

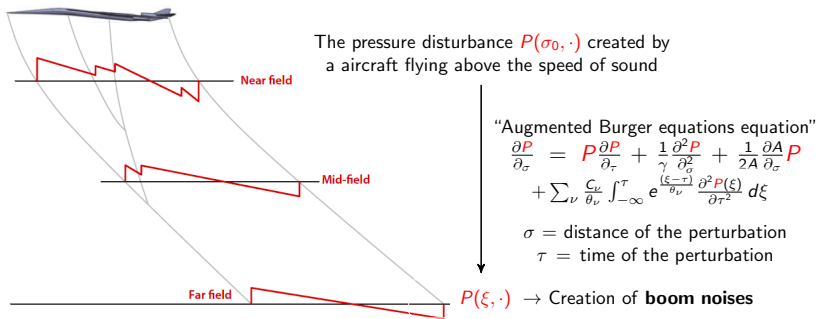
Inverse design of one-dimensional Burgers equation¹

Enrique Zuazua

1. Based on joint work with Thibault Liard : Liard, T., & Zuazua, E. (2021). Initial data identification for the one-dimensional Burgers equation. IEEE Transactions on Automatic Control, 67(6), 3098-3104.

- ① Introduction
 - Sonic boom minimization
 - Presentation of the control optimal problem under consideration
- ② Preliminaries and notations
 - Wave-front tracking algorithm
 - The backward operator S_t^-
- ③ Main result : full characterization of minimizers
- ④ Find randomly all possible minimizers using
 - a backward-forward method
 - a wave-front tracking algorithm
- ⑤ Conclusion and open problems

Sonic boom and supersonic airplanes



Objective : Tailoring the shape of the aircraft to minimize the ground sonic boom effects

The optimal control problem is $\min_{P_0 \in \mathcal{A}} d(P(\xi, \cdot), P^*(\cdot))$

The admissible set \mathcal{A} is chosen to ensure feasible aircraft design (for instance aerodynamic lift).
 $d(\cdot, \cdot)$ is chosen to be a robust and realistic metric for boom noises (Perceived loudness (PLdB)
 P^* a desired ground signature and ξ the distance of the propagation

References : [Whitham, 1952 ; Cleveland, 1995 ; Alonso-Colonno, 2012 ; Rallabhandi, 2011 ; Adimurthi-Ghoshal-Gowda, 2014 ; Allahverdi-Pozo-Zuazua, 2016]

The one-dimensional Burgers equation

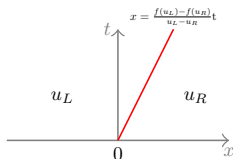
The one-dimensional Burgers equation

$$\begin{cases} u_t + f(u)_x = 0, & (t, x) \in \mathbb{R}^+ \times \mathbb{R}, \\ u(0, x) = u_0(x). & x \in \mathbb{R}, \end{cases} \quad (\text{PDE})$$

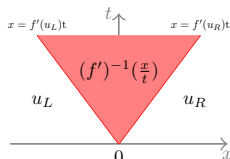
- The flux $f : u \rightarrow \frac{u^2}{2}$
- $u_0 \in BV(\mathbb{R})$

→ The function u is a **weak solution** to (PDE), for $(t, x) \in (0, +\infty) \times \mathbb{R}$, i.e for all $\varphi \in C_c^1(\mathbb{R}^2, \mathbb{R})$,

$$\int_{\mathbb{R}_+} \int_{\mathbb{R}} (u \partial_t \varphi + f(u) \partial_x \varphi) dx dt + \int_{\mathbb{R}} u_0(x) \varphi(0, x) dx = 0.$$



$u_L < u_R$. A weak solution of (PDE)



$u_L < u_R$. A weak-entropy solution of (PDE)

→ The function u is an **entropy solution** to (PDE) For every $k \in \mathbb{R}$, for all $\varphi \in C_c^1(\mathbb{R}^2, \mathbb{R}_+)$, it holds

$$\int_{\mathbb{R}_+} \int_{\mathbb{R}} (|u - k| \partial_t \varphi + \text{sgn}(u - k)(f(u) - f(k)) \partial_x \varphi) dx dt + \int_{\mathbb{R}} |u_0 - k| \varphi(0, x) dx \geq 0.$$

Optimal control problem

For any initial datum $u_0 \in BV(\mathbb{R})$ there exists a unique weak-entropy solution $S_t^+(u_0) \in L^\infty([0, T] \times \mathbb{R}) \cap C^0([0, T], L_{loc}^1(\mathbb{R}))$ of (PDE)

Our aim is to solve the following optimal control problem

$$\min_{u_0 \in \mathcal{U}_{ad}^0} J_0(u_0) := \int_{\mathbb{R}} (u^T(x) - S_T^+(u_0))^2 dx, \quad (\text{Opt-Pb})$$

Above $u^T \in BV(\mathbb{R})$ and the class of admissible initial data is defined by

$$\mathcal{U}_{ad}^0 = \{u_0 \in BV(\mathbb{R}) / \|u_0\|_{BV(\mathbb{R})} < C \text{ and } \text{Supp}(u_0) \subset K_0\}.$$

Objectives :

- Construction of a minimizer of (Opt-Pb) via a **backward-forward method**.
- **Implementation of an algorithm** to find (randomly) all possible minimizers of (Opt-Pb)

Definition : u^T is **reachable** at time T if there exists $u_0 \in BV(\mathbb{R})$ such that $S_T^+(u_0) = u^T$.

If u^T is **reachable** at time T :

→ Characterization of reachable u^T : [*Colombo-Perrollaz, 2019*],[*Gosse-Zuazua, 2017*]

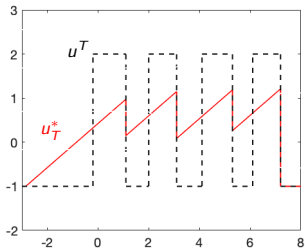
→ Fully characterization of initial data u_0 leading to u^T : [*Colombo-Perrollaz, 2019*]

If u^T is **unreachable** at time T :

→ Notion of weak-differentiability of the cost function J_0 in (Opt-Pb) :
[*Majda, 1983*; *Bardos-Pironneau, 2005*; *Bouchut-James, 1999*; *Bressan-Marson, 1995*]

→ Implementation of Gradient descent method to solve (Opt-Pb) :
[*Castro-Palacios-Zuazua, 2008-2010*; *Allahverdi-Pozo-Zuazua, 2016*; *Gosse-Zuazua, 2017*]

An amuse-bouche



A target $u^T \in \{-1, 2\}$.

Plotting of two minimizers u_0 and u_1 of (Opt-Pb) such that

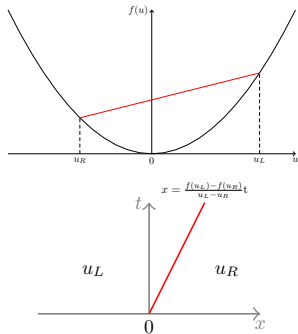
$$S_T^+(u_0) = S_T^+(u_1) = u_T^*$$

Wave-front tracking algorithm

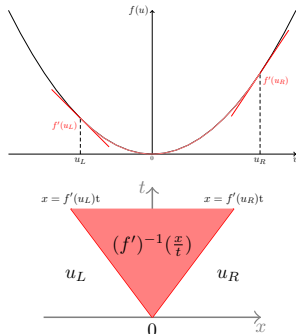
Conservation laws and Riemann solutions

The Burgers equation with Riemann type initial data

$$\partial_t \rho + \partial_x (f(\rho)) = 0, \quad (t, x) \in \mathbb{R}^+ \times \mathbb{R},$$
$$u(0, x) = \begin{cases} u_L & \text{if } x < 0 \\ u_R & \text{if } x > 0 \end{cases}, \quad x \in \mathbb{R}.$$



Riemann solution when
 $u_L > u_R$: a shock wave

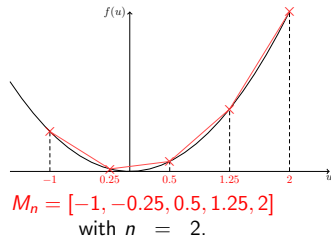


Riemann solution when
 $u_L < u_R$: a rarefaction wave

A Wave-front tracking method

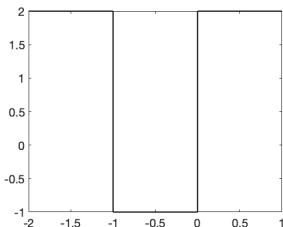
Assuming that there exists \underline{u}, \bar{u} such that $\underline{u} \leq u_0 \leq \bar{u}$.

- Construction of a state mesh
 $\mathcal{M}_n = \underline{u} + (\bar{u} - \underline{u})2^{-n}\mathbb{N} \cap [0, 1]$
- We approximate $u_0 \in BV(\mathbb{R})$ by a piecewise constant function $u_0^n \in \mathcal{M}_n$.



- We solve approximately the Riemann problem at each point of discontinuity $(x_i)_{i \in \{1, \dots, N\}}$ of u_0^n .
 - if $u_0^n(x_i-) > u_0^n(x_i+)$, a shock wave is generated with speed given by the Rankine-Hugoniot condition.
 - if $u_0^n(x_i-) < u_0^n(x_i+)$, we decompose the rarefaction wave into a fan of rarefaction shocks traveling with speed given by Rankine-Hugoniot condition.

A Wave-front tracking method



$$u_0 = 2\mathbb{1}_{(-\infty, -1)} - \mathbb{1}_{(-1, 0)} + 2\mathbb{1}_{(0, \infty)}$$

- We construct an approximate solution $u^n(t, x)$ until a time t_1 , where at least two wave fronts interact together.
- At $t = t_1^+$ a new Riemann problem arises and we repeat the previous strategy replacing $t = 0$ and u_0^n by $t = t_1$ and $u^n(t_1, \cdot)$ respectively.

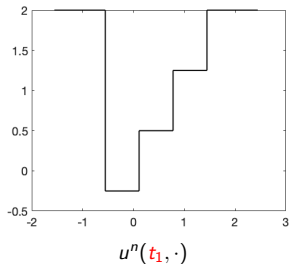
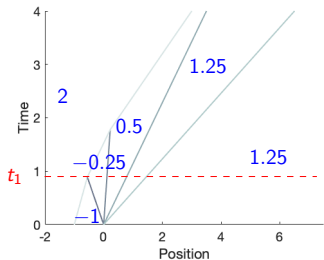
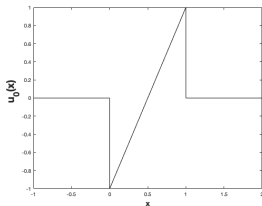
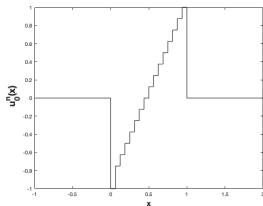


Illustration of a WFT method



Initial datum u_0



Construction of an approximate initial datum $u_0^n : x \rightarrow \mathcal{M}_n$ of u_0 with $n = 5$

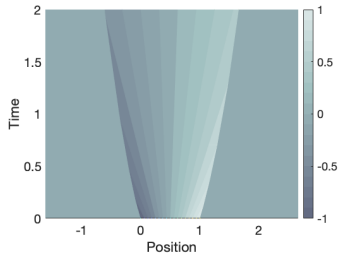
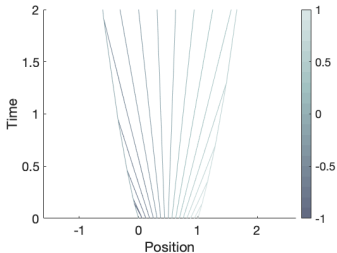


Illustration of the “wave-front” objects

Wave-front tracking methods VS Godunov scheme

Godunov scheme is a conservative three-point numerical scheme having the following form

$$u_{j+1}^n = u_j^n - \frac{\Delta t}{\Delta x} (g(u_j^n, u_{j+1}^n) - g(u_{j-1}^n, u_j^n)),$$

with g a numerical flux and
 $u(n\Delta t, j\Delta x) \approx u_j^n, n \in \mathbb{N}, j \in \mathbb{Z}$.

— WFT algorithm

— Godunov scheme

Wave-front tracking methods VS Godunov scheme

Godunov scheme :

- Discretization in space Δx and time Δt ,
- “Backward uniqueness” because of diffusion effects,
- Easy to implement,
- A CFL condition has to be satisfied ($\frac{\Delta t}{\Delta x} \max_{u \in [\underline{u}, \bar{u}]} |f'(u)| \leq \frac{1}{2}$) \rightarrow The final time T is small.

Wave-front tracking method :

- Discretization in state Δu ,
- No Backward uniqueness because shocks may be created,
- Hard to implement (creation of objects and find interaction points between objects),
- No CFL condition is imposed \rightarrow The final time T may be large.

The backward operator S_t^-

The backward operator S_t^-

The backward operator S_t^- associated to the Burgers dynamic is defined by

$$S_t^-(u^T)(x) = S_t^+(x \rightarrow u^T(-x))(-x),$$

for every $t \in [0, T]$ and for a.e $x \in \mathbb{R}$.

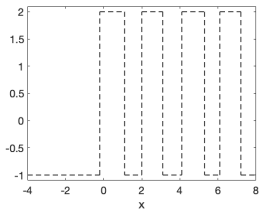
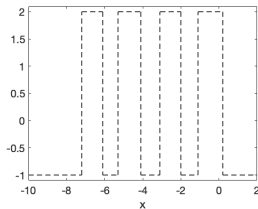
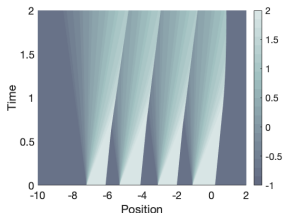
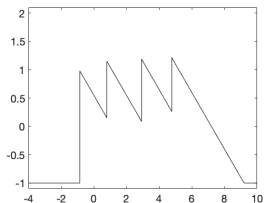
Remark : The solution $S_t^-(u^T)$ may be regarded as the zero viscosity limit of $S_T^{-,\epsilon}(u^T)$ solution of the following backward equation

$$\begin{cases} \partial_t u(t, x) + \partial_x f(u(t, x)) = -\epsilon \partial_{xx}^2 u(t, x), & (t, x) \in \mathbb{R}^+ \times \mathbb{R}, \\ u(T, \cdot) = u^T(x), & x \in \mathbb{R}. \end{cases}$$

Using the change of variable $(t, x) \rightarrow (T - t, -x)$, we notice that the backward equation above is well-defined.

Thus, $S_T^-(u^T)$ is also called the **backward entropy solution** with final target u^T .

$$u^T(x) = \begin{cases} 2 & \text{if } x \in (-0.2, 1.1) \cup (2, 3.1) \cup (4.1, 5.3) \cup (6.1, 7.2), \\ -1 & \text{otherwise.} \end{cases}$$


 u^T

 $x \rightarrow u^T(-x)$

 $(t, x) \rightarrow S_t^+(x \rightarrow u^T(-x))$

 $S_t^-(u^T) : (t, x) \rightarrow S_t^+(x \rightarrow u^T(-x))(-x)$

Main result

Our aim is to solve the following optimal control problem

$$\min_{u_0 \in \mathcal{U}_{\text{ad}}^0} J_0(u_0) := \int_{\mathbb{R}} (u^T(x) - S_T^+(u_0))^2 dx, \quad (\text{Opt-Pb})$$

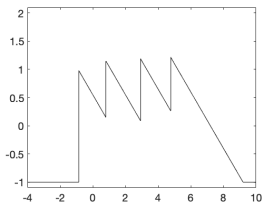
with $u^T \in BV(\mathbb{R})$ and $\mathcal{U}_{\text{ad}}^0 = \{u_0 \in BV(\mathbb{R}) / \|u_0\|_{BV(\mathbb{R})} < C \text{ and } \text{Supp}(u_0) \subset K_0\}$.

Theorem

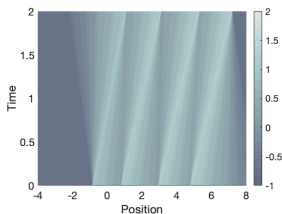
For a.e $T > 0$, the optimal control problem (Opt-Pb) admits multiple optimal solutions. Moreover, the initial datum $u_0 \in BV(\mathbb{R})$ is an optimal solution of (Opt-Pb) if and only if $u_0 \in BV(\mathbb{R})$ verifies $S_T^+(u_0) = S_T^+(S_T^-(u^T))$.

- A full characterization of the set of initial data $u_0 \in BV(\mathbb{R})$ such that $S_T^+(u_0) = S_T^+(S_T^-(u^T))$ is given in [**Colombo-Perrolaz, 2019**].
- If there exists an initial datum $u_0 \in BV(\mathbb{R})$ such that $S_T^+(u_0) = u^T$ then $S_T^+(S_T^-(u^T)) = u^T$.

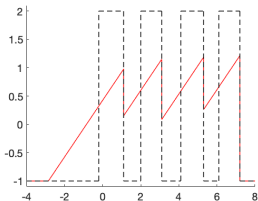
$$u^T(x) = \begin{cases} 2 & \text{if } x \in (-0.2, 1.1) \cup (2, 3.1) \cup (4.1, 5.3) \cup (6.1, 7.2), \\ -1 & \text{otherwise.} \end{cases}$$



$$x \rightarrow S_T^-(u^T)(x)$$



$$(t, x) \rightarrow S_t^+(S_T^-(u^T))(x)$$



$$u^T \text{ and } x \rightarrow S_t^+(S_T^-(u^T))(x)$$

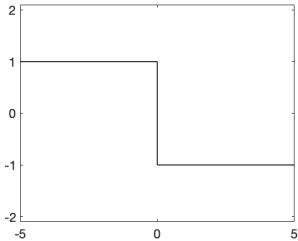
A target u^T with finite number of shocks

The two following results are given in [Colombo-Perrolaz, 2019].

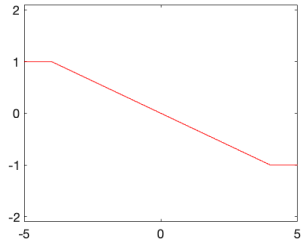
- There exists an initial datum $u_0 \in BV(\mathbb{R})$ such that $S_T^+(u_0) = u^T$ iff u^T satisfies the Oleinik condition, means that $\partial_x u^T \leq \frac{1}{T}$ in the sense of distributions.
- A map $u_0 \in BV(\mathbb{R})$ verifies $S_T^+(u_0) = u^T$ if and only if the two following statements hold :
 - For every $x \in \mathbb{R} \setminus \cup_{i=1}^N [a_i, b_i]$, $u_0(x-) = S_T^-(u^T)(x-)$.
 - For every $x \in \cup_{i=1}^N [a_i, b_i]$

$$\int_{a_i}^x u_0(s) ds \geq \int_{a_i}^x S_T^-(u^T)(s) ds,$$
$$\int_{a_i}^{b_i} u_0(s) ds = \int_{a_i}^{b_i} S_T^-(u^T)(s) ds.$$

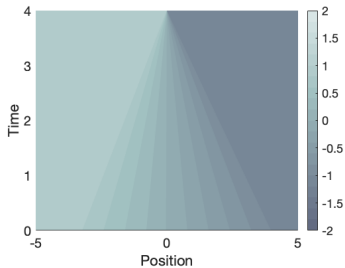
with $a_i := x_i^T - Tf'(u^T(x_i^T-))$ and $b_i := x_i^T - Tf'(u^T(x_i^T+))$ and $(x_i^T)_{i \in \{0, \dots, N\}}$ the $N \in \mathbb{N} \cup \{\infty\}$ discontinuous points of u^T such that $u^T(x_i^T+) < u^T(x_i^T-)$.



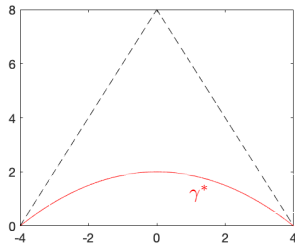
$$u^T = \mathbb{1}_{(-\infty, 0)} - \mathbb{1}_{(0, \infty)}$$



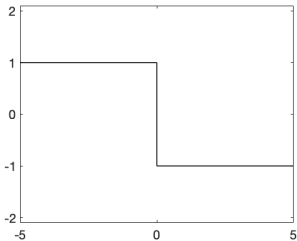
$$S_T^-(u^T) \text{ such that } S_T^+(S_T^-(u^T)) = u^T.$$



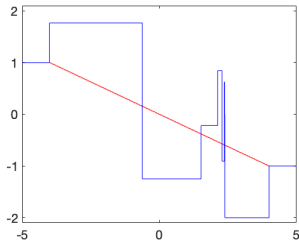
$$(t, x) \rightarrow S_t^+(S_T^-(u^T))(x)$$



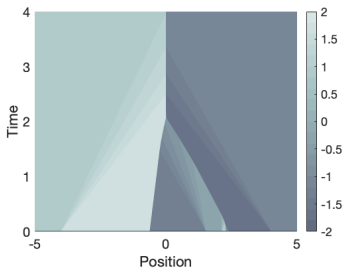
$$x \rightarrow \gamma^*(x) := \int_{-4}^x S_T^-(u^T)(s) ds.$$



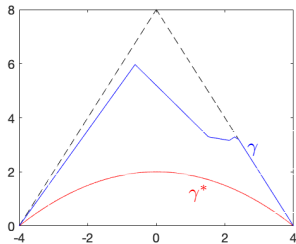
$$u^T = \mathbb{1}_{(-\infty, 0)} - \mathbb{1}_{(0, \infty)}$$



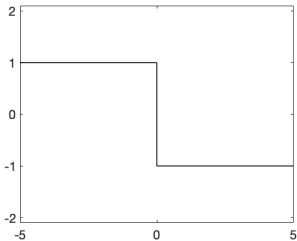
$$S_T^-(u^T) \text{ and } u_0 \text{ such that } S_T^+(S_T^-(u^T)) = S_T^+(u_0) = u^T.$$



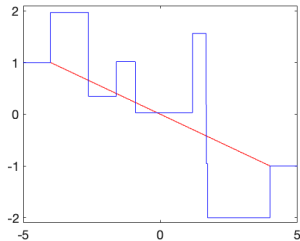
$$(t, x) \rightarrow S_t^+(u_0)(x)$$



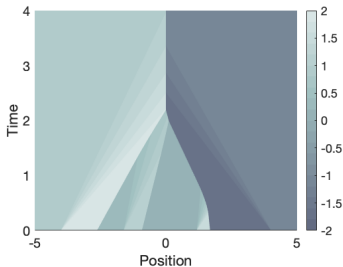
$$\gamma(x) := \int_{-4}^x u_0(s) ds \geq \gamma^*(x), \forall x \in [-4, 4]$$



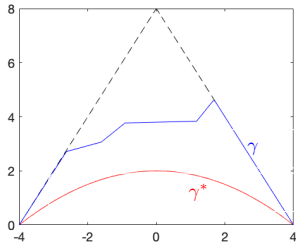
$$u^T = \mathbb{1}_{(-\infty, 0)} - \mathbb{1}_{(0, \infty)}$$



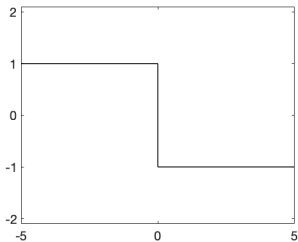
$$S_T^-(u^T) \text{ and } u_0 \text{ such that } S_T^+(S_T^-(u^T)) = S_T^+(u_0) = u^T.$$



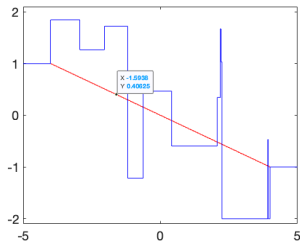
$$(t, x) \rightarrow S_t^+(u_0)(x)$$



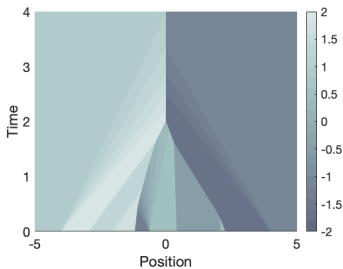
$$\gamma(x) := \int_{-4}^x u_0(s) ds \geq \gamma^*(x), \forall x \in [-4, 4]$$



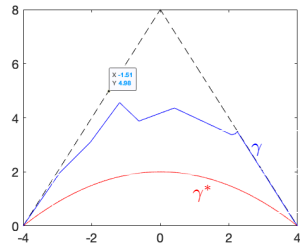
$$u^T = \mathbb{1}_{(-\infty, 0)} - \mathbb{1}_{(0, \infty)}$$



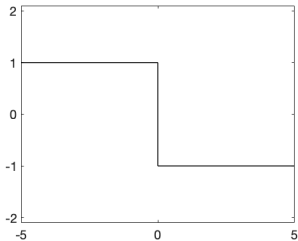
$$S_T^-(u^T) \text{ and } u_0 \text{ such that } S_T^+(S_T^-(u^T)) = S_T^+(u_0) = u^T.$$



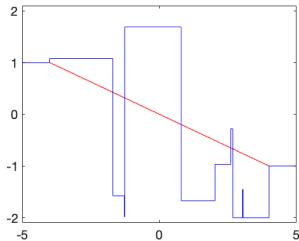
$$(t, x) \rightarrow S_t^+(u_0)(x)$$



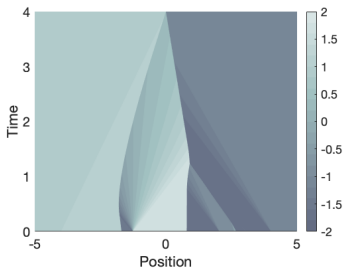
$$\gamma(x) := \int_{-4}^x u_0(s) ds \geq \gamma^*(x), \forall x \in [-4, 4]$$



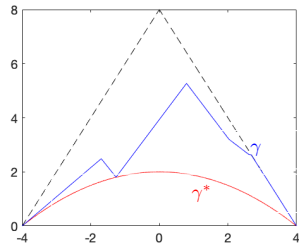
$$u^T = \mathbb{1}_{(-\infty, 0)} - \mathbb{1}_{(0, \infty)}$$



$$S_T^-(u^T) \text{ and } u_0 \text{ such that } S_T^+(S_T^-(u^T)) = S_T^+(u_0) = u^T.$$



$$(t, x) \rightarrow S_t^+(u_0)(x)$$



$$\gamma(x) := \int_{-4}^x u_0(s) ds \geq \gamma^*(x), \forall x \in [-4, 4]$$

Multiple initial data leading to a shock u^T

$$S_T^+(u_0) = u^T$$

Construction of multiple initial data leading to a shock

$$\text{Assuming that } u^T = \begin{cases} u_L & \text{if } x < \bar{x}, \\ u_R & \text{if } x > \bar{x}. \end{cases}$$

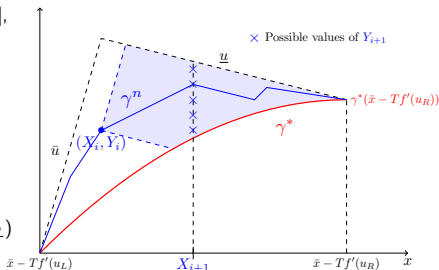
We construct a state mesh $\mathcal{M}_n = \underline{u} + (\bar{u} - \underline{u})2^{-n}\mathbb{N} \cap [0, 1]$ such that $\underline{u} \leq u_0 \leq \bar{u}$ and $u_L, u_R \in \mathcal{M}_n$

Construction of a path γ^n such that

- $\gamma^n(x) \geq \gamma^*(x), \forall x \in [\bar{x} - Tf'(u_L), \bar{x} - Tf'(u_R)],$
- $\gamma^n(\bar{x} - Tf'(u_L)) = 0,$
- $\gamma^n(\bar{x} - Tf'(u_R)) = \gamma^*(\bar{x} - Tf'(u_R)),$
- $\dot{\gamma}^n \in \mathcal{M}_n.$

Construction of u_0 such that $S_T^+(u_0) = u^T :$

$$u_0 = \begin{cases} u_L & \text{for } x < \bar{x} - Tf'(u_L) \\ \dot{\gamma}^n & \text{for a.e. } \bar{x} - Tf'(u_L) \leq x \leq \bar{x} - Tf'(u_R) \\ u_R & \text{for } \bar{x} - Tf'(u_R) < x \end{cases}$$



We consider the following optimal control problem

$$\min_{u_0 \in \mathcal{U}_{\text{ad}}^0} J_0(u_0) := \int_{\mathbb{R}} (u^T(x) - S_T^+(u_0))^2 dx, \quad (\text{Opt-Pb-1})$$

$$\mathcal{U}_{\text{ad}}^0 = \{u_0 \in BV(\mathbb{R}) / \|u_0\|_{BV(\mathbb{R})} < C \text{ and } \text{Supp}(u_0) \subset K_0\}.$$

$$\{\exists u_0 \in BV(\mathbb{R}) / S_T^+(u_0) = q\} \text{ iff } \{q \in BV(\mathbb{R}) / \partial_x q \leq \frac{1}{T}\}$$

$$\min_{q \in \mathcal{U}_{\text{ad}}^1} J_1(q) := \|u^T - q\|_{L^2(\mathbb{R})}, \quad (\text{Opt-Pb-2})$$

$$\mathcal{U}_{\text{ad}}^1 = \{q \in BV(\mathbb{R}) / \partial_x q \leq \frac{1}{T} \text{ and } \|q\|_{BV(\mathbb{R})} \leq C \text{ and } \text{Supp}(q) \subset K_1\}.$$

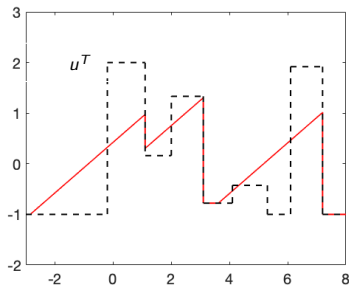
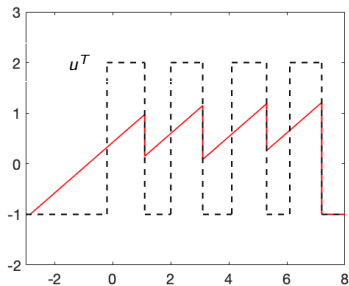
Using $S_T^-(S_T^+(S_T^-(u^T))) = S_T^-(u^T)$ and a full characterization of u_0 such that $S_T^-(u_0) = S_T^-(u^T)$

$S_T^+(S_T^-(u^T))$ is the unique critical point of (Opt-Pb-2).

Construction of an optimal solution

We consider the following optimal control problem

$$\min_{u_0} \int_{\mathbb{R}} (u^T(x) - S_T^+(u_0))^2 dx,$$

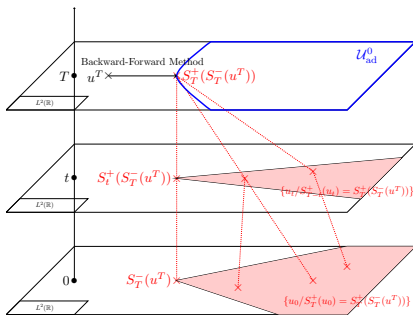


Plotting of the target u^T and $x \rightarrow S_T^+(S_T^-(u^T))(x)$
with $S_T^-(u^T)$ an optimal solution.

Plotting of multiple optimal solutions

$$S_T^+(u_0) = S_T^+(S_T^-(u^T))$$

$$\text{Optimal problem : } \min_{u_0 \in \mathcal{U}_{\text{ad}}^0} \int_{\mathbb{R}} (u^T(x) - S_T^+(u_0))^2 dx \quad (\text{Opt-Pb})$$



→ Fully characterization of minimizers for (Opt-Pb)

- Construction of the minimizer $S_T^+(S_T^-(u^T))$ of (Opt-Pb) via a **backward-forward method**
- u_0 is a minimizer of (Opt-Pb) iff $S_T^+(u_0) = S_T^-(u^T)$

→ **Implementation of a WFT algorithm** to pick up randomly one of the minimizer of (Opt-Pb)

Open problems

- 1 It would be interesting to extend this work to an “augmented Burgers equation” in order to minimize the sonic boom effects caused by supersonic aircrafts.
- 2 We may also consider a convex-concave function as a flux function in (PDE) which is for instance a more realistic choice to describe the flow of pedestrian.
- 3 We can also investigate systems of conservation laws in one dimension (Euler equations, Shallow water equations).
- 4 To finish, it would be interesting to study numerically the inverse design of multidimensional Burgers equation.