

Primal-dual interior point methods for state and input constrained optimal control

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Overview

- 1 Introduction and motivations
- 2 Problem presentation and main assumptions
- 3 First order optimality conditions and main result
- 4 Preliminary results
- 5 Interiority Analysis
- 6 Convergence analysis
- 7 Primal dual Algorithm
- 8 Conclusion and perspectives

Outline

- 1 Introduction and motivations
- 2 Problem presentation and main assumptions
 - Objective of the paper
 - Problem presentation and main assumptions
- 3 First order optimality conditions and main result
 - First order optimality conditions
 - Main result
- 4 Preliminary results
- 5 Interiority Analysis
- 6 Convergence analysis
- 7 Primal dual Algorithm
 - Primal dual TPBVP
 - Algorithm presentation
 - Numerical Example : Battery management
- 8 Conclusion and perspectives

Original and penalized problem

Original problem

$$\min_{c(x) \leq 0} f(x)$$

Penalized problem

$$\min_x f(x) - \epsilon \log(-c(x))$$

Numerical optimization problem (2/2)

Primal penalized problem

The primal problem consists in solving for x the following first order conditions

$$f'(x) - c'(x) \cdot \frac{\epsilon}{c(x)} = 0$$

Primal dual penalized problem

The primal-dual problem consists in solving for x and λ the following first order conditions

$$f'(x) + c'(x) \cdot \lambda = 0$$

$$\lambda c(x) + \epsilon = 0 \Leftrightarrow \lambda - c(x) - \sqrt{\lambda^2 + c(x)^2 + 2\epsilon} = 0$$

Outline

- 1 Introduction and motivations
- 2 Problem presentation and main assumptions
 - Objective of the paper
 - Problem presentation and main assumptions
- 3 First order optimality conditions and main result
 - First order optimality conditions
 - Main result
- 4 Preliminary results
- 5 Interiority Analysis
- 6 Convergence analysis
- 7 Primal dual Algorithm
 - Primal dual TPBVP
 - Algorithm presentation
 - Numerical Example : Battery management
- 8 Conclusion and perspectives

Outline

- 1 Introduction and motivations
- 2 Problem presentation and main assumptions
 - Objective of the paper
 - Problem presentation and main assumptions
- 3 First order optimality conditions and main result
 - First order optimality conditions
 - Main result
- 4 Preliminary results
- 5 Interiority Analysis
- 6 Convergence analysis
- 7 Primal dual Algorithm
 - Primal dual TPBVP
 - Algorithm presentation
 - Numerical Example : Battery management
- 8 Conclusion and perspectives

Objective of the paper

Objective of the paper

Adapt primal-dual interior point methods from numerical optimization to pure state and input constrained optimal control.

State of the art

- M. Weiser. *Interior Point Methods in Function Space*, SIAM Journal on Control and Optimization, 2005.
- J.F. Bonnans, Th. Guilbaud. *Using logarithmic penalties in the shooting algorithm for optimal control problems*, Optimal Control Applications and Methods, 2003.
- P. Malisani, F. Chaplais, N. Petit, *An interior penalty method for optimal control problems with state and input constraints of nonlinear systems*, Optimal Control Applications and Methods, 2014.

Outline

- 1 Introduction and motivations
- 2 Problem presentation and main assumptions
 - Objective of the paper
 - Problem presentation and main assumptions
- 3 First order optimality conditions and main result
 - First order optimality conditions
 - Main result
- 4 Preliminary results
- 5 Interiority Analysis
- 6 Convergence analysis
- 7 Primal dual Algorithm
 - Primal dual TPBVP
 - Algorithm presentation
 - Numerical Example : Battery management
- 8 Conclusion and perspectives

Problem Statement

$$\min_{u \in \mathcal{U}} J(u) = \int_0^T \ell(x_u(t), u(t)) dt + \varphi(x_u(T))$$

$$\dot{x}_u(t) = f(x_u(t), u(t))$$

$$x_u(0) = x^0$$

$$\mathcal{U} := L^\infty([0, T]; U_{ad} \subset \mathbb{R}^m)$$

$$g(x_u(t)) \leq 0$$

Where $T > 0$, x^0 are fixed and x_u is the solution of the ODE with control u . And we define the classical pre-Hamiltonian function

$H : \mathbb{R}^n \times \mathbb{R}^m \times \mathbb{R}^n \mapsto \mathbb{R}$ as follows

$$H(x, u, p) := \ell(x, u) + p \cdot f(x, u)$$

Assumptions

- (A1) The functions $\ell : \mathbb{R}^n \times \mathbb{R}^m \mapsto \mathbb{R}$, $f : \mathbb{R}^n \times \mathbb{R}^m \mapsto \mathbb{R}^n$ and $g_i : \mathbb{R}^n \mapsto \mathbb{R}$ are at least twice continuously differentiable.
- (A2) The dynamics $f : \mathbb{R}^n \times \mathbb{R}^m$ from satisfies a sublinear growth property

$$\exists D < +\infty, \forall x \in \mathbb{R}^n, \forall u \in [-1, 1]^m \text{ s.t. } \|f(x, u)\| \leq D(1 + \|x\|)$$

- (A3) Interior accessibility

$$\{u \text{ s.t. } g(x_u) \leq 0\} \subseteq \text{cl}_{L^1}(\{u \text{ s.t. } g(x_u) < 0\})$$

- (A4) The OCP has a unique solution u^* and $\exists \beta \geq 0$ and $r > 0$ such that

$$J(u) - J(u^*) \geq \beta \|u - u^*\|_{L^2}^2, \quad \forall u \in B_{L^2}(u^*, r) \cap \mathcal{U}$$

Outline

- 1 Introduction and motivations
- 2 Problem presentation and main assumptions
 - Objective of the paper
 - Problem presentation and main assumptions
- 3 First order optimality conditions and main result**
 - First order optimality conditions**
 - Main result**
- 4 Preliminary results
- 5 Interiority Analysis
- 6 Convergence analysis
- 7 Primal dual Algorithm
 - Primal dual TPBVP
 - Algorithm presentation
 - Numerical Example : Battery management
- 8 Conclusion and perspectives

Outline

- 1 Introduction and motivations
- 2 Problem presentation and main assumptions
 - Objective of the paper
 - Problem presentation and main assumptions
- 3 First order optimality conditions and main result**
 - First order optimality conditions**
 - Main result
- 4 Preliminary results
- 5 Interiority Analysis
- 6 Convergence analysis
- 7 Primal dual Algorithm
 - Primal dual TPBVP
 - Algorithm presentation
 - Numerical Example : Battery management
- 8 Conclusion and perspectives

State constrained Pontryagin Maximum Principle

Any optimal solution (\bar{x}_u, \bar{u}) of the presented COCP is a Pontryagin extremal, i.e. $(\bar{x}_u, \bar{u}, \bar{p}, \bar{\mu})$ is solution of

$$\dot{\bar{x}}_u(t) = f(\bar{x}_u(t), \bar{u}(t))$$

$$-d\bar{p}(t) = H'_x(\bar{x}_u(t), \bar{u}(t), \bar{p}(t))dt + g'(\bar{x}_u(t))d\bar{\mu}(t)$$

$$\bar{x}_u(0) = x^0$$

$$\bar{p}(T) = \varphi'(\bar{x}_u(T))$$

$$H(\bar{x}_u(t), \bar{u}(t), \bar{p}(t)) = \inf_{v \in U_{ad}} H(\bar{x}_u(t), v, \bar{p}(t))$$

$$d\bar{\mu} \geq 0$$

$$\int_0^T g(\bar{x}_u(t))d\bar{\mu}(t) = 0$$

Outline

- 1 Introduction and motivations
- 2 Problem presentation and main assumptions
 - Objective of the paper
 - Problem presentation and main assumptions
- 3 First order optimality conditions and main result**
 - First order optimality conditions
 - Main result**
- 4 Preliminary results
- 5 Interiority Analysis
- 6 Convergence analysis
- 7 Primal dual Algorithm
 - Primal dual TPBVP
 - Algorithm presentation
 - Numerical Example : Battery management
- 8 Conclusion and perspectives

Log-barrier OCP

Let $\epsilon > 0$, the penalized optimal control problem writes

$$\min_{u \in \mathcal{U}} J_\epsilon(u) = \int_0^T \ell(x_u(t), u(t)) - \epsilon \log(-g(x_u(t))) dt + \varphi(x_u(T))$$

$$\dot{x}_u(t) = f(x_u(t), u(t))$$

$$x_u(0) = x^0$$

$$\mathcal{U} := L^\infty([0, T]; U_{ad})$$

and the corresponding penalized pre-Hamiltonian is

$$H[\epsilon](x, u, p) = H(x, u, p) - \epsilon \log(-g(x))$$

Main result (1/2)

Any Pontryagin extremal $(x_{u,\epsilon}, u_\epsilon, p_\epsilon)$ of the penalized problem, i.e. solution of

$$\dot{x}_{u,\epsilon}(t) = f(x_{u,\epsilon}(t), u_\epsilon(t))$$

$$\dot{p}_\epsilon(t) = -H[\epsilon]'_x(x_{u,\epsilon}(t), u_\epsilon(t), p_\epsilon(t))$$

$$x_{u,\epsilon}(0) = x^0$$

$$p_\epsilon(T) = \varphi'(x_{u,\epsilon}(T))$$

$$H[\epsilon](x_{u,\epsilon}(t), u_\epsilon(t), p_\epsilon(t)) = \inf_{v \in U_{ad}} H[\epsilon](x_{u,\epsilon}(t), v, p_\epsilon(t))$$

Main result (2/2)

converges to $(\bar{x}_u, \bar{u}, \bar{p}, \bar{\mu})$ to a Pontryagin extremal of the original problem as follows

$$\lim_{\epsilon \downarrow 0} \|u_\epsilon - \bar{u}\|_{L^2} = 0$$

$$\lim_{\epsilon \downarrow 0} \|x_{u,\epsilon} - \bar{x}_u\|_{L^\infty} = 0$$

$$\lim_{\epsilon \downarrow 0} \|p_\epsilon - \bar{p}\|_{L^1} = 0$$

$$\lim_{\epsilon \downarrow 0} \frac{-\epsilon}{g(x_{u,\epsilon})} dt \xrightarrow{*} d\bar{\mu}$$

Outline

- 1 Introduction and motivations
- 2 Problem presentation and main assumptions
 - Objective of the paper
 - Problem presentation and main assumptions
- 3 First order optimality conditions and main result
 - First order optimality conditions
 - Main result
- 4 Preliminary results**
- 5 Interiority Analysis
- 6 Convergence analysis
- 7 Primal dual Algorithm
 - Primal dual TPBVP
 - Algorithm presentation
 - Numerical Example : Battery management
- 8 Conclusion and perspectives

Preliminary Results (1/3)

State Lipschitz continuity

From the sublinear growth property, $\forall u_1, u_2 \in \mathcal{U}$, $\exists \text{const}(f) < +\infty$ such that

$$\|x_{u_1} - x_{u_2}\|_{L^\infty} \leq \text{const}(f) \|u_1 - u_2\|_{L^1}$$

State-constraint measure

For all $u \in \mathcal{U}$ and For all $E \subseteq g \circ x[u]([0, T])$ we note $m[u, g]$ the push-forward g -measure of E defined as follows

$$m[u, g](E) := \text{meas} \left((g \circ x_u)^{-1}(E) \right)$$

Proposition

For all $u \in \mathcal{U}$, let $E \subseteq g \circ x_u([0, T]) \subset \mathbb{R}$ be a Lebesgue-measurable set, the g -measure is lower bounded as follows

$$m[u, g](E) \geq \text{const}(f, g) \text{meas}(E)$$

Preliminary results (2/3)

Set of state saturated control

Let us define the set of saturated-state control \mathcal{U}_g^0 as follows

$$\mathcal{U}_g^0 := \{u \in \mathcal{U} \text{ s.t. } \sup_t g(x_u(t)) = 0\}$$

Set of near state-saturated times

For all u s.t. $g(x_u) \leq 0$ and $\forall \delta \geq 0$ we define the set of near state-saturated times, noted $S_u(\delta)$ as follows

$$S_u(\delta) := (g \circ x_u)^{-1} ([-\delta, +\infty))$$

Preliminary results (3/3)

Proposition

There exists $\Gamma_g > 0$ and $v \in \mathcal{U}$ satisfying

$$\sup_t g(x_v(t)) \leq -2\Gamma_g \quad (1)$$

Thus, for all $u \in \mathcal{U}_g^0$ this yields

$$g(x_v(t)) \leq g(x_u(t)) - \Gamma_g, \quad \forall t \in S_u(\Gamma_g) \quad (2)$$

Proof

Let $\delta > 0$

$$\Gamma_g = -\frac{1}{2} \sup_{u \in \mathcal{U}_g^0} \left\{ \inf_{v \in B_{L^1}(u, \delta) \cap \mathcal{U}} \left\{ \sup_t g(x[v](t)) \right\} \right\}$$

Outline

- 1 Introduction and motivations
- 2 Problem presentation and main assumptions
 - Objective of the paper
 - Problem presentation and main assumptions
- 3 First order optimality conditions and main result
 - First order optimality conditions
 - Main result
- 4 Preliminary results
- 5 Interiority Analysis**
- 6 Convergence analysis
- 7 Primal dual Algorithm
 - Primal dual TPBVP
 - Algorithm presentation
 - Numerical Example : Battery management
- 8 Conclusion and perspectives

State-penalized problem

$$\begin{aligned}\min_u G_\epsilon(u) &= \int_0^T \ell(x_u(t), u(t)) - \epsilon \log(-g(x_u(t))) dt \\ \dot{x}_u(t) &= f(x_u(t), u(t)) \\ x_u(0) &= x^0 \\ u &\in \mathcal{U}\end{aligned}$$

Lemma

Any optimal solution of state penalized problem u_ϵ satisfies

$$g(x_{u_\epsilon}(t)) < 0, \forall t \in [0, T]$$

and $\exists K_\psi < +\infty$ such that $\forall \epsilon \in (0, \epsilon_0)$ one has

$$\left\| \frac{\epsilon}{g(x_{u_\epsilon})} \right\|_{L^1} \leq K_\psi$$

State Interiority proof (1/6)

$$G_\epsilon(v) - G_\epsilon(u_\epsilon) = \Delta_\ell(u_\epsilon, v) + \epsilon \Delta_{\log}(u_\epsilon, v)$$

with

$$\Delta_\ell(u_\epsilon, v) = \int \ell(x_v, v) - \ell(x_{u_\epsilon}, u_\epsilon) dt$$

$$\Delta_{\log}(u_\epsilon, v) = \int -\log(-g(x_v)) + \log(-g(x_{u_\epsilon})) dt$$

State Interiority proof (2/6)

$$\begin{aligned}\Delta_\ell(u_\epsilon, v) &= \int \ell(x_v, v) - \ell(x_{u_\epsilon}, u_\epsilon) dt \\ &\leq \int \text{const}(\ell)(\|x_v(t) - x_{u_\epsilon}(t)\| + \|v(t) - u_\epsilon(t)\|) dt \\ &\leq \text{const}(\ell, f, T, \Gamma_g)\end{aligned}$$

$$\begin{aligned}\epsilon \Delta_{\log}(u_\epsilon, v) &= \epsilon \int_{S_{u_\epsilon}(\Gamma_g)} -\log(-g(x_v)) + \log(-g(x_{u_\epsilon})) dt \\ &\quad + \epsilon \int_{[0, T] \setminus S_{u_\epsilon}(\Gamma_g)} -\log(-g(x_v)) + \log(-g(x_{u_\epsilon})) dt \\ &:= \epsilon \Delta_S + \epsilon \Delta_{S^c}\end{aligned}$$

$$\epsilon \Delta_{S^c} \leq \text{const}(g, f, T, \Gamma_g, \epsilon_0, \psi)$$

$$\begin{aligned} \epsilon \Delta_S &= \epsilon \int_{S_{u_\epsilon}(\Gamma_g)} -\log(-g(x_v)) + \log(-g(x_{u_\epsilon})) dt \\ &= \epsilon \int_{S_{u_\epsilon}(\Gamma_g)} -\frac{1}{g(x[u_\epsilon])} (g(x[v]) - g(x[u_\epsilon])) dt \\ &\leq -\epsilon \Gamma_g \int_{S_{u_\epsilon}(\Gamma_g)} -\frac{1}{g(x[u_\epsilon])} dt \end{aligned}$$

State Interiority proof (4/6)

Now, let us prove interiority of u_ϵ by contradiction. Assume u_ϵ an optimal solution such that $\forall \rho \in (0, \Gamma_g)$, $(g \circ x_{u_\epsilon})^{-1}((-\Gamma_g, -\rho]) \subset [0, T]$

$$\begin{aligned} \epsilon \int_{S_{u_\epsilon}(\Gamma_g)} -\frac{1}{g(x[u_\epsilon])} dt &\geq \epsilon \int_{(g \circ x[u_\epsilon])^{-1}((-\Gamma_g, -\rho))} -\frac{1}{g(x[u_\epsilon])} dt \\ &= \epsilon \int_{-\Gamma_g}^{-\rho} -\frac{1}{s} m[u_\epsilon, g](ds) \end{aligned}$$

Using the lower bound on the state-constraint measure

$$\begin{aligned} \epsilon \int_{S_{u_\epsilon}(\Gamma_g)} -\frac{1}{g(x[u_\epsilon])} dt &\geq \epsilon \text{const}(f, g) \int_{-\Gamma_g}^{-\rho} -\frac{1}{s} ds \\ &\geq \epsilon \text{const}(f, g) (-\log(\rho) + \log(\Gamma_g)) \end{aligned}$$

For ρ small enough

$$G_\epsilon(v) - G_\epsilon(u_\epsilon) \leq \text{const}(\ell, f, g, T, \Gamma_g, \epsilon_0, \psi) + \epsilon \text{const}(f, g, \Gamma_g) \log(\rho) < 0$$

which contradicts the optimality of u_ϵ and proves interiority.

State Interiority proof (6/6)

Let us prove the L^1 -boundedness by contradiction. Assume u_ϵ optimal such that $\forall K_\psi > 0, \exists \epsilon > 0, \|\epsilon \psi' \circ g(x[u_\epsilon])\|_{L^1} \geq K_\psi$. First, we have

$$(g \circ x[u_\epsilon])^{-1}(\{0\}) = \emptyset$$

thus

$$\left\| \frac{\epsilon}{g(x[u_\epsilon])} \right\|_{L^1} := \lim_{\rho \rightarrow 0} \int_{-\infty}^{-\rho} \frac{\epsilon}{g(x[u_\epsilon])} m[u_\epsilon, g](ds) > K_\psi + \frac{\epsilon_0 T}{\Gamma_g}$$

Finally, we have:

$$G_\epsilon(v) - G_\epsilon(u_\epsilon) \leq \text{const}(\ell, f, g, T, \Gamma_g, \psi, \epsilon_0) - \Gamma_g K_\psi$$

Then $\exists K_\psi > 0$, such that $G_\epsilon(v) - G_\epsilon(u_\epsilon) < 0$, which contradicts the optimality of u_ϵ .

Outline

- 1 Introduction and motivations
- 2 Problem presentation and main assumptions
 - Objective of the paper
 - Problem presentation and main assumptions
- 3 First order optimality conditions and main result
 - First order optimality conditions
 - Main result
- 4 Preliminary results
- 5 Interiority Analysis
- 6 Convergence analysis**
- 7 Primal dual Algorithm
 - Primal dual TPBVP
 - Algorithm presentation
 - Numerical Example : Battery management
- 8 Conclusion and perspectives

Primal variables

Primal variables convergence is a well established result

$$\lim_{\epsilon \downarrow 0} \|u^* - u_\epsilon\|_{L^2} = 0$$

$$\lim_{\epsilon \downarrow 0} \|x_{u^*} - x_{u_\epsilon}\|_{L^\infty} = 0$$

Convergence of state penalties (1/2)

Convergence of state penalties

The sequence $(\frac{\epsilon_n}{g(x_{u_{\epsilon_n}})})_n$ is uniformly L^1 -bounded by K_ψ . Thus

$$\forall \phi \in L^\infty, |T_{u_{\epsilon_n}}(\phi)| \leq K_\psi \|\phi\|_{L^\infty}$$

hence $\|T_{u_{\epsilon_n}}\|_{\mathcal{M}} \leq K_\psi$. By weak * compactity of the unit ball of $\mathcal{M}([0, T])$, $\exists \mu \in \mathcal{M}([0, T])$ such that

$$\lim_{k \rightarrow +\infty} \int -\frac{\epsilon_{n_k}}{g(x_{u_{\epsilon_{n_k}}})} dt \stackrel{*}{\rightharpoonup} \int d\mu$$

Complementarity conditions

Positivity of μ

$$-\frac{\epsilon_n}{g(x_{u_{\epsilon_n}})} \geq 0 \Rightarrow d\mu \geq 0$$

Complementarity conditions

$$\begin{aligned}\langle \mu, g(x[u^*]) \rangle &= \lim_{n \rightarrow +\infty} \int -\frac{\epsilon_n}{g(x_{u_{\epsilon_n}})} g(x_{u^*}) dt \\ &= \lim_{n \rightarrow +\infty} \int -\frac{\epsilon_n}{g(x_{u_{\epsilon_n}})} g(x_{u_{\epsilon_n}}) dt \\ &= \lim_{n \rightarrow +\infty} -\epsilon_n T \\ &= 0\end{aligned}$$

Convergence of adjoint state

The optimal adjoint state satisfies the following ODE

$$-dp^* = (\ell'_x(x[u^*], u^*) - f'_x(x[u^*], u^*) \cdot p^*) dt + g'(x[u^*])d\mu$$

with $p^*(T) = 0$ and the penalized adjoint state is solution of

$$\begin{aligned} \dot{p}[u_{\epsilon_n}] = & -\ell'_x(x[u_{\epsilon_n}], u_{\epsilon_n}) - f'_x(x[u_{\epsilon_n}], u_{\epsilon_n}) \cdot p[u_{\epsilon_n}] \\ & + \epsilon g'(x[u_{\epsilon_n}](t)) \frac{1}{g(x[u_{\epsilon_n}])} \end{aligned}$$

$$p[u_{\epsilon_n}](T) = 0$$

Then $p[u_{\epsilon_n}]$ **converges pointwise** to p^* , and the boundedness of both $p[u_{\epsilon_n}]$ and p^* proves $\lim_{n \rightarrow +\infty} \|p[u_{\epsilon_n}] - p^*\|_{L^1}$

Outline

- 1 Introduction and motivations
- 2 Problem presentation and main assumptions
 - Objective of the paper
 - Problem presentation and main assumptions
- 3 First order optimality conditions and main result
 - First order optimality conditions
 - Main result
- 4 Preliminary results
- 5 Interiority Analysis
- 6 Convergence analysis
- 7 Primal dual Algorithm**
 - Primal dual TPBVP
 - Algorithm presentation
 - Numerical Example : Battery management
- 8 Conclusion and perspectives

Outline

- 1 Introduction and motivations
- 2 Problem presentation and main assumptions
 - Objective of the paper
 - Problem presentation and main assumptions
- 3 First order optimality conditions and main result
 - First order optimality conditions
 - Main result
- 4 Preliminary results
- 5 Interiority Analysis
- 6 Convergence analysis
- 7 Primal dual Algorithm**
 - Primal dual TPBVP
 - Algorithm presentation
 - Numerical Example : Battery management
- 8 Conclusion and perspectives

Primal Dual TPBVP

Any solution $(x_{u_\epsilon}^*, u_\epsilon^*, p_\epsilon^*, \lambda_\epsilon^*)$ of the following Primal Dual TPBVP

$$\dot{x}_{u_\epsilon}^* = f(x_{u_\epsilon}^*, u_\epsilon^*)$$

$$\dot{p}_\epsilon^* = -H'_x(x_{u_\epsilon}^*, u_\epsilon^*, p_\epsilon^*) - \lambda_\epsilon^* g'(x_{u_\epsilon}^*)$$

$$H(x_{u,\epsilon}, u_\epsilon^*, p_\epsilon^*) = \inf_{v \in U_{ad}} H(x_{u,\epsilon}, v, p_\epsilon^*)$$

$$0 = \lambda_\epsilon^* - g(x_{u_\epsilon}^*) - \sqrt{\lambda_\epsilon^{*2} + g(x_{u_\epsilon}^*)^2 + 2\epsilon}$$

$$x[u_\epsilon^*](0) = x^0 ; \quad p_\epsilon^*(T) = \varphi'(x_{u_\epsilon}^*(T))$$

converges to (x_{u^*}, u^*, p^*) as follows

$$\lim_{\epsilon \downarrow 0} \|u_\epsilon^* - u^*\|_{L^2} = 0 ; \quad \lim_{\epsilon \downarrow 0} \|x_{u_\epsilon}^* - x[u^*]\|_{L^\infty} = 0 ; \quad \lim_{\epsilon \downarrow 0} \|p_\epsilon^* - p^*\|_{L^1} = 0$$

$$\lim_{\epsilon \downarrow 0} \lambda_\epsilon^* dt \xrightarrow{*} d\mu$$

Outline

- 1 Introduction and motivations
- 2 Problem presentation and main assumptions
 - Objective of the paper
 - Problem presentation and main assumptions
- 3 First order optimality conditions and main result
 - First order optimality conditions
 - Main result
- 4 Preliminary results
- 5 Interiority Analysis
- 6 Convergence analysis
- 7 Primal dual Algorithm**
 - Primal dual TPBVP
 - Algorithm presentation**
 - Numerical Example : Battery management
- 8 Conclusion and perspectives

We note

$$S(\epsilon) = (x_{u_\epsilon^*}, u_\epsilon^*, p_\epsilon^*, \lambda_\epsilon^*)$$

a solution of the primal-dual TPBVP.

Primal dual Algorithm

- 1: Define $\epsilon_0 > 0$, $\alpha \in (0, 1)$, $\text{tol} = o(1)$, $k = 0$
- 2: **while** $\epsilon_k > \text{tol}$ **do**
- 3: $S(\epsilon_{k+1}) \leftarrow$ solution of primal-dual TPBVP initialized with $S(\epsilon_k)$
- 4: $\epsilon_{k+1} \leftarrow \alpha \epsilon_k$
- 5: $k \leftarrow k + 1$
- 6: **end while**
- 7: **return** $S(\epsilon_k)$

Outline

- 1 Introduction and motivations
- 2 Problem presentation and main assumptions
 - Objective of the paper
 - Problem presentation and main assumptions
- 3 First order optimality conditions and main result
 - First order optimality conditions
 - Main result
- 4 Preliminary results
- 5 Interiority Analysis
- 6 Convergence analysis
- 7 Primal dual Algorithm**
 - Primal dual TPBVP
 - Algorithm presentation
 - **Numerical Example : Battery management**
- 8 Conclusion and perspectives

Battery management problem

Optimal control problem

$$\min_u \int_0^T \text{spot}(t) (\max\{p_{\text{compteur}}(t); 0\} - \min\{p_{\text{compteur}}(t); 0\}) dt$$

under the following constraints

$$p_{\text{compteur}} := \text{cons} - PV + \frac{1}{\rho_c} \max\{u; 0\} + \rho_d \min\{u; 0\}$$

$$\dot{x}(t) = u(t)$$

$$x(0) = \frac{x^+}{2}$$

$$x(t) \in [0; x^+]$$

$$u(t) \in [u^-; \rho_c u^+]$$

$$T := 8760h$$

Comparison with discretize then optimize approach

Discretize and optimize

- Numerical scheme : Lobatto IIIa
- Solver : IPOPT+ MUMPS
- Gradients and jacobians provided and vectorized
- Computation time $\approx 9500s$

Primal dual OCP

- Numerical scheme: Lobatto IIIa
- Residual error on ODEs: 10^{-3}
- Linear solver: SPLU + umfpack
- Gradients and jacobians provided and vectorized
- Computation time $\approx 25s$

Outline

- 1 Introduction and motivations
- 2 Problem presentation and main assumptions
 - Objective of the paper
 - Problem presentation and main assumptions
- 3 First order optimality conditions and main result
 - First order optimality conditions
 - Main result
- 4 Preliminary results
- 5 Interiority Analysis
- 6 Convergence analysis
- 7 Primal dual Algorithm
 - Primal dual TPBVP
 - Algorithm presentation
 - Numerical Example : Battery management
- 8 Conclusion and perspectives

- Extend theoretical results to mixed state and input constraints and to more general input constraints
- Provide a predictor-corrector in function space type algorithm to automatically adapt the weighting parameter of IPMs.