

FACOLTÀ DI INGEGNERIA DELL'INFORMAZIONE

ELECTIVE IN ROBOTICS

---

# Quadrotor

## Motion Planning Algorithms

**Prof. Marilena Vendittelli**  
**Prof. Jean-Paul Laumond**

**Jacopo Capolicchio**  
**Riccardo Spica**

**Academic Year 2010-2011**

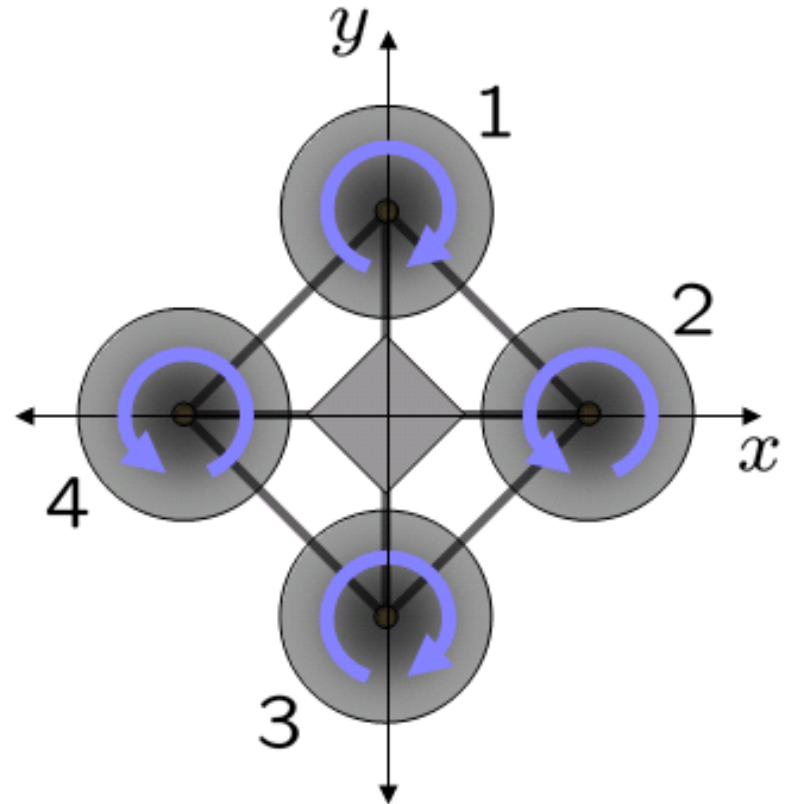


**SAPIENZA**  
UNIVERSITÀ DI ROMA

# Introduction

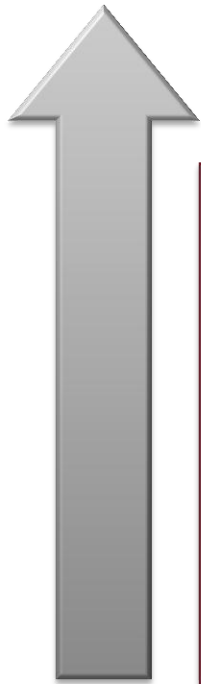
## Small-scale quadrotors

- Four **pitch-fixed rotors** disposed at the vertices of a square, whose directions of rotation are equal in pairs
- **Vertical** and **lateral displacements**, as well as **yaw rotations**, by varying the relative turning speeds of rotors



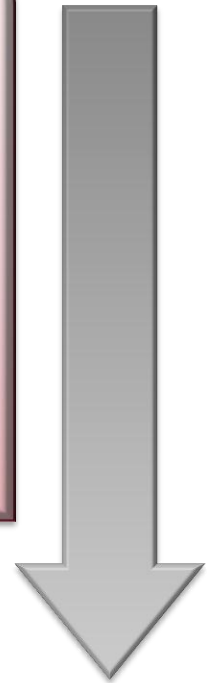
# Introduction

## Pros and Cons



Simple Mechanics  
Low cost  
Robustness  
High Maneuverability  
VTOL  
Hovering

Low Payloads  
Poor sensors  
equipment  
Low computational  
capabilities



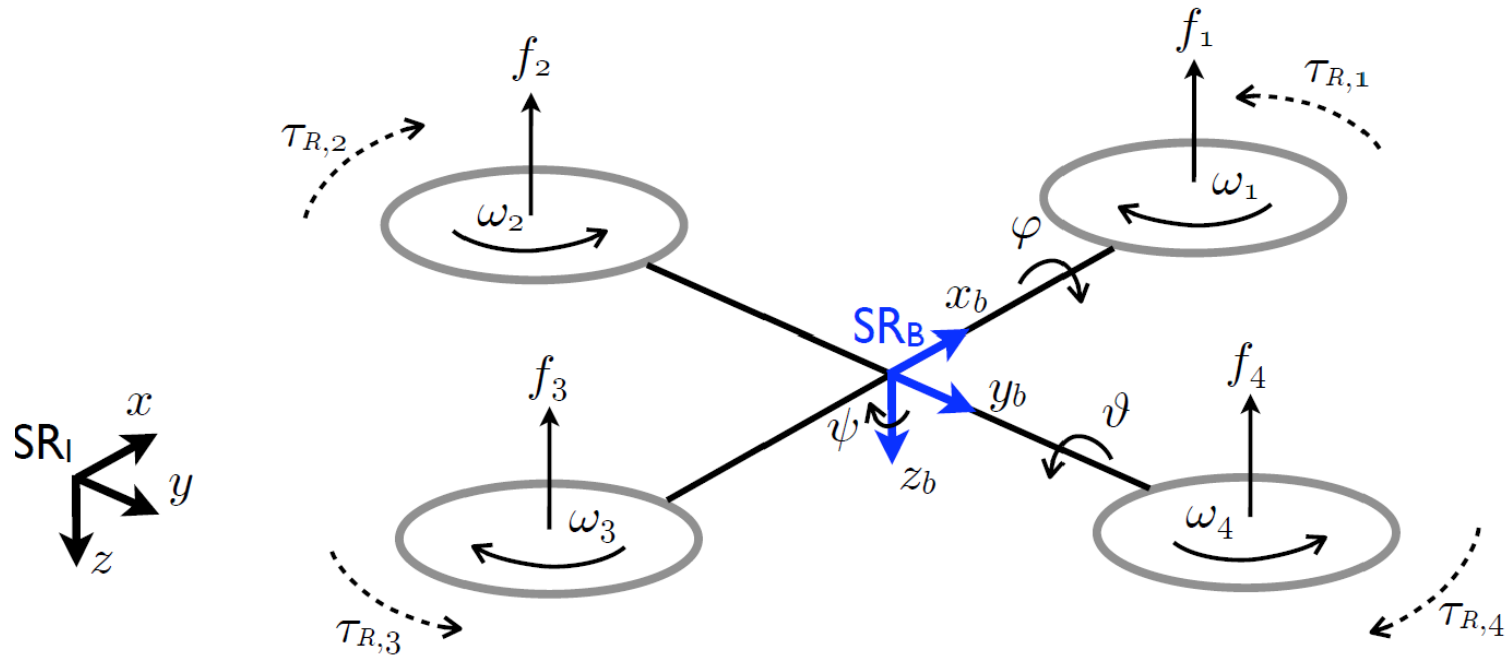
# Dynamic Model

## Preliminaries

- $SR_I$ : world inertial frame
- $SR_B$ : body-attached frame
- **Configuration** = Position + Orientation of  $SR_B$  w.r.t.  $SR_I$
- **Assumptions:**
  - Robot symmetry
  - No disturbances
  - Negligible gyroscopic and aerodynamic effects
  - Negligible motor dynamics
  - Low level controllers for blades rotational speed

# Dynamic Model

## Preliminaries (2)



$$f_i = b\omega_i^2$$

$$\tau_{R,i} = d\omega_i^2$$

$b$ : thrust factor

$d$ : drag factor.

# Dynamic Model

## Newton-Euler approach (1)

The orientation is described by using RPY angles  $(\varphi, \vartheta, \psi)$ :

$${}^I R_B = \begin{pmatrix} c_\psi c_\vartheta & c_\psi s_\vartheta s_\varphi - s_\psi c_\varphi & c_\psi s_\vartheta c_\varphi + s_\psi s_\varphi \\ s_\psi c_\vartheta & s_\psi s_\vartheta s_\varphi + c_\psi c_\varphi & s_\psi s_\vartheta c_\varphi - c_\psi s_\varphi \\ -s_\vartheta & c_\vartheta s_\varphi & c_\vartheta c_\varphi \end{pmatrix}$$

Control inputs transformation:

$$\begin{cases} T = f_1 + f_2 + f_3 + f_4 \\ \tau_\varphi = l (f_2 - f_4) \\ \tau_\vartheta = l (f_1 - f_3) \\ \tau_\psi = -\tau_{R,1} + \tau_{R,2} - \tau_{R,3} + \tau_{R,4} \end{cases}$$

# Dynamic Model

## Newton-Euler approach (2)

The translational dynamics is governed by the **Newton equation**:

$$m \begin{pmatrix} \ddot{x} \\ \ddot{y} \\ \ddot{z} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ mg \end{pmatrix} + {}^I R_B \begin{pmatrix} 0 \\ 0 \\ -T \end{pmatrix}$$

The angular acceleration is governed by the **Euler equation**:

$$I \begin{pmatrix} \dot{p} \\ \dot{q} \\ \dot{r} \end{pmatrix} + \begin{pmatrix} p \\ q \\ r \end{pmatrix} \times I \begin{pmatrix} p \\ q \\ r \end{pmatrix} = \begin{pmatrix} \tau_\varphi \\ \tau_\vartheta \\ \tau_\psi \end{pmatrix}$$

For small roll and pitch angles:

$$\begin{pmatrix} p \\ q \\ r \end{pmatrix} = \begin{pmatrix} 1 & 0 & -\sin \vartheta \\ 0 & \cos \varphi & \cos \vartheta \sin \varphi \\ 0 & -\sin \varphi & \cos \vartheta \cos \varphi \end{pmatrix} \begin{pmatrix} \dot{\varphi} \\ \dot{\vartheta} \\ \dot{\psi} \end{pmatrix} \sim \begin{pmatrix} \dot{\varphi} \\ \dot{\vartheta} \\ \dot{\psi} \end{pmatrix}$$

# Dynamic Model

## Newton-Euler approach (3)

System state:

$$\xi = [x \ y \ z \ \varphi \ \vartheta \ \psi \ \dot{x} \ \dot{y} \ \dot{z} \ \dot{\varphi} \ \dot{\vartheta} \ \dot{\psi}]^T$$

Then:

$$\left\{ \begin{array}{l} \ddot{x} = -[\cos(\psi) \sin(\vartheta) \cos(\varphi) + \sin(\psi) \sin(\varphi)] \frac{T}{m} \\ \ddot{y} = -[\sin(\psi) \sin(\vartheta) \cos(\varphi) - \cos(\psi) \sin(\varphi)] \frac{T}{m} \\ \ddot{z} = g - \cos(\vartheta) \cos(\varphi) \frac{T}{m} \\ \ddot{\varphi} = \frac{\tau_{\varphi}}{I_x} \\ \ddot{\vartheta} = \frac{\tau_{\vartheta}}{I_y} \\ \ddot{\psi} = \frac{\tau_{\psi}}{I_z} \end{array} \right.$$



# Dynamic Model

## Differential flatness (1)

Flat outputs:  $(x, y, z, \psi)$

From the dynamic model one obtains:

$$\vartheta = \operatorname{atan}\left(\frac{\ddot{x} \cos(\psi) + \ddot{y} \sin(\psi)}{\ddot{z} - g}\right)$$
$$\varphi = \operatorname{atan}\left(\frac{\ddot{x} \sin(\psi) - \ddot{y} \cos(\psi)}{\ddot{z} - g} \cos(\vartheta)\right)$$

and by derivation

$$\dot{\vartheta} = \frac{1}{1 + \tan^2 \vartheta} \frac{1}{\ddot{z} - g} \left[ (x^{(3)} + \ddot{y}\dot{\psi}) \cos \psi + (y^{(3)} - \ddot{x}\dot{\psi}) \sin \psi - z^{(3)} \tan \vartheta \right]$$
$$\dot{\varphi} = \frac{1}{1 + \tan^2 \varphi} \frac{1}{\ddot{z} - g} \left( \sin \psi (\ddot{y}\dot{\psi} + x^{(3)}) \cos \vartheta + (\ddot{y}\dot{\vartheta} \sin \vartheta) \right. \\ \left. + \cos \psi \left( (x\dot{\psi} + y^{(3)}) \cos \vartheta + (\ddot{y}\dot{\vartheta} \sin \vartheta) - \tan \varphi \right) \right)$$

# Dynamic Model

## Differential flatness (2)

$$\ddot{\vartheta} = -2\dot{\vartheta}^2 \tan \vartheta - \frac{z^{(3)}\dot{\vartheta}}{\ddot{z} - g} + \frac{1}{1 + \tan^2 \vartheta} \frac{1}{\ddot{z} - g} [(x^{(4)} + 2y^{(3)}\dot{\psi} + \ddot{y}\ddot{\psi} - \ddot{x}\dot{\psi}^2)\cos \psi + (y^{(4)} - 2x^{(3)}\dot{\psi} - \ddot{x}\ddot{\psi} - \ddot{y}\dot{\psi}^2)\sin \psi - z^{(4)} \tan \vartheta - z^{(3)}\dot{\vartheta}(1 + \tan^2 \vartheta)]$$

$$\ddot{\varphi} = -2\dot{\varphi}^2 \tan \varphi - \frac{z^{(3)}\dot{\varphi}}{\ddot{z} - g} + \frac{1}{1 + \tan^2 \varphi} \frac{1}{\ddot{z} - g} [\cos \psi (\sin \vartheta \dot{\vartheta}(y^{(3)} - \ddot{x}) + \cos \vartheta \dot{\vartheta}^2 \ddot{y} + \sin \vartheta \ddot{\vartheta} \ddot{y}) + \cos \vartheta (-y^{(4)} + x^{(3)}\dot{\psi} + \ddot{x}\ddot{\psi} + x^{(3)}\dot{\psi} + \ddot{y}\dot{\psi}^2) + \sin \vartheta \dot{\vartheta}(y^{(3)} - \ddot{x}\dot{\psi}) + \sin \psi (\sin \vartheta \dot{\vartheta}(x^{(3)} - \ddot{y}) + \cos \vartheta \dot{\vartheta}^2 \ddot{x} + \sin \vartheta \ddot{\vartheta} \ddot{x}) + \cos \vartheta (x^{(4)} + y^{(3)}\dot{\psi} + \ddot{y}\ddot{\psi} + y^{(3)}\dot{\psi} - \ddot{x}\dot{\psi}^2) + \sin \vartheta \dot{\vartheta}(x^{(3)} - \ddot{y}\dot{\psi}) - z^{(4)} \tan \varphi - z^{(3)}\dot{\varphi}(1 + \tan^2 \varphi)\dot{\varphi}]$$

# Dynamic Model

## Differential flatness (3)

Finally the torque inputs are:

$$\begin{cases} \tau_\varphi = \ddot{\phi} I_x \\ \tau_\vartheta = \ddot{\theta} I_y \\ \tau_\psi = \ddot{\psi} I_z \end{cases}$$

while the thrust is:

$$T = m\sqrt{\ddot{x}^2 + \ddot{y}^2 + (\ddot{z} - g)^2}$$

For  $\mathcal{T} \rightarrow \infty$  all the **derivatives** of the flat outputs **go to zero**, then

$$\begin{bmatrix} \varphi \\ \vartheta \end{bmatrix} \rightarrow \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad \text{and} \quad \begin{bmatrix} T \\ \tau_\varphi \\ \tau_\vartheta \\ \tau_\psi \end{bmatrix} \rightarrow \begin{bmatrix} mg \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

# Dynamic Model

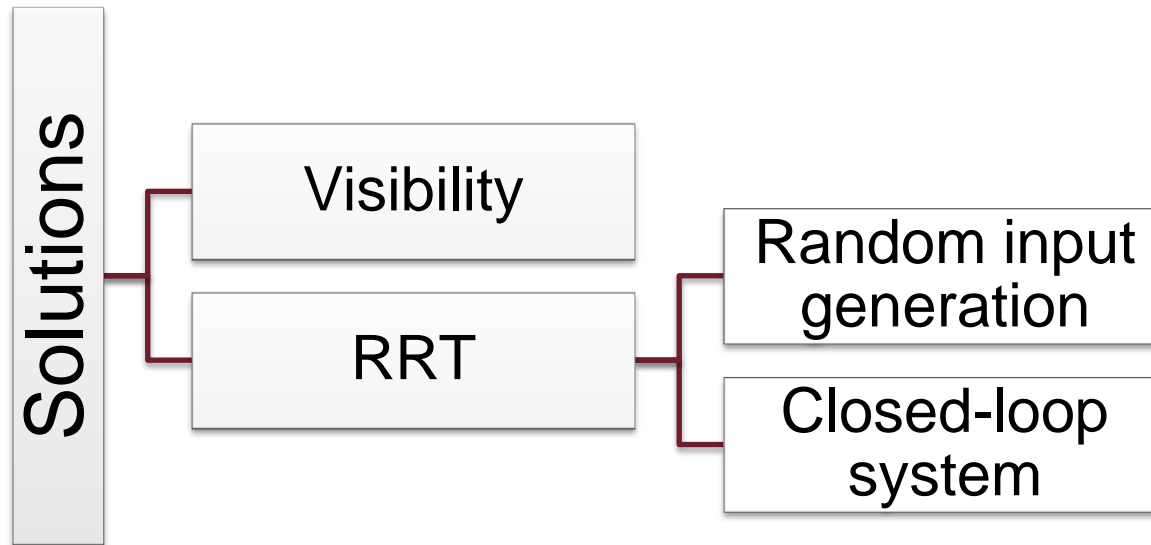
## Differential flatness (4)

- The robot is able to **track any given trajectory** of the 3D position and yaw angle provided that it is “smooth enough”
- In particular to guarantee **inputs continuity**:
  - the position trajectory has to be continue up to the fourth order derivative
  - the yaw angle trajectory has to be continue up to the second order derivative
- To satisfy **motors constraints**  $\mathcal{T}$  has to be “large enough”

# Motion Planning

## Introduction

**Aim :** *build a **feasible trajectory** going from a start to a goal, possibly considering yaw angle variations , in presence of **obstacles** and **actuators constraints***



# Visibility-Based Motion Planning

## Introduction

The motion planning has been split in **four steps**:

Creation of a **visibility-based probabilistic roadmap** in the space of flat outputs

Searching of a **piece-wise linear safe trajectory** exploiting the roadmap

Approximation of the computed trajectory through a **smooth spline**

**Time-scaling**

# Visibility-Based Motion Planning

## Visibility roadmap construction (1)

**Aim :** *build a roadmap with few nodes but still sufficient for solving the global planning problem*

- For a given **local method**  $\mathcal{L}$ , the **visibility domain** of  $q$  is the set of configurations reachable by  $q$  through  $\mathcal{L}$

$$\mathcal{V}_{\mathcal{L}}(q) = \{q' \in CS_{free} \text{ s.t. } \mathcal{L}(q, q') \subset CS_{free}\}$$

- $q$  is the **guard** of  $\mathcal{V}_{\mathcal{L}}(q)$
- A set of guards constitutes a **coverage** of  $CS_{free}$  if the union of their visibility domains covers  $CS_{free}$
- **Note:** such a finite coverage may not always exist, depending on both the shape of  $CS_{free}$  and the local method  $\mathcal{L}$

# Visibility-Based Motion Planning

## Visibility roadmap construction (2)

A free configuration  $q$  is added to the roadmap iff one of the following conditions is satisfied:

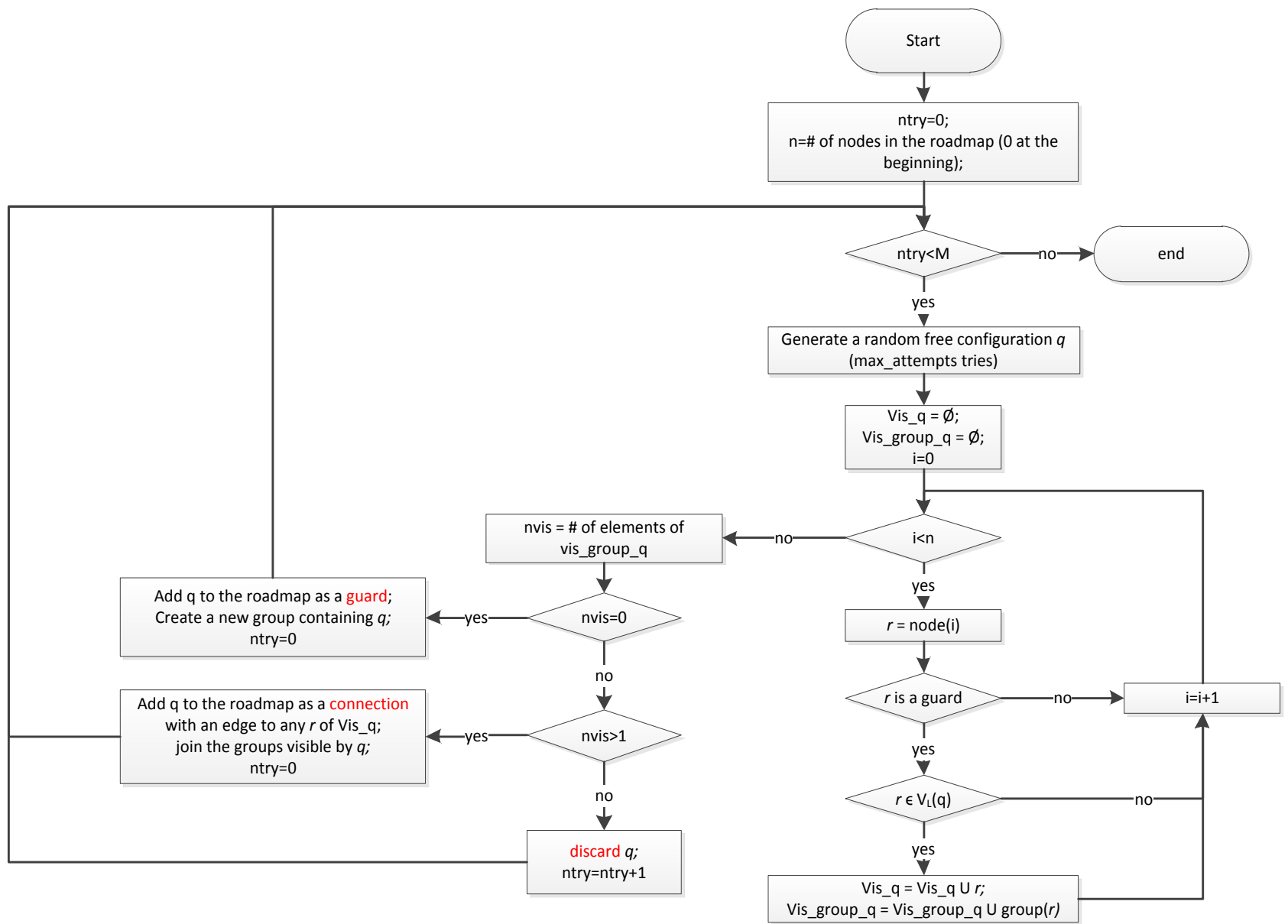
$q$  is a guard

- $q$  does not belong to the visibility domain of any other **guard** of the current roadmap → it **enlarges the coverage** of  $CS_{\text{free}}$

$q$  is a connection

- $q$  lies in the intersection of the visibility domains of at least two **guards** belonging to different connected components of the current roadmap → it **enhances the connectivity**





# Visibility-Based Motion Planning

## Visibility roadmap construction (4)

- **Operating space:** flat outputs

$$\sigma = \begin{pmatrix} x \\ y \\ z \\ \psi \end{pmatrix}$$

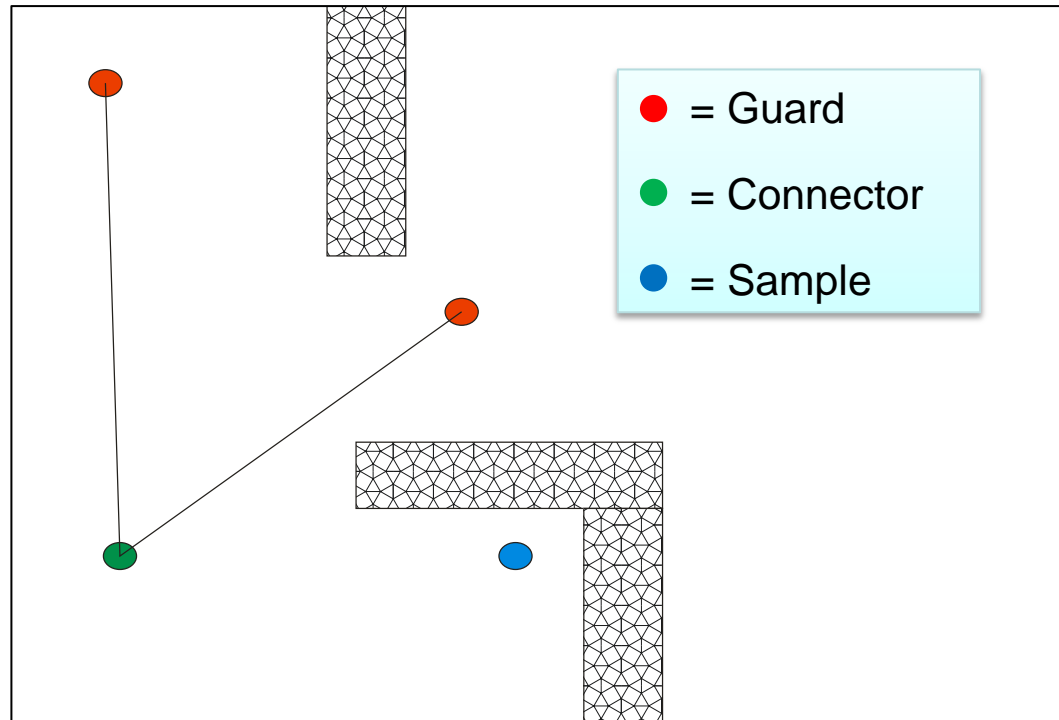
- **Collision checking:** obstacle expansion

$$\text{arm length } l = 0,25 \text{ m}$$

- **Local method:** straight lines

# Visibility-Based Motion Planning

## Visibility roadmap construction (5)



*Why not to exploit connections to increase the coverage?*

# Visibility-Based Motion Planning

## Visibility roadmap construction (6)

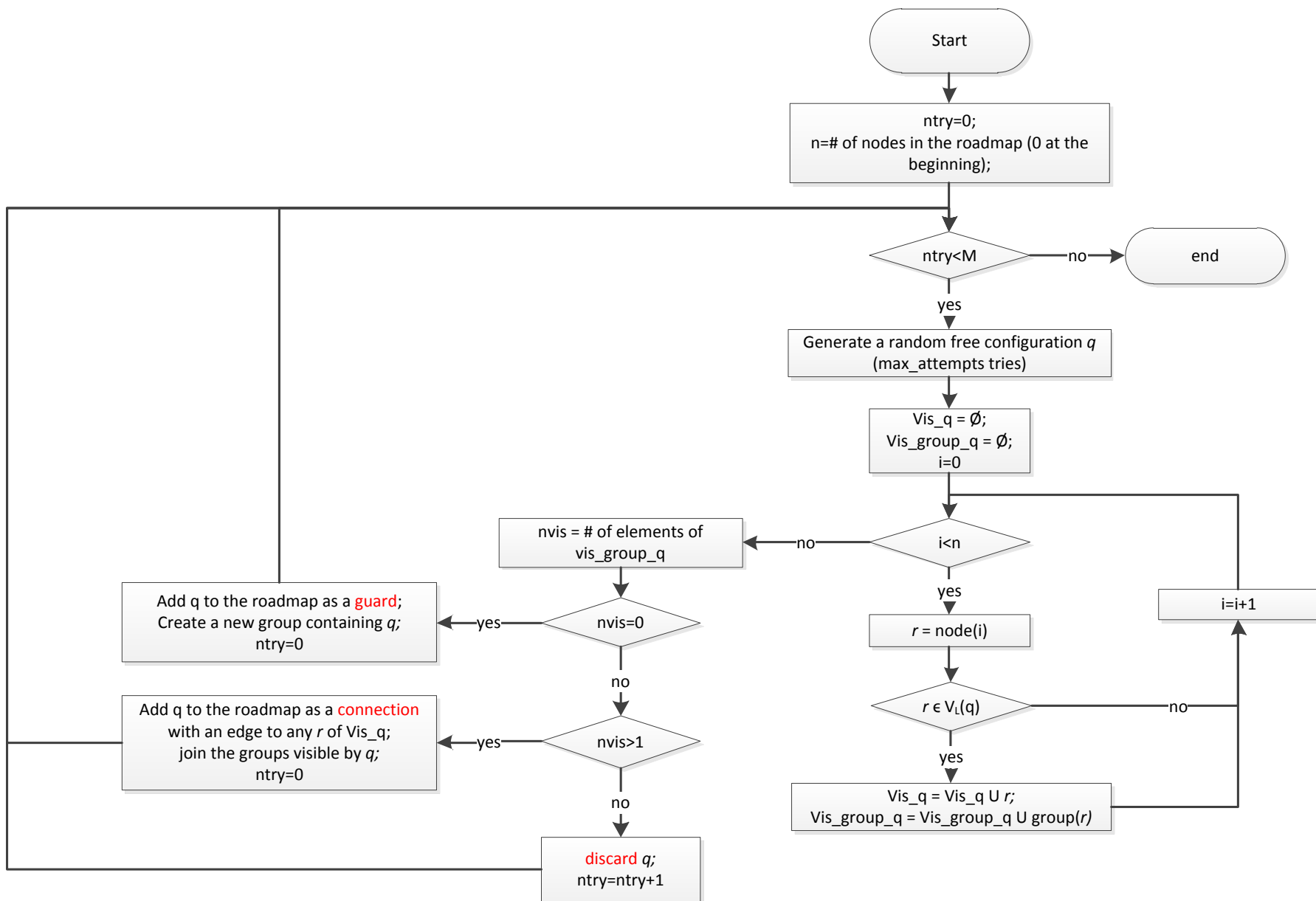
In the alternative version of algorithm a free configuration  $q$  is added to the roadmap iff one of the following conditions is satisfied:

$q$  is a guard

- $q$  does not belong to the visibility domain of any other **node** (either a guard or a connection) of the current roadmap

$q$  is a connection



- $q$  lies in the intersection of the visibility domains of at least two **nodes** (either guards or connections) belonging to two different connected components of the current roadmap



# Visibility-Based Motion Planning

## Searching a safe trajectory (1)

**Aim** : *compute (if possible) a **safe piece-wise linear path** that goes from the starting position to the goal*

- Such a trajectory requires the quadrotor to **stop at each viapoint**
- The searching of a safe solution is made iteratively:
  - try to connect (using straight lines) start and goal to two nodes  $q_s$  and  $q_g$  belonging to the same connected component of the roadmap
  - If success, a solution exists  **A\***
  - If not  expand the roadmap and try again

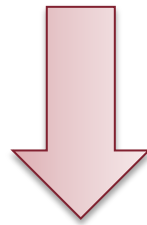


# Visibility-Based Motion Planning

## Spline interpolation (1)

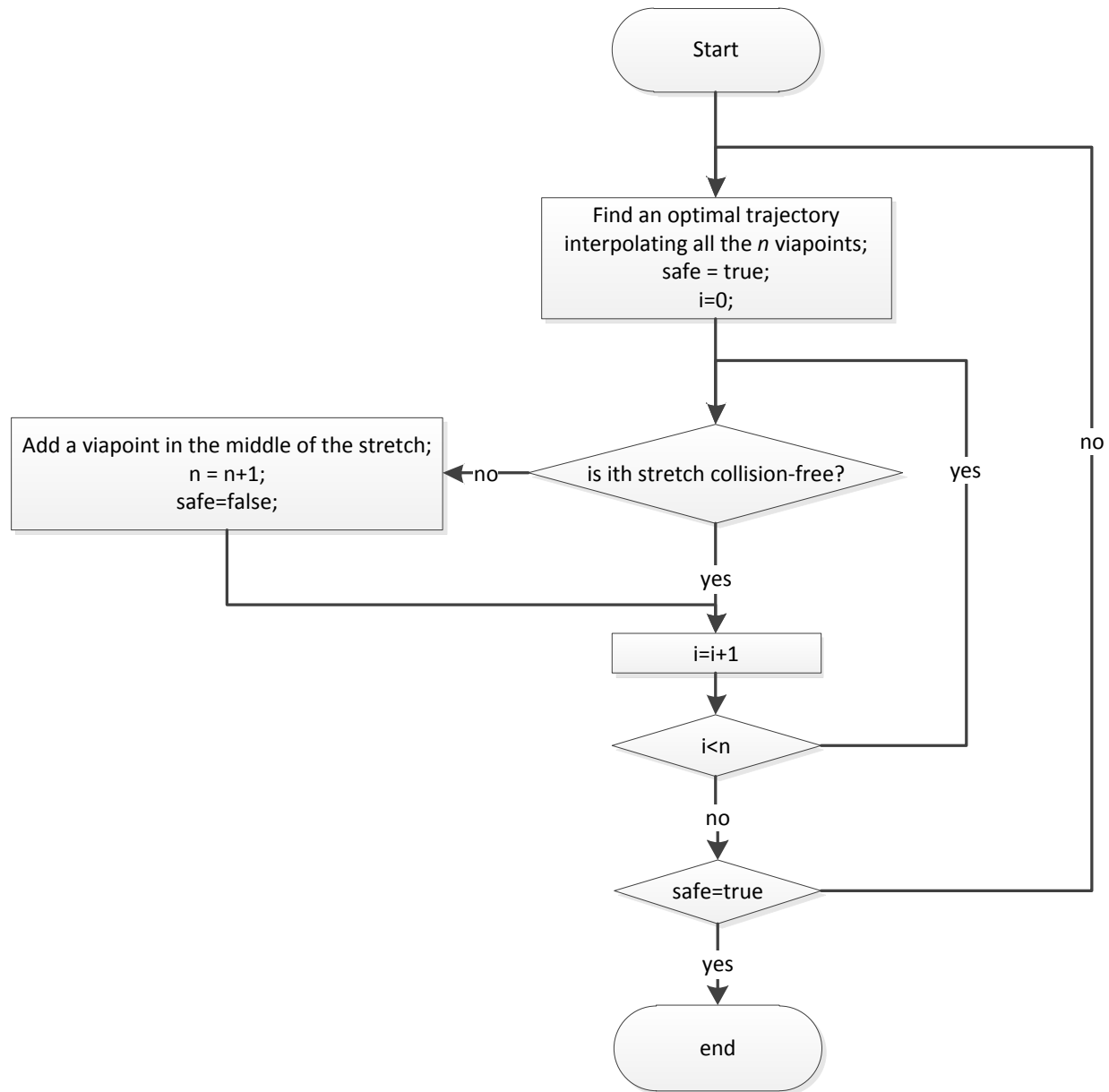
**Aim** : *compute a **smooth solution** approximating the piece-wise linear path obtained in the previous step*

- Path computed by using a **B-spline interpolating viapoints**
- This kind of trajectory may significantly diverge from the safe piece-wise linear solution and consequently become unsafe



Introduction of a **viapoint in the middle** of any unsafe stretch





# Visibility-Based Motion Planning

## Spline interpolation (3)

Split the geometrical aspect from the temporal one

$$\left. \begin{array}{l} s_x(\tau) \in \mathcal{C}^4: [0,1] \rightarrow \mathbb{R} \\ s_y(\tau) \in \mathcal{C}^4: [0,1] \rightarrow \mathbb{R} \\ s_z(\tau) \in \mathcal{C}^4: [0,1] \rightarrow \mathbb{R} \end{array} \right\} \longrightarrow 6^{th} \text{ order}$$
$$s_\psi(\tau) \in \mathcal{C}^2: [0,1] \rightarrow \mathbb{R} \longrightarrow 4^{th} \text{ order}$$

such that

$$\begin{cases} s_x(\tau_i) = x_i, & i = 1, \dots, n \\ s_y(\tau_i) = y_i, & i = 1, \dots, n \\ s_z(\tau_i) = z_i, & i = 1, \dots, n \\ s_\psi(\tau_i) = \psi_i, & i = 1, \dots, n \end{cases} \quad \text{with } 0 = \tau_1 < \tau_2 < \dots < \tau_n = 1$$

# Visibility-Based Motion Planning

## Spline interpolation (4)

$(\tau_1, \tau_2, \dots, \tau_n)$  are chosen by solving an **optimization problem**:

- **Variables:**

$$\delta_i = \tau_{i+1} - \tau_i \quad \text{for } i = 1, \dots, n - 1$$

- **Constraints:**

$$0 < \delta_i < 1 \quad \text{for } i = 1, \dots, n - 1$$

$$\sum_{i=1}^{n-1} \delta_i = 1$$

- **Objective function:**

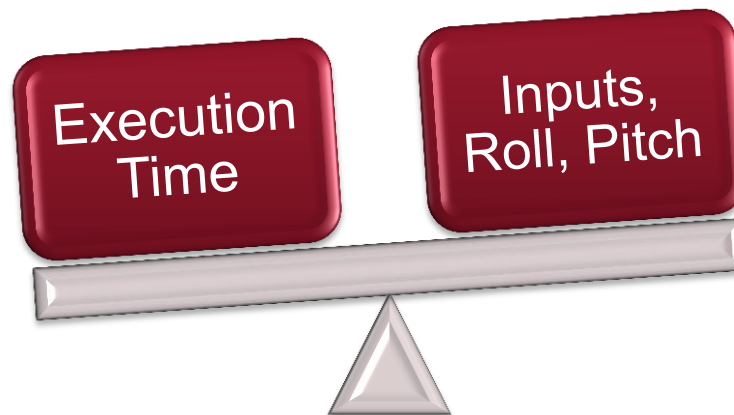
$$\int_0^1 \left[ a \left( \left\| \frac{d^4 s_x(\tau)}{d\tau^4} \right\|^2 + \left\| \frac{d^4 s_y(\tau)}{d\tau^4} \right\|^2 + \left\| \frac{d^4 s_z(\tau)}{d\tau^4} \right\|^2 \right) + b \left\| \frac{d^2 s_\psi(\tau)}{d\tau^2} \right\|^2 \right] d\tau$$

# Visibility-Based Motion Planning

## Time scaling (1)

**Aim :** *optimize the time needed to complete the trajectory*

- Changing the time to navigate through the nodes by a factor of  $k$  ( s.t.  $t_i = k\tau_i$ ) the result is simply a time-scaled version of the non-dimensional solution



# Visibility-Based Motion Planning

## Time scaling (2)

$k$  is chosen by solving an **optimization problem**:

- Variables:

$$k$$

- Constraints:


$$\begin{bmatrix} |T| \\ |\tau_\varphi| \\ |\tau_\theta| \\ |\tau_\psi| \end{bmatrix} \leq \begin{bmatrix} T_M \\ \tau_{\varphi,M} \\ \tau_{\theta,M} \\ \tau_{\psi,M} \end{bmatrix}$$

- Objective function:

$$k = \mathcal{J}$$

# Visibility-Based Motion Planning

Alternative choice for  $\psi$

$\psi$  is actuated independently  we can neglect it during the roadmap construction

Putting:

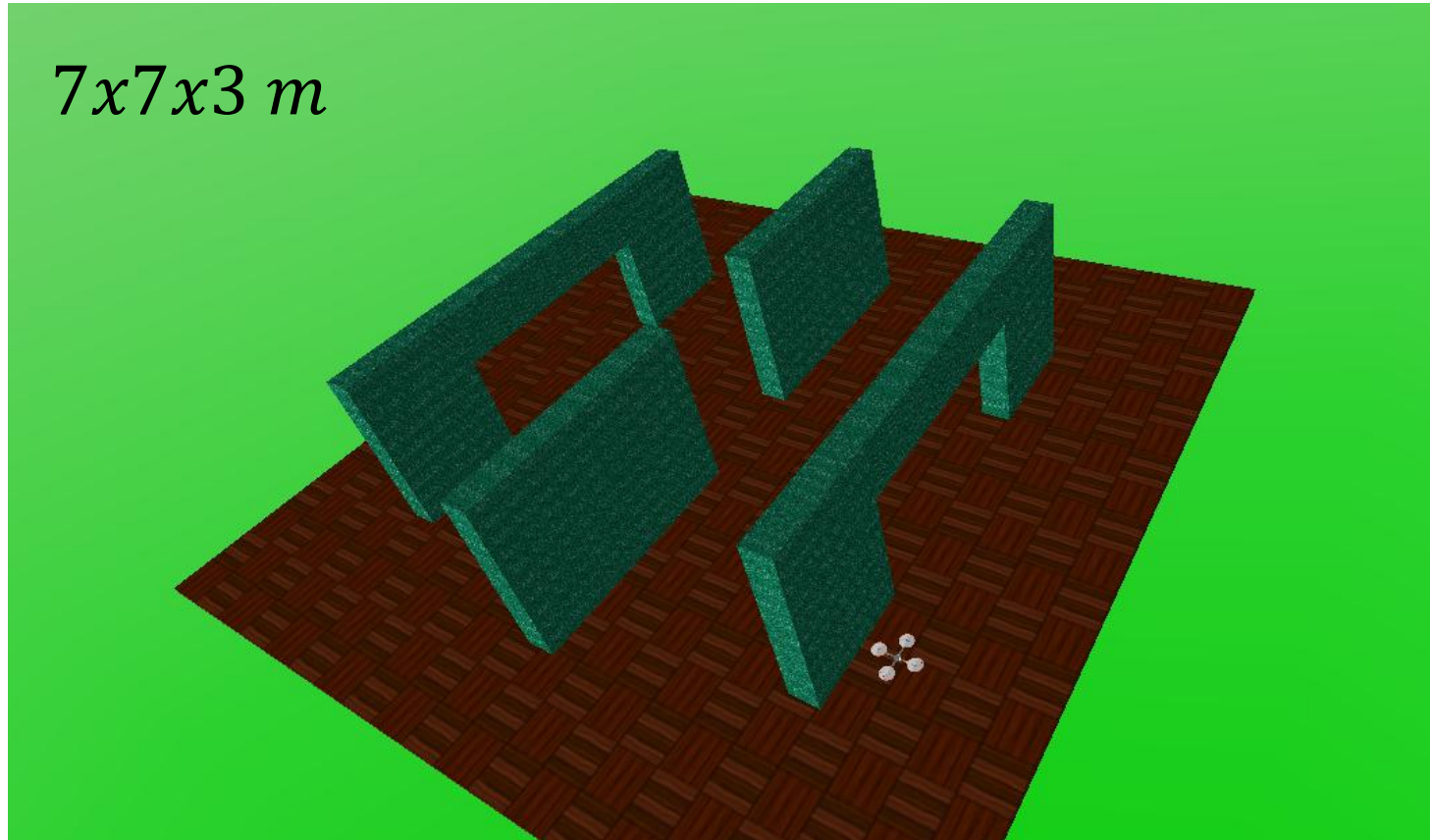
$$\psi = \text{atan2}(\dot{y}, \dot{x})$$

The robot **points in the direction of motion** and

$$\dot{\psi} = \frac{1}{1 + \left(\dot{y}/\dot{x}\right)^2} \left( \frac{\ddot{y}}{\dot{x}} - \frac{\dot{y}\ddot{x}}{\dot{x}^2} \right)$$
$$\ddot{\psi} = \frac{1}{1 + \left(\dot{y}/\dot{x}\right)^2} \left\{ \frac{1}{\dot{x}} \left[ y^{(3)} + \frac{1}{\dot{x}} \left( 2 \frac{\dot{y}\ddot{x}^2}{\dot{x}} - 2\ddot{y}\ddot{x} - \dot{y}x^{(3)} \right) \right] \right\} - 2 \frac{\dot{y}\dot{\psi}^2}{\dot{x}}$$

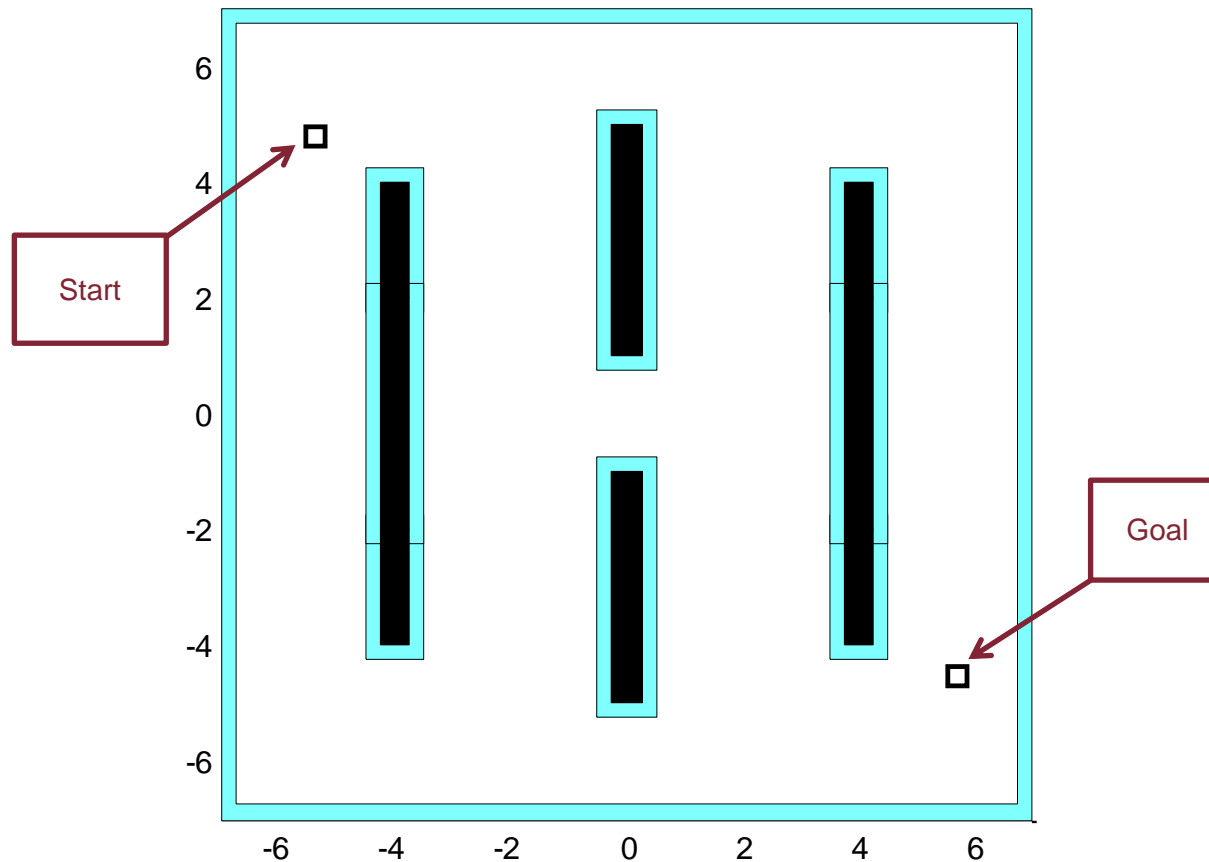
# Visibility-Based Motion Planning

Simulation scenario



# Visibility-Based Motion Planning

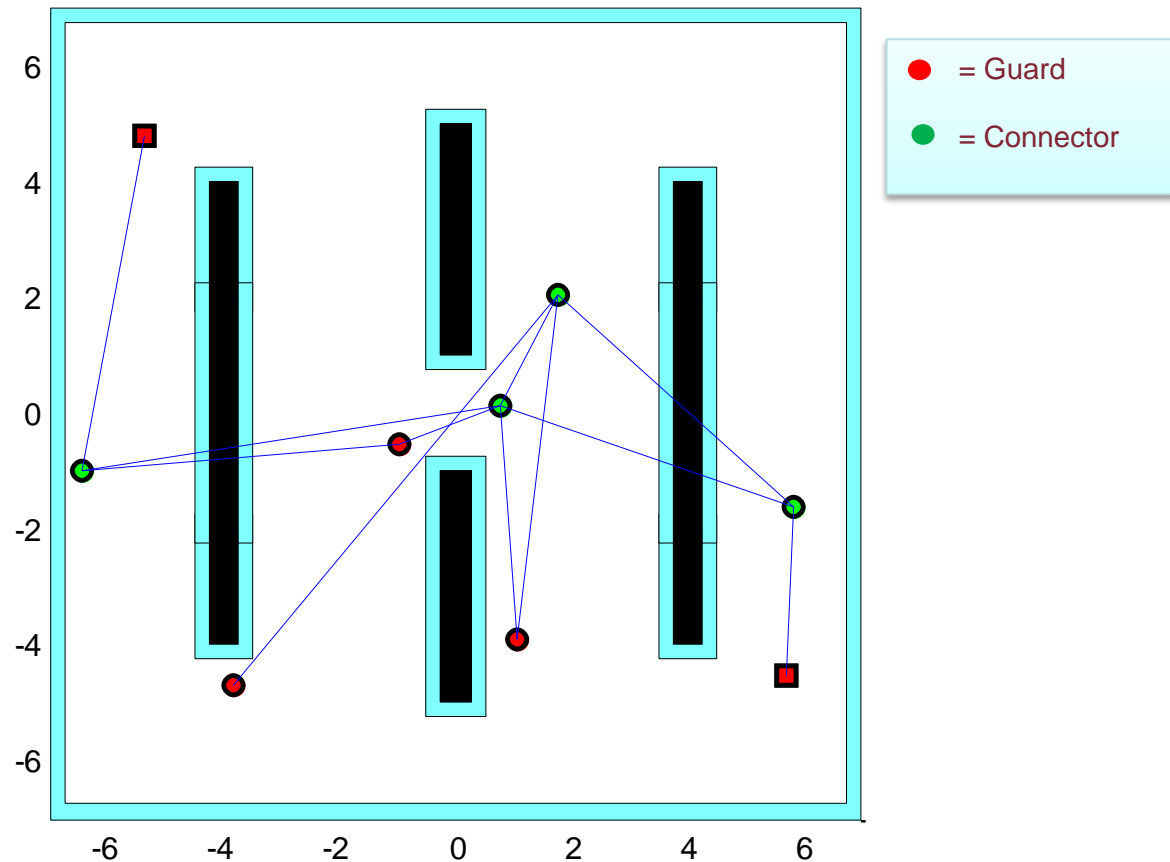
Start and Goal selection





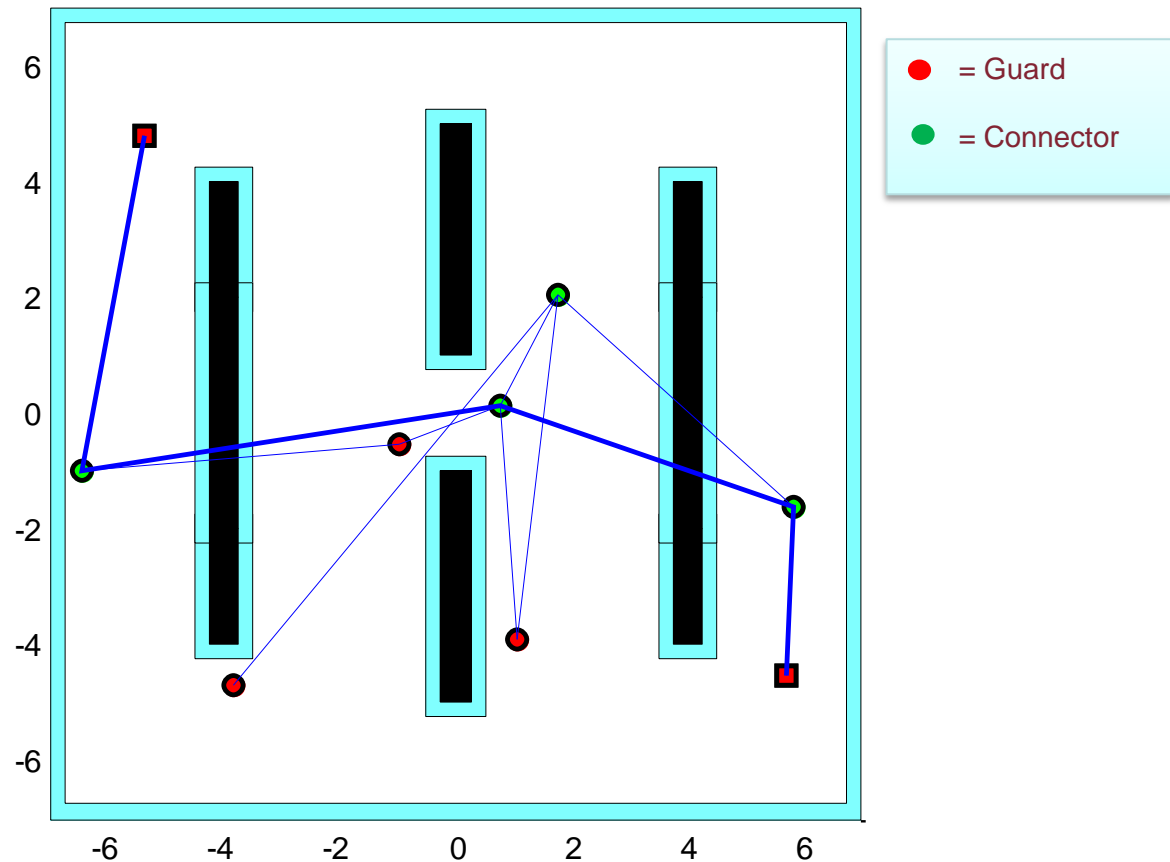
# Visibility-Based Motion Planning

## Roadmap construction



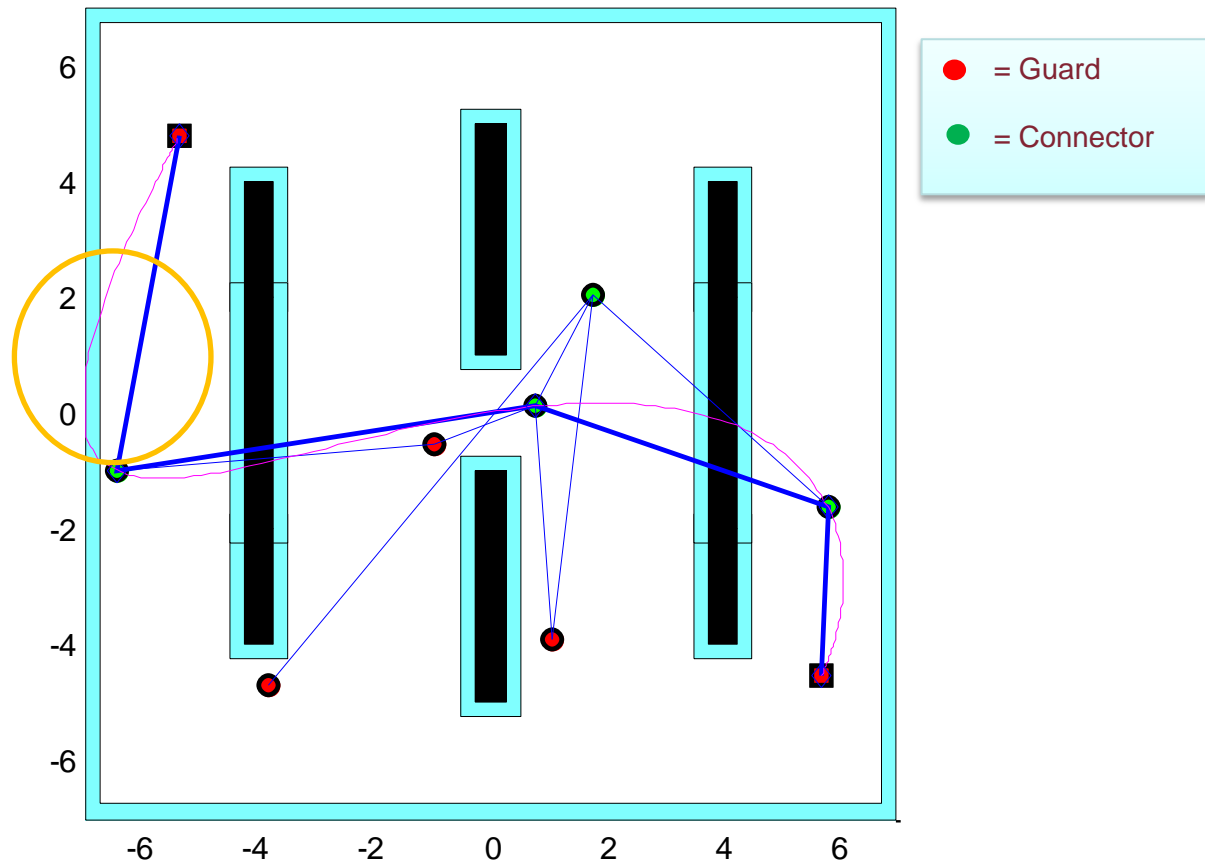
# Visibility-Based Motion Planning

Path computed by A\*



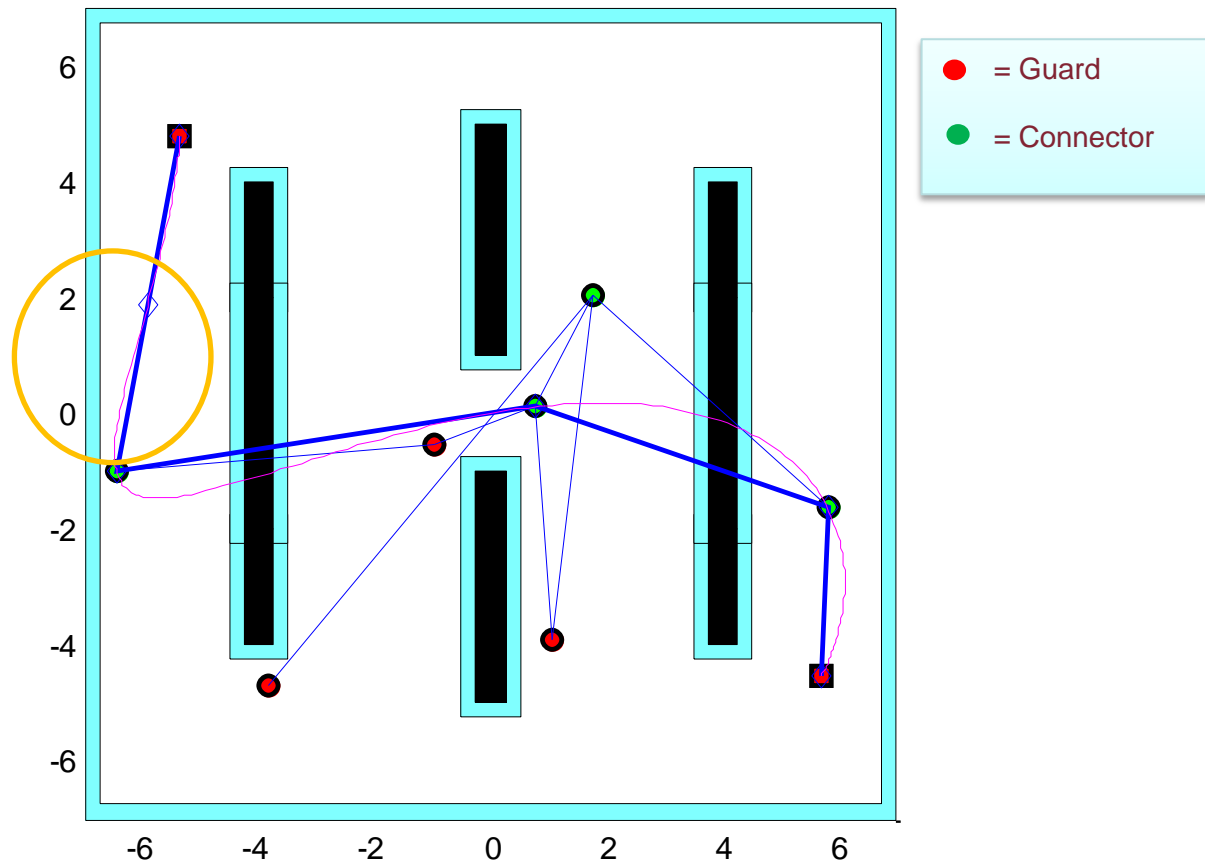
# Visibility-Based Motion Planning

Path Computed after the Optimization



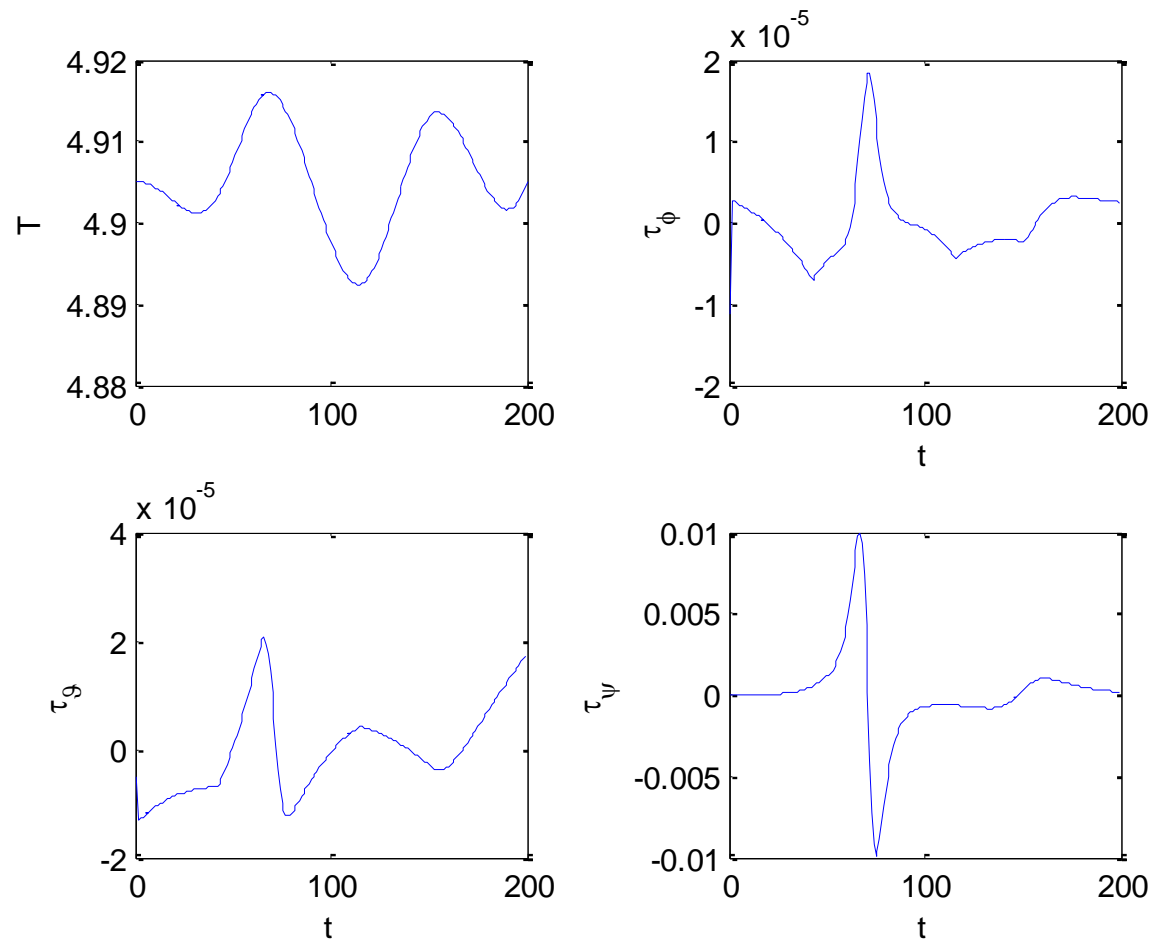
# Visibility-Based Motion Planning

Safe Path Computed after the Optimization



# Visibility-Based Motion Planning

## Time Scaling



# RRT-Based Motion Planning

## Introduction

**Key-idea:** *incrementally grow a **search tree** by applying control inputs over short time intervals to reach new nodes*

### Features:

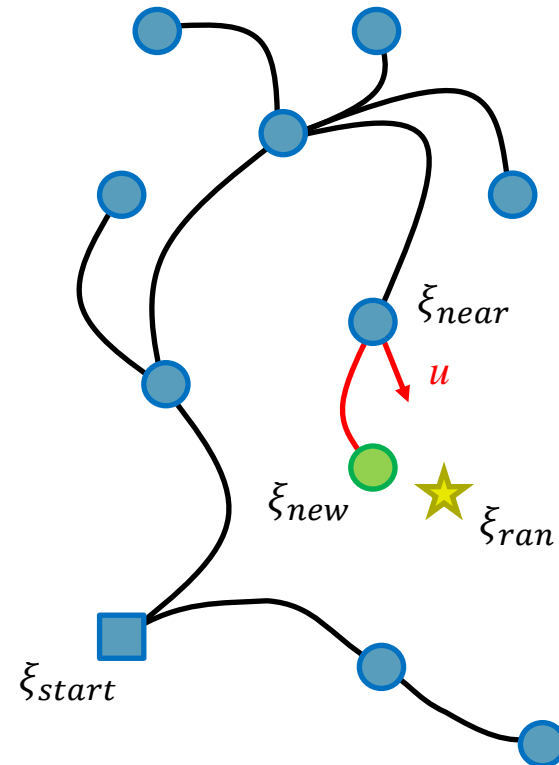
- **Exploration** biased toward unexplored portions of the space (Rapidly Exploring)
- Suitable for solving problems in **high-dimensional** spaces
- Takes into account both kinematic and dynamic **constraints**
- Generates a trajectory directly in the **state space**
- Provides control inputs needed to execute the trajectory

# RRT-Based Motion Planning

## Basic iteration

At each iteration:

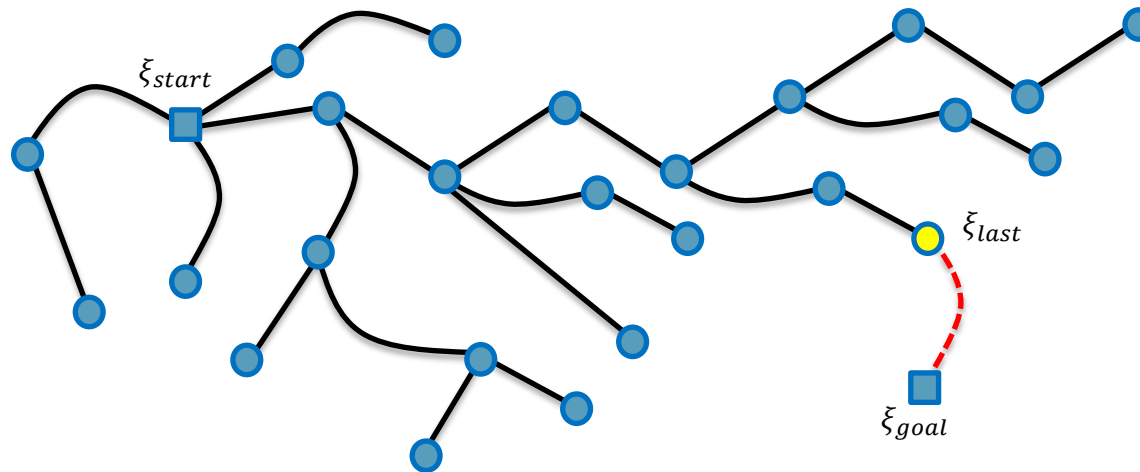
- Extract a random state  $\xi_{ran}$
- Select  $\xi_{near}$  (according to the metric)
- Compute  $u$  to steer the robot towards  $\xi_{ran}$
- Apply  $u$  for a time  $\delta t$ , reaching  $\xi_{new}$
- If the path from  $\xi_{near}$  to  $\xi_{new}$  is safe, add  $\xi_{new}$  to the tree and save  $u$
- Otherwise, discarded it



# RRT-Based Motion Planning

## Trajectory reconstruction

- Once  $\|\xi_{last} - \xi_{goal}\| < \epsilon$ , the algorithm stops
- The solution trajectory can then be found by traversing the tree backwards from  $\xi_{last}$  to  $\xi_{start}$
- The last stretch might be computed (if necessary) using any local planner





# RRT-Based Motion Planning

## Exploration vs. Exploitation

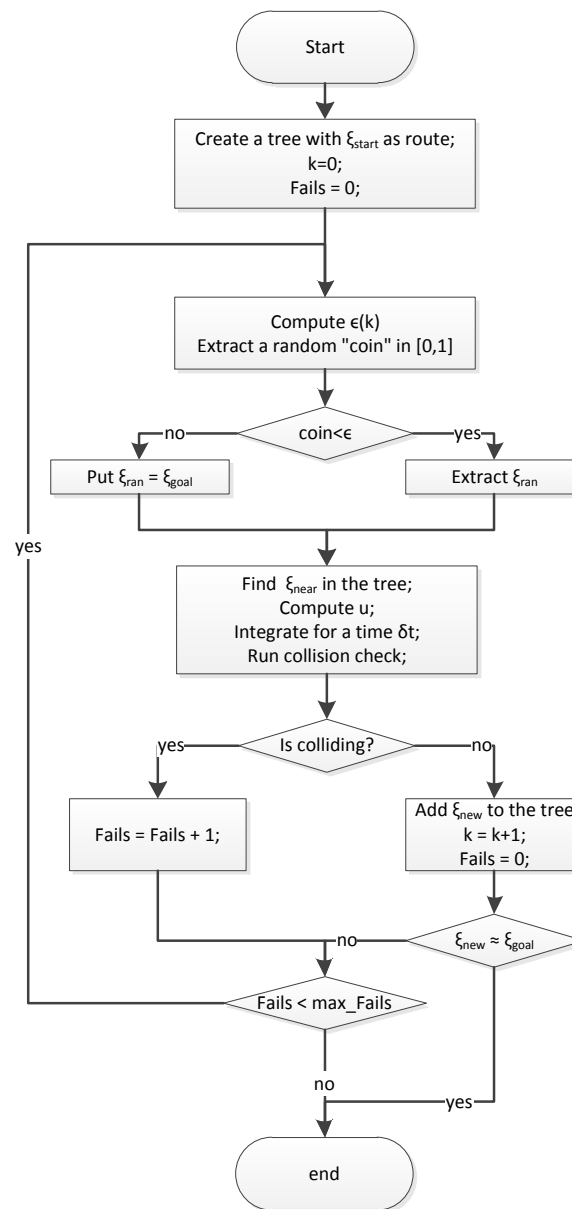
The tree will **eventually cover** the connected component of  $CS_{free}$  containing the start, coming **arbitrarily close** to any goal belonging to the same component

Convergence is often slow  switch between two phases:

- **Exploration**: the tree is expanded toward a random state
- **Exploitation**: the tree is expanded toward the goal ( $\xi_{ran} = \xi_{goal}$ )

**$\epsilon$ -greedy** strategy:

$$\epsilon = \frac{\epsilon_0}{1 + \alpha k} \Rightarrow \begin{cases} coin < \epsilon \rightarrow exploration \\ coin > \epsilon \rightarrow exploitation \end{cases}$$



# RRT-Based Motion Planning

Random inputs choice

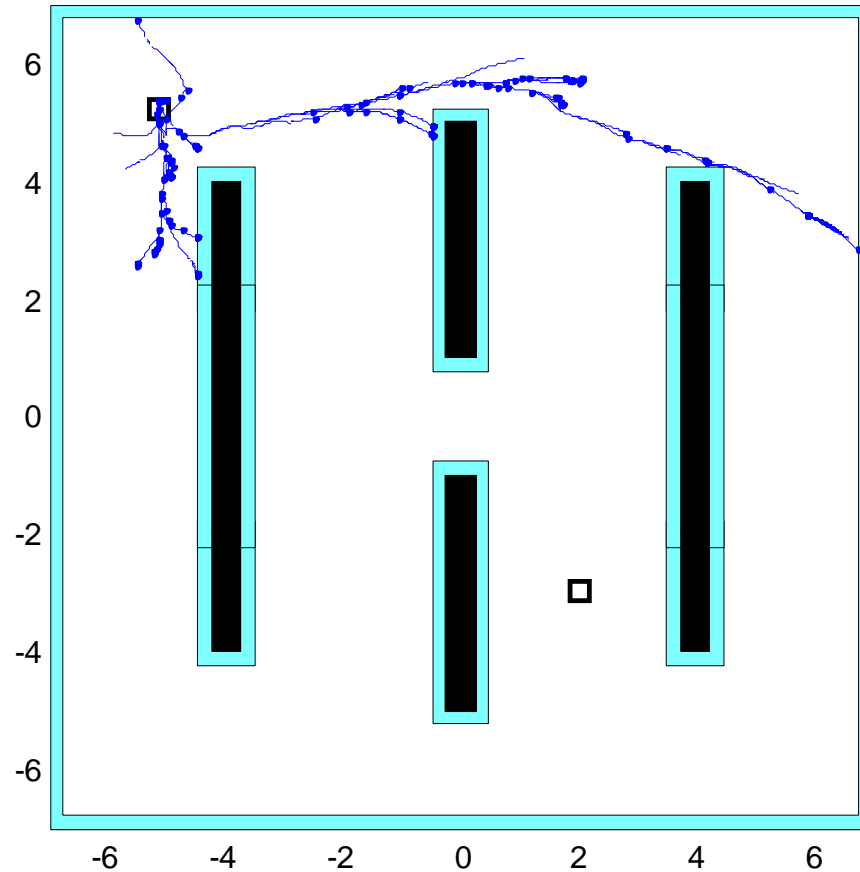
At each iteration:

- Extract 10 **random** values of the **motors velocities** between 0 and 1000 rad/s
- Compute the corresponding thrust and torques using input transformation
- Integrate the equations of motion (*ode45*) for a constant time  $\delta t = 0.1$  s
- Choose the input vector that brings (safely) the robot closer to  $\xi_{ran}$ .

# RRT-Based Motion Planning

Random inputs choice (Simulation Results)

6 hours later...



# RRT-Based Motion Planning

Closed-loop system

**Key-idea:** use a combination of the RRT planning along with a **controller** to ensure that the quadrotor moves toward  $\xi_{ran}$

Variable **LQR controller**: linearization around the current  $\xi_{near}$

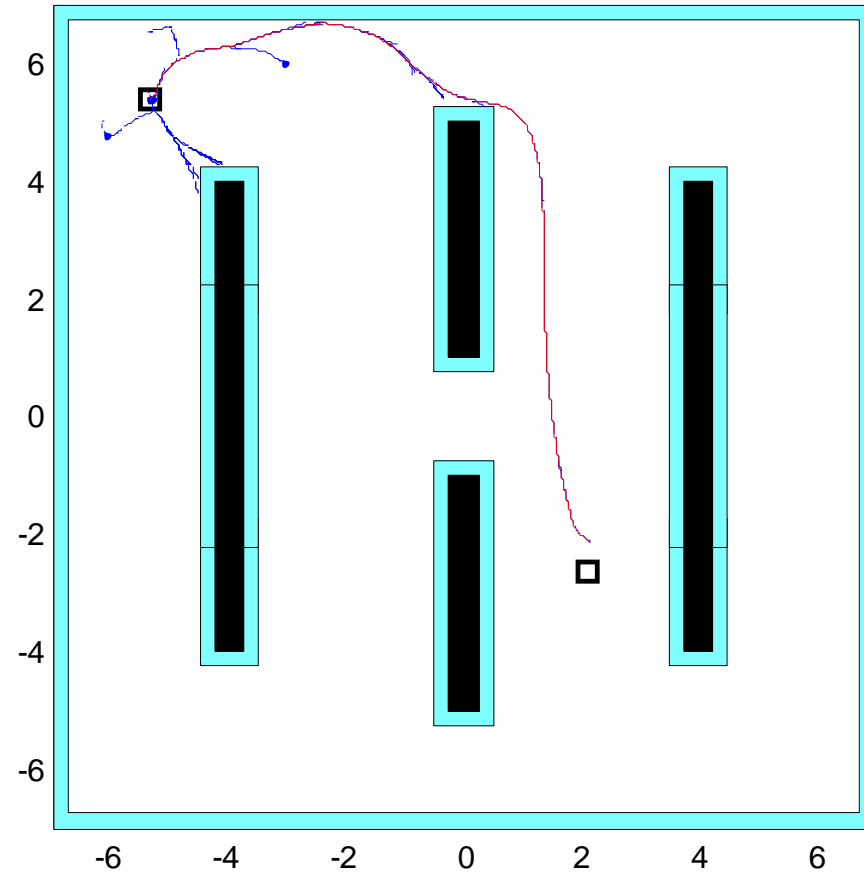
$$A = \left. \frac{\partial \dot{\xi}}{\partial \xi} \right|_{\substack{u=(mg \ 0 \ 0 \ 0)^T \\ \xi=\xi_{near}}} \quad B = \left. \frac{\partial \dot{\xi}}{\partial u} \right|_{\substack{u=(mg \ 0 \ 0 \ 0)^T \\ \xi=\xi_{near}}}$$

$K$  computed by Matlab<sup>®</sup> *lqr* function

$$u = (mg \ 0 \ 0 \ 0)^T - K(\xi - \xi_{ran})$$


# RRT-Based Motion Planning

Closed-loop system (Simulation Results)

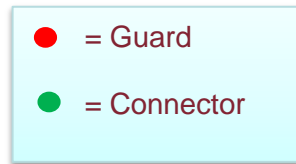
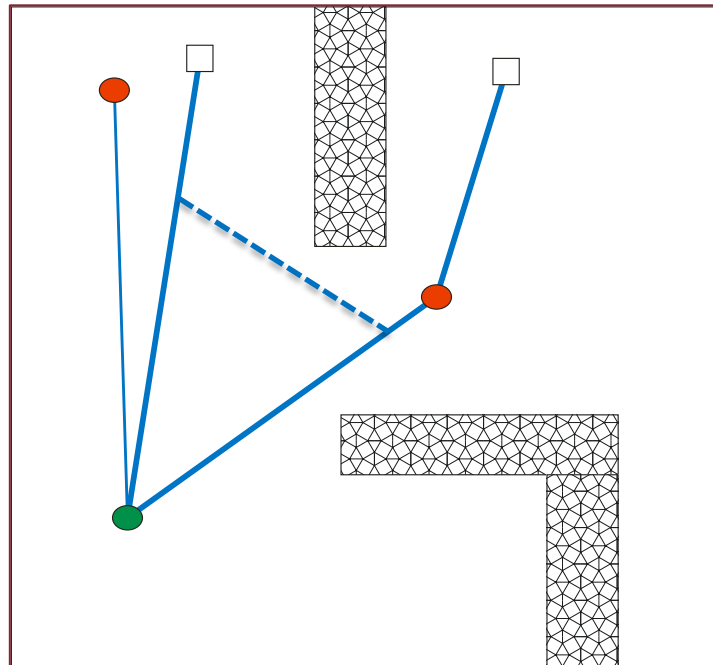


# Conclusions

(1)

- Visibility algorithm is suitable for finding a roadmap with **few nodes**  easy to handle but, possibly not containing the **shortest path**

Smart-cutting



# Conclusions

(2)

- The trajectory would be even smoother if the spline passed **near** to the nodes of the roadmap instead of interpolating them (more freedom to the optimizer)
- **RRT** planners seem to **perform worse**
- Enhancements may be achieved by:
  - **Tuning**  $\delta t$ ,  $\epsilon$  and the number of random inputs
  - Putting the system under the action of a controller
  - **Tuning** the gains  $Q$  and  $R$  of the LQR filter or by using any other controller (even non-linear)



# THANKS FOR LISTENING

