

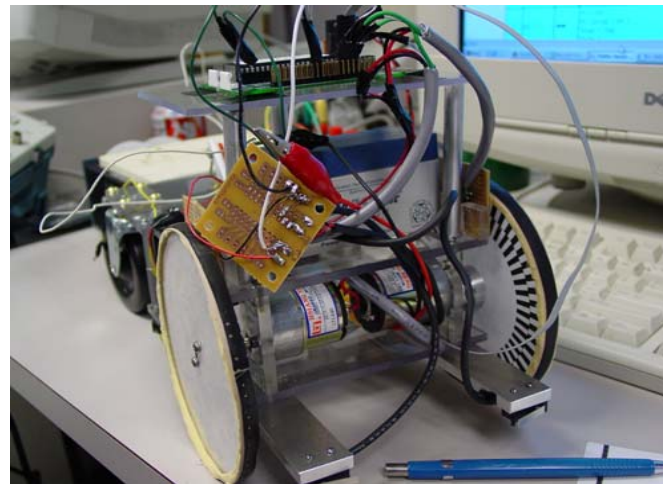
# Motion Planning

Howie CHoset

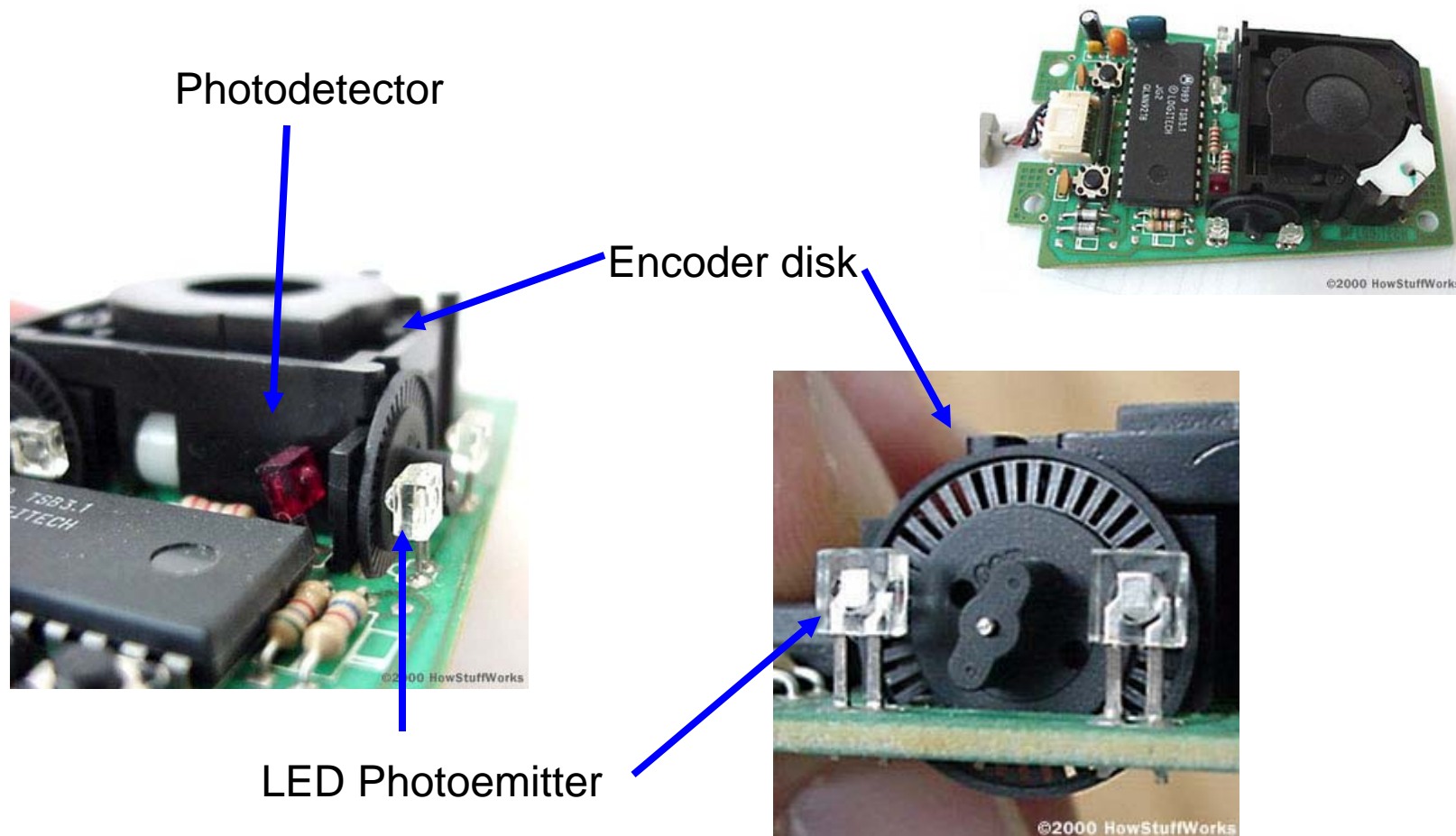
# Questions

- Where are we?
- Where do we go?
- Which is more important?

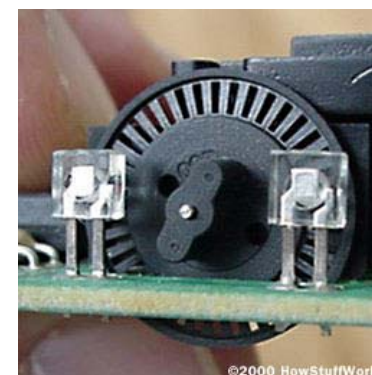
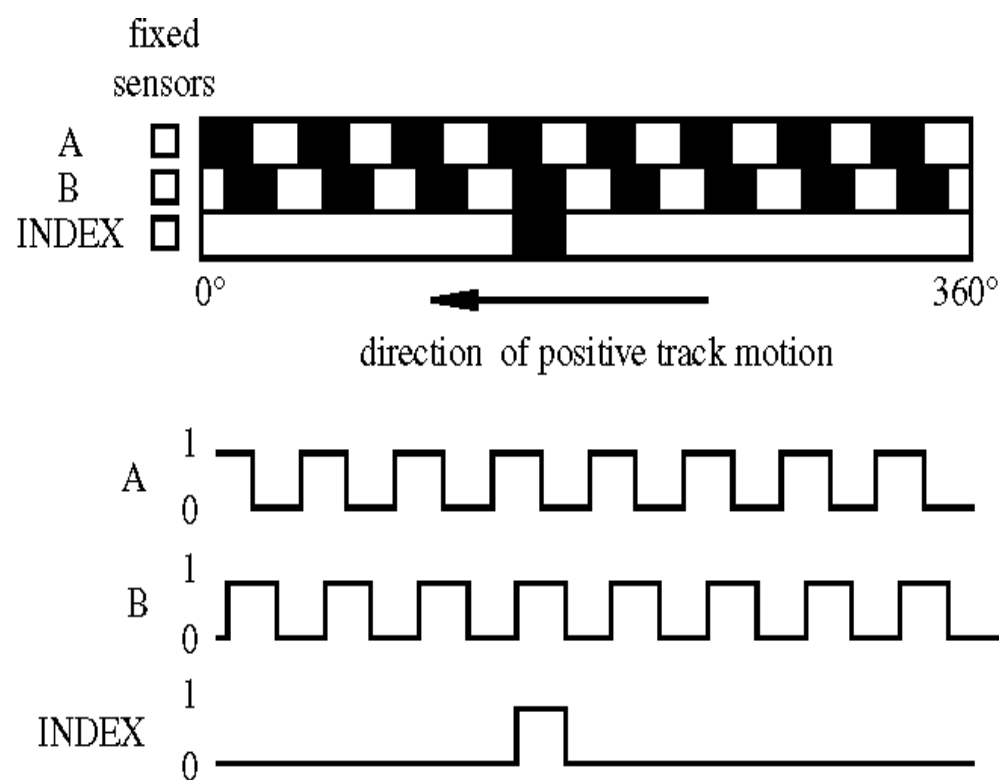
# Encoders



# Encoders – Incremental

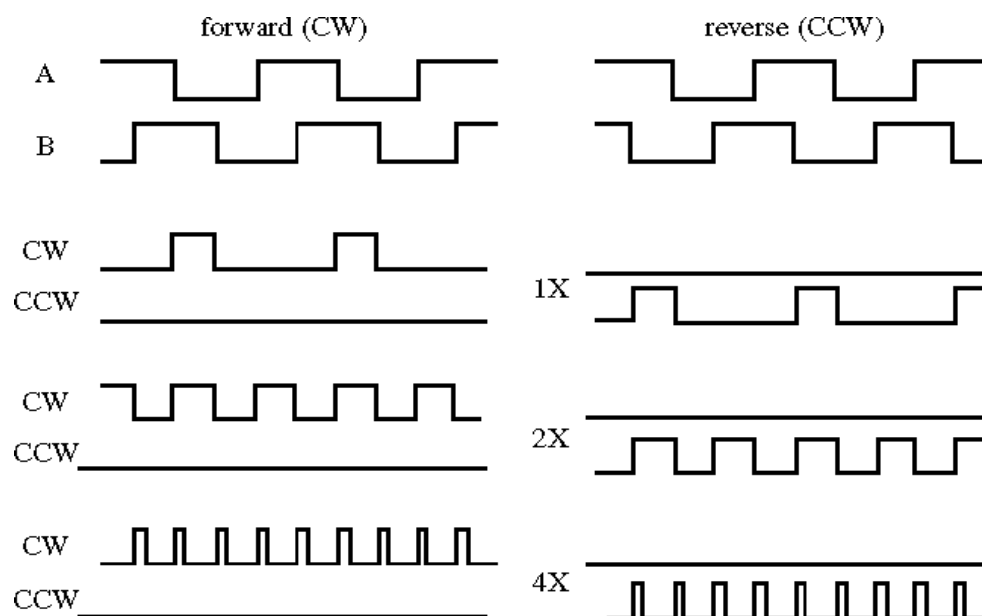


# Encoders - Incremental



# Encoders - Incremental

- Quadrature (resolution enhancing)



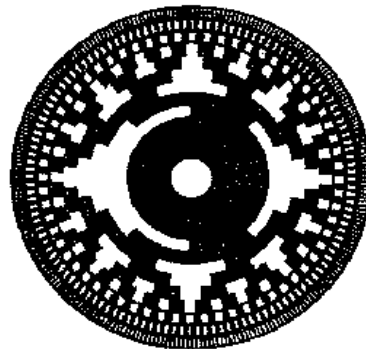
# Where are we?

- If we know our encoder values after the motion, do we know where we are?

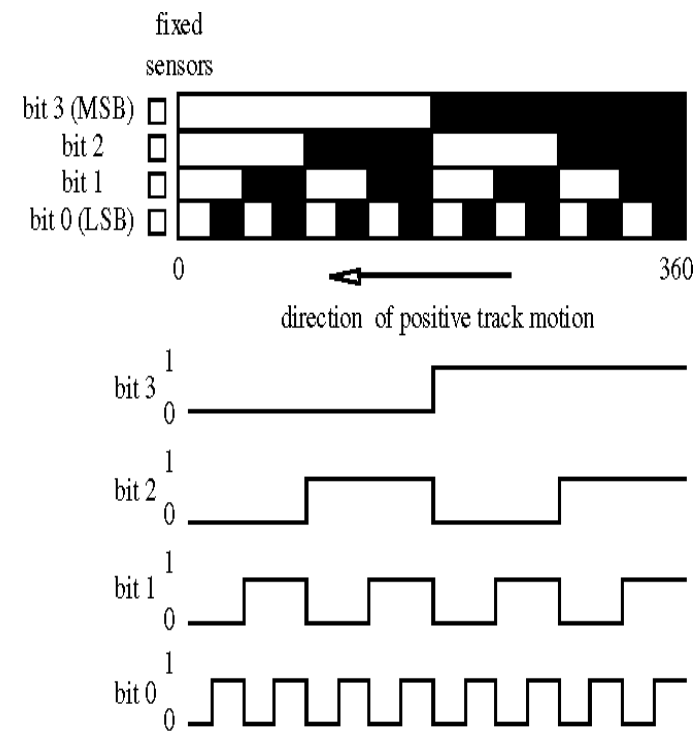
# Encoders - Absolute

- More expensive
- Resolution =  $360^\circ / 2^N$   
where N is number of tracks

## 4 Bit Example



(b) actual disk (Courtesy of Parker Compumotor Division, Rohnert Park, CA)

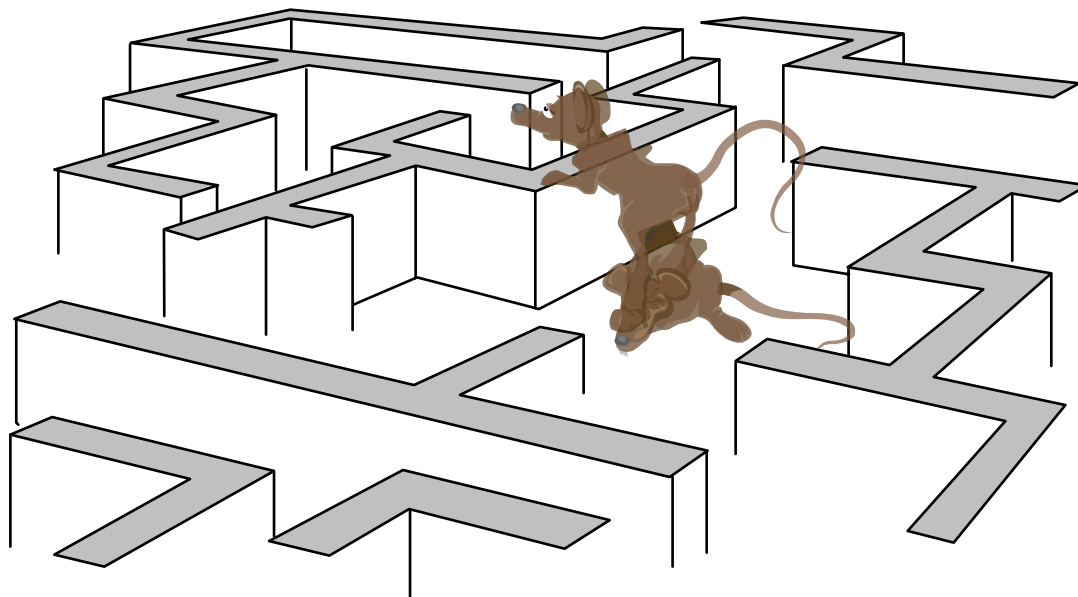


(a) schematic and signals



# What is Motion Planning?

- Determining where to go



# Overview

- The Basics
  - Motion Planning Statement
  - The World and Robot
  - Configuration Space
  - Metrics
- Path Planning Algorithms
  - Start-Goal Methods
  - Map-Based Approaches
  - Cellular Decompositions
- Applications
  - Navigating Large Spaces
  - Coverage

# The World consists of...

- Obstacles
  - Already occupied spaces of the world
  - In other words, robots can't go there
- Free Space
  - Unoccupied space within the world
  - Robots “might” be able to go here
  - To determine where a robot can go, we need to discuss what a *Configuration Space* is

## Motion Planning Statement

If  $W$  denotes the robot's workspace,

And  $C_i$  denotes the  $i$ 'th obstacle,

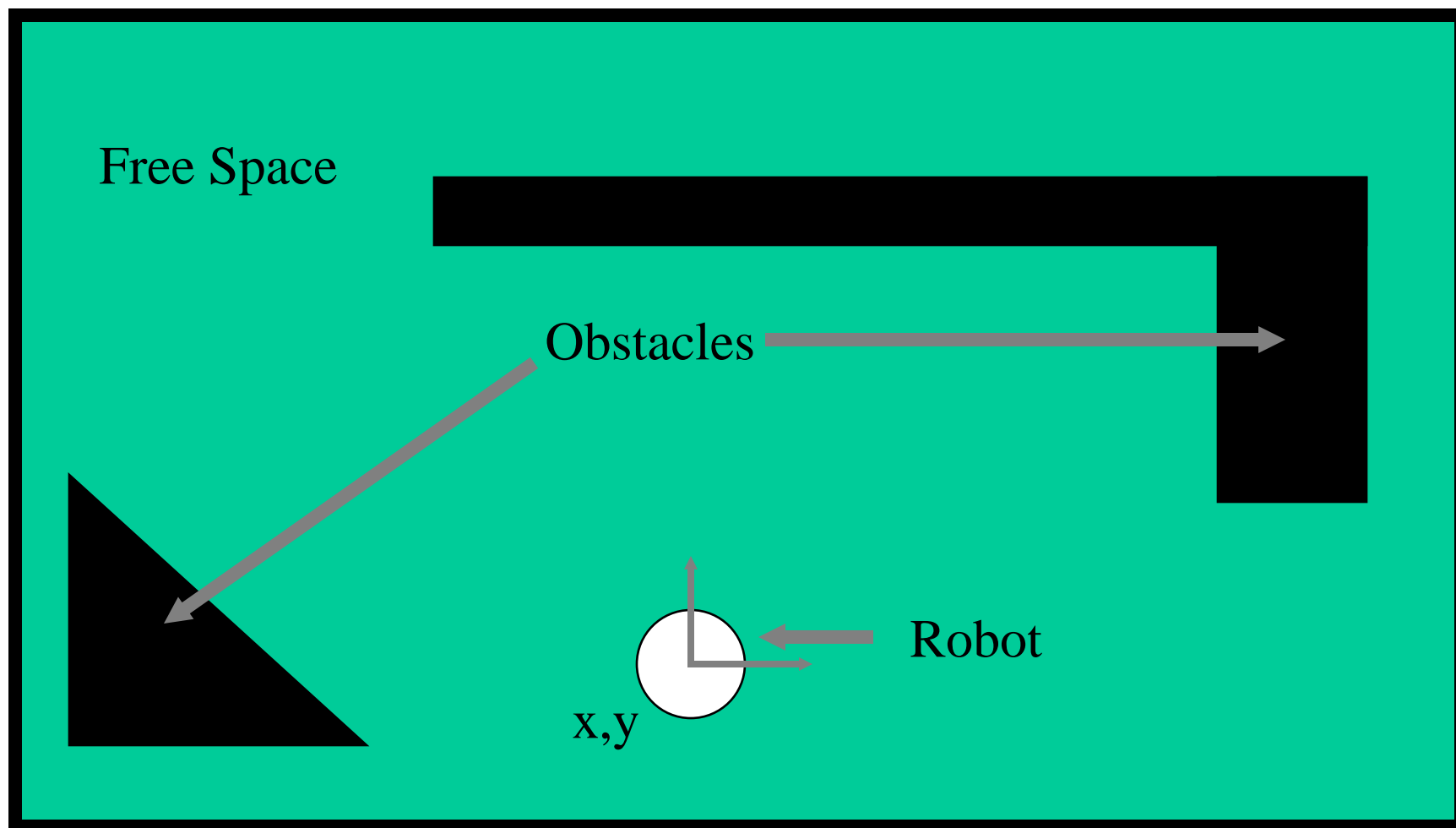
Then the robot's free space,  $FS$ , is defined as:

$$FS = W - (\cup C_i)$$

And a path  $c \in \mathcal{C}$  is  $c : [0,1] \rightarrow FS$

where  $c(0)$  is  $q_{start}$  and  $c(1)$  is  $q_{goal}$

## Example of a World (and Robot)



# What is a good path?

# Basics: Metrics

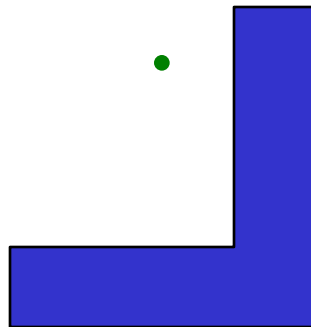
- There are many different ways to measure a path:
  - Time
  - Distance traveled
  - Expense
  - Distance from obstacles
  - Etc...



# Bug 1

But some computing power!

- known direction to goal
  - otherwise local sensing
- walls/obstacles & **encoders**



## "Bug 1" algorithm

- 1) head toward goal
- 2) if an obstacle is encountered, circumnavigate it *and* remember how close you get to the goal
- 3) return to that closest point (by wall-following) and continue



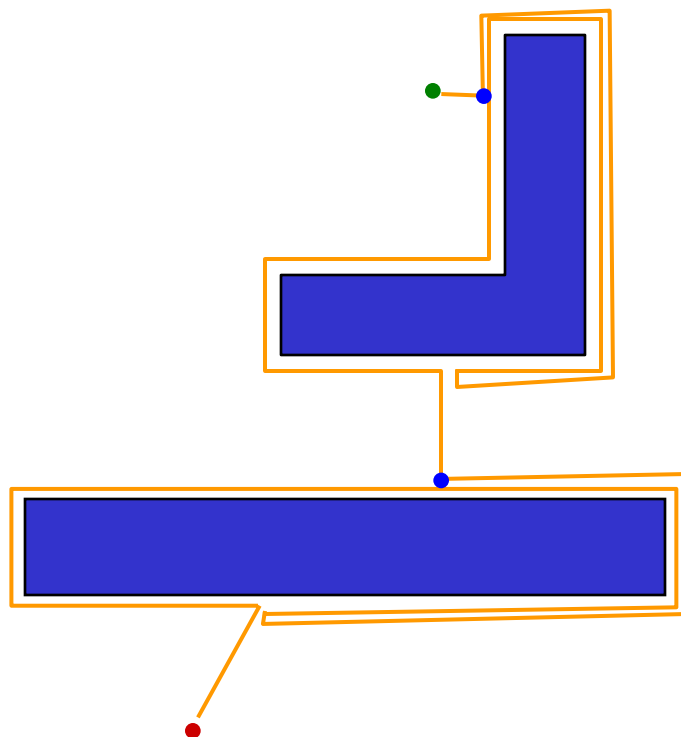


# Bug 1

But some computing power! {

- **known direction to goal**
- **otherwise local sensing**

walls/obstacles & **encoders**



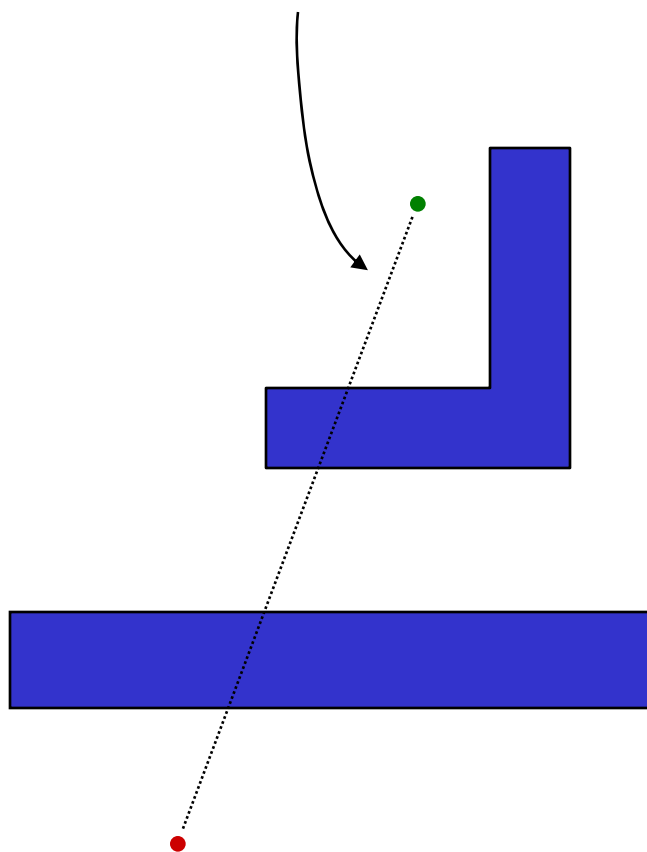
## "Bug 1" algorithm

- 1) head toward goal
- 2) if an obstacle is encountered, circumnavigate it *and* remember how close you get to the goal
- 3) return to that closest point (by wall-following) and continue

# Bug2

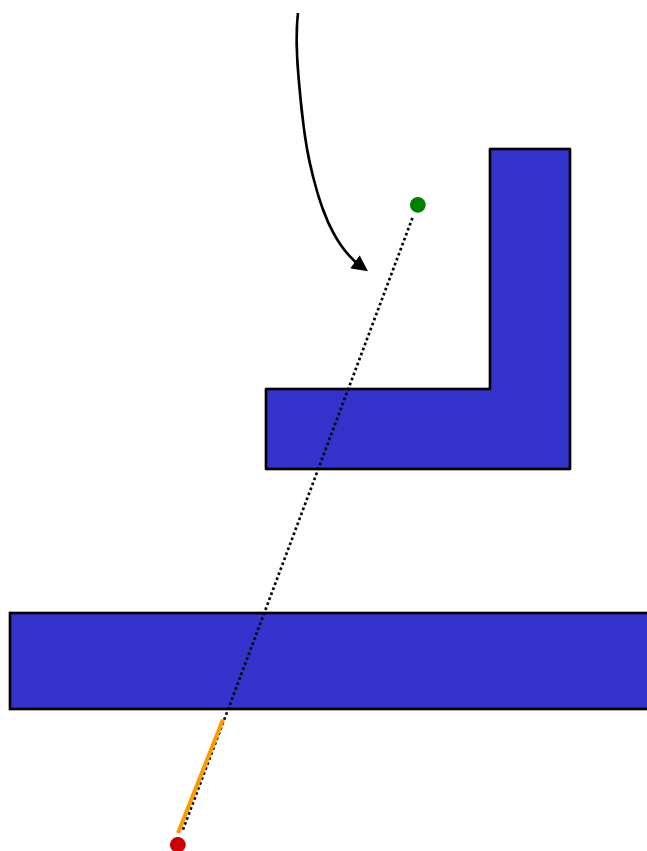
Call the line from the starting point to the goal the *m-line*

"Bug 2" Algorithm



# A better bug?

Call the line from the starting point to the goal the *m-line*

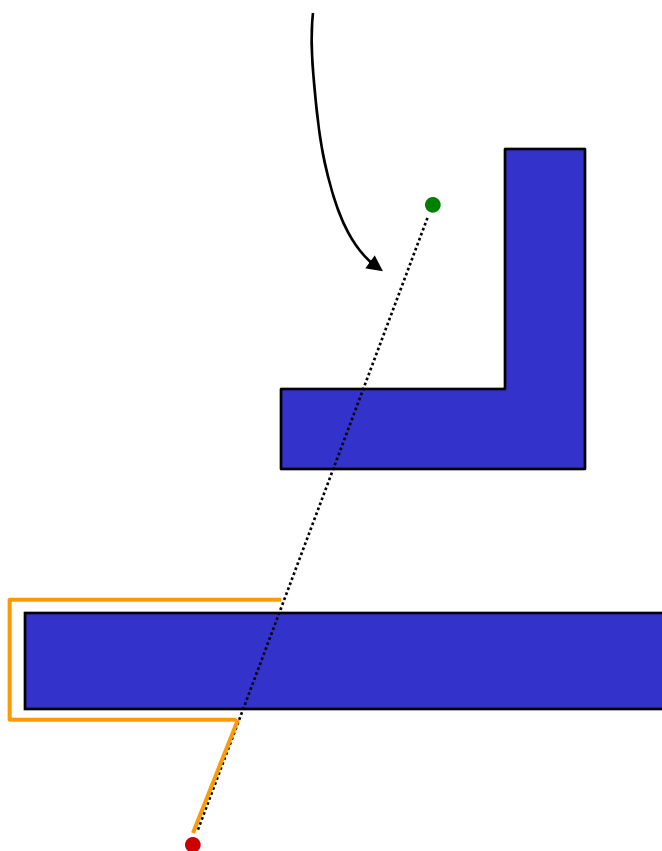


"Bug 2" Algorithm

1) head toward goal on the *m-line*

# A better bug?

Call the line from the starting point to the goal the *m-line*

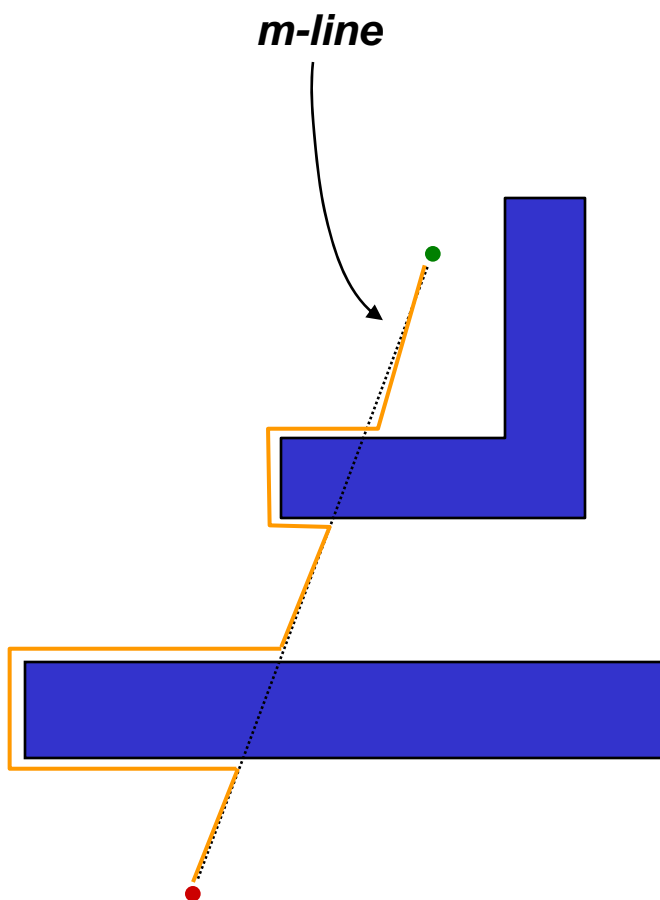


## "Bug 2" Algorithm

- 1) head toward goal on the *m-line*
- 2) if an obstacle is in the way, follow it until you encounter the *m-line* again.

# A better bug?

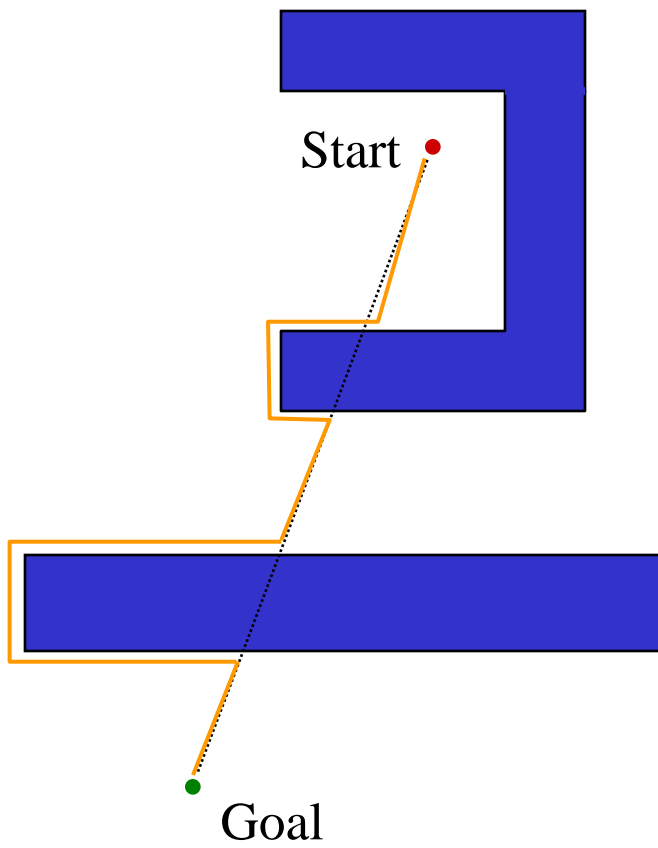
## "Bug 2" Algorithm



- 1) head toward goal on the *m-line*
- 2) if an obstacle is in the way, follow it until you encounter the *m-line* again.
- 3) Leave the obstacle and continue toward the goal

# A better bug?

## "Bug 2" Algorithm

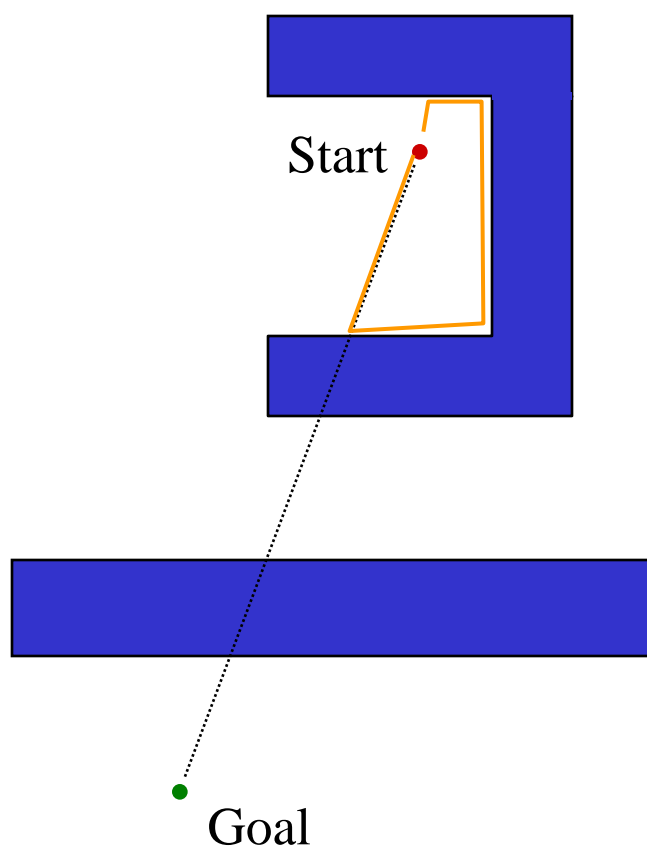


- 1) head toward goal on the *m-line*
- 2) if an obstacle is in the way, follow it until you encounter the *m-line* again.
- 3) Leave the obstacle and continue toward the goal

# Better or worse than Bug1?

# A better bug?

## "Bug 2" Algorithm



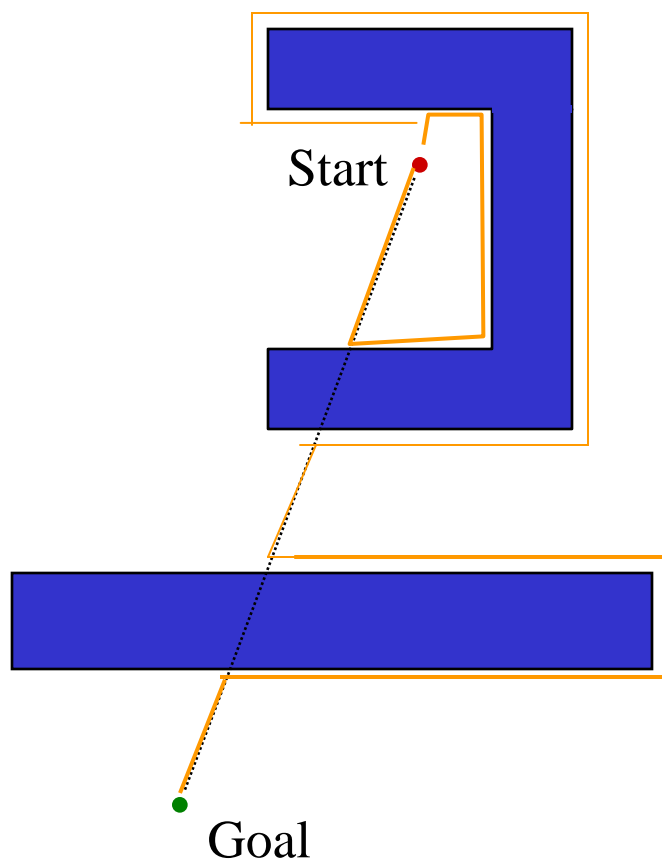
- 1) head toward goal on the *m-line*
- 2) if an obstacle is in the way, follow it until you encounter the *m-line* again.
- 3) Leave the obstacle and continue toward the goal

NO! How do we fix this?



# A better bug?

## "Bug 2" Algorithm

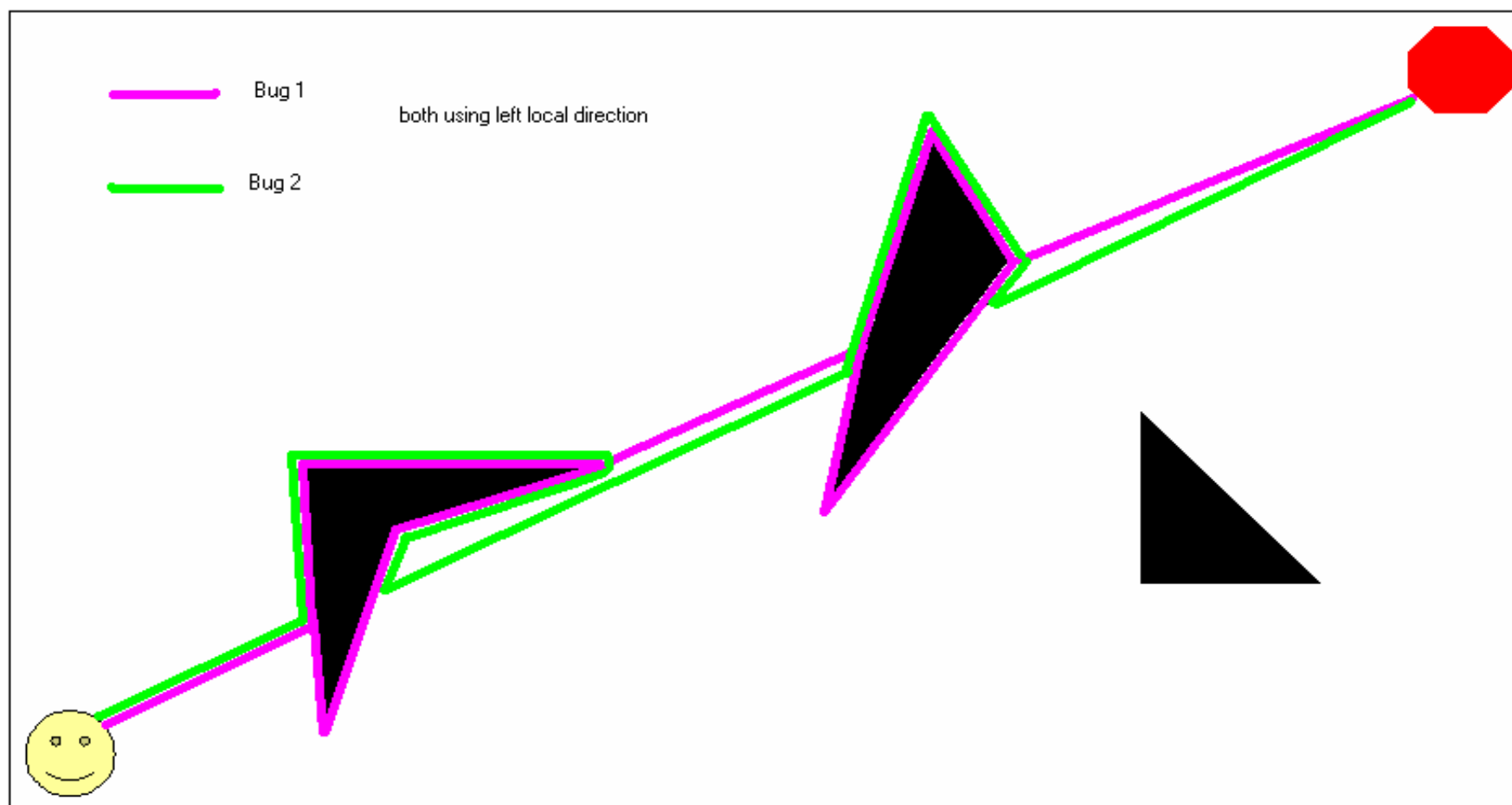


- 1) head toward goal on the *m-line*
- 2) if an obstacle is in the way, follow it until you encounter the m-line again *closer to the goal*.
- 3) Leave the obstacle and continue toward the goal

Better or worse than Bug1?



# Start-Goal Algorithm: Lumelsky Bug Algorithms



# Lumelsky Bug Algorithms

- Unknown obstacles, known start and goal.
- Simple “bump” sensors, encoders.
- Choose arbitrary direction to turn (left/right) to make all turns, called “local direction”
- Motion is like an ant walking around:
  - In Bug 1 the robot goes all the way around each obstacle encountered, recording the point nearest the goal, then goes around again to leave the obstacle from that point
  - In Bug 2 the robot goes around each obstacle encountered until it can continue on its previous path toward the goal

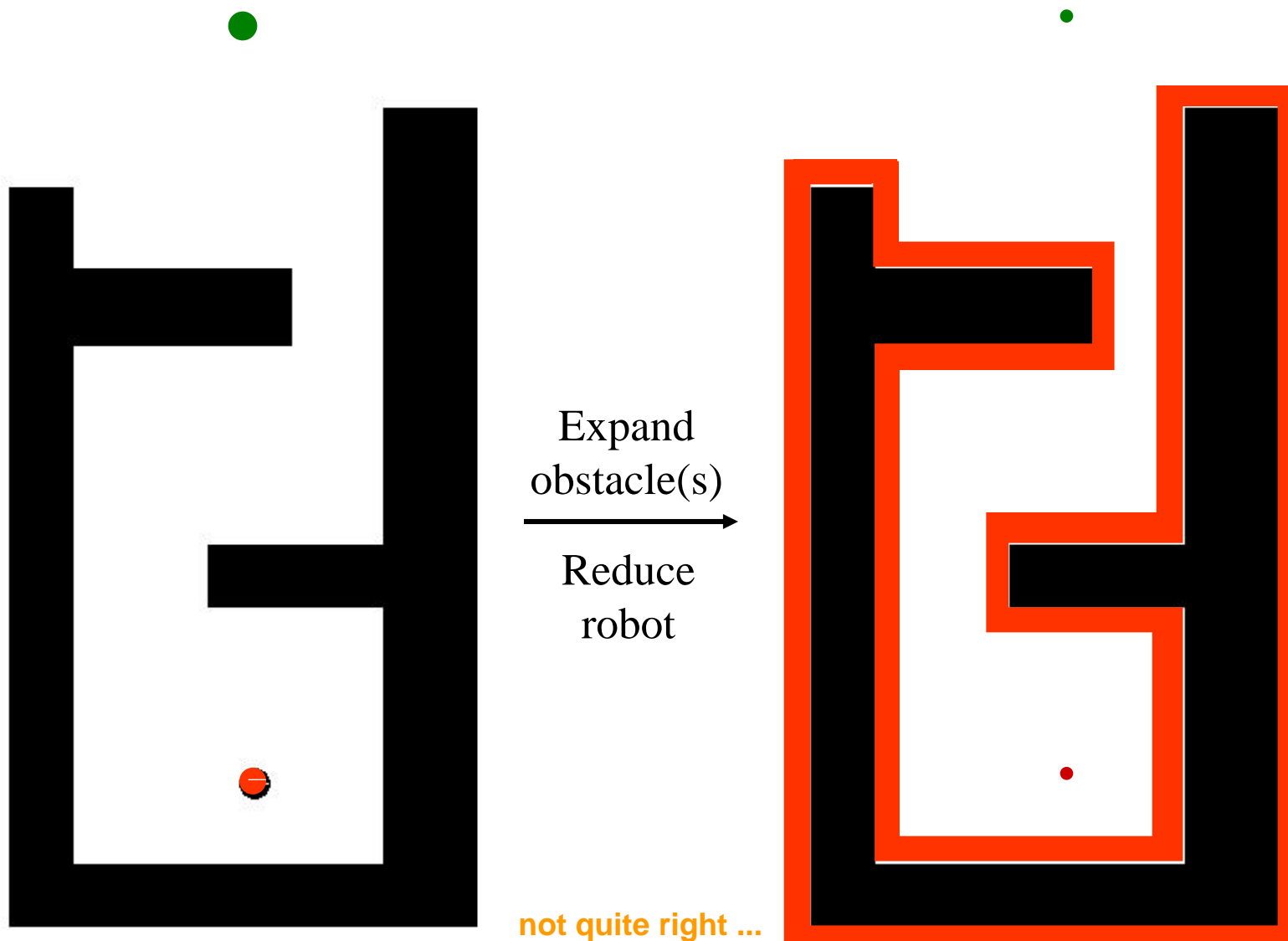
# Assumptions?

# Assumptions

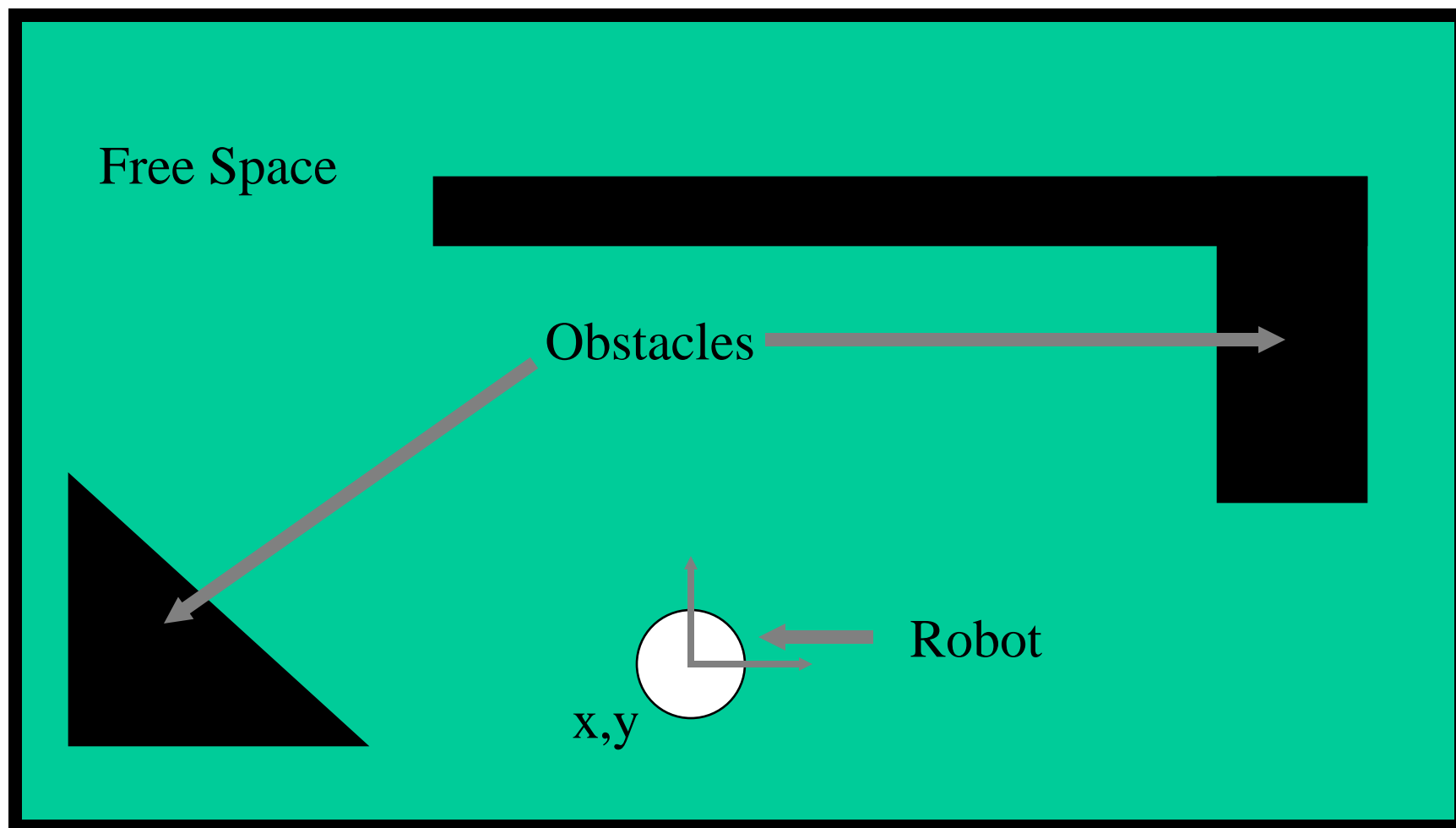
- Size of robot
- Perfect sensing
- Perfect control
- Localization (heading)

What else?

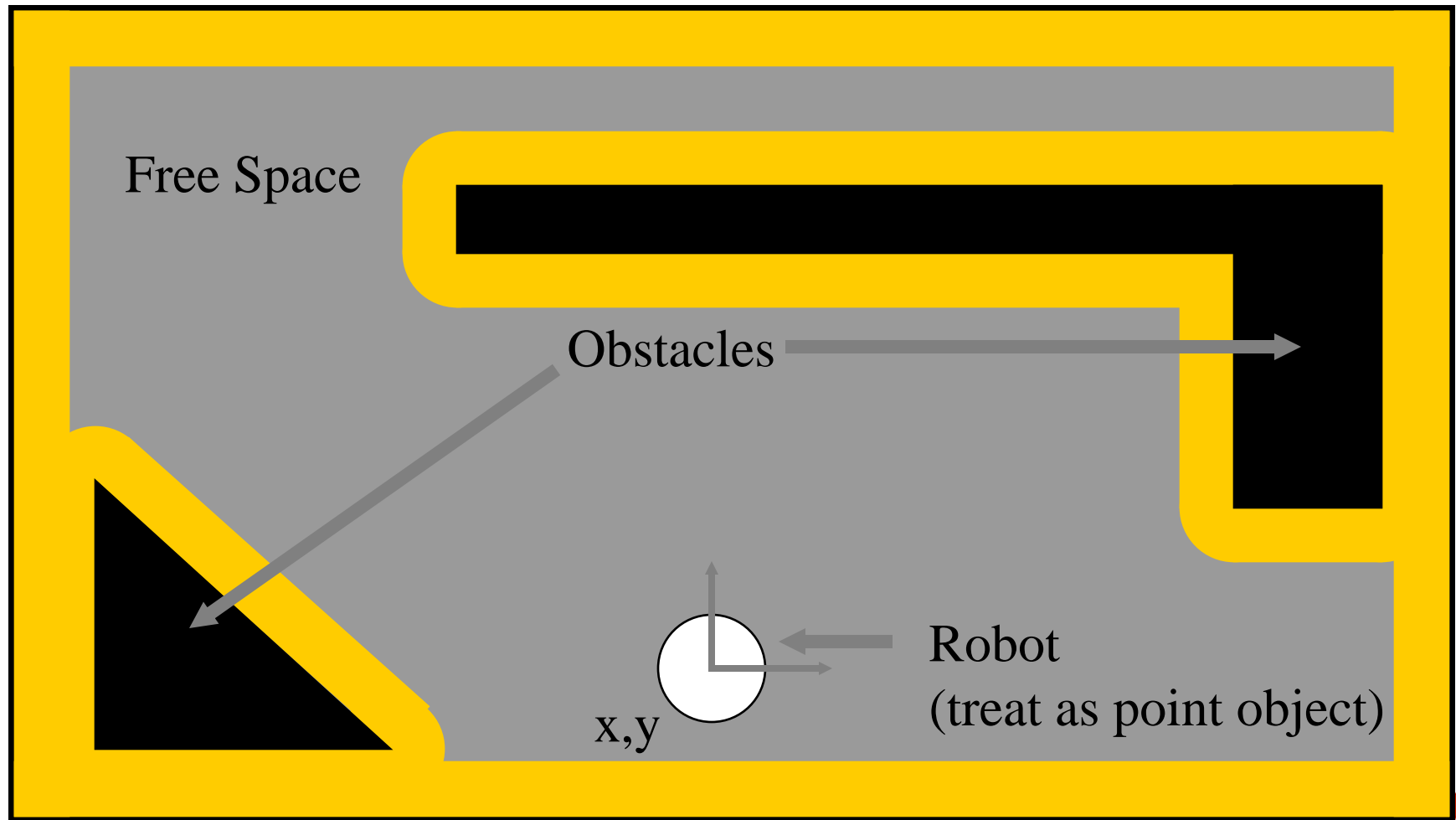
# What is the position of the robot?



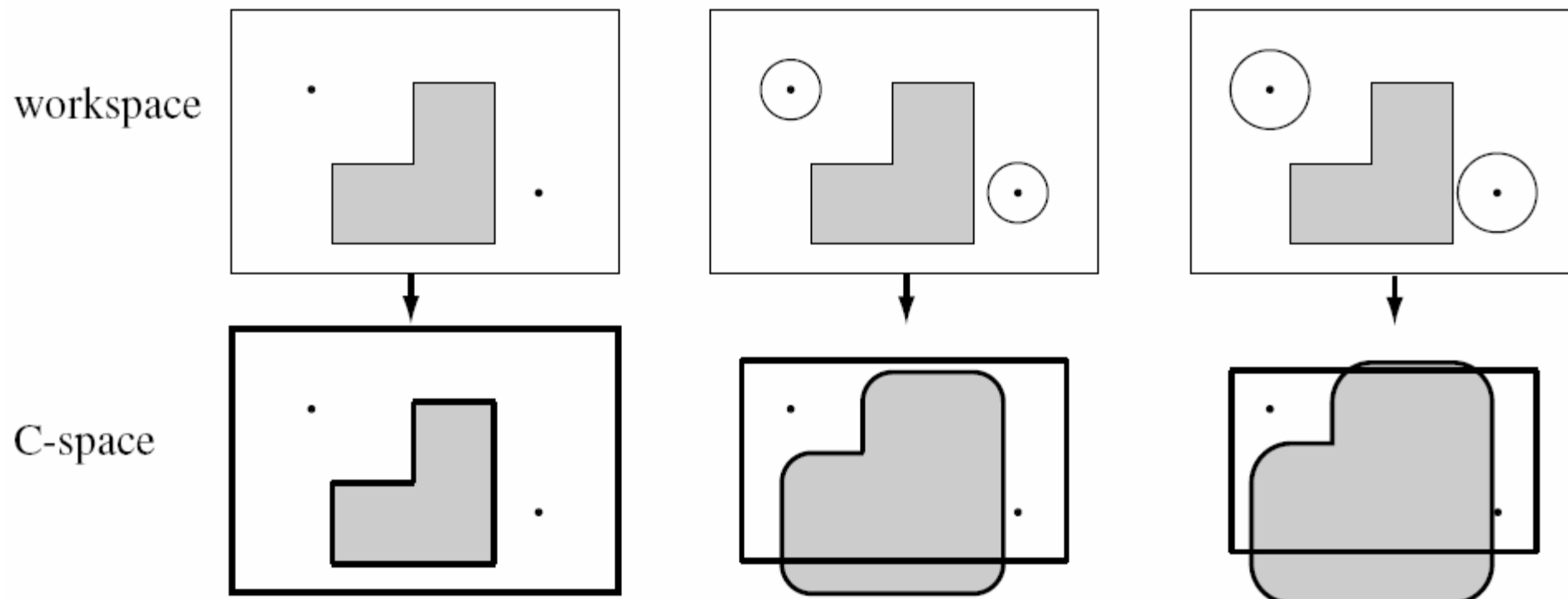
## Example of a World (and Robot)



# Configuration Space: Accommodate Robot Size



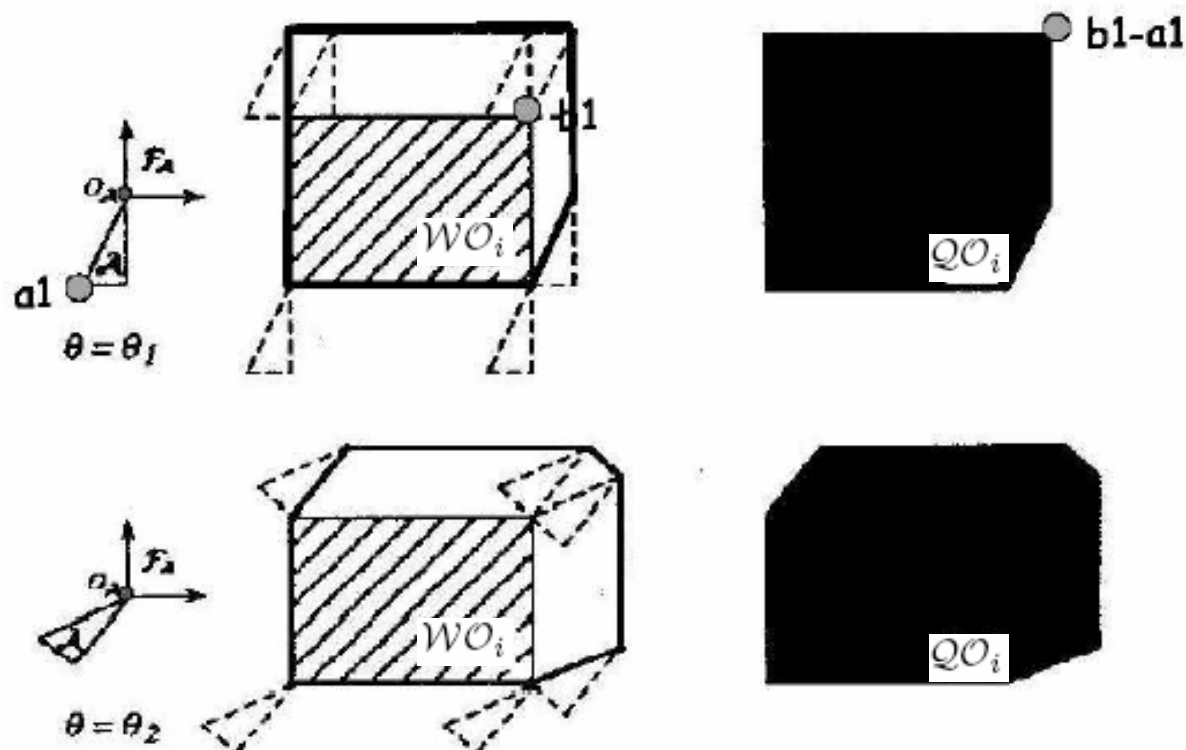
# Trace Boundary of Workspace



Pick a reference point...



# Translate-only, non-circularly



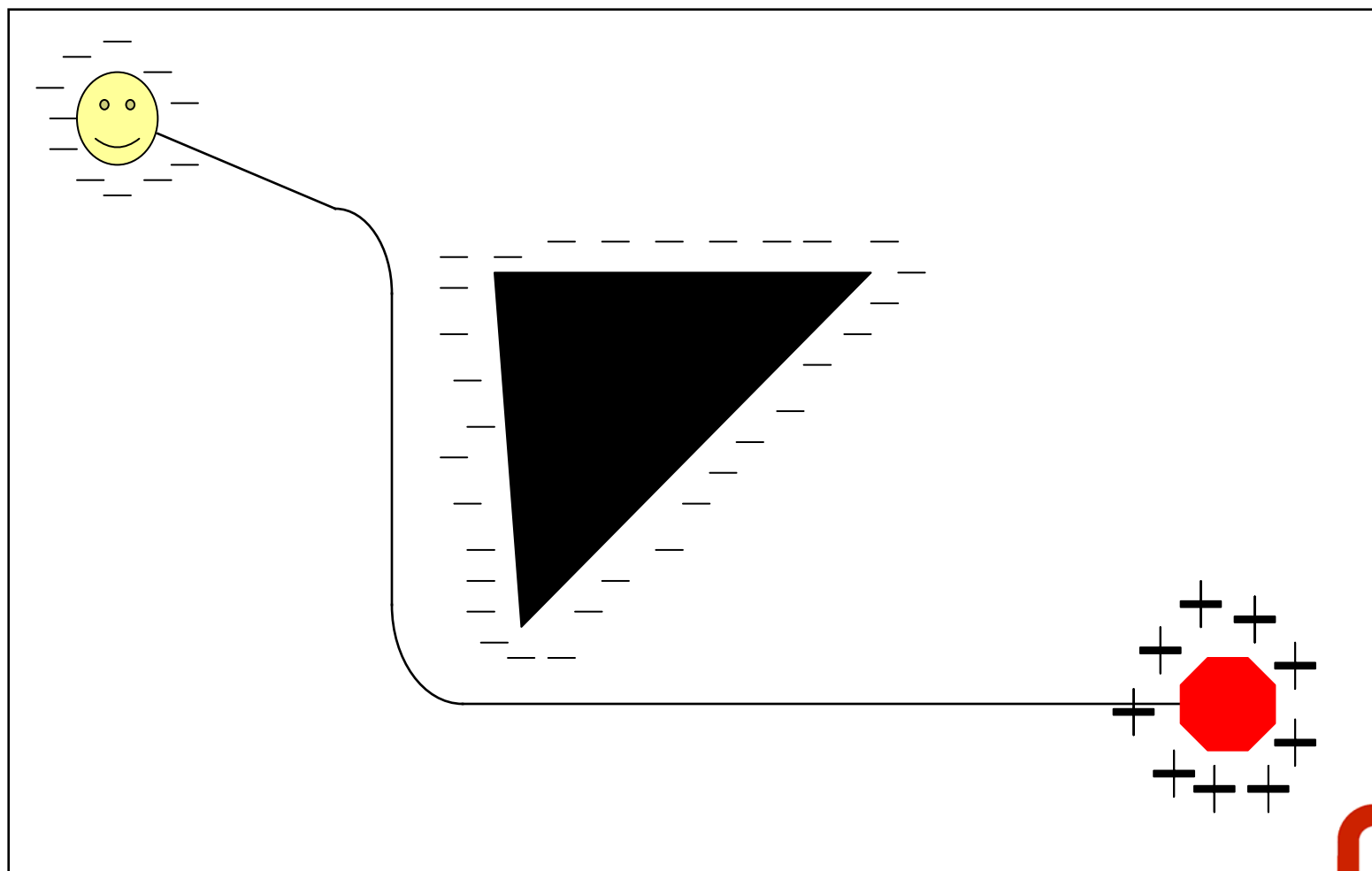
$$\mathcal{QO}_i = \{q \in \mathcal{Q} \mid R(q) \cap \mathcal{WO}_i \neq \emptyset\}.$$

Pick a reference point...

# The Configuration Space

- What it is
  - A set of “reachable” areas constructed from knowledge of both the robot and the world
- How to create it
  - First abstract the robot as a point object. Then, enlarge the obstacles to account for the robot’s footprint and degrees of freedom
  - In our example, the robot was circular, so we simply enlarged our obstacles by the robot’s radius (*note the curved vertices*)

# Start-Goal Algorithm: Potential Functions



# Attractive/Repulsive Potential Field

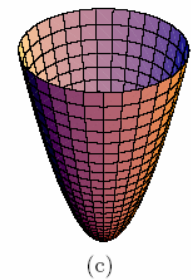
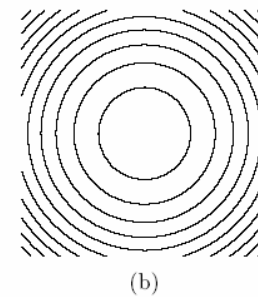
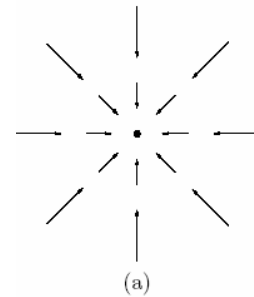
$$U(q) = U_{\text{att}}(q) + U_{\text{rep}}(q)$$

- $U_{\text{att}}$  is the “attractive” potential --- move to the goal
- $U_{\text{rep}}$  is the “repulsive” potential --- avoid obstacles

# Artificial Potential Field Methods: Attractive Potential

Quadratic Potential →

$$U_{\text{att}}(q) = \frac{1}{2}\zeta d^2(q, q_{\text{goal}}),$$



$$\begin{aligned} F_{\text{att}}(q) &= \nabla U_{\text{att}}(q) = \nabla \left( \frac{1}{2}\zeta d^2(q, q_{\text{goal}}) \right), \\ &= \frac{1}{2}\zeta \nabla d^2(q, q_{\text{goal}}), \\ &= \zeta(q - q_{\text{goal}}), \end{aligned}$$

# Distance

$$d : R^2 \times R^2 \rightarrow R$$

L1 Metric	(diamond)	$d(a, b) =  a_x - b_x  +  a_y - b_y $
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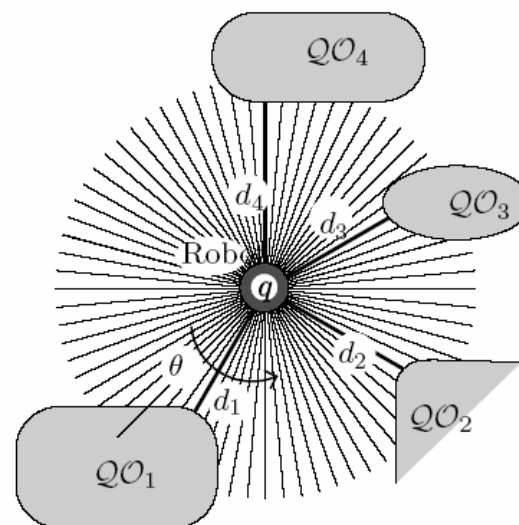
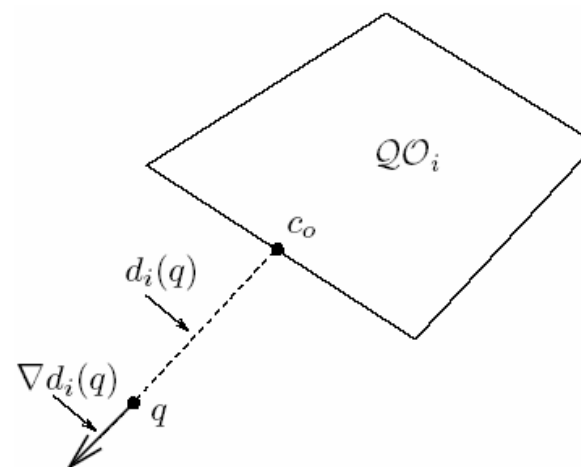
L2 Metric	(circle)	$d(a, b) = \sqrt{(a_x - b_x)^2 + (a_y - b_y)^2}$
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# Distance to Obstacle(s)

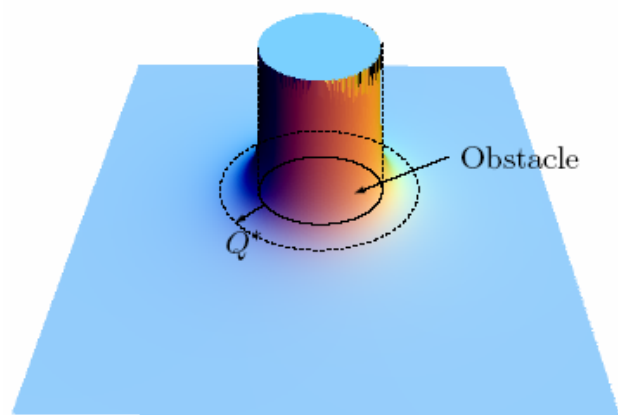
$$d_i(q) = \min_{c \in QO_i} d(q, c).$$

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

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



# The Repulsive Potential



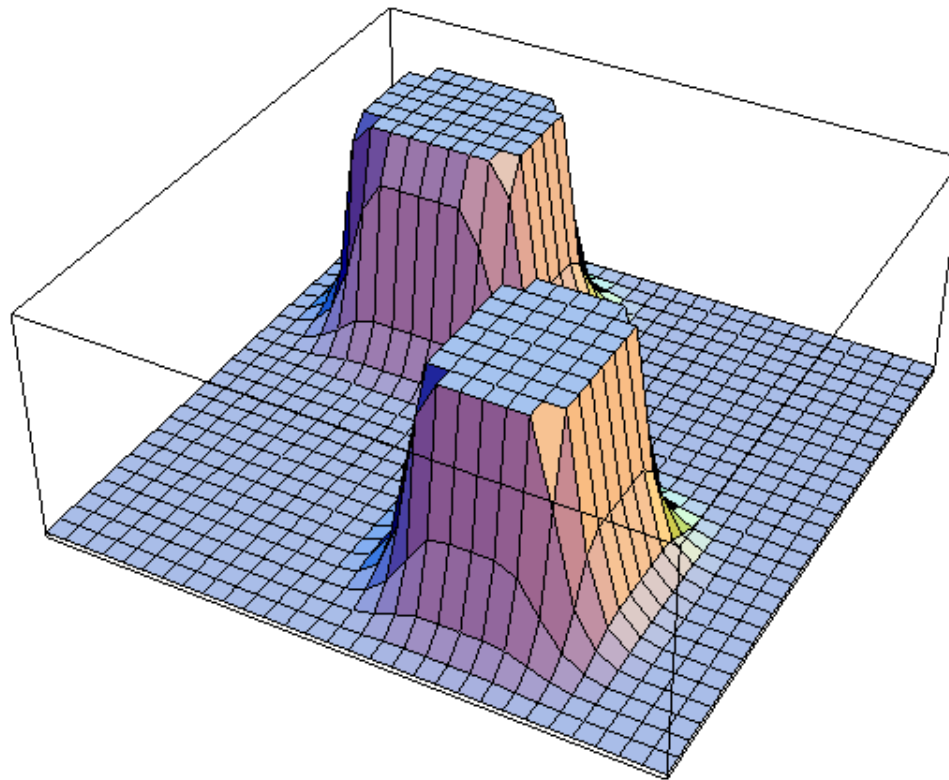
$$U_{\text{rep}}(q) = \begin{cases} \frac{1}{2}\eta\left(\frac{1}{D(q)} - \frac{1}{Q^*}\right)^2, & D(q) \leq Q^*, \\ 0, & D(q) > Q^*, \end{cases}$$

whose gradient is

$$\nabla U_{\text{rep}}(q) = \begin{cases} \eta \left( \frac{1}{Q^*} - \frac{1}{D(q)} \right) \frac{1}{D^2(q)} \nabla D(q), & D(q) \leq Q^*, \\ 0, & D(q) > Q^*, \end{cases}$$



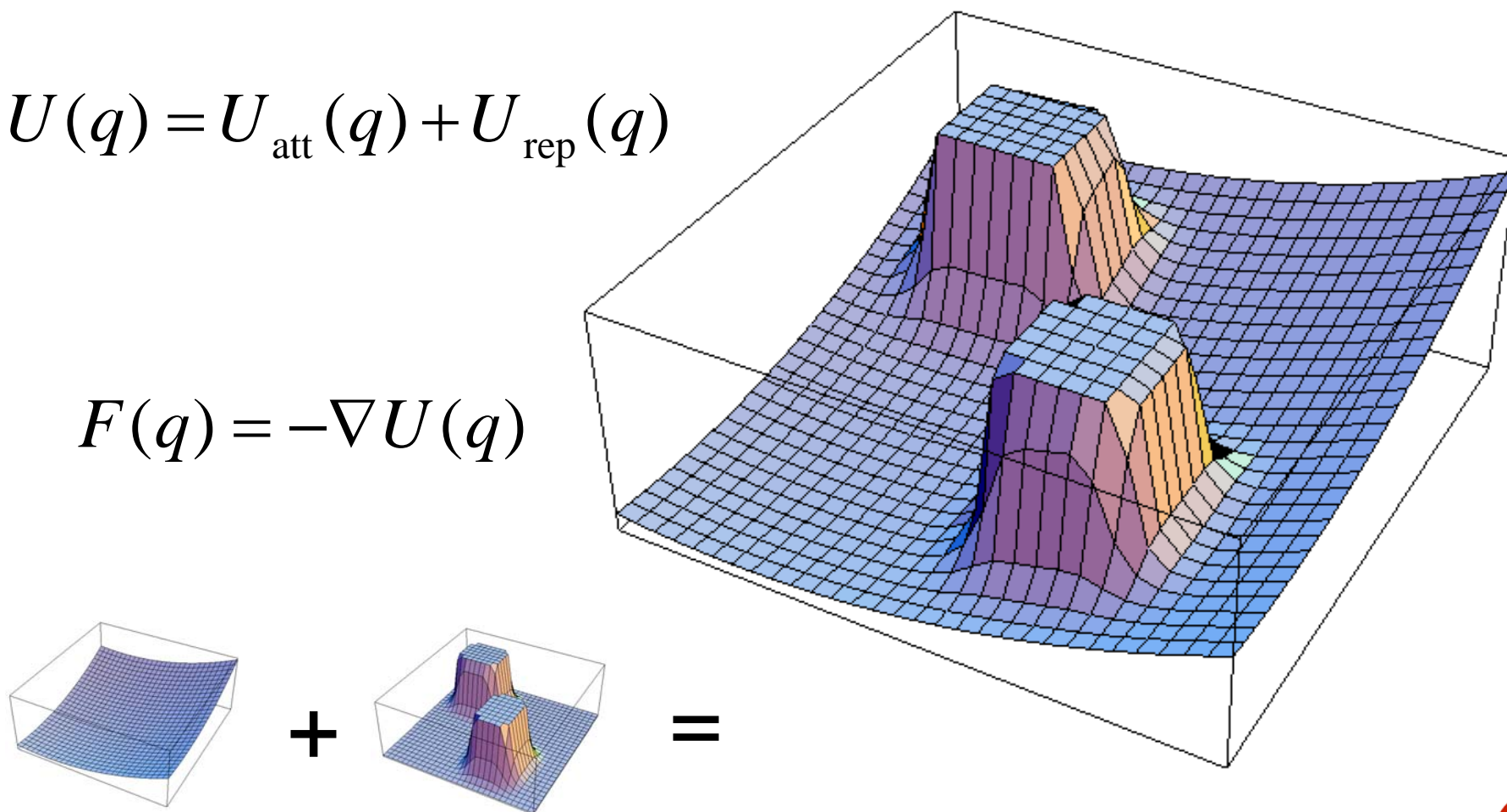
# Repulsive Potential



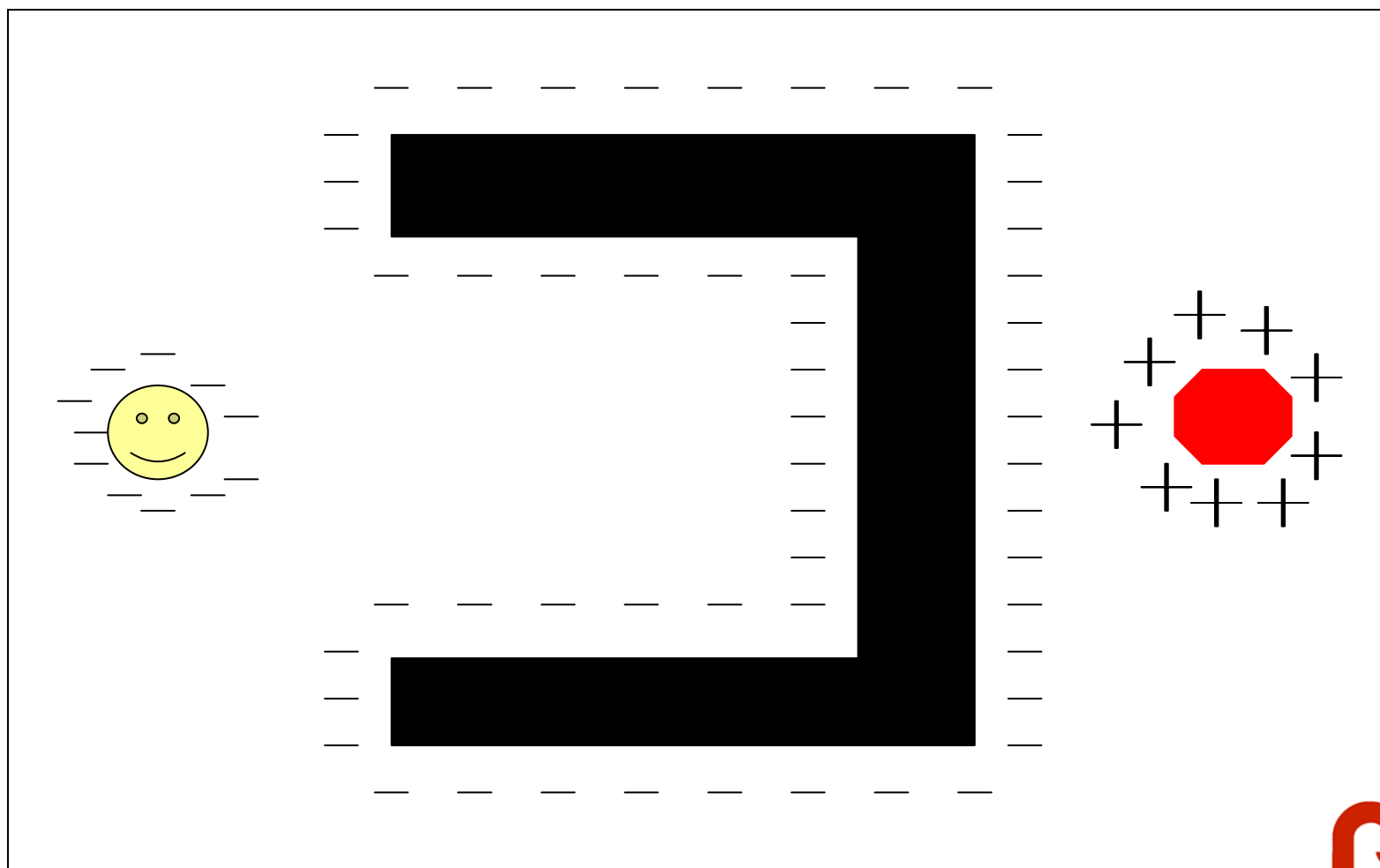
# Total Potential Function

$$U(q) = U_{\text{att}}(q) + U_{\text{rep}}(q)$$

$$F(q) = -\nabla U(q)$$



# Local Minimum Problem with the Charge Analogy

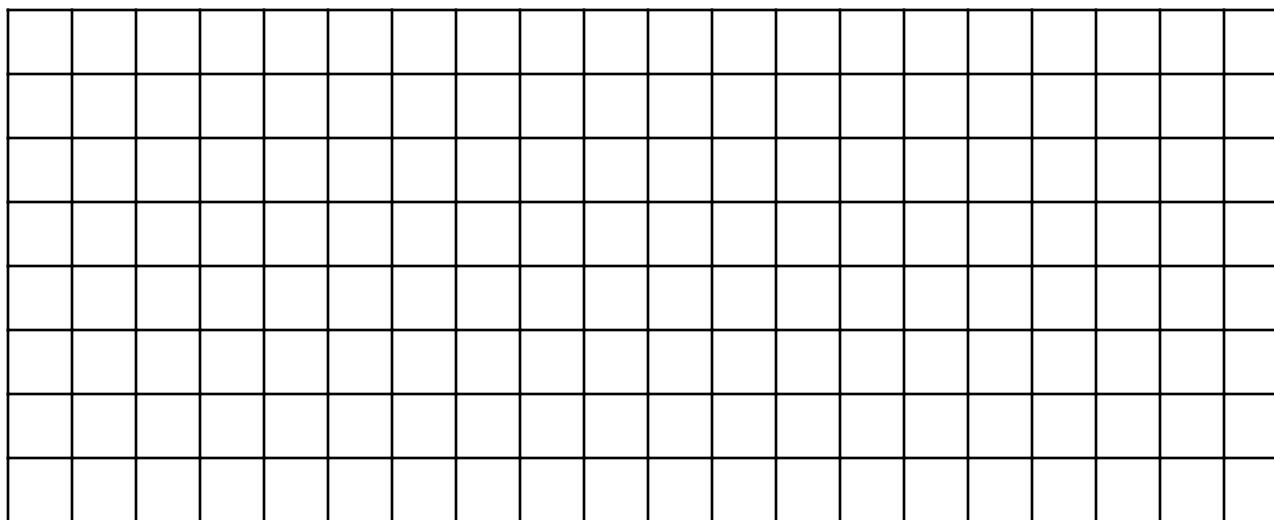


# The Wavefront Planner

- A common algorithm used to determine the shortest paths between two points
  - In essence, a breadth first search of a graph
- For simplification, we'll present the world as a two-dimensional grid
- Setup:
  - Label free space with 0
  - Label start as START
  - Label the destination as 2

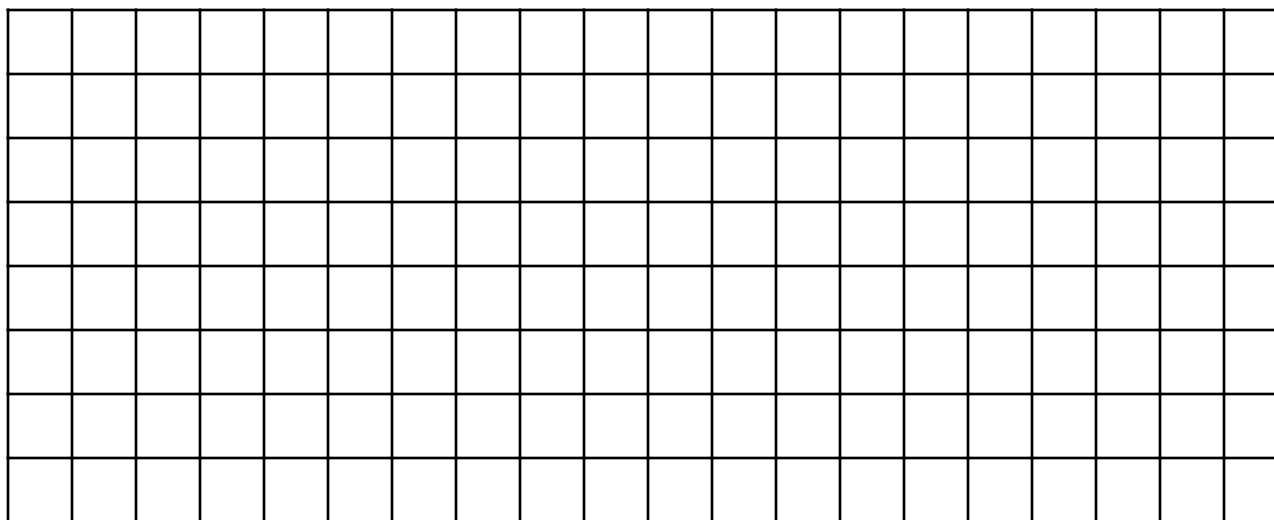
# Representations

- World Representation
  - You could always use a large region and distances
  - However, a grid can be used for simplicity



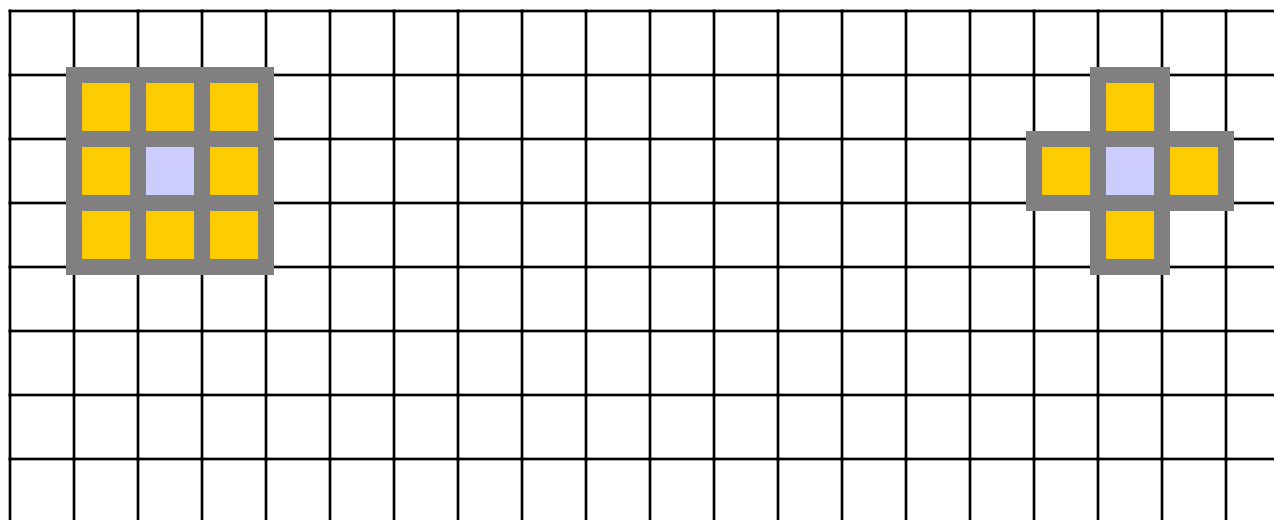
# Representations: A Grid

- Distance is reduced to discrete steps
  - For simplicity, we'll assume distance is uniform
- Direction is now limited from one adjacent cell to another
  - Time to revisit Connectivity (Remember Vision?)



# Representations: Connectivity

- 8-Point Connectivity
- 4-Point Connectivity
  - (approximation of the  $L1$  metric)



# The Wavefront Planner: Setup

7	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
6	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
5	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
4	0	0	0	0	1	1	1	1	1	1	1	1	0	0	0	
3	0	0	0	0	1	1	1	1	1	1	1	1	0	0	0	
2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	2	
	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15



# The Wavefront in Action (Part 1)

- Starting with the goal, set all adjacent cells with “0” to the current cell + 1
  - 4-Point Connectivity or 8-Point Connectivity?
  - Your Choice. We'll use 8-Point Connectivity in our example

7	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
6	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
5	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
4	0	0	0	0	1	1	1	1	1	1	1	1	0	0	0	
3	0	0	0	0	1	1	1	1	1	1	1	1	0	0	0	
2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
1	0	0	0	0	0	0	0	0	0	0	0	0	0	3	3	
0	0	0	0	0	0	0	0	0	0	0	0	0	0	3	2	
	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

# The Wavefront in Action (Part 2)

- Now repeat with the modified cells
  - This will be repeated until no 0's are adjacent to cells with values  $\geq 2$ 
    - 0's will only remain when regions are unreachable

7	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
6	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
5	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
4	0	0	0	0	1	1	1	1	1	1	1	1	0	0	0
3	0	0	0	0	1	1	1	1	1	1	1	1	0	0	0
2	0	0	0	0	0	0	0	0	0	0	0	0	0	4	4
1	0	0	0	0	0	0	0	0	0	0	0	0	0	4	3
0	0	0	0	0	0	0	0	0	0	0	0	0	0	4	2
	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14

# The Wavefront in Action (Part 3)

- Repeat again...

7	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
6	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
5	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
4	0	0	0	0	1	1	1	1	1	1	1	1	0	0	0
3	0	0	0	0	1	1	1	1	1	1	1	1	5	5	5
2	0	0	0	0	0	0	0	0	0	0	0	0	5	4	4
1	0	0	0	0	0	0	0	0	0	0	0	0	5	4	3
0	0	0	0	0	0	0	0	0	0	0	0	0	5	4	2
	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14

# The Wavefront in Action (Part 4)

- And again...

7	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
6	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
5	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
4	0	0	0	0	1	1	1	1	1	1	1	1	6	6	6
3	0	0	0	0	1	1	1	1	1	1	1	1	5	5	5
2	0	0	0	0	0	0	0	0	0	0	0	6	5	4	4
1	0	0	0	0	0	0	0	0	0	0	0	6	5	4	3
0	0	0	0	0	0	0	0	0	0	0	0	6	5	4	2
	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14

# The Wavefront in Action (Part 5)

- And again until...

7	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
6	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
5	0	0	0	0	0	0	0	0	0	0	7	7	7	7	7
4	0	0	0	0	1	1	1	1	1	1	1	1	6	6	6
3	0	0	0	0	1	1	1	1	1	1	1	1	5	5	5
2	0	0	0	0	0	0	0	0	0	0	7	6	5	4	4
1	0	0	0	0	0	0	0	0	0	0	7	6	5	4	3
0	0	0	0	0	0	0	0	0	0	0	7	6	5	4	2
	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14

# The Wavefront in Action (Done)

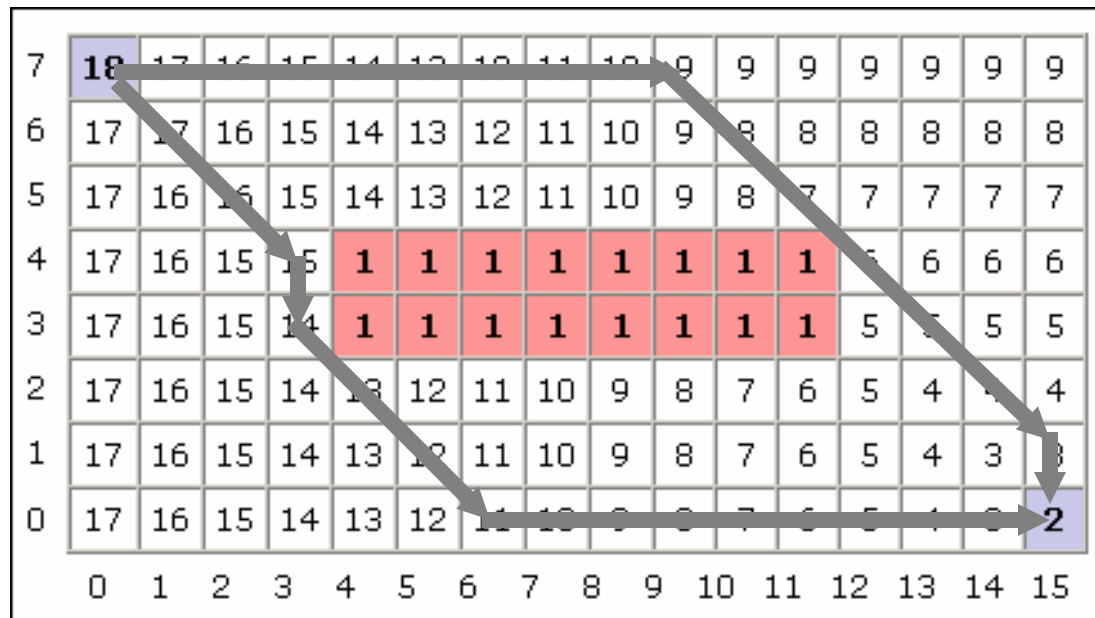
- You're done
  - Remember, 0's should only remain if unreachable regions exist

7	18	17	16	15	14	13	12	11	10	9	9	9	9	9	9
6	17	17	16	15	14	13	12	11	10	9	8	8	8	8	8
5	17	16	16	15	14	13	12	11	10	9	8	7	7	7	7
4	17	16	15	15	1	1	1	1	1	1	1	1	6	6	6
3	17	16	15	14	1	1	1	1	1	1	1	1	5	5	5
2	17	16	15	14	13	12	11	10	9	8	7	6	5	4	4
1	17	16	15	14	13	12	11	10	9	8	7	6	5	4	3
0	17	16	15	14	13	12	11	10	9	8	7	6	5	4	2
	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14

# The Wavefront, Now What?

- To find the shortest path, according to your metric, simply always move toward a cell with a lower number
  - The numbers generated by the Wavefront planner are roughly proportional to their distance from the goal

Two  
possible  
shortest  
paths  
shown



# Wavefront (Overview)

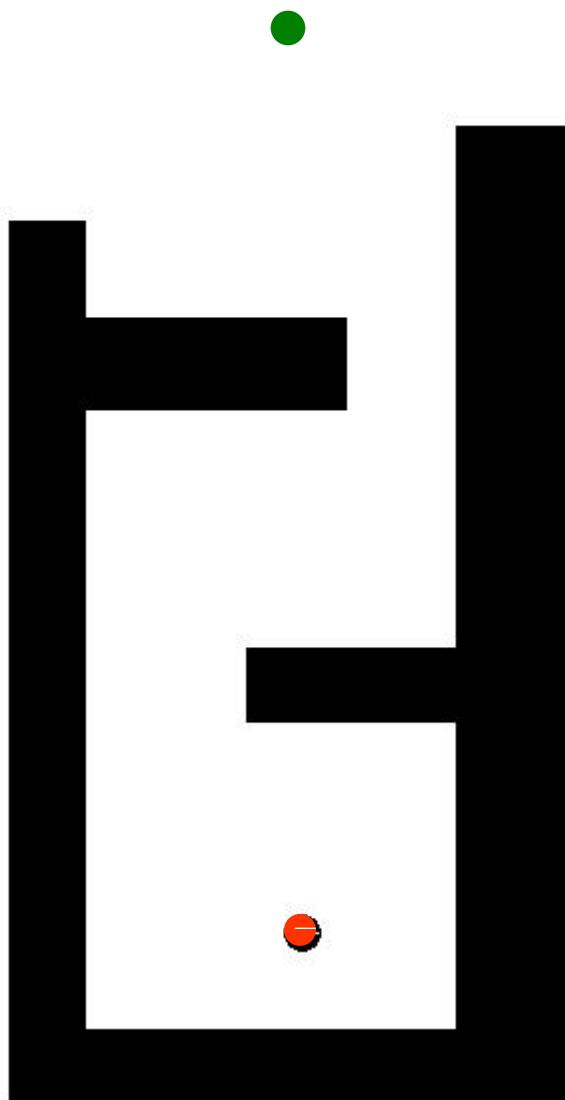
- Divide the space into a grid.
- Number the squares starting at the start in either 4 or 8 point connectivity starting at the goal, increasing till you reach the start.
- Your path is defined by any uninterrupted sequence of decreasing numbers that lead to the goal.



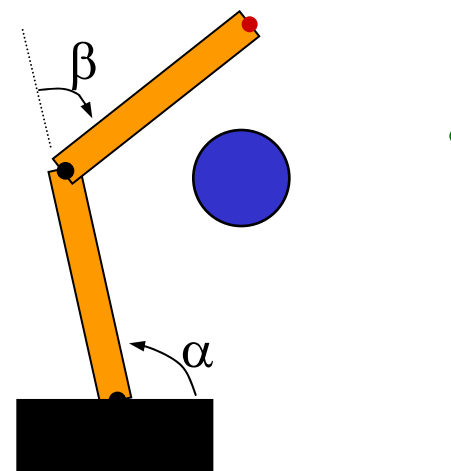
# Return to Configuration Spaces

- Non-Euclidean
- Non-Planar

# What if the robot is not a point?



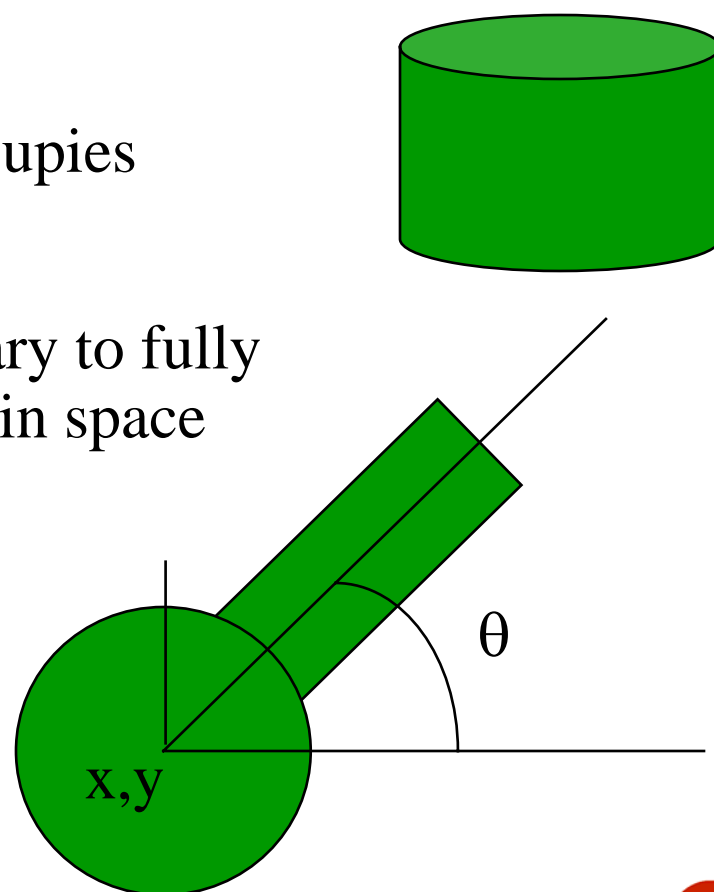
The Scout should probably not be modeled as a point...



Nor should robots with extended linkages that may contact obstacles...

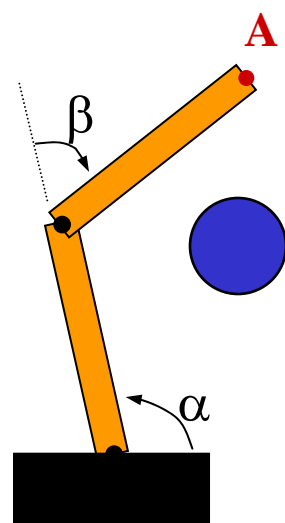
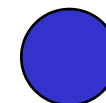
# Configuration Space: the robot has...

- A Footprint
  - The amount of space a robot occupies
- Degrees of Freedom
  - The number of variables necessary to fully describe a robot's configuration in space
    - You'll cover this more in depth later
    - *fun with non-holonomic constraints, etc*

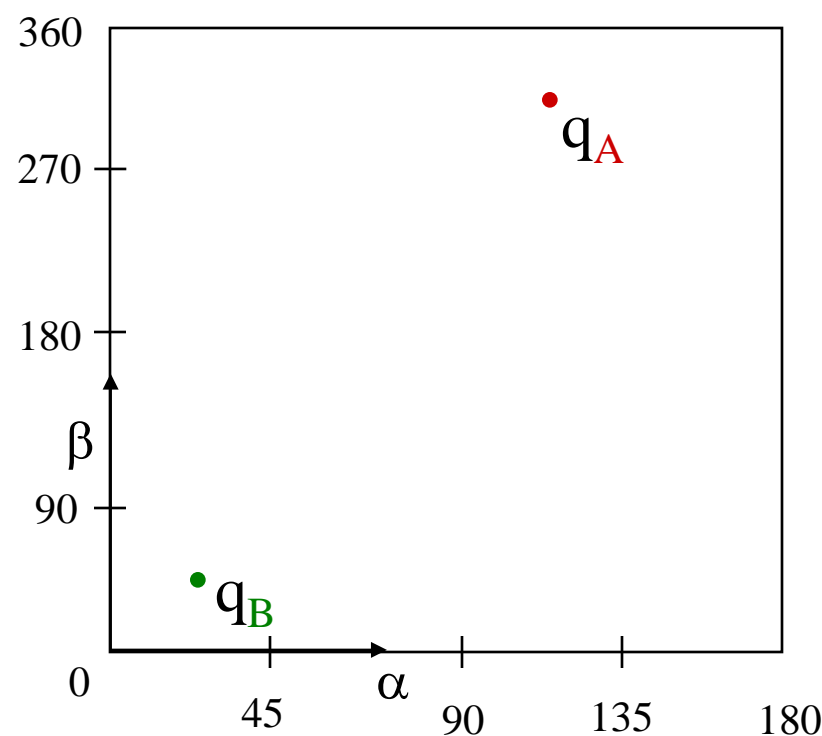


# Configuration Space “Quiz”

Where do we put ?



An obstacle in the robot's workspace

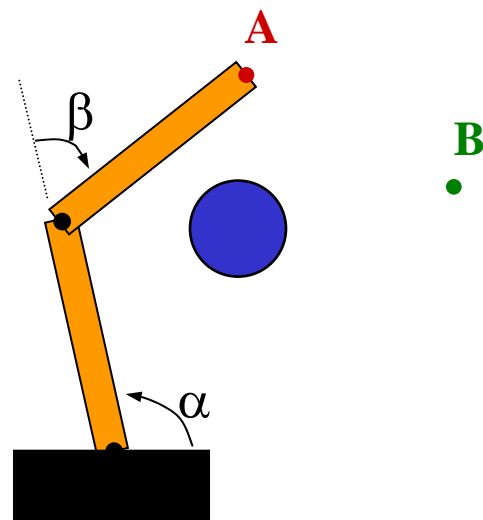


Torus

(wraps horizontally and vertically)

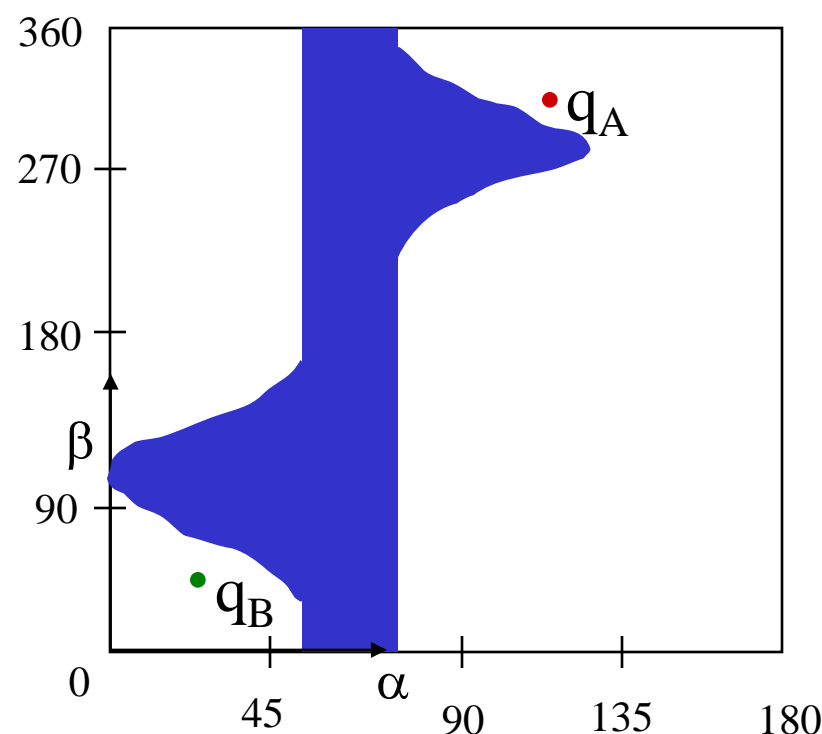
# Configuration Space Obstacle

Reference configuration



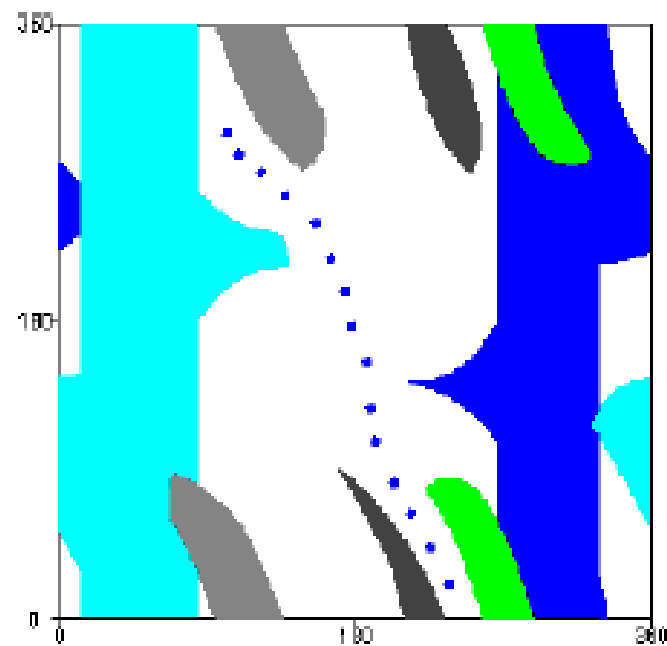
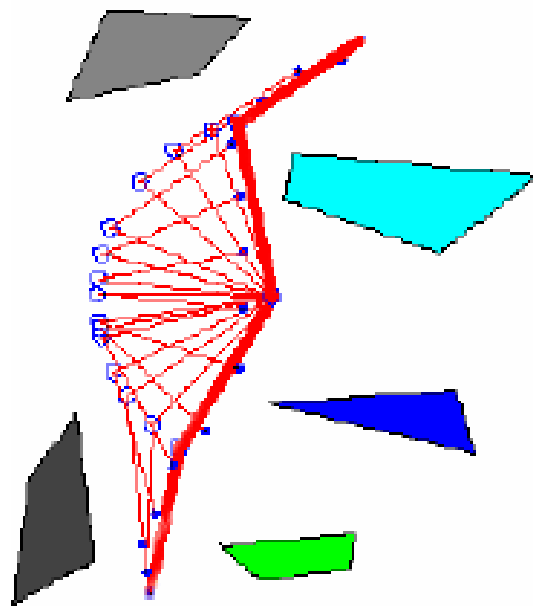
An obstacle in the robot's workspace

How do we get from A to B ?



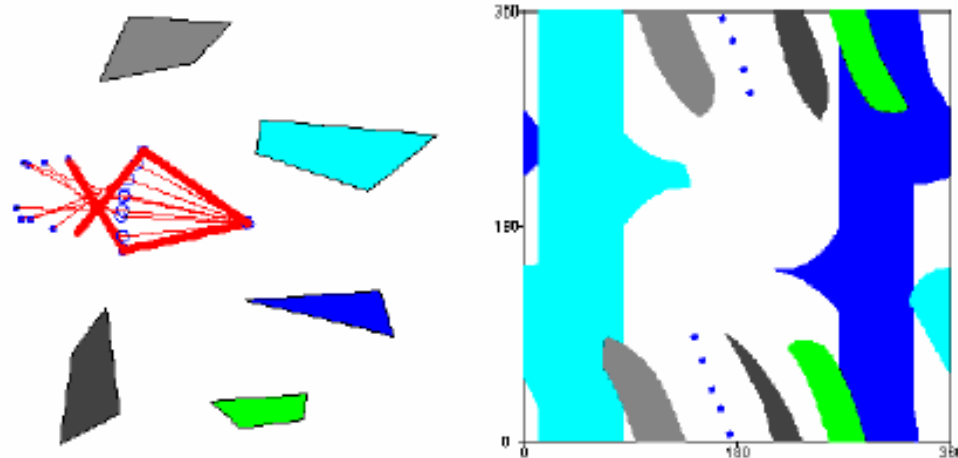
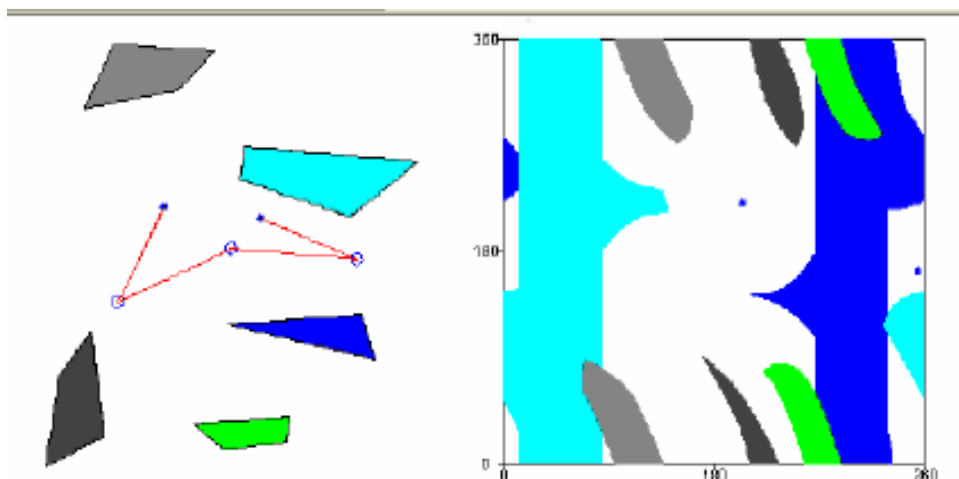
The C-space representation  
of this obstacle...

# Two Link Path



Thanks to Ken Goldberg

# Two Link Path



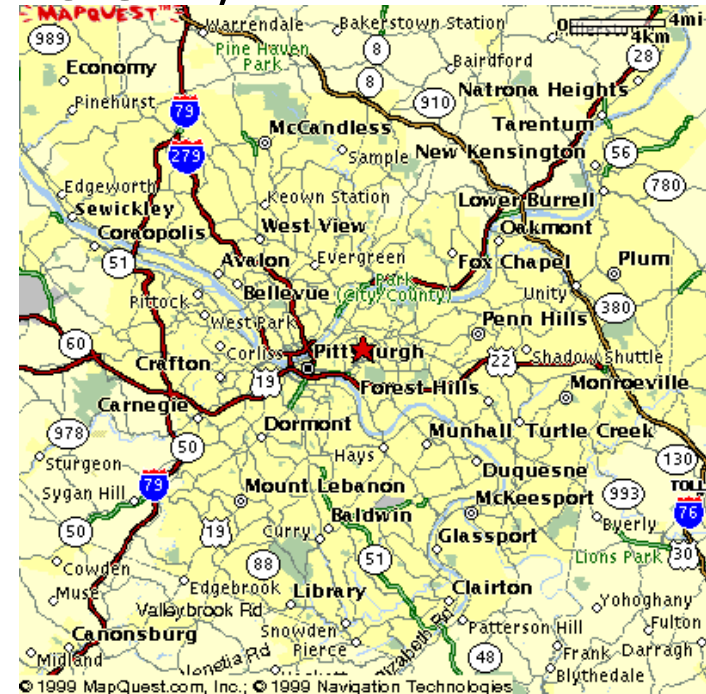
# More Example Configuration Spaces (contrasted with workspace)

- Free moving (no wheels) robot in plane:
  - workspace  $\mathbb{R}^2$
  - configuration space  $\mathbb{R}^2$
- 3-joint revolute arm in the plane
  - Workspace, a torus of outer radius  $L1 + L2 + L3$
  - configuration space  $T^3 = S^1 \times S^1 \times S^1$
- 2-joint revolute arm with a prismatic joint in the plane
  - workspace disc of radius  $L1 + L2 + L3$
  - configuration space  $T^2 \times \mathbb{R}$
- 3-joint revolute arm mounted on a mobile robot (holonomic)
  - workspace is a “sandwich” of radius  $L1 + L2 + L3$
  - $\mathbb{R}^2 \times T^3$
- 3-joint revolute arm floating in space
  - workspace is  $\mathbb{R}^3$
  - configuration space is  $SE(3) \times T^3$



# 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).

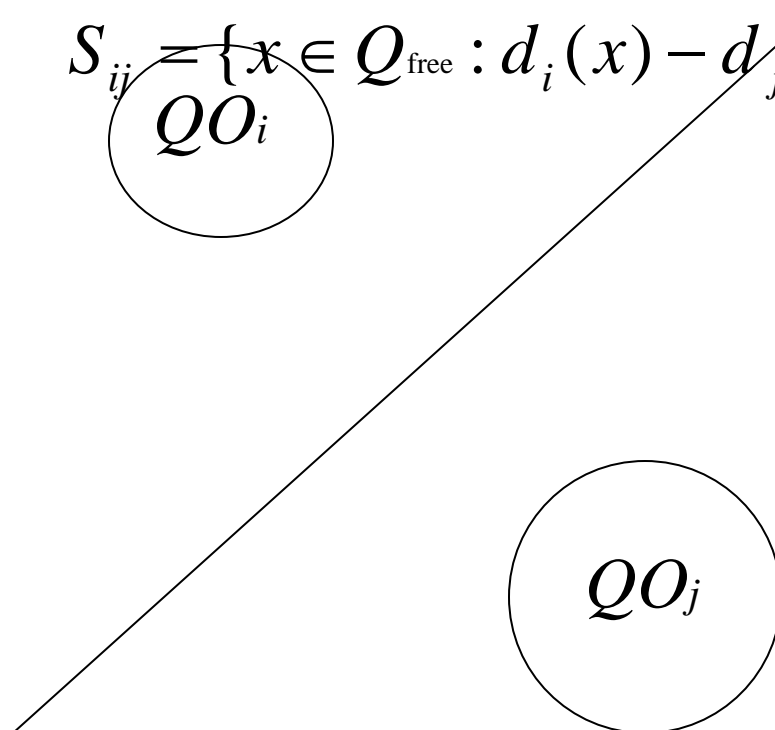


- a roadmap exists  $\Leftrightarrow$  a path exists
- Examples of Roadmaps
  - Generalized Voronoi Graph (GVG)
  - Visibility Graph

# Two-Equidistant

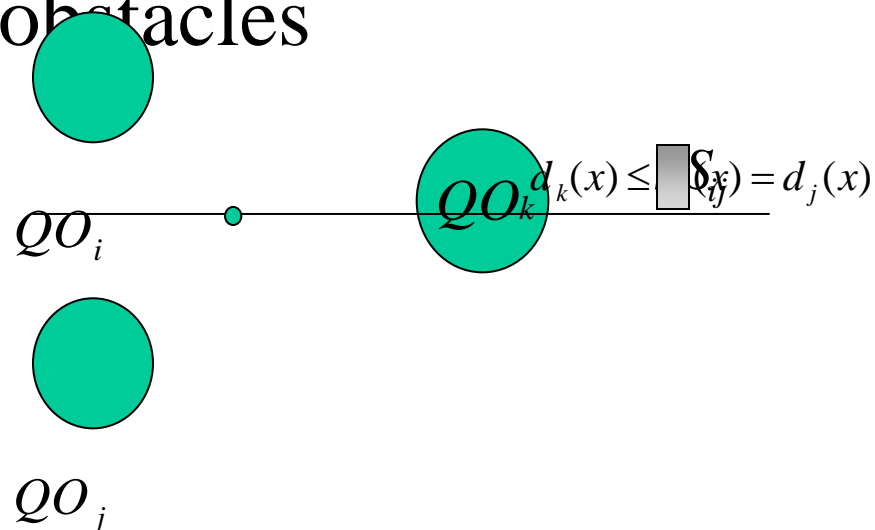
- *Two-equidistant surface*

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



# More Rigorous Definition

Going through obstacles

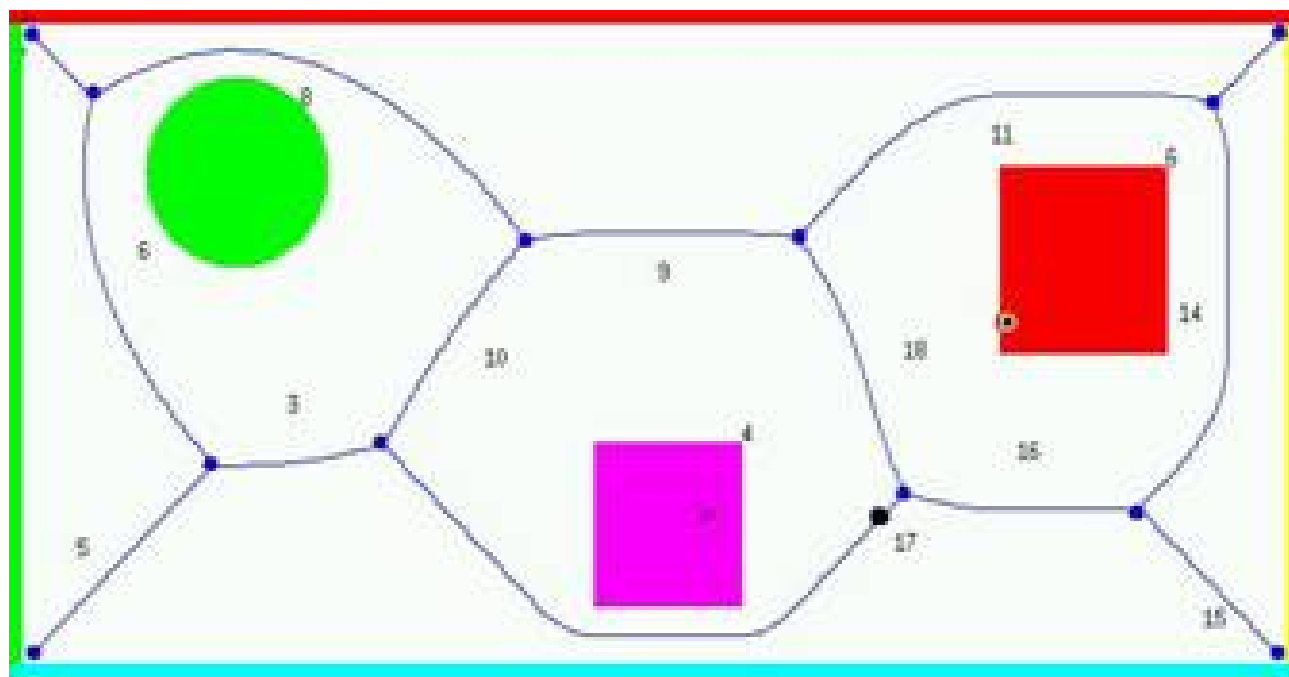


*Two-equidistant face*

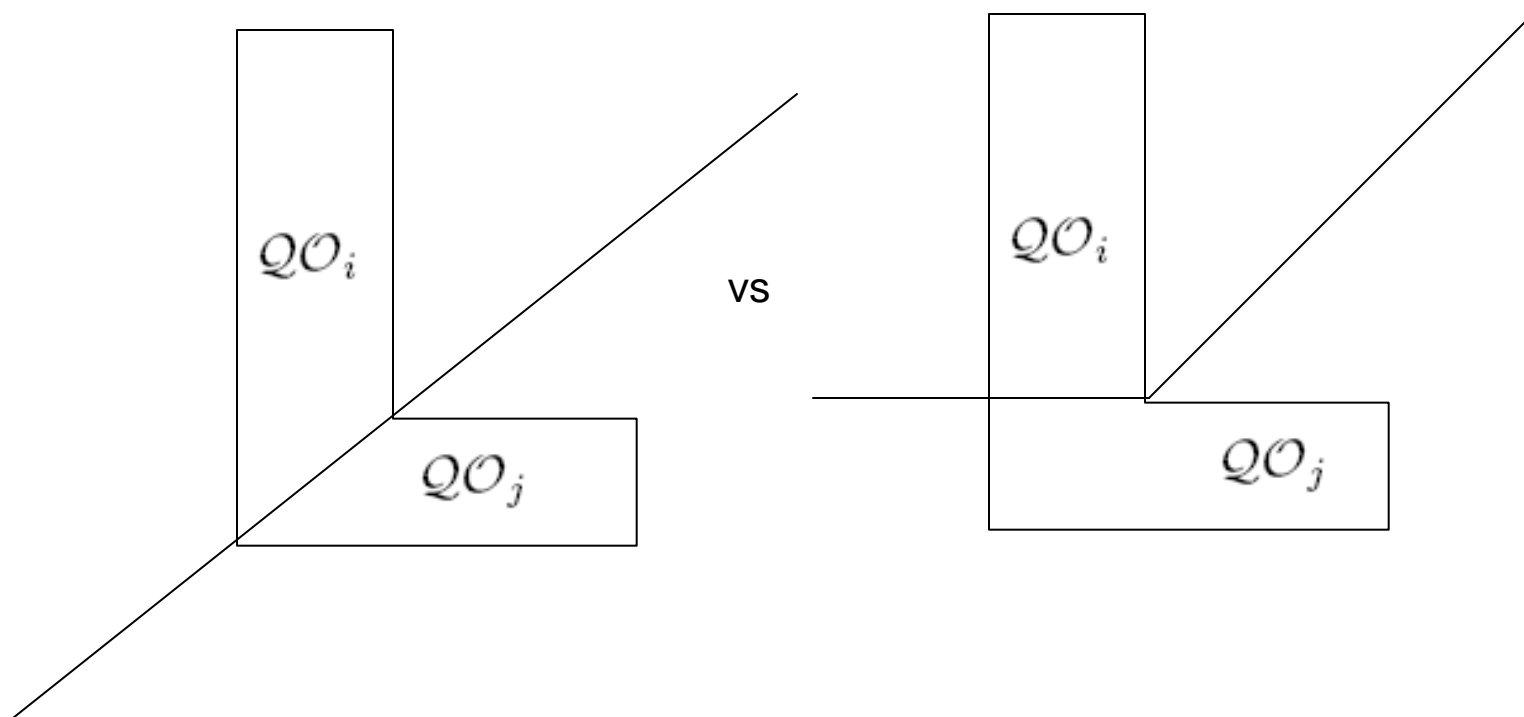
$$F_{ij} = \{x \in S_{ij} : d_i(x) = d_j(x) \leq d_h(x), \forall h \neq i, j\}$$

# General Voronoi Diagram

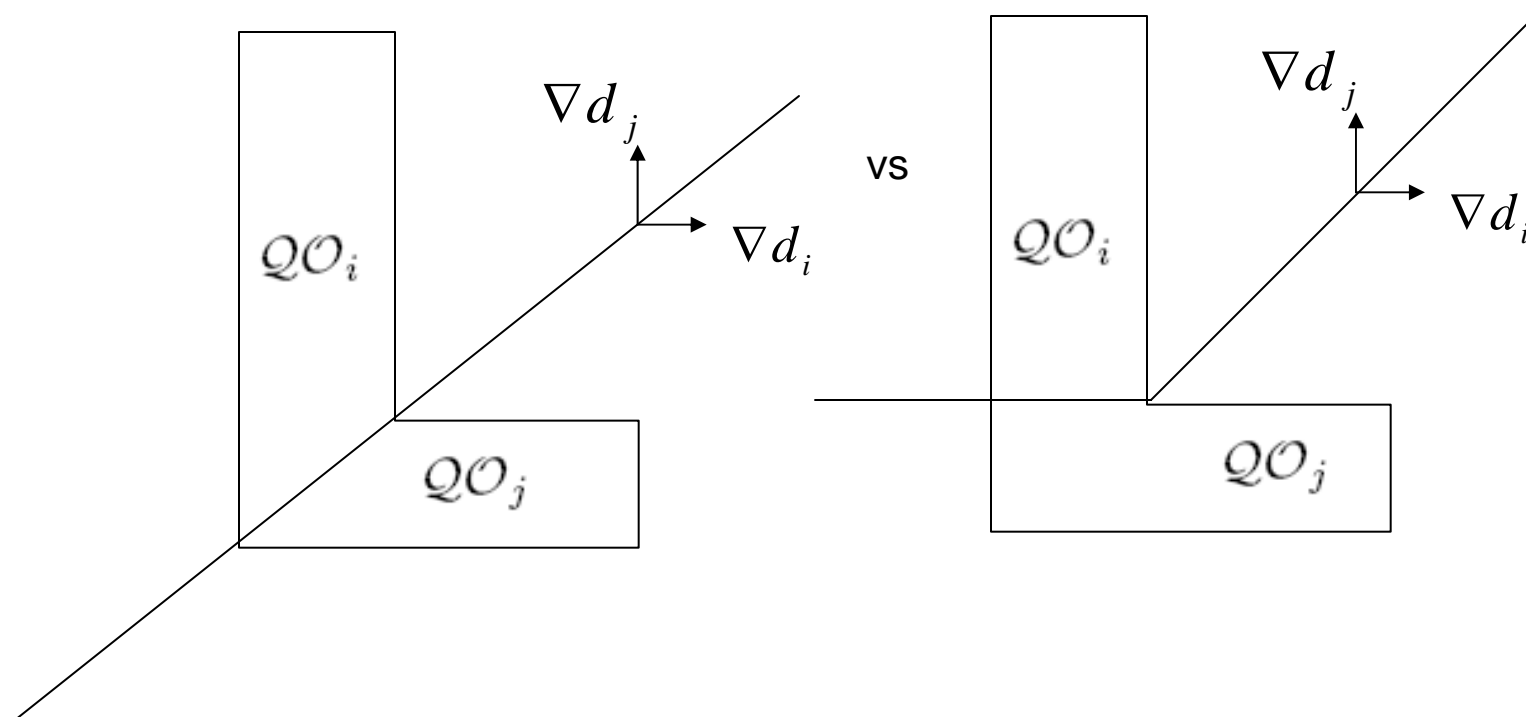
$$\text{GVD} = \bigcup_{i=1}^{n-1} \bigcup_{j=i+1}^n F_{ij}$$



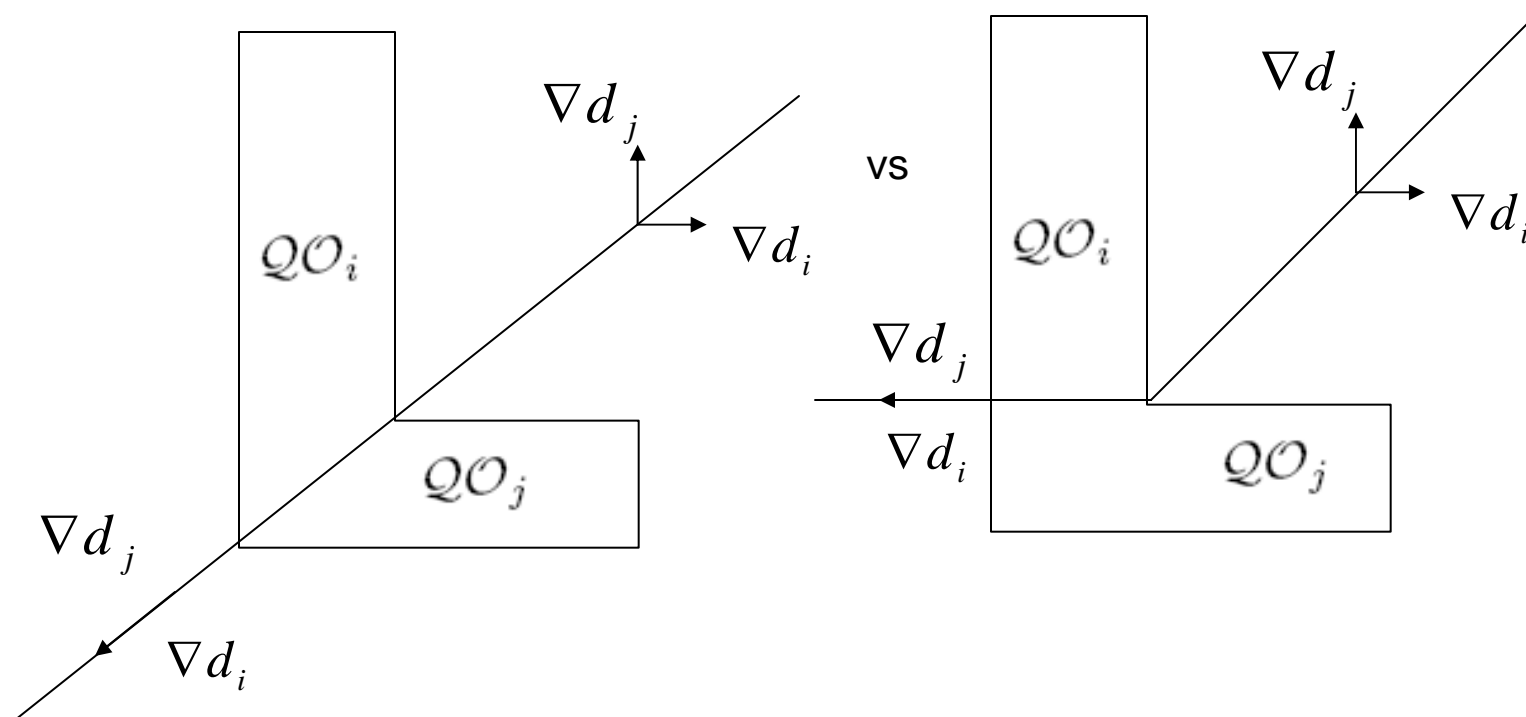
# What about concave obstacles?



# What about concave obstacles?



# What about concave obstacles?



# Two-Equidistant

- Two-equidistant surface*

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

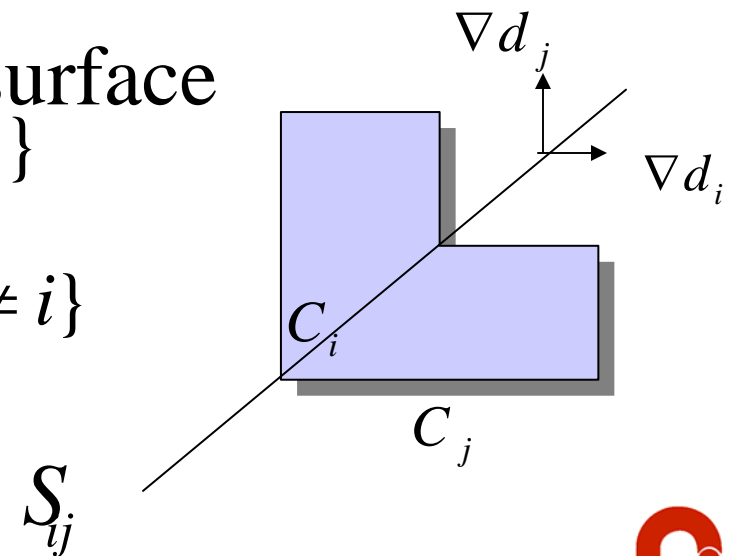
Two-equidistant surjective surface

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

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

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

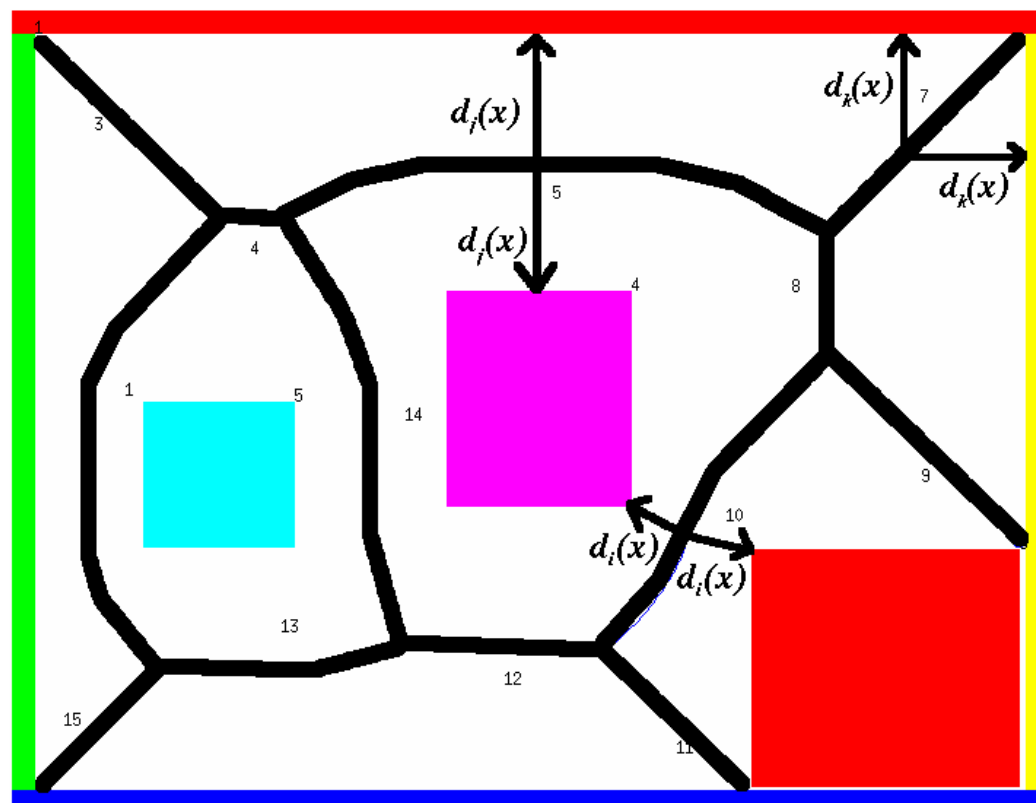
Two-equidistant Face





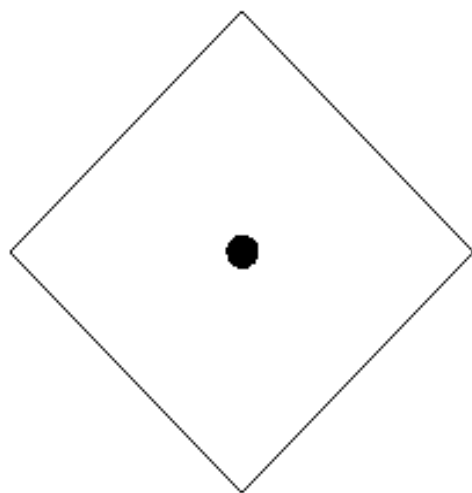
# Roadmap: GVG

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



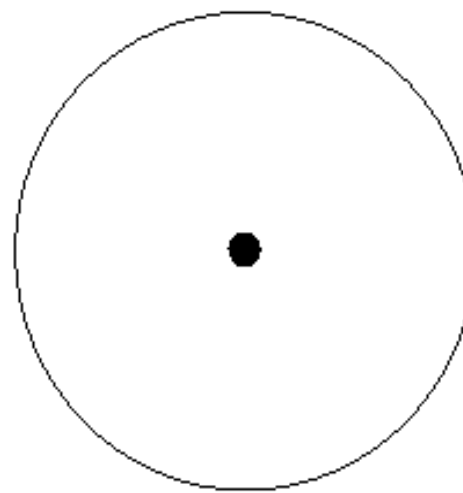
- This generates a very safe roadmap which avoids obstacles as much as possible

# Voronoi Diagram: Metrics



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

L1

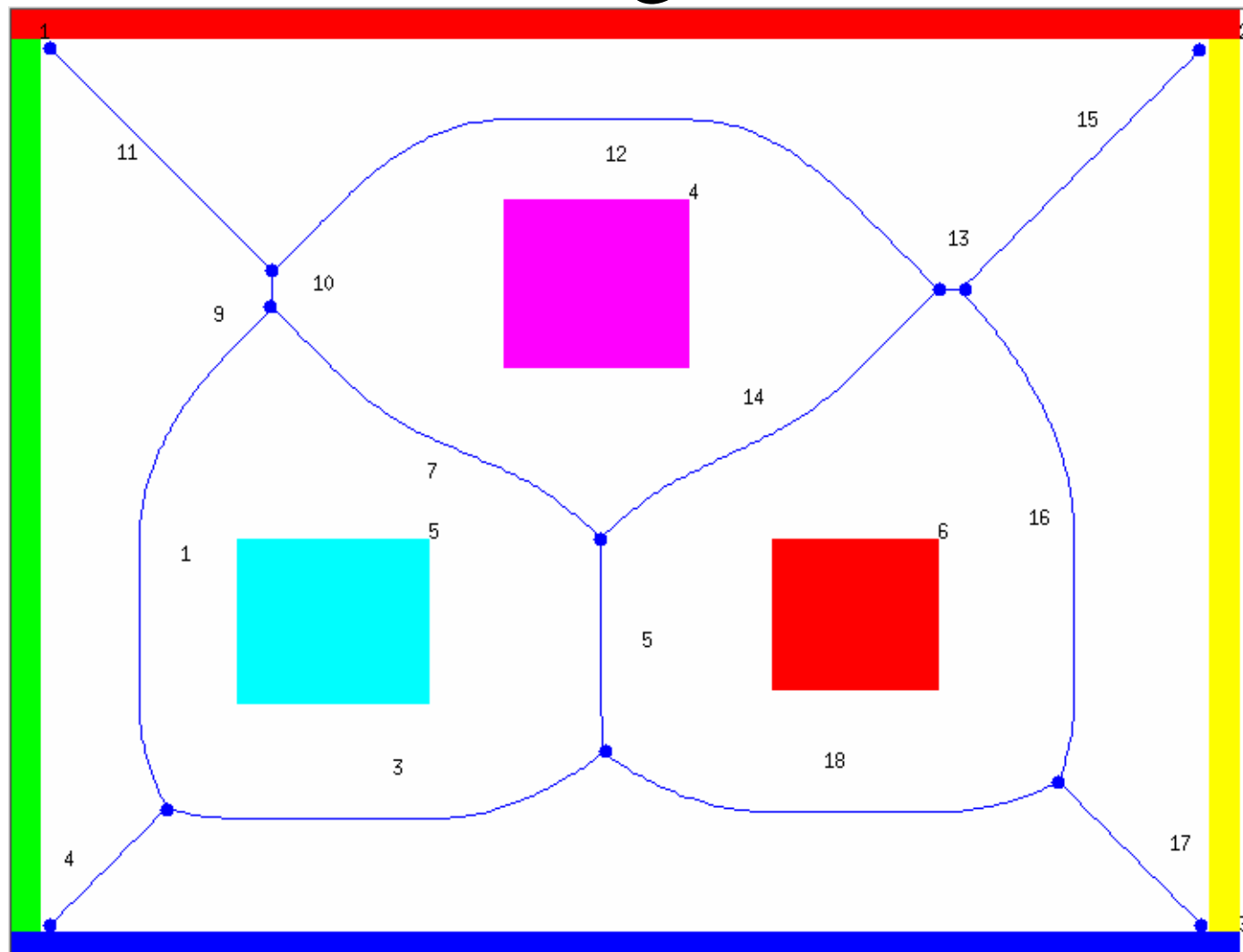


$$\{(x,y) : x^2 + y^2 = \text{const}\}$$

L2

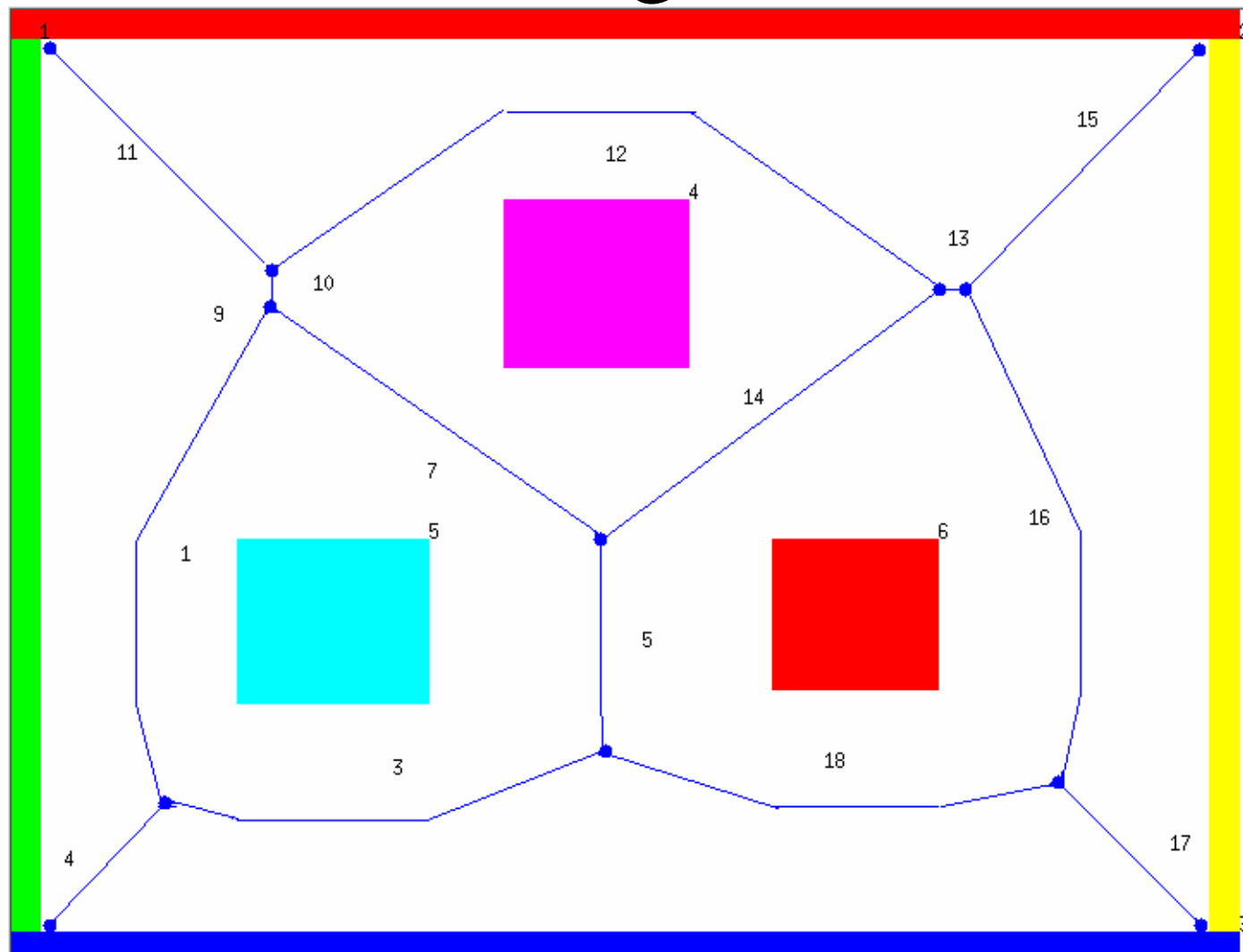
# Voronoi Diagram (L2)

Note the  
curved  
edges



# Voronoi Diagram (L1)

Note the  
lack of  
curved  
edges

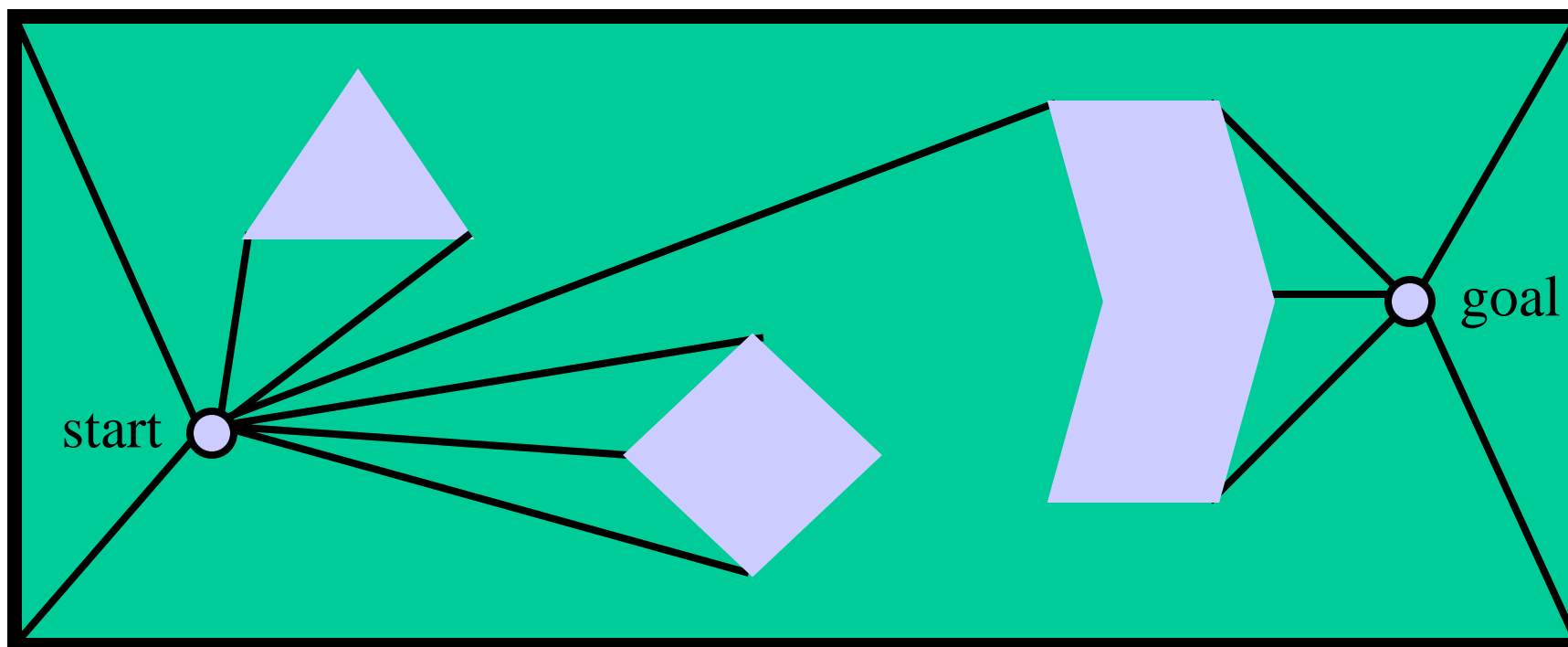


# Roadmap: Visibility Graph

- Formed by connecting all “visible” vertices, the start point and the end point, to each other
- For two points to be “visible” no obstacle can exist between them
  - Paths exist on the perimeter of obstacles
- In our example, this produces the shortest path with respect to the L2 metric. However, the close proximity of paths to obstacles makes it dangerous

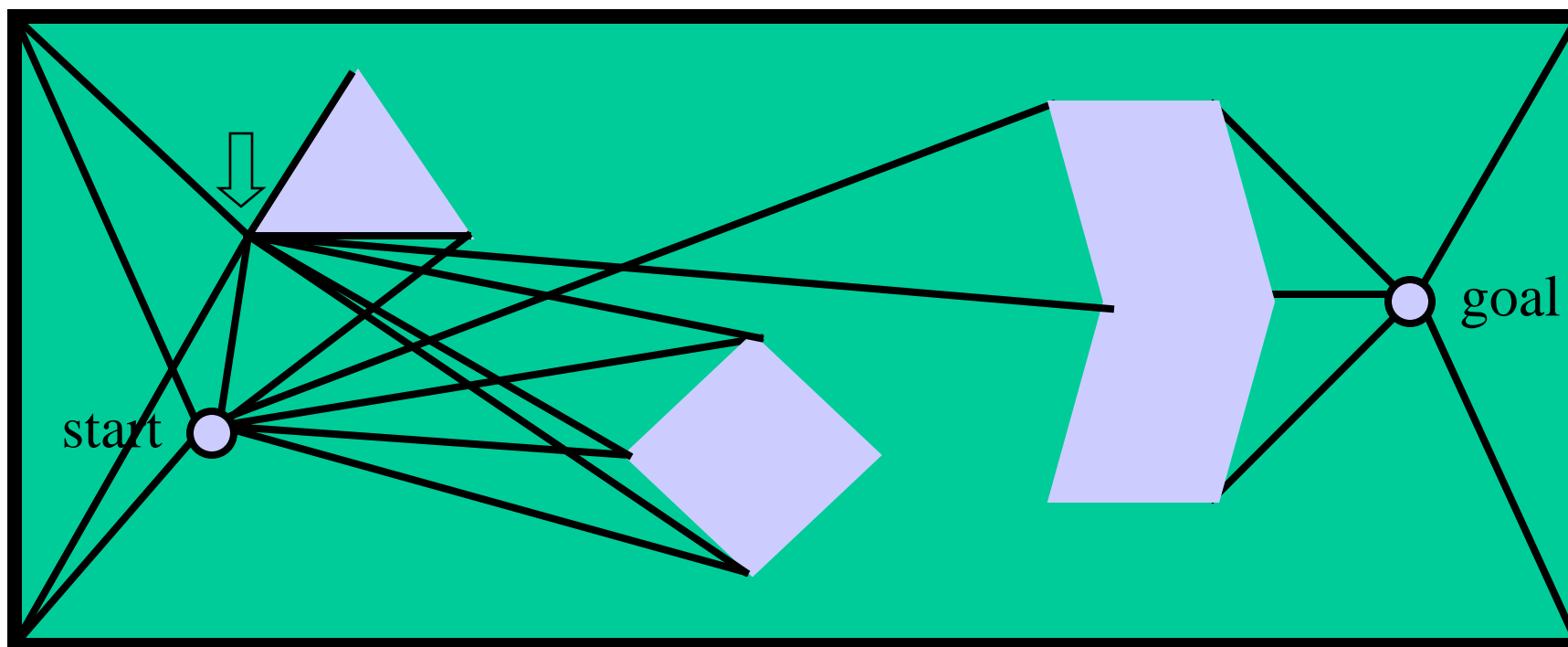
# The Visibility Graph in Action (Part 1)

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



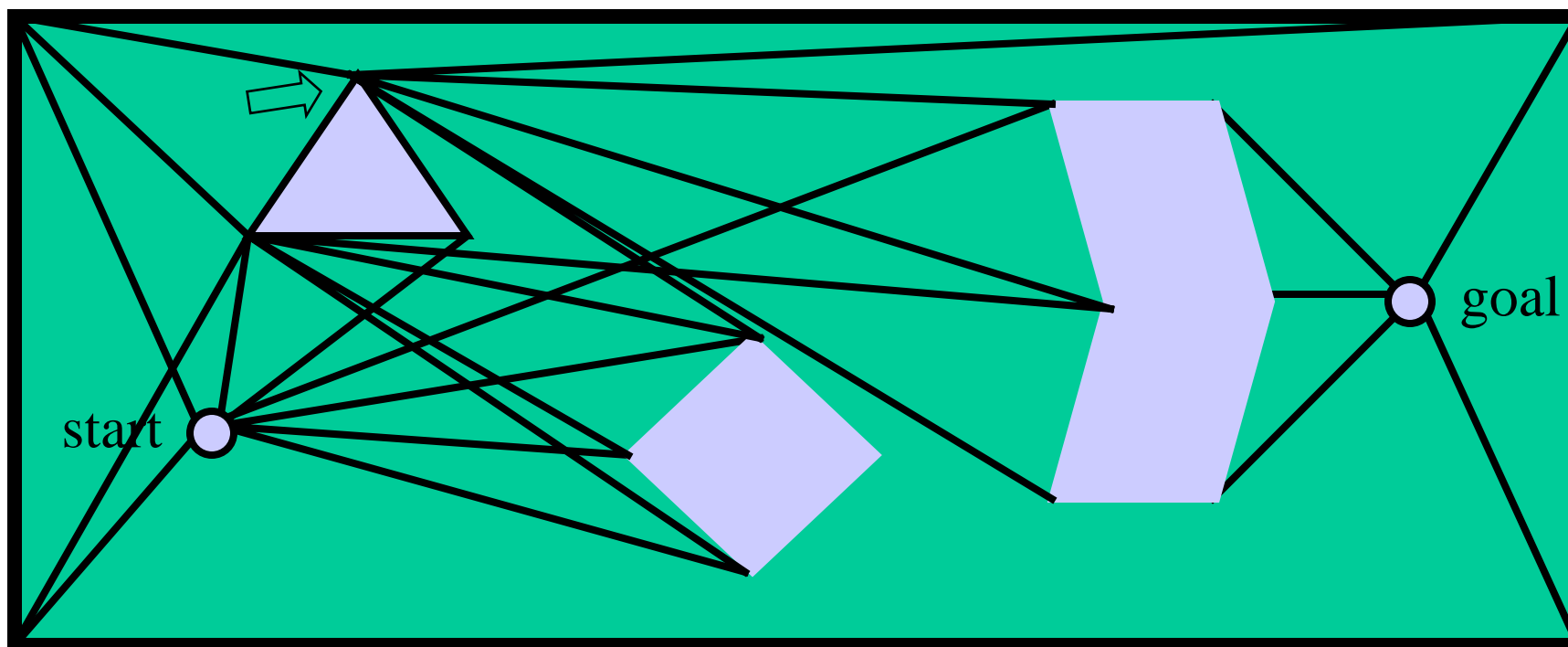
## The Visibility Graph in Action (Part 2)

- 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)

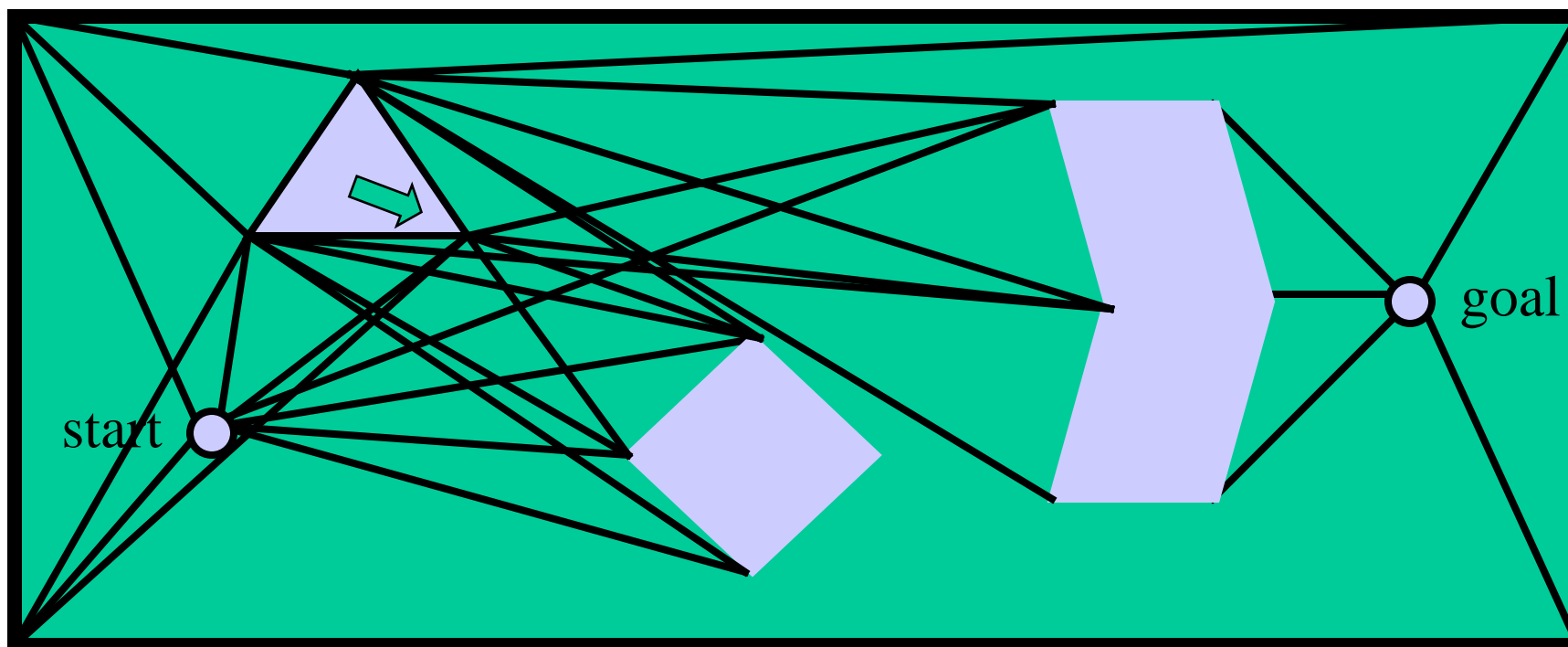
- Second, draw lines of sight from every vertex of every obstacle like before. Remember lines along edges are also lines of sight.





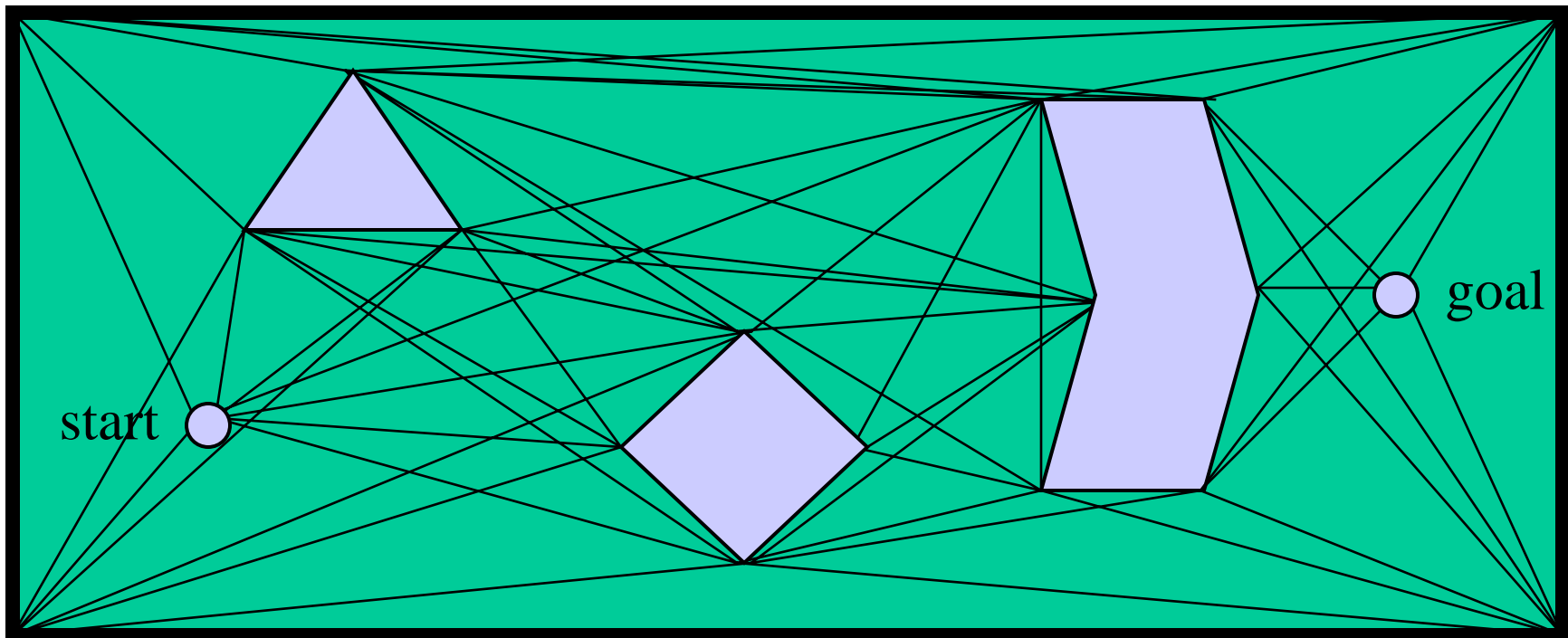
# The Visibility Graph in Action (Part 4)

- 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.

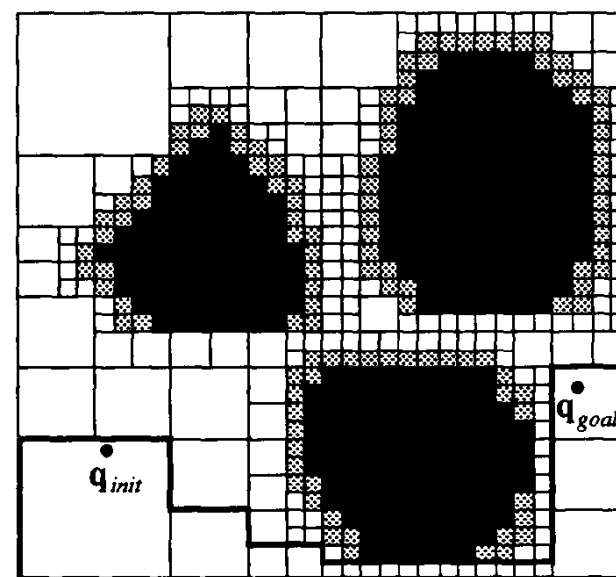
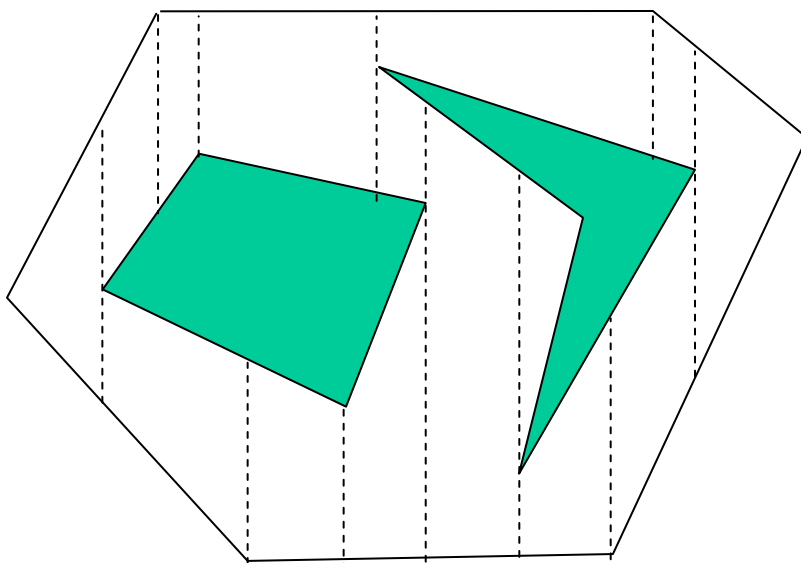


# Visibility Graph Overview

- Start with a map of the world, draw lines of sight from the start and goal to every “corner” of the world and vertex of the obstacles, not cutting through any obstacles.
- Draw lines of sight from every vertex of every obstacle like above. Lines along edges of obstacles are lines of sight too, since they don’t pass through the obstacles.
- If the map was in Configuration space, each line potentially represents part of a path from the start to the goal.

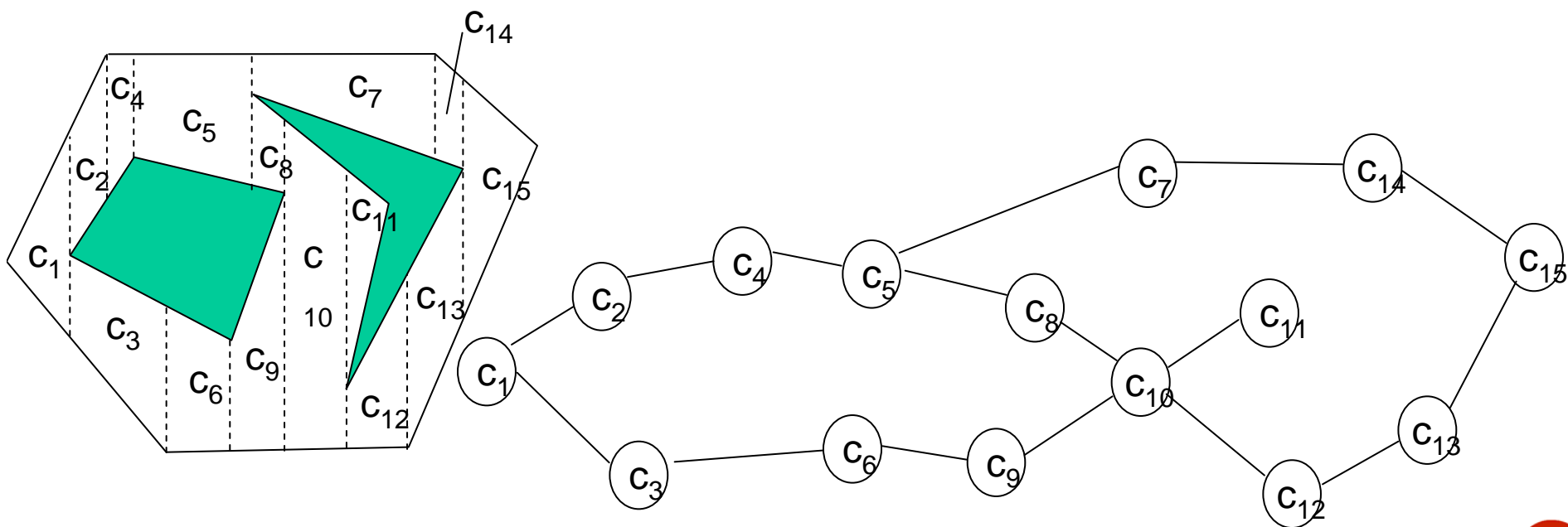
# Exact Cell vs. Approximate Cell

- 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  $A$  ( $\partial A$ ) 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

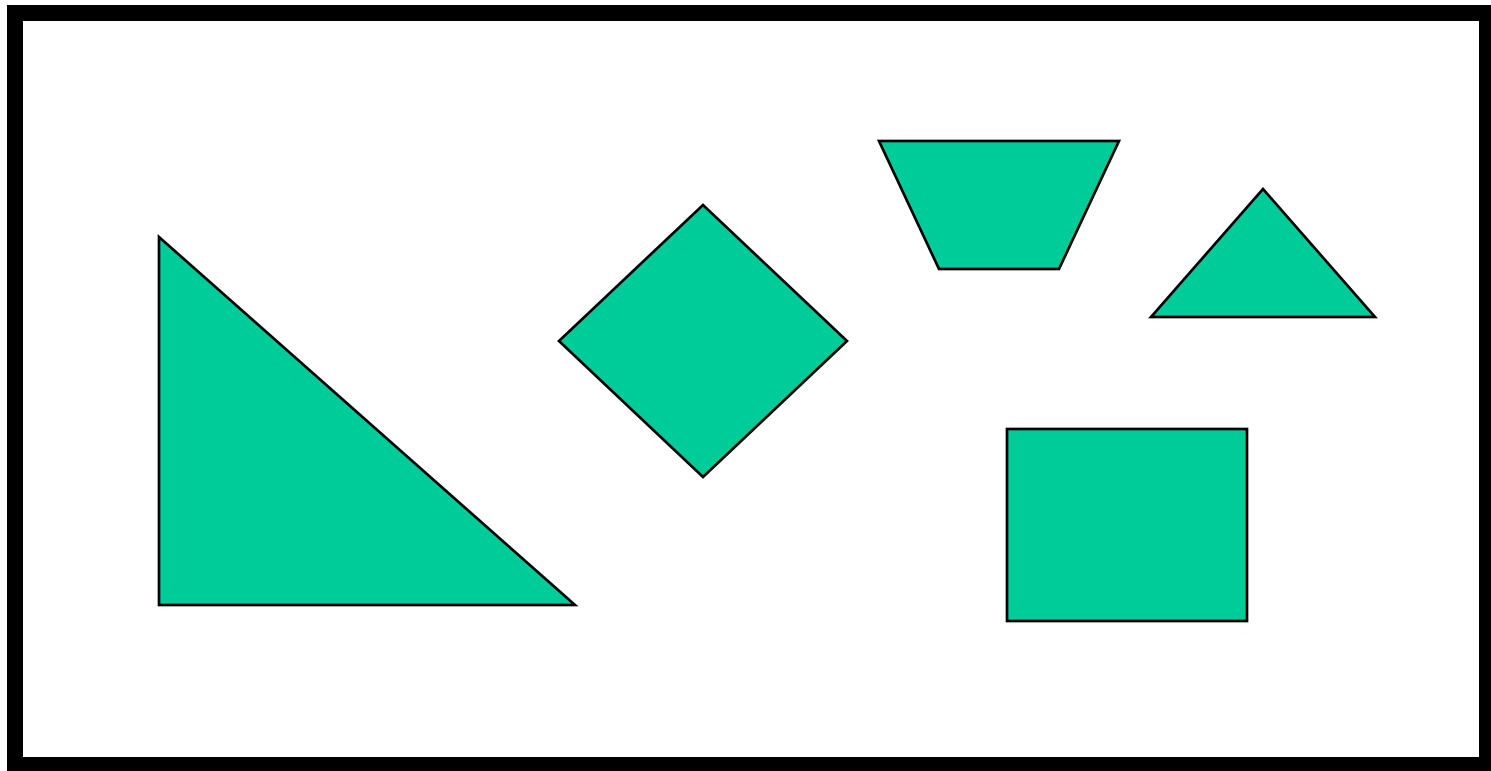
Exact Cellular Decomposition (as opposed to approximate)

- $\nu_i$  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  $\nu_i$  and  $\nu_j$  are adjacent cells
- $Fs = \cup_i (\nu_i)$



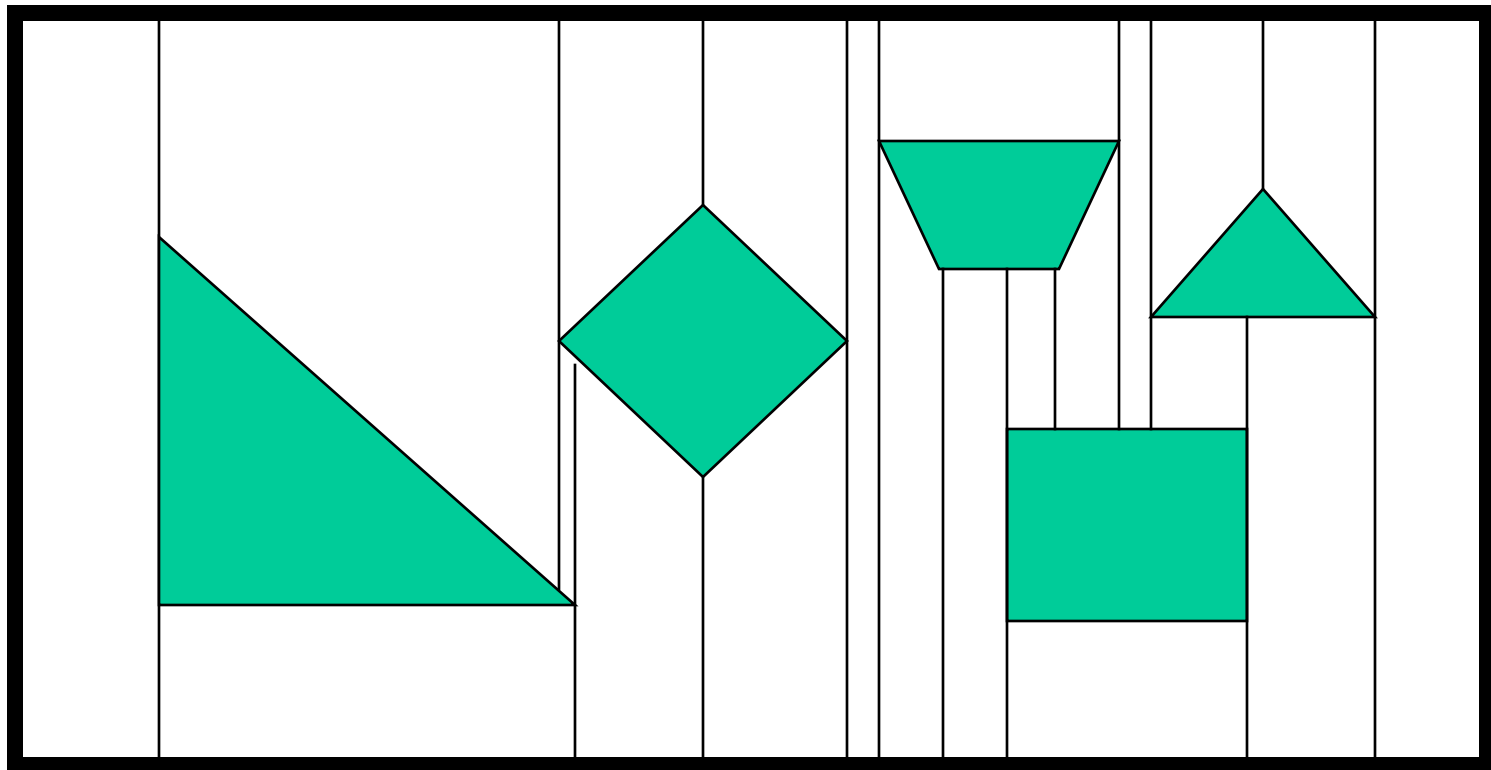
# Cell Decompositions: Trapezoidal Decomposition

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



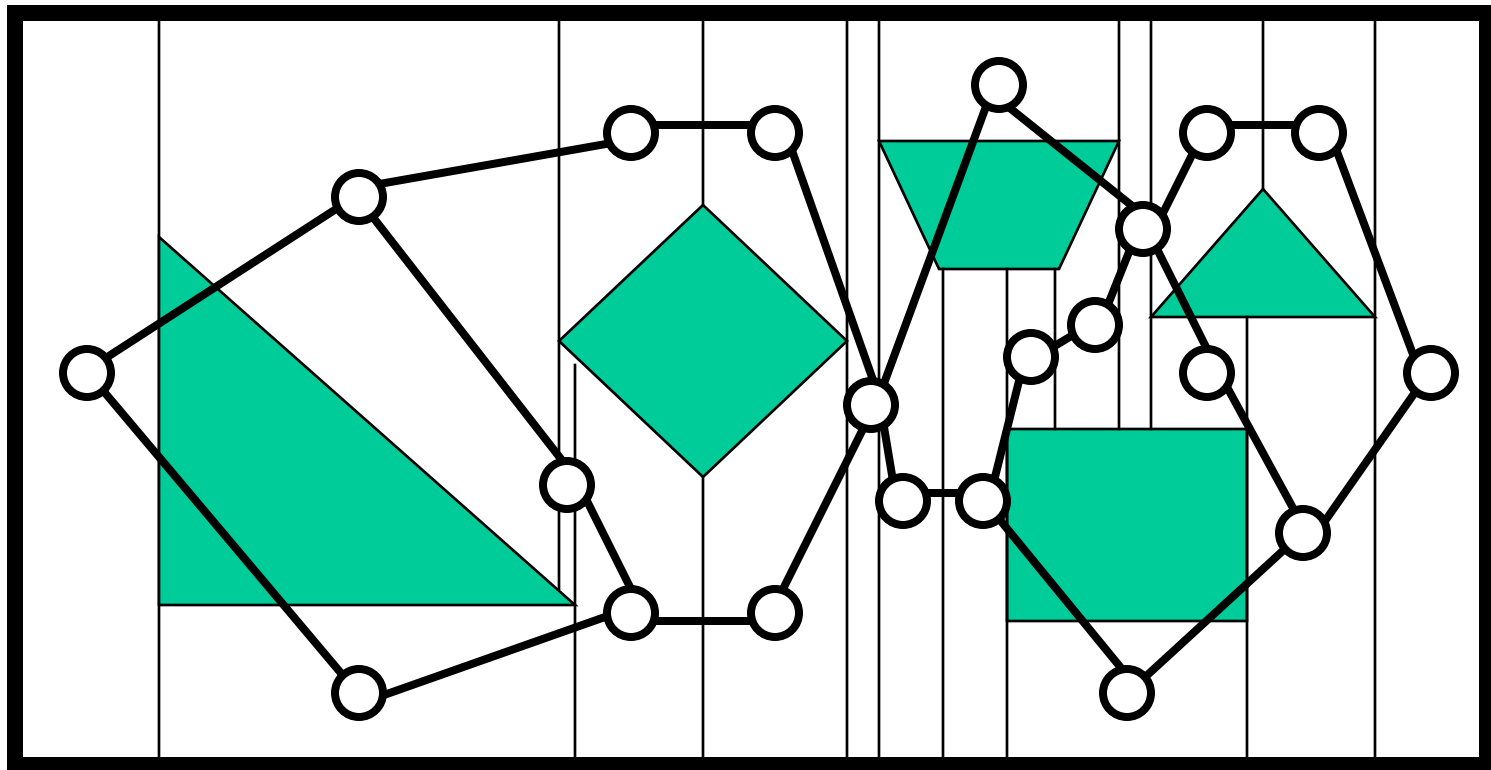
# Cell Decompositions: Trapezoidal Decomposition

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



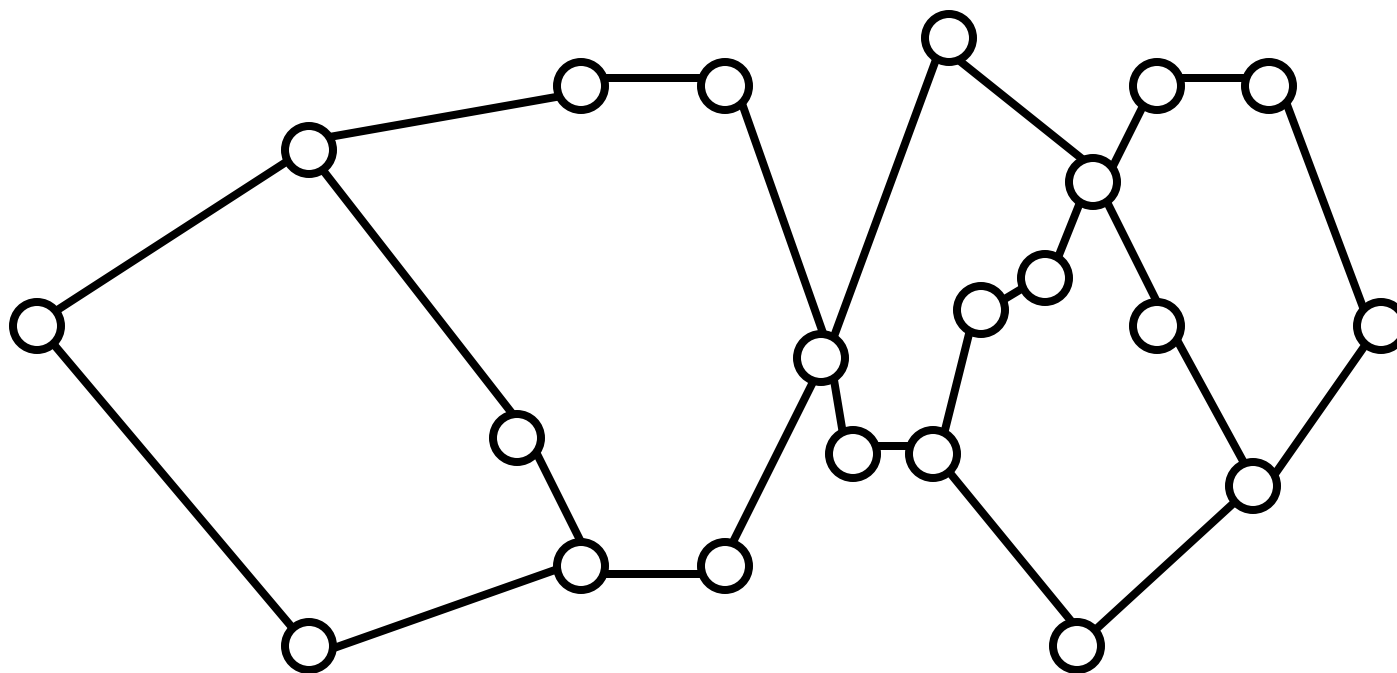
# Applications: Coverage

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



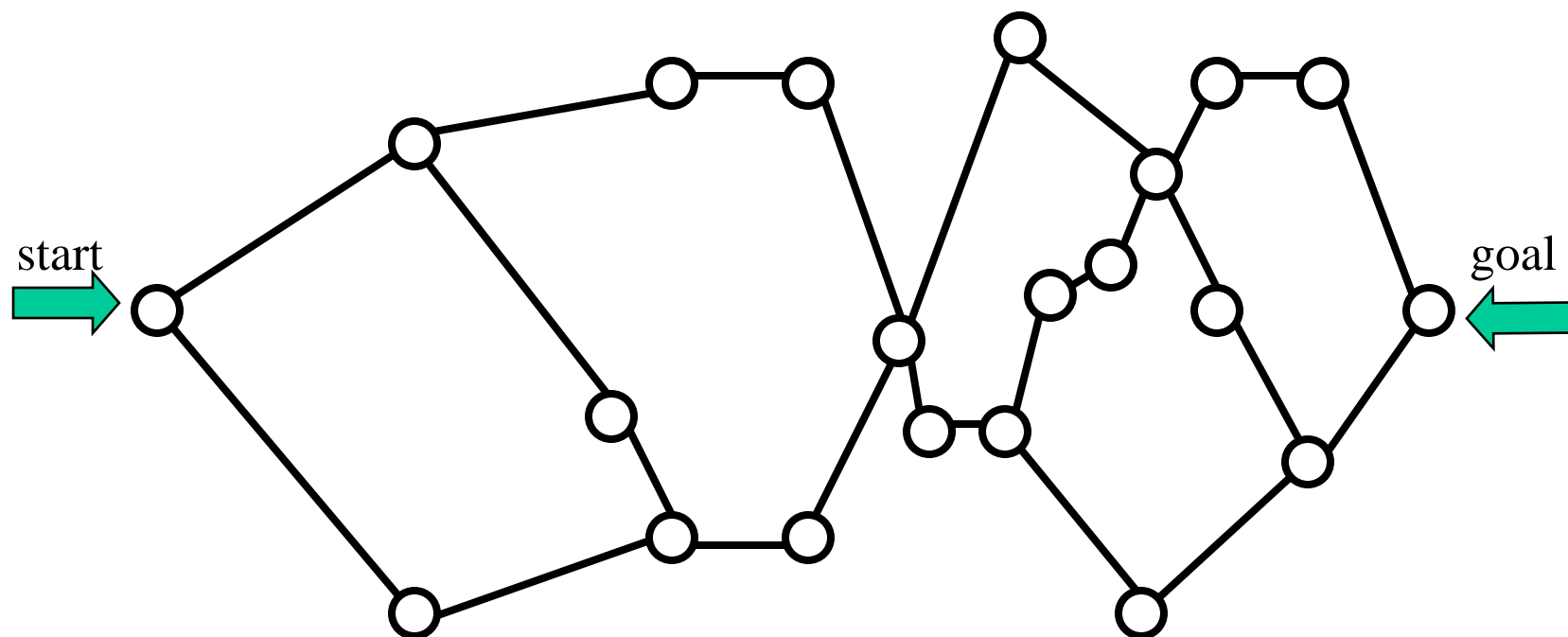
# Find a path

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



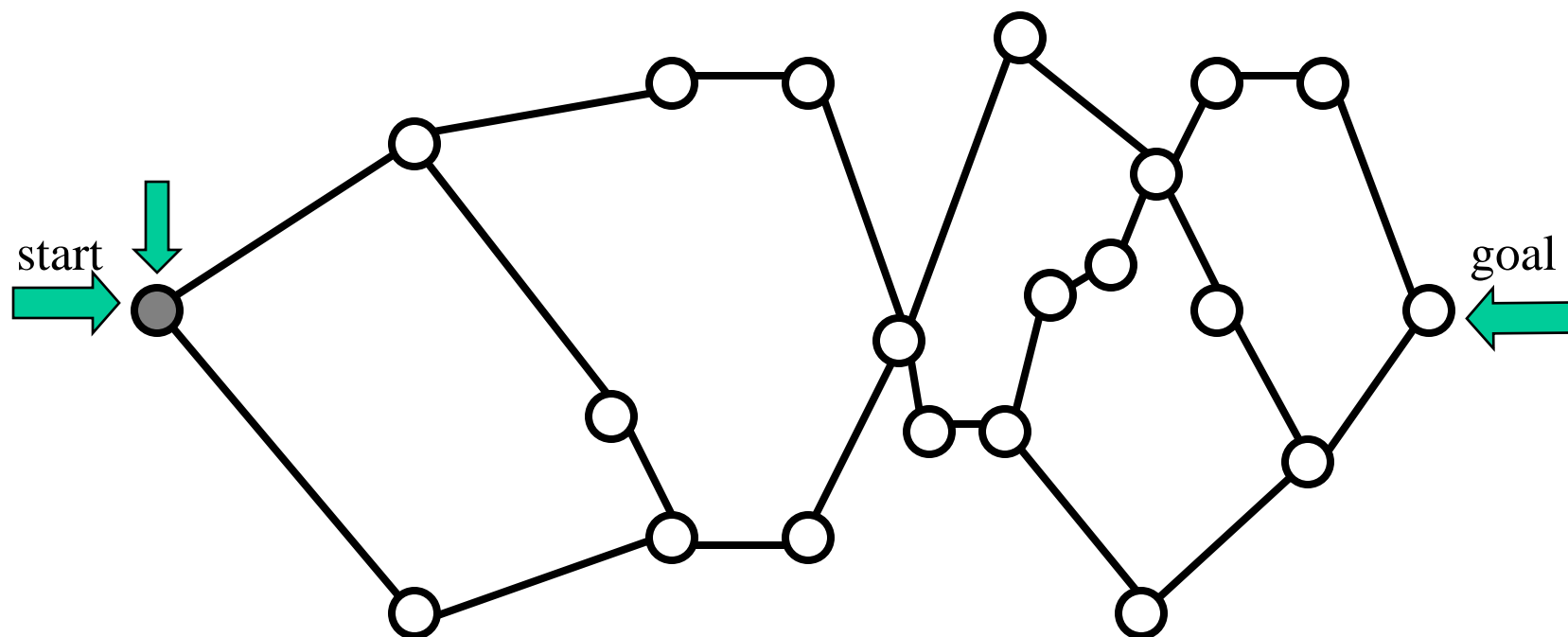
# Find a path

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



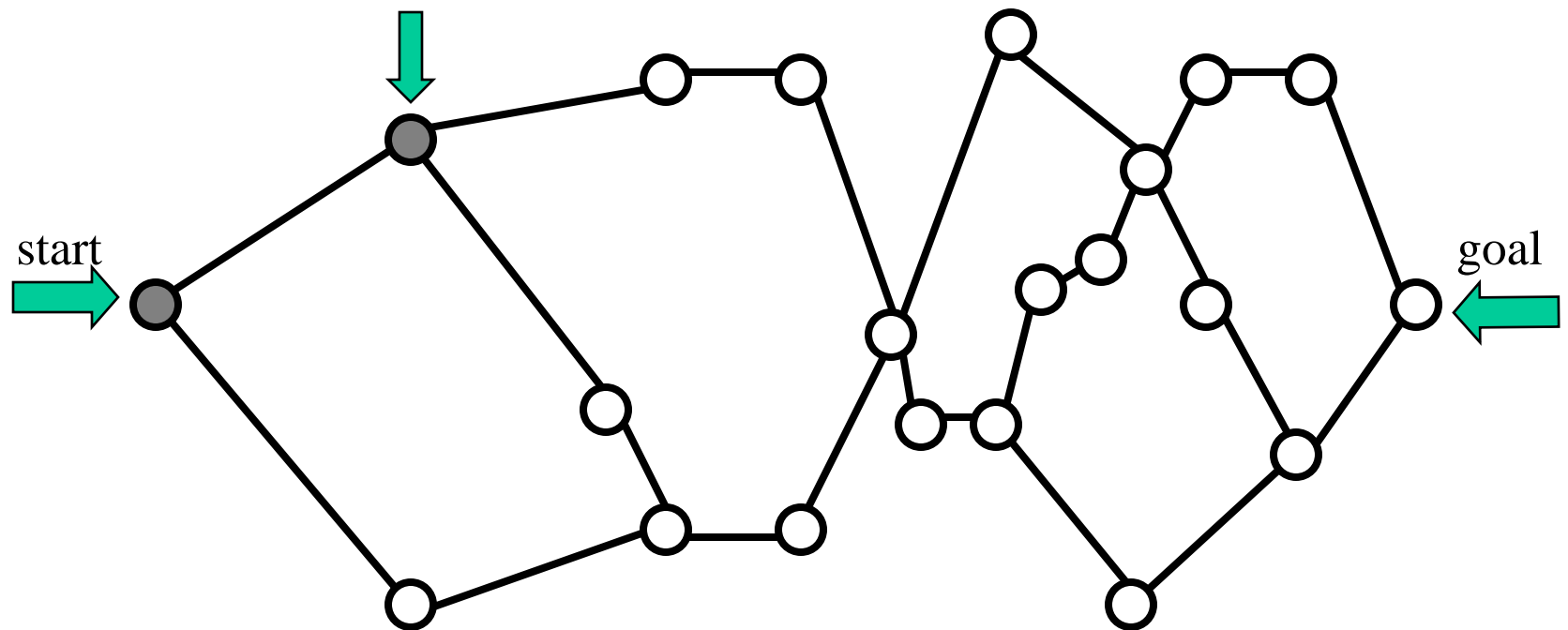
# Find a path

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



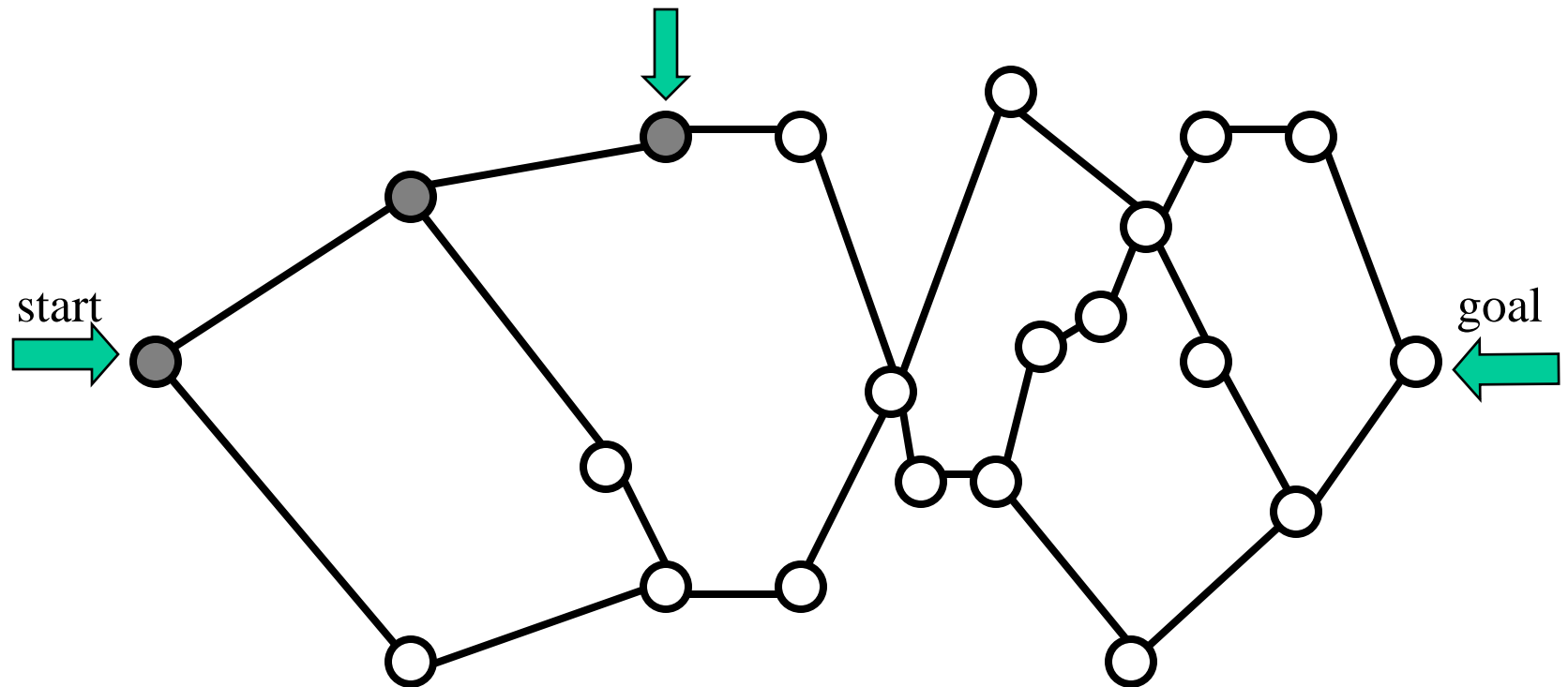
# Find a path

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



# Find a path

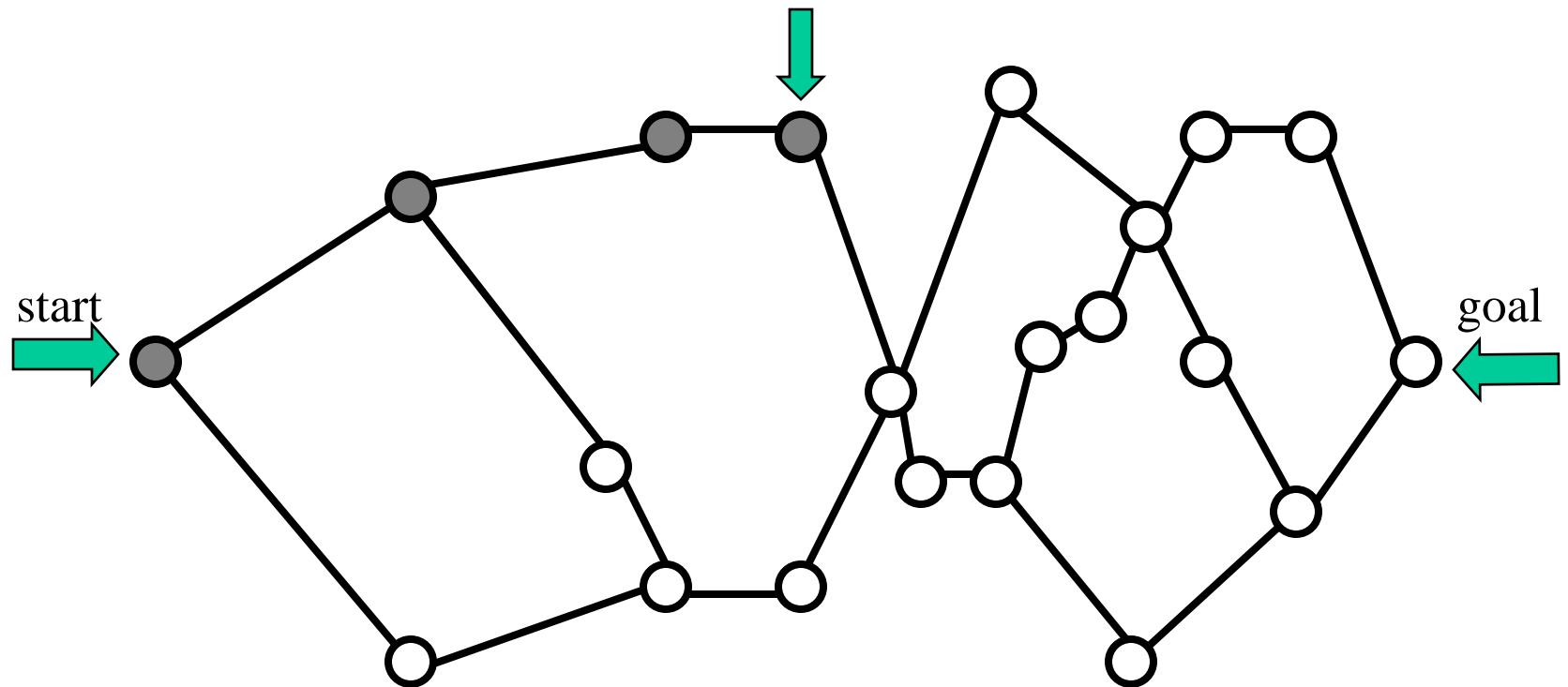
- With an adjacency graph, a path from start to goal can be found by simple traversal





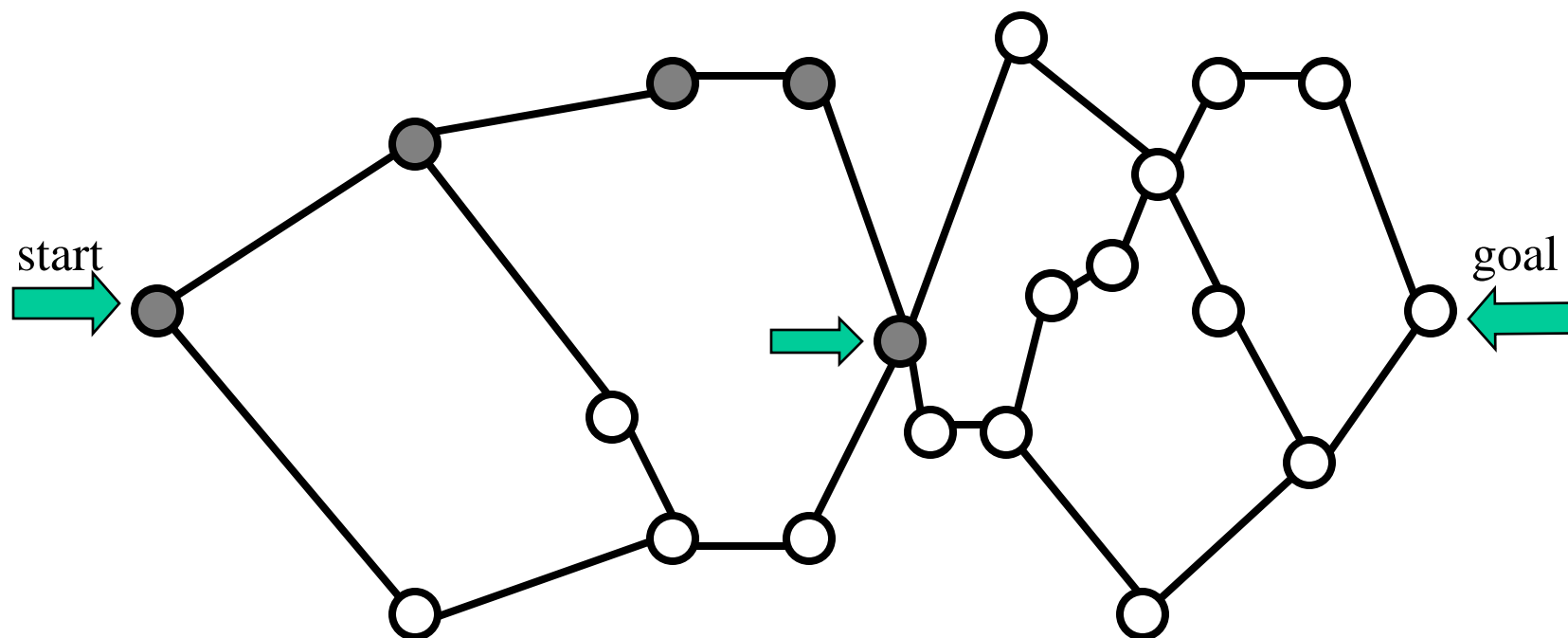
# Find a path

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



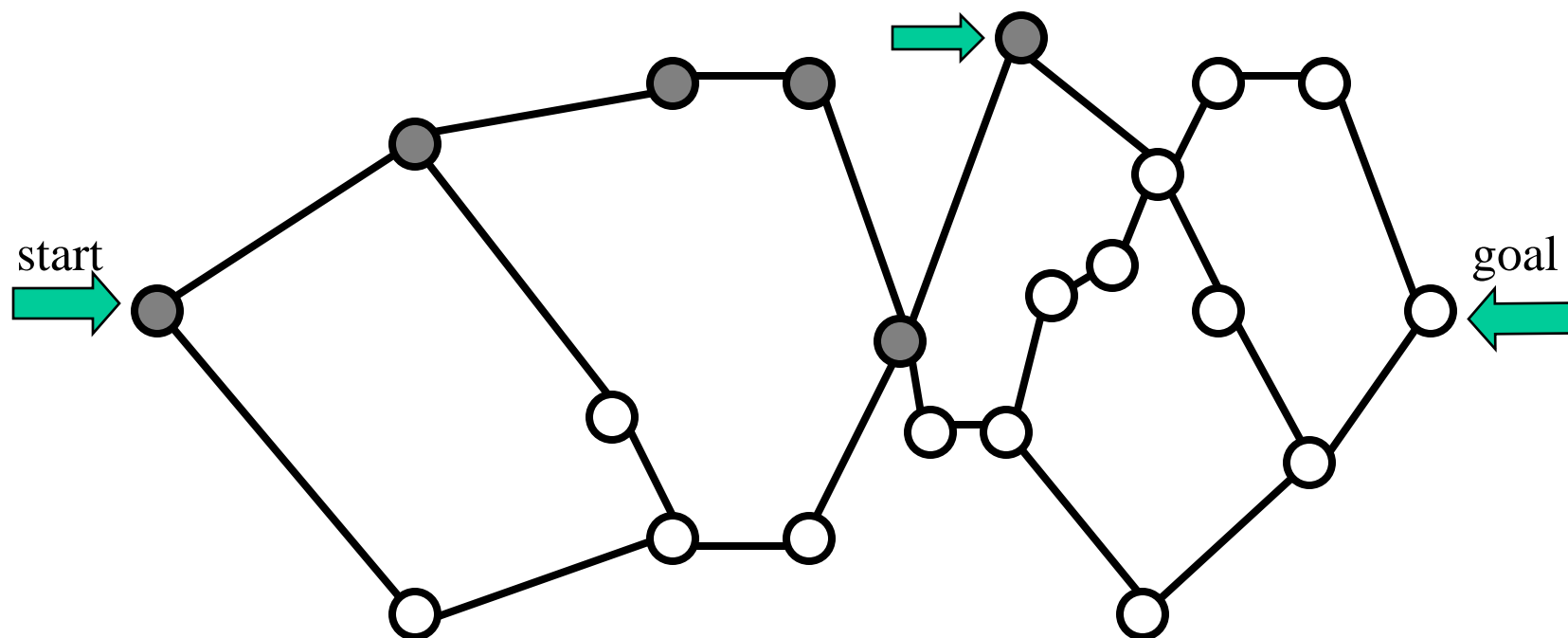
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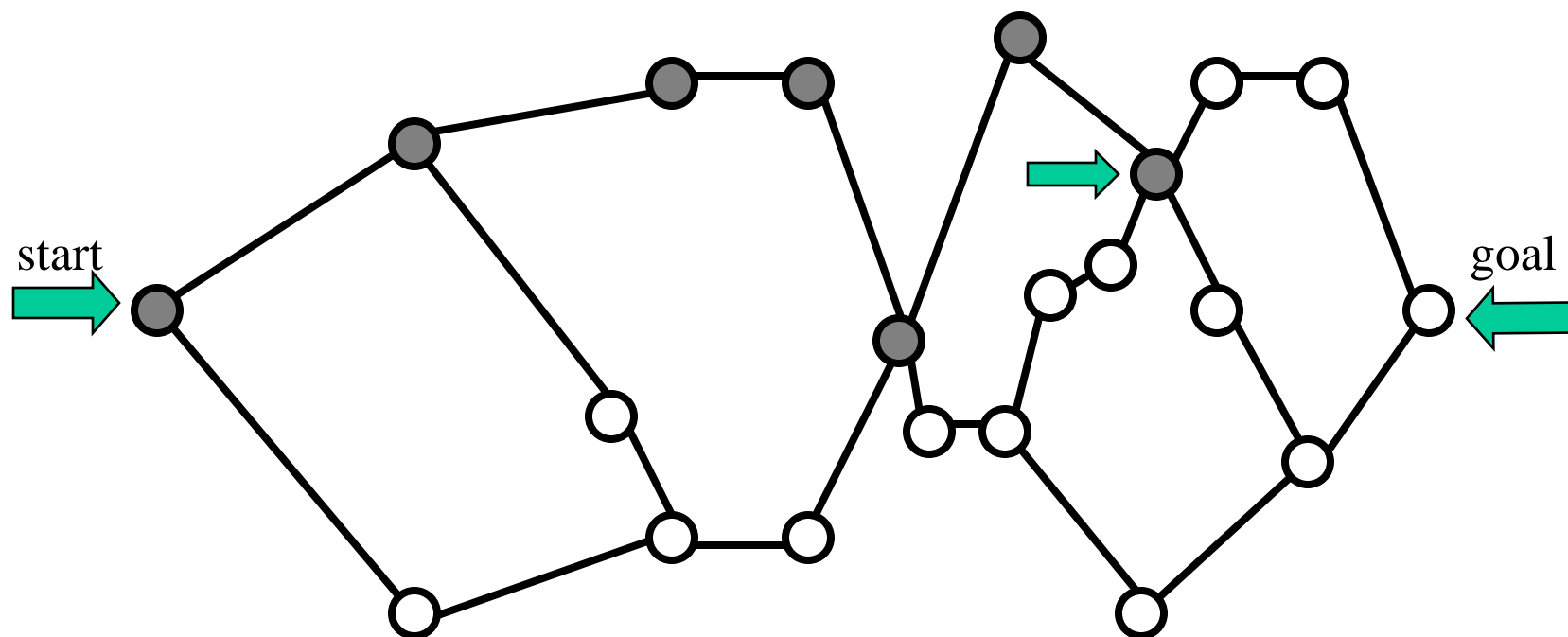
# Find a path

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



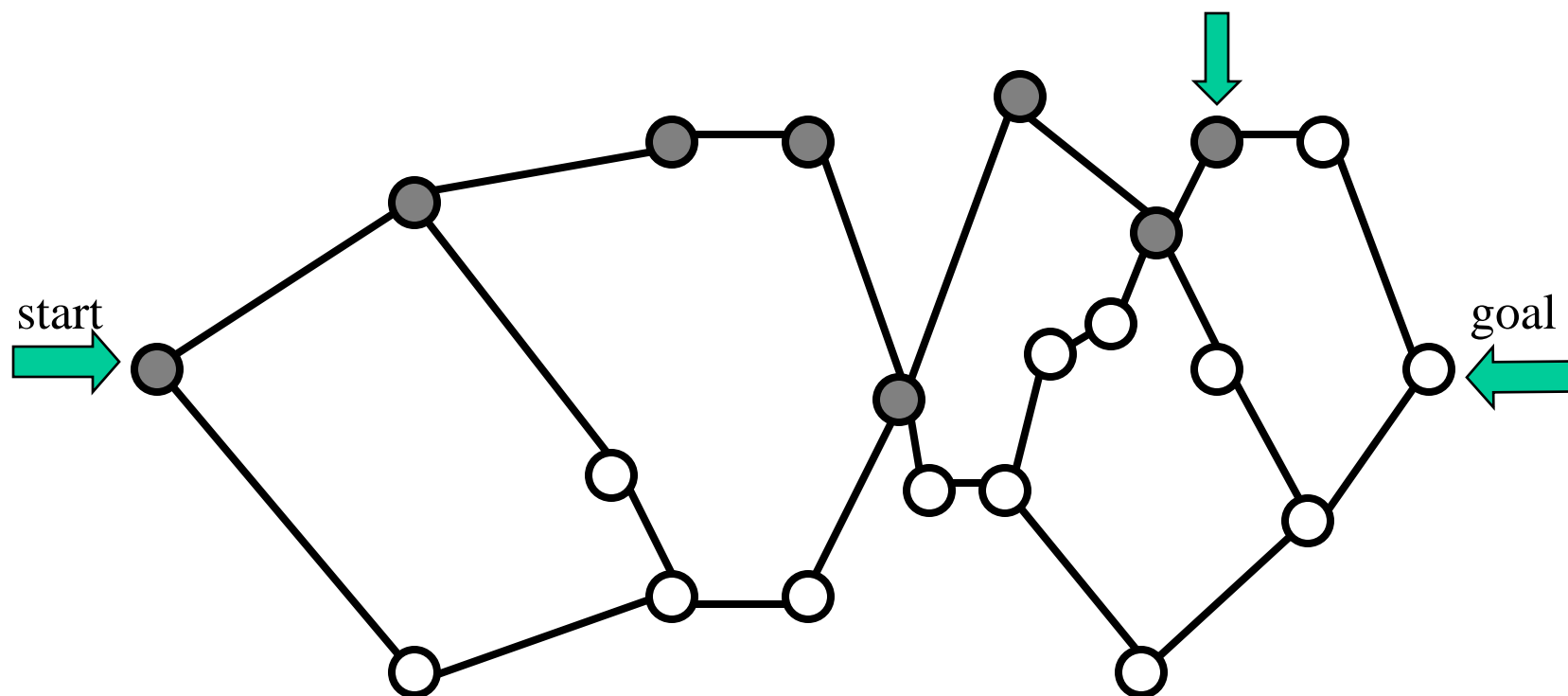
# Find a path

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



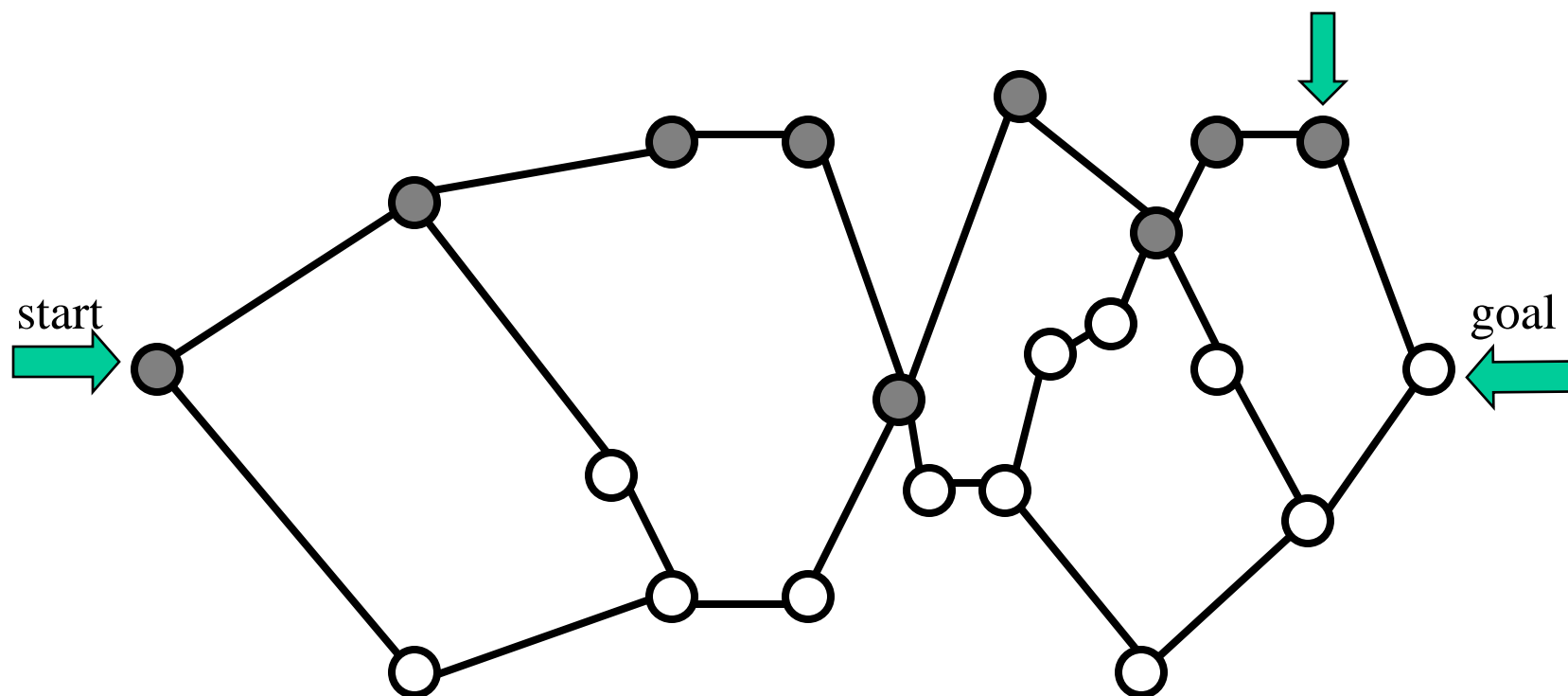
# Find a path

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



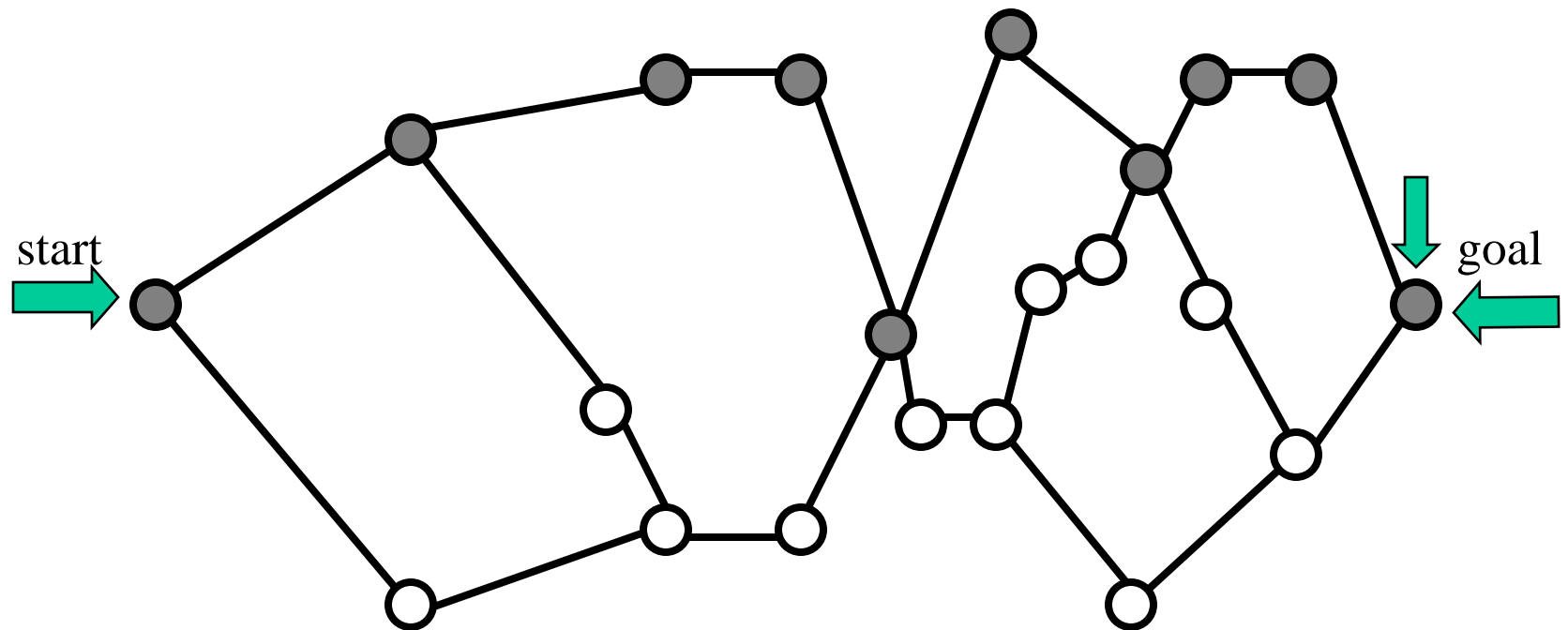
# Find a path

- 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



# Connect Midpoints of Traps

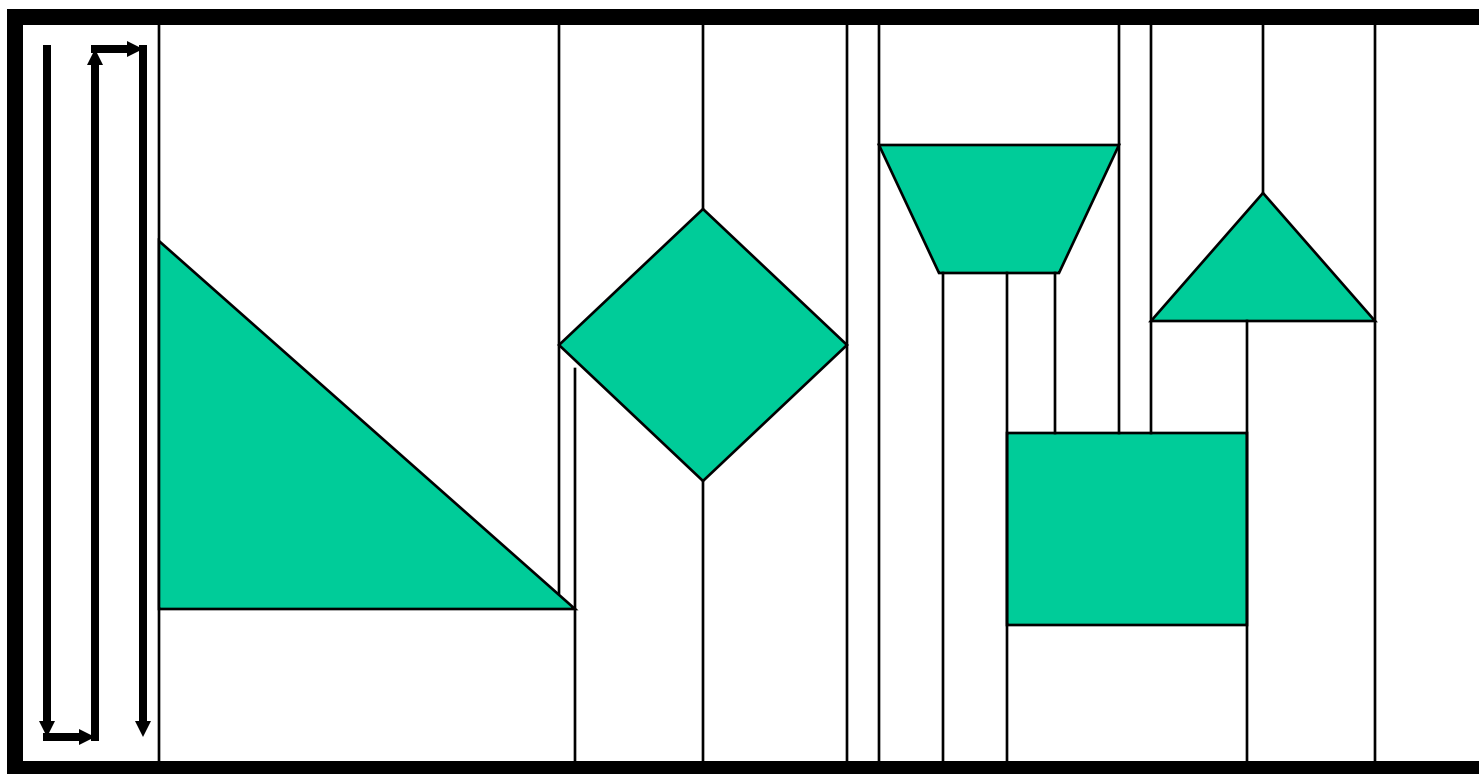


# Applications: Coverage

- 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

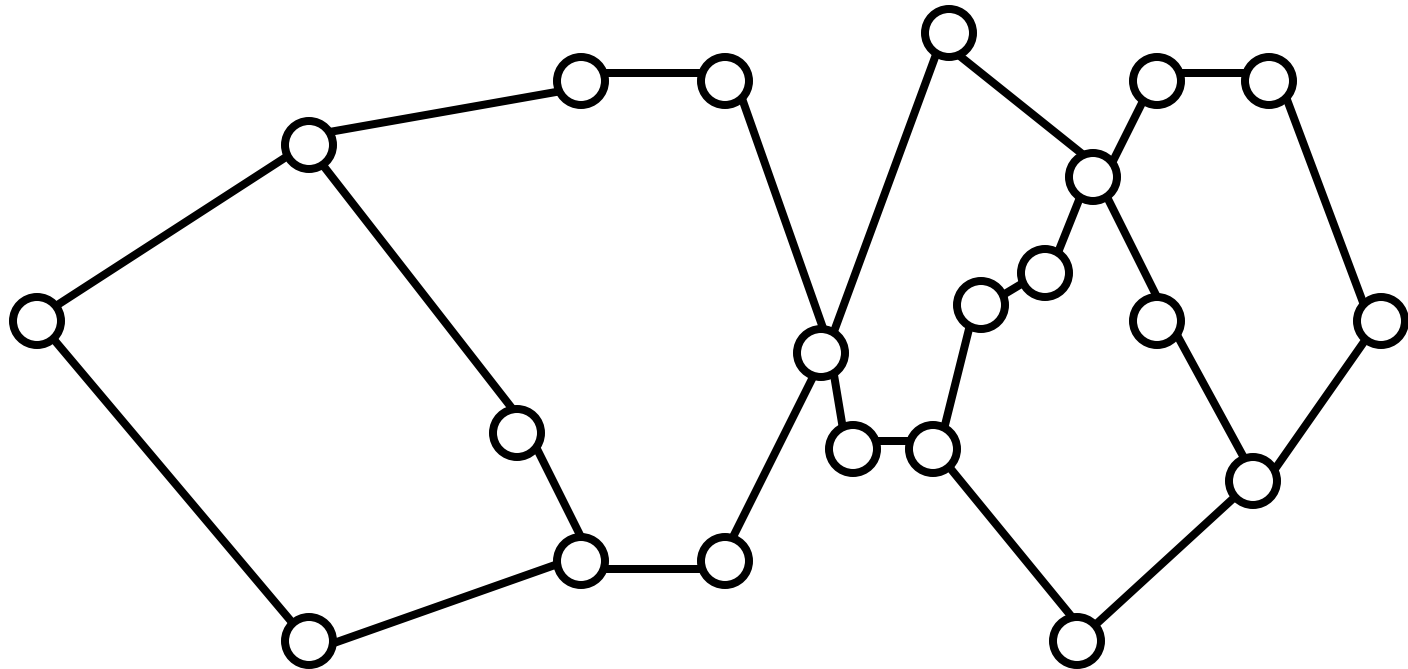
# Cell Decompositions: Trapezoidal Decomposition

- 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.

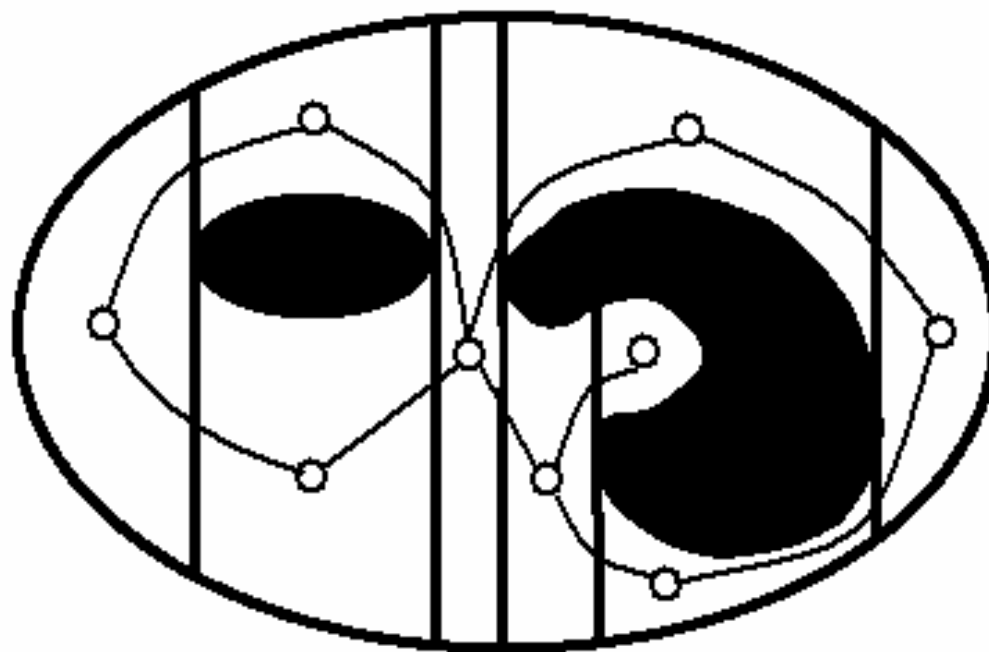


# Applications: Coverage

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



# Boustrophedon Decomposition



# 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!*