

Tutorium 1

16.November 2022

1.1 Optimal control of the Laplace equation

We expand the concept of finite-dimensional optimal control theory from the lecture to infinite dimensional optimal control problems using the Laplace equation.

Let us consider the minimization problem

$$\min J(y, u) \quad \text{s.t.} \quad \begin{cases} -\Delta y = \langle u, \xi \rangle_{\mathbb{R}^{n_u}} & \text{on } \Omega \\ y = 0 & \text{on } \partial\Omega \end{cases} \quad \text{and} \quad u \in U_{ad} \quad (1.1)$$

with $\Omega \subset \mathbb{R}^2$ a close and bounded Lipschitz domain, $J : Y \times \mathbb{R}^{n_u} \rightarrow \mathbb{R}$ with

$$J(y, u) := \frac{1}{2} \int_{\Omega} |y - y_d|^2 dx + \frac{1}{2} \|u - u_d\|_{\mathbb{R}^{n_u}}^2$$

denotes the quadratic cost functional, $Y := H_0^1(\Omega)$ the state space with the following scalar product

$$\langle y, \varphi \rangle_{H_0^1(\Omega)} := \int_{\Omega} \nabla y \nabla \varphi dx,$$

$U_{ad} \subset \mathbb{R}^{n_u}$ is the set of admissible control parameters, $y_d \in Y$ the desired state, $u_d \in \mathbb{R}^{n_u}$ the desired control and $\xi \in L^2(\Omega, \mathbb{R}^{n_u})$.

We look for a control vector $u \in \mathbb{R}^{n_u}$ and a state $y \in Y$ which solve (1.1).

Remark 1.1.1. For $u \in U_{ad}$ the function $y \in Y = H_0^1(\Omega)$ is called a weak solution of

$$\begin{cases} -\Delta y = \langle u, \xi \rangle_{\mathbb{R}^{n_u}} & \text{on } \Omega \\ y = 0 & \text{on } \partial\Omega \end{cases} \quad (1.2)$$

if the weak formulation

$$a(y, \varphi) := \int_{\Omega} \nabla y \nabla \varphi dx = \int_{\Omega} \langle u, \xi \rangle_{\mathbb{R}^{n_u}} \varphi dx =: F(u, \varphi) \quad (1.3)$$

holds for all $\varphi \in Y$ with $a : Y \times Y \rightarrow \mathbb{R}$ and $F : U_{ad} \times Y \rightarrow \mathbb{R}$. ◇

Theorem 1.1.2 (Lax and Milgram). *Let Y be a Hilbert space. Suppose that for the bilinear form $a : Y \times Y \rightarrow \mathbb{R}$ there exist $c, \alpha > 0$ with*

1. $|a(y, \varphi)| \leq \alpha \|y\|_Y \|\varphi\|_Y$ (boundedness)
2. $c \|y\|_Y^2 \leq a(y, y)$ (V-ellipticity).

Then for every $F \in Y^*$ there exists a unique $y \in Y$ which solves

$$a(y, \varphi) = F(\varphi) \quad \text{for all } \varphi \in Y$$

and there exists $c_a > 0$ with

$$\|y\|_Y \leq c_a \|F\|_{Y^*}.$$

The proof can be found in [1].

Theorem 1.1.3. *If Ω is a close and bounded Lipschitz domain, then for every $u \in U_{ad}$ and $\xi \in L^2(\Omega, \mathbb{R}^{n_u})$ there exists an unique weak solution $y \in Y$ of (1.2) with*

$$\|y\|_{H^1(\Omega)} \leq c\|u\|_2\|\xi\|_{L^2(\Omega, \mathbb{R}^{n_u})}. \quad (1.4)$$

Proof. Exercise. (Use Lax and Milgram and Poincaré inequality.) \square

Remark 1.1.4. Induced by Theorem 1.1.3 we define the solution operator $\mathcal{S} : U_{ad} \rightarrow Y$ with $u \mapsto y(u)$. With this, we can define the reduced problem

$$\min \hat{J}(u) := J(\mathcal{S}(u), u) \text{ s.t. } u \in U_{ad}. \quad (1.5)$$

\diamond

Theorem 1.1.5. *Let $\Omega \subset \mathbb{R}^2$ be a closed and bounded Lipschitz domain, $Y := H_0^1(\Omega)$, $y_d \in Y$, $U_{ad} \subset \mathbb{R}^{n_u}$ a closed, nonempty, bounded and convex set, $\xi \in L^2(\Omega, \mathbb{R}^{n_u})$ and assume that the solution operator $\mathcal{S} : U_{ad} \rightarrow Y$ is linear and continuous. Then (1.5) has an optimal control \bar{u} with an optimal state \bar{y} . If \bar{u} is strictly convex, the optimal control \bar{u} is unique.*

Proof. Exercise. \square

Lemma 1.1.6. Assume that the conditions from Theorem 1.1.5 hold. Then the reduced cost functional \hat{J} is continuously differentiable on U_{ad} and it holds

$$\nabla \hat{J}(u) = \left(\int_{\Omega} \xi_1 p \, dx, \dots, \int_{\Omega} \xi_n p \, dx \right)^{\top} + (u - u_d) \quad (1.6)$$

and $p \in Y$ solves the adjoint equation

$$\begin{cases} -\Delta p = \mathcal{S}(u) - y_d & \text{on } \Omega \\ p = 0 & \text{on } \partial\Omega \end{cases} \quad (1.7)$$

p is called the adjoint variable, associated with (y, u) .

Proof. Exercise. Use the representation of (1.6) to show

$$\lim_{\|w\|_{\mathbb{R}^{n_u}} \rightarrow 0} \frac{\hat{J}(u+w) - \hat{J}(u) - \langle \nabla \hat{J}(u), w \rangle}{\|w\|_{\mathbb{R}^{n_u}}} = 0.$$

\square

Theorem 1.1.7. *Assume that the conditions of Theorem 1.1.5 hold and assume that \bar{u} solves the reduced problem (1.5). Then the variational inequality*

$$\langle \nabla \hat{J}(\bar{u}), u - \bar{u} \rangle_{\mathbb{R}^{n_u}} \geq 0 \text{ for all } u \in U_{ad} \quad (1.8)$$

holds.

If $\tilde{u} \in U_{ad}$ is a solution of the variational inequality (1.8) and \hat{J} is convex, then \tilde{u} solves (1.5).

Proof. Exercise. (Same arguments as in Optimization II). \square

Remark 1.1.8. Utilizing the adjoint variable for the representation of the gradient of the reduced cost functional \hat{J} the variational inequality (1.8) is equivalent to

$$0 \leq \langle \nabla \hat{J}(\bar{u}), u - \bar{u} \rangle_{\mathbb{R}^{n_u}} = \int_{\Omega} \langle u - \bar{u}, \xi \rangle_{\mathbb{R}^{n_u}} p \, dx + \langle \bar{u} - u_d, u - \bar{u} \rangle_{\mathbb{R}^{n_u}} \quad \text{for all } u \in U_{ad}.$$

Now we can formulate an optimality system for the unknown variables \bar{u} , \bar{y} and \bar{p} :

$$\begin{aligned} a(\bar{y}, \varphi) &= F(\bar{u}, \varphi) && \text{for all } \varphi \in Y, \\ a(\bar{p}, \varphi) &= F_p(\bar{y}, \varphi) && \text{for all } \varphi \in Y, \\ 0 &\leq \int_{\Omega} \langle u - \bar{u}, \xi \rangle_{\mathbb{R}^{n_u}} p \, dx + \langle \bar{u} - u_d, u - \bar{u} \rangle_{\mathbb{R}^{n_u}} && \text{for all } u \in U_{ad} \end{aligned} \quad (1.9)$$

with

$$F_p(\bar{y}, \varphi) := \int_{\Omega} (\bar{y} - y_d) \varphi \, dx.$$

Every solution (\bar{y}, \bar{u}) to (1.1) must satisfy, together with the adjoint variable \bar{p} , the necessary conditions (1.9). Due to the convexity in this setting, the conditions are sufficient. \diamond

Bibliography

- [1] F. Tröltzsch. *Optimal Control of Partial Differential Equations: Theory, Methods and Applications*, volume 112 of *Graduate Studies in Mathematics*. American Mathematical Society, 2010.