

Pareto front characterization for finite horizon optimal control problems with two different objectives

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Pareto Optimality

$$\begin{array}{ll} \text{minimize} & g(x) = (g_1(x), g_2(x)) \\ \text{subject to} & x \in X \end{array}$$

- $x^* \in X$ is a **Pareto solution** if there exists no $x \in X$ such that $g(x) \neq g(x^*)$ and

$$g_i(x) \leq g_i(x^*), \quad \text{for all } i = 1, 2.$$

- $x^* \in X$ is a **weak Pareto solution** if there exists no $x \in X$ such that

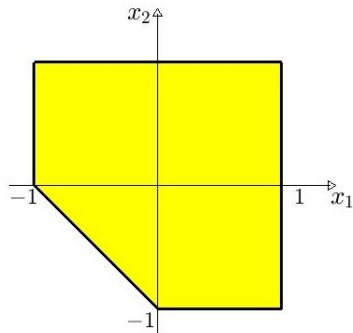
$$g_i(x) < g_i(x^*), \quad \text{for all } i = 1, 2.$$

- Set of Pareto and weak Pareto solutions: P and P_w
- Pareto and weak Pareto front:

$$\mathcal{F} = \{g(x) \mid x \in P\}, \quad \mathcal{F}_w = \{g(x) \mid x \in P_w\}$$

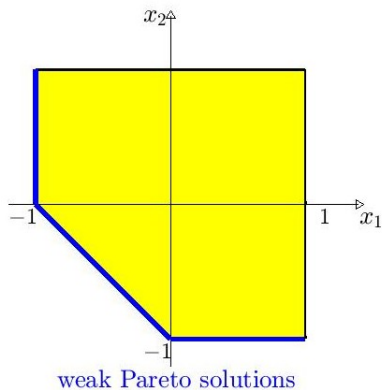
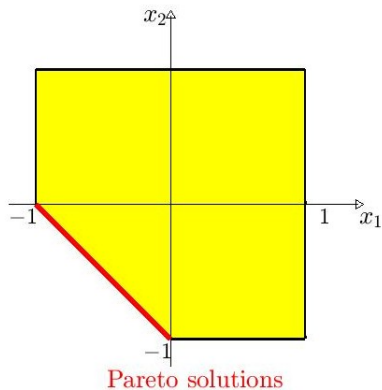
Example: Pareto optimality

Minimize $g(x) = (x_1, x_2)$



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ε -Pareto optimality

Let $\varepsilon \geq 0$. We define the following sets of ε -Pareto solutions:

- $x \in \mathcal{P}^{1,\varepsilon}$ if there is no $y \in X$ such that

$$g(y) \neq g(x) - \varepsilon \mathbf{1} \text{ and } g_i(y) \leq g_i(x) - \varepsilon, \quad \text{for all } i = 1, 2.$$

- $x \in \mathcal{P}^{2,\varepsilon}$ if there is $x^* \in \mathcal{P}$ such that $|g(x) - g(x^*)| \leq \varepsilon$.
- $x \in \mathcal{P}^{3,\varepsilon}$ if for some $y \in X$, $g(y) \neq g(x)$ and $g(y) \leq g(x)$, then

$$g(y) \geq g(x) - \varepsilon \quad \text{for all } i = 1, 2.$$

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Proposition (D.J. White, 1986)

If the feasible set X is a **compact** set and g is a continuous function, then $\mathcal{P} \subset \mathcal{P}^{3,\varepsilon} \subseteq \mathcal{P}^{2,\varepsilon} \subseteq \mathcal{P}^{1,\varepsilon}$.

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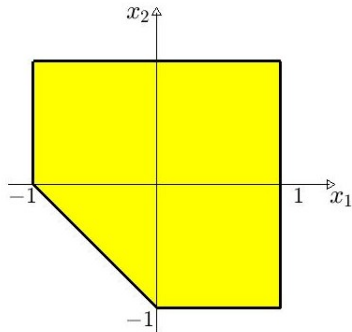
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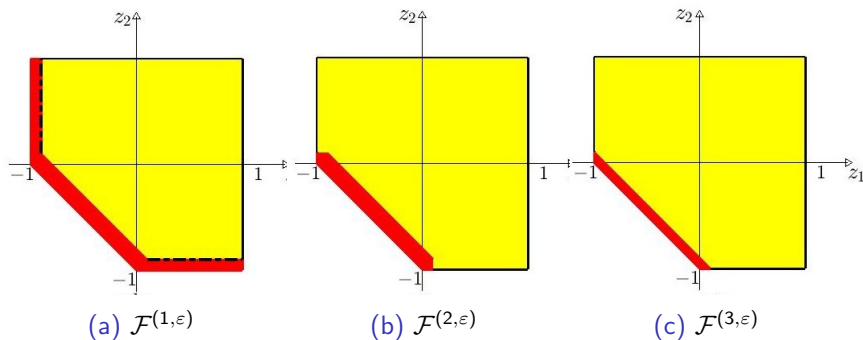
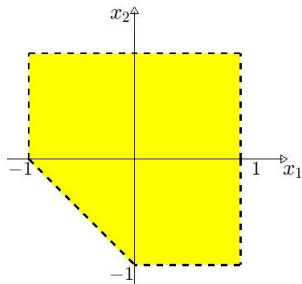


Figure: ε -Pareto fronts for $\varepsilon = 0.1$

- What happens if the feasible set is not closed?

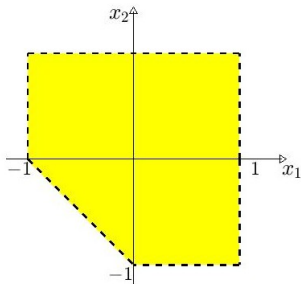
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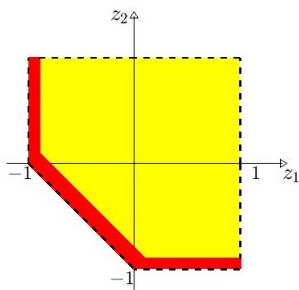
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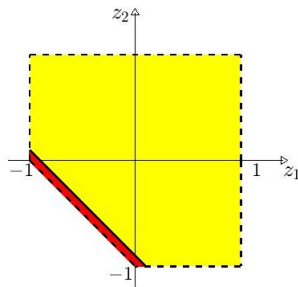
- For this example \mathcal{P} and \mathcal{P}_w are empty, so are \mathcal{F} and \mathcal{F}_w also empty.

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Minimize $g(x) = (x_1, x_2)$



(a) $\mathcal{F}(1, \varepsilon)$



(b) $\mathcal{F}(3, \varepsilon)$

Figure: ε -Pareto fronts for $\varepsilon = 0.1$

Consider the problem of minimize the objective functions over the closure of the feasible set X

$$\begin{cases} \text{Minimize } g(x) = (g_1(x), g_2(x)) \\ \text{subject to } x \in \overline{X} \end{cases}$$

- Denote the Pareto set and the weak Pareto set by $\mathcal{P}^\#$ and $\mathcal{P}_w^\#$, respectively.

Theorem [APC and Zidani, 2018]

Assume that f_i are Lipschitz continuous functions, with Lipschitz constant L_i , $i = 1, 2$.

(i) $\forall x^* \in \mathcal{P}^\#, \forall \text{varepsilon} > 0, \exists x \in \mathcal{P}^{1,\varepsilon}$ such that

$$|x^* - x| \min_i(\varepsilon/L_i) \leq \quad \text{and} \quad |g(x^*) - g(x)| \leq \varepsilon.$$

(ii) $\forall x^* \in \mathcal{P}_w^\#, \forall \text{varepsilon} > 0, \exists x \in \mathcal{P}_w^{1,\varepsilon}$ such that

$$|x^* - x| \leq \min_i(\varepsilon/L_i) \quad \text{and} \quad |g(x^*) - g(x)| \leq \varepsilon.$$

(iii) $\forall \varepsilon > 0, \forall x \in \mathcal{P}^{3,\varepsilon}$ there exists $x^* \in \mathcal{P}^\#$ such that

$$|g(x) - g(x^*)| \leq 2\varepsilon.$$

Bi-objective Optimal Control Problem

An illustrative example



Figure Source: <http://www.esalq.usp.br> and <http://www.acupunctureinwestchase.com>

- **State variables:** population of a nuisance for humans and its predator.
- **Control Variable:** spraying chemical to poison the pest or not.
- **Dynamics:** (intertwined) growth of both populations.
- **Goal:** reduce the cost of spraying the chemical. (**integral cost**)

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- **Goal 2:** keep nuisance expansion under control by minimizing the **maximum** difference of certain proportion of both species.

Objectives of different nature

$$\left\{ \begin{array}{l} \text{inf} \quad \left(\int_0^T 0.3\mathbf{u}(s)ds, \max_{0 \leq s \leq T} 0.25(\mathbf{y}_1(s) - 0.7\mathbf{y}_2(s))^2 \right), \\ \text{s.t.} \quad \dot{\mathbf{y}}_1(s) = \mathbf{y}_1(s) - \mathbf{y}_1(s)\mathbf{y}_2(s) - 0.4\mathbf{y}_1(s)\mathbf{u}(s), \\ \quad \quad \dot{\mathbf{y}}_2(s) = -\mathbf{y}_2(s) + \mathbf{y}_1(s)\mathbf{y}_2(s) - 0.2\mathbf{y}_2(s)\mathbf{u}(s), \\ \quad \quad \mathbf{y}(0) = \mathbf{x}, \\ \quad \quad \mathbf{u}(s) \in \{0, 1\} \end{array} \right.$$

Problem statement

$$\left\{ \begin{array}{l} \text{inf} \quad \left(\int_t^T \ell(\mathbf{y}(s), \mathbf{u}(s)) ds + \varphi(\mathbf{y}(T)), \max_{0 \leq s \leq T} 0.25(\mathbf{y}_1(s) - 0.7\mathbf{y}_2(s))^2 \right), \\ \text{s.t.} \quad \dot{\mathbf{y}}_1(s) = \mathbf{y}_1(s) - \mathbf{y}_1(s)\mathbf{y}_2(s) - 0.4\mathbf{y}_1(s)\mathbf{u}(s), \\ \quad \quad \dot{\mathbf{y}}_2(s) = -\mathbf{y}_2(s) + \mathbf{y}_1(s)\mathbf{y}_2(s) - 0.2\mathbf{y}_2(s)\mathbf{u}(s), \\ \quad \quad \mathbf{y}(0) = x, \\ \quad \quad \mathbf{u}(s) \in \{0, 1\} \end{array} \right.$$

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$$(MOP) \left\{ \begin{array}{l} \inf \left(\int_t^T \ell(\mathbf{y}(s), \mathbf{u}(s)) ds + \varphi(\mathbf{y}(T)), \max_{s \in [t, T]} \psi(\mathbf{y}(s)) \right), \\ \text{s. t. } \dot{\mathbf{y}}(s) = f(\mathbf{y}(s), \mathbf{u}(s)), \quad \text{a.e. } s \in [0, T], \\ \mathbf{y}(t) = \mathbf{x}, \\ \mathbf{u} \in \mathcal{U} \end{array} \right.$$

where

- $\mathcal{U} = \left\{ \mathbf{u} : [0, +\infty) \rightarrow \mathbb{R}^M \text{ measurable, } \mathbf{u}(s) \in U \text{ a.e.} \right\}$

U is a compact subset of \mathbb{R}^M , $\ell : \mathbb{R}^N \times U \rightarrow \mathbb{R}$, $\varphi : \mathbb{R}^N \rightarrow \mathbb{R}$, $\psi : \mathbb{R}^N \rightarrow \mathbb{R}$ and $f : \mathbb{R}^N \times U \mapsto \mathbb{R}^N$ are Lipschitz continuous and bounded functions.

An equivalent bi-objective problem

Define:

- $\mathcal{A}(x, u) = M_\ell - \ell(x, u)$ and the set-valued function

$$G(x) = \left\{ \begin{pmatrix} f(x, u) \\ -\ell(x, u) - a \end{pmatrix}, u \in U, 0 \leq a \leq \mathcal{A}(x, u) \right\}.$$

- set of trajectories:

$$\mathcal{S}_{[t, T]}(x, 0) = \{(\mathbf{y}, \mathbf{z}) : (\dot{\mathbf{y}}(s), \dot{\mathbf{z}}(s))^T \in G(\mathbf{y}(s)), \text{ for a.e. } s \in [t, T]; \\ (\mathbf{y}(t), \mathbf{z}(t)) = (x, 0)\}.$$

- bi-objective optimal control problem:

$$\begin{cases} \inf \left(\varphi(\mathbf{y}(T)) - \mathbf{z}(T), \max_{s \in [t, T]} \psi(\mathbf{y}(s)) \right) \\ \text{s.t } (\mathbf{y}, \mathbf{z}) \in \mathcal{S}_{[t, T]}(x, 0). \end{cases} \quad (1)$$

An equivalent bi-objective problem

- For any $(\mathbf{y}^*, \mathbf{z}^*)$ (weak) that is a Pareto optimal solution of (1) and \mathbf{u}^* , γ^* the respective controls. The pair $(\mathbf{y}^*, \mathbf{u}^*)$ is a (weak) Pareto optimal solution for problem (MOP).
- Moreover the (weak) Pareto front of two problems coincides.

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- Moreover the (weak) Pareto front of two problems coincides.

Without any additional assumption it is not possible to guarantee that the set of trajectories $\mathcal{S}_{[t, T]}(x, 0)$ is compact.

Relaxed problem

- Consider

$$\mathcal{S}_{[t, T]}^{\#}(x, 0) = \{(\mathbf{y}, \mathbf{z}) : (\dot{\mathbf{y}}(s), \dot{\mathbf{z}}(s))^{\top} \in \overline{\text{co}}(G(\mathbf{y}(s))), \text{ for a.e. } s \in [t, T]; \\ (\mathbf{y}(t), \mathbf{z}(t)) = (x, z)\},$$

where $\overline{\text{co}}(S)$ denotes the closed convex hull of the subset S , that is the minimal convex set that contains S .

- $\mathcal{S}_{[t, T]}^{\#}(x, 0)$ is compact and equal the closure of $\mathcal{S}_{[t, T]}(x, z)$ in the space of continuous functions $C(t, T)$.

[Frankowska and Rampazzo, 1999]

So we introduce the following convexified bi-objective optimal control problem

$$\begin{cases} \min \left(\varphi(\mathbf{y}(T)) - \mathbf{z}(T), \max_{s \in [t, T]} \psi(\mathbf{y}(s)) \right) \\ \text{s.t } (\mathbf{y}, \mathbf{z}) \in \mathcal{S}_{[t, T]}^{\#}(x, 0). \end{cases} \quad (\text{MORP})$$

Theorem [APC and Zidani, 2018]

Let $(t, x) \in [0, T] \times \mathbb{R}^N$.

(i) $\forall (\mathbf{y}^*, \mathbf{z}^*) \in \mathcal{P}^\#(t, x), \forall \varepsilon > 0, \exists (\mathbf{y}, \mathbf{z}) \in \mathcal{P}^{1, \varepsilon}(t, x)$ such that

$$|(\mathbf{y}^*, \mathbf{z}^*) - (\mathbf{y}, \mathbf{z})| \leq \min\left(\frac{\varepsilon}{L_\varphi}, \frac{\varepsilon}{L_\psi}\right) \quad \text{and}$$

$$\left| \left(\varphi(\mathbf{y}^*(T)) - \mathbf{z}^*(T), \max_{s \in [t, T]} \psi(\mathbf{y}^*(s)) \right) - \left(\varphi(\mathbf{y}(T)) - \mathbf{z}(T), \max_{s \in [t, T]} \psi(\mathbf{y}(s)) \right) \right| \leq \varepsilon.$$

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Auxiliary control problem

$$w(t, x, z_1, z_2) = \min_{(\mathbf{y}, \mathbf{z}) \in \mathcal{S}_{[t, T]}^\#(x, 0)} \left[\left(\varphi(\mathbf{y}(T)) - \mathbf{z}(T) \right) \bigvee \max_{s \in [t, T]} (\psi(\mathbf{y}(s)) - z_2) \right]$$

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HJB equation

The function w is the unique viscosity solution of the following HJB equation:

$$\begin{aligned} \min \left(\partial_t w(t, x, z) + \mathcal{H}^\#(x, D_x w, D_z w), w(t, x, z) - (g(x) - z_2) \right) &= 0, \\ &\forall t \in [0, T), x \in \mathbb{R}^n, z \in \mathbb{R}^2 \\ w(T, x, z) &= \left(\varphi(x) - z_1 \right) \bigvee \left(g(x) - z_2 \right) \quad \forall x \in \mathbb{R}^n, z \in \mathbb{R}^2, \end{aligned}$$

where the Hamiltonian $\mathcal{H}^\#$ is given by

$$\mathcal{H}^\#(x, p, q) = \max_{(v_x, v_z) \in \overline{\text{co}}(G(x))} \left(-v_x \cdot p - v_z \cdot q_1 \right).$$

Characterization of the Pareto front

Characterization of the weak Pareto front

Consider:

- $$z_1^*(t, x) = \inf \left\{ \zeta \in \mathbb{R} \mid \exists z \in \mathbb{R}^2 \ w(t, x, \zeta, z_2) \leq 0 \right\}$$
$$= \min_{(\mathbf{y}, \mathbf{z}) \in \mathcal{S}_{[t, T]}^\#(x, 0)} \varphi(\mathbf{y}(T)) - \mathbf{z}(T) = \varphi(\mathbf{y}_1(T)) - \mathbf{z}_1(T);$$
- $$z_2^*(t, x) = \inf \left\{ \zeta \in \mathbb{R} \mid \exists z \in \mathbb{R}^2 \ w(t, x, z_1, \zeta) \leq 0 \right\}$$
$$= \min_{(\mathbf{y}, \mathbf{z}) \in \mathcal{S}_{[t, T]}^\#(x, 0)} \max_{s \in [t, T]} \psi(\mathbf{y}(s)) = \max_{s \in [t, T]} \psi(\mathbf{y}_2(s));$$

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- $\bar{z}_1(t, x) = \inf \left\{ \zeta \in \mathbb{R} \mid w(t, x, \zeta, z_2^*(t, x)) = 0 \right\} = \varphi(\mathbf{y}_2(T)) - \mathbf{z}_2(T);$
- $\bar{z}_2(t, x) = \inf \left\{ \zeta \in \mathbb{R} \mid w(t, x, z_1^*(t, x), \zeta) = 0 \right\} = \max_{s \in [t, T]} \psi(\mathbf{y}_1(s));$

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 $= \min_{(\mathbf{y}, \mathbf{z}) \in \mathcal{S}_{[t, T]}^\#(x, 0)} \max_{s \in [t, T]} \psi(\mathbf{y}(s)) = \max_{s \in [t, T]} \psi(\mathbf{y}_2(s));$
- $\bar{z}_1(t, x) = \inf \left\{ \zeta \in \mathbb{R} \mid w(t, x, \zeta, z_2^*(t, x)) = 0 \right\} = \varphi(\mathbf{y}_2(T)) - \mathbf{z}_2(T);$
- $\bar{z}_2(t, x) = \inf \left\{ \zeta \in \mathbb{R} \mid w(t, x, z_1^*(t, x), \zeta) = 0 \right\} = \max_{s \in [t, T]} \psi(\mathbf{y}_1(s));$

Assume that the ideal vector $\beta^*(t, x) = (z_1^*(t, x), z_2^*(t, x)) \in \mathbb{R}^2$ is not feasible, otherwise the Pareto front is reduced to this point.

Characterization of the weak Pareto front

Define:

- $\Omega = [z_1^*(t, x), \bar{z}_1(t, x)] \times [z_2^*(t, x), \bar{z}_2(t, x)].$

Theorem (APC and Zidani, 2018)

Let (t, x) be in $[0, T] \times \mathbb{R}^N$

- (i) $\mathcal{F}^\#(t, x) \subset \mathcal{F}_w^\#(t, x) \cap \Omega \subset \{z \in \Omega \mid w(t, x, z) = 0\}.$
- (ii) Let $z \in \Omega$ such that $w(t, x, z) = 0$. If there exists a admissible pair $(\mathbf{y}, \mathbf{z}) \in \mathcal{S}_{[t, T]}^\#(x, 0)$ such that $\varphi(\mathbf{y}(T)) - \mathbf{z}(T) = z_1$ and $\max_{s \in [t, T]} \psi(\mathbf{y}(s)) = z_2$, then $z \in \mathcal{F}_w^\#(t, x).$

Characterization of the Pareto front

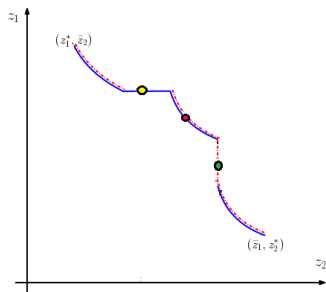
Introduce the projectors operators on \mathbb{R}^2 :

$$\pi_1(z) = z_1, \quad \pi_2(z) = z_2.$$

and the following functions

$$\eta_1 : [z_1^*, \bar{z}_1] \rightarrow [z_2^*, \bar{z}_2], \quad \eta_1(\gamma_1) := \inf\{\theta \mid w(t, x, \gamma_1, \theta) \leq 0\},$$

$$\eta_2 : [z_2^*, \bar{z}_2] \rightarrow [z_1^*, \bar{z}_1], \quad \eta_2(\gamma_2) := \inf\{\theta \mid w(t, x, \theta, \gamma_2) \leq 0\}.$$



Let (t, x) be in $[0, T] \times \mathbb{R}^N$.

$$(ii) \mathcal{F}^\#(t, x) = \left\{ (\zeta, \eta_1(\zeta)), \zeta \in \text{dom}(\eta_1) \right\} \cap \left\{ (\eta_2(\zeta), \zeta), \zeta \in \text{dom}(\eta_2) \right\}.$$

Let (t, x) be in $[0, T] \times \mathbb{R}^N$.

- (ii) $\mathcal{F}^\#(t, x) = \left\{ (\zeta, \eta_1(\zeta)), \zeta \in \text{dom}(\eta_1) \right\} \cap \left\{ (\eta_2(\zeta), \zeta), \zeta \in \text{dom}(\eta_2) \right\}$.
- (iii) For any $z \in \mathcal{F}^\#(t, x)$ let a trajectory $(\mathbf{y}, \mathbf{z}) \in \mathcal{S}_{[t, T]}^\#(x, 0)$ that is optimal for the auxiliary problem. Then (\mathbf{y}, \mathbf{z}) is a Pareto optimal solution of (MORP).

Let (t, x) be in $[0, T] \times \mathbb{R}^N$ and $\varepsilon > 0$.

$$(i) \mathcal{F}^{1,\varepsilon}(t, x) \subset \mathcal{F}_w^{1,\varepsilon}(t, x) \subset \left\{ z \in \mathbb{R}^2 \mid -\varepsilon \leq w(t, x, z) \leq 0 \right\}.$$

Let (t, x) be in $[0, T] \times \mathbb{R}^N$ and $\varepsilon > 0$.

- (i) $\mathcal{F}^{1,\varepsilon}(t, x) \subset \mathcal{F}_w^{1,\varepsilon}(t, x) \subset \left\{ z \in \mathbb{R}^2 \mid -\varepsilon \leq w(t, x, z) \leq 0 \right\}$.
- (ii) Let $z_\varepsilon \in \left\{ z \in \mathbb{R}^2 \mid -\varepsilon \leq w(t, x, z) \leq 0 \right\}$. If there exists $(\mathbf{y}_\varepsilon, \mathbf{z}_\varepsilon) \in \mathcal{S}_{[t, T]}(x, 0)$ that is optimal for the auxiliary control problem w . Then $(\mathbf{y}_\varepsilon, \mathbf{z}_\varepsilon) \in \mathcal{P}_w^{1,\varepsilon}(t, x)$ of problem (1).

Pest control example

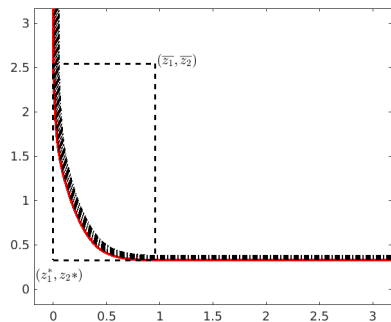
$$\left\{ \begin{array}{l} \inf \left(\int_0^T 0.3\mathbf{u}(s)ds, \max_{0 \leq s \leq T} 0.25(\mathbf{y}_1(s) - 0.7\mathbf{y}_2(s))^2 \right), \\ \text{s.t. } \dot{\mathbf{y}}_1(s) = \mathbf{y}_1(s) - \mathbf{y}_1(s)\mathbf{y}_2(s) - 0.4\mathbf{y}_1(s)\mathbf{u}(s), \\ \dot{\mathbf{y}}_2(s) = -\mathbf{y}_2(s) + \mathbf{y}_1(s)\mathbf{y}_2(s) - 0.2\mathbf{y}_2(s)\mathbf{u}(s), \\ \mathbf{y}(0) = (0.7, 0.2), \\ \mathbf{u}(s) \in \{0, 1\} \end{array} \right.$$

In order to obtain a relaxed problem we are going to consider $\mathbf{u}(s) \in [0, 1]$. The corresponding value function w is solution of the following HJB equation

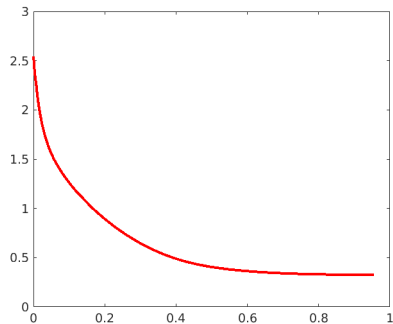
$$\begin{aligned} \min \left(\partial_t w(t, x, z) + \mathcal{H}(x, D_x w, D_z w), w(t, x, z) - (0.25(x_1 - 0.7x_2)^2 - z_2) \right) &= 0, \\ &\text{for } t \in [0, T], x, z \in \mathbb{R}^2 \\ w(T, x, z) &= -z_1 \sqrt{(0.25(x_1 - 0.7x_2)^2 - z_2)} \quad \text{for } x \in \mathbb{R}^2, z \in \mathbb{R}^2. \end{aligned}$$

- The HJB equation was solved by a finite difference method implement at C++ HJB-solver "ROC-HJ", available in *<http://uma.ensta-paristech.fr/soft/ROC-HJ>*
- Grid of 75^4 nodes on the domain $[0, 3.5] \times [0, 3.5] \times [0, 3] \times [0, 3]$.

Pest control example

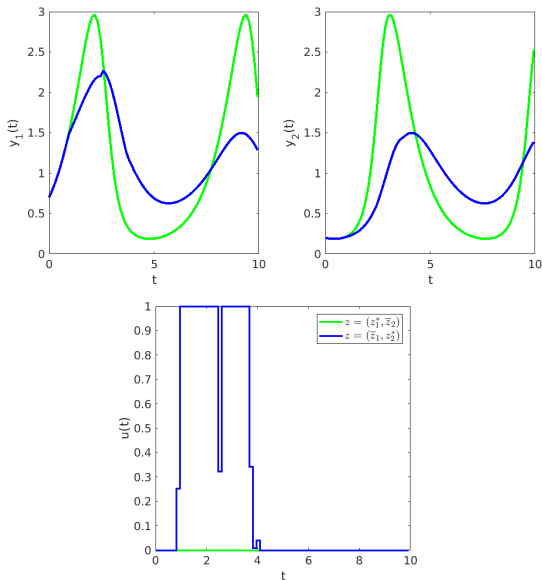


(a) Set that contains $\mathcal{F}_w^{1,\varepsilon}(t, x_0)$ and $\mathcal{F}_w^\#(t, x_0)$

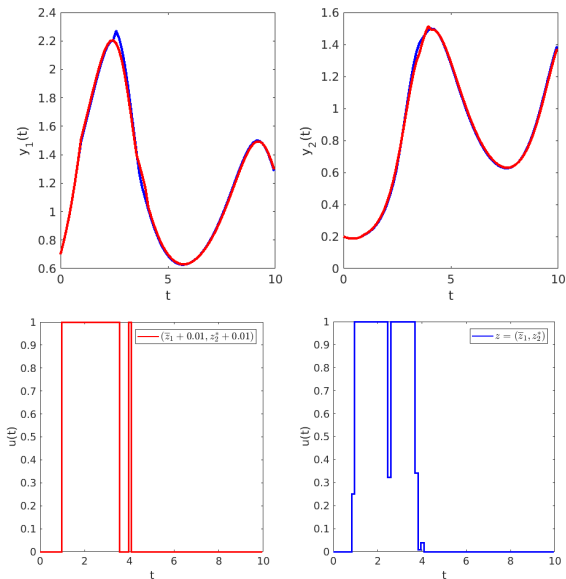


(b) zoom of $\mathcal{F}^\#(t, x_0)$

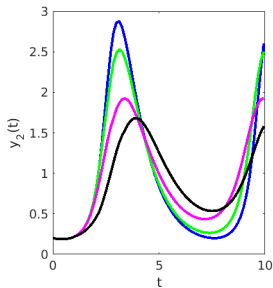
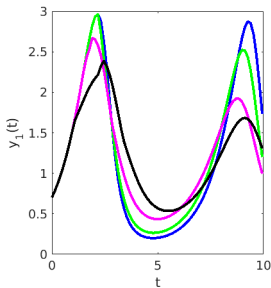
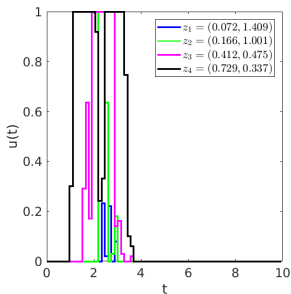
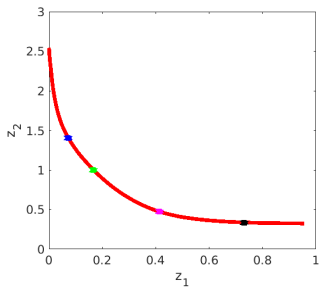
Optimal trajectories for single objective problems



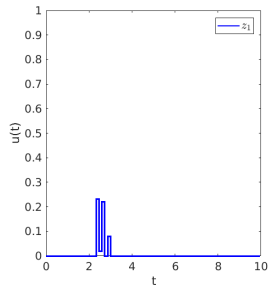
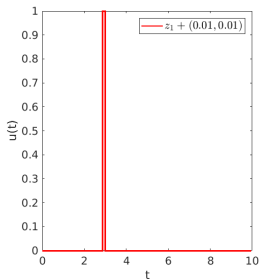
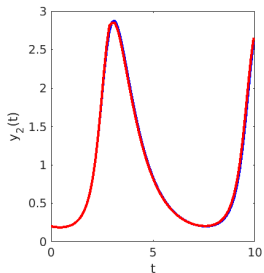
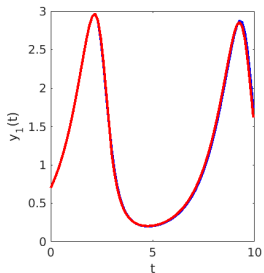
ε -Pareto Optimal trajectories for the original problem



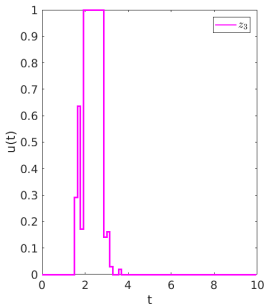
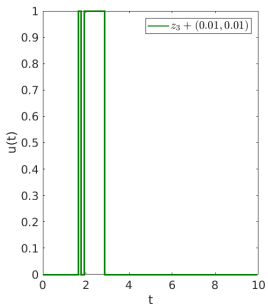
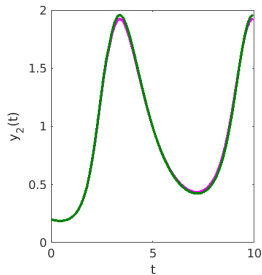
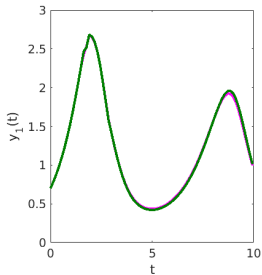
Pareto Optimal trajectories for the relaxed problem



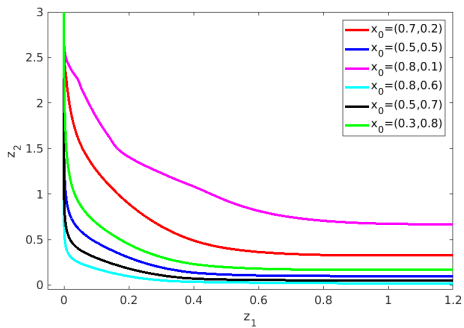
ε -Pareto Optimal trajectories for the original problem



ε -Pareto Optimal trajectories for the original problem



Different initial states



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Thank you for your attention