

Some perspectives in interactive Robotics

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Introduction - A bit of context

Observation

- Robotics is absent from many domains of human activities (industry, service)
- ↪ Complexity of the activities and their related environments

A few industrial examples

- Construction, maintenance and dismantling of industrial products
- Flexible production lines, Construction and public works, production and transportation of energy, naval construction and aeronautics, off-shore activities...

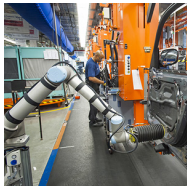


???

→



Introduction - A bit of context



Limits of “classical” industrial Robotics

- No versatility: pre-defined trajectories
- Intrinsically dangerous: high structural stiffness, high gear ratios and inertia, non back-drivability
- Perception capabilities ≈ 0



Characteristics of the potentially new application fields of Robotics

- Highly constrained and partially structured/known environments:
 - ▶ geometrically
 - ▶ mechanically
 - ▶ dynamically
- Multiple modes:
 - ▶ autonomous
 - ▶ teleoperated
 - ▶ collaborative
 - ▶ mixt
- Contacts are intrinsically present:
 - ▶ Under-actuated systems with non fixed base
 - ▶ Comanipulation
 - ▶ Task related mechanical actions applied to the environment



Introduction - A bit of context



What possibilities for the design of robots in this context ?

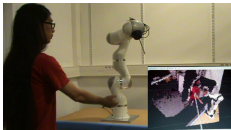
- Morphology and mechatronics properties
- Control laws
- Perception capabilities
- Control architectures

Design objectives ?

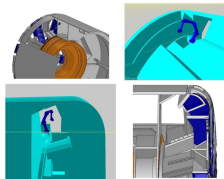
- Physical and cognitive ergonomy for the end users
- Safety wrt the environment
- Adaptation capabilities
- Performances

→ **Complex problems!**

Scope of the presentation



Multi-tasks control under constraints



A few words about the automatic design of dedicated robotics architecture



A few words about learning and adaptation

Multi-tasks control under constraints

Considered System

$$M(\mathbf{q})\ddot{\mathbf{q}} + \mathbf{n}(\mathbf{q}, \dot{\mathbf{q}}) + \epsilon(\mathbf{q}, \dot{\mathbf{q}}, \dots) = \mathbf{J}_c(\mathbf{q})^T \boldsymbol{\chi}, \quad (1)$$

with $\boldsymbol{\chi} = [\mathbf{w}_c^T \ \boldsymbol{\tau}^T]^T$:

- actuation torque ($\boldsymbol{\tau} \in \mathbb{R}^{n_a}$)
- external wrenches ($\mathbf{w}_c = [\mathbf{w}_{c,1}^T \ \dots \ \mathbf{w}_{c,n_c}^T]^T$)

Constraints

- physics: (1)
- actuation limits (max torque and speed);
- obstacles (joint limits, environment);
- contact wrenches (existence conditions, magnitude...).

$$\mathbf{G}(\mathbf{q}, \dot{\mathbf{q}})\boldsymbol{\chi} \leq \mathbf{h}(\mathbf{q}, \dot{\mathbf{q}}). \quad (2)$$

Some prerequisites

- good knowledge of a realistic model of the system: modeling + identification + adaptation/learning
- perception and signal processing capabilities

Multi-tasks control under constraints

Task ?

- A function from the joint space to the operational space with characteristics:
 - ▶ a physical frame
 - ▶ a task parametrization $\xi_i \in \mathbb{R}^{m_i}$
 - ▶ a forward model $\ddot{\xi}_i = J_i(q)\ddot{q} + \dot{J}_i(q, \dot{q})\dot{q}$
 - ▶ a desired trajectory $\xi_i^*(t)$ obtained through
 - ★ Global planning
 - ★ On-line min-jerk like planning
 - ★ Predictive control / Optimal control
 - ★ Motion primitives learned by demonstration (e.g. DMPs)
 - ★ CPGs (!)
 - ★ ...
 - ▶ a local feedback controller (P(I)D with acceleration feedforward term)
$$\ddot{\xi}_i^d = \ddot{\xi}_i^* + K_p e + K_d \dot{e} (+ K_i \int e dt)$$
 - ▶ a level of hierarchical importance α_i with respect to other tasks

Multi-tasks control under constraints

Task ?

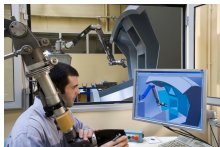
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$$\ddot{\xi}_i^d = \ddot{\xi}_i^* + K_p \mathbf{e} + K_d \dot{\mathbf{e}} + K_i \int \mathbf{e} dt$$
- ▶ a level of hierarchical importance α_i with respect to other tasks

→ Find τ at each control instant in order to maximize the task performance under constraints

Multi-tasks control under constraints: the Telemach example

Objective: safe and reactive control law in a static environment

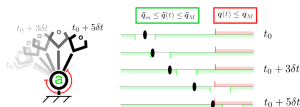
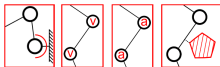


Reactive ?

- reactive \neq planned **movie**
- Operational objectives are set in real-time
- Teleoperation, exteroceptive sensor-based control,...

Safe ?

- Robot related: joint position, velocity and torque limit
- Environment related: no collision



Guarantee the existence of a solution to the control problem $t \rightarrow \infty$

- Modify the constraints expression a priori
- Acceleration capabilities of the system in operational space (!)
- Check on-line the existence of an escape trajectory

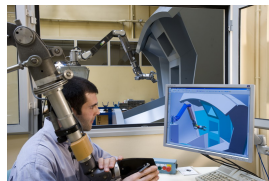
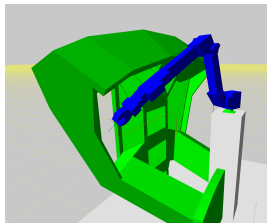
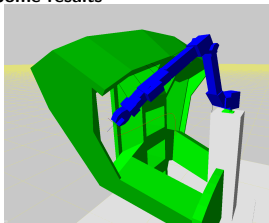
Multi-tasks control under constraints: the Telemach example

Constraints Compliant Control (CCC)

- 3-level hierarchical control law
 - ▶ Highest priority task: passive avoidance
 - ▶ Secondary task: task objective $\dot{\xi}_O = J_O(q)\dot{q}$
 - ▶ Smallest priority task: active avoidance
 - ▶ Tikhonov regularization of the Jacobian pseudoinverse
 - ▶ Heuristic iteration over the constraints to be avoided passively
- Hypothesis: constraints are compatible (a solution exists)
- $\dot{q} = J_c^+ 0 + (J_O P_{J_c})^+ \dot{\xi}_O^d + (J_{\bar{c}} P_{\begin{bmatrix} J_c \\ J_O \end{bmatrix}})^+ (\dot{\xi}_{\bar{c}}^d - J_{\bar{c}} (J_O P_{J_c})^+ \dot{\xi}_O^d)$

→ Under optimal LQP solution

Some results



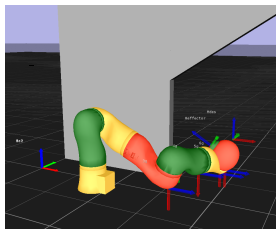
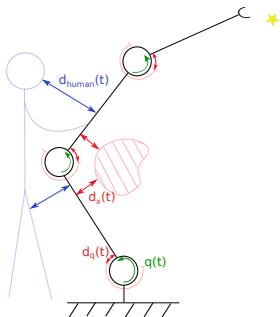
Multi-tasks control under constraints: workspace sharing

Objective: safe and reactive control law in a human/robot workspace sharing context

- ↪ Challenge 1 - Dynamic performance vs User safety
- ↪ Challenge 2 - Intuitive and continuous mode switching

! Energetic approach

- ↪ Classical obstacle avoidance (no contact) + ...
- ↪ ... monitored robot energy contact
- ↪ Considered energy for each segment: Σ_c (noo contact) + $\Sigma_{p,elastique}$ (contact)



Multi-tasks control under constraints: workspace sharing

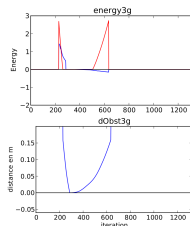
Control law

$$\min_{\chi, \ddot{q}} \frac{1}{2} \left(\beta \|\ddot{\xi}^d - J\ddot{q} + \dot{J}\dot{q}\|^2 + \delta \|E_{i/j}\|^2 \right)$$

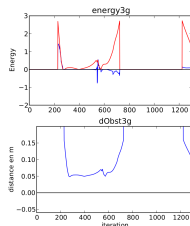
s.t. $M(q)\ddot{q} + n(q, \dot{q}) = J_c(q)^T \chi$

$$G(q, \dot{q})\chi \leq h(q, \dot{q})$$

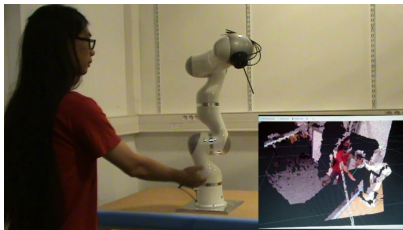
$$E_{i/j} \leq E_{limite}^i(d, d_{safe}, E_{safe}, d_{max}, k)$$



(a) Sans contrainte énergétique

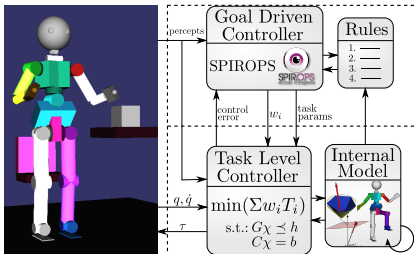


(b) Avec contrainte énergétique



Multi-tasks control under constraints: complex activities

Objective: complex activities through combination and sequencing of elementary tasks



Specificities

- Non strict task hierarchies (weighting strategy)
- Position, wrench or impedance tasks
- Fuzzy decision making engine: tasks weights and activations
- Reactive approach but ...
- ... model predictive control to generate $\ddot{\xi}_{CdG}^d$
- Task transition mechanism

Limitations

- No strict hierarchies between tasks
- Reactive approach: no anticipation of disturbances related to task incompatibilities at short term

Multi-tasks control under constraints: tasks hierarchies

Objective: Unique formalism to define both strict and soft tasks hierarchies and allowing to modify continuously priority levels

Existing approaches

- ▶ Analytical \rightarrow inequality constraints are problematic
- ▶ Numerical (through optimization, QP type)
- ▶ In both case two paradigms: strict hierarchies (SoT) or soft hierarchies but no global approach

Soft hierarchies through weighting

$$\left\{ \begin{array}{ll} \min_{\mathbf{w}_1, \mathbf{w}_2, \dot{\mathbf{q}}} & \frac{1}{2} [k_1 \|\mathbf{w}_1\|^2 + k_2 \|\mathbf{w}_2\|^2 + \varepsilon \|\dot{\mathbf{q}}\|^2] \\ \text{s.t} & \mathbf{w}_1 = \dot{\mathbf{x}}_1 - \mathbf{J}_1 \dot{\mathbf{q}} \\ & \mathbf{w}_2 = \dot{\mathbf{x}}_2 - \mathbf{J}_2 \dot{\mathbf{q}} \end{array} \right. \quad (3)$$

Strict hierarchies

$$\left\{ \begin{array}{ll} \min_{\mathbf{w}_1, \dot{\mathbf{q}}} & \frac{1}{2} [\|\mathbf{w}_1\|^2 + \varepsilon \|\dot{\mathbf{q}}\|^2] \\ \text{s.t} & \mathbf{w}_1 + \mathbf{J}_1 \dot{\mathbf{q}} = \dot{\mathbf{x}}_1 \end{array} \right. \quad (4)$$

$$\left\{ \begin{array}{ll} \min_{\mathbf{w}_2, \dot{\mathbf{q}}} & \frac{1}{2} [\|\mathbf{w}_2\|^2 + \varepsilon \|\dot{\mathbf{q}}\|^2] \\ \text{s.t} & \mathbf{w}_2 + \mathbf{J}_2 \dot{\mathbf{q}} = \dot{\mathbf{x}}_2 \\ & |\dot{\mathbf{x}}_1 - \mathbf{J}_1 \dot{\mathbf{q}}| \leq \mathbf{w}_1^* \end{array} \right. \quad (5)$$

Multi-tasks control under constraints: tasks hierarchies

$$\begin{aligned} & \arg \min_{\ddot{\mathbf{q}}', \chi} \sum_{i=1}^{n_t} \left\| \mathbf{f}_i \left(\ddot{\mathbf{q}}'_i, \xi_i^d \right) \right\|^2 + \left\| \begin{bmatrix} \ddot{\mathbf{q}}' \\ \chi \end{bmatrix} \right\|_{Q_r}^2 \\ & \text{subject to} \end{aligned}$$

$$J_c(\mathbf{q})^T \chi = M(\mathbf{q}) P \ddot{\mathbf{q}}' + \mathbf{n}(\mathbf{q}, \dot{\mathbf{q}})$$

$$G(\mathbf{q}, \dot{\mathbf{q}}) \begin{pmatrix} P \ddot{\mathbf{q}}' \\ \chi \end{pmatrix} \leq \mathbf{h}(\mathbf{q}, \dot{\mathbf{q}}),$$

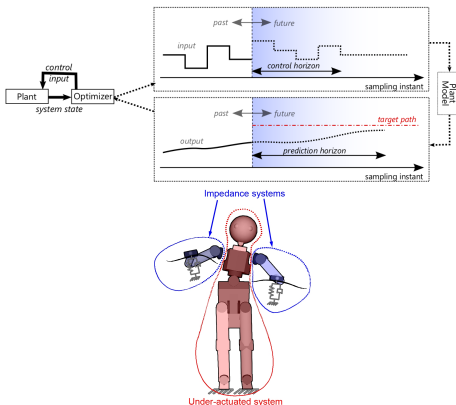
$$\text{with } \ddot{\mathbf{q}}' = \begin{bmatrix} \ddot{\mathbf{q}}'_1 \\ \vdots \\ \ddot{\mathbf{q}}'_{n_t} \end{bmatrix} \text{ and } P = [P_1(\alpha_1) \dots P_{n_t}(\alpha_{n_t})].$$

Specificites

- **Non** lexicographic notion of what priority is (no global order)
- By default, projection of task i in the kernel of others
- Continuous modification of the projector onto the kernel associated to task i : $i > j \leftrightarrow j > i$
- Implicit task insertion/deletion mechanism
- Continuous control approach

Multi-tasks control under constraints: tasks hierarchies

Objective: robustness of the solutions to the multi-tasks control under constraints problem



General principles

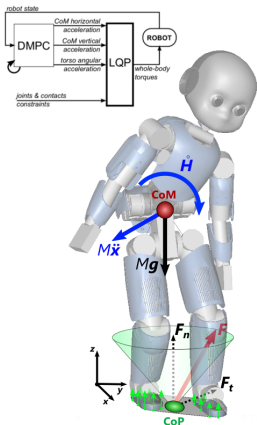
- Looking for an optimal control input horizon (sliding window)
- Use of a computationally inexpensive model of the system which dynamics has to be pre-visualized
- Locomotion: generation of reference trajectories for the center of mass position

Specificities

- Extension of the existing ZMP models to consider external disturbances
- Optimal horizon of control gains for the manipulation tasks (impedance adaptation)

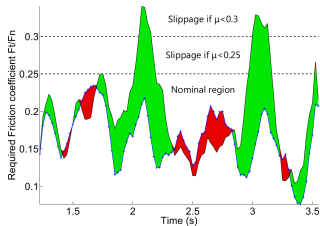
Multi-tasks control under constraints: tasks hierarchies

Objective: robustness of the solutions to the multi-tasks control under constraints problem

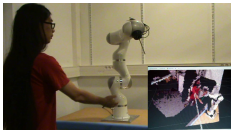


Specificities

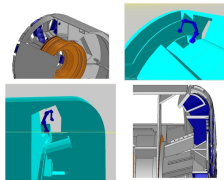
- Distributed approach
- Optimal trade-off between several contradictory objectives for the center of mass
- Minimization of the tipping-over risk and optimization of the friction conditions
- Optimal horizon of horizontal and vertical center of mass acceleration and torso angular acceleration



Scope of the presentation



Multi-tasks control under constraints



A few words about the automatic design of dedicated robotics architecture



A few words about learning and adaptation

Automatic design of dedicated robots

Objective: Optimization of the morphology of a manipulator evolving in an highly cluttered environment

Observation

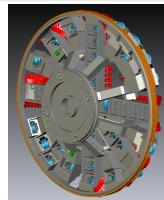
- No intuitive solution
- Combinatorics is high
- Classical design method are not adapted

↪ **MO optimization problem (MO)**

How to solve it ? ... knowing that

- looking for an arrangement of bodies and joints
- Optimization criteria of very different nature / contradictory
- Evaluation of the solution wrt specific tasks ot be performed → simulations
- The structure of the space of solutions is unknown (local minima ...)
- Large optimization problem

↪ **Stochastic optimization using MO evolutionary algorithms**

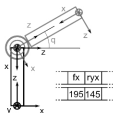


Automatic design of dedicated robots

Type	Lengths (m)
10: none	0: 0.05
11: rxx	1: 0.15
12: rxy	2: 0.25
13: rxz	3: 0.35
14: ryx	4: 0.45
15: ryy	5: 0.55
16: ryz	6: 0.65
17: rzx	7: 0.75
18: rzz	8: 0.85
19: ex	9: 0.95
20: ey	
21: ez	

ex : gene 163
ryz
length : 0.35

Fig. 3. Genotype table



2 segments

How to represent a robot

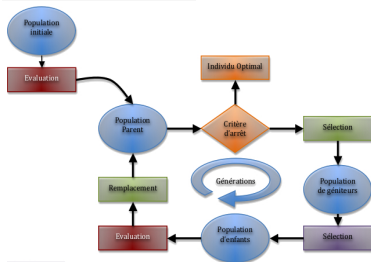
- Genotype: each gene codes for a body and the following joint
- The value of the gene accounts for the type of joint and the length of the associated body

Optimization criteria

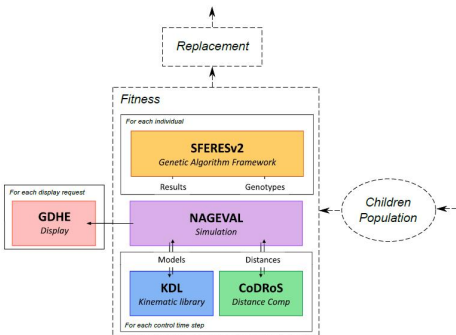
- ↘ complexity of the solution (nb. of DoFs, total length...)
- ↗ tasks performances
- ↗ ergonomy

General principle of genetic algorithms

- Initial candidate robots population
- Evaluation of the population
- Selection of the individuals of the next generation: performance and diversity
- Genetic operations (mutation, crossing,...) on the population
- Update of the population



Automatic design of dedicated robots

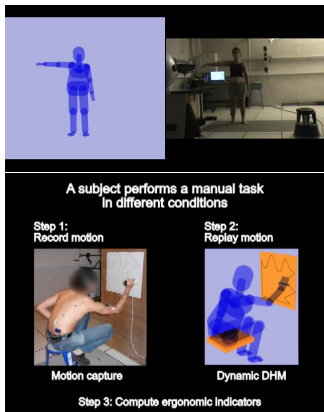


Telemach example

- NSGA2 algorithm in the *Sferes framework* (by S. Doncieux et JB. Mouret, ISIR)
- Kinematic evaluation of robots
- Trajectory tracking using of “typical” trajectories using the CCC

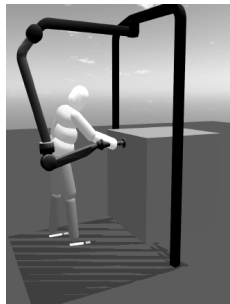
Automatic design of dedicated robots

Objective: Definition of quantitative ergonomics and performance indicators for the design of a cobot

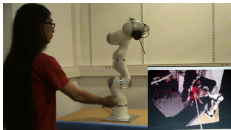


Approach

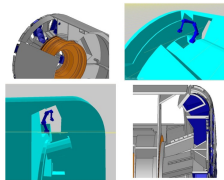
- Mocap and indicators synthesis
- Cobots architectures are evaluated through simulation



Scope of the presentation



Multi-tasks control under constraints



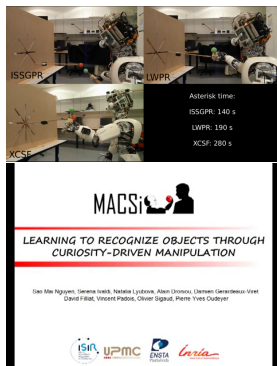
A few words about the automatic design of dedicated robotics architecture



A few words about learning and adaptation

A few words about learning and adaptation

Objective: learning and adaptation of the models and optimal policies required for the control of robots



Problem

- Infinity of possibilities of the world
- Need of adaptation in unknown contexts
- Computationally efficient and incremental acquisition of optimal motor abilities

Some on-going work

- Visuo-motor Jacobian incremental learning
- Socially guided exploration for learning in high dimension spaces
- Bridging the gap between optimal control and policy learning through reinforcement learning
- Learning globally stable controllers based on demonstrated trajectories
- Context and objective based MPC parameters tuning through evolutionary exploration (with C. Santos, S. Doncieux and José Pontes)

Conclusions

- Manipulation in constrained environment is a complex problem !
- It requires scientific contributions in several domains
 - ▶ design
 - ▶ control
 - ▶ learning and adaptation
 - ▶ ...
- From a control perspective, advanced model-based control algorithms exist in order to
 - ▶ combine tasks
 - ▶ under constraints
 - ▶ in an optimal fashion (MPC)
- ... but they require
 - ▶ realistic models of the robot and its environment: identification, learning and adaptation
 - ▶ perception capabilities

Collaborations (academia and industry)



Chaire de Robotique d'Intervention RTE/UPMC - 2011–2016




Projet ANR TELEMACH - 2008–2010



Projet FUI ROMEO2 - 2013–2017




Projet TELEMACH, ROMEO2, Simulateur physique  @CEA



Projet ANR Equipex **robotex**
Réseau "Robotique Humanoïde et Interactions Naturelles"



Projet PESSOA – 2014
avec l'Université de Minho 



Projet européen CODYCO - STREP FP7-ICT-2011.2.1 - 2013–2017
porté par l'Institut Italien de Technologie + University of Birmingham + TU Darmstadt + Institut Jožef Stefan + UPMC



Projet ANR MACSi - 2010–2013
porté par l'UPMC + INRIA Bordeaux + ENSTA + Gostai

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Thank you for your attention.
Questions ?