MT-RRT, the general purpose multi threading library for RRT. 1.0

Generated by Doxygen 1.9.1

1 Foundamental concepts	1
1.1 What is an RRT algorithm?	1
1.2 Background on RRT	1
1.2.1 Standard RRT	2
1.2.2 Bidirectional version of the RRT	3
1.2.3 Compute the optimal solution: the RRT*	3
1.3 MT-RRT pipeline	5
1.3.0.1 A) Define the problem class	5
1.3.0.2 B) Instantiate a Planner	8
1.3.0.3 C) Solve the problem calling Planner ::solve	8
1.3.0.4 D) Inspect the results after solving the problem	8
2 Customize your own planning problem	9
2.1 Planar maze problem	9
2.1.1 Sampling	9
2.1.2 Optimal trajectory and constraints	9
2.1.3 Advancement along the optimal trajectory	10
2.2 Articulated arm problem	10
2.2.1 Sampling	11
2.2.2 Optimal trajectory and constraints	11
2.2.3 Advancement along the optimal trajectory	11
2.2.3.1 Tunneled check collision	11
2.2.3.2 Bubble of free configuration	12
2.3 Navigation problem	14
2.3.1 Sampling	14
2.3.2 Optimal trajectory and constraints	14
2.3.3 Advancement along the optimal trajectory	15
3 Parallel RRT	17
3.0.1 Parallelization of the query activities	17
3.0.2 Shared tree critical regions	
3.0.3 Parallel expansions on linked trees	
3.0.4 Multi agents approach	19
4 Namespace Index	21
4.1 Namespace List	21
5 Hierarchical Index	23
5.1 Class Hierarchy	23
6 Class Index	25
6.1 Class List	25
7 Namespace Documentation	29

	7.1 mt_rrt Namespace Reference	29
	7.1.1 Detailed Description	32
	7.1.2 Function Documentation	32
	7.1.2.1 make_random_seed()	32
8	Class Documentation	33
	8.1 mt_rrt::BidirSolution Class Reference	33
	8.2 mt_rrt::Connector Class Reference	34
	8.2.1 Detailed Description	34
	8.2.2 Member Function Documentation	34
	8.2.2.1 getTrajectory()	34
	8.2.2.2 minCost2Go()	35
	8.2.2.3 minCost2GoConstrained()	35
	8.3 mt_rrt::Copiable< T > Class Template Reference	36
	8.4 mt_rrt::DescriptionAndParameters Struct Reference	36
	8.5 mt_rrt::Determinism Class Reference	36
	8.5.1 Detailed Description	37
	8.6 mt_rrt::DeterminismRegulator Class Reference	37
	8.7 mt_rrt::detail::Distribution < DistributionShape > Class Template Reference	37
	8.8 mt_rrt::EmbarassinglyParallelPlanner Class Reference	38
	8.8.1 Detailed Description	38
	8.9 mt_rrt::Error Class Reference	38
	8.10 mt_rrt::Extender Class Reference	39
	8.10.1 Detailed Description	40
	8.10.2 Member Function Documentation	40
	8.10.2.1 search()	40
	8.11 mt_rrt::ExtenderBidirectional Class Reference	40
	8.12 mt_rrt::ExtenderSingle Class Reference	41
	8.13 mt_rrt::GaussianEngine Class Reference	42
	8.13.1 Detailed Description	42
	8.13.2 Constructor & Destructor Documentation	42
	8.13.2.1 GaussianEngine()	42
	8.14 mt_rrt::HyperBox Class Reference	43
	8.14.1 Detailed Description	43
	8.14.2 Constructor & Destructor Documentation	43
	8.14.2.1 HyperBox()	43
	8.14.3 Member Function Documentation	44
	8.14.3.1 sampleSeed()	44
	8.14.3.2 sampleState()	44
	8.15 mt_rrt::Iterations Class Reference	45
	8.15.1 Detailed Description	45
	8.16 mt rrt··KeenSearchDredicate Struct Reference	15

8.17 mt_rrt::Limited < T > Class Template Reference	46
8.17.1 Detailed Description	46
8.17.2 Constructor & Destructor Documentation	46
8.17.2.1 Limited()	46
8.17.3 Member Function Documentation	47
8.17.3.1 get()	47
8.17.3.2 set()	47
8.18 mt_rrt::TunneledConnector::Line Class Reference	47
8.18.1 Member Function Documentation	48
8.18.1.1 getCumulatedCost()	48
8.18.1.2 getState()	48
8.19 mt_rrt::LinkedTreesPlanner Class Reference	49
8.19.1 Detailed Description	49
$8.20 \ \text{mt_rrt::LockFreeForwardList} < T > Class \ \text{Template Reference} \dots \dots$	49
8.21 mt_rrt::LowerLimited < T > Class Template Reference	50
8.21.1 Detailed Description	50
8.22 mt_rrt::MultiAgentPlanner Class Reference	50
8.22.1 Detailed Description	51
8.23 mt_rrt::MultiThreadedPlanner Class Reference	51
8.24 mt_rrt::NearestQuery Struct Reference	52
8.25 mt_rrt::NearSet Struct Reference	52
8.26 mt_rrt::NearSetElement Struct Reference	52
8.27 mt_rrt::NearSetQuery Struct Reference	52
8.28 mt_rrt::Node Class Reference	53
8.28.1 Detailed Description	54
8.28.2 Member Function Documentation	54
8.28.2.1 cost2Root()	54
8.28.3 Member Data Documentation	54
8.28.3.1 MAX_ITERATIONS	54
8.29 mt_rrt::NodeOwning Class Reference	54
8.30 mt_rrt::NodesAllocatorT < NodeT > Class Template Reference	55
8.31 mt_rrt::ObjectPool< T > Class Template Reference	55
8.32 mt_rrt::ParallelFor Class Reference	56
8.33 mt_rrt::ParallelizedQueriesPlanner Class Reference	56
8.33.1 Detailed Description	57
8.34 mt_rrt::Parameters Struct Reference	57
8.34.1 Detailed Description	57
8.35 mt_rrt::Planner Class Reference	57
8.35.1 Detailed Description	58
8.36 mt_rrt::PlannerSolution Struct Reference	58
8.36.1 Detailed Description	59

8.37 mt_rrt::Positive< T > Class Template Reference	59
8.37.1 Detailed Description	59
8.38 mt_rrt::ProblemAware Class Reference	60
8.38.1 Detailed Description	60
8.39 mt_rrt::ProblemDescription Struct Reference	60
8.39.1 Detailed Description	61
8.40 mt_rrt::ProblemDescriptionCloner Class Reference	61
8.41 mt_rrt::Rewire Struct Reference	61
8.42 mt_rrt::Sampler Class Reference	62
8.42.1 Detailed Description	62
8.42.2 Member Function Documentation	62
8.42.2.1 sampleSeed()	62
8.42.2.2 sampleState()	63
8.43 mt_rrt::SharedTreePlanner Class Reference	63
8.43.1 Detailed Description	64
8.44 mt_rrt::SimpleSolution Class Reference	64
8.45 mt_rrt::Solution Class Reference	64
8.46 mt_rrt::StandardPlanner Class Reference	65
8.46.1 Detailed Description	65
8.47 mt_rrt::SteerIterations Class Reference	66
8.47.1 Detailed Description	66
8.48 mt_rrt::Connector::SteerResult Struct Reference	66
8.48.1 Detailed Description	66
8.49 mt_rrt::SynchronizationAware Class Reference	67
8.50 mt_rrt::SynchronizationDegree Class Reference	67
8.50.1 Detailed Description	68
8.51 mt_rrt::Threads Class Reference	68
8.51.1 Detailed Description	68
8.52 mt_rrt::Trajectory Class Reference	69
8.52.1 Detailed Description	69
8.52.2 Member Enumeration Documentation	69
8.52.2.1 AdvanceInfo	69
8.52.3 Member Function Documentation	70
8.52.3.1 getCumulatedCost()	70
8.52.3.2 getState()	70
8.53 mt_rrt::TreeHandler Class Reference	70
8.54 mt_rrt::TreeHandlerBasic Class Reference	71
8.55 mt_rrt::TunneledConnector Class Reference	72
8.55.1 Detailed Description	73
8.55.2 Member Function Documentation	73
8.55.2.1 getTrajectory()	73

Index	77
8.58 mt_rrt::View Struct Reference	76
8.57.1 Detailed Description	75
8.57 mt_rrt::UpperLimited $<$ T $>$ Class Template Reference	75
8.56.2.1 UniformEngine()	74
8.56.2 Constructor & Destructor Documentation	74
8.56.1 Detailed Description	74
8.56 mt_rrt::UniformEngine Class Reference	74
8.55.2.2 minCost2Go()	73

Foundamental concepts

1.1 What is an RRT algorithm?

Rapidly Random exploring Tree(s), aka RRT(s), is one of the most popular technique adopted for solving planning path problems in robotics. In essence, a planning problem consists of finding a feasible trajectory or path that leads a manipulator, or more in general a dynamical system, from a starting configuration/state to an ending desired one, consistently with a series of constraints. RRTs were firstly proposed in [5]. They are able to explore a state space in an incremental way, building a search tree, even if they may require lots of iterations before terminating. They were proved be capable of always finding at least one solution to a planning problem, if a solution exists, i.e. they are probabilistic complete. RRT were also proved to perform well as kinodynamic planners, designing optimal LQR controllers driving a generic dynamical system to a desired final state, see [9] and [8].

The typical disadvantage of RRTs is that for medium-complex problems, they require thousands of iterations to get the solution. For this reason, the aim of this library is to provide multi-threaded planners implementing parallel version of RRTs, in order to speed up the planning process.

It is possible to use this library for solving any problem handled by an RRT algorithm. The only necessary thing to do when facing a new class of problem is to derive some specific objects describing the problem itself as detailed in Section 2.

At the same time, one of the most common problem to solve with RRT is a standard path planning for an articulated arm. What matters in such cases is to have a collision checker, which is not provided by this library. Anyway, the interfaces Tunneled_check_collision and Bubbles_free_ configuration allows you to integrate the collision checker you prefer for solving standard path planning problems (see also Section 2.2.3).

The next Section briefly reviews the basic mechanism of the RRT. The notations and formalisms introduced in the next Section will be also adopted by the other Sections. Therefore, the reader is strongly encouraged to read before the next Section.

Section 1.3 will describe the typical pipeline to consider when using MT-RRT, while some examples of planning problems are reported in Chapter 2. Chapter 3 will describe the possible parallelization strategy that MT-RRT offers you. ¹.

1.2 Background on RRT

¹A similar guide, but in a html format, is also available at http://www.andreacasalino.altervista.org/__MT_RRT_doxy_guide/index.html.

1.2.1 Standard RRT

RRTs explore the state space of a particular problem, in order to find a series of states connecting a starting x_o and an ending one x_f , at the same time accounting for the presence of constraints. More precisely, this is done by building at least a search tree $T(x_o)$ having x_o as root. Each node $x_i \in T$ is connected to its unique father $x_{fi} = Fath(x_f)$ by a trajectory $\tau_{fi \to i}$. The root x_o is the only node not having a father $(Fath(x_o) = \emptyset)$. The set $\mathcal{X} \subseteq \mathbb{R}^d$ will contain all the possible states x of the system whose motion must be controlled, while $\underline{\mathcal{X}} \subseteq \mathcal{X}$ is a subset describing the admissible region induced by a series of constraints. The solution we are interested in, consists clearly of a sequence of trajectories τ entirely contained in $\underline{\mathcal{X}}$. If we consider classical path planning problems, the constraints are represented by the obstacles populating the scene, which must be avoided. However, according to the nature of the problem to solve, different kind of constraints might need to be accounted. The basic version of an RRT algorithm is described by Algorithm 1, whose steps are visually represented by Figure 1.2. Essentially, the tree is randomly grown by performing several steering operations. Sometimes, the extension of the tree toward the target state x_f is tried in order to find an edge leading to that state.

```
Data: x_o, x_f
T = \{x_o\};
for k = 1: MAX\_ITERATIONS do
    sample r \sim U(0,1);
    if r < \sigma then
        x_{steered} = \mathsf{Extend}(T, x_f);
        if x_{steered} is VALID then
            if ||x_{steered} - x_f|| \le \epsilon then
                 Return Path_to_root(x_{steered})\cup x_f;
            end
        end
    end
    else
        sample a x_R \in \mathcal{X};
        Extend(T, x_R);
    end
end
```

Algorithm 1: Standard RRT. A deterministic bias is introduced for connecting the tree toward the specific target state x_f . The probability σ regulates the frequency adopted for trying the deterministic extension. The Extension procedure is described in algorithm 2.

Algorithm 2: The Extend procedure.

```
Data: T, x_R
Return \underset{x_i \in T}{\operatorname{argmin}}(C(\tau_{i \to R}));
```

Algorithm 3: The Nearest_Neighbour procedure: the node in T closest to the given state x_R is searched.

The Steer function in algorithm 2 must be problem dependent. Basically, It has the aim to extend a certain state x_i already inserted in the tree, toward another one x_R . To this purpose, an optimal trajectory $\tau_{i\to R}$,

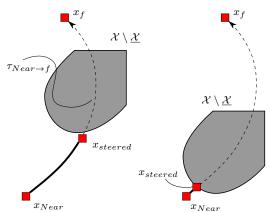


Figure 1.1 The dashed curves in both pictures are the optimal trajectories, agnostic of the constraints, connecting the pair of states x_{Near} and x_f , while the filled areas are regions of X not allowed by constraints. The steering procedure is ideally in charge of searching the furthest state to x_{Near} along $\tau_{Near \to f}$. For the example on the right, the steering is not possible: the furthest state along $\tau_{Near \to f}$ is too much closer to x_{Near} .

agnostic of the constraints, going from x_i to x_R , must be taken into account. Ideally, the steering procedure should find the furthest state from x_i that lies on $\tau_{i\to R}$ and for which the portion of $\tau_{i\to R}$ leading to that state is entirely contained in $\underline{\mathcal{X}}$. However, in real implementations the steered state returned might be not the possible farthest from x_i . Indeed, the aim is just to extend the tree toward x_R . At the same time, in case such the steered state results too closer to x_i , the steering should fails $\frac{1}{2}$.

The Nearest_Neighbour procedure relies on the definition of a cost function $C(\tau)$. Therefore, the closeness of states does not take into account the shape of $\underline{\mathcal{X}}$. Indeed $C(\tau)$ it's just an estimate agnostic of the constraints. Then, the constraints are taken into account when steering the tree. The algorithm terminates when a steered configuration x_s sufficiently close to x_f is found.

The steps involved in the standard RRT are summarized by Figure 1.2.

1.2.2 Bidirectional version of the RRT

The behaviour of the RRT can be modified leading to a bidirectional strategy [6], which expands simultaneously two different trees. Indeed, at every iteration one of the two trees is extended toward a random state. Then, the other tree is extended toward the steered state previously obtained. At the next iteration, the roles of the trees are inverted. The algorithm stops, when the two trees meet each other. The detailed pseudocode is reported in Algorithm 4.

This solution offers several advantages. For instance, the computational times absorbed by the Near-est Neighbour search is reduced since this operation is done separately for the two trees and each tree contains at an average half of the states computed. The steps involved in the bidirectional strategy are depicted in Figure 1.3.

1.2.3 Compute the optimal solution: the RRT*

For any planning problem there are infinite $\tau_{o\to f}\subset \underline{\mathcal{X}}$, i.e. infinite trajectories starting from x_o and terminating in x_f which are entirely contained in the admissible region $\underline{\mathcal{X}}$. Among the aforementioned set, we might be interested in finding the trajectory minimizing the cost $C(\tau_{o\to f})$, refer to Figure 1.4. The basic version of the RRT algorithm is proved to find with a probability equal to 1, a suboptimal solution [4]. The optimality is addressed by a variant of the RRT, called RRT* [4], whose pseudocode is contained

²This is done to avoid inserting less informative nodes in the tree, reducing the tree size.

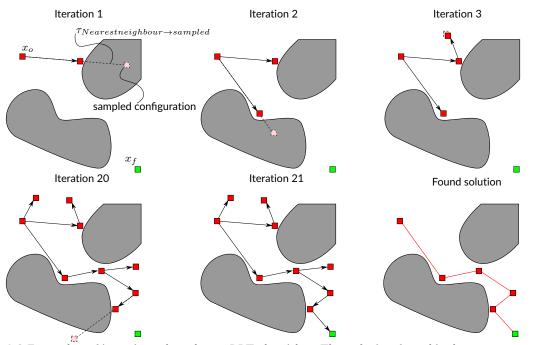


Figure 1.2 Examples of iterations done by an RRT algorithm. The solution found is the one connecting the state in the tree that reached x_f , with the root x_o .

```
Data: x_o, x_f
T_A = \{x_o\};
T_B = \{x_f\};
x_{target} = \text{root of } T_A;
x_2 = \text{root of } T_B;
T_{master} = T_A;
T_{slave} = T_B;
for k = 1: MAX\_ITERATIONS do
    sample r \sim U(0,1);
    if r < \sigma then
       x_{steered} = \mathsf{Extend}(T_{master}, x_{target});
    end
    else
         sample a x_R \in \mathcal{X};
         x_{steered} = \mathsf{Extend}(T_{master}, x_R);
    end
    if x_{steered} is VALID then
         x_{steered2} = \mathsf{Extend}(T_{slave}, x_{steered});
         if x_{steered2} is VALID then
             if ||x_{steered} - x_{steered2}|| \le \epsilon then
                  Return Path_to_root(x_{steered}) \cup Revert ( Path_to_root(x_{steered}2) );
             end
         end
    end
    Swap T_{target} and T_2;
    Swap T_{master} and T_{slave};
end
```

Algorithm 4: Bidirectional RRT. A deterministic bias is introduced for accelerating the steps. The probability σ regulates the frequency adopted for trying the deterministic extension. The Revert procedure behaves as exposed in Figure 1.3.

1.3 MT-RRT pipeline 5

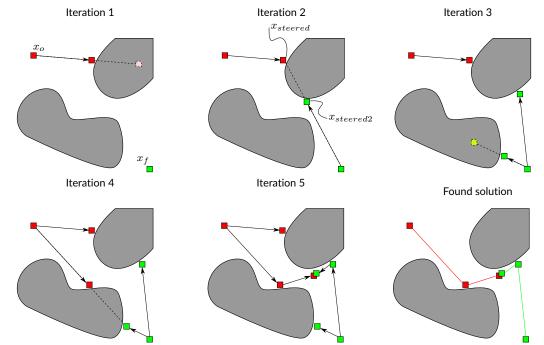


Figure 1.3 Examples of iterations done by the bidirectional version of the RRT. The path in the tree rooted at x_f is reverted to get the second part of the solution.

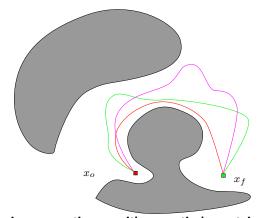


Figure 1.4 Different trajectories connecting x_o with x_f , entirely contained in $\underline{\mathcal{X}}$. If we assume as cost the length of a path, the red solution is the optimal one.

in Algorithm 5. Essentially, the RRT* after inserting in a tree a steered state, tries to undertake local improvements to the connectivity of the tree, in order to minimize the cost-to-go of the states in the Near set. This approach is proved to converge to the optimal solution after performing an infinite number of iterations 3 . There are no precise stopping criteria for the RRT*: the more iterations are performed, the more the solution found get closer to the optimal one.

1.3 MT-RRT pipeline

When solving a planning problem with MT-RRT, the pipeline of Figure 1.5 should be followed.

1.3.0.1 A) Define the problem class

Whenever you need to solve a new class of planning problem, you should define:

³In real cases, after a sufficient big number of iterations an optimizing effect can be yet appreciated.

```
Data: x_o, x_f
T = \{x_o\};
Solutions = \emptyset;
for k = 1: MAX\_ITERATIONS do
    sample r \sim U(0,1);
    if r < \sigma then
         x_{steered} = \text{Extend\_Star}(T, x_f);
         if x_{steered} is VALID then
             if ||x_{steered} - x_f|| \le \epsilon then
              Solutions = Solutions \cup x_{steered};
             end
        end
    end
    else
         sample a x_R \in \mathcal{X};
         Extend_Star(T, x_R);
    end
end
            \operatorname{argmin} (Cost_to_root(x_S));
x_{best} =
          x_S \in \widetilde{Solutions}
Return Path_to_root(x_{best}) \cup x_f;
Algorithm 5: RRT*. The Extend_Star, Rewird and Cost_to_root procedures are explained in, respec-
tively, algorithm 6, 7 and 8.
Data: T, x_R
x_{steered} = Extend(T, x_R);
if x_{steered} is VALID then
    Near = \left\{ x_i \in T \middle| C(\tau_{i \to steered}) \le \gamma \left(\frac{log(|T|)}{|T|}\right)^{\frac{1}{d}} \right\};
    Rewird(Near, x_{steered});
end
Return x_{steered};
                      Algorithm 6: The Extend_Star procedure. d is the cardinality of \mathcal{X}.
Data: Near, x_s
```

```
x_{bestfather} = Fath(x_s);
C_{min} = C(\tau_{bestfather \to s});
for x_n \in Near do
     if \tau_{n \to s} \subset \underline{\mathcal{X}} AND C(\tau_{n \to s}) < C_{min} then
         C_{min} = C(\tau_{n \to s});
          x_{best fath} = x_n;
     end
end
Fath(x_s) = x_{bestfath};
C_s = \mathsf{Cost\_to\_root}(x_s);
Near = Near \setminus x_{bestfath};
for x_n \in Near do
     if \tau_{s \to n} \subset \underline{\mathcal{X}} then
         C_n = C(\tau_{s \to n}) + C_s;
         if C_n < \mathsf{Cost\_to\_root}(x_n) then
           Fath(x_n) = x_s;
          end
     end
end
```

Algorithm 7: The Rewird procedure.

1.3 MT-RRT pipeline 7

```
\begin{array}{l} \text{Data: } x_n \\ \text{if } Fath(x_n) = \emptyset \text{ then} \\ \mid \text{ Return 0;} \\ \text{end} \\ \text{else} \\ \mid \text{ Return } C(\tau_{Fath(n) \rightarrow n}) + \text{Cost\_to\_root}(Fath(x_n)) \text{ ;} \\ \text{end} \end{array}
```

Algorithm 8: The Cost_to_root procedure computing the cost spent to go from the root of the tree to the passed node.

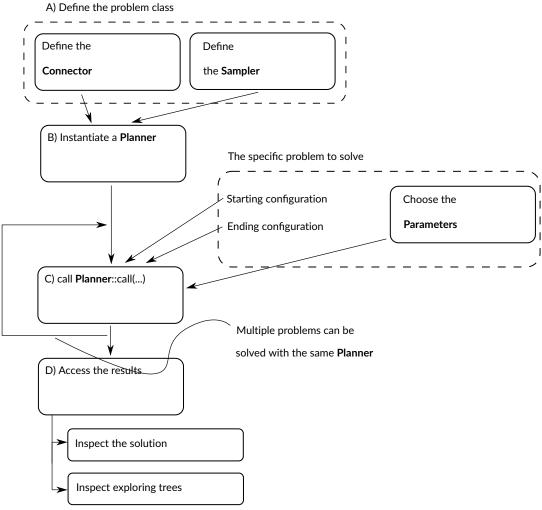


Figure 1.5 Steps to follow for consuming the MT-RRT library.

- a Sampler, whose responsability is to draw random states to allow the tree(s) growth.
- a Connector, whose responsability is generate optimal trajectories τ connecting pairs of states.

Chapter 2 reports some examples of planning problems, describing from theoretical point of view how to derive the corresponding **Sampler** and **Connector**. The concrete implementations are instead contained under the samples folder.

1.3.0.2 B) Instantiate a Planner

After having defined the problem, you need to build a **Planner**. This can be a classical mon threaded StandardPlanner or a multi threaded one. The population of such planners offered by this library is (refer also to Sections 3.0.1, 3.0.2, 3.0.3 and 3.0.4):

- EmbarassinglyParallel
- ParallelizedQueriesPlanner
- SharedTreePlanner
- LinkedTreesPlanner
- MultiAgentPlanner

Each planner support all the rrt algorithms described in Sections 1, 1.2.2 and 5, with the only exception that the MultiAgentPlanner does not support the bidirectional algorithm.

1.3.0.3 C) Solve the problem calling Planner::solve

You can use the generated **Planner** to solve a specific problem, trying to connect a starting and an ending configuration. You need also to specify additional **Parameters** like for instance the kind of strategy (Sections 1, 1.2.2 and 5).

1.3.0.4 D) Inspect the results after solving the problem

Planner::solve returns a **PlannerSolution** structure which contains all the information about the found solution, among which you can find:

- the found solution, i.e. a series of states $x_{1,2,3,\dots,M}$ that must be visited to get from the starting configuration to the ending one, by traversing the trajectories $\tau_{1\to 2}, \tau_{2\to 3}, \dots, \tau_{M-1\to M} \subset \underline{\mathcal{X}}$. In case a solution was not found, an empty option is returned.
- the tree(s) computed for finding the solution.

Additional information (the iterations spent, the computation time) regarding the solution can be accessed too.

Customize your own planning problem

MT-RRT can be deployed to solve each possible problems for which RRT can be used. The only thing to do is to derive specific concretions of Sampler and Connector containing all the problem-specific information, Section 1.3.0.1.

In order to help the user in understanding how to implement such derivations, three main kind of examples are part of the library. In the following Sections, they will be briefly reviewed.

2.1 Planar maze problem

The state space characterizing this problem is two dimensional, having $x_{1,2}$ as coordinates. The aim is to connect two 2D coordinates while avoiding the rectangular obstacles depicted in Figure 2.1. The state space is bounded by two corners describing the maximum and minimum possible x_1 and x_2 , see Figure 2.1.

2.1.1 Sampling

A sampled state x_R lies in the square delimited by the spatial bounds, i.e.:

$$x_{R} = \begin{bmatrix} x_{R1} \sim U(x_{1min}, x_{1max}) \\ x_{R2} \sim U(x_{2min}, x_{2max}) \end{bmatrix}$$
 (2.1)

2.1.2 Optimal trajectory and constraints

The optimal trajectory $\tau_{i\to k}$ between two states in \mathcal{X} is simply the segment connecting that states. The cost $C(\tau_{i\to k})$ is assumed to be the length of such segment:

$$C(\tau_{i \to k}) = ||x_i - x_k|| \tag{2.2}$$

The admissible region $\underline{\mathcal{X}}$ is obtained subtracting the points pertaining to the obstacles. In other words, the segment connecting the states in the tree should not traverse any rectangular obstacle, refer to the right part of Figure 2.1.

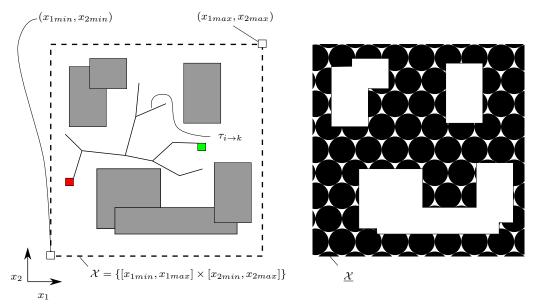


Figure 2.1 Example of maze problem.

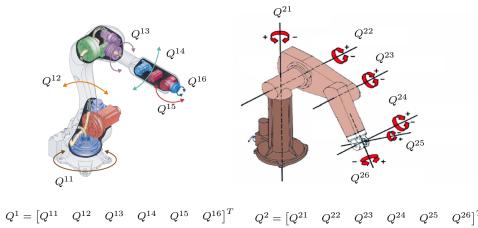


Figure 2.2 Rotating joints of two articulated manipulators.

2.1.3 Advancement along the optimal trajectory

The steering procedure is done as similarly described in Section 2.2.3.1, advancing at every steer trial of a quantum of space along τ and checking every time that the segment connecting the previously steered state and the reached one entirely lies in the admitted space.

2.2 Articulated arm problem

This is for sure one of the most common problem that can be solved using rrt algorithms. Consider a cell having a group of articulated serial robots. Q^i will denote the vector describing the configuration of the i^{th} robot, i.e. the positional values assumed by each of its joint. A generic state x_i is characterized by the series of poses assumed by all the robots in the cell:

$$x_i = Q_i = [(Q_i^1)^T \dots (Q_i^n)^T]^T$$
 (2.3)

refer also to Figure 2.2.

These kind of problems consist in finding a path in the configurational space that leads the set of robots from an initial state Q_o to and ending one Q_f , while avoiding the obstacles populating the scene, i.e. avoid

collisions between any object in the cell and any part of the robots as well as cross-collision between all the robot parts. Here the term path, refer to a series of intermediate waypoints $Q_{1,...,m}$ to traverse to lead the robot from Q_o to Q_f .

2.2.1 Sampling

The i^{th} joint of the k^{th} robot, denoted as Q^{ki} , is subjected to some kinematic limitations prescribing that its positional value must remain always within a compact interval $Q^{ki} \in [Q^{ki}_{min}, Q^{ki}_{max}]$. Therefore, the sampling of a random configuration Q_R is done as follows:

2.2.2 Optimal trajectory and constraints

Similarly to the problem described in Section 2.1.2, $\tau_{i \to k}$ is assumed to be a segment in the configurational space and the cost C is the Euclidean distance of a pair of states. The admissible region \underline{X} is made by all the configurations Q for which a collision is not present.

2.2.3 Advancement along the optimal trajectory

The trajectory going from Q_i to Q_k can be parametrized in order to characterize all the possible configurations pertaining to $\tau_{i\to k}$:

$$Q(s) = \tau_{i \to k}(s) = Q_i + s\left(Q_k - Q_i\right)$$
(2.5)

s is a parameter spanning $au_{i \to k}$ and can assume a value inside [0,1]. Ideally, the steer process has the aim of determine that state $Q(s_{steered})$ that is furthest from Q_i and at the same time contained in \underline{X} (Figure 1.1). Anyway, determine the exact value of $s_{steered}$ would be too much computationally demanding. Therefore, in real situations, two main approaches can be in principle adopted: a tunneled check collision or the bubble of free configuration. In the samples contained by this repository, the second one was preferred. Anyway, both methods will be discussed for completeness.

2.2.3.1 Tunneled check collision

This approach consider as steered state $Q_{steered}$ the following quantity:

$$Q_{steered} = \begin{cases} if(\|Q_k - Q_i\| \le \epsilon) \Rightarrow Q_k \\ else \Rightarrow Q_i + s_{\Delta}(Q_k - Q_i) \text{ s.t. } s_{\Delta} \|Q_k - Q_i\| = \epsilon \end{cases}$$
 (2.6)

with ϵ in the order of few degrees. $Q_{steered}$ is checked to be or not in \underline{X} , by checking the presence of collisions, and is consequently marked as VALID or INVALID. This library does not provide a general collision checker (some very basics geometrical functions were implemented in order to run the samples). Anyway when implementing such an approach, you can easily integrate your favourite collision checker (like for example [1] or [2]) to embed in your own Connector.

Clearly, multiple tunneled check, starting from Q_i , can be done in order to get as close as possible to Q_k . This process can be arrested when reaching Q_k or an intermediate state for which a collision check is not passed. This behaviour can be obtained by setting a value grater than 1 with Solver::setSteerTrials(...). Figure 2.3 summarizes the above considerations.

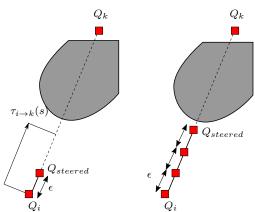


Figure 2.3 Steer extension along the segment connecting two states in the configuration space. On the left a single steer trial approach, on the right a multiple one.

2.2.3.2 Bubble of free configuration

This approach was first proposed in [7] and is based on the definition of a so called bubble of free configuration \mathcal{B} . Such a bubble is a region of the configurational space that is built around a state Q_i . More formally, $\mathcal{B}(Q_i)$ is defined as follows ¹

$$\mathcal{B}(\bar{Q} = \begin{bmatrix} \bar{Q}^{1T} & \dots & \bar{Q}^{nT} \end{bmatrix}^T) = \mathcal{B}_O(\bar{Q}) \cap \mathcal{B}_C(\bar{Q})$$
(2.7)

where \mathcal{B}_O contains describes the region containing the poses guaranteed to not manifest collisions with the fixed obstacles, while \mathcal{B}_C describes the poses for which the robots do not collide with each others. They are defined as follows:

$$\mathcal{B}_O(\bar{Q}) = \left\{ Q \middle| \forall j \in \{1, \dots, n\} \sum_{i} R^{ji} |Q^{ji} - \bar{Q}^{ji}| \le d_{min}^j \right\}$$
 (2.8)

$$\mathcal{B}_{C}(\bar{Q}) = \left\{ Q \middle| \forall j, k \in \{1, \dots, n\} \sum_{i} R^{ji} |Q^{ji} - \bar{Q}^{ji}| + \sum_{i} R^{ki} |Q^{ki} - \bar{Q}^{ki}| \le d_{min}^{jk} \right\}$$
 (2.9)

where d^j_{min} is the minimum distance between the j^{th} robot and all the obstacles in the scene, while d^{jk}_{min} is the minimum distance between the j^{th} and the k^{th} robot. R^{ki} is the distance of the furthest point of the shape of the k^{th} robot to its i^{th} axis of rotation. Refer also to Figure 2.4.

Each configuration $Q \in \mathcal{B}$ is guaranteed to be inside the admitted region \underline{X} . This fact can be exploited for performing steering operation. Indeed, we can take as $Q_{steered}$ the pose at the border of $\mathcal{B}(Q_i)$ along the segment connecting Q_i to Q_k . It is not difficult to prove that such a state is equal to:

$$Q_{steered} = \begin{bmatrix} Q_{steered}^{1T} & \dots & Q_{steered}^{nT} \end{bmatrix}^{T} = Q_{i} + s_{steered}(Q_{k} - Q_{i})$$

$$s_{steered} = min \left\{ s_{A}, s_{B} \right\}$$

$$s_{A} = min_{j \in \{1, \dots, n\}, q} \left\{ \frac{d_{min}^{j}}{\sum_{q} R^{jq} |Q_{i}^{jq} - Q_{k}^{jq}|} \right\}$$

$$s_{B} = min_{j,k \in \{1, \dots, n\}, q, q_{2}} \left\{ \frac{d_{min}^{jk}}{\sum_{q} R^{jq} |Q_{i}^{jq} - Q_{k}^{jq}| + \sum_{q_{2}} R^{kq2} |Q_{i}^{kq_{2}} - Q_{k}^{kq_{2}}|} \right\}$$
(2.10)

Also in this case a multiple steer approach is possible for this strategy, refer also to Figure 2.5. Again, you can deploy your own geometric engine in order to compute the distances d_{min}^j , d_{min}^{jk} as well as the radii R^{ki} and define your custom Connector implementing the approach described in this Section.

¹where \bar{Q}^{jT} refers to the pose of the j^{th} robot, see equation (2.5).

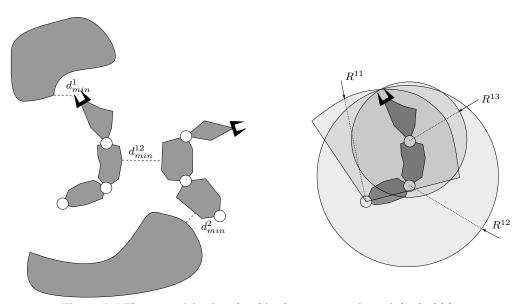


Figure 2.4 The quantities involved in the computation of the bubble \mathcal{B} .

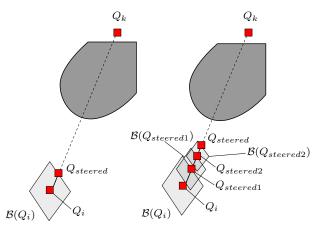


Figure 2.5 Single (left) and multiple (right) steer using the bubbles of free configurations.

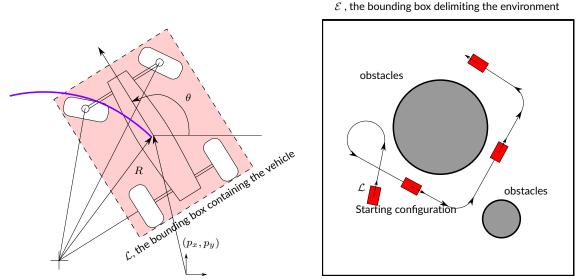


Figure 2.6 Vehicle motion in a planar environment.

2.3 Navigation problem

This problem is typical when considering autonomous vehicle. We have a 2D map in which a cart must move. In order to simplify the collision check task, a bounding box $\mathcal L$ is assumed to contain the entire shape of the vehicle, Figure 2.6. The cart moves at a constant velocity when advancing on a straight line and cannot change instantaneously its cruise direction. Indeed, the cart has a steer, which allows to do a change direction by moving on a portion of a circle, refer to Figure 2.6. We assume that possible the steering radius R in a compact interval $[R_{min}, R_{max}]$.

Since the cart is a rigid body, its position and orientation in the plane can be completely described using three quantities: the coordinates p_x, p_y of its center of gravity and the absolute angle θ . Therefore, a configuration $x_i \in \mathcal{X}$ is a vector defined as follows: $x_j = \begin{bmatrix} p_{xi} & p_{yi} & \theta_i \end{bmatrix}$. The admitted region $\underline{\mathcal{X}}$ is made by all the configurations x for which the vehicle results to be not in collision with any obstacles populating the scene.

2.3.1 Sampling

The environment where the vehicle can move is assumed to be finite and equal to a bounding box \mathcal{E} with certain sizes, right portion of Figure 2.6. The sampling of a random configuration for the vehicle is done in this way:

$$x_R = egin{bmatrix} (p_{xi}, p_{yi}) \sim \mathcal{E} \\ heta \sim U(-\pi, \pi) \end{bmatrix}$$
 (2.11)

2.3.2 Optimal trajectory and constraints

The optimal trajectory connecting two configurations x_i, x_j is made of three parts ² (refer to the examples in the right part of Figure 2.6 and the top part of Figure 2.7):

• a straight line starting from x_i

²Except when the starting and ending configuration have the same orientation and lies on the same line. In that case the trejectory is a simple segment connecting the 2 states.

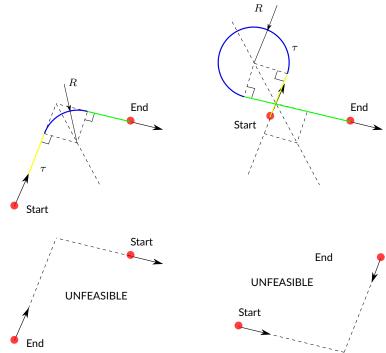


Figure 2.7 Examples of feasible, top, and non feasible trajectories, bottom. The different parts of the feasible trajectories are highlighted with different colors.

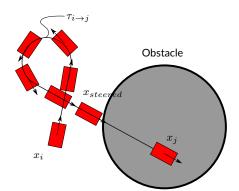


Figure 2.8 Steering procedure for a planar navigation problem.

- a circular portion motion used to get from θ_i to θ_j
- a straight line ending in x_i

The cost $C(\tau)$ is assumed to be the total length of τ . It is worthy to remark that not for every pair of configurations exists a trajectory connecting them, refer to Figure 2.7. Therefore, in case the trajectory $\tau_{i \to j}$ does not exists, $C(\tau_{i \to j})$ is assumed equal to $+\infty$.

2.3.3 Advancement along the optimal trajectory

The steering from a state x_i toward another x_j is done by moving along the trajectory $\tau_{i \to j}$, advancing every time of a little quantity of space (also when traversing the circular part of the trajectory). The procedure is arrested when a configuration not lying in $\underline{\mathcal{X}}$ is found or x_j is reached. Figure 2.8 summarizes the steering procedure.

Parallel RRT

This Chapter will provide details about the multi-threaded strategies provided by MT-RRT. The original implementation was proposed in [3], which is the publication were for the first time MT-RRT was presented. That document has to be taken with a grain of salt, has the current implementation was improved with a few more optimizations.

3.0.1 Parallelization of the guery activities

All the RRT versions spend a significant time in performing query operations on the tree, i.e. operations that require to traverse all the tree. Such operations are mainly the nearest neighbour search, algorithm 3, and the determination of the near set, algorithm 6.

The key idea is to perform the above query operations by making use of a thread pool implementing parallel for regions, where at an average all the threads process the same amount of nodes in the tree, computing their distances for determine the nearest neighbour or the near set. All the threads in the pool are spawn when a new planning problem must be solved and remain active and ready to perform the parallel for described before. All the operations of the RRT (regardless the version considered) are done by the main thread, which notifies at the proper time when a new query operation must be process collectively by all the threads. Figure 3.1.a summarizes the approach.

The object implementing this approach is ParallelizedQueriesPlanner.

3.0.2 Shared tree critical regions

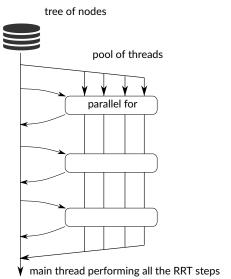
Another way to obtain a parallelization is to simultaneously do every step of the RRT expansions. Therefore, we can imagine having threads sharing a common tree (or two trees in the case of a bidirectional strategy), executing in parallel every step of the expansion process. The data structure representing the shared tree can be a lock free list that does not require the adoption of any critical regions for adding elements at the end of the list as well as iterate the list elements. Therefore, in case of the classical RRT, threads will be simply perform in parallel the expansions steps: sample a random state, find the nearest neighbour, steer a state and then push back the newly added state to the shared lock free list. At the same time, in case of RRT* some critical regions are required for synchronizing the threads when doing rewires. Notice that such critical regions are achieved by making use of a spin lock instead of a traditional heavy duty mutex in order to save as much as possible time. Figure 3.1.b summarizes the approach. The object implementing this approach is SharedTreePlanner.

3.0.3 Parallel expansions on linked trees

To limit as much as possible the overheads induced by the presence of critical sections, we can consider a version similar to the one proposed in the previous Section, but for which every thread has a private copy of the search tree. After a new node is added by a thread to its own tree, P-1 copies are computed and dispatched 1 to the other threads, were P is the number of working threads. Sporadically, all the threads

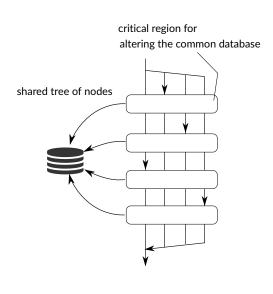
¹They are dispatched into proper buffer, but not directly inserted in the private copies of the other trees.

18 Parallel RRT



(a) Schematic

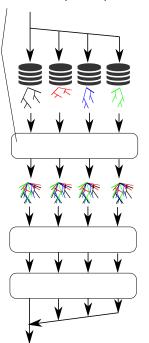
representation of the parallelization of the query activities approach.



Schematic representation of the parallel extensions of a common tree approach.

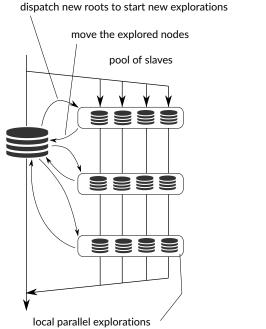
(b)

add to the local tree the nodes explored by the others



(c) Schematic repre-

sentation of the parallel expansions of copied trees approach.



Schematic representation of the multi agent approach.

Figure 3.1 Approaches adopted for parallelize RRT.

take into account the list of nodes received from the others and insert them into their private trees. This mechanism is able to avoid the simultaneous modification of a tree by two different threads, avoiding the use of critical sections.

When considering the bidirectional RRT, the mechanism is analogous but introducing for every thread a private copy of both the involved trees.

Instead, the RRT* version is slightly modified. Indeed, the rewirds done by a thread on its own tree are not dispatched to the others. At the same time, each thread consider all the nodes produced and added to its own tree when doing their own rewirds. When searching the best solution at the end of all the iterations, the best connections among all the trees in every threads are taken into account. Indeed, the predecessor of a node is assumed to be the parent with the lowest cost to go among the ones associated to each clones. Figure 3.1.c summarizes the approach.

Clearly, the amount memory required by this approach is significantly high, since multiple copies of a node must live in the different threads. This can be a problem to account for.

The object implementing this approach is LinkedTreesPlanner.

3.0.4 Multi agents approach

The strategy described in this Section aims at exploiting a significant number of threads, with both a reduced synchronizing need and allocation memory requirements. To this purpose, a variant of the RRT was developed for which every exploring thread has not the entire knowledge of the tree, but it is conscious of a small portion of it. Therefore, we can deploy many threads to simultaneously explore the state space $\mathcal X$ (ignoring the results found by the other agents) for a certain amount of iterations. After completing this sub-exploration task, all data incoming from the agents are collected and stored in a centralized data base while the agents wait to begin a new explorative batch, completely forgetting the nodes found at the previous iteration. The described behaviour resembles one of many exploring ants, which reports the exploring data to a unique anthill.

Notice that there is no need to physically copy the states computed by the agents when inserting them into the central database, since threads share a common memory: the handler of the node is simply moved. When considering this approach a bidirectional search is not implementable, while the RRT* can be extended as reported in the following. Essentially, the agents perform a standard non-optimal exploration, implementing the steps of a canonical RRT, Section 1.2.1. Then, at the time of inserting the nodes into the common database, the rewirds are done.

The described multi agent approach is clearly a modification of the canonical RRT versions, since the agents start exploring every time from some new roots, ignoring all the previously computed nodes. However, it was empirically found that the global behaviour of the path search is not deteriorated and the optimality properties of the RRT* seems to be preserved.

Before concluding this Section it is worthy to notice that the mean time spent for the querying operations is considerably lower, since such operations are performed by agents considering only their own local reduced size trees.

Figure 3.1.d summarizes the approach. The object implementing this approach is MultiAgentPlanner.

20 Parallel RRT

Namespace Index

4.1 Namespace List

Here is a list of all documented namespaces with brief descriptions:	
mt_rrt	29

22 Namespace Index

Hierarchical Index

5.1 Class Hierarchy

This inheritance list is sorted roughly, but not completely, alphabetically:

$mt_rrt::Copiable < T > \dots \dots$
mt_rrt::Copiable < Connector >
mt_rrt::Connector
mt_rrt::TunneledConnector
mt_rrt::Copiable < Sampler >
mt_rrt::Sampler
mt_rrt::HyperBox
mt_rrt::DescriptionAndParameters
mt_rrt::DeterminismRegulator
$mt_rrt:: detail:: Distribution < Distribution Shape > $
$mt_rrt:: detail:: Distribution < std:: normal_distribution < float >> \dots \dots$
mt_rrt::GaussianEngine
$mt_rrt:: detail:: Distribution < std::uniform_real_distribution < float >> \dots \dots \dots \dots \dots 37$
mt_rrt::UniformEngine
mt_rrt::HyperBox
mt_rrt::KeepSearchPredicate
mt_rrt::Limited< T >
$mt_rrt::LowerLimited < std::size_t > \dots $
mt_rrt::Iterations
mt_rrt::SteerIterations
mt_rrt::Threads
mt_rrt::LowerLimited < T >
mt_rrt::Positive < float >
$mt_rrt::$ Positive $<$ T $>$
$mt_rrt::UpperLimited < T > \dots $
mt_rrt::Limited < float >
mt_rrt::Determinism
mt_rrt::SynchronizationDegree
$mt_rrt::LockFreeForwardList < T > \dots $
mt_rrt::NearestQuery
mt_rrt::NearSet
mt_rrt::NearSetElement
mt_rrt::NearSetQuery

24 Hierarchical Index

mt_rrt::Node
mt_rrt::NodeOwning
$mt_rrt::NodesAllocatorT < NodeT > \ \ . \ \ \ . \ \ \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ \ \ \ \ . \ \ \ \ . \ \ . \ \ . \ \ . \ \ . \ \ . \ \ \ \ . \ \ \ \ . \ \ \ \ . \ \ \ \ \ . \ \ \ \ . \ \ \ \ . \ \ \ \ \ . \ \ \ \ . \ \ \ \ . \ \ \ \ . \ \ \ . \ \ \ \ . \ \ \ \ . \ \ \ \ . \ \ \ \ \ $
$mt_rrt::Nodes Allocator T < Node > \dots $
$mt_rrt::ObjectPool < T > \dots \dots$
$mt_rrt:: Object Pool < float > \dots $
$mt_rrt::ObjectPool < NodeT > \dots $
mt_rrt::ParallelFor
mt_rrt::Parameters
mt_rrt::PlannerSolution
mt_rrt::ProblemAware
mt_rrt::Extender
mt_rrt::ExtenderBidirectional
mt_rrt::ExtenderSingle
mt_rrt::Planner
mt_rrt::MultiThreadedPlanner
mt_rrt::EmbarassinglyParallelPlanner
mt_rrt::LinkedTreesPlanner
mt_rrt::MultiAgentPlanner
mt_rrt::ParallelizedQueriesPlanner
mt_rrt::SharedTreePlanner
mt_rrt::StandardPlanner
mt_rrt::TreeHandler
mt_rrt::TreeHandlerBasic
mt_rrt::ProblemDescription
mt_rrt::ProblemDescriptionCloner
mt_rrt::MultiThreadedPlanner
mt_rrt::Rewire
std::runtime error
mt_rrt::Error
mt_rrt::Solution
mt_rrt::BidirSolution
mt_rrt::SimpleSolution
mt_rrt::LinkedTreesPlanner
mt_rrt::MultiAgentPlanner
mt_rrt::Trajectory
mt_rrt::TunneledConnector::Line
mt rrt::View

Class Index

6.1 Class List

Here are the classes, structs, unions and interfaces with brief descriptions:

mt_rrt::BidirSolution	33
mt_rrt::Connector	34
Factory of optimal trajectories, refer to Section 1.3.0.1 of the documentation mt rrt::Copiable < T >	36
mt_rrt::DescriptionAndParameters	36
mt_rrt::Determinism	30
Used to regulate the deterministic bias, see documentation at Section "Background on	
	36
RRT"	37
mt_rrt::detail::Distribution < DistributionShape >	37
mt_rrt::EmbarassinglyParallelPlanner	37
Into N different threads, N different solution are found running the classical version of	
the planner. The best solution among thosw found by all threads is returned	38
mt_rrt::Error	38
mt rrt::Extender	50
Used to extend one or two connected search trees	39
mt rrt::ExtenderBidirectional	40
mt_rrt::ExtenderSingle	41
mt_rrt::GaussianEngine	71
Used to draw samples from a normal distribution	42
mt rrt::HyperBox	72
A sampler drawing samples inside an n-dimensioned hypercube described by 2 corners.	
For example, corners [11, 12, 13, 14] and [u1, u2, u3, u4], describe an hyperbox whose	
points $[x1,x2,x3,x4]$ are all such that: $ x < xi < yi$ ui	43
mt rrt::Iterations	
Iterations limits to find a solution	45
mt rrt::KeepSearchPredicate	45
mt_rrt::Limited< T >	
A quantity whose value should always remain between defined bounds	46
mt rrt::TunneledConnector::Line	47
mt_rrt::LinkedTreesPlanner	
Approach described at Section "Parallel expansions on linked trees" of the documenta-	
tion	49
mt_{rrt} ::LockFreeForwardList $\ldots\ldots\ldots\ldots\ldots$	49
mt_rrt::LowerLimited< T >	
A @Limited quantity, having infinity as upper bound	50

26 Class Index

mt_rrt::MultiAgentPlanner	
Approach described at Section "Multi agents approach" of the documentation	50
mt_rrt::MultiThreadedPlanner	51
mt_rrt::NearestQuery	52
mt_rrt::NearSet	52
mt_rrt::NearSetElement	52
mt_rrt::NearSetQuery	52
mt rrt::Node	-
Used for representing a state x \in \underline{\mathcal{X}}, Section 1.2 of the documen-	
tation. This class is used internally to extend search trees. The user is not expected to	
consume it	53
mt rrt::NodeOwning	54
-	
mt_rrt::NodesAllocatorT< NodeT >	55
mt_rrt::ObjectPool< T >	55
mt_rrt::ParallelFor	56
mt_rrt::ParallelizedQueriesPlanner	
Approach described at Section "Parallelization of the query activities" of the documen-	
tation	56
mt_rrt::Parameters	
Groups together all the parameters that a @Planner neeeds to know to solve a specific	
problem for connecting 2 pair of states	57
mt_rrt::Planner	
In order to solve any kind of problem you need to build a kind of Planner, passing for	
sure the @ProblemDescription. The same @Planner, can be used to solve (one at a time)	
multiple problems, calling many times Planner::solve(), possibly using different (or not)	
@Parameters every time	57
mt rrt::PlannerSolution	٥,
Groups all the information characterizing a found solution	58
mt rrt::Positive < T >	50
A @LowerLimited quantity, having 0.0 as lower bound	59
mt rrt::ProblemAware	37
-	40
Someone aware of the static description of the class of problems to solve	60
mt_rrt::ProblemDescription	
Groups together all the static information characterizing the class of problems to solve	60
mt_rrt::ProblemDescriptionCloner	61
mt_rrt::Rewire	61
mt_rrt::Sampler	
Interface for a sampler of states	62
mt_rrt::SharedTreePlanner	
Approach described at Section "Shared tree critical regions" of the documentation	63
mt_rrt::SimpleSolution	64
mt rrt::Solution	64
mt rrt::StandardPlanner	
Classical mon-thread solver described in Section "Background on RRT" of the documen-	
tation	65
mt rrt::SteerIterations	00
NUmber of times to try steer, refer to documentation at Section "Background on RRT"	66
mt rrt::Connector::SteerResult	00
-	
Performs a steering operation, Sections 1.2.1, 1.2.2 and 1.2.3 of the documentation,	,,
from a staring node to a target one	66
mt_rrt::SynchronizationAware	67
mt_rrt::SynchronizationDegree	
Synchronization degree used by MultiAgentPlanner and LinkedTreesPlanner. This regu-	
lates how often the threads stop expansions and share info about explored states with	
other threads	67
mt_rrt::Threads	
Number of threads used by a MultiThreadedPlanner	68

6.1 Class List

mt_rrt::Trajectory	
Interface describing an optimal trajectory \tau connecting 2 states, Section 1.2 of the documentation, in a particular problem to solve. After construction, the state on the	
trajectory is assumed at the beginning of the trajectory itself. Calling advance(), shift	
the cursor along the trajectory	69
mt_rrt::TreeHandler	70
mt_rrt::TreeHandlerBasic	71
mt_rrt::TunneledConnector	
Implements the approach described at Section "Tunneled check collision" of the documentation. Optimal trajectories \tau are assumed to be simple line in the space of configuration. Every advancement move the cursor of a qunatum of space along the trajectory and only the reached state is checked to be in the admitted space (and not	
the segment going from the previous state to the reached one)	72
mt_rrt::UniformEngine	
Used to draw sample whithin a compact inverval [I, U]	74
mt_rrt::UpperLimited< T >	
A @Limited quantity, having negative infinity as lower bound	75
mt_rrt::View	76

28 Class Index

Chapter 7

Namespace Documentation

7.1 mt_rrt Namespace Reference

Classes

class EmbarassinglyParallelPlanner

Into N different threads, N different solution are found running the classical version of the planner. The best solution among thosw found by all threads is returned.

class LinkedTreesPlanner

Approach described at Section "Parallel expansions on linked trees" of the documentation.

- class LockFreeForwardList
- class MultiAgentPlanner

Approach described at Section "Multi agents approach" of the documentation.

- class ProblemDescriptionCloner
- class Threads

Number of threads used by a MultiThreadedPlanner.

- class MultiThreadedPlanner
- class ParallelFor
- class ParallelizedQueriesPlanner

Approach described at Section "Parallelization of the query activities" of the documentation.

• class SharedTreePlanner

Approach described at Section "Shared tree critical regions" of the documentation.

class SynchronizationDegree

Synchronization degree used by MultiAgentPlanner and LinkedTreesPlanner. This regulates how often the threads stop expansions and share info about explored states with other threads.

- class SynchronizationAware
- class Copiable
- class Error
- class Limited

A quantity whose value should always remain between defined bounds.

class LowerLimited

A @Limited quantity, having infinity as upper bound.

class Positive

A @LowerLimited quantity, having 0.0 as lower bound.

class UpperLimited

A @Limited quantity, having negative infinity as lower bound.

- class ObjectPool
- class GaussianEngine

Used to draw samples from a normal distribution.

class UniformEngine

Used to draw sample whithin a compact inverval [I, U].

- struct View
- class Connector

Factory of optimal trajectories, refer to Section 1.3.0.1 of the documentation.

- struct KeepSearchPredicate
- class DeterminismRegulator
- class Extender

Used to extend one or two connected search trees.

- class BidirSolution
- class ExtenderBidirectional
- class SimpleSolution
- class ExtenderSingle
- struct NearestQuery
- struct NearSetQuery
- class Node

Used for representing a state $x \in \mathbb{X}$, section 1.2 of the documentation. This class is used internally to extend search trees. The user is not expected to consume it.

- class NodesAllocatorT
- class NodeOwning
- struct PlannerSolution

Groups all the information characterizing a found solution.

class Planner

In order to solve any kind of problem you need to build a kind of Planner, passing for sure the @ProblemDescription. The same @Planner, can be used to solve (one at a time) multiple problems, calling many times Planner::solve(...), possibly using different (or not) @Parameters every time.

struct ProblemDescription

Groups together all the static information characterizing the class of problems to solve.

- struct DescriptionAndParameters
- class ProblemAware

Someone aware of the static description of the class of problems to solve.

class Sampler

Interface for a sampler of states.

class HyperBox

- class Solution
- class StandardPlanner

the classical mon-thread solver described in Section "Background on RRT" of the documentation.

class Trajectory

Interface describing an optimal trajectory \t au connecting 2 states, Section 1.2 of the documentation, in a particular problem to solve. After construction, the state on the trajectory is assumed at the beginning of the trajectory itself. Calling advance(), shift the cursor along the trajectory.

- struct NearSetElement
- struct NearSet
- struct Rewire
- class TreeHandler
- class TreeHandlerBasic
- class TunneledConnector

Implements the approach described at Section "Tunneled check collision" of the documentation. Optimal trajectories \tau are assumed to be simple line in the space of configuration. Every advancement move the cursor of a qunatum of space along the trajectory and only the reached state is checked to be in the admitted space (and not the segment going from the previous state to the reached one).

class Determinism

used to regulate the deterministic bias, see documentation at Section "Background on RRT"

class Iterations

iterations limits to find a solution.

class SteerIterations

NUmber of times to try steer, refer to documentation at Section "Background on RRT".

struct Parameters

Groups together all the parameters that a @Planner neeeds to know to solve a specific problem for connecting 2 pair of states.

Typedefs

- using **Seed** = unsigned
- using ConnectorPtr = std::unique_ptr< Connector >
- using ExtenderPtr = std::unique_ptr< Extender >
- using NodesAllocator = NodesAllocatorT < Node >
- using ProblemDescriptionPtr = std::shared_ptr< const ProblemDescription >
- using SamplerPtr = std::unique_ptr< Sampler >
- using Solutions = std::vector< std::shared ptr< Solution > >
- using TrajectoryPtr = std::unique_ptr < Trajectory >
- using TreeHandlerPtr = std::unique_ptr < TreeHandler >
- using Cost = Positive < float >

Enumerations

• enum class ExpansionStrategy { Single , Bidir , Star }

The kind of strategy to use, refer to documentation at Sections 1.2.1, 1.2.2 and 1.2.3.

Functions

- Seed make random seed ()
- template<typename... Args> std::string merge (Args &&...slices_to_merge)

put all the passed slices all together into a single string, that is returned.

• template<typename lterT >

const Node * nearest_neighbour (const View &state, IterT tree_begin, IterT tree_end, const DescriptionAndParameters &context)

template<typename IterT >

std::vector< NearSetElement > near_set (const View &state, IterT tree_begin, IterT tree_end, std
::size_t size, const DescriptionAndParameters &context)

- std::vector< Rewire > compute_rewires (Node &candidate, NearSet &&near_set, const DescriptionAndParameters &context)
- std::optional < Connector::SteerResult > extend (const View & target, TreeHandler & tree_handler, const bool is deterministic)
- std::optional< Connector::SteerResult > extend_star (const View &target, TreeHandler &tree_← handler, const bool is_deterministic, std::vector< Rewire > &rewires)
- void apply_rewires_if_better (const Node &parent, const std::vector < Rewire > &rewires)
- std::vector< std::vector< float > > sequence_from_root (const Node &subject)
- void **sort_solutions** (Solutions & subject)
- std::vector < std::vector < float > > find_best_solution (const Solutions &subject)
- float near_set_ray (std::size_t tree_size, std::size_t problem_size, float gamma)
- float euclidean_distance (const View &a, const View &b)

7.1.1 Detailed Description

Author: Andrea Casalino Created: 16.05.2019

report any bug to andrecasa91@gmail.com.

Author: Andrea Casalino Created: 16.02.2021

report any bug to andrecasa91@gmail.com.

7.1.2 Function Documentation

7.1.2.1 make_random_seed()

```
Seed mt_rrt::make_random_seed ( )
```

Returns

time since start of the process is used to generated the seed

Chapter 8

Class Documentation

8.1 mt_rrt::BidirSolution Class Reference

Inheritance diagram for mt_rrt::BidirSolution:



Public Member Functions

- BidirSolution (const Node *byPassFront, const Node *byPassBack, float cost2Back)
- std::vector< std::vector< float > > getSequence () const final
- float cost () const final

Public Attributes

- const Node * byPassFront
- const Node * byPassBack
- float cost2Back

The documentation for this class was generated from the following file:

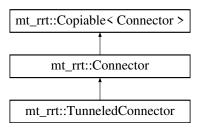
• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/ExtenderBidir.h

8.2 mt_rrt::Connector Class Reference

Factory of optimal trajectories, refer to Section 1.3.0.1 of the documentation.

```
#include <Connector.h>
```

Inheritance diagram for mt_rrt::Connector:



Classes

struct SteerResult

Performs a steering operation, Sections 1.2.1, 1.2.2 and 1.2.3 of the documentation, from a staring node to a target one.

Public Member Functions

- Connector (const Connector &)=delete
- Connector & operator= (const Connector &)=delete
- virtual float minCost2Go (const View &start, const View &end) const =0
- float minCost2GoConstrained (const View &start, const View &end) const
- virtual TrajectoryPtr getTrajectory (const View &start, const View &end) const =0
- std::optional< SteerResult > steer (const Node &start, const View &target, const SteerIterations &trials) const

8.2.1 Detailed Description

Factory of optimal trajectories, refer to Section 1.3.0.1 of the documentation.

8.2.2 Member Function Documentation

8.2.2.1 getTrajectory()

Parameters

the	starting state
the	ending state

Returns

the optimal trajectory connecting the passed states. nullptr is returned in case a feasible trajectory does not exist

Implemented in mt_rrt::TunneledConnector.

8.2.2.2 minCost2Go()

Parameters

the	starting node in the trajectory whose cost is to evaluate
the	ending node in the trajectory whose cost is to evaluate

Returns

the cost C(\tau), Section 1.2 of the documentation, to traverse the optimal trajectory connecting the passed pair of states. This cost does not account for constraints, as this is done by $minCost2 \leftarrow GoConstrained(...)$

Implemented in mt_rrt::TunneledConnector.

8.2.2.3 minCost2GoConstrained()

Parameters

the	starting node in the trajectory whose cost is to evaluate
the	ending node in the trajectory whose cost is to evaluate

Returns

the cost C(\tau), Section 1.2 of the documentation, of the trajectory \tau going from the starting node to the ending one, but accounting for the constraints, Section 1.2 of the documentation. COST_MAX is returned in this case returned when a feasible trajectory exists, but is not entirely contained in the admitted set. Section 1.2 of the documentation.

The documentation for this class was generated from the following file:

• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Connector.h

8.3 mt_rrt::Copiable < T > Class Template Reference

Public Member Functions

virtual std::unique_ptr< T > copy () const =0
 A deep copy needs to be implemented.

The documentation for this class was generated from the following file:

• /home/andrea/Scrivania/GitProj/MT-RRT/src/carpet/header/MT-RRT/Copiable.h

8.4 mt_rrt::DescriptionAndParameters Struct Reference

Public Attributes

- const ProblemDescription & description
- const Parameters & parameters

The documentation for this struct was generated from the following file:

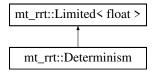
/home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/ProblemDescription.h

8.5 mt_rrt::Determinism Class Reference

used to regulate the deterministic bias, see documentation at Section "Background on RRT"

```
#include <Types.h>
```

Inheritance diagram for mt_rrt::Determinism:



Public Member Functions

Determinism (const float val)

Additional Inherited Members

8.5.1 Detailed Description

used to regulate the deterministic bias, see documentation at Section "Background on RRT"

The documentation for this class was generated from the following file:

• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Types.h

8.6 mt_rrt::DeterminismRegulator Class Reference

Public Member Functions

- DeterminismRegulator (const Seed &seed, const Determinism &determinism)
- bool doDeterministicExtension () const

The documentation for this class was generated from the following file:

• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Extender.h

8.7 mt_rrt::detail::Distribution < DistributionShape > Class Template Reference

Public Member Functions

- float sample () const
- **Distribution** (const Distribution &o)
- Distribution & operator= (const Distribution &o)=delete
- **Distribution** (Distribution &&o)=delete
- Distribution & operator= (Distribution &&o)=delete
- Seed sampleSeed () const

Protected Member Functions

Distribution (DistributionShape &&shape, const std::optional < Seed > &seed)

The documentation for this class was generated from the following file:

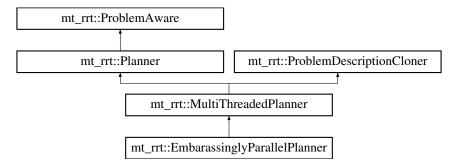
• /home/andrea/Scrivania/GitProj/MT-RRT/src/carpet/header/MT-RRT/Random.h

8.8 mt_rrt::EmbarassinglyParallelPlanner Class Reference

Into N different threads, N different solution are found running the classical version of the planner. The best solution among thosw found by all threads is returned.

#include <EmbarassinglyParallel.h>

Inheritance diagram for mt_rrt::EmbarassinglyParallelPlanner:



Public Member Functions

template<typename... Args>
 MultiThreadedPlanner (Args &&...args)

default number of threads is assumed equal to 4. This can be changed after construction by using MultiThreaded \leftarrow Planner::setThreads or MultiThreadedPlanner::setMaxThreads

Protected Member Functions

• void **solve_** (const std::vector< float > &start, const std::vector< float > &end, const Parameters ¶meters, PlannerSolution &recipient) final

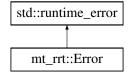
8.8.1 Detailed Description

Into N different threads, N different solution are found running the classical version of the planner. The best solution among thosw found by all threads is returned.

The documentation for this class was generated from the following file:

8.9 mt_rrt::Error Class Reference

Inheritance diagram for mt_rrt::Error:



Public Member Functions

- Error (const std::string &what)
- template<typename T1, typename T2, typename... Args>
 Error (const T1 &first, const T2 &second, Args &&...slices to merge)

The documentation for this class was generated from the following file:

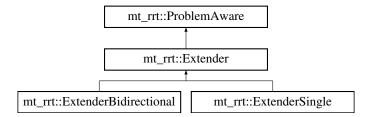
/home/andrea/Scrivania/GitProj/MT-RRT/src/carpet/header/MT-RRT/Error.h

8.10 mt_rrt::Extender Class Reference

Used to extend one or two connected search trees.

#include <Extender.h>

Inheritance diagram for mt_rrt::Extender:



Public Member Functions

• std::size_t search ()

Perform the specified number of estensions on the wrapped tree(s). This function may be called multiple times, for performing batch of extensions. All the solutions found while extending are saved and stored in this object.

- virtual std::vector< TreeHandlerPtr > dumpTrees ()=0
- const Solutions & getSolutions () const
- Solutions & getSolutions ()

Protected Member Functions

- Extender (const TreeHandler &handler)
- virtual void search_iteration ()=0

Protected Attributes

- const Parameters & parameters
- Solutions solutions
- std::optional < DeterminismRegulator > determinism_manager

8.10.1 Detailed Description

Used to extend one or two connected search trees.

8.10.2 Member Function Documentation

8.10.2.1 search()

```
std::size_t mt_rrt::Extender::search ( )
```

Perform the specified number of estensions on the wrapped tree(s). This function may be called multiple times, for performing batch of extensions. All the solutions found while extending are saved and stored in this object.

Parameters

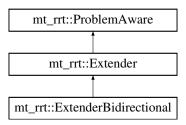
```
the | number of extension to perform
```

The documentation for this class was generated from the following file:

• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Extender.h

8.11 mt_rrt::ExtenderBidirectional Class Reference

Inheritance diagram for mt_rrt::ExtenderBidirectional:



Public Member Functions

- ExtenderBidirectional (TreeHandlerPtr front, TreeHandlerPtr back)
- std::vector< TreeHandlerPtr > dumpTrees () final

Public Attributes

- TreeHandlerPtr front_handler
- TreeHandlerPtr back_handler

Protected Member Functions

• void search_iteration () final

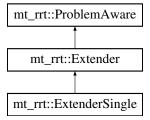
Additional Inherited Members

The documentation for this class was generated from the following file:

/home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/ExtenderBidir.h

8.12 mt_rrt::ExtenderSingle Class Reference

Inheritance diagram for mt_rrt::ExtenderSingle:



Public Member Functions

- ExtenderSingle (TreeHandlerPtr handler, const std::vector< float > &target)
- std::vector< TreeHandlerPtr > dumpTrees () final

Public Attributes

- std::vector< float > target
- TreeHandlerPtr tree_handler

Protected Member Functions

• void search_iteration () final

Additional Inherited Members

The documentation for this class was generated from the following file:

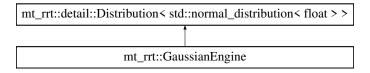
/home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/ExtenderSingle.h

8.13 mt_rrt::GaussianEngine Class Reference

Used to draw samples from a normal distribution.

```
#include <Random.h>
```

Inheritance diagram for mt_rrt::GaussianEngine:



Public Member Functions

GaussianEngine (const float &mean=0, const float &stdDeviation=1.f, const std::optional < Seed > &seed=std::nullopt)

Additional Inherited Members

8.13.1 Detailed Description

Used to draw samples from a normal distribution.

8.13.2 Constructor & Destructor Documentation

8.13.2.1 GaussianEngine()

Parameters

t	the	mean of the normal distribution
t	the	standard deviation of the normal distribution

The documentation for this class was generated from the following file:

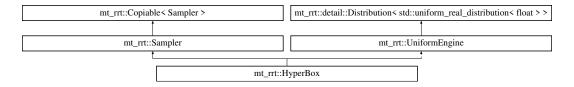
/home/andrea/Scrivania/GitProj/MT-RRT/src/carpet/header/MT-RRT/Random.h

8.14 mt_rrt::HyperBox Class Reference

A sampler drawing samples inside an n-dimensioned hypercube described by 2 corners. For example, corners [I1, I2, I3, I4] and [u1, u2, u3, u4], describe an hyperbox whose points [x1,x2,x3,x4] are all such that: Ii <= xi <= ui.

```
#include <Sampler.h>
```

Inheritance diagram for mt_rrt::HyperBox:



Public Member Functions

- HyperBox (const std::vector< float > &lowerCorner, const std::vector< float > &upperCorner, const std::optional< Seed > &seed=std::nullopt)
- std::unique_ptr< Sampler > copy () const override

A deep copy needs to be implemented.

• std::vector< float > sampleState () const override

Returns a state randomly sampled in the \mathcal{X} space, Sections 1.2.1, 1.2.2 and 1.2.3 of the documentation. This random state are used for randomly growing searching trees.

- Seed sampleSeed () const final
- const auto & minCorner () const
- std::vector< float > maxCorner () const

Additional Inherited Members

8.14.1 Detailed Description

A sampler drawing samples inside an n-dimensioned hypercube described by 2 corners. For example, corners [I1, I2, I3, I4] and [u1, u2, u3, u4], describe an hyperbox whose points [x1,x2,x3,x4] are all such that: Ii <= xi <= ui.

8.14.2 Constructor & Destructor Documentation

8.14.2.1 HyperBox()

Parameters

the	lower corner of the hyperbox
the	upper corner of the hyperbox

Exceptions

if

lowerCorner and upperCorner size mismatch or some of the values inside lowerCorner are greater than ones in upperCorner

8.14.3 Member Function Documentation

8.14.3.1 sampleSeed()

```
Seed mt_rrt::HyperBox::sampleSeed ( ) const [inline], [final], [virtual]
```

Returns

a random seed to use for intializing another Sampler.

Implements mt_rrt::Sampler.

8.14.3.2 sampleState()

```
std::vector<float> mt_rrt::HyperBox::sampleState ( ) const [override], [virtual]
```

Returns a state randomly sampled in the \mathcal{X} space, Sections 1.2.1, 1.2.2 and 1.2.3 of the documentation. This random state are used for randomly growing searching trees.

Returns

a drawn random state.

Implements mt_rrt::Sampler.

The documentation for this class was generated from the following file:

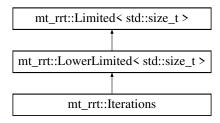
• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Sampler.h

8.15 mt_rrt::Iterations Class Reference

iterations limits to find a solution.

```
#include <Types.h>
```

Inheritance diagram for mt_rrt::Iterations:



Public Member Functions

• Iterations (const std::size_t iters)

Additional Inherited Members

8.15.1 Detailed Description

iterations limits to find a solution.

The documentation for this class was generated from the following file:

• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Types.h

8.16 mt_rrt::KeepSearchPredicate Struct Reference

Public Member Functions

• bool operator() (std::size_t iter) const

Public Attributes

- bool best_effort
- std::size_t max_iterations
- ExpansionStrategy strategy
- std::atomic_bool one_solution_was_found = false

The documentation for this struct was generated from the following file:

• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Extender.h

8.17 mt_rrt::Limited < T > Class Template Reference

A quantity whose value should always remain between defined bounds.

```
#include <Limited.h>
```

Inheritance diagram for mt_rrt::Limited< T >:



Public Member Functions

- Limited (const T &lowerBound, const T &upperBound, const T &initialValue)
- template<bool LowerOrUpper = true>

Limited (const T &lowerBound, const T &upperBound)

similar to Limited::Limited(const T& lowerBound, const T& upperBound, const T& initialValue), assuming lower← Bound as initial value

- Limited (const Limited &)=default
- Limited & operator= (const Limited &)=default
- T getLowerBound () const
- T getUpperBound () const
- T get () const
- void set (const T &newValue)

Protected Attributes

- T value
- T lowerBound
- T upperBound

8.17.1 Detailed Description

```
template<typename T> class mt rrt::Limited< T>
```

A quantity whose value should always remain between defined bounds.

8.17.2 Constructor & Destructor Documentation

8.17.2.1 Limited()

Parameters

lower	bound for the value
upper	bound for the value
initial	value to set

8.17.3 Member Function Documentation

8.17.3.1 get()

```
template<typename T >
T mt_rrt::Limited< T >::get ( ) const [inline]
```

Returns

the current value

8.17.3.2 set()

Parameters

the new value to assumed

Exceptions

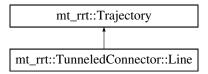
if the value is not consistent with the bounds

The documentation for this class was generated from the following file:

• /home/andrea/Scrivania/GitProj/MT-RRT/src/carpet/header/MT-RRT/Limited.h

8.18 mt_rrt::TunneledConnector::Line Class Reference

Inheritance diagram for mt_rrt::TunneledConnector::Line:



Public Member Functions

- Line (const View &start, const View &end, const TunneledConnector &caller)
- AdvanceInfo advance () final

Advance along the trajectory.

- View getState () const final
- float getCumulatedCost () const final

Additional Inherited Members

8.18.1 Member Function Documentation

8.18.1.1 getCumulatedCost()

```
float mt_rrt::TunneledConnector::Line::getCumulatedCost ( ) const [final], [virtual]
```

Returns

the cost to go from the beginning of the trajectory to the current reached state. IMPORTANT: it is a no-sense value in case last advance() returned blocked

Implements mt_rrt::Trajectory.

8.18.1.2 getState()

```
View mt_rrt::TunneledConnector::Line::getState ( ) const [inline], [final], [virtual]
```

Returns

the current state on the trajectory. IMPORTANT: it is a no-sense value in case last this->advance() returned blocked

Implements mt_rrt::Trajectory.

The documentation for this class was generated from the following file:

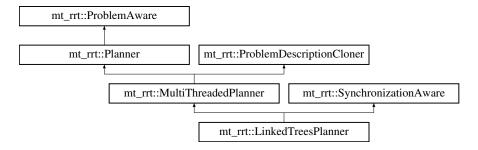
• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/TunneledConnector.h

8.19 mt_rrt::LinkedTreesPlanner Class Reference

Approach described at Section "Parallel expansions on linked trees" of the documentation.

#include <LinkedTreesPlanner.h>

Inheritance diagram for mt_rrt::LinkedTreesPlanner:



Public Member Functions

template<typename... Args>
 MultiThreadedPlanner (Args &&...args)

default number of threads is assumed equal to 4. This can be changed after construction by using MultiThreaded \leftarrow Planner::setThreads or MultiThreadedPlanner::setMaxThreads

Protected Member Functions

 void solve_ (const std::vector< float > &start, const std::vector< float > &end, const Parameters ¶meters, PlannerSolution &recipient) final

8.19.1 Detailed Description

Approach described at Section "Parallel expansions on linked trees" of the documentation.

The documentation for this class was generated from the following file:

8.20 mt_rrt::LockFreeForwardList< T > Class Template Reference

Public Member Functions

- template<typename... Args>
 LockFreeForwardList (Args &&...args_front)
- template<typename... Args>
 void emplace_back (Args &&...args)
- template<typename Pred > void forEach (Pred &&pred)
- std::size_t size () const
- T * getFront () const

The documentation for this class was generated from the following file:

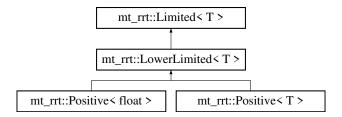
/home/andrea/Scrivania/GitProj/MT-RRT/src/multi-threaded/header/MT-RRT/LockFreeForward
 List.h

8.21 mt_rrt::LowerLimited < T > Class Template Reference

A @Limited quantity, having infinity as upper bound.

#include <Limited.h>

Inheritance diagram for mt_rrt::LowerLimited< T >:



Public Member Functions

- LowerLimited (const T &lowerBound, const T &initialValue)
- LowerLimited (const T &lowerBound)

Additional Inherited Members

8.21.1 Detailed Description

```
template<typename T> class mt_rrt::LowerLimited< T >
```

A @Limited quantity, having infinity as upper bound.

The documentation for this class was generated from the following file:

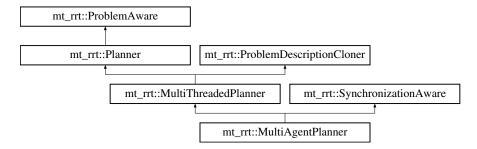
• /home/andrea/Scrivania/GitProj/MT-RRT/src/carpet/header/MT-RRT/Limited.h

8.22 mt_rrt::MultiAgentPlanner Class Reference

Approach described at Section "Multi agents approach" of the documentation.

```
#include <MultiAgentPlanner.h>
```

Inheritance diagram for mt_rrt::MultiAgentPlanner:



Public Member Functions

template<typename... Args>
 MultiThreadedPlanner (Args &&...args)

 $default number of threads is assumed equal to 4. This can be changed after construction by using MultiThreaded \\ \leftarrow Planner::setThreads or MultiThreadedPlanner::setMaxThreads$

Protected Member Functions

• void **solve_** (const std::vector< float > &start, const std::vector< float > &end, const Parameters ¶meters, PlannerSolution &recipient) final

8.22.1 Detailed Description

Approach described at Section "Multi agents approach" of the documentation.

The documentation for this class was generated from the following file:

8.23 mt rrt::MultiThreadedPlanner Class Reference

Inheritance diagram for mt_rrt::MultiThreadedPlanner:



Public Member Functions

template<typename... Args>
 MultiThreadedPlanner (Args &&...args)

default number of threads is assumed equal to 4. This can be changed after construction by using MultiThreaded \leftarrow Planner::setThreads or MultiThreadedPlanner::setMaxThreads

- void setThreads (const Threads &threads_to_use)
- void setMaxThreads ()

the maximum possible threads available on this machine is used

std::size_t getThreads () const

Additional Inherited Members

The documentation for this class was generated from the following file:

8.24 mt_rrt::NearestQuery Struct Reference

Public Member Functions

void operator() (const Node &candidate, float cost2Go)

Public Attributes

- const Node * closest = nullptr
- float closestCost = COST MAX

The documentation for this struct was generated from the following file:

/home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/ExtenderUtils.h

8.25 mt_rrt::NearSet Struct Reference

Public Attributes

- float cost2RootSubject
- std::vector< NearSetElement > set

The documentation for this struct was generated from the following file:

• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/TreeHandler.h

8.26 mt_rrt::NearSetElement Struct Reference

Public Attributes

- bool isRoot
- Node * element
- float cost2Root
- float cost2go

The documentation for this struct was generated from the following file:

/home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/TreeHandler.h

8.27 mt_rrt::NearSetQuery Struct Reference

Public Member Functions

• void operator() (Node &subject)

Public Attributes

- float ray
- View state_pivot
- Connector * connector
- std::vector < NearSetElement > set

The documentation for this struct was generated from the following file:

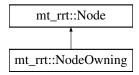
/home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/ExtenderUtils.h

8.28 mt_rrt::Node Class Reference

Used for representing a state $x \in \mathbb{X}$, Section 1.2 of the documentation. This class is used internally to extend search trees. The user is not expected to consume it.

```
#include <Node.h>
```

Inheritance diagram for mt_rrt::Node:



Public Member Functions

- Node (const View &state)
- const View & state () const
- const Node * getParent () const
 position of the parent node inside the owning Tree
- float cost2Go () const
- void setParent (const Node &parent, float cost2Go)
- virtual float cost2Root () const

Protected Attributes

- View state_
- Positive< float > cost2Go_ = Positive<float>{0}

The cost to spend to go from the parent to this node.

const Node * parent_ = nullptr

Static Protected Attributes

static const std::size_t MAX_ITERATIONS

8.28.1 Detailed Description

Used for representing a state $x \in \mathbb{X}$, Section 1.2 of the documentation. This class is used internally to extend search trees. The user is not expected to consume it.

8.28.2 Member Function Documentation

8.28.2.1 cost2Root()

```
virtual float mt_rrt::Node::cost2Root ( ) const [virtual]
```

Returns

Computes the cost to get from the root to this node, see 1.2.

Exceptions

when the root is not reached, cause loopy connections were made for some reason.

8.28.3 Member Data Documentation

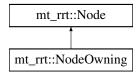
8.28.3.1 MAX_ITERATIONS

The documentation for this class was generated from the following file:

/home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Node.h

8.29 mt_rrt::NodeOwning Class Reference

Inheritance diagram for mt_rrt::NodeOwning:



Public Member Functions

- NodeOwning (std::vector< float > &&state)
- NodeOwning (const Node &o)

Additional Inherited Members

The documentation for this class was generated from the following file:

/home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Node.h

8.30 mt_rrt::NodesAllocatorT < NodeT > Class Template Reference

Public Member Functions

NodeT & emplace_back (const View &state)

The documentation for this class was generated from the following file:

• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Node.h

8.31 mt_rrt::ObjectPool < T > Class Template Reference

Public Member Functions

- template<typename... Args>
 T & emplace_back (Args &&...args)
- T * emplace_back_multiple (const T *source, std::size_t how_many)

Static Public Attributes

• static const std::size_t INITIAL_CAPACITY = 100

The documentation for this class was generated from the following file:

/home/andrea/Scrivania/GitProj/MT-RRT/src/carpet/header/MT-RRT/ObjectPool.h

8.32 mt_rrt::ParallelFor Class Reference

Public Member Functions

- ParallelFor (const Threads &pool_size)
- template < typename T , typename Pred > void process (const std::vector < T > &elements, Pred &&pred)
- std::size_t size () const

The documentation for this class was generated from the following file:

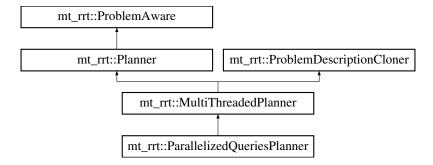
/home/andrea/Scrivania/GitProj/MT-RRT/src/multi-threaded/header/MT-RRT/ParallelFor.h

8.33 mt_rrt::ParallelizedQueriesPlanner Class Reference

Approach described at Section "Parallelization of the query activities" of the documentation.

#include <ParallelizedOueriesPlanner.h>

Inheritance diagram for mt_rrt::ParallelizedQueriesPlanner:



Public Member Functions

template<typename... Args>
 MultiThreadedPlanner (Args &&...args)

default number of threads is assumed equal to 4. This can be changed after construction by using MultiThreaded \leftarrow Planner::setThreads or MultiThreadedPlanner::setMaxThreads

Protected Member Functions

• void **solve**_ (const std::vector< float > &start, const std::vector< float > &end, const Parameters ¶meters, PlannerSolution &recipient) final

8.33.1 Detailed Description

Approach described at Section "Parallelization of the query activities" of the documentation.

The documentation for this class was generated from the following file:

8.34 mt rrt::Parameters Struct Reference

Groups together all the parameters that a @Planner needs to know to solve a specific problem for connecting 2 pair of states.

#include <Types.h>

Public Attributes

- ExpansionStrategy expansion_strategy = ExpansionStrategy::Star
- SteerIterations steer trials
- Iterations iterations
- Determinism determinism = Determinism {0.35f}
- bool best_effort = true

If true, the expansion of the tree(s) is arrested as soon as a solution is found. Otherwise, the search is kept on possibly finding additional solutions.

8.34.1 Detailed Description

Groups together all the parameters that a @Planner needs to know to solve a specific problem for connecting 2 pair of states.

The documentation for this struct was generated from the following file:

• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Types.h

8.35 mt_rrt::Planner Class Reference

In order to solve any kind of problem you need to build a kind of Planner, passing for sure the @Problem ← Description. The same @Planner, can be used to solve (one at a time) multiple problems, calling many times Planner::solve(...), possibly using different (or not) @Parameters every time.

#include <Planner.h>

Inheritance diagram for mt rrt::Planner:



Public Member Functions

- Planner (ProblemDescription &&problem)
 steals the connector and the sampler contained in the passed problem
- PlannerSolution **solve** (const std::vector< float > &start, const std::vector< float > &end, const Parameters ¶meters)

Protected Member Functions

- Planner (std::shared_ptr< ProblemDescription > problem) steals the connector and the sampler contained in the passed problem
- virtual void **solve_** (const std::vector< float > &start, const std::vector< float > &end, const Parameters ¶meters, PlannerSolution &recipient)=0

8.35.1 Detailed Description

In order to solve any kind of problem you need to build a kind of Planner, passing for sure the @Problem → Description. The same @Planner, can be used to solve (one at a time) multiple problems, calling many times Planner::solve(...), possibly using different (or not) @Parameters every time.

When enabling SHOW_PLANNER_PROGRESS, all instantiated planners will display the iterations in the console while running. This could be done mainly for debug purpose and you should be aware that this of course affects performances.

The documentation for this class was generated from the following file:

• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Planner.h

8.36 mt_rrt::PlannerSolution Struct Reference

Groups all the information characterizing a found solution.

```
#include <Planner.h>
```

Public Types

• using **TimeUnit** = std::chrono::microseconds

Public Attributes

- TimeUnit time
 - computation time spent for obtaining the solution, or trying to get one.
- std::size_t iterations
 - iterations spent for obtaining the solution, or trying to get one.
- std::vector< std::vector< float >> solution
 - The sequence of states forming the solution to the planning problem. Is empty in case a solution was not found.
- std::vector< TreeHandlerPtr > trees

The search trees produced while trying to solve the problem.

8.36.1 Detailed Description

Groups all the information characterizing a found solution.

The documentation for this struct was generated from the following file:

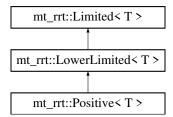
• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Planner.h

8.37 mt_rrt::Positive < T > Class Template Reference

A @LowerLimited quantity, having 0.0 as lower bound.

```
#include <Limited.h>
```

Inheritance diagram for mt_rrt::Positive < T >:



Public Member Functions

Positive (const T &initialValue=static_cast< T >(0))

Additional Inherited Members

8.37.1 Detailed Description

```
template<typename T> class mt_rrt::Positive< T >
```

A @LowerLimited quantity, having 0.0 as lower bound.

The documentation for this class was generated from the following file:

• /home/andrea/Scrivania/GitProj/MT-RRT/src/carpet/header/MT-RRT/Limited.h

8.38 mt_rrt::ProblemAware Class Reference

Someone aware of the static description of the class of problems to solve.

#include <ProblemDescription.h>

Inheritance diagram for mt_rrt::ProblemAware:



Public Member Functions

- ProblemAware (const ProblemDescriptionPtr &description)
- ProblemAware (const ProblemAware &o)
- const ProblemDescription & problem () const

Protected Member Functions

• ProblemDescriptionPtr problemPtr () const

8.38.1 Detailed Description

Someone aware of the static description of the class of problems to solve.

The documentation for this class was generated from the following file:

• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/ProblemDescription.h

8.39 mt_rrt::ProblemDescription Struct Reference

Groups together all the static information characterizing the class of problems to solve.

#include <ProblemDescription.h>

Public Attributes

bool simmetry

when true, the path connecting start to end, can be traversed in the opposite direction to connect end to start.

Positive < float > gamma

\gamma involved in the near set computation, refer to Section 1.2.3 of the documentation

- SamplerPtr sampler
- ConnectorPtr connector

8.39.1 Detailed Description

Groups together all the static information characterizing the class of problems to solve.

The documentation for this struct was generated from the following file:

/home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/ProblemDescription.h

8.40 mt_rrt::ProblemDescriptionCloner Class Reference

Inheritance diagram for mt_rrt::ProblemDescriptionCloner:



Protected Member Functions

- ProblemDescriptionCloner (const ProblemDescriptionPtr &problem)
- ProblemDescriptionPtr at (const std::size_t thread_id)
- const std::vector< ProblemDescriptionPtr > & getAllDescriptions () const
- void resizeDescriptions (const std::size_t size)

The documentation for this class was generated from the following file:

 /home/andrea/Scrivania/GitProj/MT-RRT/src/multi-threaded/header/MT-RRT/MultiThreaded← Planner.h

8.41 mt_rrt::Rewire Struct Reference

Public Attributes

- Node * involved_node
- float new_cost_from_father

The documentation for this struct was generated from the following file:

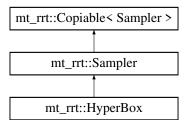
• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/TreeHandler.h

8.42 mt_rrt::Sampler Class Reference

Interface for a sampler of states.

```
#include <Sampler.h>
```

Inheritance diagram for mt_rrt::Sampler:



Public Member Functions

- virtual std::vector < float > sampleState () const =0

 Returns a state randomly sampled in the \mathcal{X} space, Sections 1.2.1, 1.2.2 and 1.2.3 of the documentation.

 This random state are used for randomly growing searching trees.
- virtual Seed sampleSeed () const =0

8.42.1 Detailed Description

Interface for a sampler of states.

8.42.2 Member Function Documentation

8.42.2.1 sampleSeed()

```
virtual Seed mt_rrt::Sampler::sampleSeed ( ) const [pure virtual]
```

Returns

a random seed to use for intializing another Sampler.

Implemented in mt_rrt::HyperBox.

8.42.2.2 sampleState()

```
virtual std::vector<float> mt_rrt::Sampler::sampleState ( ) const [pure virtual]
```

Returns a state randomly sampled in the \mathcal{X} space, Sections 1.2.1, 1.2.2 and 1.2.3 of the documentation. This random state are used for randomly growing searching trees.

Returns

a drawn random state.

Implemented in mt_rrt::HyperBox.

The documentation for this class was generated from the following file:

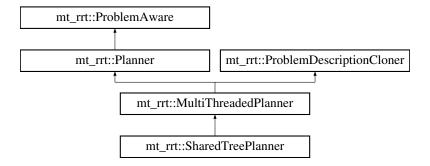
• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Sampler.h

8.43 mt_rrt::SharedTreePlanner Class Reference

Approach described at Section "Shared tree critical regions" of the documentation.

#include <SharedTreePlanner.h>

Inheritance diagram for mt_rrt::SharedTreePlanner:



Public Member Functions

template<typename... Args>
 MultiThreadedPlanner (Args &&...args)

default number of threads is assumed equal to 4. This can be changed after construction by using MultiThreaded \leftarrow Planner::setThreads or MultiThreadedPlanner::setMaxThreads

Protected Member Functions

• void **solve**_ (const std::vector< float > &start, const std::vector< float > &end, const Parameters ¶meters, PlannerSolution &recipient) final

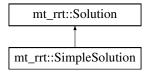
8.43.1 Detailed Description

Approach described at Section "Shared tree critical regions" of the documentation.

The documentation for this class was generated from the following file:

8.44 mt_rrt::SimpleSolution Class Reference

Inheritance diagram for mt_rrt::SimpleSolution:



Public Member Functions

- SimpleSolution (const Node *byPassNode, float cost2Target, const View &target)
- std::vector< std::vector< float > > getSequence () const final
- float cost () const final

Public Attributes

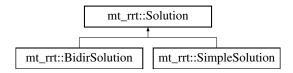
- const Node * byPassNode
- float cost2Target
- View target

The documentation for this class was generated from the following file:

• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/ExtenderSingle.h

8.45 mt_rrt::Solution Class Reference

Inheritance diagram for mt_rrt::Solution:



Public Member Functions

- virtual std::vector< std::vector< float > > getSequence () const =0
- virtual float cost () const =0

The documentation for this class was generated from the following file:

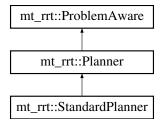
• /home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Solution.h

8.46 mt_rrt::StandardPlanner Class Reference

the classical mon-thread solver described in Section "Background on RRT" of the documentation.

```
#include <StandardPlanner.h>
```

Inheritance diagram for mt_rrt::StandardPlanner:



Public Member Functions

- Planner (ProblemDescription &&problem)
 - steals the connector and the sampler contained in the passed problem
- Planner (std::shared_ptr < ProblemDescription > problem)
 steals the connector and the sampler contained in the passed problem

Protected Member Functions

• void **solve**_ (const std::vector< float > &start, const std::vector< float > &end, const Parameters ¶meters, PlannerSolution &recipient) final

8.46.1 Detailed Description

the classical mon-thread solver described in Section "Background on RRT" of the documentation.

The documentation for this class was generated from the following file:

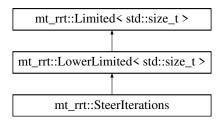
/home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/StandardPlanner.h

8.47 mt_rrt::SteerIterations Class Reference

NUmber of times to try steer, refer to documentation at Section "Background on RRT".

```
#include <Types.h>
```

Inheritance diagram for mt_rrt::SteerIterations:



Public Member Functions

• SteerIterations (const std::size_t trials)

Additional Inherited Members

8.47.1 Detailed Description

NUmber of times to try steer, refer to documentation at Section "Background on RRT".

The documentation for this class was generated from the following file:

/home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Types.h

8.48 mt_rrt::Connector::SteerResult Struct Reference

Performs a steering operation, Sections 1.2.1, 1.2.2 and 1.2.3 of the documentation, from a staring node to a target one.

```
#include <Connector.h>
```

Public Attributes

- NodeOwning node
- bool target_is_reached

8.48.1 Detailed Description

Performs a steering operation, Sections 1.2.1, 1.2.2 and 1.2.3 of the documentation, from a staring node to a target one.

Parameters

starting	configuration
target	configuration to reach

Returns

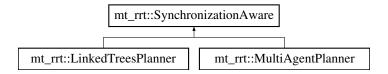
the steered configuration and a flag specifying whether target was reached. Is a nullopt when the steering was not possible at all

The documentation for this struct was generated from the following file:

/home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Connector.h

8.49 mt_rrt::SynchronizationAware Class Reference

Inheritance diagram for mt_rrt::SynchronizationAware:



Public Member Functions

- SynchronizationDegree & synchronization ()
- const SynchronizationDegree & synchronization () const

The documentation for this class was generated from the following file:

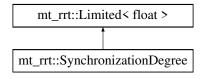
/home/andrea/Scrivania/GitProj/MT-RRT/src/multi-threaded/header/MT-RRT/Synchronization.h

8.50 mt_rrt::SynchronizationDegree Class Reference

Synchronization degree used by MultiAgentPlanner and LinkedTreesPlanner. This regulates how often the threads stop expansions and share info about explored states with other threads.

#include <Synchronization.h>

Inheritance diagram for mt_rrt::SynchronizationDegree:



Public Member Functions

• SynchronizationDegree (float initial_value)

Additional Inherited Members

8.50.1 Detailed Description

Synchronization degree used by MultiAgentPlanner and LinkedTreesPlanner. This regulates how often the threads stop expansions and share info about explored states with other threads.

The documentation for this class was generated from the following file:

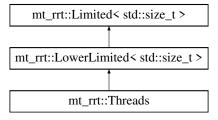
/home/andrea/Scrivania/GitProj/MT-RRT/src/multi-threaded/header/MT-RRT/Synchronization.h

8.51 mt_rrt::Threads Class Reference

Number of threads used by a MultiThreadedPlanner.

#include <MultiThreadedPlanner.h>

Inheritance diagram for mt_rrt::Threads:



Public Member Functions

• Threads (const std::size_t threads)

Additional Inherited Members

8.51.1 Detailed Description

Number of threads used by a MultiThreadedPlanner.

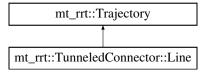
The documentation for this class was generated from the following file:

8.52 mt_rrt::Trajectory Class Reference

Interface describing an optimal trajectory \tau connecting 2 states, Section 1.2 of the documentation, in a particular problem to solve. After construction, the state on the trajectory is assumed at the beginning of the trajectory itself. Calling advance(), shift the cursor along the trajectory.

```
#include <Trajectory.h>
```

Inheritance diagram for mt_rrt::Trajectory:



Public Types

enum class AdvanceInfo { blocked , advanced , targetReached }
 explanation:

Public Member Functions

- Trajectory (const Trajectory &)=delete
- Trajectory & operator= (const Trajectory &)=delete
- virtual AdvanceInfo advance ()=0

Advance along the trajectory.

- virtual View getState () const =0
- virtual float getCumulatedCost () const =0

8.52.1 Detailed Description

Interface describing an optimal trajectory \tau connecting 2 states, Section 1.2 of the documentation, in a particular problem to solve. After construction, the state on the trajectory is assumed at the beginning of the trajectory itself. Calling advance(), shift the cursor along the trajectory.

8.52.2 Member Enumeration Documentation

8.52.2.1 AdvanceInfo

```
enum mt_rrt::Trajectory::AdvanceInfo [strong]
```

explanation:

- blocked -> when the advancement is not anymore possible, i.e. last state reached is not admitted by constraints
- advanced -> normal advancement. The state reached is admitted by constraints.
- targetReached -> similar to advanced, but in case in the reached state is the ending one

8.52.3 Member Function Documentation

8.52.3.1 getCumulatedCost()

```
virtual float mt_rrt::Trajectory::getCumulatedCost ( ) const [pure virtual]
```

Returns

the cost to go from the beginning of the trajectory to the current reached state. IMPORTANT: it is a no-sense value in case last advance() returned blocked

Implemented in mt_rrt::TunneledConnector::Line.

8.52.3.2 getState()

```
virtual View mt_rrt::Trajectory::getState ( ) const [pure virtual]
```

Returns

the current state on the trajectory. IMPORTANT: it is a no-sense value in case last this->advance() returned blocked

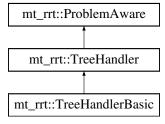
Implemented in mt_rrt::TunneledConnector::Line.

The documentation for this class was generated from the following file:

/home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/Trajectory.h

8.53 mt_rrt::TreeHandler Class Reference

Inheritance diagram for mt_rrt::TreeHandler:



Public Types

using DeterministicSteerRegister = std::unordered_map< const Node *, std::unordered_set< const float * > >

Public Member Functions

- virtual const Node * nearestNeighbour (const View &state) const =0
- virtual NearSet nearSet (const Node &subject) const =0
- virtual Node * internalize (const Node &subject)=0
- virtual void applyRewires (const Node &new_father, const std::vector < Rewire > &rewires)=0

Public Attributes

- std::vector< Node * > nodes
- Parameters parameters
- DeterministicSteerRegister deterministic_steer_register

Protected Member Functions

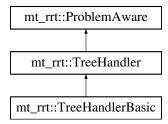
TreeHandler (const ProblemDescriptionPtr &problem, const Parameters ¶meters)

The documentation for this class was generated from the following file:

/home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/TreeHandler.h

8.54 mt rrt::TreeHandlerBasic Class Reference

Inheritance diagram for mt_rrt::TreeHandlerBasic:



Public Member Functions

- TreeHandlerBasic (const View &root, const ProblemDescriptionPtr &problem, const Parameters ¶meters)
- TreeHandlerBasic (Node &root, const ProblemDescriptionPtr &problem, const Parameters ¶meters)
- const Node * nearestNeighbour (const View &state) const override
- NearSet nearSet (const Node &subject) const override
- Node * internalize (const Node &subject) override
- void applyRewires (const Node &new_father, const std::vector < Rewire > &rewires) override

Protected Attributes

• NodesAllocator allocator

Additional Inherited Members

The documentation for this class was generated from the following file:

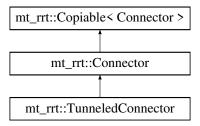
/home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/TreeHandler.h

8.55 mt_rrt::TunneledConnector Class Reference

Implements the approach described at Section "Tunneled check collision" of the documentation. Optimal trajectories \tau are assumed to be simple line in the space of configuration. Every advancement move the cursor of a qunatum of space along the trajectory and only the reached state is checked to be in the admitted space (and not the segment going from the previous state to the reached one).

#include <TunneledConnector.h>

Inheritance diagram for mt rrt::TunneledConnector:



Classes

• class Line

Public Member Functions

- TunneledConnector (const std::size_t size)
- float minCost2Go (const View &start, const View &end) const override
- TrajectoryPtr getTrajectory (const View &start, const View &end) const override
- std::size_t getStateSpaceSize () const
- float getSteerDegree () const
- void **setSteerDegree** (const Positive < float > °ree)
- virtual bool checkAdvancement (const View &previous_state, const View &advanced_state) const
 =0

Returns true is the advanced_state is not admitted.

Protected Member Functions

• TunneledConnector (const TunneledConnector &o)

8.55.1 Detailed Description

Implements the approach described at Section "Tunneled check collision" of the documentation. Optimal trajectories \tau are assumed to be simple line in the space of configuration. Every advancement move the cursor of a qunatum of space along the trajectory and only the reached state is checked to be in the admitted space (and not the segment going from the previous state to the reached one).

8.55.2 Member Function Documentation

8.55.2.1 getTrajectory()

Parameters

the	starting state
the	ending state

Returns

the optimal trajectory connecting the passed states. nullptr is returned in case a feasible trajectory does not exist

Implements mt_rrt::Connector.

8.55.2.2 minCost2Go()

Parameters

the	starting node in the trajectory whose cost is to evaluate
the	ending node in the trajectory whose cost is to evaluate

Returns

the cost C(\tau), Section 1.2 of the documentation, to traverse the optimal trajectory connecting the passed pair of states. This cost does not account for constraints, as this is done by $minCost2 \leftarrow GoConstrained(...)$

Implements mt_rrt::Connector.

The documentation for this class was generated from the following file:

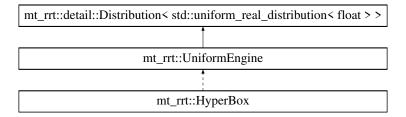
/home/andrea/Scrivania/GitProj/MT-RRT/src/core/header/MT-RRT/TunneledConnector.h

8.56 mt_rrt::UniformEngine Class Reference

Used to draw sample whithin a compact inverval [I, U].

```
#include <Random.h>
```

Inheritance diagram for mt_rrt::UniformEngine:



Public Member Functions

UniformEngine (const float &lowerBound=0, const float &upperBound=1.f, const std::optional
 Seed > &seed=std::nullopt)

Additional Inherited Members

8.56.1 Detailed Description

Used to draw sample whithin a compact inverval [I, U].

8.56.2 Constructor & Destructor Documentation

8.56.2.1 UniformEngine()

Parameters

the	lower bound of the compact interval
the	upper bound of the compact interval

The documentation for this class was generated from the following file:

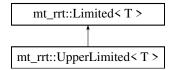
• /home/andrea/Scrivania/GitProj/MT-RRT/src/carpet/header/MT-RRT/Random.h

8.57 mt_rrt::UpperLimited < T > Class Template Reference

A @Limited quantity, having negative infinity as lower bound.

```
#include <Limited.h>
```

Inheritance diagram for mt_rrt::UpperLimited< T >:



Public Member Functions

- UpperLimited (const T &upperBound, const T &initialValue)
- UpperLimited (const T &upperBound)

Additional Inherited Members

8.57.1 Detailed Description

```
template<typename T> class mt_rrt::UpperLimited< T >
```

A @Limited quantity, having negative infinity as lower bound.

The documentation for this class was generated from the following file:

/home/andrea/Scrivania/GitProj/MT-RRT/src/carpet/header/MT-RRT/Limited.h

8.58 mt_rrt::View Struct Reference

Public Member Functions

- View (const float *data, std::size_t size)
- **View** (const std::vector< float > &owner)
- View trim (std::size_t size, std::size_t from=0) const
- std::vector< float > convert () const

Public Attributes

- std::size_t size = 0
- const float * data = nullptr

The documentation for this struct was generated from the following file:

• /home/andrea/Scrivania/GitProj/MT-RRT/src/carpet/header/MT-RRT/View.h

Bibliography

- [1] Bullet3. https://github.com/bulletphysics/bullet3. 11
- [2] ReactPhysics3D. https://www.reactphysics3d.com. 11
- [3] Andrea Casalino, Andrea Maria Zanchettin, and Paolo Rocco. Mt-rrt: a general purpose multithreading library for path planning. In *IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS 2019)*, pages 1510–1517. IEEE, 2019. 17
- [4] Sertac Karaman and Emilio Frazzoli. Sampling-based algorithms for optimal motion planning. *The international journal of robotics research*, 30(7):846–894, 2011. 3
- [5] James J Kuffner and Steven M LaValle. Rrt-connect: An efficient approach to single-query path planning. In Robotics and Automation, 2000. Proceedings. ICRA'00. IEEE International Conference on, volume 2, pages 995–1001. IEEE, 2000. 1
- [6] James J Kuffner and Steven M LaValle. Rrt-connect: An efficient approach to single-query path planning. In Robotics and Automation, 2000. Proceedings. ICRA'00. IEEE International Conference on, volume 2, pages 995–1001. IEEE, 2000. 3
- [7] Sean Quinlan. *Real-time modification of collision-free paths*. Number 1537. Stanford University Stanford, 1994. 12
- [8] Adnan Tahirovic and Faris Janjoš. A class of sdre-rrt based kinodynamic motion planners. In *American Control Conference (ACC)*, volume 2018, 2018. 1
- [9] Dustin J Webb and Jur van den Berg. Kinodynamic rrt*: Optimal motion planning for systems with linear differential constraints. 2012. 1