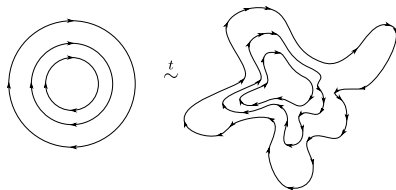


On Topological Equivalence in Linear Quadratic Optimal Control



15th International Young Researchers Workshop on Geometry,
Mechanics, and Control — November 30, 2020,

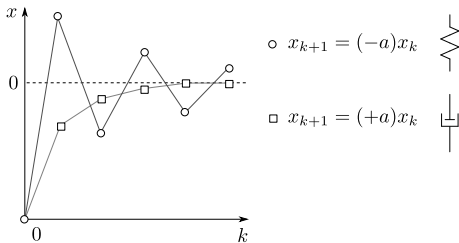
Wouter Jongeneel, Daniel Kuhn

École polytechnique fédérale de Lausanne | RAO
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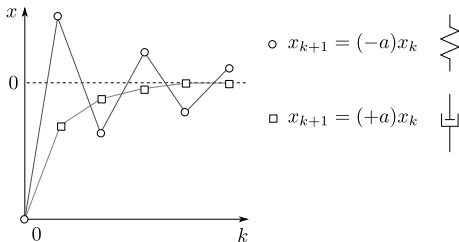
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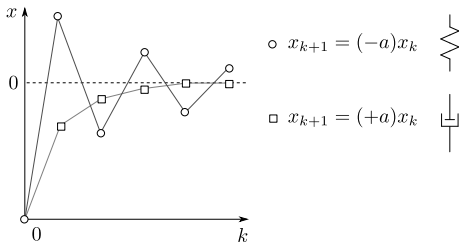
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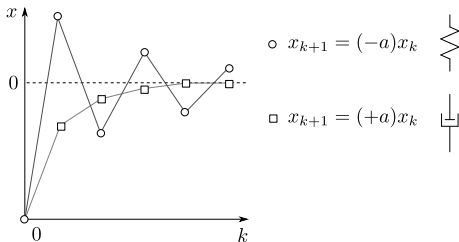
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Regardless of choice of coordinates, we see a structural difference.

Structure in Optimal Control

We will look at (discrete-time) dynamical systems of the form

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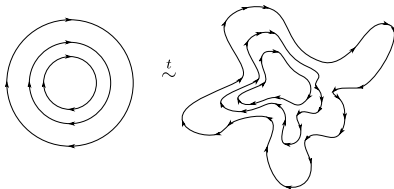
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We will observe structural equivalences (or the lack thereof) in optimal control problems:

$$\arg \min_{f \in \mathcal{F}} J_1(f) \sim \arg \min_{f \in \mathcal{F}} J_2(f) \quad \forall J_1, J_2 \in \mathcal{J}.$$

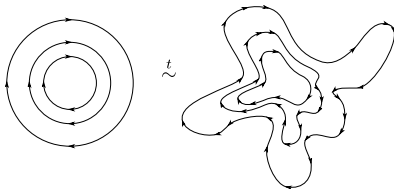
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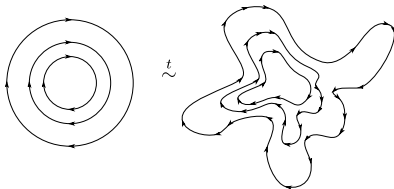
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- ▶ Hard in continuous-time, since for $\dot{x} = f(x)$ and $\dot{y} = g(y)$ one needs to work with their *flows*.
- ▶ In discrete-time, for $x_{k+1} = f(x_k)$, $y_{k+1} = g(y_k)$ seek φ such that $y = \varphi(x)$ relates trajectories, that is, $f = \varphi^{-1} \circ g \circ \varphi$ must hold.

Topological Equivalence (2/2)

Definition (Topological Equivalence): Two endomorphisms $f : \mathcal{V} \rightarrow \mathcal{V}$ and $g : \mathcal{W} \rightarrow \mathcal{W}$ over topological vector spaces \mathcal{V} and \mathcal{W} are topologically equivalent (conjugate), denoted $f \overset{t}{\sim} g$, if and only if there exists a homeomorphism $\varphi : \mathcal{V} \rightarrow \mathcal{W}$ such that $g \circ \varphi = \varphi \circ f$, that is, the diagram

$$\begin{array}{ccc} \mathcal{V} & \xrightarrow{f} & \mathcal{V} \\ \varphi \downarrow & & \downarrow \varphi \\ \mathcal{W} & \xrightarrow{g} & \mathcal{W} \end{array}$$

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$f \stackrel{t}{\sim} g$ if a C^0 change of coordinates relates their orbits.

The Scalar Setting, 7 classes

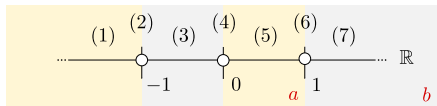
Consider for some scalar a and b

$$x \mapsto f(x), \quad f(x) := ax, \quad y \mapsto g(y), \quad g(y) := by.$$

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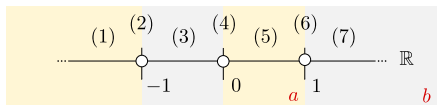
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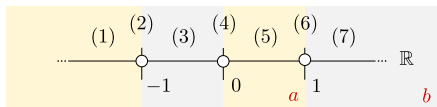
Proposition (Topological Equivalence of Scalar Systems [Kuiper and Robbin 1973, Proposition 1.5]): *Let a and b be members of the same class in \mathbb{R} (see Figure), then $g = \varphi \circ f \circ \varphi^{-1}$ for*

$$\varphi(x) = x|x|^{c-1}, \quad c = \log(|b|)/\log(|a|).$$

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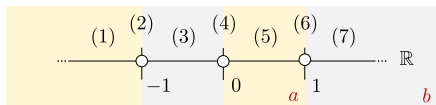
$$\varphi(x) = x|x|^{c-1}, \quad c = \log(|b|)/\log(|a|).$$

So for example $f(x) = 2x$ and $g(y) = 8y$ are in class (7) and related by the C^ω map $\varphi(x) = x^3$ while $\varphi^{-1} = x^{1/3}$, which is merely C^0 over \mathbb{R} .

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Similarity transformations would yield a continuum of systems!

A Remark

Building upon Poincaré, Birkhoff, ... The concept of topological equivalence shows, for example, up in the Hartman-Grobman theorem and is key in bifurcation theory. The hope is to study a *finite* amount of classes instead of a continuum.

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Is this true? "*... the classification of all phase portraits on a given manifold, ... up to equivalence under homeomorphisms ... Although some results have been obtained ... it became clear rather early that this program was too ambitious.*" Abraham and Marsden 1978.

For Today

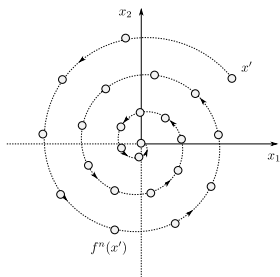
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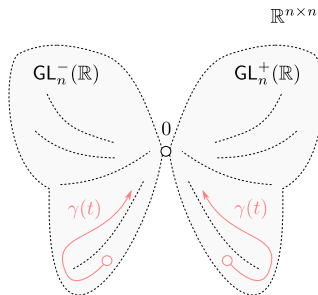
Better yet, focus on the *automorphic* part of $f(x)$, we will look at *invertible* maps $f(x) = Fx$.



The Real General Linear Group

Let $GL_n(\mathbb{R}) := \{A \in \mathbb{R}^{n \times n} : \det(A) \neq 0\}$, with

- ▶ n^2 -dim C^ω manifold, dense in $\mathbb{R}^{n \times n}$
- ▶ $GL_n(\mathbb{R}) = GL_n^+(\mathbb{R}) \overset{\circ}{\cup} GL_n^-(\mathbb{R})$, e.g., $\det(X) > 0 \iff X \in GL_n^+(\mathbb{R})$.
- ▶ Exponential map not surjective.
- ▶ $GL_n^{(i)}(\mathbb{R}), \forall i \in \{+, -\}$ path-connected, yet, not simply-connected.

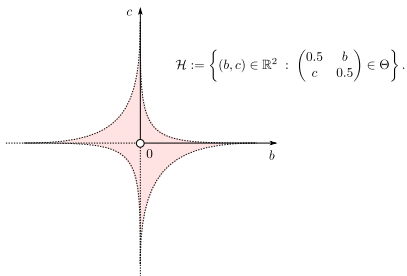


The set of **Asymptotically stable matrices**

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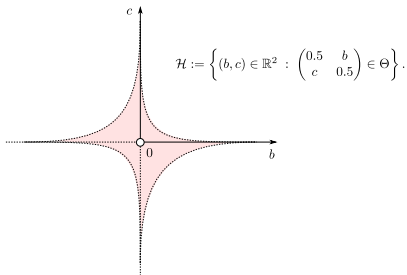
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Let $\Theta := \{\theta \in \mathbb{R}^{n \times n} : \rho(\theta) < 1\}$ be the set of *asymptotically stable* matrices. Θ is a semi-algebraic non-convex (star-convex) set.

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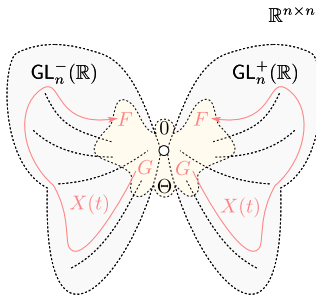


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We will usually talk about $F \in \text{GL}_n(\mathbb{R}) \cap \Theta$, that is $\lim_{n \rightarrow \infty} f^n(x) = 0 \forall x$.

Characterizing Equivalence

Theorem (Topological Equivalence of Asymptotically Stable Systems [Robinson 1995, Theorem 9.2 page 117]): *Let $f(x) := Fx$ and $g(y) := Gy$ be asymptotically stable linear automorphisms on \mathbb{R}^n . Moreover, let $X(t)$ parametrize a path in $GL_n(\mathbb{R})$, continuously depending on $t \in [0, 1]$, such that $X(0) = F$ and $X(1) = G$, then, $f \stackrel{t}{\sim} g$.*



Linear Quadratic Optimal Control (1/3)

The general (discrete-time) Linear Quadratic Optimal Control problem is given by

$$\begin{aligned} & \underset{\{u_k\}_{k \geq 0}}{\text{minimize}} && \sum_{k=0}^{\infty} \begin{pmatrix} x_k \\ u_k \end{pmatrix}^{\top} \begin{pmatrix} Q & S \\ S^{\top} & R \end{pmatrix} \begin{pmatrix} x_k \\ u_k \end{pmatrix} \\ & \text{subject to} && x_{k+1} = Ax_k + Bu_k, \quad x_0 = x' \end{aligned}$$

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If (A, B, C) , $C^T C := Q$, is a *minimal* realization, $R \succ 0$, then, $u_k^* = K^* x_k$

$$\begin{aligned} P &= Q + A^T (P - PB(R + B^T PB)^{-1} B^T P) A, \\ K^* &= -(R + B^T PB)^{-1} B^T P A. \end{aligned}$$

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“*Tuning*” the pair (Q, R) ?

Linear Quadratic Optimal Control (2/3)

If in

$$c(x, u) = \begin{pmatrix} x \\ u \end{pmatrix}^\top \begin{pmatrix} Q & S \\ S^\top & R \end{pmatrix} \begin{pmatrix} x \\ u \end{pmatrix}.$$

$S \neq 0$ “diagonalize” via $v := R^{-1}S^\top x + u$, $Q' = Q - SR^{-1}S^\top$ and $A' := A - BR^{-1}S^\top$, obtain the standard LQ problem:

$$\begin{aligned} & \text{minimize}_{\{v_t\}_{t \geq 0}} \sum_{k=0}^{\infty} \underbrace{x_k^\top Q' x_k + v_k R v_k}_{\text{stage cost } c'(x_k, v_k)} \\ & \text{subject to } \underbrace{x_{k+1} = A' x_k + B v_k}_{x_k \mapsto \sigma'(x_k, v_k)}, \quad x_0 = x' \end{aligned}$$

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For a sensible solution we need the same conditions as before, yet, additionally $\text{rank}(Q) = \text{rank}(Q')$, otherwise we solve a different problem.

Linear Quadratic Optimal Control (3/3)

Let $\underbrace{(A, B)}_{x \mapsto \sigma(x, u)}$ be a stabilizable pair, then regarding the cost, we look at the set

$$\mathcal{C}(\sigma) := \left\{ (Q, R, S) \in \mathcal{S}_{\geq 0}^n \times \mathcal{S}_{> 0}^m \times \mathbb{R}^{n \times m} : \begin{array}{l} \text{rank}(Q) = \text{rank}(Q'), \\ \exists C \in \mathbb{R}^{p \times n} : C^T C = Q, \\ (A, C) \text{ detectable} \end{array} \right\}.$$

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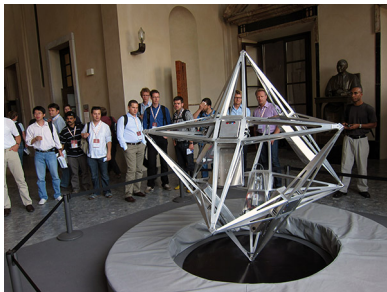
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“*Tuning*” $(Q, R, S) \in \mathcal{C}(\sigma)$? Example: Trimpe and D’Andrea 2012, their motivation was to penalize $\|u_k - u_{k-1}\|$



idsc.ethz.ch/research-dandrea/research-projects/archive/balancing-cube

Topological Perspective on Tuning (1/2)

Introduce an orientation-dependent version of $\mathcal{C}(\sigma)$ Given a $\sigma \in \Sigma$ such that $A \in \text{GL}(n, \mathbb{R})$ and define $\mathcal{C}^{(i)}(\sigma)$, $(i) \in \{+, -\}$ by

$$\mathcal{C}^{(i)}(\sigma) := \left\{ (Q, R, S) \in \mathcal{C}(\sigma) : \begin{array}{l} A \in \text{GL}_n^{(i)}(\mathbb{R}), \\ \underbrace{A'}_{A - BR^{-1}S^T} \in \text{GL}_n^{(i)}(\mathbb{R}) \end{array} \right\}.$$

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Theorem (Topological Equivalence in LQ regulation, [JK20]): Let $A \in \text{GL}_n^{(i)}(\mathbb{R})$, $(i) \in \{+, -\}$

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- ▶ Consequences for Inverse Optimal Control, given any $K' \in \mathbb{R}^{m \times n}$:

$$x^T (K - K')^T (K - K') x = \begin{pmatrix} x \\ Kx \end{pmatrix}^T \begin{pmatrix} K'^T K' & -K'^T \\ -K' & I_m \end{pmatrix} \begin{pmatrix} x \\ Kx \end{pmatrix}.$$

The Real Symplectic Group

A different point of view.

Let the bilinear form $\omega : \mathbb{R}^{2n} \times \mathbb{R}^{2n} \rightarrow \mathbb{R}$ be defined as $\omega(x, y) = x^T \Omega y$ for $\Omega \in \mathbb{R}^{2n \times 2n}$ as given by

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Then, define the real *Symplectic group* by

$$\mathrm{Sp}(2n, \mathbb{R}) := \{M \in \mathbb{R}^{2n \times 2n} : M^T \Omega M = \Omega\}.$$

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Define $M \in \text{Sp}(2n, \mathbb{R})$ (Hamiltonian) by

$$M := \begin{bmatrix} A' + BR^{-1}B^T A'^{-T} Q' & -BR^{-1}B^T A'^{-T} \\ -A'^{-T} Q' & A'^{-T} \end{bmatrix}.$$

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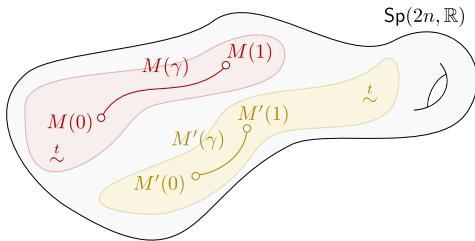
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Symplectic Perspective (2/2)

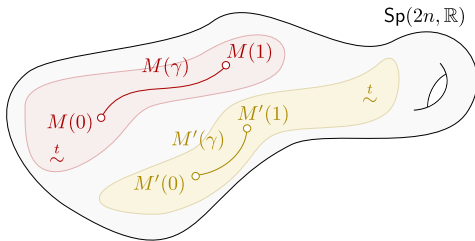
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The asymptotically stable automorphic case.

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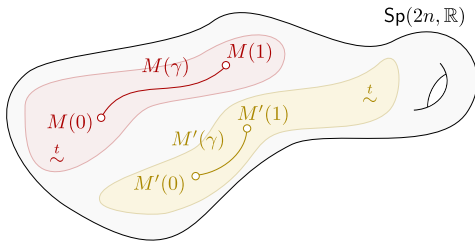


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Can this be generalized to other Hamiltonians?

Example: Bifurcation by tuning

Consider the LQR problem for $B = I_2$, $R = I_2$, $Q = 10 \cdot I_2$ and

$$A = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}, \quad S(\varepsilon) = \begin{pmatrix} \varepsilon & 0 \\ 0 & -\varepsilon \end{pmatrix}, \quad \varepsilon \in \mathbb{R}_{\geq 0}.$$

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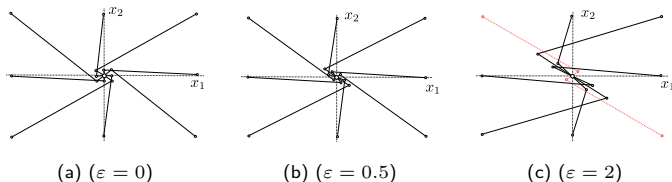


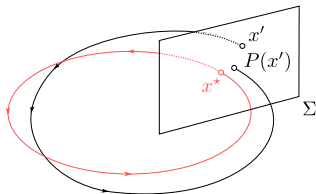
Figure: A few closed-loop trajectories as a function of $S(\varepsilon)$.

Example: Poincaré maps (1/3)

In continuous-time: $\dot{x} = Ax$, $x(0) = x'$ leads to $x(t) = e^{At}x'$. Sampling ($\mu > 0$) then leads to $x \mapsto e^{A\mu}x$: *orientation preserving* map.

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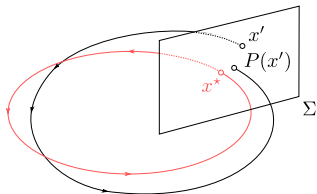
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What about controlled Poincaré maps?

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Consider the affine dynamical control system given by:

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Since $1 = P(1, 0)$, linearize around $r = 1$, that is, for $\xi := r - 1$ obtain the local linear model:

$$\xi_{k+1} = e^{4\pi} \xi_k + \left(\frac{1}{2}e^{4\pi} - \frac{1}{2}\right) u_k =: A_\xi \xi_k + B_\xi u_k.$$

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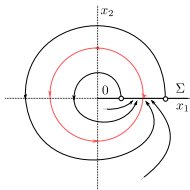
Design for $(Q, R) = (1, 1)$ the LQR gain K_ξ^* . Compare with another *stabilizing* gain \tilde{K} , satisfying, $|K_\xi^* - \tilde{K}| < 3.5 \cdot 10^{-6}$.

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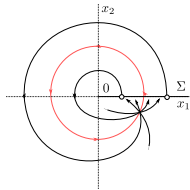
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(b) Trajectories under \tilde{K} .

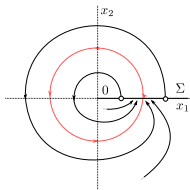
Figure: One controlled cycle for both K_ξ^* and \tilde{K} .

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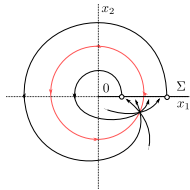
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Periodic version of damper vs. spring.

More family members

The result extends to the whole *family* of Linear Quadratic (LQ) optimal control problems, that is, dynamic games, LEQR, H_∞ -control, etc.

After the Ending

Based on:

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[JK20] Wouter Jongeneel and Daniel Kuhn (2020), *On Topological Equivalence in Linear Quadratic Optimal Control* <http://wjongeneel.nl/pub/TopoEquiv.pdf>.

For more information, see wjongeneel.nl or rao.epfl.ch