

# Greedy Algorithm for Parameter Dependent Operator Lyapunov Equations

Martin Lazar, Jerome Weston

University of Dubrovnik

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# Background and Motivation

Consider a family of parameter-dependent linear control problems

$$\begin{cases} \frac{d}{dt}x_\nu(t) &= A_\nu x_\nu(t) + B_\nu u_\nu(t), & 0 \leq t \leq T \\ x_\nu(0) &= x_{0,\nu} \end{cases} \quad (1)$$

where parameter  $\nu \in \mathcal{N} \subset \mathbb{R}^n$  and  $\mathcal{N}$  is compact.

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where parameter  $\nu \in \mathcal{N} \subset \mathbb{R}^n$  and  $\mathcal{N}$  is compact.

If  $x_1$  is a desired final state, for  $\epsilon > 0$ , we wish to design  $u$  such that  $\|x_1 - x(T)\| < \epsilon$ .

# Background and Motivation

One way to construct  $u$  in order to achieve this result is to set  $u = B^* \phi$ , where  $\phi$  is the solution to the adjoint problem

$$\begin{cases} -\frac{d}{dt}\phi(t) = A^*\phi(t), & 0 \leq t \leq T \\ \phi(T) = \phi_0 \end{cases} \quad (2)$$

where  $\phi_0$  is the solution to the following associated optimal control problem

$$\min_{z \in R_T \cap \partial B_\epsilon(x^1)} J(z) = \left\{ \frac{1}{2} \|B^* \phi\|^2 : \phi(T) = \Lambda_T^{-1}(z - e^{TA}x_0) \right\}. \quad (3)$$

# Background and Motivation

We compute  $\Lambda_T$  as

$$\Lambda_T = \int_0^T e^{tA} B B^* e^{tA^*} dt.$$

It is known that for sufficiently large  $T$  and  $A$  stable, the controllability Gramian can be approximated by

$$\Lambda_\infty = \int_0^\infty e^{tA} B B^* e^{tA^*} dt.$$

This approximation is the solution to the (algebraic) operator Lyapunov equation (OLE)

$$AP + PA^* = -BB^*.$$

## Background and Motivation

We consider the related family of parameter-dependent operator Lyapunov equations (OLE)

$$A_\nu P_\nu + P_\nu A_\nu^* = -B_\nu B_\nu$$

and we wish to establish an efficient greedy algorithm for constructing an approximate controllability Gramian to use to build the necessary optimal control with minimum  $L^2$ -norm.

## Framework: Required Spaces

For Hilbert space  $X$  and unbounded linear operator  $A$  let  $X_1$  be the  $D(A)$  equipped with the norm

$$\|x\|_1 = \|(\beta I - A)x\|, \quad x \in D(A), \beta \in \rho(A)$$

and  $X_{-1}$  be  $X$  completed with the norm

$$\|x\|_{-1} = \|(\beta I - A)^{-1}x\|, \quad x \in X, \beta \in \rho(A).$$

Similar spaces  $X_1^d$  and  $X_{-1}^d$  can be made by replacing  $A$  with its adjoint.



# Framework: Assumptions on $\mathcal{A} = \{A_\nu\}$ , $\mathcal{A}^* = \{A_\nu^*\}$

For each operator  $A_\nu : X_{1,\nu} \rightarrow X$ :

- 1)  $A_\nu$  is closed.
- 2)  $A_\nu$  is stable.
- 3)  $A_\nu$  is sectorial<sup>1</sup>, i.e. there exists constants  $\omega \in \mathbb{R}$ ,  $\theta \in (\frac{\pi}{2}, \pi)$ ,  $M > 0$  such that

$$\begin{cases} \rho(A_\nu) \supset \mathbf{S}_{\theta,\omega} = \{\lambda \in \mathbb{C} : \lambda \neq \omega, |\arg(\lambda - \omega)| < \theta\}, \\ \|(\lambda I - A_\nu)^{-1}\|_{\mathcal{L}(X)} \leq \frac{M}{|\lambda - \omega|}, \quad \forall \lambda \in \mathbf{S}_{\theta,\omega}. \end{cases}$$

- 4)  $D(A_\nu)$  is dense in  $X$ .
- 5)  $D(A_{\nu_1}) = D(A_{\nu_2})$  and  $D(A_{\nu_1}^*) = D(A_{\nu_2}^*)$  for  $\nu_1, \nu_2 \in \mathcal{N}$ .

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<sup>1</sup>Lunardi, A, *Analytic Semigroups and Optimal Regularity in Parabolic Problems*,

## Framework: Consequences

### Lemma

*If  $A_\nu$  has properties 1-4, then  $\mathbb{T} = \{e^{tA_\nu}\}_{t \geq 0}$  is a strongly continuous, strongly stable, and analytic semigroup.*

## Framework: Assumptions on $\mathcal{B} = \{B_\nu\}$

Each operator  $B_\nu : U \rightarrow X_{-1,\nu}$  is an **infinite-time admissible** control operator for semigroup  $\mathbb{T}$ , i.e. for every  $u \in L^2([0, \infty), U)$ , the mapping  $\tau \rightarrow \Phi_\tau u$  is bounded in  $X$ . The map  $\Phi$  is defined as

$$\Phi_\tau u := \int_0^\tau \mathbb{T}_t B_\nu u(t) dt.$$

## Framework: Sufficient Conditions

### Theorem (Hansen-Weiss (1997))

Let  $\mathbb{T}$  be a strongly continuous semigroup on the Hilbert space  $X$ , with generator  $A$ . If  $B$  is an infinite-time admissible control operator for  $\mathbb{T}$ , then there exists an operator  $P \in \mathcal{L}(X)$  such that

1) For any  $x \in X_1^d$ ,  $Px = \lim_{\tau \rightarrow \infty} \int_0^\tau \mathbb{T}_t B B^* \mathbb{T}_t^* x \, dt$ .

2)  $P \geq 0$  satisfies (OLE) with  $B B^*$ ,  $AP + PA^* \in \mathcal{L}(X_1^d, X_{-1})$ .

Moreover, if  $\mathbb{T}^*$  is strongly stable, then  $P$  is the unique self-adjoint solution of (OLE)

## Greedy Algorithm: Idea

Approximate (compact) solution manifold  $\mathcal{P}$  in a Banach space  $\mathbb{P}$  by a sequence of finite dimensional subspaces  $\mathcal{P}_k$  of dimension  $k$ .

Offline procedure generates approximation subspace within given precision error; Online routine calculates approximations for any element in  $\mathcal{P}$ .

# Greedy Algorithm: Offline Outline

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## Algorithm 1 (Weak) Greedy Algorithm

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**Initialize:** Fix a constant  $\gamma \in (0, 1]$  and  $\epsilon > 0$ ;

1: In the first step, choose  $P_1 \in \mathcal{P}$  such that

$$\|P_1\|_{\mathbb{P}} \geq \gamma \max_{P \in \mathcal{P}} \|P\|_{\mathbb{P}}.$$

2: At the general step, having found  $P_1, \dots, P_k$ , denote

$$\mathcal{P}_k = \text{span}\{P_1, \dots, P_k\} \text{ and } \sigma_k(\mathcal{P}) := \max_{P \in \mathcal{P}} \text{dist}(P, \mathcal{P}_k);$$

3: **repeat**

4:     choose  $P_{k+1}$  such that

$$\text{dist}(P_{k+1}, \mathcal{P}_k) \geq \gamma \sigma_k(\mathcal{P});$$

5: **until**  $\sigma_k(\mathcal{P}) < \epsilon$ ;

# Greedy Algorithm: Distance Approximation

Theorem ((DeVore, Petrova, Wojtaszczyk - 2013))

*For the weak greedy algorithm with constant  $\gamma$  in a Hilbert space  $\mathbb{P}$ , we have the following: If the compact set  $\mathcal{P}$  is such that, for some  $\alpha > 0$  and  $C_0 > 0$*

$$d_k(\mathcal{P}) \leq C_0 n^{-\alpha}, \quad n \in \mathbb{N},$$

*then*

$$\sigma_k(\mathcal{P}) \leq C_1 n^{-\alpha}, \quad n \in \mathbb{N},$$

*where  $C_1 := \gamma^{-2} 2^{5\alpha+1} C_0$ .*

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For our purposes, we choose  $d_k(\mathcal{P})$  as the *Kolmogorov  $n$ -width* of  $\mathcal{P}$  defined as

$$d_k(\mathcal{P}) := \inf_{\dim Y = k} \sup_{P \in \mathcal{P}} \inf_{\Gamma \in Y} \|P - \Gamma\|_{\mathbb{P}}$$



# Implementation: Residual Analysis

In computation we take

$$d_k(\mathcal{P}) = \max_{\nu \in \tilde{\mathcal{N}}} \inf_{P \in \mathcal{P}_k} \|R_\nu(P)\|_{\mathcal{L}(X_{1,\nu}^d, X_{-1,\nu})}$$

where

- $\tilde{\mathcal{N}}$  is a discrete subset of  $\mathcal{N}$  such that for all  $\nu \in \mathcal{N}$ ,  $\text{dist}(\nu, \tilde{\mathcal{N}}) < \delta$  for some  $\delta > 0$ .
- $R_\nu(P) := A_\nu P - PA_\nu + B_\nu B_\nu^*$

## Implementation: Residual Analysis

We can easily show that  $R_\nu(P_j) \in \mathcal{L}(X_{1,\nu}^d, X_{-1,\nu})$  for each  $\nu$ . As mentioned previously, the domains of all operators in  $\mathcal{A}^*$  are the same. Thus  $X_{1,\nu_1}^d$  and  $X_{1,\nu_2}^d$  differ only with respect to their norms.

If the respective norms are equivalent, we drop parameter dependence on  $\mathcal{L}(X_{1,\nu}^d, X_{-1,\nu})$

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### Theorem

$$\|R\|_{\mathcal{L}(X_1^d, X_{-1})} \sim \|P - P_\nu\|_{\mathcal{L}(X_1^d, X)}$$

## Implementation: Online Routine

For specific  $\nu \in \mathcal{N}$

- Project  $-B_\nu B_\nu^*$  to  $L_\nu(\mathcal{P}_k) := \text{span} \{L_\nu(P_j)\}_{j=1}^k$ .

Here  $L_\nu(P_j) = A_\nu P_j + P_j A_\nu^*$ .

- Define approximating Gramian operator as  $\Lambda_{\nu,a} := \Sigma \alpha_j P_j$

The corresponding approximate control is computed as before.

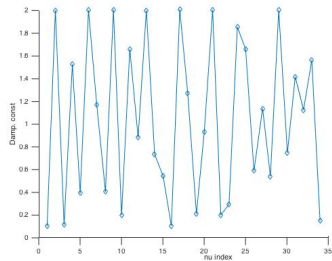
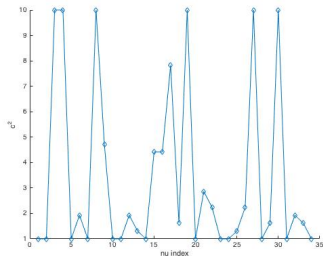
## Example: 1D Damped Wave Equation

$$\left\{ \begin{array}{ll} \frac{\partial^2}{\partial t^2} z - c^2 \frac{\partial^2}{\partial x^2} z = -\beta \frac{\partial}{\partial t} z & \text{in } (0, 1) \times (0, T), \\ z(0, t) = 0, & z(1, t) = u_\nu(t) \\ z(x, 0) = 0, & \frac{\partial}{\partial t} z(x, 0) = 0. \end{array} \right. \quad (4)$$

Take final state  $z_1 = \sin(\pi x)$ .

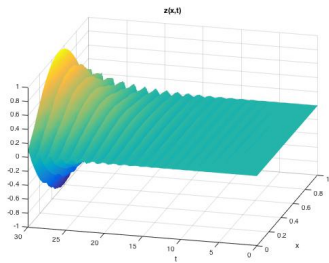
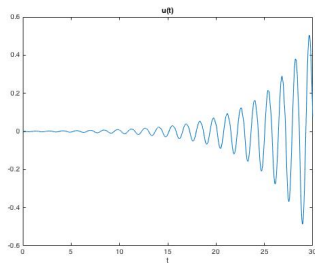
Here  $N = 40$  with  $\epsilon = 0.01$ , and the discretization of  $\mathcal{N} = [1, 10] \times [0.1, 2]$  of  $l = 1200$ .

# Example: Plots



**Figure:** Distribution of selected parameter values for a) velocity squared and b) dampening constant.

# Example: Plots



**Figure:** Evolution of a) the approximate control and b) the solution of semi-discretized example problem governed by the approximate control for  $\nu = (2, 0.4)$  in time  $T = 30$ .

## References

- DeVore, R., G. Petrova, and P. Wojtaszczyk, *Greedy Algorithms for Reduced Bases in Banach Spaces*, *Constr. Approx.*, 37 (2013) 455-466.
- Hansen, S. and G. Weiss, *New results on the operator Carleson measure criterion*, *IMA J. Math. Control Inform.*, 14 (1997), pp. 3-32. Distributed parameter systems: analysis, synthesis and applications, Part 1.
- Lunardi, A, *Analytic Semigroups and Optimal Regularity in Parabolic Problems*, Birkhauser-Verlag, Switzerland, 1995
- Tucsnak, M. and G. Weiss, *Observation and Control for Operator Semigroups*, Birkhauser-Verlag AG, Berlin, 2009.