

Robotic Motion Planning: Bug Algorithms

Robotics Institute 16-735

<http://voronoi.sbp.ri.cmu.edu/~motion>

Howie Choset

<http://voronoi.sbp.ri.cmu.edu/~choset>

What's Special About Bugs

- Many planning algorithms assume global knowledge
- Bug algorithms assume only *local* knowledge of the environment and a global goal
- Bug behaviors are simple:
 - 1) Follow a wall (right or left)
 - 2) Move in a straight line toward goal
- Bug 1 and Bug 2 assume essentially tactile sensing
- Tangent Bug deals with finite distance sensing

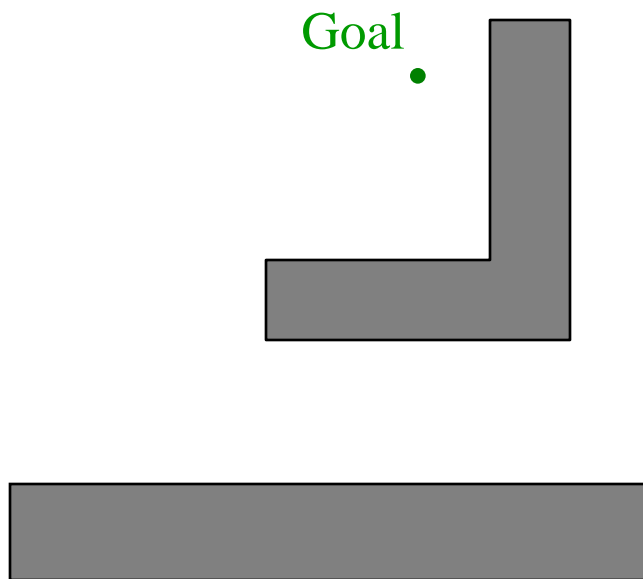
A Few General Concepts

- Workspace W
 - $\mathbb{R}(2)$ or $\mathbb{R}(3)$ depending on the robot
 - could be infinite (open) or bounded (closed/compact)
- Obstacle WO_i
- Free workspace $W_{free} = W \setminus \cup_i WO_i$

The *Bug* Algorithms

provable results...

Insect-inspired



• Start

- **known direction to goal**

- **robot can measure distance $d(x,y)$ between pts x and y**

- **otherwise local sensing**

- walls/obstacles & encoders

- **reasonable world**

- 1) finitely many obstacles in any finite area
- 2) a line will intersect an obstacle finitely many times
- 3) Workspace is bounded

$$W \subset B_r(x), r < \infty$$

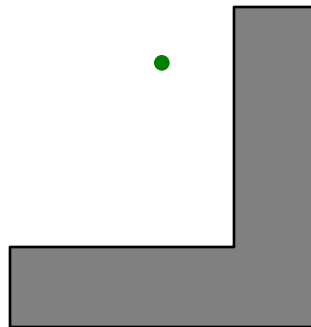
$$B_r(x) = \{ y \in \mathcal{R}(2) \mid d(x,y) < r \}$$

Buginner Strategy

"Bug 0" algorithm

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

walls/obstacles & encoders



Some notation:

q_{start} and q_{goal}

"hit point" q_i^H

"leave point" q_i^L

A *path* is a sequence of hit/leave pairs bounded by q_{start} and q_{goal}

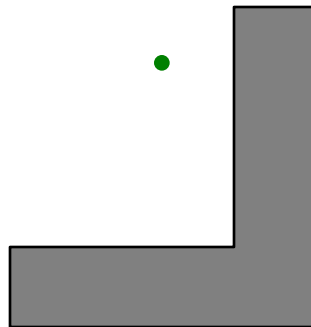
Buginner Strategy

"Bug 0" algorithm

- **known direction to goal**

- **otherwise local sensing**

walls/obstacles & encoders



1) head toward goal

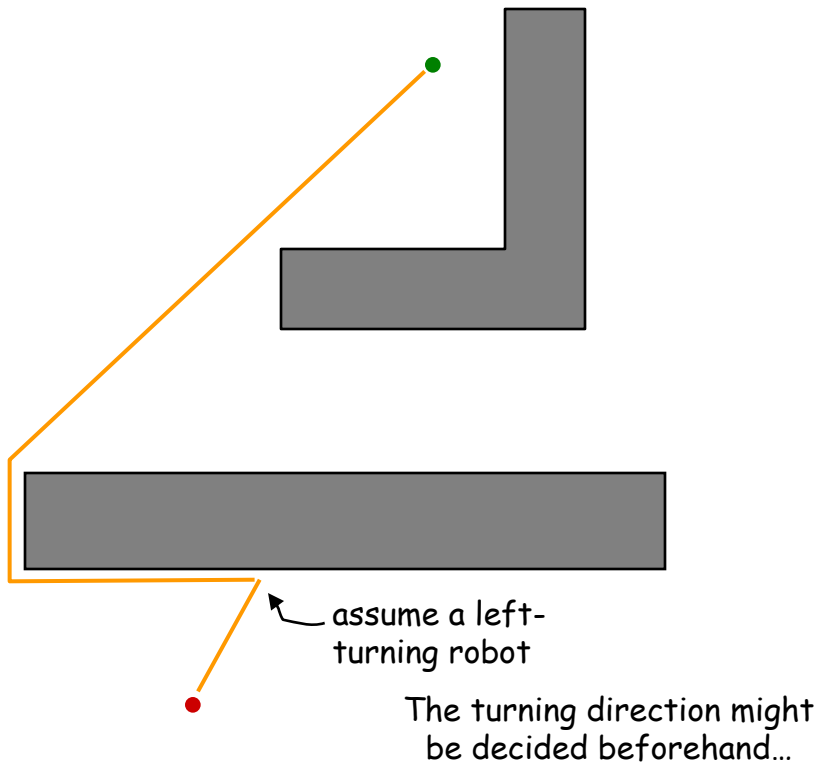
2) follow obstacles until you can head toward the goal again

3) continue



"Bug 0" algorithm

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



Bug Zapper

What map will foil Bug 0 ?

"Bug 0" algorithm

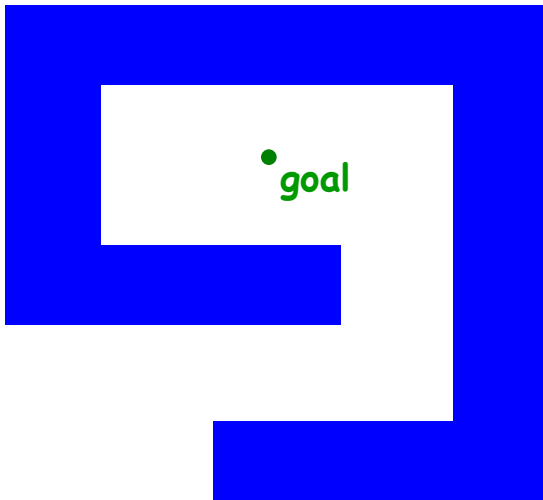
- 1) head toward goal
- 2) follow obstacles until you can head toward the goal again
- 3) continue

Bug Zapper

What map will foil Bug 0 ?

"Bug 0" algorithm

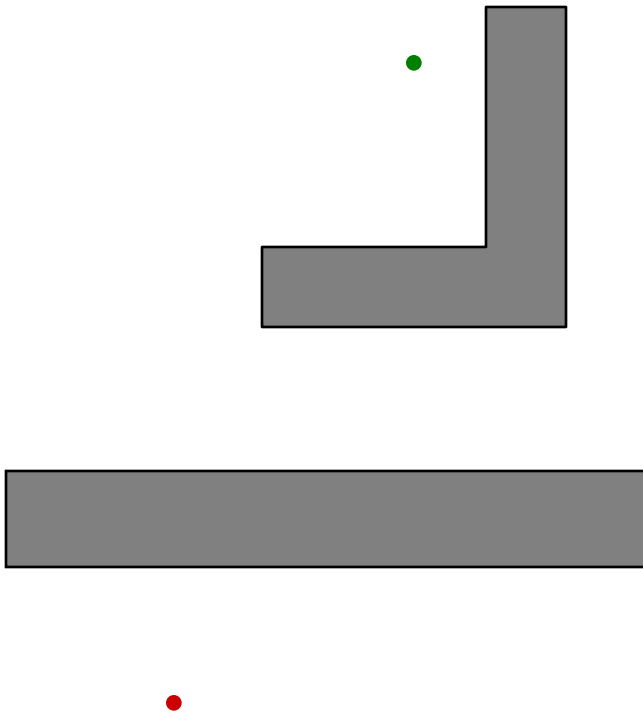
- 1) head toward goal
- 2) follow obstacles until you can head toward the goal again
- 3) continue



A better bug?

But add some memory!

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

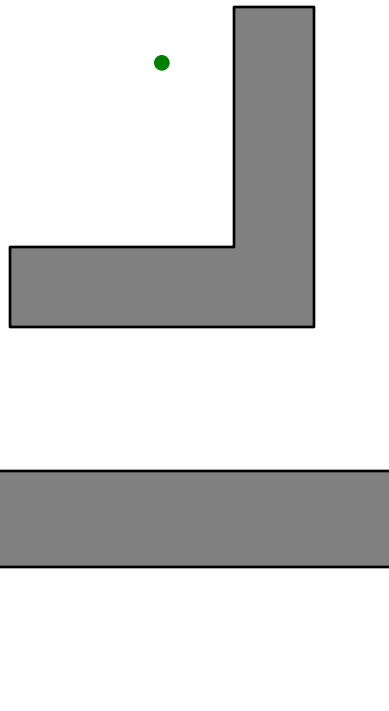




Bug 1

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

Vladimir Lumelsky & Alexander Stepanov: Algorithmica 1987

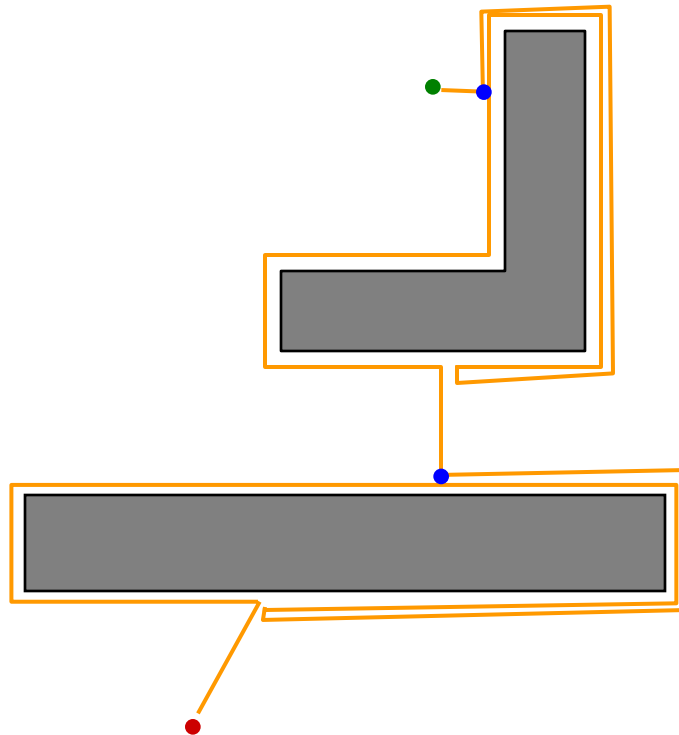
16-735, Howie Choset with slides from G.D. Hager and Z. Dodds



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 More formally

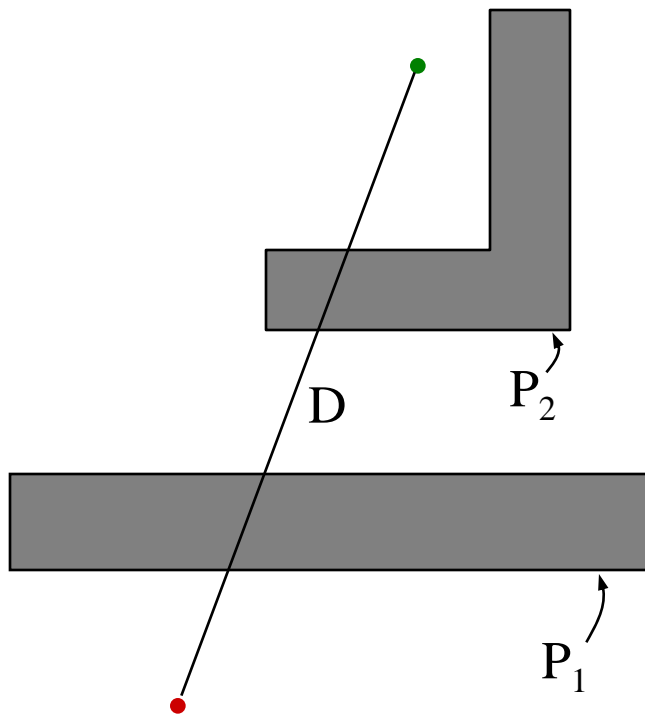
- Let $q_0^L = q_{\text{start}}$; $i = 1$
- repeat
 - repeat
 - from q_{i-1}^L move toward q_{goal}
 - until goal is reached or obstacle encountered at q_i^H
 - if goal is reached, exit
 - repeat
 - follow boundary recording pt q_i^L with shortest distance to goal
 - until q_{goal} is reached or q_i^H is re-encountered
 - if goal is reached, exit
 - Go to q_i^L
 - if move toward q_{goal} moves into obstacle
 - exit with failure
 - else
 - $i=i+1$
 - continue

“Quiz”

Bug 1 analysis

Bug 1: Path Bounds

What are upper/lower 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 bound:

What's the shortest distance it might travel?

Upper bound:

What's the longest distance it might travel?

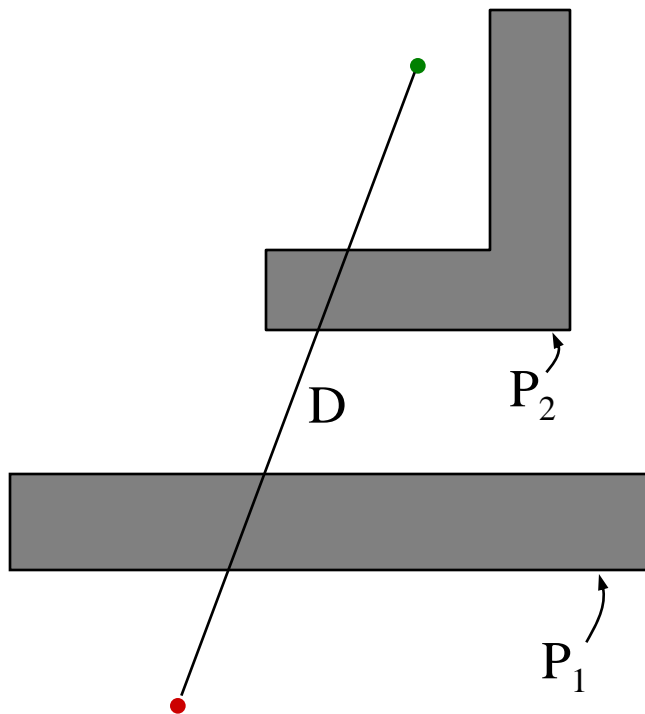
What is an environment where your upper bound is required?

“Quiz”

Bug 1 analysis

Bug 1: Path Bounds

What are upper/lower 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 bound:

What's the shortest distance it might travel?

D

Upper bound:

What's the longest distance it might travel?

$$D + 1.5 \sum_i P_i$$

What is an environment where your upper bound is required?

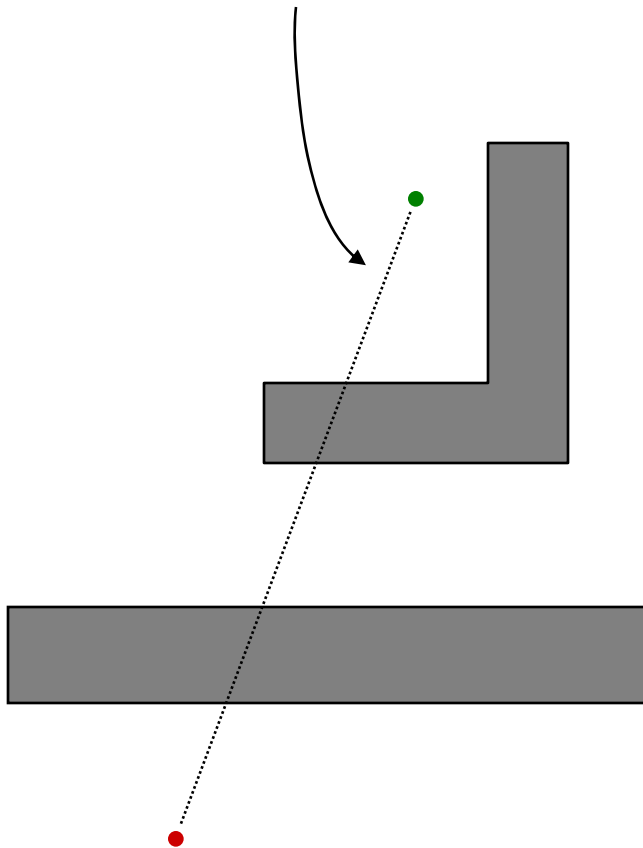
How Can We Show Completeness?

- An algorithm is *complete* if, in finite time, it finds a path if such a path exists or terminates with failure if it does not.
- Suppose BUG1 were incomplete
 - Therefore, there is a path from start to goal
 - By assumption, it is finite length, and intersects obstacles a finite number of times.
 - BUG1 does not find it
 - Either it terminates incorrectly, or, it spends an infinite amount of time
 - Suppose it never terminates
 - but each leave point is closer to the obstacle than corresponding hit point
 - Each hit point is closer than the last leave point
 - Thus, there are a finite number of hit/leave pairs; after exhausting them, the robot will proceed to the goal and terminate
 - Suppose it terminates (incorrectly)
 - Then, the closest point after a hit must be a leave where it would have to move into the obstacle
 - But, then line from robot to goal must intersect object even number of times (Jordan curve theorem)
 - But then there is another intersection point on the boundary closer to object. Since we assumed there is a path, we must have crossed this pt on boundary which contradicts the definition of a leave point.

Another step forward?

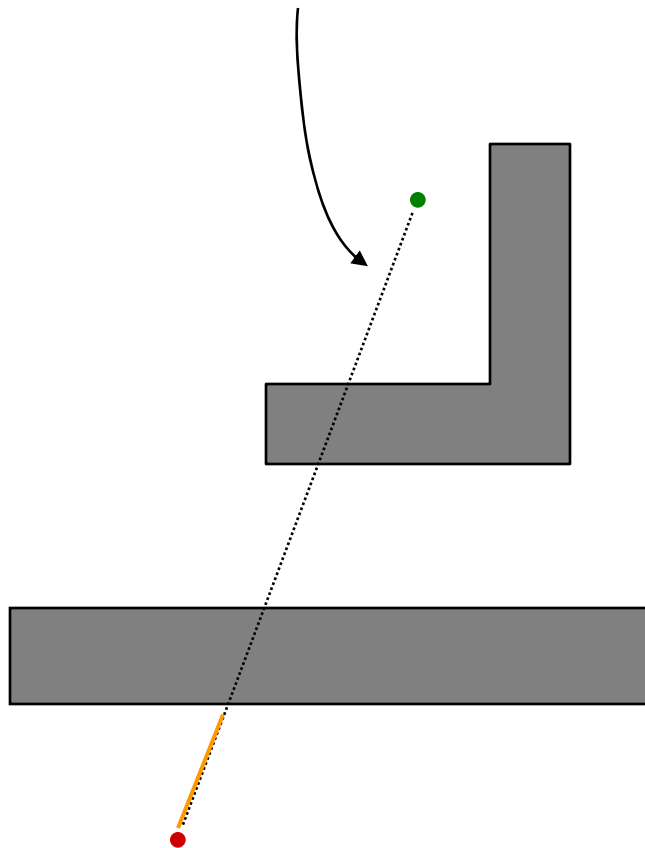
Call the line from the starting point to the goal the *m-line*

"Bug 2" Algorithm



A better bug?

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

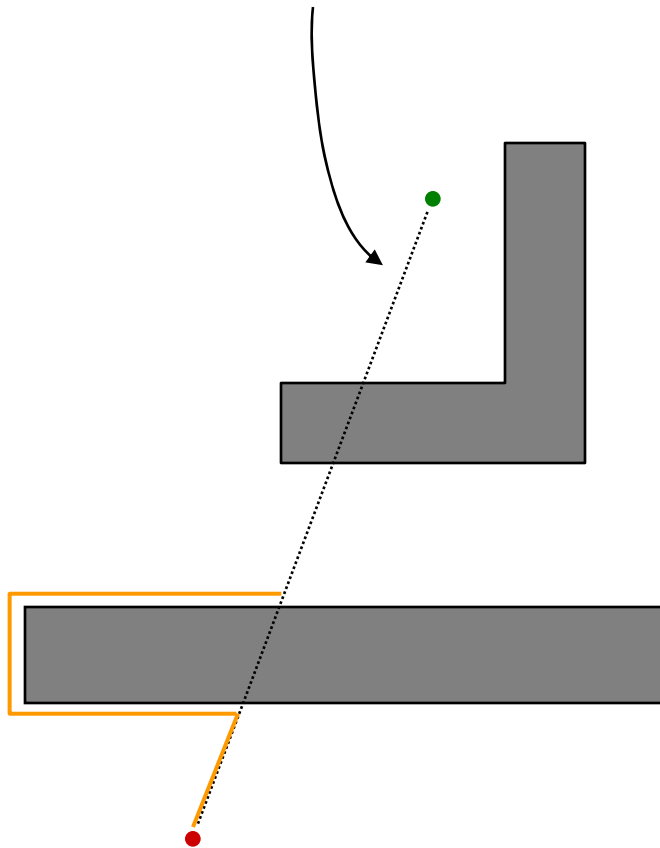


"Bug 2" Algorithm

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

A better bug?

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

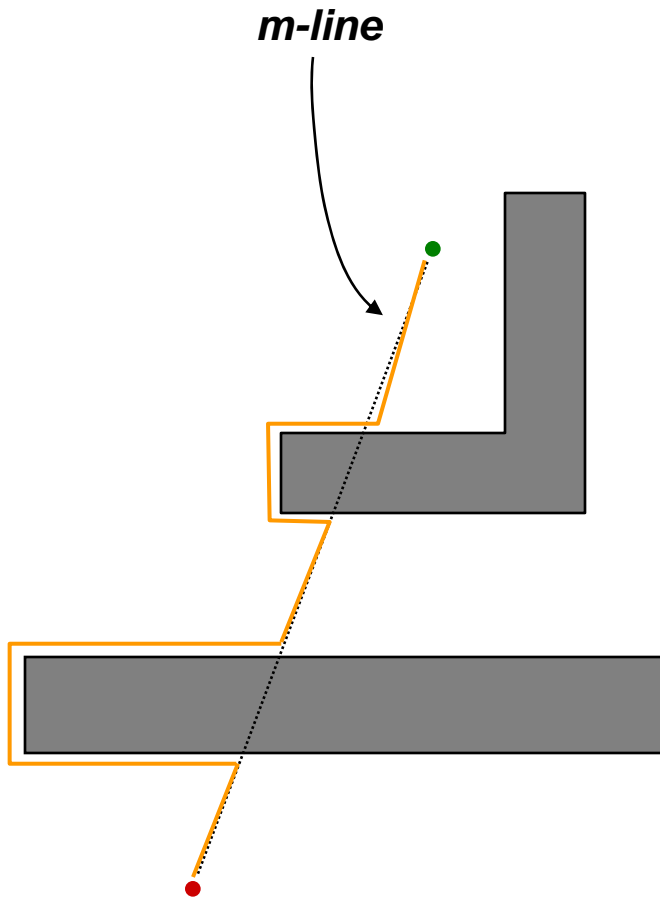


"Bug 2" Algorithm

- 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?

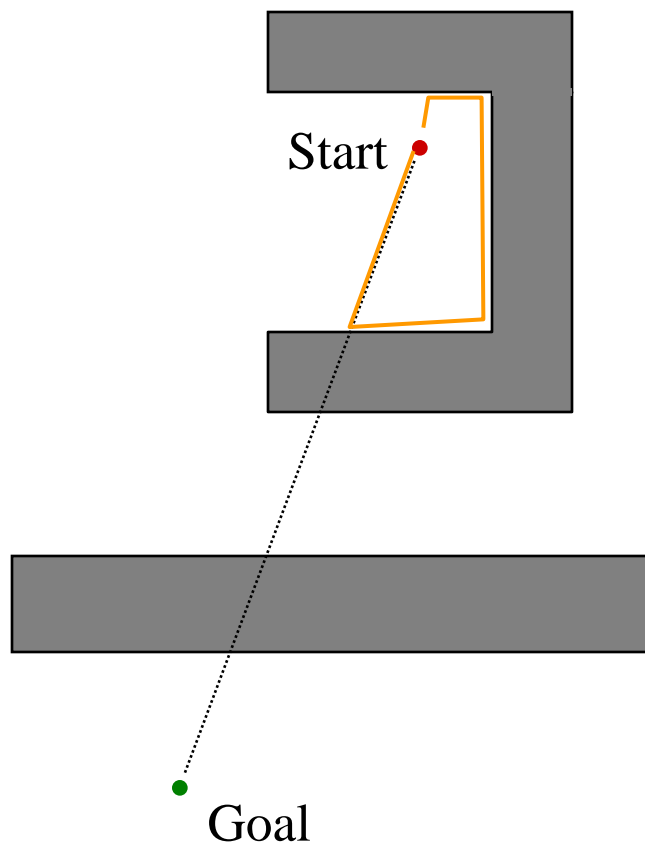
"Bug 2" Algorithm



- 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?

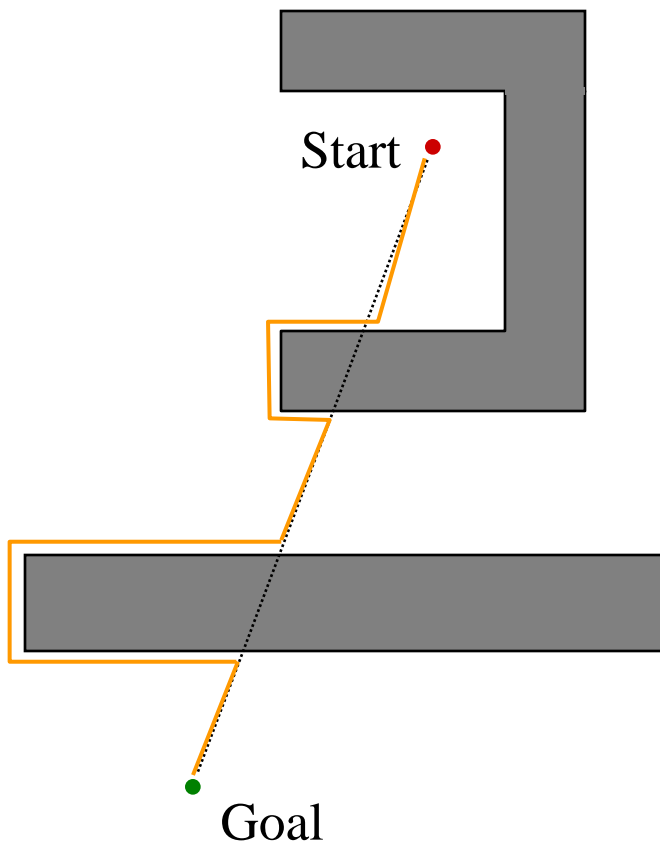
"Bug 2" Algorithm



- 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

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?

BUG 2 More formally

- Let $q_0^L = q_{\text{start}}$; $i = 1$
- repeat
 - repeat
 - from q_{i-1}^L move toward q_{goal} along the m-line
 - until goal is reached or obstacle encountered at q_i^H
 - if goal is reached, exit
 - repeat
 - follow boundary
 - until q_{goal} is reached or q_i^H is re-encountered or m-line is re-encountered, x is not q_i^H , $d(x, q_{\text{goal}}) < d(q_i^H, q_{\text{goal}})$ and way to goal is unimpeded
 - if goal is reached, exit
 - if q_i^H is reached, return failure
 - else
 - $q_i^L = m$
 - $i = i + 1$
 - continue

head-to-head comparison

or thorax-to-thorax, perhaps

Draw worlds in which Bug 2 does better than Bug 1 (and vice versa).

Bug 2 beats **Bug 1**

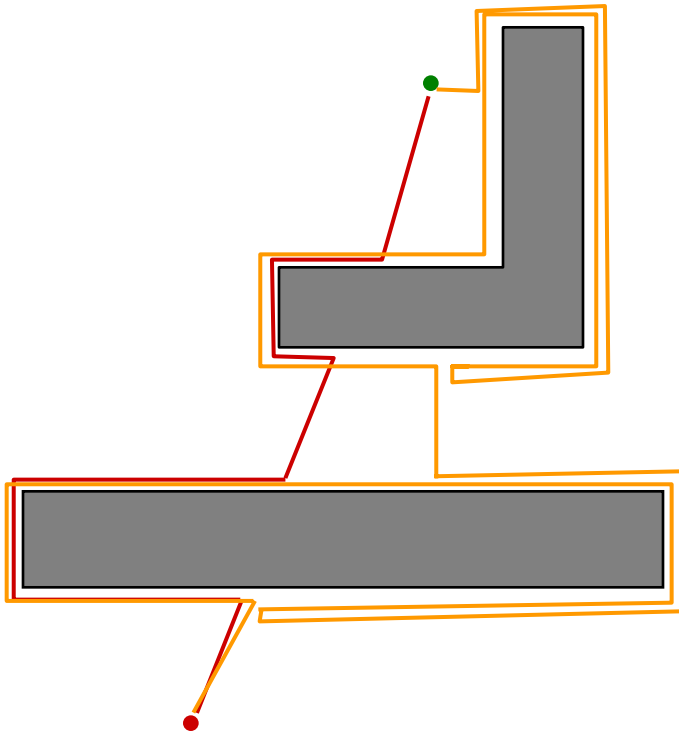
Bug 1 beats **Bug 2**

head-to-head comparison

or thorax-to-thorax, perhaps

Draw worlds in which Bug 2 does better than Bug 1 (and vice versa).

Bug 2 beats **Bug 1**

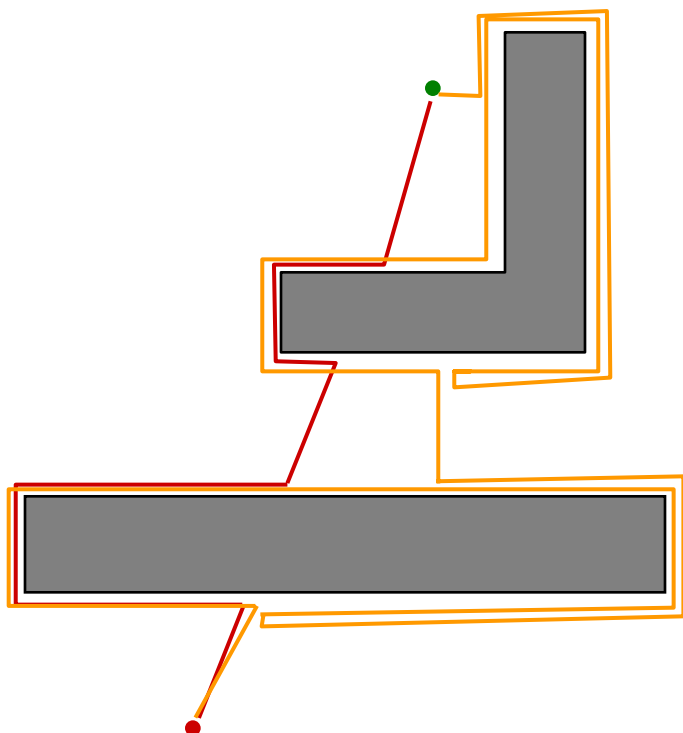


Bug 1 beats **Bug 2**

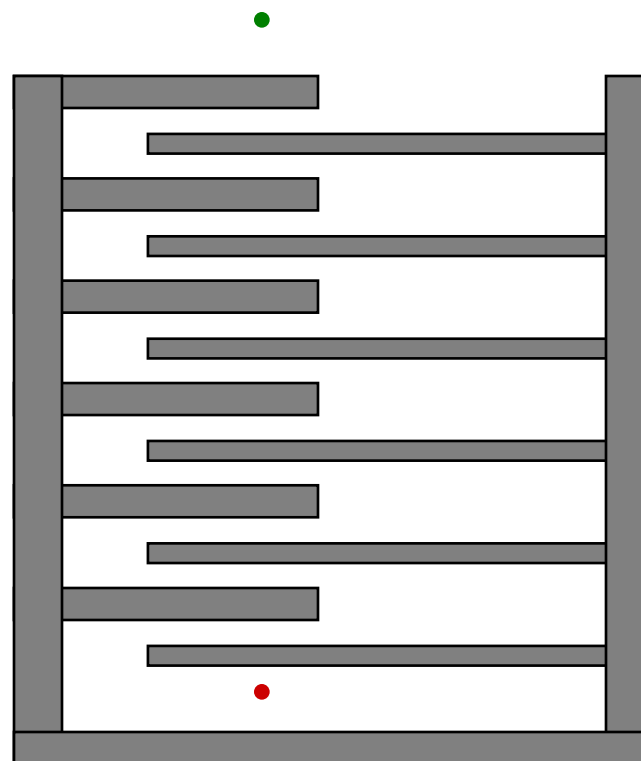
?

or thorax-to-thorax, perhaps

Bug 2 beats Bug 1



Bug 1 beats Bug 2



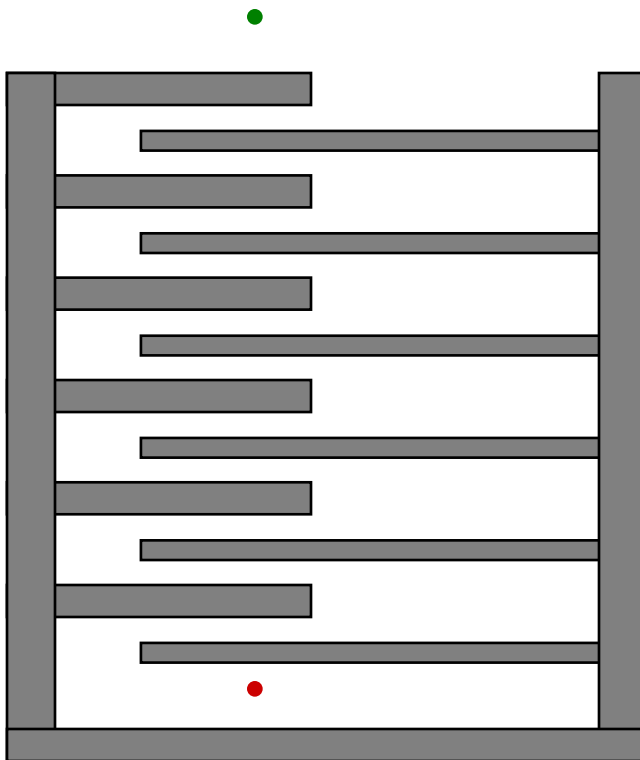
BUG 1 vs. BUG 2

- BUG 1 is an *exhaustive search algorithm*
 - it looks at all choices before committing
- BUG 2 is a *greedy* algorithm
 - it takes the first thing that looks better
- In many cases, BUG 2 will outperform BUG 1, but
- BUG 1 has a more predictable performance overall

“Quiz”

Bug 2 analysis

Bug 2: Path Bounds



What are upper/lower 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 bound:

What's the shortest distance it might travel?

D

Upper bound:

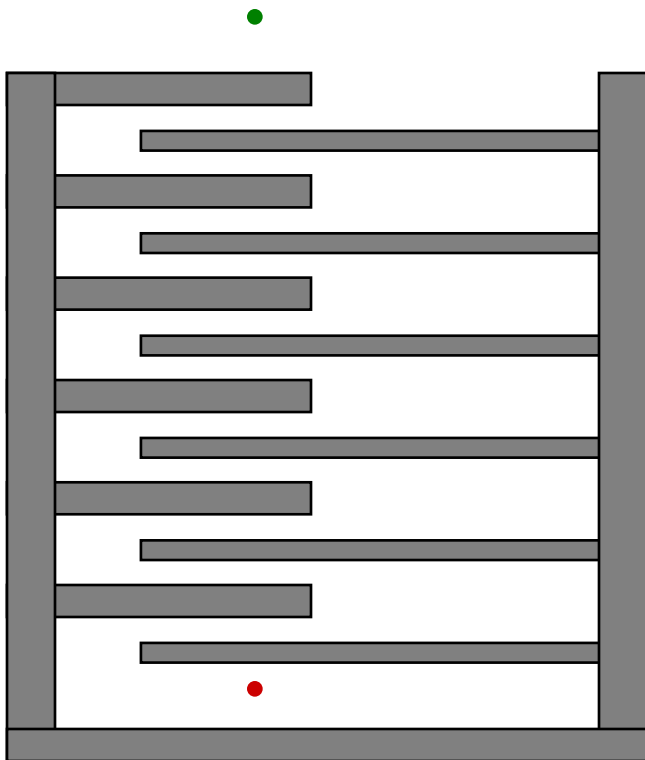
What's the longest distance it might travel?

What is an environment where your upper bound is required?

“Quiz”

Bug 2 analysis

Bug 2: Path Bounds



What are upper/lower 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 bound:

What's the shortest distance it might travel?

D

Upper bound:

What's the longest distance it might travel?

$$D + \sum_i \frac{n_i}{2} P_i$$

n_i = # of s-line intersections of the i th obstacle

What is an environment where your upper bound is required?

“Quiz”

Bug 2 analysis

Bug 2: Path Bounds



What are upper/lower 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 bound:

What's the shortest distance it might travel?

D

Upper bound:

What's the longest distance it might travel?

$$D + \sum_i \frac{n_i}{2} P_i$$

n_i = # of s-line intersections of the i th obstacle

What is an environment where your upper bound is required?

A More Realistic Bug

- As presented: global beacons plus contact-based wall following
- The reality: we typically use some sort of range sensing device that lets us look ahead (but has finite resolution and is noisy).
- Let us assume we have a range sensor

Raw Distance Function

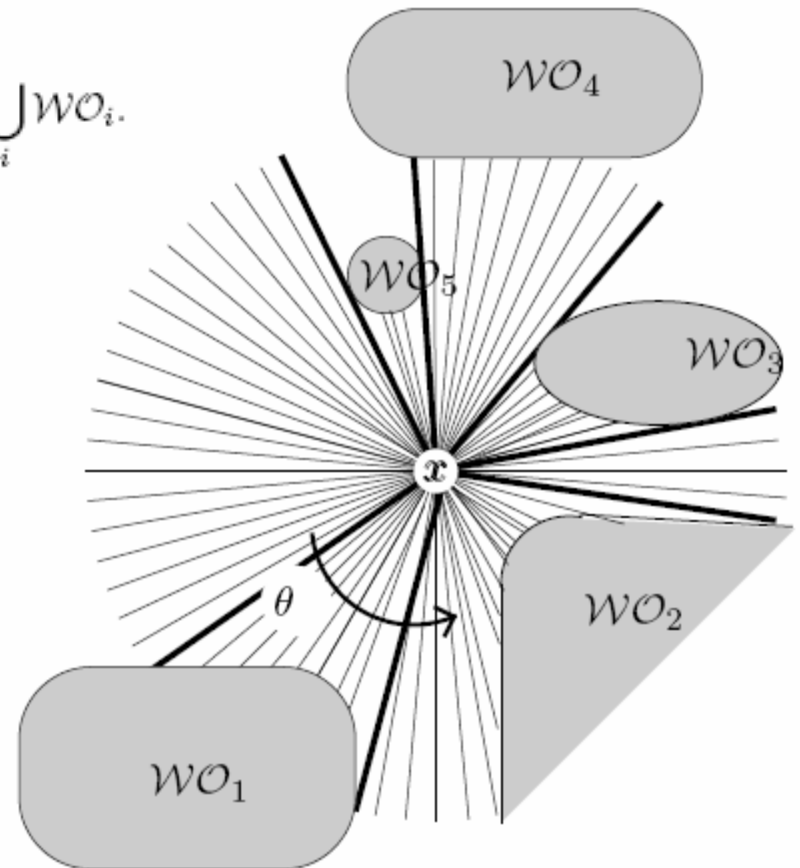
$$\rho(x, \theta) = \min_{\lambda \in [0, \infty]} d(x, x + \lambda [\cos \theta, \sin \theta]^T),$$

such that $x + \lambda [\cos \theta, \sin \theta]^T \in \bigcup_i \mathcal{WO}_i$.

$$\rho: \mathbb{R}^2 \times S^1 \rightarrow \mathbb{R}$$

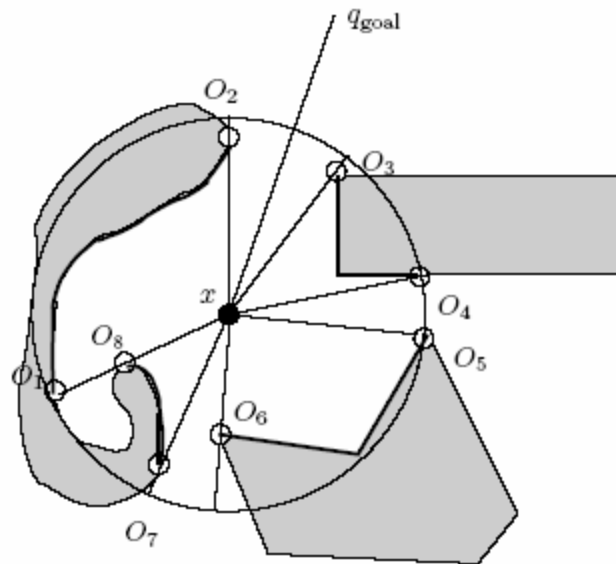
Saturated raw distance function

$$\rho_R(x, \theta) = \begin{cases} \rho(x, \theta), & \text{if } \rho(x, \theta) < R \\ \infty, & \text{otherwise.} \end{cases}$$



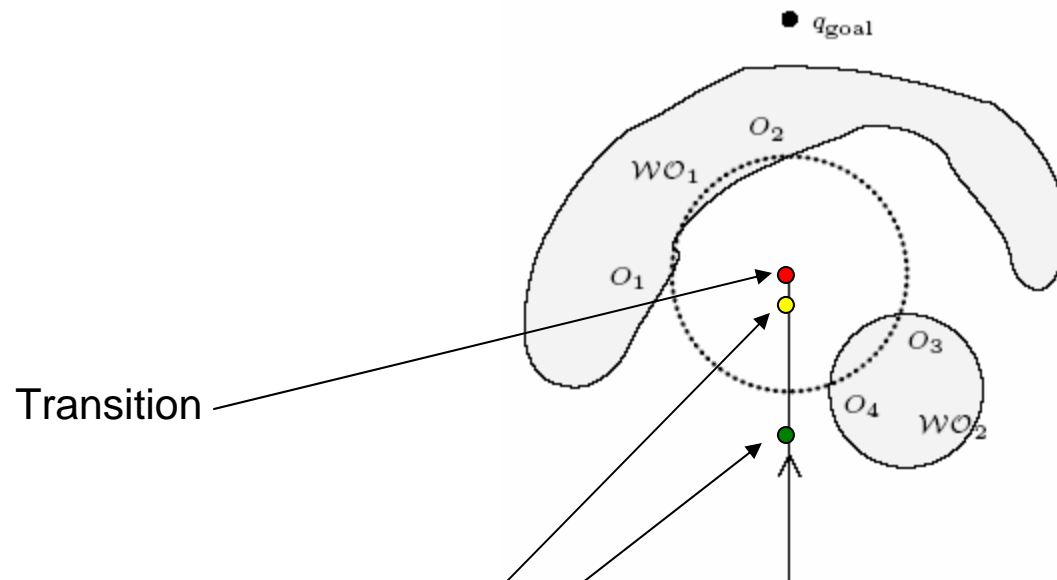
Intervals of Continuity

- Tangent Bug relies on finding endpoints of finite, conts segments of ρ_R



Motion-to-Goal Transition

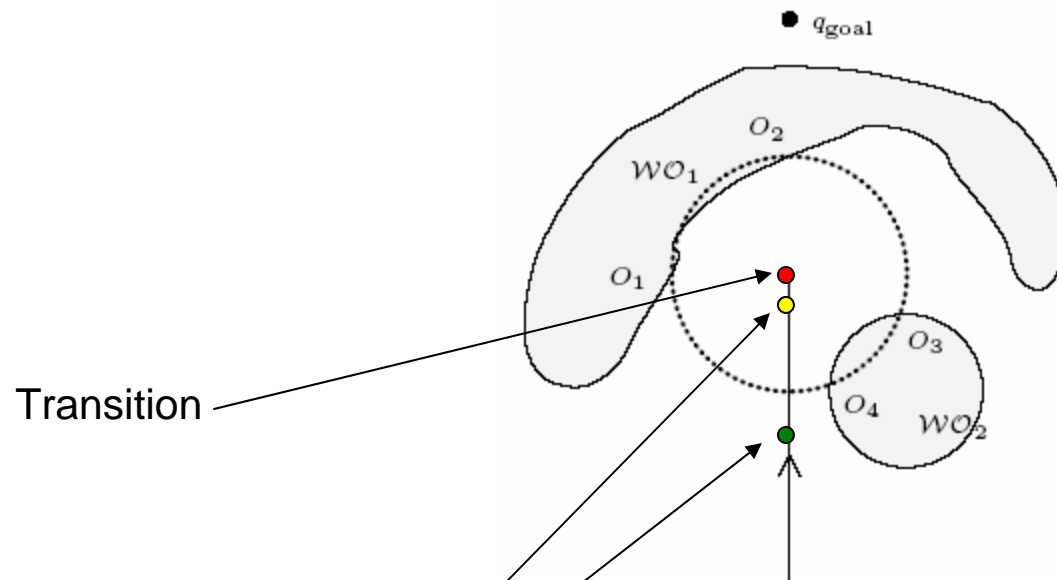
from Moving Toward goal to “following obstacles”



Currently, the motion-to-goal behavior “thinks” the robot can get to the goal

Motion-to-Goal Transition

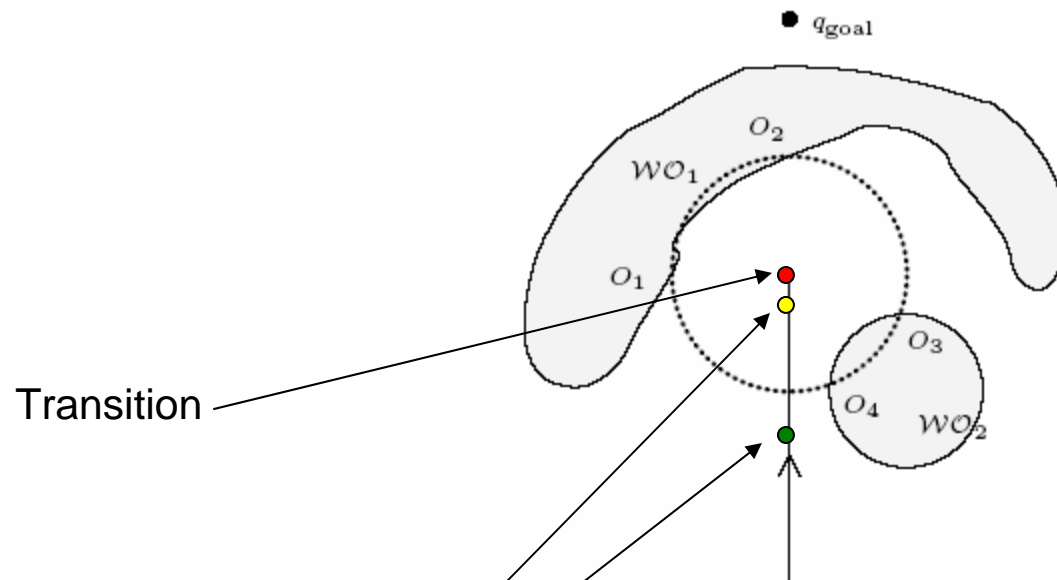
from Moving Toward goal to “following obstacles”



Currently, the motion-to-goal behavior “thinks” the robot can get to the goal

Motion-to-Goal Transition

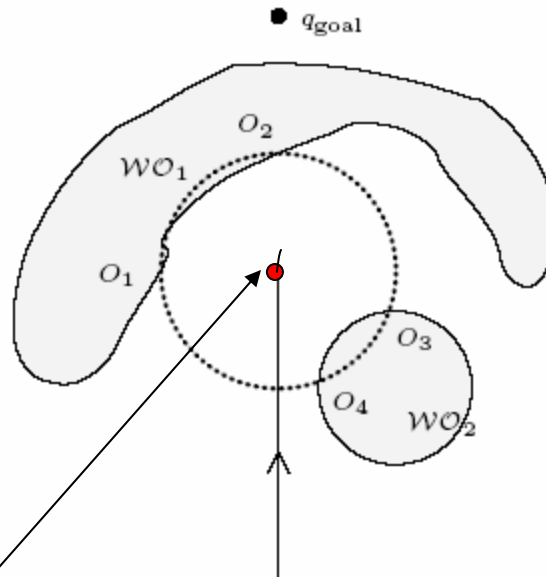
Among Moving Toward goal to “following obstacles”



Currently, the motion-to-goal behavior “thinks” the robot can get to the goal

Motion-to-Goal Transition

Minimize Heuristic

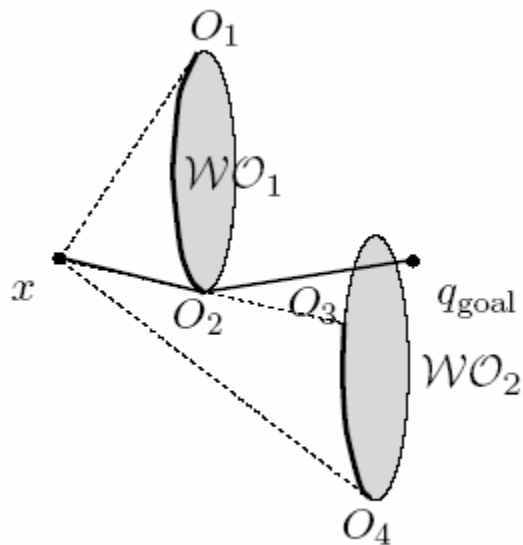


Now, it starts to see something --- what to do?

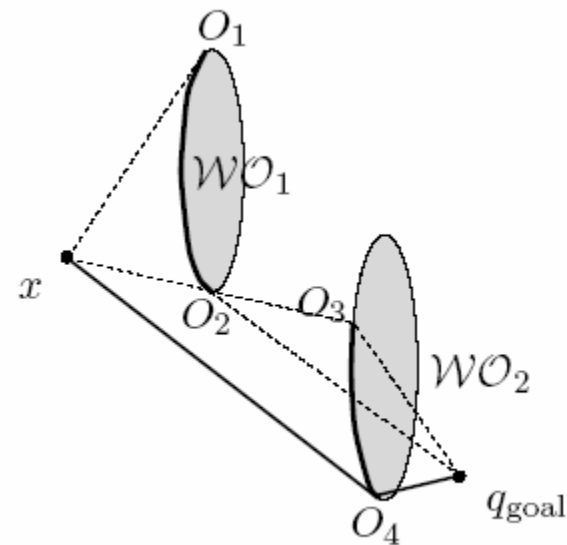
Ans: Choose the pt O_i that minimizes $d(x, O_i) + d(O_i, q_{goal})$

Minimize Heuristic Example

At x , robot knows only what it sees and where the goal is,



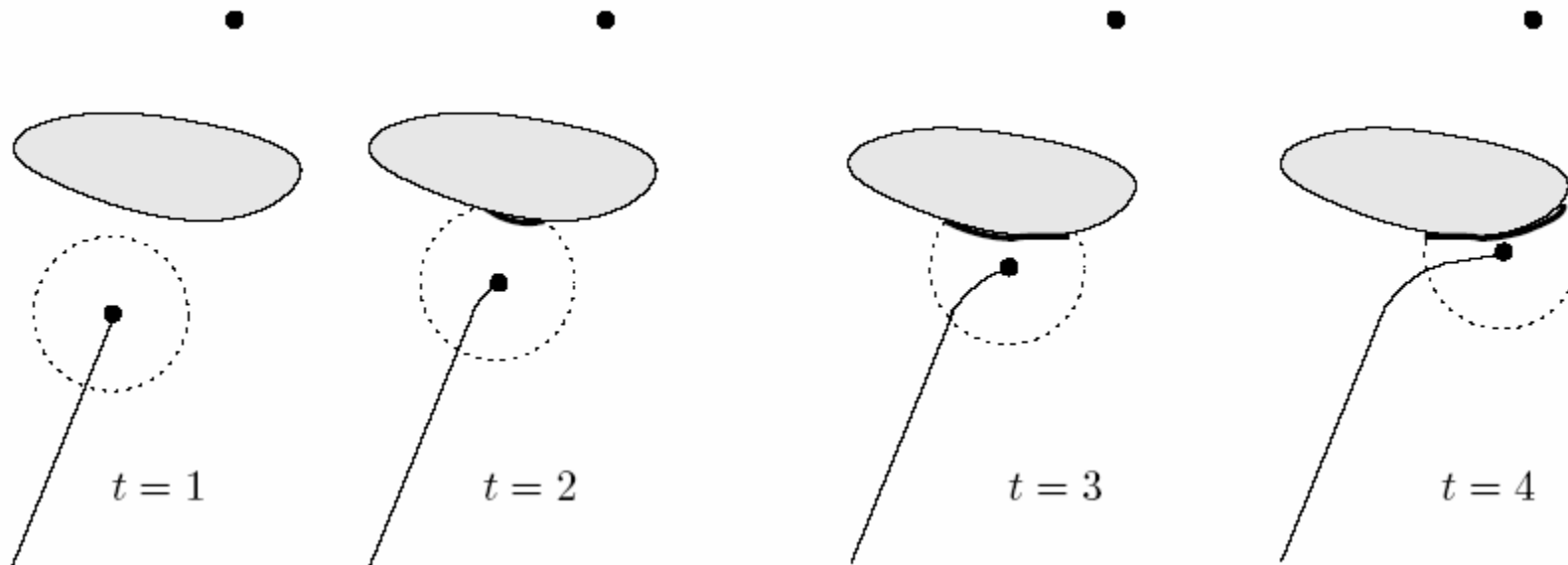
so moves toward O_2 . Note the line connecting O_2 and goal pass through obstacle



so moves toward O_4 . Note some "thinking" was involved and the line connecting O_4 and goal pass through obstacle

Choose the pt O_i that minimizes $d(x, O_i) + d(O_i, q_{\text{goal}})$

Motion To Goal Example



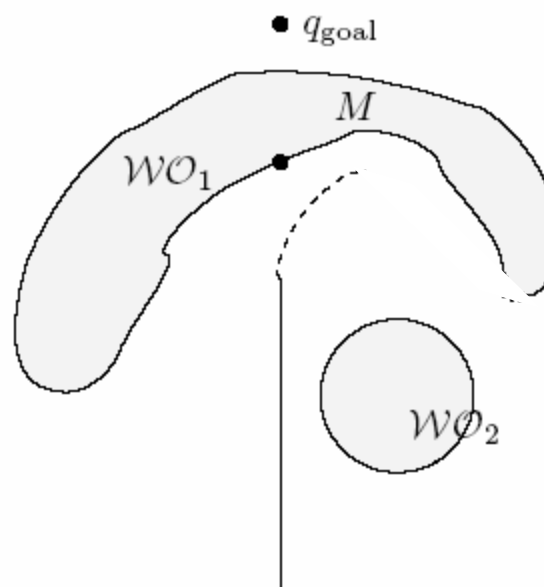
Choose the pt O_i that minimizes $d(x, O_i) + d(O_i, q_{\text{goal}})$

Transition *from* Motion-to-Goal

Choose the pt O_i that minimizes
 $d(x, O_i) + d(O_i, q_{\text{goal}})$

Problem: what if this distance
starts to go up?

Ans: start to act like a BUG and
follow boundary



M is the point on the “sensed”
obstacle which has the shortest
distance to the goal

Followed obstacle: the obstacle
that we are currently sensing

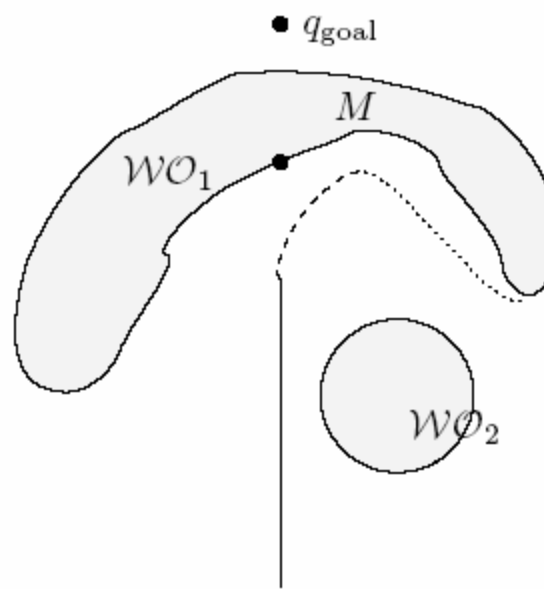
Blocking obstacle: the obstacle
that intersects the segment
 $(1 - \lambda)x + \lambda q_{\text{goal}} \quad \forall \lambda \in [0, 1]$

They start as the same

Boundary Following

Move toward the O_i on the followed obstacle in the “chosen” direction

Maintain d_{followed} and d_{reach}



M is the point on the “sensed” obstacle which has the shortest distance to the goal

Followed obstacle: the obstacle that we are currently sensing

Blocking obstacle: the obstacle that intersects the segment

They start as the same

d_{followed} and d_{reach}

- d_{followed} is the shortest distance between the sensed boundary and the goal
- d_{reach} is the shortest distance between *blocking* obstacle and goal (or my distance to goal if no blocking obstacle visible)

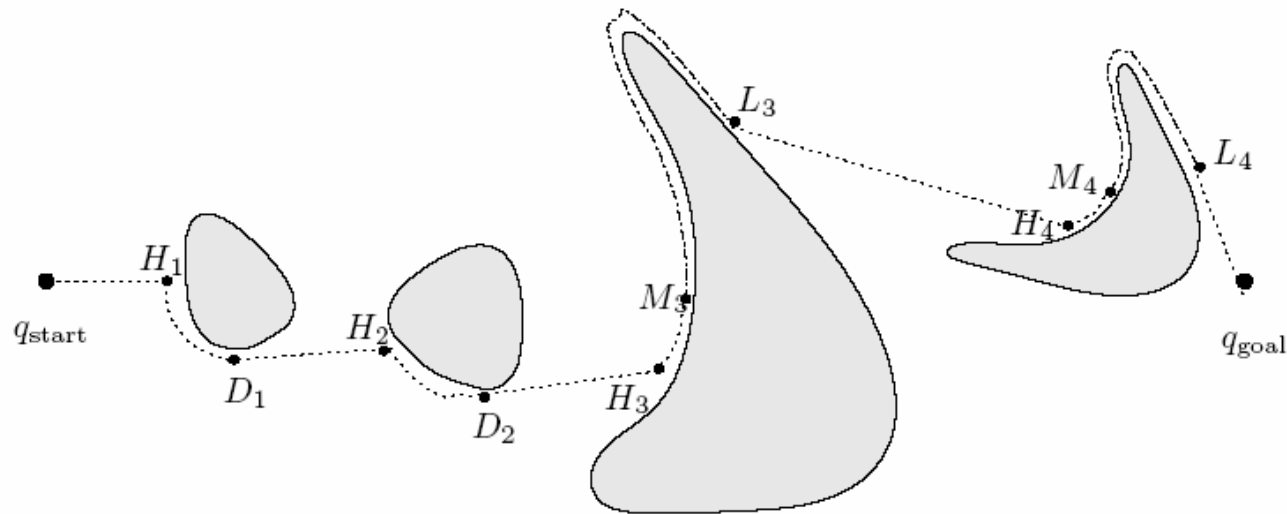
$$\Lambda = \{y \in \partial \mathcal{W} \mathcal{O}_b : \lambda x + (1 - \lambda)y \in \mathcal{Q}_{\text{free}} \quad \forall \lambda \in [0, 1]\}.$$

$$d_{\text{reach}} = \min_{c \in \Lambda} d(q_{\text{goal}}, c)$$

- Terminate boundary following behavior when $d_{\text{reach}} < d_{\text{followed}}$
- Initialize with $x = q_{\text{start}}$ and $d_{\text{leave}} = d(q_{\text{start}}, q_{\text{goal}})$

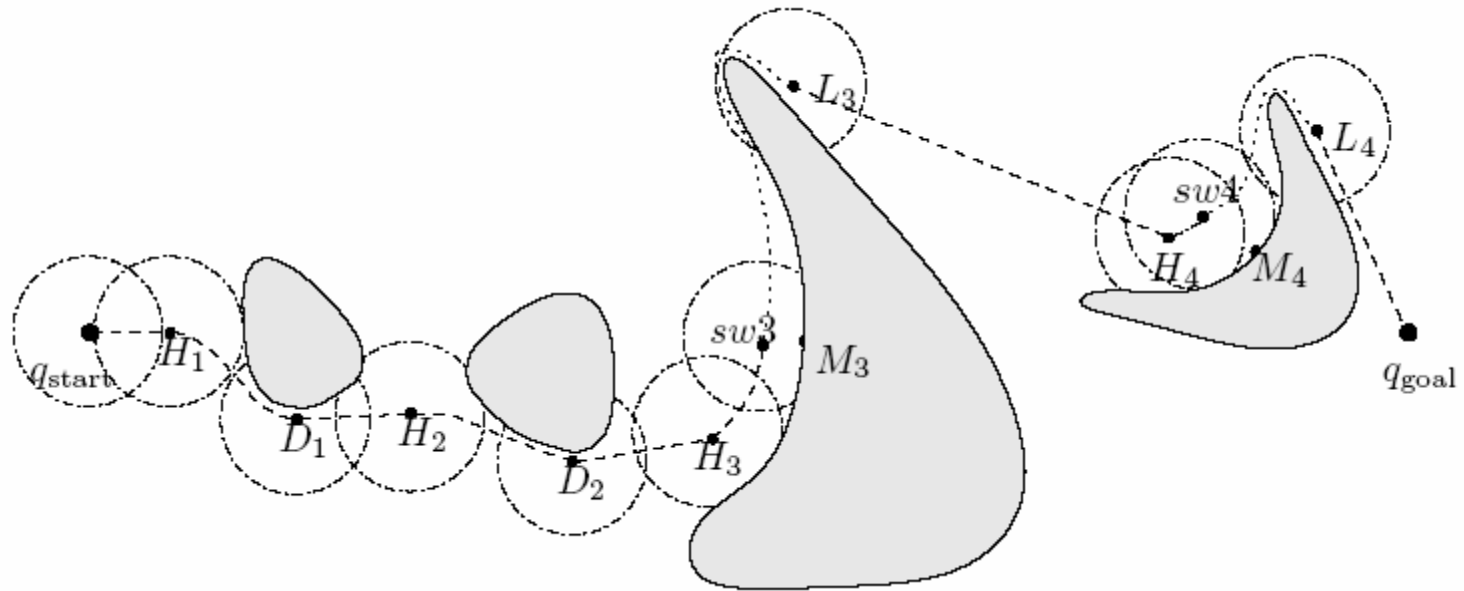
typo

Example: Zero Sensor Range

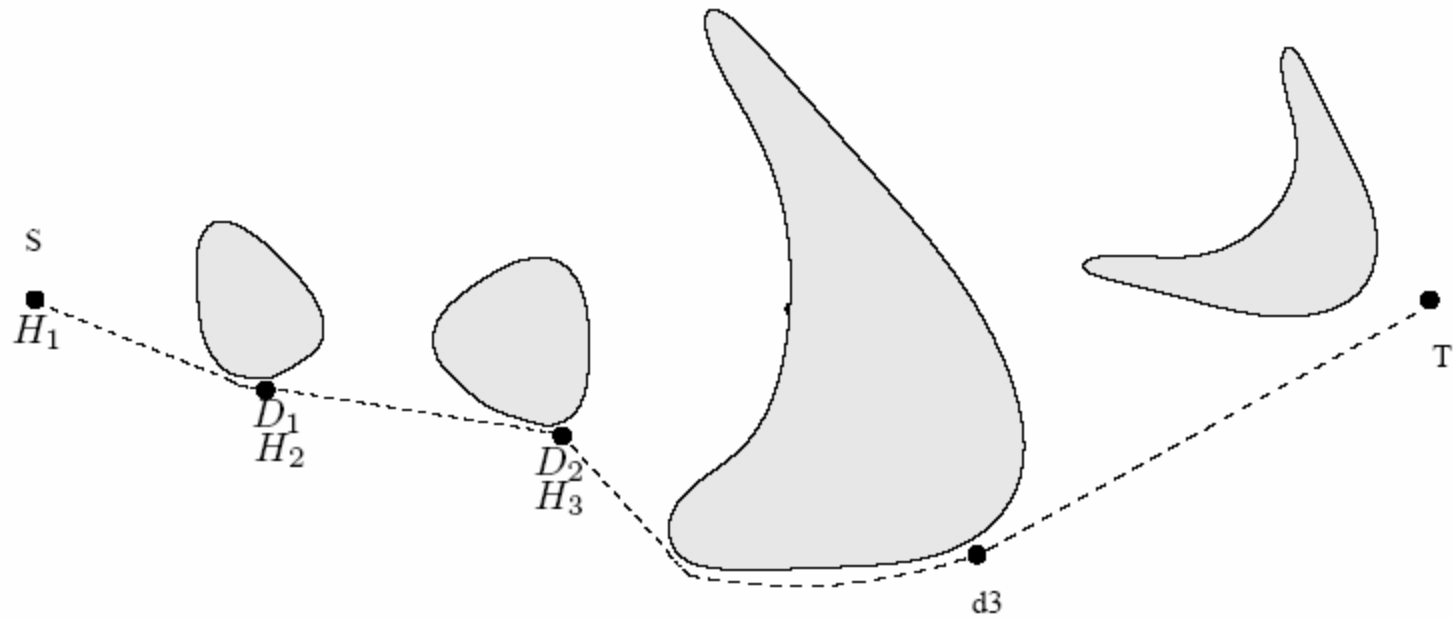


1. Robot moves toward goal until it hits obstacle 1 at H_1
2. Pretend there is an infinitely small sensor range and the O_i which minimizes the heuristic is to the right
3. Keep following obstacle until robot can go toward obstacle again
4. Same situation with second obstacle
5. At third obstacle, the robot turned left until it could not increase heuristic
6. $D_{followed}$ is distance between M_3 and goal, d_{reach} is distance between robot and goal because sensing distance is zero

Example: Finite Sensor Range

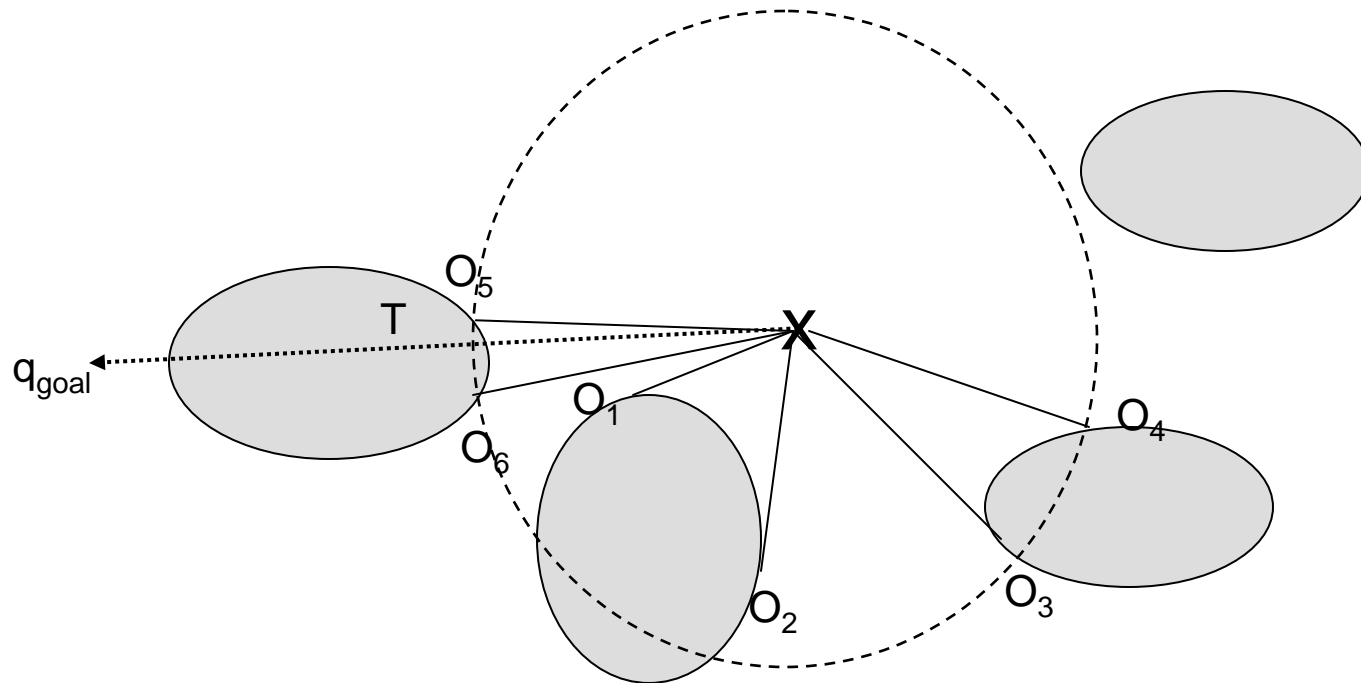


Example: Infinite Sensor Range



Tangent Bug

- Tangent Bug relies on finding endpoints of finite, conts segments of ρ_R



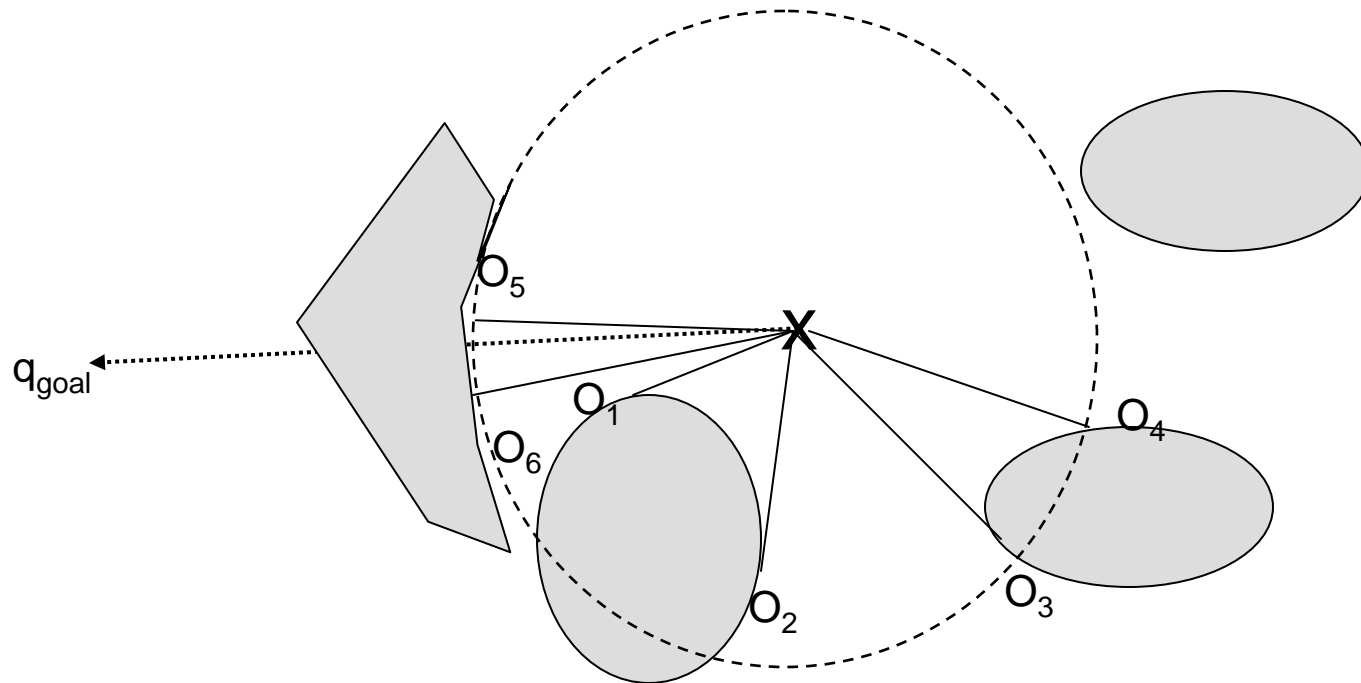
Now, it starts to see something --- what to do?

Ans: Choose the pt O_i that minimizes $d(x, O_i) + d(O_i, q_{goal})$

“Heuristic distance”

Tangent Bug

- Tangent Bug relies on finding endpoints of finite, conts segments of ρ_R



Problem: what if this distance starts to go up?

Ans: start to act like a BUG and follow boundary

The Basic Ideas

- A motion-to-goal behavior as long as way is clear or there is a visible obstacle boundary pt that decreases heuristic distance
- A boundary following behavior invoked when heuristic distance increases.
- A value d_{followed} which is the shortest distance between the sensed boundary and the goal
- A value d_{reach} which is the shortest distance between *blocking* obstacle and goal (or my distance to goal if no blocking obstacle visible)
- Terminate boundary following behavior when $d_{\text{reach}} < d_{\text{followed}}$

Tangent Bug Algorithm

- 1) repeat
 - a) Compute continuous range segments in view
 - b) Move toward $n \in \{T, O_i\}$ that minimizes $h(x, n) = d(x, n) + d(n, q_{\text{goal}})$until
 - a) goal is encountered, or
 - b) the value of $h(x, n)$ begins to increase
- 2) follow boundary continuing in same direction as before repeating
 - a) update $\{O_i\}$, d_{reach} and d_{followed}until
 - a) goal is reached
 - b) a complete cycle is performed (goal is unreachable)
 - c) $d_{\text{reach}} < d_{\text{followed}}$

Note the same general proof reasoning as before applies, although the definition of hit and leave points is a little trickier.

Implementing Tangent Bug

- Basic problem: compute tangent to curve forming boundary of obstacle at any point, and drive the robot in that direction
- Let $D(x) = \min_c d(x,c) \quad c \in \cup_i WO_i$
- Let $G(x) = D(x) - W^* \leftarrow$ some safe following distance
- Note that $\nabla G(x)$ points radially away from the object
- Define $T(x) = (\nabla G(x))$ the tangent direction
 - in a real sensor (we'll talk about these) this is just the tangent to the array element with lowest reading
- We could just move in the direction $T(x)$
 - open-loop control
- Better is $\delta x = \mu (T(x) - \lambda (\nabla G(x)) G(x))$
 - closed-loop control (predictor-corrector)

Sensors!

Robots' link to the external world...

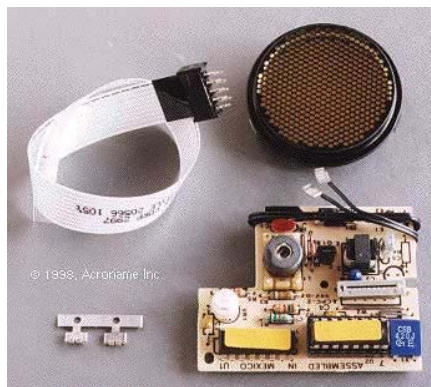


gyro

Sensors, sensors, sensors!
and tracking what is sensed: world models



IR rangefinder



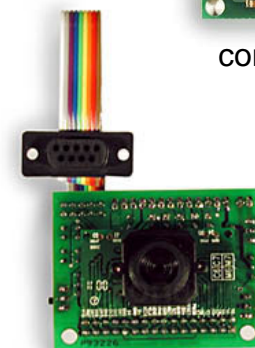
sonar rangefinder



sonar rangefinder



compass

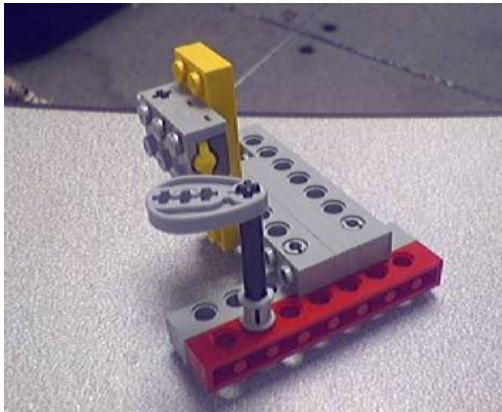


CMU cam with on-board processing

odometry...

16-735, Howie Choset with slides from G.D. Hager and Z. Dodds

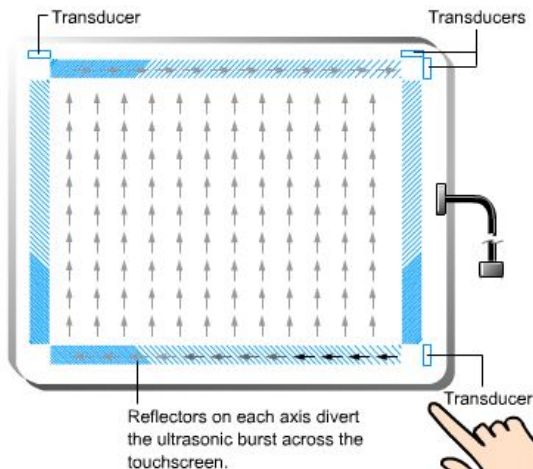
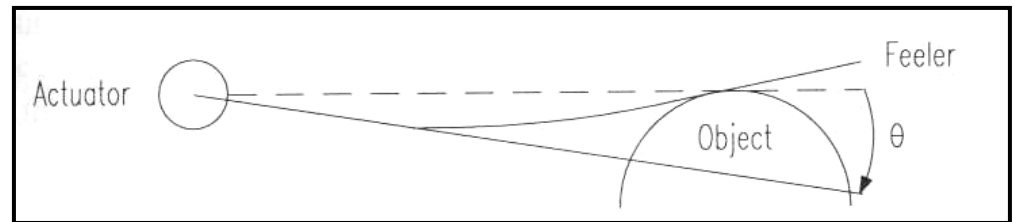
Tactile sensors



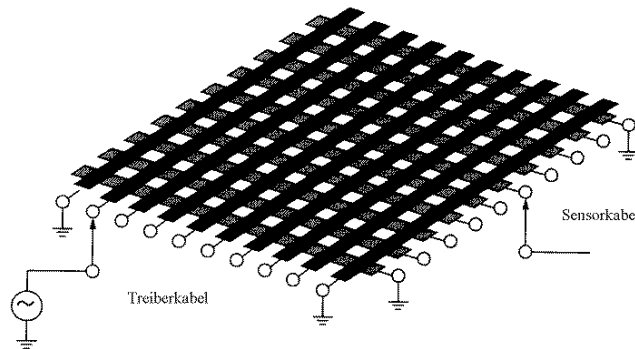
on/off switch

as a low-resolution encoder...

analog input: “Active antenna”



Surface acoustic waves



Capacitive array sensors



Resistive sensors

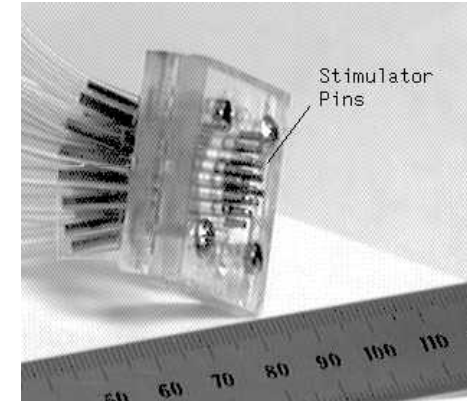
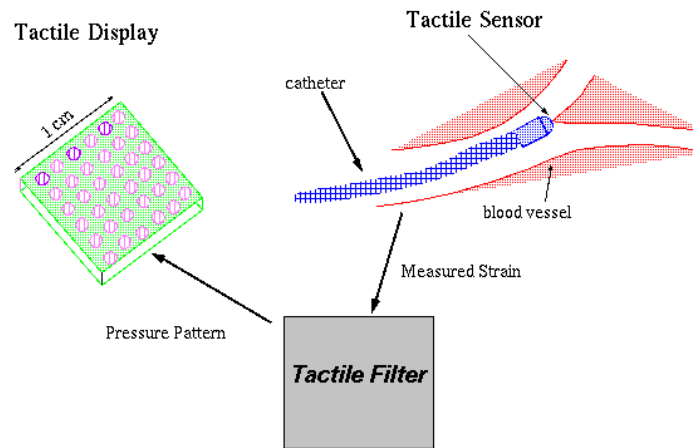
100% of light passes through 16-735 Howie Chose with slides from G.D. Hager and Z. Dodos 90% of light passes through 75% of light passes through

Tactile applications

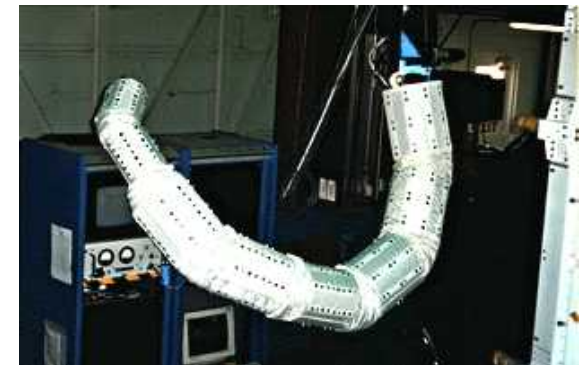
Medical teletaction interfaces



daVinci medical system



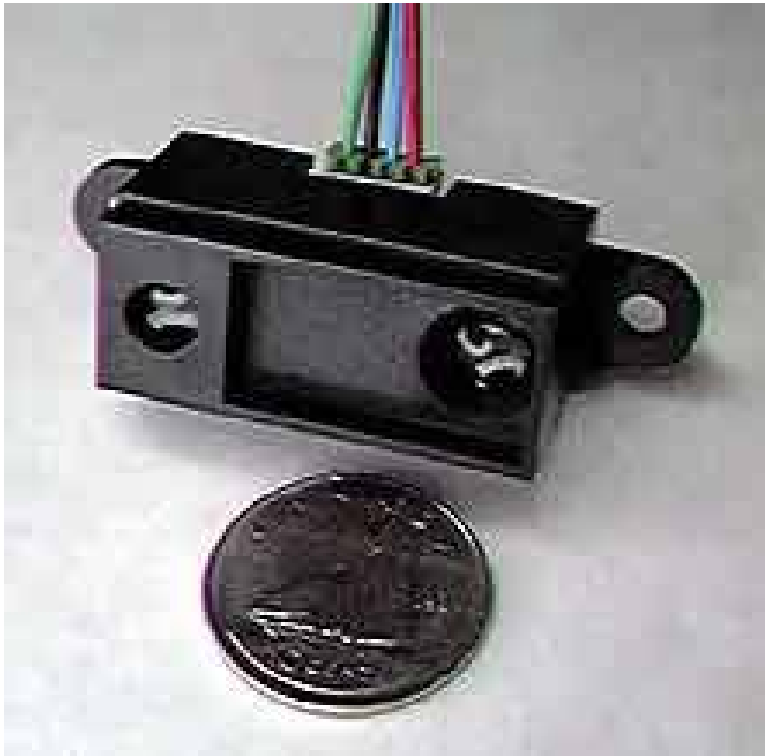
haptics



Robotic sensing Merritt systems, FL

Infrared sensors

“Noncontact bump sensor”



IR emitter/detector pair

IR detector

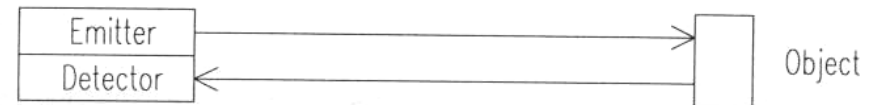
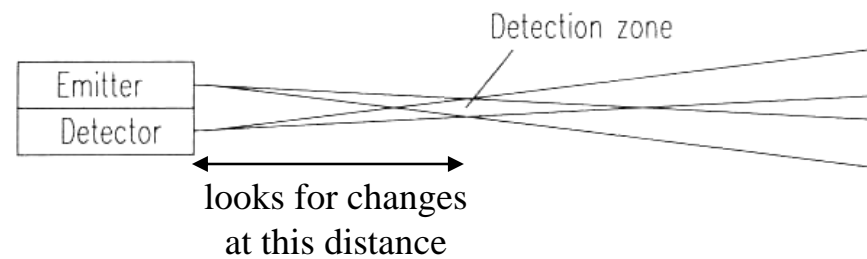


16-735, Howie Chos

rom G.I

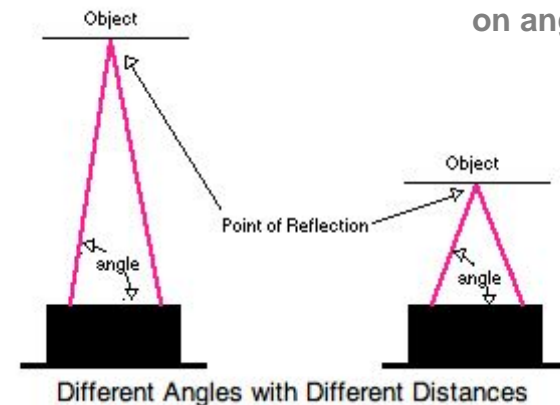
(1) sensing is based on light intensity.

“object-sensing” IR



diffuse distance-sensing IR

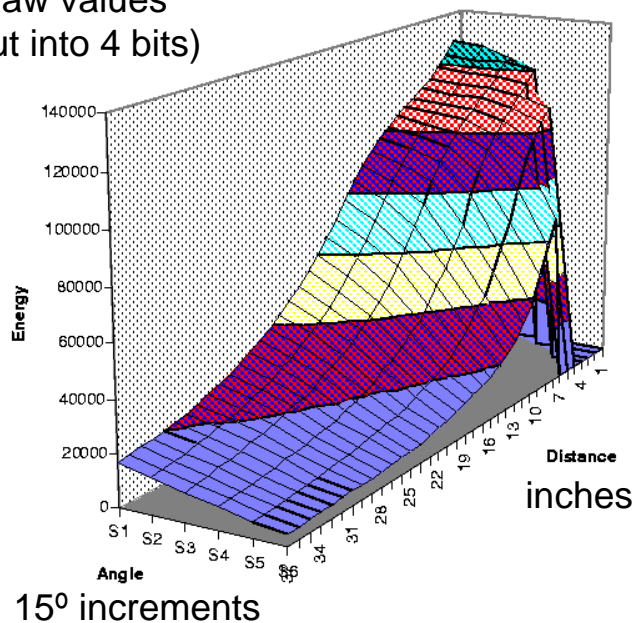
(2) sensing is based on angle received.



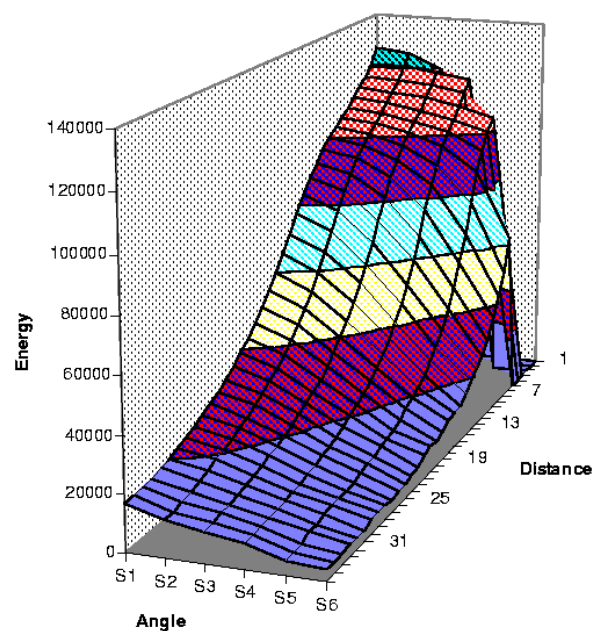
Infrared calibration

The response to white copy paper
(a dull, reflective surface)

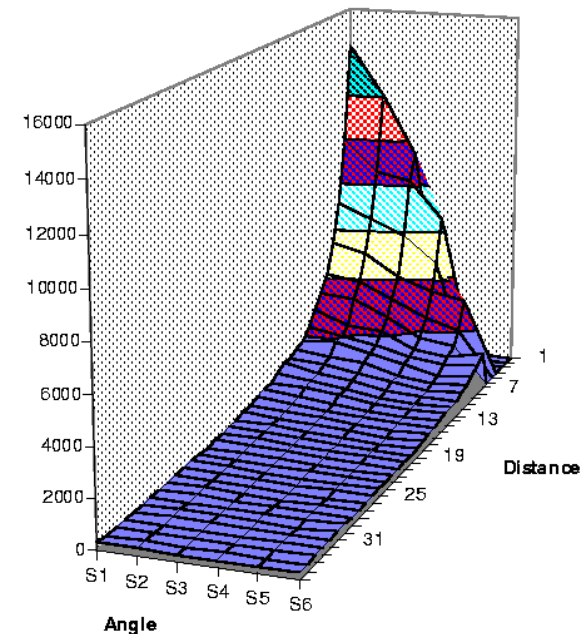
raw values
(put into 4 bits)



in the dark

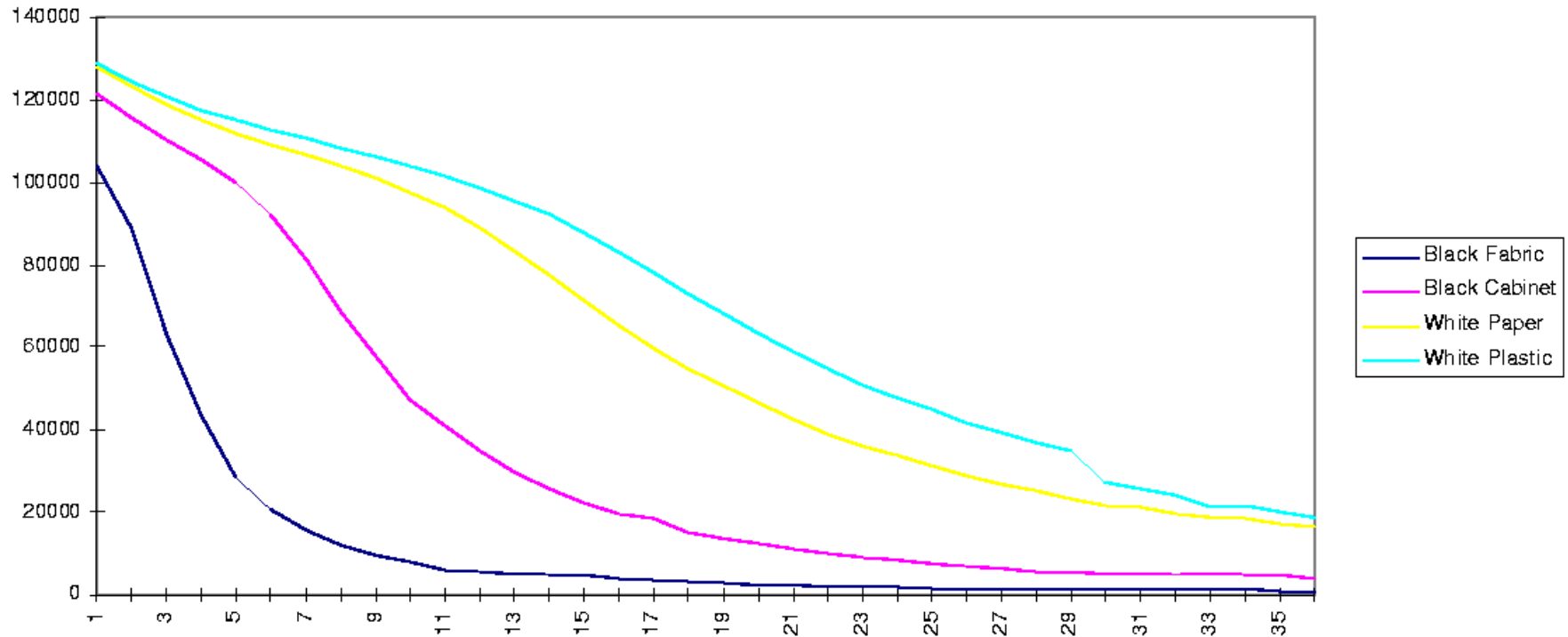


fluorescent light



incandescent light

Infrared calibration

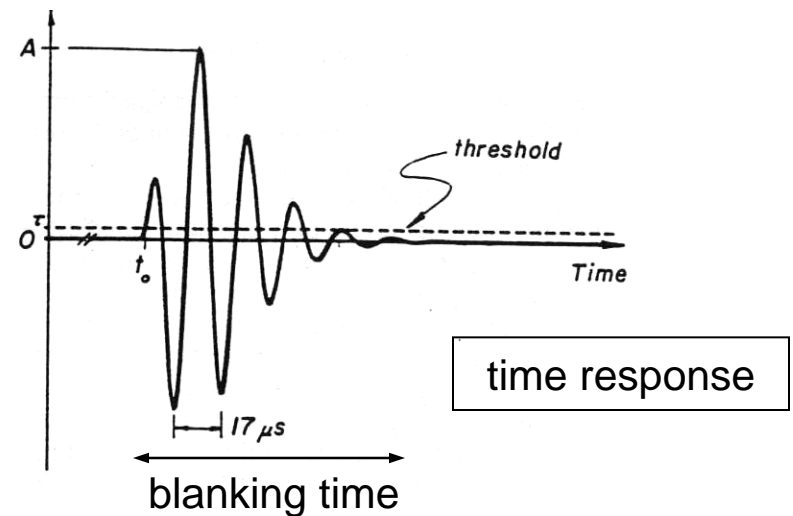
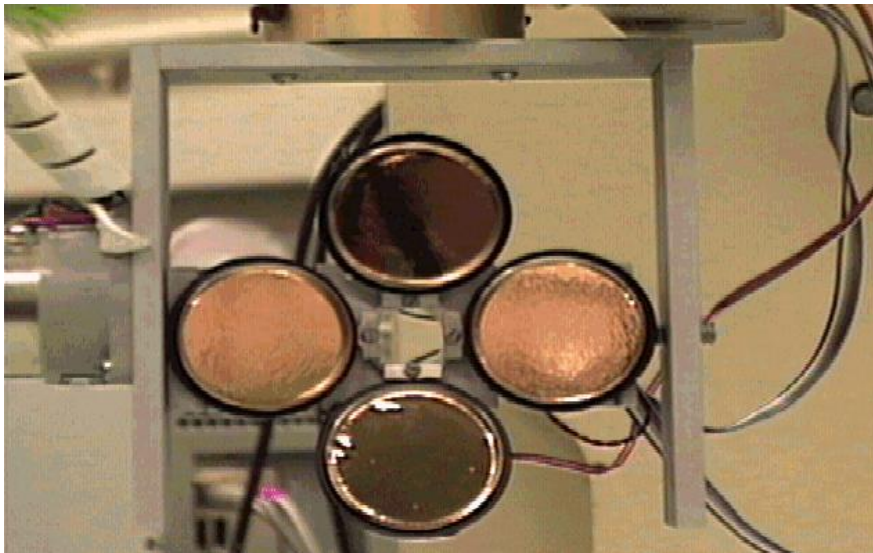
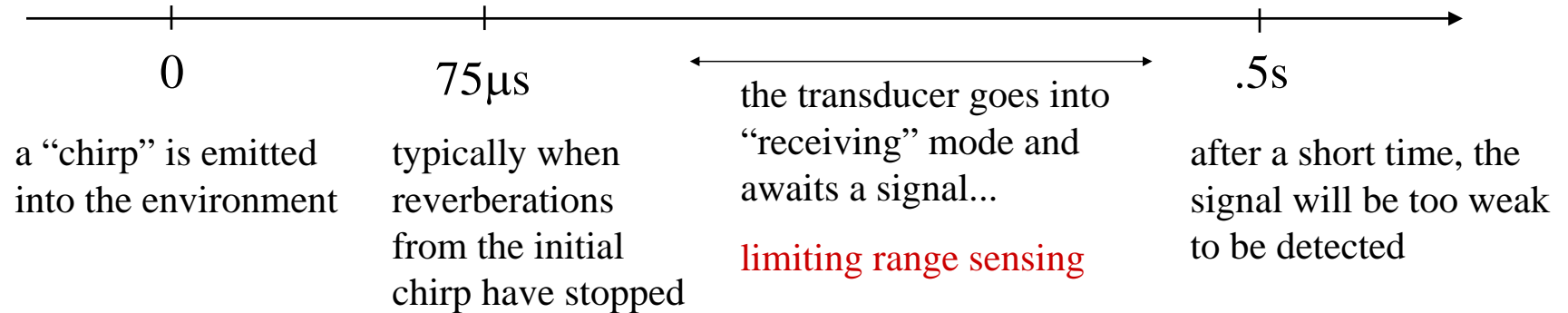


energy vs. distance for various materials
(the incident angle is 0° , or head-on)
(with no ambient light)



Sonar sensing

single-transducer sonar timeline

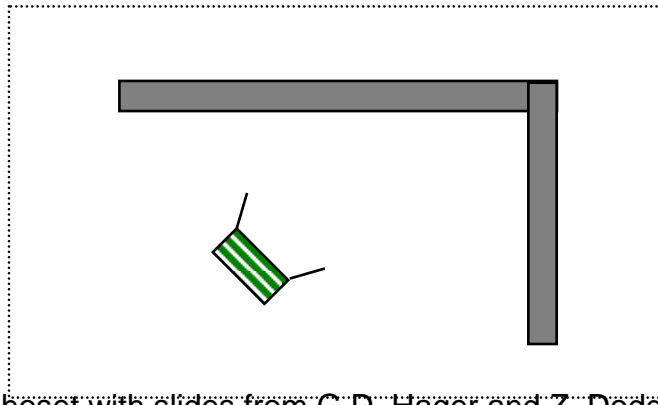
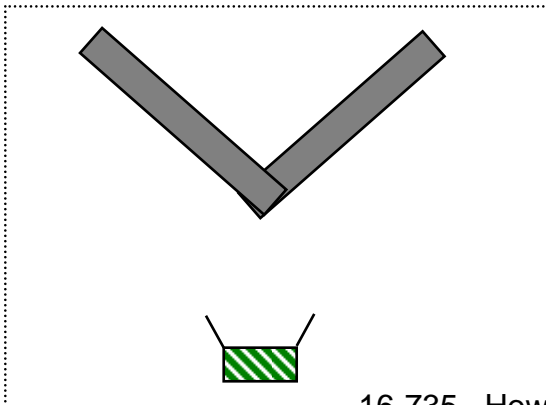
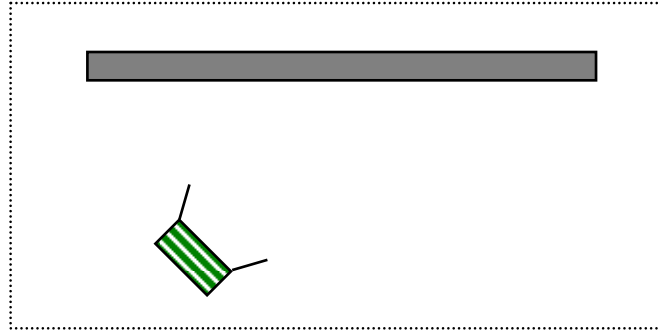
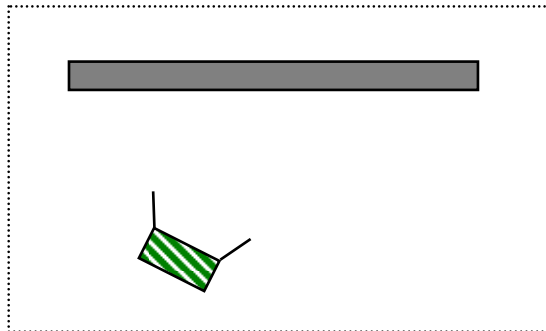
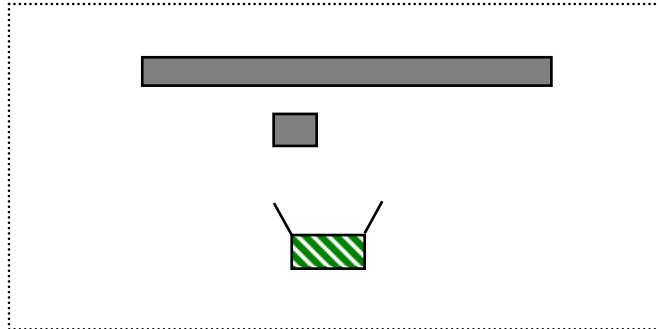
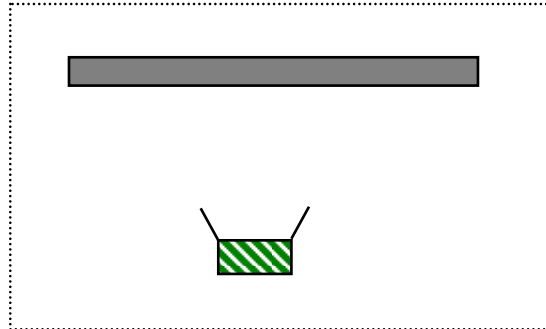
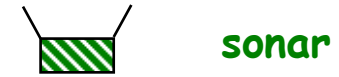


Polaroid sonar emitter/receivers

No lower range limit for paired sonars...



Sonar effects



Draw the range reading that the sonar will return in each case...

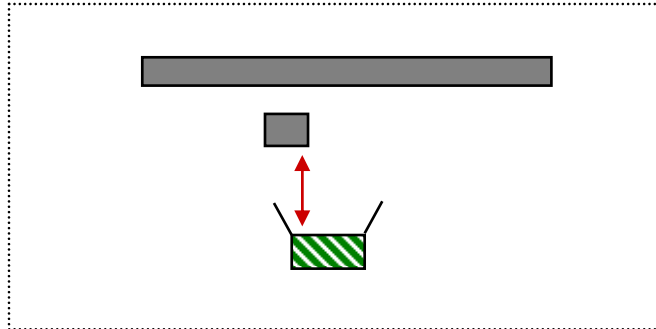
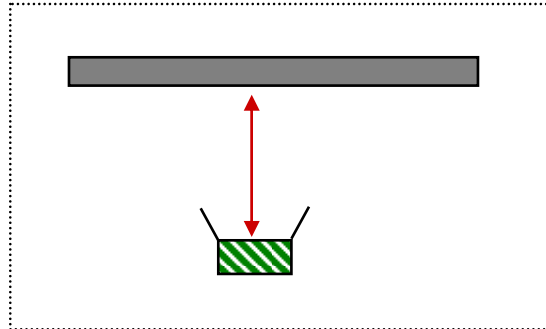


walls
(obstacles)

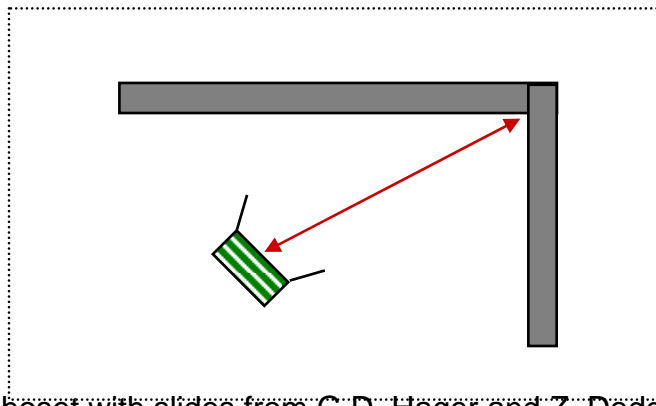
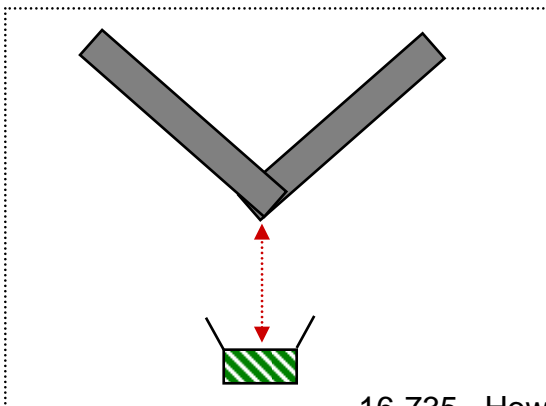
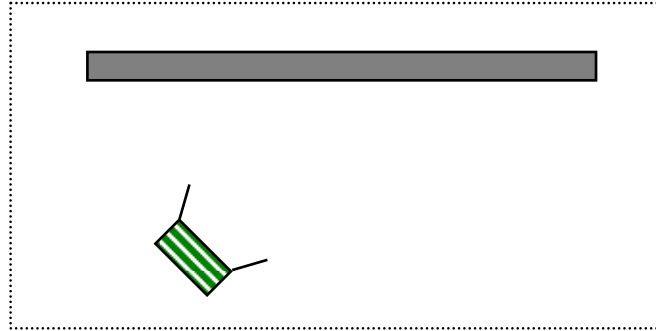
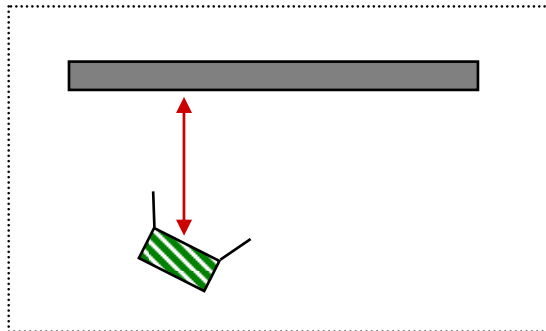
Sonar effects



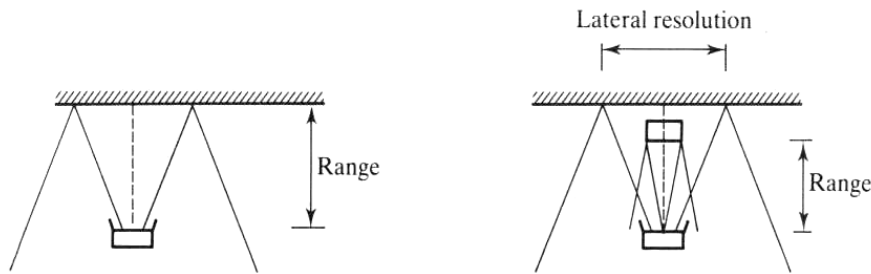
sonar



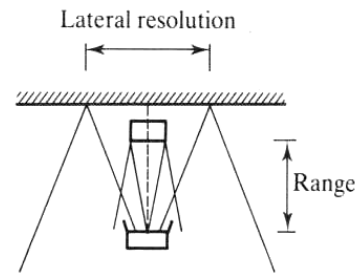
Draw the range
reading that the
sonar will return
in each case...



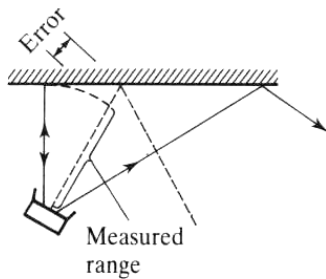
Sonar effects



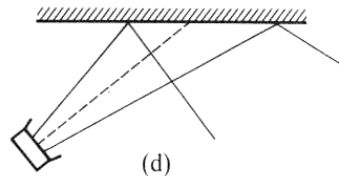
(a)



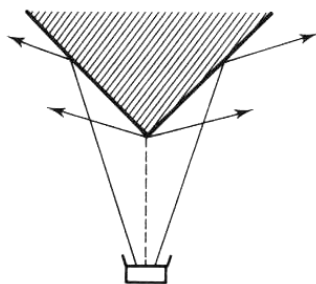
(b)



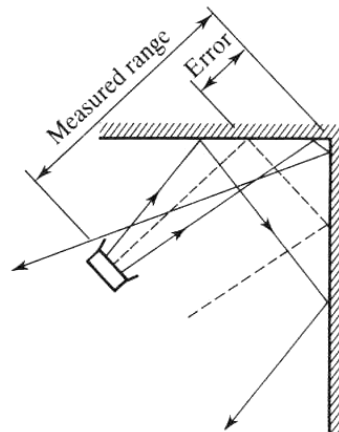
(c)



(d)



(e)



(f)

(a) Sonar providing an accurate range measurement

(b-c) Lateral resolution is not very precise; the closest object in the beam's cone provides the response

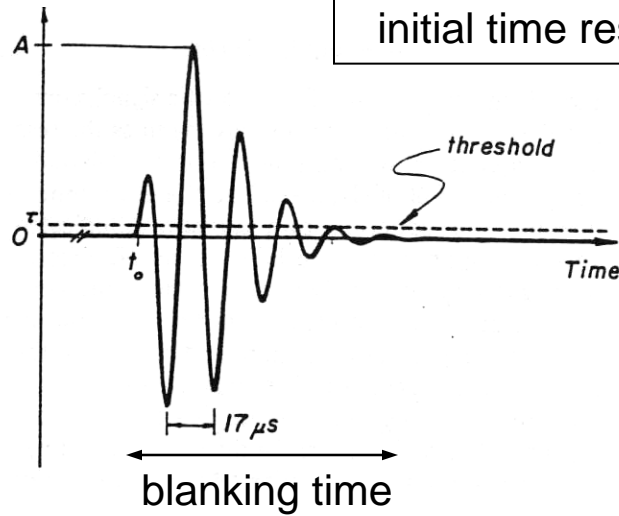
(d) Specular reflections cause walls to disappear

(e) Open corners produce a weak spherical wavefront

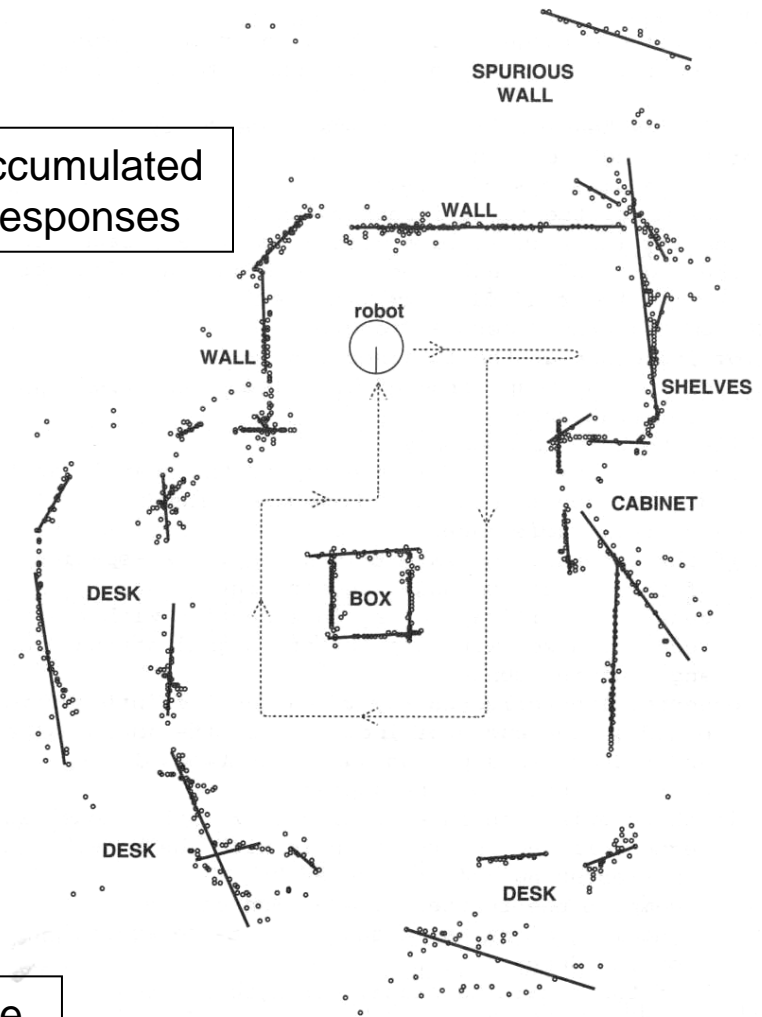
(f) Closed corners measure to the corner itself because of multiple reflections --> sonar ray tracing

Sonar modeling

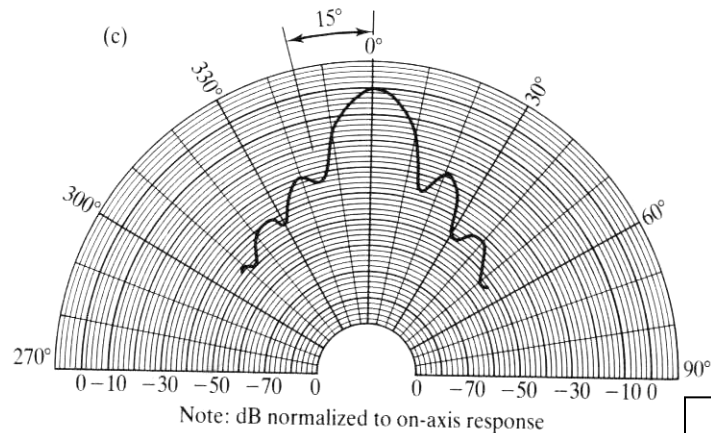
initial time response



accumulated responses



cone width

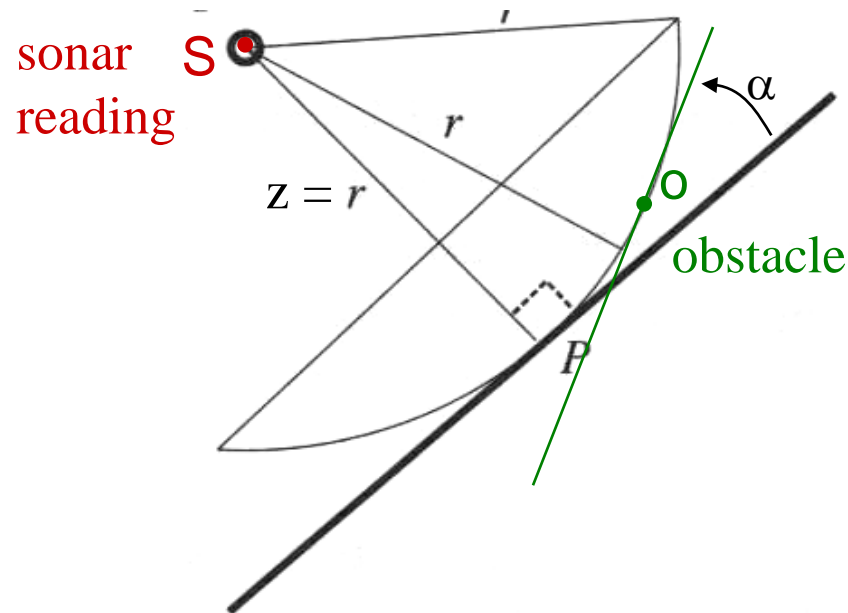


spatial response

Sonar Modeling

response model (Kuc)

$$h_R(t, z, a, \alpha) = \frac{2c \cos \alpha}{\pi a \sin \alpha} \sqrt{1 - \frac{c^2(t - 2z/c)^2}{a^2 \sin^2 \alpha}}$$



- Models the response, h_R , with

c = speed of sound

a = diameter of sonar element

t = time

z = orthogonal distance

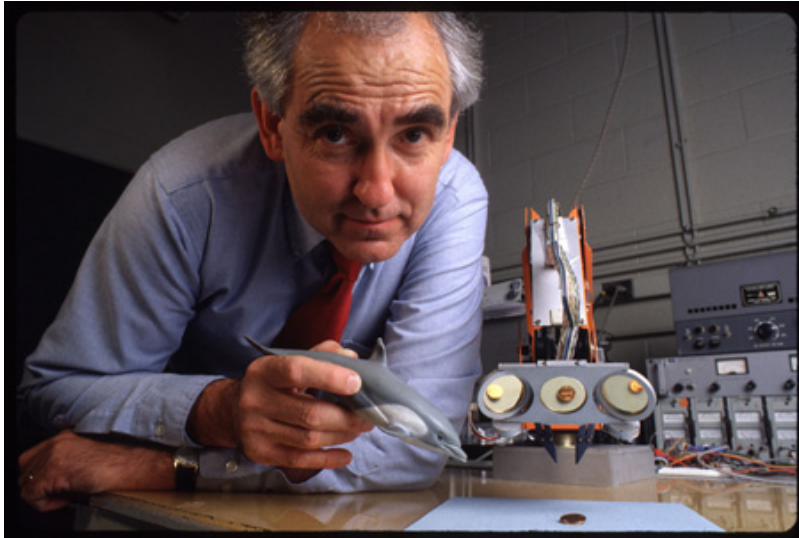
α = angle of environment surface

- Then, allow uncertainty in the model to obtain a probability:

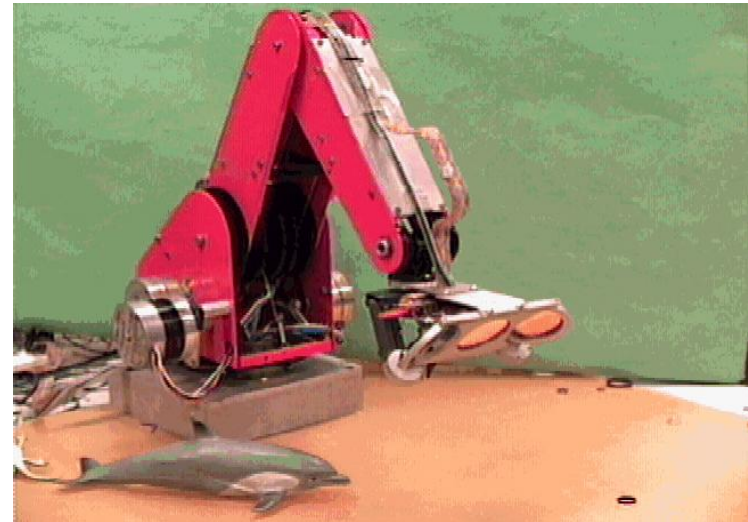
$$p(S | o)$$

chance that the sonar reading is S , given an obstacle at location O

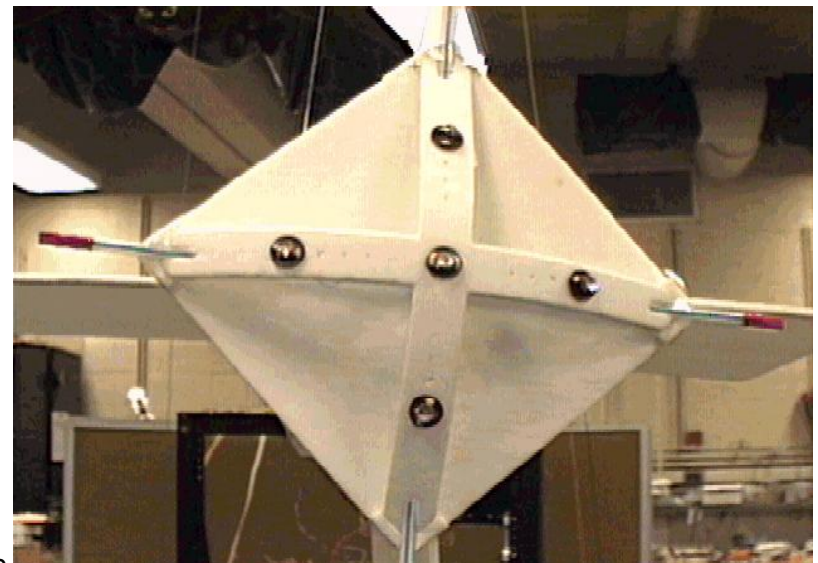
Roman Kuc's animals



Rodolph



Robat



10-755, Flowie Choset with slides from G.D. Hager and Z. Bouas

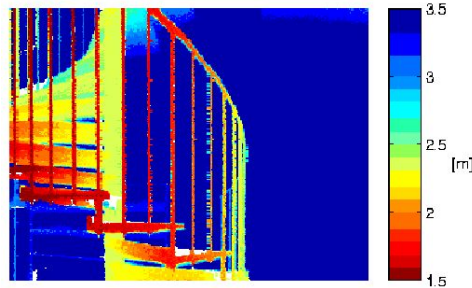
Laser Ranging



LIDAR



Sick Laser



LIDAR map

Summary

- Bug 1: safe and reliable
- Bug 2: better in some cases; worse in others
- Should understand the basic completeness proof
- Tangent Bug: supports range sensing
- Sensors and control
 - should understand basic concepts and know what different sensors are