

IN PARTNERSHIP WITH: CNRS

Ecole Polytechnique

Activity Report 2017

Project-Team COMMANDS

Control, Optimization, Models, Methods and Applications for Nonlinear Dynamical Systems

IN COLLABORATION WITH: Centre de Mathématiques Appliquées (CMAP), Unité de Mathématiques Appliquées (UMA - ENSTA)

RESEARCH CENTER **Saclay - Île-de-France**

THEME Optimization and control of dynamic systems

Table of contents

1.	Personnel	. 1
2.	Overall Objectives	. 1
	2.1. Scientific directions	1
	2.2. Industrial impact	2
3.	Research Program	. 2
	3.1. Historical aspects	2
	3.2. Trajectory optimization	2
	3.3. Hamilton-Jacobi-Bellman approach	3
4.	Application Domains	. 3
	4.1. Fuel saving by optimizing airplanes trajectories	3
	4.2. Hybrid vehicles	3
	4.3. Biological systems	3
5.	Highlights of the Year	. 4
6.	New Software and Platforms	. 4
	6.1. BOCOP	4
	6.2. Bocop HJB	4
	6.3. Bocop Avion	4
	6.4. Bocop HJB Avion	5
7.	New Results	. 5
	7.1. Deterministic Optimal Control	5
	7.1.1. Galerkin approximations of nonlinear optimal control problems in Hilbert spaces	5
	7.1.2. Galerkin approximations for the optimal control of nonlinear delay differential equations	5
	7.2. Stochastic Control	6
	7.2.1. On the time discretization of stochastic optimal control problems: the dynamic program-	-
	ming approach	6
	7.2.2. Variational analysis for options with stochastic volatility and multiple factors	6
	7.2.3. Infinite Horizon Stochastic Optimal Control Problems with Running Maximum Cost	6
	7.3. Applications	6
8.	Bilateral Contracts and Grants with Industry	. 7
	8.1.1. Safety Line	7
	8.1.2. IFPEN	7
9.	Partnerships and Cooperations	10
	9.1. Regional Initiatives	10
	9.2. National Initiatives	10
	9.2.1.1. Cosy	10
	9.2.1.2. Algae in Silico	10
	9.3. International Research Visitors	10
10.	Dissemination	11
	10.1. Promoting Scientific Activities	11
	10.1.1. Scientific Events Selection	11
	10.1.2. Journal	11
	10.1.2.1. Member of the Editorial Boards	11
	10.1.2.2. Reviewer - Reviewing Activities	11
	10.1.3. Invited Talks	11
	10.1.4. Leadership within the Scientific Community	11
	10.2. Teaching - Supervision - Juries	11
	10.2.1. Teaching	11
	10.2.2. Supervision	11
	10.3. Popularization	11

11.	Bibliography		1	2
-----	--------------	--	---	---

Project-Team COMMANDS

Creation of the Project-Team: 2009 January 01

Keywords:

Computer Science and Digital Science:

A6.2.1. - Numerical analysis of PDE and ODE

A6.2.6. - Optimization

A6.2.7. - High performance computing

A6.3.2. - Data assimilation

A6.4.1. - Deterministic control

A6.4.2. - Stochastic control

Other Research Topics and Application Domains:

B4.4.1. - Smart grids

B7.1.2. - Road traffic

- B7.1.3. Air traffic
- B7.2.1. Smart vehicles

1. Personnel

Research Scientists

Joseph Frederic Bonnans [Team leader, Inria, Senior Researcher, HDR] Axel Kroner [Inria, Starting Research Position, until Mar 2017] Pierre Martinon [Inria, Researcher]

PhD Students

Arthur Le Rhun [Ifpen] Cedric Rommel [Safety Line]

Technical staff

Jinyan Liu [Inria]

Intern

Joao Miguel Machado [Inria, from Sep 2017]

Administrative Assistants

Hanadi Dib [Inria] Jessica Gameiro [Inria]

Visiting Scientist

Justina Gianatti [U. Rosario, from Nov 2017]

External Collaborator

Axel Kroner [U. Humboldt, from Apr 2017]

2. Overall Objectives

2.1. Scientific directions

Commands is a team devoted to dynamic optimization, both for deterministic and stochastic systems. This includes the following approaches: trajectory optimization, deterministic and stochastic optimal control, stochastic programming, dynamic programming and Hamilton-Jacobi-Bellman equation.

Our aim is to derive new and powerful algorithms for solving numerically these problems, with applications in several industrial fields. While the numerical aspects are the core of our approach it happens that the study of convergence of these algorithms and the verification of their well-posedness and accuracy raises interesting and difficult theoretical questions, such as, for trajectory optimization: qualification conditions and second-order optimality condition, well-posedness of the shooting algorithm, estimates for discretization errors; for the Hamilton-Jacobi-Bellman approach: accuracy estimates, strong uniqueness principles when state constraints are present, for stochastic programming problems: sensitivity analysis.

2.2. Industrial impact

For many years the team members have been deeply involved in various industrial applications, often in the framework of PhD theses. The Commands team itself has dealt since its foundation in 2009 with several types of applications:

- Space vehicle trajectories, in collaboration with CNES, the French space agency.
- Aeronautics, in collaboration with the startup Safety Line.
- Production, management, storage and trading of energy resources, in collaboration with EDF, GDF and TOTAL.
- Energy management for hybrid vehicles, in collaboration with Renault and IFPEN.

We give more details in the Bilateral contracts section.

3. Research Program

3.1. Historical aspects

The roots of deterministic optimal control are the "classical" theory of the calculus of variations, illustrated by the work of Newton, Bernoulli, Euler, and Lagrange (whose famous multipliers were introduced in [32]), with improvements due to the "Chicago school", Bliss [24] during the first part of the 20th century, and by the notion of relaxed problem and generalized solution (Young [37]).

Trajectory optimization really started with the spectacular achievement done by Pontryagin's group [36] during the fifties, by stating, for general optimal control problems, nonlocal optimality conditions generalizing those of Weierstrass. This motivated the application to many industrial problems (see the classical books by Bryson and Ho [28], Leitmann [34], Lee and Markus [33], Ioffe and Tihomirov [31]).

Dynamic programming was introduced and systematically studied by R. Bellman during the fifties. The HJB equation, whose solution is the value function of the (parameterized) optimal control problem, is a variant of the classical Hamilton-Jacobi equation of mechanics for the case of dynamics parameterized by a control variable. It may be viewed as a differential form of the dynamic programming principle. This nonlinear first-order PDE appears to be well-posed in the framework of *viscosity solutions* introduced by Crandall and Lions [29]. The theoretical contributions in this direction did not cease growing, see the books by Barles [22] and Bardi and Capuzzo-Dolcetta [21].

3.2. Trajectory optimization

The so-called *direct methods* consist in an optimization of the trajectory, after having discretized time, by a nonlinear programming solver that possibly takes into account the dynamic structure. So the two main problems are the choice of the discretization and the nonlinear programming algorithm. A third problem is the possibility of refinement of the discretization once after solving on a coarser grid.

In the *full discretization approach*, general Runge-Kutta schemes with different values of control for each inner step are used. This allows to obtain and control high orders of precision, see Hager [30], Bonnans [25]. In the *indirect* approach, the control is eliminated thanks to Pontryagin's maximum principle. One has then to solve the two-points boundary value problem (with differential variables state and costate) by a single or multiple shooting method. The questions are here the choice of a discretization scheme for the integration of the boundary value problem, of a (possibly globalized) Newton type algorithm for solving the resulting finite dimensional problem in IR^n (*n* is the number of state variables), and a methodology for finding an initial point.

3.3. Hamilton-Jacobi-Bellman approach

This approach consists in calculating the value function associated with the optimal control problem, and then synthesizing the feedback control and the optimal trajectory using Pontryagin's principle. The method has the great particular advantage of reaching directly the global optimum, which can be very interesting when the problem is not convex.

Optimal stochastic control problems occur when the dynamical system is uncertain. A decision typically has to be taken at each time, while realizations of future events are unknown (but some information is given on their distribution of probabilities). In particular, problems of economic nature deal with large uncertainties (on prices, production and demand). Specific examples are the portfolio selection problems in a market with risky and non-risky assets, super-replication with uncertain volatility, management of power resources (dams, gas). Air traffic control is another example of such problems.

For solving stochastic control problems, we studied the so-called Generalized Finite Differences (GFD), that allow to choose at any node, the stencil approximating the diffusion matrix up to a certain threshold [27]. Determining the stencil and the associated coefficients boils down to a quadratic program to be solved at each point of the grid, and for each control. This is definitely expensive, with the exception of special structures where the coefficients can be computed at low cost. For two dimensional systems, we designed a (very) fast algorithm for computing the coefficients of the GFD scheme, based on the Stern-Brocot tree [26].

4. Application Domains

4.1. Fuel saving by optimizing airplanes trajectories

We have a collaboration with the startup Safety Line on the optimization of trajectories for civil aircrafts. Key points include the reliable identification of the plane parameters (aerodynamic and thrust models) using data from the flight recorders, and the robust trajectory optimization of the climbing and cruise phases. We use both local (quasi-Newton interior-point algorithms) and global optimization tools (dynamic programming). The local method for the climb phase is in production and has been used for several hundreds of actual plane flights.

4.2. Hybrid vehicles

We have a collaboration with IFPEN on the energy management for hybrid vehicles. A significant direction is the analysis and classification of traffic data. More specifically, we focus on the traffic probability distribution in the (speed,torque) plane, with a time / space subdivision (road segments and timeframes).

4.3. Biological systems

We renewed in 2017 our interest in (micro)biological systems, joining projects Cosy and Algae in silico on the topic of the optimization of micro-organisms populations.

5. Highlights of the Year

5.1. Suboptimal feedback control of PDEs

In [13], J. Garcke (SCAI-Fraunhofer I.) and A. Kröner were able to solve finite time horizon suboptimal feedback control problems for partial differential equations is proposed by solving dynamic programming equations on adaptive sparse grids. The approach is illustrated for the wave equation and an extension to equations of Schrödinger type is discussed. A semi-discrete optimal control problem is introduced and the feedback control is derived from the corresponding value function. A semi-Lagrangian scheme is combined with spatially adaptive sparse grids. An adaptive grid refinement procedure is explored. We present several numerical examples studying the effect the parameters characterizing the sparse grid have on the accuracy of the value function and the optimal trajectory. Problems with dimensions up to eight were solved.

6. New Software and Platforms

6.1. BOCOP

Boite à Outils pour le Contrôle OPtimal

KEYWORDS: Dynamic Optimization - Identification - Biology - Numerical optimization - Energy management - Transportation

FUNCTIONAL DESCRIPTION: Bocop is an open-source toolbox for solving optimal control problems, with collaborations with industrial and academic partners. Optimal control (optimization of dynamical systems governed by differential equations) has numerous applications in transportation, energy, process optimization, energy and biology. Bocop includes a module for parameter identification and a graphical interface, and runs under Linux / Windows / Mac.

RELEASE FUNCTIONAL DESCRIPTION: Handling of delay systems Alternate automatic differentiation tool: CppAD Update for CMake and MinGW (windows version)

- Participants: Benjamin Heymann, Virgile Andreani, Jinyan Liu, Joseph Frédéric Bonnans and Pierre Martinon
- Contact: Pierre Martinon
- URL: http://bocop.org

6.2. Bocop HJB

KEYWORDS: Optimal control - Stochastic optimization - Global optimization FUNCTIONAL DESCRIPTION: Toolbox for stochastic or deterministic optimal control, dynamic programming / HJB approach.

RELEASE FUNCTIONAL DESCRIPTION: User interface State jumps for switched systems Explicit handling of final conditions Computation of state probability density (fiste step to mean field games)

- Participants: Benjamin Heymann, Jinyan Liu, Joseph Frédéric Bonnans and Pierre Martinon
- Contact: Joseph Frédéric Bonnans
- URL: http://bocop.org

6.3. Bocop Avion

KEYWORDS: Optimization - Aeronautics

FUNCTIONAL DESCRIPTION: Optimize the climb speeds and associated fuel consumption for the flight planning of civil airplanes.

NEWS OF THE YEAR: Improved atmosphere model 2D interpolations for temperature and wind data

- Participants: Gregorutti Baptiste, Cindie Andrieu, Anamaria Lupu, Joseph Frédéric Bonnans, Karim Tekkal, Pierre Jouniaux and Pierre Martinon
- Partner: Safety Line
- Contact: Pierre Martinon
- URL: http://www.safety-line.fr

6.4. Bocop HJB Avion

KEYWORDS: Optimization - Aeronautics

FUNCTIONAL DESCRIPTION: Optimize the climb and cruising trajectory of flight by a HJB approach. NEWS OF THE YEAR: First demonstrator for cruise flight deployed at Safety Line

- Participants: Pierre Martinon, Joseph Frédéric Bonnans, Jinyan Liu, Gregorutti Baptiste and Anamaria Lupu
- Partner: Safety Line
- Contact: Pierre Martinon
- URL: http://www.safety-line.fr

7. New Results

7.1. Deterministic Optimal Control

7.1.1. Galerkin approximations of nonlinear optimal control problems in Hilbert spaces

Participant: Axel Kroner.

With Mickaël D. Chekroun (UCLA), and Honghu Liu (Virginia Tech). Nonlinear optimal control problems in Hilbert spaces are considered for which we derive approximation theorems for Galerkin approximations. Approximation theorems are available in the literature. The originality of our approach relies on the identification of a set of natural assumptions that allows us to deal with a broad class of nonlinear evolution equations and cost functionals for which we derive convergence of the value functions associated with the optimal control problem of the Galerkin approximations. This convergence result holds for a broad class of nonlinear control strategies as well. In particular, we show that the framework applies to the optimal control of semilinear heat equations posed on a general compact manifold without boundary. The framework is then shown to apply to geoengineering and mitigation of greenhouse gas emissions formulated for the first time in terms of optimal control of energy balance climate models posed on the sphere S^2 . See [12].

7.1.2. Galerkin approximations for the optimal control of nonlinear delay differential equations

Participant: Axel Kroner.

With Mickaël D. Chekroun (UCLA), and Honghu Liu (Virginia Tech).

Optimal control problems of nonlinear delay differential equations (DDEs) are considered for which we propose a general Galerkin approximation scheme built from Koornwinder polynomials. Error estimates for the resulting Galerkin-Koornwinder approximations to the optimal control and the value function, are derived for a broad class of cost function-als and nonlinear DDEs. The approach is illustrated on a delayed logistic equation set not far away from its Hopf bifurcation point in the parameter space. In this case, we show that low-dimensional controls for a standard quadratic cost functional can be efficiently computed from Galerkin-Koornwinder approximations to reduce at a nearly optimal cost the oscillation amplitude displayed by the DDE's solution. Optimal controls computed from the Pontryagin's maximum principle (PMP) and the Hamilton-Jacobi-Bellman equation (HJB) associated with the corresponding ODE systems, are shown to provide numerical solutions in good agreement. It is finally argued that the value function computed from the corresponding reduced HJB equation provides a good approximation of that obtained from the full HJB equation. See [16].

7.2. Stochastic Control

7.2.1. On the time discretization of stochastic optimal control problems: the dynamic

programming approach

Participant: Frederic Bonnans.

With Justina Gianatti (U. Rosario) and Francisco J. Silva (U. Limoges) In this work we consider the time discretization of stochastic optimal control problems. Under general assumptions on the data, we prove the convergence of the value functions associated with the discrete time problems to the value function of the original problem. Moreover, we prove that any sequence of optimal solutions of discrete problems is minimizing for the continuous one. As a consequence of the Dynamic Programming Principle for the discrete problems, the minimizing sequence can be taken in discrete time feedback form. See [17].

7.2.2. Variational analysis for options with stochastic volatility and multiple factors

Participants: Frederic Bonnans, Axel Kroner.

We perform a variational analysis for a class of European or American options with stochastic volatility models, including those of Heston and Achdou-Tchou. Taking into account partial correlations and the presence of multiple factors, we obtain the well-posedness of the related partial differential equations, in some weighted Sobolev spaces. This involves a generalization of the commutator analysis introduced by Achdou and Tchou. See [18].

7.2.3. Infinite Horizon Stochastic Optimal Control Problems with Running Maximum Cost Participant: Axel Kroner.

With Athena Picarelli (U. Oxford) and Hasna Zidani (ENSTA).

An infinite horizon stochastic optimal control problem with running maximum cost is considered. The value function is characterized as the viscosity solution of a second-order HJB equation with mixed boundary condition. A general numerical scheme is proposed and convergence is established under the assumptions of consistency, monotonicity and stability of the scheme. A convergent semi-Lagrangian scheme is presented in detail. See [19].

7.3. Applications

7.3.1. On the Design of Optimal Health Insurance Contracts under Ex Post Moral Hazard Participant: Pierre Martinon.

With Pierre Picard and Anasuya Raj (Ecole Polytechnique, Econ. dpt).

We analyze in [20] the design of optimal medical insurance under ex post moral hazard, i.e., when illness severity cannot be observed by insurers and policyholders decide on their health expenditures. We characterize the trade-off between ex ante risk sharing and ex post incentive compatibility, in an optimal revelation mechanism under hidden information and risk aversion.

We establish that the optimal contract provides partial insurance at the margin, with a deductible when insurers rates are affected by a positive loading, and that it may also include an upper limit on coverage. We show that the potential to audit the health state leads to an upper limit on out-of-pocket expenses. Numerical simulations indicate that these qualitative results tend to be robust with respect to the health parameter.

8. Bilateral Contracts and Grants with Industry

8.1. Bilateral Contracts with Industry

8.1.1. Safety Line

In the framework of an Ilab with startup Safety Line (http://www.safety-line.fr), we design tools for the optimization of fuel consumption for civil planes. A first part is devoted to the identification of the aerodynamic and thrust characteristics of the plane, using recorded data from hundreds of flights. As an illustration, Fig. 1 shows the drag and lift coefficients for a Boeing 737, as functions of Mach and angle of attack. Latest results have been presented by Cedric Rommel at [15].

A second part is optimizing the fuel consumption during the climb and cruise phases. Fig. 2 shows a simulated climb phase, along with recorded data from the actual flight. This collaboration relies significantly on the toolboxes BOCOP and BOCOPHJB developed by Commands since 2010. The resulting commercial tool OptiClimb is currently under testing in several airplane companies, totalling about a hundred actual optimized flights per day. Recent improvements include better atmosphere models and more accurate data for temperature and wind, as well as a first demonstrator for cruise flight optimization, see Fig. 3.



Figure 1. Lift and drag aerodynamic forces for a Boeing 737.

8.1.2. IFPEN

This study is presently conducted in the framework of the PhD of Arthur Le Rhun, started in Fall 2016. The main axis is to design a traffic model suitable for optimizing the fuel consumption of a hybrid vehicle following a given route. The first step was to develop a new traffic model in which the consumption is infered only on the functionning points in the (speed,torque) plane. More precisely, we are interested in the probability distribution of these functionning points when considering a space/time subdivision into road segments and timeframes (see Fig.4). In order to reduce the huge number of distributions obtained, we perform a clustering step using k-means (Fig.5). Since the objects to be clustered are distributions, we choose to use the Wasserstein distance based on optimal transport. The task of computing these Wasserstein barycenters was done by Sinkhorn iterations, and we also developed a variant of stochastic gradient that scales better for huge data sets.



Figure 2. Simulated climb phase vs actual flight data



Figure 3. Simulated cruise flight (altitude in black, mach speed in red, wind speed in background)



Figure 4. Distributions for all timeframes for a given road segment



Figure 5. Barycenters after clustering (k=3)

In order to obtain the data for our traffic analysis, we work with a traffic simulator called SUMO, with the LUST scenario modeling the city of Luxembourg (http://sumo.dlr.de).

9. Partnerships and Cooperations

9.1. Regional Initiatives

• Gaspard Monge Program for Optimization and Operational Research (Fondation Jacques Hadamard)

Title	:	Optimal control of partial differential equations using parameterizing manifolds,
		model reduction, and dynamic programming,
Funding	:	10,000 Euro (for 2016-17), 7,000 Euro (for 2017-2018)
PI	:	Axel Kröner, U. Humboldt and Inria
Period	:	2015 - 2018
Members	:	Frédéric Bonnans (Inria Saclay and CMAP, École Polytechnique),
		Mickaël Chekroun (UCLA, Los Angeles), Martin Gubisch (U. of Konstanz),
		Honghu Liu (Virginia Tech),
		Karl Kunisch (University of Graz), Hasnaa Zidani (ENSTA ParisTech).

9.2. National Initiatives

9.2.1. IPL

9.2.1.1. Cosy

Inria Project Lab COSY (started in 2017) aims at exploiting the potential of state-of-art biological modelling, control techniques, synthetic biology and experimental equipment to achieve a paradigm shift in control of microbial communities. More precisely, we plan to determine and implement control strategies to make heterogeneous communities diversify and interact in the most profitable manner. Study of yeast cells has started in collaboration with team Lifeware (G. Batt) in the framework of the PhD of V. Andreani.

9.2.1.2. Algae in Silico

Inria Project Lab ALGAE IN SILICO (started in 2014) is dedicated to provide an integrated platform for numerical simulation of microalgae "from genes to industrial process". The project has now reached a stage where we can tackle the optimization aspects. Commands is currently joining the IPL, in the following of our previous collaborations with teams Modemic and Biocore on bioreactors, see [35], [23]

9.3. International Research Visitors

9.3.1. Internships

Joao Miguel Machado, from FGV (Rio de Janeiro), spent his master internship in our team from sept-dec 2017, working with F. Bonnans and M.S. Aronna (EMAP-FGV) on the second order necessary and sufficient optimality conditions for optimal control problems of ODEs with broken extremals, i.e., with discontinuous control. We are currently extending the classical theory to the case of a jump between interior and boundary values for the control.

10. Dissemination

10.1. Promoting Scientific Activities

10.1.1. Scientific Events Selection

- 10.1.1.1. Member of the Conference Program Committees
 - F. Bonnans: PGMO Days 2017.

10.1.2. Journal

10.1.2.1. Member of the Editorial Boards

• F. Bonnans: Associate Editor of "Applied Mathematics and Optimization" and of "Series on Mathematics and its Applications, Annals of The Academy of Romanian Scientists".

10.1.2.2. Reviewer - Reviewing Activities

Reviews in 2017 for major journals in the field: Applied Mathematics and Optimization, Automatica, Int. J. of Control, Inverse problems, J. Convex Analysis, J. Diff. Equations, J. of Optimization Theory and Applications, Optimization Set Valued and Variational Analysis, SIAM J. Optimization, SIAM J. Control and Optimization, several conference proceedings.

10.1.3. Invited Talks

• F. Bonnans: Forecasting and risk management for renewable energy, June 7-9, Paris; Numoc, June 19-23, Roma; NHOC2017, July 3-5, Porto; Optimal Control of Partial Differential Equations, Sept, Castro Urdiales.

10.1.4. Leadership within the Scientific Community

- F. Bonnans: French representative to the IFIP-TC7 committee (International Federation of Information Processing; TC7 devoted to System Modeling and Optimization).
- F. Bonnans: member of the PGMO board and Steering Committee (Gaspard Monge Program for Optimization and Operations Research, EDF-FMJH).

10.2. Teaching - Supervision - Juries

10.2.1. Teaching

Master :

F. Bonnans: *Numerical analysis of partial differential equations arising in finance and stochastic control*, 24h, M2, Ecole Polytechnique and U. Paris 6, France.

F. Bonnans: *Optimal control*, 15h, M2, Optimization master (U. Paris-Saclay) and Ensta, France. F. Bonnans: *Stochastic optimization*, 15h, M2, Optimization master (U. Paris-Saclay), France.

The Bolinais Stochastic Optimization, 191, 192, Optimization master (U. Faris-Saciay), France.

A. Kröner : Optimal control of partial differential equations, 20h, M2, Optimization master (U. Paris-Saclay), France.

10.2.2. Supervision

PhD in progress : Cédric Rommel, Data exploration for the optimization of aircraft trajectories. Started November 2015 (CIFRE fellowship with Safety Line), F. Bonnans and P. Martinon.

PhD in progress : Arthur Le Rhun, Optimal and robust control of hybrid vehicles. Started September 2016 (IFPEN fellowship), F. Bonnans and P. Martinon.

10.3. Popularization

The collaboration with startup Safety Line was presented at events "Vivatech" (17/06/2017, https:// vivatechnology.com/) and "Rencontre Inria Industrie" (17/10/2017) in Paris.

11. Bibliography

Major publications by the team in recent years

- [1] A. AFTALION, J. F. BONNANS. Optimization of running strategies based on anaerobic energy and variations of velocity, in "SIAM Journal of Applied Mathematics", October 2014, vol. 74, n^o 5, pp. 1615-1636, https:// hal.inria.fr/hal-00851182
- [2] M. S. ARONNA, J. F. BONNANS, B. S. GOH. Second order analysis of state-constrained control-affine problems, in "Mathematical Programming, Series A", January 2016, vol. 160, n^o 1, pp. 115-147, https:// hal.inria.fr/hal-01081111
- [3] M. S. ARONNA, J. F. BONNANS, A. KRÖNER. Optimal Control of Infinite Dimensional Bilinear Systems: Application to the Heat and Wave Equations, in "Mathematical Programming B", November 2016, 32 p., https://hal.inria.fr/hal-01273496
- [4] M. S. ARONNA, J. F. BONNANS, P. MARTINON. A Shooting Algorithm for Optimal Control Problems with Singular Arcs, in "Journal of Optimization Theory and Applications", August 2013, vol. 158, n^o 2, pp. 419-459 [DOI: 10.1007/s10957-012-0254-8], https://hal.inria.fr/inria-00631332
- [5] J. F. BONNANS, X. DUPUIS, L. PFEIFFER. Second-order necessary conditions in Pontryagin form for optimal control problems, in "SIAM Journal on Control and Optimization", 2014, vol. 52, n^o 6, pp. 3887-3916 [DOI: 10.1137/130923452], https://hal.inria.fr/hal-00825273
- [6] J. F. BONNANS, X. TAN. A model-free no-arbitrage price bound for variance options, in "Applied Mathematics and Optimization", July 2013, vol. 68, n^o 1, pp. 43-73 [DOI : 10.1007/s00245-013-9197-1], https://hal. inria.fr/inria-00634387
- [7] B. BONNARD, M. CLAEYS, O. COTS, P. MARTINON. Geometric and numerical methods in the contrast imaging problem in nuclear magnetic resonance, in "Acta Applicandae Mathematicae", February 2015, vol. 135, n^o 1, pp. 5-45 [DOI: 10.1007/s10440-014-9947-3], https://hal.inria.fr/hal-00867753
- [8] L. GIRALDI, P. MARTINON, M. ZOPPELLO. Optimal Design for Purcell Three-link Swimmer, in "Physical Review", February 2015, vol. 91, n^o 2, 023012, https://hal.archives-ouvertes.fr/hal-01098501
- [9] A. KRÖNER, K. KUNISCH, B. VEXLER. Semismooth Newton methods for optimal control of the wave equation with control constraints, in "SIAM Journal on Control and Optimization", 2011, vol. 49, n^o 2, pp. 830–858, http://dx.doi.org/10.1137/090766541
- [10] A. KRÖNER, K. KUNISCH, H. ZIDANI. Optimal feedback control of undamped wave equations by solving a HJB equation, in "ESAIM: Control, Optimisation and Calculus of Variations", 2014, vol. 21, n^o 2, pp. 442 -464 [DOI: 10.1051/COCV/2014033], https://hal.archives-ouvertes.fr/hal-00924089

Publications of the year

Articles in International Peer-Reviewed Journals

- [11] J. F. BONNANS, A. FESTA. Error estimates for the Euler discretization of an optimal control problem with first-order state constraints, in "SIAM Journal on Numerical Analysis", December 2017, vol. 55, n^o 2, pp. 445–471, https://hal.inria.fr/hal-01093229
- [12] M. D. CHEKROUN, A. KRÖNER, H. LIU. Galerkin approximations of nonlinear optimal control problems in Hilbert spaces, in "Electronic Journal of Differential Equations", 2017, vol. 2017, n^o 189, pp. 1-40, https://hal.archives-ouvertes.fr/hal-01501178
- [13] J. GARCKE, A. KRÖNER. Suboptimal feedback control of PDEs by solving HJB equations on adaptive sparse grids, in "Journal of Scientific Computing", January 2017, vol. 70, n^O 1, pp. 1-28, https://hal.archives-ouvertes. fr/hal-01185912
- [14] B. HEYMANN, J. F. BONNANS, P. MARTINON, F. SILVA, F. LANAS, G. JIMENEZ. Continuous Optimal Control Approaches to Microgrid Energy Management, in "Energy Systems", 2017, https://hal.inria.fr/hal-01129393

International Conferences with Proceedings

[15] J. F. BONNANS, B. GREGORUTTI, P. MARTINON, C. ROMMEL. Aircraft Dynamics Identification for Optimal Control, in "EUCASS 2017", Milan, Italy, R. MARTINEZ-VAL, C. BONNAL, D. KNIGHT (editors), July 2017 [DOI: 10.13009/EUCASS2017-179], https://hal.inria.fr/hal-01639731

Research Reports

[16] M. D. CHEKROUN, A. KRÖNER, H. LIU. Galerkin approximations for the optimal control of nonlinear delay differential equations, Inria, 2017, https://hal.archives-ouvertes.fr/hal-01534673

Other Publications

- [17] J. F. BONNANS, J. GIANATTI, F. J. SILVA. On the time discretization of stochastic optimal control problems: the dynamic programming approach, February 2017, working paper or preprint, https://hal.inria. fr/hal-01474285
- [18] J. F. BONNANS, A. KRÖNER. Variational analysis for options with stochastic volatility and multiple factors, April 2017, working paper or preprint, https://hal.inria.fr/hal-01516011
- [19] A. KRÖNER, A. PICARELLI, H. ZIDANI. Infinite Horizon Stochastic Optimal Control Problems with Running Maximum Cost, October 2017, working paper or preprint, https://hal.archives-ouvertes.fr/hal-01585766
- [20] P. MARTINON, P. PICARD, A. RAJ. On the Design of Optimal Health Insurance Contracts under Ex Post Moral Hazard, 2017, working paper or preprint, https://hal-polytechnique.archives-ouvertes.fr/hal-01348551

References in notes

- [21] M. BARDI, I. CAPUZZO-DOLCETTA. Optimal control and viscosity solutions of Hamilton-Jacobi-Bellman equations, Systems and Control: Foundations and Applications, Birkhäuser, Boston, 1997
- [22] G. BARLES. Solutions de viscosité des équations de Hamilton-Jacobi, Mathématiques et Applications, Springer, Paris, 1994, vol. 17

- [23] T. BAYEN, F. MAIRET, P. MARTINON, M. SEBBAH. Analysis of a periodic optimal control problem connected to microalgae anaerobic digestion, in "Optimal Control Applications and Methods", 2014 [DOI: 10.1002/OCA.2127], https://hal.archives-ouvertes.fr/hal-00860570
- [24] G. BLISS. Lectures on the Calculus of Variations, University of Chicago Press, Chicago, Illinois, 1946
- [25] J. F. BONNANS, J. LAURENT-VARIN. Computation of order conditions for symplectic partitioned Runge-Kutta schemes with application to optimal control, in "Numerische Mathematik", 2006, vol. 103, n^o 1, pp. 1–10
- [26] J. F. BONNANS, E. OTTENWAELTER, H. ZIDANI. Numerical schemes for the two dimensional second-order HJB equation, in "ESAIM: M2AN", 2004, vol. 38, pp. 723-735
- [27] J. F. BONNANS, H. ZIDANI. Consistency of generalized finite difference schemes for the stochastic HJB equation, in "SIAM J. Numerical Analysis", 2003, vol. 41, pp. 1008-1021
- [28] A. E. BRYSON, Y.-C. HO. Applied optimal control, Hemisphere Publishing, New-York, 1975
- [29] M. CRANDALL, P. LIONS. Viscosity solutions of Hamilton Jacobi equations, in "Bull. American Mathematical Society", 1983, vol. 277, pp. 1–42
- [30] W. HAGER. Runge-Kutta methods in optimal control and the transformed adjoint system, in "Numerische Mathematik", 2000, vol. 87, n^o 2, pp. 247–282
- [31] A. IOFFE, V. TIHOMIROV. Theory of Extremal Problems, North-Holland Publishing Company, Amsterdam, 1979, Russian Edition: Nauka, Moscow, 1974
- [32] J.-L. LAGRANGE. Mécanique analytique, Paris, 1788, reprinted by J. Gabay, 1989
- [33] E. LEE, L. MARKUS. Foundations of optimal control theory, John Wiley, New York, 1967
- [34] G. LEITMANN. An introduction to optimal control, Mc Graw Hill, New York, 1966
- [35] R. MUÑOZ-TAMAYO, P. MARTINON, G. BOUGARAN, F. MAIRET, O. BERNARD. Getting the most out of it: Optimal experiments for parameter estimation of microalgae growth models, in "Journal of Process Control", 2014, vol. 24, n^o 6, pp. 991 - 1001 [DOI : 10.1016/J.JPROCONT.2014.04.021], http://www.sciencedirect. com/science/article/pii/S095915241400122X
- [36] L. PONTRYAGIN, V. BOLTYANSKI, R. GAMKRELIDZE, E. MICHTCHENKO. *The Mathematical Theory of Optimal Processes*, Wiley Interscience, New York, 1962
- [37] L. YOUNG. Lectures on the calculus of variations and optimal control theory, W. B. Saunders Co., Philadelphia, 1969