

### **CSCE 774 ROBOTIC SYSTEMS**

**Path Planning** 

### **Outline**

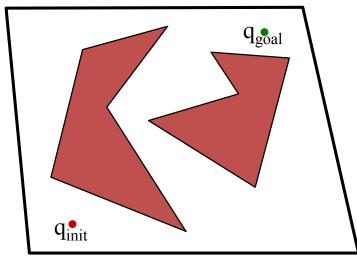
- Path Planning
  - Visibility Graph
  - Potential Fields
  - Bug Algorithms
  - Skeletons/Voronoi Graphs
  - C-Space

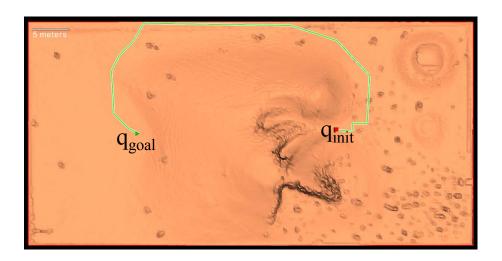


## **Motion Planning**

- The ability to go from A to B
  - Known map Off-line planning
  - Unknown Environment –Online planning
  - Static/Dynamic Environment

 $\bullet q_{init}$   $\bullet q_{goal}$ 







World

Robot

Мар



World

- •Indoor/Outdoor
- •2D/2.5D/3D
- •Static/Dynamic
- •Known/Unknown
- Abstract (web)

Мар

Robot

World

Robot

Мар

- Mobile
  - ➤ Indoor/Outdoor
  - ➤ Walking/Flying/Swimming
- Manipulator
- Humanoid
- Abstract



World

Robot

Мар

- Topological
- Metric
- •Feature Based
- •1D,2D,2.5D,3D



### World

- Indoor/Outdoor
- •2D/2.5D/3D
- Static/Dynamic
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### Robot

- Mobile
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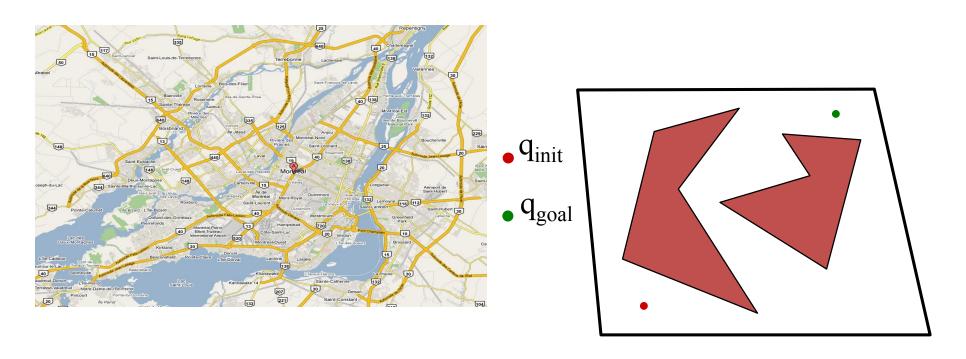
### Map

- Topological
- Metric
- Feature Based
- •1D,2D,2.5D,3D



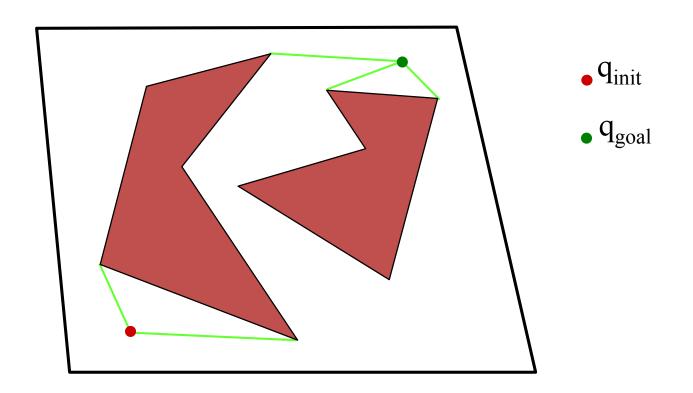
## **Path Planning: Assumptions**

- Known Map
- Roadmaps (Graph representations)
- Polygonal Representation



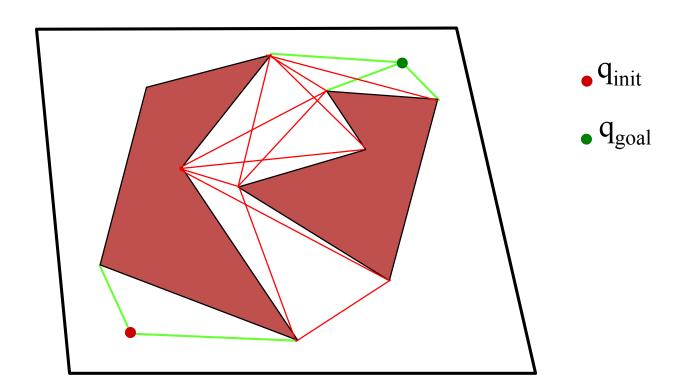


Connect Initial and goal locations with all the visible vertices

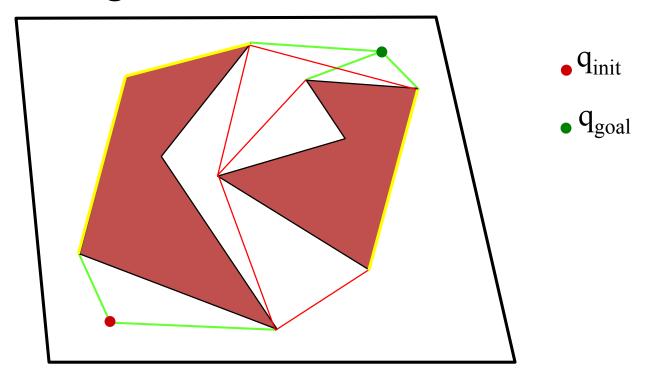




- Connect initial and goal locations with all the visible vertices
- Connect each obstacle vertex to every visible obstacle vertex



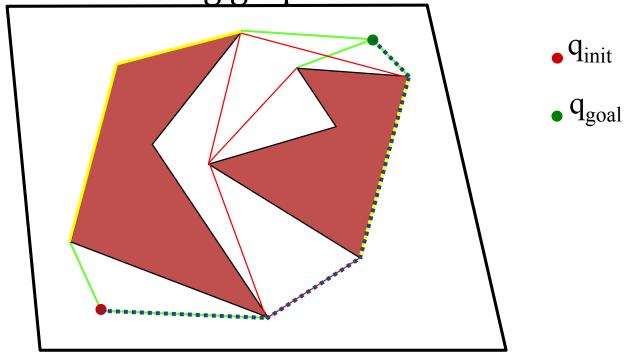
- Connect initial and goal locations with all the visible vertices
- Connect each obstacle vertex to every visible obstacle vertex
- Remove edges that intersect the interior of an obstacle





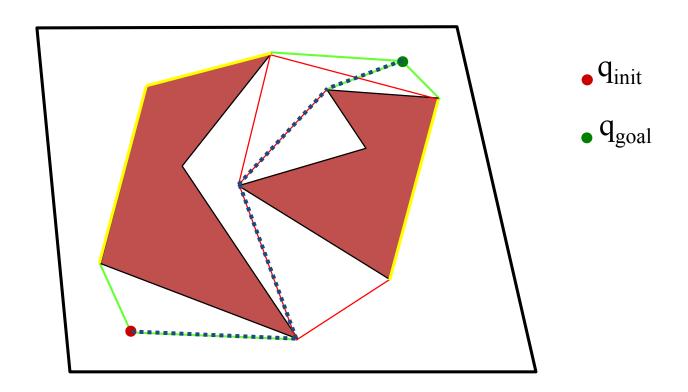
- Connect initial and goal locations with all the visible vertices
- Connect each obstacle vertex to every visible obstacle vertex
- Remove edges that intersect the interior of an obstacle

Plan on the resulting graph





- An alternative path
- Alternative name: "Rubber band algorithm"





## **Major Fault**

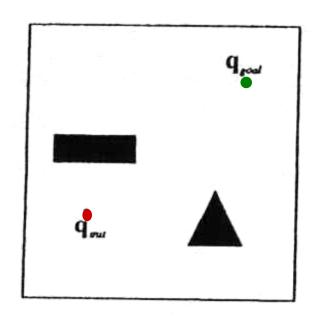
- Point robot
- Path planning like that guarantees to hit the obstacles

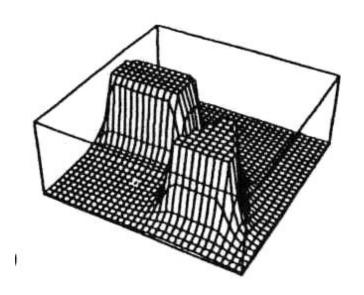




#### Potential Field methods

• compute a repulsive force away from obstacles



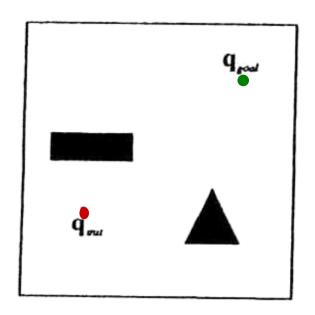


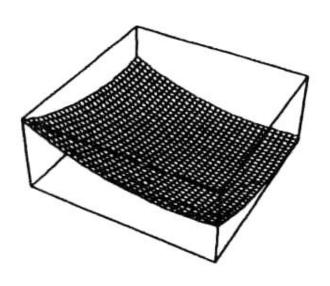


# Local techniques

#### Potential Field methods

- compute a repulsive force away from obstacles
- compute an attractive force toward the goal



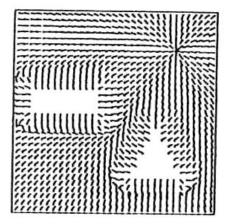


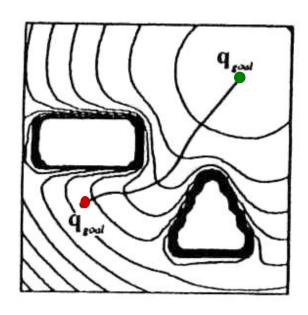


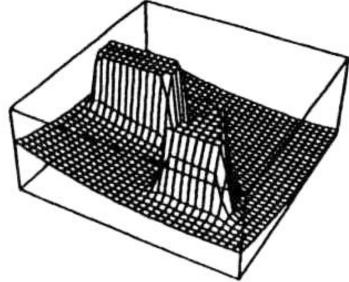
# Local techniques

#### Potential Field methods

- compute a repulsive force away from obstacles
- compute an attractive force toward the goal
- → let the sum of the forces control the robot





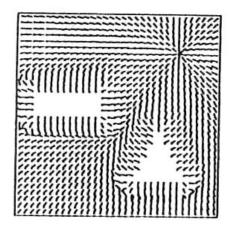


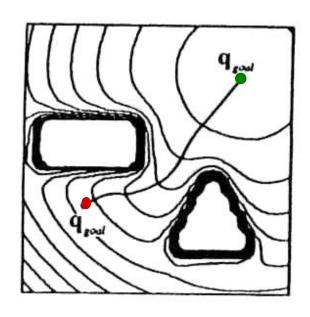


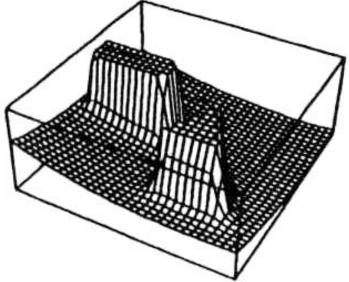
# Local techniques

#### Potential Field methods

- compute a repulsive force away from obstacles
- compute an attractive force toward the goal
- → let the sum of the forces control the robot

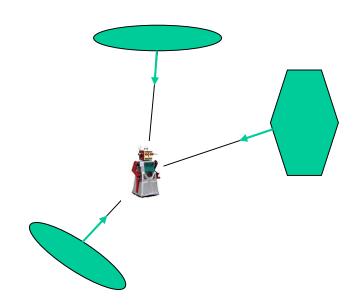








### Sensor Based Calculations

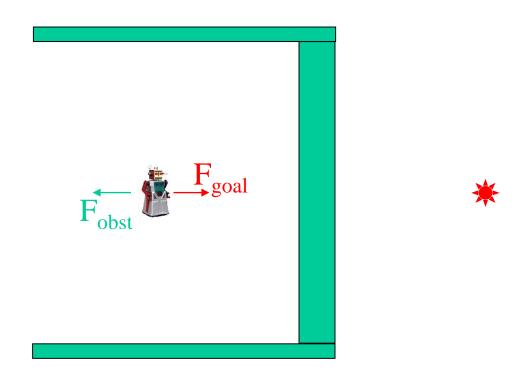




# Major Problem?



## Local Minima!





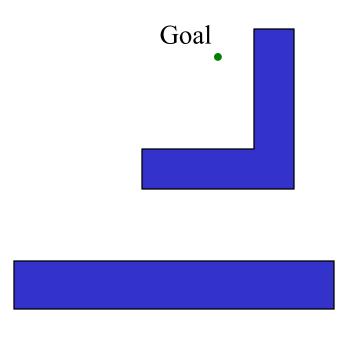
# Simulated Annealing

• Every so often add some random force



# Limited-knowledge path planning

- Path planning with limited knowledge
  - Insect-inspired "bug" algorithms



Start

- known direction to goal
- otherwise local sensing walls/obstacles encoders
- •"reasonable" world
- 1. finitely many obstacles in any finite disc
- a line will intersect an obstacle finitely many times

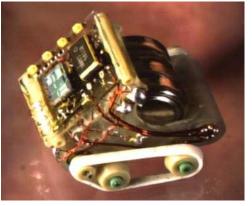


## Not truly modeling bugs...

Insects do use several cues for navigation:



visual landmarks
polarized light
chemical sensing



neither are the current bugsized robots

they're not ears...

Other animals use information from

magnetic fields

electric currents

temperature



CSCE-774 Robotic Systems bacteria

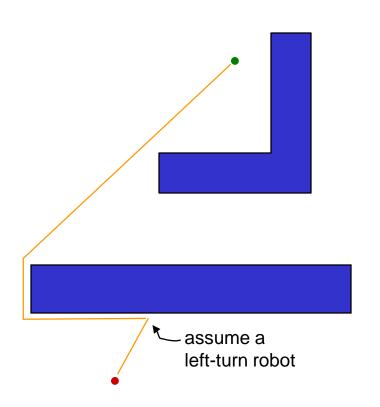


migrating bobolink



# Bug Strategy

#### Insect-inspired "bug" algorithms



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

#### "Bug 0" algorithm

- 1) head toward goal
- 2) follow obstacles until you can head toward the goal again
- 3) continue



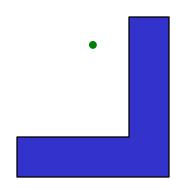
### Does It Work?



# Bug 1

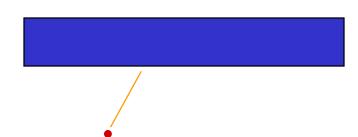
Insect-inspired "bug" algorithms

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



"Bug 1" algorithm

1) head toward goal



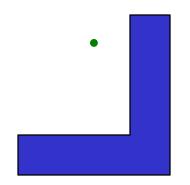


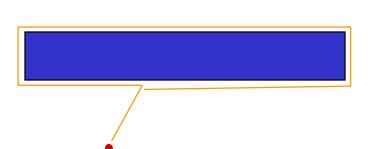
# Bug 1

Insect-inspired "bug" algorithms



 otherwise only local sensing walls/obstacles encoders





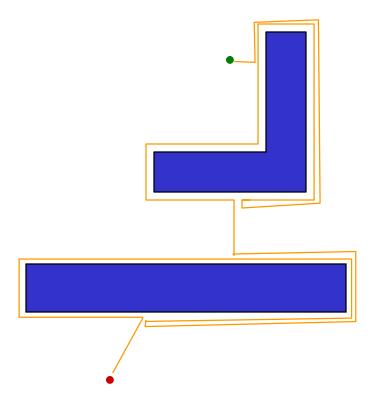
### "Bug 1" algorithm

- 1) head toward goal
- 2) if an obstacle is encountered, circumnavigate it *and* remember how close you get to the goal



# Bug 1

#### Insect-inspired "bug" algorithms



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

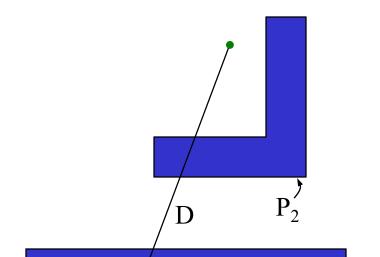
#### "Bug 1" algorithm

- 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 analysis

Distance Traveled



 $P_1$ 

What are bounds on the path length that the robot takes?

Available Information:

D = straight-line distance from start to goal

 $P_i$  = perimeter of the *i* th obstacle

Lower and upper bounds?

Lower bound:

Upper bound:



# Bug 1 analysis

Distance Traveled

What are bounds on the path length that the robot takes?



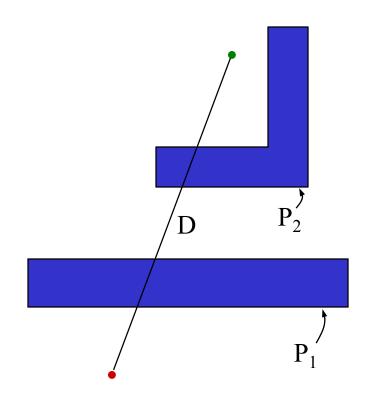
D = straight-line distance from start to goal

 $P_i$  = perimeter of the *i* th obstacle

Lower and upper bounds?

Lower bound: D

Upper bound:





# Bug 1 analysis

#### Distance Traveled

 $P_1$ 

What are bounds on the path length that the robot takes?

Available Information:

D = straight-line distance from start to goal

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Lower and upper bounds?

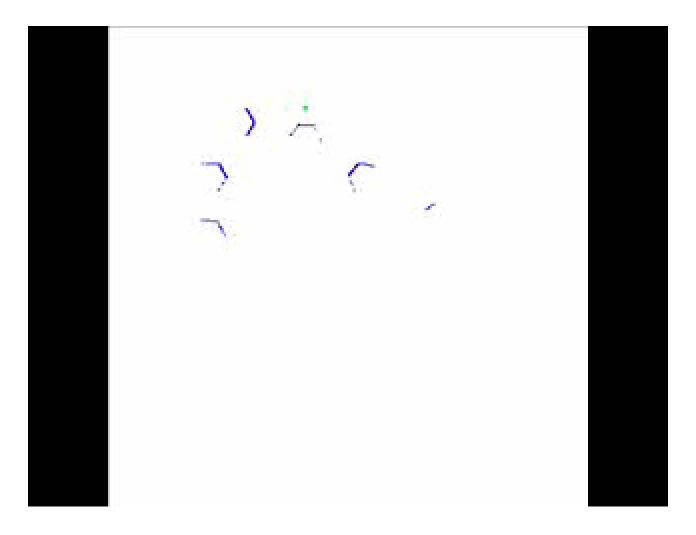
Lower bound: D

Upper bound:  $D + 1.5 \sum_{i} P_{i}$ 

How good a bound?

How good an algorithm?

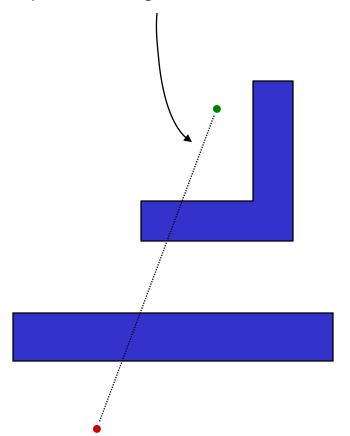
# Bug Mapping





# A better bug?

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

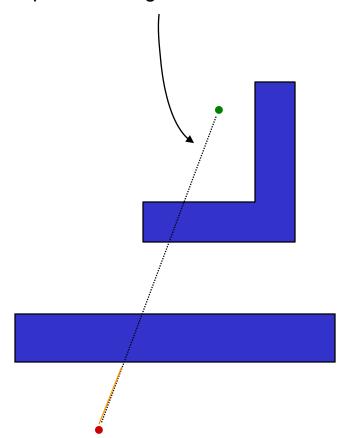


"Bug 2" algorithm



# A better bug?

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



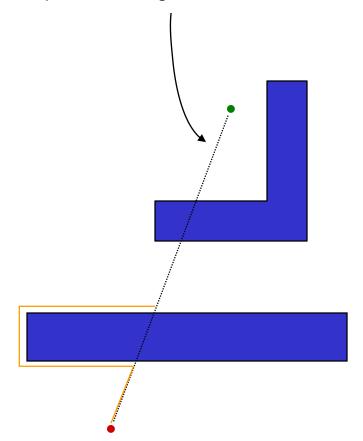
"Bug 2" algorithm

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



# A better bug?

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



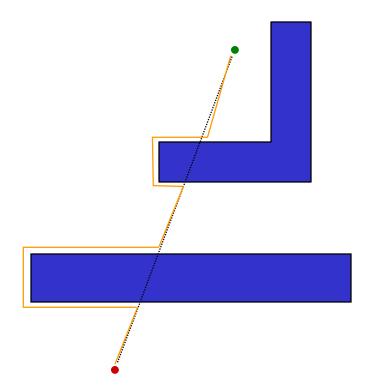
"Bug 2" algorithm

- 1) head toward goal on the *s-line*
- 2) if an obstacle is in the way, follow it until encountering the sline again.



# A better bug?

s-line

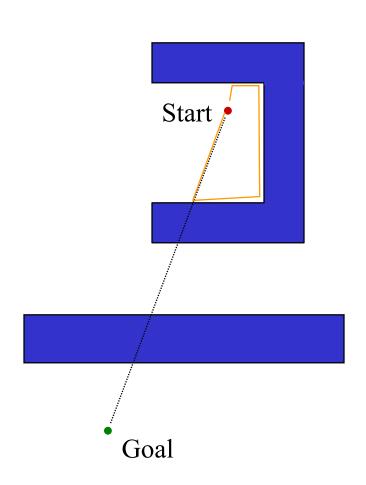


## "Bug 2" algorithm

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



# A better bug?

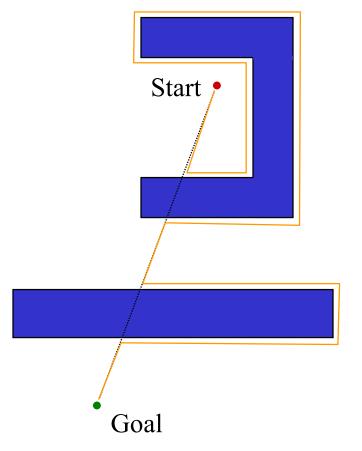


## "Bug 2" algorithm

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



Distance Traveled



What are bounds on the path length that the robot takes?

Available Information:

D = straight-line distance from start to goal

 $P_i$  = perimeter of the *i* th obstacle

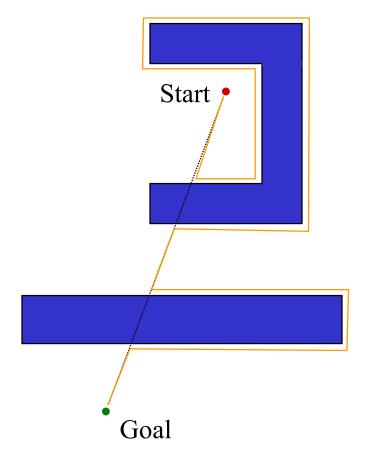
Lower and upper bounds?

Lower bound:

Upper bound:



Distance Traveled



What are bounds on the path length that the robot takes?

Available Information:

D = straight-line distance from start to goal

 $P_i$  = perimeter of the *i* th obstacle

 $N_i$  = number of s-line intersections with the i th obstacle

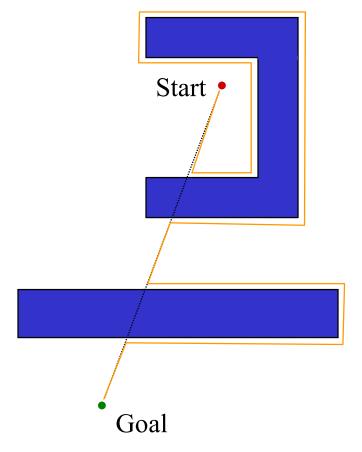
Lower and upper bounds?

Lower bound:

Upper bound:



Distance Traveled



What are bounds on the path length that the robot takes?

Available Information:

D = straight-line distance from start to goal

 $P_i$  = perimeter of the *i* th obstacle

 $N_i$  = number of s-line intersections with the i th obstacle

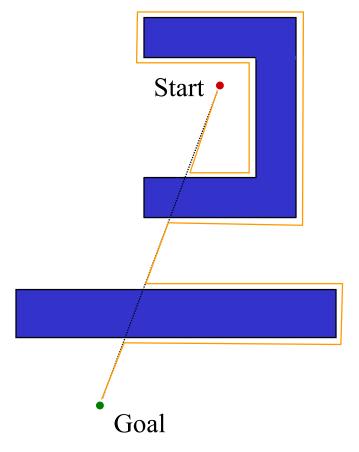
Lower and upper bounds?

Lower bound: D

Upper bound:



### Distance Traveled



What are bounds on the path length that the robot takes?

Available Information:

D = straight-line distance from start to goal

 $P_i$  = perimeter of the *i* th obstacle

 $N_i$  = number of s-line intersections with the i th obstacle

Lower and upper bounds?

Lower bound: D

Upper bound:  $D + 0.5 \sum_{i} N_i P_i$ 



# head-to-head comparison

or thorax-to-thorax, perhaps

What are worlds in which Bug 2 does better than Bug 1 (and vice versa)?

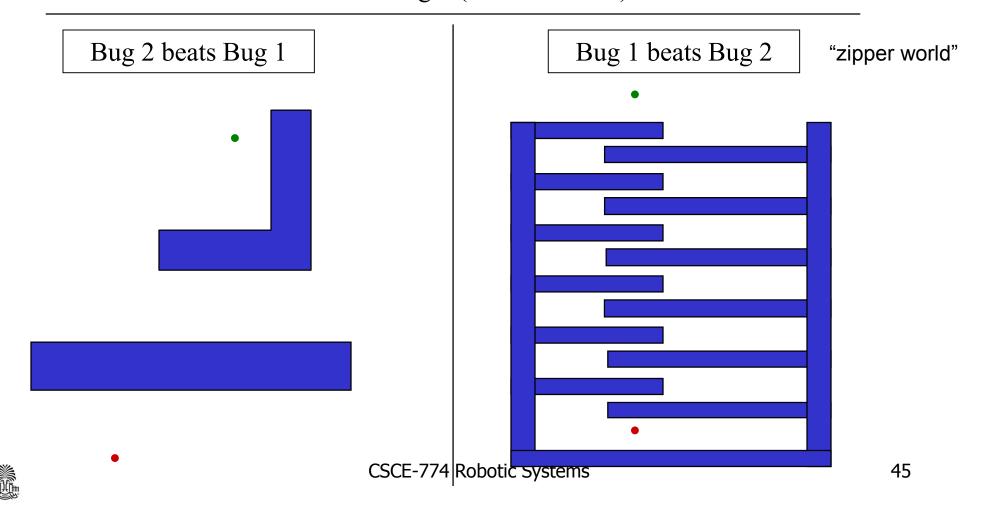
Bug 2 beats Bug 1 Bug 1 beats Bug 2



# head-to-head comparison

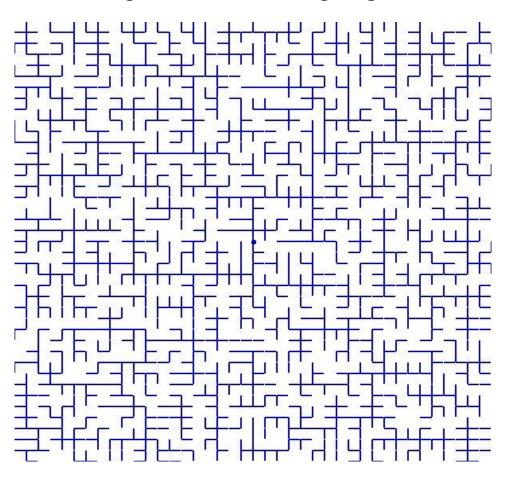
or thorax-to-thorax, perhaps

What are worlds in which Bug 2 does better than Bug 1 (and vice versa)?



# Other bug-like algorithms

## The Pledge maze-solving algorithm

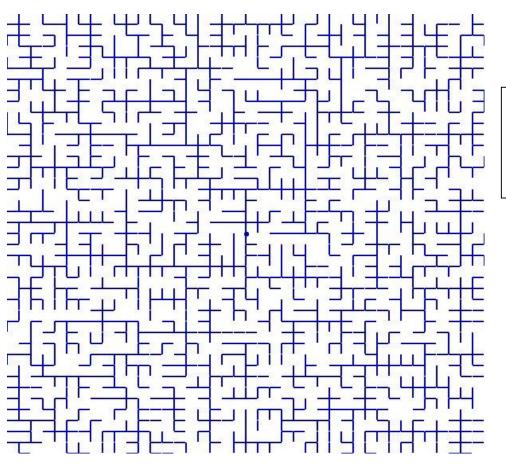


- 1. Go to a wall
- 2. Keep the wall on your right
- 3. Continue until out of the maze



# Other bug-like algorithms

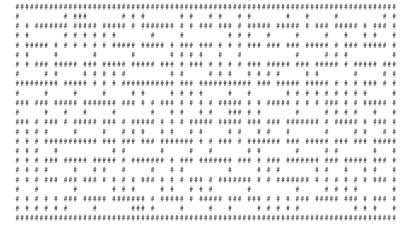
### The Pledge maze-solving algorithm



- 1) Go to a wall
- 2) Keep the wall on your right
- 3) Continue until out of the maze

int a[1817]; main(z,p,q,r){for(p=80;q+p-80;p=2\*a[p]) for(z=9;z--;)q=3&(r=time(0)+r\*57)/7,q=q?q-1?q-2?1-p%79?-1:0:p%79-77?1:0:p<1659?79:0:p>158?-79:0,q?!a[p+q\*2]?a[p+=a[p+=q]=q]=q:0:0;for(;q++-1817;)printf(q%79?"%c":"%c\n"," #"[!a[q-1]]);}

#### IOCCC random maze generator



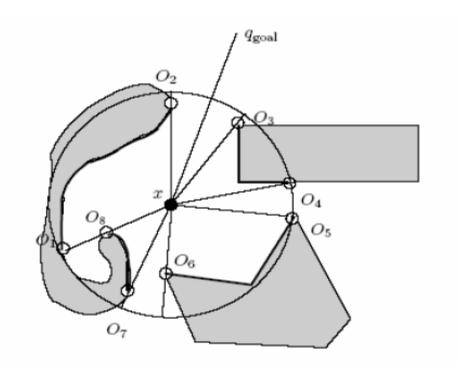


discretized RRT

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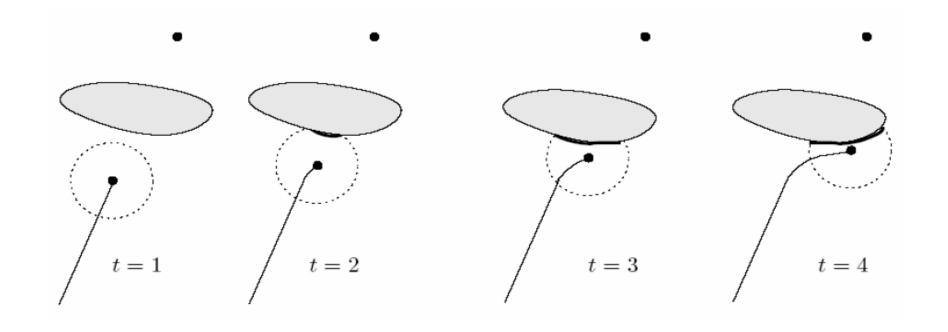
# Tangent Bug

- Limited Range Sensor
- Tangent Bug relies on finding endpoints of finite, continues segments of the obstacles



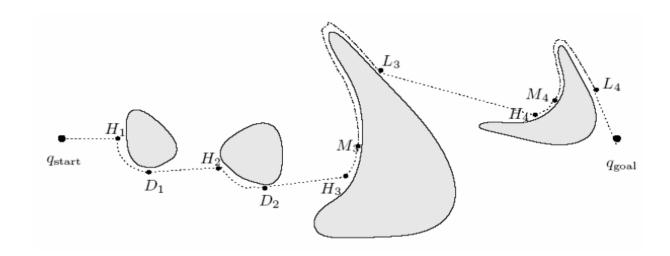


# Tangent Bug





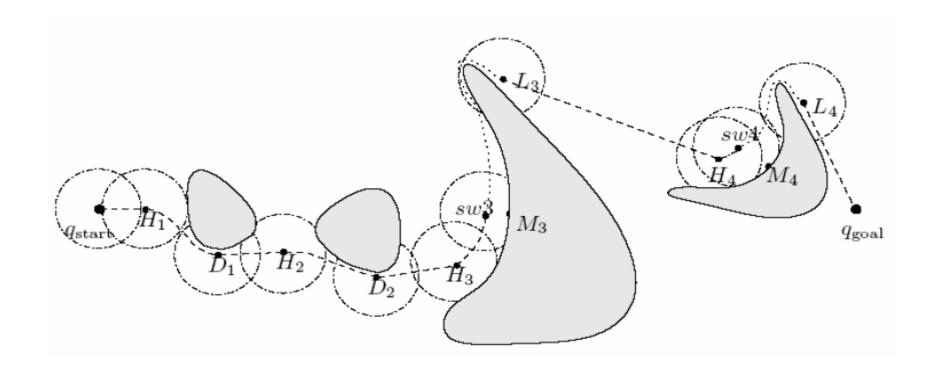
# Contact Sensor Tangent Bug



- 1. Robot moves toward goal until it hits obstacle 1 at H1
- 2. Pretend there is an infinitely small sensor range and the direction which minimizes the heuristic is to the right
- 3. Keep following obstacle until robot can go toward obstacle again
- 4. Same situation with second obstacle
- 5. At third obstacle, the robot turned left until it could not increase heuristic
- D\_followed is distance between M3 and goal, d\_reach is distance between robot and goal because sensing distance is zero

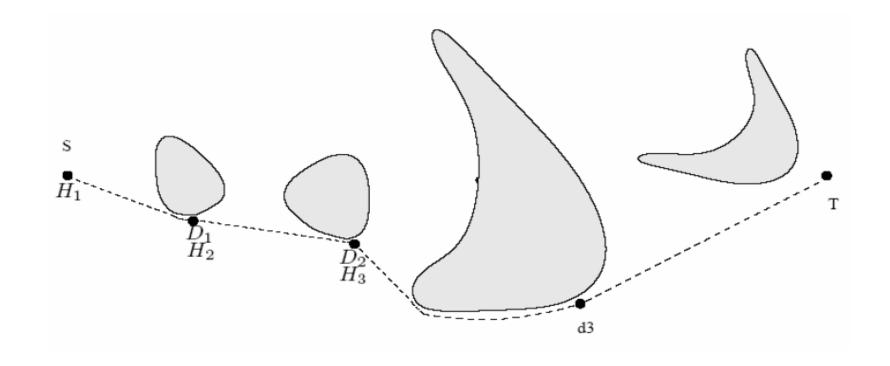


# Limited Sensor Range Tangent-Bug





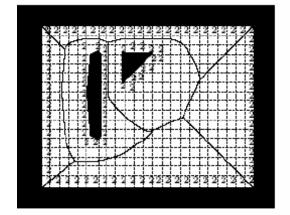
# Infinite Sensor Range Tangent Bug

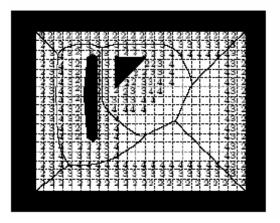


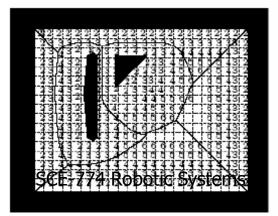


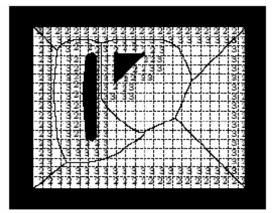
# Known Map

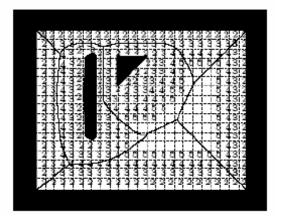
**Brushfire Transform** 

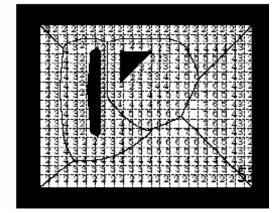






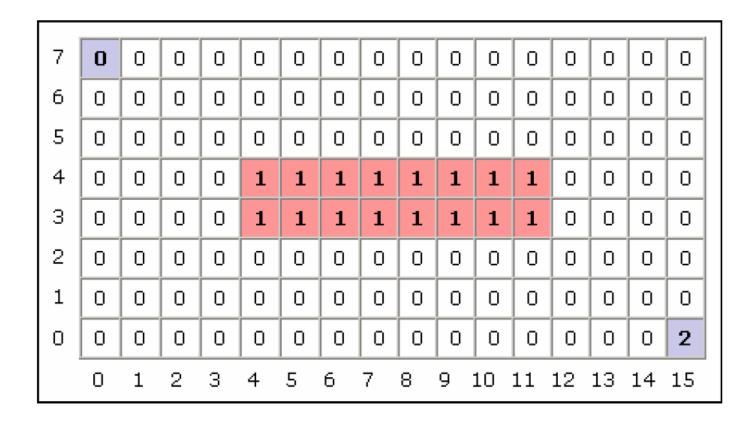








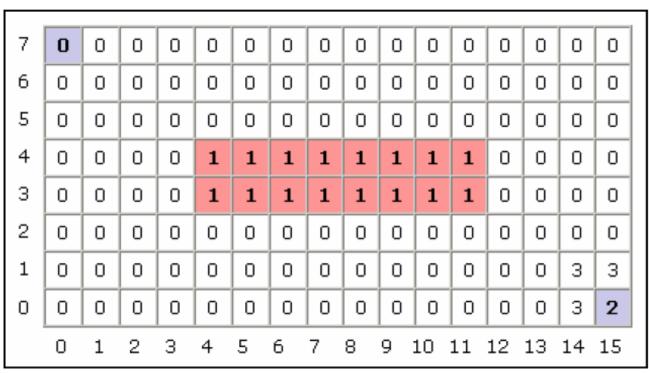
# The Wavefront Planner: Setup





## The Wavefront in Action (Part 1)

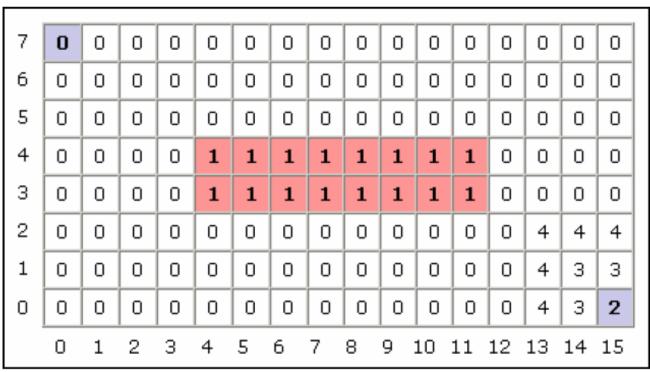
- Starting with the goal, set all adjacent cells with "0" to the current cell + 1
  - 4-Point Connectivity or 8-Point Connectivity?
  - Your Choice. We'll use 8-Point Connectivity in our example





## The Wavefront in Action (Part 2)

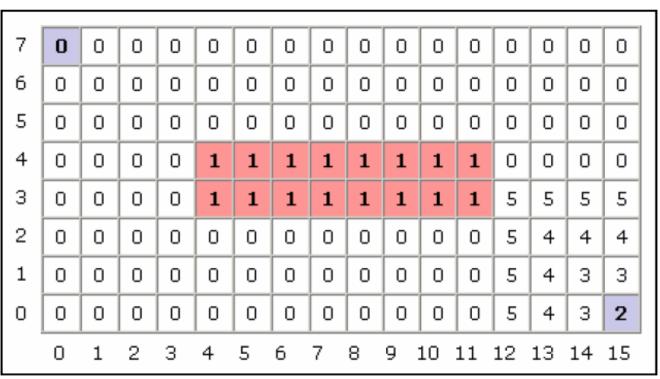
- Now repeat with the modified cells
  - This will be repeated until no 0's are adjacent to cells with values  $\geq 2$
- 0's will only remain when regions are unreachable





# The Wavefront in Action (Part 3)

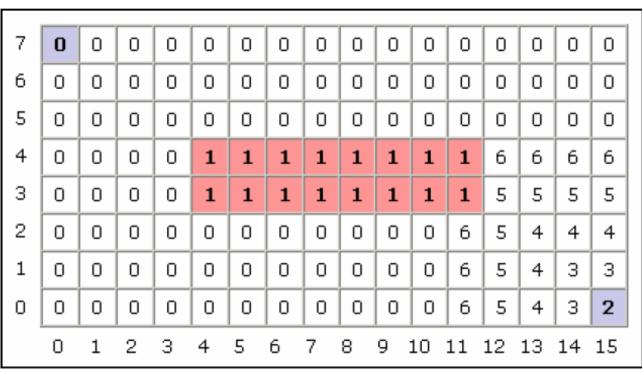
Repeat





# The Wavefront in Action (Part 3)

Repeat





# The Wavefront in Action (Part 3)

#### Until Done

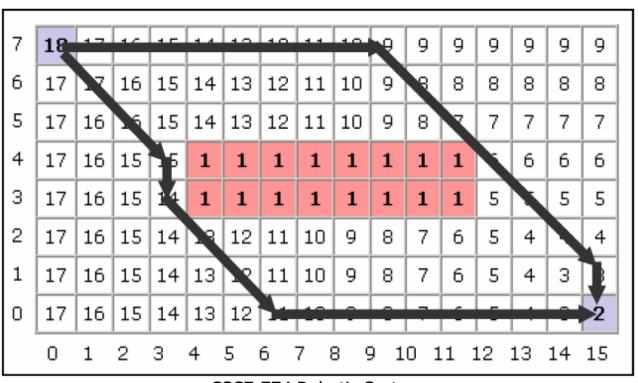
- 0's would only remain in the unreachable areas

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



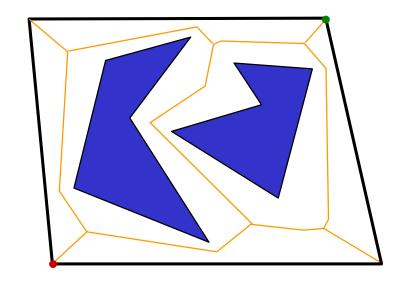
## The Wavefront in Action

- 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

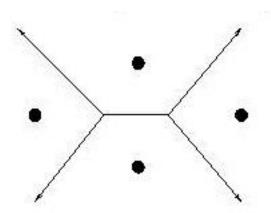




# An alternative roadmap



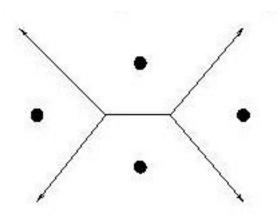




These line segments make up the **Voronoi diagram** for the four points shown here.

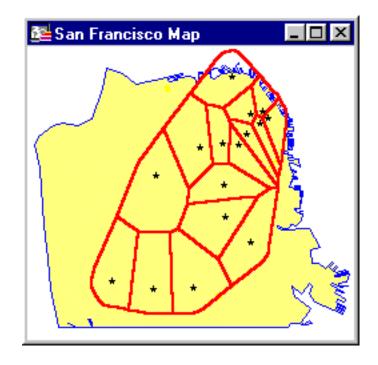
Solves the "Post Office Problem"





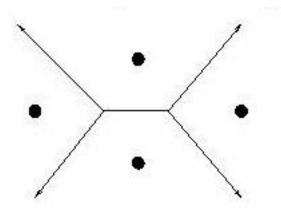
These line segments make up the **Voronoi diagram** for the four points shown here.

Solves the "Post Office Problem"

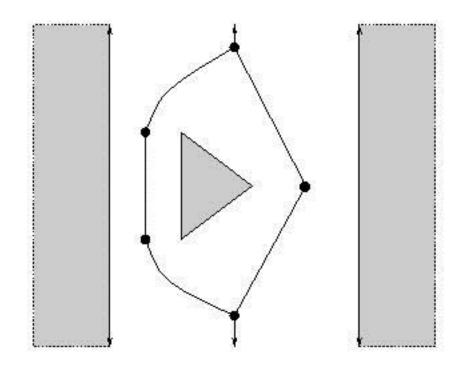


or, perhaps, more important problems...





"true" Voronoi diagram (isolates a set of points)

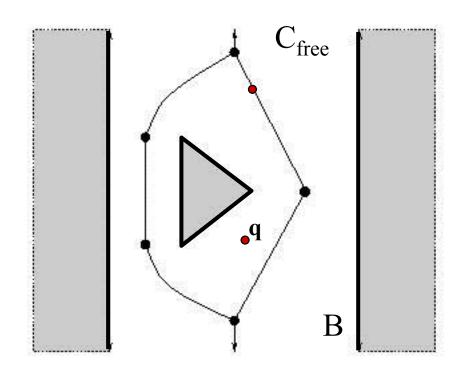


generalized Voronoi diagram What is it?



Let B =the boundary of  $C_{free}$ .

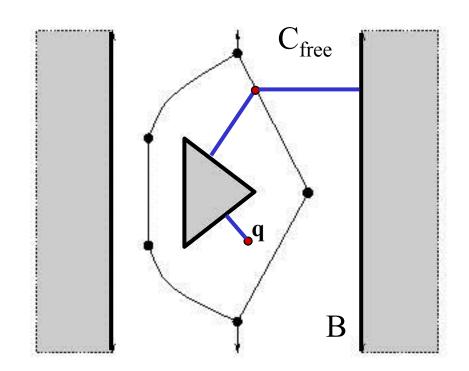
Let **q** be a point in  $C_{free}$ . ( • )





Let B = the boundary of  $C_{free}$ .

Let  $\mathbf{q}$  be a point in  $C_{free}$ .

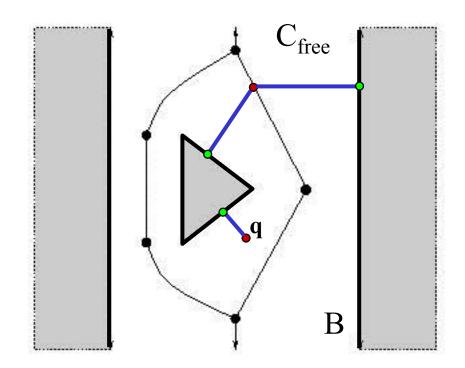


Define *clearance*(q) = min  $\{ | q - p | \}$ , for all  $p \in B$ 



Let B = the boundary of  $C_{free}$ .

Let  $\mathbf{q}$  be a point in  $C_{free}$ .



Define *clearance*(q) = min  $\{ | q - p | \}$ , for all  $p \in B$ 

Define  $near(q) = \{ p \in B \text{ such that } | q - p | = clearance(q) \}$ 

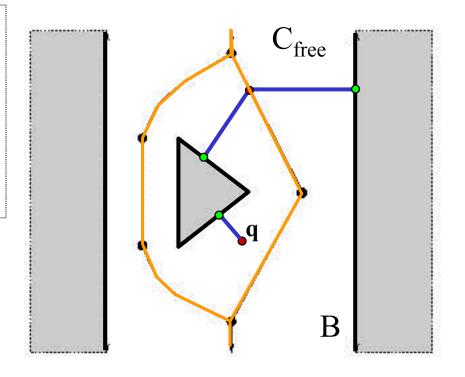


#### **Evaluation**

- + maximizes distance from obstacles
- + reduces to graph search
- + can be used in higher-dimensions
- nonoptimal
- real diagrams tend to be noisy

Let B =the boundary of  $C_{free}$ .

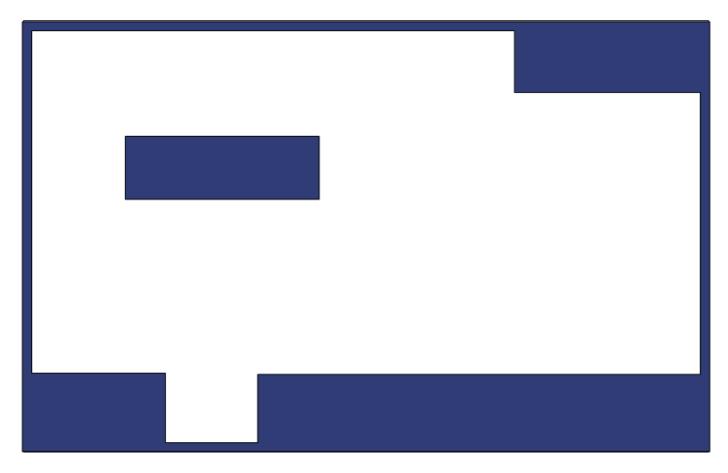
Let  $\mathbf{q}$  be a point in  $C_{\text{free}}$ .



Define 
$$clearance(q) = min \{ | q - p | \}$$
, for all  $p \in B$   
Define  $near(q) = \{ p \in B \text{ such that } | q - p | = clearance(q) \}$   
 $q$  is in the  $Coronoi\ diagram\ of\ C_{free}\ if | near(q) | > 1$ 

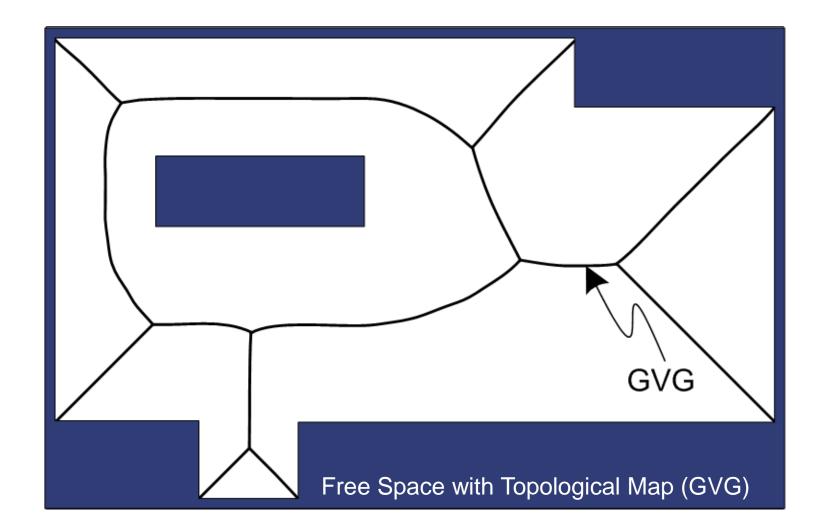
$$CSCE-774\ Robotic\ Systems$$
number of set elements

## Generalized Voronoi Graph (GVG)



# CSCE-774 Robotic Systems

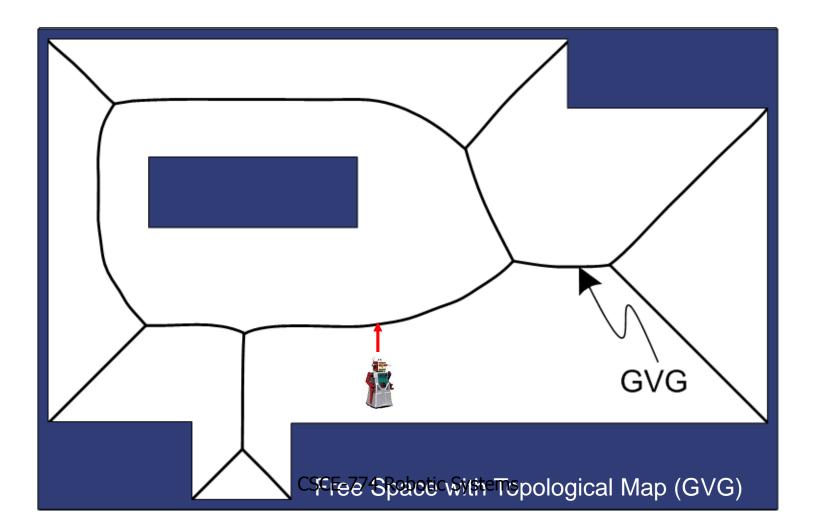
# Generalized Voronoi Graph (GVG)





# Generalized Voronoi Graph (GVG)

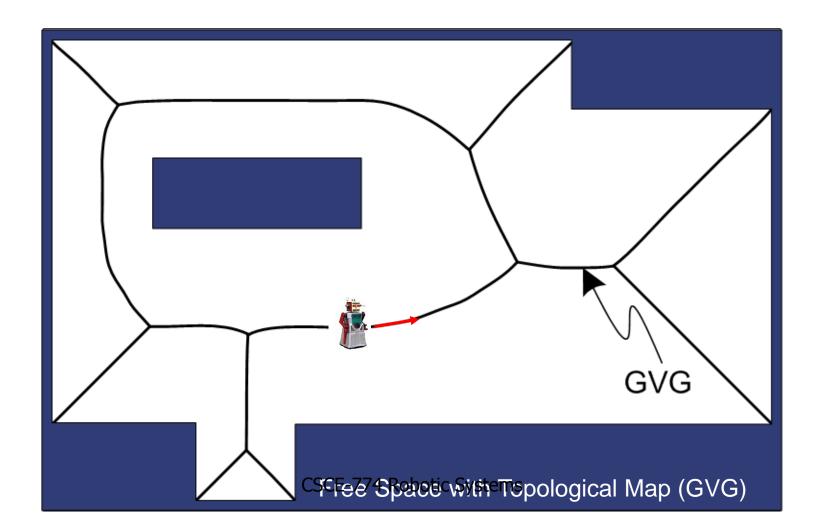
Access GVG





# Generalized Voronoi Graph (GVG)

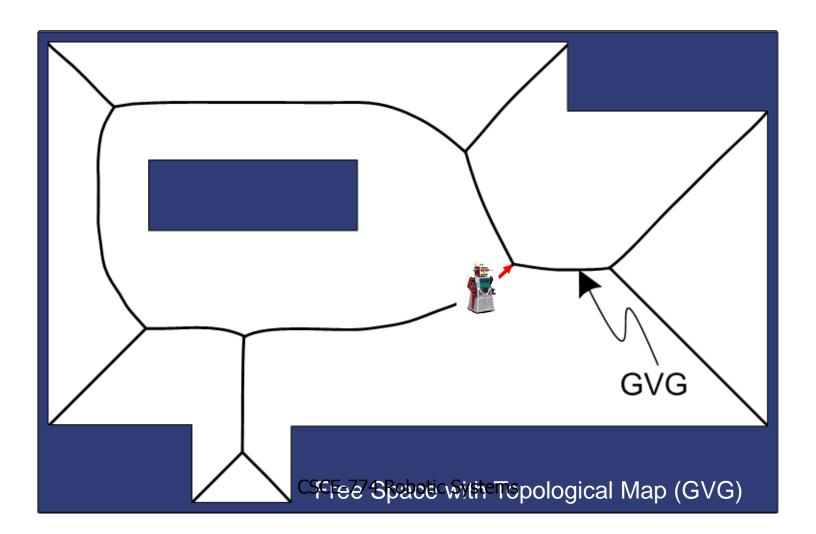
- Access GVG
- •Follow Edge





# Generalized Voronoi Graph (GVG)

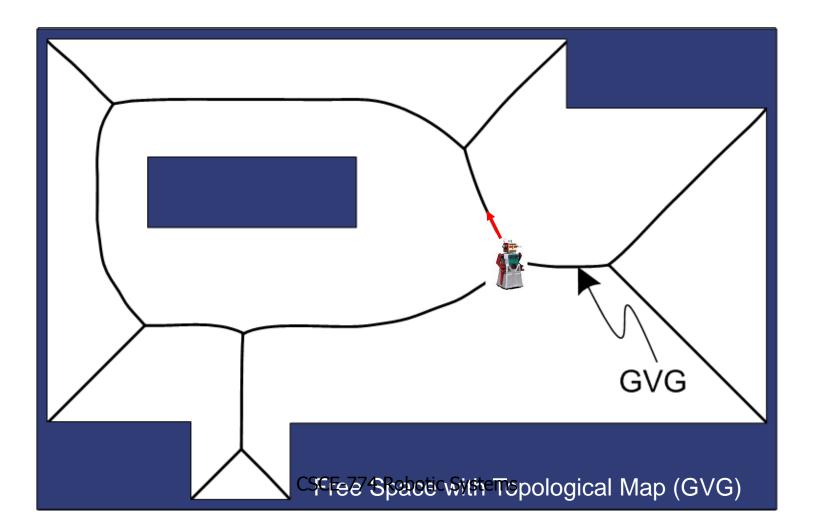
- •Access GVG •Home to the MeetPoint
- •Follow Edge





# Generalized Voronoi Graph (GVG)

- •Access GVG •Home to the MeetPoint
- •Follow Edge •Select Edge





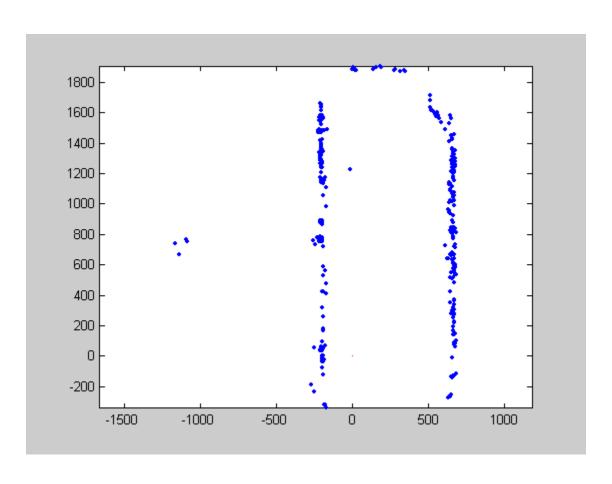
# GVG construction using sonar



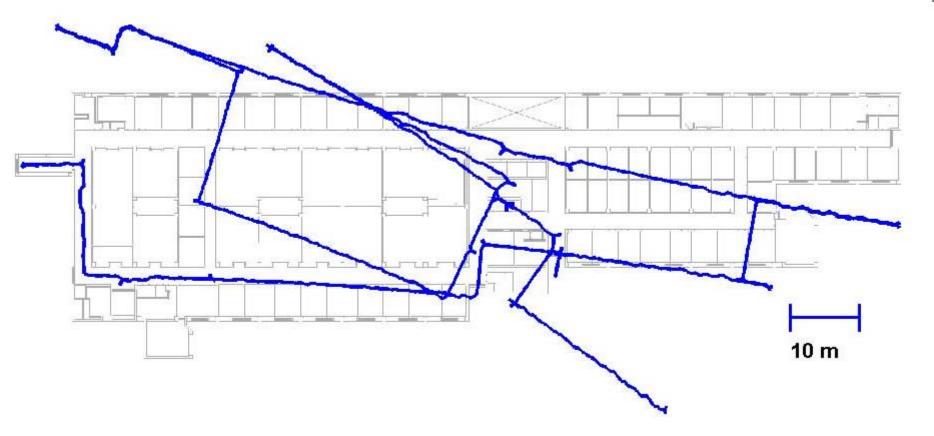
- Nomadic Scout
- Sonar (GVG navigation)
- Camera with omni-directional mirror (feature detection)
- Onboard 1.2 GHz processor



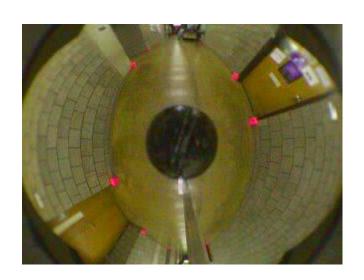
# GVG construction using sonar



# GVG construction using sonar



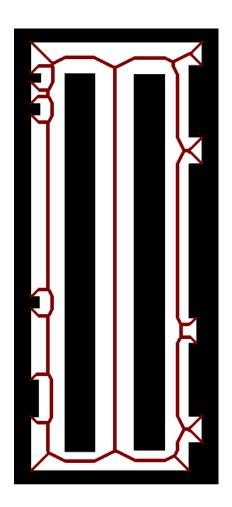
## Slammer in Action

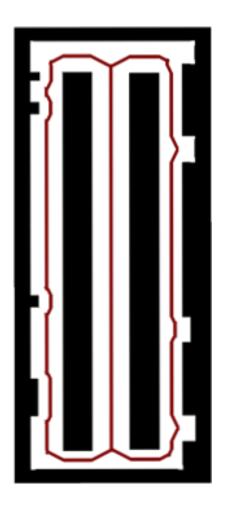






# Removing Edges



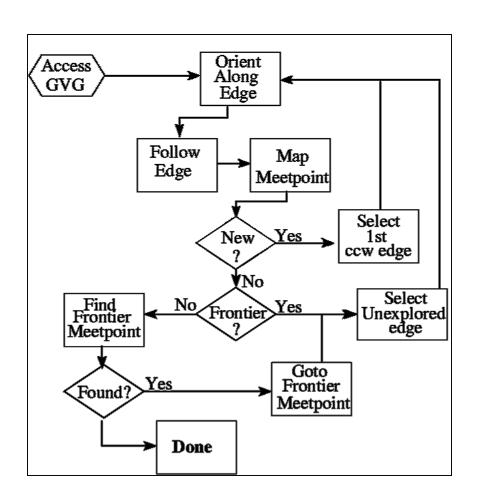


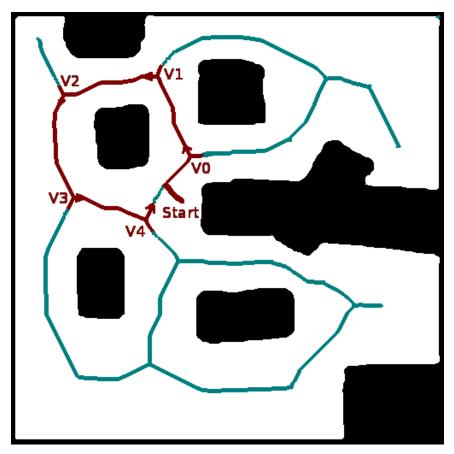


# Meetpoint Detection

- 3σ uncertainty ellipse of explored meetpoints
- Meetpoint degree (branching factor)
- Distances to local obstacles
- Relative angle bearings
- Edge signature
  - Edge length
  - Edge Curvature
- Vertex signal

# Ear-based Exploration

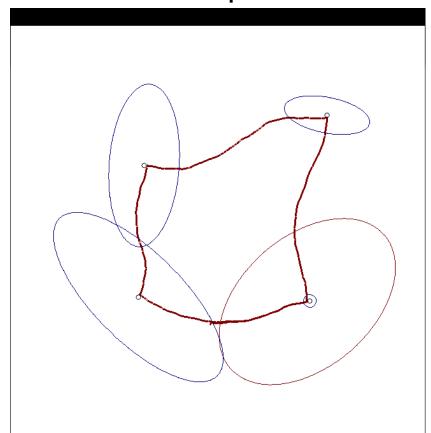




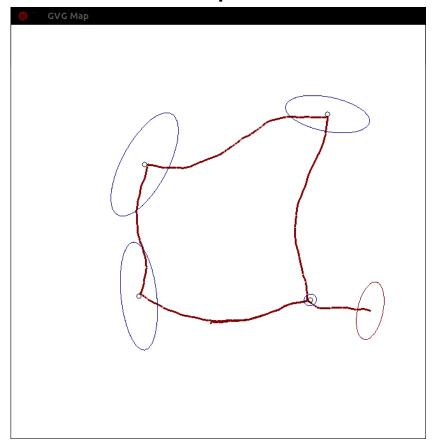


# Uncertainty Reduction

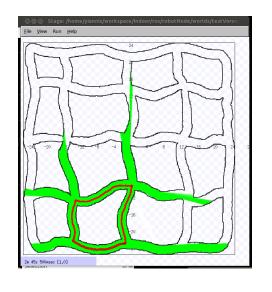
#### Before Loop-closure

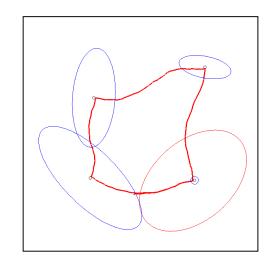


#### After Loop-closure

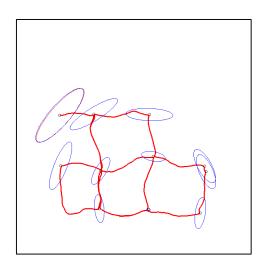


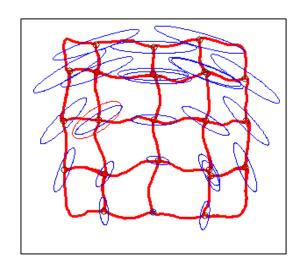






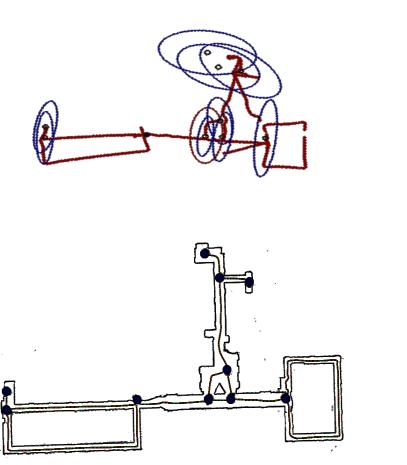
## Simulation

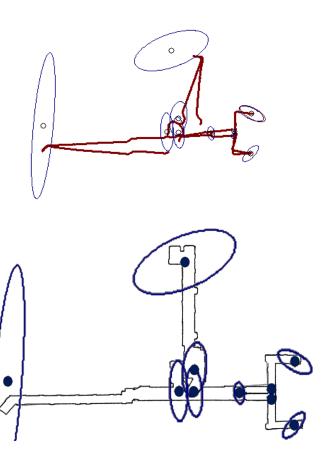






## Real Environment







# Ear-based Exploration on Hybrid Metric/Topological Maps

#### Ear-based Exploration on Hybrid Metric/Topological Maps

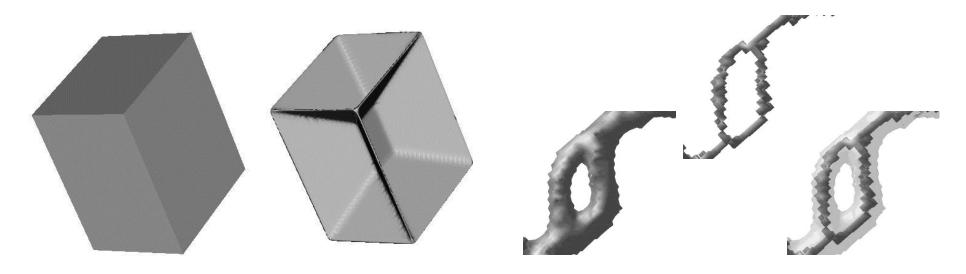


Q. Zhang, D. Whitney, F. Shkurti, and I. Rekleitis School of Computer Science, McGill University

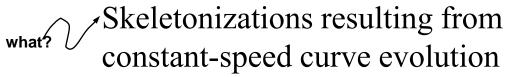




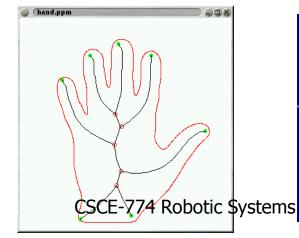
# Voronoi applications

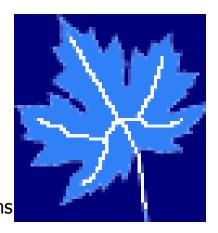


A retraction of a 3d object == "medial surface"





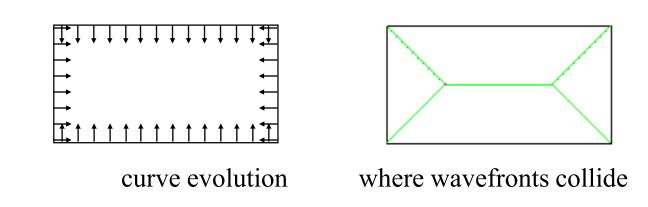


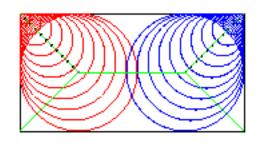


in 2d, it's called a medial axis



# skeleton → shape



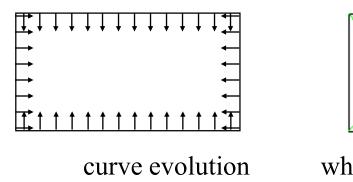


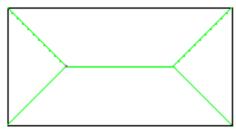
centers of maximal disks

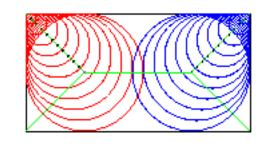
again reduces a 2d (or higher) problem to a question about graphs...



# skeleton → shape





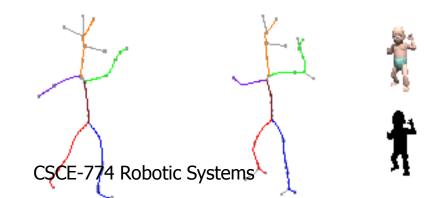


where wavefronts collide

centers of maximal disks

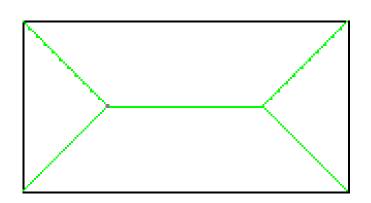
again reduces a 2d (or higher) problem to a question about graphs...

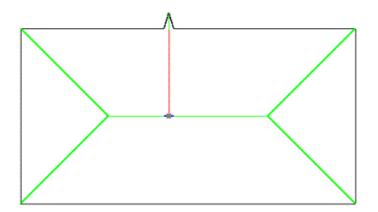






### **Problems**





The skeleton is sensitive to small changes in the object's boundary.



# Roadmap problems

If an obstacle decides to roll away... (or wasn't there to begin with)

