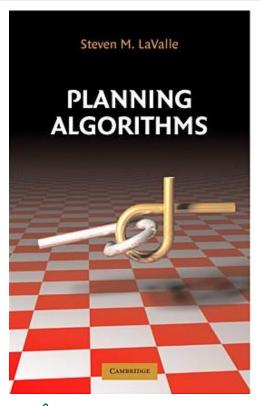


Motion Planning: Rapidly-Exploring Random Trees

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Book Recommendation



Available for free - online!

http://planning.cs.uiuc.edu/

Chapter:

- Sampling-based
- Motion Planning

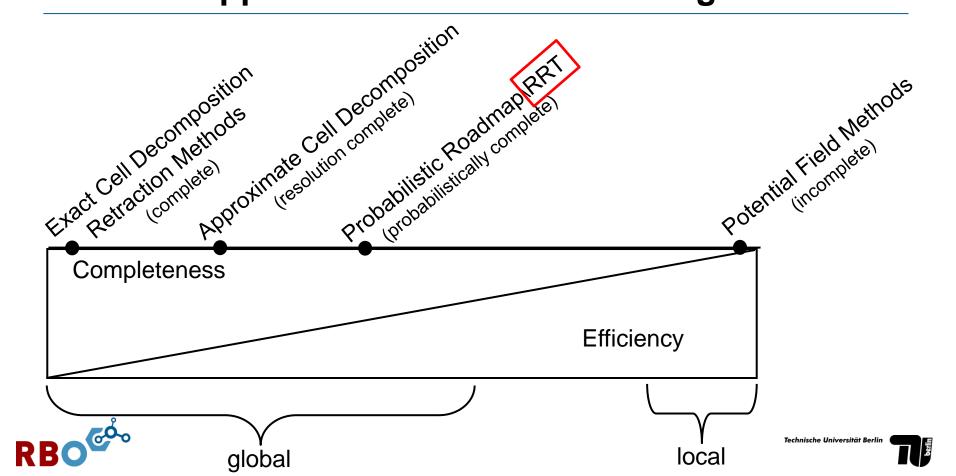


The Motion Planning Problem

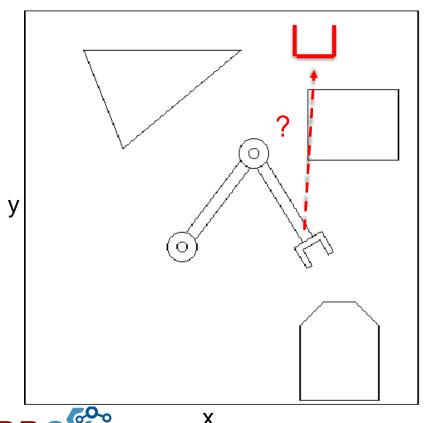
- R: robot with n degrees of freedom
- $B_{1,...,m}$: obstacles, rigid objects
- W: workspace R² or R³
- C: configuration space
- $R, B_{1,\ldots,m} \in W$
- $q_{\text{initial}}, q_{\text{final}}$
- Find a free path τ so that R moves from $q_{\rm initial}$ to $q_{\rm final}$ without touching any $B_{\rm i}$

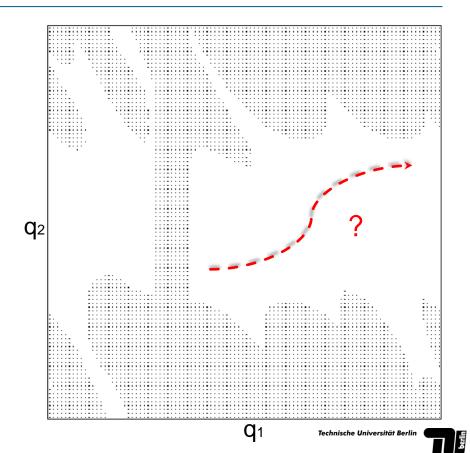


Different Approaches to Motion Planning

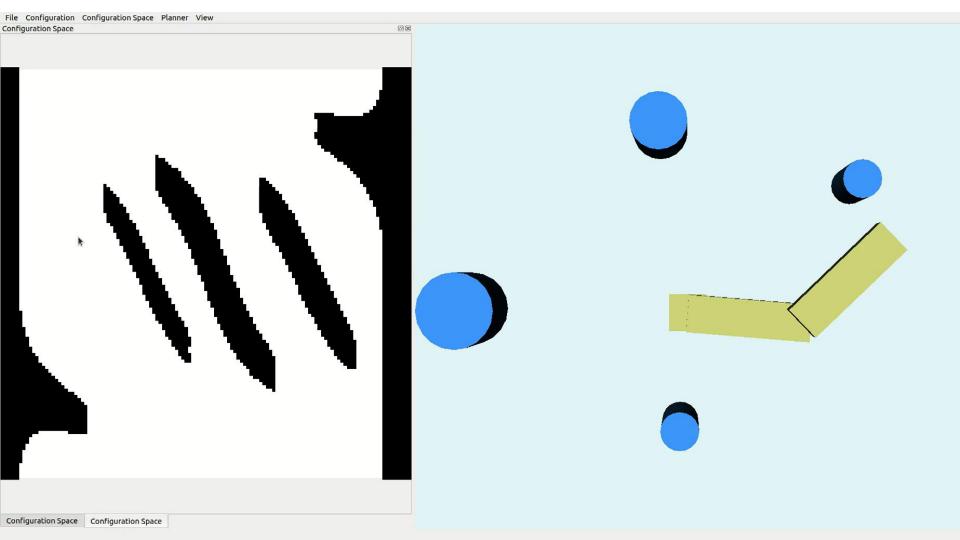


Motion Planning - Illustrated



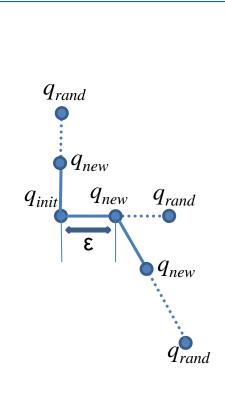


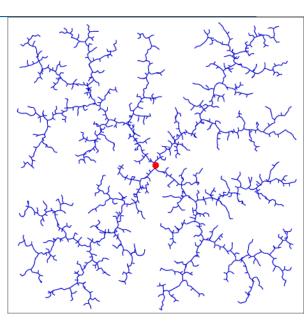




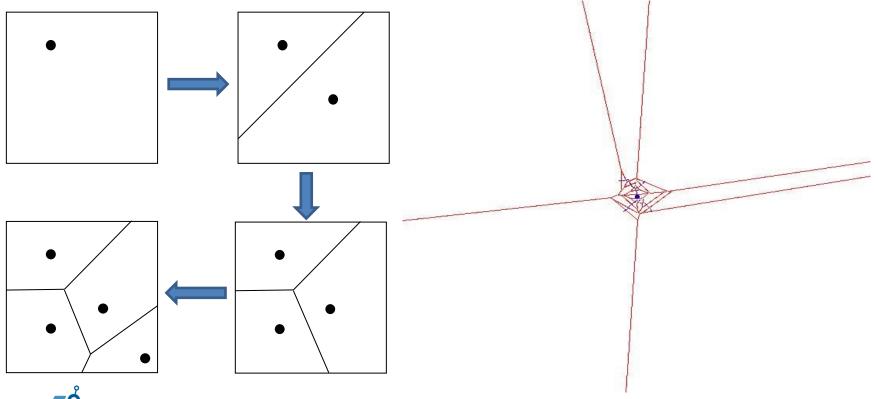
Rapidly-Exploring Random Trees

```
BUILD_RRT(q_{init})
      T.\operatorname{init}(q_{init});
      for k = 1 to K do
           q_{rand} \leftarrow \text{RANDOM\_CONFIG()};
           \text{EXTEND}(\mathcal{T}, q_{rand});
      Return \mathcal{T}
\text{EXTEND}(\mathcal{T}, q)
      q_{near} \leftarrow \text{NEAREST\_NEIGHBOR}(q, T);
      if NEW_CONFIG(q, q_{near}, q_{new}) then
           T.add\_vertex(q_{new});
           T.add\_edge(q_{near}, q_{new});
           if q_{new} = q then
                Return Reached;
           else
                 Return Advanced;
      Return Trapped;
```



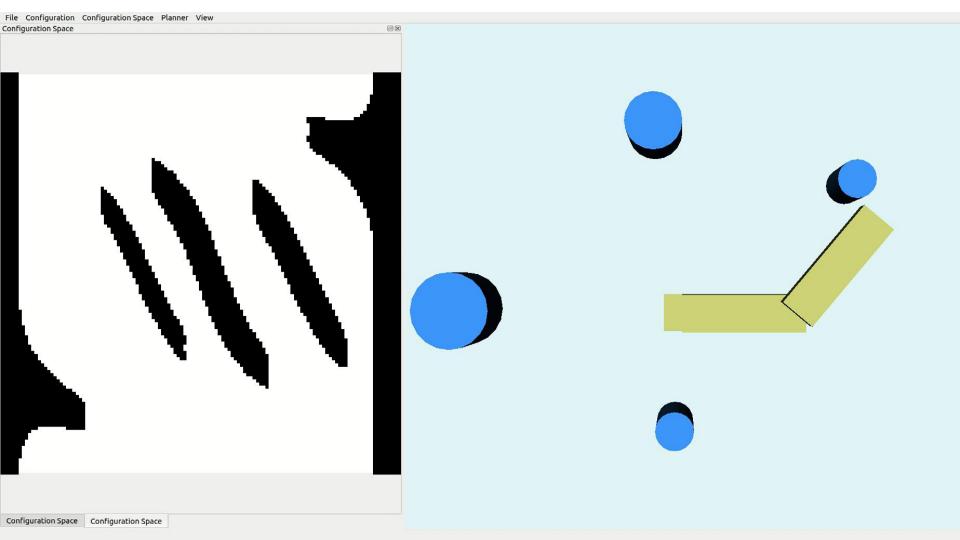


The Voronoi Bias







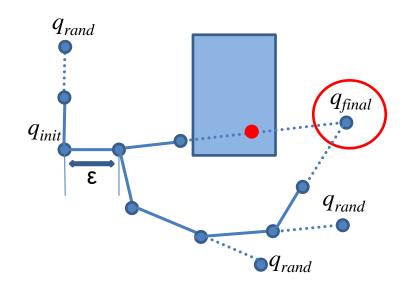


RRT Extensions (1) – Goal Bias

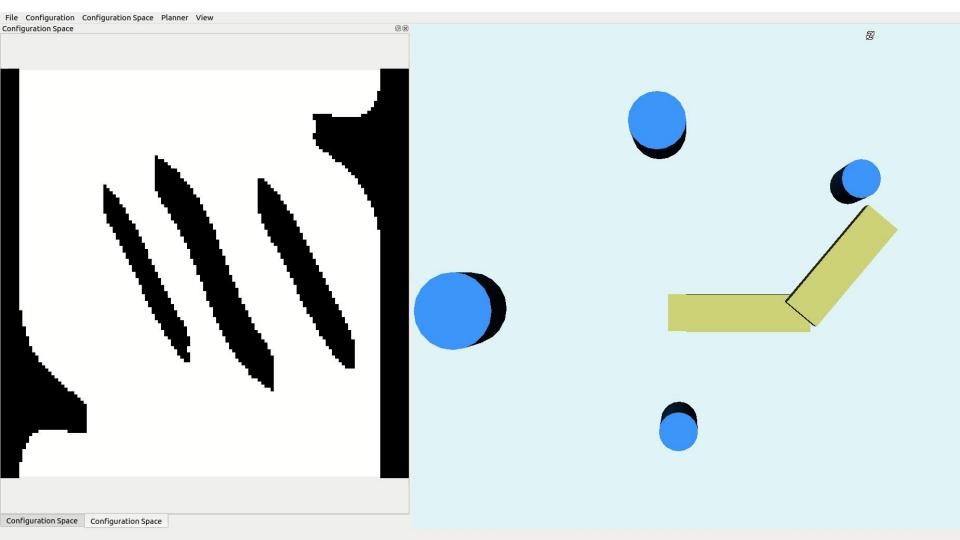
```
BUILD_RRT(q_{init}, q_{final}, p)
1 T.init(q_{init});
2 for k = 1 to K do
3 q_{rand} \leftarrow RANDOM\_GOAL\_BIAS(p);
4 EXTEND(T, q_{rand});
5 Return T
```



- Sample random configuration with probability 1-p
- Sample goal configuration with probability p







RRT Extensions (2) – Connect

```
BUILD_RRT(q_{init}, q_{final}, p)

1 T.init(q_{init});

2 for k = 1 to K do

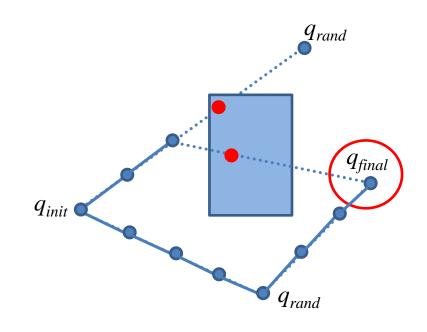
3 q_{rand} \leftarrow RANDOM\_GOAL\_BIAS(p);

4 CONNECT(T, q_{rand})

5 Return T
```

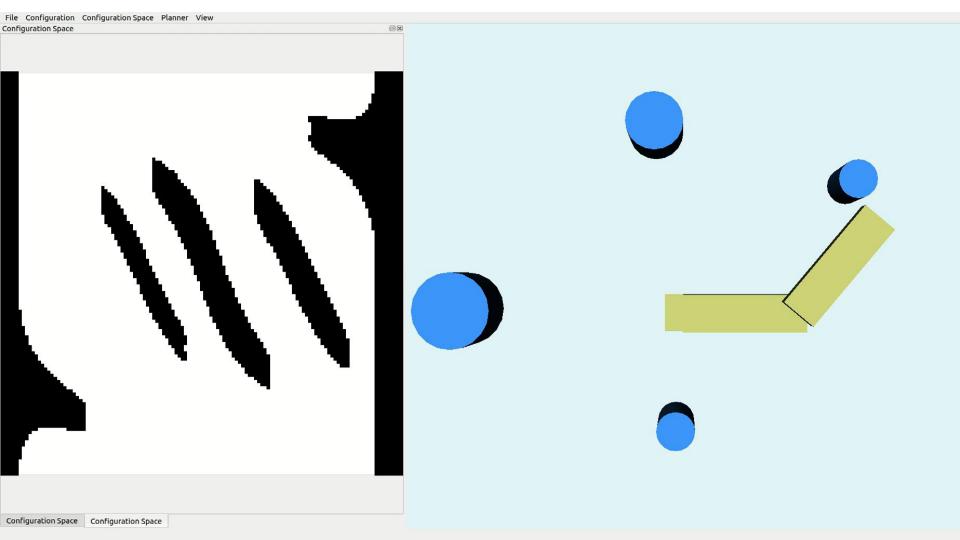


```
CONNECT(T, q)
1 repeat
2 S \leftarrow \text{EXTEND}(T, q);
3 until not (S = Advanced)
4 Return S;
```



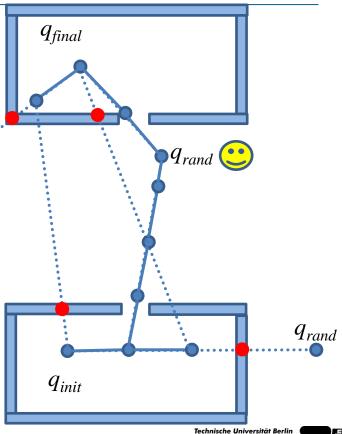




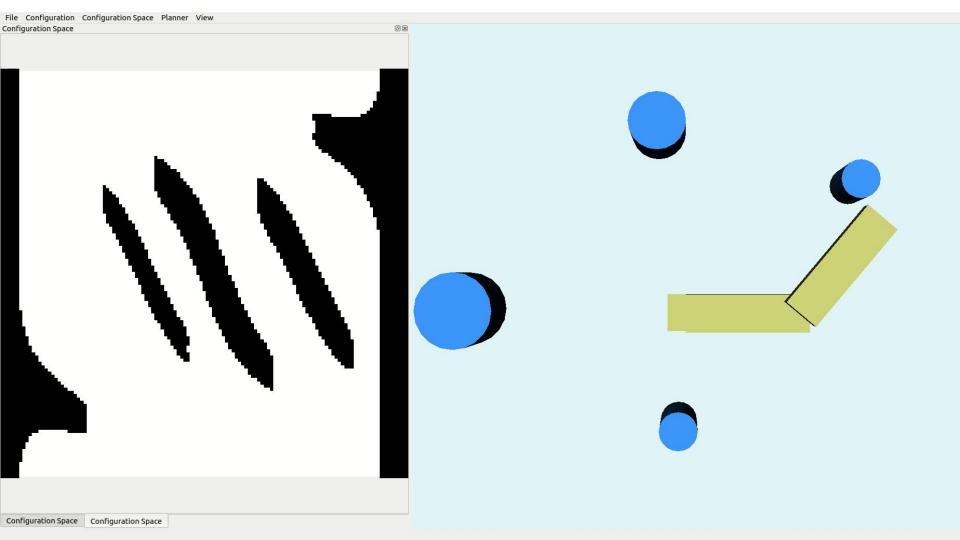


RRT Extensions (3) – Using Two Trees

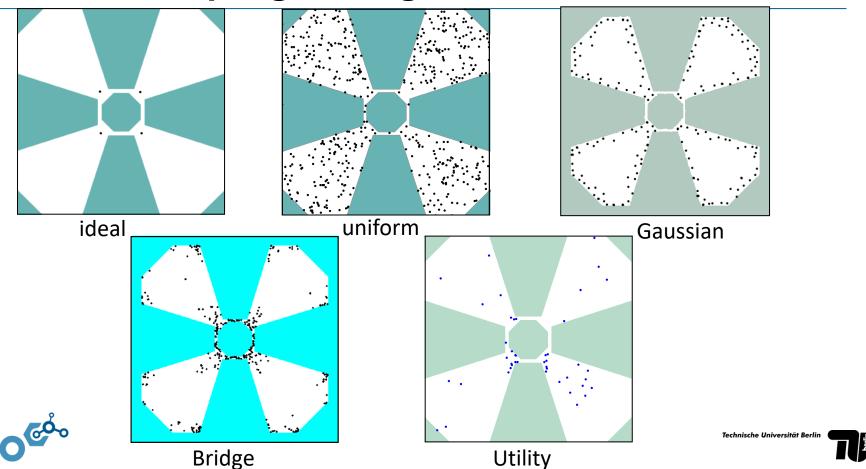
```
1: function RRT_BIDIRECTIONAL(q_{init}, q_{final})
        \mathcal{T}_a.init(q_{init}); \mathcal{T}_b.init(q_{final});
        for k = 1 to K do
             q_{rand} \leftarrow \texttt{RANDOM\_CONFIGURATION()};
             CONNECT(\mathcal{T}_a, q_{rand});
             if q_{new} AND CONNECT(\mathcal{T}_b, q_{new}) == Reached then
6:
                  Return PATH(\mathcal{T}_a, \mathcal{T}_b);
             end if
                                                                                     q_{rand}
             SWAP(\mathcal{T}_a, \mathcal{T}_b);
         end for
10:
        Return Failure;
11:
12: end function
```







Different Sampling Strategies



Exploration versus Exploitation

Exploration seeks **understanding of the state space**, irrespective of a particular task. In motion planning, the process **exploration** seeks to understand the connectivity of the configuration space, irrespective of solving a particular motion planning problem.

Guided exploration seeks **efficient understanding of the state space**, irrespective a particular task, by **leveraging available information**.

Exploitation strives to **accomplish a particular task** as **efficiently as possible** by **leveraging available information**.

In motion planning, **exploitation** seeks a valid path for a **particular task**, based on available information.

