

# Approximation of nonlinear optimal control problems in infinite dimension

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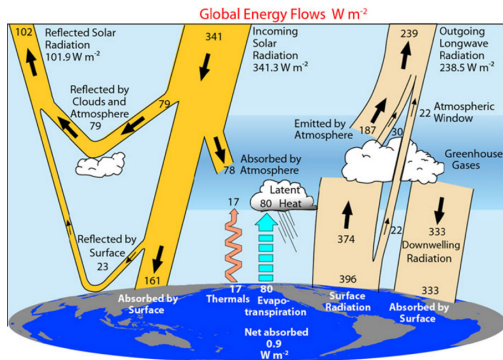
# Outline

- 1 A motivating example: Optimal control of climate?
- 2 Functional framework
- 3 Galerkin approximation: convergence of value functions
- 4 Applications

# A motivating example: Optimal control of climate?

*“Probably intervention in atmospheric and climate matters will come in a few decades, and will unfold on a scale difficult to imagine at present.”*

(John von Neumann, “Can we survive technology?”, *Fortune* (June, 1955).)



The global annual mean Earth's energy budget for the Mar 2000 to May 2004 period. (K. E. Trenberth, J. T. Fasullo, and J. Kiehl, *Earth's Global Energy Budget*, **AMS**, March 2009.) [See also “Heat stored in the Earth system 1960–2020: where does the energy go?”, *ESSD* article, 2023]

The consideration of climate engineering (a.k.a. geoengineering) is raising in the scientific community. (See e.g. National Research Council, *Climate intervention: Carbon dioxide removal and reliable sequestration*, National Academies Press, 2015.)

# Functional Framework

# Class of Equations considered and Functional Framework

We consider the following class of controlled **nonlinear evolution equations**:

$$\frac{dy}{ds} = Ly + F(y) + \mathfrak{C}(u(s)), \quad s \in (t, T], \quad u \in \mathcal{U}_{ad}[t, T],$$

- $(\mathcal{H}, \|\cdot\|)$  a separable Hilbert space;
- $L : D(L) \subset \mathcal{H} \rightarrow \mathcal{H}$  a linear operator generating a  $C_0$ -semigroup,  $\{T(s)\}_{s \geq 0}$ ;
- $F : \mathcal{H} \rightarrow \mathcal{H}$  is locally Lipschitz;
- $\mathfrak{C} : V \rightarrow \mathcal{H}$  is allowed to be nonlinear and  $V$  is a separable Hilbert space;
- The set of admissible controls is taken to be:

$$\mathcal{U}_{ad} := \{f \in L^q(0, T; V) : f(s) \in U \text{ for a.e. } s \in [0, T]\}, \quad q \geq 1,$$

with  $U$  a bounded set in  $V$ .

- $u \in \mathcal{U}_{ad}[t, T] := \{v|_{[t, T]} : v \in \mathcal{U}_{ad}\}$ .

**Note:** This framework covers a range of equations, including **semilinear parabolic PDEs** and **delay differential equations**; and allows for **a broad class of nonlinear control laws**.

# The optimal control problem

**Cost functional:** For each  $(t, x)$  in  $[0, T] \times \mathcal{H}$ , we consider the following type of cost functional  $J_{t,x} : \mathcal{U}_{ad}[t, T] \rightarrow \mathbb{R}^+$ :

$$J_{t,x}(u) := \int_t^T [\mathcal{G}(y_{t,x}(s, u(s))) + \mathcal{E}(u(s))] ds.$$

- $\mathcal{G} : \mathcal{H} \rightarrow \mathbb{R}^+$  is locally Lipschitz;
- $\mathcal{E} : V \rightarrow \mathbb{R}^+$  is continuous;
- $y_{t,x}(\cdot, u)$  denotes the mild solution of the state equation with  $y(t) = x$ .

We consider the following family of **optimal control problems**:

$$\begin{aligned} \min J_{t,x}(u) \quad \text{subject to} \quad & (y, u) \in L^2(t, T; \mathcal{H}) \times \mathcal{U}_{ad}[t, T] \quad \text{solves} \\ & \begin{cases} \frac{dy}{ds} = L_\lambda y + F(y) + \mathfrak{C}(u(s)), & s \in (t, T], \\ y(t) = x \in \mathcal{H}. \end{cases} \end{aligned}$$

# Galerkin Approximation: Convergence of Value Functions

## Galerkin approximation

- Let  $\{\mathcal{H}_N : N \in \mathbb{N}^*\}$  be a sequence of finite-dimensional subspaces of  $\mathcal{H}$  associated with *orthogonal projectors*  $\Pi_N : \mathcal{H} \rightarrow \mathcal{H}_N$ , such that

$$\|(\Pi_N - \text{Id})x\| \xrightarrow{N \rightarrow \infty} 0, \quad \forall x \in \mathcal{H}, \quad \mathcal{H}_N \subset D(L), \text{ for all } N.$$

- The corresponding Galerkin approximation of the state equation reads:

$$\frac{dy_N}{ds} = L_N y_N + \Pi_N F(y_N) + \Pi_N \mathfrak{C}(u(s)), \quad s \in (t, T], \quad u \in \mathcal{U}_{ad}[t, T], \quad (4.1)$$

$$y_N(t) = \Pi_N x, \quad x \in \mathcal{H},$$

where  $L_N := \Pi_N L \Pi_N : \mathcal{H} \rightarrow \mathcal{H}_N$ .

- The associated cost functional is:

$$J_{t,x_N}^N(u) := \int_t^T [\mathcal{G}(y_{t,x_N}^N(s; u)) + \mathcal{E}(u(s))] ds,$$

where  $y_{t,x_N}^N(\cdot, u)$  denotes the solution of the Galerkin approximation (4.1).



# Value functions and the main convergence result

The value functions corresponding to the above optimal control problems and their Galerkin approximations are defined by:

$$v(t, x) := \inf_{u \in \mathcal{U}_{ad}[t, T]} J_{t, x}(u), \quad \forall (t, x) \in [0, T) \times \mathcal{H} \quad \text{and} \quad v(T, x) := 0,$$

$$v_N(t, x_N) := \inf_{u \in \mathcal{U}_{ad}[t, T]} J_{t, x_N}^N(u), \quad \forall (t, x_N) \in [0, T) \times \mathcal{H}_N \quad \text{and} \quad v_N(T, x_N) := 0.$$

We identify **checkable conditions** that guarantee the following convergence result:

$$\lim_{N \rightarrow \infty} \sup_{t \in [0, T]} |v_N(t, \Pi_N x) - v(t, x)| = 0, \quad \forall x \in \mathcal{H}.$$

## Reference:

[CKL17] M. D. Chekroun, A. Kröner and H. Liu, *Galerkin approximations of nonlinear optimal control problems in Hilbert spaces*, EJDE, 1–40, 2017. [[arXiv link](#)]

## Sufficient conditions for the convergence of value functions

The conditions identified in [CKL17] to ensure the above convergence result can be put into three groups.

**Group I:** Conditions on the linear operator  $L$  and its Galerkin approximation  $L_N$ .

- $L : D(L) \subset \mathcal{H} \rightarrow \mathcal{H}$  generates a  $C_0$ -semigroup of bounded linear operators  $\{T(t)\}_{t \geq 0}$  on  $\mathcal{H}$ . In particular, there are constants  $M > 0$  and  $\omega \in \mathbb{R}$ , such that  $\|T(t)\| \leq Me^{\omega t}$  for all  $t \geq 0$ .
- The linear flow  $e^{L_N t} : \mathcal{H}_N \rightarrow \mathcal{H}_N$  extends to a  $C_0$ -semigroup  $T_N(t)$  on  $\mathcal{H}$  for each  $N \geq 1$ . The following **stability condition** is satisfied by the family  $\{T_N(t)\}_{N \geq 1, t \geq 0}$

$$\|T_N(t)\| \leq Me^{\omega t}, \quad N \geq 1, \quad t \geq 0.$$

- The following **consistency condition** holds:

$$\lim_{N \rightarrow \infty} \|L_N \phi - L \phi\|_{\mathcal{H}} = 0, \quad \forall \phi \in D(L).$$

## Sufficient conditions (cont'd)

**Group II: Local Lipschitz conditions** on  $F$  and  $\mathcal{G}$ , **continuity** of  $\mathfrak{C}$  and  $\mathcal{E}$ , as well as **compactness** of  $U$  in  $V$  are required.

**Group III:** A uniform in  $u$  **a priori bound** as well as a condition on the **residual energy** are required:

(III-1) For each  $x \in \mathcal{H}$  and  $T > 0$ , there exists a constant  $\mathcal{C} := \mathcal{C}(T, x)$  such that

$$\begin{aligned} \|y(s; x, u)\|_{\mathcal{H}} &\leq \mathcal{C}, & \forall s \in [0, T], u \in \mathcal{U}_{ad}, \\ \|y_N(s; \Pi_N x, u)\|_{\mathcal{H}} &\leq \mathcal{C}, & \forall s \in [0, T], u \in \mathcal{U}_{ad}, N \in \mathbb{N}^*, \end{aligned}$$

where  $y(\cdot; x, u) := y_{0,x}(\cdot, u)$  and  $y_N(\cdot; x, u) := y_{0,x}^N(\cdot, u)$ .

(III-2) It is required that the residual energy of the solution  $y(\cdot; x, u)$  satisfies

$$\lim_{N \rightarrow \infty} \sup_{u \in \mathcal{U}_{ad}} \sup_{s \in [0, T]} \|\Pi_N^\perp y(s; x, u)\|_{\mathcal{H}} = 0,$$

where  $\Pi_N^\perp := \text{Id}_{\mathcal{H}} - \Pi_N$ .

**Note:** Sufficient conditions to ensure (III-2) are also identified in [CKL17] for Galerkin approx. based on eigenbasis. It requires essential  $L$  to be self-adjoint with compact resolvent, and (III-1) is satisfied.

# Main result and sketch of the proof

## Theorem (Chekroun, Kröner, L., 2017)

Assume that

- the conditions given in Groups I, II, III above hold;
- there exists for each pair  $(t, x)$  a minimizer  $u_{t,x}^*$  (resp.  $u_{t,x}^{N,*}$ ) in  $\mathcal{U}_{ad}[t, T]$  for the the value function  $v(t, x)$  (resp.  $v_N(t, \Pi_N x)$ ).

Then, it holds that

$$\lim_{N \rightarrow \infty} \sup_{t \in [0, T]} |v_N(t, \Pi_N x) - v(t, x)| = 0, \quad \forall x \in \mathcal{H}.$$

**Sketch of the proof:** Recall that

$v(t, x) := \inf_{u \in \mathcal{U}_{ad}[t, T]} J_{t,x}(u)$ ,  $v_N(t, x_N) := \inf_{u \in \mathcal{U}_{ad}[t, T]} J_{t,x_N}^N(u)$ , with

$$J_{t,x}(u) = \int_t^T [\mathcal{G}(y_{t,x}(s, u)) + \mathcal{E}(u(s))] ds, \quad J_{t,x_N}^N(u) = \int_t^T [\mathcal{G}(y_{t,x_N}^N(s; u)) + \mathcal{E}(u(s))] ds.$$

We only need to estimate  $\|y_{t,x}(s, u) - y_{t,x_N}^N(s; u)\|_{\mathcal{H}}$ . But, since  $L$  and  $F$  are time-independent, it suffices to estimate  $\|y(s; x, u) - y_N(s; x_N, u)\|_{\mathcal{H}}$ .

**Preparatory Lemma I:** Assume that

- $F : \mathcal{H} \rightarrow \mathcal{H}$  is locally Lipschitz;
- Conditions in Group III (i.e., (III-1) and (III-2)) hold.

Then,

$$\lim_{N \rightarrow \infty} \sup_{u \in \mathcal{U}_{ad}} \sup_{t \in [0, T]} \|\Pi_N^\perp F(y(t; x, u))\|_{\mathcal{H}} = 0.$$

**Proof:** Note that

$$\begin{aligned} \|\Pi_N^\perp F(y(t; x, u))\|_{\mathcal{H}} &\leq \underbrace{\|\Pi_N^\perp (F(y(t; x, u)) - F(\Pi_{N_0} y(t; x, u)))\|_{\mathcal{H}}}_{=: I_1(N, N_0; u)} \\ &\quad + \underbrace{\|\Pi_N^\perp F(\Pi_{N_0} y(t; x, u))\|_{\mathcal{H}}}_{=: I_2(N, N_0; u)}. \end{aligned}$$

Denoting  $\mathfrak{B} := B(0, \mathcal{C}) \subset \mathcal{H}$ , for any  $N \in \mathbb{N}^*$ , we have

$$I_1(N, N_0; u) \leq \text{Lip}(F|_{\mathfrak{B}}) \|\Pi_{N_0}^\perp y(t; x, u)\|_{\mathcal{H}} \xrightarrow{N_0 \rightarrow \infty} 0, \quad \forall t \in [0, T], u \in \mathcal{U}_{ad}.$$

Note also for each fixed  $N_0 \in \mathbb{N}^*$ , by compactness of  $\Pi_{N_0} \mathfrak{B}$ , we have

$$I_2(N, N_0; u) \xrightarrow{N \rightarrow \infty} 0, \quad \forall t \in [0, T], u \in \mathcal{U}_{ad}. \quad \square$$

**Preparatory Lemma II:** Assume all the conditions in Groups I–III hold. Then, for any  $(x, u)$  in  $\mathcal{H} \times \mathcal{U}_{ad}$ , the following uniform convergence result holds:

$$\lim_{N \rightarrow \infty} \sup_{u \in \mathcal{U}_{ad}} \sup_{t \in [0, T]} \|y_N(t; \Pi_N x, u) - y(t; x, u)\|_{\mathcal{H}} = 0.$$

**Proof:** Let  $w_N(t; u) := y(t; u) - y_N(t; u)$ . We have

$$\begin{aligned} w_N(t; u) &= \underbrace{T(t)x - e^{L_N t} \Pi_N x}_{J_1(t)} + \int_0^t \underbrace{(T(t-s) - e^{L_N(t-s)} \Pi_N) F(y(s; u))}_{J_2(t, s; u)} ds \\ &\quad + \int_0^t \underbrace{e^{L_N(t-s)} \Pi_N (F(y(s; u)) - F(y_N(s; u)))}_{J_3(t, s; u)} ds \\ &\quad + \int_0^t \underbrace{(T(t-s) - e^{L_N(t-s)} \Pi_N) \mathfrak{C}(u(s))}_{J_4(t, s; u)} ds. \end{aligned}$$

For  $J_3(t, s; u)$ , we have

$$\begin{aligned}\|J_3(t, s; u)\|_{\mathcal{H}} &= \|e^{L_N(t-s)}\Pi_N(F(y(s; u)) - F(y_N(s; u)))\|_{\mathcal{H}} \\ &\leq MLip(F|_{\mathfrak{B}})e^{\omega(t-s)}\|w_N(s; u)\|_{\mathcal{H}}\end{aligned}$$

By Gronwall's inequality, we get for all  $t$  in  $[0, T]$ :

$$\begin{aligned}\|w_N(t; u)\|_{\mathcal{H}} &\leq \left( \|J_1(t)\|_{\mathcal{H}} + \int_0^T \sup_{t \in [s, T]} \|J_2(t, s; u)\|_{\mathcal{H}} ds \right. \\ &\quad \left. + \int_0^T \sup_{t \in [s, T]} \|J_4(t, s; u)\|_{\mathcal{H}} ds \right) \exp(MLip(F|_{\mathfrak{B}})e^{\omega T}T).\end{aligned}$$

Conditions in Group I ensures (Trotter-Kato theorem [Pazy83, Thm. 4.5, p.88]):

$$\|J_1(t)\|_{\mathcal{H}} = \|T(t)x - e^{L_N t}\Pi_N x\|_{\mathcal{H}} \rightarrow 0, \quad \forall t \in [0, T].$$

For  $J_2(t, s; u)$ , we have

$$\begin{aligned} \|J_2(t, s; u)\|_{\mathcal{H}} &\leq \underbrace{\|(T(t-s) - e^{L_N(t-s)}\Pi_N)\Pi_{N_0}F(y(s; x, u))\|_{\mathcal{H}}}_{K_1(N, N_0; u)} \\ &\quad + \underbrace{\|(T(t-s) - e^{L_N(t-s)}\Pi_N)\Pi_{N_0}^\perp F(y(s; x, u))\|_{\mathcal{H}}}_{K_2(N, N_0; u)}. \end{aligned}$$

Note that by **Lemma I**,

$$\sup_{t \in [s, T]} K_2(N, N_0; u) \leq 2Me^{\omega T} \|\Pi_{N_0}^\perp F(y(s; x, u))\|_{\mathcal{H}} \xrightarrow{N_0 \rightarrow \infty} 0, \quad \forall s \in [0, T], \quad u \in \mathcal{U}_{ad}.$$

By again the compactness of  $\Pi_{N_0}\mathfrak{B}$  and the Trotter-Kato theorem, we have

$$\sup_{t \in [s, T]} K_1(N, N_0; u) \xrightarrow{N \rightarrow \infty} 0, \quad \forall s \in [0, T], \quad u \in \mathcal{U}_{ad}.$$

It follows that  $\sup_{t \in [s, T]} \|J_2(t, s; u)\|_{\mathcal{H}} \xrightarrow{N \rightarrow \infty} 0, \forall s \in [0, T], u \in \mathcal{U}_{ad}$ .

The  $\sup_{t \in [s, T]} \|J_4(t, s; u)\|_{\mathcal{H}}$  term can be dealt with in the same way.  $\square$



## Application to Optimal Control of Delay Differential Equations

### Reference:

[CKL18] M. D. Chekroun, A. Kröner and H. Liu, *Galerkin approximations for optimal control of nonlinear delay differential equations*, Chapter 4 in *Hamilton-Jacobi-Bellman Equations: Numerical Methods and Applications in Optimal Control*, Edited by D. Kalise, K. Kunisch, and Z. Rao, De Gruyter, 2018. [[arXiv link](#)]

## Application to the Wright equation

For any  $u$  in  $L^2(0, T; \mathbb{R})$  and  $T > 0$ , we consider

$$\frac{dm}{dt} = -m(t - \tau)(1 + m(t)) + u(t), \quad t \in (0, T), \quad (5.1)$$

supplemented with

$$\begin{aligned} m(t) &= \phi(t), \quad t \in [-\tau, 0), \quad \text{where } \phi \in L^2(-\tau, 0; \mathbb{R}), \\ m(0) &= m_0 \in \mathbb{R}. \end{aligned} \quad (5.2)$$

**Cost functional:**

$$J(m, u) := \int_0^T \left[ \frac{1}{2} m(t)^2 + \frac{\mu}{2} u(t)^2 \right] dt, \quad \mu > 0$$

**Optimal control problem:**

$\min J(m, u)$  subject to  $(m, u) \in L^2(0, T; \mathbb{R}) \times L^2(0, T; \mathbb{R})$   
solves the problem (5.1)–(5.2).

## Recasting into an evolution equation in a Hilbert space

The reformulation is classical. Denote by  $m_t$  the time evolution of the history segments of a solution

$$m_t(\theta) := m(t + \theta), \quad t \geq 0, \quad \theta \in [-\tau, 0],$$

we introduce then a new variable

$$y(t, \theta) := (m_t(\theta), m_t(0)), \quad t \geq 0,$$

and take the state space to be:

$$\mathcal{H} := L^2([-\tau, 0]; \mathbb{R}) \times \mathbb{R}.$$

The problem (5.1)–(5.2) can be rewritten as:

$$\frac{dy}{dt} = \mathcal{A}y + \mathcal{F}(y) + \mathfrak{C}u(t), \quad y(0) = \Phi := (\phi, m_0).$$

The linear operator  $\mathcal{A}: D(\mathcal{A}) \rightarrow \mathcal{H}$  is defined by

$$[\mathcal{A}\Psi](\theta) := \begin{cases} \frac{d^+ \Psi^D}{d\theta}, & \theta \in [-\tau, 0), \\ -\Psi^D(-\tau), & \theta = 0, \end{cases}$$

with domain

$$D(\mathcal{A}) = \left\{ (\Psi^D, \Psi^S) \in \mathcal{H} : \Psi^D \in H^1([-\tau, 0); \mathbb{R}), \lim_{\theta \rightarrow 0^-} \Psi^D(\theta) = \Psi^S \right\}.$$

The nonlinearity  $\mathcal{F}: \mathcal{H} \rightarrow \mathcal{H}$  is defined by

$$[\mathcal{F}(\Psi)](\theta) := \begin{cases} 0, & \theta \in [-\tau, 0), \\ -\Psi^D(-\tau)\Psi^S, & \theta = 0, \end{cases} \quad \text{for all } \Psi = (\Psi^D, \Psi^S) \in \mathcal{H}.$$

The control operator  $\mathfrak{C}: V \rightarrow \mathcal{H}$  is taken here to be linear and given by

$$[\mathfrak{C}v](\theta) := \begin{cases} 0, & \theta \in [-\tau, 0) \\ v, & \theta = 0 \end{cases}, \quad v \in V,$$

where  $V = \mathbb{R}$ .

## Galerkin approximation

We take  $\mathcal{H}_N$  to be spanned by the Koornwinder polynomials  $\{K_j^\tau : [-\tau, 0] \rightarrow \mathbb{R}\}$ :

$$\mathcal{H}_N := \text{span}\{K_j^\tau := (K_j^\tau, K_j^\tau(0)) : j = 0, \dots, N-1\}.$$

Using such polynomials to build Galerkin approx. of DDEs is first introduced in [CGLW16], to which we refer for its **theoretical advantages** and **numerical efficiencies**.

The corresponding Galerkin approximation is an ODE system given by:

$$\boxed{\frac{d\xi_N}{dt} = M\xi_N + G(\xi_N) + \mathfrak{C}_N u(t), \quad \xi_N(0) = \zeta_N \in \mathbb{R}^N, \quad \text{with}}$$

$$(M)_{j,n} = \frac{1}{\|\mathcal{K}_j\|_{\mathcal{E}}^2} \left( -K_n(-1) + \frac{2}{\tau} \sum_{k=0}^{n-1} a_{n,k} (\delta_{j,k} \|\mathcal{K}_j\|_{\mathcal{E}}^2 - 1) \right), \quad 0 \leq j, n \leq N-1,$$

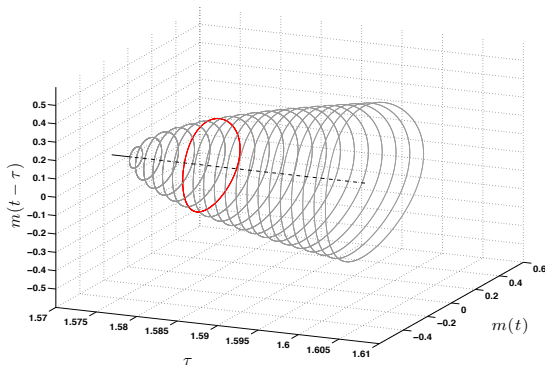
$$G_j(\xi_N) = -\frac{1}{\|\mathcal{K}_j\|_{\mathcal{E}}^2} \left[ \sum_{n=0}^{N-1} \xi_n^N(t) \right] \left[ \sum_{n=0}^{N-1} \xi_n^N(t) K_n(-1) \right], \quad 0 \leq j \leq N-1,$$

$$\mathfrak{C}_N v = \left( \frac{1}{\|\mathcal{K}_0^\tau\|_{\mathcal{H}}^2}, \dots, \frac{1}{\|\mathcal{K}_{N-1}^\tau\|_{\mathcal{H}}^2} \right)^{\text{tr}} v.$$

[CGLW16] M. D. Chekroun, M. Ghil, H. Liu & S. Wang, *Low-dimensional Galerkin approximations of nonlinear DDEs*. DCDS-A, Vol. 36, pp 4133–4177, 2016. [[arXiv link](#)]

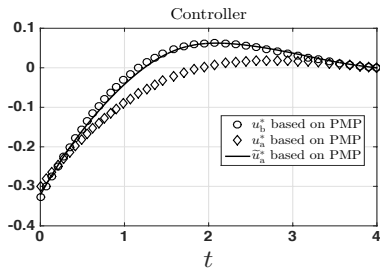
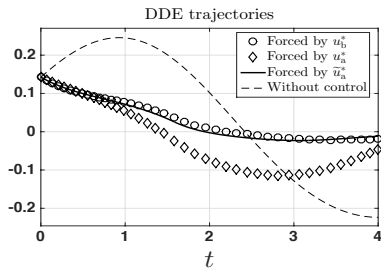
## Numerical setup: Amplitude oscillation reduction

The uncontrolled equation experiences a supercritical Hopf bifurcation when  $\tau$  crosses the critical delay  $\tau_c = \frac{\pi}{2}$  from below:



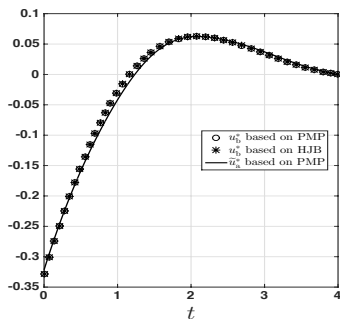
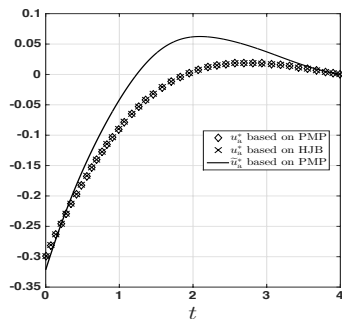
Our goal is to show that close to the criticality (i.e.  $\tau \approx \tau_c$ ), these amplitudes can be reduced at a nearly optimal cost, by solving efficiently a low-dimensional optimal control problem.

# Numerical results via Pontryagin Maximum Principle



- $\tilde{u}_a^*$ : from a 6-dim Galerkin-Koornwinder approximation (as a benchmark);
- $u_a^*$ : from a 2-dim Galerkin-Koornwinder approximation;
- $u_b^*$ : from another 2-dim ODE system obtained by projecting the 6-dim Galerkin-Koornwinder approximation onto the eigen-subspace spanned by the first two eigenvectors of the matrix  $M$ .

# Numerical results via Dynamic Programming

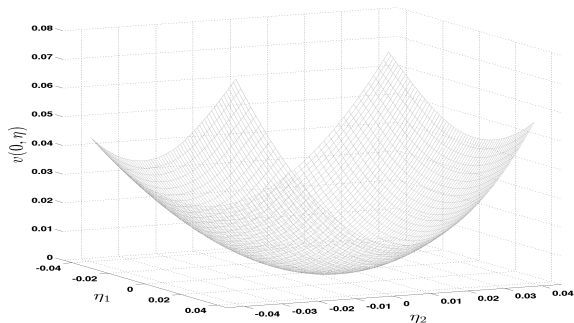


- **Left panel:** Control obtained without basis transformation (HJB vs PMP).
- **Right panel:** Control obtained after basis transformation (HJB vs PMP).



# Numerical results via Dynamic Programming (cont'd)

- Value function:



## Handling non-polynomial nonlinearities

There are applications for which non-polynomial nonlinearities can arise, and the conditions in the developed framework can still be checked. Performing **Taylor expansion** of such nonlinearity **may require high degree polynomials**.

**Example [CL20]:** Optimal management of harvested population

$$\min_{u \in L^2(t, T; L^2(\Omega))} \left( \frac{1}{2} \int_t^T |y(s) - p\delta'|_{L^2(\Omega)}^2 ds + \frac{\kappa}{2} \int_t^T |u(s)|_{L^2(\Omega)}^2 ds \right)$$

where  $y(s, x)$  solves the Kolmogorov-Petrovsky-Piskunov Eqn

$$\frac{\partial y}{\partial s} = D\nabla^2 y + \mu(x)y - \nu(x)y^2 - \delta\rho_\epsilon(y) + u(s, x), \quad (s, x) \in [t, T] \times \Omega,$$

$$\frac{\partial y}{\partial n} = 0, \quad (s, x) \in [t, T] \times \partial\Omega, \quad \text{with } y(t, x) = y_0(x).$$

The harvest function  $\rho_\epsilon$  takes the form

$$\rho_\epsilon(y) = \begin{cases} 1, & \text{if } y \geq \epsilon, \\ 0.5 \sin(\pi(y - 0.5\epsilon)/\epsilon) + 0.5, & \text{if } 0 < y < \epsilon, \\ 0, & \text{otherwise.} \end{cases}$$

# Summary

- **Convergence results** for Galerkin approximations of optimal control of nonlinear evolution equations are obtained.
- **Checkable conditions** are delineated. Error estimates for the value function and the optimal control can also be obtained under additional conditions [CKL17].
- The framework is **general and flexible**, which covers not only **semilinear parabolic PDEs** but also **delay differential equations**; and allows for a **broad class of nonlinear control laws**.

# References

[CKL17] M. D. Chekroun, A. Kröner and H. Liu, *Galerkin approximations of nonlinear optimal control problems in Hilbert spaces*, EJDE, 1–40, 2017. [[arXiv link](#)]

[CKL18] M. D. Chekroun, A. Kröner and H. Liu, *Galerkin approximations for optimal control of nonlinear delay differential equations*, Chapter 4 in *Hamilton-Jacobi-Bellman Equations: Numerical Methods and Applications in Optimal Control*, Edited by D. Kalise, K. Kunisch, and Z. Rao, De Gruyter, 2018. [[arXiv link](#)]

[CL20] M. D. Chekroun and H. Liu, *Optimal management of harvested population at the edge of extinction*, Chapter 2 in *Advances in Nonlinear Biological Systems: Modeling and Optimal Control*, Edited by J. Kotas, AIMS, 2020. [[arXiv link](#)]

## Some related references

- Galerkin approximations for the case of linear evolution equations
  - Ferretti (1997).
- Feedback control using POD and dynamic programming
  - Kunisch, Volkwein, Xie (2004), Kunisch, Xie (2005), Alla, Falcone (2013), Alla, Falcone, Volkwein (2015), Alla, Falcone, Kalise (2016)