

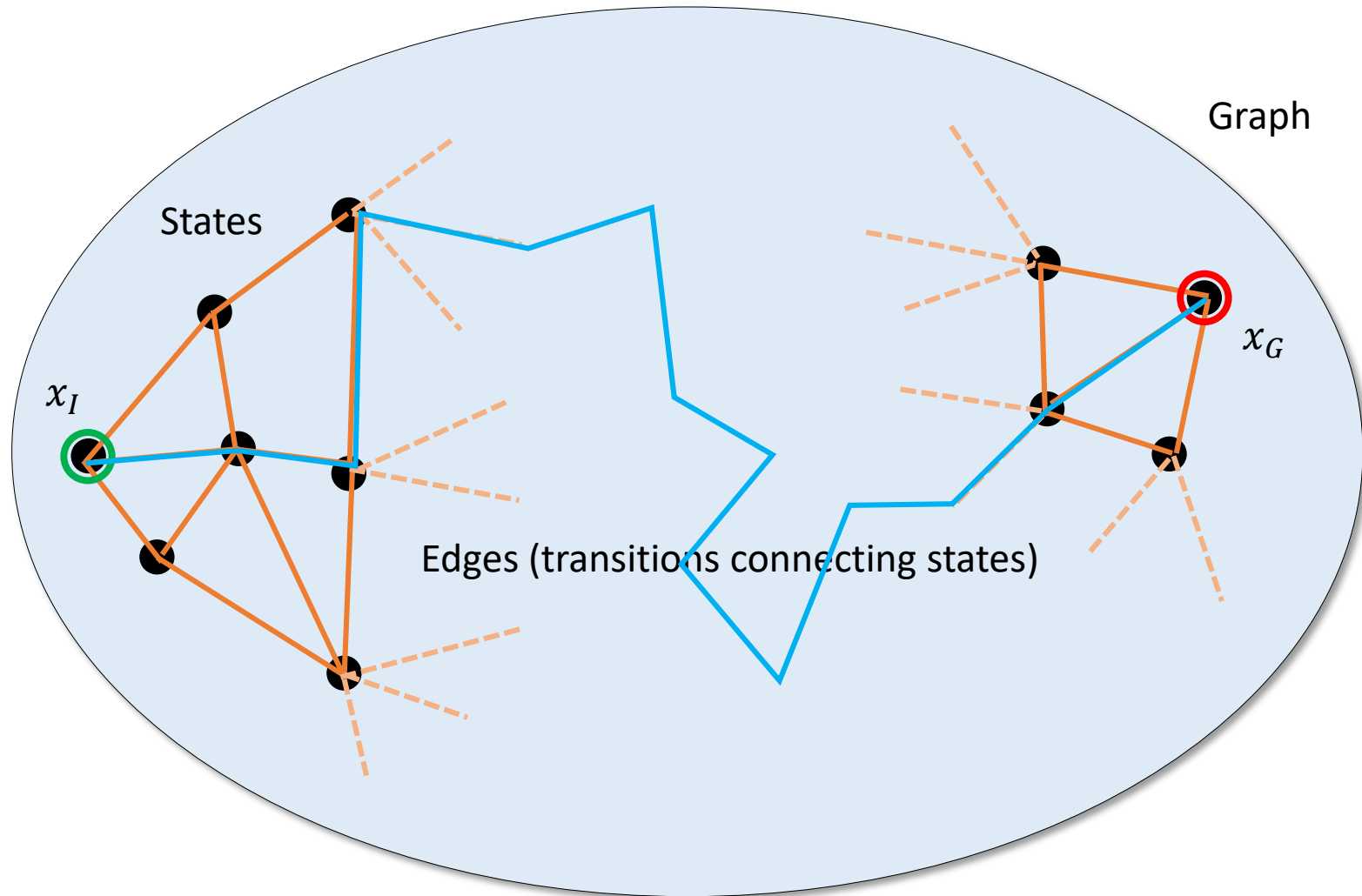
Review: A Generic Graph Search Algorithm

```
input:  $G = (V, E)$ ,  $x_I$ ,  $x_G$ 
AddToQueue( $x_I$ , Queue); // Add  $x_I$  to a queue of nodes to be expanded
while (!IsEmpty(Queue))
     $x \leftarrow \text{Front}(\text{Queue})$ ; // Retrieve the front of the queue
    if( $x.\text{expanded} == \text{true}$ ) continue; // Do not expand a node twice
     $x.\text{expanded} = \text{true}$ ; // Mark  $x$  as expanded
    if( $x == x_G$ ) return solution; // Return if goal is reached
    for each neighbor  $n_i$  of  $x$  // Add all neighbors of to the queue
        if( $n_i.\text{expanded} == \text{false}$ ) AddToQueue( $n_i$ , Queue)
return failure;
```

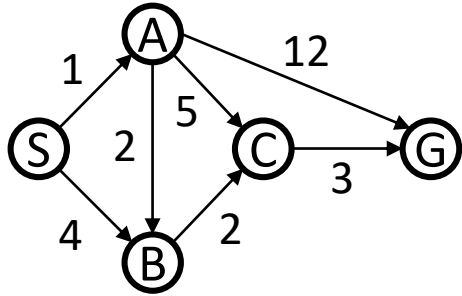
Different graph search algorithms (breadth first, depth-first, uniform-cost, ...) differ at the function AddToQueue

To retrieve the actual path, use **back pointers**

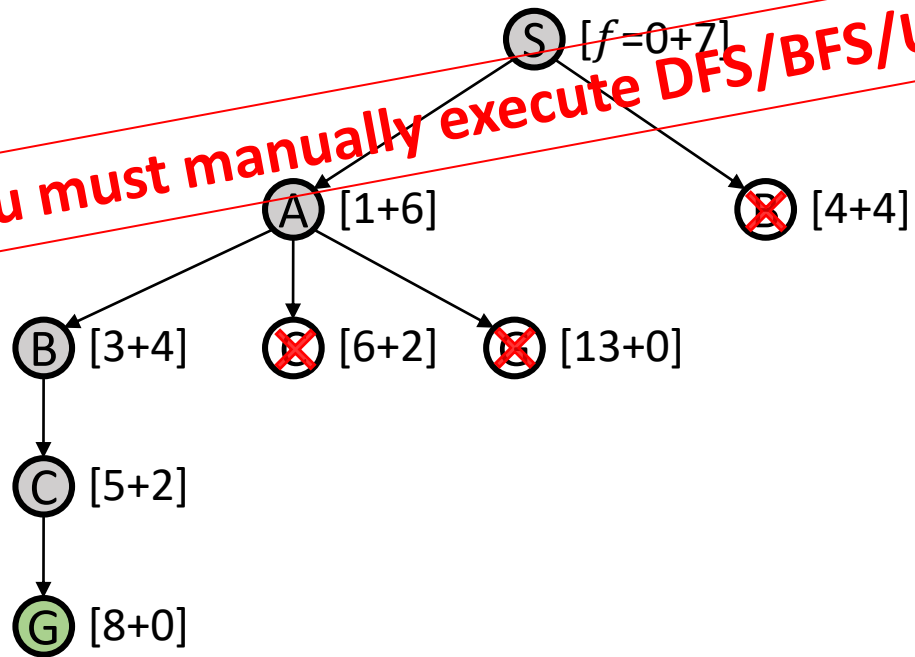
Review: Graph View of Search



Very Important: Manual Execution



For a good understanding, you must manually execute DFS/BFS/UC/A*!



State	$h(x)$
S	7
A	6
B	4
C	2
G	0

CS 460/560

Introduction to Computational Robotics
Fall 2019, Rutgers University

Lecture 11

Combinatorial Planning: In the Plane

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

Instructor: Jingjin Yu

Outline

Holonomic robots

Convex shapes, revisited

Combinatorial planning

- ⇒ Computing intersection and visibility
- ⇒ Vertical cell decomposition
- ⇒ Shortest-path roadmaps
- ⇒ Maximum clearance roadmaps

Holonomic Robots

Formally, a robot is **holonomic** if its constraints are of the form

$$f(x_1, \dots, x_n, t) = 0$$

Important: constraints cannot contain \dot{x}_i , \ddot{x}_i , and so on

Essentially, a **holonomic robot** has its controllable DOFs (i.e., number of independent directions along which it can move) equal to its configuration space dimension

Examples

⇒ Holonomic robot: a vehicle with Mecanum wheels

⇒ Configuration space: $SE(2)$, 3 dimensions: x, y, θ

⇒ Such robots can move and rotate at the same time

⇒ Non-holonomic: a regular car

⇒ Configuration space: $SE(2)$, 3 dimensions: x, y, θ

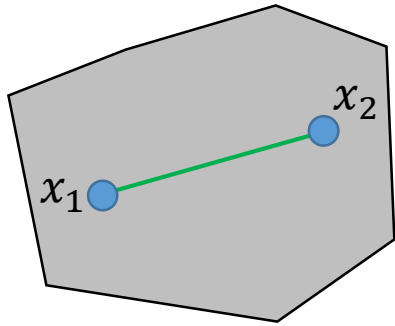
⇒ Can only move along its heading direction: two controllable degrees



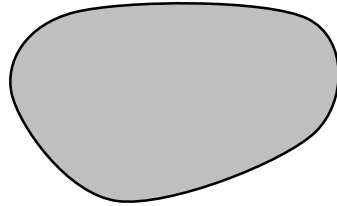
For this week, we work with holonomic robots unless otherwise stated

Convex Shapes, Revisited

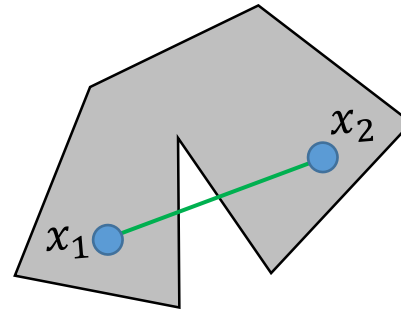
Recall: in a Euclidean space, a set X is **convex** if given any $x_1, x_2 \in X$, all points on the straight-line segment x_1x_2 belong to X .



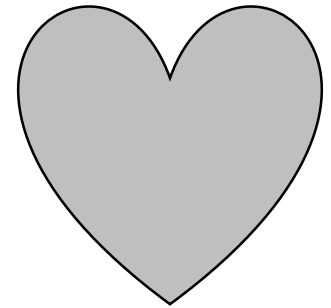
✓



✓



✗



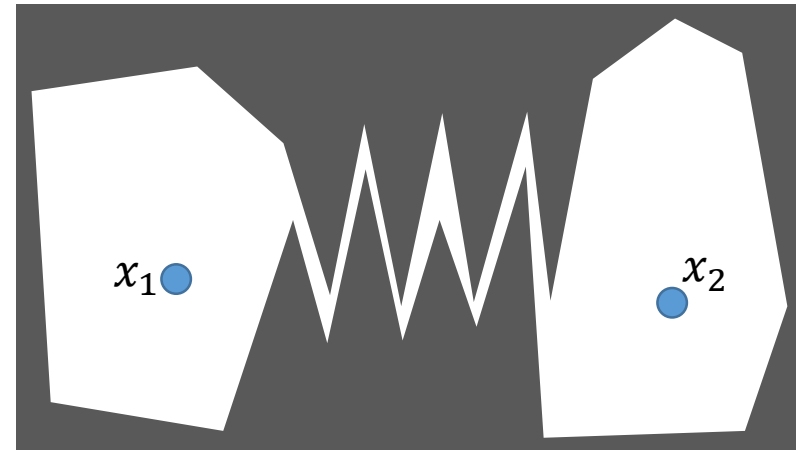
✗

For holonomic robots, planning inside a convex shape is easy

- ⇒ Why?
- ⇒ Simply connect the two points!
- ⇒ Not true for general robots (e.g., cars)

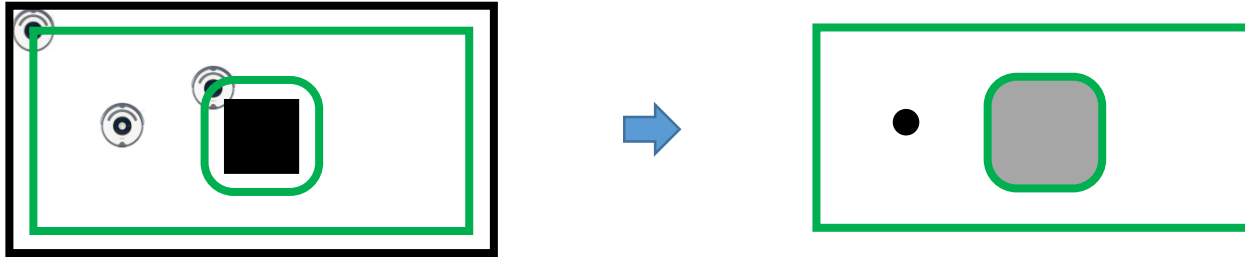
It's not as easy in non-convex shapes

Remember: we do not have
a “God’s view” in computer!



Combinatorial Planning

Recall that the C-space abstraction “shrinks” the robot into a point



With C_{free} computed and the robot shrunk into a point, we can attempt path planning from x_I to x_G

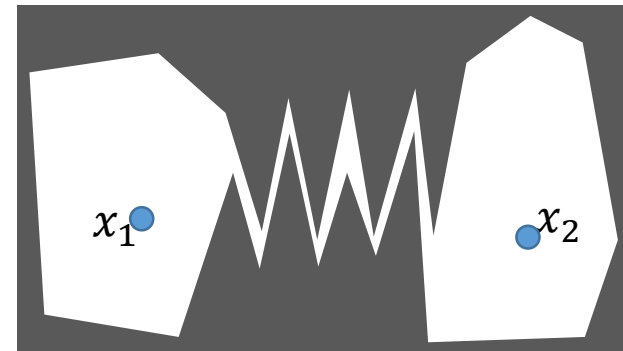
- ⇒ Now we are looking for a **simple path** connecting x_I and x_G
- ⇒ No need to consider the geometry of the robot anymore!

Challenge: the configuration space may be **non-convex**

- ⇒ So the path is not a straight line in C_{free}

Classical (2D) methods

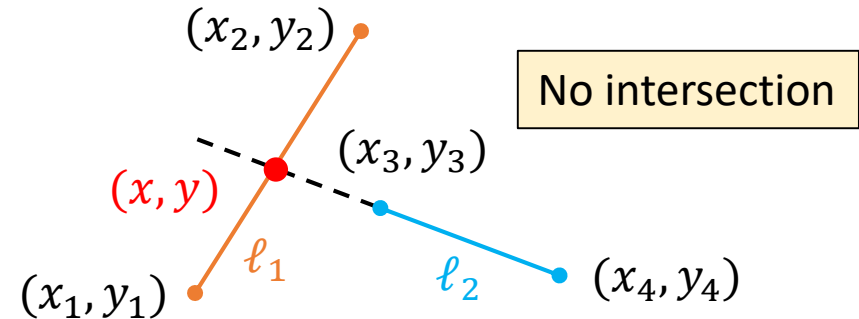
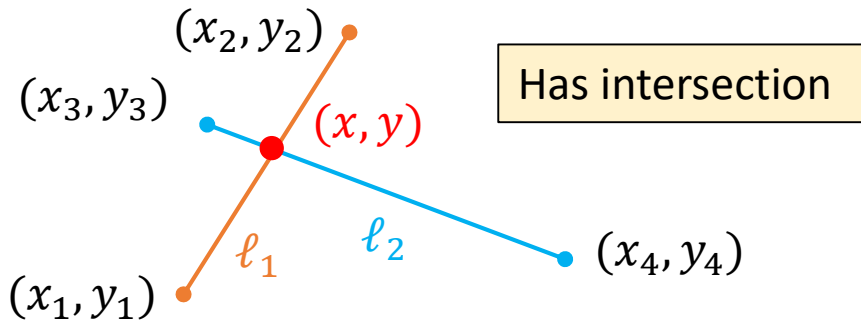
- ⇒ Cell decomposition
- ⇒ Shortest-path roadmaps
- ⇒ Maximum clearance roadmaps



Finding Intersection and Checking Visibility

These methods need to **find intersections** and **check visibility**

Finding intersection of two line segments

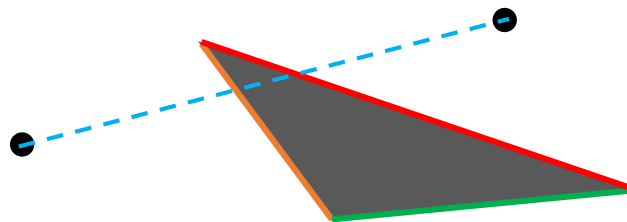


⇒ Equations: $\ell_1: a_1x + b_1y + c_1 = 0$, $\ell_2: a_2x + b_2y + c_2 = 0$

⇒ Two equations, two unknowns x, y , can solve for (x, y)

⇒ Check whether (x, y) belongs to both ℓ_1 and ℓ_2

Checking visibility is similar

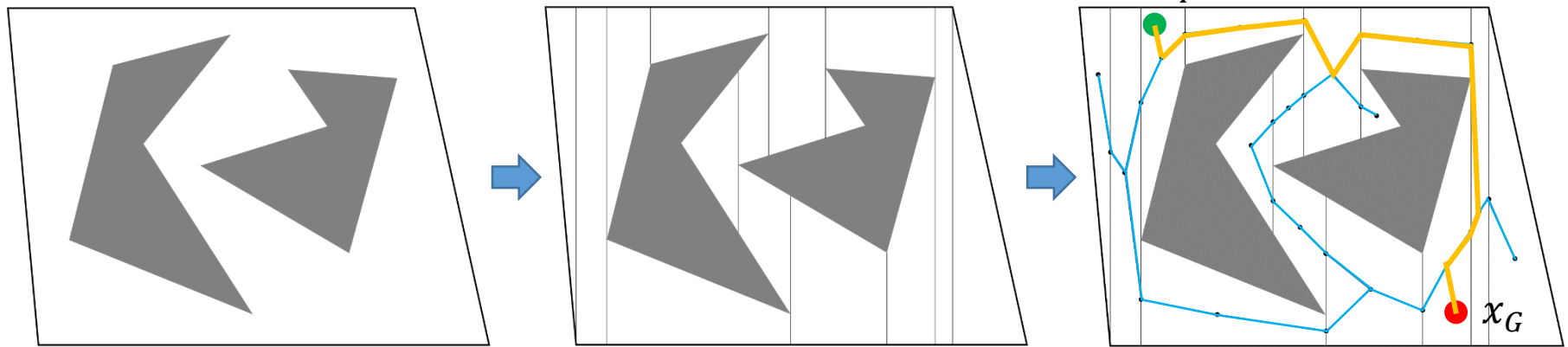


Cell Decomposition Methods

Cell decomposition breaks down C_{free} into simpler pieces that allow graph search to be carried out

Example: vertical cell decomposition

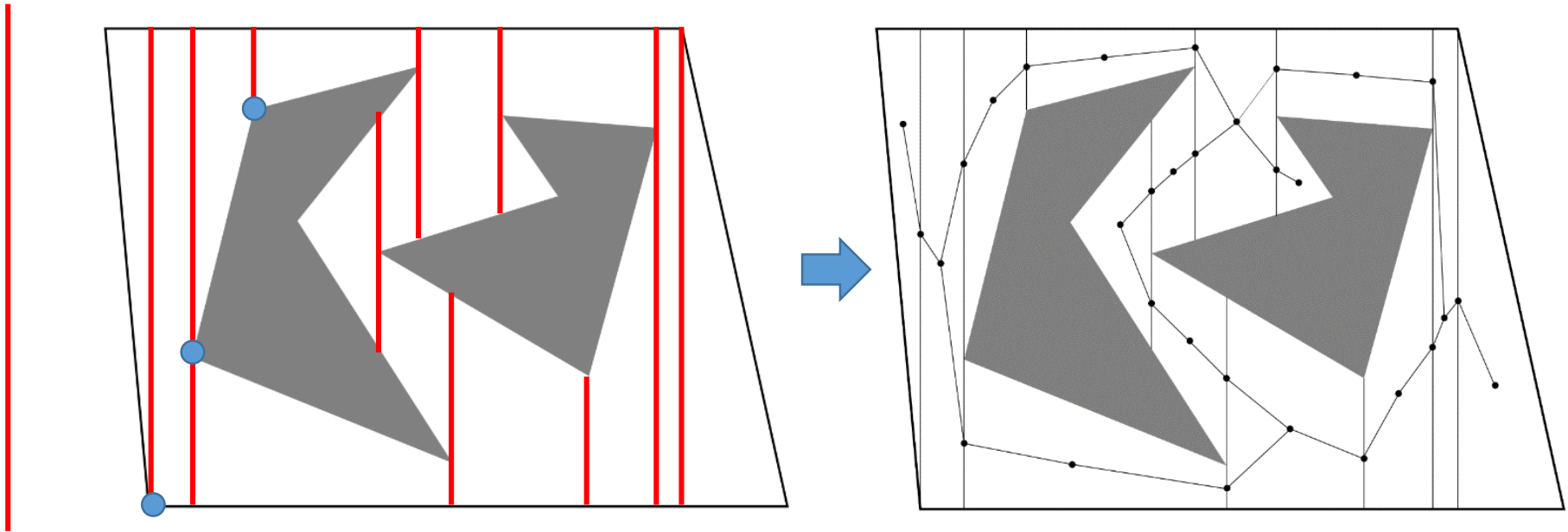
- ⇒ Break the free space into **convex** pieces (cells)
- ⇒ Build a roadmap (graph) connecting the cells
- ⇒ Plug in x_I and x_G and connect to cell centers
- ⇒ Search over the roadmap for a path



Many different types of cell decomposition methods

Vertical Cell Decomposition in More Detail

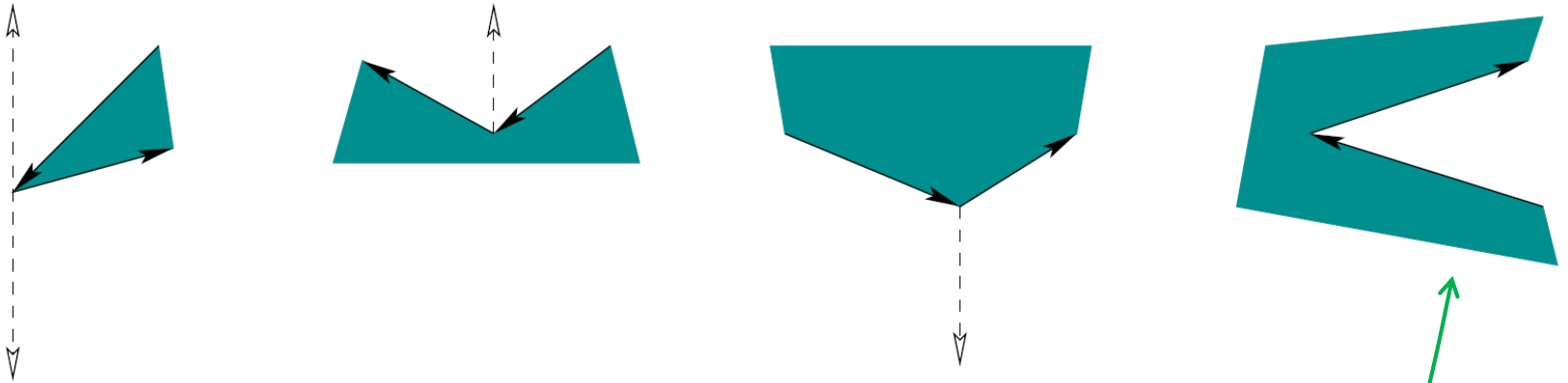
Vertical cell decomposition uses a “vertical sweep-line” for doing the decomposition



- ⇒ Move the sweep-line from left to right
- ⇒ For each vertex it touches, extend from the vertex along the line
- ⇒ Stop when the line hits other features of the environment
- ⇒ Compute the centers and boundary centers and connect them

Vertical Cell Decomposition in More Detail

There are four types of vertices



The second type is actually not needed

⇒ With it, we get trapezoids

⇒ Otherwise, we get convex polygons which is still OK

The last type does not generate any vertical lines

Vertical Cell Decomposition in More Detail

Algorithm details

⇒ A simple algorithm

⇒ Sort the x coordinates of input points - $O(n \log n)$

⇒ Order the points from left to right

⇒ For each point, check whether the vertical scan line intersects each line segment of the environment, pick the closest intersection point(s) - $O(n)$

⇒ Overall computation time is $O(n^2)$

⇒ Smarter algorithm

⇒ If we take more care to track the line segments in the environment we can cut down the computation time to $O(n \log n)$

The algorithm comes from **plane-sweep** or **line-sweep** algorithms from computational geometry

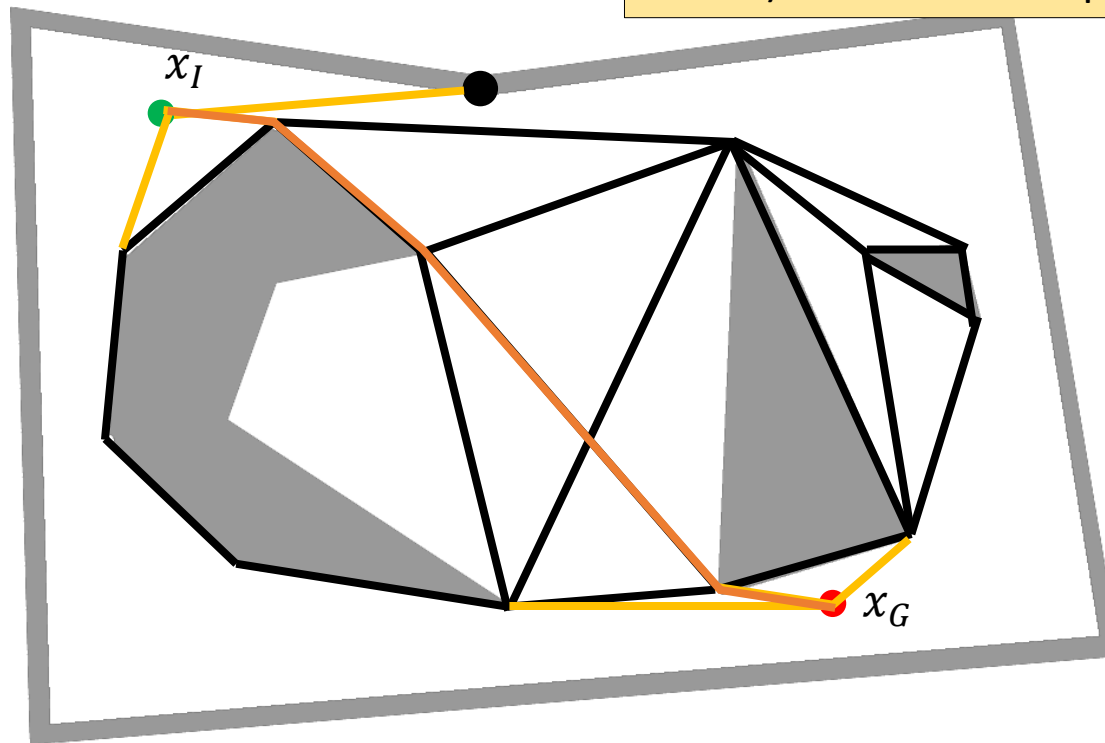
⇒ Can be found in many computational geometry textbooks

Shortest-Path Roadmaps

Vertical cell decomposition is simple but is suboptimal – it does not provide shortest paths connecting x_I and x_G

Shortest-path roadmaps solve the problem (also known as the **reduced visibility graph**)

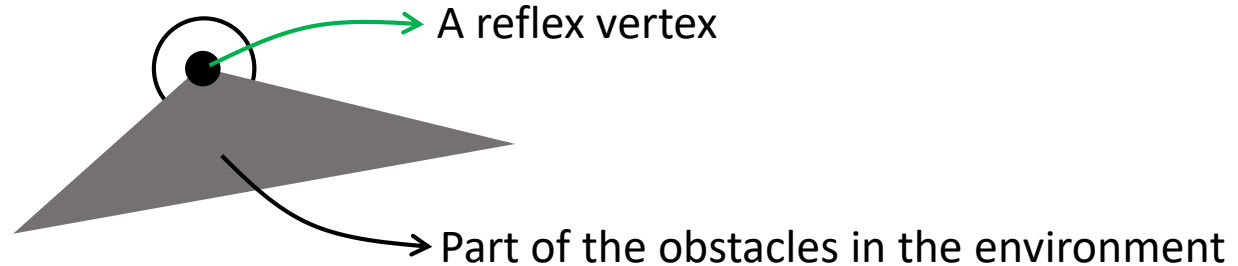
The black structure (including the isolated vertex) is the shortest-path roadmap



Building Shortest-Path Roadmap

Reflex vertices

⇒ A **reflex vertex** is one where the angle through the vertex in the environment is greater than π



Building the roadmap

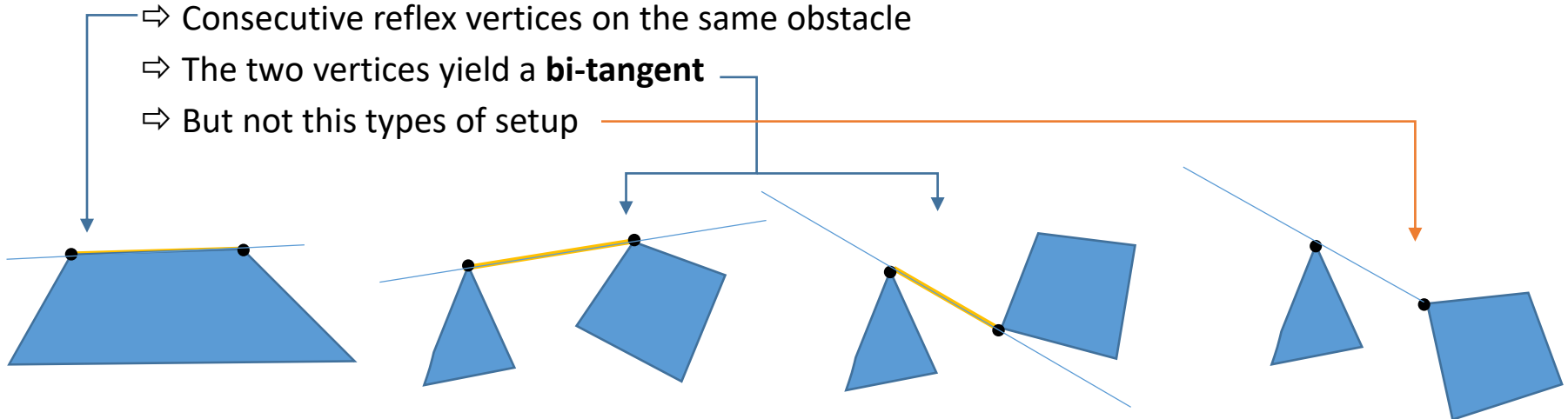
⇒ Add all reflex vertices to the roadmap

⇒ For two reflex vertices that are visible to each other, add an edge if

⇒ Consecutive reflex vertices on the same obstacle

⇒ The two vertices yield a **bi-tangent**

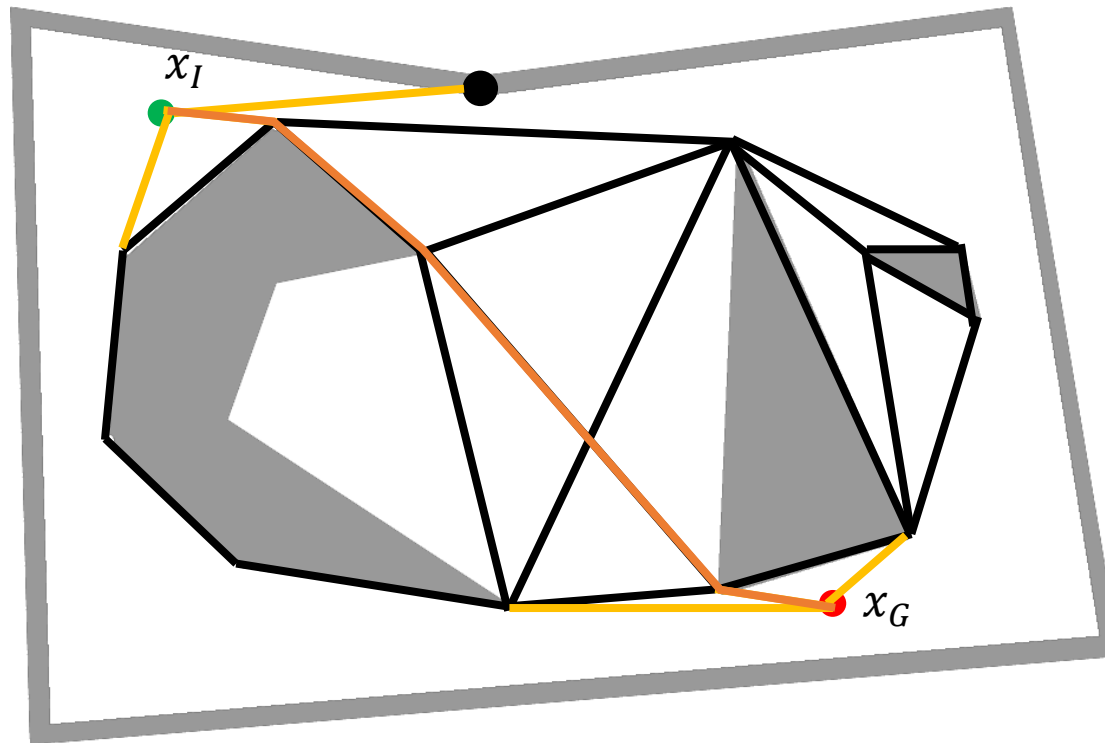
⇒ But not this types of setup



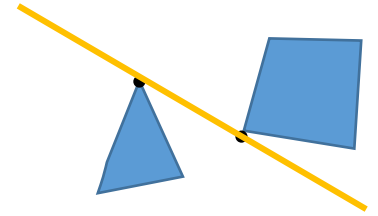
Search on the Shortest-Path Roadmaps

To search for a path from x_I and x_G on the roadmap

- ⇒ Add x_I and x_G to the roadmap
- ⇒ Connect x_I to visible roadmap vertices; same for x_G
- ⇒ Searching for a shortest path through the connected roadmap containing x_I and x_G



Running Time for Roadmap Building



Naïve algorithm

- ⇒ There are $O(n)$ reflex vertices so n^2 pairs to check for bi-tangents
- ⇒ For each pair, check whether the pair is visible - $O(n)$
- ⇒ This yields an $O(n^3)$ algorithm

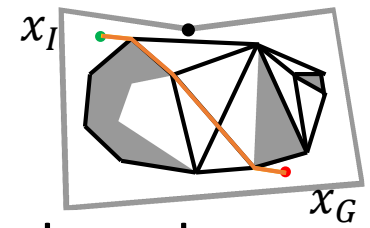
Radial sweep algorithm

- ⇒ For a reflex vertex, compute all bi-tangents by sweeping from 0 to 2π radially
- ⇒ This allows the computation of all bi-tangents from the vertex in $O(n \log n)$ time
- ⇒ Total complexity is $O(n^2 \log n)$

Even faster algorithms

- ⇒ Output sensitive algorithm $O(n \log n + m)$ where m is the number of edges in the output roadmap

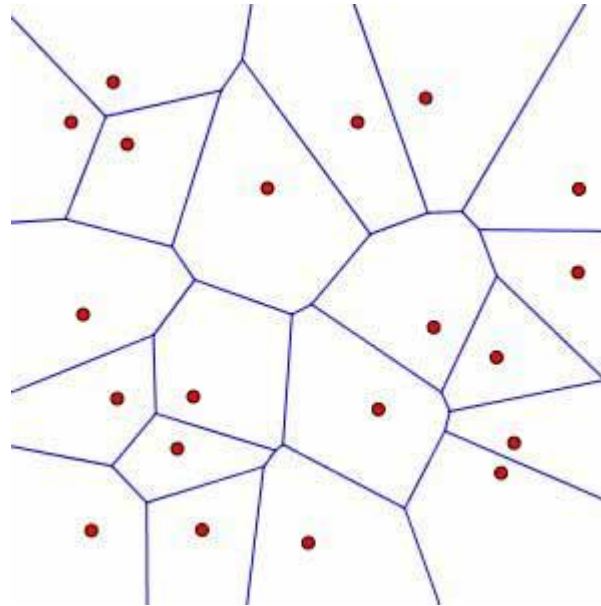
Maximum Clearance Roadmap



The shortest-path roadmap requires robots to go along obstacle boundaries, which can be unsafe

Maximum clearance roadmap does the opposite

⇒ Based on the idea of Voronoi diagrams



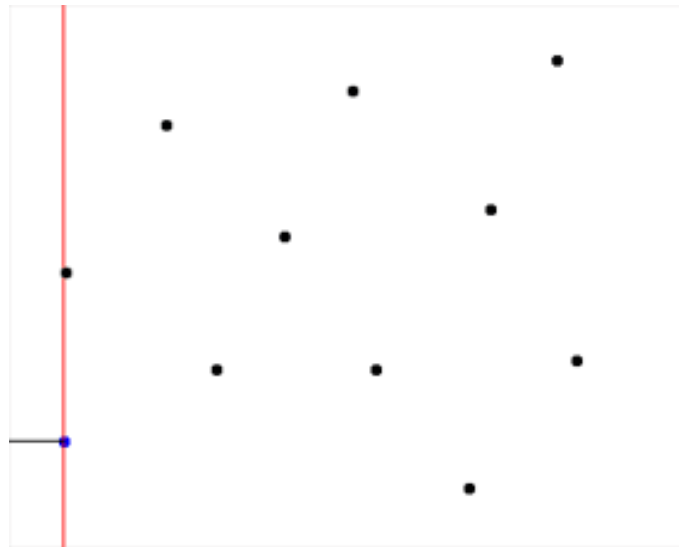
⇒ In a Voronoi diagram, the lines are as far away from all points as possible

Computing the Voronoi Diagram

Voronoi diagram can be computed in the plane in $O(n \log n)$ time using Fortune's algorithm

⇒ Also a sweep-line algorithm!

⇒ Maintains a “beach front”

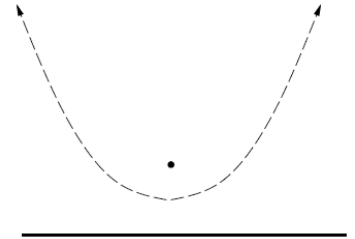
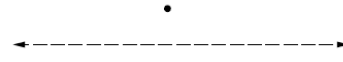
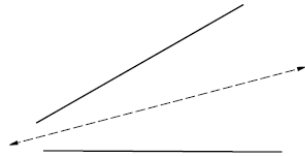


For higher dimensions, Bowyer-Watson computes a Delaunay triangulation which can be turned into a Voronoi diagram

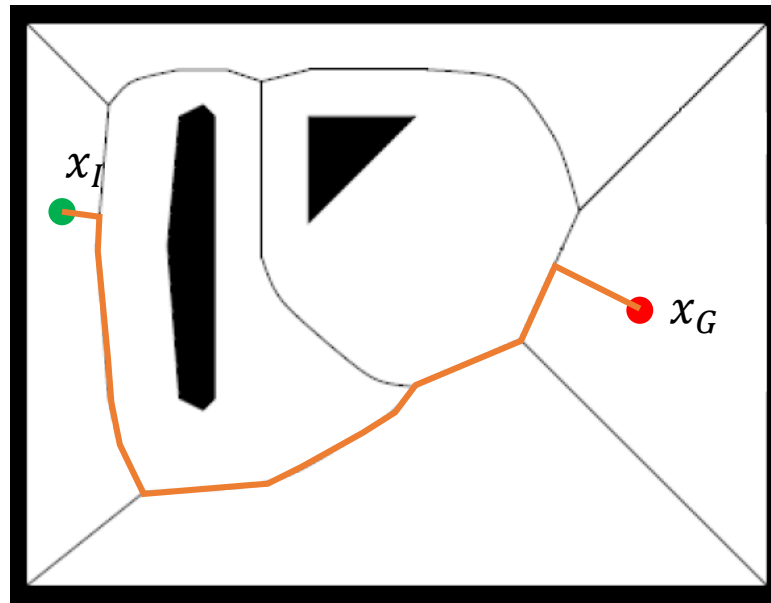
Building Maximum Clearance Roadmap

Instead of vertex-vertex interactions, now with edge interactions

- ⇒ Edge-edge
- ⇒ Vertex-vertex
- ⇒ Vertex-edge



Example



Running Time for Roadmap Building

Naïve algorithm

- ⇒ For each pair of features, compute the line as shown above - n^2 of these
- ⇒ For each pair of these lines, compute their intersections – $O(n^4)$

Using a better algorithm, can get the running time to $O(n \log n)$

- ⇒ This can be done by modifying the sweep line algorithm