

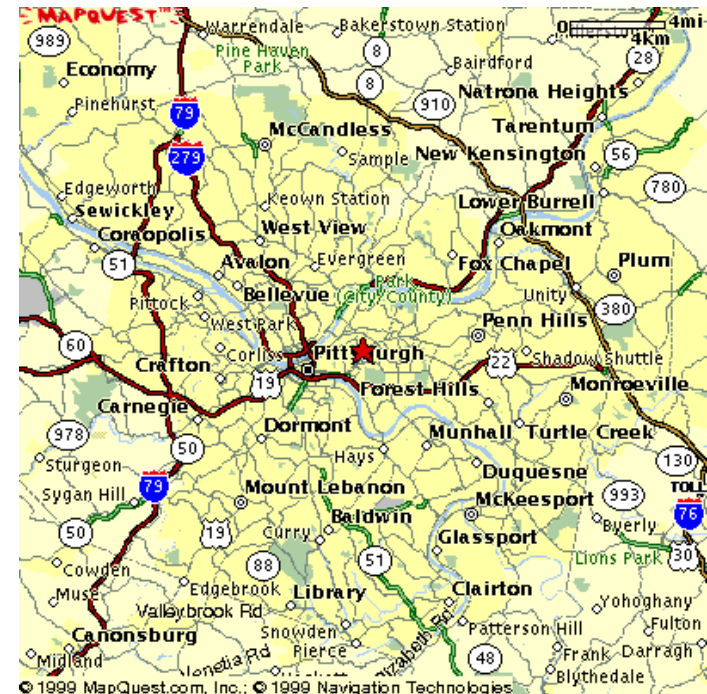
Motion Planning, Part IV

Graph Search Part II

Howie Choset

Map-Based Approaches: Roadmap Theory

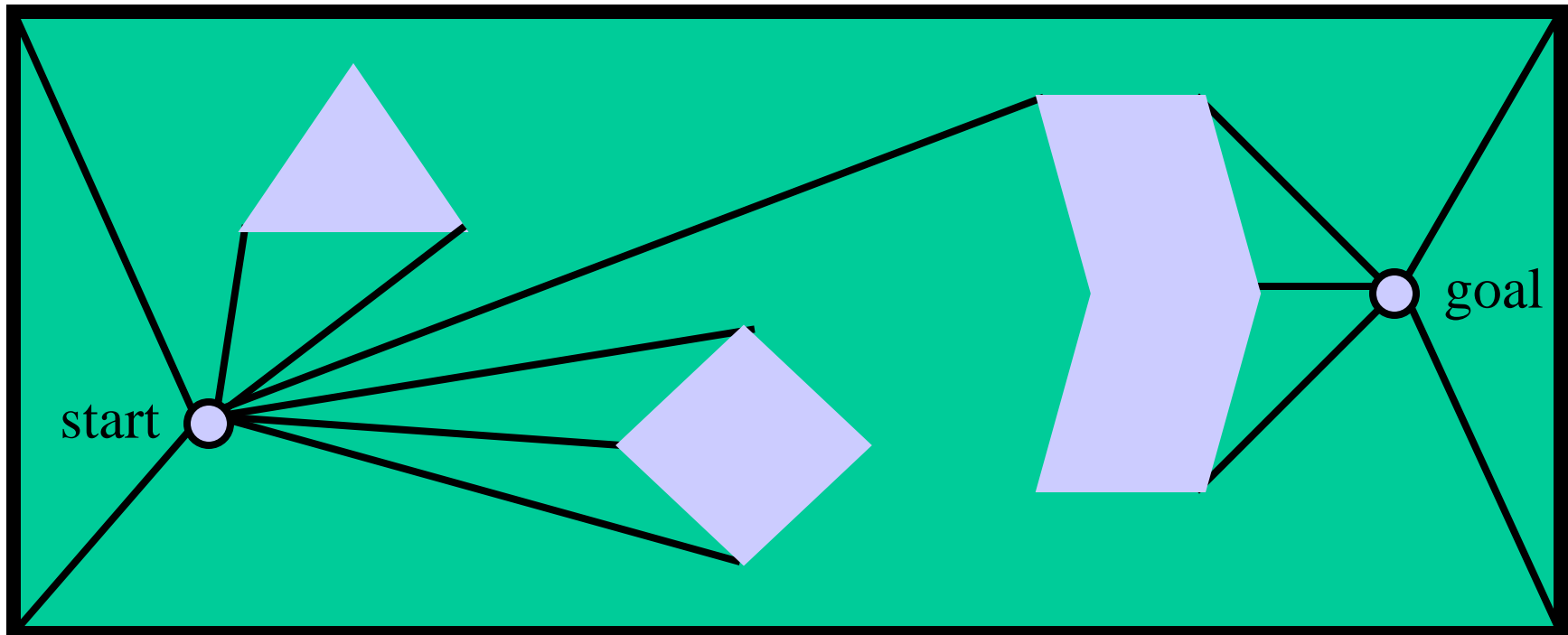
- Properties of a roadmap:
 - Accessibility: there exists a collision-free path from the start to the road map
 - Departability: there exists a collision-free path from the roadmap to the goal.
 - Connectivity: there exists a collision-free path from the start to the goal (on the roadmap).



- a roadmap exists \Leftrightarrow a path exists
- Examples of Roadmaps
 - Generalized Voronoi Graph (GVG)
 - Visibility Graph

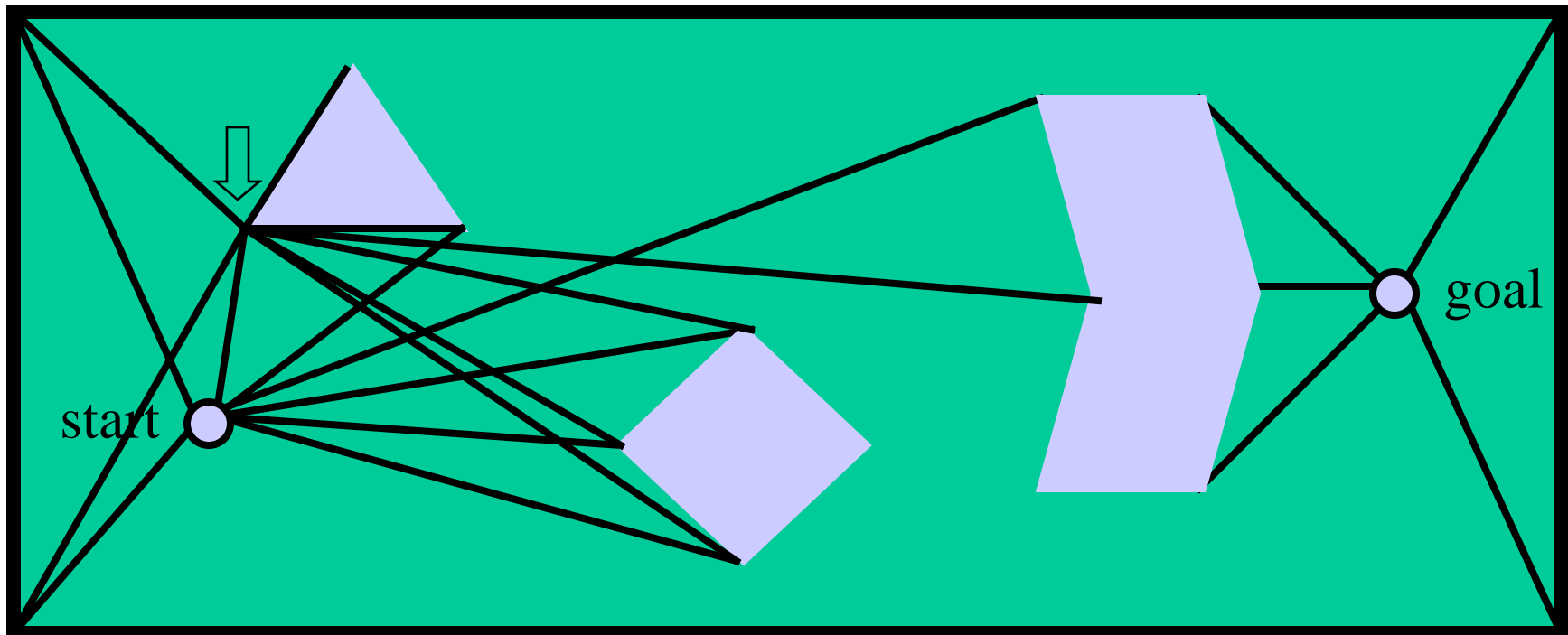
The Visibility Graph in Action (Part 1)

- First, draw lines of sight from the start and goal to all “visible” vertices and corners of the world.



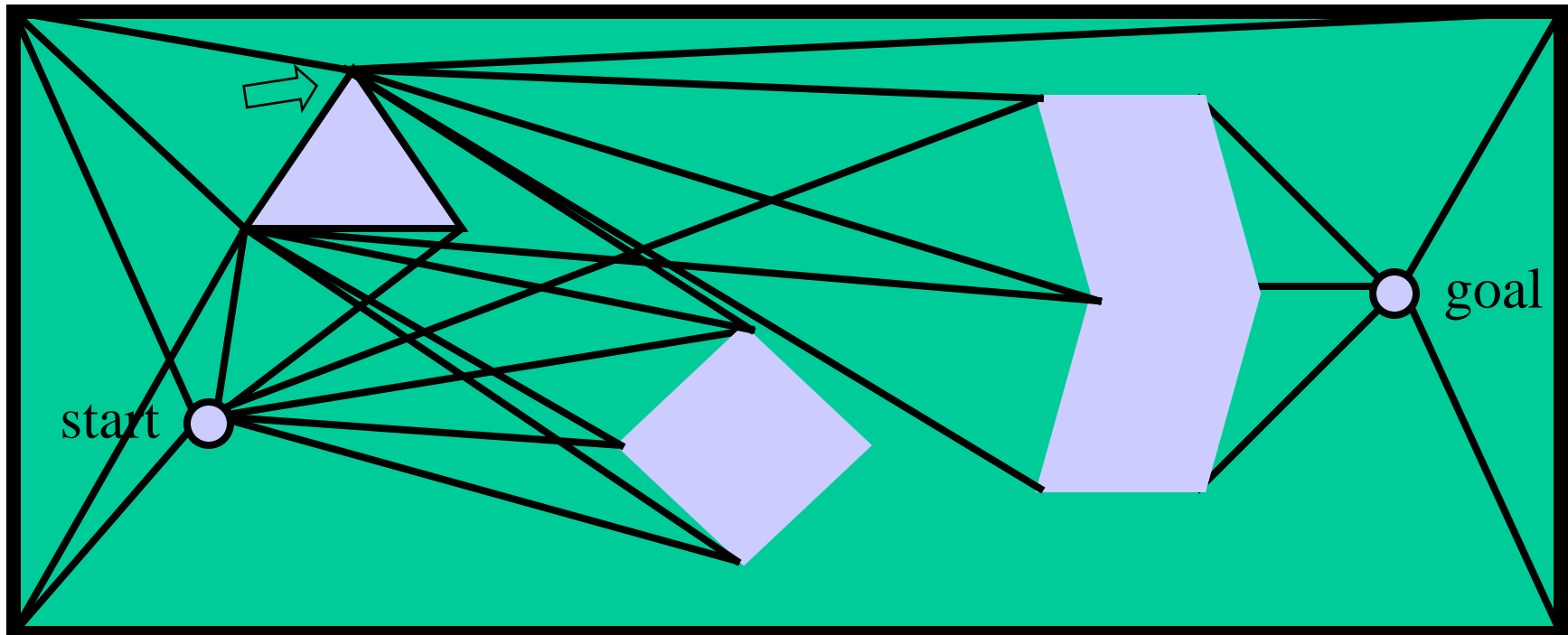
The Visibility Graph in Action (Part 2)

- Second, draw lines of sight from every vertex of every obstacle like before. Remember lines along edges are also lines of sight.



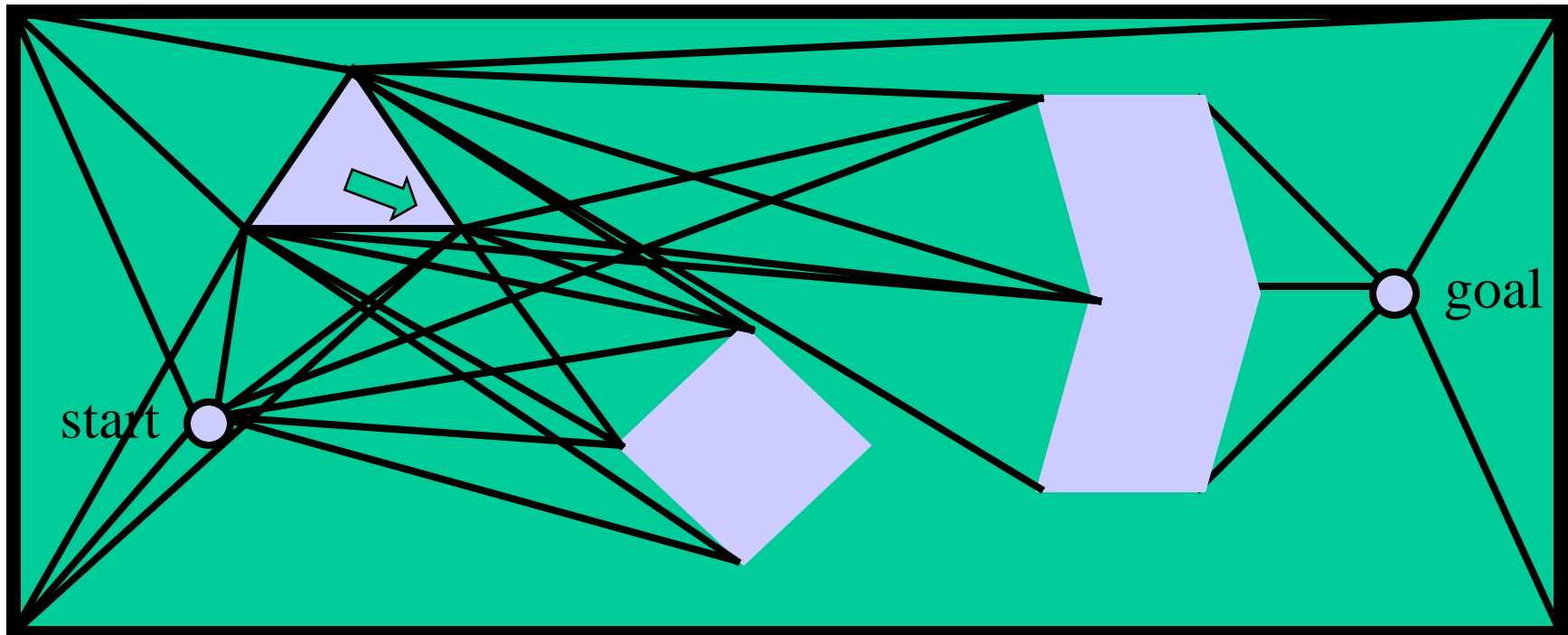
The Visibility Graph in Action (Part 3)

- Second, draw lines of sight from every vertex of every obstacle like before. Remember lines along edges are also lines of sight.



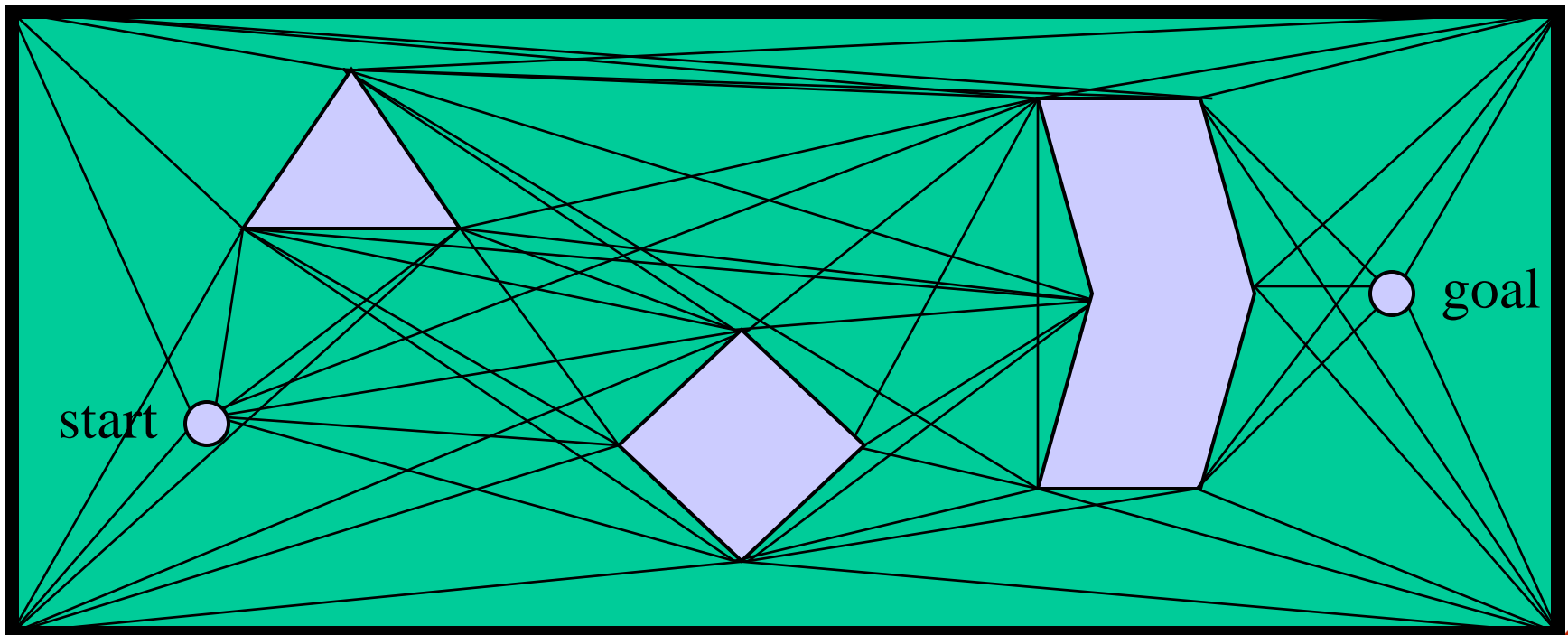
The Visibility Graph in Action (Part 4)

- Second, draw lines of sight from every vertex of every obstacle like before. Remember lines along edges are also lines of sight.



The Visibility Graph (Done)

- Repeat until you're done.



Graph Search

Howie Choset

16-311

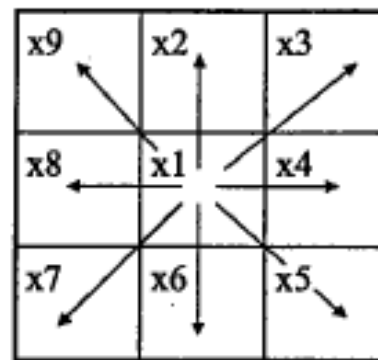
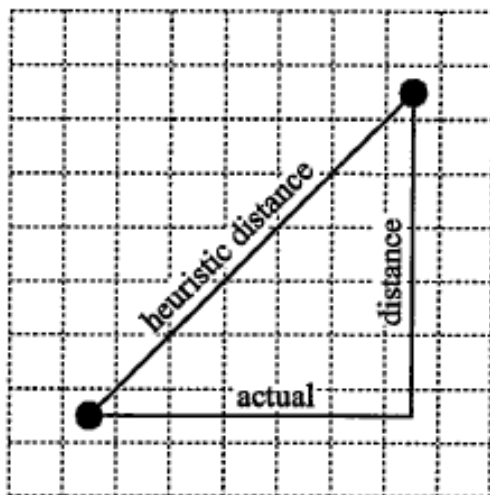
Informed Search: A*

Notation

- $n \rightarrow$ node/state
- $c(n_1, n_2) \rightarrow$ the length of an edge connecting between n_1 and n_2
- $b(n_1) = n_2 \rightarrow$ backpointer of a node n_1 to a node n_2 .

Informed Search: A*

- Evaluation function, $f(n) = g(n) + h(n)$
- Operating cost function, $g(n)$
 - Actual operating cost having been already traversed
- Heuristic function, $h(n)$
 - Information used to find the promising node to traverse
 - Admissible \rightarrow never overestimate the actual path cost



$$c(x1, x2) = 1$$

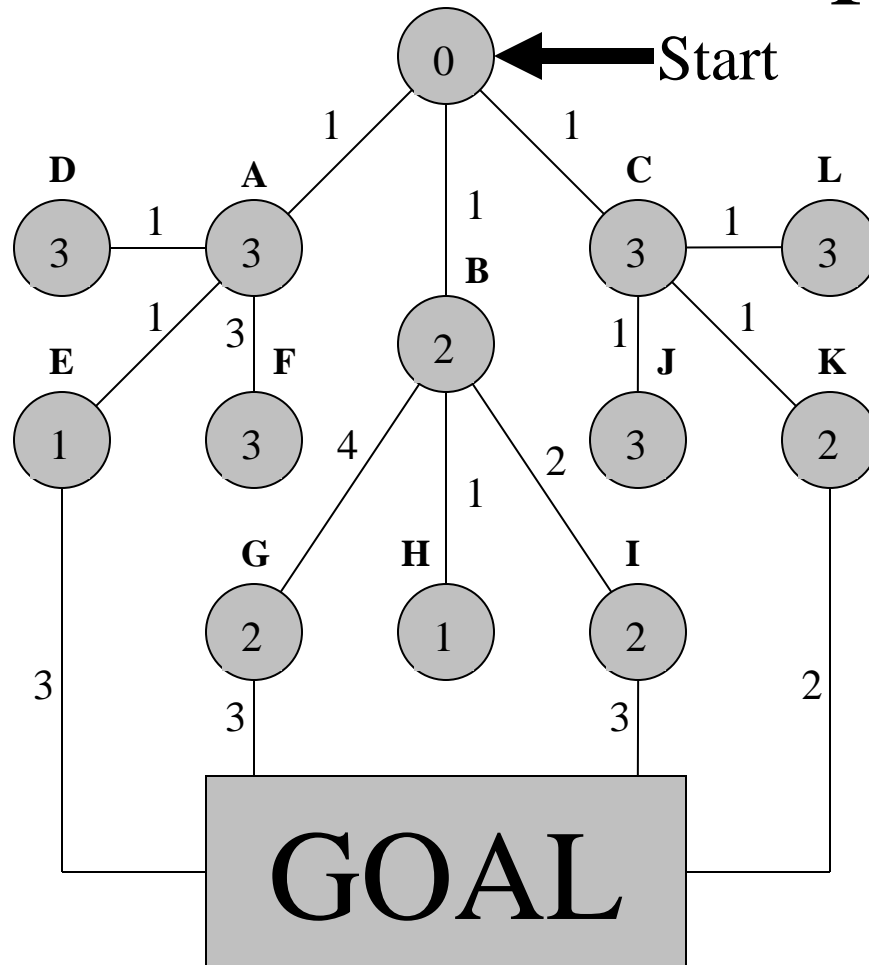
$$c(x1, x9) = 1.4$$

$$c(x1, x8) = 10000, \text{ if } x8 \text{ is in obstacle, } x1 \text{ is a free cell}$$

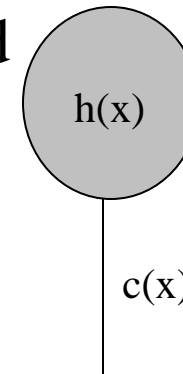
$$c(x1, x9) = 10000.4, \text{ if } x9 \text{ is in obstacle, } x1 \text{ is a free cell}$$

Cost on a grid

Example (1/5)



Legend



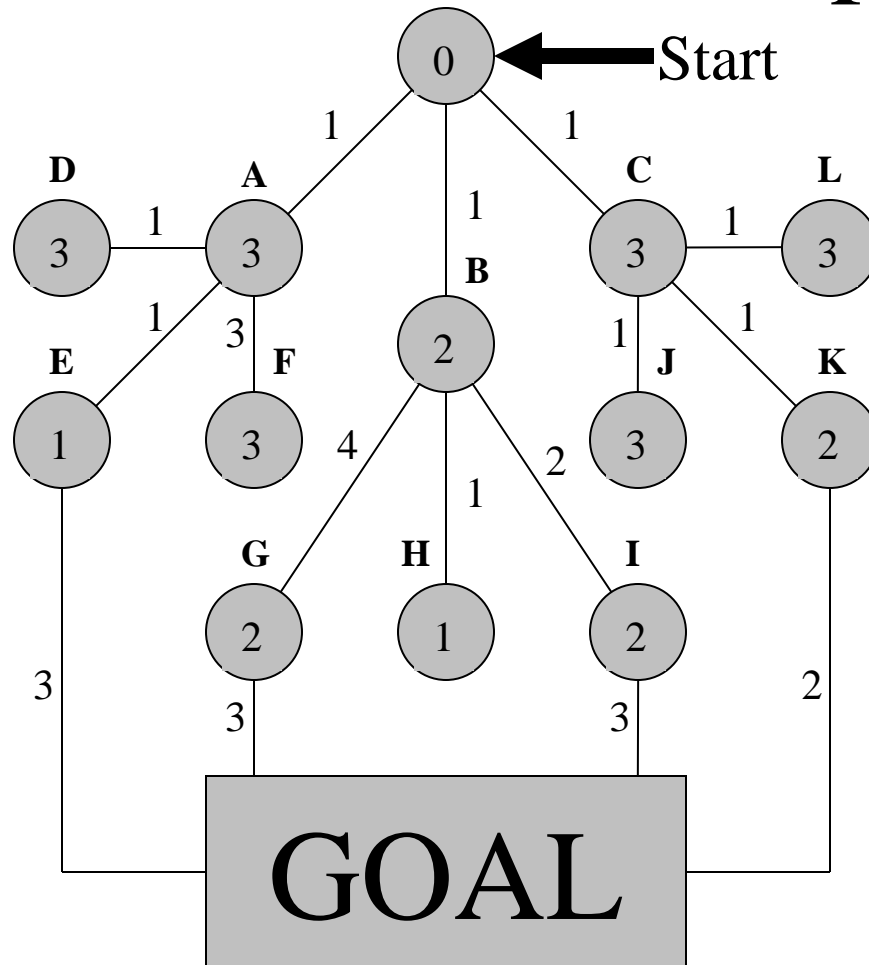
$$\text{Priority} = g(x) + h(x)$$

Note:

$g(x)$ = sum of all previous arc costs, $c(x)$,
from start to x

Example: $c(H) = 2$

Example (2/5)



First expand the start node

B (3)
A (4)
C (4)

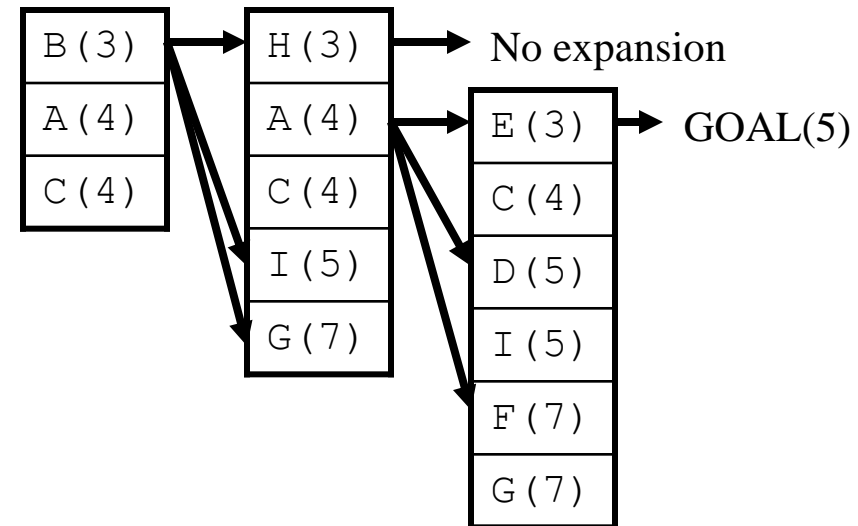
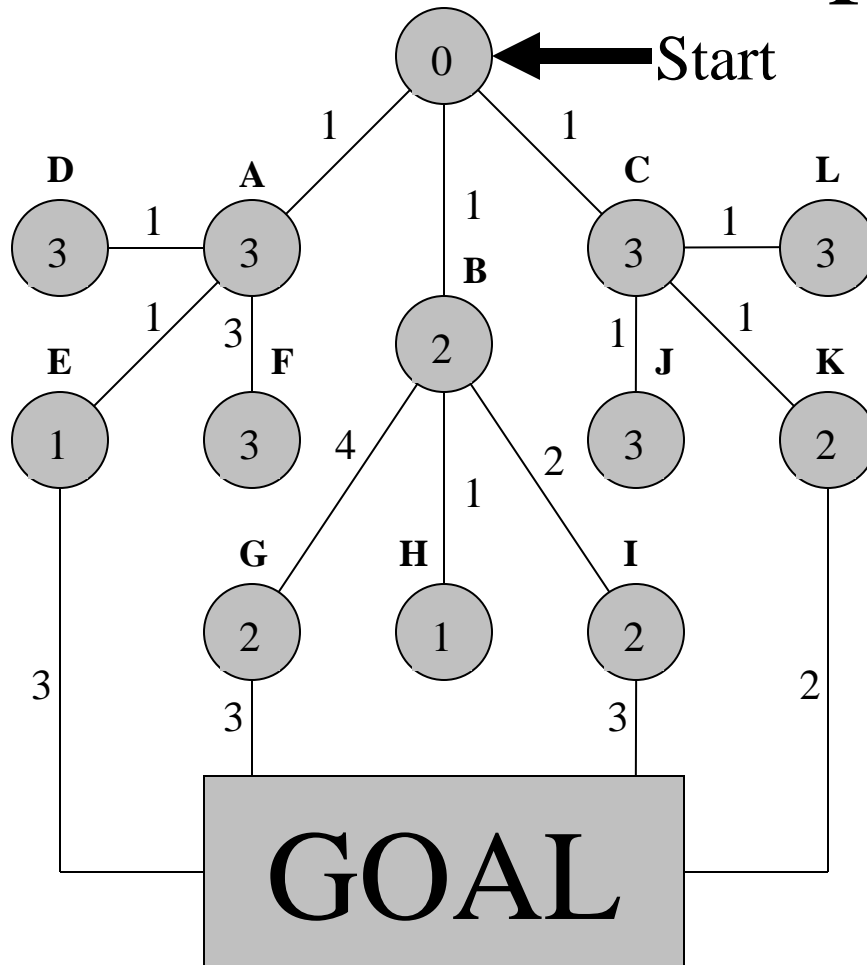
If goal not found,
expand the first node
in the priority queue
(in this case, B)

H (3)
A (4)
C (4)
I (5)
G (7)

Insert the newly expanded
nodes into the priority queue
and continue until the goal
is found, or the priority queue
is empty (in which case no path

Note: for each expanded node,
you also need a pointer to its respective
parent. For example, nodes A, B and C
point to Start

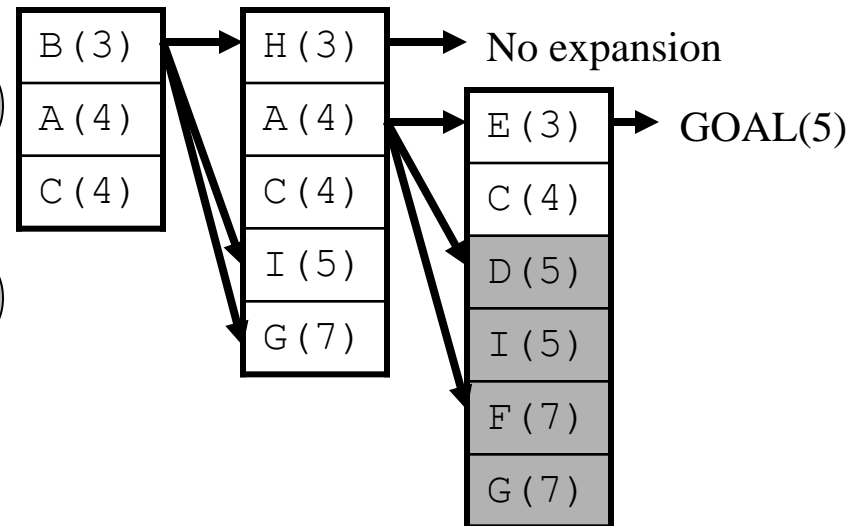
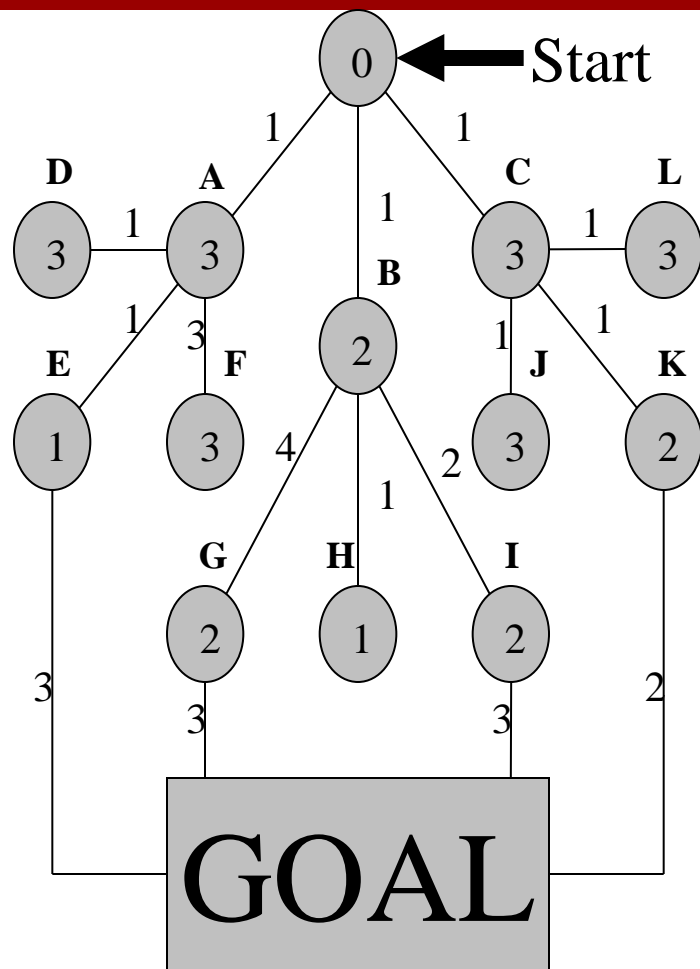
Example (3/5)



We've found a path to the goal:
 Start \Rightarrow A \Rightarrow E \Rightarrow Goal
(from the pointers)

Are we done?

Example (4/5)

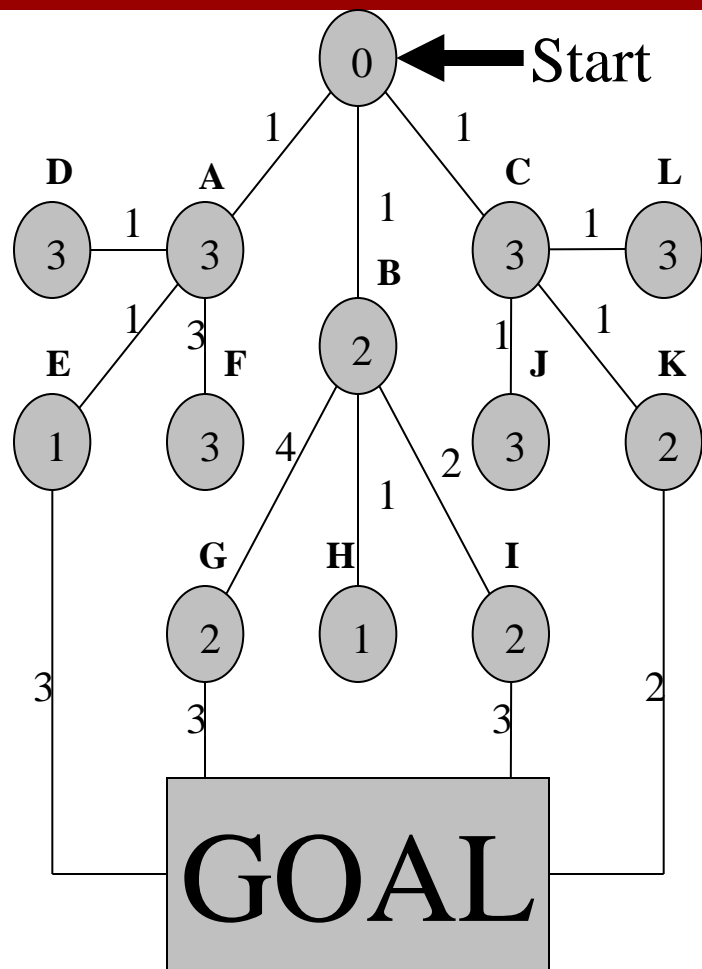


There might be a shorter path, but assuming non-negative arc costs, nodes with a lower priority than the goal cannot yield a better path.

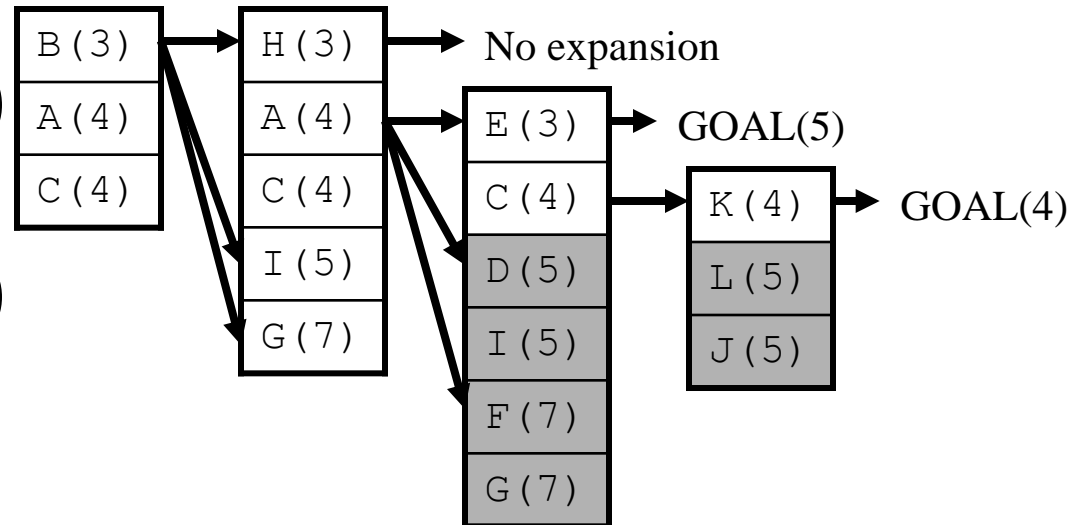
In this example, nodes with a priority greater than or equal to 5 can be pruned.

Why don't we expand nodes with an equivalent priority (why not expand nodes D and I?)

Example (5/5)



If the priority queue still wasn't empty, we would continue expanding while throwing away nodes with priority lower than 4.
(remember, lower numbers = higher priority)



We can continue to throw away nodes with priority levels lower than the lowest goal found.

As we can see from this example, there was a shorter path through node K. To find the path, simply follow the back pointers.

Therefore the path would be:
Start => C => K => Goal

Monotonic

- never overestimates the cost of getting from a node to its neighbor.

- for all paths x, y where y is a successor of x , i.e.,

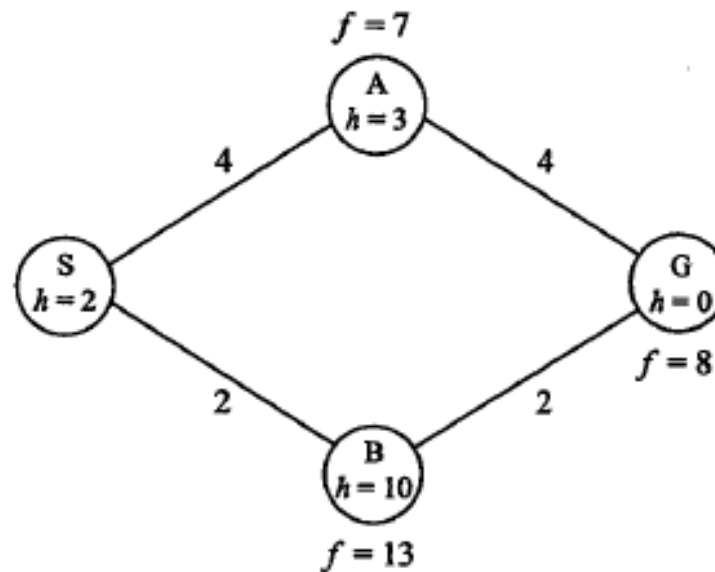
$$h(x) \leq g(y) - g(x) + h(y)$$

- $h(A) = 3 \quad g(A) = 1 \quad h(E) = 1 \quad g(E) = 2$

$$h(A) = 3 \not\leq g(E) - g(A) + h(E) = 2 - 1 + 1 = 2$$

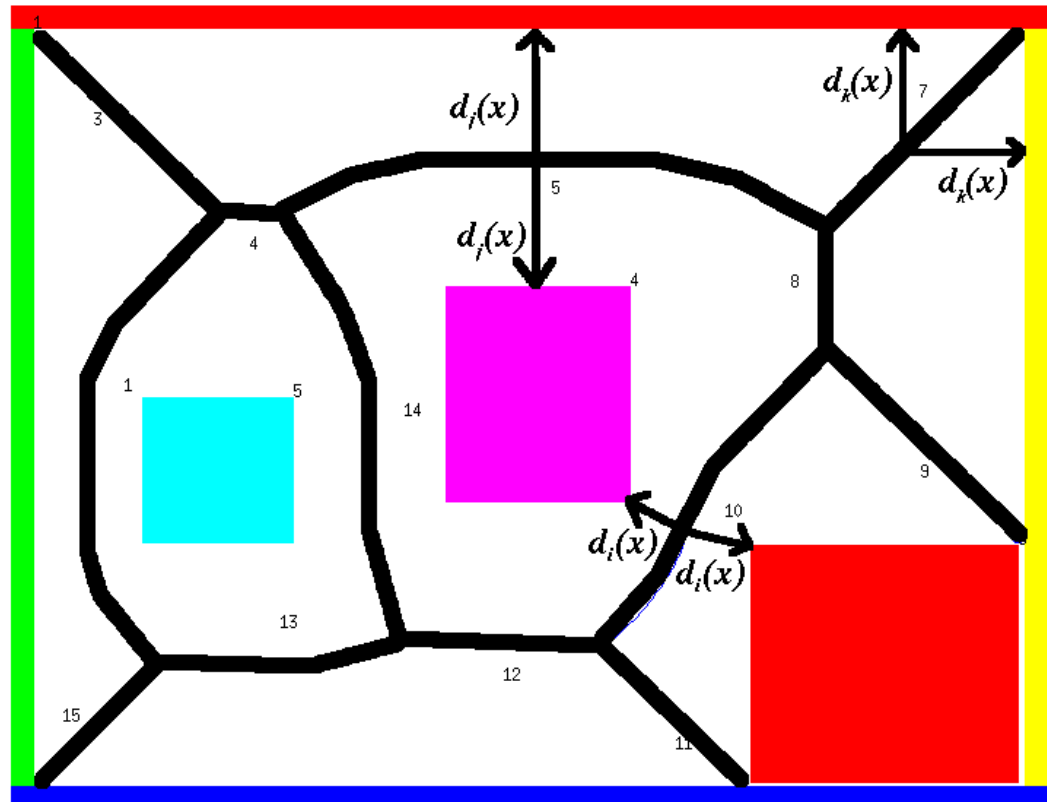
Non-opportunistic

1. Put S on priority Q and expand it
2. Expand A because its priority value is 7
3. The goal is reached with priority value 8
4. This is less than B's priority value which is 13



Roadmap: GVG

- A GVG is formed by paths equidistant from the two closest objects
- *Remember “spokes”, start and goal*

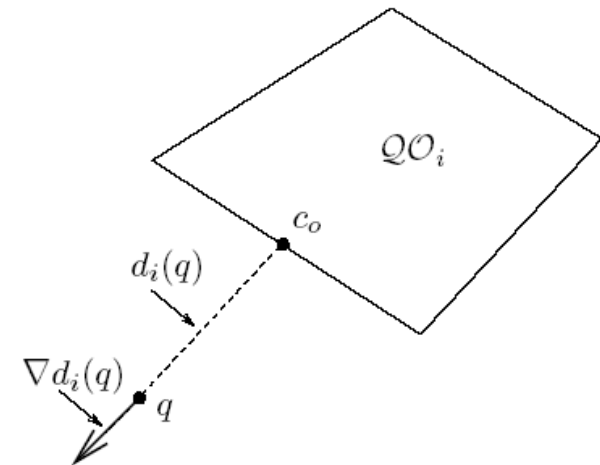


- This generates a very safe roadmap which avoids obstacles as much as possible

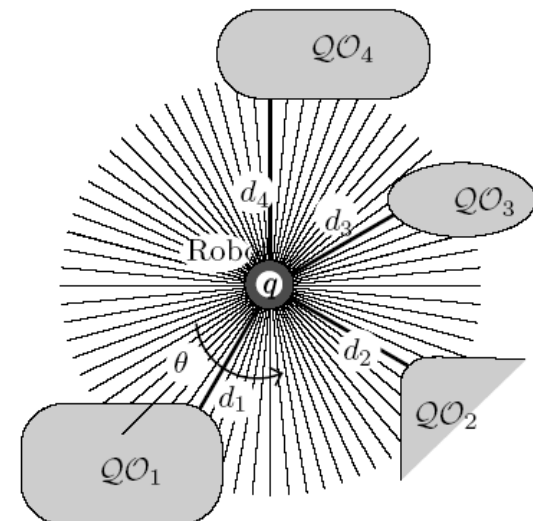
Distance to Obstacle(s)

$$d_i(q) = \min_{c \in \mathcal{QO}_i} d(q, c).$$

$$\nabla d_i(q) = \frac{q - c}{d(q, c)}$$



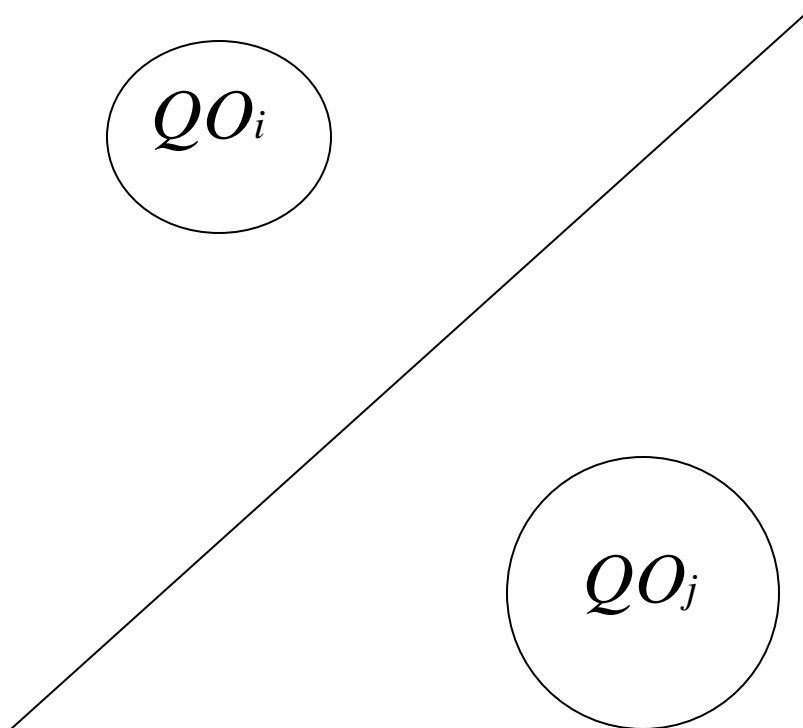
$$D(q) = \min d_i(q)$$



Two-Equidistant

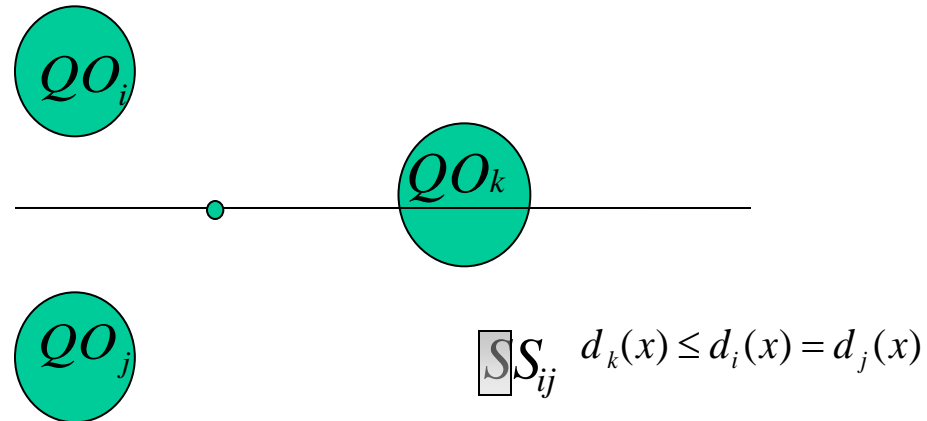
- *Two-equidistant surface*

$$S_{ij} = \{x \in Q_{\text{free}} : d_i(x) - d_j(x) = 0\}$$



More Rigorous Definition

Going through obstacles

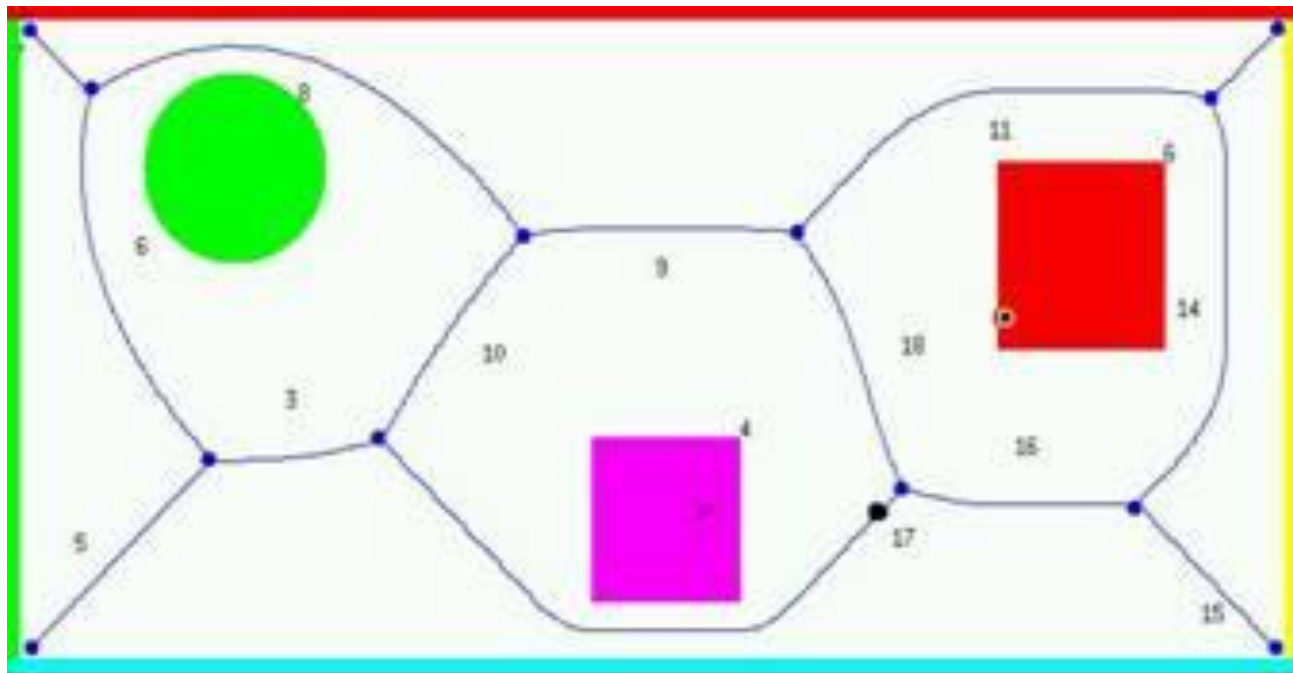


Two-equidistant face

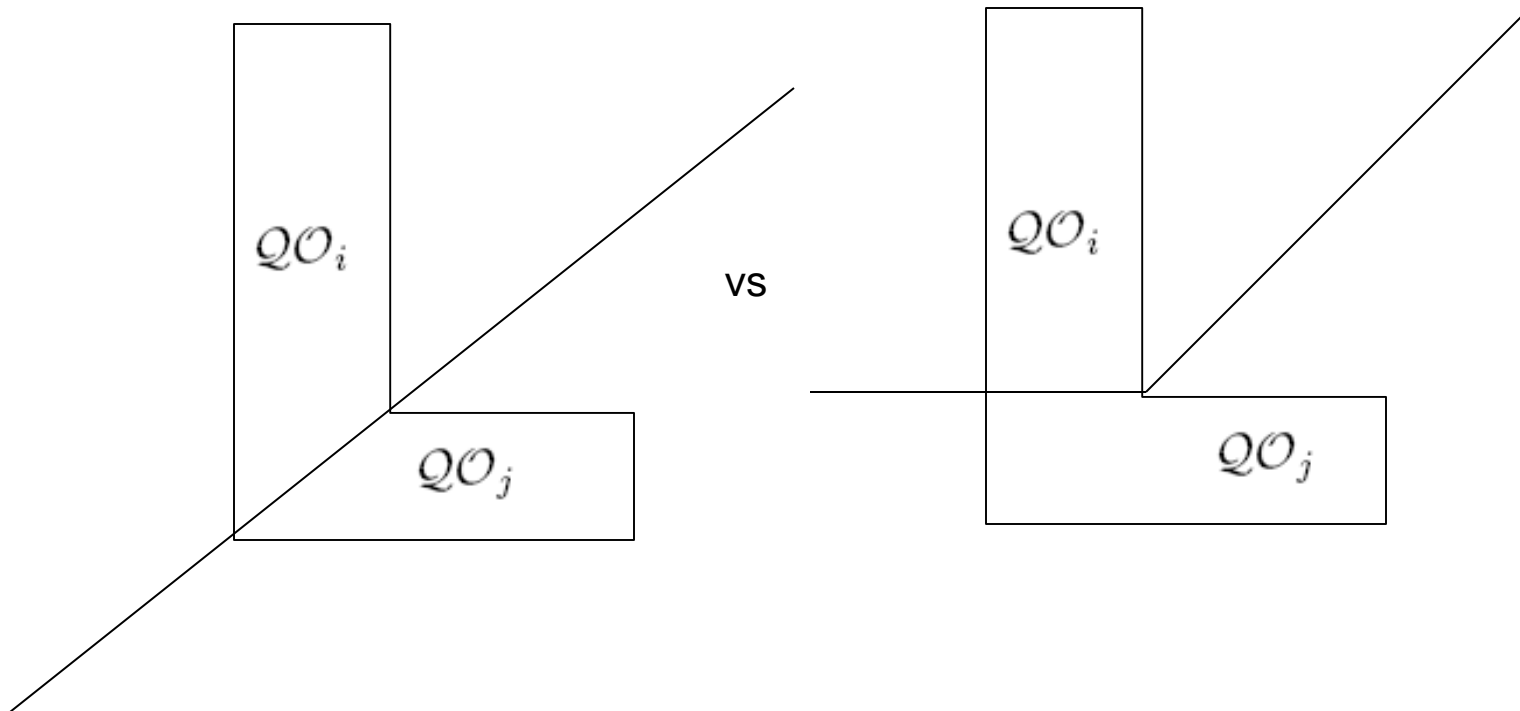
$$F_{ij} = \{x \in \mathbb{S}S_{ij} : d_i(x) = d_j(x) \leq d_h(x), \forall h \neq i, j\}$$

General Voronoi Diagram

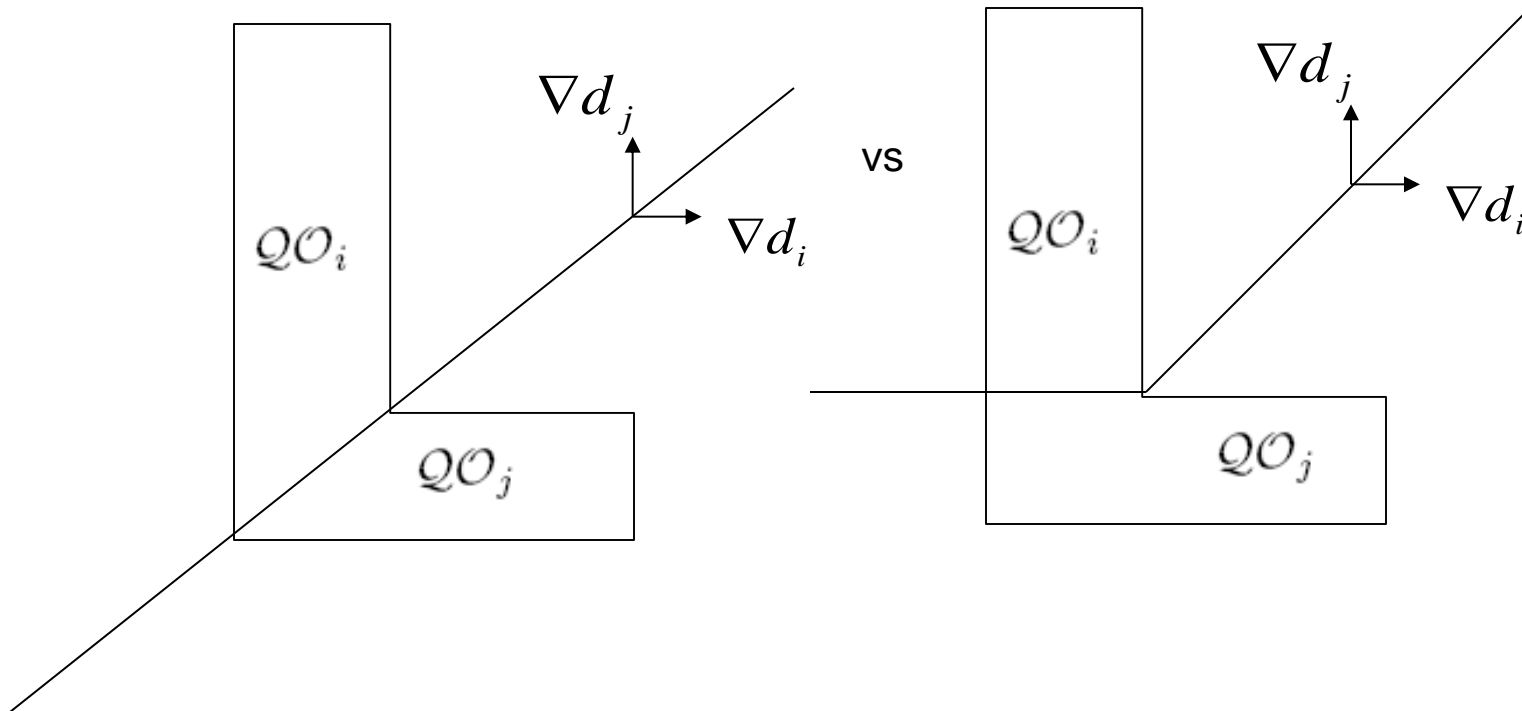
$$\text{GVD} = \bigcup_{i=1}^{n-1} \bigcup_{j=i+1}^n F_{ij}$$



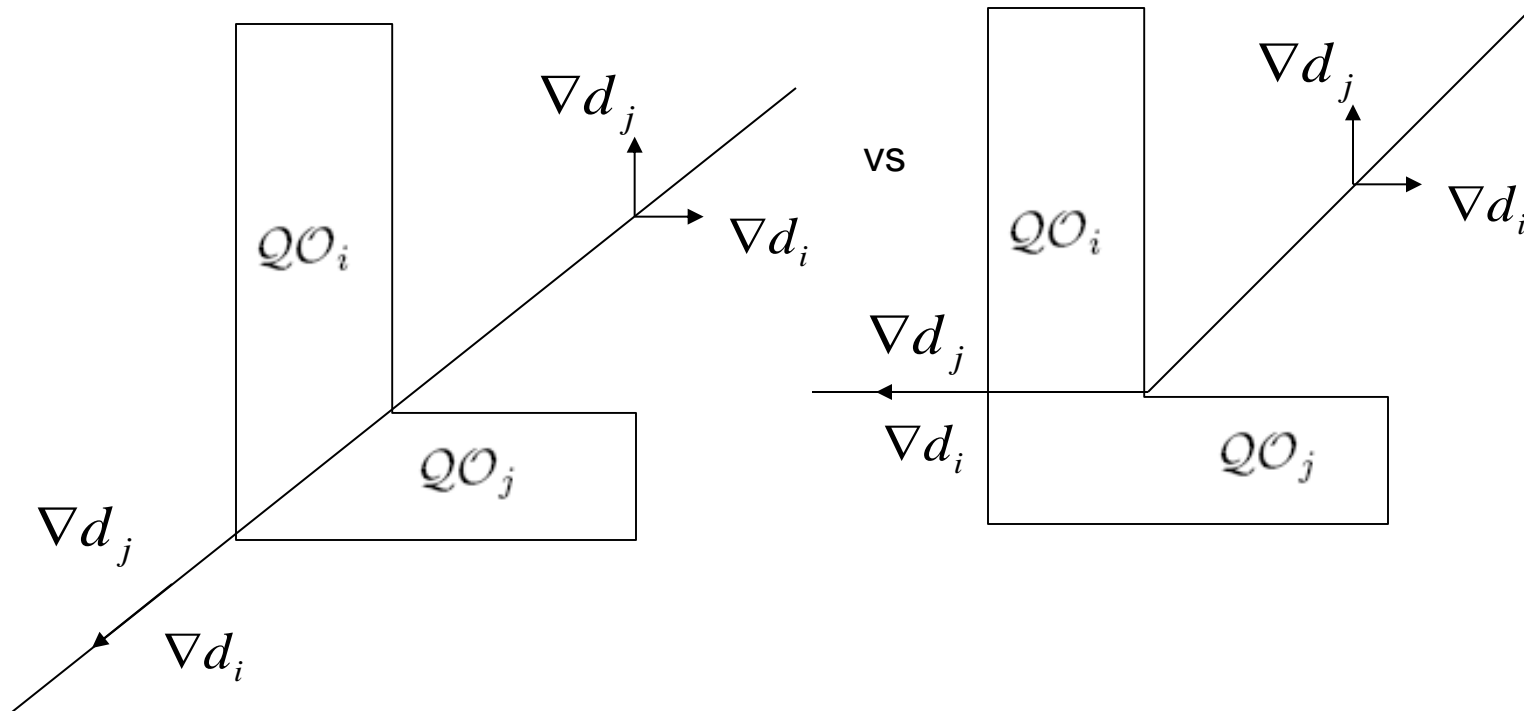
What about concave obstacles?



What about concave obstacles?



What about concave obstacles?



Two-Equidistant

- *Two-equidistant surface*

$$S_{ij} = \{x \in Q_{\text{free}} : d_i(x) - d_j(x) = 0\}$$

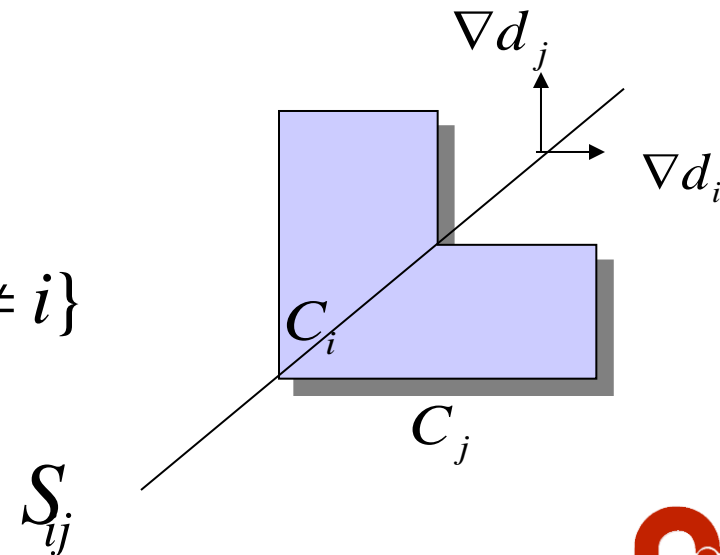
Two-equidistant surjective surface

$$SS_{ij} = \{x \in S_{ij} : \nabla d_i(x) \neq \nabla d_j(x)\}$$

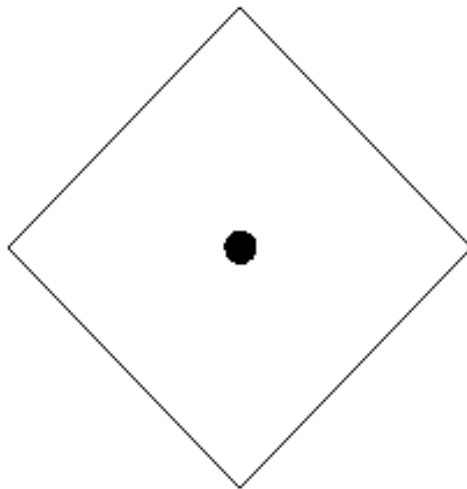
Two-equidistant Face

$$F_{ij} = \{x \in SS_{ij} : d_i(x) \leq d_h(x), \forall h \neq i\}$$

$$\text{GVD} = \bigcup_{i=1}^{n-1} \bigcup_{j=i+1}^n F_{ij}$$

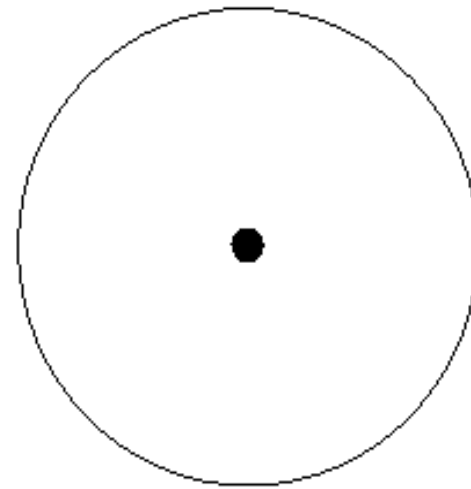


Voronoi Diagram: Metrics



$$\{(x,y) : |x| + |y| = \text{const}\}$$

L1

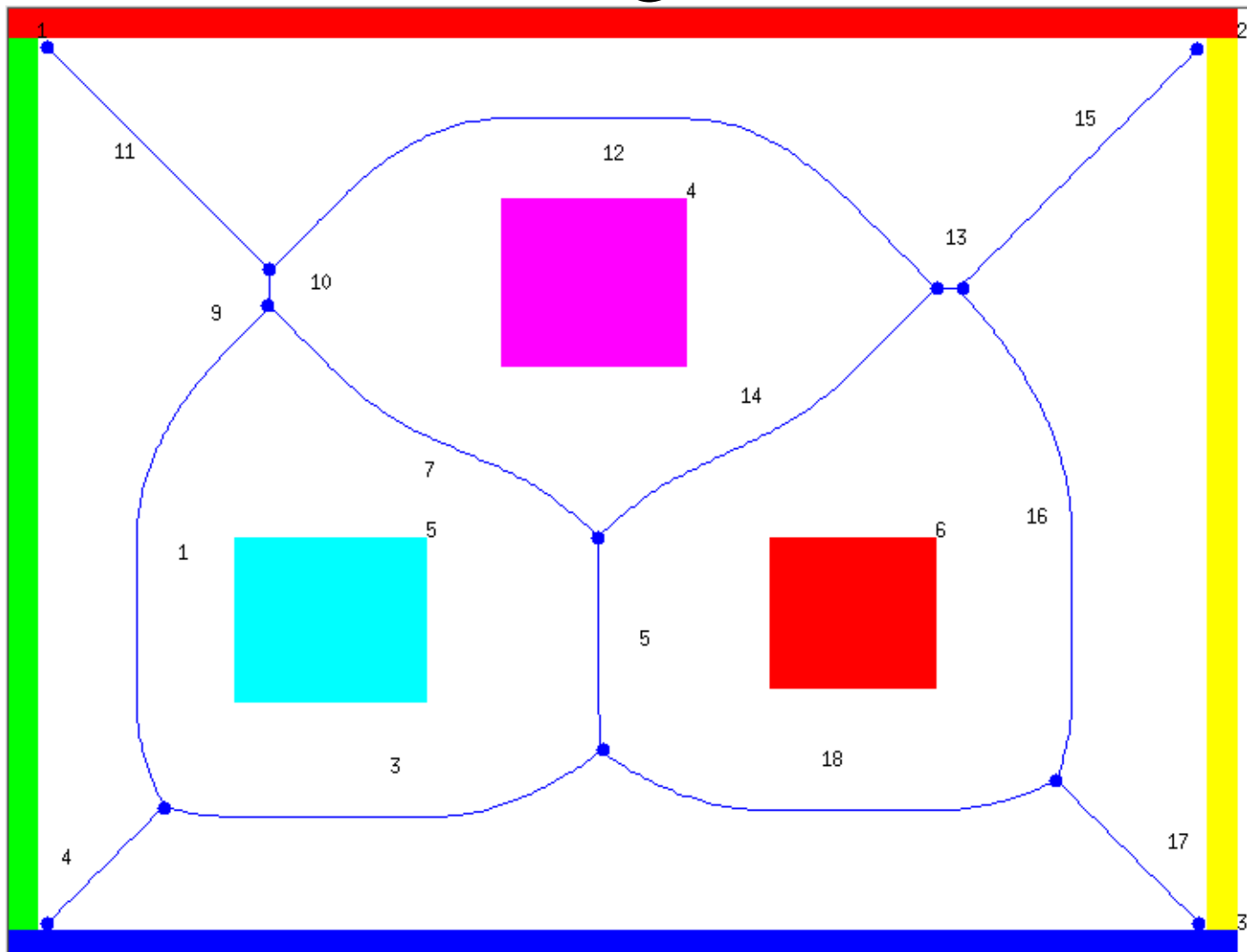


$$\{(x,y) : x^2 + y^2 = \text{const}\}$$

L2

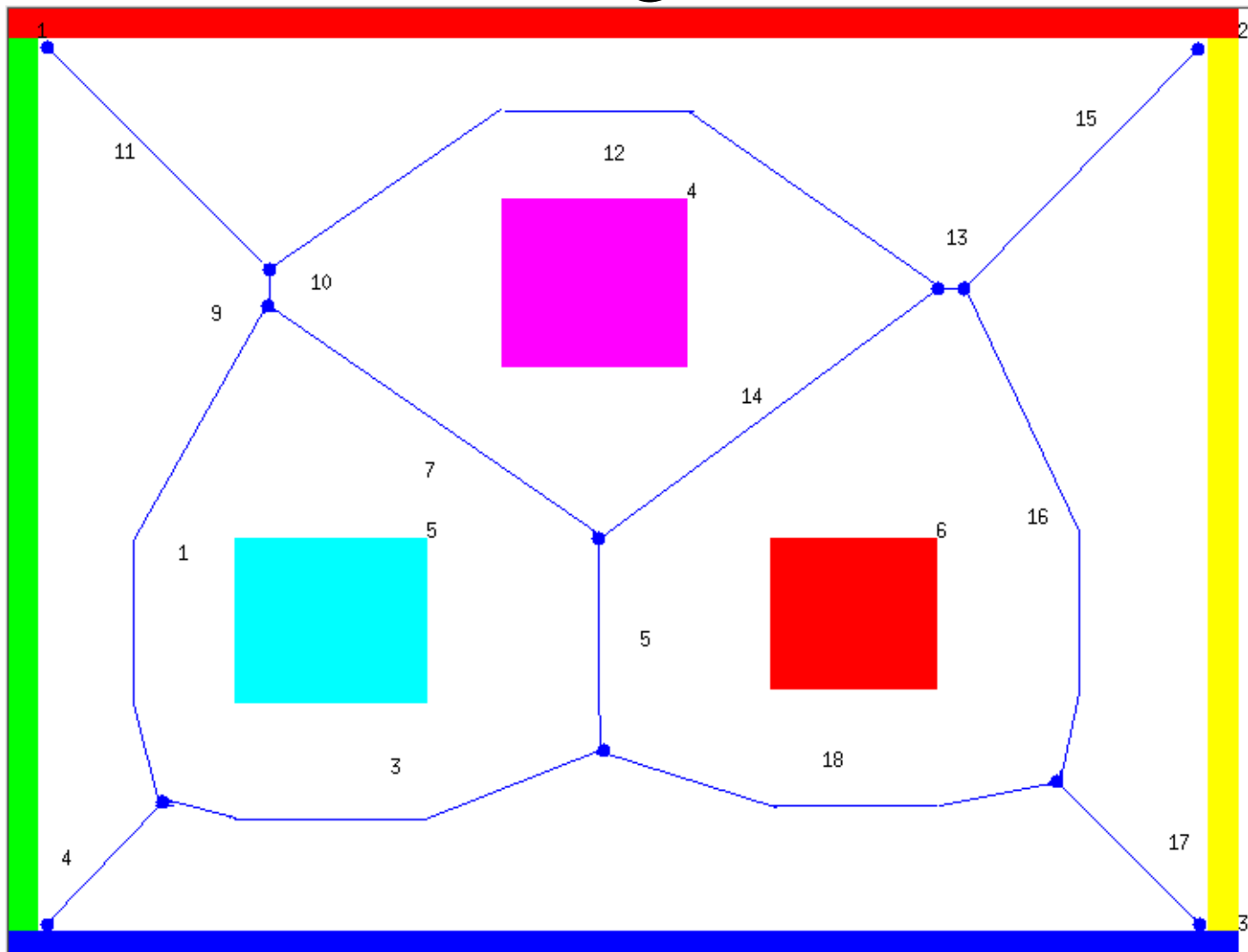
Voronoi Diagram (L2)

Note the
curved
edges



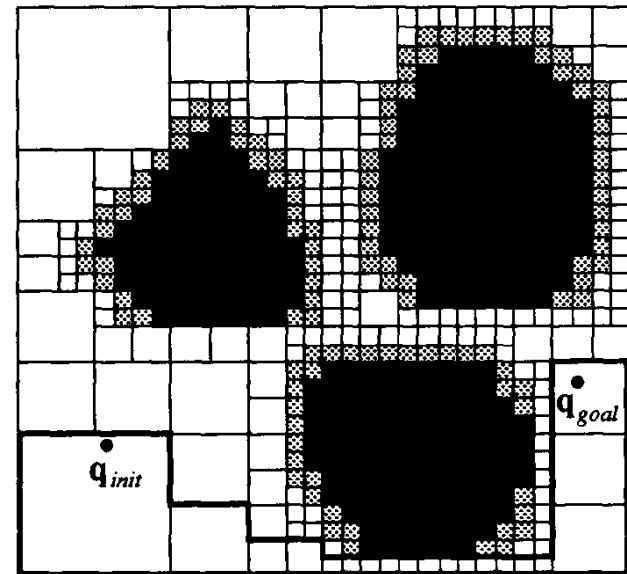
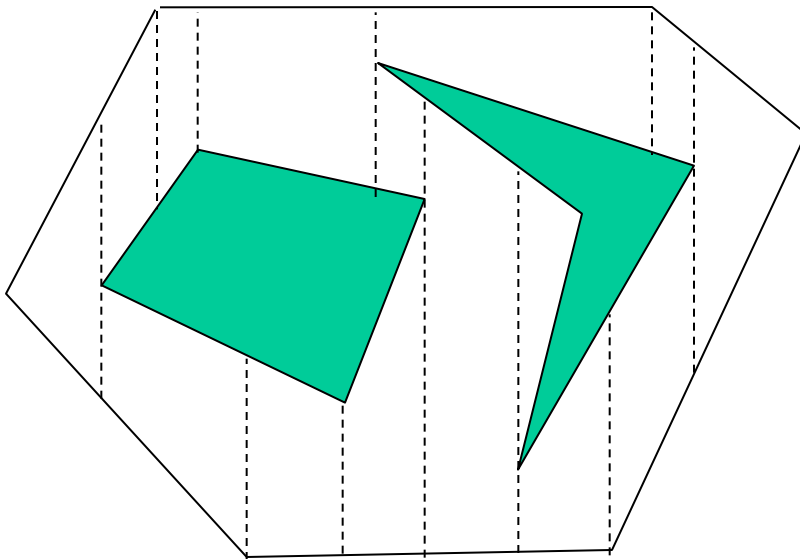
Voronoi Diagram (L1)

Note the
lack of
curved
edges



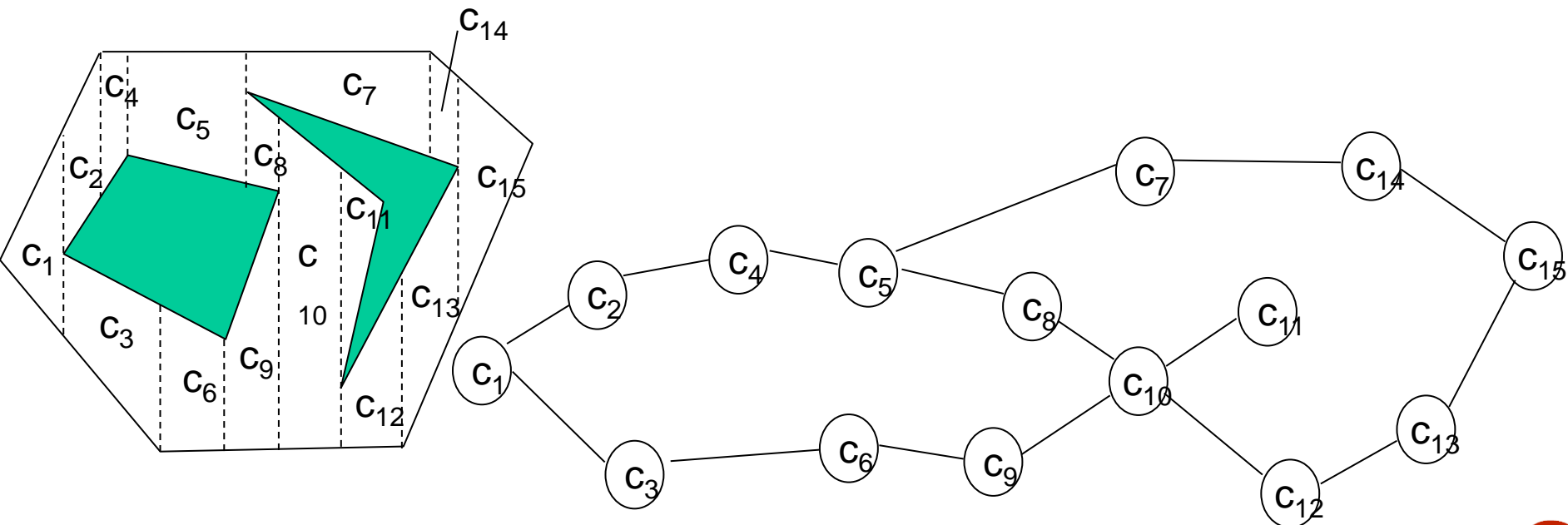
Exact Cell vs. Approximate Cell

- Cell: simple region



Adjacency Graph

- Node correspond to a cell
- Edge connects nodes of adjacent cells
- Two cells are *adjacent* if they share a common boundary



Set Notation

Some set notation

- Interior of A ($\text{int}(A)$) is the largest open subset of A
- Closure of A ($\text{cl}(A)$) is the smallest closed set that contains A
- Complement of A (\bar{A}) is everything not in A .
- Boundary of A (∂A) is the closure of A take away its interior.

Examples

Examples

- $\text{int}[0, 1] = (0, 1), \text{int}(0, 1) = (0, 1)$
- $\text{cl}[0, 1] = [0, 1], \text{cl}(0, 1) = [0, 1]$
- $\bar{[0, 1]} = (-\infty, 0) \cup (1, \infty)$
- $\partial[0, 1] = \partial(0, 1) = \{0, 1\}$

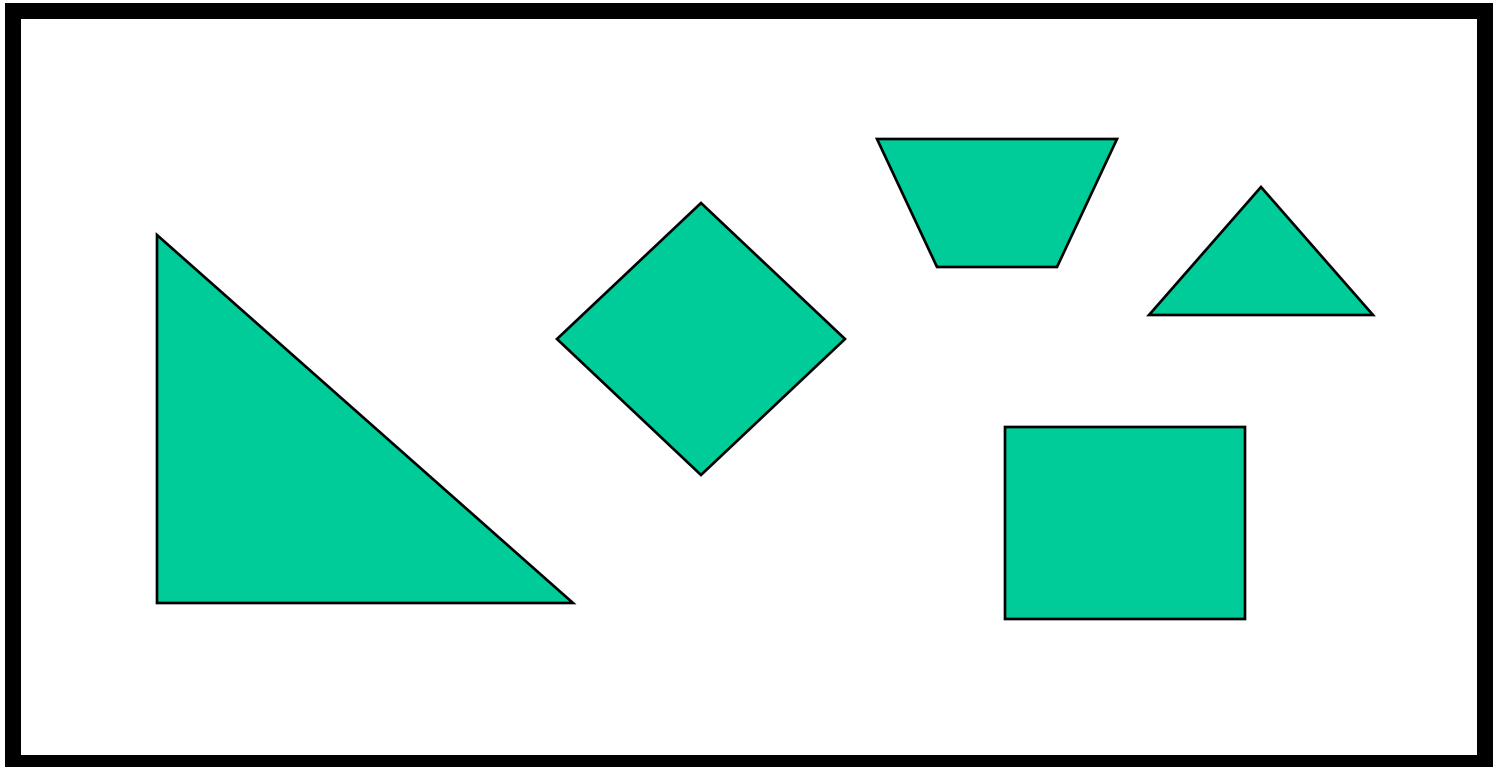
Definition

Exact Cellular Decomposition (as opposed to approximate)

- ν_i is a cell
- $\text{int}(\nu_i) \cap \text{int}(\nu_j) = \emptyset$ if and only if $i \neq j$
- $Fs \cap (\text{cl}(\nu_i) \cap \text{cl}(\nu_j)) \neq \emptyset$ if ν_i and ν_j are adjacent cells
- $Fs = \cup_i (\nu_i)$

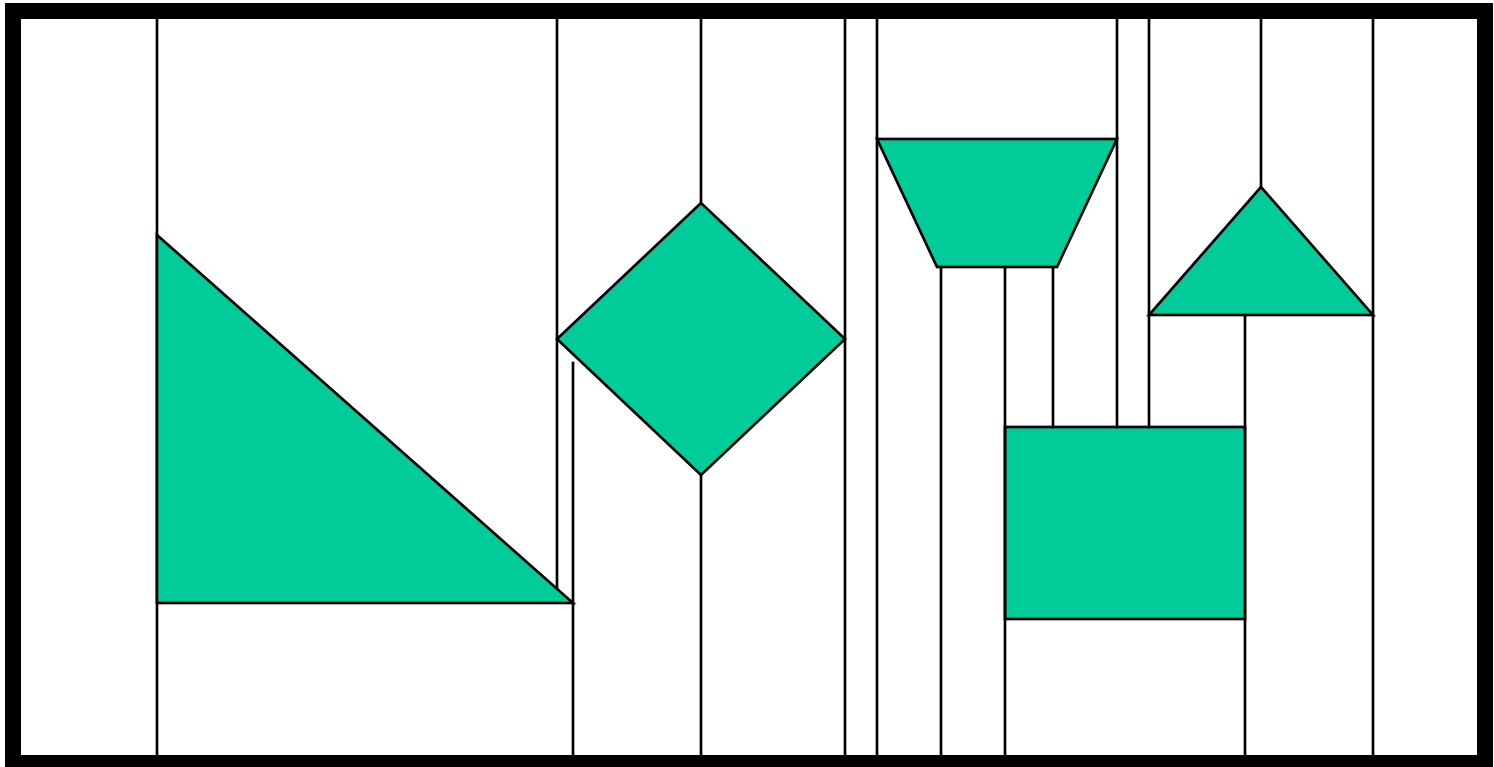
Cell Decompositions: Trapezoidal Decomposition

- A way to divide the world into smaller regions
- Assume a polygonal world



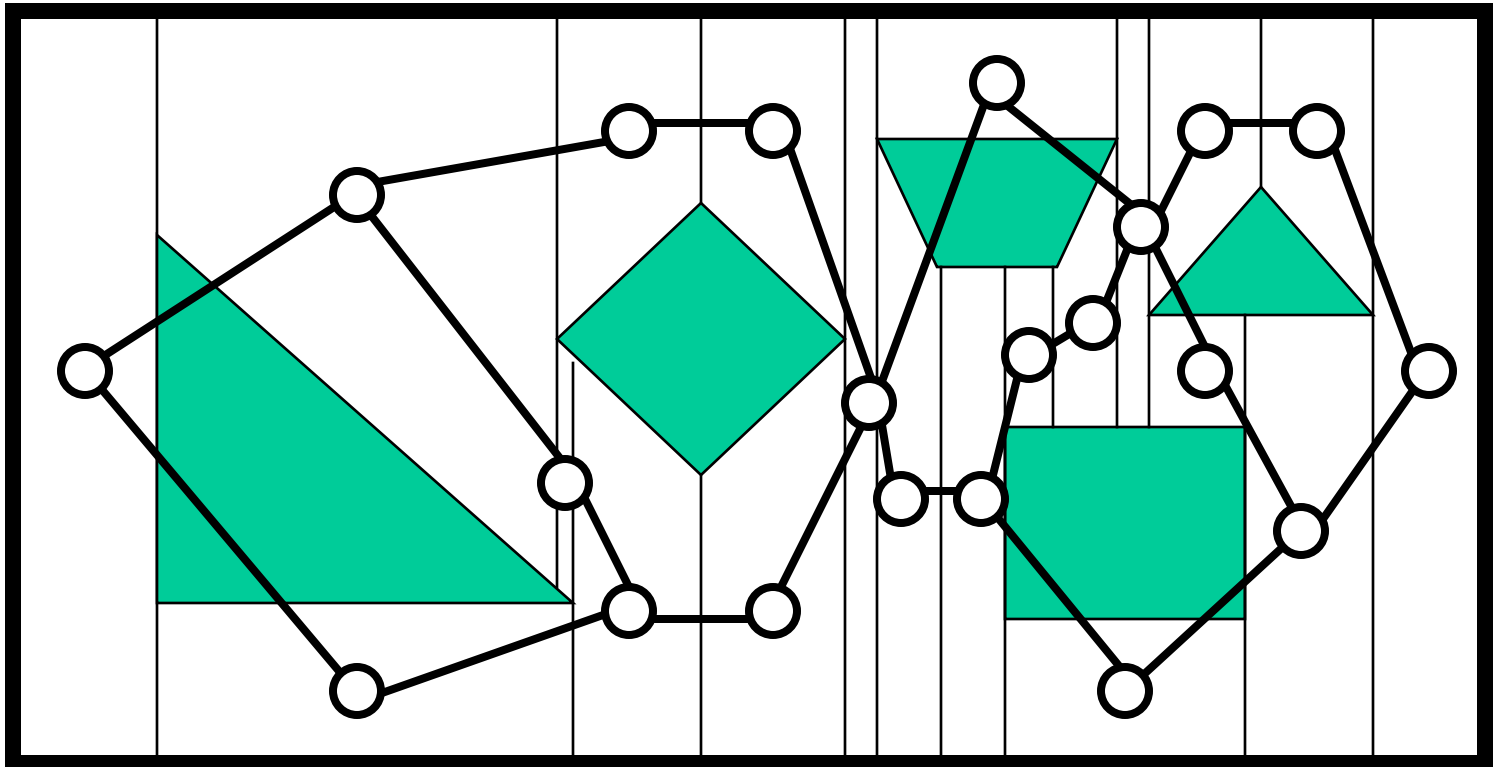
Cell Decompositions: Trapezoidal Decomposition

- Simply draw a vertical line from each vertex until you hit an obstacle. This reduces the world to a union of trapezoid-shaped cells



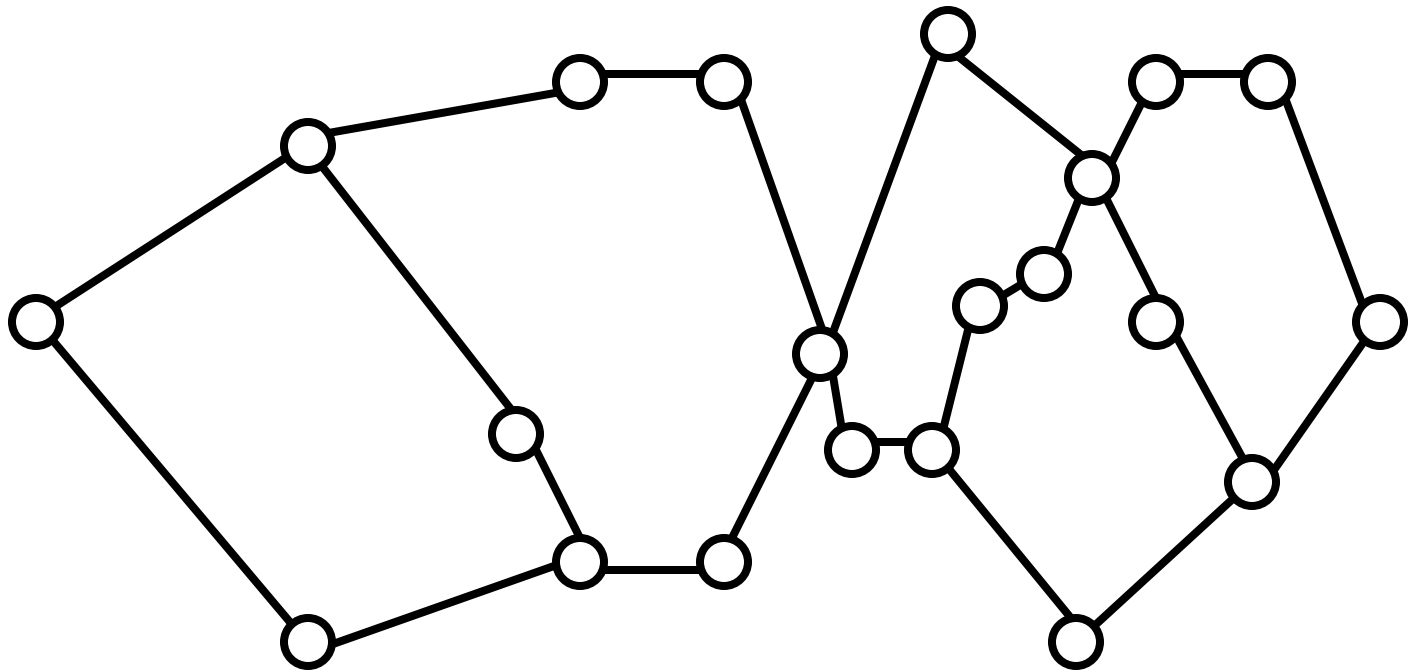
Applications: Coverage

- By reducing the world to cells, we've essentially abstracted the world to a graph.



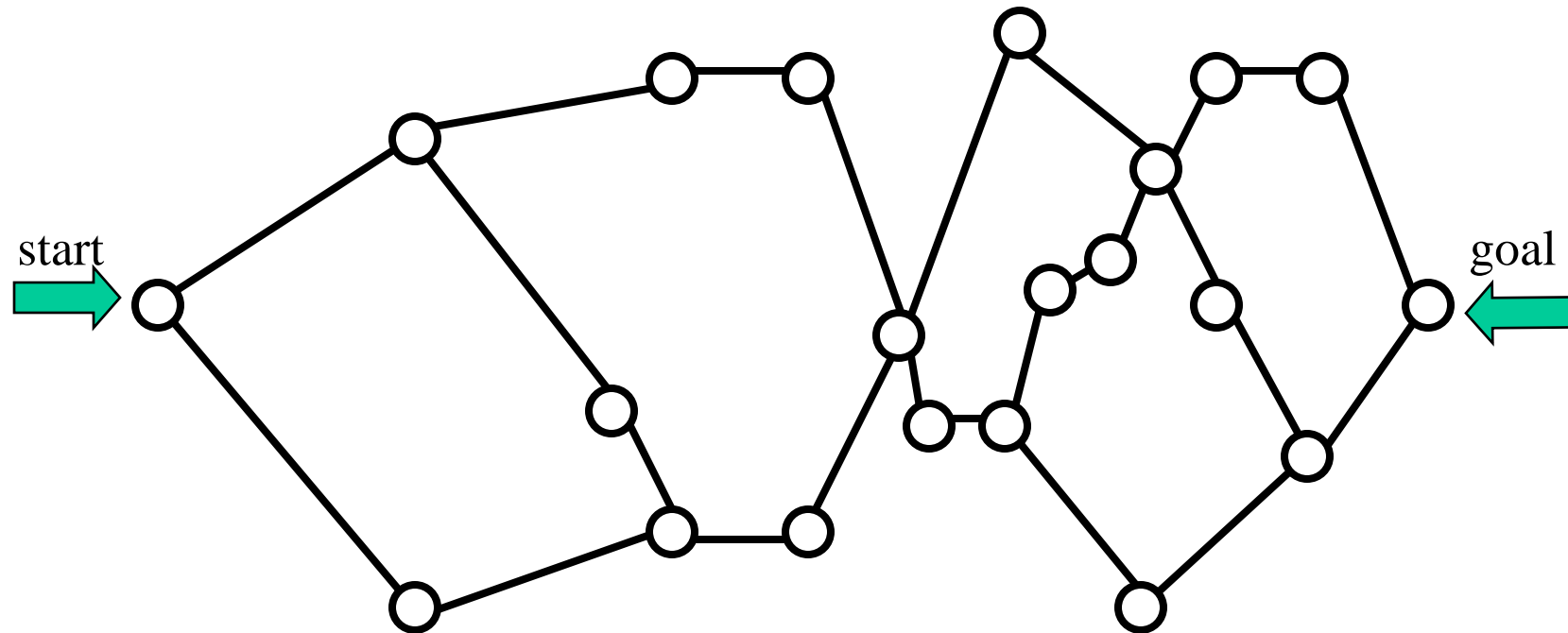
Find a path

- By reducing the world to cells, we've essentially abstracted the world to a graph.



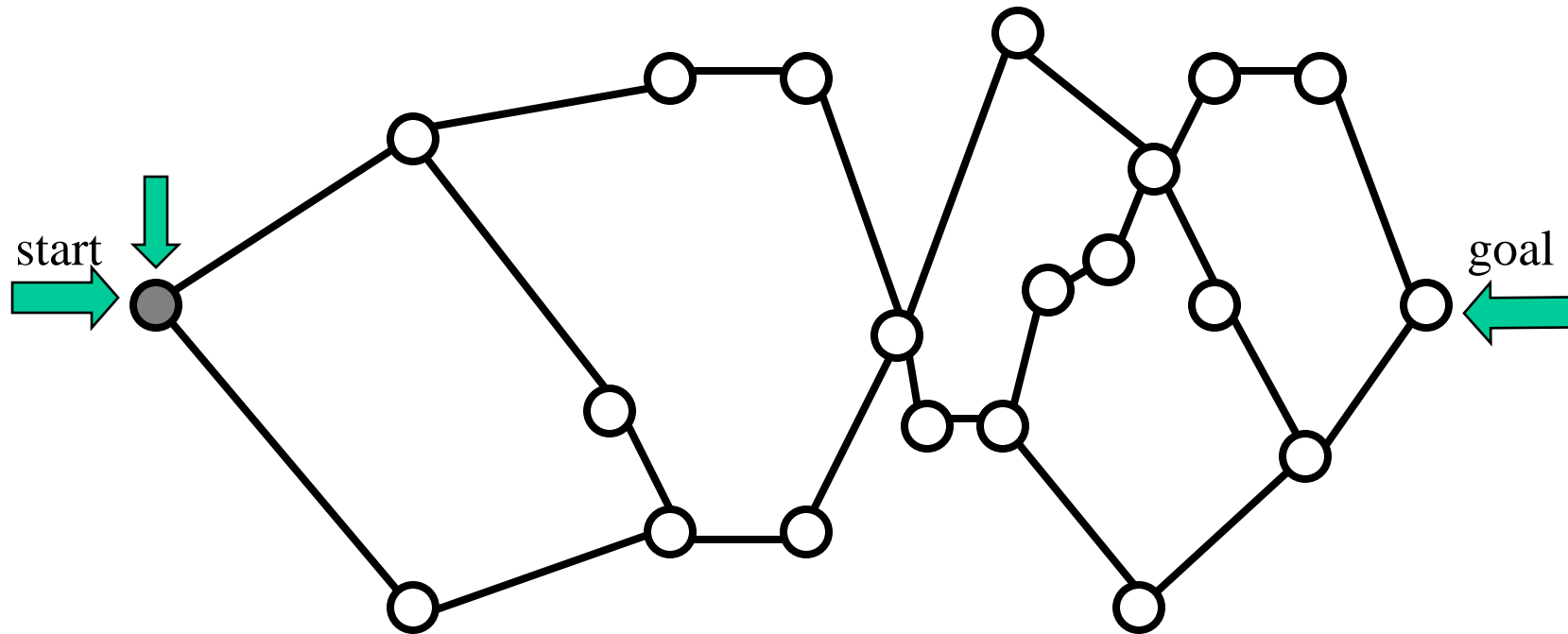
Find a path

- With an adjacency graph, a path from start to goal can be found by simple traversal



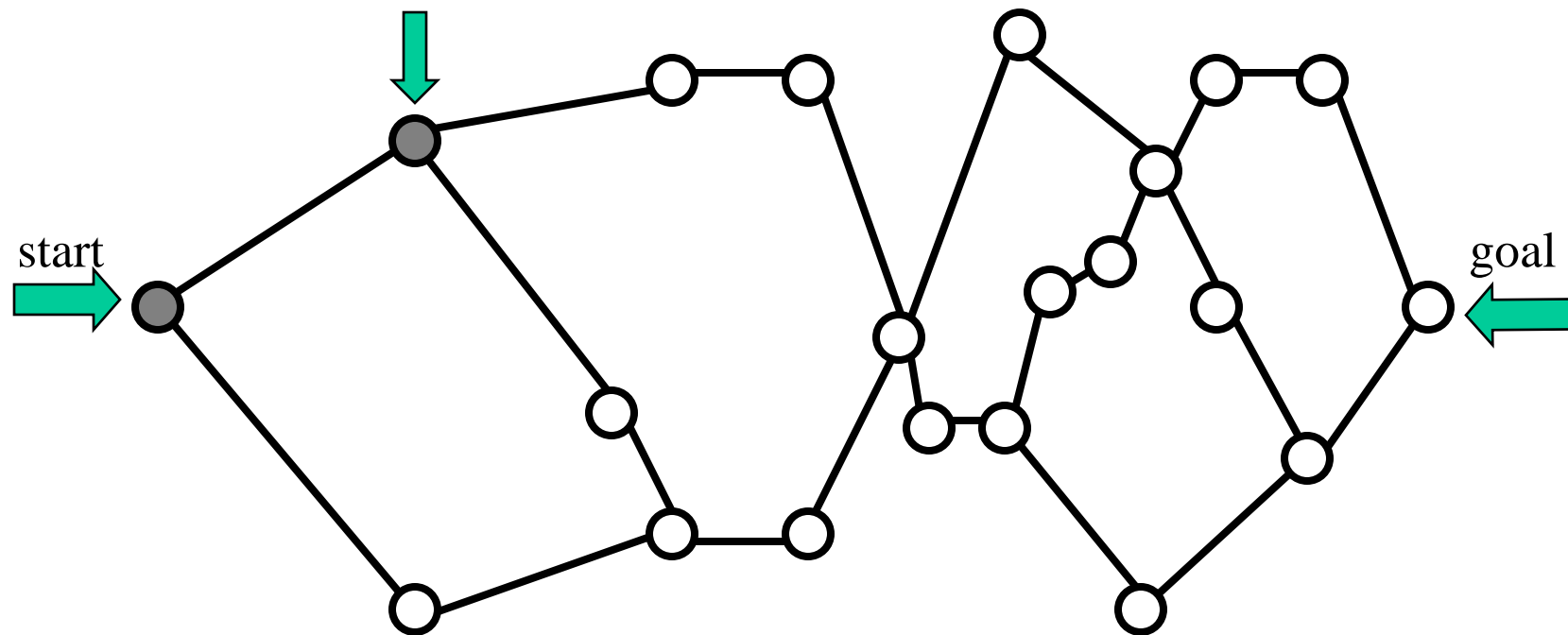
Find a path

- With an adjacency graph, a path from start to goal can be found by simple traversal



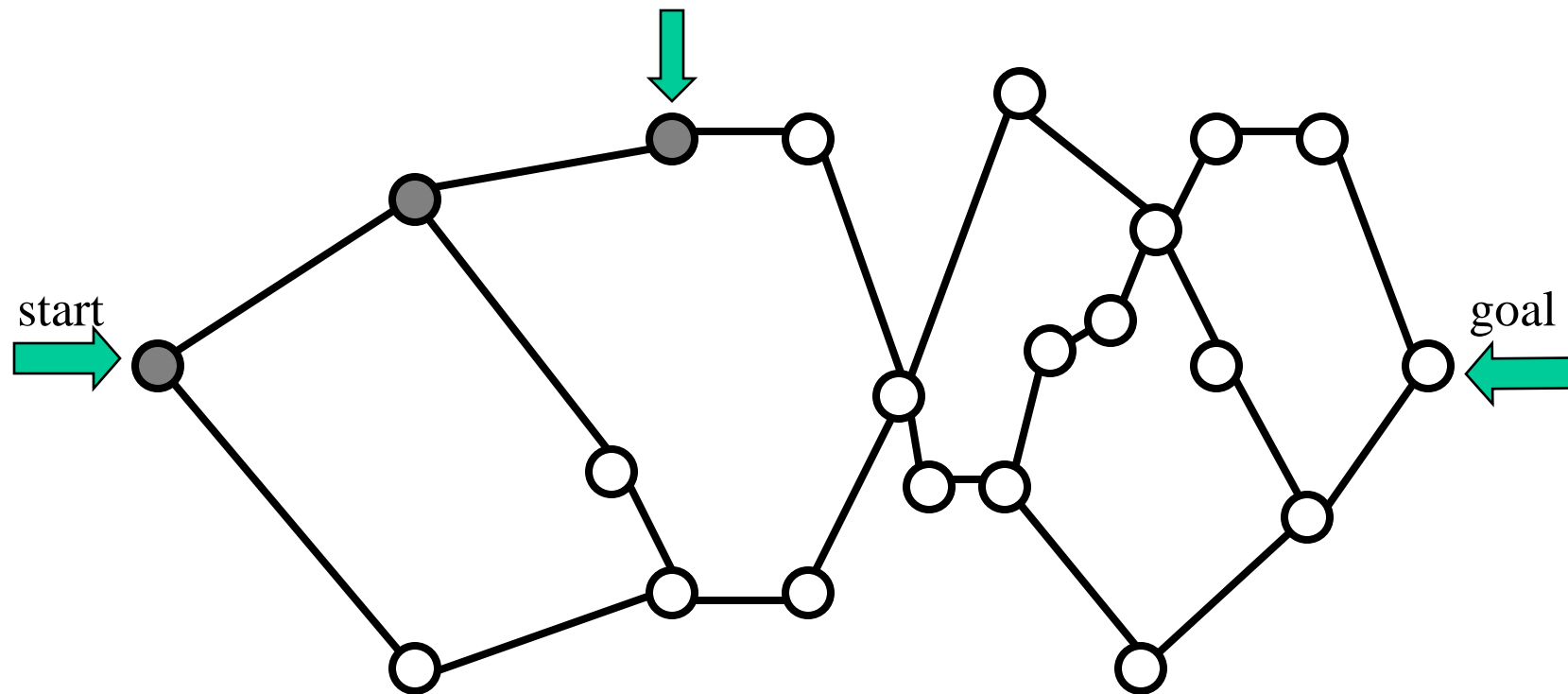
Find a path

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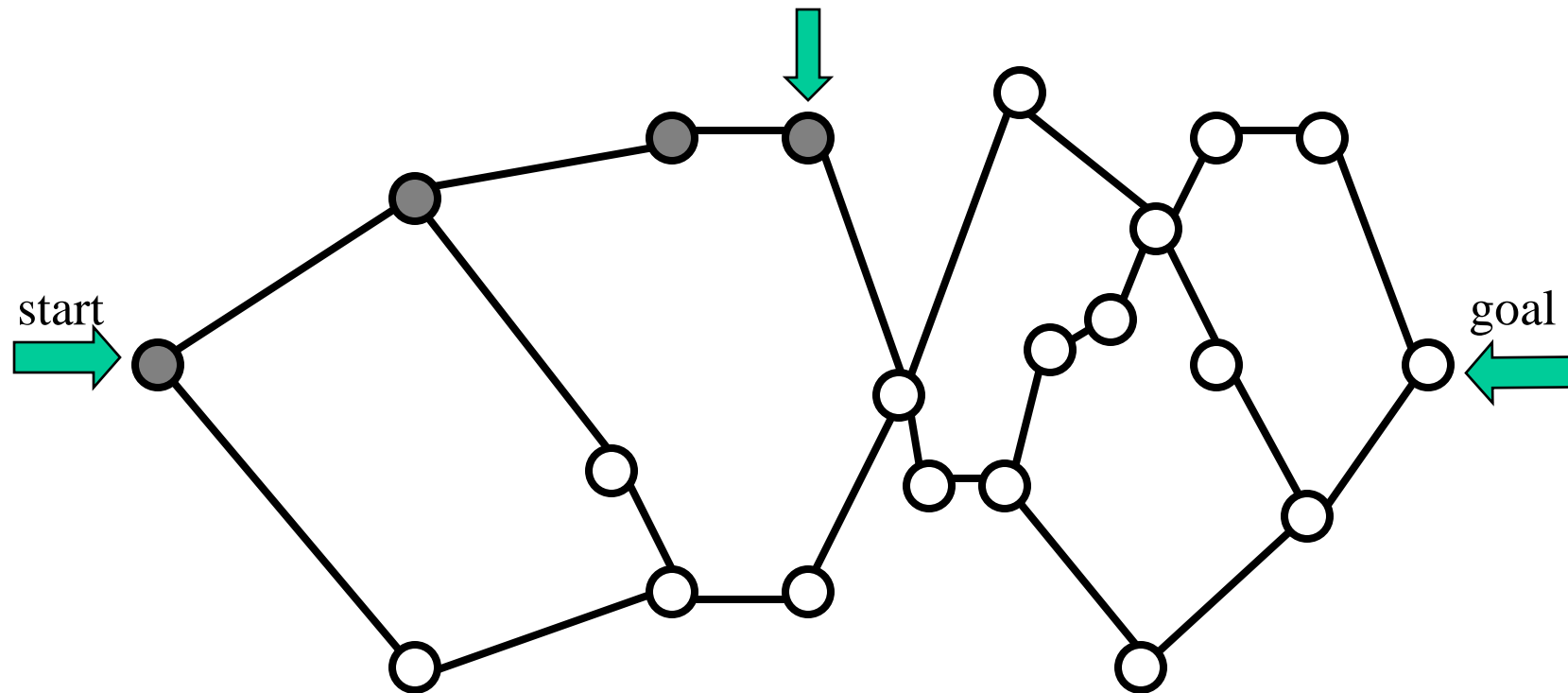
Find a path

- With an adjacency graph, a path from start to goal can be found by simple traversal



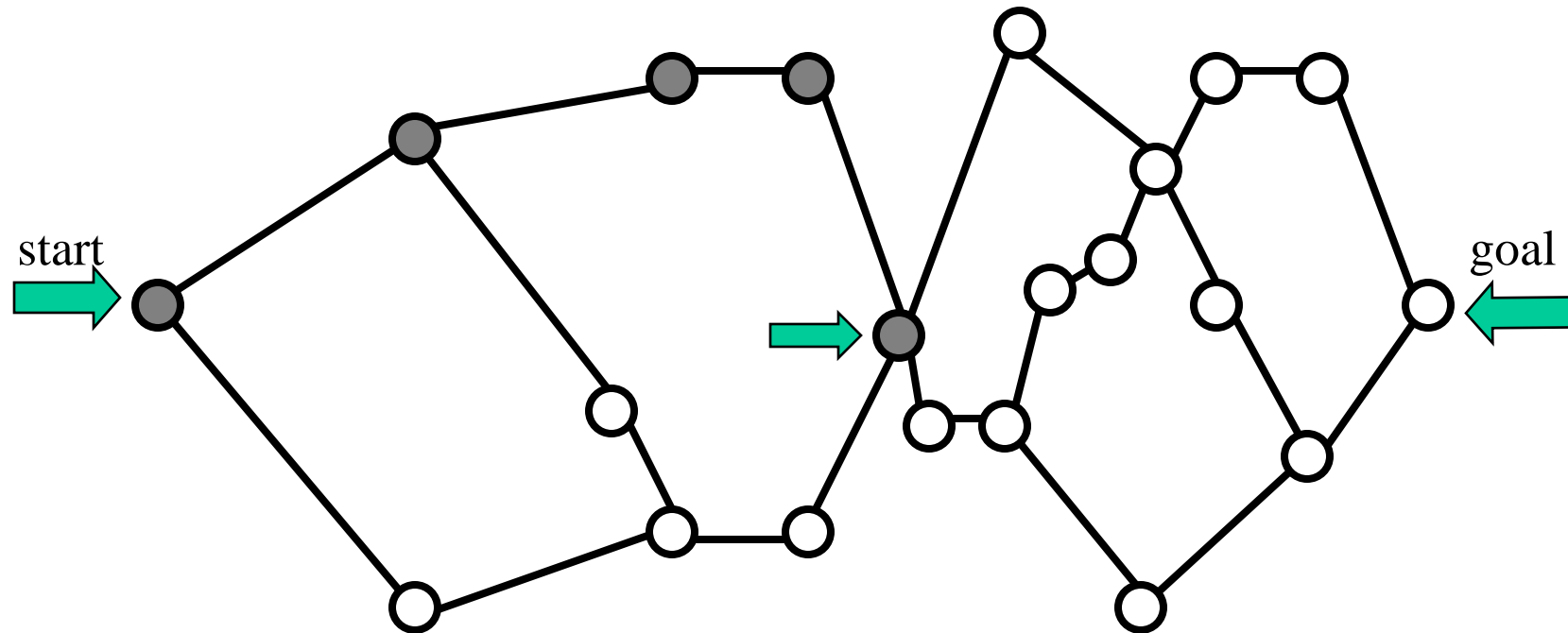
Find a path

- With an adjacency graph, a path from start to goal can be found by simple traversal



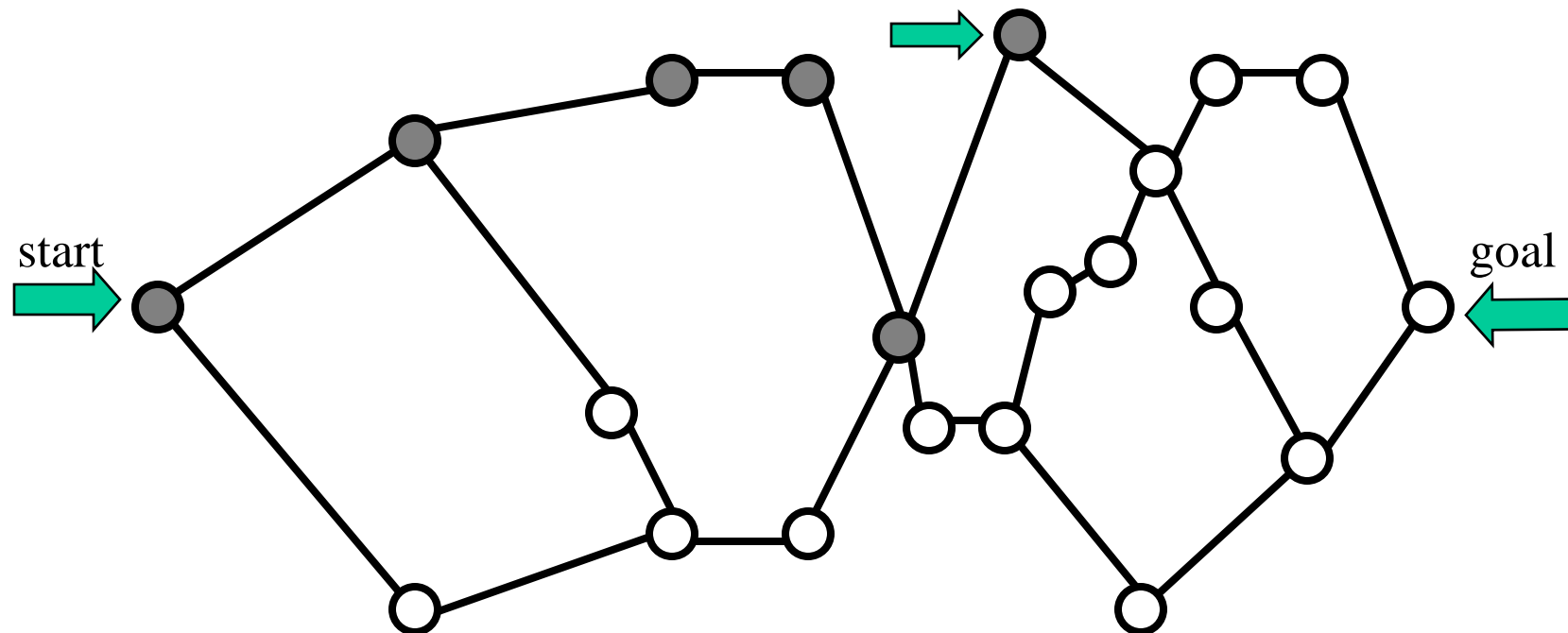
Find a path

- With an adjacency graph, a path from start to goal can be found by simple traversal



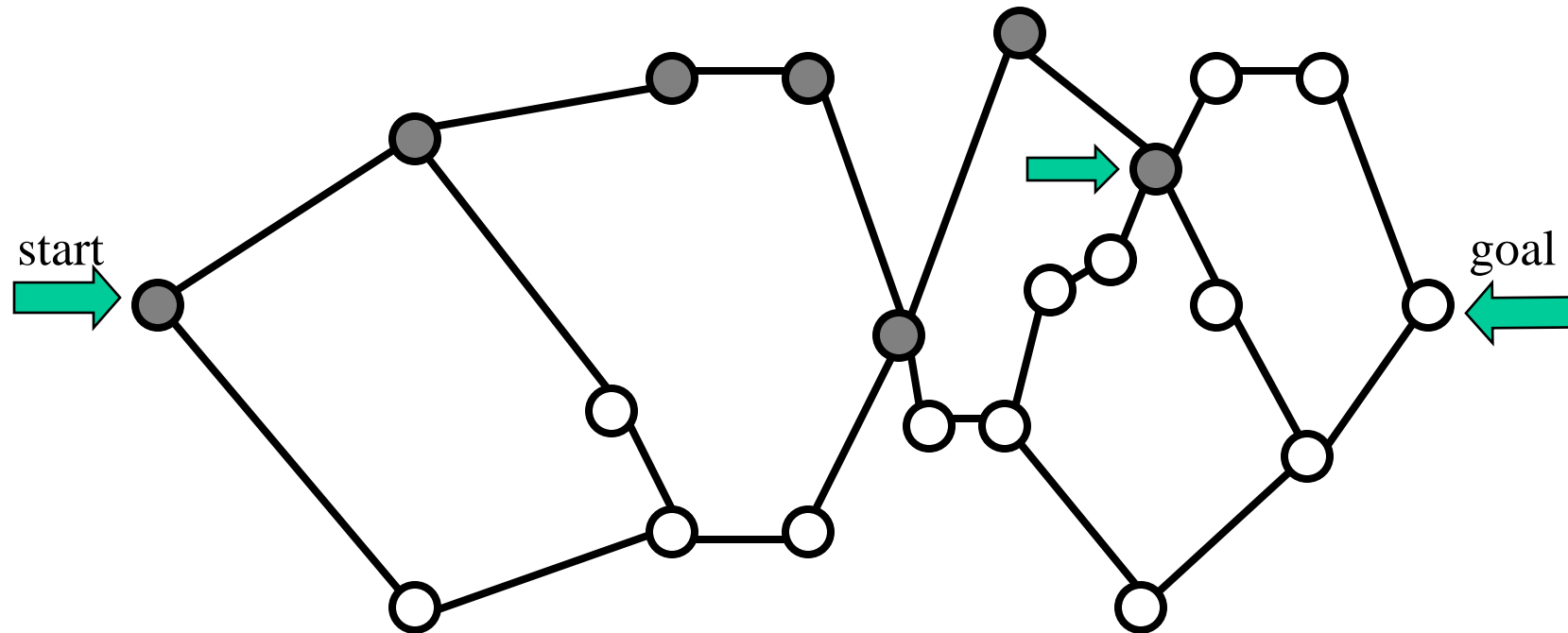
Find a path

- With an adjacency graph, a path from start to goal can be found by simple traversal



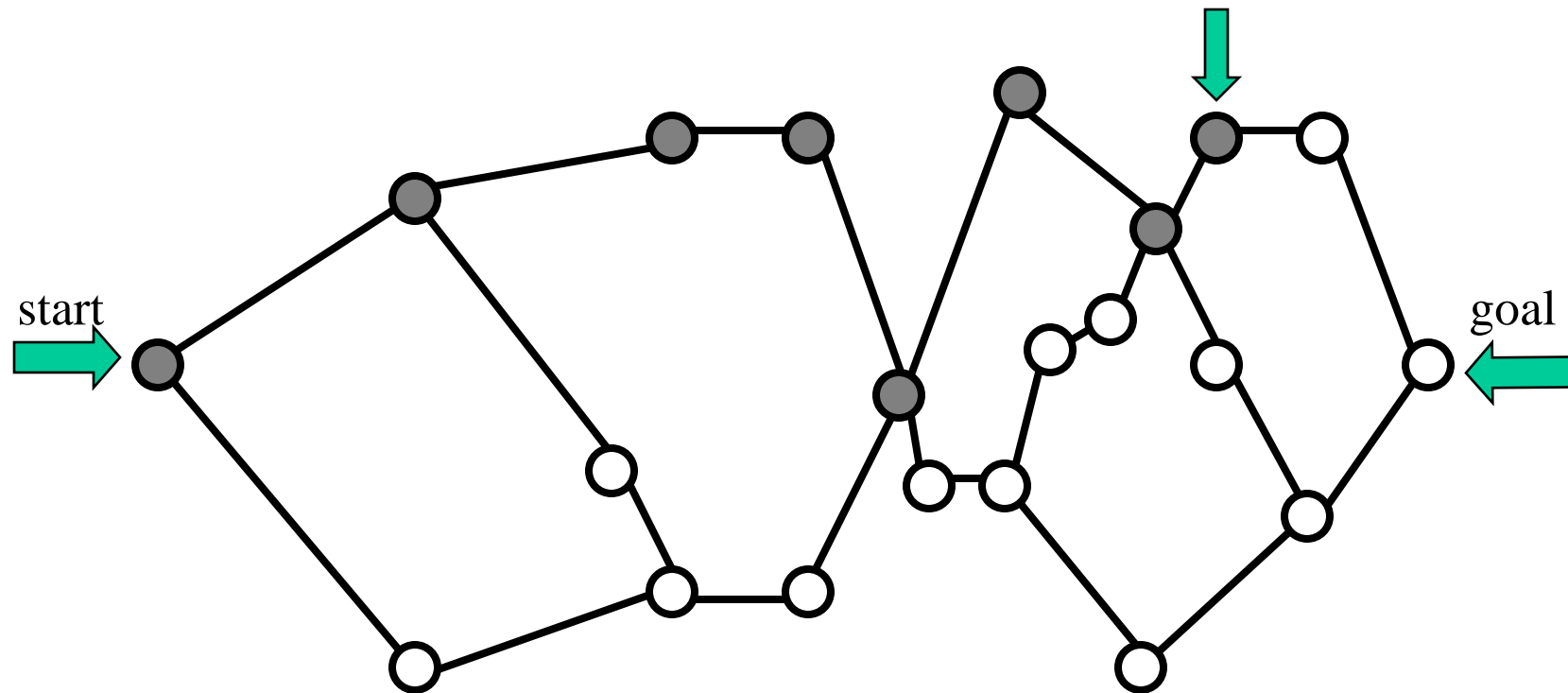
Find a path

- With an adjacency graph, a path from start to goal can be found by simple traversal



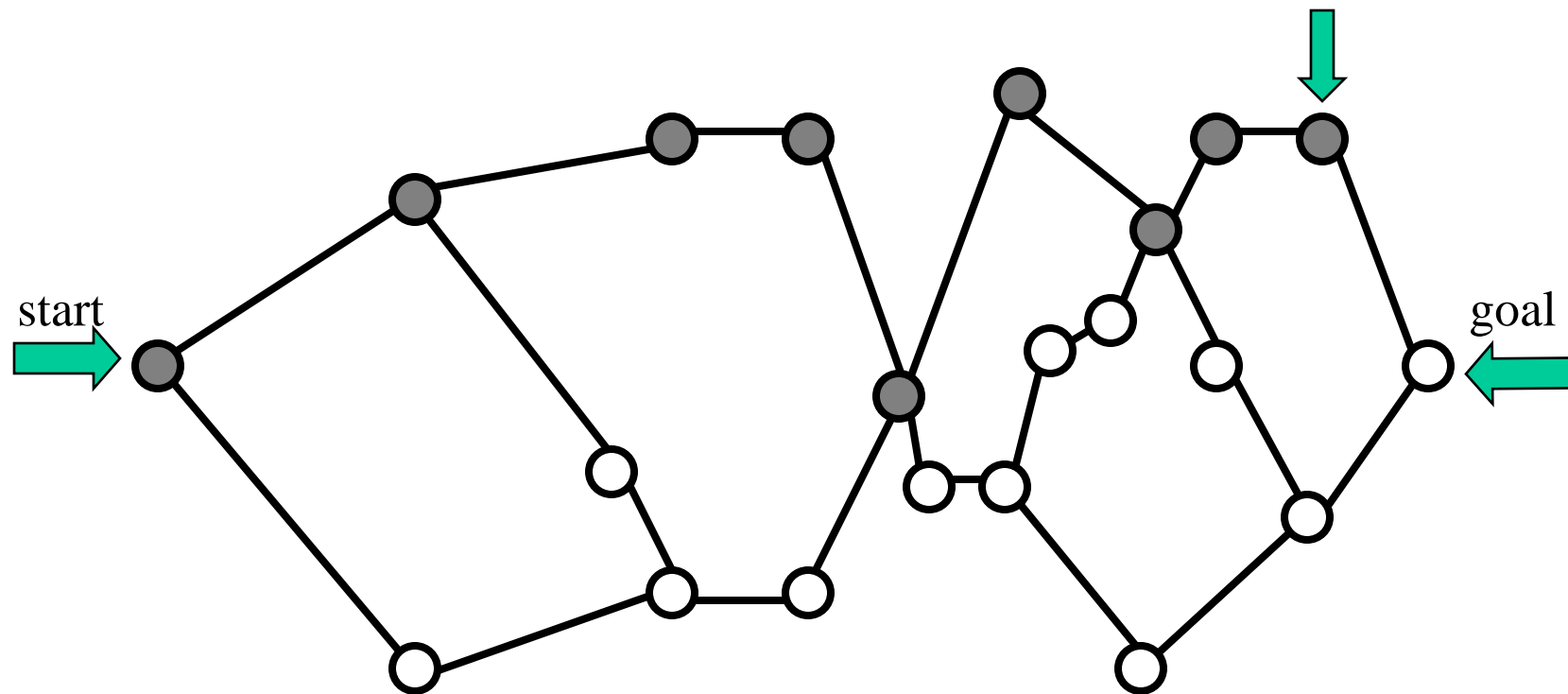
Find a path

- With an adjacency graph, a path from start to goal can be found by simple traversal



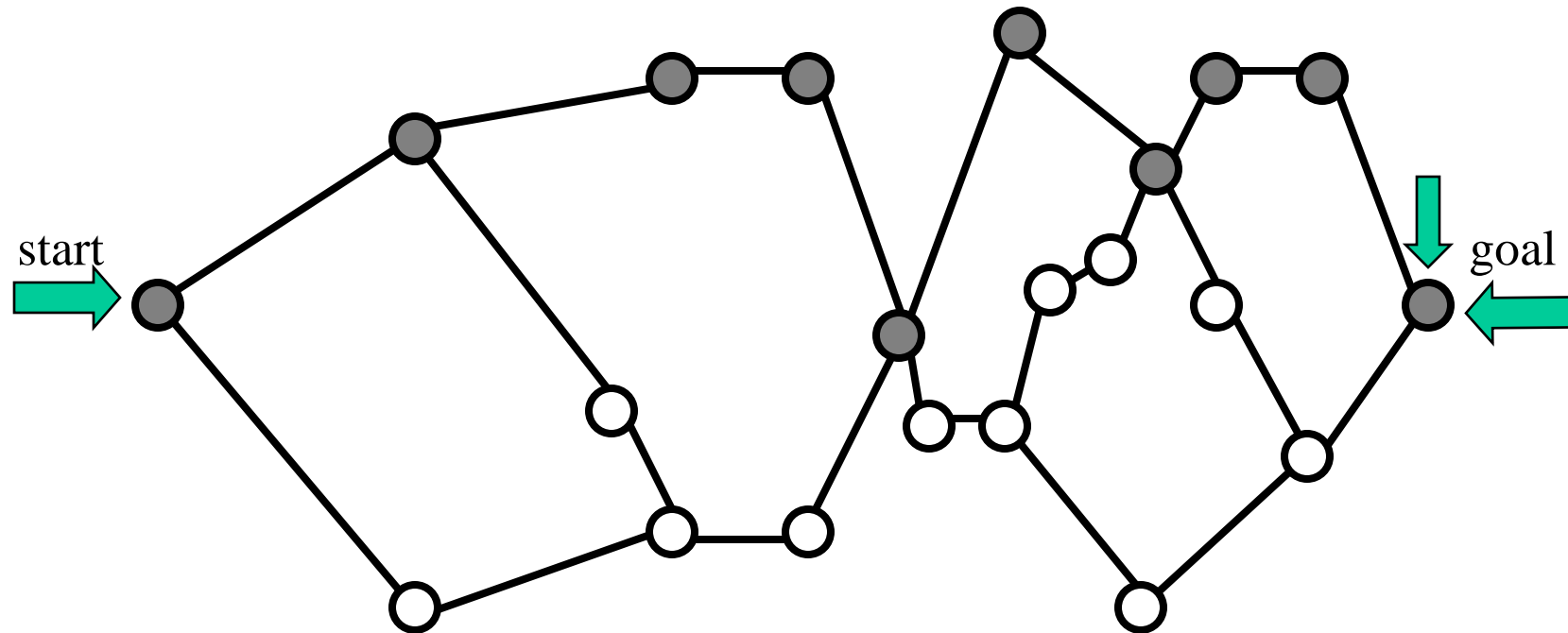
Find a path

- With an adjacency graph, a path from start to goal can be found by simple traversal



Find a path

- With an adjacency graph, a path from start to goal can be found by simple traversal



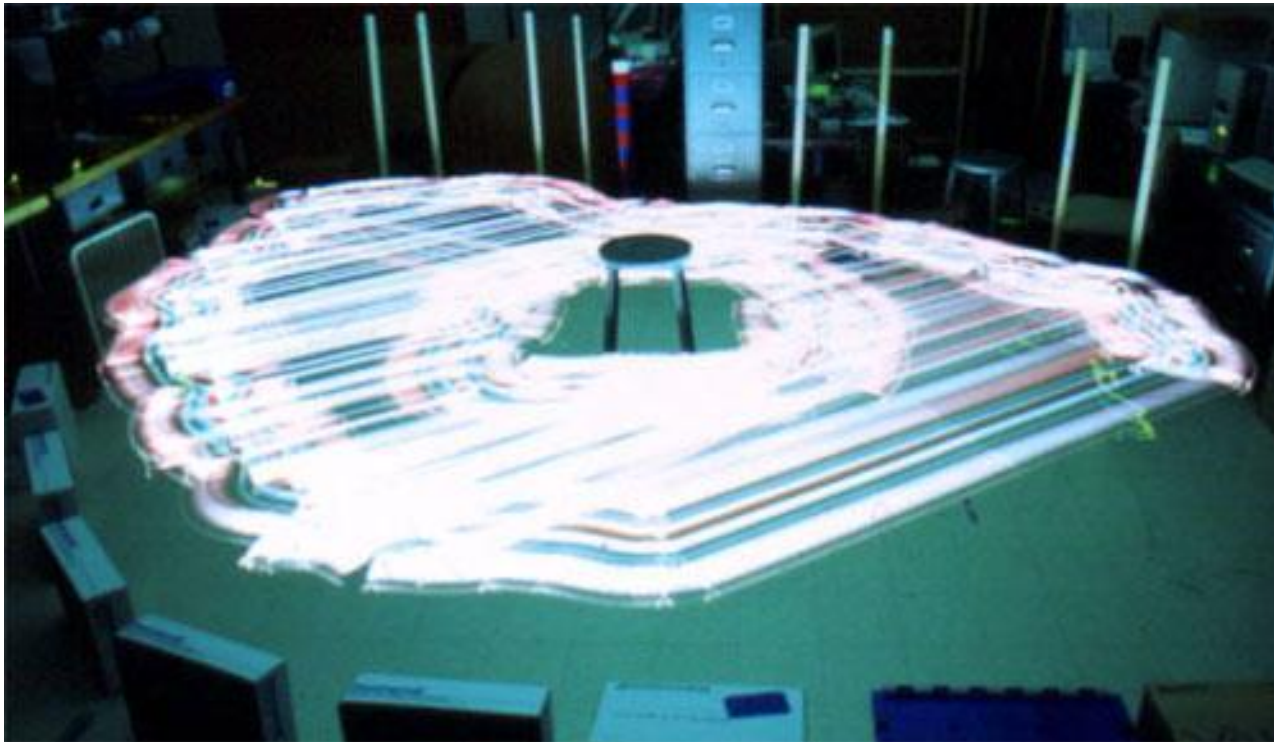
Connect Midpoints of Traps

Applications: Coverage

- First, a distinction between sensor and detector must be made
- *Sensor*: Senses obstacles
- *Detector*: What actually does the coverage
- We'll be observing the simple case of having an omniscient sensor and having the detector's footprint equal to the robot's footprint

Howie Choset

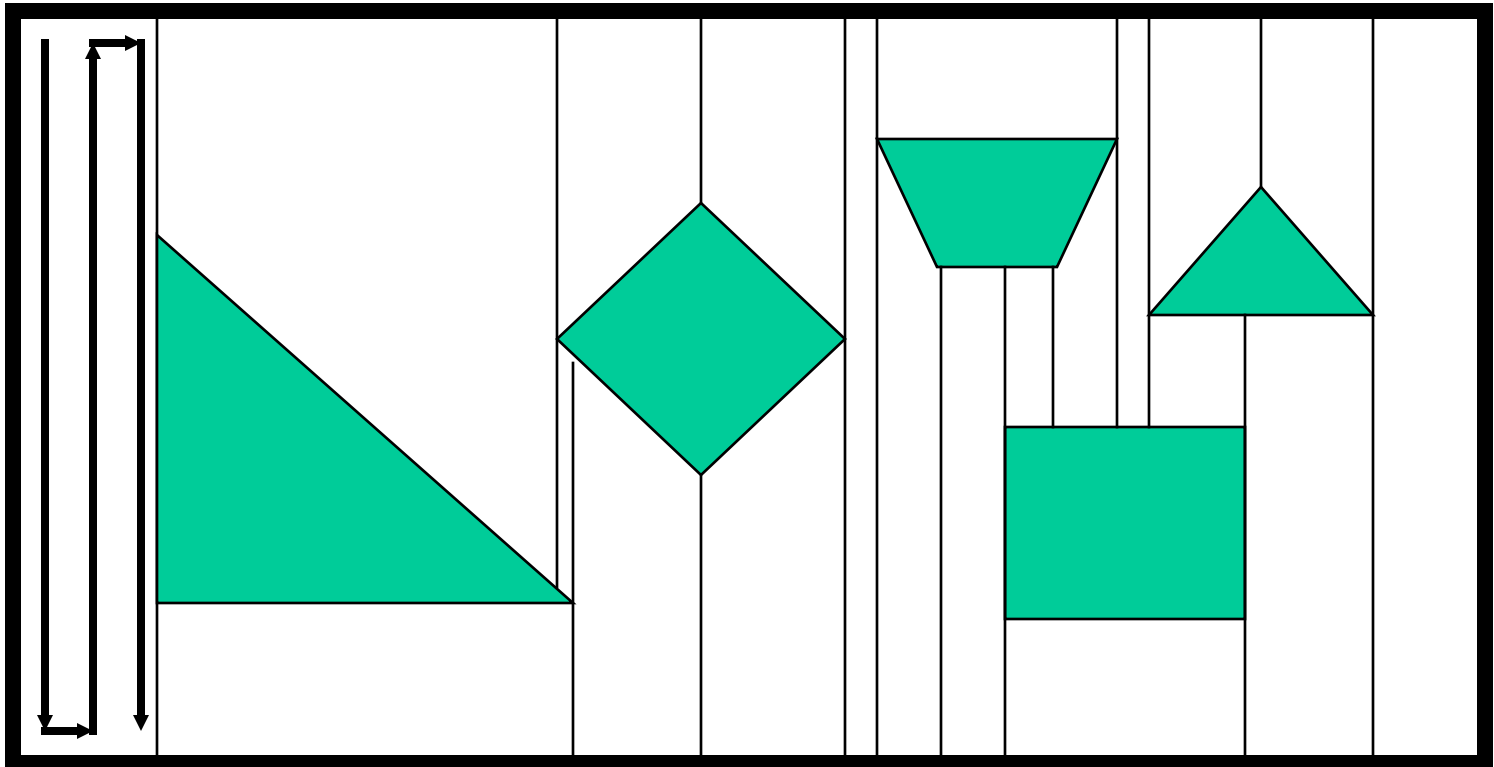
Robotics Institute



Coverage
(snake robots)

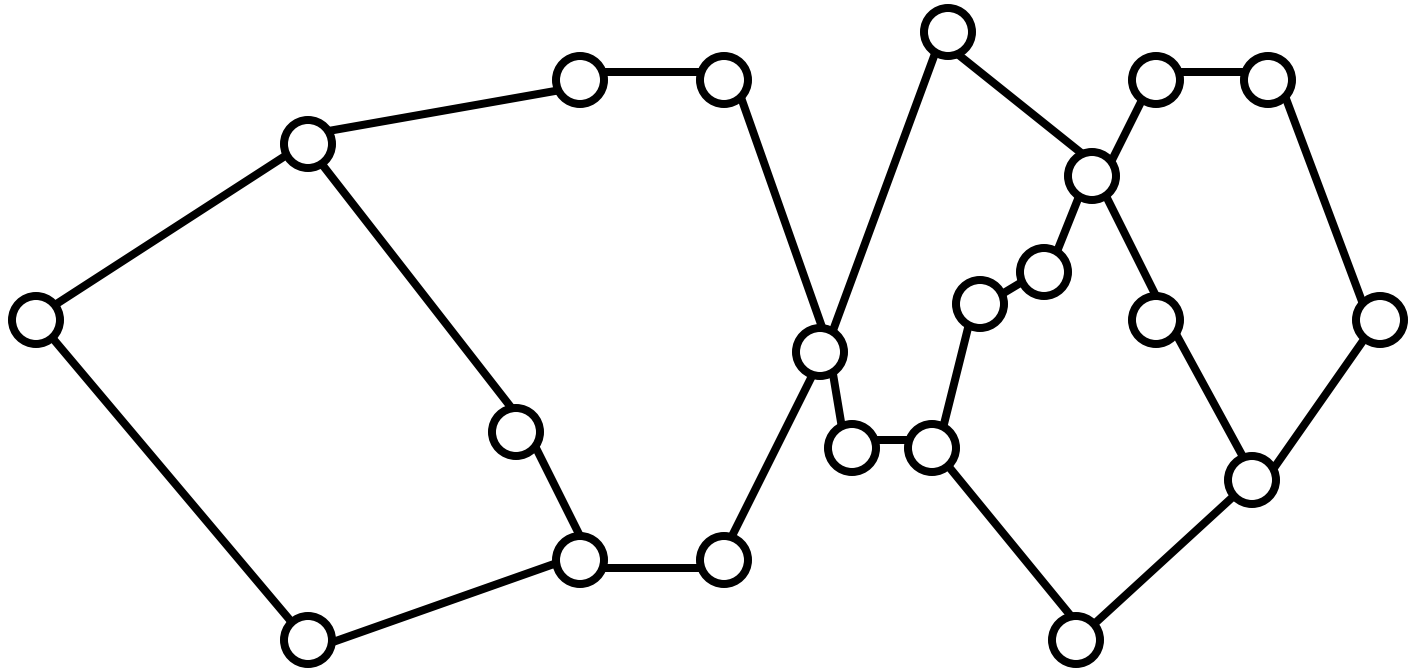
Cell Decompositions: Trapezoidal Decomposition

- How is this useful? Well, trapezoids can easily be covered with simple back-and-forth sweeping motions. If we cover all the trapezoids, we can effectively cover the entire “reachable” world.

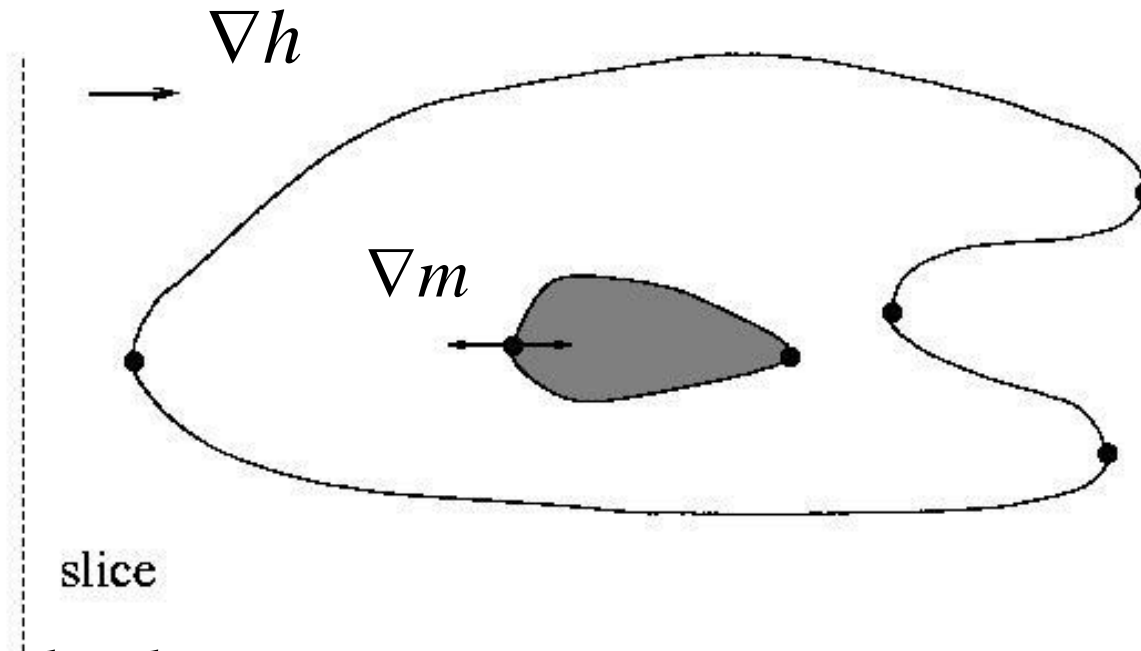


Applications: Coverage

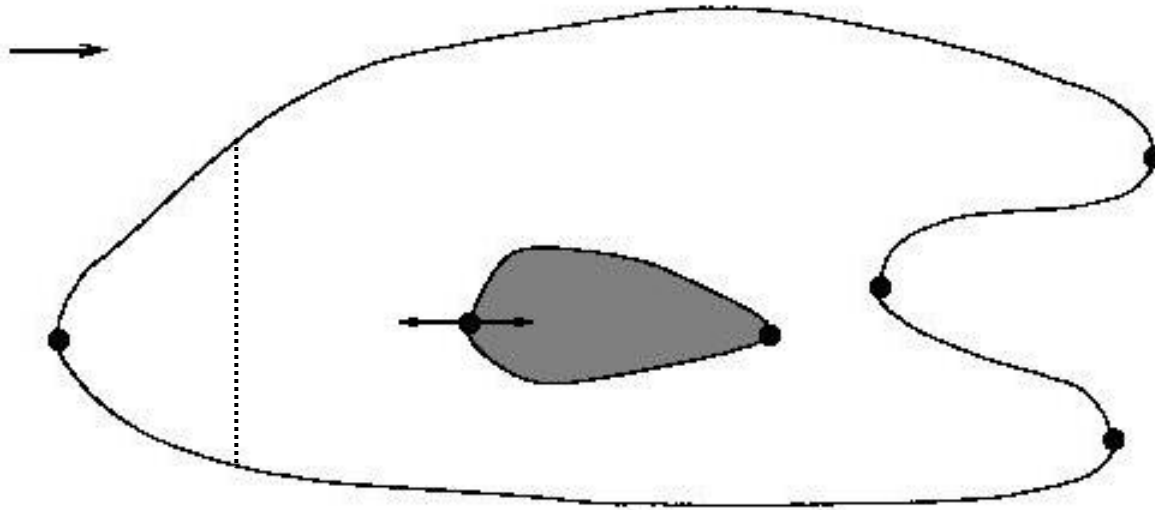
- Simply visit all the nodes, performing a sweeping motion in each, and you're done.



Cell Decomposition in Terms of Critical Points

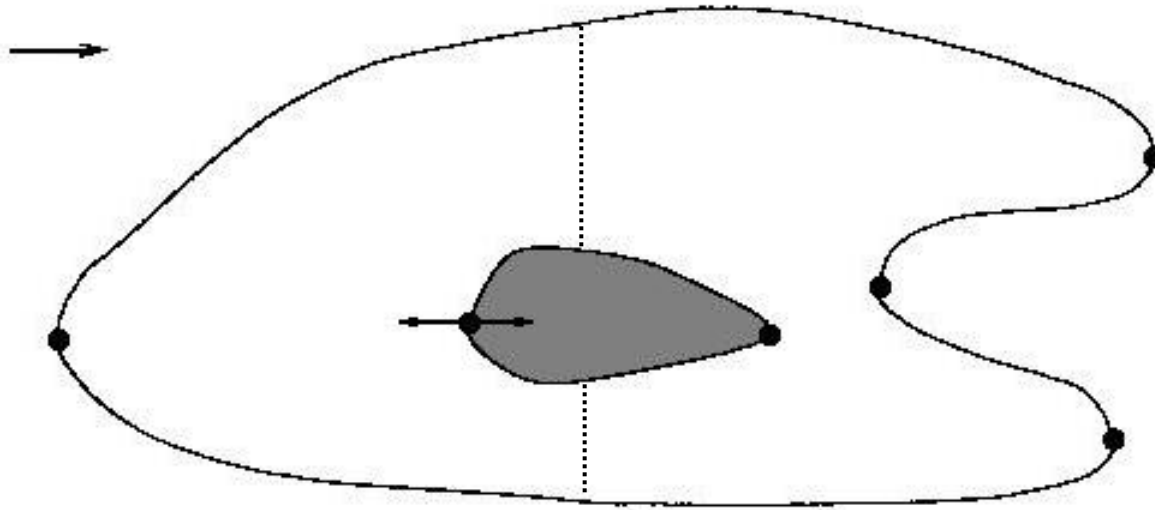


- *Slice is a level set*
- *Slice function: $h(x,y) = x$, $slice = \{(x,y) / h(x,y) = \lambda\}$*
- *At a critical point x of $h|_M$, $\nabla h(x) = \nabla m(x)$ where $M = \{x / m(x) = 0\}$*



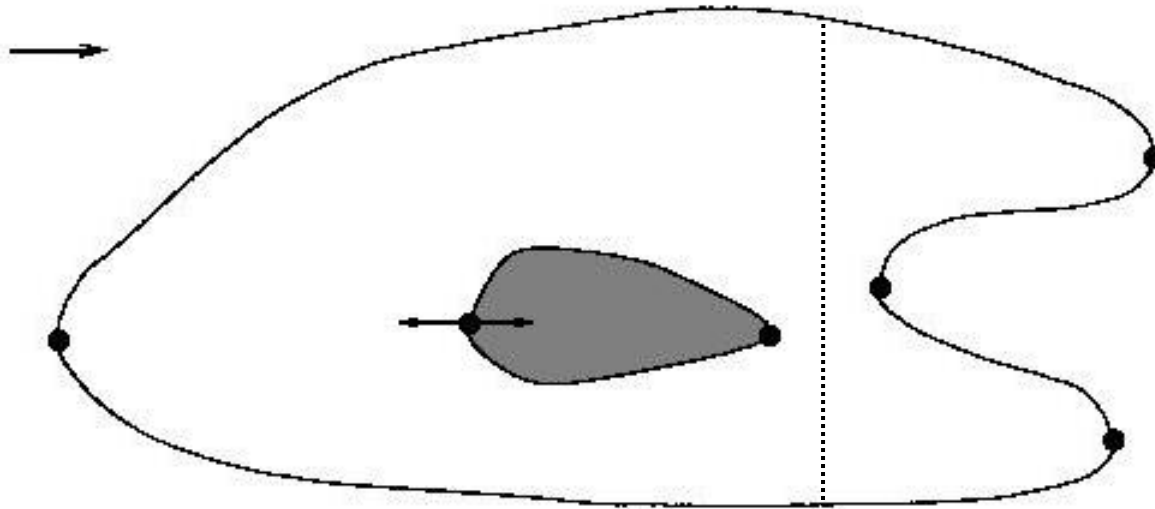
1-connected

$$h(x,y) = a_1$$



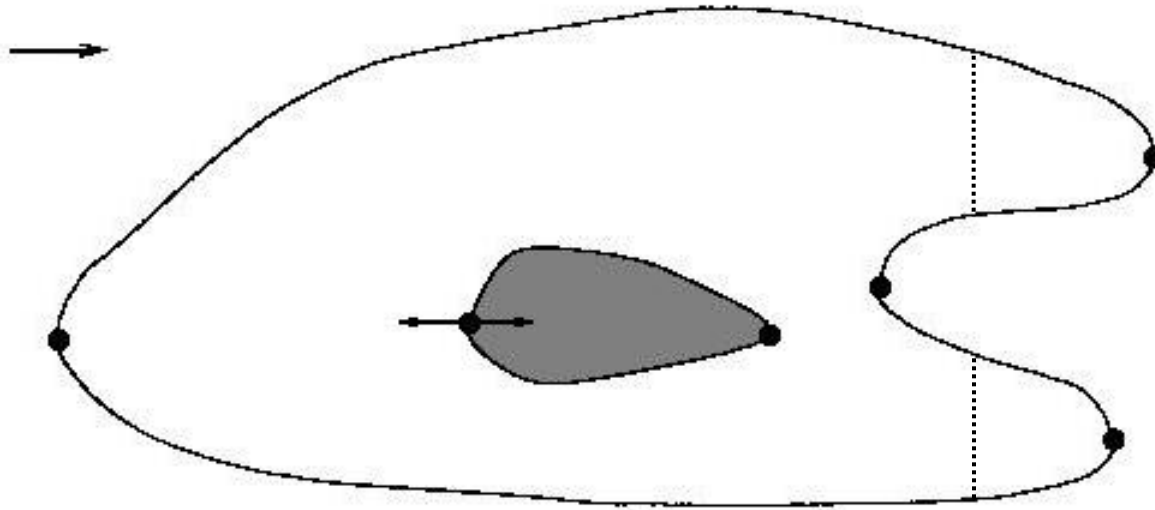
2-connected

$$h(x,y) = a_2$$



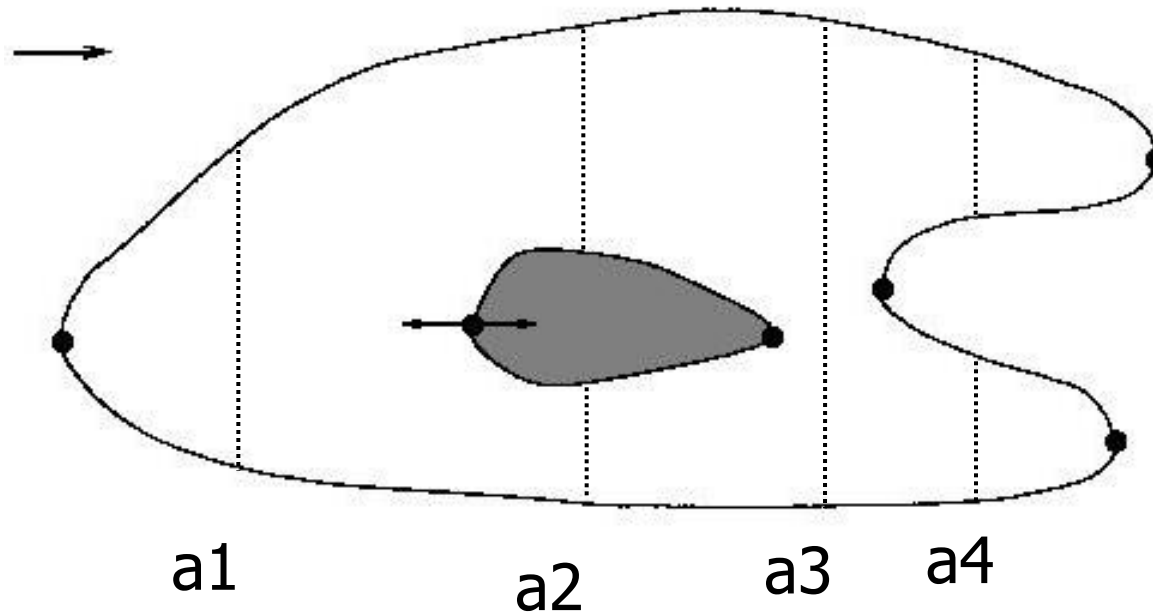
1-connected

$$h(x,y) = a_3$$

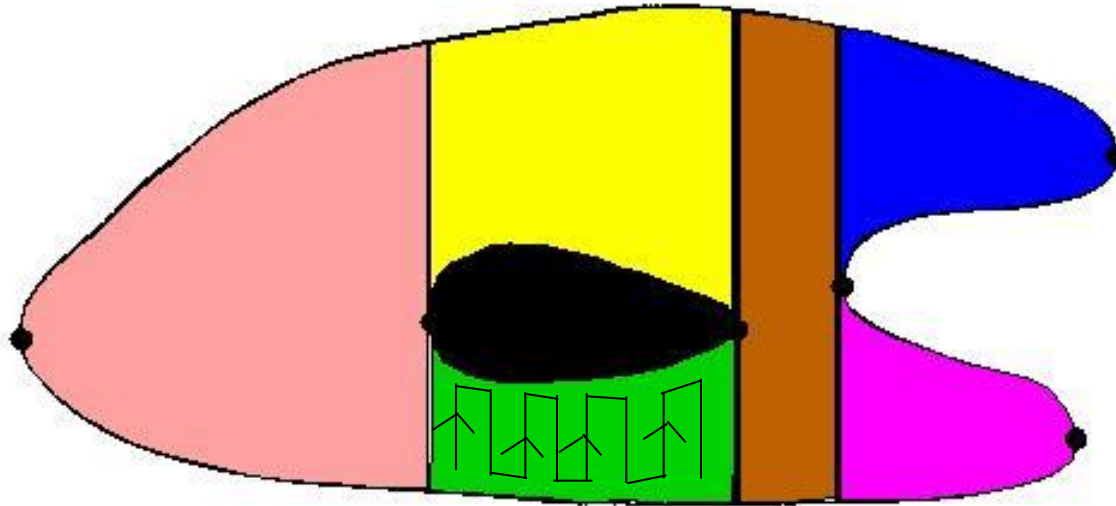


2-connected

$$h(x,y) = a_4$$

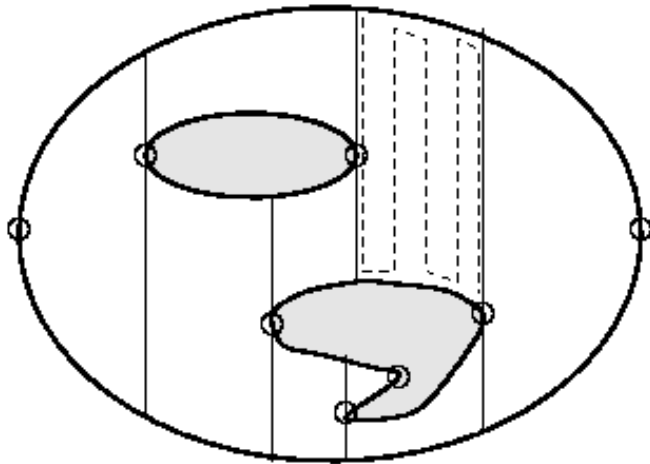


- *Connectivity of the slice in the free space changes at the critical points (Morse theory)*



- *Each cell can be covered by back and forth motions*

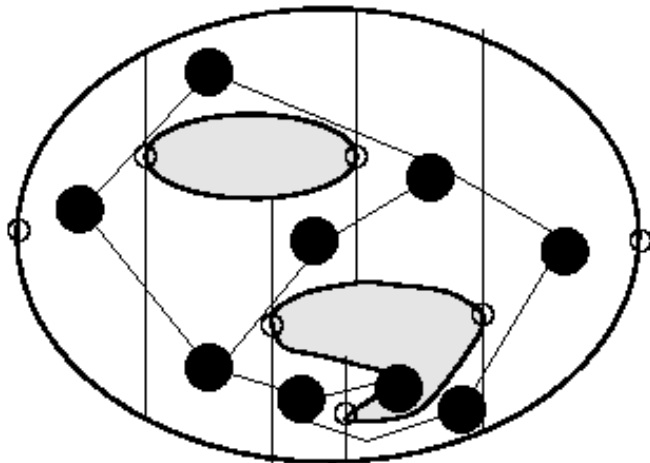
Cell-Decomposition Approach



Cell
Decomp

Define Decomposition

- Completeness



Adjacency
Graph

Sensor-based Construction

Define Other
Decompositions

- Other patterns
- Extended detector

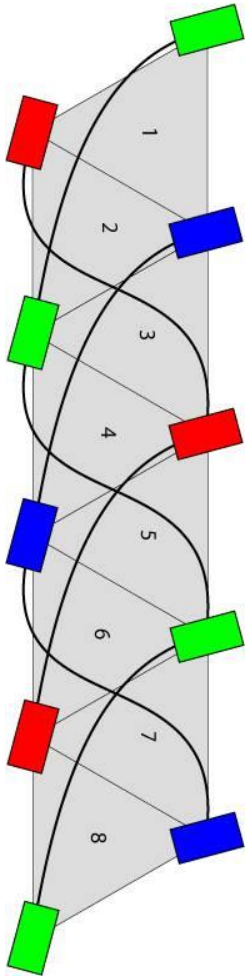
Provably completeness \neq guaranteed completeness



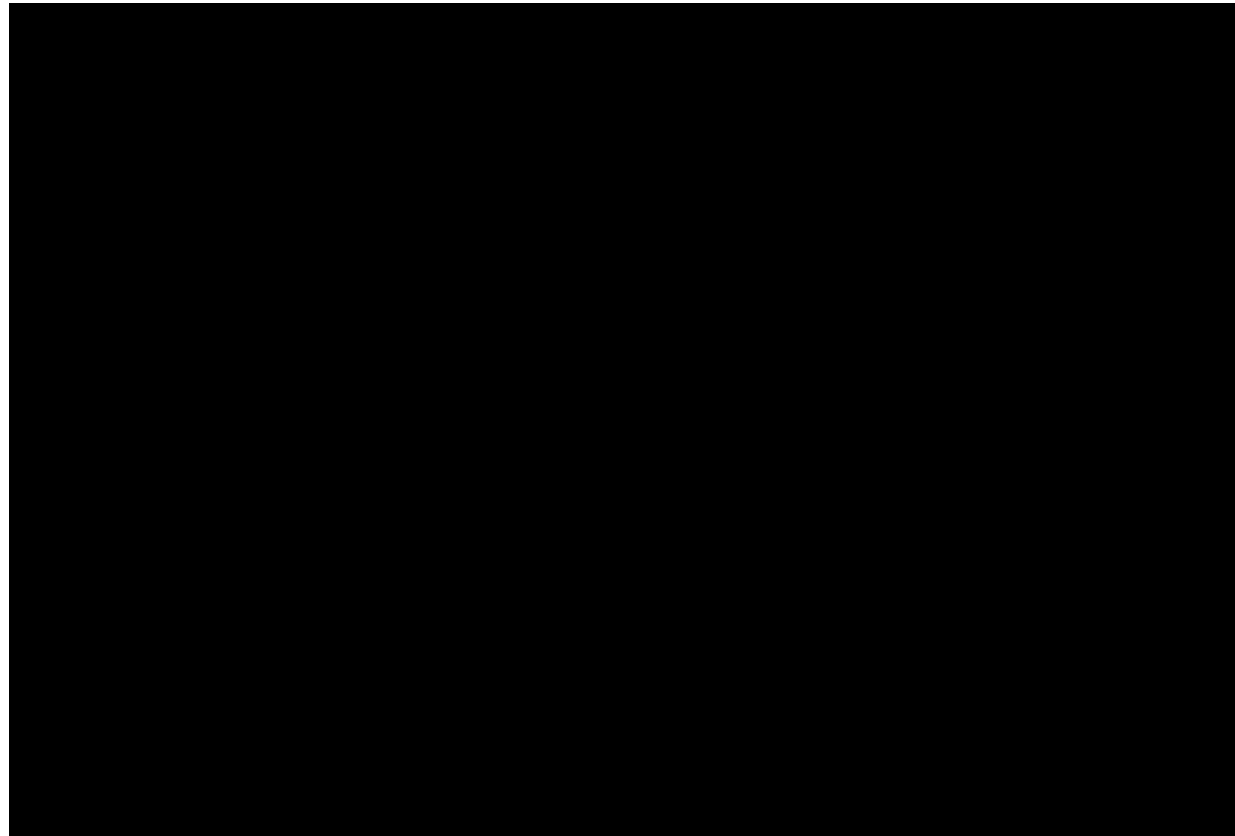
jump to swear



Localization



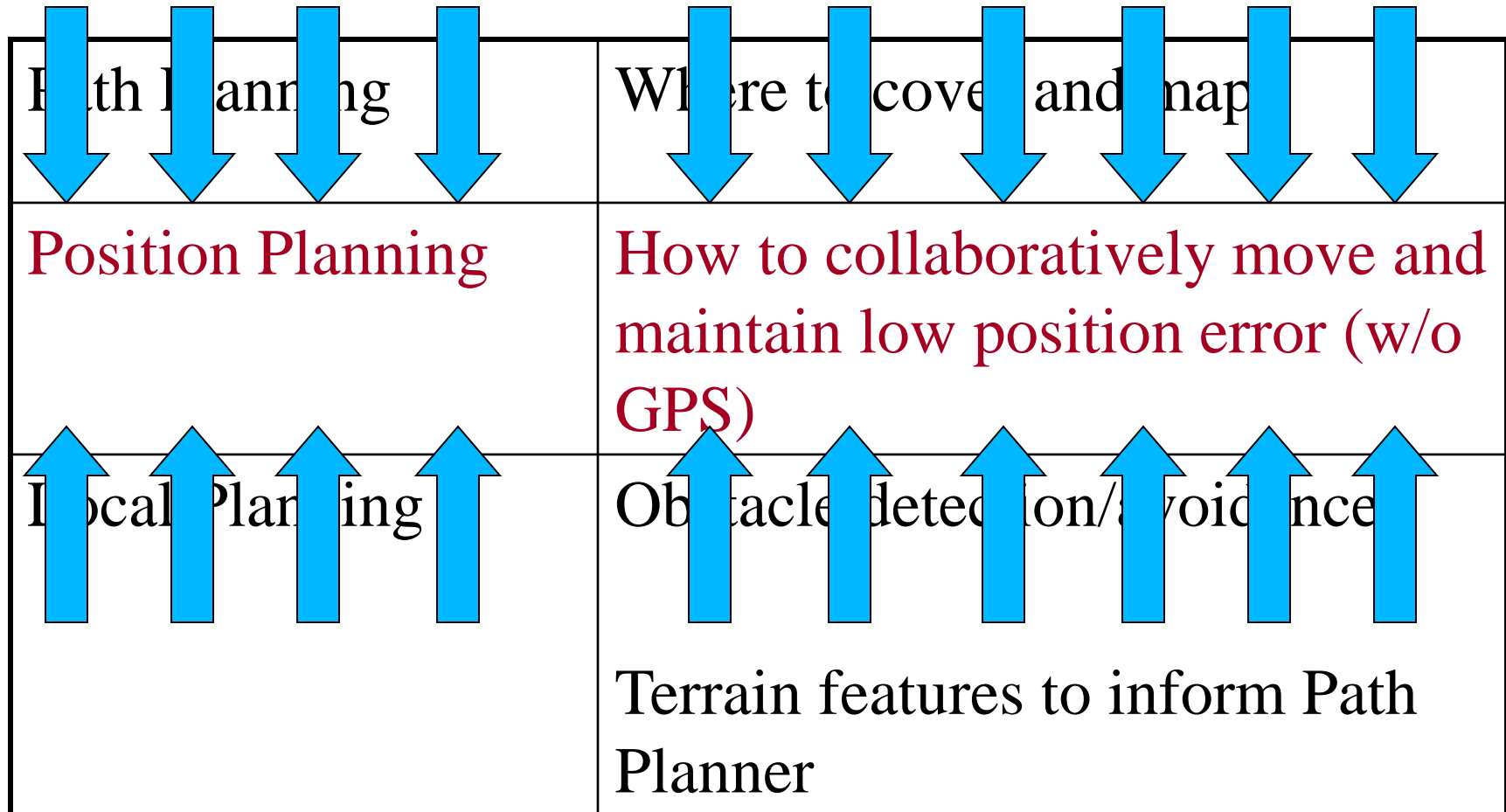
*mapping too



25m x 30m

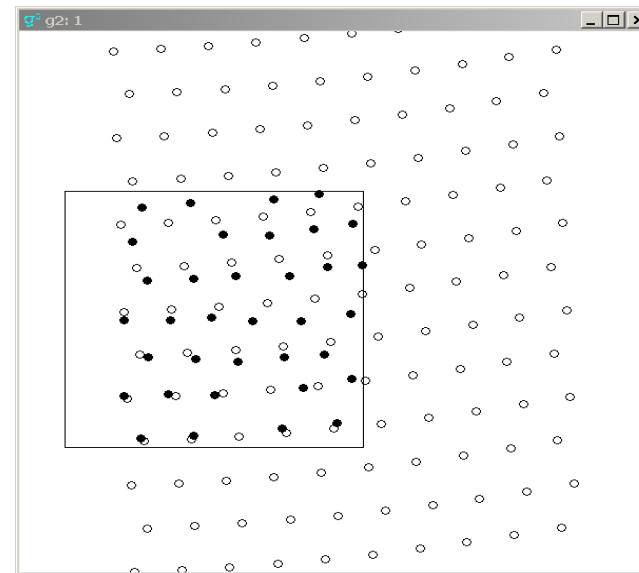
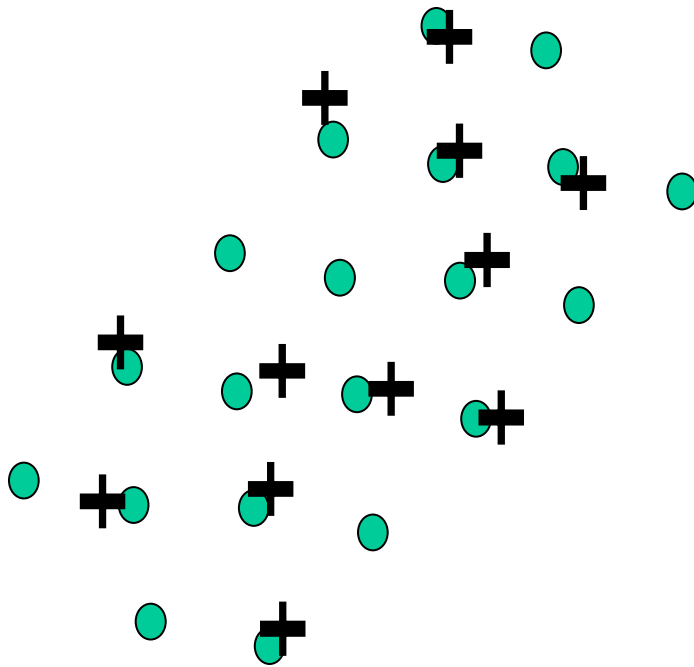
Successful Experiment:
Stopped because of robot battery
limitations

Operational Hierarchy



Calibrate robots' initial location and go

Probabilistic Coverage






Surface Deposition

- Process Variables
 - Uniformity
 - Waste
 - Positioning
- Cycle-time
 - Time-to-completion
 - Programming time



Conclusion: Complete Overview

-  • The Basics
 - Motion Planning Statement
 - The World and Robot
 - Configuration Space
 - Metrics
-  • Path Planning Algorithms
 - Start-Goal Methods
 - Lumelsky Bug Algorithms
 - Potential Charge Functions
 - The Wavefront Planner
 - Map-Based Approaches
 - Generalized Voronoi Graphs
 - Visibility Graphs
 - Cellular Decompositions => Coverage
-  • *Done with Motion Planning!*