- A **convex polygonal decomposition**  $K$  of  $C_{free}$  is a finite collection of convex polygons, called cells, such that the interiors of any two cells do not intersect and the union of all cells is  $C_{free}$ .
  - Two cells  $k$  and  $k' \in K$  are adjacent iff  $k \cap k'$  is a line segment of non-zero length (i.e., not a single point)
- The **connectivity graph** associated with a convex polygonal decomposition  $K$  of  $C_{free}$  is an undirected graph  $G$  where
  - nodes in  $G$  correspond to cells in  $K$
  - nodes connected by edge in  $G$  iff corresponding cells adjacent in  $K$



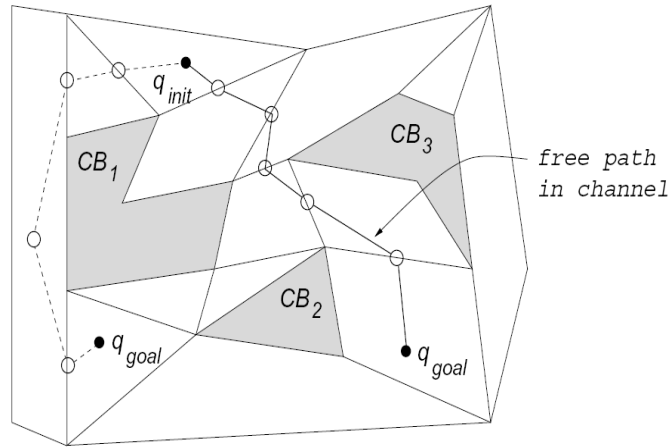
# Convex Decomposition



# Convex Decomposition



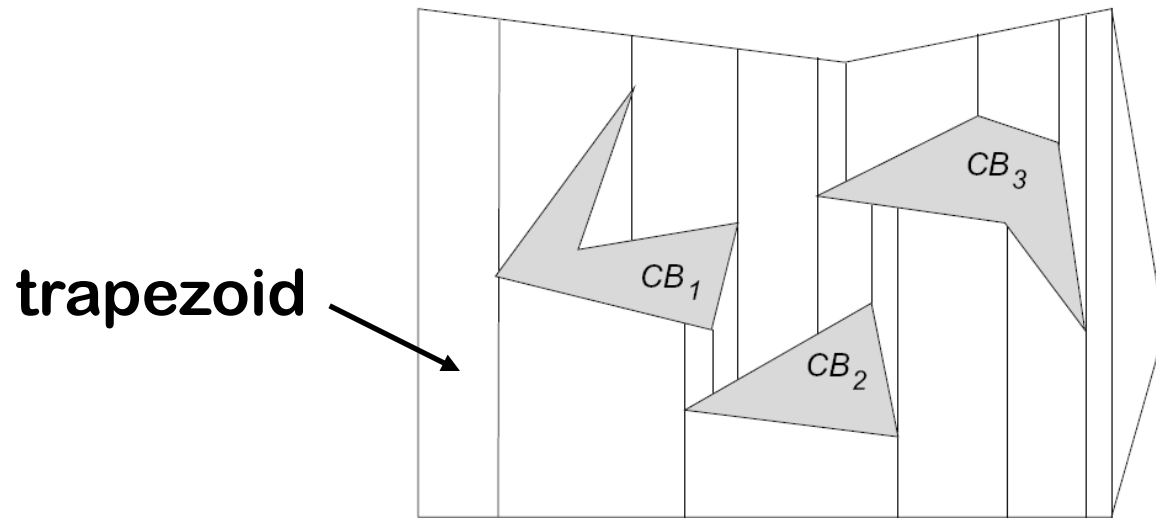
# Convex Decomposition



**Bad news:** Computing convex decomposition is not easy nor can be done efficiently. In fact the problem is NP hard to generate minimum number of convex components for polygon with holes

# Trapezoidal Decomposition

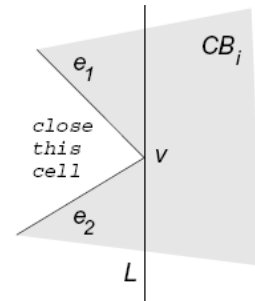
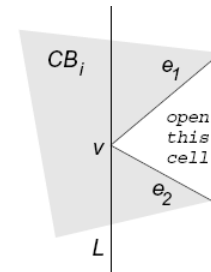
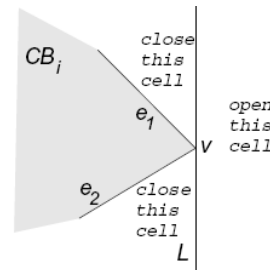
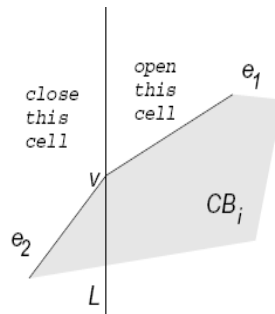
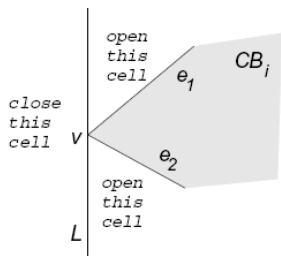
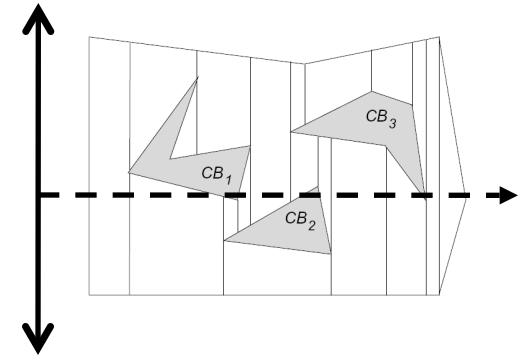
- Basic Idea: at every vertex of C-obstacle, extend a vertical line up and down in  $C_{free}$  until it touches a C-obstacle or the boundary of  $C_{free}$





# Trapezoidal Decomposition

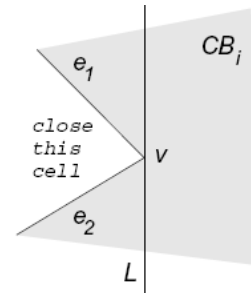
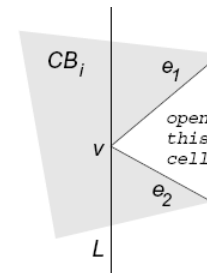
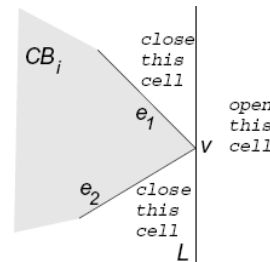
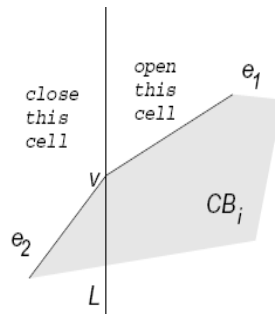
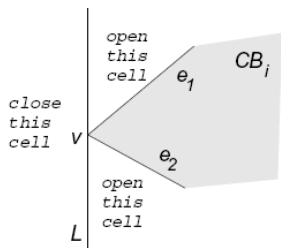
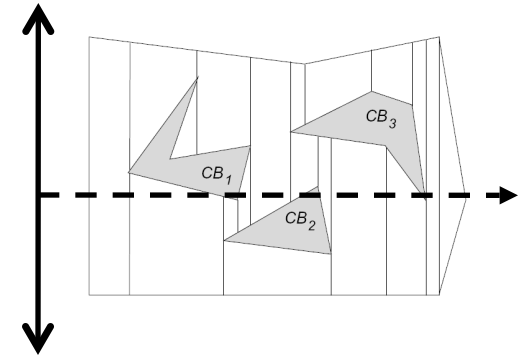
- Sweep line algorithm
  - Add vertical lines as we sweep from left to right
  - Events need to be handled accordingly



# Trapezoidal Decomposition

- Sweep line algorithm

- Add vertical lines as we sweep from left to right
- Events need to be handled accordingly



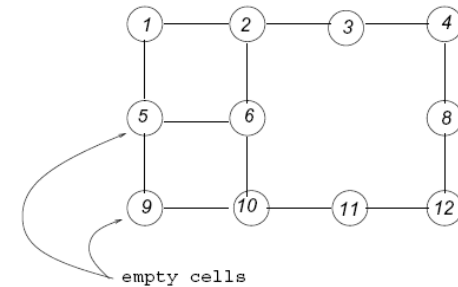
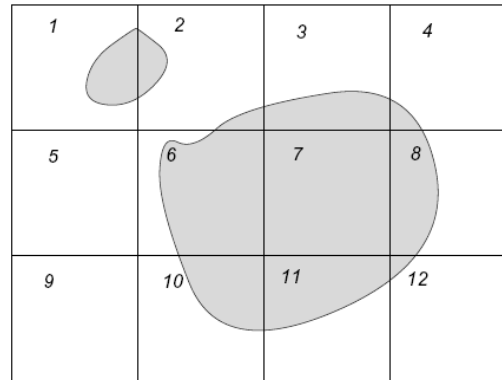
trapezoidal decomposition can be built in  $O(n \log n)$  time

# Approx. Cell Decomposition

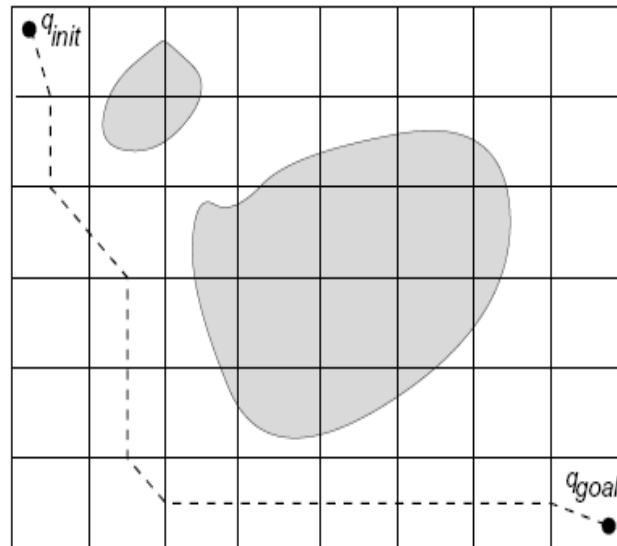
- Construct a collection of non-overlapping cells such that the union of all the cells **approximately** covers the free C-space!
- Cell characteristics
  - Cell should have simple shape
  - Easy to test adjacency of two cells
  - Easy to find path across two adjacent cells

# Approx. Cell Decomposition

- Each cell is
  - Empty
  - Full
  - Mixed

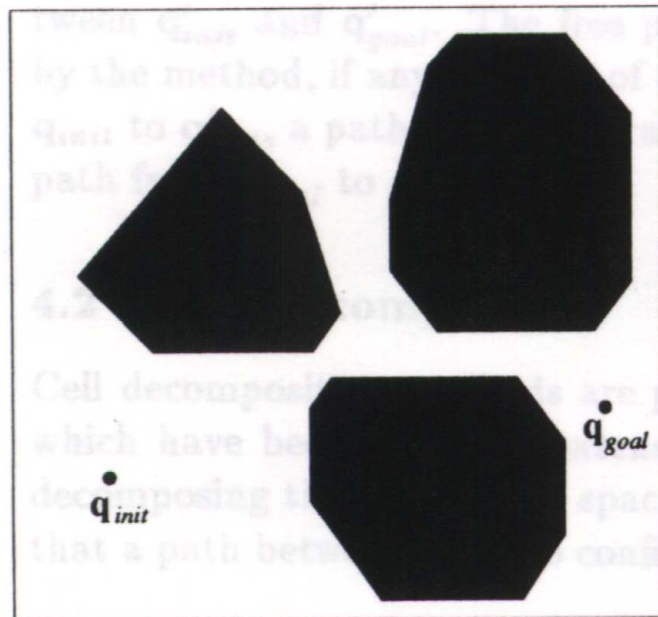


- Different resolution
  - Different roadmap



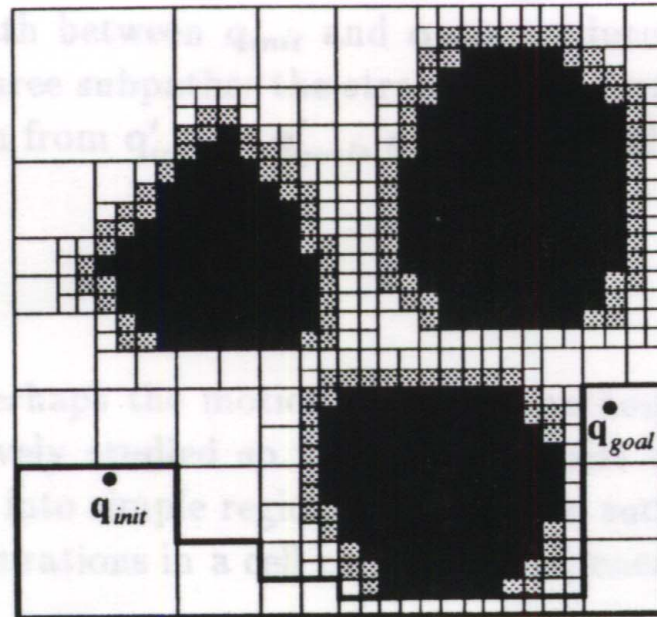
# Approx. Cell Decomposition

- Higher resolution around CBs



$R$

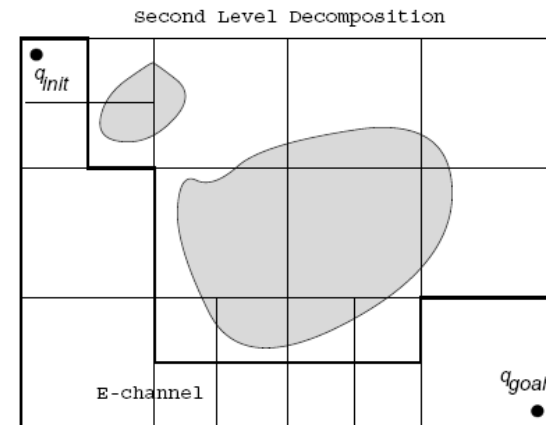
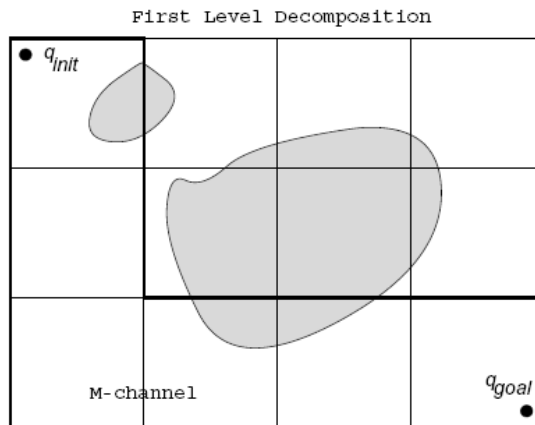
(a)



(b)

# Approx. Cell Decomposition

- Hierarchical approach
  - Find path using empty and mixed cells
  - Further decompose mixed cells into smaller cells



# Approx. Cell Decomposition

- Advantages:
  - simple, uniform decomposition
  - easy implementation
  - adaptive
- Disadvantages:
  - large storage requirement
  - Lose completeness
- **Bottom line 1:** We sacrifice exactness for simplicity and efficiency
- **Bottom line 2:** Approx. cell decomposition methods are practically for lower dimension  $C$ , i.e., dof  $< 5$ , b/c they generate too many cells, i.e.  $(N^d)$  cells in  $d$  dimension

# Potential Field Methods

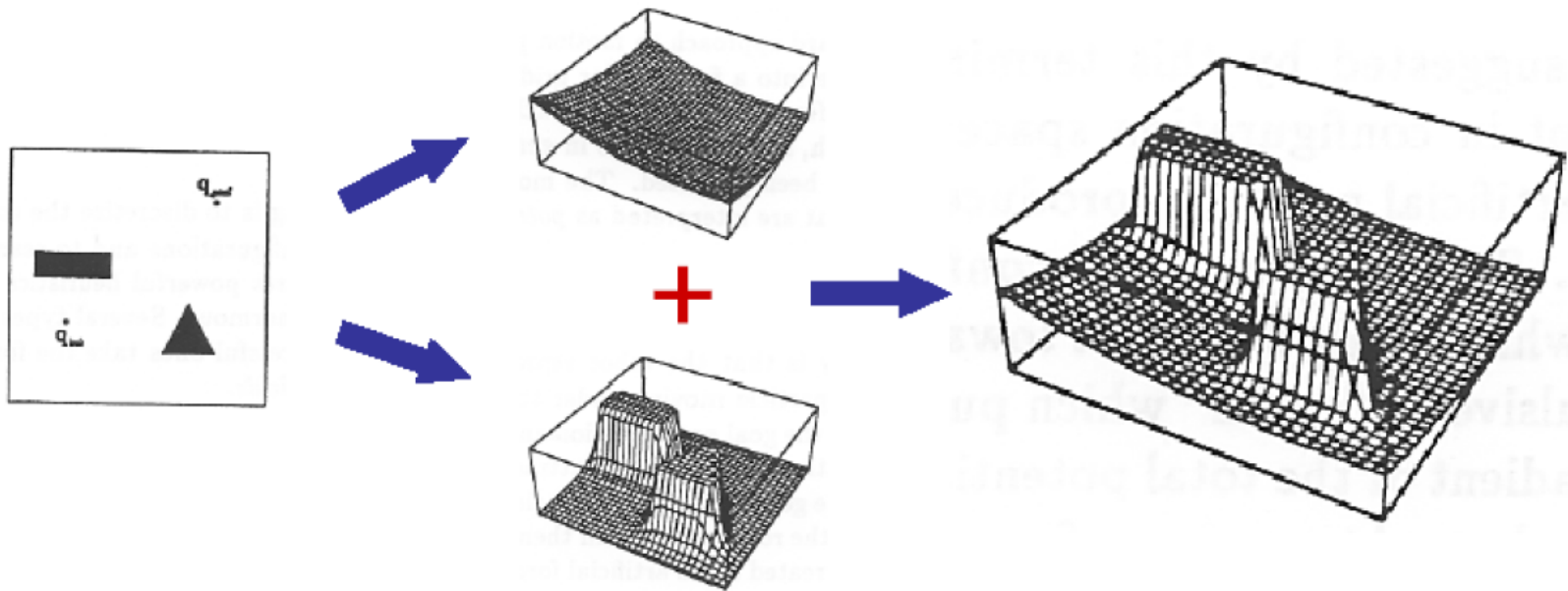
- Approach initially proposed for real-time collision avoidance [Khatib, 86].
  - Hundreds of papers published on it

$$F_{Goal} = -k_p (x - x_{Goal})$$

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



# Potential Field Methods



# Potential Field+Grid Search

- Superimpose a grid over C-space
- Each cell has a potential value
- Search from start to goal on the grid using **best-first search or A\* search**

# Potential Field Methods

- At each step move an increment in the direction that minimizes the energy
  - + Good heuristic for high DOF
- Can get trapped in local minima
  - use some probabilistic motion to escape
- Oscillations can also occur