

Strict Dissipativity for Multiobjective Optimal Control via Weighted Sums

Lars Grüne, Lisa Krügel, Matthias A. Müller

Mathematisches Institut, Universität Bayreuth
Institut für Regelungstechnik, Leibniz Universität Hannover

supported by  Deutsche
Forschungsgemeinschaft

Workshop on
“Trends on dissipativity in systems and control”
Brig, Switzerland, 23-25 May 2022

Setting

Control system in discrete time

$$x_{\mathbf{u}}(k+1) = f(x_{\mathbf{u}}(k), u(k)), \quad f : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^n, \quad n \in \mathbb{N}_0,$$

$\mathbf{u} = (u(0), u(1), \dots)$, initial condition $x_{\mathbf{u}}(0) = x_0$

Setting

Control system in discrete time

$$x_{\mathbf{u}}(k+1) = f(x_{\mathbf{u}}(k), u(k)), \quad f : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^n, \quad n \in \mathbb{N}_0,$$

$\mathbf{u} = (u(0), u(1), \dots)$, initial condition $x_{\mathbf{u}}(0) = x_0 \rightsquigarrow x_{\mathbf{u}}(k, x_0)$,

Setting

Control system in discrete time

$$x_{\mathbf{u}}(k+1) = f(x_{\mathbf{u}}(k), u(k)), \quad f : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^n, \quad n \in \mathbb{N}_0,$$

$\mathbf{u} = (u(0), u(1), \dots)$, initial condition $x_{\mathbf{u}}(0) = x_0 \rightsquigarrow x_{\mathbf{u}}(k, x_0)$,

admissible state and control spaces $\mathbb{X} \subseteq \mathbb{R}^n$, $\mathbb{U} \subseteq \mathbb{R}^m$

Setting

Control system in **discrete time**

$$x_{\mathbf{u}}(k+1) = f(x_{\mathbf{u}}(k), u(k)), \quad f : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^n, \quad n \in \mathbb{N}_0,$$

$\mathbf{u} = (u(0), u(1), \dots)$, initial condition $x_{\mathbf{u}}(0) = x_0 \rightsquigarrow x_{\mathbf{u}}(k, x_0)$,

admissible state and control spaces $\mathbb{X} \subseteq \mathbb{R}^n$, $\mathbb{U} \subseteq \mathbb{R}^m$

$$J_i^N(x_0, \mathbf{u}) := \sum_{k=0}^{N-1} \ell_i(x_{\mathbf{u}}(k, x_0), u(k)) + F_i(x_{\mathbf{u}}(N, x_0)), \quad i = 1, \dots, s,$$

stage costs $\ell_i : \mathbb{X} \times \mathbb{U} \rightarrow \mathbb{R}$, **terminal costs** $F_i : \mathbb{X}_0 \rightarrow \mathbb{R}$

Setting

Control system in discrete time

$$x_{\mathbf{u}}(k+1) = f(x_{\mathbf{u}}(k), u(k)), \quad f : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^n, \quad n \in \mathbb{N}_0,$$

$\mathbf{u} = (u(0), u(1), \dots)$, initial condition $x_{\mathbf{u}}(0) = x_0 \rightsquigarrow x_{\mathbf{u}}(k, x_0)$,

admissible state and control spaces $\mathbb{X} \subseteq \mathbb{R}^n$, $\mathbb{U} \subseteq \mathbb{R}^m$

$$J_i^N(x_0, \mathbf{u}) := \sum_{k=0}^{N-1} \ell_i(x_{\mathbf{u}}(k, x_0), u(k)) + F_i(x_{\mathbf{u}}(N, x_0)), \quad i = 1, \dots, s,$$

stage costs $\ell_i : \mathbb{X} \times \mathbb{U} \rightarrow \mathbb{R}$, terminal costs $F_i : \mathbb{X}_0 \rightarrow \mathbb{R}$

$$\begin{array}{l} \text{Task: } \min_{\mathbf{u}} J^N(x_0, \mathbf{u}) = \min_{\mathbf{u}} (J_1^N(x_0, \mathbf{u}), \dots, J_s^N(x_0, \mathbf{u})) \\ \text{s.t. } \left. \begin{array}{l} x_{\mathbf{u}}(k, x_0) \in \mathbb{X}, \quad k = 0, \dots, N-1, \\ x_{\mathbf{u}}(N, x_0) \in \mathbb{X}_0, \\ u(k) \in \mathbb{U}, \quad k = 0, \dots, N-1 \end{array} \right\} \mathbf{u} \in \mathbb{U}^N(x_0) \end{array}$$

Optimality in Multiobjective (MO) Optimization

What does 'min' $\mathbf{u} \in \mathbb{U}^N(x_0)$ $(J_1^N(x_0, \mathbf{u}), \dots, J_s^N(x_0, \mathbf{u}))$ mean?

\rightsquigarrow Concept of optimality:

A sequence $\mathbf{u}^* \in \mathbb{U}^N(x_0)$ is called **efficient** (or **Pareto optimal**) if there is no $\mathbf{u} \in \mathbb{U}^N(x_0)$ such that

$$\begin{aligned} \forall i \in \{1, \dots, s\} : J_i^N(x_0, \mathbf{u}) &\leq J_i^N(x_0, \mathbf{u}^*) \text{ and} \\ \exists i \in \{1, \dots, s\} : J_i^N(x_0, \mathbf{u}) &< J_i^N(x_0, \mathbf{u}^*) \end{aligned}$$

Optimality in Multiobjective (MO) Optimization

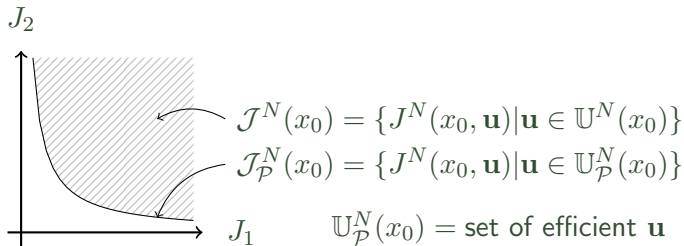
What does 'min' $\mathbf{u} \in \mathbb{U}^N(x_0) (J_1^N(x_0, \mathbf{u}), \dots, J_s^N(x_0, \mathbf{u}))$ mean?

\rightsquigarrow Concept of optimality:

A sequence $\mathbf{u}^* \in \mathbb{U}^N(x_0)$ is called **efficient** (or **Pareto optimal**) if there is no $\mathbf{u} \in \mathbb{U}^N(x_0)$ such that

$$\forall i \in \{1, \dots, s\} : J_i^N(x_0, \mathbf{u}) \leq J_i^N(x_0, \mathbf{u}^*) \text{ and}$$

$$\exists i \in \{1, \dots, s\} : J_i^N(x_0, \mathbf{u}) < J_i^N(x_0, \mathbf{u}^*)$$



Multiobjective Model Predictive Control

The multiobjective optimisation problem “choose $\mathbf{u}^* \in \mathbb{U}_{\mathcal{P}}^N(x)$ ” can be used as a building block for **multiobjective MPC**:

0. Choose $\mathbf{u}_{x(0)}^* \in \mathbb{U}_{\mathcal{P}}^N(x(0))$, set $n := 0$ and go to 2.
1. Measure $x(n)$ and choose $\mathbf{u}_{x(n)}^* \in \mathbb{U}_{\mathcal{P}}^N(x(n))$ with suitable properties
2. Apply the feedback $\mu^N(x(n)) := \mathbf{u}_{x(n)}^*(0)$, set $n := n + 1$ and go to 1.

If the “suitable properties” are properly chosen, approximate infinite horizon efficiency can be shown

[Stieler '18, Gr./Stieler '19, Eichfelder/Gr./Krügel/Schießl '22]

Multiobjective Model Predictive Control

The multiobjective optimisation problem “choose $\mathbf{u}^* \in \mathbb{U}_{\mathcal{P}}^N(x)$ ” can be used as a building block for **multiobjective MPC**:

0. Choose $\mathbf{u}_{x(0)}^* \in \mathbb{U}_{\mathcal{P}}^N(x(0))$, set $n := 0$ and go to 2.
1. Measure $x(n)$ and choose $\mathbf{u}_{x(n)}^* \in \mathbb{U}_{\mathcal{P}}^N(x(n))$ with suitable properties
2. Apply the feedback $\mu^N(x(n)) := \mathbf{u}_{x(n)}^*(0)$, set $n := n + 1$ and go to 1.

If the “suitable properties” are properly chosen, approximate infinite horizon efficiency can be shown

[Stieler '18, Gr./Stieler '19, Eichfelder/Gr./Krügel/Schießl '22]

This approach gives **full control** over the choice of the particular efficient solution

Multiobjective Model Predictive Control

The multiobjective optimisation problem “choose $\mathbf{u}^* \in \mathbb{U}_{\mathcal{P}}^N(x)$ ” can be used as a building block for **multiobjective MPC**:

0. Choose $\mathbf{u}_{x(0)}^* \in \mathbb{U}_{\mathcal{P}}^N(x(0))$, set $n := 0$ and go to 2.
1. Measure $x(n)$ and choose $\mathbf{u}_{x(n)}^* \in \mathbb{U}_{\mathcal{P}}^N(x(n))$ with suitable properties
2. Apply the feedback $\mu^N(x(n)) := \mathbf{u}_{x(n)}^*(0)$, set $n := n + 1$ and go to 1.

If the “suitable properties” are properly chosen, approximate infinite horizon efficiency can be shown

[Stieler '18, Gr./Stieler '19, Eichfelder/Gr./Krügel/Schiebl '22]

This approach gives **full control** over the choice of the particular efficient solution, but **requires strong assumptions**, for instance appropriate terminal conditions

Multiobjective Model Predictive Control

The multiobjective optimisation problem “choose $\mathbf{u}^* \in \mathbb{U}_{\mathcal{P}}^N(x)$ ” can be used as a building block for **multiobjective MPC**:

0. Choose $\mathbf{u}_{x(0)}^* \in \mathbb{U}_{\mathcal{P}}^N(x(0))$, set $n := 0$ and go to 2.
1. Measure $x(n)$ and choose $\mathbf{u}_{x(n)}^* \in \mathbb{U}_{\mathcal{P}}^N(x(n))$ with suitable properties
2. Apply the feedback $\mu^N(x(n)) := \mathbf{u}_{x(n)}^*(0)$, set $n := n + 1$ and go to 1.

If the “suitable properties” are properly chosen, approximate infinite horizon efficiency can be shown

[Stieler '18, Gr./Stieler '19, Eichfelder/Gr./Krügel/Schießl '22]

This approach gives **full control** over the choice of the particular efficient solution, but **requires strong assumptions**, for instance appropriate terminal conditions

↪ search for alternative approaches

Weighted sum approach

The simplest way to solve a **multiobjective** optimal control problem with stage costs l_i , $i = 1, \dots, s$, is to solve a standard optimal control problem with stage cost

$$l = \sum_{i=1}^s \nu_i l_i,$$

with weights $\nu_i \geq 0$, $\sum \nu_i = 1$

Weighted sum approach

The simplest way to solve a **multiobjective** optimal control problem with stage costs l_i , $i = 1, \dots, s$, is to solve a standard optimal control problem with stage cost

$$l = \sum_{i=1}^s \nu_i l_i,$$

with weights $\nu_i \geq 0$, $\sum \nu_i = 1$

If the efficient solution set is convex, **all efficient solutions** can be obtained by varying ν_i

Weighted sum approach

The simplest way to solve a **multiobjective** optimal control problem with stage costs l_i , $i = 1, \dots, s$, is to solve a standard optimal control problem with stage cost

$$l = \sum_{i=1}^s \nu_i l_i,$$

with weights $\nu_i \geq 0$, $\sum \nu_i = 1$

If the efficient solution set is convex, **all efficient solutions** can be obtained by varying ν_i

For non-convex efficient solution sets, at least **a subset** can be realised this way

Weighted sum approach

Idea: If we know that

$$\ell = \sum_{i=1}^s \nu_i l_i$$

satisfies the usual assumptions for stability and performance of (economic) MPC schemes, then we can apply **known standard results for MPC** e.g. [Faulwasser/Gr./Müller '18]

Weighted sum approach

Idea: If we know that

$$\ell = \sum_{i=1}^s \nu_i l_i$$

satisfies the usual assumptions for stability and performance of (economic) MPC schemes, then we can apply **known standard results for MPC** e.g. [Faulwasser/Gr./Müller '18]

Most important property: strict dissipativity

Weighted sum approach

Idea: If we know that

$$\ell = \sum_{i=1}^s \nu_i \ell_i$$

satisfies the usual assumptions for stability and performance of (economic) MPC schemes, then we can apply **known standard results for MPC** e.g. [Faulwasser/Gr./Müller '18]

Most important property: strict dissipativity

Question: If strict dissipativity holds for the ℓ_i , does it also hold for ℓ ?

Weighted sum approach

Idea: If we know that

$$\ell = \sum_{i=1}^s \nu_i \ell_i$$

satisfies the usual assumptions for stability and performance of (economic) MPC schemes, then we can apply **known standard results for MPC** e.g. [Faulwasser/Gr./Müller '18]

Most important property: **strict dissipativity**

Question: If strict dissipativity holds for the ℓ_i , does it also hold for ℓ ?

We present our results for $s = 2$ cost functions

$$\rightsquigarrow \ell = \nu \ell_1 + (1 - \nu) \ell_2, \quad \nu \in [0, 1]$$

Strict dissipativity

Dynamics: $x^+ = f(x, u)$

Strict dissipativity

Dynamics: $x^+ = f(x, u)$ $x \in X = \mathbb{R}^n, u \in U = \mathbb{R}^m$

Strict dissipativity

Dynamics: $x^+ = f(x, u)$ $x \in X = \mathbb{R}^n$, $u \in U = \mathbb{R}^m$

Constraints: $(x(k), u(k)) \in \mathbb{Y}$

Strict dissipativity

Dynamics: $x^+ = f(x, u) \quad x \in X = \mathbb{R}^n, u \in U = \mathbb{R}^m$

Constraints: $(x(k), u(k)) \in \mathbb{Y}$

$\mathbb{X} := \{x \in X \mid \text{there is } u \in U \text{ with } (x, u) \in \mathbb{Y}\}$

Strict dissipativity

Dynamics: $x^+ = f(x, u)$ $x \in X = \mathbb{R}^n$, $u \in U = \mathbb{R}^m$

Constraints: $(x(k), u(k)) \in \mathbb{Y}$

$\mathbb{X} := \{x \in X \mid \text{there is } u \in U \text{ with } (x, u) \in \mathbb{Y}\}$

A pair $(x, u) \in \mathbb{Y}$ is an **equilibrium**, if $f(x, u) = x$

An equilibrium $(x^e, u^e) \in \mathbb{Y}$ is **optimal**, if $\ell(x^e, u^e) \leq \ell(x, u)$
for all equilibria $(x, u) \in \mathbb{Y}$

An equilibrium $(x^e, u^e) \in \mathbb{Y}$ is **strictly optimal**, if $\ell(x^e, u^e) < \ell(x, u)$
for all equilibria $(x, u) \in \mathbb{Y}$ with $(x, u) \neq (x^e, u^e)$

Strict dissipativity

Dynamics: $x^+ = f(x, u)$ $x \in X = \mathbb{R}^n$, $u \in U = \mathbb{R}^m$

Constraints: $(x(k), u(k)) \in \mathbb{Y}$

$\mathbb{X} := \{x \in X \mid \text{there is } u \in U \text{ with } (x, u) \in \mathbb{Y}\}$

A pair $(x, u) \in \mathbb{Y}$ is an **equilibrium**, if $f(x, u) = x$

An equilibrium $(x^e, u^e) \in \mathbb{Y}$ is **optimal**, if $\ell(x^e, u^e) \leq \ell(x, u)$
for all equilibria $(x, u) \in \mathbb{Y}$

An equilibrium $(x^e, u^e) \in \mathbb{Y}$ is **strictly optimal**, if $\ell(x^e, u^e) < \ell(x, u)$
for all equilibria $(x, u) \in \mathbb{Y}$ with $(x, u) \neq (x^e, u^e)$

Strict dissipativity: there exists a **storage function**

$\lambda : \mathbb{X} \rightarrow \mathbb{R}$, bounded from below, and $\alpha \in \mathcal{K}_\infty$ with

$$\lambda(f(x, u)) \leq \lambda(x) + \ell(x, u) - \ell(x^e, u^e) - \alpha(\|x - x^e\|)$$

for all $(x, u) \in \mathbb{Y}$ with $f(x, u) \in \mathbb{X}$

Strict dissipativity

Dynamics: $x^+ = f(x, u)$ $x \in X = \mathbb{R}^n$, $u \in U = \mathbb{R}^m$

Constraints: $(x(k), u(k)) \in \mathbb{Y}$

$\mathbb{X} := \{x \in X \mid \text{there is } u \in U \text{ with } (x, u) \in \mathbb{Y}\}$

A pair $(x, u) \in \mathbb{Y}$ is an **equilibrium**, if $f(x, u) = x$

An equilibrium $(x^e, u^e) \in \mathbb{Y}$ is **optimal**, if $\ell(x^e, u^e) \leq \ell(x, u)$
for all equilibria $(x, u) \in \mathbb{Y}$

An equilibrium $(x^e, u^e) \in \mathbb{Y}$ is **strictly optimal**, if $\ell(x^e, u^e) < \ell(x, u)$
for all equilibria $(x, u) \in \mathbb{Y}$ with $(x, u) \neq (x^e, u^e)$

Strict dissipativity: there exists a **storage function**

$\lambda : \mathbb{X} \rightarrow \mathbb{R}$, bounded from below, and $\alpha \in \mathcal{K}_\infty$ with

$$\lambda(f(x, u)) \leq \lambda(x) + \ell(x, u) - \ell(x^e, u^e) - \alpha(\|x - x^e\|)$$

for all $(x, u) \in \mathbb{Y}$ with $f(x, u) \in \mathbb{X}$ [$-\alpha(\|u - u^e\|)$]

Strict dissipativity

Dynamics: $x^+ = f(x, u)$ $x \in X = \mathbb{R}^n, u \in U = \mathbb{R}^m$

Constraints: $(x(k), u(k)) \in \mathbb{Y}$

$\mathbb{X} := \{x \in X \mid \text{there is } u \in U \text{ with } (x, u) \in \mathbb{Y}\}$

A pair $(x, u) \in \mathbb{Y}$ is an **equilibrium**, if $f(x, u) = x$

An equilibrium $(x^e, u^e) \in \mathbb{Y}$ is **optimal**, if $\ell(x^e, u^e) \leq \ell(x, u)$
for all equilibria $(x, u) \in \mathbb{Y}$

An equilibrium $(x^e, u^e) \in \mathbb{Y}$ is **strictly optimal**, if $\ell(x^e, u^e) < \ell(x, u)$
for all equilibria $(x, u) \in \mathbb{Y}$ with $(x, u) \neq (x^e, u^e)$

Strict $[(x, u)-]$ **dissipativity**: there exists a **storage function**
 $\lambda : \mathbb{X} \rightarrow \mathbb{R}$, bounded from below, and $\alpha \in \mathcal{K}_\infty$ with

$$\lambda(f(x, u)) \leq \lambda(x) + \ell(x, u) - \ell(x^e, u^e) - \alpha(\|x - x^e\|)$$

for all $(x, u) \in \mathbb{Y}$ with $f(x, u) \in \mathbb{X}$ $[-\alpha(\|u - u^e\|)]$

Linear quadratic problems

We start with problems with **linear dynamics**

$$x^+ = Ax + Bu$$

and generalised **quadratic costs**

$$\ell_i(x, u) = x^T Q_i x + u^T R_i u + s_i^T x + v_i^T u,$$

$i = 1, 2$, with $Q_i \geq 0$ and $R_i > 0$

Linear quadratic problems

We start with problems with **linear dynamics**

$$x^+ = Ax + Bu$$

and generalised **quadratic costs**

$$\ell_i(x, u) = x^T Q_i x + u^T R_i u + s_i^T x + v_i^T u,$$

$i = 1, 2$, with $Q_i \geq 0$ and $R_i > 0$

Theorem: Assume that \mathbb{Y} is either convex and compact or $\mathbb{Y} = \mathbb{R}^n \times \mathbb{R}^m$. Assume strict dissipativity for both ℓ_1 and ℓ_2 , with optimal equilibria in the interior of \mathbb{Y}

Then **strict dissipativity** holds for $\ell_\nu = \nu \ell_1 + (1 - \nu) \ell_2$ for all $\nu \in [0, 1]$

Idea of proof

Idea of proof: Strict dissipativity holds in the LQ setting if and only if it holds with **linear-quadratic** storage function

$$\lambda_i(x) = x^T P_i x + p_i x$$

with P_i satisfying

$$Q_i + P_i - A^T P_i A > 0.$$

[Gr./Guglielmi '20]

Idea of proof

Idea of proof: Strict dissipativity holds in the LQ setting if and only if it holds with **linear-quadratic** storage function

$$\lambda_i(x) = x^T P_i x + p_i x$$

with P_i satisfying

$$Q_i + P_i - A^T P_i A > 0.$$

[Gr./Guglielmi '20]

This can be used to **build an LQ-storage function** λ_ν with

$$P_\nu = \nu P_1 + (1 - \nu) P_2$$

and p_ν the **Lagrange multiplier** of the problem

$$\min \ell_\nu(x, u) \quad \text{s.t.} \quad f(x, u) = x, \quad (x, u) \in \mathbb{Y}$$

Idea of proof

Idea of proof: Strict dissipativity holds in the LQ setting if and only if it holds with **linear-quadratic** storage function

$$\lambda_i(x) = x^T P_i x + p_i x$$

with P_i satisfying

$$Q_i + P_i - A^T P_i A > 0.$$

[Gr./Guglielmi '20]

This can be used to **build an LQ-storage function** λ_ν with

$$P_\nu = \nu P_1 + (1 - \nu) P_2$$

and p_ν the **Lagrange multiplier** of the problem

$$\min \ell_\nu(x, u) \quad \text{s.t.} \quad f(x, u) = x, \quad (x, u) \in \mathbb{Y}$$

Interestingly, while P_ν is a convex combination of P_1 and P_2 , the vector p_ν in general **depends nonlinearly on ν**

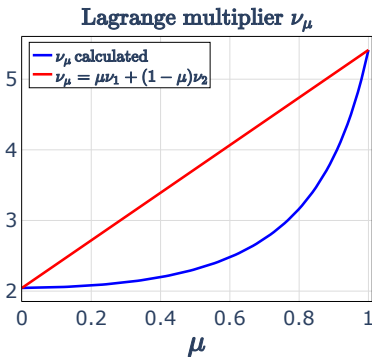
Example

Consider the 1d dynamics $x^+ = 2x + 4u$ with cost functions

$$\ell_1(x, u) = 0.1x^2 + 10u^2 + 6x + 7u$$

and

$$\ell_2(x, u) = 4x^2 + 3u^2 + 3x + 8u$$



Convex problems

Next we consider linear dynamics $x^+ = Ax + Bu$ with nonlinear and (strictly) convex costs

Convex problems

Next we consider linear dynamics $x^+ = Ax + Bu$ with nonlinear and (strictly) convex costs

Theorem: Consider linear dynamics, strictly convex costs ℓ_1 and ℓ_2 , and convex and compact constraint set \mathbb{Y} with optimal equilibria in the interior of \mathbb{Y}

Then the optimal control problem is strictly dissipative for $l_\nu = \nu\ell_1 + (1 - \nu)\ell_2$ for all $\nu \in [0, 1]$

Convex problems

Next we consider **linear dynamics** $x^+ = Ax + Bu$ with nonlinear and **(strictly) convex costs**

Theorem: Consider linear dynamics, strictly convex costs l_1 and l_2 , and convex and compact constraint set \mathbb{Y} with optimal equilibria in the interior of \mathbb{Y}

Then the optimal control problem is **strictly dissipative** for $l_\nu = \nu l_1 + (1 - \nu)l_2$ for all $\nu \in [0, 1]$

Idea of proof: Known: problems with **linear dynamics** and **strictly convex costs** are strictly dissipative. This is the case for $l_\nu = \nu l_1 + (1 - \nu)l_2$.

Convex problems

Next we consider **linear dynamics** $x^+ = Ax + Bu$ with nonlinear and **(strictly) convex costs**

Theorem: Consider linear dynamics, strictly convex costs ℓ_1 and ℓ_2 , and convex and compact constraint set \mathbb{Y} with optimal equilibria in the interior of \mathbb{Y}

Then the optimal control problem is **strictly dissipative** for $\ell_\nu = \nu\ell_1 + (1 - \nu)\ell_2$ for all $\nu \in [0, 1]$

Idea of proof: Known: problems with **linear dynamics** and **strictly convex costs** are strictly dissipative. This is the case for $\ell_\nu = \nu\ell_1 + (1 - \nu)\ell_2$. The storage function is $\lambda(x) = p_\nu x$ with p_ν the Lagrange multiplier as before

Convex problems

Next we consider **linear dynamics** $x^+ = Ax + Bu$ with nonlinear and **(strictly) convex costs**

Theorem: Consider linear dynamics, strictly convex costs ℓ_1 and ℓ_2 , and convex and compact constraint set \mathbb{Y} with optimal equilibria in the interior of \mathbb{Y}

Then the optimal control problem is **strictly dissipative** for $\ell_\nu = \nu\ell_1 + (1 - \nu)\ell_2$ for all $\nu \in [0, 1]$

Idea of proof: Known: problems with **linear dynamics** and **strictly convex costs** are strictly dissipative. This is the case for $\ell_\nu = \nu\ell_1 + (1 - \nu)\ell_2$. The storage function is $\lambda(x) = p_\nu x$ with p_ν the Lagrange multiplier as before

Note: Strict convexity of either ℓ_1 or ℓ_2 can be **relaxed to mere convexity** if only $\nu \in (0, 1)$ is considered

Fully nonlinear problems - sufficient conditions

Now we turn to fully nonlinear problems $x^+ = f(x, u)$

Fully nonlinear problems - sufficient conditions

Now we turn to fully nonlinear problems $x^+ = f(x, u)$

Theorem: Assume strict dissipativity for the cost functions ℓ_1 and ℓ_2 at the same equilibrium x^e

Then the optimal control problem is strictly dissipative for $\ell_\nu = \nu\ell_1 + (1 - \nu)\ell_2$ for all $\nu \in [0, 1]$

Fully nonlinear problems - sufficient conditions

Now we turn to fully nonlinear problems $x^+ = f(x, u)$

Theorem: Assume strict dissipativity for the cost functions ℓ_1 and ℓ_2 at the same equilibrium x^e

Then the optimal control problem is strictly dissipative for $\ell_\nu = \nu\ell_1 + (1 - \nu)\ell_2$ for all $\nu \in [0, 1]$

Idea of proof: In this particular case one checks that $\lambda_\nu = \nu\lambda_1 + (1 - \nu)\lambda_2$ is a storage function for ℓ_ν

However, if the optimal equilibrium (x_ν^e, u_ν^e) “moves” with ν , then this will not work

Fully nonlinear problems - sufficient conditions

Now we turn to fully nonlinear problems $x^+ = f(x, u)$

Theorem: Assume strict dissipativity for the cost functions ℓ_1 and ℓ_2 at the same equilibrium x^e

Then the optimal control problem is strictly dissipative for $\ell_\nu = \nu\ell_1 + (1 - \nu)\ell_2$ for all $\nu \in [0, 1]$

Idea of proof: In this particular case one checks that $\lambda_\nu = \nu\lambda_1 + (1 - \nu)\lambda_2$ is a storage function for ℓ_ν

However, if the optimal equilibrium (x_ν^e, u_ν^e) “moves” with ν , then this will not work

For LQ problems, in this case the ansatz

$$\lambda_\nu = \nu\lambda_1 + (1 - \nu)\lambda_2 + \text{“linear correction”}$$

was successful.

Fully nonlinear problems - sufficient conditions

Now we turn to fully nonlinear problems $x^+ = f(x, u)$

Theorem: Assume strict dissipativity for the cost functions ℓ_1 and ℓ_2 at the same equilibrium x^e

Then the optimal control problem is strictly dissipative for $\ell_\nu = \nu\ell_1 + (1 - \nu)\ell_2$ for all $\nu \in [0, 1]$

Idea of proof: In this particular case one checks that $\lambda_\nu = \nu\lambda_1 + (1 - \nu)\lambda_2$ is a storage function for ℓ_ν

However, if the optimal equilibrium (x_ν^e, u_ν^e) “moves” with ν , then this will not work

For LQ problems, in this case the ansatz

$$\lambda_\nu = \nu\lambda_1 + (1 - \nu)\lambda_2 + \text{“linear correction”}$$

was successful. Does this work for nonlinear problems?

Linear correction

It is known from [Faulwasser/Zanon '18] that under our standard assumptions the **derivative of the storage function** in the equilibrium satisfies

$$D\lambda(x^e) = p,$$

where p is (as before) the **Lagrange multiplier** of the problem

$$\min \ell_\nu(x, u) \quad \text{s.t.} \quad f(x, u) = x, \quad (x, u) \in \mathbb{Y}$$

Linear correction

It is known from [Faulwasser/Zanon '18] that under our standard assumptions the **derivative of the storage function** in the equilibrium satisfies

$$D\lambda(x^e) = p,$$

where p is (as before) the **Lagrange multiplier** of the problem

$$\min \ell_\nu(x, u) \quad \text{s.t.} \quad f(x, u) = x, \quad (x, u) \in \mathbb{Y}$$

This **uniquely determines** the “linear correction” in the formula

$$\lambda_\nu = \nu\lambda_1 + (1 - \nu)\lambda_2 + \text{“linear correction”}$$

Linear correction

It is known from [Faulwasser/Zanon '18] that under our standard assumptions the **derivative of the storage function** in the equilibrium satisfies

$$D\lambda(x^e) = p,$$

where p is (as before) the **Lagrange multiplier** of the problem

$$\min \ell_\nu(x, u) \quad \text{s.t.} \quad f(x, u) = x, \quad (x, u) \in \mathbb{Y}$$

This **uniquely determines** the “linear correction” in the formula

$$\lambda_\nu = \nu\lambda_1 + (1 - \nu)\lambda_2 + \text{“linear correction”}$$

In other words: In order to check whether this ansatz yields a valid storage function, we only need to **check one linear correction**, not many of them

Linear correction — example

$$x^+ = f(x, u) = 2x - x^2 + u + u^2 + u^3$$

with cost functions

$$\ell_1(x, u) = 2x^2 + 0.0001u^2 \quad \text{and} \quad \ell_2(x, u) = 2x^2 + 0.9999u^2 + 2u.$$

Linear correction — example

$$x^+ = f(x, u) = 2x - x^2 + u + u^2 + u^3$$

with cost functions

$$\ell_1(x, u) = 2x^2 + 0.0001u^2 \quad \text{and} \quad \ell_2(x, u) = 2x^2 + 0.9999u^2 + 2u.$$

Strict dissipativity holds for both ℓ_i with linear storage functions λ_i , $i = 1, 2$

Linear correction — example

$$x^+ = f(x, u) = 2x - x^2 + u + u^2 + u^3$$

with cost functions

$$\ell_1(x, u) = 2x^2 + 0.0001u^2 \quad \text{and} \quad \ell_2(x, u) = 2x^2 + 0.9999u^2 + 2u.$$

Strict dissipativity holds for both ℓ_i with linear storage functions λ_i , $i = 1, 2$

For $\nu = 0.5$: Lagrange multiplier $p_\nu = 1.111667$

Linear correction — example

$$x^+ = f(x, u) = 2x - x^2 + u + u^2 + u^3$$

with cost functions

$$\ell_1(x, u) = 2x^2 + 0.0001u^2 \quad \text{and} \quad \ell_2(x, u) = 2x^2 + 0.9999u^2 + 2u.$$

Strict dissipativity holds for both ℓ_i with linear storage functions λ_i , $i = 1, 2$

For $\nu = 0.5$: Lagrange multiplier $p_\nu = 1.111667$

↪ if a storage function $\lambda_\nu = 0.5\lambda_1 + 0.5\lambda_2 +$ “linear correction” exists, then it must be linear, hence $\lambda_\nu = 1.111667x$

Linear correction — example

$$x^+ = f(x, u) = 2x - x^2 + u + u^2 + u^3$$

with cost functions

$$\ell_1(x, u) = 2x^2 + 0.0001u^2 \quad \text{and} \quad \ell_2(x, u) = 2x^2 + 0.9999u^2 + 2u.$$

Strict dissipativity holds for both ℓ_i with linear storage functions λ_i , $i = 1, 2$

For $\nu = 0.5$: Lagrange multiplier $p_\nu = 1.111667$

↪ if a storage function $\lambda_\nu = 0.5\lambda_1 + 0.5\lambda_2 +$ “linear correction” exists, then it must be linear, hence $\lambda_\nu = 1.111667x$

However, for this storage function one checks that dissipativity is violated at $x = x^e$

Fully nonlinear problems — sufficient conditions

Theorem: Under suitable uniform lower bounds on the second derivatives of

$$\tilde{\ell}_\nu(x, u) = \nu \tilde{\ell}_1(x, u) + (1 - \nu) \tilde{\ell}_2(x, u),$$

in $(x, u) = (x_\nu^e, u_\nu^e)$ with rotated costs

$$\tilde{\ell}_i(x, u) := \ell_i(x, u) - \ell_i(x_i^e, u_i^e) + \lambda_i(x) - \lambda_i(f(x, u))$$

and uniform upper bounds on the second derivatives of

$$\tilde{p}_\nu f(x, u)$$

for the linear corrections \tilde{p}_ν , strict dissipativity holds for $\ell_\nu = \nu \ell_1 + (1 - \nu) \ell_2$ for all $\nu \in [0, 1]$

Fully nonlinear problems — sufficient conditions

Theorem: Under suitable uniform lower bounds on the second derivatives of

$$\tilde{\ell}_\nu(x, u) = \nu \tilde{\ell}_1(x, u) + (1 - \nu) \tilde{\ell}_2(x, u),$$

in $(x, u) = (x_\nu^e, u_\nu^e)$ with rotated costs

$$\tilde{\ell}_i(x, u) := \ell_i(x, u) - \ell_i(x_i^e, u_i^e) + \lambda_i(x) - \lambda_i(f(x, u))$$

and uniform upper bounds on the second derivatives of

$$\tilde{p}_\nu f(x, u)$$

for the linear corrections \tilde{p}_ν , strict dissipativity holds for $\ell_\nu = \nu \ell_1 + (1 - \nu) \ell_2$ for all $\nu \in [0, 1]$ with storage function

$$\lambda_\nu(x) = \nu \lambda_1(x) + (1 - \nu) \lambda_2(x) + \tilde{p}_\nu x$$

Fully nonlinear problems — sufficient conditions

Theorem: Under suitable uniform lower bounds on the second derivatives of

$$\tilde{\ell}_\nu(x, u) = \nu \tilde{\ell}_1(x, u) + (1 - \nu) \tilde{\ell}_2(x, u),$$

in $(x, u) = (x_\nu^e, u_\nu^e)$ with rotated costs

$$\tilde{\ell}_i(x, u) := \ell_i(x, u) - \ell_i(x_i^e, u_i^e) + \lambda_i(x) - \lambda_i(f(x, u))$$

and uniform upper bounds on the second derivatives of

$$\tilde{p}_\nu f(x, u)$$

for the linear corrections \tilde{p}_ν , strict dissipativity holds for $\ell_\nu = \nu \ell_1 + (1 - \nu) \ell_2$ for all $\nu \in [0, 1]$ with storage function

$$\lambda_\nu(x) = \nu \lambda_1(x) + (1 - \nu) \lambda_2(x) + \tilde{p}_\nu x$$

Idea of proof: Use KKT conditions

Example

For $x^+ = x^3 - 2x^2 + u$ and the two stage costs

$$\ell_1(x, u) = \ln(5x^{0.34} - u),$$

$$\ell_2(x, u) = \ln(3x^{0.2} - u),$$

we can show that the lower bounds requested in the theorem hold

⇒ **strict dissipativity** holds for all weights $\nu \in [0, 1]$

Fully nonlinear problems — sufficient conditions

Furthermore, we can show by means of the implicit function theorem that under suitable regularity conditions on the optimisation problem for determining the optimal equilibrium, strict (x, u) -dissipativity **persists for small changes** in ν

Fully nonlinear problems — sufficient conditions

Furthermore, we can show by means of the implicit function theorem that under suitable regularity conditions on the optimisation problem for determining the optimal equilibrium, strict (x, u) -dissipativity **persists for small changes** in ν

Can we also find situations where we can prove that **strict dissipativity is lost**?

Fully nonlinear problems — necessary condition

Theorem: Assume strict dissipativity for the cost function $\ell_\nu = \nu\ell_1 + (1 - \nu)\ell_2$ for all $\nu \in [\underline{\nu}, \bar{\nu}] \subseteq [0, 1]$ and that the corresponding optimal equilibria (x_ν^e, u_ν^e) are contained in a compact set $\widehat{Y} \subset Y$. Then the map

$$\nu \mapsto x_\nu^e$$

is continuous on $[\underline{\nu}, \bar{\nu}]$.

Fully nonlinear problems — necessary condition

Theorem: Assume strict dissipativity for the cost function $l_\nu = \nu l_1 + (1 - \nu)l_2$ for all $\nu \in [\underline{\nu}, \bar{\nu}] \subseteq [0, 1]$ and that the corresponding optimal equilibria (x_ν^e, u_ν^e) are contained in a compact set $\hat{\mathbb{Y}} \subset \mathbb{Y}$. Then the map

$$\nu \mapsto x_\nu^e$$

is continuous on $[\underline{\nu}, \bar{\nu}]$.

Idea of proof: Strict dissipativity implies the existence of a strictly globally optimal equilibrium x^e ,

Fully nonlinear problems — necessary condition

Theorem: Assume strict dissipativity for the cost function $l_\nu = \nu l_1 + (1 - \nu)l_2$ for all $\nu \in [\underline{\nu}, \bar{\nu}] \subseteq [0, 1]$ and that the corresponding optimal equilibria (x_ν^e, u_ν^e) are contained in a compact set $\hat{\mathbb{Y}} \subset \mathbb{Y}$. Then the map

$$\nu \mapsto x_\nu^e$$

is continuous on $[\underline{\nu}, \bar{\nu}]$.

Idea of proof: Strict dissipativity implies the existence of a strictly globally optimal equilibrium x^e . However, at any point of discontinuity there are two different optimal equilibria with identical objective value

Fully nonlinear problems — necessary condition

Theorem: Assume strict dissipativity for the cost function $l_\nu = \nu l_1 + (1 - \nu)l_2$ for all $\nu \in [\underline{\nu}, \bar{\nu}] \subseteq [0, 1]$ and that the corresponding optimal equilibria (x_ν^e, u_ν^e) are contained in a compact set $\hat{\mathbb{Y}} \subset \mathbb{Y}$. Then the map

$$\nu \mapsto x_\nu^e$$

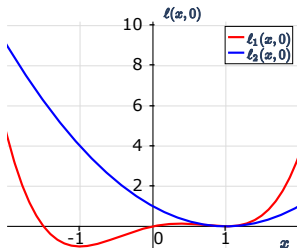
is continuous on $[\underline{\nu}, \bar{\nu}]$.

Idea of proof: Strict dissipativity implies the existence of a strictly globally optimal equilibrium x^e . However, at any point of discontinuity there are two different optimal equilibria with identical objective value. Hence, x^e cannot exist

Example

Consider the dynamics $x^+ = x + u$ and the cost functions

$$\ell_1(x, u) = \frac{1}{2}x^4 - \frac{1}{4}x^3 - x^2 + \frac{3}{4}x + u^2 \quad \text{and} \quad \ell_2(x, u) = (x-1)^2 + u^2$$



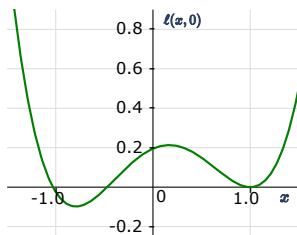
$\ell_1(x, 0)$ and $\ell_2(x, 0)$

Example

Consider the dynamics $x^+ = x + u$ and the cost functions

$$\ell_1(x, u) = \frac{1}{2}x^4 - \frac{1}{4}x^3 - x^2 + \frac{3}{4}x + u^2 \quad \text{and} \quad \ell_2(x, u) = (x-1)^2 + u^2$$

$$\mu = 33/41$$



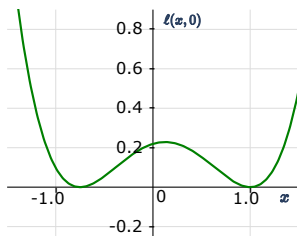
$$\ell_\nu(x, 0) \text{ for } \nu = 33/41$$

Example

Consider the dynamics $x^+ = x + u$ and the cost functions

$$\ell_1(x, u) = \frac{1}{2}x^4 - \frac{1}{4}x^3 - x^2 + \frac{3}{4}x + u^2 \quad \text{and} \quad \ell_2(x, u) = (x-1)^2 + u^2$$

$$\mu = 32/41$$

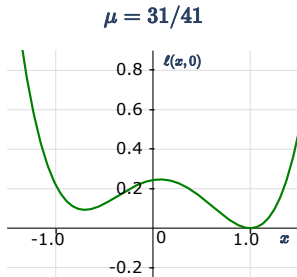


$$\ell_\nu(x, 0) \text{ for } \nu = 32/41$$

Example

Consider the dynamics $x^+ = x + u$ and the cost functions

$$\ell_1(x, u) = \frac{1}{2}x^4 - \frac{1}{4}x^3 - x^2 + \frac{3}{4}x + u^2 \quad \text{and} \quad \ell_2(x, u) = (x-1)^2 + u^2$$



$\ell_\nu(x, 0)$ for $\nu = 31/41$

Conclusion

- **Strict dissipativity** of a weighted sum of cost functions is a desirable property for ensuring **near optimal performance** of **multiobjective MPC**

Conclusion

- **Strict dissipativity** of a weighted sum of cost functions is a desirable property for ensuring **near optimal performance** of **multiobjective MPC** (as well as for ensuring other nice properties, such as turnpike behaviour)

Conclusion

- **Strict dissipativity** of a weighted sum of cost functions is a desirable property for ensuring **near optimal performance** of **multiobjective MPC** (as well as for ensuring other nice properties, such as turnpike behaviour)
- For **linear dynamics** and **linear-quadratic** or **strictly convex** costs **strict dissipativity** persists for all weights

Conclusion

- **Strict dissipativity** of a weighted sum of cost functions is a desirable property for ensuring **near optimal performance** of **multiobjective MPC** (as well as for ensuring other nice properties, such as turnpike behaviour)
- For **linear dynamics** and **linear-quadratic** or **strictly convex** costs **strict dissipativity** persists for all weights, but the corresponding storage functions are **more complicated** than expected

Conclusion

- **Strict dissipativity** of a weighted sum of cost functions is a desirable property for ensuring **near optimal performance** of **multiobjective MPC** (as well as for ensuring other nice properties, such as turnpike behaviour)
- For **linear dynamics** and **linear-quadratic** or **strictly convex** costs **strict dissipativity** persists for all weights, but the corresponding storage functions are **more complicated** than expected
- For nonlinear systems, the problem is **surprisingly complicated**. We could find **a couple of interesting insights**

Conclusion

- **Strict dissipativity** of a weighted sum of cost functions is a desirable property for ensuring **near optimal performance** of **multiobjective MPC** (as well as for ensuring other nice properties, such as turnpike behaviour)
- For **linear dynamics** and **linear-quadratic** or **strictly convex** costs **strict dissipativity** persists for all weights, but the corresponding storage functions are **more complicated** than expected
- For nonlinear systems, the problem is **surprisingly complicated**. We could find **a couple of interesting insights** but not a simple condition that would apply to a large class of systems