

Cooperative mobile robots.

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OUTLINE

1 Introduction

2 Collaborative robotics

3 Collaborative path planning

4 Control

Robotics is a cross fertilizing area which aims at designing and using concrete physical devices with the following capabilities :

- action, (actuators)
- perception, (sensors)
- decision,
- interaction with the environment,

in order to fulfill a task with or without a human.
(The case << not >> : human-robot interactions)

Classification des robots :

- **Robots mobiles** : à roues, à chenilles, à pattes, selon le type de locomotion, aérien, sous-marin, terrestre ou spatial, avec une attention particulière pour la robotique humanoïde
- Robots fixes : manipulateurs, interface haptique, etc ...

Introduction

Collaborative robotics

Collaborative path planning

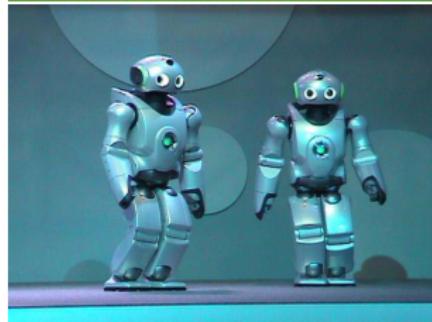
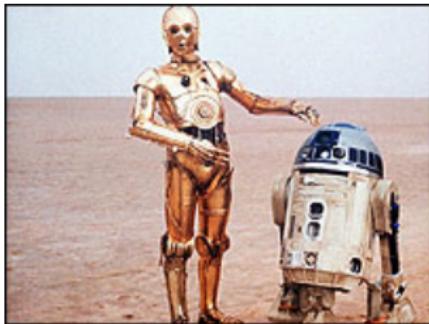
Control

Bibliography

Un Robot, c'est quoi ?

Robots en réseau





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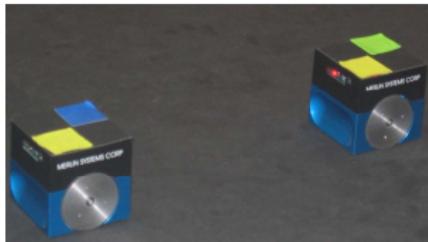
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La robotique concerne l'intégration de quatre composantes :

- **Conception, modélisation, analyse** : nouveaux besoins
- **Chaine de perception** : **capteurs** qui << nourrissent >> les robots d'informations de nature diverse (signaux analogiques, numériques (par exemple image), etc . . .).
- **Chaine d'action et de décision** qui comporte plusieurs chainons : cognitif, la planification de tâches, la planification de mouvements, la commande et se terminant par les actionneurs.
- **L'interaction** avec l'environnement : collaboration robots/robots et/ou hommes et/ou monde physique.

Réseaux informatiques notamment sans fils ont permis d'entrevoir la séparation de l'ensemble capteurs-commande-actionneurs (CCA).

Conséquences :

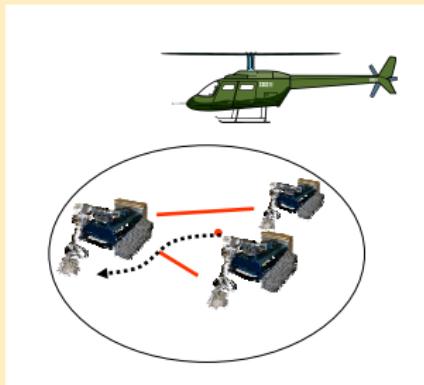
- téléopération de robots, (nouveaux enjeux).
- robots en réseaux : ce sont des dispositifs robotisés (manipulateurs, véhicules mobiles, robots humanoïdes, etc ...) qui sont connectés *via* un réseau de communication tel qu'un réseau local (LAN) ou le réseau internet (WAN) → faire coopérer un ensemble de robots.

Nouveaux problèmes : pertes de paquets, retards, QoS etc ...

☞ Robocoop project : <http://syner.ec-lille.fr/robocoop>

Goals

- Deployment of large scale networks of cooperative mobile robots

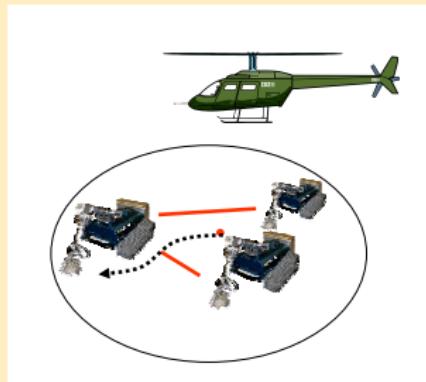


- to get complex behaviors by using simple agent based behaviors

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Applicative fields

- health (tele-robotics, ...)
- transportation (plane fleet, drones, mobile robots, heterogeneous robots (mobile of different type, planes, underwater robots, ...))
- security (fire, data collection for “spying”, ...)
- ...

Challenges

- local information and decision process,
- constrained communication + delays,
- large scale system,
- uncertain and hostile dynamic environment,
- ...

Framework : multidisciplinary research

- modeling, path planning and control (constraints, nonlinear models, time delays, hierarchical aspects, hybrid system aspect, quantization . . .)
- graph theory,
- communication protocols,
- logical decision making, scheduling,
- . . .

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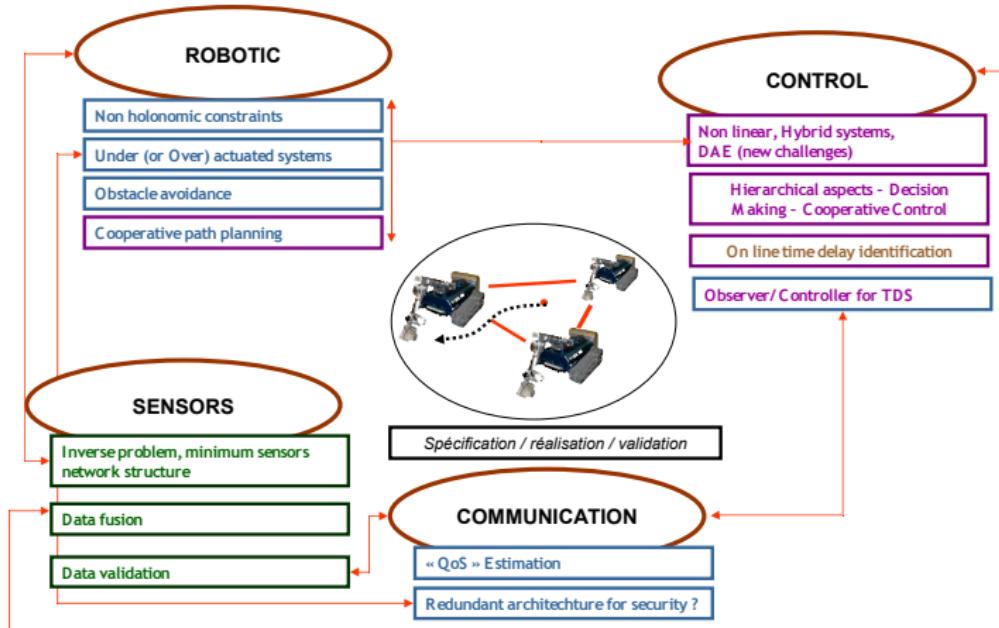
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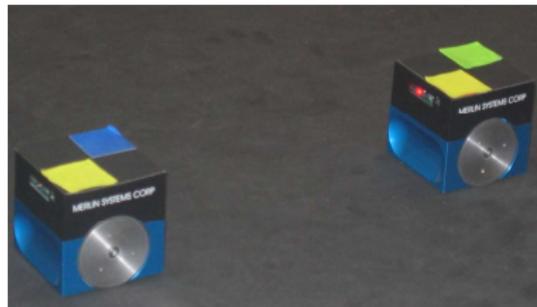
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Snap shot of Robocoop project / Big picture



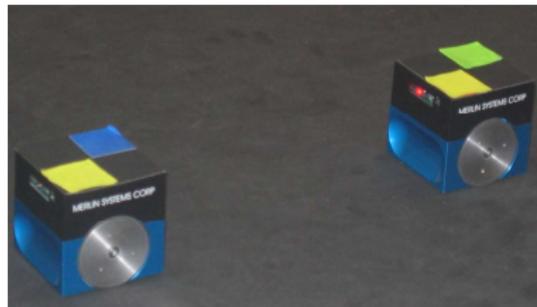
Our Goals

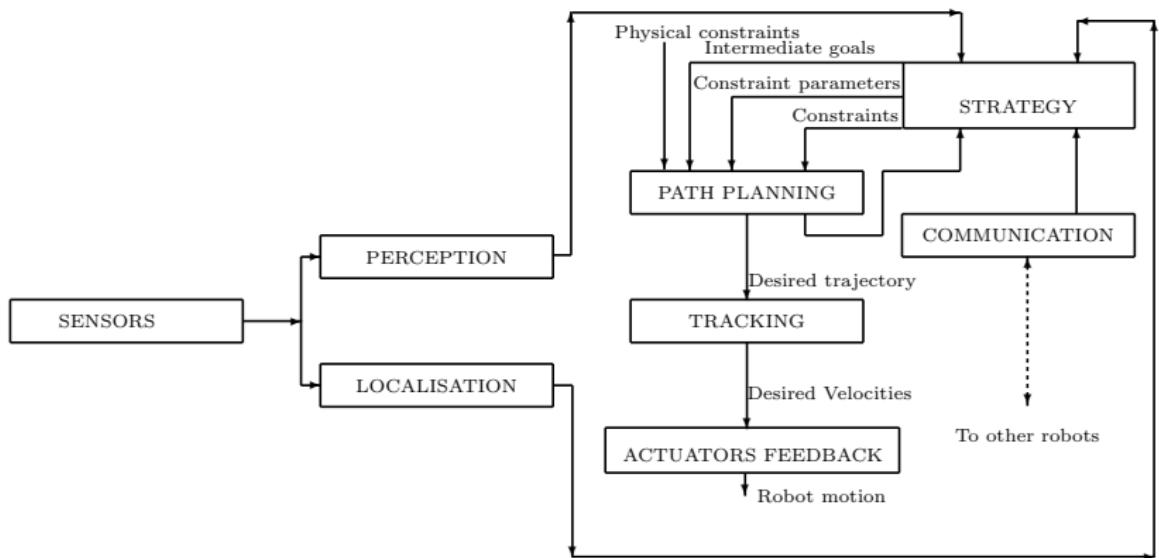
- ✓ path planning and path tracking
 - ✓ test on benchmarks



Our Goals

- ✓ path planning and path tracking
 - ✓ test on benchmarks





Leader or not ?

Within a group of mobile robots, some of them may play a particular role : **leaders**. Distinguish between fleets :

- ① **with** leader : the leader drive the whole fleet or a part of it.
- ② **without** leader : need of a **local/global** coordination : decision rules must use local informations (most of the time **neighbors**) or global informations

Questions

- ☞ How to **collect** such informations ?
- ☞ What happen if this robot dedicated to data collection is out of order, destroy, or not reliable ?

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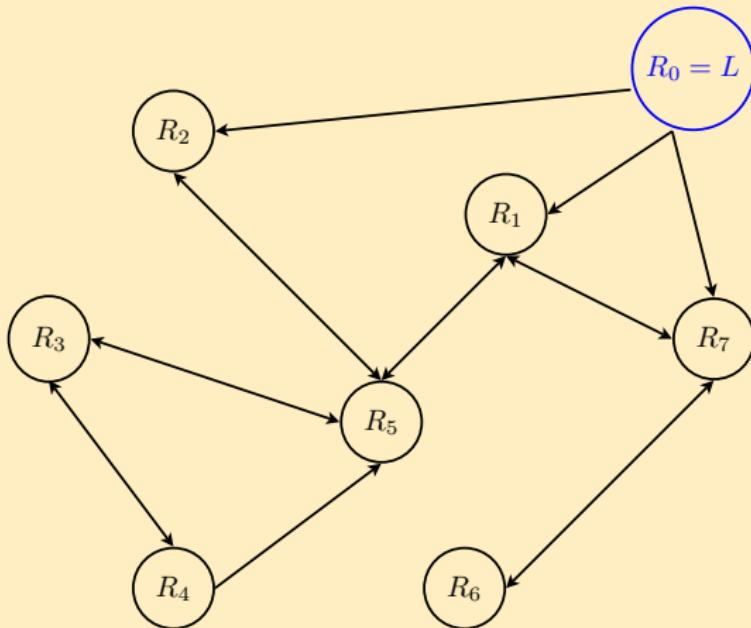
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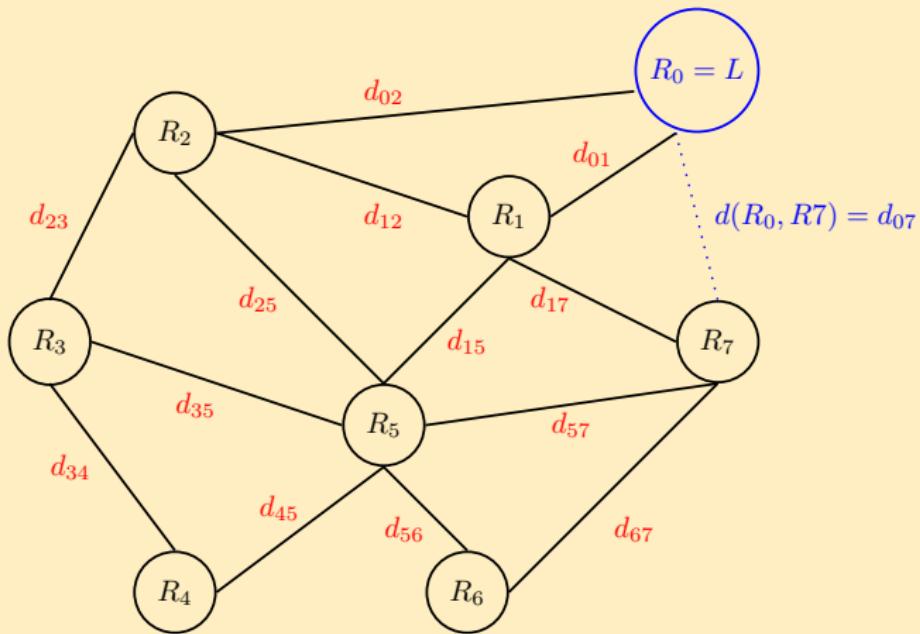
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Communications graph \mathcal{G}^c



Formation graph \mathcal{G}^f



Questions

- ☞ How to extract from a graph a **minimal representation** ensuring some properties (communications, geometric forms of the formation, . . .) ?
- ☞ According to some mission how to choose an **initial graph** which induces some good properties ?
- ☞ These graphs are **time varying** (dynamical graphs) :

Open questions : analyse, how to control ? . . .

Hierarchical structure

To achieve computational tractability :

- “Strategic layer” (higher level) : goal planning (for example choose an appropriate functional cost), task scheduling (for example use a petri net for description),
- “Tactical layer” (mid level) : guidance, navigation
- “Reflexive layer” (low level) : (control) state observation or estimation, trajectory tracking, ...

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How can we get an “integrated layer” ?

☞ Solve an optimisation problem which integrate some of these facts (gives a path) and then use a good “trajectory tracking”

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On the way to integration

Philosophy **integrated layer** :

“Strategic layer” (goal planning) + “Tactical layer” (guidance, navigation) + “Reflexive layer” (obstacle avoidance)

Solution

☞ Generate and execute a (sub)-optimal path planning which satisfy :

- geometric formation and communications constraints,
- obstacle avoidance constraints,
- given boundary conditions,
- other constraints : time constraints (rescue missions), energy constraints (batteries duration, ...)

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General constraints

We would like to include in the path planning the following constraints :

- ① some **constraints due to physics** (energy limitation, maximal velocity and acceleration of the robots)
- ② **obstacle avoidance**,
- ③ **collision avoidance** with the robots and other mobile objects,
- ④ **distances** between robots (communications),
- ⑤ **geometry** of the formation
- ⑥ ...

Somme settings

Group of N_i ^a mobile robots related to the i^{th} mobile robot evolving in a partially known space with N_o obstacles.

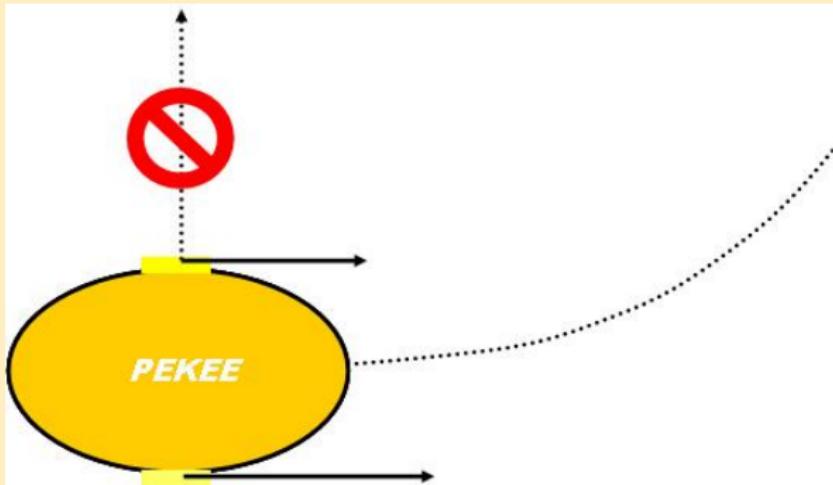
- each obstacle O_m ($m \in \{1, \dots, N_o\}$) is covered by a disc centered at (x_m^o, y_m^o) with radius r_m^o (if complex geometry use a covering of discs).
- I_N the set $\{1, \dots, N\}$,
- the $n^{\text{th}} \in I_N$ mobile robot denoted by A_n and located at (x_n, y_n) occupies a space modeled by a disc of radius r_n centered at (x_n, y_n) .
- robot A_n : X_n and U_n denotes respectively the state variables and the control variables.

 - a. Index i (dropped sometimes) will refer to some properties or known objects linked to the i^{th} mobile robot. $N = N_i$!



Robot de type unicycle

Non Holonomy (Admissible path)



Kinematic constraint : a robot should not be reduced to a single point (x, y).



Types of models : see [5] "Theory of Robot Control", C. Canudas de Wit, B. Siciliano and G. Bastin (Eds).

- ① Kinematic model (take into account non holonomic constraints)

Posture $=(x, y, \theta)$ in most of the case.

- ② Dynamical model (KM + dynamics induced by actuators (most of the time electrical motors))

Kinematic model is enough

Kinematic model which includes *non holonomic constraints* this is non integrable constraints of the form

$$\dot{q} = B(q)u, \quad (1)$$

where $u \in R^m, q \in R^n$ ($n > m$).

☞ If not under-actuated (with respect to the mobility degree) then one can perform a **feedback linearization** :

$$J(q)\dot{u} + C(q, u)u + G(q) = B^T(q)D(q)\Gamma,$$

leads to

$$\dot{u} = v$$

by using $\Gamma = (B^T(q)D(q))^{-1} (J(q)v + C(q, u)u + G(q)).$



Kinematic model : flatness is the key point

Flatness (see works from M. Fliess, J.Lévine, Ph.Martin, et P.Rouchon details in [11, 12, 13, 15, 16])

$\dot{x} = f(x, u)$, $x \in \mathbb{R}^n$, $u \in \mathbb{R}^m$ flat \Leftrightarrow il existe m fonctions qui paramétrisent tout !

Donc fixer ces fonctions c'est fixer le comportement dynamique du système !

Kinematic model : flatness is the key point

- ☞ Thus the PKM and PDM are **flat**.
- ☞ Thus it implies that they are **controllable**.
- ☞ But from Brockett's theorem (see [4]) they are **not stabilizable by a continuous static time-invariant state feedback**.

Kinematic model : flatness is the key point

① Unicycle mobile robot (type (2,0))

$$\begin{aligned}\dot{x} &= v \cos \theta \\ \dot{y} &= v \sin \theta \\ \dot{\theta} &= w\end{aligned}\tag{2}$$

② Car-like mobile robot (type (1,1))

$$\begin{aligned}\dot{x} &= v \cos \theta \\ \dot{y} &= v \sin \theta \\ \dot{\theta} &= v \frac{\tan(\phi)}{l} \\ \dot{\phi} &= w\end{aligned}\tag{3}$$

Kinematic model : flatness is the key point

Flat Outputs : (x, y) .

Indeed :

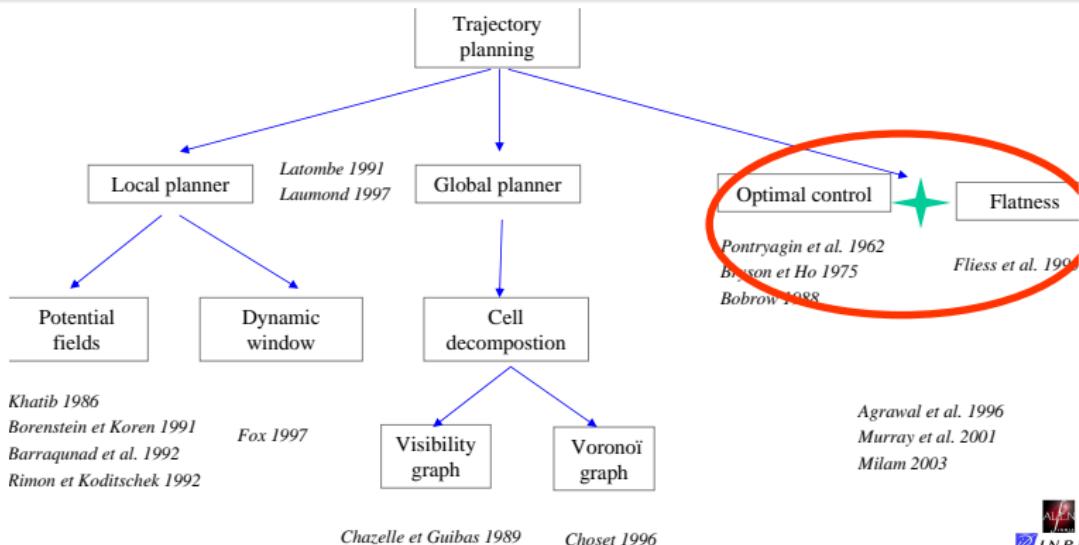
① for (2) : $\theta = \arctan\left(\frac{\dot{y}}{\dot{x}}\right), v = \pm\sqrt{\dot{x}^2 + \dot{y}^2}, w = \frac{\dot{x}\ddot{y} - \dot{y}\ddot{x}}{\dot{x}^2 + \dot{y}^2}$

② for (3) : $\theta = \arctan\left(\frac{\dot{y}}{\dot{x}}\right), v = \pm\sqrt{\dot{x}^2 + \dot{y}^2}, \phi = \arctan\left(l \frac{\dot{x}\ddot{y} - \dot{y}\ddot{x}}{(\dot{x}^2 + \dot{y}^2)^{3/2}}\right), w = \dot{\phi}$.

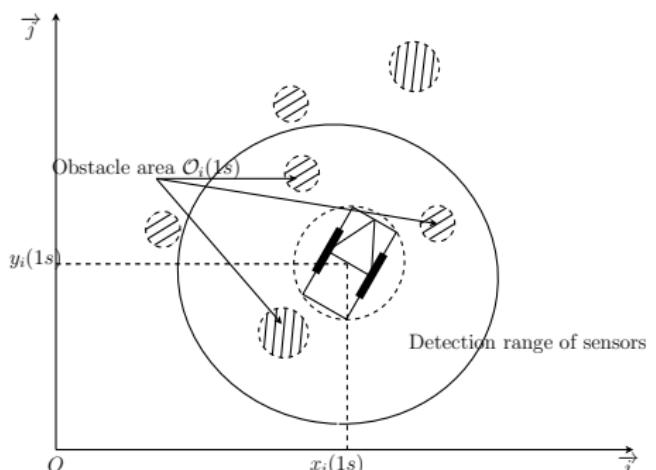
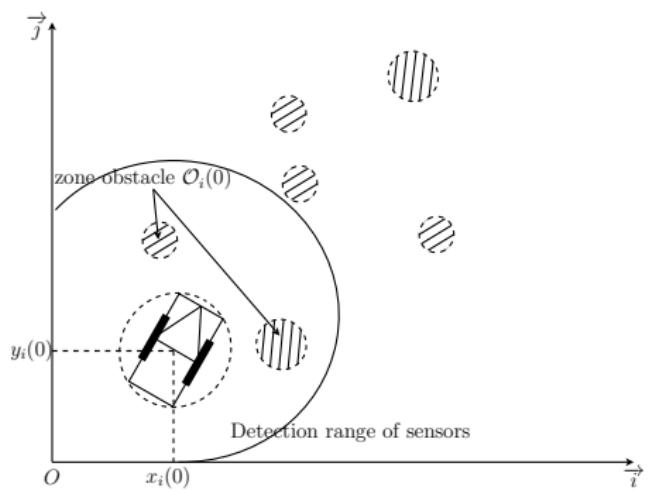
Path planning for a single robot

Motion planning

Computation of an executable collision-free trajectory for a robot between an initial given configuration and a final given configuration



Motion planning for a single robot : problem setup



Single robot : off-line algorithm

Dynamic optimisation based on flatness



Criteria :

$$J = \int_{t_{initial}}^{t_{final}} L_i(q_i, u_i, t) dt$$

wrt : $\forall t \in [t_{initial}, t_{final}],$

- $\dot{q}_i(t) = f_i(q_i(t), u_i(t))$
- $$\begin{cases} q_i(t_{initial}) = q_{i,initial} \\ q_i(t_{final}) = q_{i,final} \\ u_i(t_{initial}) = u_{i,initial} \\ u_i(t_{final}) = u_{i,final} \end{cases}$$
- $u_i(t) \in \mathcal{U}_i$
- $\forall O_{m_i} \in \mathcal{O}_i(t_{initial})$
 $d(q_i(t), O_{m_i}) \geq \rho_i + r_{m_i}$



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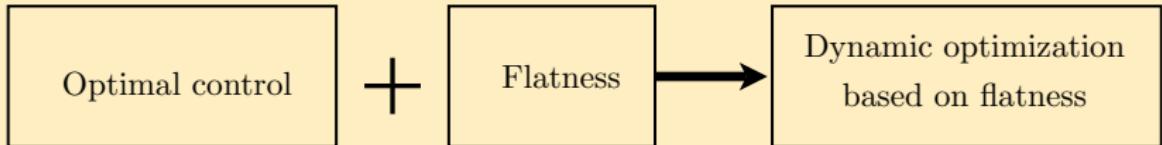
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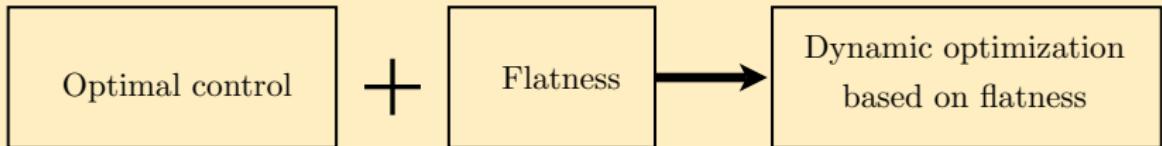
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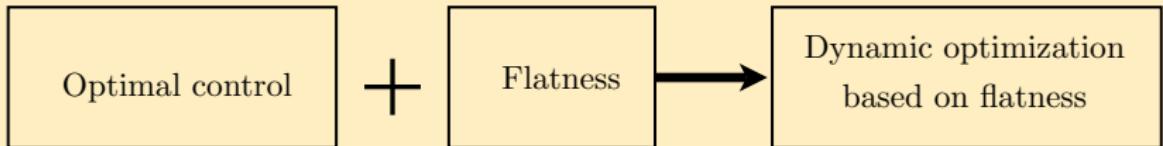
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$$\int_{t_{initial}}^{t_{final}} L_i(\varphi_1(z_i, \dot{z}_i, \ddot{z}_i), \varphi_2(z_i, \dot{z}_i, \ddot{z}_i), t) dt$$

wrt : $\forall t \in [t_{initial}, t_{final}]$,

- ...
- $\begin{cases} \varphi_1(z_i(t_{initial}), \dot{z}_i(t_{initial})) = q_{i,initial} \\ \varphi_1(z_i(t_{final}), \dot{z}_i(t_{final})) = q_{i,final} \\ \varphi_2(z_i(t_{initial}), \dot{z}_i(t_{initial})) = u_{i,initial} \\ \varphi_2(z_i(t_{final}), \dot{z}_i(t_{final})) = u_{i,final} \end{cases}$
- etc.



Single robot : off-line algorithm

Dynamic optimisation based on flatness

Optimal control



Flatness

Dynamic optimization
based on flatness

Resolution of optimal control problems

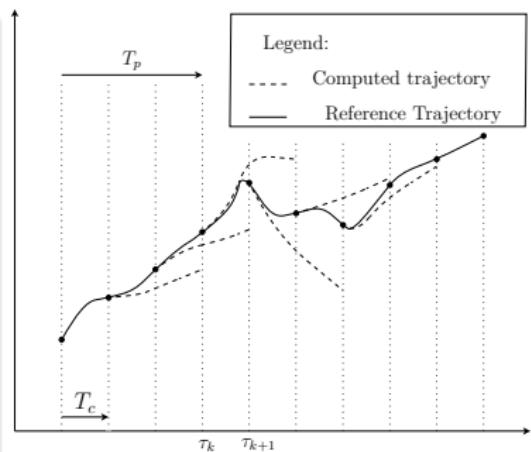
- ☞ Transformation into a **nonlinear programming** problem, using **B-spline** functions in order to approximate the trajectory of the flat output
- ☞ Computation of optimal control points using an optimisation procedure (**CFSQP**)
- ☞ Computation of the corresponding control inputs using the flatness properties of the system

Single robot : on-line algorithm

Main Principle

- ☞ To relax the constraint that the final point is reached during the planning horizon, allowing the use of an on-line receding horizon motion planner

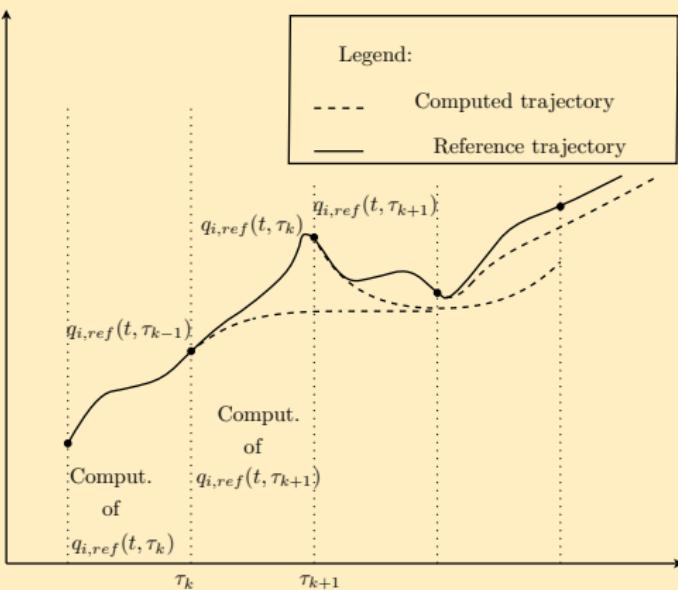
- $T_p (> 0)$: planning horizon
 - $T_c (> 0)$: update period
 - $\tau_k (k \in \mathbb{N}, \tau_k = t_{initial} + kT_c)$: updates



Single robot : on-line algorithm

Implementation

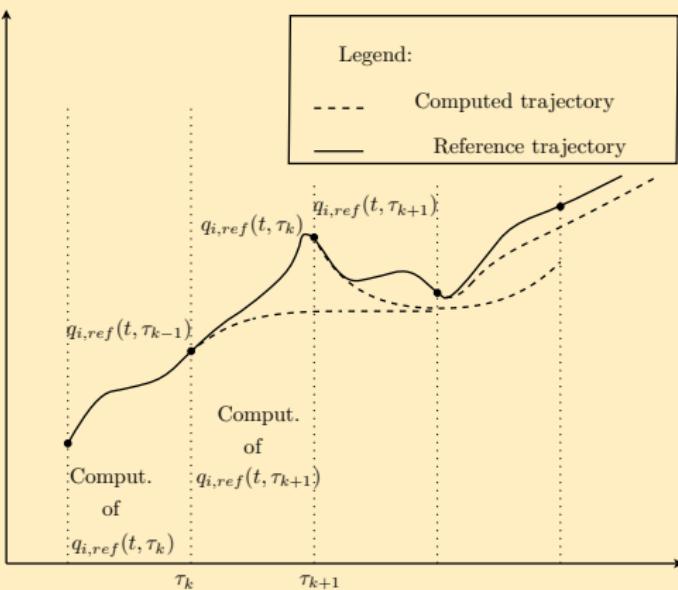
- ☞ **initialisation** step : computations before the movement of the robot
- ☞ step of iterative computations : computations over any interval $[\tau_{k-1}, \tau_k)$



Single robot : on-line algorithm

Implementation

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Multi-robots coordination

Objective

- ☞ To generate a (sub) optimal trajectory for each robot which satisfy :
 - terminal constraints
 - physical constraints (nonholonomic, maximum velocities, ...)
 - obstacle avoidance
 - **minimum distances between robots** (collision avoidance)
 - **maximum distances between robots** (respect of the broadcasting range)

Communication graph $(\mathcal{N}, \mathcal{A}, \mathcal{S})$

- Robots $\mathcal{N} = \{1, \dots, N_a\}$
- Edges $\mathcal{A} \subset \mathcal{N} \times \mathcal{N} \leftrightharpoons$ communication links
- Constraints of the edges

$d_{i,com} \in \mathbb{R}^+$: broadcasting range of robot i

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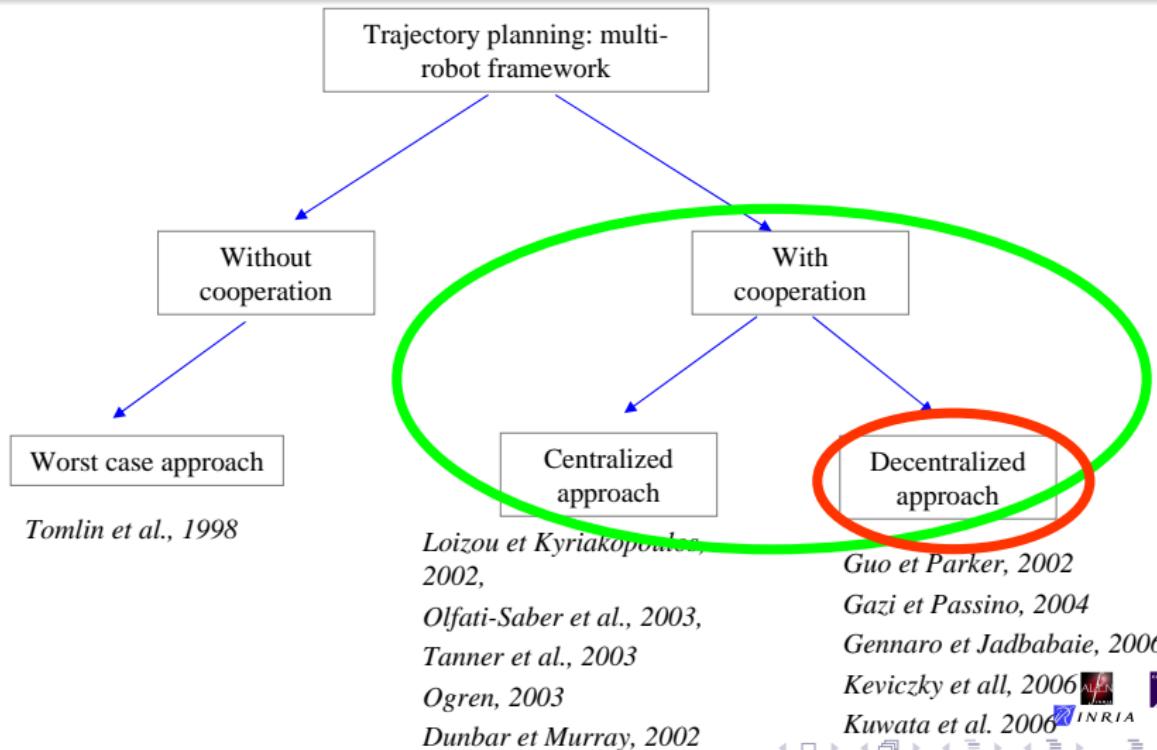
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Multi-robots coordination



Multi-robots coordination : centralized approach

Limitation to direct extension via a supervisor

- ① Prohibitive computation time (centralized approche))
- ② Problems due to the supervisor (if destroyed ...)

Solution to 1

Step of simplification of the initial problem :

☞ Motion planning of a virtual robot which is located at the centre of gravity of the formation

Multi-robots coordination : centralized approach

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Multi-robots coordination : decentralized approach

Desired objectives

- low computation time
- high performances
- use of available local information
- no supervisor

Solution

Distributed optimisation based on local information

- ❖ Each vehicle i only takes into account the intentions of the robots belonging to the conflict set $\mathcal{C}_i(\tau_k)$ (may produce a collision $\mathcal{C}_{i,collision}(\tau_k)$ or may lost the communication $\mathcal{C}_{i,com}(\tau_k)$)

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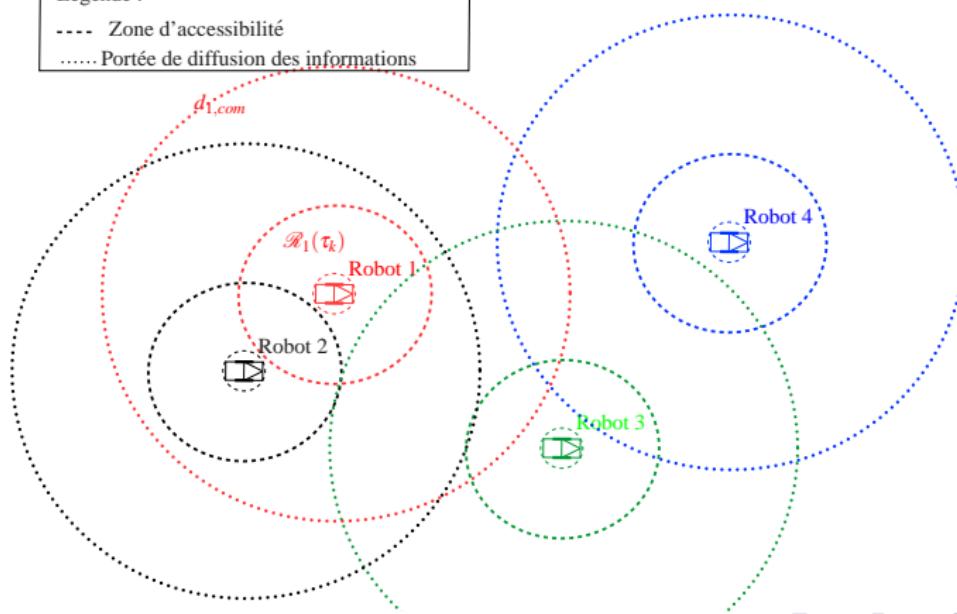
Conflicts with robot 1 :

$$\mathcal{C}_{1,collision}(\tau_k) = \{2\} \quad \mathcal{C}_{1,com}(\tau_k) = \{4\}$$

Légende :

----- Zone d'accessibilité

..... Portée de diffusion des informations



Multi-robots coordination : decentralized approach

Difficulties

Knowledge of the intentions of robots $p \in \mathcal{C}_i(\tau_k)$

- uniqueness of the presumed trajectory
- coherence between the presumed trajectory and the optimal planned trajectory

Solution

☞ Decomposition of the algorithm into 2 steps :

- * determination of the presumed trajectory (which only satisfy the individual constraints)
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Multi-robots coordination : decentralized approach

Few notations of robot i

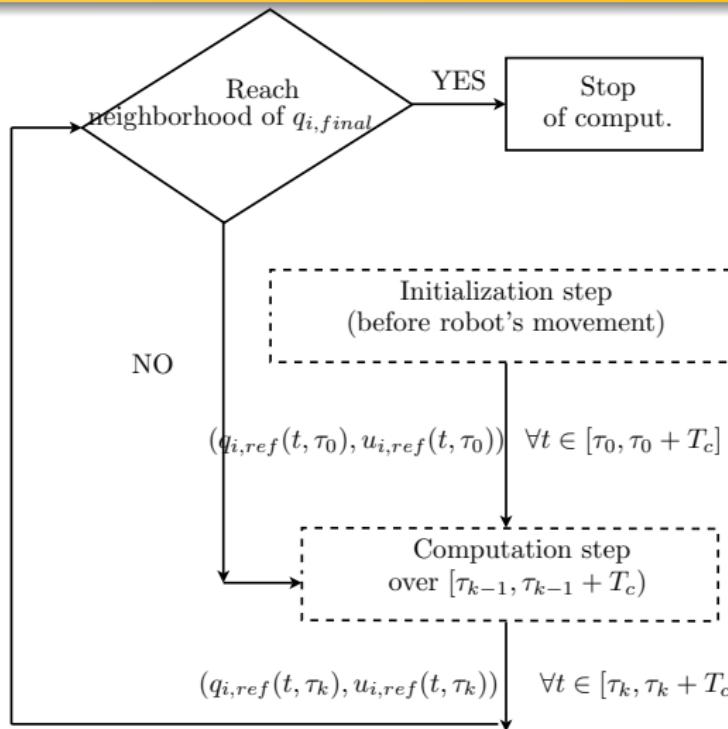
- Intuitive horizon $T_d \in \mathbb{R}^+$
- Planning horizon $T_p \in \mathbb{R}^+ (T_p \leq T_d)$
- Update horizon $T_c \in \mathbb{R}^+ (T_c \leq T_p)$
- $\hat{q}_i(t, \tau_k), \hat{u}_i(t, \tau_k)$: presumed trajectory of robot i beginning at τ_k with $t \in [\tau_k, \tau_k + T_d]$ and corresponding control inputs
- $q_{i,ref}(t, \tau_k), u_{i,ref}(t, \tau_k)$: optimal planned trajectory of robot i beginning at τ_k with $t \in [\tau_k, \tau_k + T_p]$ and corresponding control inputs

Multi-robots coordination : decentralized approach

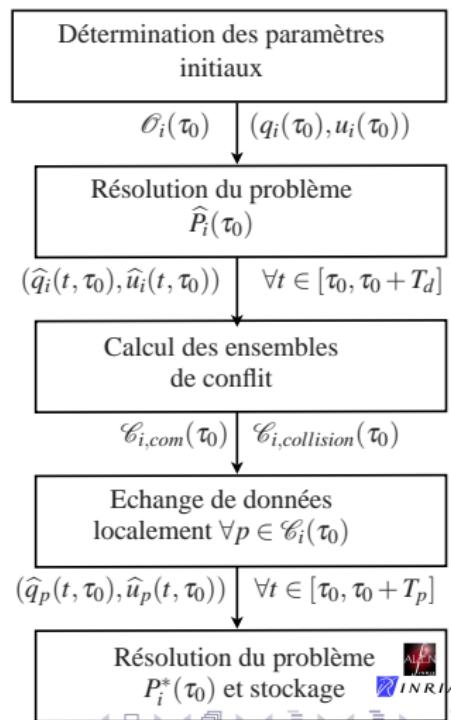
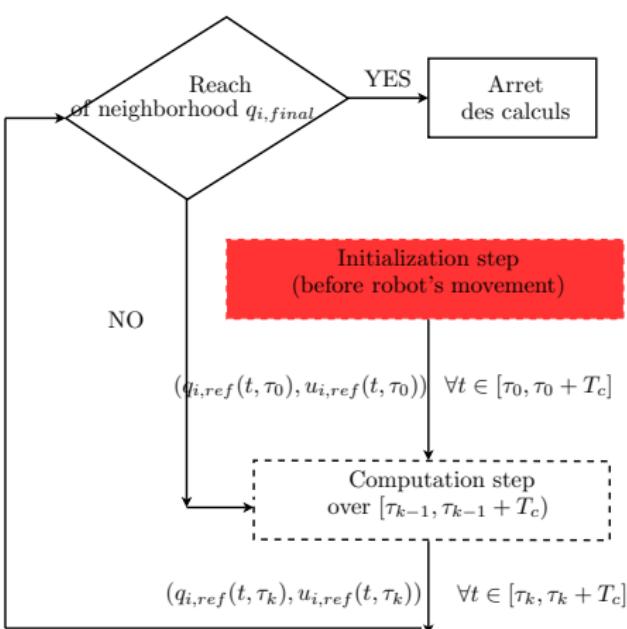
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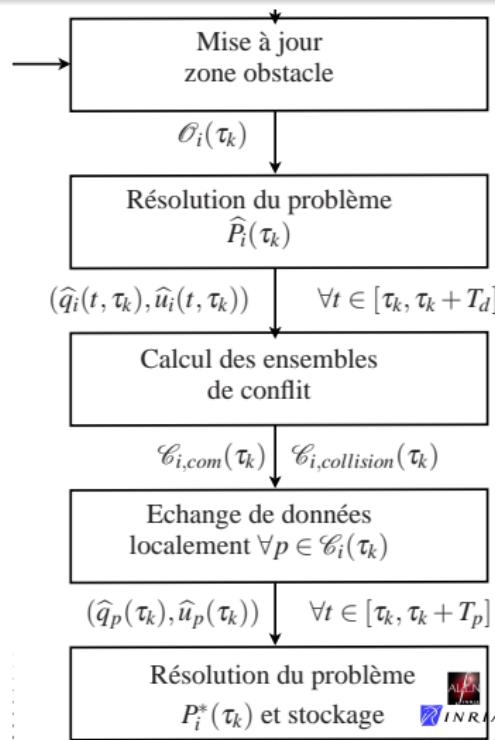
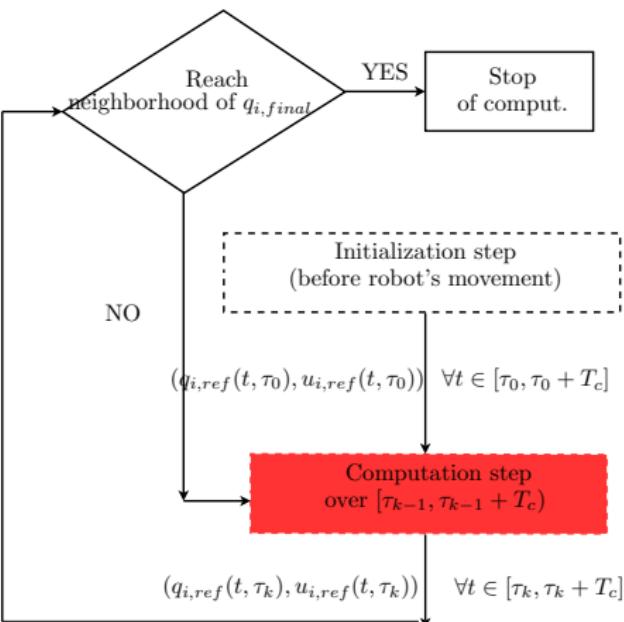
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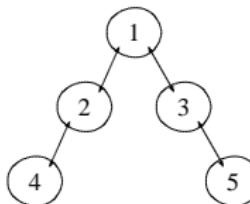
Multi-robots coordination : decentralized approach



Multi-robots coordination : decentralized approach



Multi-robots coordination : comparative results (2)



Number of robots N_a	5
Maximum linear velocity $v_{i,max}$	0.5m/s
Maximum angular velocity $\omega_{i,max}$	5rad/s
Radius of robot ρ_i	0.2m
Broadcasting range $d_{i,com}$	2.5m
Planning horizon T_p	2s
Update horizon T_c	0.5s
Intuitive horizon T_d	2.5s
Maximum deformation ξ	0.25

Multi-robots coordination : comparative results (2)

Approach	Cent.	Leader/Follower	Weakly decent.	Strongly decent.
Maxi time of conflict resolution	2050ms	313ms	703ms	121ms
Exchanged Info.	global	local	local	local
Implém.	-- if $N_a \gg 1$	++ sequential resolution	- if conflict with a lot of robots	+
Time reaching goal	35s	39s	36s	36.5s

Strongly decentralized

Video

Strongly decentralized

Video

Challenges

- Nonlinear dynamics,
- Presence of perturbations (unmodelled dynamics, sensor noise, external disturbances)
- How deal with the stabilization problem at low or zero velocity ?
- How to integrate cooperation into the control design ?
- Leader or not ?

Facts :

- 90 percent of the job is done by nominal control (path planning from which the open loop control is obtained thanks to differential flatness),
- 10 by feedback !

Several solutions were proposed, a **challenging problem being control design taking into account cooperation.**

Nouvelle architecture de commande (robots communicants) :

- nouveaux pb en Automatique (présence de retards **variables**, effet de quantification des données, aspects distribués de la commande . . .)
- partage des **capteurs extéroceptifs** (une caméra associée à un robot, un télémètre à un second, un compteur Geiger-Müller à un troisième, etc.) : traitement de l'information (conditionnement des signaux, fusion de données, etc...)

Quelques points durs :

- la conception et la commande de robots mobiles *via* des réseaux : dynamiques limitées (quelques m/s) ;
- capteurs distants : pbs liés à une identification performante + mesure temps réel précise de l'état du réseaux (perturbations électromagnétiques, pertes d'informations, gigue, retards, etc.) ;
- les QoS (actuelles) sur des réseaux hétérogènes asynchrones filaires et/ou sans fil ne sont pas adaptées en termes de temps de réponse, d'accès au médium pour la commande de processus ;
- les architectures de commandes logicielles et matérielles temps réel devront être adaptées à des systèmes communicants et distribués dont les systèmes d'exploitations seront certainement hétérogènes ainsi que les supports matériels.

La conclusion pour RECAP voici un nouveau terrain de jeux où les capteurs "bougent" et donnent naissance à de nouveaux problèmes pour la communauté scientifique

- topologie du réseau change dans le **temps et l'espace**,
- les entités qui bougent sont plus ou moins connues (modèles cinématique + dynamique : on peut avec les techniques de l'automatique **estimer** certaines variables).... il faut utiliser ces connaissances,
- il y a déjà eu, du temps des RTP, des interaction automatique/ informatique (commande à travers les réseaux, commande pour la gestion de congestion etc...) : let's do it again...
- ces problèmes d'optimisations en décentralisé ne peuvent-ils pas répondre à d'autres problèmes de RECAP ?

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