

Optimal control of parabolic equations: a spectral calculus based approach

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The problem

Initial condition or starting optimal control: solve the following problem

$$\min_{u \in \mathcal{H}} \{ J(u) : \|y(T) - y^*\| \leq \varepsilon \},$$

where

$$\begin{cases} \dot{y}(t) + Ay(t) = f(t) & \text{for } 0 \leq t \leq T, \\ y(0) = u, \end{cases}$$

$$J(u) = \frac{\alpha}{2} \|u\|^2 + \frac{1}{2} \int_0^T \beta(t) \|y(t) - w(t)\|^2 dt.$$

The problem

Here we assume:

- ▶ $A \geq \kappa$ selfadjoint operator on a Hilbert space \mathcal{H} , given as a partial differential operator on the domain $\Omega \subset \mathbb{R}^d$,
- ▶ $f \in L^2((0, T); \mathcal{H})$.

Parameters are:

- ▶ y^* the target state,
- ▶ $\varepsilon > 0$ the tolerance,
- ▶ $\alpha > 0$ weight,
- ▶ $\beta \in L^\infty((0, T); [0, \infty))$ weight, and
- ▶ $w \in L^2((0, T); \mathcal{H})$ the desired trajectory of the system.

Solving the problem

The problem is ill-posed, thus numerically challenging.

We use *first optimize, then discretize* approach. We want

- ▶ to construct a closed form expression of the solution of the problem, and
- ▶ use a discretization scheme which is tailored to the particular problem.

The solution

The solution u^{opt} of the problem is given by

$$u^{\text{opt}} = (\mu^\varepsilon S_{2T} + \Psi)^{-1}(\mu^\varepsilon S_T y^* + \psi),$$

$$\Psi = \alpha I + \int_0^T \beta(t) S_{2t} dt, \quad \psi = \int_0^T \beta(t) S_t w(t) dt.$$

Here

- ▶ $\{S_t\}$ is the semigroup generated by $-A$
- ▶ $\mu^\varepsilon \geq 0$ is
 - ▶ the unique solution of $\Phi(\mu) = \varepsilon$ if $\varepsilon < \|\Psi^{-1} S_T \psi - y^*\|$,
 - ▶ and zero otherwise.

The solution

- ▶ the function $\Phi: [0, \infty) \rightarrow [0, \infty)$ is defined by

$$\Phi(\mu) = \|y^* - (\mu S_{2T} + \Psi)^{-1}(\mu S_{2T}y^* + S_T\psi)\|$$

and is a decreasing function.

- ▶ The solution of the unconstrained problem ($\varepsilon = \infty$) is given by the same formula with $\mu^\varepsilon = 0$.

Practical solution

- ▶ Let $\phi(\mu) = y^* - (\mu S_{2T} + \Psi)^{-1}(\mu S_{2T}y^* + S_T\psi)$, hence $\Phi(\mu) = \|\phi(\mu)\|$.
Then $\phi(\mu) = y^* - x$, where x is the solution of the equation

$$(\mu S_{2T} + \Psi) x = \mu S_{2T}y^* + S_T\psi,$$

hence the calculation of $\Phi(\mu)$ reduces to solving a linear equation.

- ▶ Note also that the optimal initial state u^{opt} is the solution of the equation

$$(\mu^\varepsilon S_{2T} + \Psi) x = \mu^\varepsilon S_Ty^* + \psi.$$

How to calculate Ψ

To calculate Ψ we use the fact that it can be written as a function of A by using

$$\int_0^T \beta(t) \mathcal{S}_{2t} dt = \int_{-\infty}^{\infty} \int_0^T \beta(t) e^{-2t\lambda} dt dE(\lambda) = \tilde{\beta}_0(A),$$

where $\tilde{\beta}_0$ is a function given by $\tilde{\beta}_0(\lambda) = \int_0^T \beta(t) e^{-2t\lambda} dt$ and E is the spectral measure of A .

How to calculate ψ

We can always find a good approximant for ψ . Let $\tilde{w}(t) = \sum_{i=1}^N w_i \chi_{[t_{i-1}, t_i]}$ be an approximation of w , where $0 = t_0 < t_1 < \dots < t_N = T$, $w_i \in \mathcal{H}$, $i = 1, \dots, N$, and χ_S is the characteristic function of the set S . Then

$$\tilde{\psi} = \sum_{i=1}^N \tilde{\beta}_i(A) w_i, \text{ where } \tilde{\beta}_i(\lambda) = \int_{t_{i-1}}^{t_i} \beta(t) e^{-t\lambda} dt,$$

is an approximation of ψ .

How to *really* calculate Ψ and ψ : 1/5

Both Ψ and ψ can be represented as functions of operator A , so we want to *efficiently* calculate $g(A)$ for a given function g .

If g has the form $u_0(x) + u_1(x)e^{ax}$, where u_0, u_1 are rational functions, we use the fact that such functions can be very efficiently approximated: there exists a rational function $r_{n,n+k}$ such that

$$\|g - r_{n,n+k}\|_{\infty} \leq C \cdot 9.28903^{-n}$$

How to *really* calculate Ψ and ψ : 2/5

We use `rkfit` algorithm to find a rational function

$$r_{RK}(x) = r_0 + \frac{r_1}{x - \zeta_1} + \dots + \frac{r_d}{x - \zeta_d}$$

such that

$$\|r_{RK} - g\|_{\infty} \leq \text{tol} \|g\|_2.$$

Now the operator $r_{RK}(A) := r_0 I + \sum_{i=1}^d r_i (A - \zeta_i)^{-1}$ satisfies

$$\|g(A) - r_{RK}(A)\|_{L(\mathcal{H})} \leq \text{tol} \|g\|_2.$$

So it remains to approximate the resolvent of the operator A .

How to *really* calculate Ψ and ψ : 3/5

We approximate the (action of the) resolvent by selecting a finite dimensional subspace $\mathcal{V}_h \subset \text{Dom}(A^{1/2})$ and then forming the Galerkin projection of A onto \mathcal{V}_h . The Galerkin projection $A_h : \mathcal{V}_h \rightarrow \mathcal{V}_h$ is given by the formula (assuming $A \leq 0$)

$$A_h = (A^{1/2}P_h)^*(A^{1/2}P_h),$$

where P_h is the orthogonal projection onto \mathcal{V}_h .

How to *really* calculate Ψ and ψ : 4/5

Let \mathcal{V}_h be the space of piece-wise linear, for a given triangular tessellation of Ω , and continuous functions on Ω . Let

- ▶ h be the maximal diameter of a triangle in the chosen tessellation, and
- ▶ ν be a parameter depending on the regularity of the functions in $\text{Dom}(A)$.

The resolvent estimate for A using the Galerkin projection A_h for $h < h_0$ and $v \in V_h$ is

$$\|(A - z)^{-1}v - (A_h - z)^{-1}v\|_2 \leq Ch^{2\nu} \|v\|_2.$$

Here h_0 denotes the minimal level of refinement from which the estimate holds.

How to *really* calculate Ψ and ψ : 5/5

Finally, let $g(x) = u_0(x) + u_1(x)e^{ax}$ be a perturbed exponential function. For a given rational function r_{RK} and $v \in V_h$ we have the estimate

$$\|g(A)v - r_{RK}(A_h)v\|_2 \leq \|g - r_{RK}\|_\infty \|v\|_2 + dCh^{2\nu} \|v\|_2.$$

By choosing suitable r_{RK} and h , this estimate ensures a good approximation of $g(A)v$ based on a finite dimensional approximation of the operator A .

The theorem from the following slide ensures that the numerical solution is a good approximation of the solution of the optimal problem.

Sensitivity of the problem

Let us perturb all the parameters of the problem with perturbations $< \nu$, in respective norms, such that that the perturbed problem has the same structure (A still selfadjoint etc.). For the perturbation of the operator A , we allow δA which is relatively bounded with respect to A . Then we have

Theorem

For small enough $\nu > 0$ the optimal solutions of the original and the perturbed problem differ (in norm) by $< C\nu$ where C does not depend on ν .

An example

Let $A < 0$, $f = 0$, $\beta = \chi_{[T/3, 2T/3]}$ and assume that w does not depend on time.

Then

$$\Psi = \alpha I + \frac{1}{2} A^{-1} S_{2T/3} (I - S_{2T/3}),$$

$$\psi = A^{-1} S_{T/3} (I - S_{T/3}) w.$$

Hence

$$\Psi = \alpha I + \int_{\mathbb{R}} \underbrace{e^{\lambda 2T/3} (1/\lambda - e^{\lambda 2T/3} / \lambda)}_{:=g(\lambda)} dE(\lambda).$$

An example

The value of the function $\Phi(\mu)$ is computed using the rational approximation and the spectral calculus

$$\begin{aligned}
 (\mu S_{2T} + \Psi)^{-1} \mu S_{2T} y^* &= \int_{-\infty}^0 \frac{\mu e^{2T\lambda}}{\underbrace{\mu e^{2T\lambda} + \alpha + (1/\lambda - e^{\lambda 2T/3}/\lambda)}_{:=g(\lambda)}} dE(\lambda) y^* \\
 &\approx r_0 y^* + \sum_{i=1}^{18} r_i (A - \zeta_i)^{-1} y^*,
 \end{aligned}$$

where $\text{tol} = 10^{-15}$.

The solution of $\Phi(\mu) = \varepsilon$ is obtain using the Brent method.

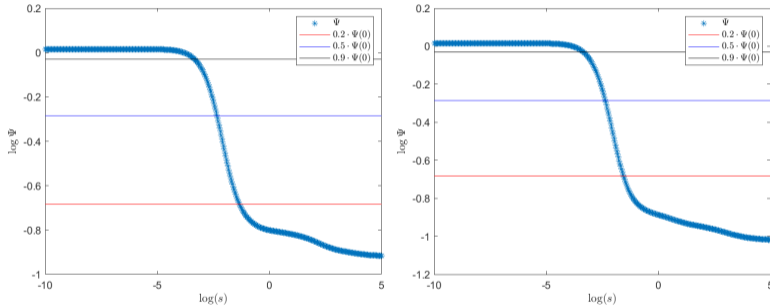
1D heat equation

Let

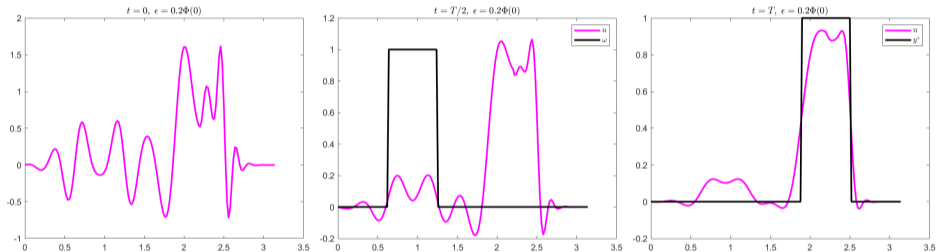
- ▶ $A = -\partial_x((1 + a\chi_{[2.2,\pi]})\partial_x)$ on $\Omega = [0, \pi]$, $a \leq 0$,
- ▶ $\alpha = 10^{-4}$,
- ▶ $T = 0.01$,
- ▶ $w = \chi_{[\pi/5, 2\pi/5]}$,
- ▶ $y^* = \chi_{[3\pi/5, 4\pi/5]}$,
- ▶ $\varepsilon = [0.2, 0.5, 0.9]\Phi(0)$.

$\Phi(0)$ is the distance between the target y^* and $y(T)$ in the case $y(0) = u^{\min}$, u^{\min} being the solution of the corresponding unconstrained problem.

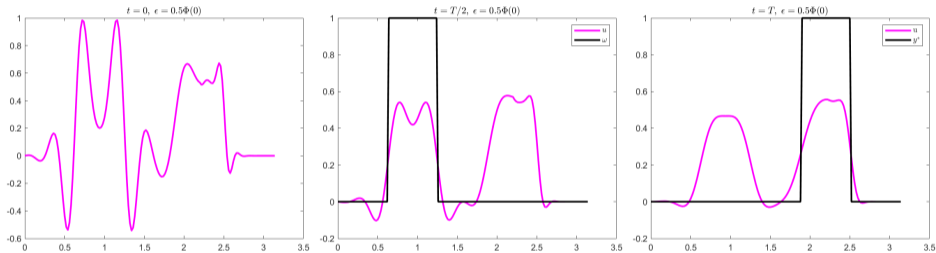
$a = 0$ and $a = -0.8$, corresponding μ_ε



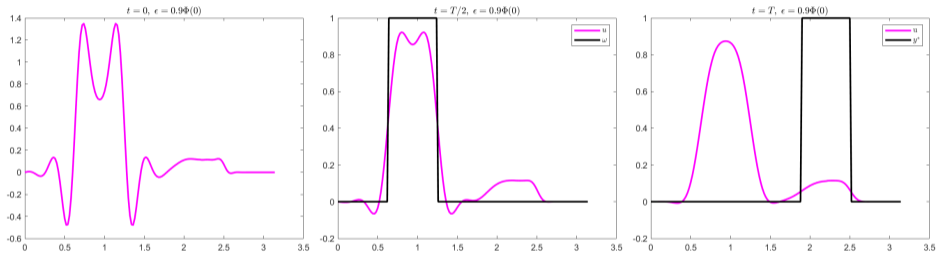
$$a = -0.8, \varepsilon = 0.2\Phi(0)$$



$$a = -0.8, \varepsilon = 0.5\Phi(0)$$



$$a = -0.8, \varepsilon = 0.9\Phi(0)$$

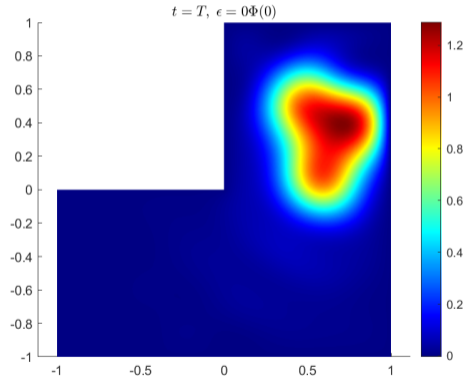


2D heat equation on irregular domain

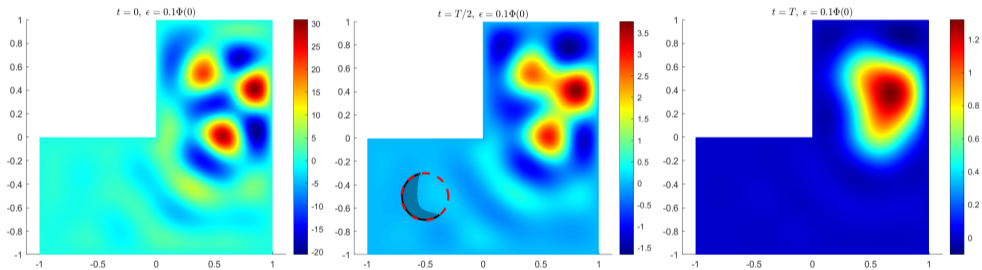
Let

- ▶ A be Dirichlet Laplacian on $\Omega = [-1, 1]^2 \setminus ([-1, 0] \times [0, 1])$,
- ▶ $\alpha = 10^{-4}$,
- ▶ $T = 0.01$,
- ▶ $w(x) = \chi_{\|x-x_0\|_1 \leq 0.2}$,
- ▶ $y^*(x) = e^{-20\|x-x_1\|^2} + e^{-20\|x-x_2\|^2} + e^{-30\|x-x_3\|^2}$, with $x_0 = (-0.5, -0.5)$, $x_1 = (0.5, 0.5)$, $x_2 = (0.6, 0.1)$ and $x_3 = (0.8, 0.4)$.

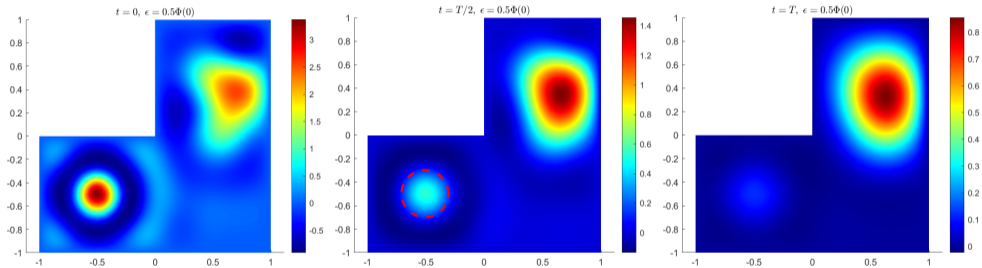
The prescribed final target y^* .



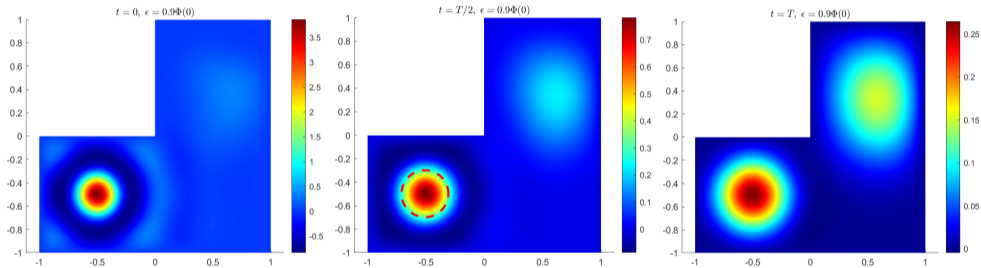
$$\varepsilon = 0.1\Phi(0)$$



$$\varepsilon = 0.5\Phi(0)$$



$$\varepsilon = 0.9\Phi(0)$$



Non-homogeneous boundary condition

Suppose we have

$$\begin{cases} \dot{y}(t) + Ly(t) = 0 & \text{for } 0 \leq t \leq T, \\ Gy(t) = g(t), \\ y(0) = u, \end{cases}$$

where (L, G) form a well-posed *boundary control system* on \mathcal{U} , \mathcal{Z} and \mathcal{X} .

The optimal control problem is given by

$$J(u) = \frac{\alpha}{2} \|u\|^2 + \frac{1}{2} \int_0^T \beta(t) \|y(t) - w(t)\|_{\mathcal{Z}}^2 dt,$$

$$\min_{u \in \mathcal{Z}} \{J(u) : \|y(T) - y^*\|_{\mathcal{Z}} \leq \varepsilon\}.$$

Non-homogeneous b. c.

We can also solve optimal control problems of this type.

Our approach is to lift the boundary condition to obtain

$$\dot{y} + \hat{A}y = Bg(t), \quad (*)$$

where $\hat{A} = (L|_{\text{Ker } G})_{-1}$ and $B \in \mathcal{L}(U, \mathcal{X}_{-1})$.

One cannot directly apply the previous result as the the constraints are given in $\mathcal{Z} \subset \mathcal{X}$ and the equation (*) lives in \mathcal{X}_{-1} .

Recapitulation

- ▶ We constructed and implemented a numerical algorithm for a constrained optimal control problem.
- ▶ For the numerical implementation we explored efficient Krylov subspace techniques that allowed us to approximate a complex function of an operator by a series of linear problems. We provided a-priori estimates for the approximation that are not sensitive to any particular spatial discretization, and neither to a matrix representation of the operator A .
- ▶ We provided a complete quantified sensitivity analysis of the solution with respect to all the data entering the problem.
- ▶ This approach can be generalised to other optimal control problems.