

Sparse optimal control for a nonlinear parabolic equation with mixed control-state constraints

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Control and stabilization issues for PDE

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Chemnitz-Klaffenbach 2011



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Outline

- 1 The optimal control problem
- 2 The state equation
- 3 Necessary optimality conditions
- 4 Sparsity of the optimal control
- 5 Application of duality theory for "bottleneck type" linear programs

Sparse optimal control problem

$$\min J(y, u) := \frac{1}{2} \int_Q |y - y_Q|^2 dxdt + \frac{\nu}{2} \int_{\Sigma} |u|^2 d\sigma dt + \kappa \int_{\Sigma} |u| d\sigma dt$$

subject to

$$\partial_t y - \Delta y + d(y) = 0 \quad \text{in } Q := \Omega \times (0, T)$$

$$\partial_n y + b(y) = u \quad \text{in } \Sigma := \Gamma \times (0, T)$$

$$y(x, 0) = y_0 \quad \text{in } \Omega,$$

$$u_a \leq u(x, t) \leq u_d + y(x, t)$$

for a.a. $(x, t) \in \Sigma$.

Data

$y_Q \in L^\infty(Q)$, $T > 0$, $\Omega \subset \mathbb{R}^N$, $N \leq 3$, bounded Lipschitz domain; ∂_n : outward normal derivative; $\nu > 0$, $\kappa > 0$, $u_a < 0 < u_d$, assume $y_0 = 0$ for the talk.

State: $y \in W(0, T) \cap L^\infty(Q)$

Control: $u \in L^\infty(\Sigma)$.

Main difficulties

$$\min J(y, u) := \frac{1}{2} \int_Q |y - y_Q|^2 dx dt + \frac{\nu}{2} \int_{\Sigma} |u|^2 d\sigma dt + \kappa \int_{\Sigma} |u| d\sigma dt$$

subject to

$$\partial_t y - \Delta y + d(y) = 0 \quad \text{in } Q := \Omega \times (0, T)$$

$$\partial_n y + b(y) = u \quad \text{in } \Sigma := \Gamma \times (0, T)$$

$$y(x, 0) = 0 \quad \text{in } \Omega,$$

$$u_a \leq u(x, t) \leq u_d + y(x, t).$$

- Nondifferentiability of $\|u\|_{L^1(\Sigma)}$
- Nonlinearity of the control-to-state mapping $u \mapsto y$ (nonconvex problem)
- Existence of a Lagrange multiplier μ to $u - y - u_d \leq 0$

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Well-posedness of the state equation

Assumption (Simplified)

- d and b are real functions of class C^1 with locally Lipschitz derivative
- There is $c_0 \in \mathbb{R}$: $d'(y) \geq c_0$ and $b'(y) \geq c_0 \quad \forall y \in \mathbb{R}$.

Example: $d(y) = (y + 1)(y - 1)(y - 2), \quad b(y) = y^3|y|$
(FHN equations, Stefan-Boltzmann radiation condition)

Theorem

For every $u \in L^p(\Sigma)$ with $p > N + 1$, the state equation has a unique solution $y_u \in W(0, T) \cap L^\infty(Q)$.

We denote by y_u the state associated with u .

Inverse isotony

Notation: For $u, v \in L^2(\Sigma)$, we write $u \leq v$ if $u(x, t) \leq v(x, t)$ holds a.e. in Σ .

Theorem (E. Casas, F.T. 2018)

For every function $g \in L^p(\Sigma)$ with $p > N + 1$, the equation

$$v - y_v = g \quad \text{in } L^p(\Sigma)$$

has a unique solution $v \in L^p(\Sigma)$.

If $u \in L^p(\Sigma)$ satisfies the inequality

$$u - y_u \leq g \quad \text{in } L^p(\Sigma),$$

then $u \leq v$ holds.

Control-to-state mapping: $G : u \mapsto y_u$, $G(u) = y_u$, $G_\Sigma : u \mapsto y_u|_\Sigma$.

The theorem above in a nutshell:

$$u - G_\Sigma(u) \leq v - G_\Sigma(v) \implies u \leq v \quad (\text{inverse isotony})$$

Existence of an optimal control

Lemma

There is some constant $M > 0$ such that

$$u_a \leq u \leq M$$

holds for all feasible controls of the optimal control problem.

The proof is based on inverse isotony:

$$u \leq u_d + y_u \quad \Leftrightarrow \quad u - G_\Sigma(u) \leq u_d$$

$\Rightarrow u \leq z$, where $z \in L^\infty(\Sigma)$ solves $z - G_\Sigma(z) = u_d$; take $M := \|z\|_{L^\infty(\Sigma)}$.

Theorem

Assume that there exists a feasible control, i.e. a control with $u_a \leq u \leq u_d + y_u$. Then there exists at least one optimal control \bar{u} .

Theorem

The mapping G is of class C^1 . The derivative of G at $u \in L^p(\Sigma)$ in the direction $v \in L^p(\Sigma)$ is given by

$$G'(u)v = z_v,$$

where $z_v \in W(0, T) \cap L^\infty(Q)$ is the unique solution to the linearized equation

$$\partial_t z - \Delta z + d'(y_u)z = 0$$

$$\partial_n z + b'(y_u)z = v$$

$$z(x, 0) = 0.$$

By inverse isotony, we show that $G'(u)$ is a nonnegative operator, i.e.

$$v \geq 0 \Rightarrow G'(u)v \geq 0.$$

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Expected optimality conditions

$$\min J(u) := \frac{1}{2} \int_Q \underbrace{|y_u - y_Q|^2}_{G(u)} dxdt + \frac{\nu}{2} \int_{\Sigma} |u|^2 d\sigma dt + \kappa \int_{\Sigma} |u| d\sigma dt$$

subject to

$$\partial_t y - \Delta y + d(y) = 0$$

$$\partial_n y + b(y) = u$$

$$y(x, 0) = 0,$$

$$u_a \leq u \leq u_d + y_u.$$

Definition:

$$f(u) = \frac{1}{2} \int_Q |G(u) - y_Q|^2 dxdt + \frac{\nu}{2} \int_{\Sigma} |u(x, t)|^2 d\sigma dt,$$

$$j(u) = \|u\|_{L^1(\Sigma)}.$$

Reduced optimal control problem

The optimal control problem is equivalent to

$$\min_{u \in F} J(u) := f(u) + \kappa j(u) \quad (\text{P})$$

with the *feasible (in general non-convex) set* $F \subset L^\infty(\Sigma)$ defined by

$$F := \{u \in L^\infty(\Sigma) : u_a \leq u \leq u_d + G_\Sigma(u)\}.$$

Lagrangian function

$$\mathcal{L}(u, \mu) := f(u) + \kappa j(u) + \int_\Sigma (u - G_\Sigma(u) - u_d) \mu \, d\sigma dt.$$

If a Lagrange multiplier $0 \leq \mu \in L^2(\Sigma)$ exists then an optimal solution \bar{u} of (P) should satisfy the optimality conditions of the problem without state constraint

$$\min_{u \geq u_a} \mathcal{L}(u, \mu).$$

"Sum of a nonconvex differentiable and a convex nondifferentiable functional"

Roadmap for optimality conditions

- Linearization of the equation \longrightarrow Convex control problem with linear mixed control-state-constraint \leftarrow Linearization theorem
- Subdifferential calculus \longrightarrow Linearized control problem with linear mixed control-state-constraint
- Transfer of the problem to an L^2 -setting
- Duality theory of linear programming to prove the existence of a Lagrange multiplier μ
- Proof of a pointwise minimum principle
- Sparsity theorem (optimal control vanishes in some parts of Σ)

Linearization theorem

Abstract optimization problem: $\min \{f(u) + g(u) : u \in K \text{ and } H(u) \in C\}$.

- U, Y : normed vector spaces,
- $K \subset U, C \subset Y$: convex sets
- $H : U \rightarrow Y, f : U \rightarrow \mathbb{R}$, and $g : U \rightarrow (-\infty, +\infty]$: given mappings.

Theorem (E. Casas, F.T. 2018)

Let \bar{u} be an associated local solution. Assume that f and H are Fréchet differentiable at \bar{u} , g is convex, and $\text{int } C \neq \emptyset$. If the linearized Slater condition

$$\exists u_0 \in K : H(\bar{u}) + H'(\bar{u})(u_0 - \bar{u}) \in \text{int } C$$

is satisfied, then

$$f'(\bar{u})(u - \bar{u}) + g(u) - g(\bar{u}) \geq 0$$

holds for all $u \in K$ that satisfy

$$H(\bar{u}) + H'(\bar{u})(u - \bar{u}) \in C.$$

Linearized optimal control problem

We apply the theorem in $U = Y = L^\infty(\Sigma)$; **notice:** $\text{int } L^\infty(\Sigma)_+ \neq \emptyset$.

Assumption (Linearized Slater condition)

Let $\bar{u} \in L^\infty(\Sigma)$ be fixed. Assume the existence of $u_0 \in L^\infty(\Sigma)$ and $\delta > 0$ such that

$$u_0 \geq u_a \quad \text{and} \quad u_0 - u_d - y_{\bar{u}} - G'_\Sigma(\bar{u})(u_0 - \bar{u}) \leq -\delta.$$

From the abstract linearization theorem, we obtain:

Theorem (Linearization of the problem, E.Casas, F.T. 2018)

Let a locally optimal $\bar{u} \in L^\infty(\Sigma)$ satisfy the linearized Slater condition. Then

$$f'(\bar{u})(u - \bar{u}) + \kappa j(u) - \kappa j(\bar{u}) \geq 0$$

for all $u \in L^\infty(\Sigma)$ with

$$u_a \leq u \leq u_d + y_{\bar{u}} + G'_\Sigma(\bar{u})(u - \bar{u}).$$

Advantage: This is a convex problem, where subdifferential calculus applies.

Drawback: It is yet posed in $L^\infty(\Sigma)$ and $\bar{\mu} \in L^\infty(\Sigma)^*$ is not a good choice.

Roadmap for completing the necessary conditions

- We consider the linearized control problem in $L^2(\Sigma)$ instead of $L^\infty(\Sigma)$.
- Also in the linearized problem, we have uniform boundedness of all feasible controls. The feasible set does not change by extending it to L^2 !
- However, this brings a new difficulty: The interior of the cone of nonnegative functions of $L^2(\Sigma)$ is empty.
- We find a Lagrange multiplier $\bar{\mu} \in L^2(\Sigma)$ as solution of a dual linear programming problem posed in $L^2(\Sigma)$.
- We can even show $\bar{\mu} \in L^\infty(\Sigma)$ and uniform boundedness of $\bar{\mu}$ w.r. to κ .

Definition

The operator $G'_\Sigma(\bar{u}) : L^\infty(\Sigma) \rightarrow L^\infty(\Sigma)$ can be continuously extended to $L^2(\Sigma)$. We denote this extension by S ; in this extended sense we have

$$Su := G'_\Sigma(\bar{u})u \quad \forall u \in L^2(\Sigma).$$

Necessary conditions with subdifferential calculus

The control \bar{u} solves $\min_{u_a \leq u_d + y_{\bar{u}} + S(u - \bar{u})} \{f'(\bar{u})(u - \bar{u}) + g(u)\}$

Lemma (Use of subdifferential calculus)

Under the linearized Slater condition, there is some $\bar{\lambda} \in \partial j(\bar{u})$ such that

$$f'(\bar{u})(u - \bar{u}) + \kappa \int_{\Sigma} \bar{\lambda}(x, t)(u(x, t) - \bar{u}(x, t)) d\sigma dt \geq 0$$

is satisfied for all controls u of $L^2(\Sigma)$ that obey the linearized restrictions

$$u \geq u_a \quad \text{and} \quad u \leq u_d + y_{\bar{u}} + S(u - \bar{u}).$$

Notice that this is a linear problem!

We recall that $\bar{\lambda}$ belongs to the subdifferential $\partial j(\bar{u})$ iff

$$\bar{\lambda}(x, t) = \begin{cases} 1 & \text{if } \bar{u}(x, t) > 0 \\ \in [-1, 1] & \text{if } \bar{u}(x, t) = 0 \\ -1 & \text{if } \bar{u}(x, t) < 0. \end{cases}$$

Next step: Express $f'(\bar{u})(u - \bar{u})$ by an adjoint state.

Adjoint equation

$$\begin{aligned} -\partial_t \bar{\varphi} - \Delta \bar{\varphi} + \mathbf{d}'(\bar{\mathbf{y}}) \bar{\varphi} &= \bar{\mathbf{y}} - \mathbf{y}_Q && \text{in } Q \\ \partial_n \bar{\varphi} + \mathbf{b}'(\bar{\mathbf{y}}) \bar{\varphi} &= 0 && \text{in } \Sigma \\ \bar{\varphi}(x, T) &= 0 && \text{in } f\Omega. \end{aligned}$$

With this **adjoint state** $\bar{\varphi}$, we have

$$f'(\bar{\mathbf{u}})(\mathbf{u} - \bar{\mathbf{u}}) = \int_{\Sigma} (\bar{\varphi} + \nu \bar{\mathbf{u}})(\mathbf{u} - \bar{\mathbf{u}}) \, d\sigma dt.$$

Preliminary optimality conditions

Theorem (Preliminary necessary condition)

Let \bar{u} be a local solution to the optimal control problem that satisfies the linearized Slater condition. Then there are $\bar{\lambda} \in \partial j(\bar{u})$ and a unique adjoint state $\bar{\varphi} \in L^\infty(\Sigma)$ such that

$$\int_{\Sigma} (\bar{\varphi}(x, t) + \nu \bar{u}(x, t) + \kappa \bar{\lambda}(x, t))(u(x, t) - \bar{u}(x, t)) d\sigma dt \geq 0$$

holds for all $u \in L^2(\Sigma)$ that satisfy

$$u_a \leq u \leq u_d + y_{\bar{u}} + S(u - \bar{u}).$$

Why only preliminary?

The upper constraint for u is not a pointwise one, it is still an "operator inequality".

Existence of a Lagrange multiplier

Assumption (Feasibility of u_a)

The lower bound u_a is assumed to be feasible for the linearized problem, i.e. there holds

$$u_a \leq u_d + y_{\bar{u}} + S(u_a - \bar{u}).$$

Theorem (Proof by duality theory)

Let the assumption above be fulfilled and let $\bar{u} \in L^2(\Sigma)$ solve the linearized control problem. Denote by $\bar{y} := y_{\bar{u}}$ the associated state. Then there exists a non-negative **Lagrange multiplier** $\bar{\mu} \in L^\infty(\Sigma)$ associated with the mixed control-state constraint. This is a function $\bar{\mu}$ such that

$$\int_{\Sigma} (\bar{\varphi} + \nu \bar{u} + \kappa \bar{\lambda} + \bar{\mu} - S^* \bar{\mu})(u - \bar{u}) d\sigma dt \geq 0 \quad \forall u \geq u_a$$

and the following complementarity conditions are satisfied

$$\bar{\mu} \geq 0, \quad \bar{u} - \bar{y} - u_d \leq 0, \quad \int_{\Sigma} (\bar{u} - \bar{y} - u_d) \bar{\mu} d\sigma dt = 0.$$

Pointwise minimum principle

Yet, the optimality condition cannot be directly applied to proving sparsity properties. We slightly reformulate them to get a pointwise version:

Theorem (Pointwise minimum principle)

Let the last assumption be satisfied. If \bar{u} is a locally optimal control and $\bar{\mu} \in L^\infty(\Sigma)$ is an associated Lagrange multiplier that exists according to the last theorem, then for almost all $(x, t) \in \Sigma$ the solution u of the problem

$$\min_{\underbrace{(\bar{\varphi} - \mathbf{S}^* \bar{\mu} + \nu \bar{u} + \kappa \bar{\lambda})}_{\bar{\psi}}}(x, t) u$$

subject to $u \in \mathbb{R}$ and

$$u_a \leq u \leq u_d + \bar{y}(x, t)$$

is attained by the real number $u = \bar{u}(x, t)$.

Extended adjoint state

We have $\bar{\varphi}|_{\Sigma} - \mathbf{S}^* \bar{\mu} = \bar{\psi}|_{\Sigma}$ where the extended adjoint state $\bar{\psi}$ solves

$$-\partial_t \psi - \Delta \psi + \mathbf{d}'(\bar{y}) \psi = \bar{y} - y_Q$$

$$\partial_n \psi + \mathbf{b}'(\bar{y}) \psi = -\bar{\mu}$$

$$\psi(x, T) = 0.$$

Pointwise minimum principle

$$\min_{u_a \leq u \leq u_d + \bar{y}(x, t)} (\bar{\psi}(x, t) + \nu \bar{u}(x, t) + \kappa \bar{\lambda}(x, t)) u$$

is attained by $u = \bar{u}(x, t)$ a.e. on Σ .

For sparsity, we need $u_a < 0$ and $u_d + \bar{y}(x, t) > 0 \dots$

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Lemma

Let $u = 0$ be feasible, denote by \bar{u}_κ a family of optimal controls corresponding to $\kappa \geq 0$, and let $\bar{y}_\kappa = G(\bar{u}_\kappa)$ be the associated states. Then

$$\lim_{\kappa \rightarrow \infty} \|\bar{u}_\kappa\|_{L^p(\Sigma)} = 0 \quad \forall p \in [1, \infty)$$

and hence $\bar{y}_\kappa \rightarrow y^0$ in $L^2(0, T; H^1(\Omega)) \cap L^\infty(Q)$, where $y^0 = G(0)$.

Main idea: Assume for simplicity that d and b are monotone; $u = 0$ is feasible for all κ , and then $y = 0$ is the associated state. Therefore,

$$\begin{aligned} f(\bar{u}_\kappa) + \kappa j(\bar{u}_\kappa) &= \frac{1}{2} \|\bar{y}_\kappa - y_Q\|_{L^2(Q)}^2 + \frac{\nu}{2} \|\bar{u}_\kappa\|_{L^2(\Sigma)}^2 + \kappa \|\bar{u}_\kappa\|_{L^1(\Sigma)} \\ &\leq f(0) + \kappa j(0) = \frac{1}{2} \|y_Q\|_{L^2(Q)}^2. \end{aligned}$$

This immediately yields $\|\bar{u}_\kappa\|_{L^1(\Sigma)} \rightarrow 0$. Uniform boundedness $\Rightarrow \|\bar{u}_\kappa\|_{L^p(\Sigma)} \rightarrow 0$, hence $(\bar{y}_\kappa)|_\Sigma \rightarrow 0$ in $L^\infty(\Sigma)$. \square

This shows that $(u_d + \bar{y}_\kappa)(x, t) \geq \delta > 0 \quad \forall \kappa \geq \kappa_0$, a.e. in Σ .

Theorem (Sparsity)

- Let all of our assumptions be satisfied and let \bar{u} be a locally optimal control. Then a Lagrange multiplier $\bar{\mu} \in L^\infty(\Sigma)$ and some $\kappa_0 > 0$ exist such that a.e. in Σ

$$\bar{u}(x, t) = 0 \Leftrightarrow |\bar{\psi}(x, t)| \leq \kappa \quad \forall \kappa \geq \kappa_0.$$

- There is $\kappa_1 > 0$, such that $\bar{u} = 0 \quad \forall \kappa \geq \kappa_1$. ← $\|\bar{\mu}_\kappa\|_\infty \leq K \forall \kappa$
- The element $\bar{\lambda}$ of the subdifferential $\partial j(\bar{u})$ is given by

$$\bar{\lambda}(x, t) = \mathbb{P}_{[-1,1]} \left\{ -\frac{1}{\kappa} \bar{\psi}(x, t) \right\}.$$

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Duality theory and existence of $\bar{\mu}$

$$-a(x, t) := \bar{\varphi}(x, t) + \nu \bar{u}(x, t) + \bar{\lambda}(x, t)$$

$\bar{v} := \bar{u} - u_a$ solves the *primal problem*

$$\max \int_{\Sigma} a(x, t) v(x, t) d\sigma dt$$

$$v \leq b + Sv$$

$$v \geq 0.$$

$\bar{\mu}$ solves the *dual problem*

$$\min \int_Q b(x, t) \mu(x, t) d\sigma dt$$

$$\mu \geq a + S^* \mu$$

$$\mu \geq 0.$$

(provided that a solution exists)

Have $b := u_d + \bar{y} + S(u_a - \bar{u}) - u_a \geq 0!$

Main steps:

- Equality of primal and dual optimal value ← boundedness condition, separation theorem
- Existence of an optimal $\bar{\mu}$ of the dual problem ← monotone behaviour, $b \geq 0$
- Uniform boundedness of μ w.r. to κ ← a bit technical

Existence of $\bar{\mu}$

Dual problem:

$$\begin{aligned} \min \int_Q b(x, t) \mu(x, t) d\sigma dt \\ \mu \geq a + S^* \mu & \quad \text{a.e. in } \Sigma \\ \mu \geq 0. & \quad \text{a.e. in } \Sigma. \end{aligned}$$

Since $b \geq 0$ and $S^* \geq 0$, we find the solution $\bar{\mu}$ from

$$\mu = (a + S^* \mu)_+.$$

Dual linear programming problems in Hilbert space

Given:

- Real Hilbert spaces $\{U, \|\cdot\|_U\}, \{V, \|\cdot\|_V\}$,
- $A: U \rightarrow V$ linear and continuous,
- $K_U \subset U$ and $K_V \subset V$ nonempty, convex and closed "nonnegative" cones,
- partial orderings $u \geq_U 0$ iff $u \in K_U$, analogously $v \geq_V 0$ iff $v \in K_V$,
- elements $a \in U$ and $c \in V$.

Primal problem

$$\max (a, u)_U \quad \text{subject to} \quad Au \leq_V c, \quad u \geq_U 0. \quad (\mathcal{PP})$$

Dual problem

$$\min (c, v)_V \quad \text{subject to} \quad A^*v \geq_U^* a, \quad v \geq_V^* 0. \quad (\mathcal{DP})$$

Boundedness condition

Define for (variable) $d \in V$ and $e \in U$ the set

$$P(d) = \{u \in U : u \geq_U 0, Au \leq_V d\},$$

Boundedness condition

$\exists \eta > 0$ independent of d and, for all $d \in V$, a closed set $K(d) \subset P(d)$ such that:

$$\|u\|_U \leq \eta \|d\|_V \quad \forall u \in K(d)$$

$$\forall u \in P(d) \exists \hat{u} \in K(d) : (a, u)_U \leq (a, \hat{u})_U.$$

Duality theorem

In our case: $U = V = L^2(\Omega)$, $A = I - S$.

$$P(d) = \{u \in L^2(\Omega) : u \geq 0, \quad u - Su \leq d\}$$

The boundedness condition is easily verified.

Theorem (Strong duality)

If the feasible set $P(c)$ of the primal problem (\mathcal{PP}) is nonempty and the boundedness condition is satisfied, then the primal problem has an optimal solution. Moreover, the strong duality relation

$$\max_{u \in P(c)} (a, u)_U = \inf_{v \in D(a)} (c, v)_V$$

is fulfilled.

If the boundedness condition is satisfied, then the following cone is closed:

$$K = \{(d, \delta) \in V \times \mathbb{R} : \exists u \in P(d) \text{ with } (a, u)_U \geq \delta\}.$$

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Best wishes, Jean-Pierre!



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