



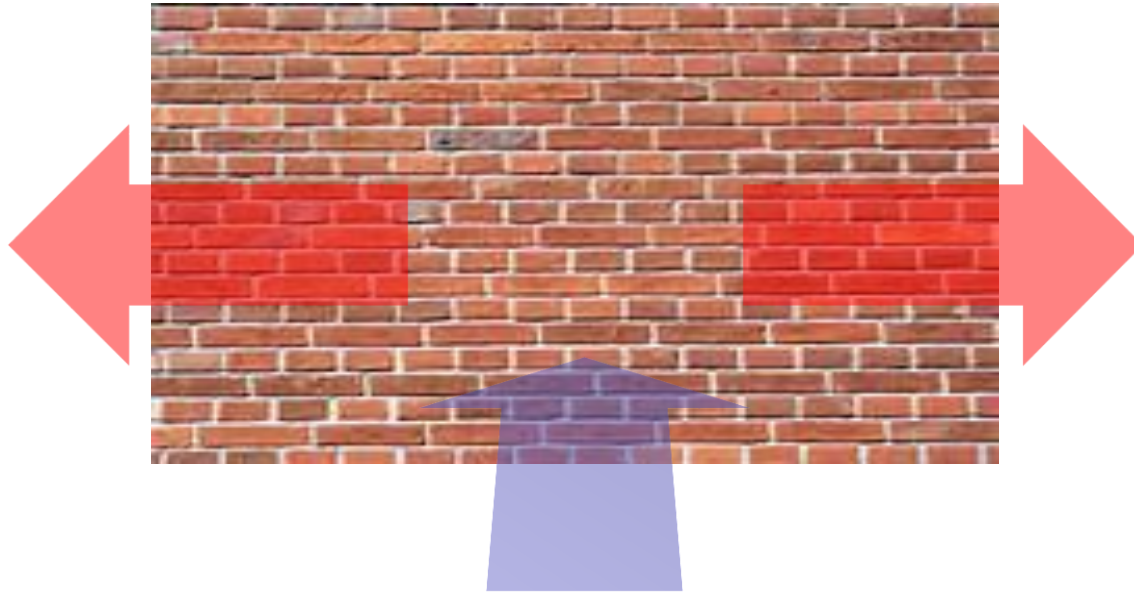
# Robotics Programming Laboratory

Bertrand Meyer  
Jiwon Shin

## Lecture 5: Obstacle Avoidance

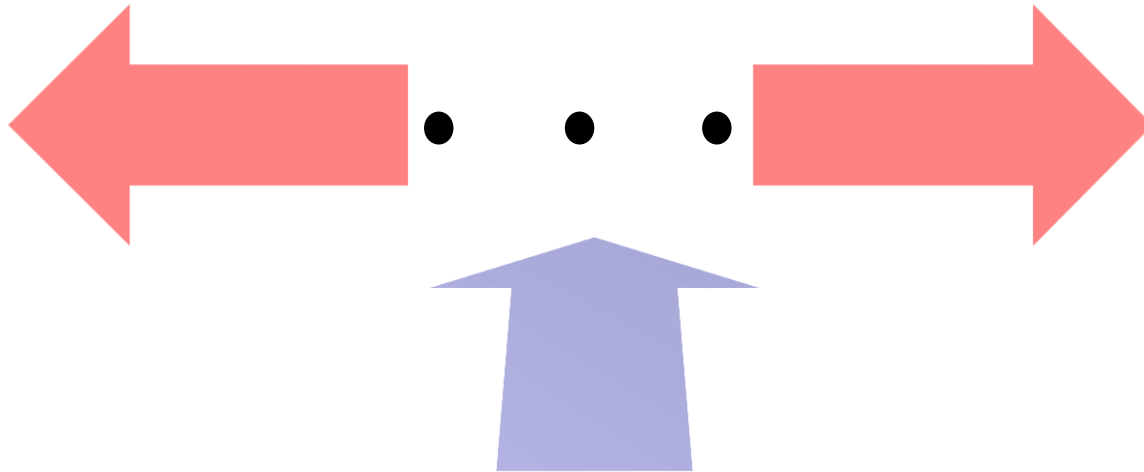
# Obstacle avoidance: our perspective

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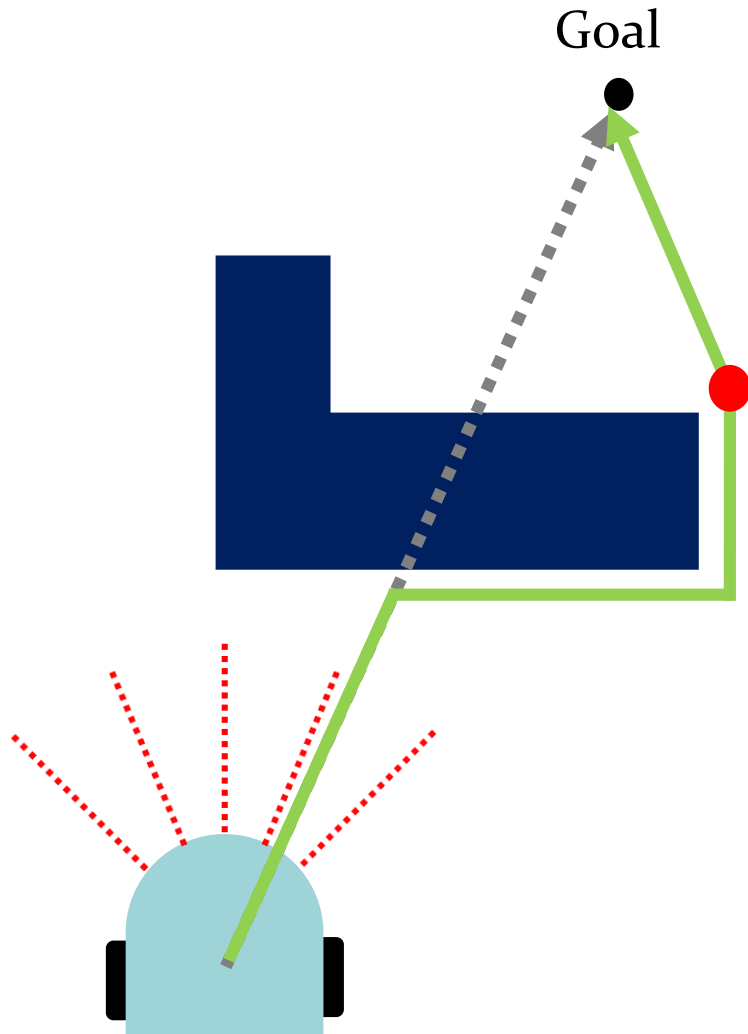
# Obstacle avoidance: robot's perspective

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- Known:
  - Goal position
  - Current position
  - Sensing ability to detect nearby obstacles
- Sense -> Act: does not store any past information
- Sensor:
  - Bug 0, Bug 1, Bug 2: tactile sensor
  - Tangent Bug: range sensor



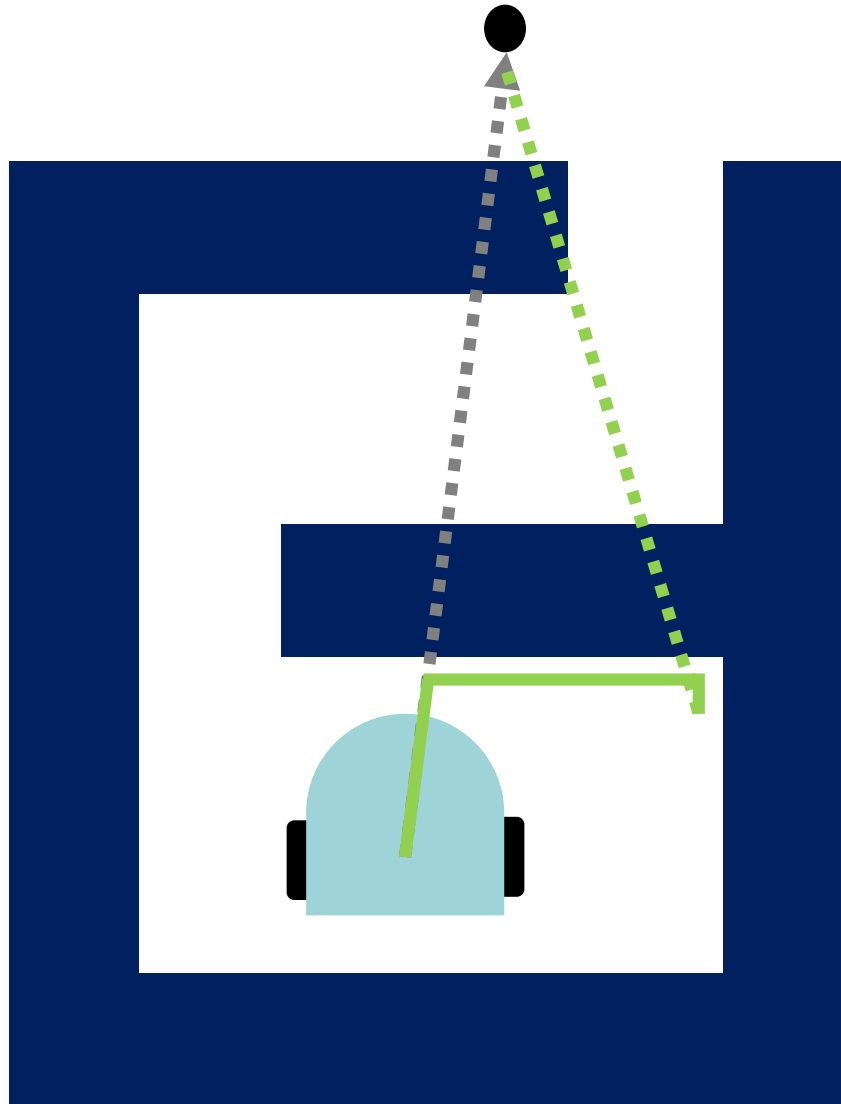
1. Move toward the goal:
  1. If the goal is reached: Stop
  2. If an obstacle is in the way: Go to step 2
2. Follow the obstacle boundary:
  1. If no obstacle in the way, go back to step 1.

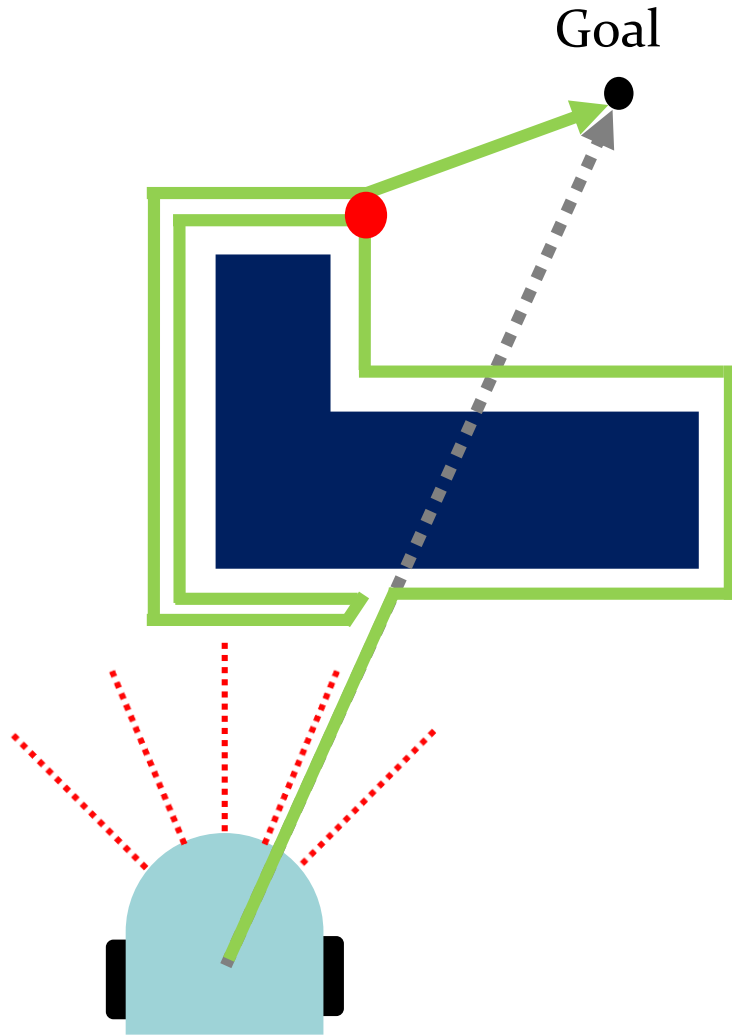
# When does Bug 0 fail?

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Goal





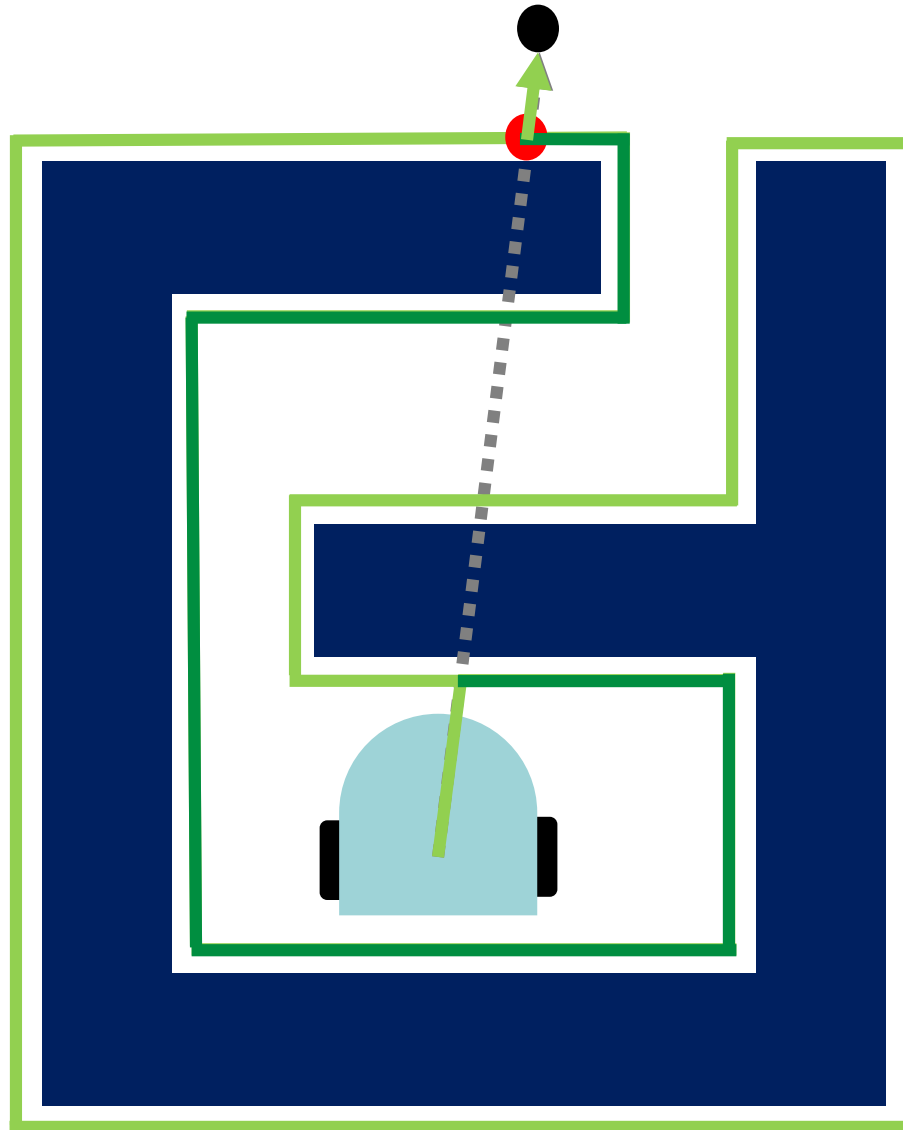
1. Move toward the goal:
  1. If the goal is reached: Stop
  2. If an obstacle is in the way: Go to step 2
2. Follow the obstacle boundary:
  1. Mark the closest
  2. After a complete loop: Go to the closest point to the goal then go back to step 1.

# Will Bug 1 fail?

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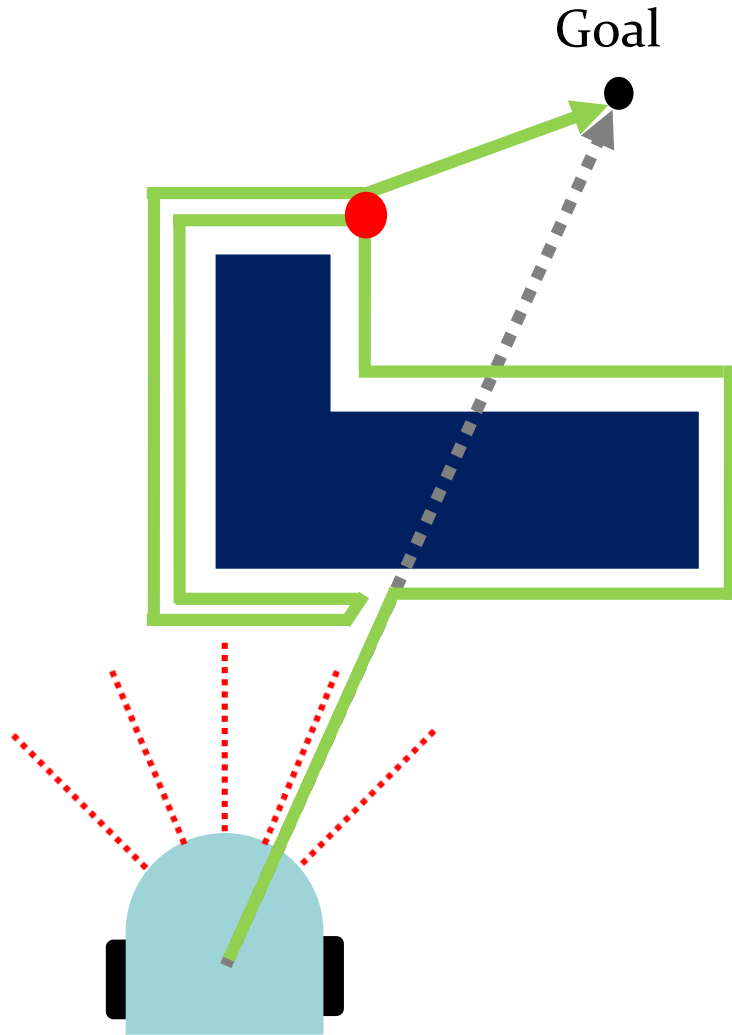


Goal





# How much would Bug 1 travel?



Given

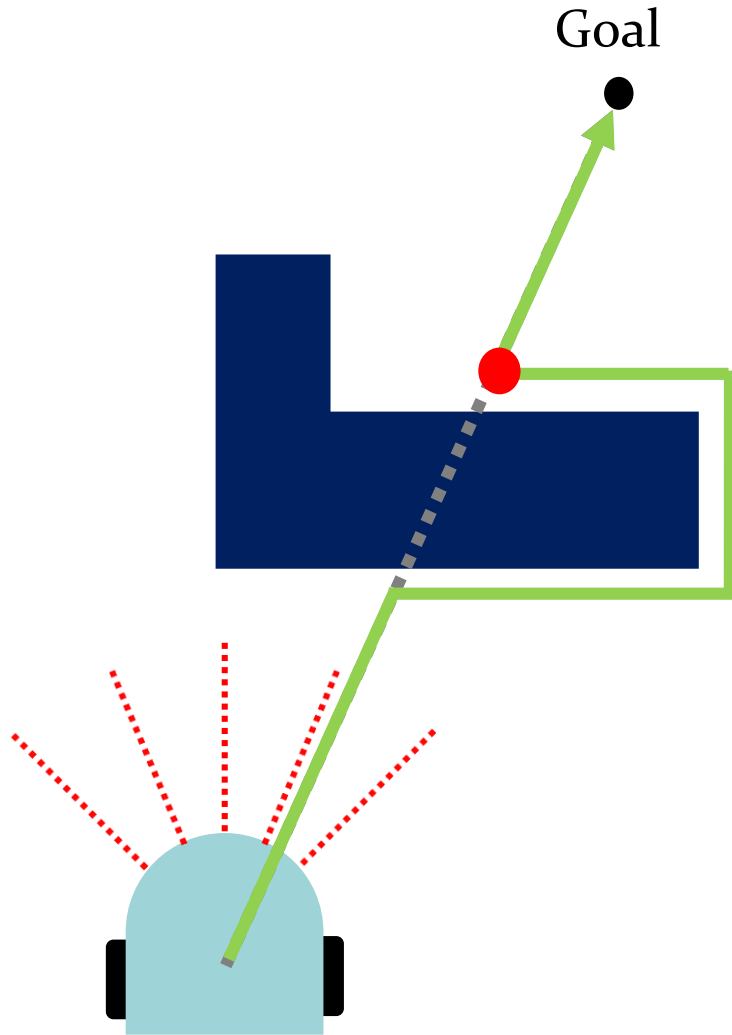
- $D$ : distance between start and goal
- $P_i$ : Perimeter of  $i$ 'th obstacle

Shortest travel distance?

- $D$

Longest travel distance?

- $D + 1.5 \sum_i P_i$

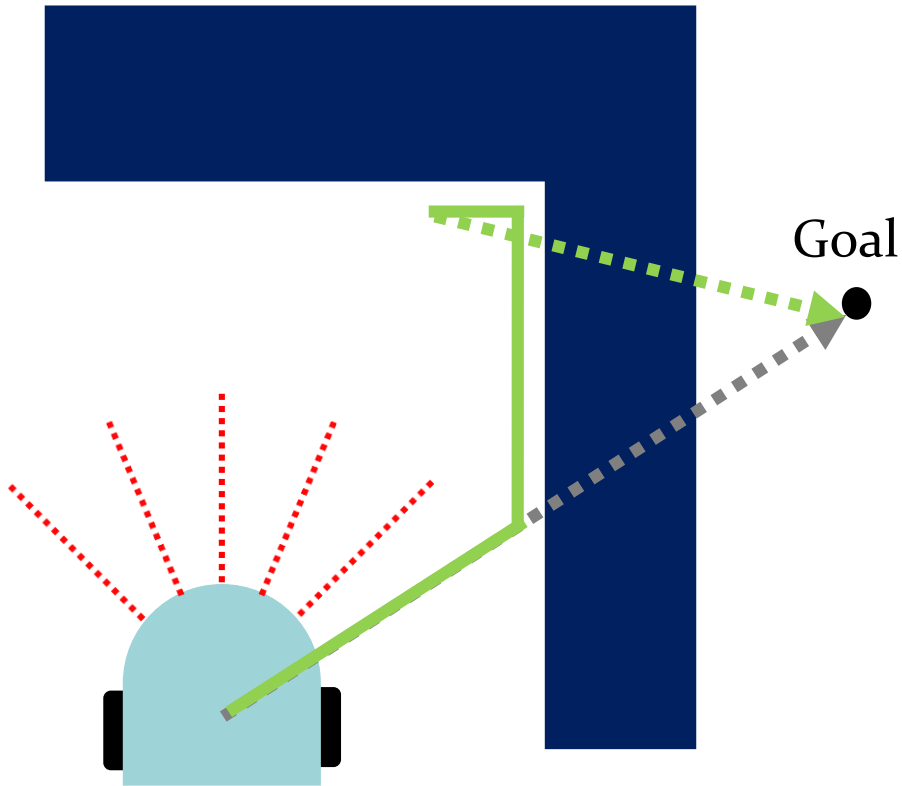


1. Move toward the goal:
  1. If the goal is reached: Stop
  2. If an obstacle is in the way: Go to step 2
2. Follow the obstacle boundary:
  1. If the goal line is crossed and is closer to the goal: Go to step 1.

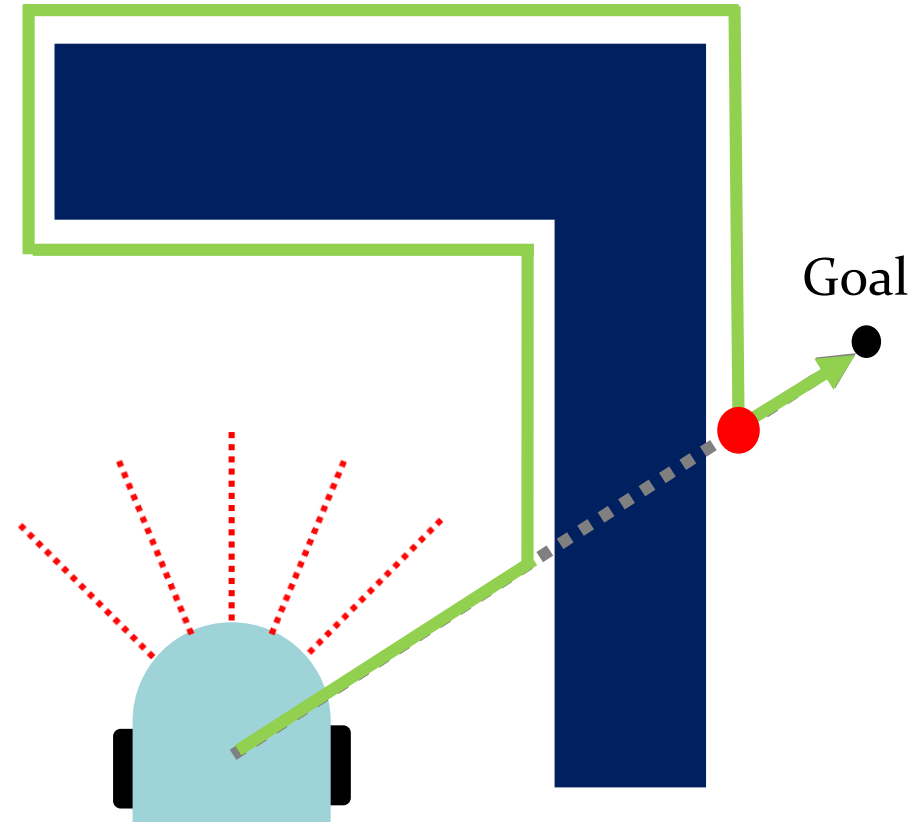
# Is crossing the goal line important?



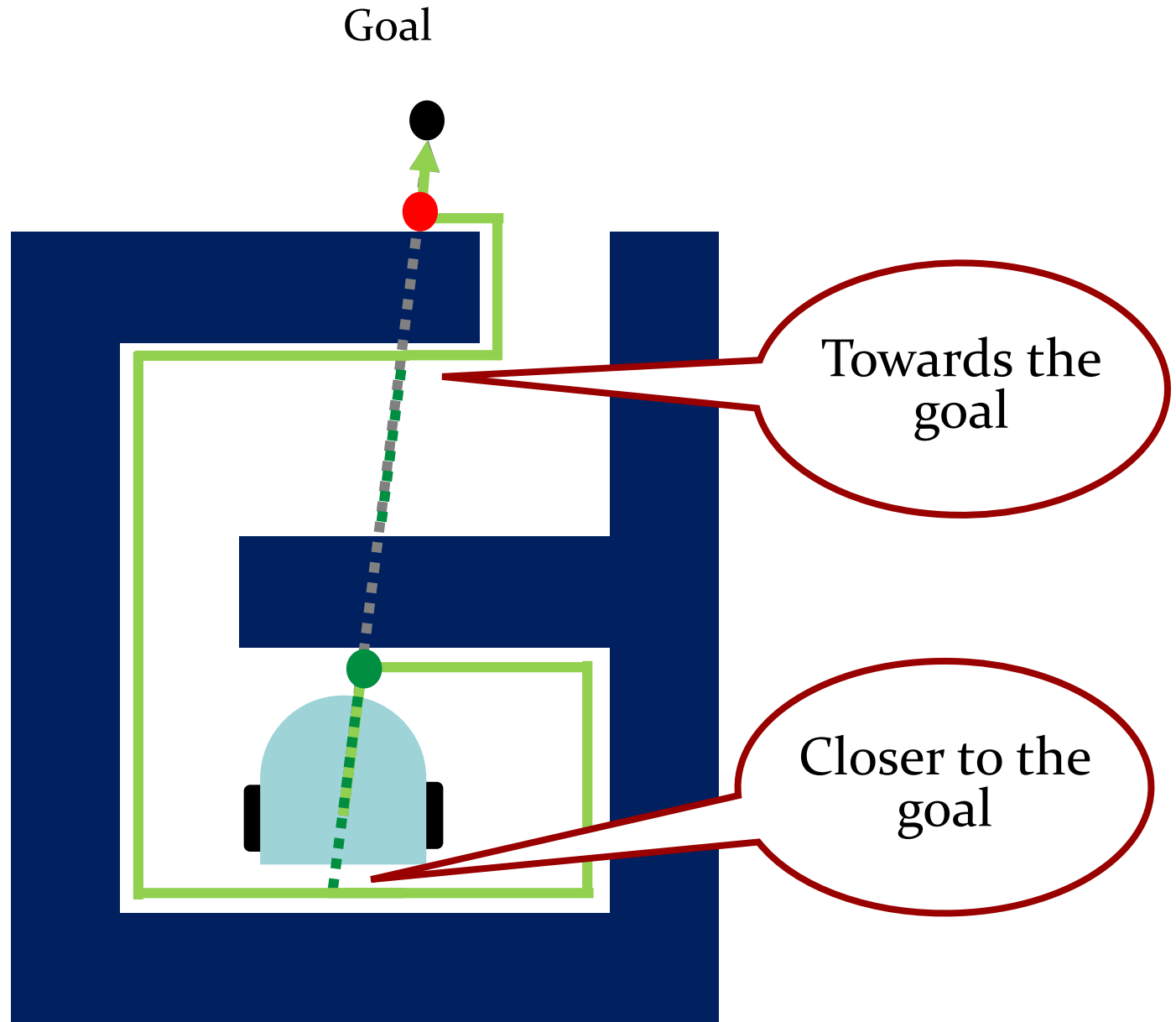
Bug 0



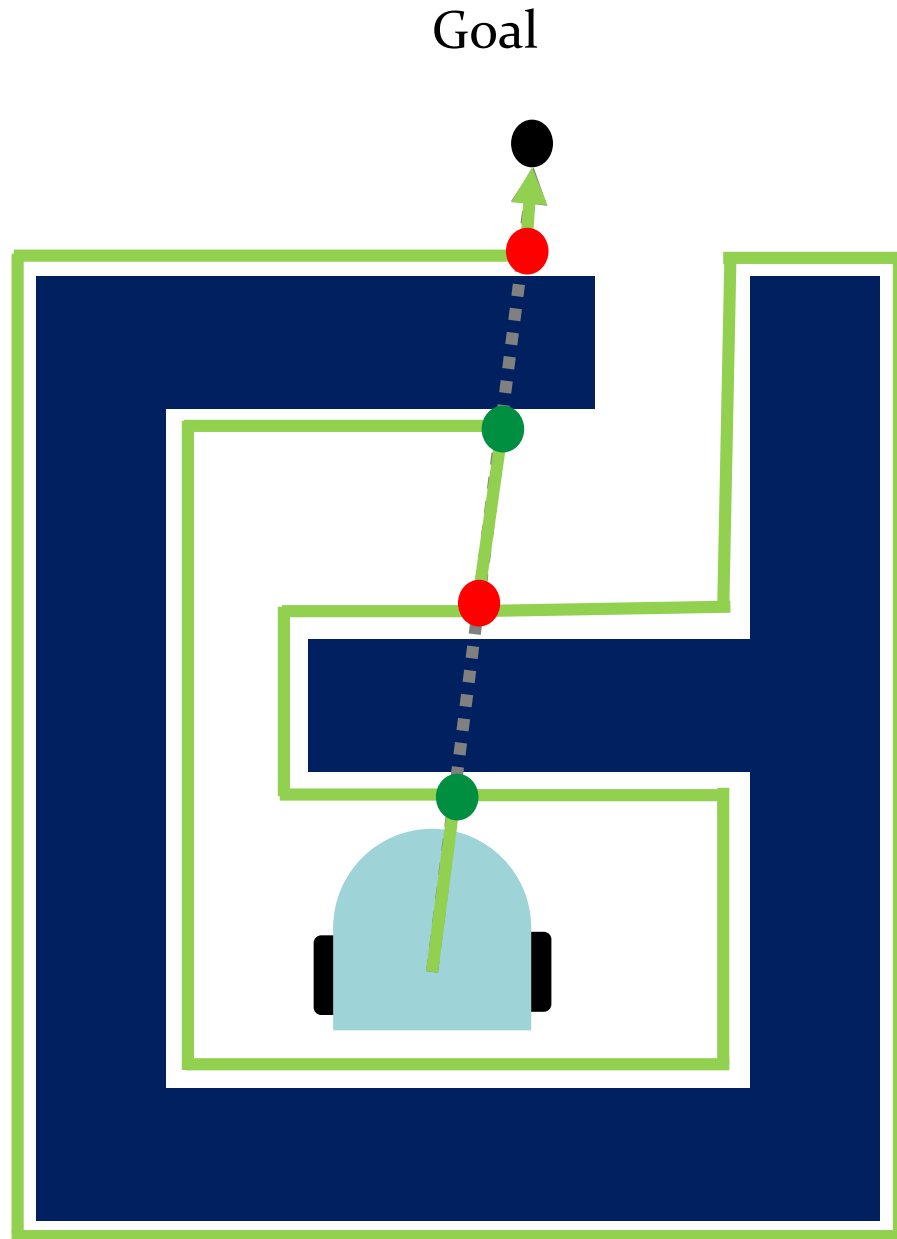
Bug 2



# How well does Bug 2 work?

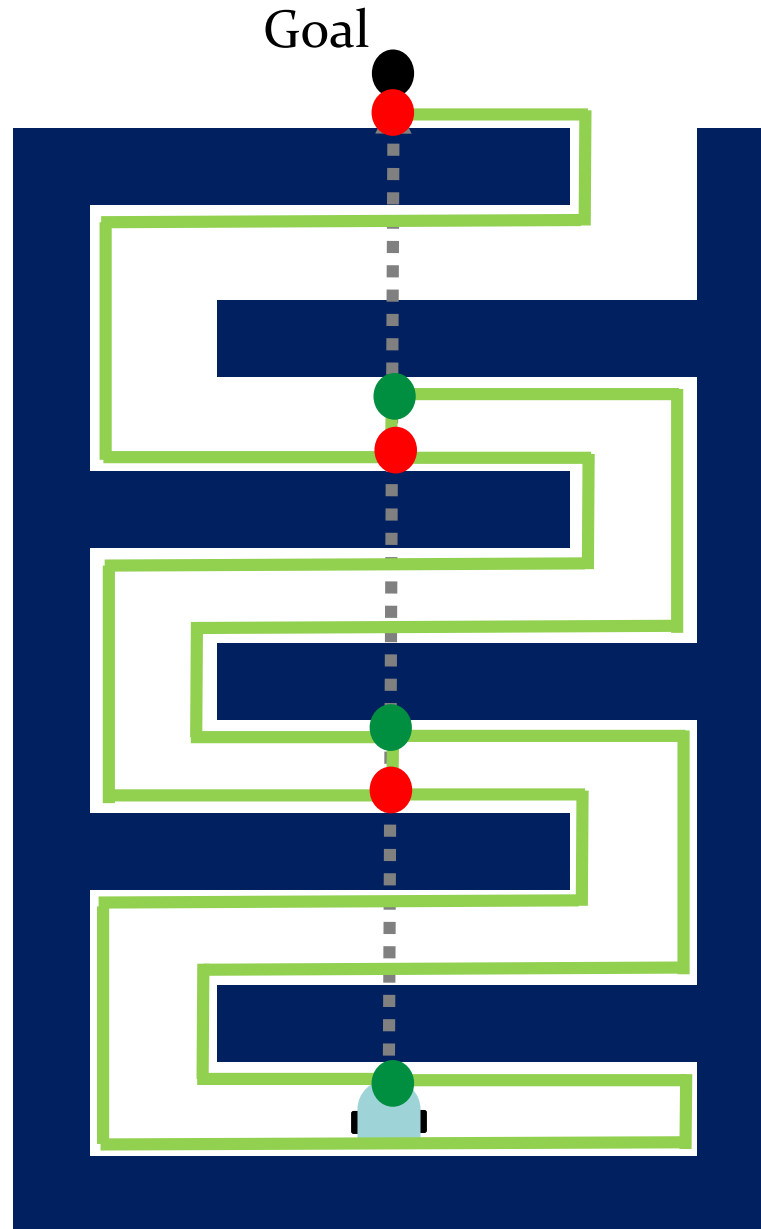


# How well does Bug 2 work?

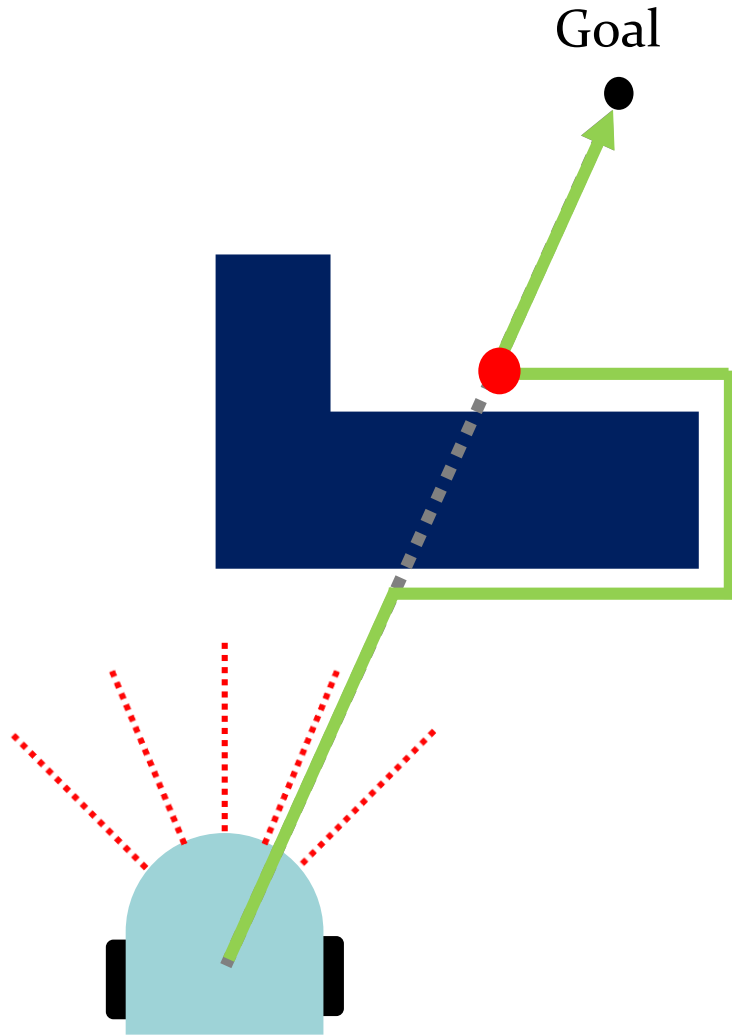


# How well does Bug 2 work?

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# How much would Bug 2 travel?



Given

- $D$ : distance between start and goal
- $P_i$ : Perimeter of  $i$ 'th obstacle
- $n_i$ : number of times  $i$ 'th obstacle crosses the goal line

Shortest travel distance?

- $D$

Longest travel distance?

- $D + 1/2 \sum_i n_i P_i$



## Bug 1

- Exhaustive search: analyze all choices before committing
- More predictable performance

## Bug 2

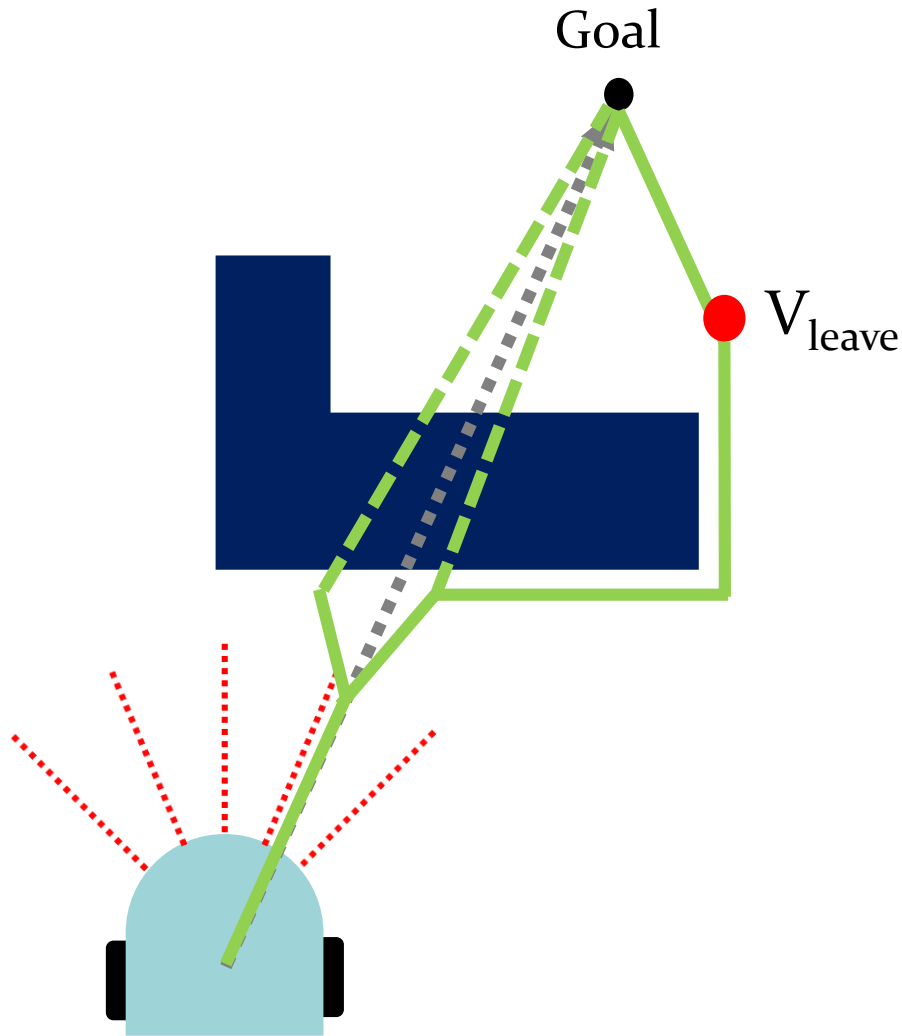
- Greedy search: take the first viable choice
- Generally outperforms Bug 1 but could be worse if the obstacles are complex



# Can we do better if we can see more?

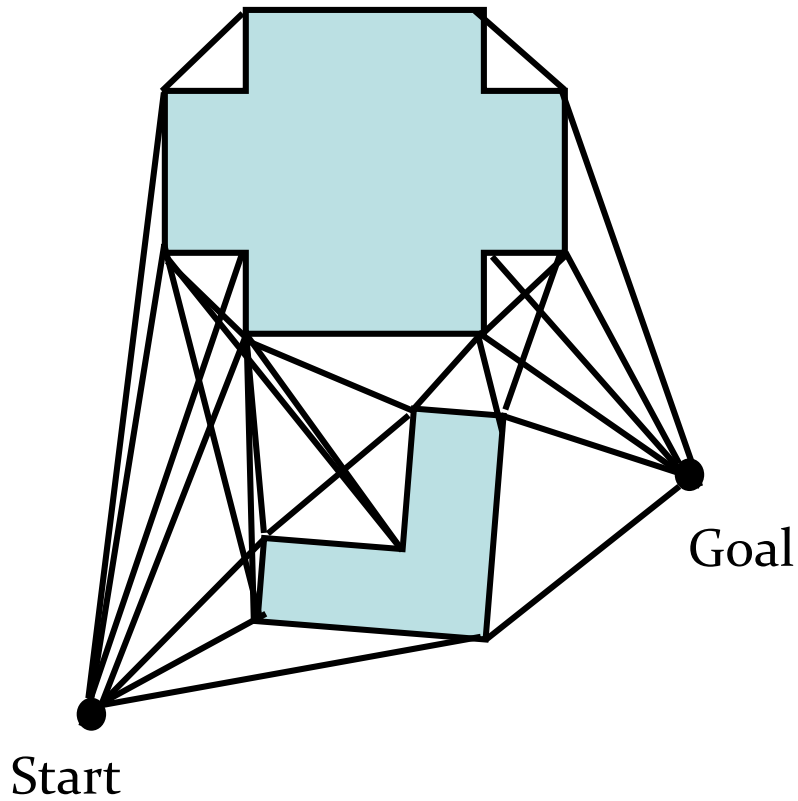
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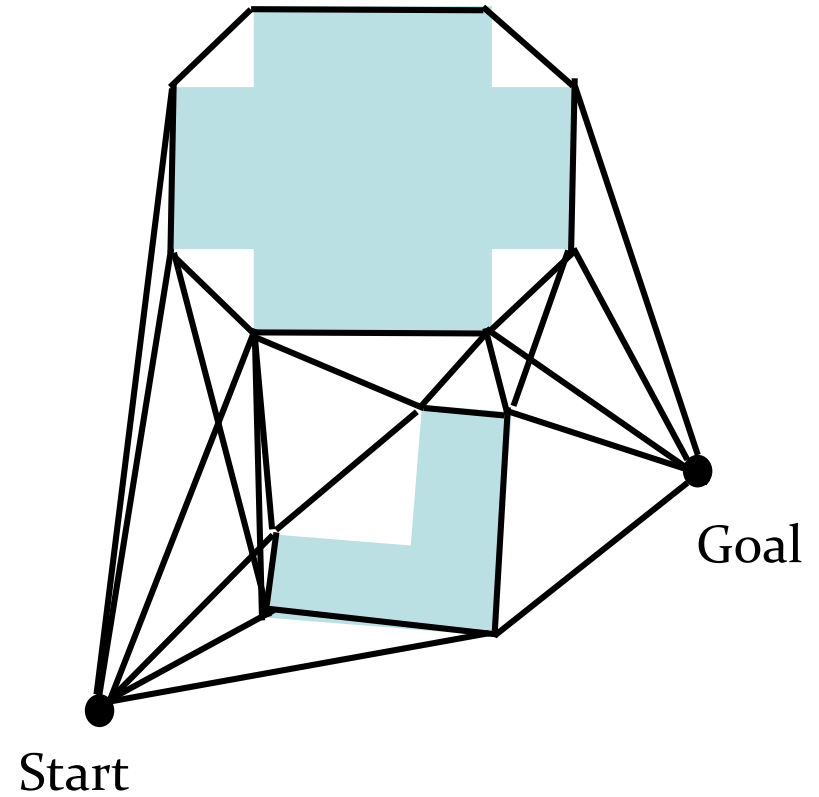


1. Move toward the goal:
  1. If the goal is reached: Stop
  2. If a local minimum is detected: Go to step 2
2. Move along the boundary marking  $d_{\min}$ :
  1. If the goal is reached: Stop
  2. If  $d(V_{\text{leave}}, \text{goal}) < d_{\min}$  : Go to step 3
3. Perform the transition phase:
  1. Move directly towards  $V_{\text{leave}}$  until Z, where  $d(Z, \text{goal}) < d_{\min}$ : Go to step 1

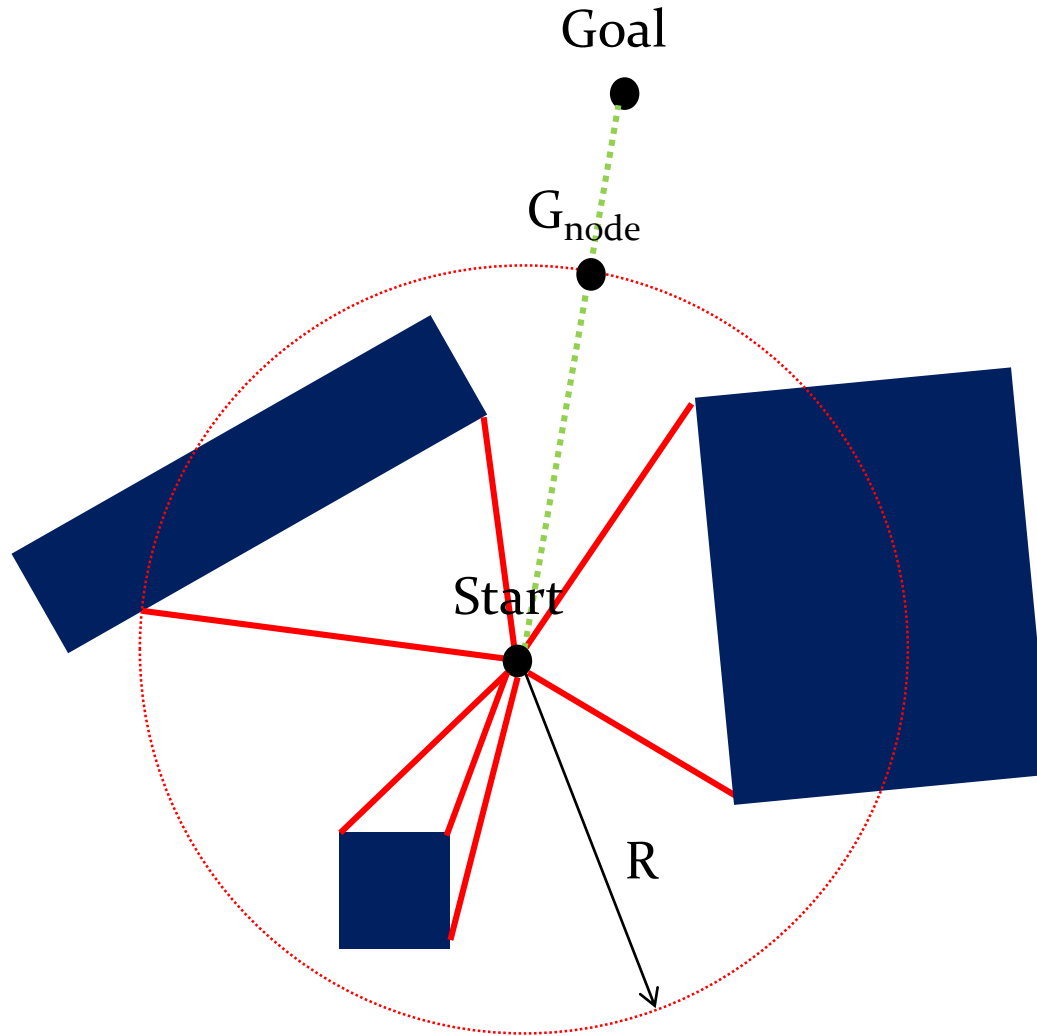
## Visibility graph



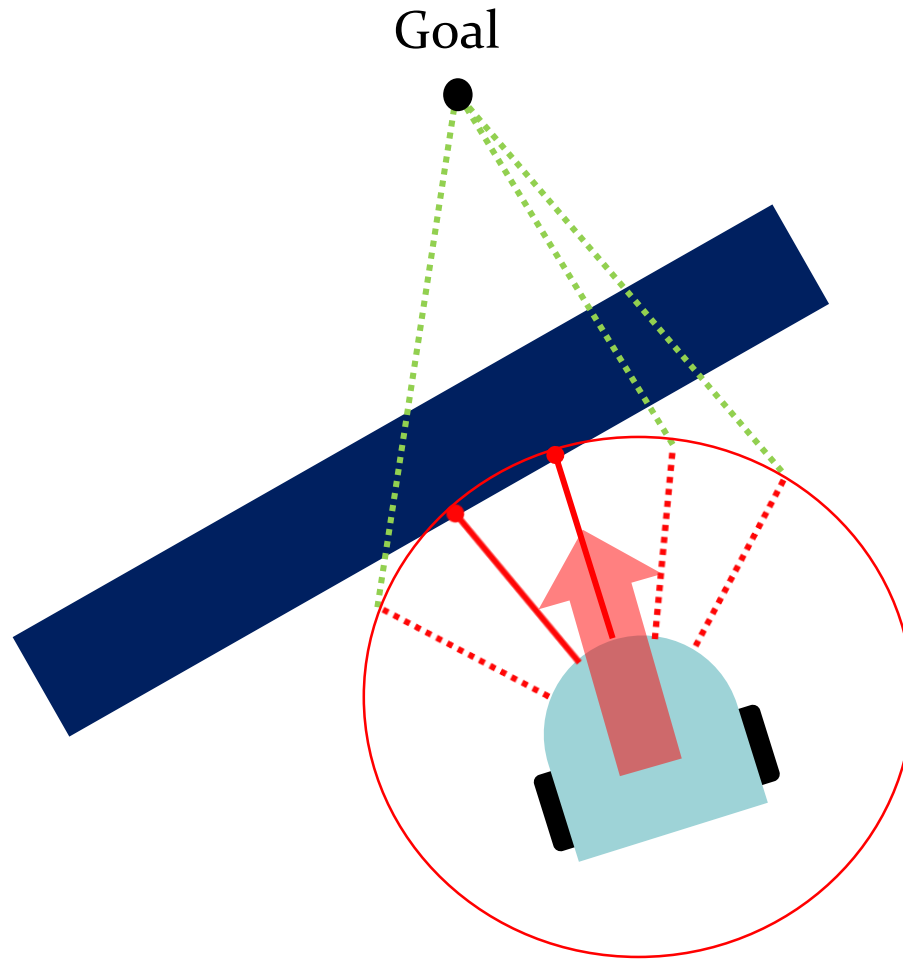
## Tangent graph



# Local tangent graph

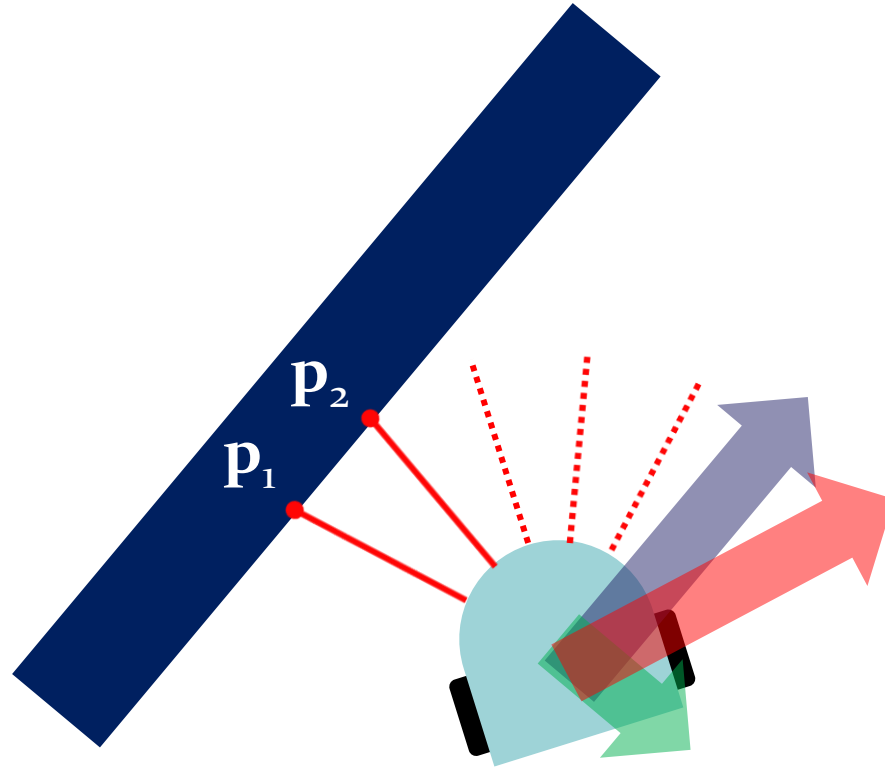


# Local minimum detection



$$d(x, \text{goal}) \leq d(V, \text{goal}) \text{ for all } V$$

# Wall Following

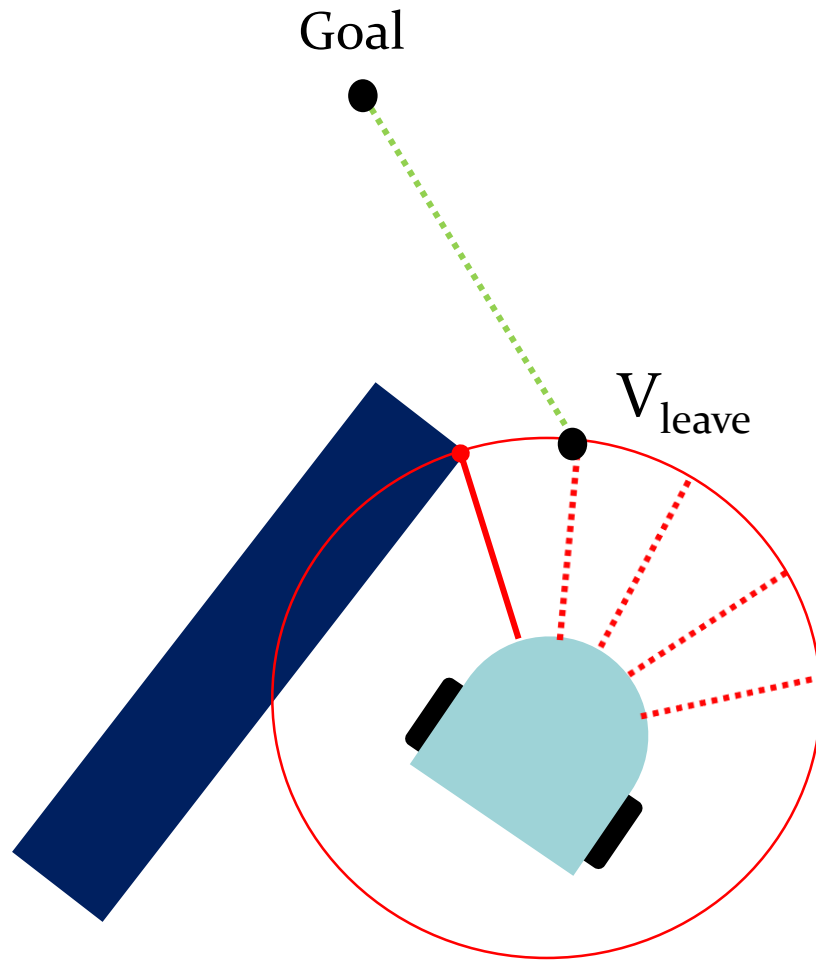


$$\mathbf{v}_{\text{wall}} := \mathbf{p}_2 - \mathbf{p}_1$$

$$\mathbf{v}_{\text{distance}} := (d_{\text{current}} - d_{\text{desired}}) * \mathbf{v}_{\text{perpendicular}}$$

$$\mathbf{v}_{\text{robot}} := d_{\text{desired}} * \mathbf{v}_{\text{wall}} + \mathbf{v}_{\text{distance}}$$

# Leave condition detection



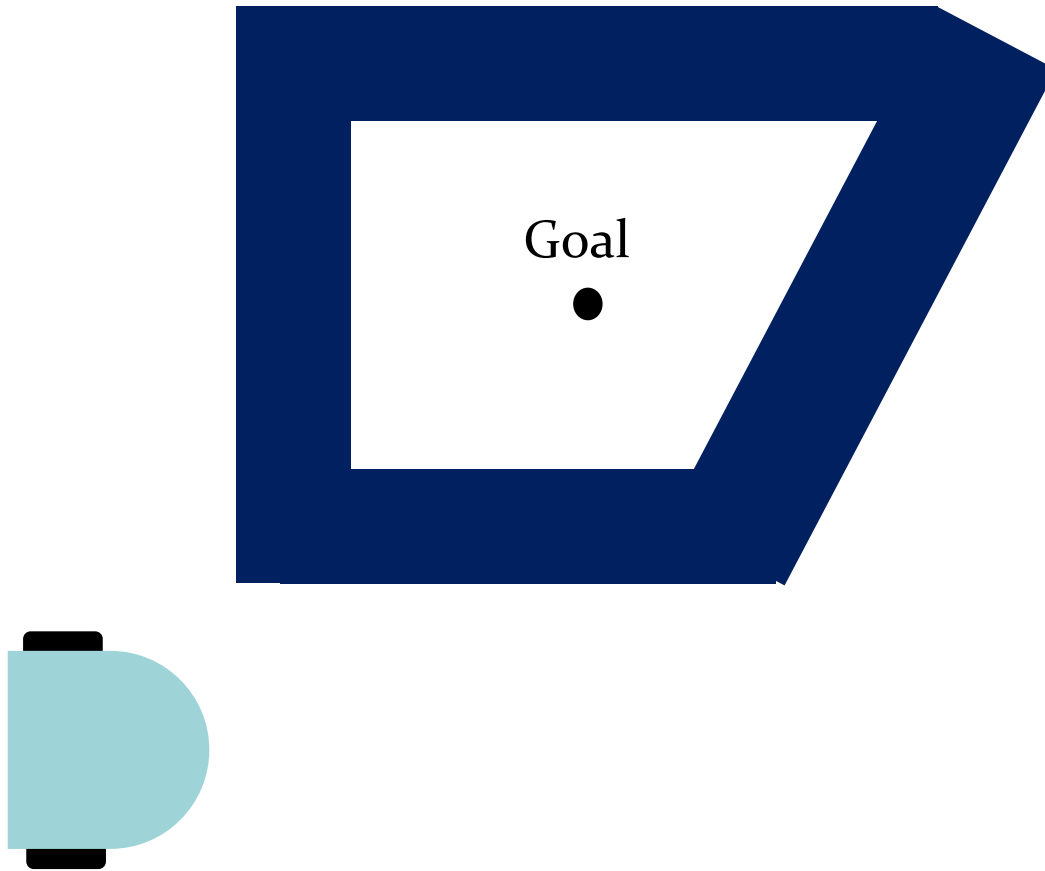
$$d(V_{\text{leave}}, \text{goal}) < d_{\text{min}}$$

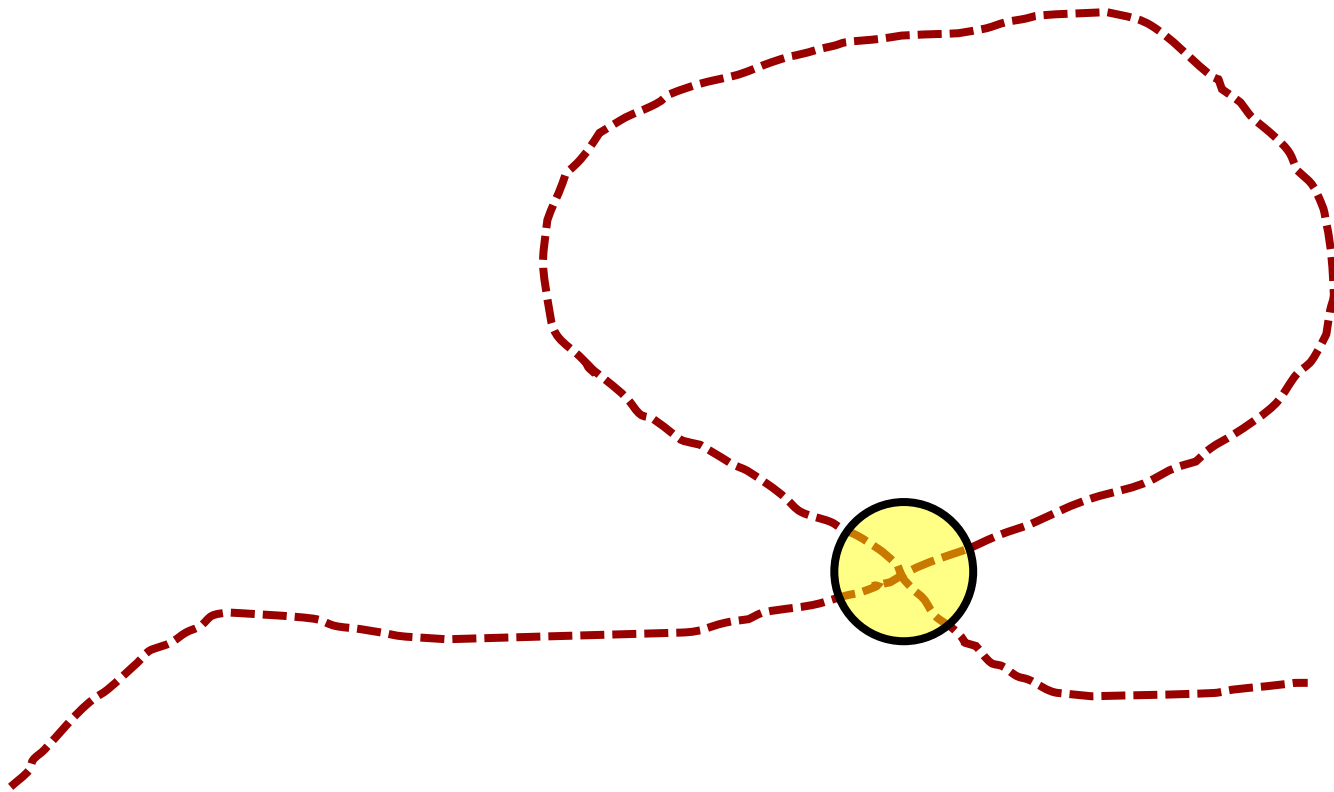




# Unreachable goal

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Challenging!

- Drift
- Limited sensor information



```
class
  TANGENT_BUG
feature
  update_velocity ( ... )
  do
    if state = go_to_goal_s then
      go_to_goal ( ... )
    elseif state = wall_following_s then
      follow_wall ( ... )
    elseif state = transition_s then
      transition_to_goal ( ... )
    ...
  end
```

```
class
  TANGENT_BUG
feature
  update_velocity ( ... )
  do
    current_state.update_velocity ( ... )
    ...
  end
current_state: STATE
```

```
deferred class
  STATE
feature
  update_velocity ( ... )
```

```
class
  GO_TO_GOAL_STATE
inherit
  STATE
```