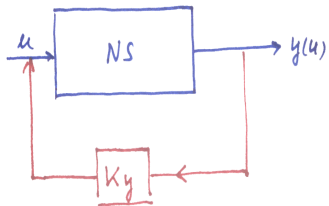


Some Practical Considerations for Closed Loop Control of Distributed Parameter Systems

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$$J(y, u) = \int_0^\infty \left(\frac{1}{2} \int_\Omega |y(t, x)|^2 dx + \frac{\beta}{2} |u(t)|^2 \right) e^{-\mu t} dt,$$

$$y_t - \nu y_{xx} + yy_x = 0 \quad \text{in } (0, \infty) \times (0, 1),$$

$$\nu y_x(\cdot, 0) + \sigma_0 y(\cdot, 0) = u \quad \text{in } (0, \infty)$$

$$\nu y_x(\cdot, 1) + \sigma_1 y(\cdot, 1) = g \quad \text{in } (0, \infty)$$

$$y(0, \cdot) = y_0 \quad \text{in } (0, 1),$$

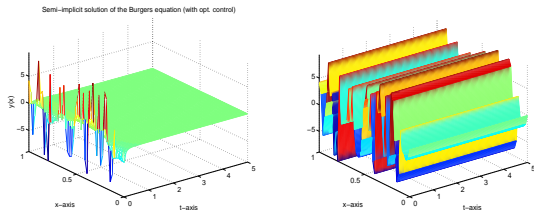


FIGURE 1. Optimal state with random noise (9.0) in the initial condition: feedback design (left) and open-loop design (right).

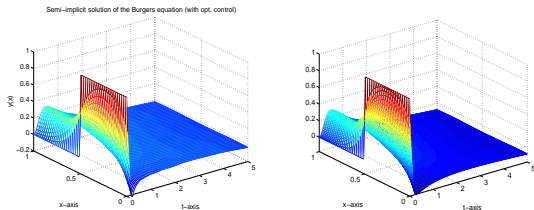


FIGURE 2. Optimal state with random noise (0.25) in the RHS: feedback design (left) and open-loop design (right).

$$\left\{ \begin{array}{l} \min E \left\{ \int_0^T ((y, Dy) + (u, Cu)) dt + (y(T), Gy(T)) \right\} \\ y_t = Ay + Bu + w \\ y(0) = y_0 \end{array} \right.$$

w white noise with intensity V , $Q_0 = E(y_0, \cdot)y_0$

$$u^* = -\mathcal{F}(t)y(t), \quad \mathcal{F}(t) = C^{-1}B^*\Pi(t)$$

$$\left\{ \begin{array}{l} -\Pi_t(t) = A^*\Pi(t) + \Pi(t)A^* - \Pi(t)BC^{-1}B^*\Pi(t) + D \\ \Pi(T) = G \end{array} \right.$$

$$E_{\min} = \text{tr}(\Pi(0)Q_0) + \int_0^T \Pi(t)V(t) dt$$

close the loop: $y_t = Ay(t) - BC^{-1}B^*\Pi(t)y(t)$

Dynamic Programming Principle

$$\begin{cases} \min J(y_0, u) = \int_0^\infty L(y(t), u(t))e^{-\mu t} dt \\ \dot{y}(t) = f(y(t), u(t)), \quad y(0) = y_0, u \in \mathcal{U}_{ad}. \end{cases}$$

value function $v(y_0) = \min_{u \in \mathcal{U}_{ad}} J(y_0, u)$.

$$(DPP) \quad v(y_0) = \min_{u \in \mathcal{U}_{ad}} \left\{ \int_0^T L(y(t; y_0, u), u(t))e^{-\mu t} dt + v(y(T; y_0, u))e^{-\mu T} \right\}$$

$$(HJB) \quad \mu v(y_0) = \min_{u \in \mathcal{U}_{ad}} \{ (\nabla v(y_0), f(y_0, u)) + L(y_0, u) \}$$

replace y_0 by $y^*(t)$.

$$u^*(t) = \mathcal{F}(y^*(t)) \quad \text{where}$$

$$\mathcal{F}(y^*(t)) \in \operatorname{argmin}_{u \in \mathcal{U}_{ad}} \{ (\nabla v(y^*(t)), f(y^*(t), u)) + L(y^*(t), u) \}$$

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$$(P) \begin{cases} \min_{u \in U_{ad}} \int_0^T e^{-\mu t} h(y) + \frac{\alpha}{2} \int_0^T |u|^2 \\ \dot{y} = f(y) + Bu(t), \quad y(0) = y_0 \end{cases}$$

$v = v(t, y)$... value functional $\nabla = \nabla_y$

$$\begin{cases} v_t + \mu v = \min_{u \in U_{ad}} ((f(t, y) + B^T u, \nabla v) + h(y) + \frac{\alpha}{2} |u|^2) \\ v(T, \cdot) = 0 \end{cases}$$

$$u = P_{U_{ad}} \left(-\frac{1}{\alpha} B^T \nabla v \right)$$

t-discretisation, infinite time horizon

$$t_j = jh$$

$$\begin{cases} y_{j+1} = y_j + hf(y_j, u_j) \\ y_0 \text{ given} \end{cases}$$

$$J_h(y_0, u_h) = \frac{h}{2} [L(y_0, u_0) + \sum_{j=1}^{\infty} e^{-\mu hj} (L(y_j, u_{j-1}) + L(y_j, u_j))].$$

$$v_h(y_0) = \inf_{u_h \in \mathcal{U}_{ad}^h} J_h(y_0, u_h)$$

$$(HJB_h) \quad v_h(y_0) = \inf_{u \in \mathcal{U}_{ad}^h} \left\{ \frac{h}{2} [L(y_0, u) + e^{-\mu h} L(y_0 + hf(y_0, u), u)] + e^{-\mu h} v_h(y_0 + hf(y_0, u)) \right\}$$

$$y_{j+1}^* = y_j^* + hf(y_j^*, \mathcal{F}_h(y_j^*)), j \geq 0$$

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$$y_{j+1}^* = y_j^* + hf(y_j^*, \mathcal{F}_h(y_j^*)), j \geq 0$$

numerically still infeasible, joint with L. Xie

discretization, finite time horizon

in time: TVD Runge-Kutta scheme

in space: local Lax-Friedrich scheme

POD-Galerkin Ansatz (for Navier Stokes):

$$y^l(t, \mathbf{x}) = \sum_{i=1}^l \alpha_i(t) \varphi_i(\mathbf{x}), \quad \mathbf{u}(t, \mathbf{x}) = \sum u_i(t) \hat{\varphi}_i(\mathbf{x}),$$

$\{\varphi_i\}$ POD basis for velocities, $\{\hat{\varphi}_i\}$ POD basis for controls

$$\dot{\alpha} = -\mathbf{A}\alpha - \alpha^T \mathbf{H}\alpha + \mathbf{G}(u)$$

$$\mathbf{A}_{ij} = \frac{1}{Re} \langle \nabla \varphi_i, \nabla \varphi_j \rangle, \quad \mathbf{H}_{ijk} = \langle \varphi_j \cdot \nabla \varphi_k, \varphi_i \rangle,$$

$$\mathbf{G}_i = \langle \mathbf{u}, \varphi_i \rangle$$

Implementation Aspects

for NS: finite difference upwind scheme on staggered nonuniform grid, fractional ϑ -scheme for time-integration.

$$(HJB_h) v_h(\alpha) = \inf_{u \in \mathcal{U}_{ad}^h} \left\{ \frac{h}{2} \left(L(\alpha, u) + e^{-\mu h} L(\alpha + hf(\alpha, u), u) \right) + e^{-\mu h} v(\alpha + hf(\alpha, u)) \right\}$$

hypercube Γ_h in \mathbb{R}^n such that $\alpha^*(t) \in \Gamma_h$ for all t ,
 α_j grid points in Γ_h , $\alpha = \sum \lambda_j \alpha_j$, $\sum \lambda_j = 1$

$$v_h(\alpha) = \sum \lambda_j v_h(\alpha_j)$$

Solve (HJB_h) as fixed point equation,
(Bardi, Capuzzo-Dolcetta; Gonzales-Rofman)
nonuniform grid \sim decay of POD eigenvalues.

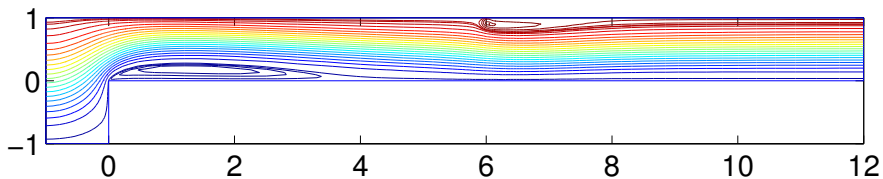
Constrained nonlinear programming problem at each grid point of Γ_h : in parallel.

Validation

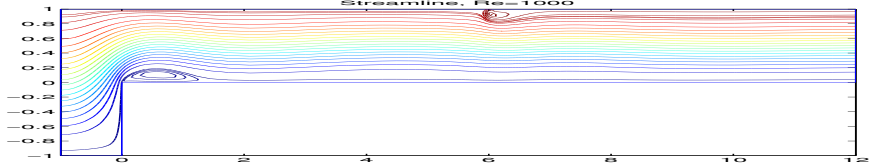
Reynolds number, Re	1000
Time horizon, T	10
No. of basis functions	6
No. of controls	2
Bounds on controls	$[-0.5, 0.5]$
Grid system	$12 \times 6 \times 6 \times 4 \times 3 \times 3$
Discount rate, μ	2
Observe region, Ω_o	$[-1, 12] \times [0, 1]$

Table: Parameter Settings

Streamline, Re=1000



Streamline, Re=1000



2-point BVP-solutions

$\mu = 0$, no control constraints

$$(OS) \begin{cases} \dot{y} = f(y) + Bu, & y(0) = y_0 \\ \dot{\lambda} + f'(y)\lambda = -h'(y), & \lambda(T) = 0 \\ u = \frac{1}{\alpha} \lambda \end{cases}$$

$$\begin{aligned} u(t) &= -\frac{1}{\alpha} B^T \nabla_y v(t, y(t)) = \mathcal{F}_T(t, y(t)) = \frac{1}{\alpha} \lambda(t; y_0) \\ &= \mathcal{F}_{T-t}(0, y(t)) \sim \mathcal{F}_T(0, y(t)) = \frac{1}{\alpha} \lambda(0; y(t)) \end{aligned}$$

finite dimensional realization

- (i) choose grid $\Sigma \subset \mathbb{R}^n$ containing optimal trajectory
- (ii) calculate $(y(\cdot, \bar{y}_0), \lambda(\cdot, \bar{y}_0))$ for all $\bar{y}_0 \in \Sigma$, thus $\mathcal{F}_T(0, \bar{y}_0)$ available
- (iii) use interpolation for $\mathcal{F}_T(0, y)$, $y \in \mathbb{R}^n$ arbitrary.

Remark: (OS) must be solved often ! (gridpoints in one direction)^{# basis} → POD use for open loop optimal control to estimate values in Σ .

Example

(Control-to-zero of Burgers equation).

	FE	POD	closed loop
cost	0.1694	0.1704	0.1733

Optimal Vortex Reduction by Non-Stationary Flows

Cost functionals:

▶ Tracking functional: $J_1(y) = \int_0^T \int_{\Omega} |y(t, x) - y_{des}(t, x)|^2 dx dt$

▶ Curl functional: $J_2(y) = \int_0^T \int_{\Omega} |\text{curl } y(t, x)|^2 dx dt$

▶ Galilean functional: $J_3(y) = \int_0^T \int_{\Omega} |\det \nabla y| dx dt$

▶ Objective functional: $J_4(y, p) = \int_0^T \int_{\Omega} (|r(y, p)| - 1)^+ dx dt,$

$$r(y, p) = \frac{\omega}{\sigma} - \frac{\sigma_s(p_{x_1 x_1} - p_{x_2 x_2}) - 2\sigma_n p_{x_1 x_2}}{\sigma^{\frac{3}{2}}},$$

$$\omega = (y_2)_{x_1} - (y_1)_{x_2}, \quad \sigma_s = (y_2)_{x_1} + (y_1)_{x_2}, \quad \sigma_n = (y_1)_{x_1} - (y_2)_{x_2}.$$

jointly with B. Vexler

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jointly with B. Vexler

$$\begin{aligned}
 y &= 0 && \text{on } \partial\Omega \setminus (\Gamma_{in} \cup \Gamma_{out}), \\
 y &= g(u) \hat{y}_{in} && \text{on } \Gamma_{in},
 \end{aligned}$$

\hat{y}_{in} parabolic profile

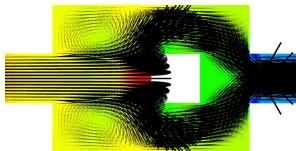
$$g(u)(t) = \frac{c_0}{T} + \sum_{k=1}^n (u_{2k-1} \sin(2\pi kt/T) + u_{2k} \cos(2\pi kt/T)).$$

$$U = \mathbb{R}^{2n}.$$

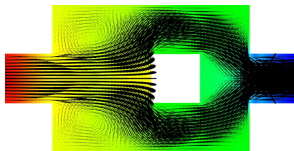
$$\int_0^T \int_{\Gamma_{in}} y \cdot n \, ds \, dt = c_0 \int_{\Gamma_{in}} \hat{y}_{in} \cdot n \, ds$$

independently of u .

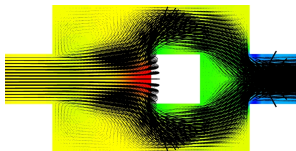
Optimal Vortex Reduction by Non-Stationary Flows



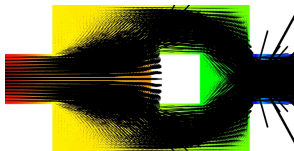
tracking functional



curl functional

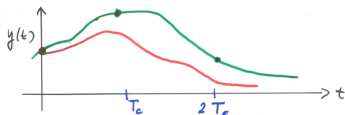


Galilean functional



Objective functional

Receding horizon control



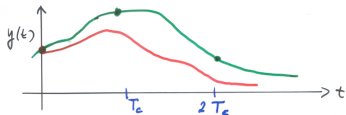
$T_c \dots$ control-interval (stabilization, $y = 0$ steady state).

$$(P) \begin{cases} \min \int_0^{\infty} h^\circ(y(t), u(t)) dt & \sim \int_0^{T_c} h^\circ(y, u) dt + \int_{T_c}^{\infty} h^\circ(y, u) dt \\ \frac{d}{dt} y(t) = f(y(t), u(t)), & y(0) = y_0, \quad \text{in } Y. \end{cases}$$

$$f(0, 0) = h^\circ(0, 0) = 0$$

end-point restriction: $y(T_c) = 0$ or $y(T_c) \in S$.

Receding horizon control



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Alternative: replace $\int_T^\infty h^\circ(y(t), u(t))dt$ by $G(y(T))$

$$(P_k) \begin{cases} \inf_u \int_{(k-1)T}^{kT} h^\circ(y(t), u(t))dt + G(y(kT)) \\ \dot{y}(t) = f(y(t), u(t)), \\ y((k-1)T) = \bar{y}((k-1)T) \end{cases}$$

global solution by concatenation: (\bar{y}_k, \bar{u}_k) .

Definition

$G \in C(Y, \mathbb{R})$, $G(0) = 0$, is a **control - Ljapunov - function**, if

$\forall y_0 \in Y$, $T > 0 \exists u = u(\cdot, y_0, T)$:

$$\int_0^T h^\circ(y(t), u(t))dt + G(y(T)) \leq G(y_0)$$

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JUSTIFY

$$V_T(y_0) = \inf_u \int_0^T h^\circ(y(t), u(t)) dt + G(y(T)).$$

Theorem (H)

- ▶ $T \rightarrow V_T(y_0)$ monotonically decreasing.
- ▶ There exists $\rho_T < 1$: $G(\bar{y}(kT)) \leq \rho_T^k G(y_0)$ for all k .

CONSTRUCT

$$G(y) = \frac{\alpha}{2} |y|^2 \dots \text{CLF ?}$$

Definition

(P) is called **closed loop dissipative**, if \exists feedback law $u = \Phi(y) \in U$ and $\alpha > 0$:

$$\alpha(f(y, u), y)_Y + h^\circ(y, u) \leq 0,$$

CLD $\Rightarrow \frac{\alpha}{2}|y|^2$ is CLF. (domains)

Examples

Navier–Stokes equations, Boussinesq equations, geophysical flows, etc.

more generally: G requires to solve a Liapunov or Riccati equation.

K.Ito-K.

Applications

(i) dissipative Systems:

$$\frac{d}{dt}y(t) = A_0y(t) + f(y(t)) + Bu(t)$$

A_0 dissipative, f : locally Lipschitz, $(f(y), y)_Y = 0$

e.g. Navier–Stokes equations, Boussinesq equation, geophysical flows, Lorenz systems, etc.

$$h^\circ(y, u) = \frac{1}{2}|y|^2 + \frac{\beta}{2}|u|^2, \quad u = -B^*y,$$

$G(y) = \frac{\alpha}{2}|y|^2$ ist CLF if α sufficiently large.

(ii) wave equation with damping

$$\left(\frac{d}{dt}\right)^2 z(t) + \partial\psi\left(\frac{d}{dt}z(t)\right) + \Lambda z(t) \ni Bu(t),$$

$$\psi: \mathbb{R}^n \rightarrow \mathbb{R} \text{ convex}, \partial\psi(0) \ni 0, \Lambda \in \mathbb{R}^{n \times n}.$$

Coulomb friction: $\psi(z) = \sum_{i=1}^n \gamma_i |z_i|$, $\gamma_i \geq 0$.

CLF–function G is solution of algebraic Riccati equation.

(iii) Semilinear wave equation

$$\begin{cases} y_{tt} + \varphi(y_t) - y_{xx} = u \chi_I, & x \in (0, 1), t > 0 \\ y(t, 0) = 0, y_x(t, 1) = 0 \end{cases}$$

$I = (x_1, x_2) \subset (0, 1)$, φ Lipschitz, $\varphi(s)s \geq 0$,

$$\int_0^T h^\circ(y, u) dt = \frac{\gamma}{2} \int_0^T \left(|y_x|_{L^2}^2 + |y_t|_{L^2}^2 + |u|_{L^2(I)}^2 \right) dt$$

$$G(y(T)) = \int_0^1 \left(|y_x(T)|^2 + |y_t(T)|^2 \right) dx +$$

$$\int_0^1 a(x) y_x(T) y_t(T) dx + b \int_0^1 y(T) y_t(T) dx$$

Feedback Control of Schrödinger Equation

$$i \frac{\partial}{\partial t} \Psi(x, t) = (\mathcal{H}_0 + \epsilon(t)\mu)\Psi(x, t), \quad \Psi(x, 0) = \Psi_0(x).$$

\mathcal{H}_0 positive, closed, self-adjoint operator in X ,
 μ self-adjoint dipole operator, $\epsilon(t) \in \mathbb{R}$ amplitude

$$i \frac{d}{dt} \mathcal{O}(t) = \mathcal{H}_0 \mathcal{O}(t),$$

$$\mathcal{O}(t) = e^{-i(\lambda t - \theta)} \psi_{k_0},$$

$$V_1(\Psi(t), \mathcal{O}(t)) = \frac{1}{2} |\Psi(t) - \mathcal{O}(t)|_X^2$$

$$V_2(\Psi(t), \mathcal{O}(t)) = \frac{1}{2} (1 - |(\mathcal{O}(t), \Psi(t))_X|^2)$$

$$V_2(\Psi, \mathcal{O}) = 0 \quad \text{if and only if} \quad \Psi = e^{i\theta} \mathcal{O},$$

observe $\Psi(t) = \Psi_0 = 1$ for all t .

$$\frac{d}{dt} V_1(\Psi(t), \mathcal{O}(t)) = \epsilon(t) \operatorname{Im}(\mathcal{O}(t), \mu\Psi(t))_X.$$

$$\epsilon(t) = -\frac{1}{\beta} \operatorname{Im}(\mathcal{O}(t), \mu\Psi(t))_X = F_1(\Psi(t), \mathcal{O}(t)),$$

then

$$\frac{d}{dt} V_1(\Psi(t), \mathcal{O}(t)) = -\beta |\epsilon(t)|^2.$$

Similarly

$$\frac{d}{dt} V_2(\Psi(t), \mathcal{O}) = \epsilon(t) \operatorname{Im}\left(\overline{(\mathcal{O}, \Psi(t))_X} (\mathcal{O}, \mu\Psi(t))_X\right).$$

$$\epsilon(t) = -\frac{1}{\beta} \operatorname{Im}\left(\overline{(\mathcal{O}, \Psi(t))_X} (\mathcal{O}, \mu\Psi(t))_X\right) = F_2(\Psi(t), \mathcal{O}),$$

then

$$\frac{d}{dt} V_2(\Psi(t), \mathcal{O}) = -\beta |\epsilon(t)|^2.$$

Assume that



$\{e^{i(\lambda_k - \lambda_{k_0})\tau}\} \cup \{e^{-i(\lambda_k - \lambda_{k_0})\tau}\}_{k \neq k_0}$ is ω -independent in $L^2(0, T)$,

for some $T > 0$



$\mu_{k_0}^k = (\psi_{k_0}, \mu \psi_k)_X \neq 0$ for all $k = 1, 2, \dots$



$\{S(t)\Psi_0, t \geq 0\}$ compact in X .

Theorem

$\lim_{t \rightarrow \infty} V_1(\Psi(t), \mathcal{O}(t)) = 0$, for the feedback law F_1 .

If, in addition, $(\Psi_0, \psi_{k_0})_X \neq 0$, then $\lim_{t \rightarrow \infty} V_2(\Psi(t), \mathcal{O}(t)) = 0$, for the feedback law F_2 .

If $|\lambda_k + \lambda_\ell - 2\lambda_{k_0}| > \delta$ is violated: $\tilde{\mu}(t) = \sum_{j=1}^m \epsilon_j(t) \mu_j$,

$$V_1(t, \Psi) = 1 - (\mathcal{O}(t), \Psi)_{H \times H}$$

satisfies the Hamilton Jacobi equation

$$\frac{\partial V_1}{\partial t} + \min_{\epsilon} \left[\frac{\beta}{2} |\epsilon|^2 + (V_1)_{\Psi}(A_0 \Psi + \epsilon B \Psi) \right] + \frac{1}{2\beta} |(\mathcal{O}(t), B \Psi)_{H \times H}|^2 = 0,$$

where $A_0 = \begin{pmatrix} 0 & \mathcal{H}_0 \\ -\mathcal{H}_0 & 0 \end{pmatrix}$, $B = \begin{pmatrix} 0 & \mu \\ -\mu & 0 \end{pmatrix}$

$$\epsilon^* = \frac{1}{\beta} (\mathcal{O}(t), B \Psi)_{H \times H} = F_1(\Psi, \mathcal{O}(t))$$

minimizes

$$J(\epsilon) = \int_0^T \left(\frac{\beta}{2} |\epsilon(t)|^2 + \frac{1}{2\beta} |(\mathcal{O}(t), B \Psi(t))_{H \times H}|^2 \right) dt + V_1(\Psi(T), \mathcal{O}(T)),$$

over $\epsilon \in L^2(0, T)$.