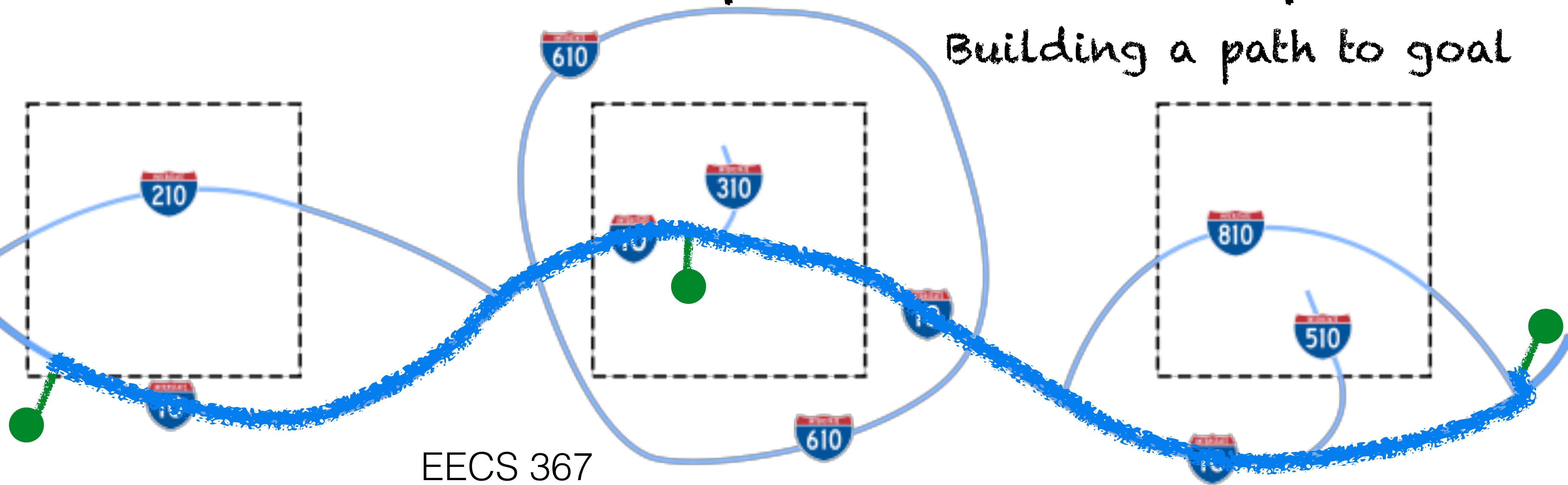


Planning with Sampling Roadmaps

Building a path to goal



EECS 367

Intro. to Autonomous Robotics

ROB 320
Robot Operating Systems

Winter 2022

autorob.org

Approaches to motion planning

- Bug algorithms: Bug[0-2], Tangent Bug
- Graph Search (fixed graph)
 - Depth-first, Breadth-first, Dijkstra, A-star, Greedy best-first
- **Sampling-based Search (build graph):**
 - **Probabilistic Road Maps, Rapidly-exploring Random Trees**
 - Optimization and local search:
 - Gradient descent, Potential fields, Simulated annealing, Wavefront

Where is this?



What is the configuration of this location?



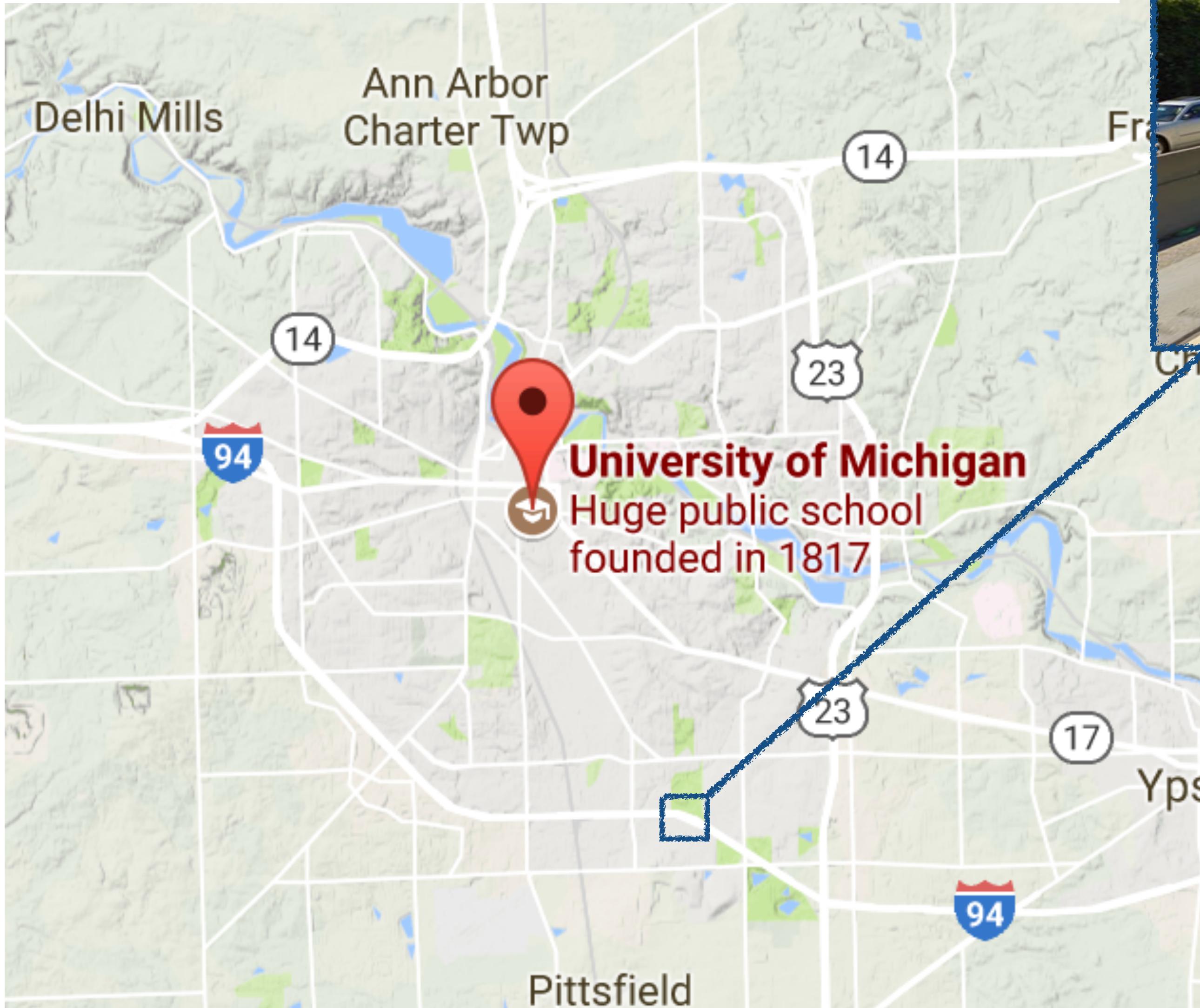
What is the configuration of this location?



Configuration (lat,lon): [42.237631, -83.7147289]

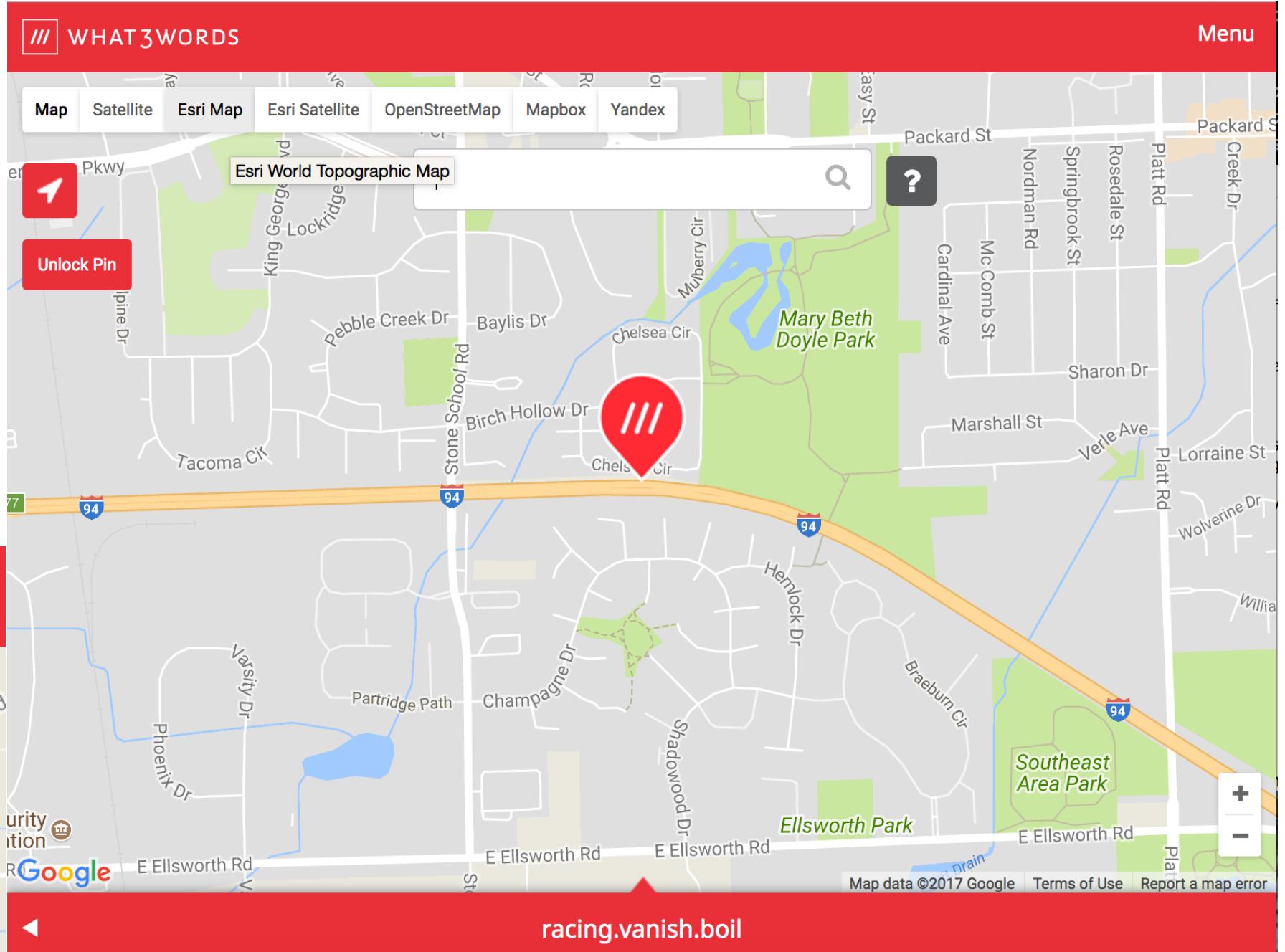
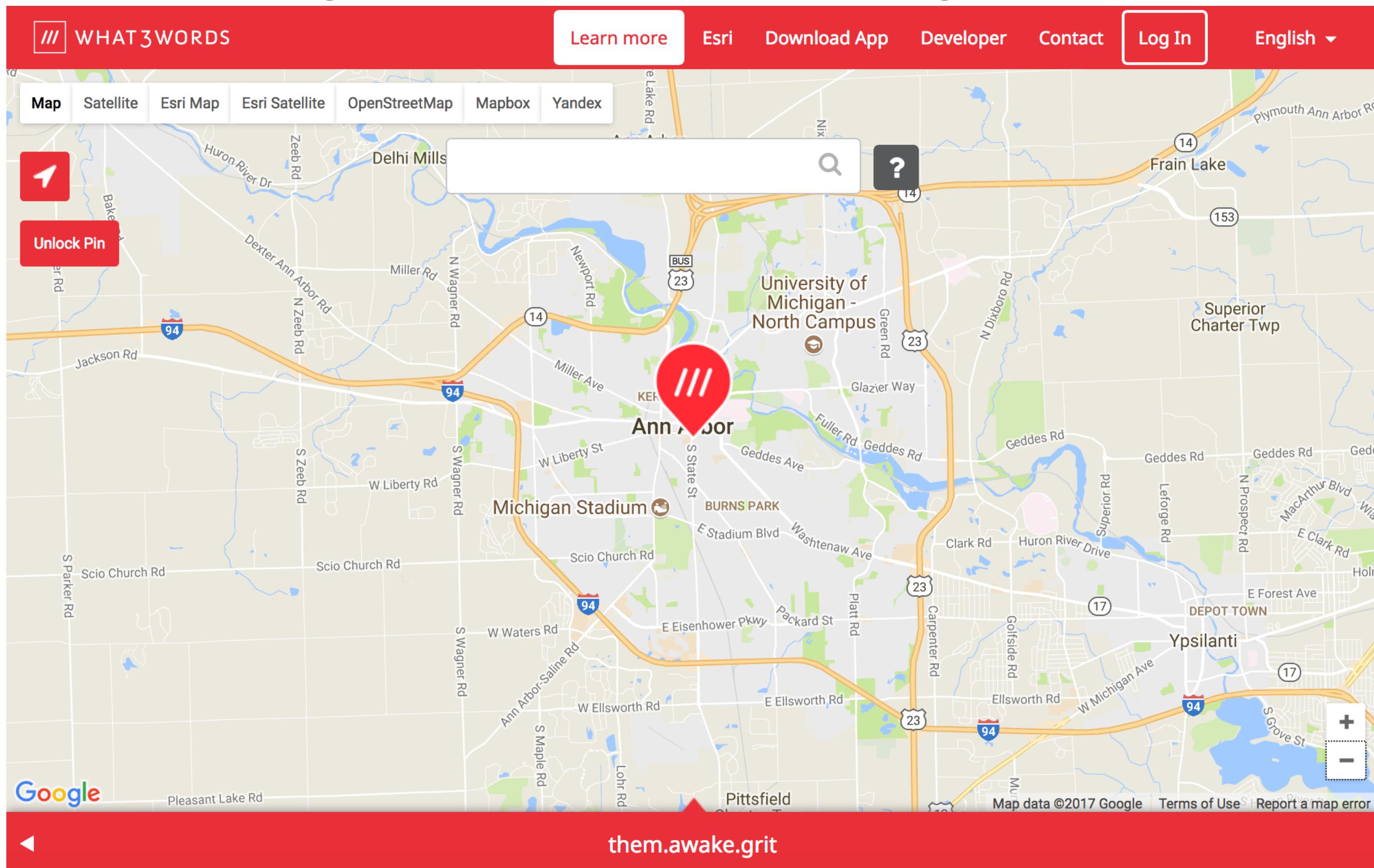
Michigan Robotics 367/320 - autorob.org

Configuration: [42.2780436,-83.7404128]



Alternatively...

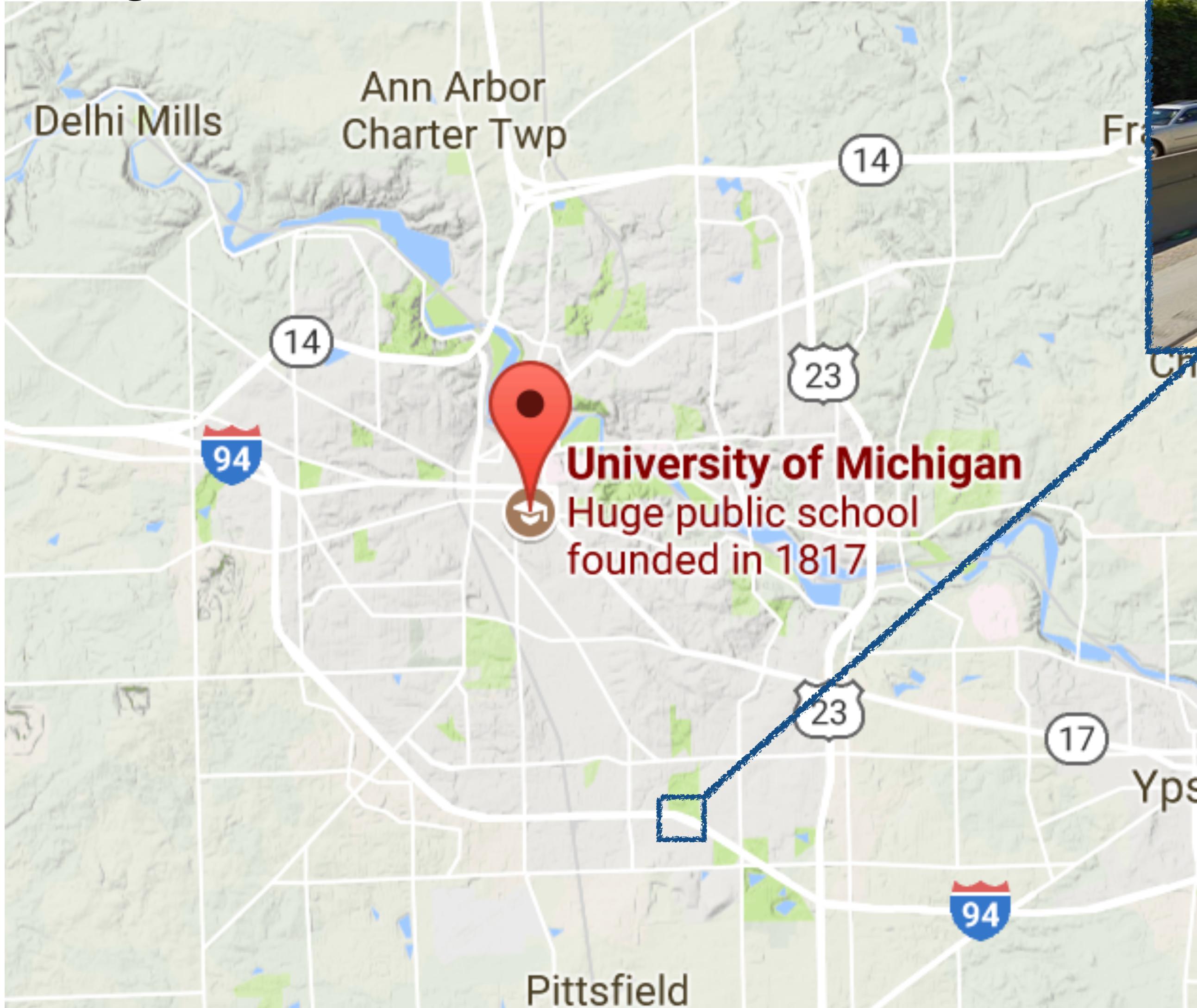
Configuration: [them.awake.grit]

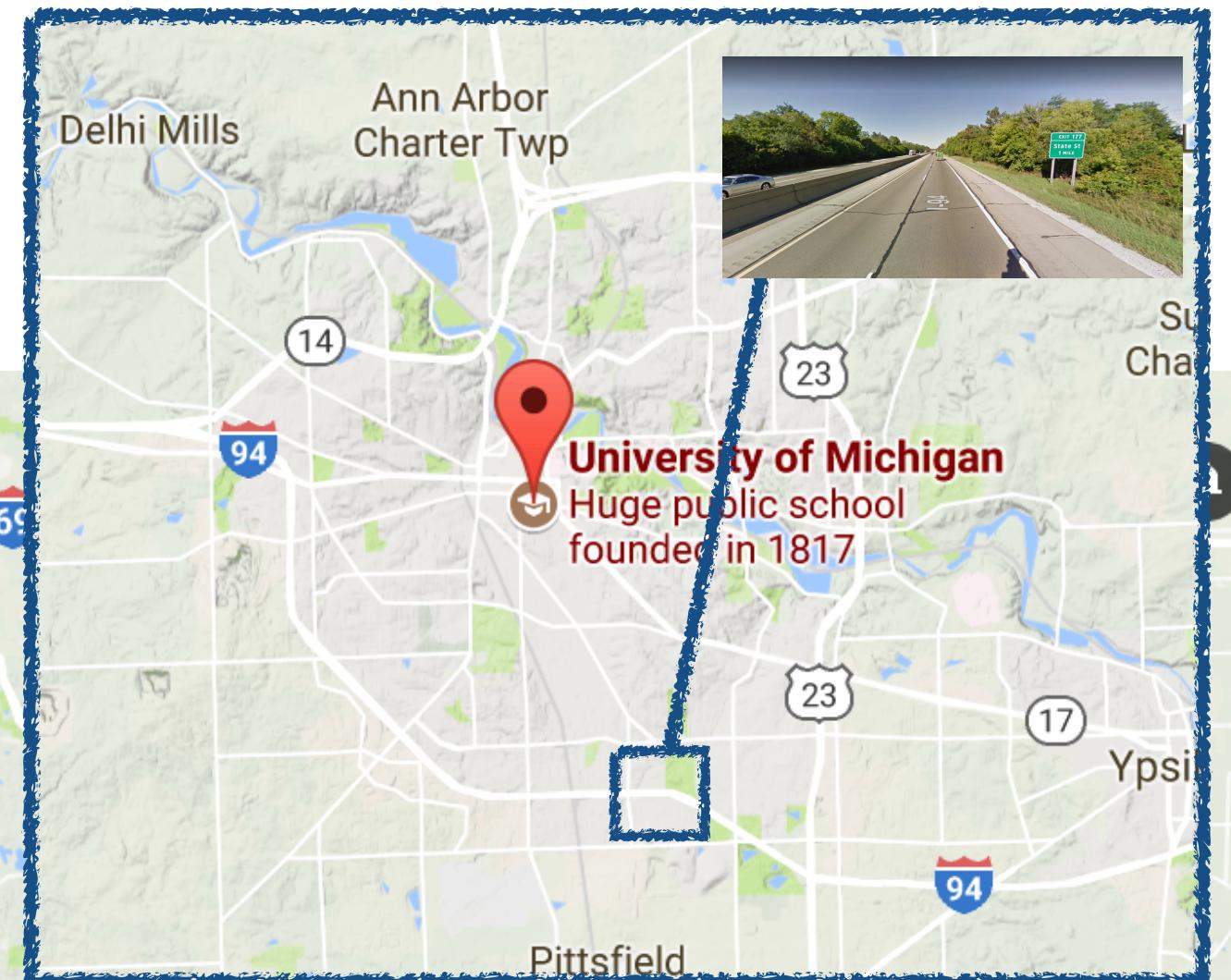
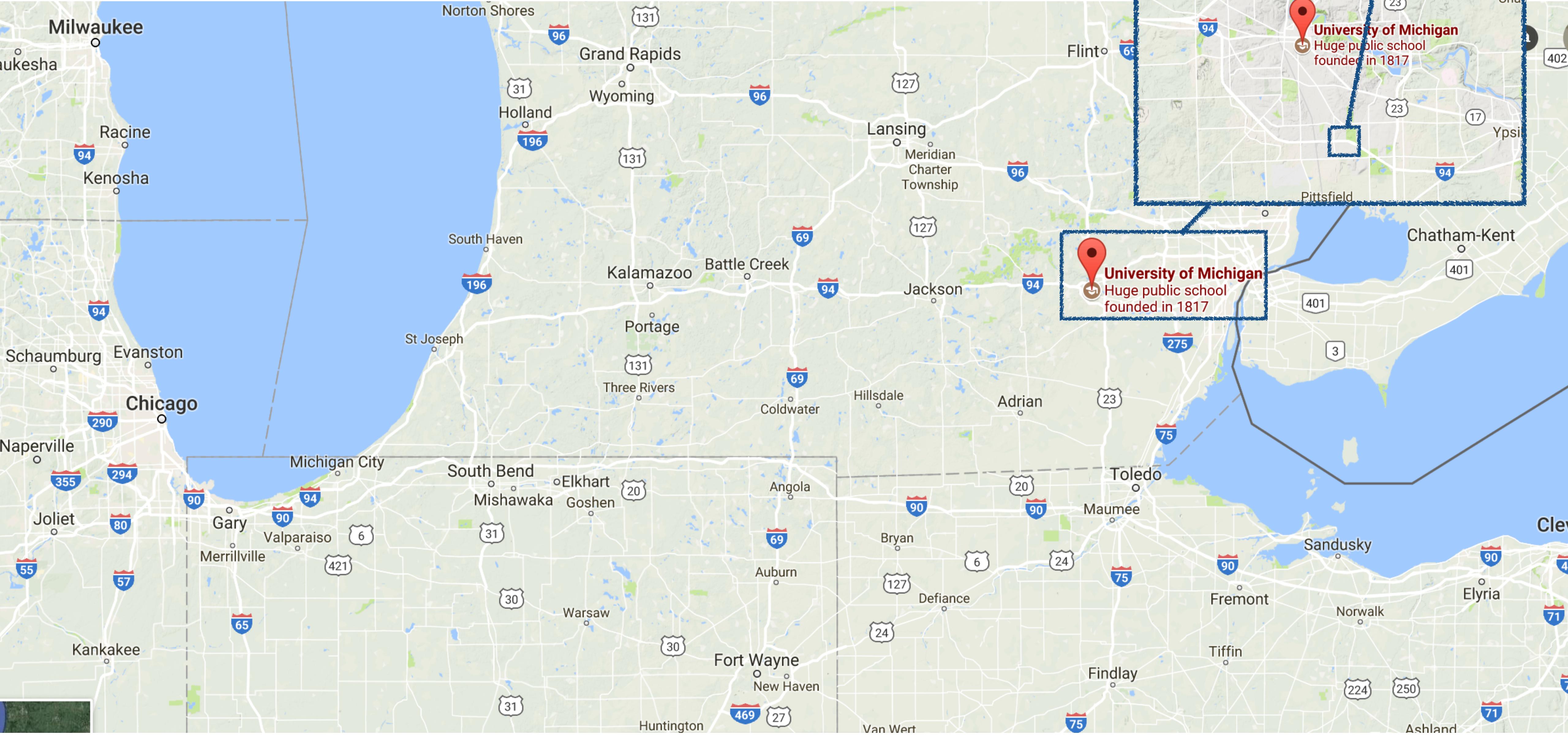


Configuration: [racing.vanish.boil]

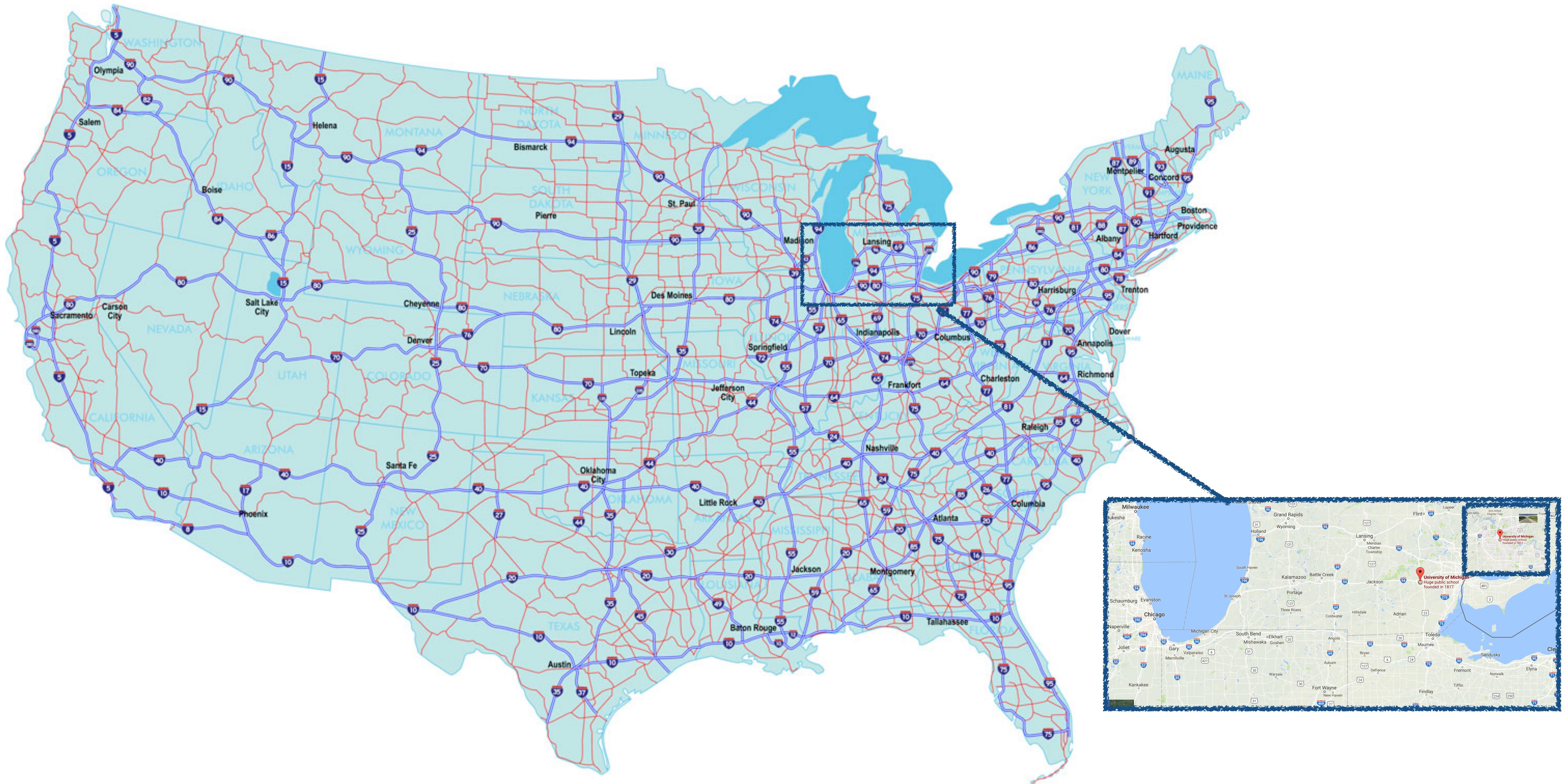
Is this a usable
representation for
configuration?

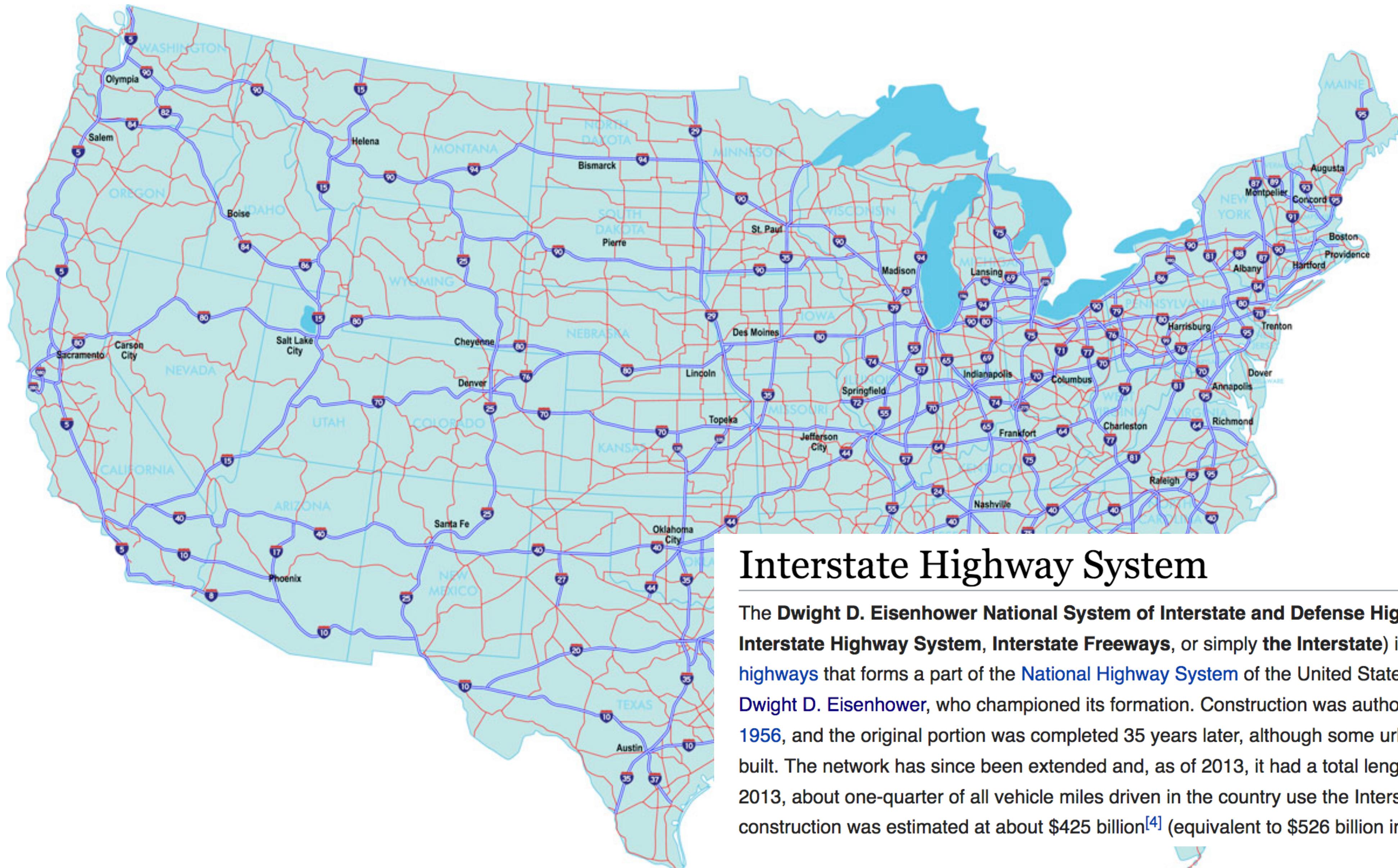
Configuration: [42.2780436,-83.7404128]





University of Michigan
Huge public school
founded in 1817



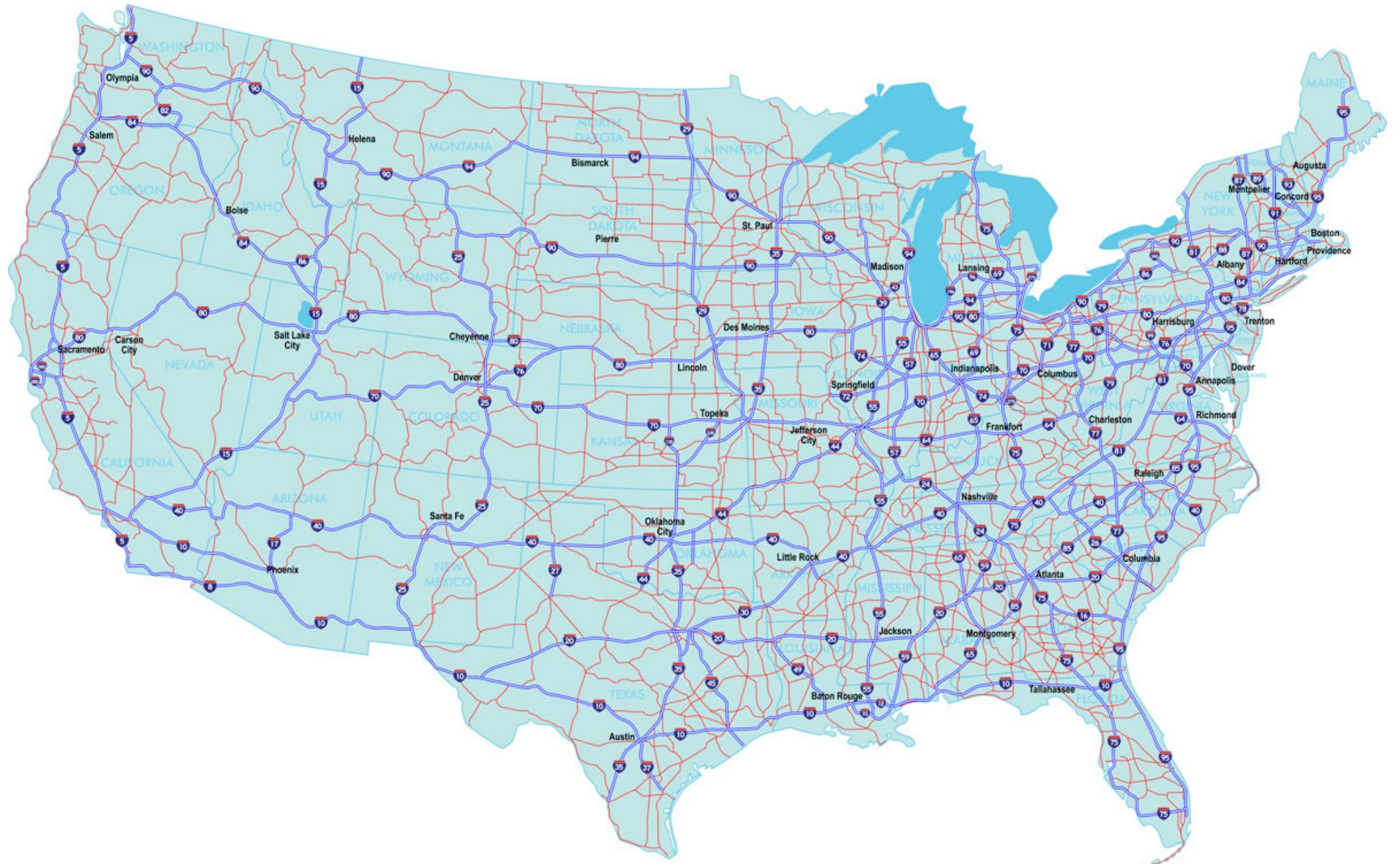


Interstate Highway System

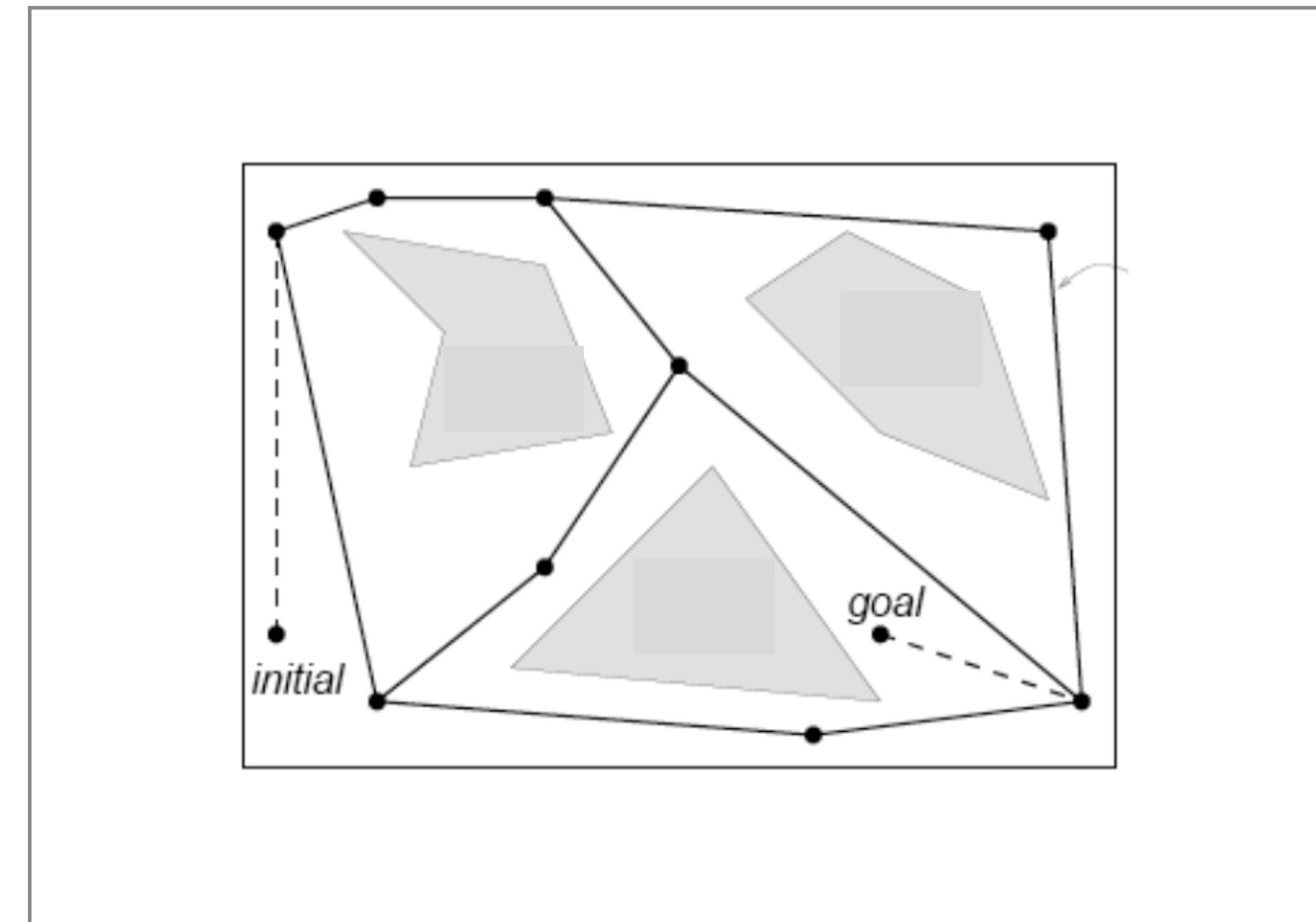
The **Dwight D. Eisenhower National System of Interstate and Defense Highways** (commonly known as the **Interstate Highway System**, **Interstate Freeways**, or simply **the Interstate**) is a network of [controlled-access highways](#) that forms a part of the [National Highway System](#) of the United States. The system is named for President Dwight D. Eisenhower, who championed its formation. Construction was authorized by the [Federal Aid Highway Act of 1956](#), and the original portion was completed 35 years later, although some urban routes were cancelled and never built. The network has since been extended and, as of 2013, it had a total length of 47,856 miles (77,017 km).^[2] As of 2013, about one-quarter of all vehicle miles driven in the country use the Interstate system.^[3] In 2006, the cost of construction was estimated at about \$425 billion^[4] (equivalent to \$526 billion in 2016^[5]).



Roadmaps



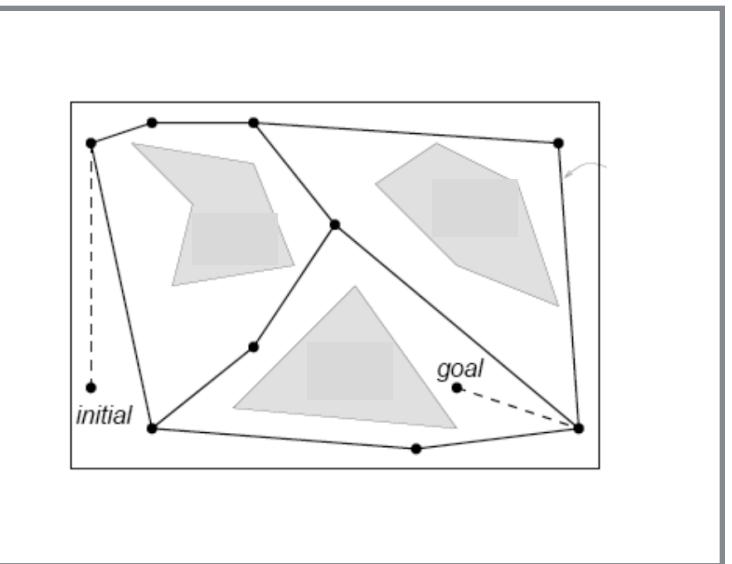
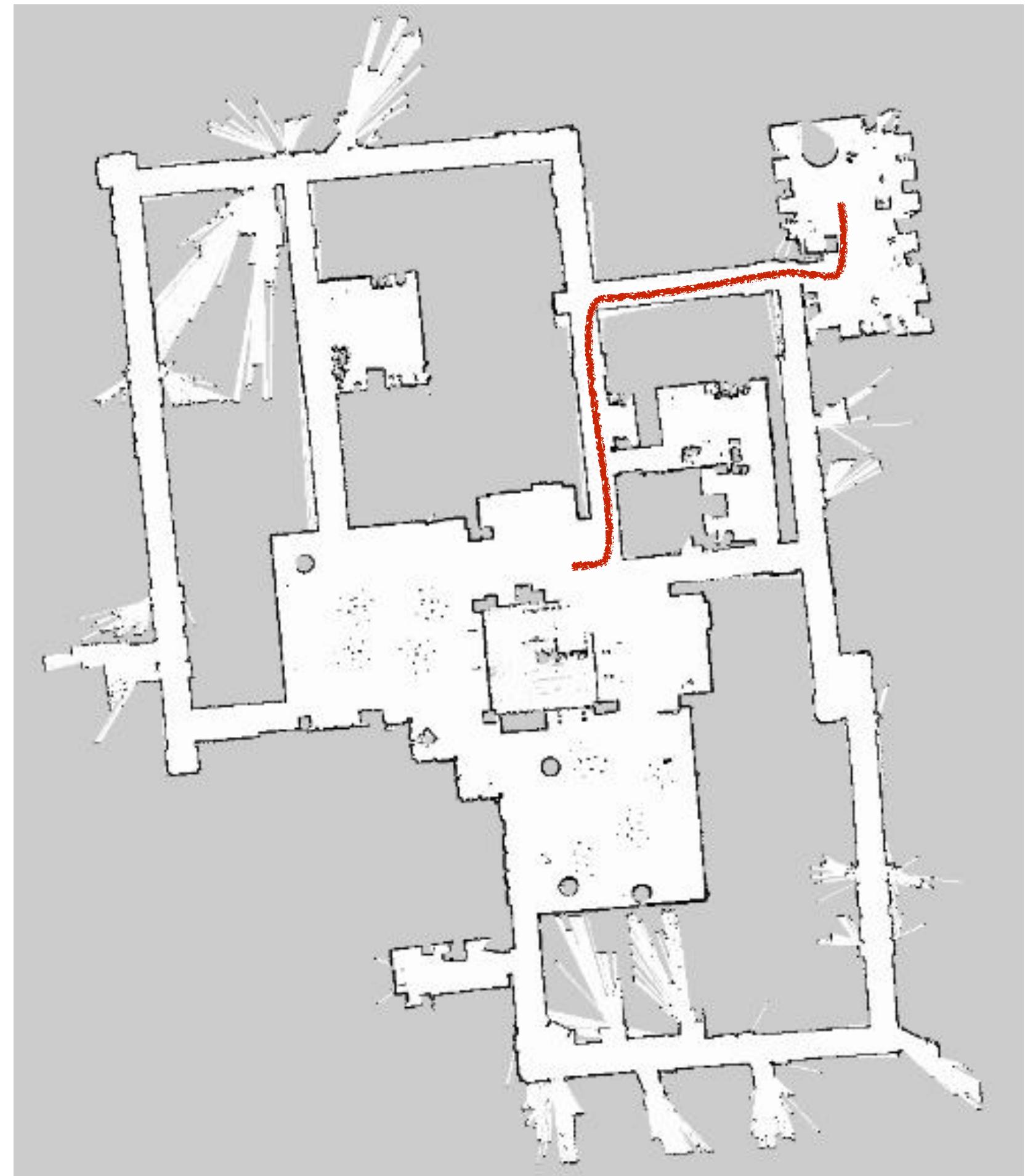
Roadmap over geolocations



Roadmap over robot configurations

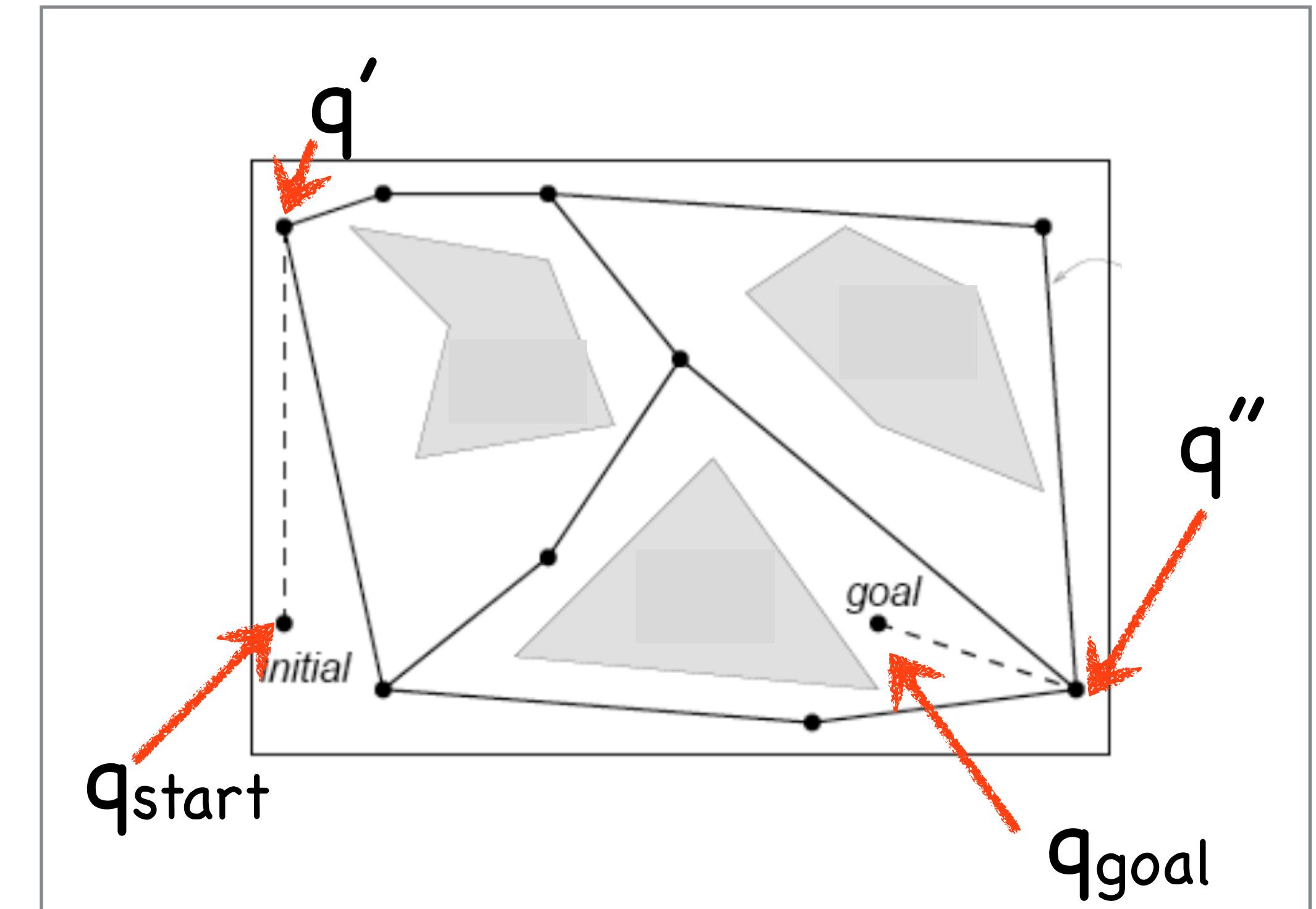
Roadmaps

- Graph search assumed C-space as a fixed uniform grid
 - finite set of discretized cells
- How does this scale beyond planar navigation?
 - curse of dimensionality
- Roadmaps are a more general notion of graphs in C-space



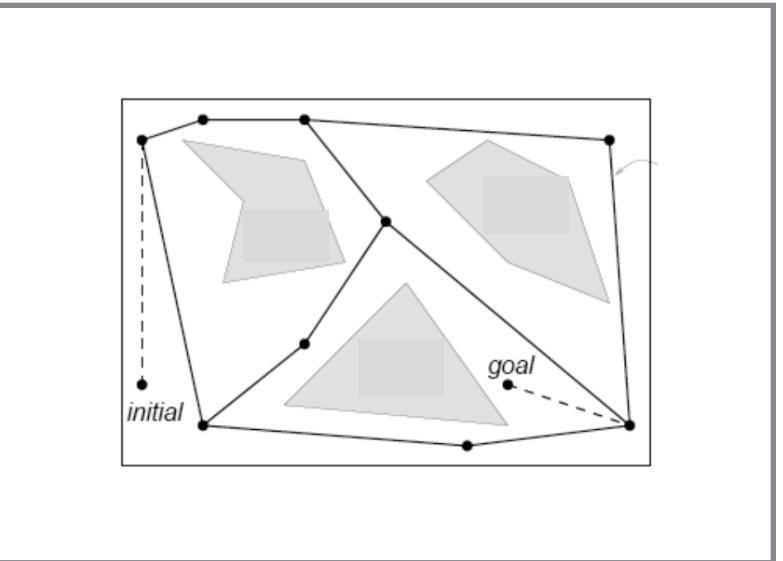
Roadmap Definition

- A roadmap RM is a union of curves s.t. all start and goal points in C-space (Q_{free}) can be connected by a path
- Roadmap properties:



- **Accessibility:** There is a path from $q_{start} \in Q_{free}$ to some $q' \in RM$
- **Departability:** There is a path from $q'' \in RM$ to $q_{goal} \in Q_{free}$
- **Connectivity:** there exists a path in RM between q' and q''

Basic Roadmap Planner



- 1) Build the roadmap RM as graph $G(V,E)$
 - V : nodes are “valid” in C-space in Q_{free}
 - a configuration q is valid if it is not in collision and within joint limits
 - E : an edge $e(q_1, q_2)$ connects two nodes if a free path connects q_1 and q_2
 - all configurations along edge assumed to be valid
- 2) Connect start and goal configurations to RM at q' and q'' , respectively
- 3) Find path in RM between q' and q''

How to build a roadmap?

How to build a roadmap?

2 Approaches

2 Approaches to Roadmaps

Deterministic:

complete algorithms

- Visibility Graph
 - trace lines connecting obstacle polygon vertices
- Voronoi Planning
 - trace edges equidistant from obstacles

Probabilistic:

C-space sampling

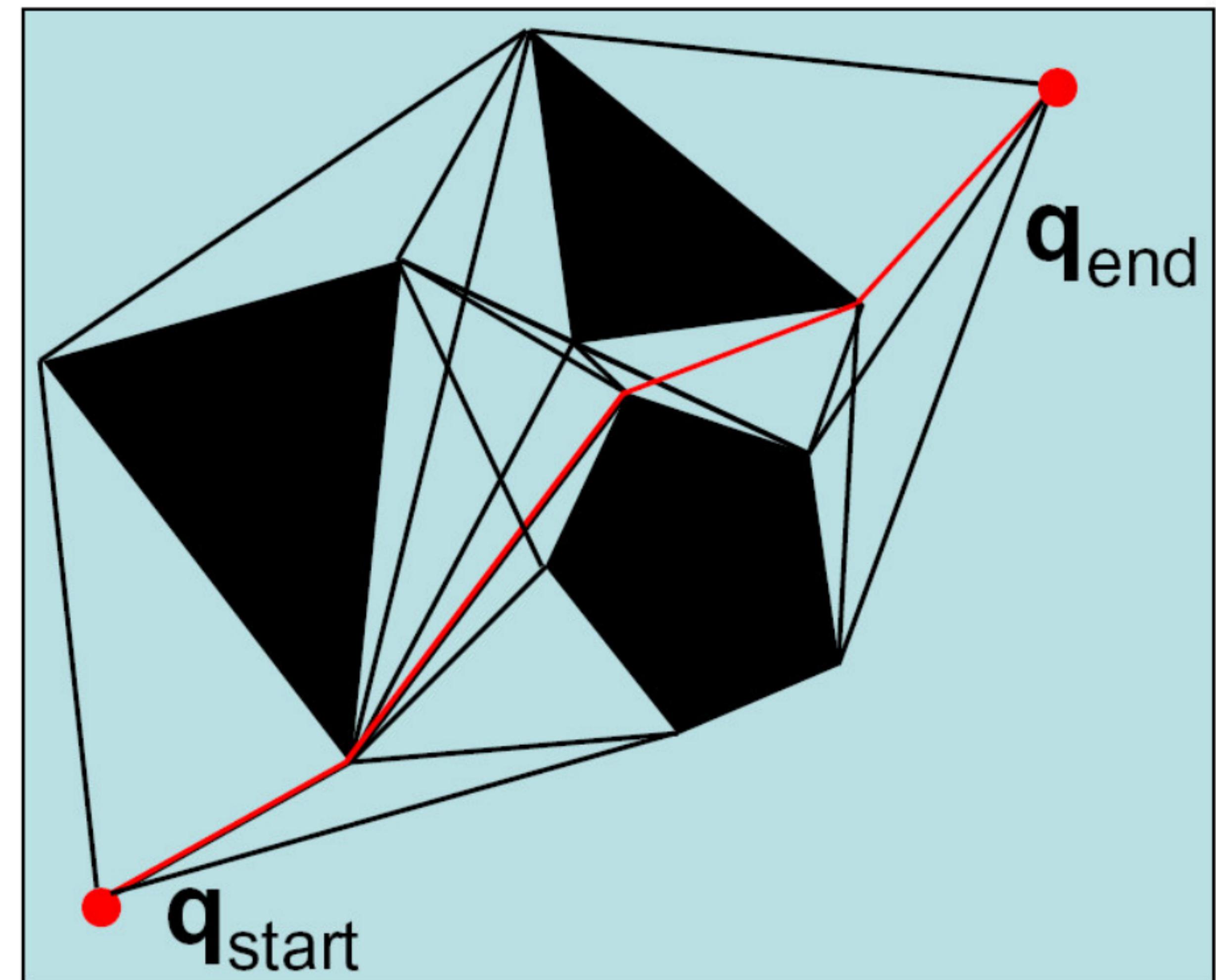
- Probabilistic Roadmap (PRM)
 - sample and connect vertices in graph for multiple planning queries
- Rapidly-exploring Random Tree (RRT)
 - sample and connect vertices in trees rooted at start and goal configuration

2 Approaches to Roadmaps

Deterministic:

complete algorithms

- Visibility Graph
 - trace lines connecting obstacle polygon vertices
- Voronoi Planning
 - trace edges equidistant from obstacles

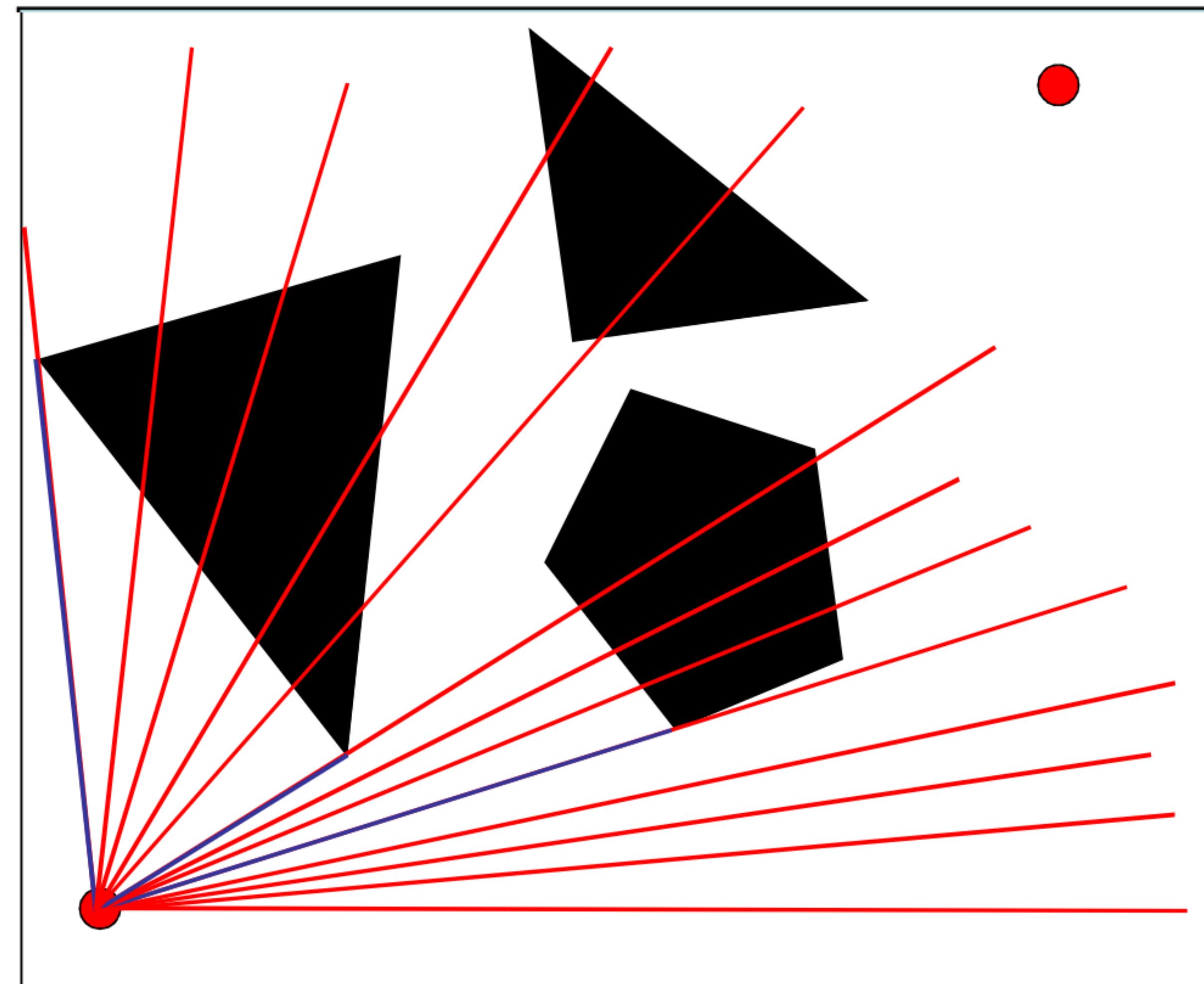


2 Approaches to Roadmaps

Deterministic:

complete algorithms

- Visibility Graph
 - trace lines connecting obstacle polygon vertices
- Voronoi Planning
 - trace edges equidistant from obstacles

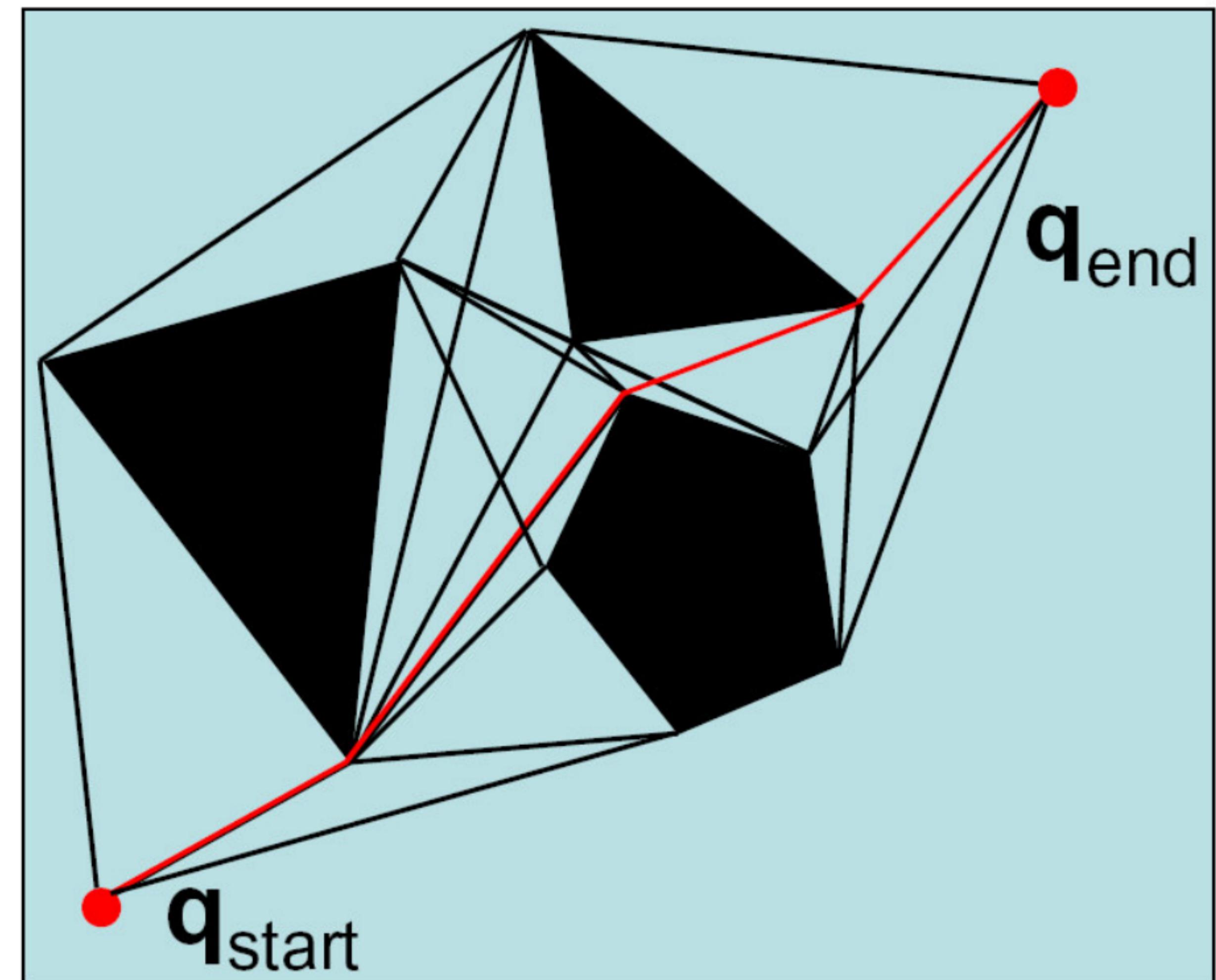


2 Approaches to Roadmaps

Deterministic:

complete algorithms

- Visibility Graph
 - trace lines connecting obstacle polygon vertices
- Voronoi Planning
 - trace edges equidistant from obstacles

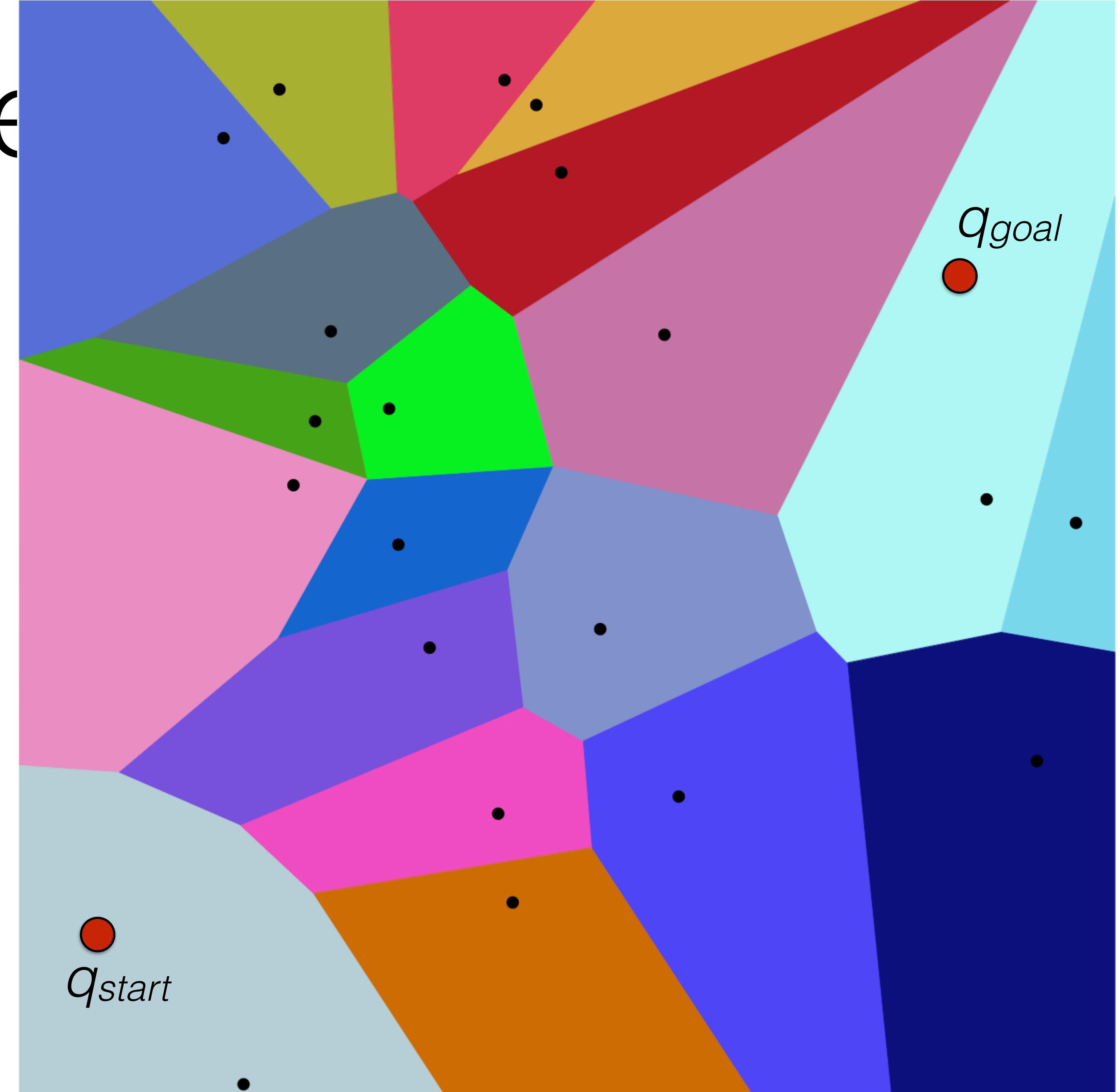


2 Approaches

Deterministic:

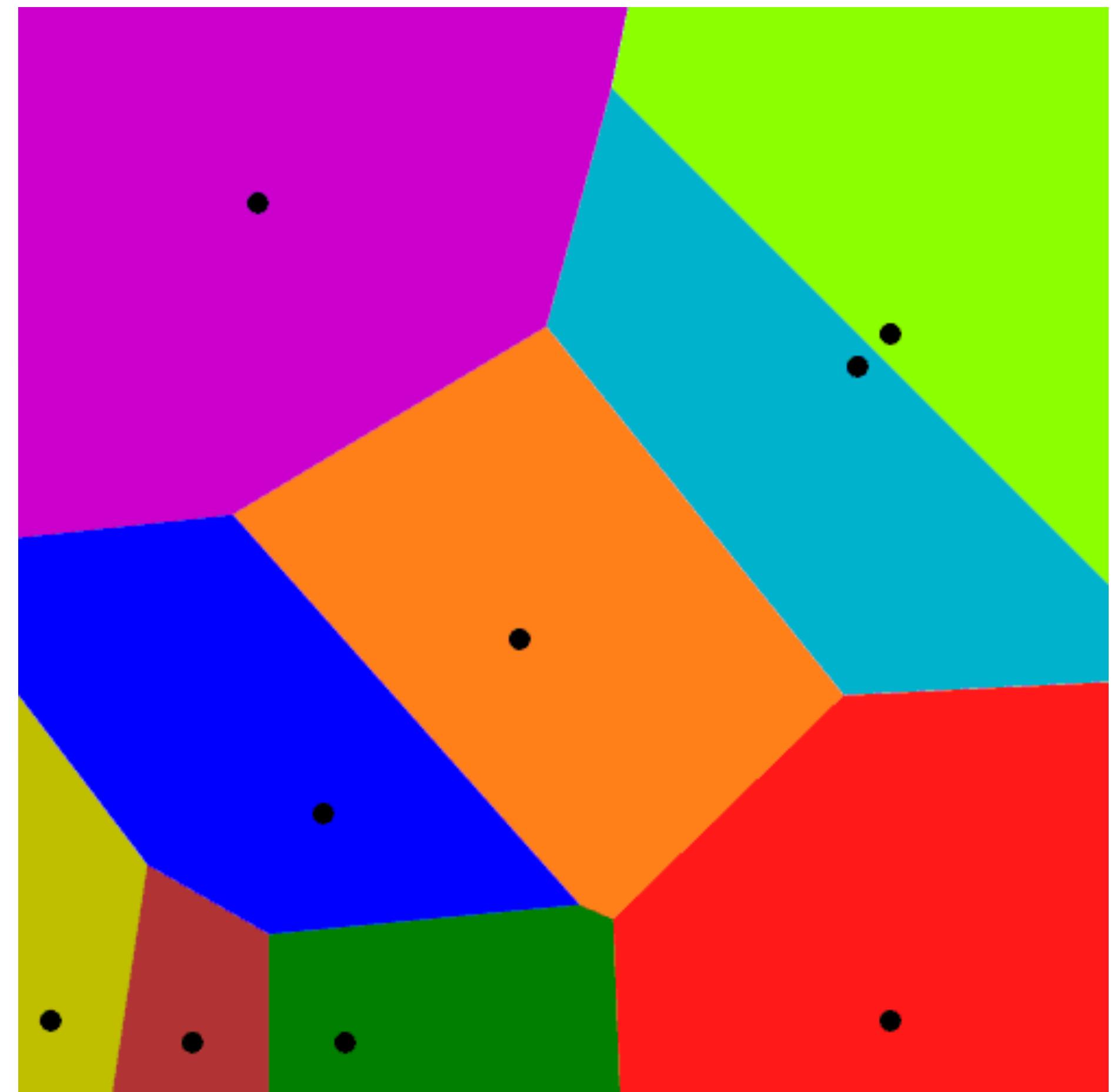
complete algorithms

- Visibility Graph
 - trace lines connecting obstacle polygon vertices
- Voronoi Planning
 - trace edges equidistant from obstacles



Voronoi Diagram

- Given N input points in a d dimensional space
- Find region boundaries such that each point on a boundary are equidistant to two or more input points
- Delaunay triangulation is a dual to the Voronoi diagram



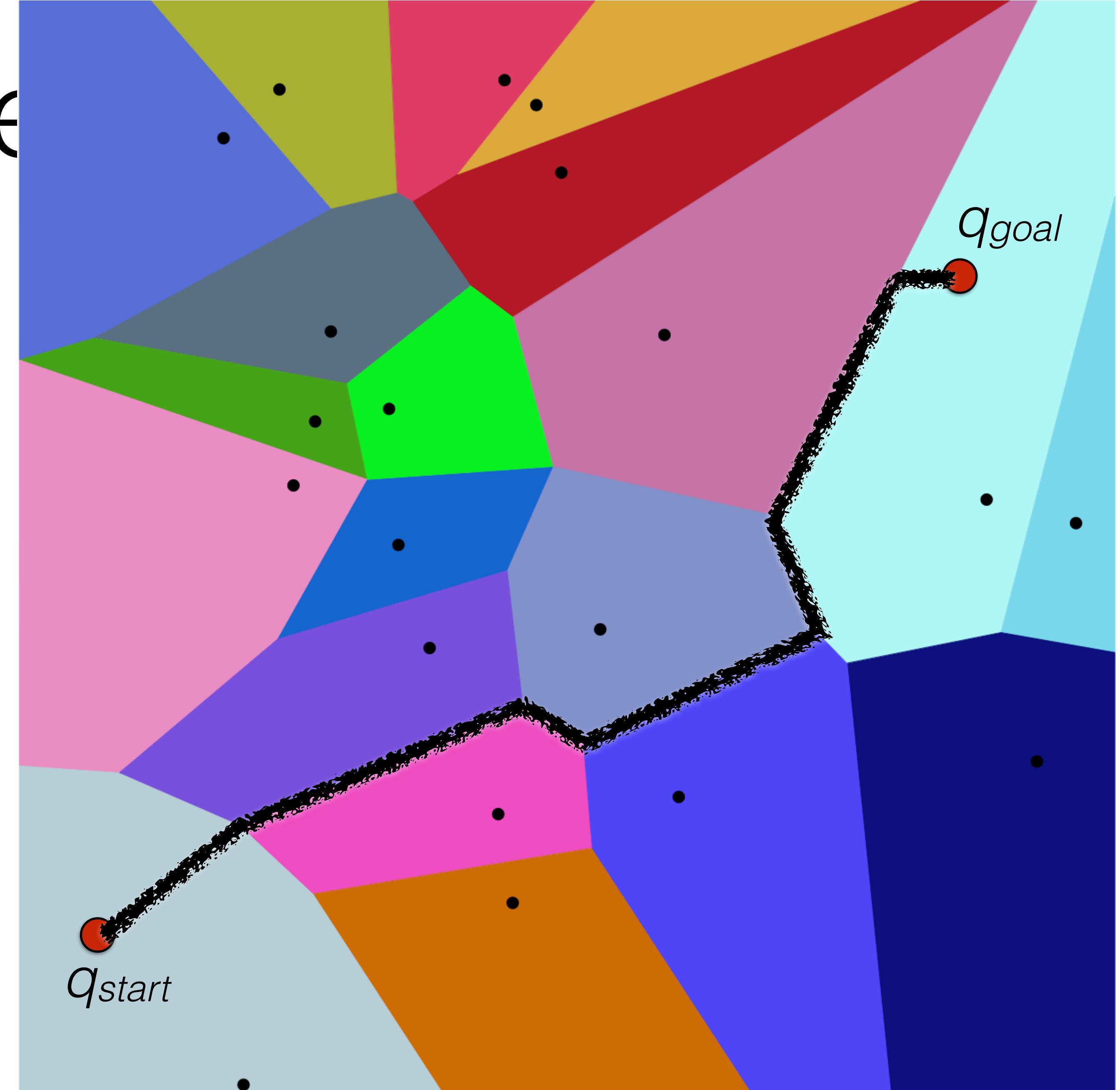
https://en.wikipedia.org/wiki/Voronoi_diagram#/media/File:Voronoi_growth_euclidean.gif

2 Approaches

Deterministic:

complete algorithms

- Visibility Graph
 - trace lines connecting obstacle polygon vertices
- Voronoi Planning
 - trace edges equidistant from obstacles



2 Approaches to Roadmaps

Deterministic:

complete algorithms

- Visibility Graph
 - trace lines connecting obstacle polygon vertices
- Voronoi Planning
 - trace edges equidistant from obstacles

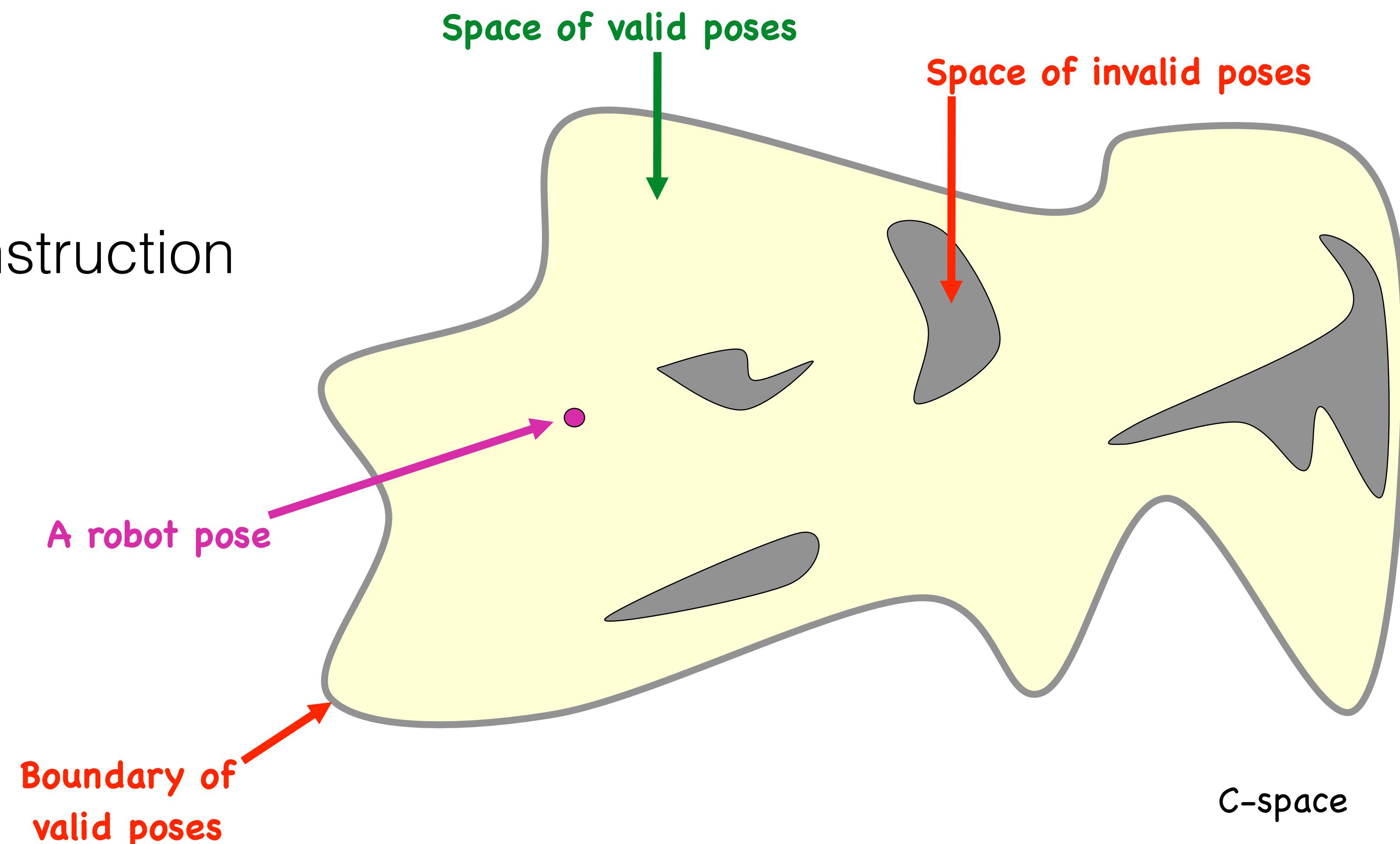
Probabilistic:

C-space sampling

- Probabilistic Roadmap (PRM)
 - sample and connect vertices in graph for multiple planning queries
- Rapidly-exploring Random Tree (RRT)
 - sample and connect vertices in trees rooted at start and goal configuration

Probabilistic road maps

- Two phases
 - Roadmap construction
 - Path Query

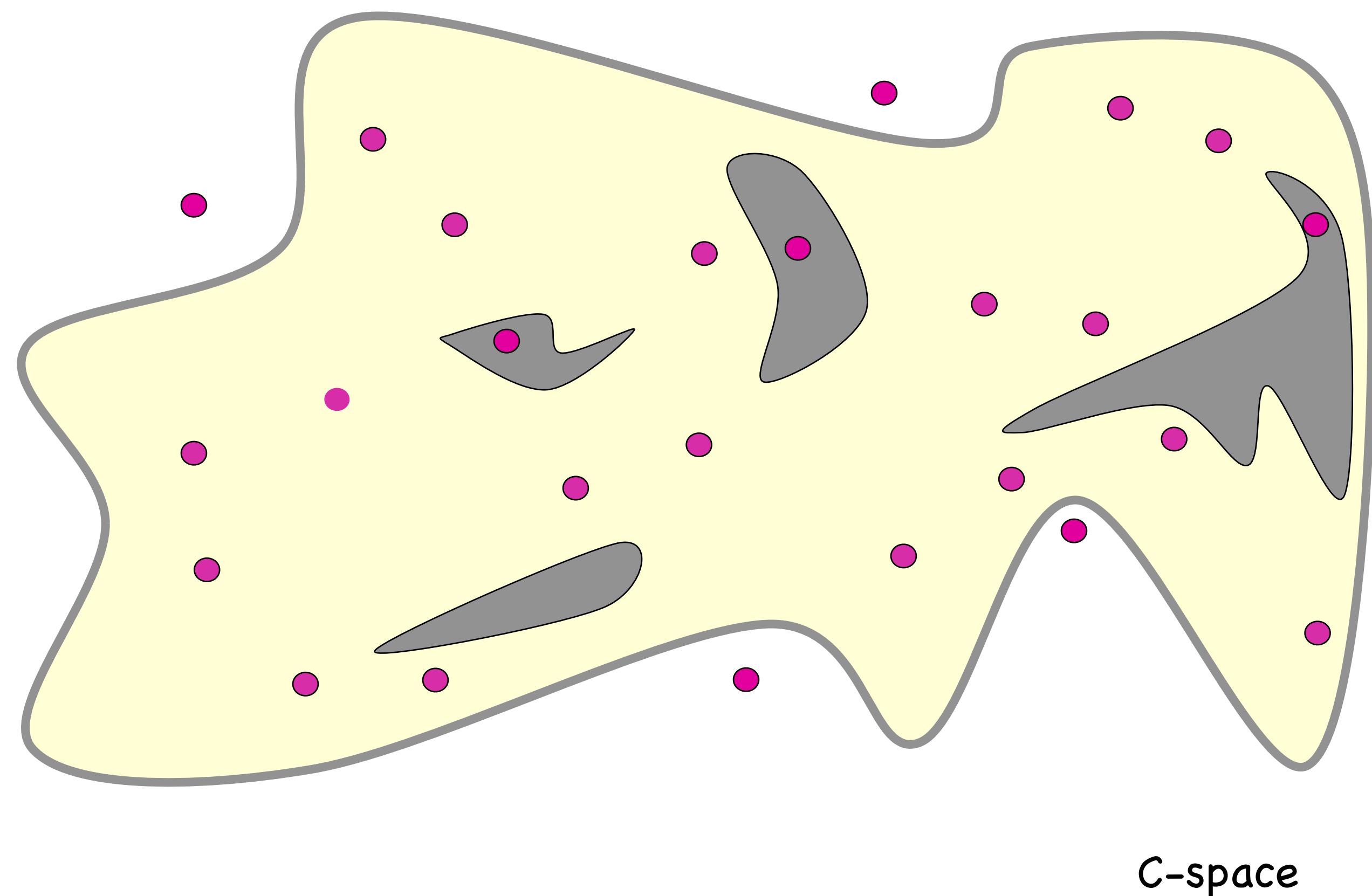


PRM: construction phase

1) Select N sample poses at random

2) Eliminate invalid poses

3) Connect neighboring poses

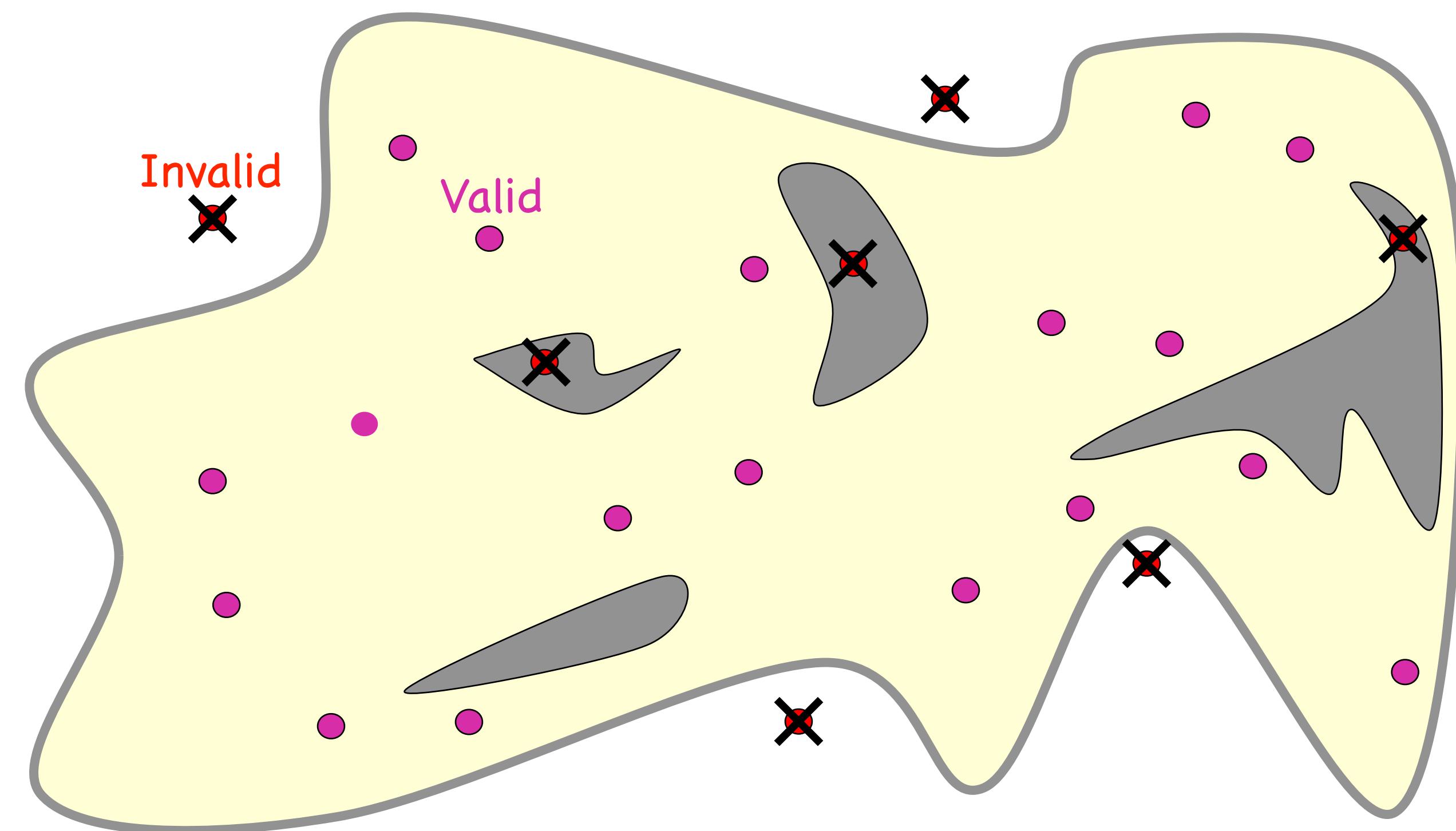


PRM: construction phase

1) Select N sample poses at random

2) Eliminate invalid poses

3) Connect neighboring poses

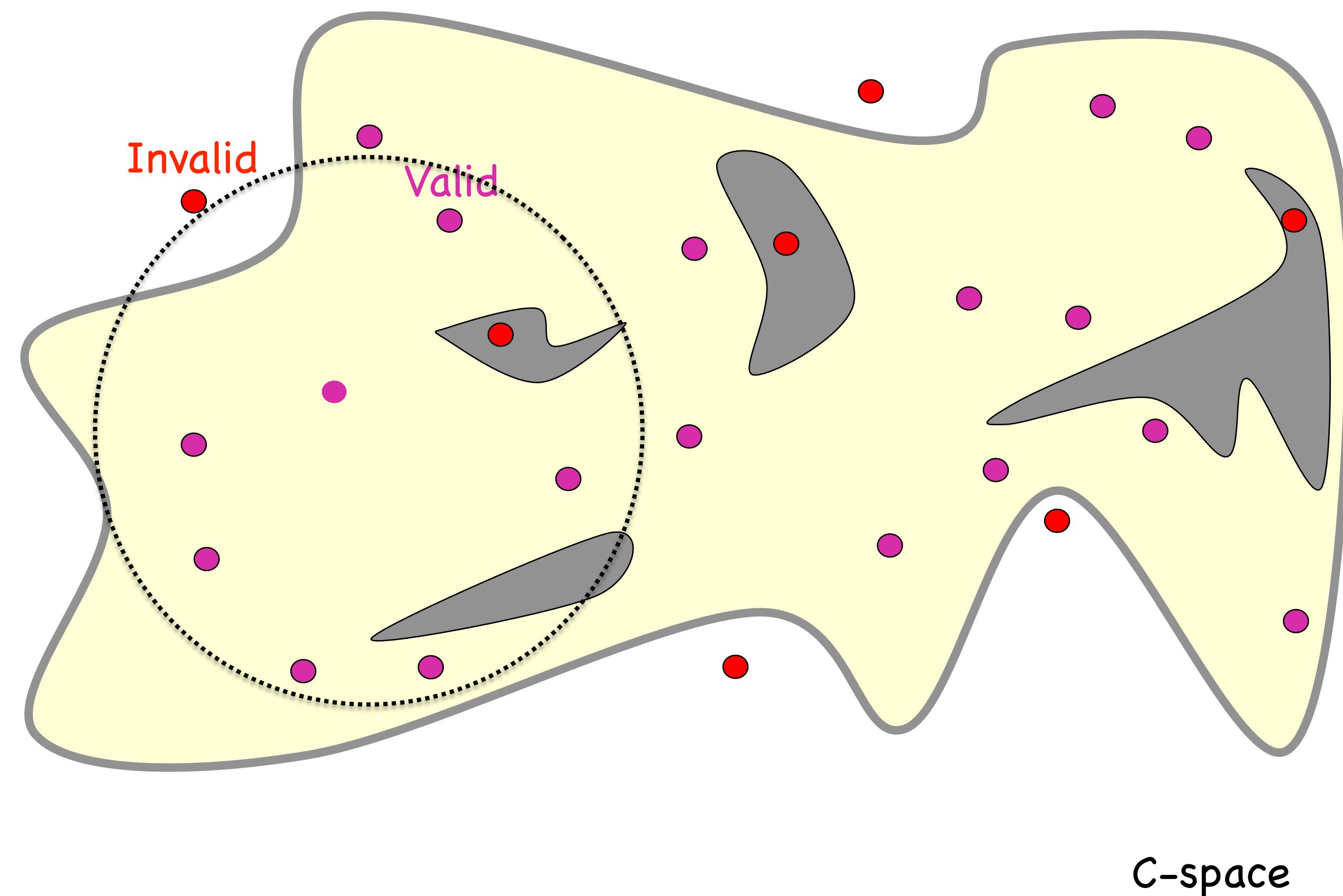


*Collision detection
will be covered later*

C-space

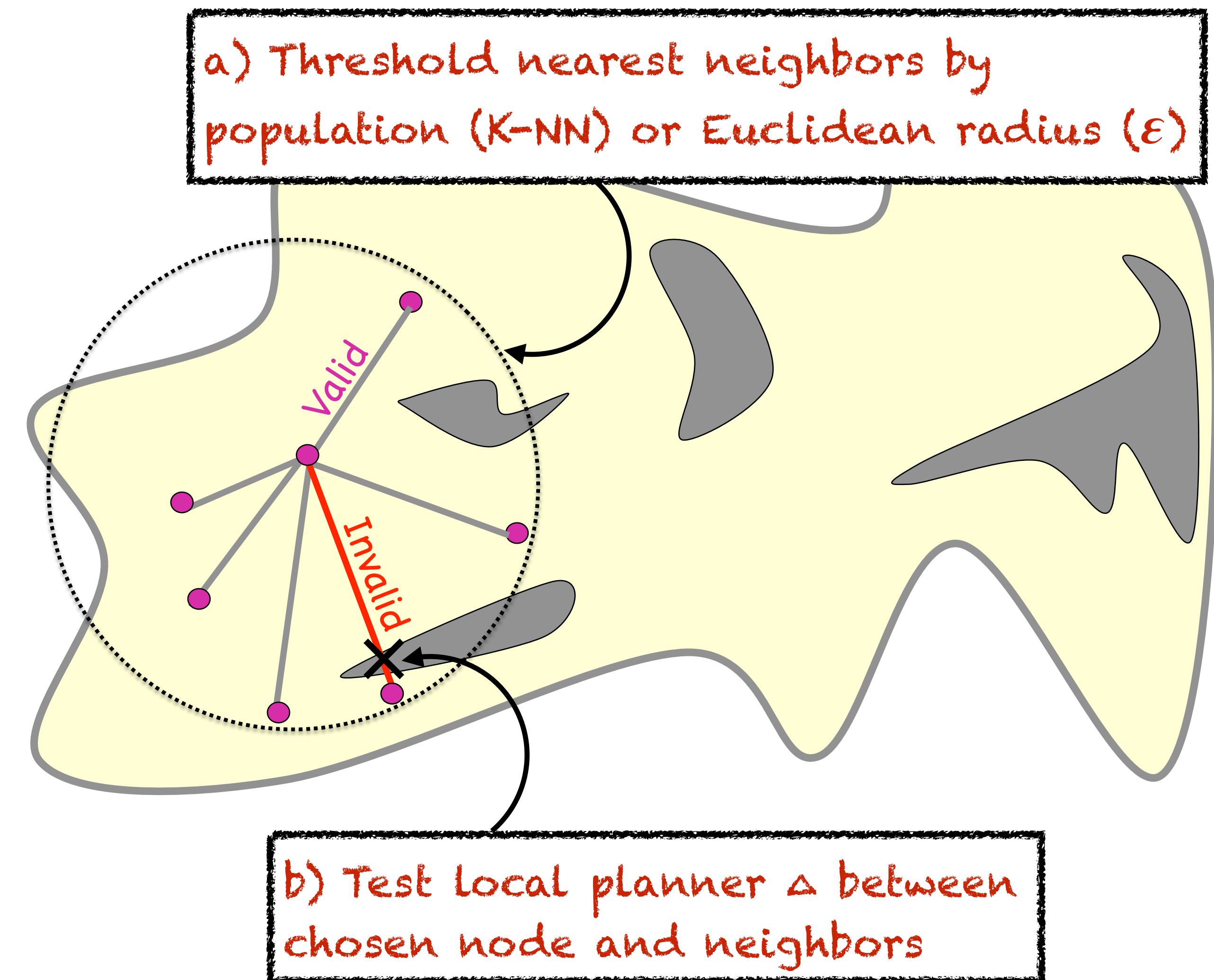
PRM: construction phase

- 1) Select N sample poses at random
- 2) Eliminate invalid poses
- 3) Connect neighboring poses



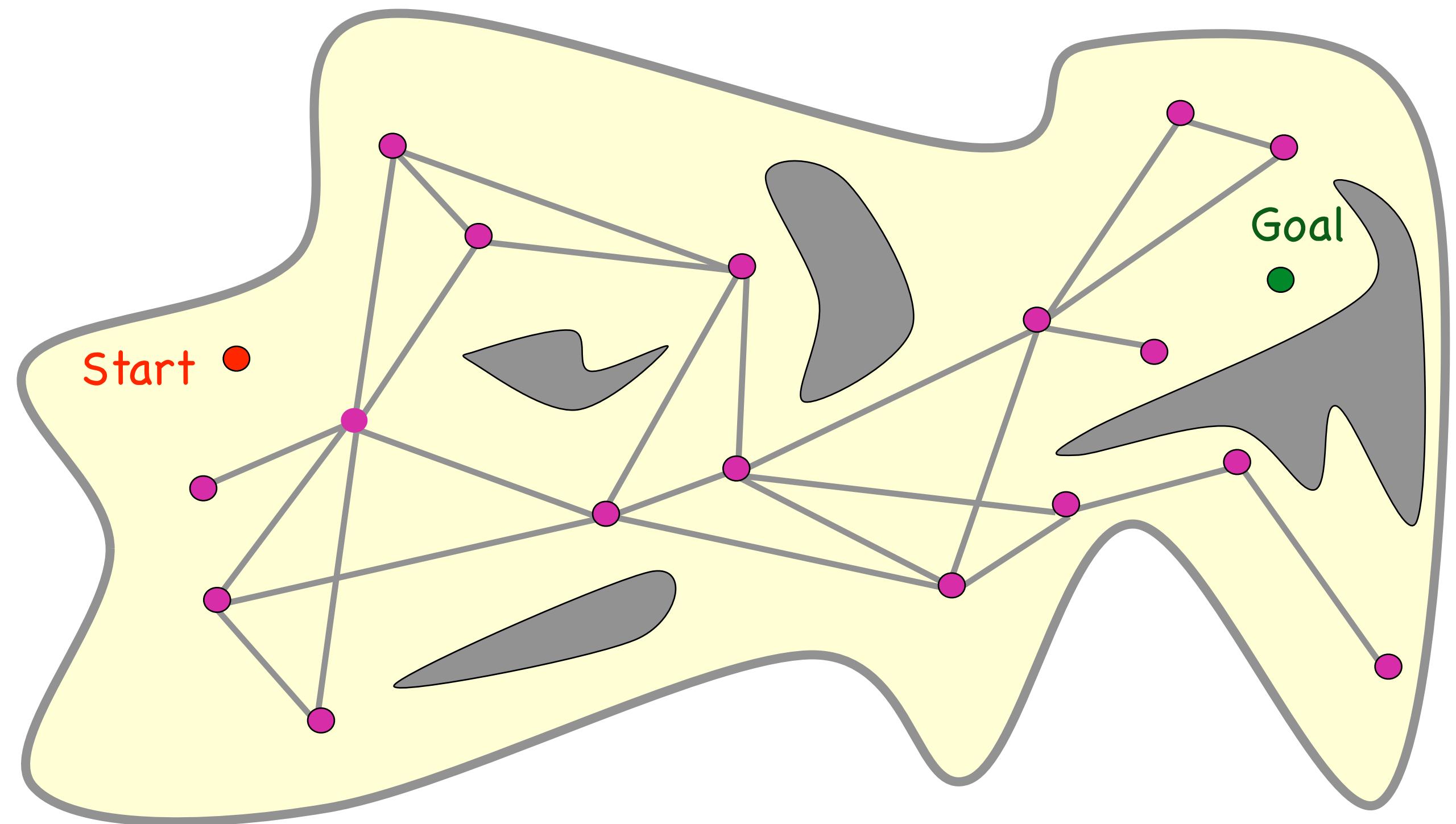
PRM: construction phase

- 1) Select N sample poses at random
- 2) Eliminate invalid poses
- 3) Connect neighboring poses



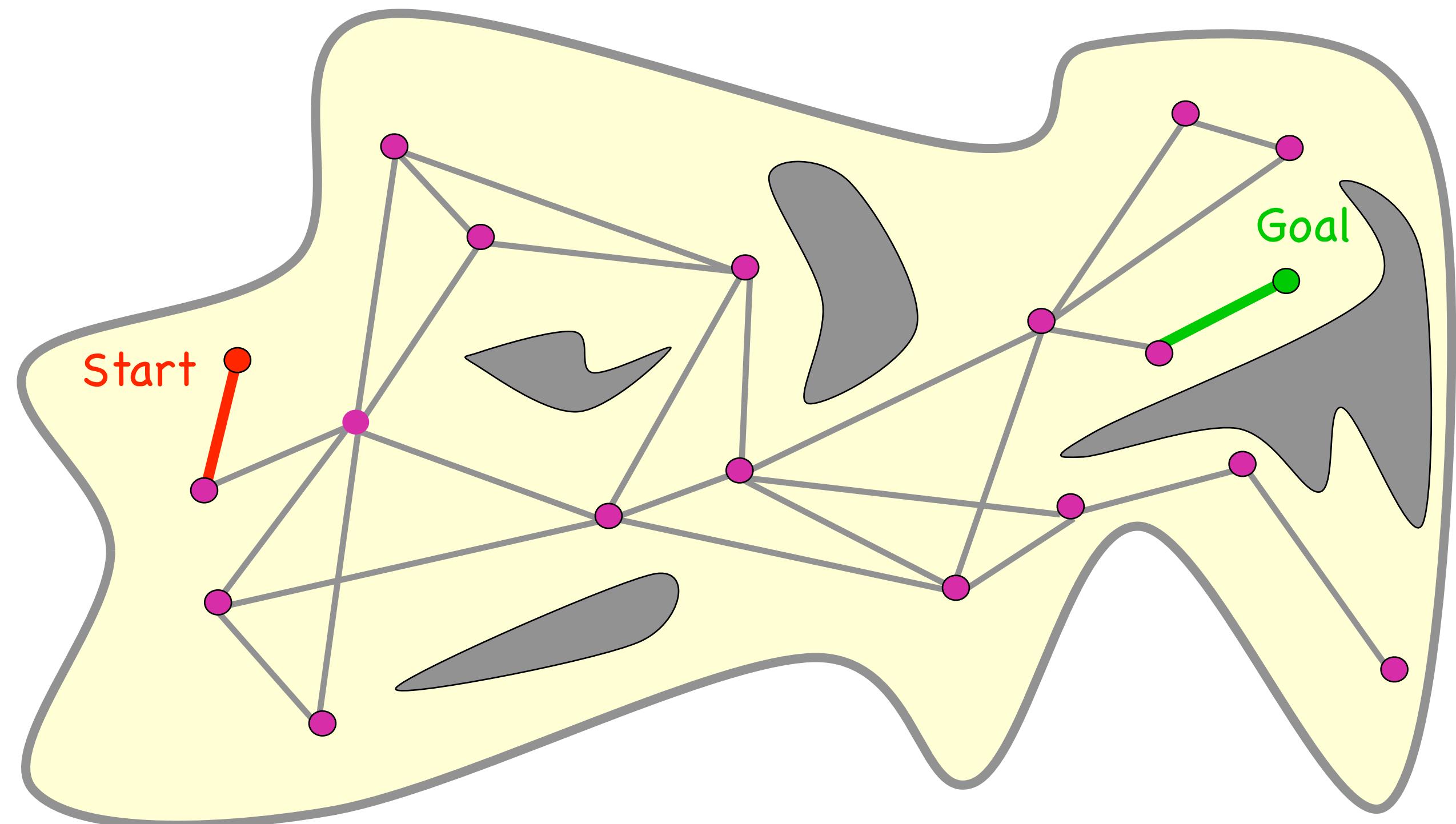
PRM: query phase

- 1) Given constructed roadmap, start pose, and goal pose
- 2) Attach goal and start to nearest roadmap entry nodes
- 3) Search for path between roadmap entry nodes
- 4) Return path with entry and departure edges



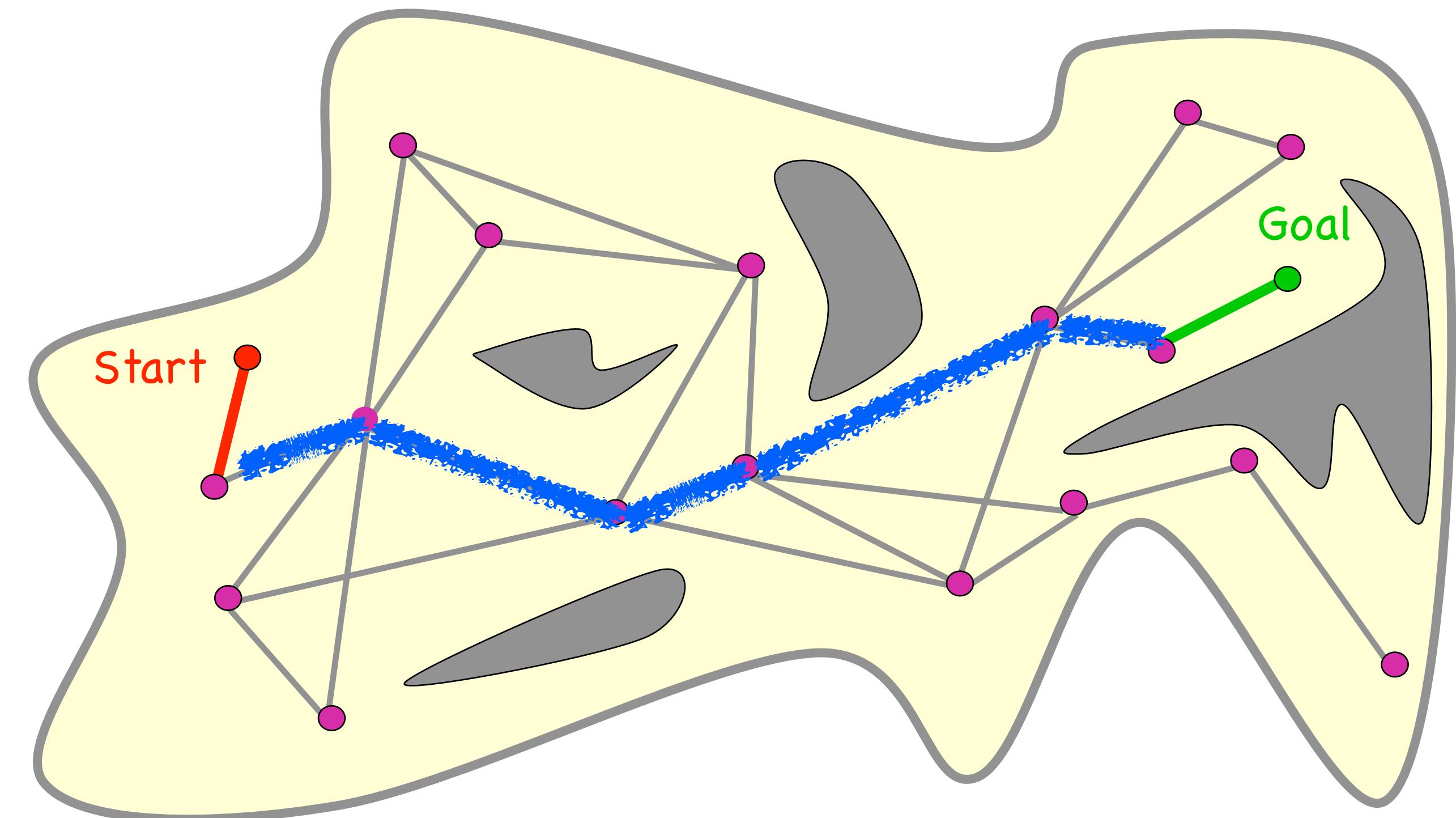
PRM: query phase

- 1) Given constructed roadmap, start pose, and goal pose
- 2) **Attach goal and start to nearest roadmap entry nodes**
- 3) Search for path between roadmap entry nodes
- 4) Return path with entry and departure edges



PRM: query phase

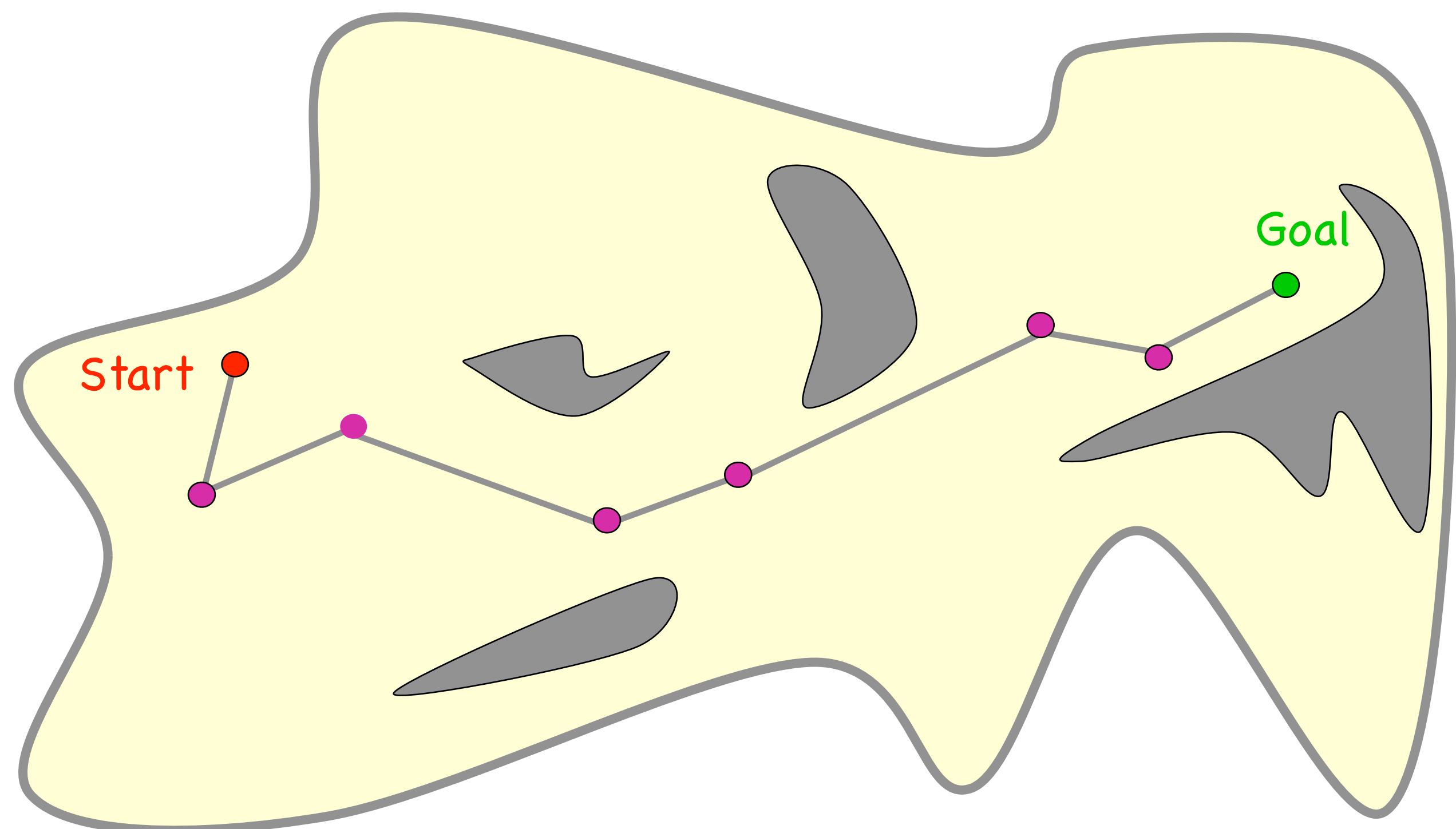
- 1) Given constructed roadmap, start pose, and goal pose
- 2) Attach goal and start to nearest roadmap entry nodes
- 3) **Search for path between roadmap entry nodes**
- 4) Return path with entry and departure edges



*Remember: graph search algorithms
A*, Dijkstra, BFS, DFS*

PRM: query phase

- 1) Given constructed roadmap, start pose, and goal pose
- 2) Attach goal and start to nearest roadmap entry nodes
- 3) Search for path between roadmap entry nodes
- 4) Return path with entry and departure edges



Multi-query planning: Considerations

- Number of samples wrt. C-space dimensionality
- Balanced sampling over C-space
- Choice of distance (e.g., Euclidean)
- Choice of local planner (e.g., line subdivision)
- Selecting neighbors: (e.g., K-NN, kd-tree, cell hashing)

2 Approaches to Roadmaps

Deterministic:

complete algorithms

- Visibility Graph
 - trace lines connecting obstacle polygon vertices
- Voronoi Planning
 - trace edges equidistant from obstacles

Probabilistic:

C-space sampling

- Probabilistic Roadmap (PRM)
 - sample and connect vertices in graph for multiple planning queries
- Rapidly-exploring Random Tree (RRT)
 - sample and connect vertices in trees rooted at start and goal configuration

Single Query Planning

- Given specific start and goal configurations
- Grow trees from start and goal towards each other
- Path is found once trees connect
- Focus sampling in unexplored areas of C-space and moving towards start/goal
- Common algorithms:
 - ESTs (expansive space trees)
 - **RRTs (rapidly exploring random trees)**

RRT Algorithm

RRT Algorithm

Extend graph towards a random configuration and repeat

```
BUILD_RRT( $q_{init}$ )
1    $T.init(q_{init})$ ;
2   for  $k = 1$  to  $K$  do
3        $q_{rand} \leftarrow \text{RANDOM\_CONFIG}()$ ;
4       EXTEND( $T, q_{rand}$ );
5   Return  $T$ 
```

RRT Algorithm

Extend graph towards a random configuration and repeat

```
BUILD_RRT( $q_{init}$ )
1    $T.init(q_{init})$ ;
2   for  $k = 1$  to  $K$  do
3        $q_{rand} \leftarrow \text{RANDOM\_CONFIG}()$ ;
4       EXTEND( $T, q_{rand}$ );
5   Return  $T$ 
```



Figure 3: The EXTEND operation.

RRT Algorithm

Extend graph towards a random configuration and repeat

```
BUILD_RRT( $q_{init}$ )
1    $T.init(q_{init})$ ;
2   for  $k = 1$  to  $K$  do
3        $q_{rand} \leftarrow \text{RANDOM\_CONFIG}()$ ;
4       EXTEND( $T, q_{rand}$ );
5   Return  $T$ 
```

```
EXTEND( $T, q$ )
1    $q_{near} \leftarrow \text{NEAREST\_NEIGHBOR}(q, T)$ ;
2   if NEW_CONFIG( $q, q_{near}, q_{new}$ ) then
3        $T.add\_vertex(q_{new})$ ;
4        $T.add\_edge(q_{near}, q_{new})$ ;
5       if  $q_{new} = q$  then
6           Return Reached;
7       else
8           Return Advanced;
9   Return Trapped;
```

Extend graph towards a random configuration

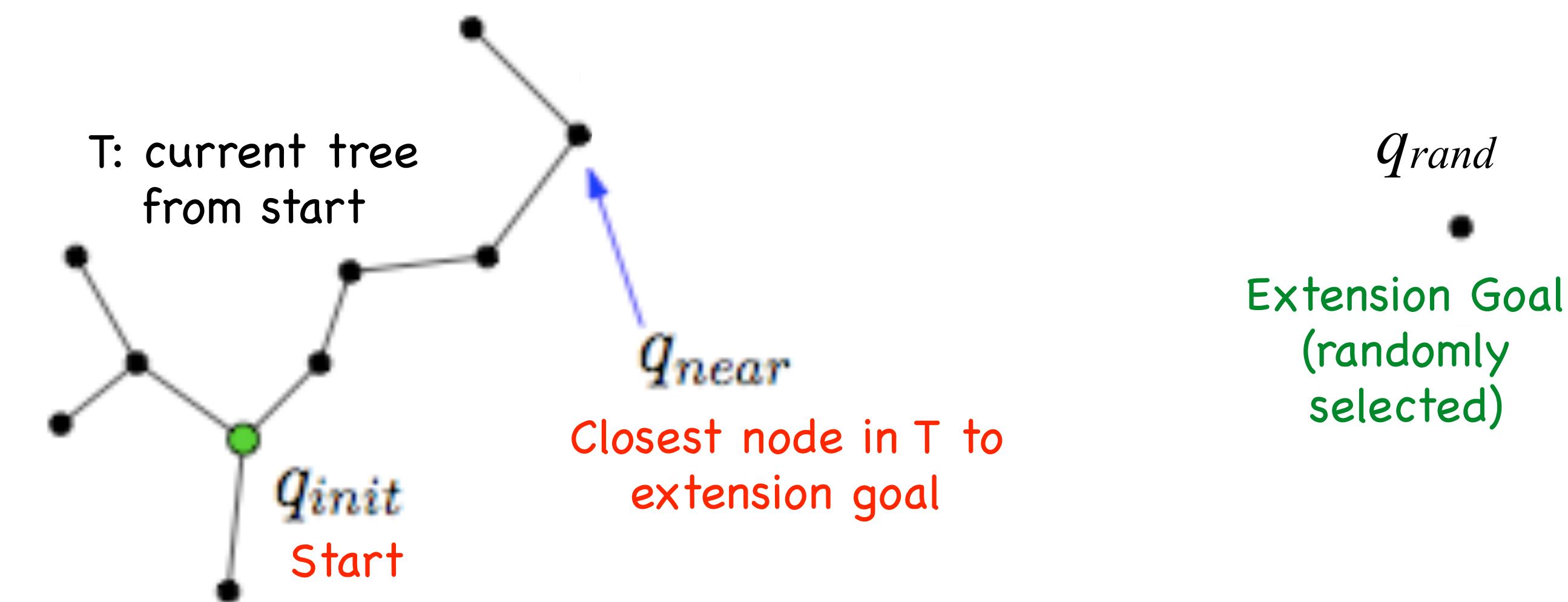


Figure 3: The EXTEND operation.

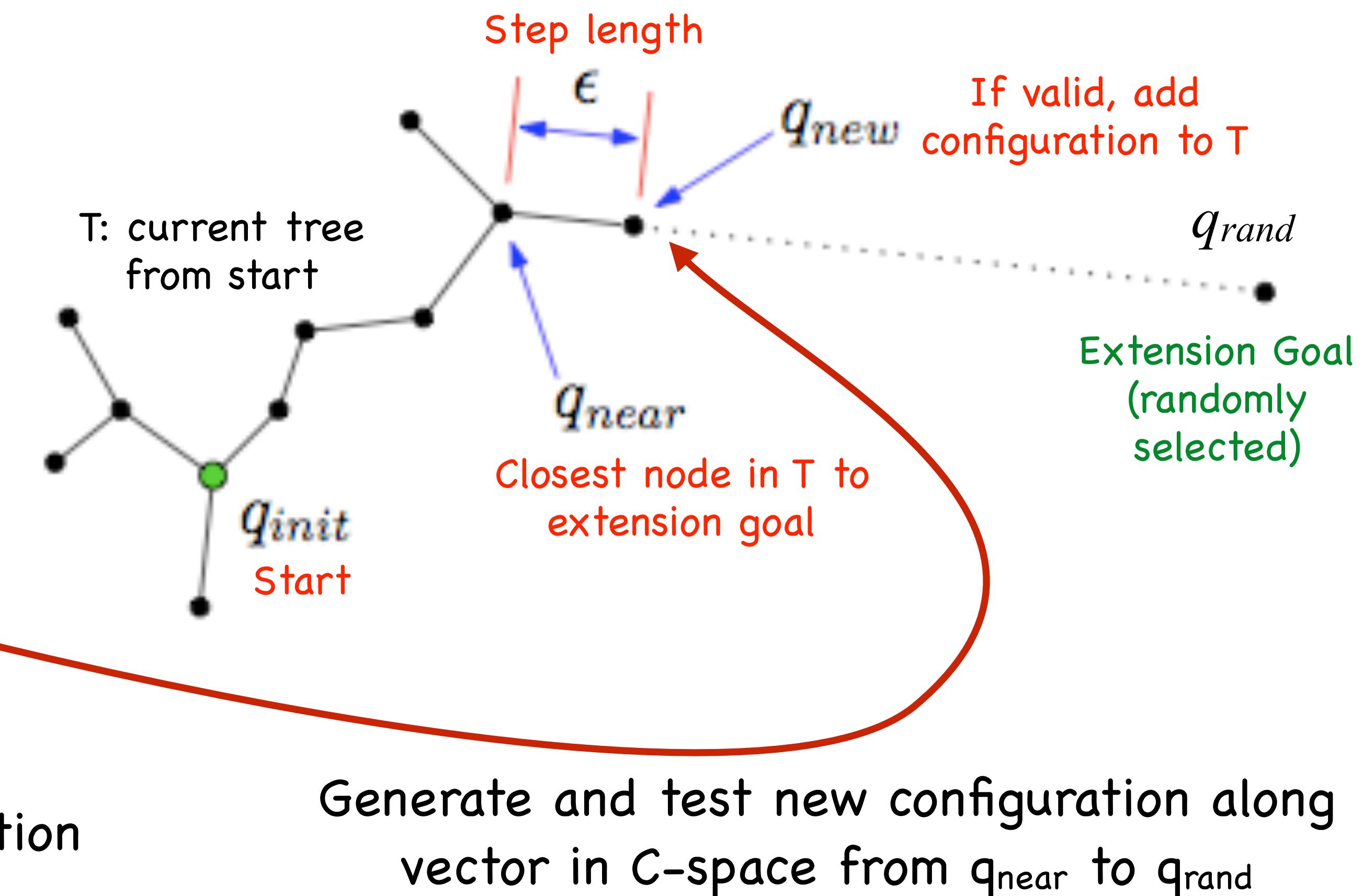
RRT Algorithm

Extend graph towards a random configuration and repeat

```
BUILD_RRT( $q_{init}$ )
1  $T.init(q_{init})$ ;
2 for  $k = 1$  to  $K$  do
3    $q_{rand} \leftarrow RANDOM\_CONFIG()$ ;
4   EXTEND( $T, q_{rand}$ );
5 Return  $T$ 
```

```
EXTEND( $T, q$ )
1  $q_{near} \leftarrow NEAREST\_NEIGHBOR(q, T)$ ;
2 if NEW_CONFIG( $q, q_{near}, q_{new}$ ) then
3    $T.add\_vertex(q_{new})$ ;
4    $T.add\_edge(q_{near}, q_{new})$ ;
5   if  $q_{new} = q$  then
6     Return Reached;
7   else
8     Return Advanced;
9 Return Trapped;
```

Extend graph towards a random configuration



RRT Extend animation

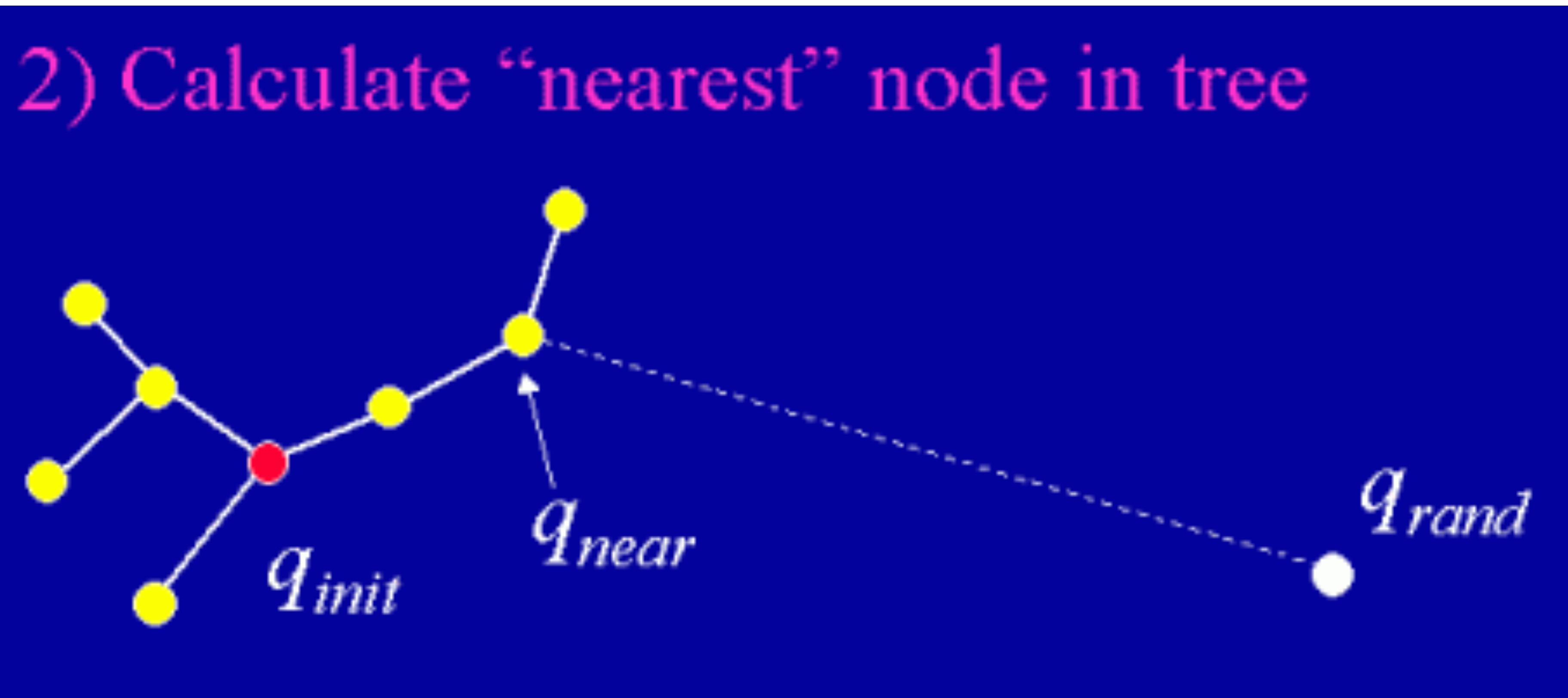


RRT Extend animation

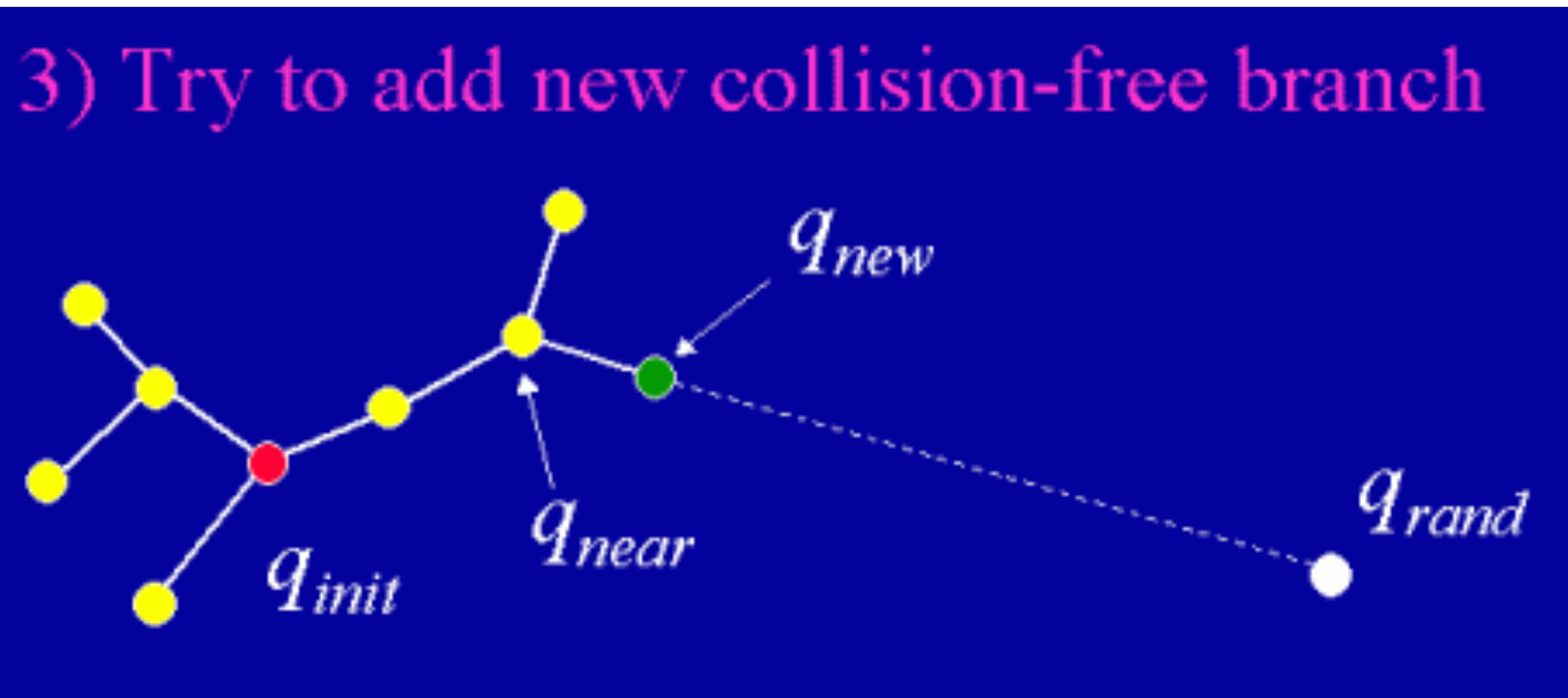
1) Select a random “target” node



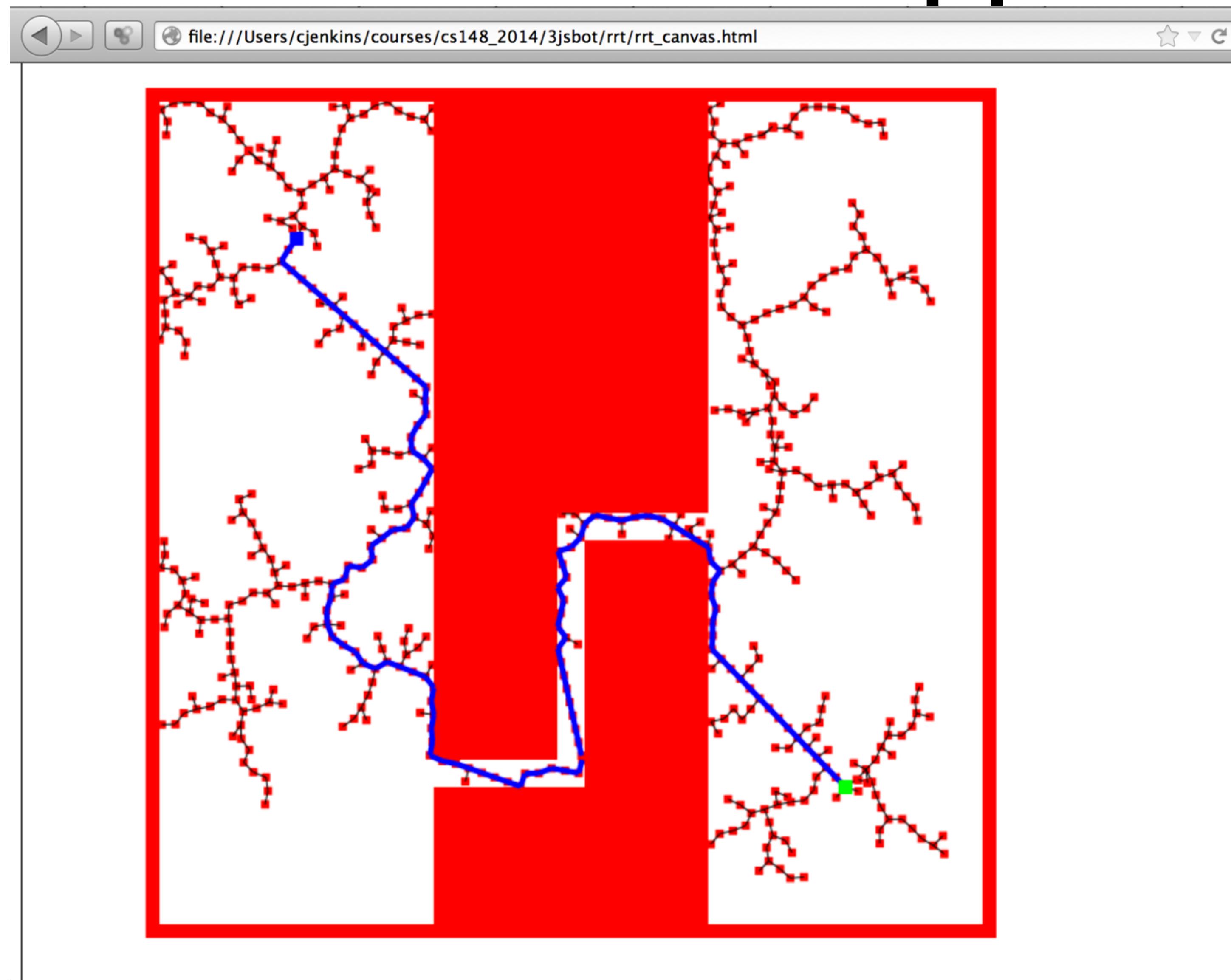
RRT Extend animation



RRT Extend animation



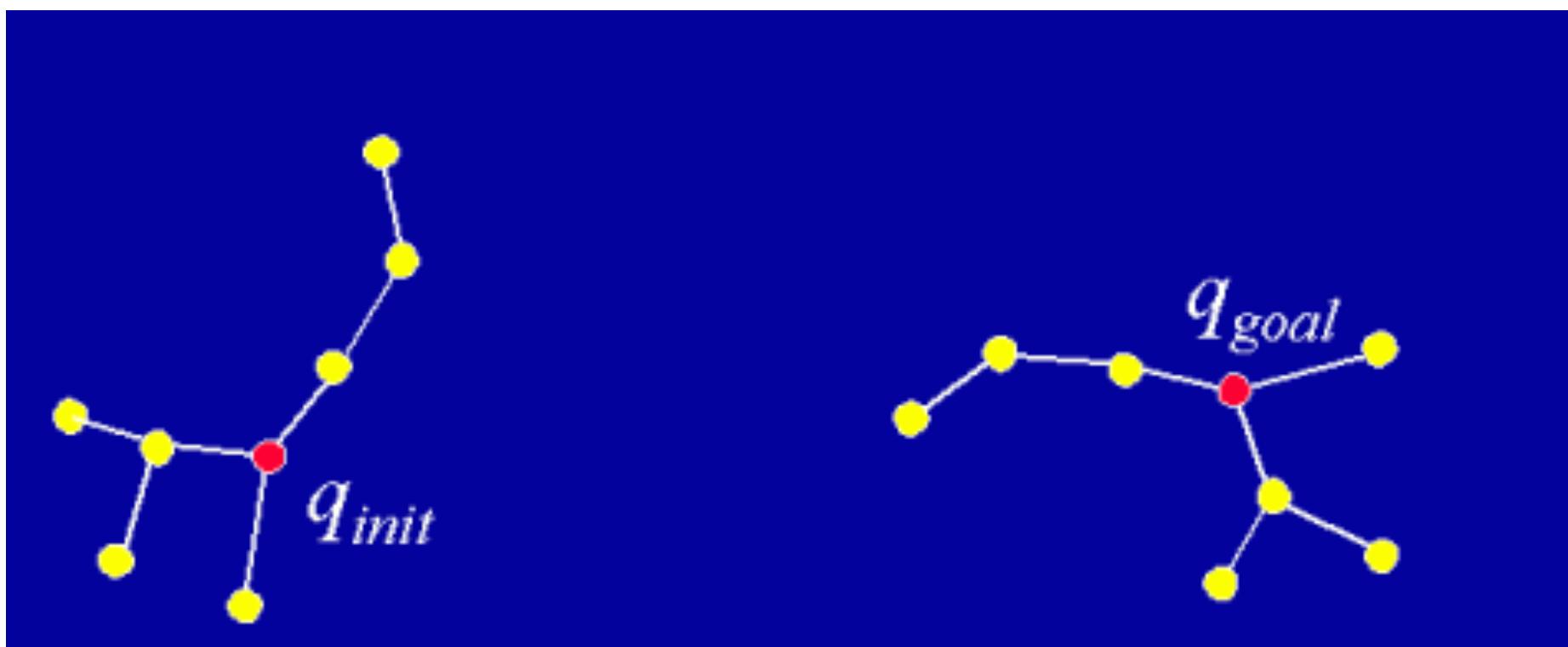
Let's see what happens



RRT Connect

0) Use 2 trees (A and B) rooted at start and goal configurations

```
RRT_CONNECT_PLANNER( $q_{init}, q_{goal}$ )
1    $\mathcal{T}_a.init(q_{init}); \mathcal{T}_b.init(q_{goal});$ 
2   for  $k = 1$  to  $K$  do
3        $q_{rand} \leftarrow \text{RANDOM\_CONFIG}();$ 
4       if not ( $\text{EXTEND}(\mathcal{T}_a, q_{rand}) = \text{Trapped}$ ) then
5           if ( $\text{CONNECT}(\mathcal{T}_b, q_{new}) = \text{Reached}$ ) then
6               Return PATH( $\mathcal{T}_a, \mathcal{T}_b$ );
7           SWAP( $\mathcal{T}_a, \mathcal{T}_b$ );
8   Return Failure
```



RRT Connect

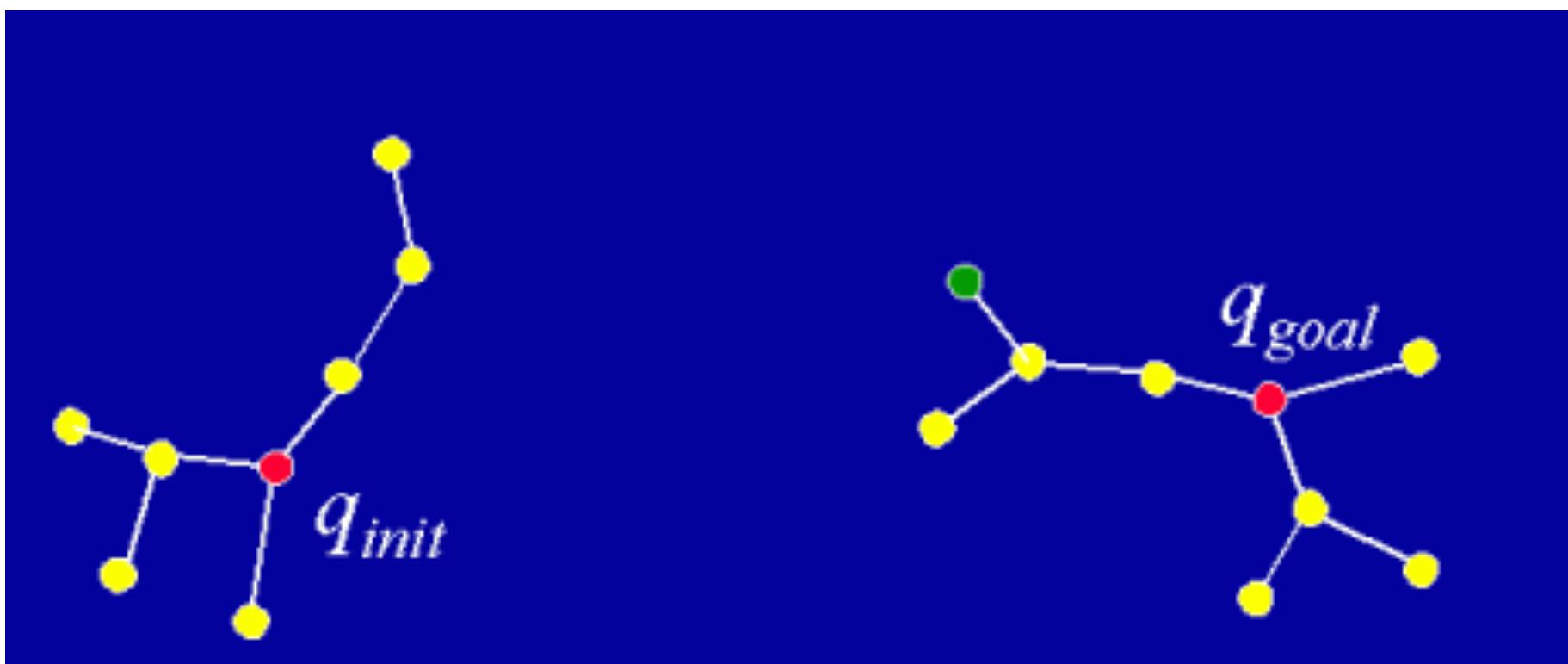
0) Use 2 trees (A and B) rooted at start and goal configurations

```
RRT_CONNECT_PLANNER( $q_{init}, q_{goal}$ )
1    $\mathcal{T}_a$ .init( $q_{init}$ );  $\mathcal{T}_b$ .init( $q_{goal}$ );
2   for  $k = 1$  to  $K$  do
3        $q_{rand} \leftarrow$  RANDOM_CONFIG();
4       if not (EXTEND( $\mathcal{T}_a, q_{rand}$ ) = Trapped) then
5           if (CONNECT( $\mathcal{T}_b, q_{new}$ ) = Reached) then
6               Return PATH( $\mathcal{T}_a, \mathcal{T}_b$ );
7           SWAP( $\mathcal{T}_a, \mathcal{T}_b$ );
8   Return Failure
```

EXTEND(\mathcal{T}, q)

```
1    $q_{near} \leftarrow$  NEAREST_NEIGHBOR( $q, \mathcal{T}$ );
2   if NEW_CONFIG( $q, q_{near}, q_{new}$ ) then
3        $\mathcal{T}$ .add_vertex( $q_{new}$ );
4        $\mathcal{T}$ .add_edge( $q_{near}, q_{new}$ );
5       if  $q_{new} = q$  then
6           Return Reached;
7       else
8           Return Advanced;
9   Return Trapped;
```

1) Extend tree A towards a random configuration



RRT Connect

0) Use 2 trees (A and B) rooted at start and goal configurations

```
RRT_CONNECT_PLANNER( $q_{init}, q_{goal}$ )
1  $\mathcal{T}_a.init(q_{init}); \mathcal{T}_b.init(q_{goal});$ 
2 for  $k = 1$  to  $K$  do
3      $q_{rand} \leftarrow \text{RANDOM\_CONFIG}();$ 
4     if not (EXTEND( $\mathcal{T}_a, q_{rand}$ ) = Trapped) then
5         if (CONNECT( $\mathcal{T}_b, q_{new}$ ) = Reached) then
6             Return PATH( $\mathcal{T}_a, \mathcal{T}_b$ );
7         SWAP( $\mathcal{T}_a, \mathcal{T}_b$ );
8     Return Failure
```

EXTEND(\mathcal{T}, q)

```
1  $q_{near} \leftarrow \text{NEAREST\_NEIGHBOR}(q, \mathcal{T});$ 
2 if NEW_CONFIG( $q, q_{near}, q_{new}$ ) then
3      $\mathcal{T}.add\_vertex(q_{new});$ 
4      $\mathcal{T}.add\_edge(q_{near}, q_{new});$ 
5     if  $q_{new} = q$  then
6         Return Reached;
7     else
8         Return Advanced;
9 Return Trapped;
```

1) Extend tree A towards a random configuration



CONNECT(\mathcal{T}, q)

```
1 repeat
2      $S \leftarrow \text{EXTEND}(\mathcal{T}, q);$ 
3 until not ( $S = \text{Advanced}$ )
4 Return  $S$ ;
```

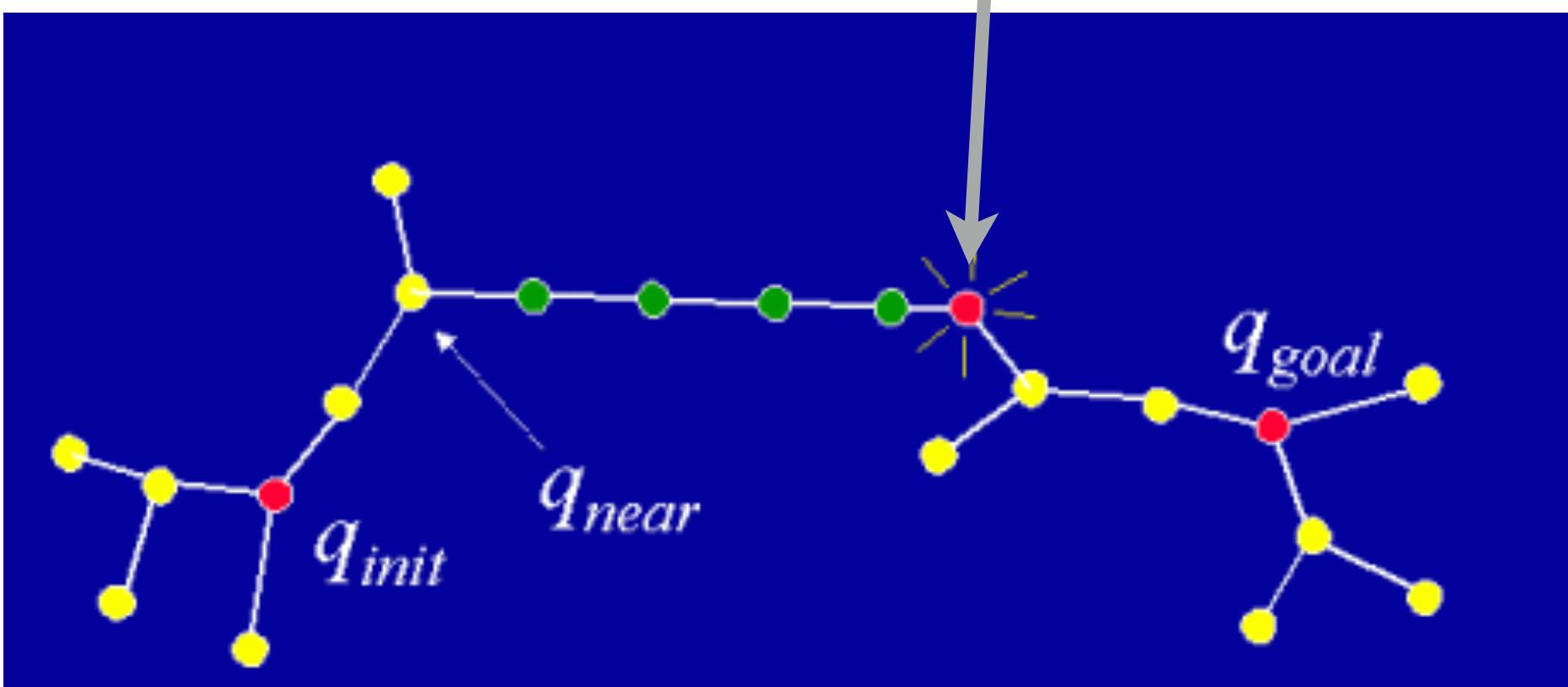
2) Try to connect tree B to tree A by extending repeatedly from its nearest neighbor

RRT Connect

0) Use 2 trees (A and B) rooted at start and goal configurations

```
RRT_CONNECT_PLANNER( $q_{init}, q_{goal}$ )
1  $\mathcal{T}_a.init(q_{init}); \mathcal{T}_b.init(q_{goal});$ 
2 for  $k = 1$  to  $K$  do
3      $q_{rand} \leftarrow \text{RANDOM\_CONFIG}();$ 
4     if not (EXTEND( $\mathcal{T}_a, q_{rand}$ ) = Trapped) then
5         if (CONNECT( $\mathcal{T}_b, q_{new}$ ) = Reached) then
6             Return PATH( $\mathcal{T}_a, \mathcal{T}_b$ );
7         SWAP( $\mathcal{T}_a, \mathcal{T}_b$ );
8     Return Failure
```

search succeeds if trees connect



EXTEND(\mathcal{T}, q)

```
1  $q_{near} \leftarrow \text{NEAREST\_NEIGHBOR}(q, \mathcal{T});$ 
2 if NEW_CONFIG( $q, q_{near}, q_{new}$ ) then
3      $\mathcal{T}.add\_vertex(q_{new});$ 
4      $\mathcal{T}.add\_edge(q_{near}, q_{new});$ 
5     if  $q_{new} = q$  then
6         Return Reached;
7     else
8         Return Advanced;
9 Return Trapped;
```

1) Extend tree A towards a random configuration

CONNECT(\mathcal{T}, q)

```
1 repeat
2      $S \leftarrow \text{EXTEND}(\mathcal{T}, q);$ 
3 until not ( $S = \text{Advanced}$ )
4 Return  $S$ ;
```

2) Try to connect tree B to tree A by extending repeatedly from its nearest neighbor

RRT Connect

- 0) Use 2 trees (A and B) rooted at start and goal configurations

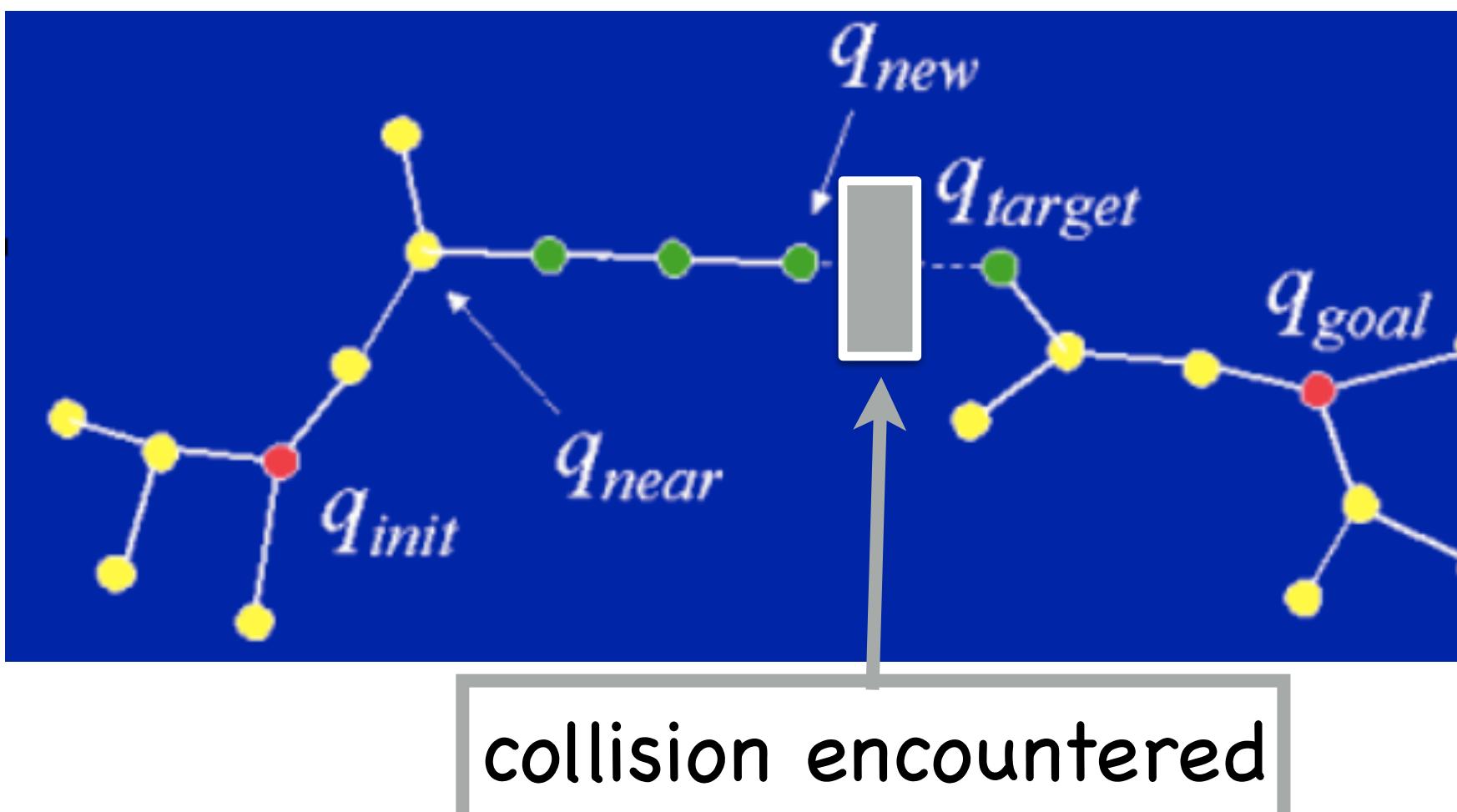
RRT_CONNECT_PLANNER(q_{init}, q_{goal})

```

1  $\mathcal{T}_a.init(q_{init}); \mathcal{T}_b.init(q_{goal});$ 
2 for  $k = 1$  to  $K$  do
3    $q_{rand} \leftarrow \text{RANDOM\_CONFIG}();$ 
4   if not (EXTEND( $\mathcal{T}_a, q_{rand}$ ) = Trapped) then
5     if (CONNECT( $\mathcal{T}_b, q_{new}$ ) = Reached) then
6       Return PATH( $\mathcal{T}_a, \mathcal{T}_b$ );
7     SWAP( $\mathcal{T}_a, \mathcal{T}_b$ );
8   Return Failure

```

- 3) reverse roles for trees A and B and repeat



EXTEND(\mathcal{T}, q)

```

1  $q_{near} \leftarrow \text{NEAREST\_NEIGHBOR}(q, \mathcal{T});$ 
2 if NEW_CONFIG( $q, q_{near}, q_{new}$ ) then
3    $\mathcal{T}.add\_vertex(q_{new});$ 
4    $\mathcal{T}.add\_edge(q_{near}, q_{new});$ 
5   if  $q_{new} = q$  then
6     Return Reached;
7   else
8     Return Advanced;
9 Return Trapped;

```

- 1) Extend tree A towards a random configuration

CONNECT(\mathcal{T}, q)

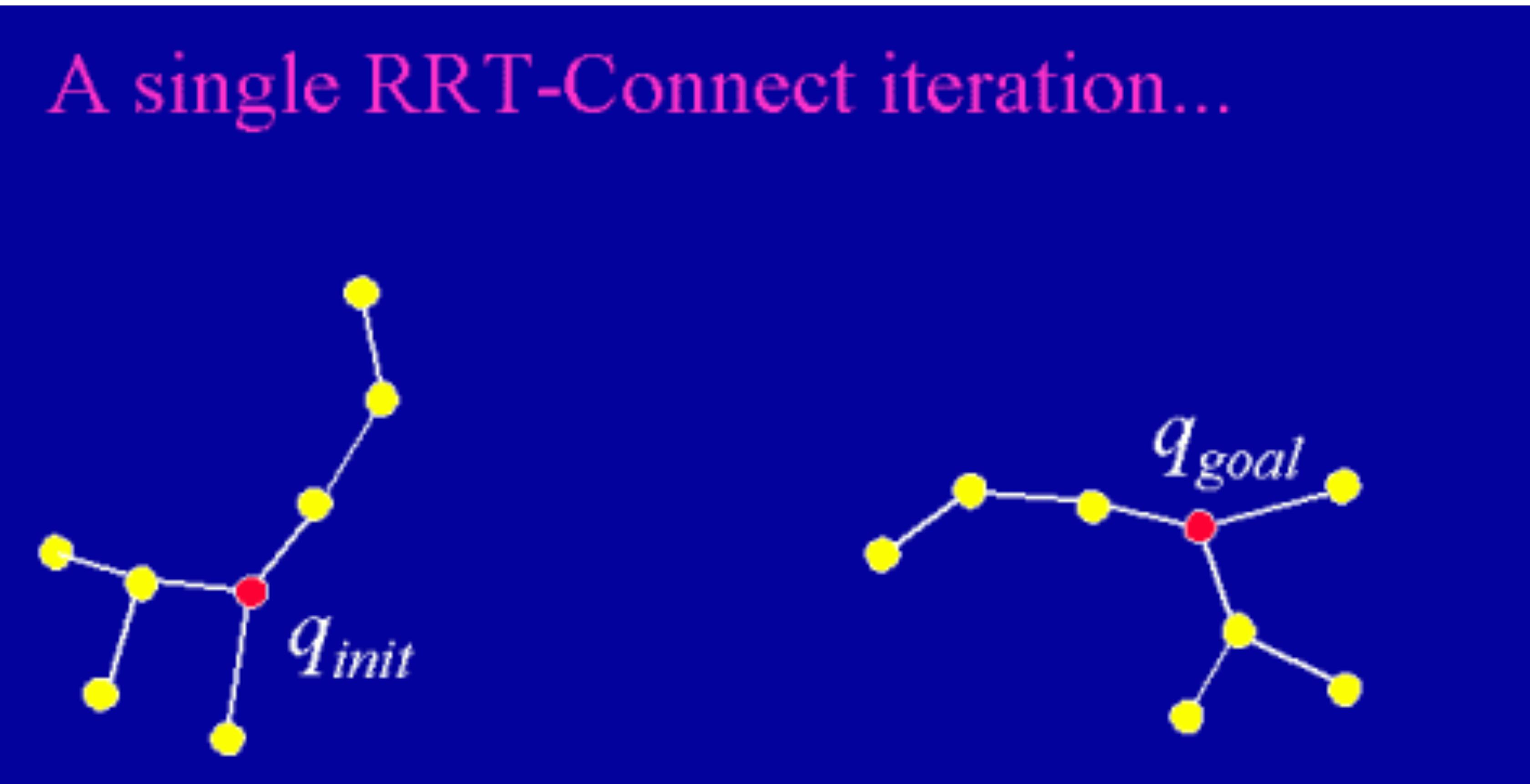
```

1 repeat
2    $S \leftarrow \text{EXTEND}(\mathcal{T}, q);$ 
3 until not ( $S = \text{Advanced}$ )
4 Return  $S;$ 

```

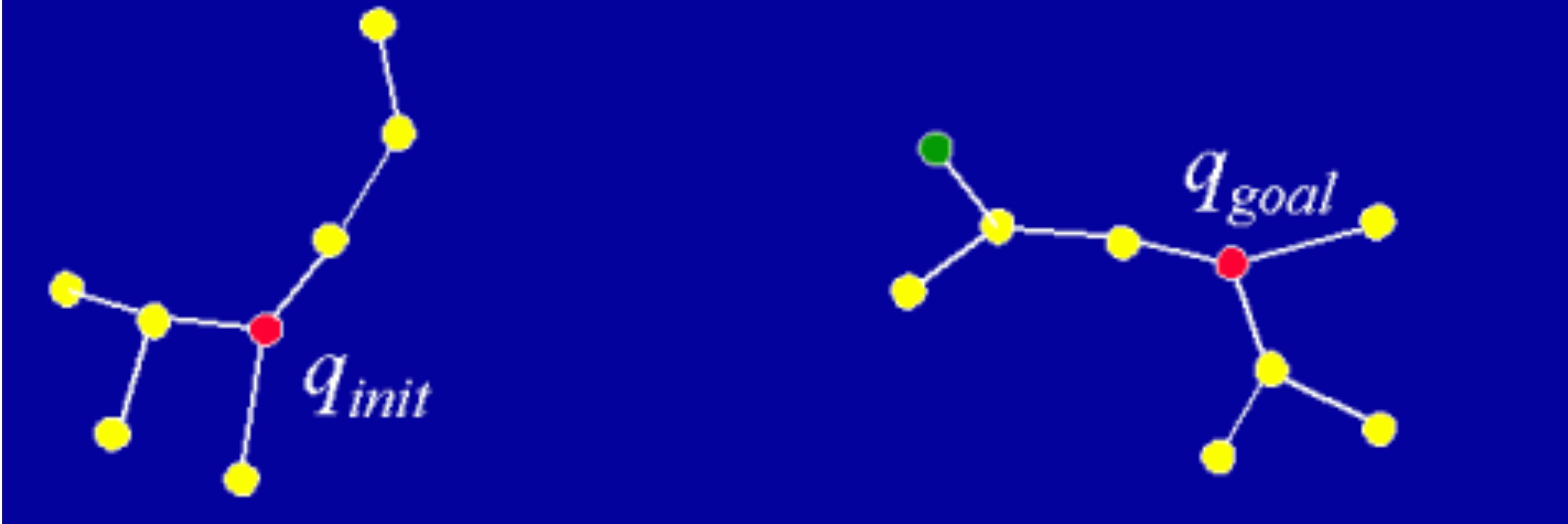
- 2) Try to connect tree B to tree A by extending repeatedly from its nearest neighbor

RRT-Connect animation

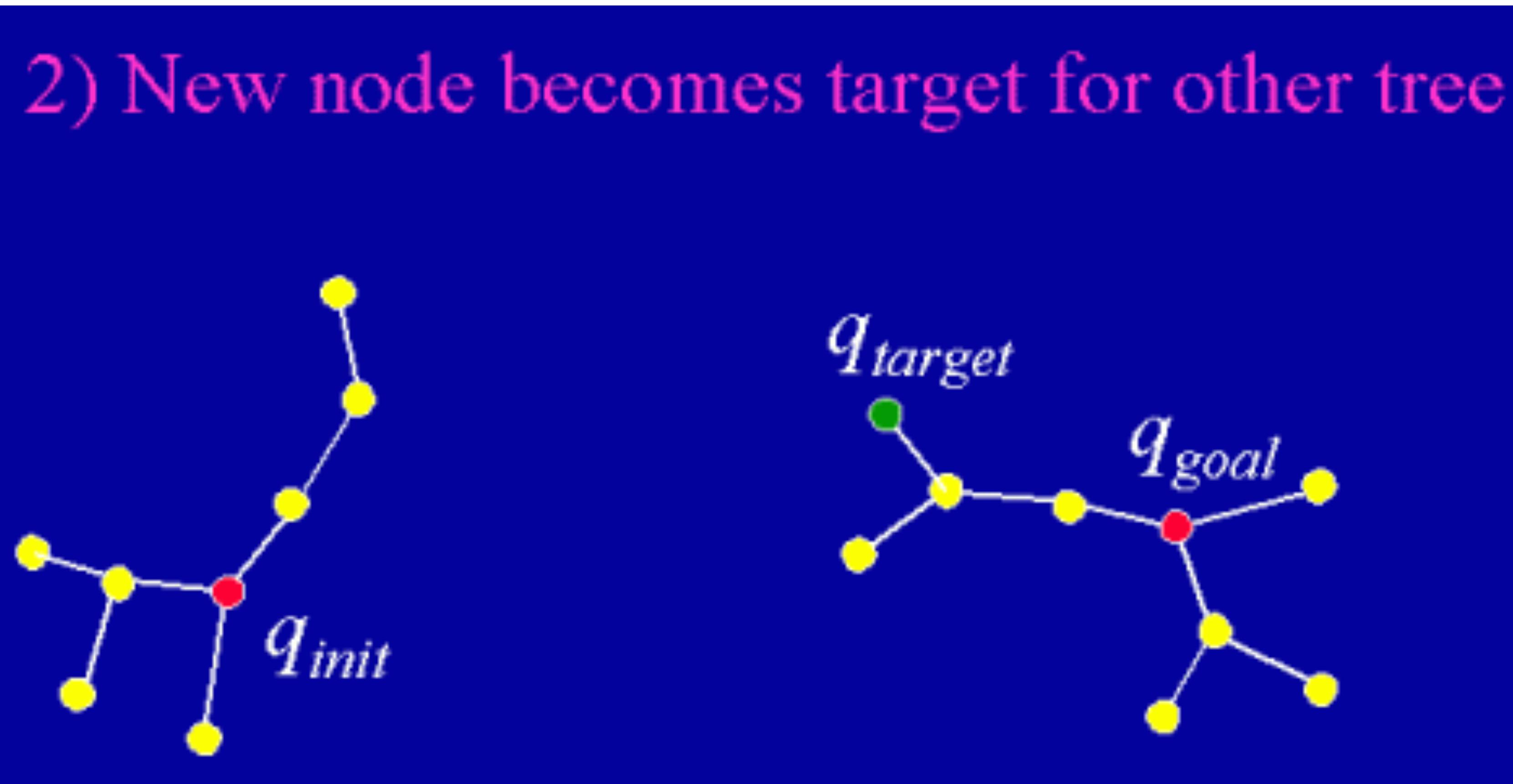


RRT-Connect animation

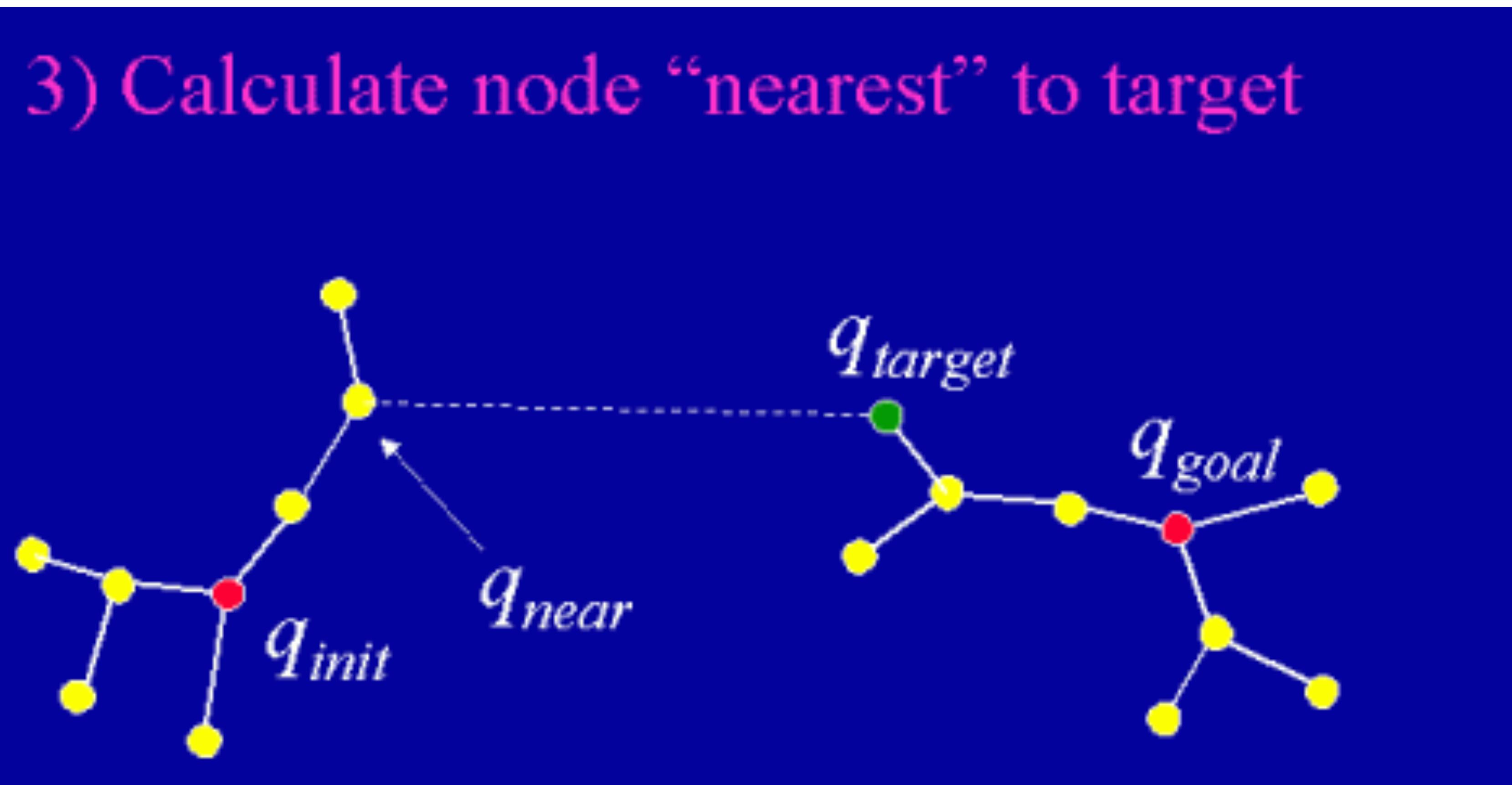
1) One tree grown using random target



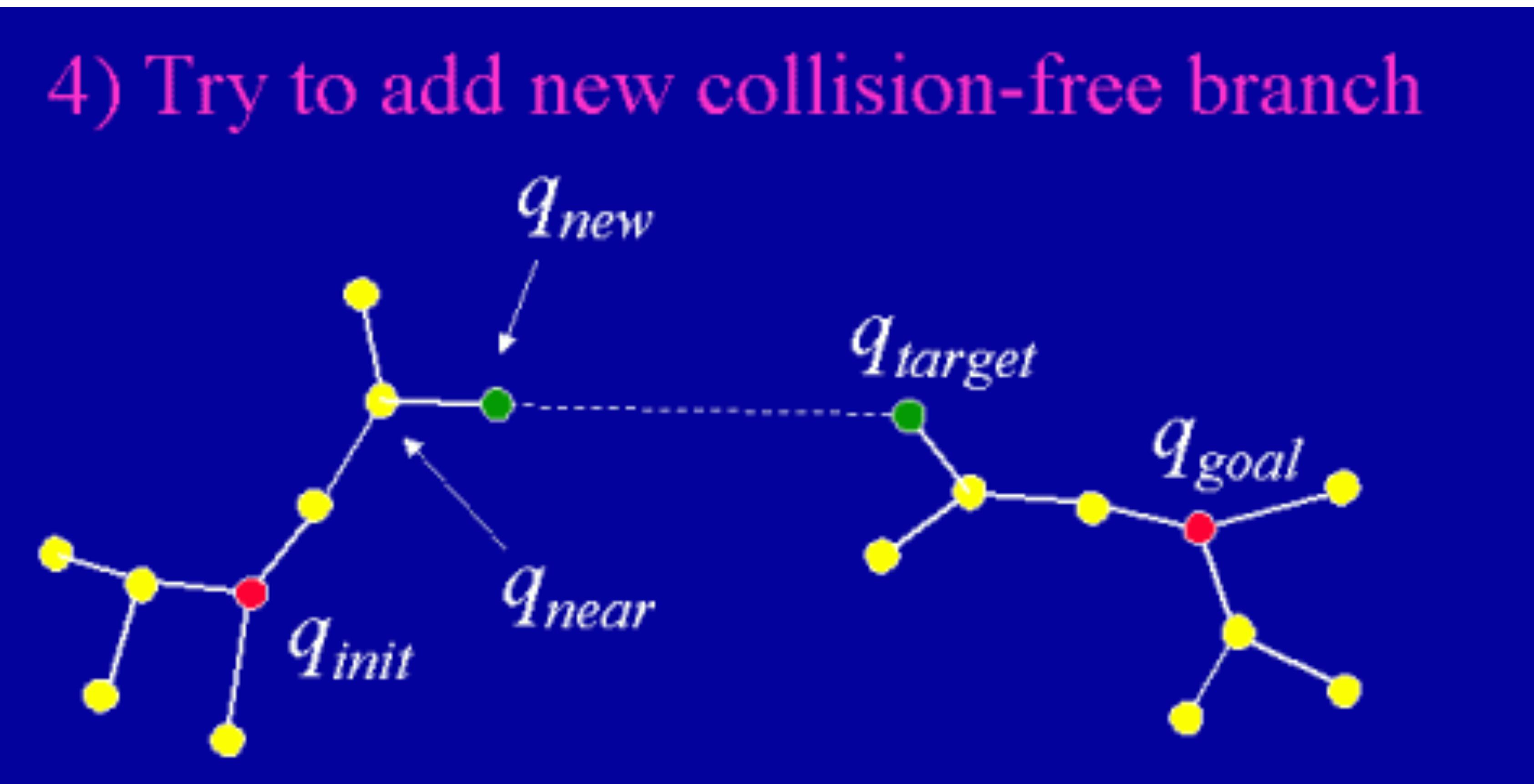
RRT-Connect animation



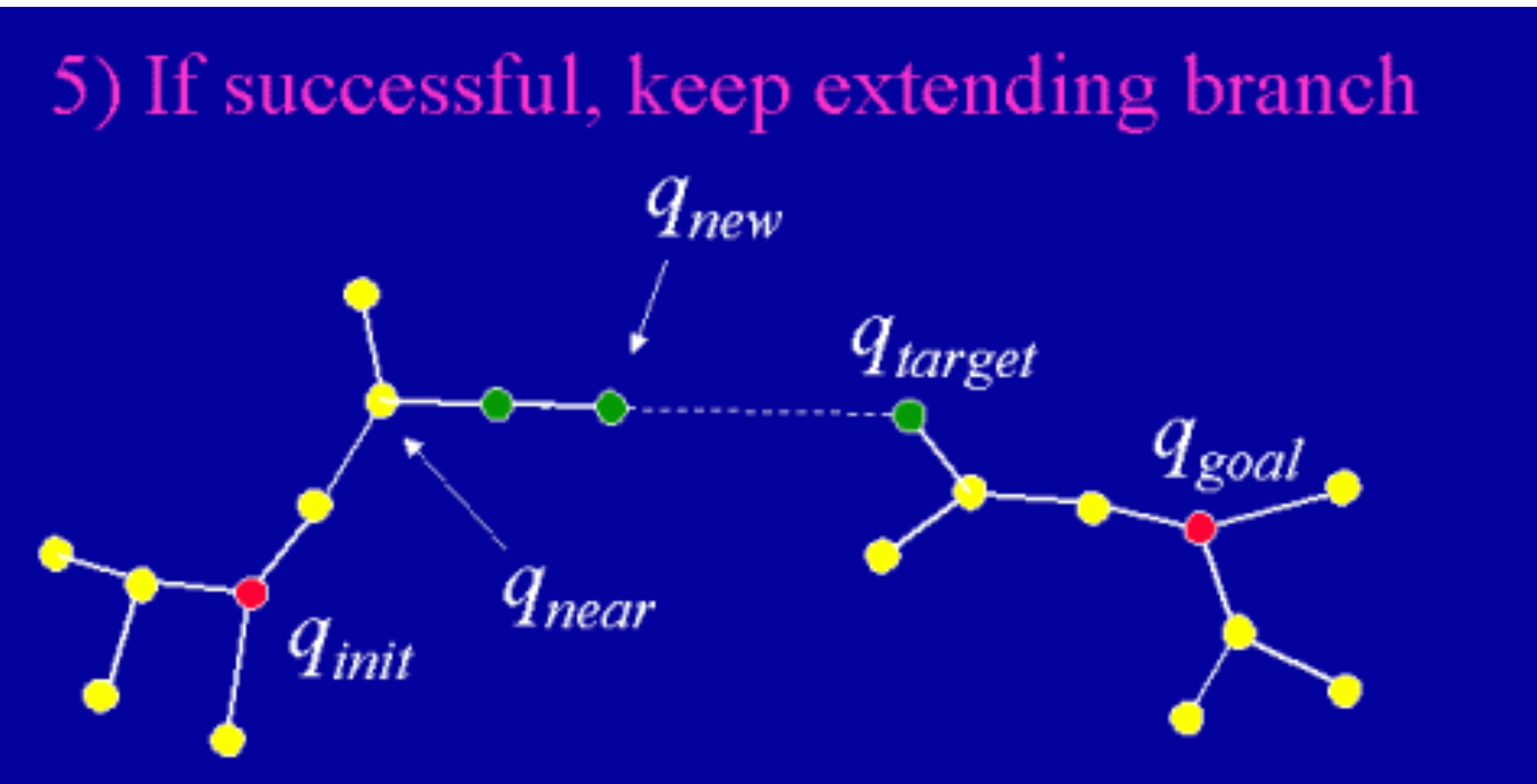
RRT-Connect animation



RRT-Connect animation



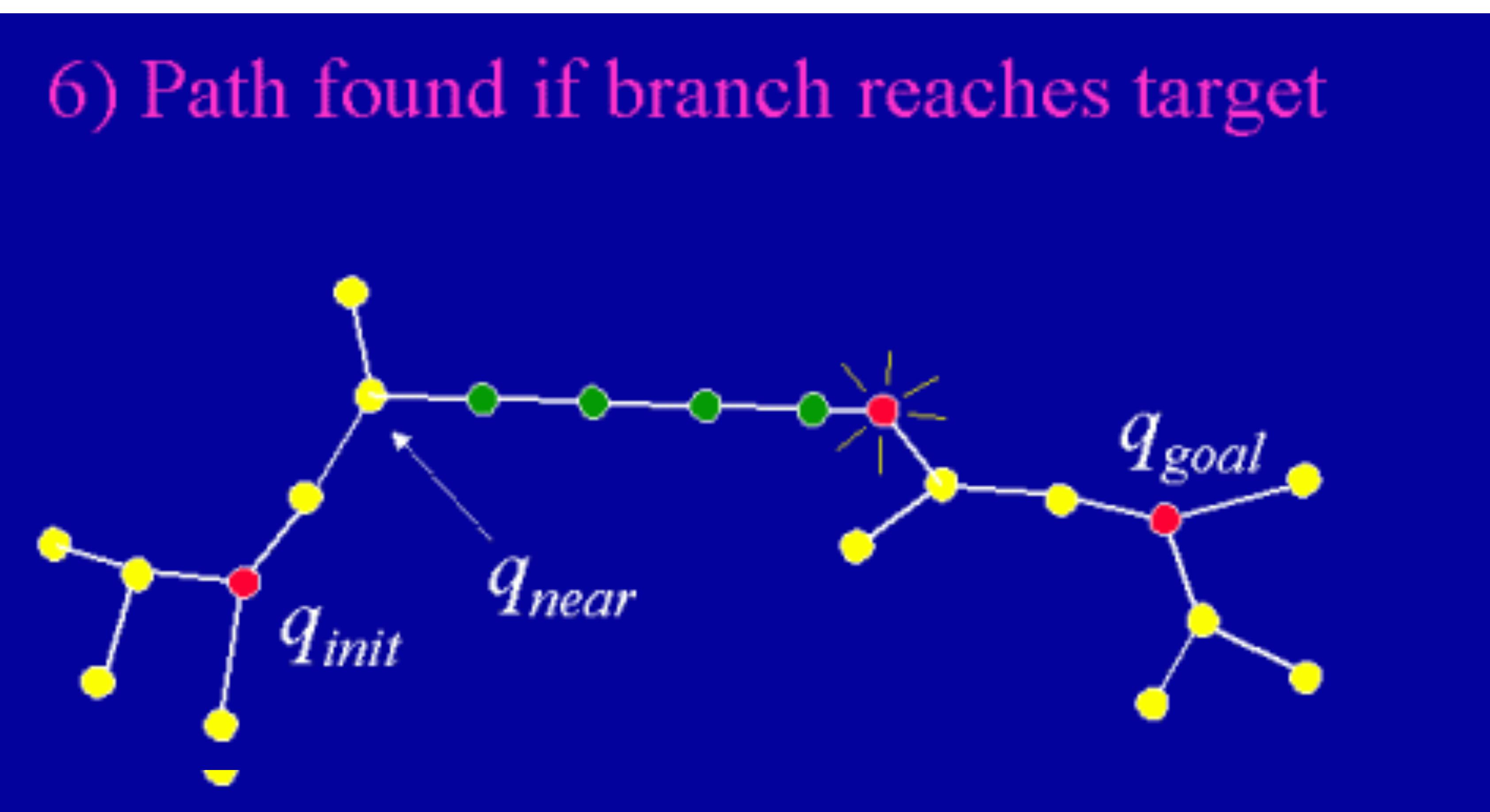
RRT-Connect animation



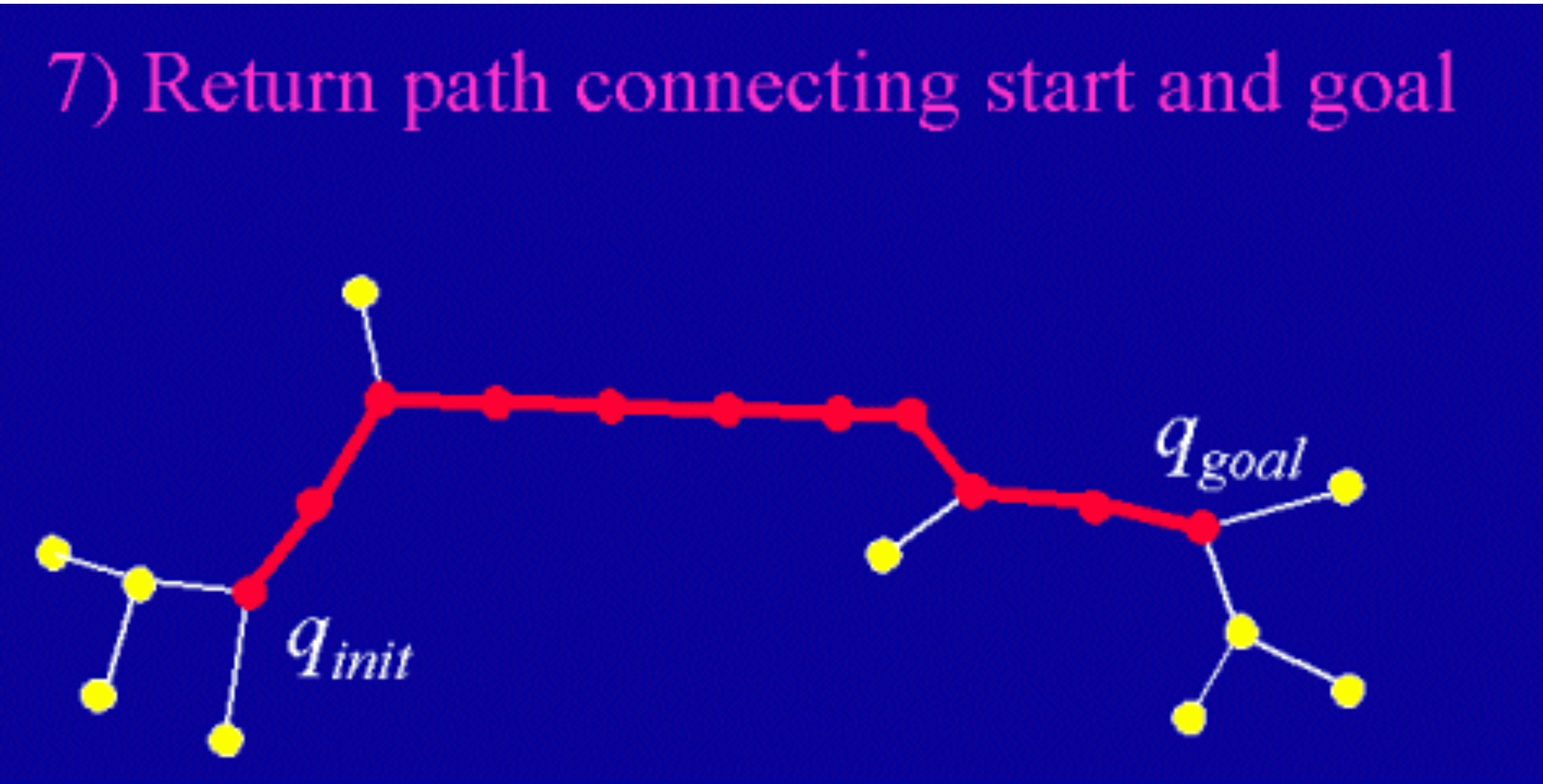
RRT-Connect animation



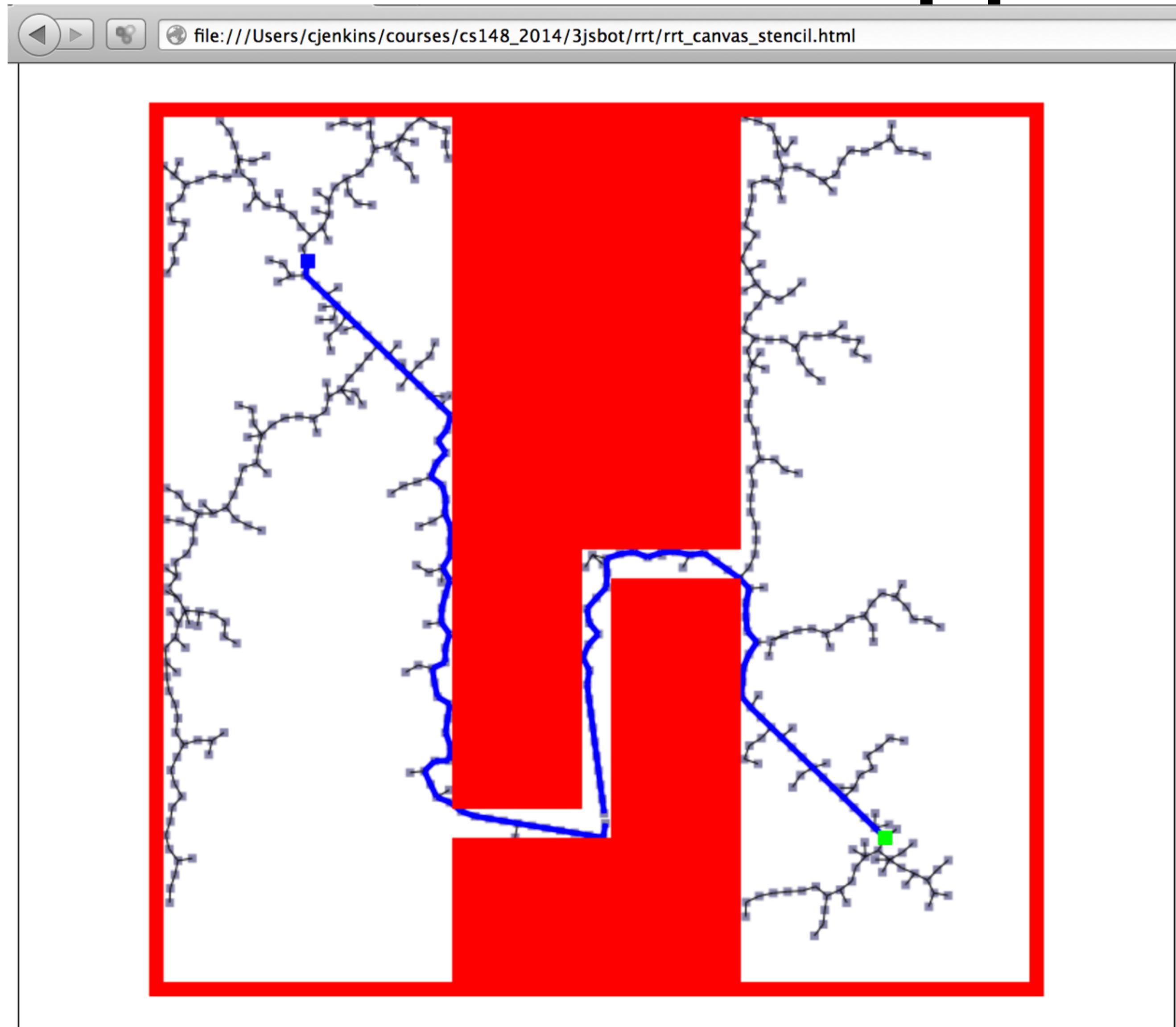
RRT-Connect animation



RRT-Connect animation

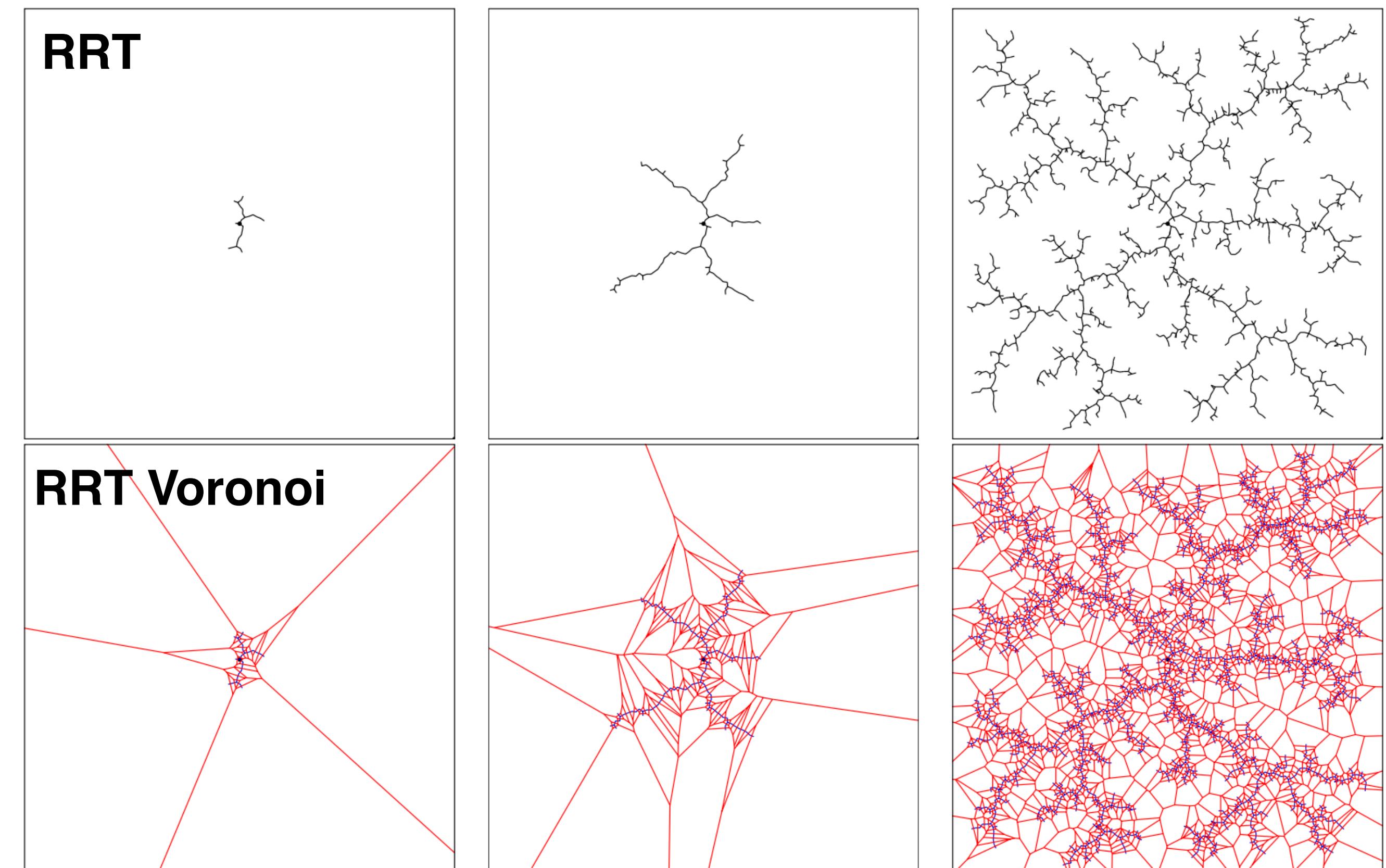


Let's see what happens

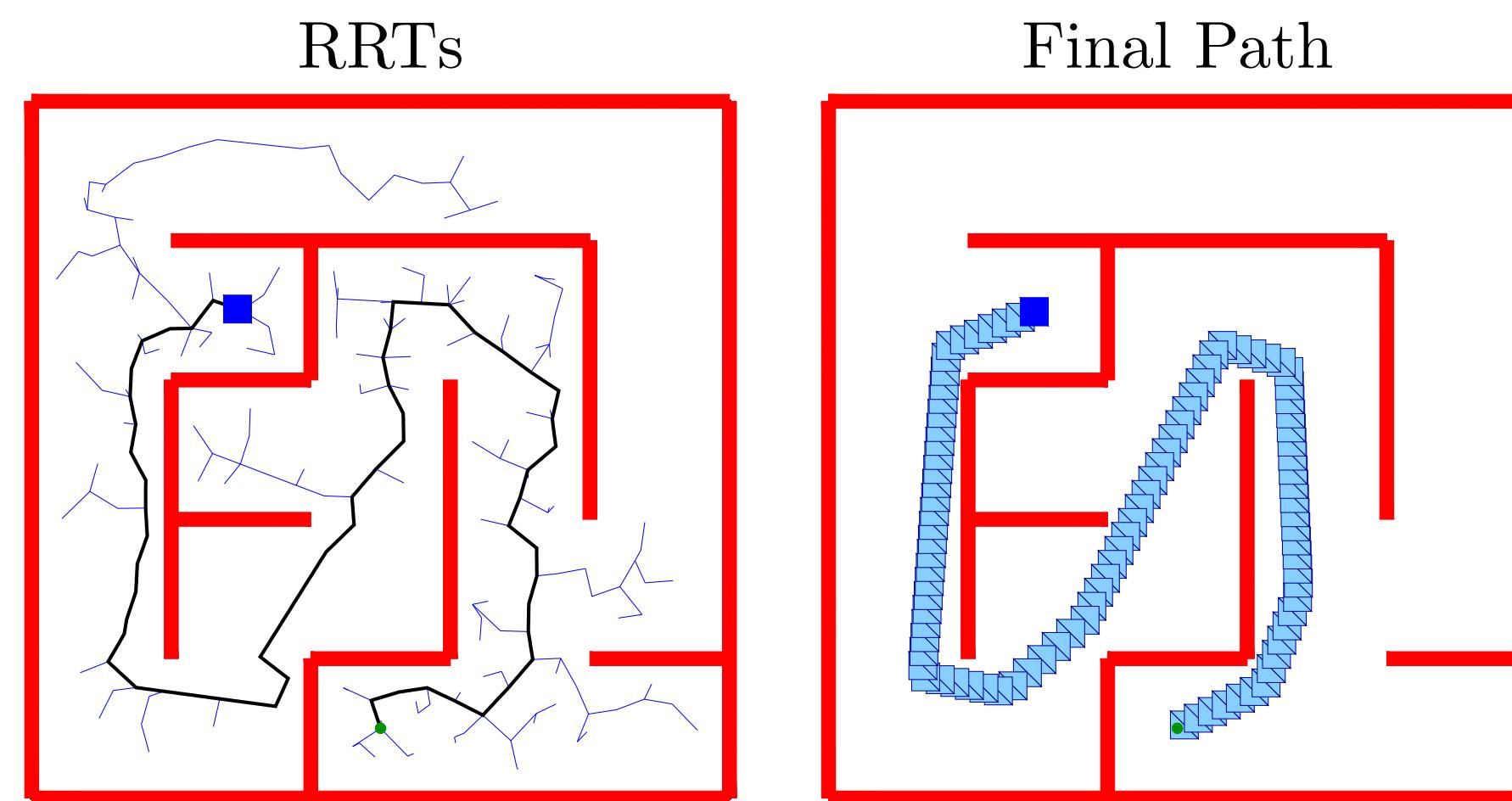


RRT Probabilistic Completeness

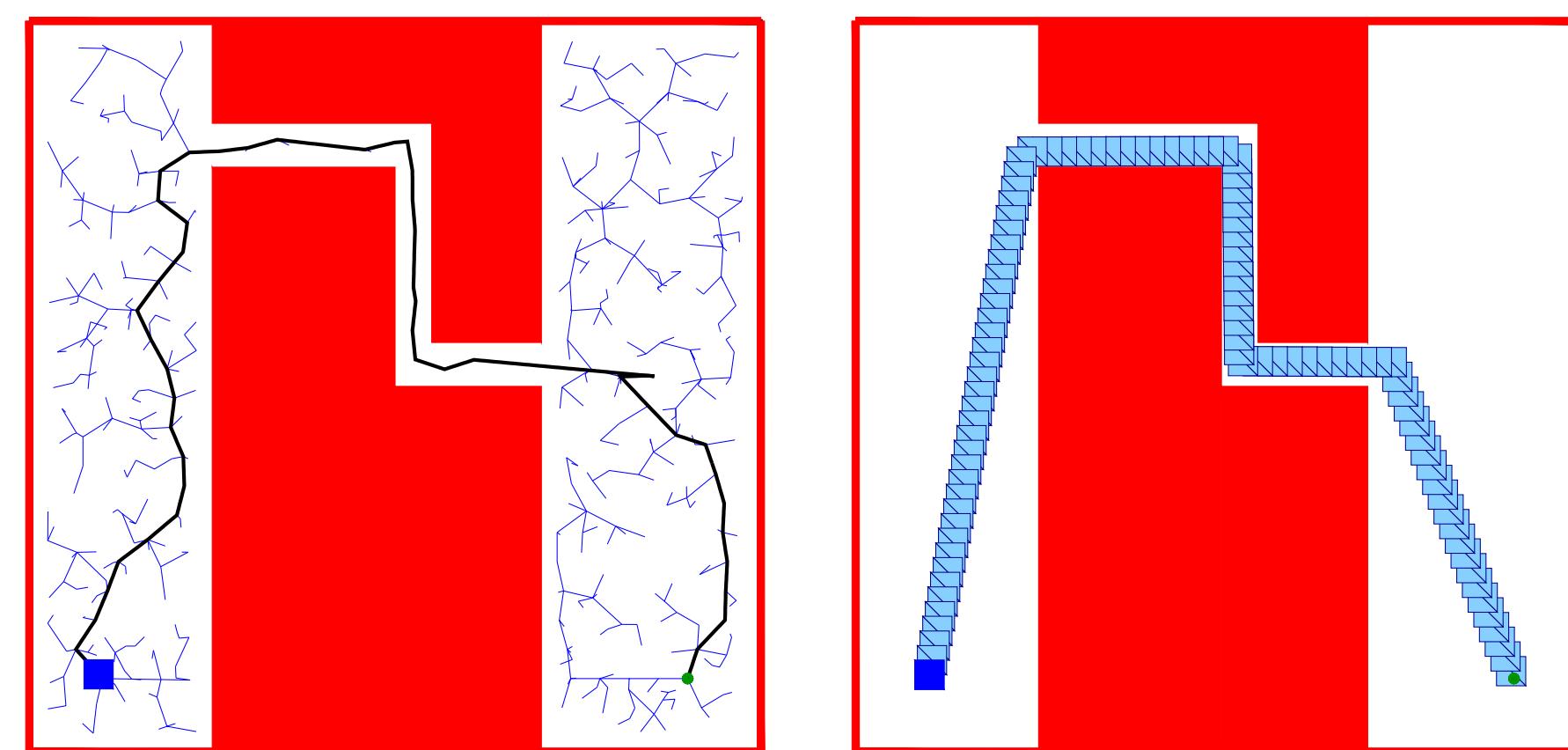
- Probability a vertex is selected for extension is proportional to its area in Voronoi diagram
- RRTs converge to a uniform coverage of C-space as the number of samples increases



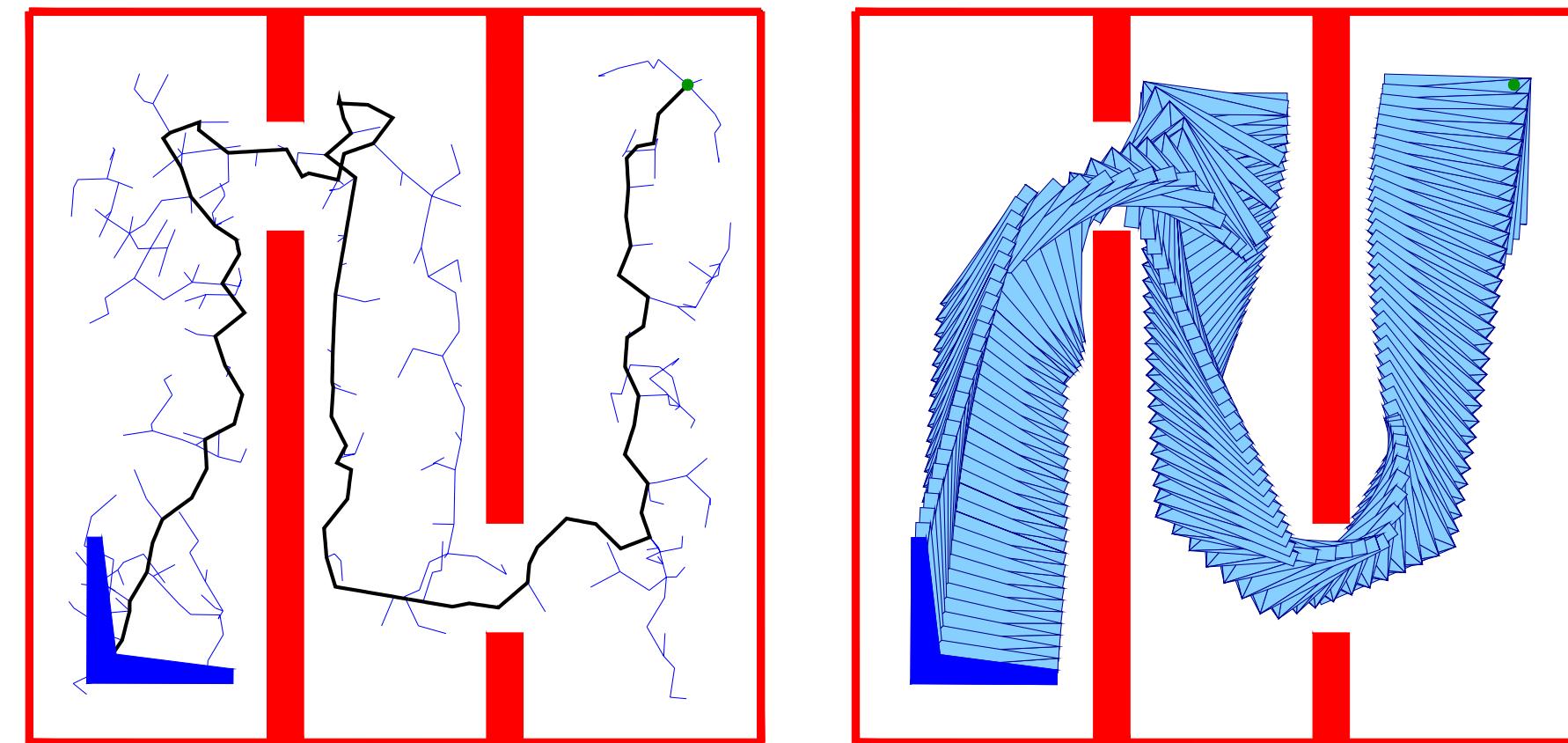
Examples of RRT-Connect



2 DOF maze

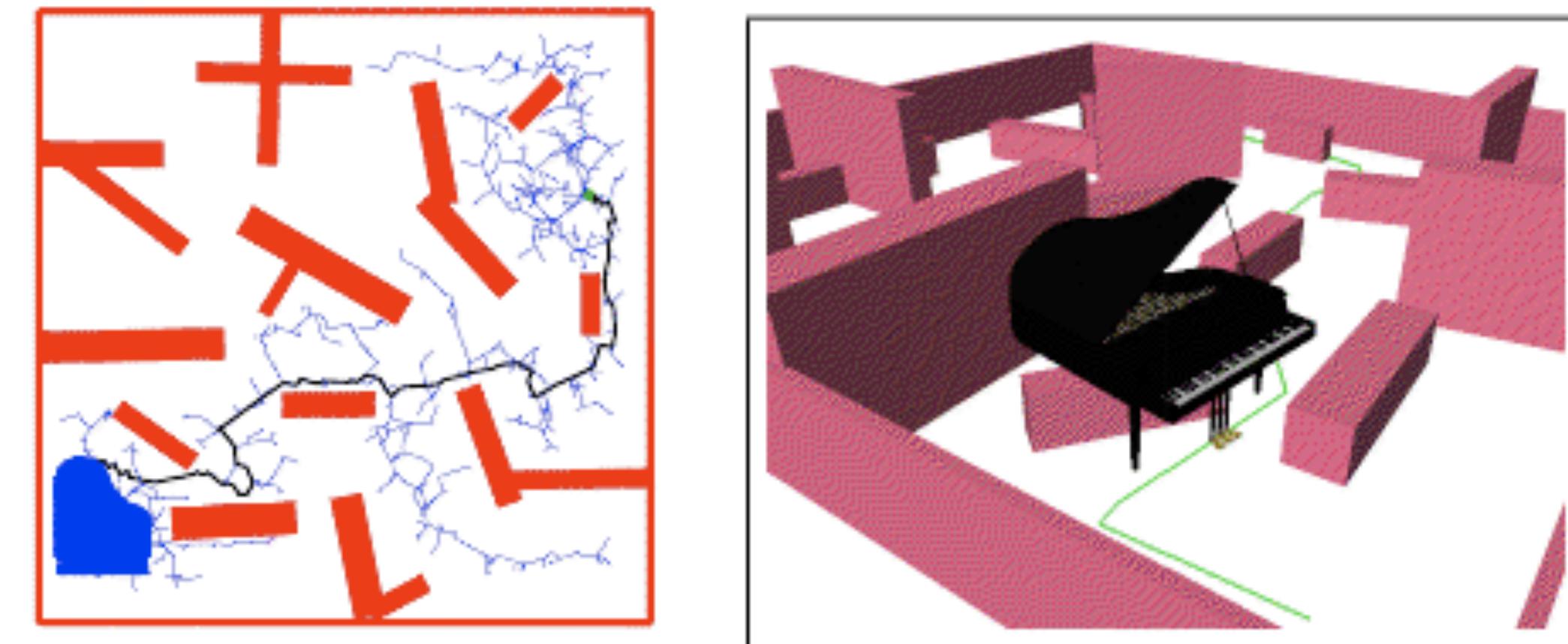


2 DOF single passway

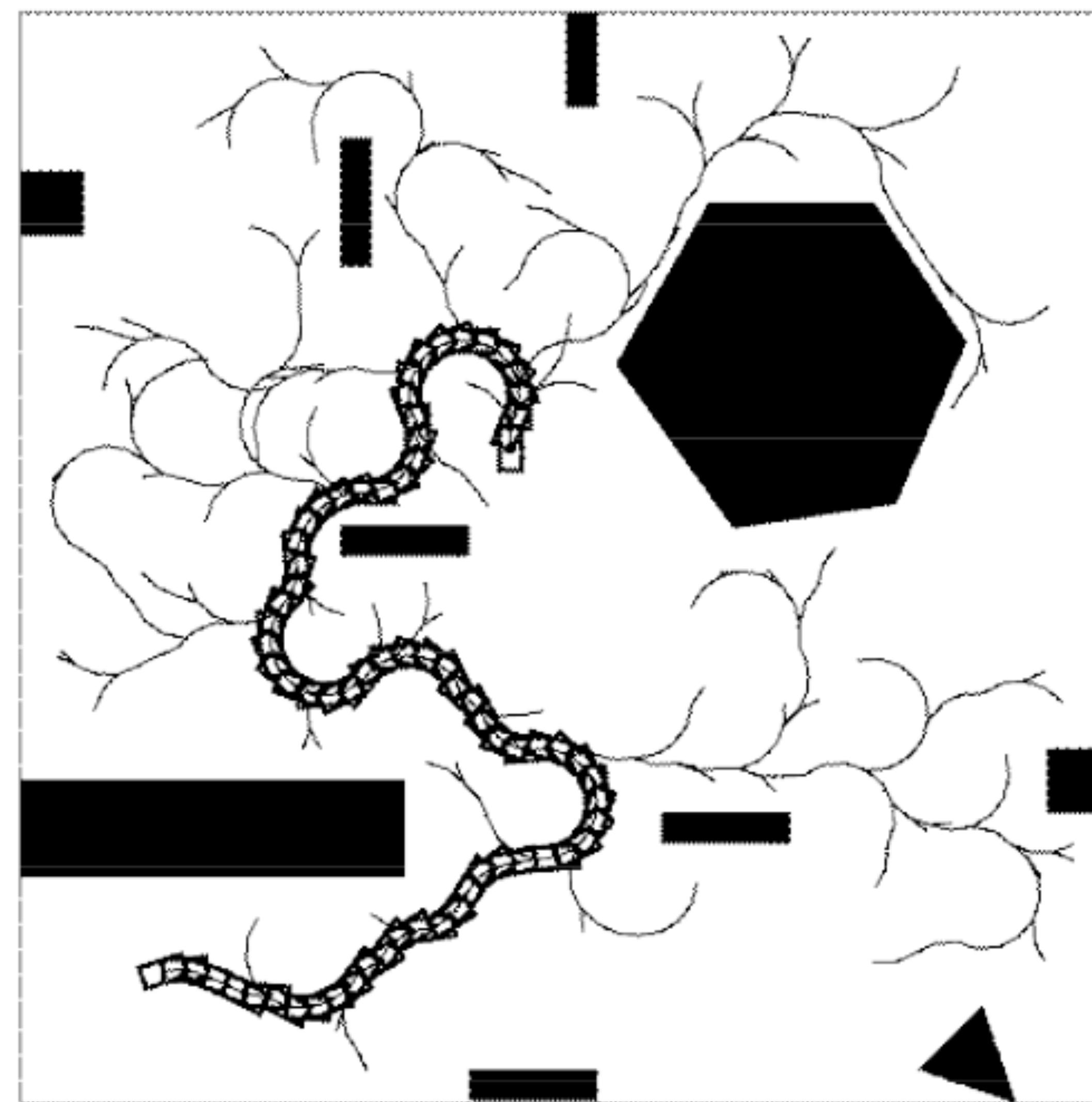


3 DOF single passway
(with non-point geometry)

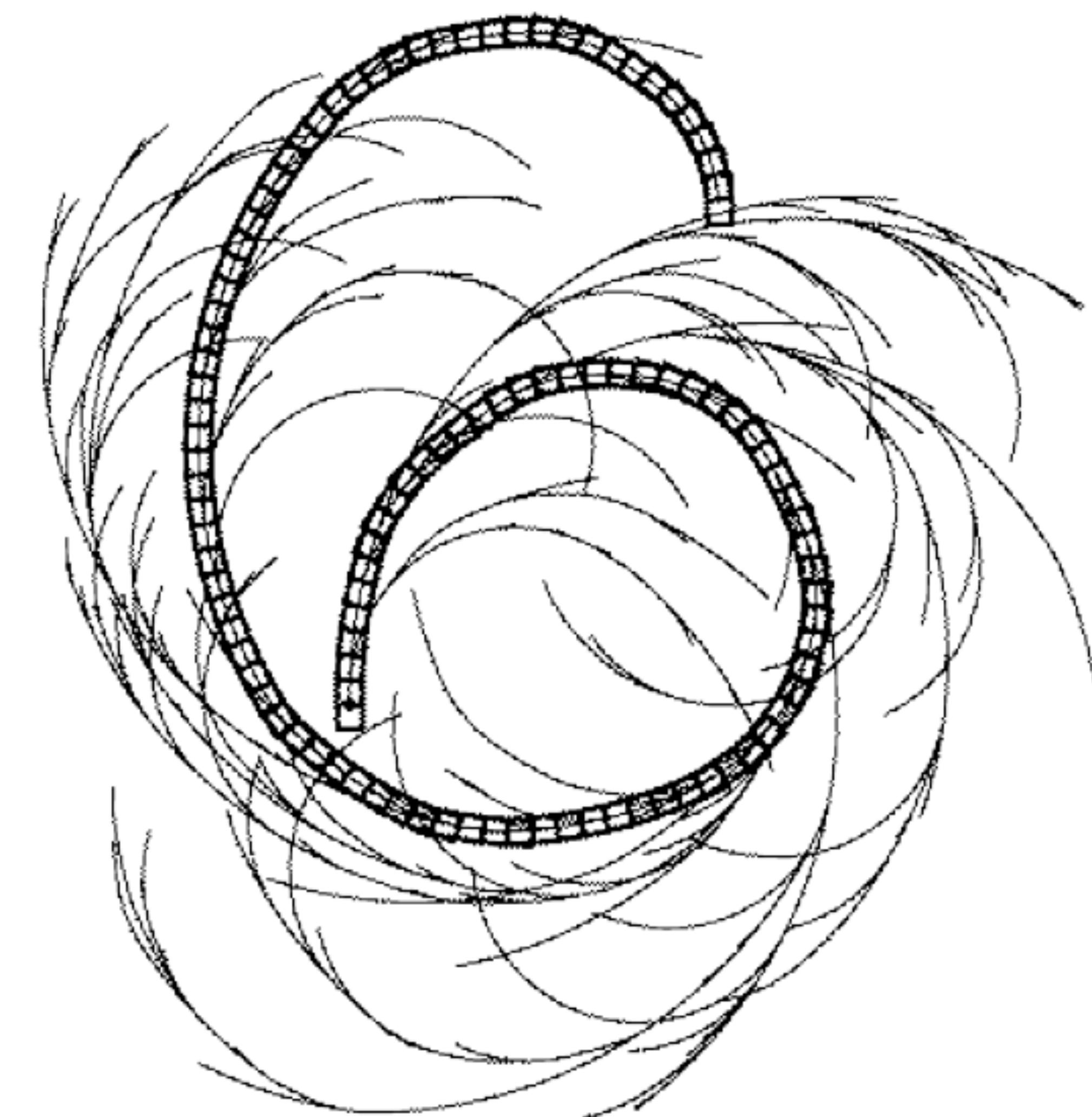
Piano Mover's Problem



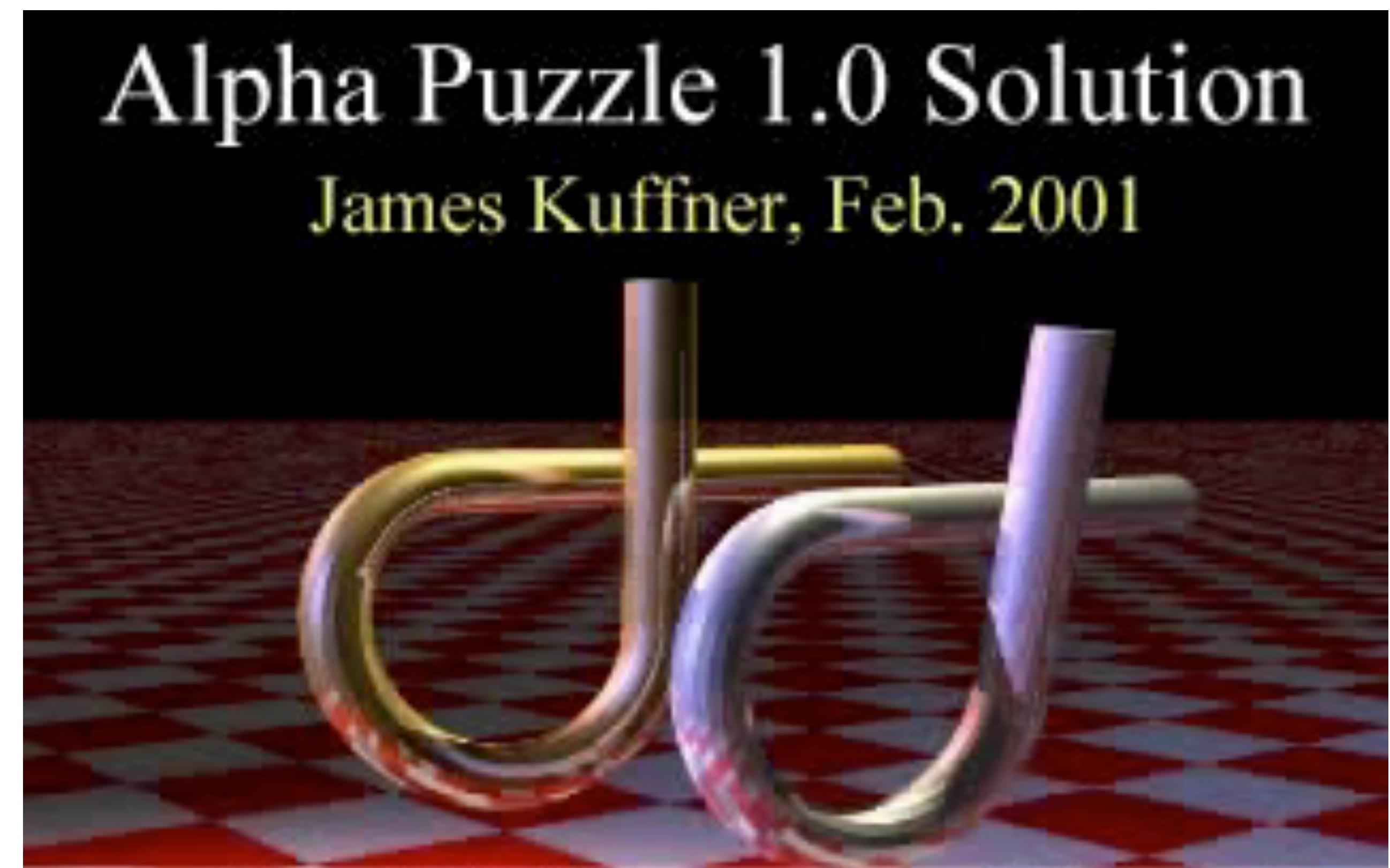
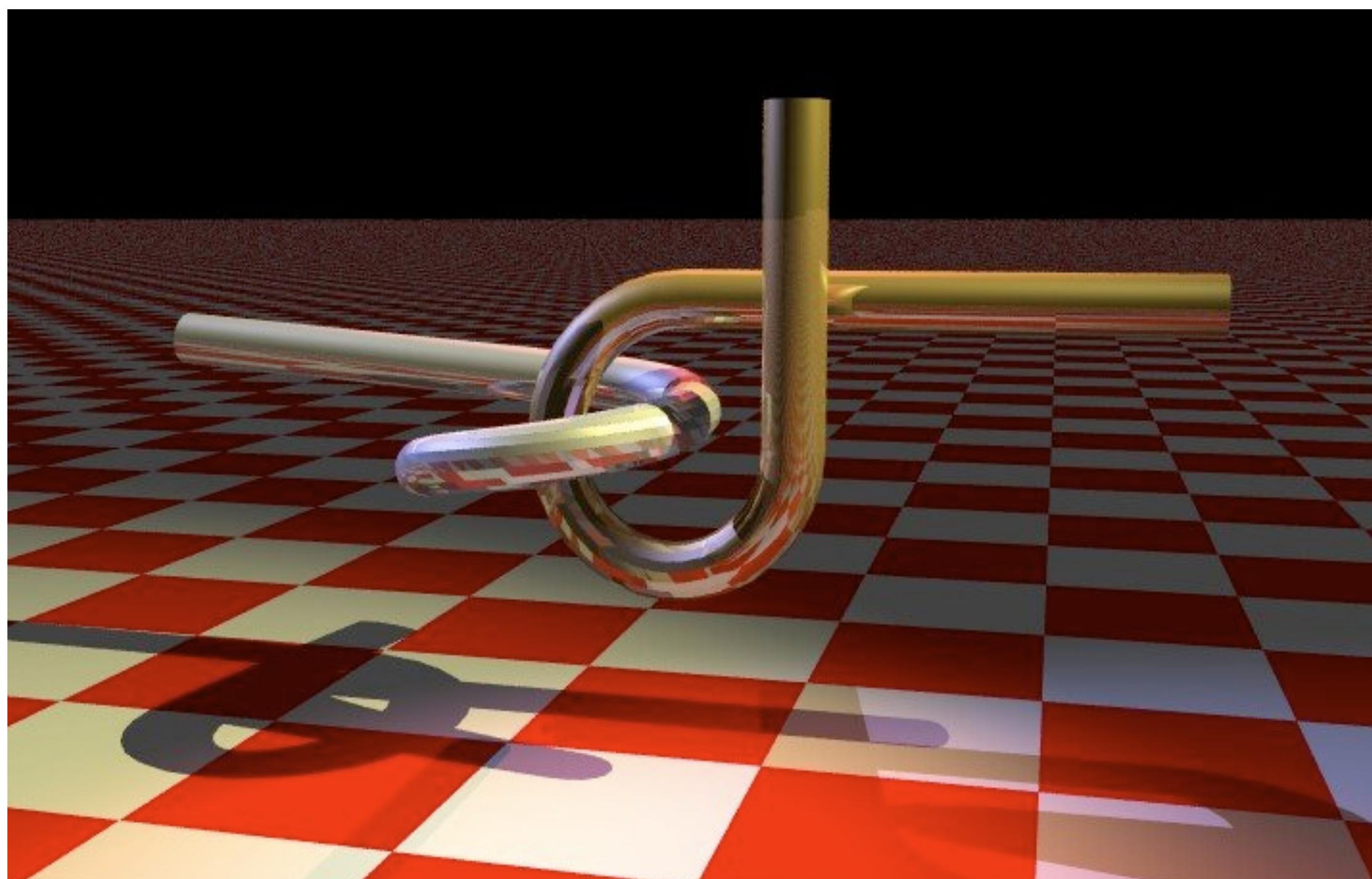
A Car-Like Robot



A Right-Turn Only Car



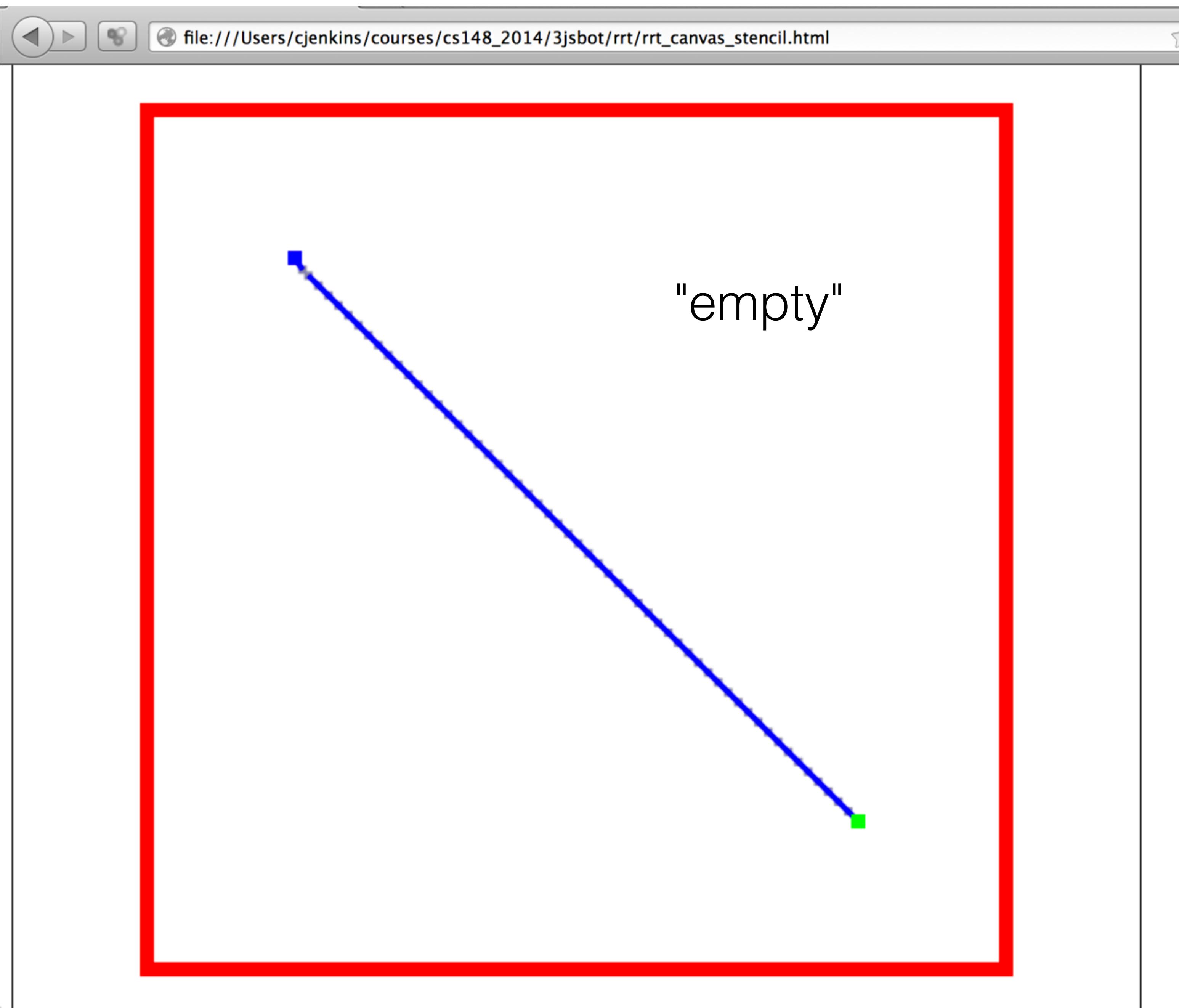
Alpha Puzzle

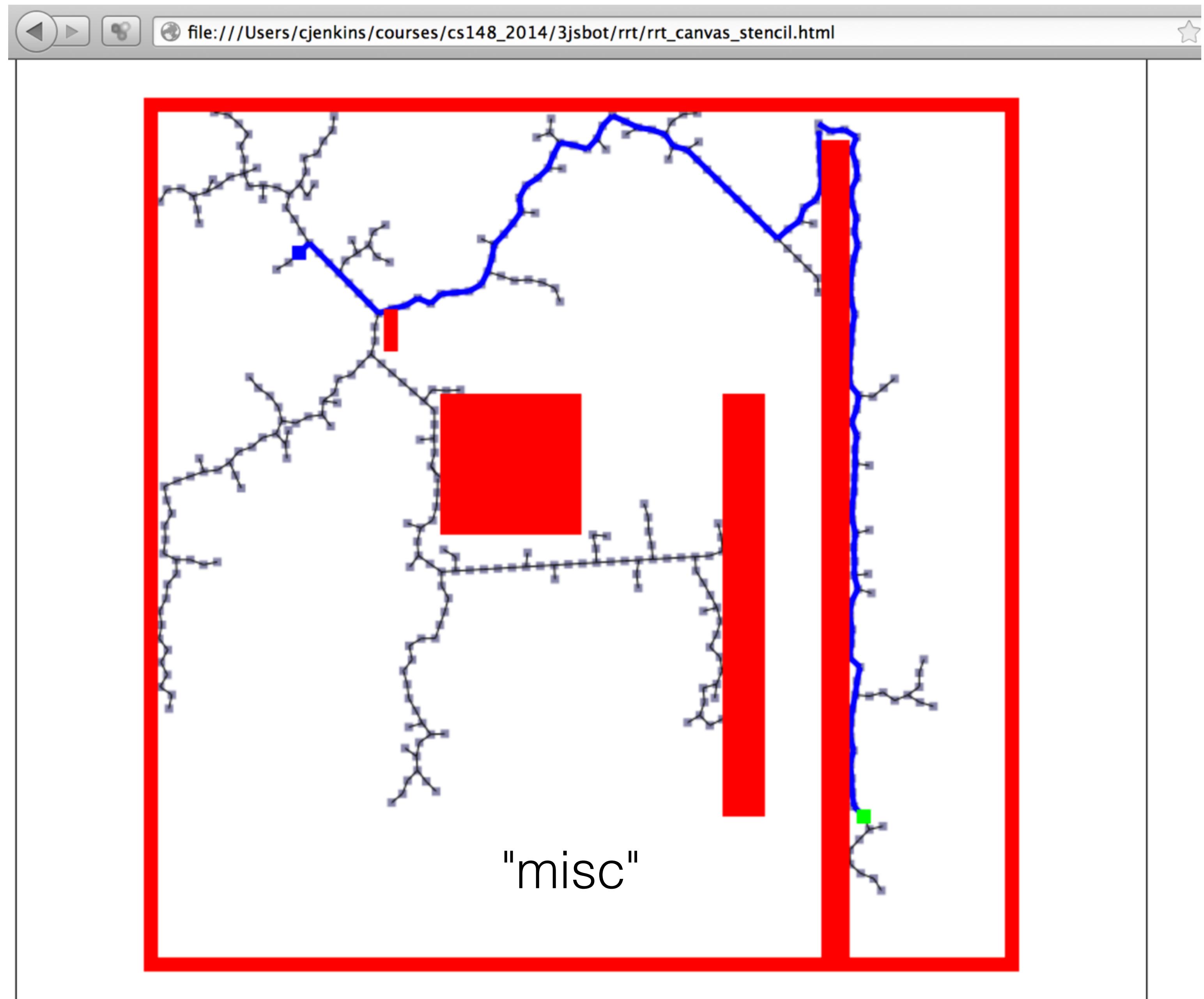


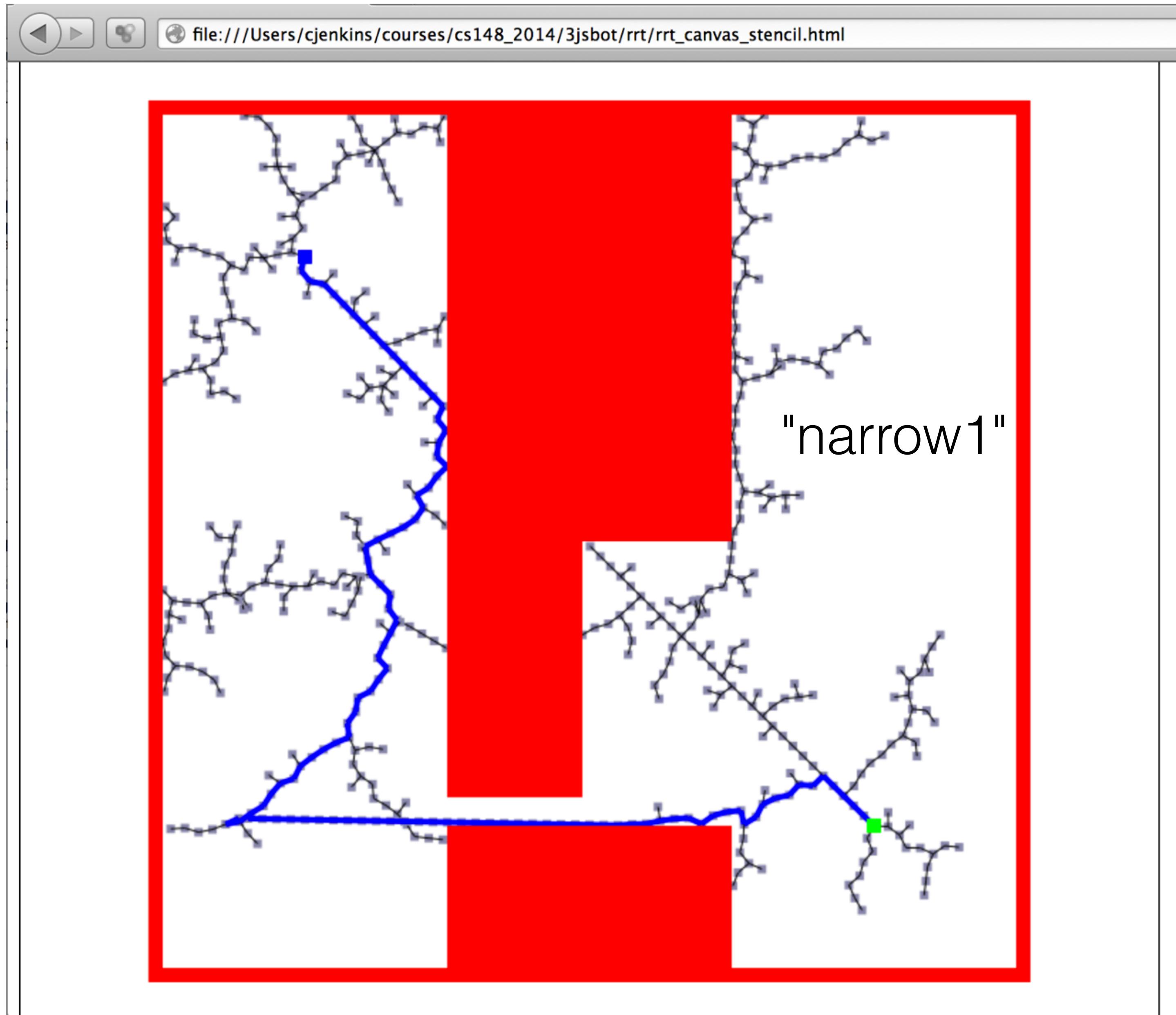
Alpha Puzzle 1.0 Solution
James Kuffner, Feb. 2001

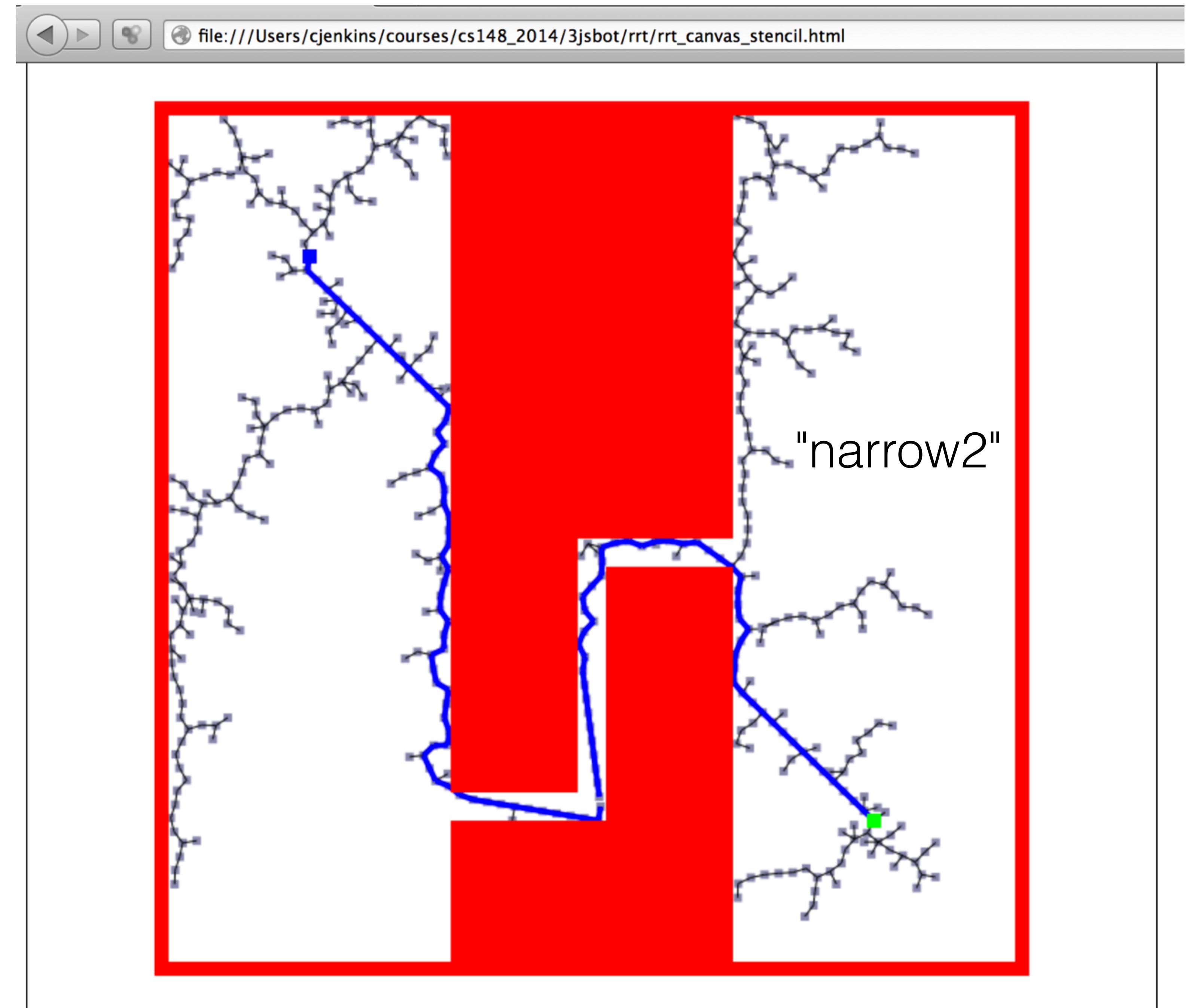
model by DSMFT group, Texas A&M Univ.
original model by Boris Yamrom, GE

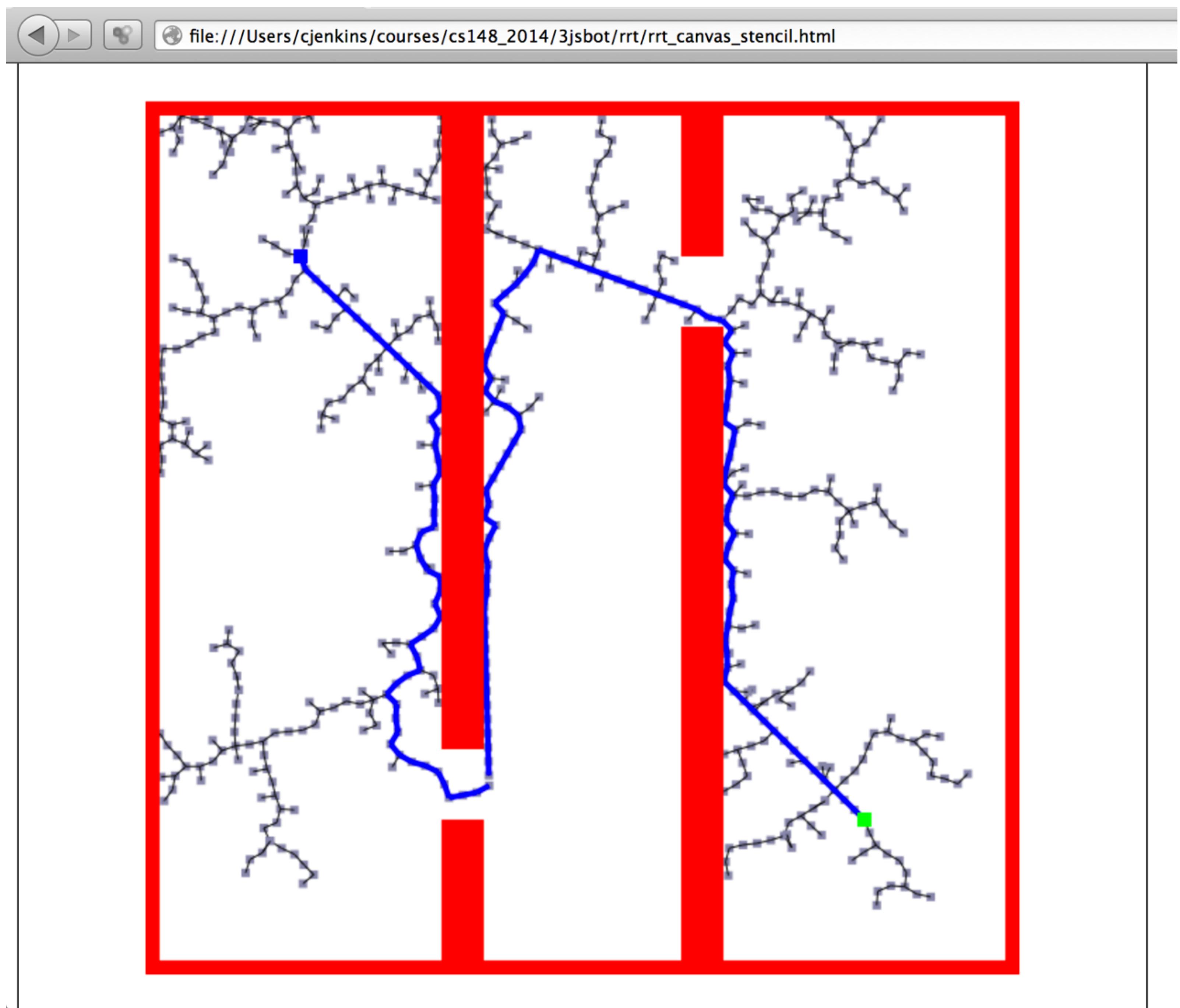
Canvas Stencil Examples





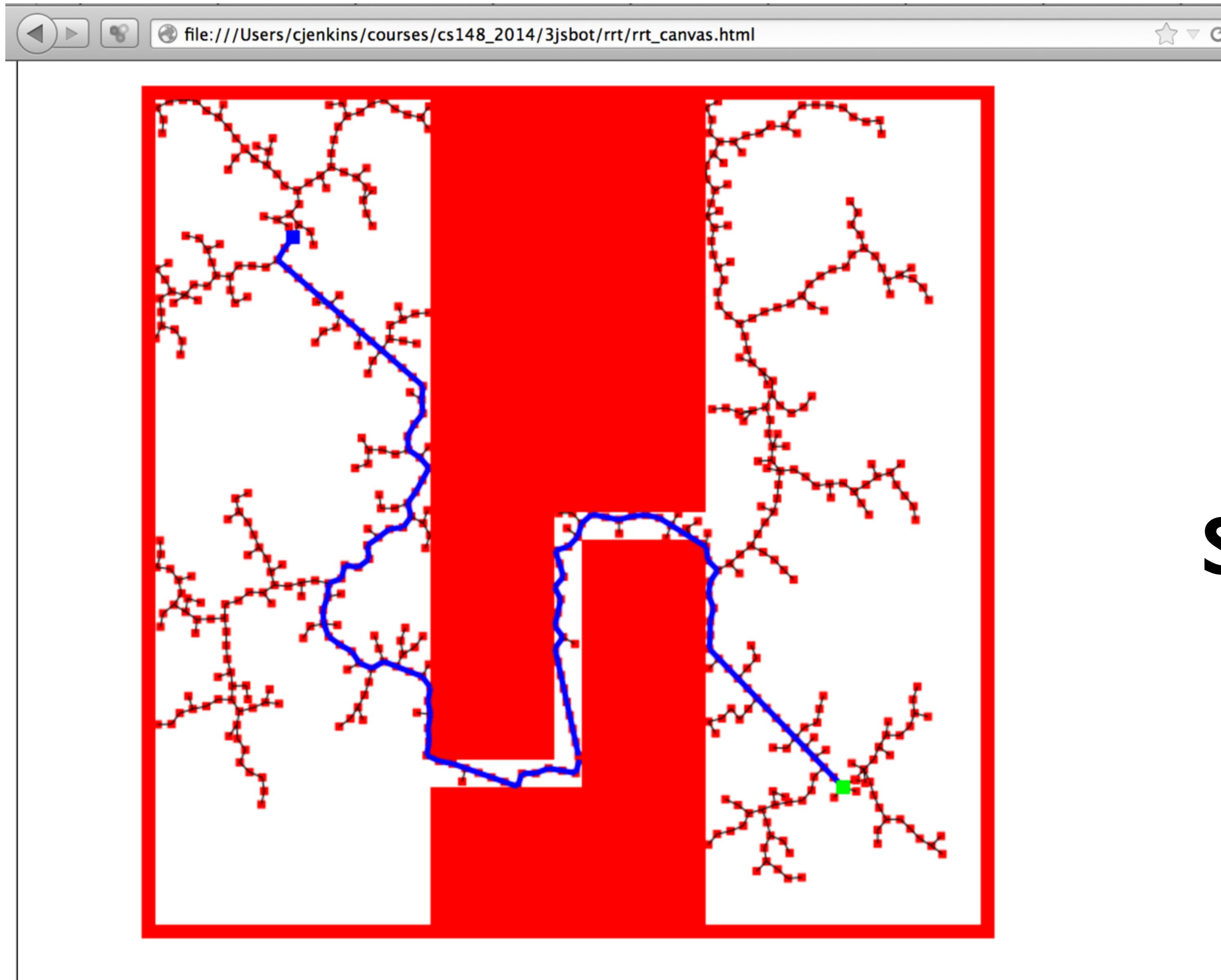






"three_sections"

RRTs can take a lot of time...



Is there a
simpler way?



Potential fields

follow your
potential