

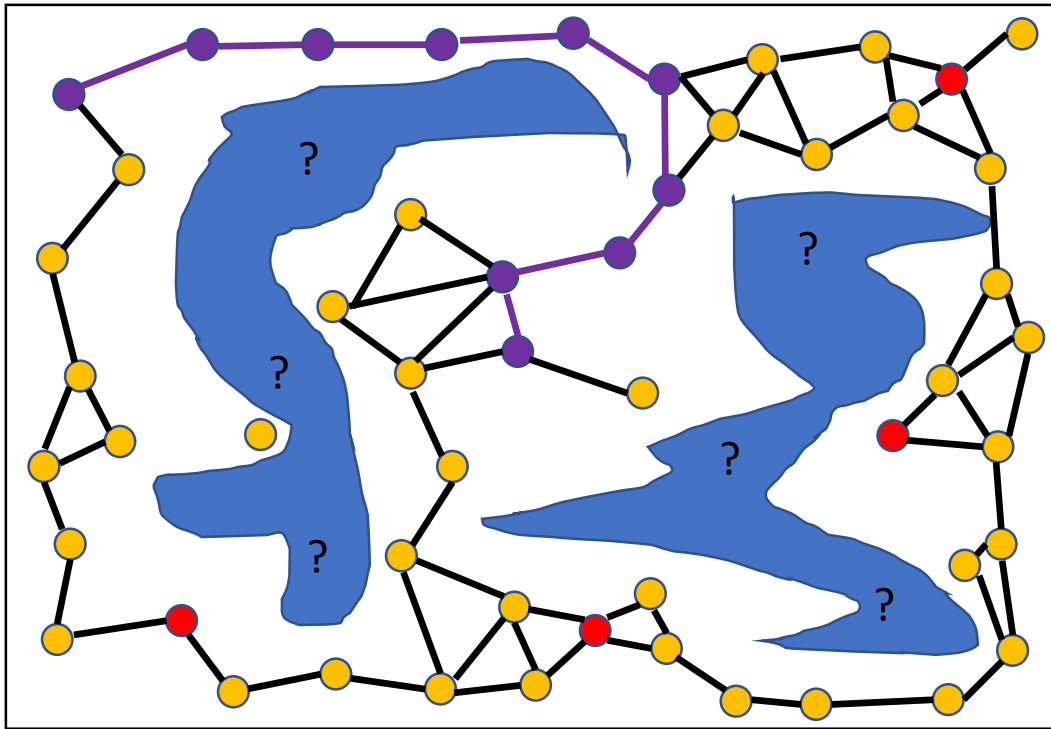
The single-query motion planners

Algorithms and Data Structures 2 – Motion Planning and its applications

University of Applied Sciences Stuttgart

Dr. Daniel Schneider

Motion Planning Algorithms



The **sPRM** is not the only motion planning algorithm.

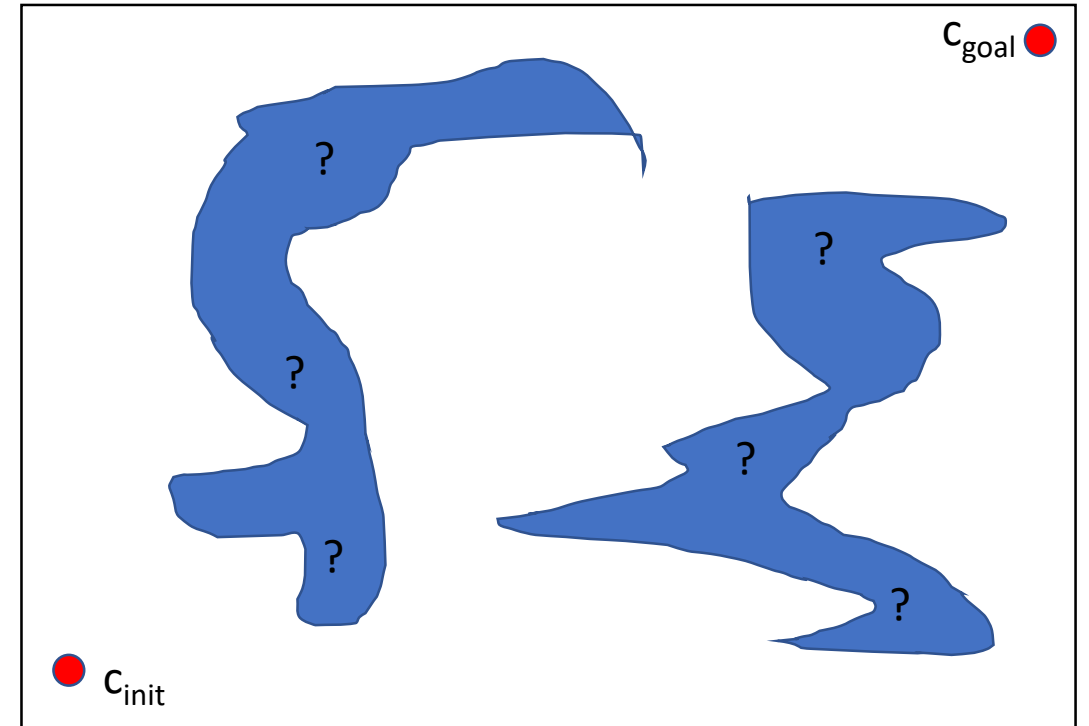
- The roadmap-based motion planner are used for multi-query motion planning.
- If you only need to compute a single path then there are better approaches.

→ These will be covered in this lection.

The Rapidly-Exploring Random Tree

Algorithm 1: RRT (c_{init} , c_{goal} , $range_{max}$, $time_{max}$)

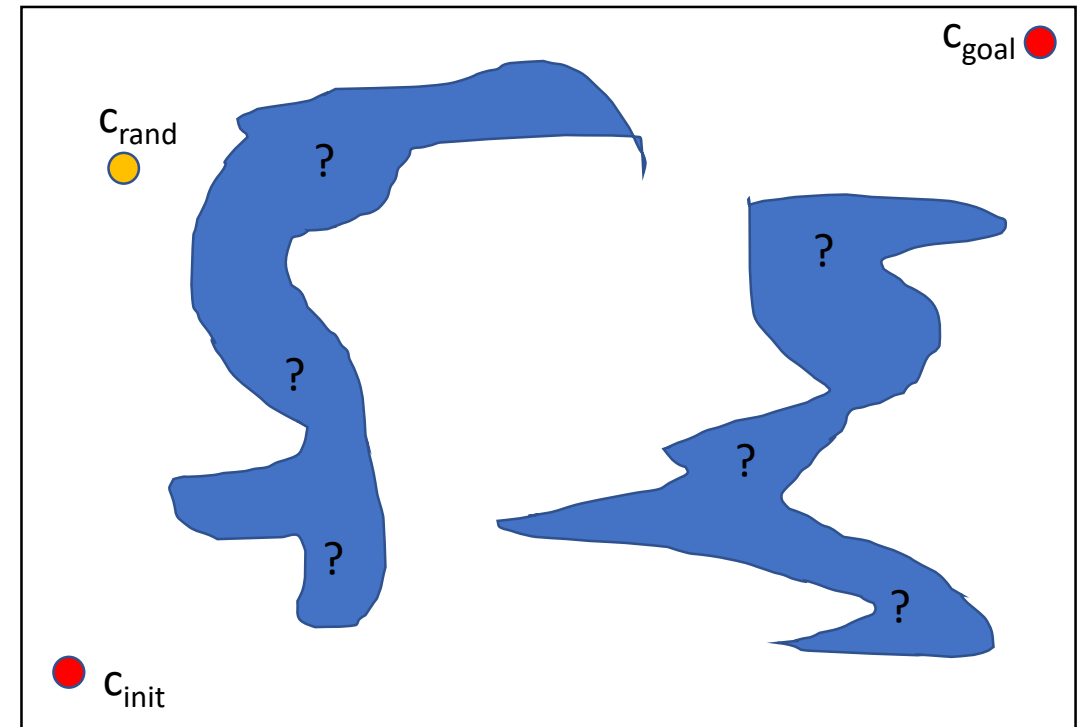
```
1   $T.init(c_{init})$ 
2
3  while ( $\neg TimeElapsed(T_{max})$ ) do
4       $c_{rand} \leftarrow RandomState()$ 
5       $c_{near} \leftarrow NearestN(c_{rand}, T)$ 
6       $c_{new} \leftarrow getValidEdge(c_{near}, c_{rand}, range_{max})$ 
7      if ( $\neg Trapped$ ) then
8           $T.addVertex(c_{new})$ 
9           $T.addEdge(c_{near}, c_{new})$ 
10         if  $goalReached(c_{new}, c_{goal})$  then
11             return  $Path(T)$ 
12
13  return  $ApproxPath(T)$ 
```



The Rapidly-Exploring Random Tree

Algorithm 1: RRT (c_{init} , c_{goal} , $range_{max}$, $time_{max}$)

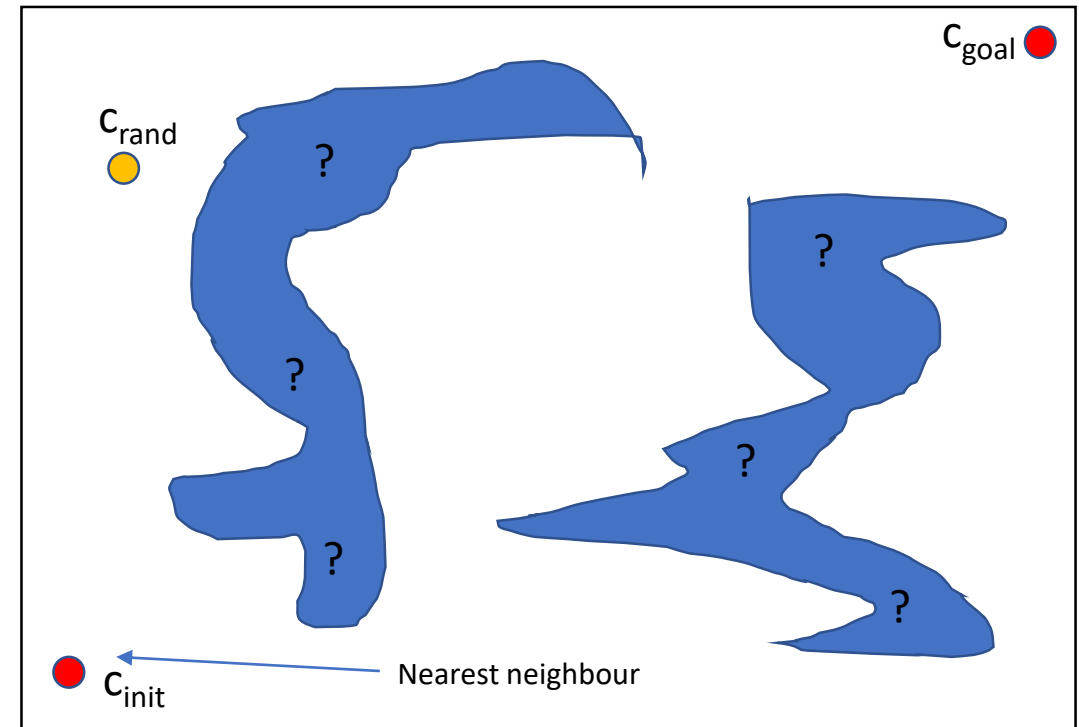
$T.init(c_{init})$	1
while ($\neg TimeElapsed(T_{max})$) do	2
$c_{rand} \leftarrow RandomState()$	3
$c_{near} \leftarrow NearestN(c_{rand}, T)$	4
$c_{new} \leftarrow getValidEdge(c_{near}, c_{rand}, range_{max})$	5
if ($\neg Trapped$) then	6
$T.addVertex(c_{new})$	7
$T.addEdge(c_{near}, c_{new})$	8
if $goalReached(c_{new}, c_{goal})$ then	9
return $Path(T)$	10
return $ApproxPath(T)$	11



The Rapidly-Exploring Random Tree

Algorithm 1: RRT (c_{init} , c_{goal} , $range_{max}$, $time_{max}$)

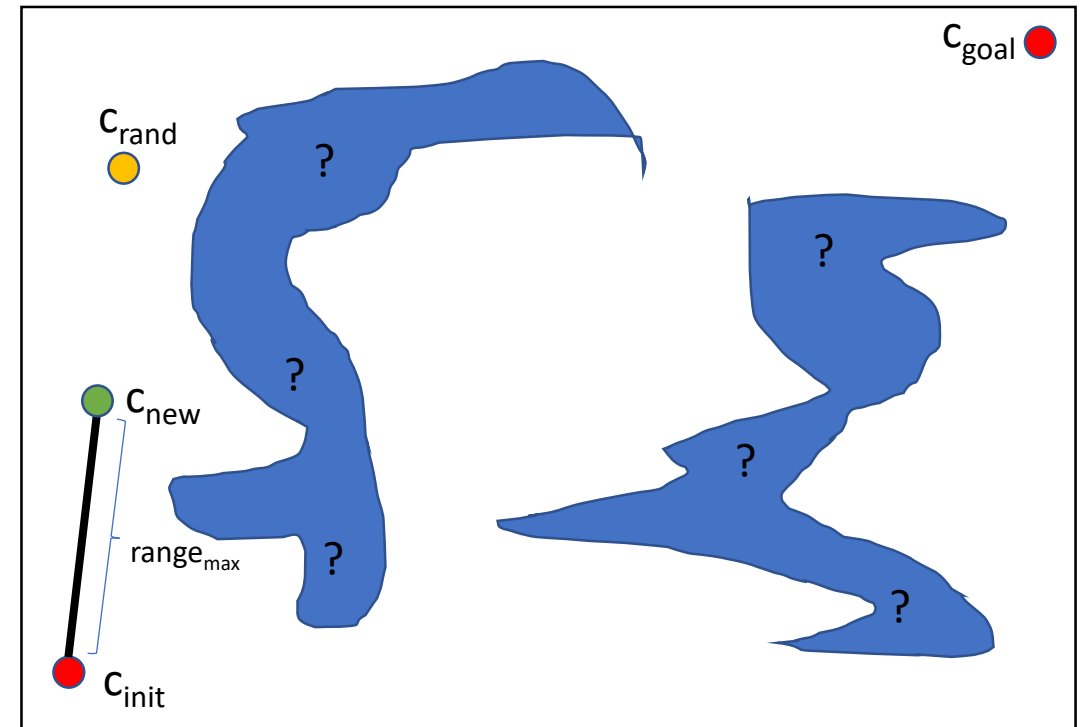
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2  while (! TimeElapsed( $T_{max}$ )) do  
3       $c_{rand} \leftarrow \text{RandomState}()$   
4       $c_{near} \leftarrow \text{NearestN}(c_{rand}, T)$   
5       $c_{new} \leftarrow \text{getValidEdge}(c_{near}, c_{rand}, range_{max})$   
6      if (! Trapped) then  
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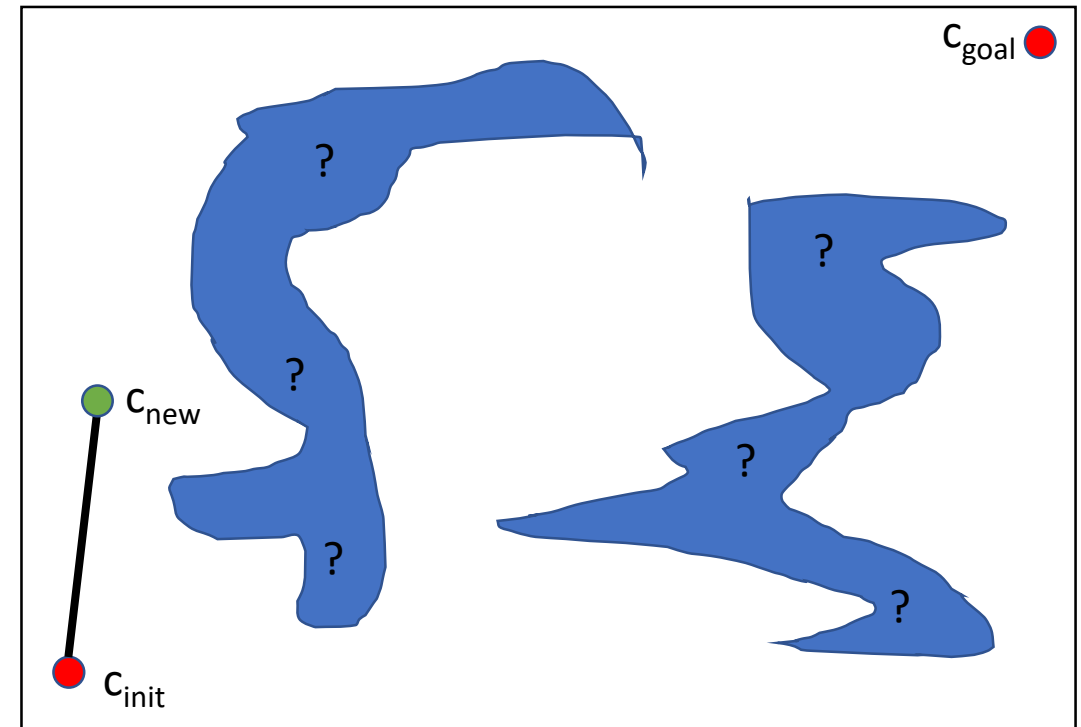
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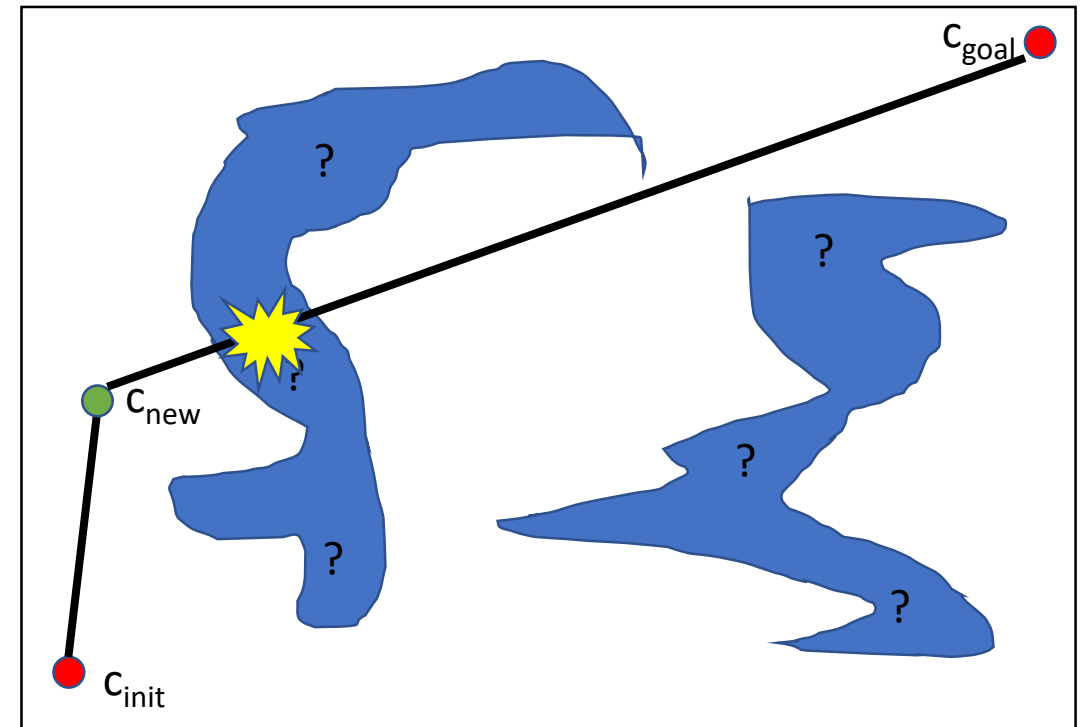


The Rapidly-Exploring Random Tree

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7      if (! Trapped) then
8          T.addVertex( $c_{new}$ )
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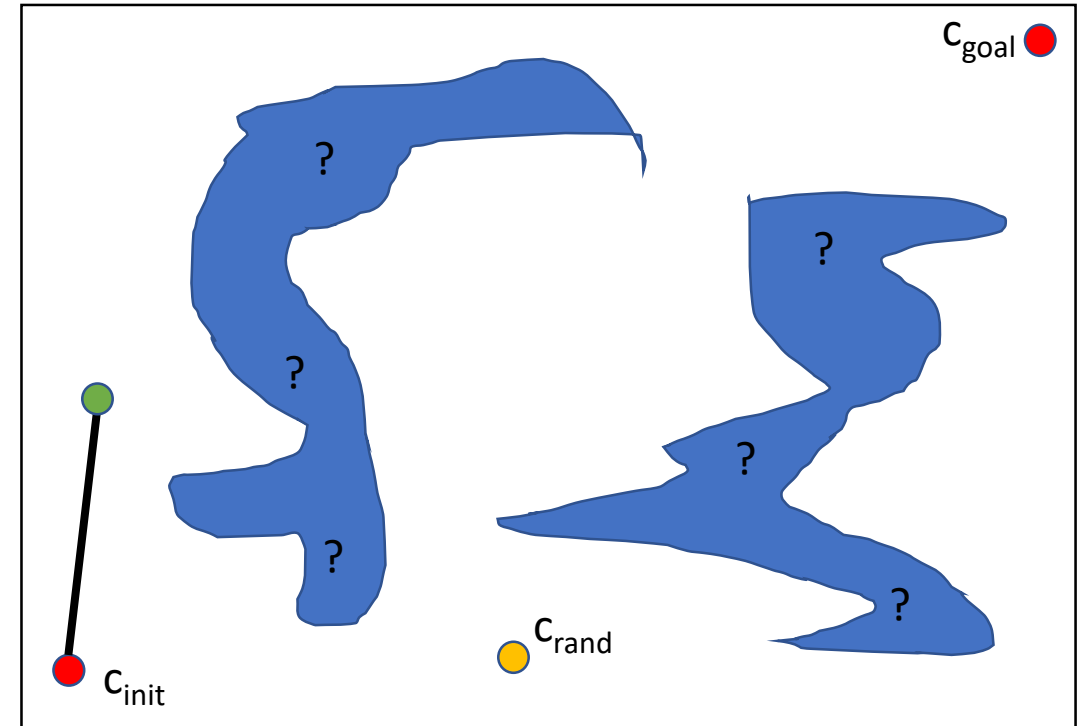
return $\text{ApproxPath}(T)$



The Rapidly-Exploring Random Tree

Algorithm 1: RRT (c_{init} , c_{goal} , $range_{max}$, $time_{max}$)

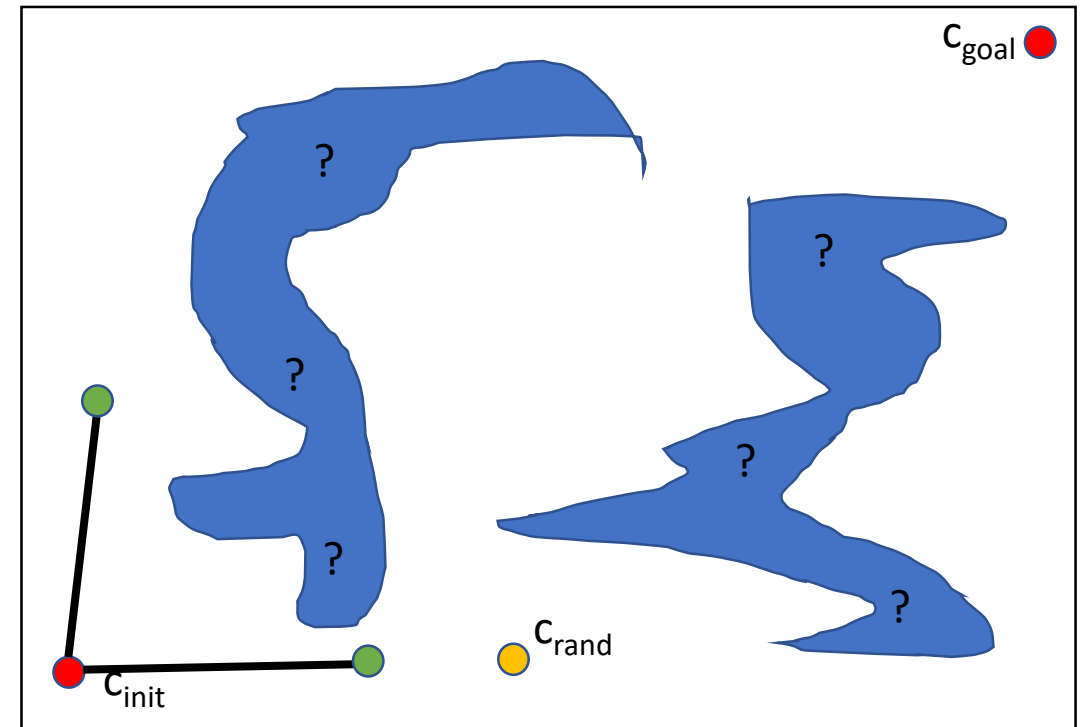
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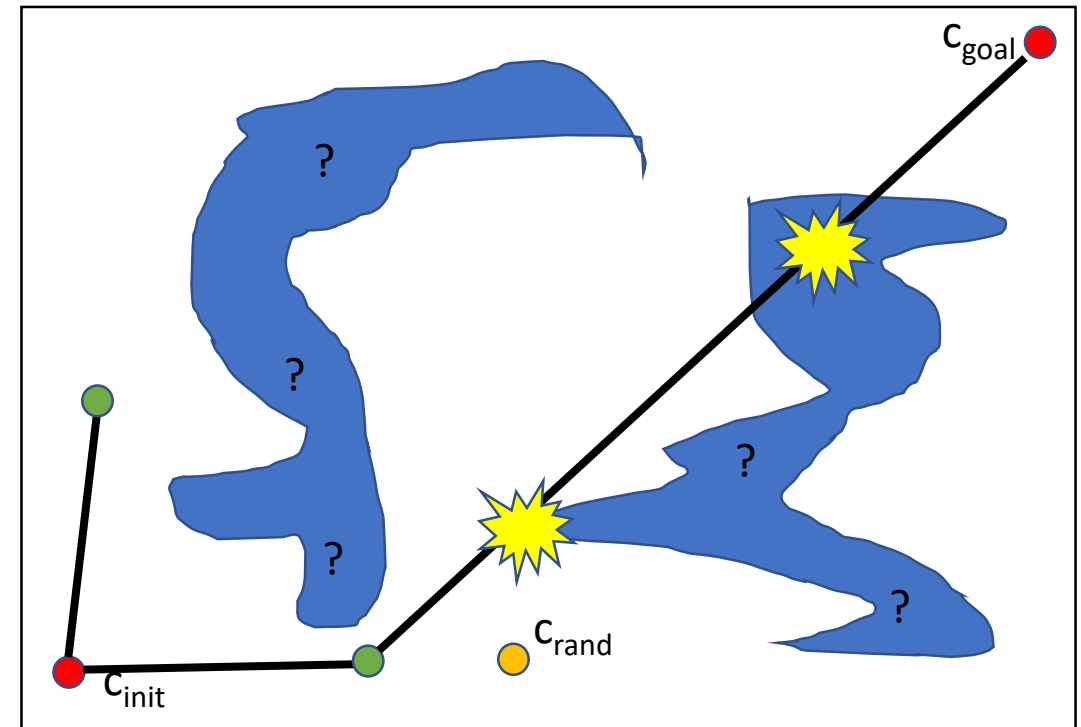
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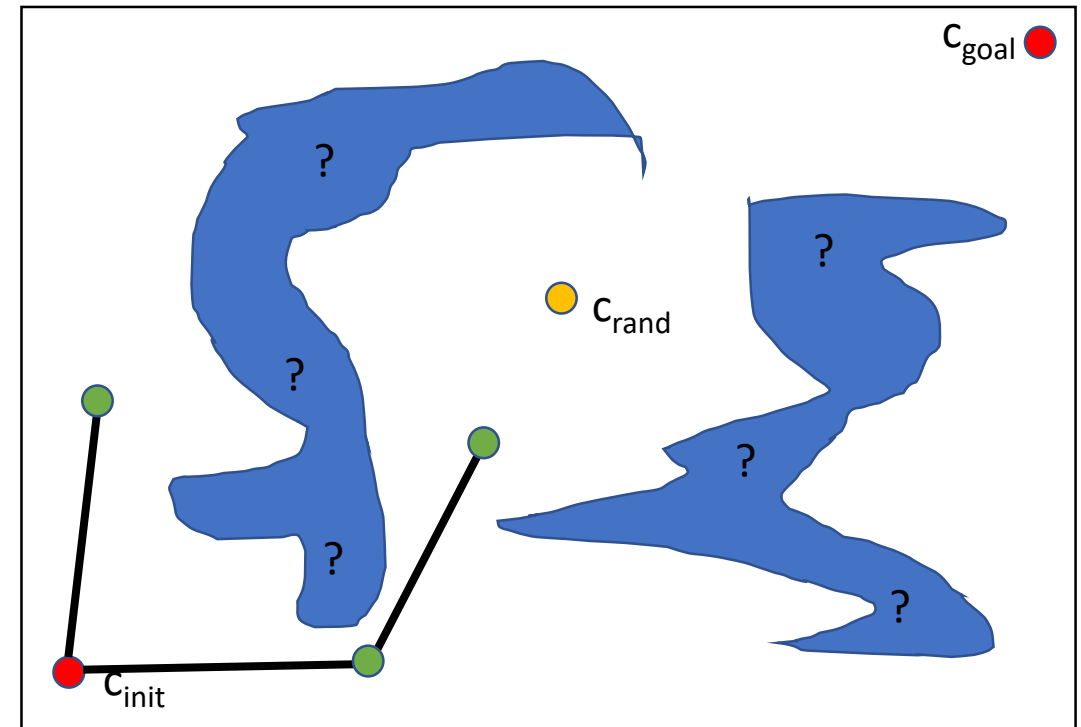
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Algorithm 1: RRT (c_{init} , c_{goal} , $range_{max}$, $time_{max}$)

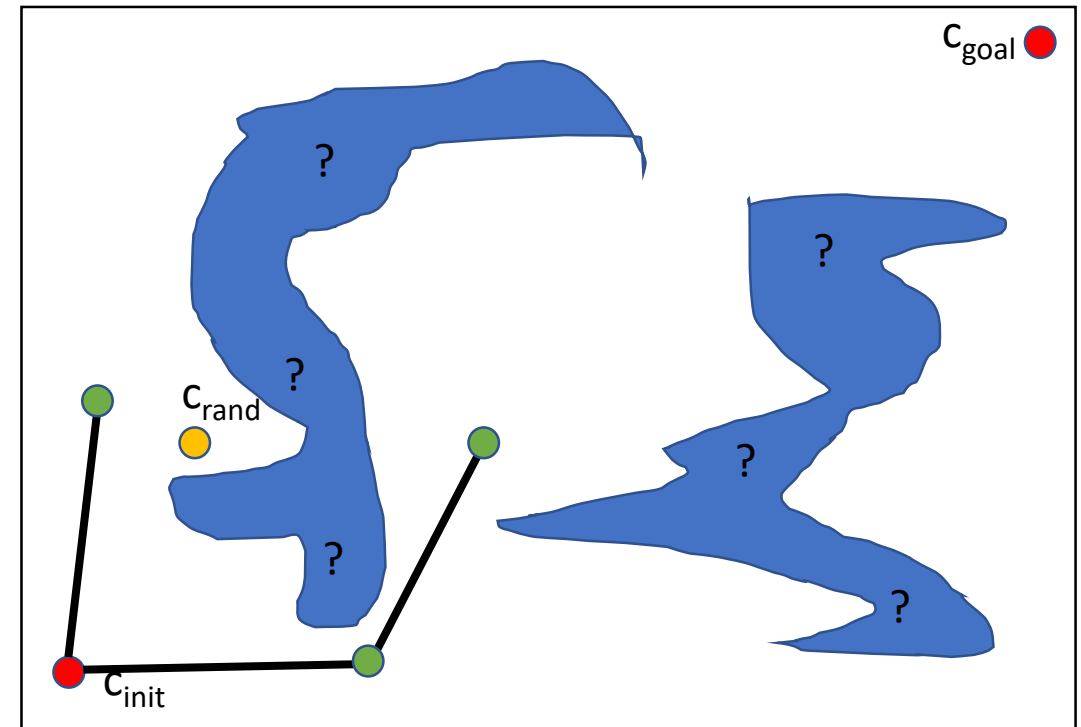
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1  T.init( $c_{init}$ )  
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```



The Rapidly-Exploring Random Tree

Algorithm 1: RRT ($c_{init}, c_{goal}, range_{max}, time_{max}$)
$$T.\text{init}(c_{init})$$

```

while (! TimeElapsed( $T_{max}$ )) do

```

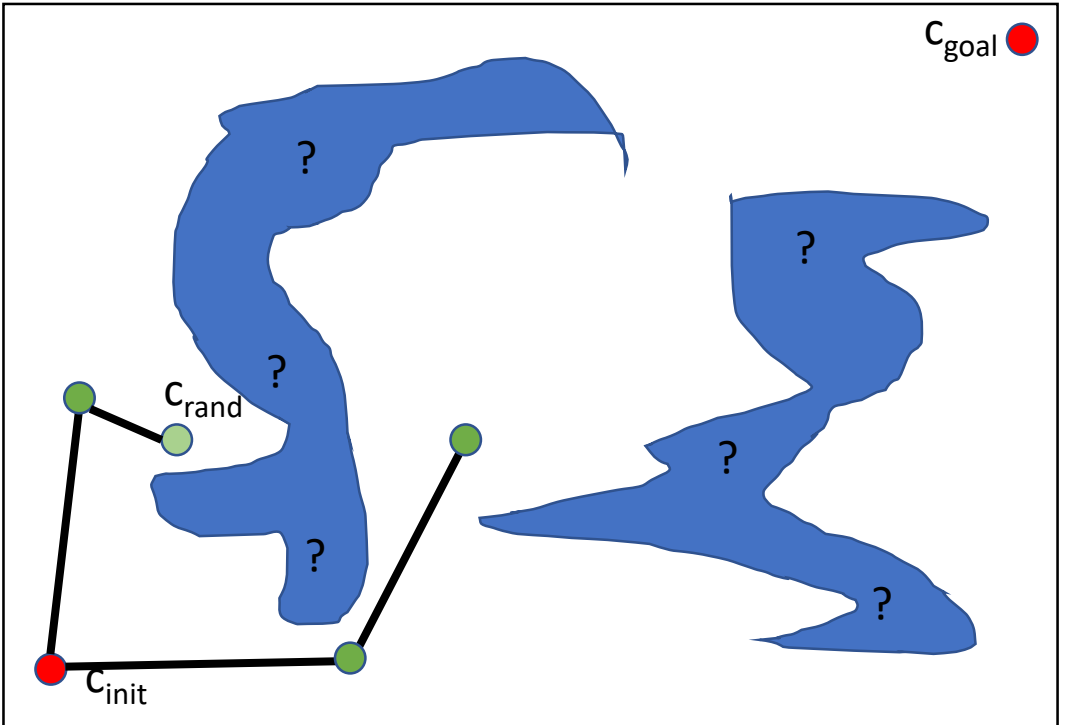
$$c_{rand} \leftarrow \text{RandomState}()$$
$$c_{near} \leftarrow \text{NearestN}(c_{rand}, T)$$
$$c_{new} \leftarrow \text{getValidEdge}(c_{near}, c_{rand}, range_{max})$$

if ($\neg Trapped$) **then**

$$T.addVertex(c_{new})$$
$$T.addEdge(c_{near}, c_{new})$$

if $goalReached(c_{new}, c_{goal})$ **then**

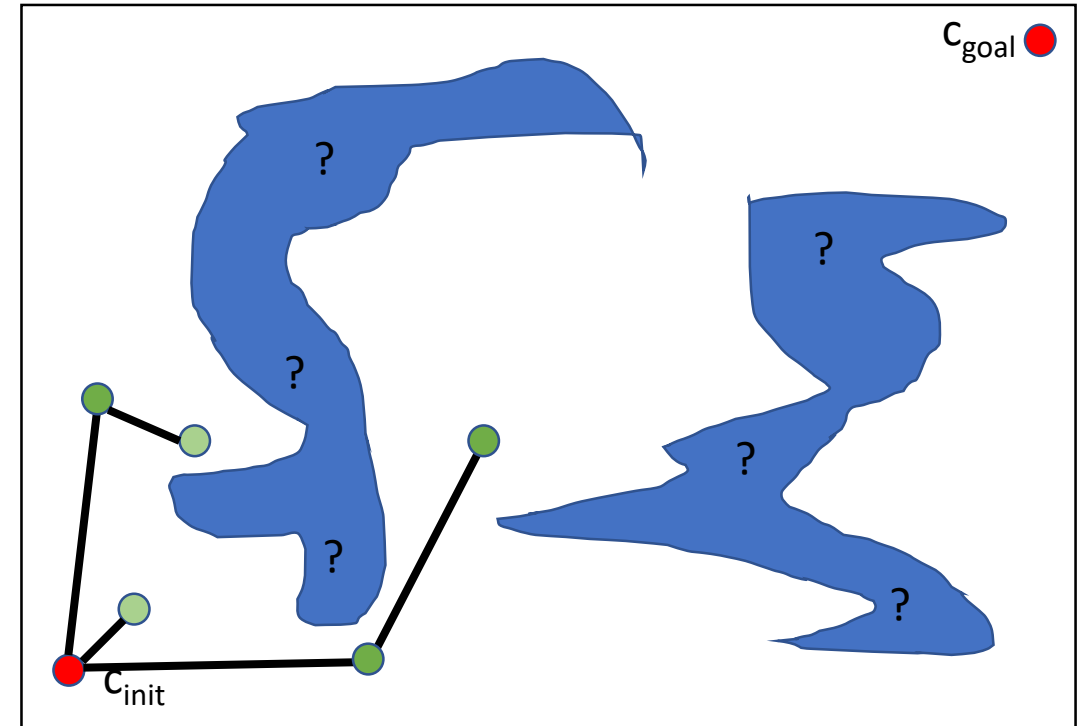
```
return  $Path(T)$ 
```

return $ApproxPath(T)$ 

The Rapidly-Exploring Random Tree

Algorithm 1: RRT (c_{init} , c_{goal} , $range_{max}$, $time_{max}$)

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while ($\neg TimeElapsed(T_{max})$) **do**

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$c_{near} \leftarrow NearestN(c_{rand}, T)$

$c_{new} \leftarrow getValidEdge(c_{near}, c_{rand}, range_{max})$

if ($\neg Trapped$) **then**

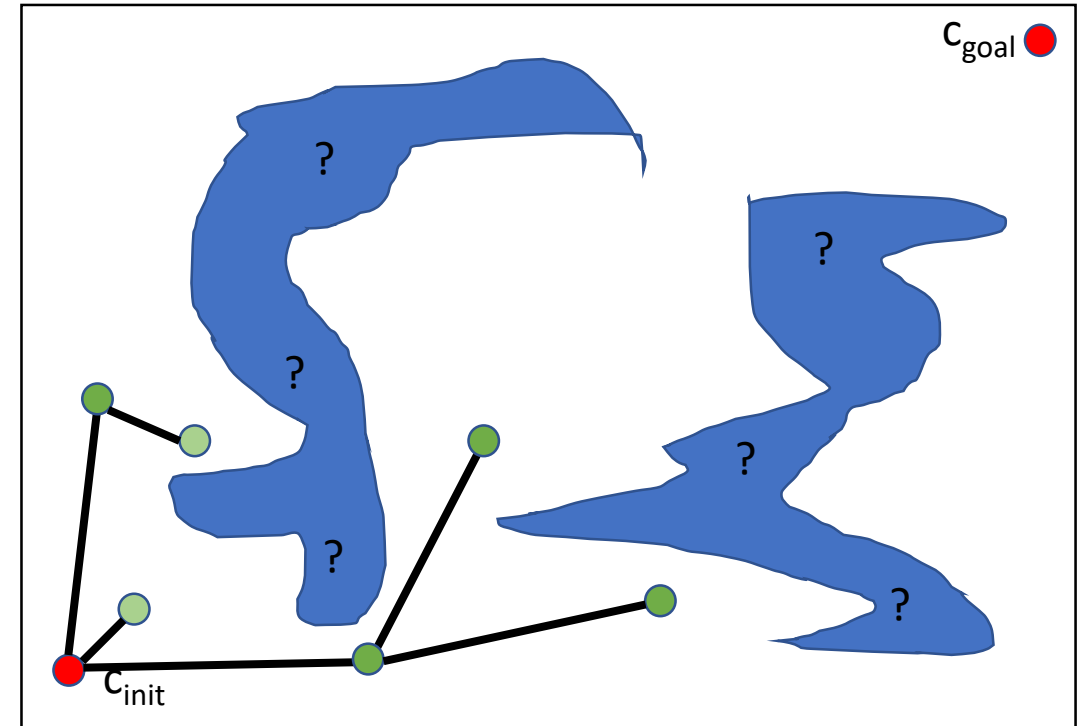
$T.addVertex(c_{new})$

$T.addEdge(c_{near}, c_{new})$

if $goalReached(c_{new}, c_{goal})$ **then**

return $Path(T)$

return $ApproxPath(T)$



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Algorithm 1: RRT (c_{init} , c_{goal} , $range_{max}$, $time_{max}$)

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$c_{near} \leftarrow NearestN(c_{rand}, T)$

$c_{new} \leftarrow getValidEdge(c_{near}, c_{rand}, range_{max})$

if ($\neg Trapped$) **then**

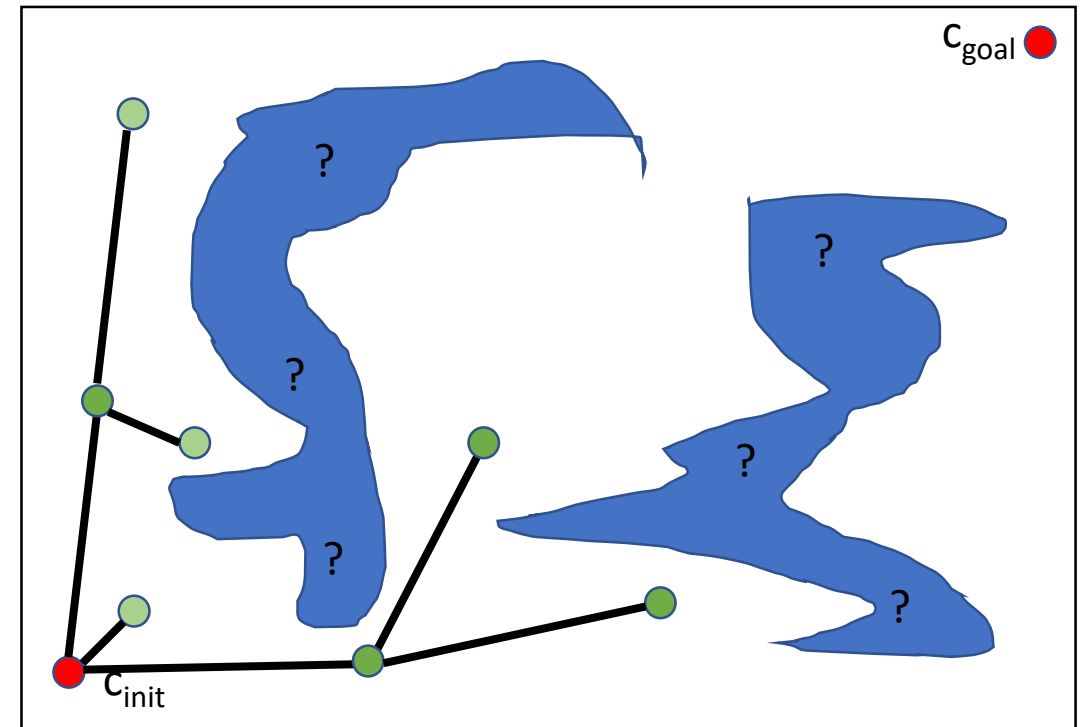
$T.addVertex(c_{new})$

$T.addEdge(c_{near}, c_{new})$

if $goalReached(c_{new}, c_{goal})$ **then**

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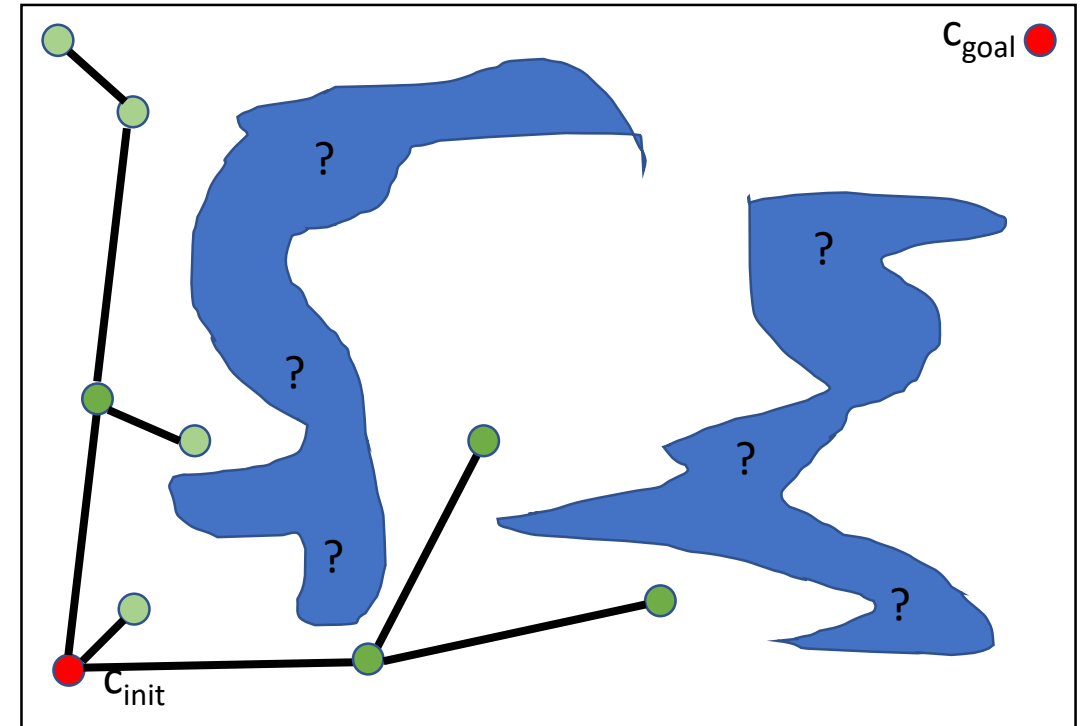
$T.addVertex(c_{new})$

$T.addEdge(c_{near}, c_{new})$

if $goalReached(c_{new}, c_{goal})$ **then**

return $Path(T)$

return $ApproxPath(T)$



The Rapidly-Exploring Random Tree

Algorithm 1: RRT (c_{init} , c_{goal} , $range_{max}$, $time_{max}$)
$$T.\text{init}(c_{init})$$

```

while (! TimeElapsed( $T_{max}$ )) do

```

$$c_{rand} \leftarrow \text{RandomState}()$$
$$c_{near} \leftarrow \text{NearestN}(c_{rand}, T)$$
$$c_{new} \leftarrow \text{getValidEdge}(c_{near}, c_{rand}, range_{max})$$

if ($\neg Trapped$) **then**

$$T.addVertex(c_{new})$$
$$T.addEdge(c_{near}, c_{new})$$

if $goalReached(c_{new}, c_{goal})$ **then**

```

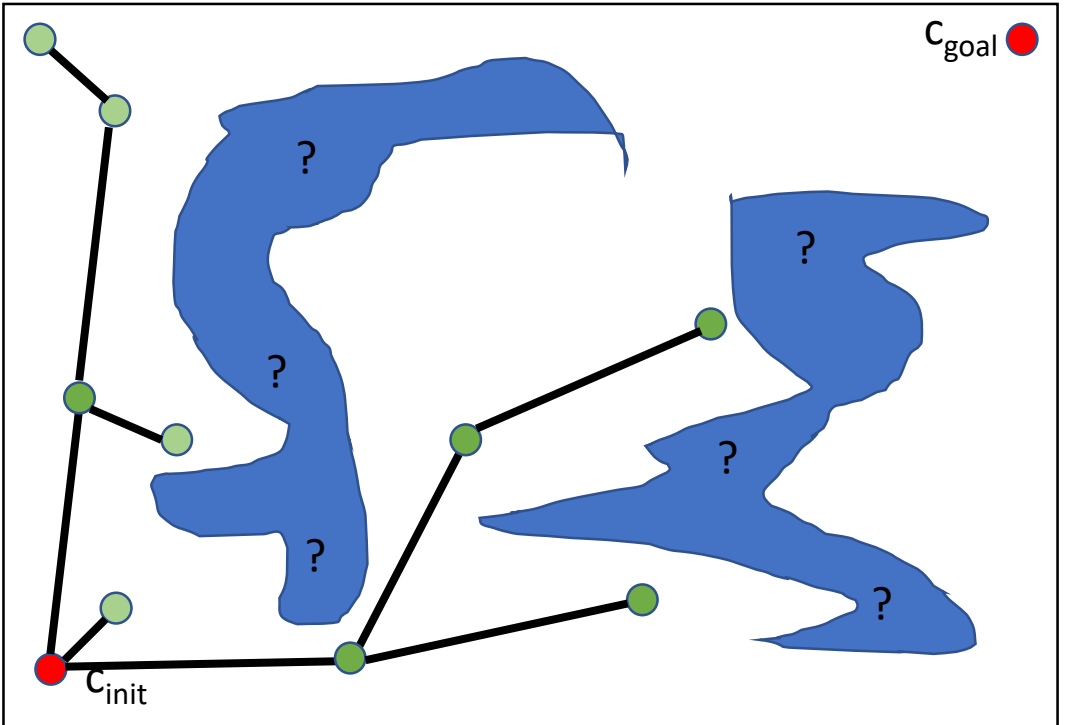
return  $Path(T)$ 

```

```

return ApproxPath(T)

```



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$c_{new} \leftarrow getValidEdge(c_{near}, c_{rand}, range_{max})$

if ($\neg Trapped$) **then**

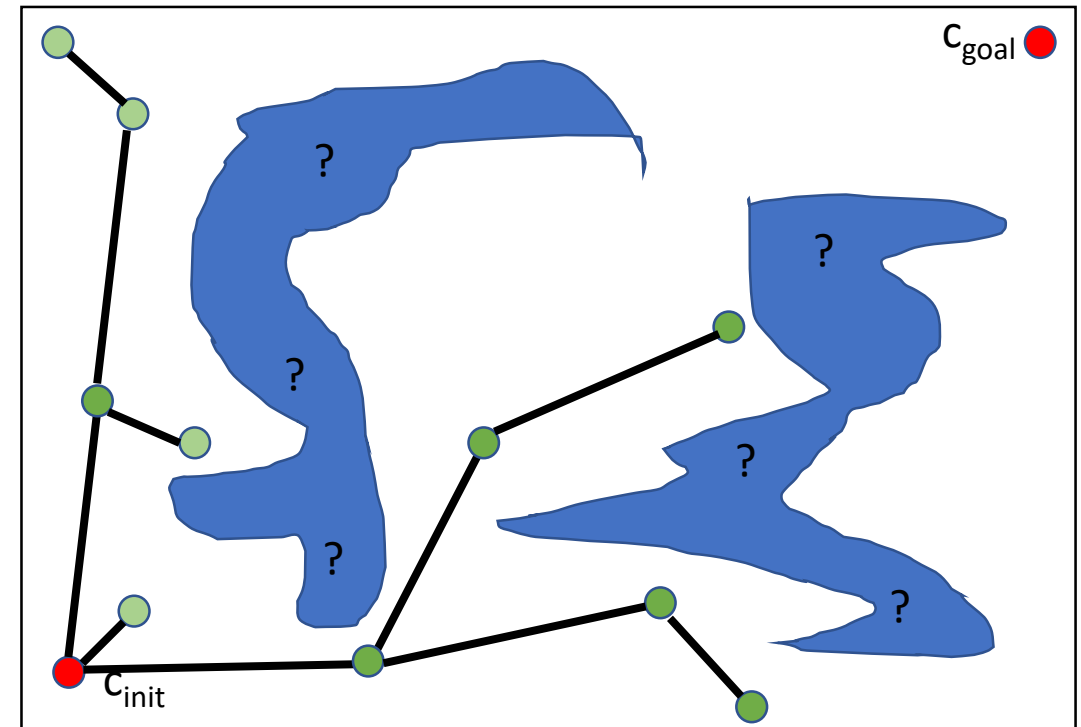
$T.addVertex(c_{new})$

$T.addEdge(c_{near}, c_{new})$

if $goalReached(c_{new}, c_{goal})$ **then**

return $Path(T)$

return $ApproxPath(T)$



The Rapidly-Exploring Random Tree

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$T.init(c_{init})$

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$c_{rand} \leftarrow RandomState()$

$c_{near} \leftarrow NearestN(c_{rand}, T)$

$c_{new} \leftarrow getValidEdge(c_{near}, c_{rand}, range_{max})$

if ($\neg Trapped$) **then**

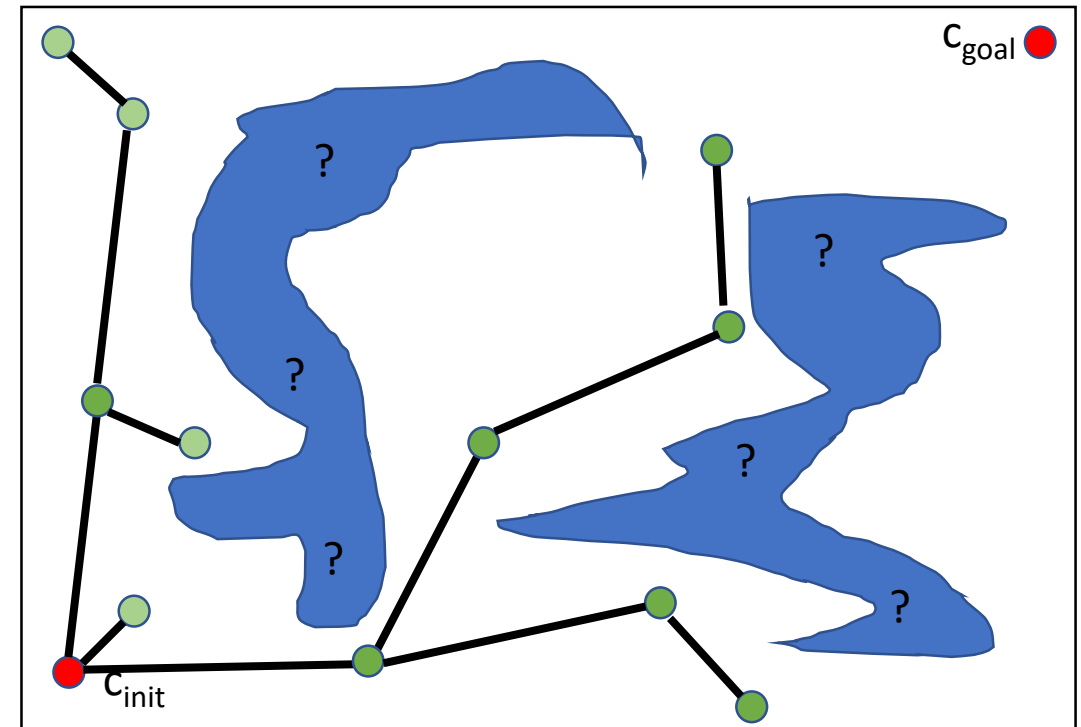
$T.addVertex(c_{new})$

$T.addEdge(c_{near}, c_{new})$

if $goalReached(c_{new}, c_{goal})$ **then**

return $Path(T)$

return $ApproxPath(T)$



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$c_{rand} \leftarrow RandomState()$

$c_{near} \leftarrow NearestN(c_{rand}, T)$

$c_{new} \leftarrow getValidEdge(c_{near}, c_{rand}, range_{max})$

if ($\neg Trapped$) **then**

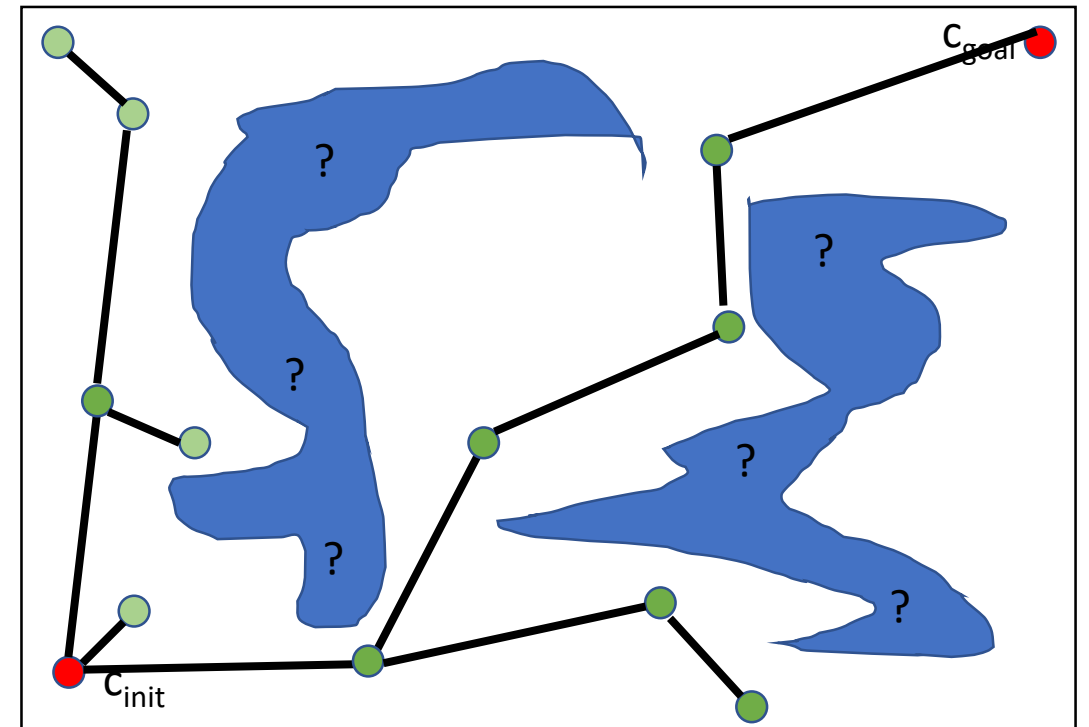
$T.addVertex(c_{new})$

$T.addEdge(c_{near}, c_{new})$

if $goalReached(c_{new}, c_{goal})$ **then**

return $Path(T)$

return $ApproxPath(T)$



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Algorithm 1: RRT (c_{init} , c_{goal} , $range_{max}$, $time_{max}$)

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$c_{near} \leftarrow NearestN(c_{rand}, T)$

$c_{new} \leftarrow getValidEdge(c_{near}, c_{rand}, range_{max})$

if ($\neg Trapped$) **then**

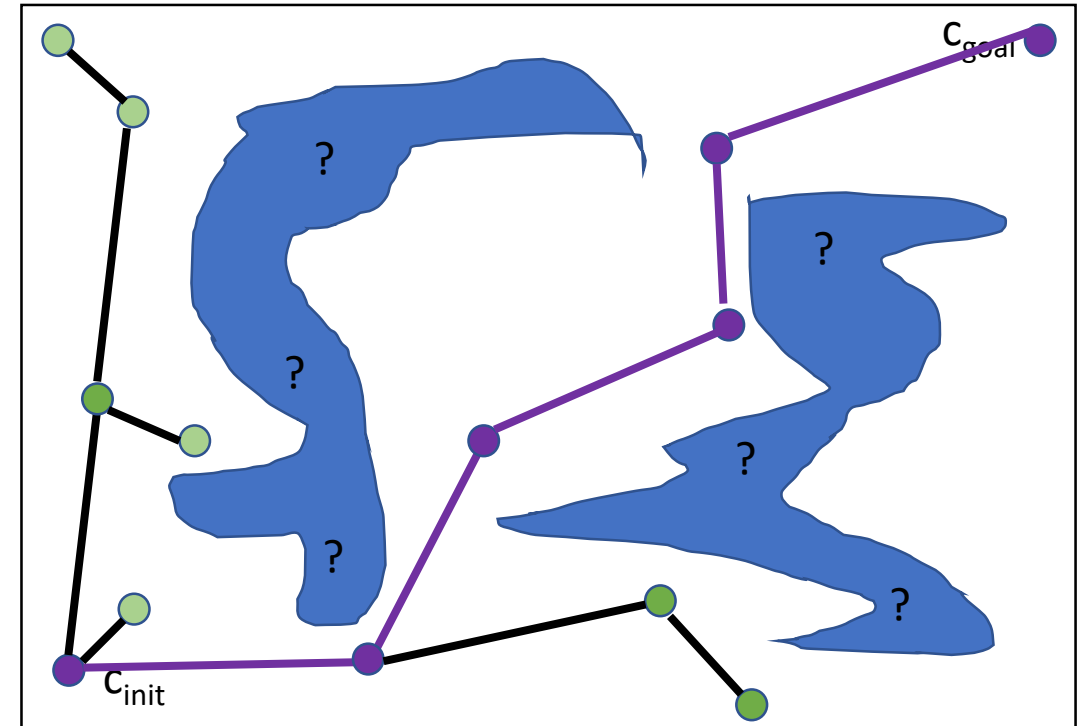
$T.addVertex(c_{new})$

$T.addEdge(c_{near}, c_{new})$

if $goalReached(c_{new}, c_{goal})$ **then**

return $Path(T)$

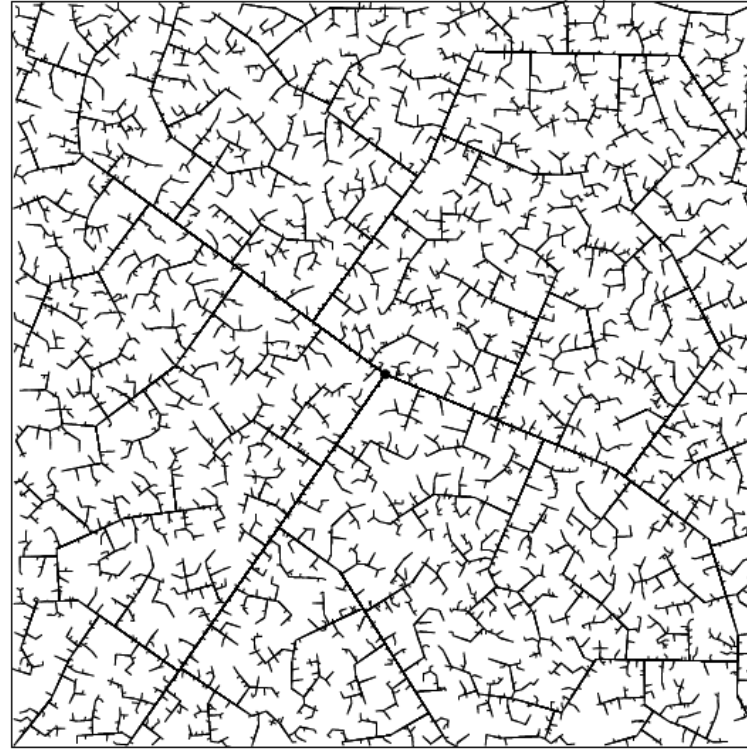
return $ApproxPath(T)$



On a Large scale



45 iterations



2345 iterations

This is why it is called
rapidly exploring:
→ it covers the c-space fast

Sources:

Motion Planning: The Essentials – LaValle - <http://msl.cs.illinois.edu/~lavalle/papers/Lav11b.pdf>

The Rapidly-Exploring Random Tree - Connect

Algorithm 2: RRTC ($c_{init}, c_{goal}, range_{max}$)

$T_1.init(c_{init}), T_2.init(c_{goal})$	1
while ($! TimeElapsed(T_{max})$) do	2
$T_{current}, T_{other} \leftarrow Swap(T_1, T_2)$	3
$c_{rand} \leftarrow RandomState()$	4
$c_{near} \leftarrow NearestN(c_{rand}, T_{current})$	
$c_{new} \leftarrow getExtend(c_{near}, range_{max})$	
if ($! Trapped$) then	
$T_{current}.addVertex(c_{new})$	8
$T_{current}.addEdge(c_{near}, c_{new})$	9
$c_{near'} \leftarrow NearestN(c_{new}, T_{other})$	10
$c_{new'} \leftarrow getExtend(c_{near'}, range_{max})$	11
if ($! Trapped$) then	12
$T_{other}.addVertex(c_{new'})$	13
$T_{other}.addEdge(c_{near'}, c_{new'})$	14
if ($Reached$) then	15
return $Path(T_1, T_2)$	16
return $ApproxPath(T_1, T_2)$	17

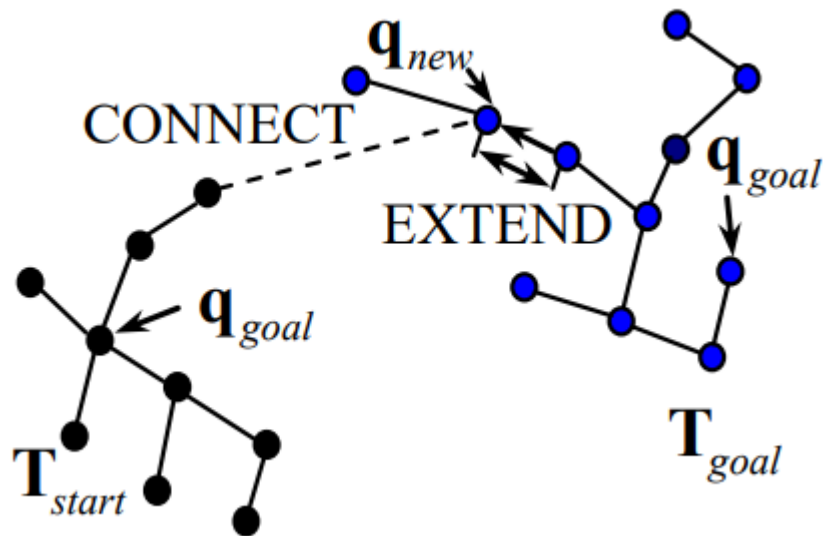
Assume $getExtend = getValidEdge$

$c_{new} \leftarrow getValidEdge(c_{near}, c_{rand}, range_{max})$

Exercise:

- Analyse the algorithm shown on the left.
- Write down an example and try to understand the difference.

Solution



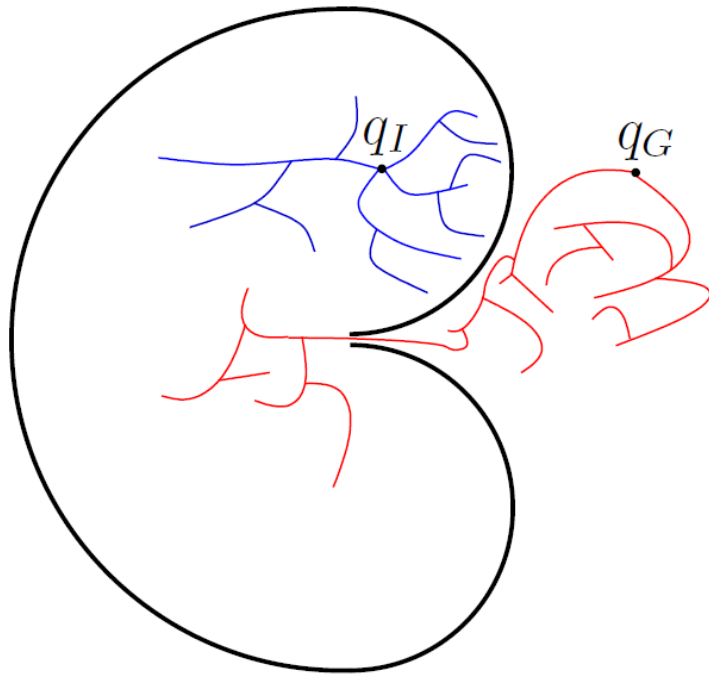
Sources:

Bidirectional RRT Algorithm for Collision Avoidance Motion Planning of FFSR
– Huazhong Li, Yongsheng Liang

- The algorithm builds up two trees.
- In each iteration the algorithm tries to connect both trees and as soon as the trees are connected the path is found.
- Difference *extend* and *validEdge*: The extend function add also configurations on the edge of the tree.

Why two trees?

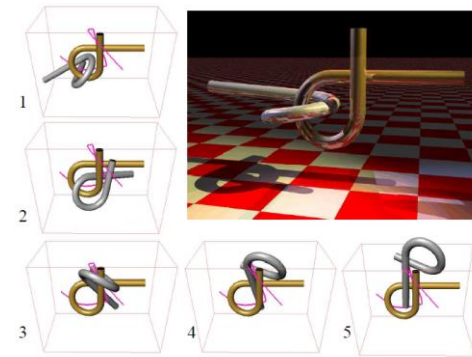
The bugtrap problem



Sources:

Planning Algorithms– LaValle - <http://planning.cs.uiuc.edu/>

- There are multiple scenarios that one tree can be stuck. Two trees can help to solve this issues.
- The RRT-Connect was the first algorithm to solve the famous alpha puzzle.

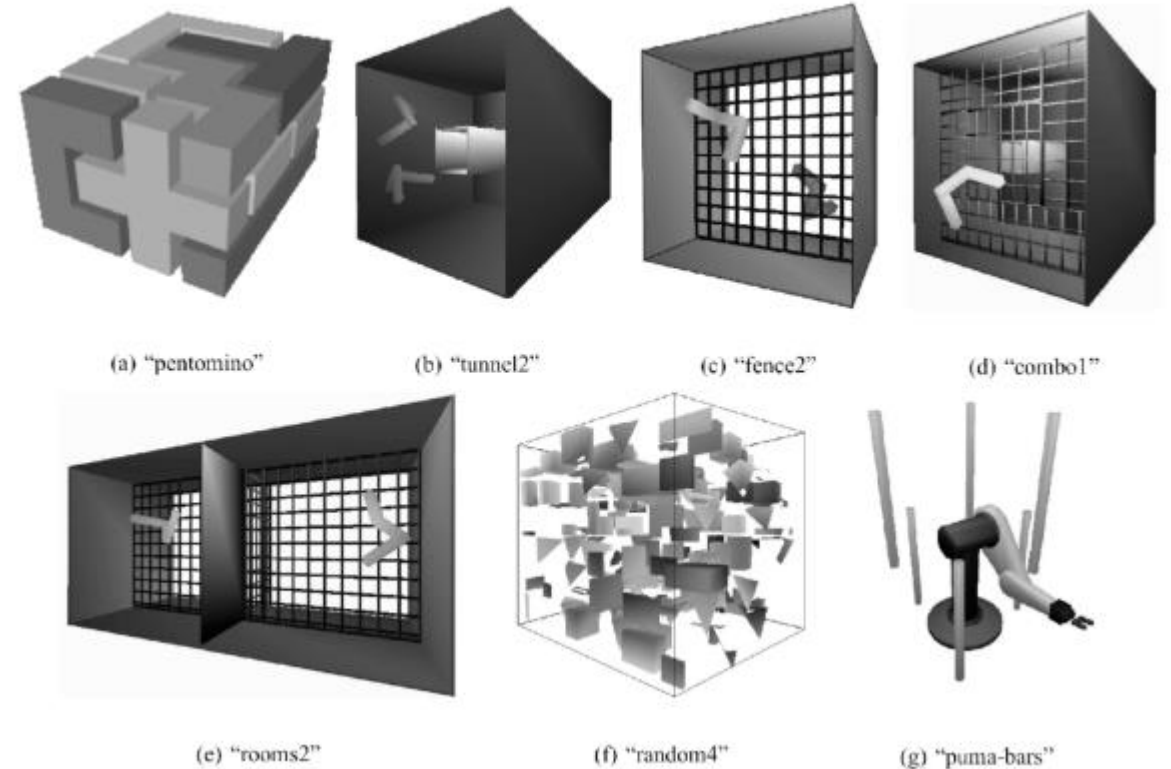


Sources:

Planning Algorithms– LaValle - <http://planning.cs.uiuc.edu/>

More than two?

- Yes, it makes sense to go for more than two trees.
- One can even build up a “roadmap of Trees”. (Combination of sPRM and RRT)
- These algorithms are good for parallel programming on the CPU.
- With this one can solve algorithms in even more dimensions.



Sources:

Sampling-Based Roadmap of Trees for Parallel Motion Planning -Lydia E. Kavraki et. al.

- <https://www.clear.rice.edu/comp450/papers/plaku2005srt-parallel-motion-planning.pdf>

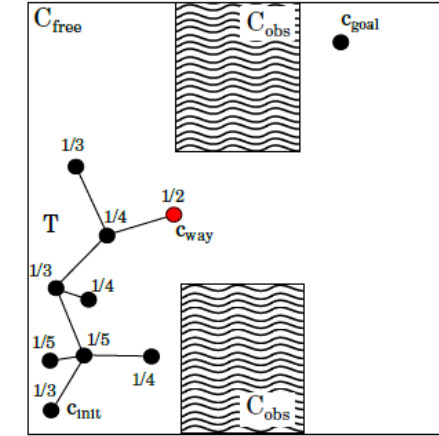
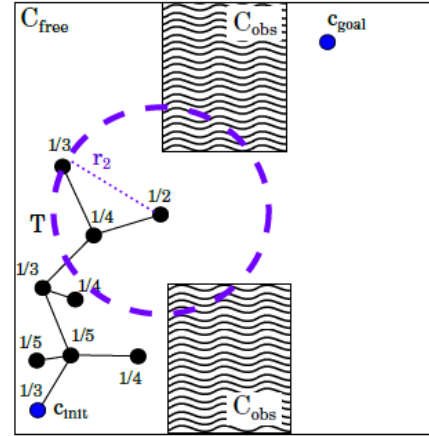
Expansive Space Tree

Algorithm 4: EST ($c_{init}, c_{goal}, r_1, r_2$)

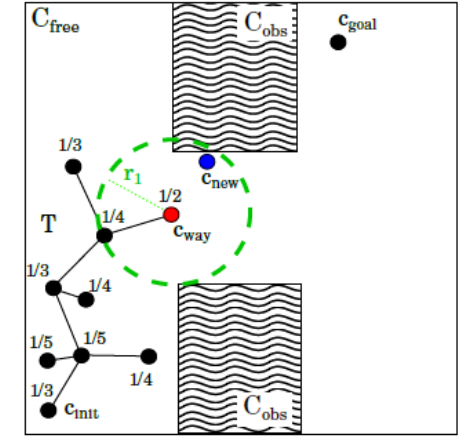
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1  T.init( $c_{init}$ )
2  while (! TimeElapsed( $T_{max}$ )) do
3       $c_{way} \leftarrow \text{chooseWaypoint}(T)$ 
4       $c_{new} \leftarrow \text{expandWaypoint}(c_{way}, r_1)$ 
5      if (! Trapped) then
6          T.addVertex( $c_{new}$ )
7          T.addEdge( $c_{way}, c_{new}$ )
8          T.updateWeights( $r_2$ )
9          if goalReached( $c_{new}, c_{goal}$ ) then
10             return Path(T)
11  return ApproxPath(T)

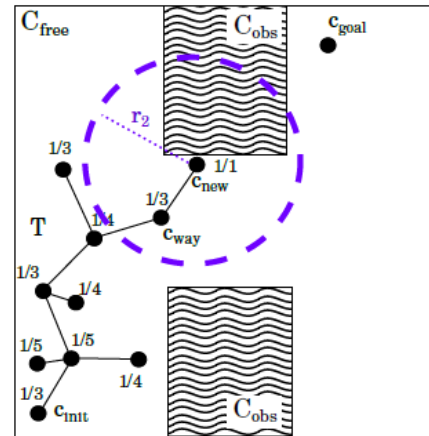
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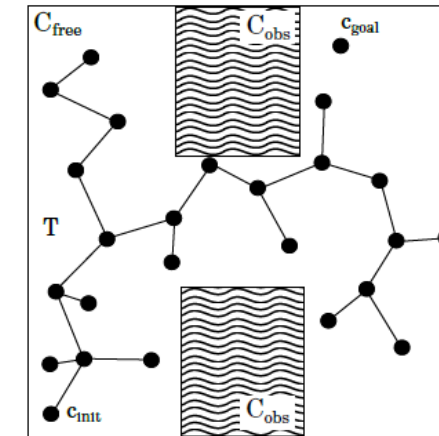
line 3



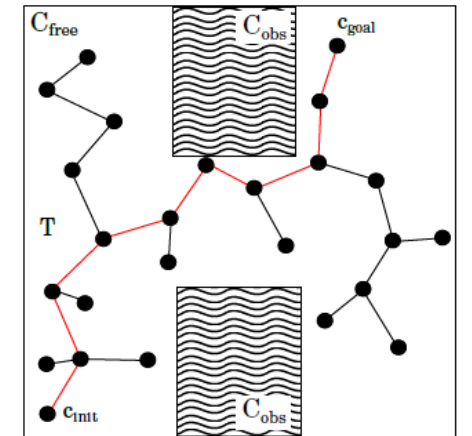
line 4



line 6-8



...iterate...



line 9-11

Fig. 11: An example of an Expansive Space Tree approach.