Introduction to Autonomous and Intelligent Systems¹ Obstacle Avoidance

Justus Piater

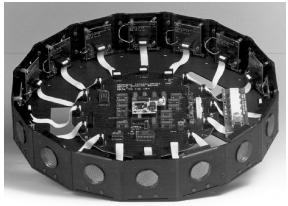
Table of Contents

1.	Range Sensors	1
2.	Obstacle Avoidance	2
2	Dafarances	16

1. Range Sensors

1.1. Sonar

- Time of flight
- Typically ultrasonic





1.2. Laser Range Finder

- Time of flight
- LIDAR ("Light detection and ranging")

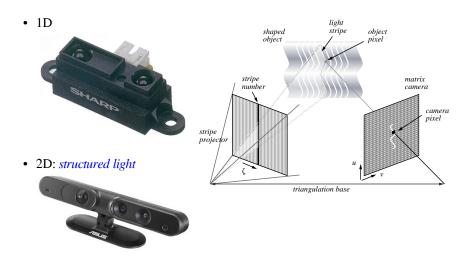


¹ https://iis.uibk.ac.at/courses/2019s/703030

1.3. ToF Cameras



1.4. Optical Triangulation



2. Obstacle Avoidance

2.1. Problem

Given:

- goal location in world coordinates
- robot pose in world coordinates
- range sensors

Sought:

• a (reactive) path to the goal in the presence of obstacles

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

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

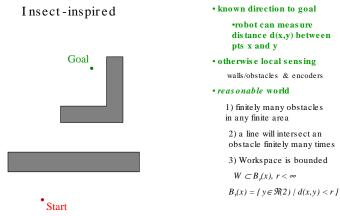
A Few General Concepts

- Workspace W
 - $\Re(2)$ or $\Re(3)$ depending on the robot
 - could be infinite (open) or bounded (closed/compact)
- Obstacle WOi
- Free works pace $W_{free} = W \setminus \bigcup_{i} WO_{i}$

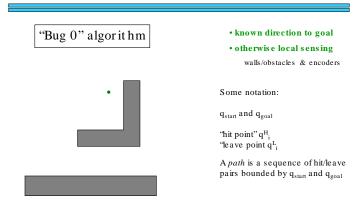
16-735, Howie Choset with slides from G.D. Hager and Z. Dodds

The **Bug** Algorithms

provable results...



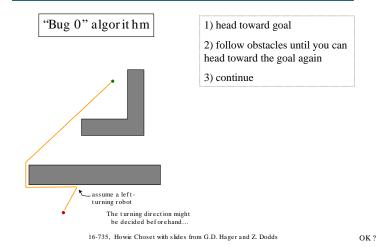
Buginner Strategy



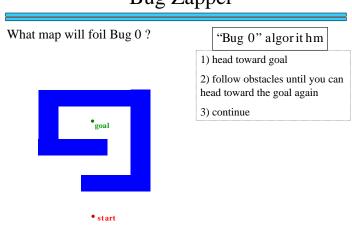
16-735, Howie Choset with slides from G.D. Hager and Z. Dodds

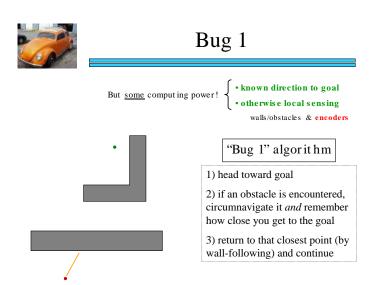
how?

Buginner Strategy



Bug Zapper



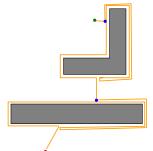


Vladimir Lumelsky & Alexander Stepanov: Algorithmica 1987 16-735, Howie Choset with slides from G.D. Hager and Z. Dodds



Bug 1

• known direction to goal But some computing power! • otherwise local sensing



"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 More formally

- $\begin{array}{ll} \bullet & \text{Let } q^L_{0} = q_{start}; i = 1 \\ \bullet & \text{repeat} \end{array}$
- - 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 $\boldsymbol{q}_{\mathrm{goal}}$ is reached or $\boldsymbol{q}^{H}_{\ i}$ is re-encountered
 - if goal is reached, exit
 - Go to q^L_i
 - if move toward $\boldsymbol{q}_{goal}\, move\, s$ into obstacle
 - · exit with failure
 - else
 - i=i+1
 - continue

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:

Upper bound:

What is an environment where your upper bound is required?

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

Quiz"

Bug 1 analysis

Bug 1: Path Bounds

D

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

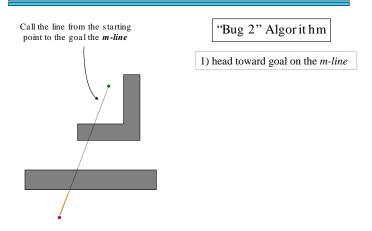
How Can We Show Completeness?

- 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

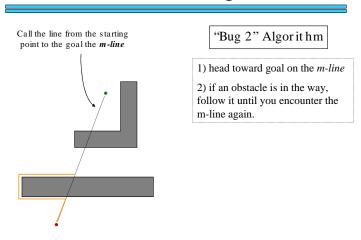
 - But, the line from robot to goal must intersect object even number of times (Jordan curve theorem)
 But then there is another intersection point on the boundary closer to object. Since we assumed there is a path, we must have crossed this pt on boundary which contradicts the definition of a leave point.

A better bug?



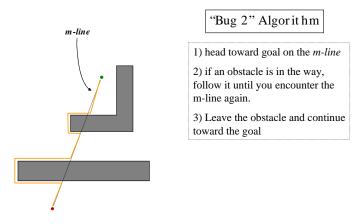
16-735, Howie Choset with slides from G.D. Hager and Z. Dodds

A better bug?



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

A better bug?



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

OK?

A better bug?

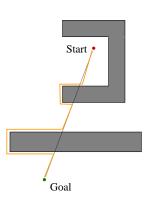
Start Goal

"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

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

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

BUG 2 More formally

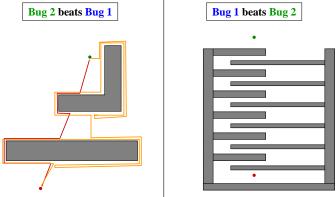
- $\begin{array}{ll} \bullet & \text{Let } q^L_{0} = q_{start;} \, i = 1 \\ \bullet & \text{repeat} \end{array}$
- - - from qLi-1 move toward qgoal along the m-line
 - until goal is reached or obstacle encountered at q_{i}^{H}
 - if goal is reached, exit
 - repeat
 - · follow boundary
 - $\begin{array}{l} \text{until } q_{goal} \text{ is reached or } q_i^H \text{ is re-encountered or} \\ \text{m-line is re-encountered}, x \text{ is not } q_i^H \text{, } d(x,q_{goal}) < d(q_i^H,q_{goal}) \text{ and way} \end{array}$ to goal is unimpeded
 - if goal is reached, exit
 - if q_{i}^{H} is reached, return failure
 - else
 - q^L_i = 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).



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

BUG 1 vs. BUG 2

- BUG 1 is an exhaustive search algorithm
 - it looks at all choices before commiting
- 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

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

"Quiz" Bug 2 analysis 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? p_i = p_i of s-line intersections of the i th obstacle

 $What \ is \ an environment \ where \ your \ upper \ bound \ is \ required?$ 16-735, Howie Choset with slides from G.D. Hager and Z. Dodds

To reach the upper bound, the robot needs to circumnavigate all obstacles (almost) completely for each hit point, minimally a sole obstacle once; see the looping obstacle.

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

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

2.2. Using A Range Sensor

• DistBug [Kamon and Rivlin 1997]

Senses the distance F in freespace from the current location X to the nearest obstacle in the direction of the goal.

Minor but effective extension of Bug2; allows the robot to leave the boundary early during wall following.

Easy to implement.

• TangentBug [Kamon et al. 1996]

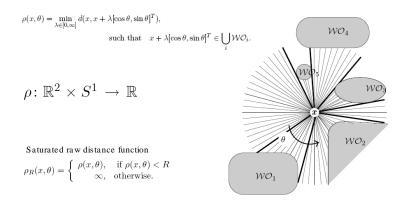
Senses the distance to the nearest obstacle all around the robot.

State of the art; good approximations to globally optimal paths in many realistic scenarios.

Slightly more involved to implement. (It is typically described in terms of the *local tangent graph*, but the following explanation succeeds without it.)

Both algorithms can handle zero, finite or infinite sensor ranges.

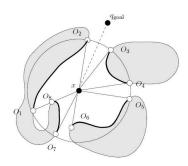
Raw Distance Function



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

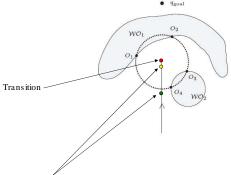
Intervals of Continuity

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



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

Motion-to-Goal Transition from Moving Toward goal to "following obstacles"

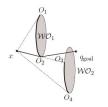


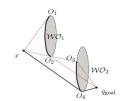
Currently, the motion-to-goal behavior "thinks" the robot can get to the goal

Now, it starts to see something --- what to do? Ans: For any O_i such that $d(O_pq_{goal}) < d(x,q_{goal})$, choose the pt O_i that minimizes $d(x,O_i) + d(O_pq_{goal})$ 16-735, Howie Choset with slides from G.D. Hager and Z. Dodds

Minimize Heuristic Example

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





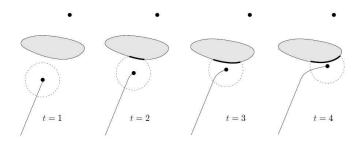
so moves toward O_2 . Note the line connecting O_2 and goal pass through obstacle

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

For any O_i such that $d(O_i, q_{goal}) < d(x, q_{goal})$, choose the pt O_i that minimizes $d(x, O_i) + d(O_i, q_{goal})$

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

Motion To Goal Example



For any O_i such that $d(O_i,q_{goal}) < d(x,q_{goal}),$ choose the pt O_i that minimizes $d(x,O_i) + d(O_i,q_{goal})$

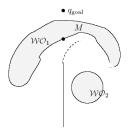
16-735, Howie Choset with slides from G.D. Hager and Z. Dodds

Transition from Motion-to-Goal

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

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 shorted distance to the goal

Followed obstacle: the obstacle that we are currently sensing

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

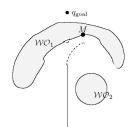
They start as the same

Transition from Motion-to-Goal

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

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 shorted distance to the goal

Followed obstacle: the obstacle that we are currently sensing

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

They start as the same

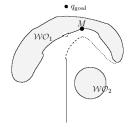
For any O_i such that $d(O_{i^i}q_{goal}) < d(x,q_{goal}),$ choose the pt O_i that minimizes $d(x,O_i) + d(O_{i^i}q_{goal})$

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

Boundary Following

Move toward the O_i on the followed obstacle in the "chosen" direction

Maintain d_{min} and d_{leave}



M is the point on the "sensed" obstacle which has the shorted 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_{min} and d_{leave}

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

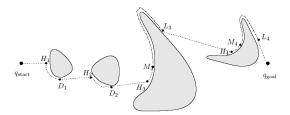
- d_{min} is the shortest distance, observed thus far, between the sensed boundary of the obstacle and the goal
- $\mathbf{d}_{\mathrm{leave}}$ is the shortest distance between any point in the currently sensed environment and the goal

 $V_R(x) = \{y \in Q_{\text{free}} | d(x, y) < R \text{and} \lambda x + (1 - \lambda)y \in W_{\text{free}} \text{for all } \lambda \in [0, 1] \}$

$$d_{\text{leave}}(x) = \min_{y \in V(x)} d(x, y).$$

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

Example: Zero Sensor Range

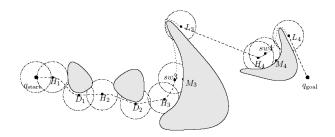


- Robot moves toward goal until it hits obstacle 1 at H1 Pretend there is an infinitely small sensor range and the Oi which minimizes the heuristic is to the right
- Keep following obstacle until robot can go toward obstacle again

- 4. Same situation with second obstacle
 5. At third obstacle, the robot turned left until it could not increase heuristic
 6. dmin is distance between M3 and goal, dleave is distance between robot and goal because sensing distance is zero

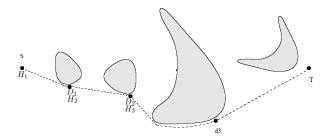
16-735, Howie Choset with slides from G.D. Hager and Z. Dodds

Example: Finite Sensor Range

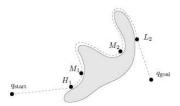


16-735, Howie Choset with slides from G.D. Hager and Z. Dodds

Example: Infinite Sensor Range



d_{min} is constantly updated



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

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_{min} which is the shortest distance observed thus far between the sensed boundary of the obstacle and the goal
- A value d_{leave} which is the shortest distance between any point in the currently sensed environment and the goal
- Terminate boundary following behavior when $d_{\text{leave}} < d_{\text{min}}$

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

Tangent Bug Algorithm

- 1) repeat
 - a) Compute continuous range segments in view
 - b) Move toward $n \in \{T, O_i\}$ that minimizes $h(x, n) = d(x, n) + d(n, q_{goal})$ antil
 - 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_{leave} and d_{min} until
 - a) goal is reached
 - b) a complete cycle is performed (goal is unreachable)
 - c) $d_{leave} < d_{min}$

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 $D(x) = \min_{c} d(x,c)$ $c \in \bigcup_{i} WO_{i}$
- Let G(x) = D(x) $W^* \leftarrow$ some safe following distance
- Note that ∇ 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)

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

2.3. Remarks

- Contact sensors as a last resort: react if you bump into something.
- What if the obstacle moves?

3. References

3.1. References

- I. Kamon, E. Rivlin, E. Rimon, "A new range-sensor based globally convergent navigation algorithm for mobile robots¹". *IEEE International Conference on Robotics and Automation*, pp. 429–435, 1996.
- I. Kamon, E. Rivlin, "Sensory-based motion planning with global proofs²". *IEEE Transactions on Robotics and Automation* 13(6), pp. 814–822, 1997.
- V. Lumelsky, A. Stepanov, "Path-planning strategies for a point mobile automaton moving amidst unknown obstacles of arbitrary shape³". *Algorithmica* 2(4), pp. 403–430, 1987.

¹ http://www.cs.technion.ac.il/~ehudr/publications/pdf/KamonRR96i.pdf

² http://www.cs.technion.ac.il/~ehudr/publications/pdf/KamonR97a.pdf

http://link.springer.com/content/pdf/10.1007%2FBF01840369.pdf