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Taylor Expansions of the Value Function Associated with Stabilization Problems

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We consider the following bilinear optimal control problem:

$$\inf_{\substack{u \in L^{2}(0,\infty)}} \mathcal{J}(u, y_{0}) := \int_{0}^{\infty} \frac{1}{2} \|y(t)\|_{Y}^{2} + \frac{\beta}{2} |u(t)|^{2} dt,$$

where:
$$\begin{cases} \dot{y}(t) = Ay(t) + Ny(t)u(t) + Bu(t), \\ y(0) = y_{0} \in Y, \end{cases}$$
 (P(y_{0}))

with associated value function: $\mathcal{V}(y_0) := \inf_{u \in L^2(0,\infty)} \mathcal{J}(u, y_0)$. Key ideas:

- The derivatives D^jV(0) are characterized by a sequence of equations.
- This allows for the numerical approximation of V and the optimal feedback law (locally, around 0).

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Assumptions			

Functional framework:

- V ⊂ Y ⊂ V* is a Gelfand triple of real Hilbert spaces, where the embedding of V into Y is dense and compact
- $W(0,\infty) = \{y \in L^2(0,\infty; V) \mid \dot{y} \in L^2(0,\infty; V^*)\}.$

Assumptions:

- (A1) The operator -A can be associated with a V-Y coercive bilinear form $a: V \times V \to \mathbb{R}$ such that $\exists \lambda \in \mathbb{R}$ and $\delta > 0$ satisfying $a(v, v) \ge \delta \|v\|_{V}^{2} \lambda \|v\|_{Y}^{2}$, for all $v \in V$.
- (A2) The operator N is such that $N \in \mathcal{L}(V, Y)$ and $N^* \in \mathcal{L}(V, Y)$.
- (A3) **[Stabilizability]** There exists an operator $F \in \mathcal{L}(Y, \mathbb{R})$ such that the semigroup $e^{(A+BF)t}$ is exponentially stable on Y.

Another technical assumption is also needed.

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Roadmap			

The **Taylor expansion** of order k, denoted V_k is of the form:

$$\mathcal{V}_k(y_0) = rac{1}{2}\mathcal{T}_2(y_0,y_0) + rac{1}{3!}\mathcal{T}_3(y_0,y_0,y_0) + ... + rac{1}{k!}\mathcal{T}_k(y_0,...,y_0),$$

where $\mathcal{T}_j = D^j \mathcal{V}(0)$ is a **bounded multilinear form** from Y^j to \mathbb{R} . *Remark:* $\mathcal{V}(0) = 0$, $D\mathcal{V}(0) = 0$.

We formally show that

- *T*₂ is the unique solution to an **algebraic Riccati equation** (ARE)
- *T*₃, *T*₄,... are the unique solutions to (linear) generalized Lyapunov equations (GLE).

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HJB equation

Proposition

Assume that there exists a neighborhood Y_0 of 0 such that

- **1** Problem $P(y_0)$ has a continuous solution $u, \forall y_0 \in \mathcal{D}(A) \cap Y_0$
- **2** The value function is continuously differentiable on Y_0 .

Then, for all $y_0 \in \mathcal{D}(A) \cap Y_0$,

 $D\mathcal{V}(y_0)Ay_0 + \frac{1}{2}||y_0||_Y^2 - \frac{1}{2\beta} (D\mathcal{V}(y_0)(Ny_0 + B))^2 = 0.$ (HJB)

Moreover, for all continuous solutions \bar{u} to problem $P(y_0)$,

$$ar{u}(t) = -rac{1}{eta} D \mathcal{V}(ar{y}(t)) (N ar{y}(t) + B), ext{ for a.e. } t.$$

Control in feedback form!

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The equations characterizing $(\mathcal{T}_j)_{j=2,3,...}$ are then obtained by successive **differentiation** of the HJB equation.

First differentiation of (HJB) w.r.t. *y* in some direction $z_1 \in \mathcal{D}(A)$:

$$D^{2}\mathcal{V}(y)(Ay, z_{1}) + D\mathcal{V}(y)Az_{1} + \langle y, z_{1} \rangle_{Y}$$

- $\frac{1}{\beta} (D^{2}\mathcal{V}(y)(Ny + B, z_{1}) + D\mathcal{V}(y)Nz_{1}) (D\mathcal{V}(y)(Ny + B)) = 0.$

Note: $y_0 \rightarrow y$.

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Second differentiation of (HJB):

$$D^{3}\mathcal{V}(y)(Ay, z_{1}, z_{2}) + D^{2}\mathcal{V}(y)(Az_{2}, z_{1}) + D^{2}\mathcal{V}(y)(Az_{1}, z_{2}) + \langle z_{1}, z_{2} \rangle_{Y}$$

- $\frac{1}{\beta} (D^{2}\mathcal{V}(y)(Ny + B, z_{1}) + D\mathcal{V}(y)Nz_{1}) (D^{2}\mathcal{V}(y)(Ny + B, z_{2}) + D\mathcal{V}(y)Nz_{2})$
- $\frac{1}{\beta} (D^{3}\mathcal{V}(y)(Ny + B, z_{1}, z_{2})) (D\mathcal{V}(y)(Ny + B))$
- $\frac{1}{\beta} (D^{2}\mathcal{V}(y)(Nz_{2}, z_{1}) + D^{2}\mathcal{V}(y)(Nz_{1}, z_{2})) (D\mathcal{V}(y)(Ny + B)) = 0.$

For y = 0, using the representation $D^2 \mathcal{V}(0)(z_1, z_2) = \langle z_1, \Pi z_2 \rangle$, where $\Pi: Y \to Y$, we obtain an algebraic Riccati equation:

$$A^*\Pi + \Pi A + \mathsf{Id} - \frac{1}{\beta}\Pi BB^*\Pi = 0.$$
 (ARE)

It has a unique self-adjoint and non-negative solution.

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Taylor expansion			

Third differentiation of (HJB), at y = 0:

$$D^{3}\mathcal{V}(0)(Az_{3}, z_{1}, z_{2}) + D^{3}\mathcal{V}(0)(Az_{2}, z_{1}, z_{3}) + D^{3}\mathcal{V}(0)(Az_{1}, z_{2}, z_{3})$$

- $\frac{1}{\beta}(D^{3}V(0)(B, z_{1}, z_{3}) + D^{2}\mathcal{V}(0)(Nz_{3}, z_{1}) + D^{2}\mathcal{V}(0)(Nz_{1}, z_{3}))D^{2}\mathcal{V}(0)(B, z_{2})$
- $\frac{1}{\beta}(D^{3}\mathcal{V}(0)(B, z_{2}, z_{3}) + D^{2}\mathcal{V}(0)(Nz_{3}, z_{2}) + D^{2}\mathcal{V}(0)(Nz_{2}, z_{3}))D^{2}\mathcal{V}(0)(B, z_{1})$
- $\frac{1}{\beta}(D^{3}\mathcal{V}(0)(B, z_{1}, z_{2}) + D^{2}\mathcal{V}(0)(Nz_{2}, z_{1}) + D^{2}\mathcal{V}(0)(Nz_{1}, z_{2}))D^{2}\mathcal{V}(0)(B, z_{3}) = 0.$

We set: $A_{\Pi} = A - \frac{1}{\beta}BB^*\Pi$, we obtain:

$$egin{aligned} \mathcal{T}_3(egin{aligned} & \mathcal{T}_3(egin{aligned} & \mathcal{T}_3(egin{aligned} & \mathcal{T}_1, egin{aligned} & \mathcal{T}_3(egin{aligned} & \mathcal{T}_1, egin{aligned} & \mathcal{T}_2, egin{aligned} & \mathcal{T}_3(egin{aligned} & \mathcal{T}_1, egin{aligned} & \mathcal{T}_3(egin{aligned} & \mathcal{T}_3, eg$$

where the trilinear form $\mathcal{R}_3 \colon Y^3 \to \mathbb{R}$ is determined by Π , N, and B.

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Taylor expansion			

Differentiation of order j of (HJB), at y = 0:

$$\mathcal{T}_{j}(A_{\Pi}z_{1}, z_{2}, ..., z_{k}) + ... + \mathcal{T}_{j}(z_{1}, ..., z_{k-1}, A_{\Pi}z_{k})$$

= $\frac{1}{2\beta}\mathcal{R}_{j}(z_{1}, ..., z_{j}), \quad \forall (z_{1}, ..., z_{j}) \in \mathcal{D}(A)^{j}.$ (GLE(j))

Properties of the derived generalized Lyapunov equations:

- linear equation
- computable right-hand side: the multilinear form *R_j*: *Y^j* → ℝ is explicitely determined by Π, *D³V*(0),...,*D^{j-1}V*(0), *N*, and *B*.

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Theorem

There exists a unique sequence $(\mathcal{T}_j)_{j=3,4,...}$ of symmetric bounded multilinear forms such that $\mathcal{T}_j: Y^j \to \mathbb{R}$ is a solution to GLE(j).

Proof. Representation formula:

$$\mathcal{T}_j(z_1,...,z_k) = -\int_0^\infty \mathcal{R}_j(e^{A_\pi t}z_1,...,e^{A_\pi}z_k)\,\mathrm{d}t.$$

Remark: the well-posedness of the GLEs can be established without knowledge regarding the differentiability of V.

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Feedback law			

Polynomial \mathcal{V}_k of degree *k*:

$$\mathcal{V}_k(y) = \sum_{k=2}^k \frac{1}{j!} \mathcal{T}_j(y, ..., y).$$

Feedback law u_k of order k:

$$\mathbf{u}_k$$
: $y \in Y \mapsto \mathbf{u}_k(y) = -\frac{1}{\beta} D\mathcal{V}_k(y)(Ny+B).$

Closed-loop system of order k:

 $\dot{y}_k(t) = Ay_k(t) + (Ny_k(t) + B)\mathbf{u}_k(y_k(t)), \quad y_k(0) = y_0.$

Open-loop control U_k(y_0) generated by the feedback **u**_k and y_0 :

 $\mathbf{U}_k(y_0;t)=\mathbf{u}_k(y_k(t)).$

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Numerical approach			

- **Discretize** the operators *A*, *N*, and *B* in such a way that the bilinear structure is preserved (e.g. with finite differences)
- Find a reduced-order model with a generalization of the balanced truncation method:

$$\inf_{u \in L^{2}(0,\infty)} J(u, y_{0}) := \int_{0}^{\infty} \frac{1}{2} \|C_{r}y_{r}(t)\|_{\mathbb{R}^{n}}^{2} + \frac{\beta}{2} |u(t)|^{2} dt,$$

where:
$$\begin{cases} \dot{y}_{r}(t) = A_{r}y_{r}(t) + N_{r}y_{r}(t)u(t) + B_{r}u(t), \\ y_{r}(0) = y_{0,r} \in Y. \end{cases}$$

3 Solve the reduced GLE with a **tensor-calculus technique**.

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 Lyapunov equations

The associated reduced GLE of order k:

$$T_{k,r}(A_{\Pi,r}z_1, z_2, ..., z_k) + ... + T_{k,r}(z_1, ..., z_{k-1}, A_{\Pi,r}z_k)$$

= $\frac{1}{2\beta}R_{k,r}(z_1, ..., z_k)$

is equivalent to a **linear system with** r^k variables. Solution:

$$T_{k,r}(z_1,...,z_k) = -\int_0^\infty R_{k,r}(e^{A_{\Pi,r}t}z_1,...,e^{A_{\Pi,r}t}z_k) dt.$$

An approximation is given by:

$$\sum_{i=-\ell}^{\ell} w_i R_{k,r}(e^{A_{\Pi,r}t_i}z_1,...,e^{A_{\Pi,r}t_i}z_k),$$

for an appropriate choice of points t_i and weights w_i .

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Fokker-Planck equation

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Controlled Fokker-Planck equation:

$$\begin{aligned} \frac{\partial \rho}{\partial t} &= \nu \Delta \rho + \nabla \cdot (\rho \nabla G) + u \nabla \cdot (\rho \nabla \alpha_j) & \text{in } \Omega \times (0, \infty), \\ 0 &= (\nu \nabla \rho + \rho \nabla G) \cdot \vec{n} & \text{on } \Gamma \times (0, \infty), \\ \rho(x, 0) &= \rho_0(x) & \text{in } \Gamma, \end{aligned}$$

where $\Omega \in \mathbb{R}^d$ denotes a bounded domain with smooth boundary Γ . For all t, $\rho(\cdot, t)$ is the probability density function of X_t , sol. to

$$\mathsf{d}X(t) = -
abla_{\mathsf{X}} \mathsf{V}(\mathsf{X}(t), t) \mathsf{d}t + \sqrt{2
u} \mathsf{d}W_t,$$

where the **potential** V is controlled by u:

 $V(x,t) = G(x) + u(t)\alpha(x), \quad \forall x \in \Omega, \ \forall t \ge 0.$

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Fokker-Planck equation

The uncontrolled Fokker-Planck equation is known to converge to its stationary distribution ρ_{∞} .



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Fokker-Planck equation

Optimal control problem:

 $\inf_{u\in L^2(0,\infty)}\int_0^\infty \frac{1}{2}\|\rho(\cdot,t)-\rho_\infty(\cdot)\|_{L^2(\Omega)}^2+\beta|u(t)|^2\mathrm{d}t,$

where ρ satisfies the Fokker-Planck equation.

Under regularity assumptions on G and α , the problem can be reformulated, so that it falls in the abstract framework.

- Control shape function $\alpha(x) \approx x/12$.
- Discretization of $\Omega = (-6, 6)$: n = 100.
- Reduction: r = 21 (selection of singular values above 10^{-6}).
- Results for two initial values (a close one/a further one), different values of β.

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Numerical results (test case 1)		



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Numerical results	(test case 1)		

β	$J(u_2)$	$J(u_3)$	$J(u_4)$	$J(u_5)$	$J(u_6)$	$J(u_{opt})$
1e ⁻³	0.156	0.155	0.155	0.155	0.155	0.154
1e ⁻⁴	0.138	0.122	0.120	0.120	0.120	0.119
1e ⁻⁵	0.205	0.194	0.104	0.111	0.113	0.095

(a) Cost of the controls u_k

в	$\ u_k - u_{opt}\ _{L^2(0,T)}$				
ρ	<i>p</i> = 2	<i>p</i> = 3	<i>p</i> = 4	<i>p</i> = 5	<i>p</i> = 6
1e ⁻³	1.149	0.169	0.119	0.034	0.031
$1e^{-4}$	18.50	7.02	3.16	4.01	1.52
$1e^{-5}$	90.5	78.0	39.0	42.6	34.3

(b) L^2 -distance between the controls u_k and the optimal control u_{opt}

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β	$J(u_2)$	$J(u_3)$	$J(u_4)$	$J(u_5)$	$J(u_6)$	$J(u_{opt})$
1e ⁻²	0.788	0.788	0.788	0.788	0.788	0.787
$1e^{-3}$	0.525	0.511	0.511	0.512	0.510	0.507
1e ⁻⁴	0.381	0.368	2.689	∞	$ \infty$	0.246

(a) Cost of the controls u_k

в	$\ u_k - u_{opt}\ _{L^2(0,T)}$				
Ρ	k = 2	<i>k</i> = 3	<i>k</i> = 4	<i>k</i> = 5	<i>k</i> = 6
$1e^{-2}$	0.19	0.15	0.15	0.15	0.15
$1e^{-3}$	4.88	1.50	1.77	2.31	1.52
$1e^{-4}$	46.34	35.36	57.08	∞	∞

(b) L^2 -distance between the controls u_k and the optimal control u_{opt}

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Theorem

There exists $\delta > 0$ such that

- for all $y_0 \in B(\delta)$, problem $P(y_0)$ has a unique solution \bar{u} ,
- the value function \mathcal{V} is infinitely differentiable on $B(\delta)$.

For all $k \ge 2$, there exist $\delta > 0$ and C > 0 such that:

- The closed-loop system (of order k) is well-posed and generates an open-loop control in L²(0,∞).
- The following estimates hold true:

 $\mathcal{J}(\mathbf{U}_{k}(y_{0}), y_{0}) \leq \mathcal{V}(y_{0}) + C \|y_{0}\|_{Y}^{2k}$ $\|\bar{u} - \mathbf{U}_{k}(y_{0})\|_{L^{2}(0,\infty)} \leq C \|y_{0}\|_{Y}^{k}.$

Remark: **local result**, δ and *C* depend on *k*.

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Result 1 (optimality conditions for the original problem). For all solutions \bar{u} with trajectory \bar{y} , there exists $\bar{p} \in W(0,\infty)$ such that

 $\dot{\bar{p}} + (A + \bar{u}N)^* \bar{p} + \bar{y} = 0, \quad \beta \bar{u} + (Ny + B)^* \bar{p} = 0.$

Result 2 (optimality conditions for the closed loop system). For the control u_k and the trajectory y_k generated by the feedback of order k, there exists $p_k \in L^2(0, \infty; V)$ such that

 $\dot{p}_k + (A + u_k N)^* p_k + y_k = w_k, \quad \beta u_k + (Ny_k + B)^* p_k = 0,$

where $||w_k|| \le C ||y_0||_Y^k$.

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Result 3 (sensitivity analysis).

The mapping $\Phi \colon (y, u, p) \in W(0, \infty) \times L^2(0, \infty) \times L^2(0, \infty; V) \mapsto$

$$\Phi(y, u, p) = \begin{pmatrix} y(0) \\ \dot{y} - (Ay + Nyu + Bu) \\ -\dot{p} - (A + uN)^*p - y \\ \beta u + (Ny + B)^*p \end{pmatrix}$$

is locally invertible around (0,0,0), with a ${\ensuremath{C^\infty}}$ inverse.

Proof: application of the inverse mapping theorem.

$$D\Phi(0,0,0)(\delta y, \delta u, \delta p) = (\omega_1, \omega_2, \omega_3, \omega_4)$$

$$\iff \begin{pmatrix} \delta y(0) = \omega_1 \\ \delta \dot{y} = A \delta y + B \delta u + \omega_2 \\ -\delta \dot{p} = A^* \delta p + \delta y + \omega_3 \\ \beta \delta u + B^* \delta p = \omega_4 \end{pmatrix}$$

$$\iff (\delta y, \delta u) \text{ unique sol. of a LQ problem.}$$

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Conclusion (for $||y_0||$ small enough).

• (\bar{y}, \bar{u}) is a solution to $P(y_0)$ with costate \bar{p} implies

 $\Phi(\bar{y},\bar{u},\bar{p}) = (y_0,0,0,0) \Longleftrightarrow (\bar{y},\bar{u},\bar{p}) = \Phi^{-1}(y_0,0,0,0).$

Uniqueness and smoothness of \mathcal{V} follow.

• (y_k, u_k, p_k) is as in Step 2 implies

 $\Phi(y_k, u_k, p_k) = (y_0, 0, w_k, 0) \iff (y_k, u_k, p_k) = \Phi^{-1}(y_0, 0, w_k, 0).$

Error estimate:

$$\begin{split} \|(y_k, u_k, p_k) - (\bar{y}, \bar{u}, \bar{p})\| &= \|\Phi^{-1}(y_0, 0, w_k, 0) - \Phi^{-1}(y_0, 0, 0, 0)\| \\ &\leq C \|w_k\| \leq C \|y_0\|_Y^k. \end{split}$$

Taylor expansions and feedback laws	Numeric results	Elements of analysis	Receding-horizon algorithm

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1 Taylor expansions and feedback laws

2 Numeric results

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4 Receding-horizon algorithm

Taylor expansions and feedback laws	Numeric results	Elements of analysis 0000	Receding-horizon algorithm ●000
Introduction			

Main result: an upper bound of

 $\|y_{RH} - \bar{y}\|_{W(0,\infty)} + \|u_{RH} - \bar{u}\|_{L^2(0,\infty)},$ where:

- (\bar{y}, \bar{u}) is the solution to $P(y_0)$
- (y_{RH}, u_{RH}) is an approximate solution obtained with the Receding-Horizon method (= Model Predictive Control).

We aim at analyzing the effect of

- the sampling time au
- the prediction horizon T
- the penalty function ϕ .

Taylor expansions and feedback laws	Numeric results	Elements of analysis	Receding-horizon algorithm 0●00
Algorithm			

Main idea of the RHC method: replace $P(y_0)$ by a sequence of (tractable) finite-horizon problems.

For a given terminal cost function $\phi \colon Y \to \mathbb{R}$, consider the truncated problem

$$\inf_{u \in L^{2}(0,\infty)} \int_{0}^{T} \frac{1}{2} \|y(t)\|_{Y}^{2} + \frac{\beta}{2} |u(t)|^{2} dt + \phi(y(T)),$$

$$\text{where:} \quad \left\{ \begin{array}{l} \dot{y}(t) = Ay(t) + Ny(t)u(t) + Bu(t), \\ y(0) = y_{\text{init}} \in Y, \end{array} \right.$$

$$(P_{T,\phi}(y_{\text{init}}))$$

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Taylor expansions and feedback laws	Numeric results	Elements of analysis	Receding-horizon algorithm 00●0
Algorithm			

Method.

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- **1** Set n = 0.
- **2** Compute a solution (y, u) to $P_{T,\phi}(y_n)$.
- **3** Set $u_{RH}(t) = u(n\tau + t)$, $y_{RH}(t) = y(n\tau + t)$ for $t \in (0, \tau)$.
- 4 Set $y_{n+1} = y_{RH}((n+1)\tau)$, n = n + 1, and go back to Step 2.

Remark

- If V is used as a terminal cost, then by the dynamic programming principle, the RH-algorithm generates the exact solution to the problem.
- Limit case when (τ, T) → 0: Feedback control. Limit case when (τ, T) → ∞: Open-loop control.

Taylor	expansions	and	feedback	

Numeric results

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Result

Theorem

For all $k \ge 1$, there exist $\tau_0 > 0$, $\delta > 0$, and M > 0 such that for all $\tau \ge \tau_0$, for all $T \ge \tau$, and all $y_0 \in B_Y(\delta)$, the RHC method with

 $\phi = \mathcal{V}_k$

is well-posed. Moreover,

 $\|y_{RH} - \bar{y}\|_{W_{\infty}} + \|u_{RH} - \bar{u}\|_{L^{2}(0,\infty)} \le Me^{-\lambda(T-\tau)-\lambda kT}\|y_{0}\|_{Y}^{k}$

where \bar{u} is the unique solution to the problem with trajectory \bar{y} .

Proof: based on a sensitivity analysis.

Taylor expansions and feedback laws	Numeric results	Elements of analysis	Receding-horizon algorithm
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Conclusion			

Summary:

- General method for deriving polynomial feedback laws
- Implementation for an infinite-dimensional problem thanks to model reduction
- Good results, but only locally.
- Theoretical result for the RHC method.

Extensions:

- Other systems, with different non-linearities.
- Analysis of other kind of feedback mechanisms (e.g. SDRE).

 Analysis of other kind of problems (e.g. problems with turnpike property).

Taylor expansions and feedback laws	Numeric results	Elements of analysis 0000	Receding-horizon algorithm 0000
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Thank you for your attention!