



**Weierstrass Institute for
Applied Analysis and Stochastics**



**Leibniz
Gemeinschaft**

Mini-Course “Introduction to Optimal Control Problems for PDEs”

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Lecture 1: Basic Concepts

Lecture 2: Elliptic Control Problems

Lecture 3: Linear-quadratic Parabolic Control Problems

Lecture 4: Necessary Optimality Conditions for General Optimization Problems in Banach
Spaces

1. What is an optimal control problem (OCP)?

A typical OCP consists of 4 ingredients:

- a **control function** u ;
- a **state equation**, which associates with every control u a **state function** y ;
- a **cost functional**, which depends on u and y , to be minimized;
- various **constraints** to be obeyed by the control u and the state y .

Constraints on u are called **control constraints**, constraints on y **state constraints**. Very often, the control and state constraints are **pointwise constraints**. In this summer school, the state equation will be given by an initial-boundary value problem for a PDE or by a variational inequality.

Steps to be taken in the analysis on an OCP:

- Existence of a solution. Is it unique?
- Derivation of first-order necessary and second-order sufficient optimality conditions
- Numerical approximation and analysis of the discretized problem. Do we have convergence?
- Implementation of the discretized problem and calculation of an approximating solution

2. Examples: Linear state equations

Example 1: Optimal stationary heating

Let $\Omega \subset \mathbb{R}^3$ be the spatial location of a body to be heated/cooled at its boundary Γ . We apply a heat source u (the control) to Γ that is constant in time: $u = u(x)$. We aim to choose u in such a way that the corresponding temperature $y = y(x)$ (the state) is the “best approximation” to some desired $y_\Omega = y_\Omega(x)$ in Ω . Model:

$$\min J(y, u) := \frac{1}{2} \int_{\Omega} |y(x) - y_\Omega(x)|^2 dx + \frac{\lambda}{2} \int_{\Gamma} |u(x)|^2 ds(x),$$

subject to the **state equation**

$$\begin{array}{rcl} -\Delta y & = & 0 \quad \text{in } \Omega \\ \frac{\partial y}{\partial \nu} & = & \alpha(u - y) \quad \text{on } \Gamma \end{array}$$

and the **pointwise control constraints**

$$u_a(x) \leq u(x) \leq u_b(x) \quad \text{on } \Gamma.$$

Here, $\lambda > 0$ is a parameter, and u_a, u_b are given.

This is a **linear-quadratic elliptic boundary control problem**.

Examples: Linear state equations

Example 2: Optimal heat source

Assume here that Ω is heated by a distributed heat source u acting in Ω (e.g., by electromagnetic induction). We then get:

$$\min J(y, u) := \frac{1}{2} \int_{\Omega} |y(x) - y_{\Omega}(x)|^2 dx + \frac{\lambda}{2} \int_{\Omega} |u(x)|^2 dx,$$

subject to

$$\begin{array}{rcl} -\Delta y & = & \beta u \quad \text{in } \Omega \\ \frac{\partial y}{\partial \nu} & = & \alpha (y_a - y) \quad \text{on } \Gamma \end{array}$$

and

$$u_a(x) \leq u(x) \leq u_b(x) \quad \text{in } \Omega.$$

Here, $y_a = \text{ext. temperature}$, $\beta = \beta(x)$ given, e.g., $\beta = \chi_D$ for $D \subset \Omega$.

This is a **linear-quadratic elliptic control problem with distributed control**.

Example 3: Optimal nonstationary boundary control

This time we assume that $\Omega \subset \mathbb{R}^3$ is heated over some time $[0, T]$, $T > 0$. We want to apply a control $u = u(x, t)$ in such a way that a desired temperature distribution y_Ω is reached at $t = T$. We put $Q := \Omega \times (0, T)$, $\Sigma := \Gamma \times (0, T)$. Problem:

$$\min J(y, u) := \frac{1}{2} \int_{\Omega} |y(x, T) - y_\Omega(x)|^2 dx + \frac{\lambda}{2} \int_0^T \int_{\Gamma} |u(x, t)|^2 ds(x) dt,$$

subject to

$y_t - \Delta y$	$=$	0	in Q
$\frac{\partial y}{\partial \nu}$	$=$	$\alpha(u - y)$	on Σ
$y(x, 0)$	$=$	$y_0(x)$	in Ω

and

$$u_a(x, t) \leq u(x, t) \leq u_b(x, t) \quad \text{on } \Sigma.$$

This is a **linear-quadratic parabolic boundary control problem**.

Example 4: Control of nonstationary flows

Nonstationary flows of incompressible media in $\Omega \subset \mathbb{R}^2$ (or \mathbb{R}^3) are described by the **Navier–Stokes equations**

$$\begin{aligned} u_t - \frac{1}{Re} \Delta u + (u \cdot \nabla) u + \nabla p &= f && \text{in } Q \\ \operatorname{div} u &= 0 && \text{in } Q \\ u &= 0 && \text{on } \Sigma \\ u(\cdot, 0) &= u_0 && \text{in } \Omega. \end{aligned}$$

Here, typically, the velocity field u (which is in this case the state variable) is to be controlled by the application of electromagnetic fields. In this case, $f = \text{gravity } g + \text{Lorentz force } j \times B$.

One then obtains a control problem with a nonlinear state equation.

Example 5: Control of phase-field equations (C. Lefter + J. S., AMSA 17 (2007))

Consider a liquid-solid phase transition, and let $\chi \in [0, 1]$ denote the **order parameter**: $\{\chi = 0\} \Leftrightarrow$ solid phase, $\{\chi = 1\} \Leftrightarrow$ liquid phase, $\{0 < \chi < 1\} \Leftrightarrow$ “mush”. A typical model is ($\theta =$ absolute temperature)

$$\mu(\theta)\chi_t = -F_1'(\chi) - \left(\frac{\beta_1}{\theta} + \beta_2\right)F_2'(\chi) - \frac{F_3'(\chi)}{\theta} \quad \text{in } Q,$$

$$C_V\theta_t + (\beta_1F_2'(\chi) + F_3'(\chi))\chi_t - \Delta\theta = 0 \quad \text{in } Q,$$

$$\frac{\partial\theta}{\partial\nu} = u - \theta \quad \text{on } \Sigma,$$

$$\chi(\cdot, 0) = \chi_0, \theta(\cdot, 0) = \theta_0, \quad \text{in } \Omega.$$

F_1, F_3 : smooth
 $F_2(\chi) = \kappa(\chi \ln(\chi) + (1 - \chi) \ln(1 - \chi))$,
 $\kappa > 0$.

Control problem:

$$\min J(\chi, \theta, u) := \frac{1}{2} \int_{\Omega} \left(|\theta(x, T) - \theta_{\Omega}(x)|^2 + |\chi(x, T) - \chi_{\Omega}(x)|^2 \right) dx + \frac{\lambda}{2} \int_{\Sigma} |u(x, t)|^2 ds(x) dt$$

subject to the IBVP and the control constraints

$$0 < u_a \leq u(x, t) \leq u_b \quad \text{on } \Sigma.$$

Example 6: A control-into-coefficients problem

Consider a clamped plate $\Omega \subset \mathbb{R}^2$, which is described by the BVP

$$\begin{aligned} \Delta(bu^3(x)\Delta y) &= f && \text{in } \Omega \\ y = \frac{\partial y}{\partial \nu} &= 0 && \text{on } \Gamma. \end{aligned}$$

Here, $b > 0$ and the load f are given; $u \in L^\infty(\Omega)$ stands for the thickness ($> 0!$) of the plate. A typical problem is:

$$\min J(y, u) := \int_{\Omega} u(x) dx \quad (= \text{the weight of the plate})$$

subject to the BVP and the control and state constraints (= safety requirements)

$$\begin{aligned} 0 < m &\leq u(x) \leq M && \text{in } \Omega \\ y(x) &\geq -\tau && \text{in } \Omega, \end{aligned}$$

where $m, M, \tau > 0$ are given.

Example 7: Optimal layout of materials

Let $\Omega \subset \mathbb{R}^3$ be a body composed of m different materials M_i having thermal conductivities k_i , $i = 1, \dots, m$. Then the total thermal conductivity k of Ω is

$k(x) = \sum_{i=1}^m \chi_i(x) k_i$, where χ_i is the characteristic function of the region occupied by M_i .

Problem: Given a heat source f , what is the optimal distribution of the materials in Ω that maximizes the temperature y in a given subdomain $\omega \subset \Omega$? Model:

$$\min_k \left\{ - \int_{\omega} y(x) dx \right\}$$

subject to

$$\begin{aligned} -\nabla \cdot (k(x) \nabla y) &= f && \text{in } \Omega \\ \frac{\partial y}{\partial \nu} &= 0 && \text{on } \Gamma. \end{aligned}$$

The optimization parameters are the subsets of Ω occupied by the various materials.

Example 8: Electrochemical machining

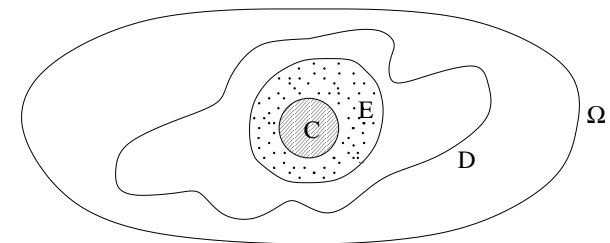
This is an OCP with a variational inequality as state equation:

$$\min_{E \subset D \subset \Omega} J(D) = \frac{1}{2} \int_{E \setminus C} |y(x)|^2 dx,$$

subject to

$$\int_{D \setminus C} \nabla y(x) \cdot \nabla (y - z)(x) dx \leq \int_{D \setminus C} f(x) (y(x) - z(x)) dx,$$

$$\forall z \in S = \{w \in H^1(D \setminus C) : w|_{\partial C} = 0, \\ w|_{\partial D} = 1, w \geq 0 \text{ a.e. in } D \setminus C\},$$



C : core

D : machine

$\partial C, \partial D$: electrodes

$E \setminus C$: desired shape
(choice of E)

$E_y = \{x \in D \setminus C : y(x) = 0\}$:
final shape obtained

To guarantee: $E_y \supset E \setminus C$!

3. Review of finite-dimensional theory

1. Let: $J = J(y, u) : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}$ cost functional to be minimized
 $A \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{n \times m}$, $U_{ad} \subset \mathbb{R}^m$ nonempty. We consider:

$$\begin{array}{l} \min J(y, u) \\ Ay = Bu, u \in U_{ad} \end{array}$$

Convention: All vectors are column vectors!

Example: $J(y, u) = \frac{1}{2}|y - y_d|^2 + \frac{\lambda}{2}|u|^2$, $y_d \in \mathbb{R}^n$ given; $|\cdot| =$ Euclidean norm

Assume: $\exists A^{-1} \implies y = Su$, with the solution matrix $S := A^{-1}B$.

\implies **reduced cost functional** $f(u) := J(y, u) = J(Su, u)$.

In the example: $f(u) = \frac{1}{2}|Su - y_d|^2 + \frac{\lambda}{2}|u|^2$.

\implies **reduced problem**

$$\min f(u), \quad u \in U_{ad}$$

2. Existence:

Def.: If $\bar{u} \in U_{ad}$ satisfies $f(\bar{u}) \leq f(u) \quad \forall u \in U_{ad}$, then \bar{u} is called **optimal control**, and $\bar{y} := S\bar{u}$ is called associated **optimal state**.

Theorem: Let $J : \mathbb{R}^n \times U_{ad} \rightarrow \mathbb{R}$ be continuous, $U_{ad} \neq \emptyset$, closed, bounded. If $\exists A^{-1}$, then \exists an optimal pair (\bar{u}, \bar{y}) .

Proof: f is continuous on the compact set U_{ad} .

3. First-order necessary optimality conditions:

We use the notation: for differentiable $f : \mathbb{R}^m \rightarrow \mathbb{R}$ we write

$f'(u)h = \nabla f(u) \cdot h$, $h \in \mathbb{R}^m$. Let J be continuously differentiable w.r.t. y, u .

$\implies f(u) = J(Su, u)$ is C^1

Theorem: Let U_{ad} be convex. Then any optimal control \bar{u} satisfies the variational inequality: $f'(u)(u - \bar{u}) \geq 0 \quad \forall u \in U_{ad}$.

Chain rule: $f' = D_y J S + D_u J$, and thus

$$\begin{aligned} f'(\bar{u})h &= D_y J(S\bar{u}, \bar{u})Sh + D_u J(S\bar{u}, \bar{u})h \\ &= (\nabla_y J(\bar{y}, \bar{u}), A^{-1}Bh)_{\mathbb{R}^n} + (\nabla_u J(\bar{y}, \bar{u}), h)_{\mathbb{R}^m} \\ &= (B^\top (A^\top)^{-1} \nabla_y J(\bar{y}, \bar{u}) + \nabla_u J(\bar{y}, \bar{u}), h)_{\mathbb{R}^m}. \end{aligned}$$

\implies Variational inequality becomes

$$(B^\top (A^\top)^{-1} \nabla_y J(\bar{y}, \bar{u}) + \nabla_u J(\bar{y}, \bar{u}), u - \bar{u})_{\mathbb{R}^m} \geq 0 \quad \forall u \in U_{ad}.$$

This requires to evaluate the inverse of A^\top , which is usually a bad idea. We therefore introduce the **adjoint state** p , given by $p = (A^\top)^{-1} \nabla_y J(\bar{y}, \bar{u})$, which solves the **adjoint state equation**

$$A^\top p = \nabla_y J(\bar{y}, \bar{u}).$$

Example: $J(y, u) = \frac{1}{2}|y - y_d|^2 + \frac{\lambda}{2}|u|^2 \implies A^\top p = y - y_d.$

Advantages: First-order condition simplifies, use of A^{-1} is avoided. Also the form of $\nabla f(\bar{u})$ simplifies: $\nabla f(\bar{u}) = B^\top p + \nabla_u J(\bar{y}, \bar{u}).$

$$\implies f'(\bar{u})h = (B^\top p + \nabla_u J(\bar{y}, \bar{u}), h)_{\mathbb{R}^m}.$$

Theorem 1: Suppose $\exists A^{-1}$. Let (\bar{y}, \bar{u}) be an optimal pair. Then there exists an associated adjoint state \bar{p} such that $(y, u, p) = (\bar{y}, \bar{u}, \bar{p})$ solves

$$Ay = Bu, \quad u \in U_{ad}$$

$$A^\top p = \nabla_y J(y, u)$$

$$(B^\top p + \nabla_u J(y, u), v - u)_{\mathbb{R}^m} \geq 0 \quad \forall v \in U_{ad}.$$

In the case $U_{ad} = \mathbb{R}^m$ follows: $B^\top \bar{p} + \nabla_u J(\bar{y}, \bar{u}) = 0.$

Example: $J(y, u) = \frac{1}{2} |Cy - y_d|^2 + \frac{\lambda}{2} |u|^2, \quad C \in \mathbb{R}^{n \times n}.$

$$\implies \nabla_y J(y, u) = C^\top (Cy - y_d), \quad \nabla_u J(y, u) = \lambda u$$

\implies Optimality system:

$$Ay = Bu, \quad u \in U_{ad}$$

$$A^\top p = C^\top (Cy - y_d)$$

$$(B^\top p + \lambda u, v - u)_{\mathbb{R}^m} \geq 0 \quad \forall v \in U_{ad}.$$

If $U_{ad} = \mathbb{R}^m$, then $B^\top \bar{p} + \lambda \bar{u} = 0$, and if $\lambda > 0$, then $\bar{u} = -\frac{1}{\lambda} B^\top \bar{p}$. We obtain the optimality system

$$Ay = -\frac{1}{\lambda} BB^\top p$$

$$A^\top p = C^\top (Cy - y_d)$$

4. Lagrangians

Def.: The function $L : \mathbb{R}^{2n+m} \rightarrow \mathbb{R}$, $L(y, u, p) := J(y, u) - (Ay - Bu, p)_{\mathbb{R}^n}$, is called the **Lagrangian function** of the OCP.

Using L , we can eliminate the equality constraint $Ay = Bu$ from the problem. The second and third optimality conditions read:

$$\begin{aligned}\nabla_y L(\bar{y}, \bar{u}, \bar{p}) &= 0 \\ (\nabla_u L(\bar{y}, \bar{u}, \bar{p}), u - \bar{u})_{\mathbb{R}^m} &\geq 0 \quad \forall u \in U_{ad}.\end{aligned}$$

$\implies (\bar{y}, \bar{u})$ solves the necessary optimality conditions of a minimization problem **without** equality constraints:

$$\min_{(y, u)} L(y, u, p), \quad u \in U_{ad}, y \in \mathbb{R}^n.$$

Remark: p plays the role of a **Lagrange multiplier** to $Ay = Bu$.

5. Discussion of the variational inequality

We now assume **box constraints** for u :

$$U_{ad} = \{u \in \mathbb{R}^m : u_a \leq u \leq u_b\}, \quad u_a, u_b \in \mathbb{R}^m \quad (\text{componentwise})$$

Variational inequality yields

$$(B^\top \bar{p} + \nabla_u J(\bar{y}, \bar{u}), \bar{u})_{\mathbb{R}^m} \leq (B^\top \bar{p} + \nabla_u J(\bar{y}, \bar{u}), u)_{\mathbb{R}^m} \quad \forall u \in U_{ad}$$

$\implies \bar{u}$ solves

$$\min_{u \in U_{ad}} (B^\top \bar{p} + \nabla_u J(\bar{y}, \bar{u}), u)_{\mathbb{R}^m} = \min_{u \in U_{ad}} \sum_{i=1}^m (B^\top \bar{p} + \nabla_u J(\bar{y}, \bar{u}))_i u_i$$

Since u_i are independent of each other \implies

$$(B^\top \bar{p} + \nabla_u J(\bar{y}, \bar{u}))_i \bar{u}_i = \min_{u_{a,i} \leq u_i \leq u_{b,i}} (B^\top \bar{p} + \nabla_u J(\bar{y}, \bar{u}))_i u_i, \quad 1 \leq i \leq m.$$

Hence, we must have:

$$\bar{u}_i = \begin{cases} u_{b,i} & \text{if } (B^\top \bar{p} + \nabla_u J(\bar{y}, \bar{u}))_i < 0 \\ u_{a,i} & \text{if } (B^\top \bar{p} + \nabla_u J(\bar{y}, \bar{u}))_i > 0. \end{cases}$$

We have no direct information if $(B^\top \bar{p} + \nabla_u J(\bar{y}, \bar{u}))_i = 0$.

Notice that we only need information about $(B^\top \bar{p} + \nabla_u J(\bar{y}, \bar{u}))$, not about \bar{p}, \bar{u} directly!

Moreover, putting (with $z_+ := \frac{1}{2}(|z| + z)$, $z_- := \frac{1}{2}(|z| - z)$)

$$\begin{aligned} \mu_a &:= (B^\top \bar{p} + \nabla_u J(\bar{y}, \bar{u}))_+ \\ \mu_b &:= (B^\top \bar{p} + \nabla_u J(\bar{y}, \bar{u}))_-, \end{aligned}$$

We find that

$$\begin{array}{l} \mu_a \geq 0, \quad u_a - \bar{u} \leq 0, \quad (u_a - \bar{u}, \mu_a)_{\mathbb{R}^m} = 0 \\ \mu_b \geq 0, \quad \bar{u} - u_b \leq 0, \quad (\bar{u} - u_b, \mu_b)_{\mathbb{R}^m} = 0 \end{array}$$

These are the so-called **complementary slackness conditions**.

Remark: We obviously have $\mu_a - \mu_b = \nabla_u J(\bar{y}, \bar{u}) + B^\top \bar{p}$

$$(1) \quad \implies \nabla_u J(\bar{y}, \bar{u}) + B^\top \bar{p} - \mu_a + \mu_b = 0$$

6. The Karush–Kuhn–Tucker conditions

Introduce the **extended Lagrangian** \mathcal{L} by adding the inequality constraints in the form

$$\mathcal{L}(y, u, p, \mu_a, \mu_b) := J(y, u) - (Ay - Bu, p)_{\mathbb{R}^n} + (u_a - u, \mu_a)_{\mathbb{R}^m} + (u - u_b, \mu_b)_{\mathbb{R}^m}.$$

Then, using (1),

$$\nabla_u \mathcal{L}(\bar{y}, \bar{u}, \bar{p}, \mu_a, \mu_b) = 0.$$

Moreover, since $\nabla_y \mathcal{L} = \nabla_y L$, the adjoint equation can be written as

$$\nabla_y \mathcal{L}(\bar{y}, \bar{u}, \bar{p}, \mu_a, \mu_b) = 0.$$

Hence, μ_a, μ_b are the Lagrange multipliers corresponding to the inequality constraints, and we have:

Theorem 2 (Karush–Kuhn–Tucker conditions)

Suppose $\exists A^{-1}$. Let $U_{ad} = \{u \in \mathbb{R}^m : u_a \leq u \leq u_b\}$. If (\bar{y}, \bar{u}) is an optimal pair, then \exists multipliers $p \in \mathbb{R}^n$ and $\mu_a, \mu_b \in \mathbb{R}^m$ such that:

$$\nabla_y \mathcal{L}(\bar{y}, \bar{u}, \bar{p}, \mu_a, \mu_b) = 0$$

$$\nabla_u \mathcal{L}(\bar{y}, \bar{u}, \bar{p}, \mu_a, \mu_b) = 0$$

$$\mu_a \geq 0, \quad \mu_b \geq 0$$

$$(u_a - \bar{u}, \mu_a) = (\bar{u} - u_b, \mu_b) = 0$$

4. General existence results

We assume:

- (A1) V, U are reflexive Banach spaces with duals V^*, U^* ; H is a Hilbert space with $V \subset H$ with continuous embedding
- (A2) U_{ad} is nonempty, closed and convex in U ;
 C is nonempty, closed and convex in H
- (A3) $A \in \mathcal{L}(V, V^*)$, $B \in \mathcal{L}(U, V^*)$, $f \in V^*$
- (A4) $J: V \times U \rightarrow (-\infty, +\infty]$ is proper, convex, l.s.c.

We consider the control and state constrained OCP:

$$\min J(y, u) \text{ subject to } Ay = Bu + f, \quad u \in U_{ad}, \quad y \in C.$$

We assume **admissibility**:

- (A5) $\exists (y_0, u_0) \in C \times U_{ad}$ such that $Ay_0 = Bu_0 + f$ and $J(y_0, u_0) < +\infty$.

Theorem 3: Suppose $\exists A^{-1} \in \mathcal{L}(V^*, V)$. If U_{ad} is bounded or J is uniformly coercive in u , i.e.,

$$(2) \quad \lim_{\|u\|_U \rightarrow \infty} J(y, u) = +\infty \quad \text{uniformly in } y,$$

then OCP has at least one optimal pair $(\bar{y}, \bar{u}) \in C \times U_{ad}$.

Proof: If U_{ad} is bounded, redefine J outside $V \times U_{ad}$ by $+\infty$. We thus may assume that (2) holds. By **(A5)**, \exists minimizing sequence $\{u_n\}$ such that, with $y_n = A^{-1} B u_n \in C$, we have

$$\lim_{n \rightarrow \infty} J(y_n, u_n) = \inf(\text{OCP}) =: \delta.$$

By (2) $\{u_n\}$ is bounded in U , and therefore $\{y_n\}$ is bounded in V . We thus may assume that $(y_n, u_n) \rightarrow (\bar{y}, \bar{u})$ weakly in $V \times U$, and hence in $H \times U$. Consequently, by **(A2)**, we have $(\bar{y}, \bar{u}) \in C \times U_{ad}$. Moreover, $\bar{y} = A^{-1} B \bar{u}$. Since J is weakly l.s.c. on $V \times U$, we find that

$$J(\bar{y}, \bar{u}) \leq \liminf_{n \rightarrow \infty} J(y_n, u_n) = \delta.$$

□

We now consider the more general state equation

$$(3) \quad A(u)y + \partial\varphi(y) \ni Bu + f.$$

Here, we assume $B \in \mathcal{L}(U, V^*)$, $f \in V^*$ (as above) and:

(A6) $\varphi : V \rightarrow (-\infty, +\infty]$ is proper, convex, l.s.c., with subdifferential

$$\partial\varphi(x) = \{w \in V^*; \varphi(x) - \varphi(v) \leq (w, x - v)_{V^* \times V} \quad \forall v \in V\}$$

(A7) (i) $A(u) \in \mathcal{L}(V, V^*) \quad \forall u \in U_{ad}$

(ii) $u_n \rightarrow u$ strongly in $U \implies A(u_n) \rightarrow A(u)$ strongly in $\mathcal{L}(V, V^*)$

(iii) $(A(u)y, y)_{V^* \times V} \geq m\|y\|_V^2 \quad \forall y \in V$, for some $m > 0$.

J, U_{ad}, C have the same properties as above.

Theorem 4: Let the above conditions be satisfied, and suppose $\exists(y_0, u_0) \in C \times U_{ad}$ such that $A(u_0)y_0 + \partial\varphi(y_0) \ni Bu_0 + f$ and $L(y_0, u_0) < +\infty$ (admissibility).

Then the OCP

$$\min J(y, u), \text{ subject to } (y, u) \in C \times U_{ad} \text{ and (3)}$$

has a solution provided that one of the following conditions is satisfied:

- (a) $U_{ad} \subset U$ is compact (but not necessarily convex!)
- (b) $C \subset V$ is compact (but not necessarily convex!), $U_{ad} \subset U$ is bounded, and the mapping $u \mapsto A(u)$ is linear and bounded.

Proof (only of (a)): Pick a minimizing sequence $\{u_n\} \subset U_{ad}$, $J(y_n, u_n) \rightarrow \inf(\text{OCP})$, where $\{y_n\} \subset C$ and $A(u_n)y_n + \partial\varphi(y_n) \ni Bu_n + f$.

Since U_{ad} is compact, we may assume that $\|u_n - u\|_U \rightarrow 0$ for some $u \in U_{ad}$.

Then $Bu_n \rightarrow Bu$ strongly in V^* , and $A(u_n) \rightarrow A(u)$ strongly in $\mathcal{L}(V, V^*)$.

Now, $\forall n \in \mathbb{N} \exists v_n \in \partial\varphi(y_n) : A(u_n)y_n + v_n = Bu_n + f$.

The monotonicity of the graph $\partial\varphi$ (φ is convex!) yields:

$$\begin{aligned} & (Bu_n + f, y_n - y_1)_{V^* \times V} \\ &= (A(u_n)y_n, y_n - y_1)_{V^* \times V} + (v_n - v_1, y_n - y_1)_{V^* \times V} + (v_1, y_n - y_1)_{V^* \times V} \\ &\geq (A(u_n)(y_n - y_1), y_n - y_1)_{V^* \times V} + (A(u_n)y_1, y_n - y_1)_{V^* \times V} - \|v_1\|_{V^*} \|y_n - y_1\|_V. \end{aligned}$$

By **(A7)**, we thus can find some $C_1 > 0$ such that

$$m \|y_n - y_1\|_V^2 \leq C_1 \|y_n - y_1\|_V \quad \forall n \in \mathbb{N}.$$

$\implies \{y_n\}$ is bounded in V , and we may assume there is some subsequence,

again indexed by n , such that

$$y_n \rightarrow y \text{ weakly in } V, \text{ and clearly } y \in C.$$

Denoting by $A(u)^* : V \simeq V^{**} \rightarrow V^*$ the dual operator of $A(u)$, we find $\forall v \in V$:

$$\begin{aligned} & |(A(u_n)y_n, v)_{V^* \times V} - (A(u)y, v)_{V^* \times V}| \\ & \leq |(A(u_n)y_n, v)_{V^* \times V} - (A(u)y_n, v)_{V^* \times V}| \\ & \quad + |(A(u)y_n, v)_{V^* \times V} - (A(u)y, v)_{V^* \times V}| \\ & \leq |v|_V |A(u_n)y_n - A(u)y_n|_{V^*} + |(y_n - y, A(u)^* v)_{V \times V^*}| \\ & \leq |v|_V |A(u_n) - A(u)|_{L(V, V^*)} |y_n|_V + |(y_n - y, A(u)^* v)_{V \times V^*}| \rightarrow 0 \end{aligned}$$

as $n \rightarrow \infty$.

Moreover, it follows from **(A7)** that

$$\begin{aligned} (4) \quad & \liminf_{n \rightarrow \infty} (A(u_n)y_n, y_n - y)_{V^* \times V} \\ &= \liminf_{n \rightarrow \infty} [(A(u_n)(y_n - y), y_n - y)_{V^* \times V} + (A(u_n)y, y_n - y)_{V^* \times V}] \\ &\geq \liminf_{n \rightarrow \infty} (A(u_n)y, y_n - y)_{V^* \times V} \\ &= \lim_{n \rightarrow \infty} (A(u_n)y, y_n)_{V^* \times V} - \lim_{n \rightarrow \infty} (A(u_n)y, y)_{V^* \times V} = 0. \end{aligned}$$

Now, the definition of $\partial \varphi$ implies that

$$\begin{aligned} \forall v \in V : \varphi(v) - \varphi(y_n) &\geq (Bu_n + f - A(u_n)y_n, v - u_n)_{V^* \times V} \\ \xrightarrow{v=y} (A(u_n)y_n, y_n - y)_{V^* \times V} + \varphi(y_n) &\leq (Bu_n + f, y_n - y)_{V^* \times V} + \varphi(y) \end{aligned}$$

But then, owing to the lower semicontinuity of φ ,

$$(5) \quad \limsup_{n \rightarrow \infty} (A(u_n)y_n, y - y_n)_{V^* \times V} \leq \varphi(y) - \liminf_{n \rightarrow \infty} \varphi(y_n) \leq 0$$

From (4), (5), we conclude that

$$\lim_{n \rightarrow \infty} (A(u_n)y_n, y_n - y) = 0.$$

On the other hand,

$$(A(u_n)y_n - A(u_n)y, y_n - y)_{V^* \times V} \geq m \|y_n - y\|_V^2,$$

that is, $y_n \rightarrow y$ strongly in V , for a subsequence indexed by n .

Moreover, since $A(u_n) \rightarrow A(u)$ strongly in $\mathcal{L}(V, V^*)$ and $y_n \rightarrow y$ strongly in V , we have

$$A(u_n)y_n \rightarrow A(u)y \quad \text{weakly in } V^*.$$

Finally, observe that the **maximal monotone** operator $\partial\varphi$ is **demiclosed**, i.e., we have: whenever $v_n \in \partial\varphi(z_n)$, $\|z_n - z\|_V \rightarrow 0$ and $v_n \rightarrow v$ weakly in V^* , then $v \in \partial\varphi(z)$.

Now,

$$v_n = Bu_n + f - A(u_n)y_n \rightarrow v = Bu + f - A(u)y \quad \text{weakly in } V^*.$$

Then,

$$v = Bu + f - A(u)y \in \partial\varphi(y),$$

and the pair $(y, u) \in C \times U_{ad}$ is admissible. The minimality follows again from the lower semicontinuity of J . □