

# Multilevel Selective Harmonic Modulation via Optimal Control

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## Abstract

We consider the *Selective Harmonic Modulation* (SHM) problem, consisting in the design of a staircase control signal with some prescribed frequency components. In this work, the SHM problem is addressed as an optimal control one in which the admissible controls are piece-wise constant functions, taking values only in a given finite set. In order to fulfill this constraint, we introduce a cost functional **with piece-wise linear penalization** which, by means of Pontryagin's maximum principle, makes the optimal control have the desired staircase form. An advantage of our approach, **relevant in practical electrical engineering applications**, is that the number of commutations and the waveform need not be specified a priori. Indeed, our algorithm provides an admissible waveform as well as the location of the switches. Up to the best of our knowledge, this approach to the SHM problem via optimal control is new. Moreover, our methodology may be applicable to other optimal control problems with a finite-set constraint on the control. We also provide several numerical examples in which the SHM problem is solved by using our approach.

*Key words:* Selective Harmonic Modulation; Optimal Control Theory; Finite-Set Control; Pontryagin's maximum principle; Piece-wise linear penalization.

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## 1 Introduction and motivations

Selective Harmonic Modulation (SHM) [Rodríguez et al., 2002] is a well-known methodology in electrical engineering, employed to improve the performances of a converter by controlling the phase and amplitude of the harmonics in its output voltage. As a matter of fact, this technique allows to increase the power of the converter and, at the same time, to reduce its losses.

Because of the growing complexity of modern electrical networks, consequence for instance of the high penetration of renewable energy sources, the demand in power of electronic converters is day by day increasing. For this and other reasons, SHM has been a preeminent research interest in the electrical engineering community, and a plethora of SHM-based techniques has been developed in

recent years. An incomplete bibliography includes [Duranay and Guldemir, 2017, Janabi et al., 2020, Yang et al., 2017].

In broad terms, the technique consists in generating a *control signal* with a desired harmonic spectrum by modulating or eliminating some specific lower-order Fourier coefficients. In practice, the signal is constructed as a step function with a finite number of switches, taking values only in a given finite set. Such a signal can be fully characterized by two features (see Figure 1):

1. The *waveform*, i.e. the sequence of values that the function takes **in its domain**.
2. The *switching angles*, i.e. the sequence of **points** where the signal switches from one value to following one.

Using this simple characterization of the signal, in many practical situations the SHM problem is reduced to a finite-dimensional optimization one in which, given a suitable waveform, the aim is to find the optimal location of the switching angles. This formulation is in analogy

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with what is known in the control literature as *switching controls* (see [Zuazua, 2011]), whose goal is to control the dynamics of a system by switching from an actuator to another in a systematic way so that, at each instant of time, only one actuator is active.

One of the principal difficulties in addressing the SHM problem through finite-dimensional optimization is that the choice of a suitable waveform may not always be evident. In addition to that, also determining the optimal number of switching angles may be a quite cumbersome task (see Section 3).

To overcome these difficulties, in this work we propose a new approach to SHM based on optimal control theory (see Section 4): the Fourier coefficients in the SHM problem are identified with the terminal state of a controlled dynamical system, where the control is actually the signal, solution to the SHM problem.

The main difficulty in our approach arises from the fact that we look for piece-wise constant solutions, taking values only in a given finite set.

This kind of constraint on the control prevents one from applying the classical tools in optimal control theory, such as Pontryagin's maximum principle. In order to bypass this difficulty, we introduce a penalization term for the control, which guarantees that the optimal control will have the desired staircase structure. The optimization is then carried out over the controls in  $L^\infty$ , and then, standard computational algorithms in optimal control theory can be implemented to solve the SHM problem.

The main contributions of the present paper are as follows:

1. We reformulate the SHM problem as an optimal control problem with a staircase constraint on the control.
2. We do not need to determine a priori the waveform of the solution nor the number of switching angles.
3. We introduce a penalization term for the control which implicitly induces the staircase property on the solution. Here we distinguish two cases: when the control set contains only two elements (bi-level or bang-bang control) and when it contains more than two elements (multilevel control).
4. Our approach allows one to compute solutions to the SHM problem with different waveforms.
- 5 Nuestra aproximacin nos permite encontrar soluciones en una regin amplia de los vectores target.

Revisar la estructura cuando el paper est terminado.

In this document, we follow the next structure. In the Section 2, we introduce a mathematical formulation of a general SHE problem. In the Section 3, we show the

classical formulation in SHE literature, and we show two main problems because of this approach. In the Section 4, we show our contributions in this problem, where explain the SHE problem as control problem and the strategies to obtain a piece-wise function as optimal control. In the Section 5 we show a concrete examples solve by control theory. In the Section 6, we summaries the proof associated to our contribution. By last, we finished with some conclusion and open problems.

## 2 Preliminaries

This section is devoted to the mathematical formulation of the SHM problem and to introduce the notation that will be used throughout the paper. Let

$$\mathcal{U} = \{u_1, \dots, u_L\} \quad (2.1)$$

be a given set of  $L \geq 2$  real numbers satisfying

$$u_1 = -1, u_L = 1 \text{ and } u_k < u_{k+1} \quad \forall k \in \{1, \dots, L\}.$$

The goal is to construct a step function

$$u(t) : [0, 2\pi) \rightarrow \mathcal{U},$$

with a finite number of switches, such that some of its lower-order Fourier coefficients take specific values prescribed a priori.

Due to applications in power converters, it is typical to only consider functions with *half-wave symmetry*, i.e. satisfying

$$u(t + \pi) = -u(t) \quad \text{for all } t \in [0, \pi). \quad (2.2)$$

In this way, we only need to determine  $u$  in the interval  $[0, \pi)$ . Moreover, as a consequence of this symmetry, the Fourier series of  $u$  only involves the odd terms (as the even terms just vanish), i.e.

$$u(t) = \sum_{\substack{j \in \mathbb{N} \\ j \text{ odd}}} a_j \cos(jt) + \sum_{\substack{j \in \mathbb{N} \\ j \text{ odd}}} b_j \sin(jt),$$

with

$$\begin{aligned} a_j &= \frac{2}{\pi} \int_0^\pi u(\tau) \cos(j\tau) d\tau, \\ b_j &= \frac{2}{\pi} \int_0^\pi u(\tau) \sin(j\tau) d\tau. \end{aligned} \quad (2.3)$$

In view of the half-wave symmetry (2.2), in what follows, we will only work with the restriction  $u|_{[0, \pi)}$ , which, with some abuse of notation, we still denote by  $u$ .

As we anticipated, we are only considering piece-wise constant functions with a finite number of switches

and taking values in  $\mathcal{U}$ . More precisely, we look for  $u : [0, \pi) \rightarrow \mathcal{U}$  of the form

$$u(t) = \sum_{m=0}^M s_m \chi_{[\phi_m, \phi_{m+1})}(t), \quad M \in \mathbb{N} \quad (2.4)$$

for some  $\mathcal{S} = \{s_m\}_{m=0}^M$  satisfying

$$s_m \in \mathcal{U} \text{ and } s_m \neq s_{m+1} \quad \forall m \in \{0, \dots, M\}$$

and  $\Phi = \{\phi_m\}_{m=1}^M$  such that

$$0 = \phi_0 < \phi_1 < \dots < \phi_M < \phi_{M+1} = \pi.$$

In (2.4),  $\chi_{[\phi_m, \phi_{m+1})}$  denotes the characteristic function of the interval  $[\phi_m, \phi_{m+1})$ . With these notations we just introduced, we can now define the waveform and the switching angles as follows.

**Definition 2.1** For a function  $u : [0, \pi) \rightarrow \mathcal{U}$  of the form (2.4), we refer to  $\mathcal{S}$  as the waveform, and we refer to  $\Phi$  as the switching angles.

Observe that any  $u$  of the form (2.4) is fully characterized by its waveform  $\mathcal{S}$  and switching angles  $\Phi$ . An example of such a function is displayed in Figure 1.

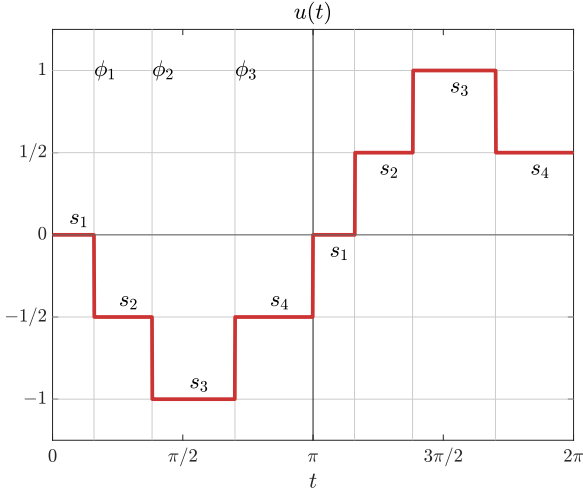


Fig. 1. A possible solution to the SHM Problem, where we considered the finite set of control  $\mathcal{U} = \{-1, -1/2, 0, 1/2, 1\}$ . We show the switching angles  $\Phi$  and the waveform  $\mathcal{S}$  (see Definition 2.1). The function  $u(t)$  is displayed on the whole interval  $[0, 2\pi)$  to highlight the half-wave symmetry introduced in (2.2).

In the practical engineering applications that motivated our study, due to technical limitations, it is preferable to employ signals taking consecutive values in  $\mathcal{U}$ . In the sequel, we will refer to this property of the waveform as the *staircase property*. We can rigorously formulate this property as follows: we say that a signal  $u$  of the

form (2.4) fulfills the staircase property if its waveform  $\mathcal{S}$  satisfies

$$(s_m^{\min}, s_m^{\max}) \cap \mathcal{U} = \emptyset, \quad \forall m \in \{0, \dots, M-1\}, \quad (2.5)$$

where

$$s_m^{\min} = \min\{s_m, s_{m+1}\} \\ s_m^{\max} = \max\{s_m, s_{m+1}\}.$$

Note that when  $\mathcal{U} = \{-1, 1\}$  (the bi-level case), this property is satisfied for any  $u$  of the form (2.4).

We can now formulate the SHM problem as follows:

**Problem 2.1 (SHM)** Let  $\mathcal{U}$  be given as in (2.1), and let  $\mathcal{E}_a$  and  $\mathcal{E}_b$  be two finite sets of odd numbers of cardinality  $|\mathcal{E}_a| = N_a$  and  $|\mathcal{E}_b| = N_b$  respectively. For any two given vectors  $\mathbf{a}_T \in \mathbb{R}^{N_a}$  and  $\mathbf{b}_T \in \mathbb{R}^{N_b}$ , construct a function  $u : [0, \pi) \rightarrow \mathcal{U}$  of the form (2.4), satisfying (2.5), such that the vectors  $\mathbf{a} \in \mathbb{R}^{N_a}$  and  $\mathbf{b} \in \mathbb{R}^{N_b}$ , defined as

$$\mathbf{a} = (a_j)_{j \in \mathcal{E}_a} \quad \text{and} \quad \mathbf{b} = (b_j)_{j \in \mathcal{E}_b} \quad (2.6)$$

satisfy

$$\mathbf{a} = \mathbf{a}_T \quad \text{and} \quad \mathbf{b} = \mathbf{b}_T,$$

where the coefficients  $a_j$  and  $b_j$  in (2.6) are given by (2.3).

**Remark 2.1 (SHE)** In Problem 2.1, we gave a very general formulation of SHM. This formulation contains also the so-called Selective Harmonic Elimination (SHE) problem (see [Sun et al., 1996]), in which the target vectors are such that

$$(a_T)_1 \neq 0 \quad (a_T)_{i \neq 1} = 0 \quad \forall i \in \mathcal{E}_a \\ (b_T)_1 \neq 0 \quad (b_T)_{j \neq 1} = 0 \quad \forall j \in \mathcal{E}_b.$$

SHE is of great relevance in the electric engineering literature. Its objective is to generate a signal with amplitude  $m_1 = \sqrt{a_1^2 + b_1^2}$  and phase  $\varphi_1 = \arctan(b_1/a_1)$ , removing the amplitude of the successive frequencies. De esta manera SHE se puede entender como un generador de modos de Fourier limpios mediante un seal escalonada.

### 3 SHM as a finite-dimensional optimization problem

A typical approach to address the SHM Problem 2.1 (see [Yang et al., 2015, Konstantinou and Agelidis, 2010, Sun et al., 1996]) consists in looking for solutions having a specific waveform determined a priori and looking for

suitable switching angles  $\Phi$  allowing to reach the desired target.

Note that, for a fixed waveform  $\mathcal{S}$ , the Fourier coefficients of a function  $u$  of the form (2.4) can be written in terms of the switching angles  $\Phi$  in the following way:

$$a_j = a_j(\Phi) = \frac{2}{j\pi} \sum_{m=0}^M s_m \left[ \sin(j\phi_{m+1}) - \sin(j\phi_m) \right]$$

$$b_j = b_j(\Phi) = \frac{2}{j\pi} \sum_{m=0}^M s_m \left[ \cos(j\phi_m) - \cos(j\phi_{m+1}) \right]$$

Hence, for two sets of odd numbers  $\mathcal{E}_a$  and  $\mathcal{E}_b$  as in Problem 2.1, and any fixed waveform  $\mathcal{S}$ , we can define the functions

$$\mathbf{a}_{\mathcal{S}}(\Phi) := (a_j(\Phi))_{j \in \mathcal{E}_a} \in \mathbb{R}^{N_a}$$

and

$$\mathbf{b}_{\mathcal{S}}(\Phi) := (b_j(\Phi))_{j \in \mathcal{E}_b} \in \mathbb{R}^{N_b}.$$

In this framework, SHM can then be cast as a finite-dimensional optimization process consisting in minimizing, over the choice of  $\Phi = \{\phi_m\}_{m=1}^M$ , the euclidean distance between the vectors  $(\mathbf{a}, \mathbf{b})$  and  $(\mathbf{a}_T, \mathbf{b}_T)$  as defined in Problem 2.1.

### Problem 3.1 (Optimization problem for SHM)

Let  $\mathcal{E}_a$ ,  $\mathcal{E}_b$  and the targets  $\mathbf{a}_T$  and  $\mathbf{b}_T$  be given as in Problem 2.1. Let  $\mathcal{S} := \{s_m\}_{m=0}^M$  be a fixed waveform satisfying (2.5). Find the switching angles  $\Phi$  solution to the following minimization problem:

$$\min_{\Phi \in [0, \pi]^M} \left( \|\mathbf{a}_{\mathcal{S}}(\Phi) - \mathbf{a}_T\|^2 + \|\mathbf{b}_{\mathcal{S}}(\Phi) - \mathbf{b}_T\|^2 \right)$$

$$\text{subject to: } 0 = \phi_0 < \phi_1 < \dots < \phi_M < \phi_{M+1} = \pi.$$

At this regard, it is important to notice that the optimization Problem 3.1 solves the original Problem 2.1 only when the optimal cost is zero.

cuando el coste ptimo es nulode manera que es necesario caracterizar el espacio de targets  $(\mathbf{a}_T, \mathbf{b}_T)$  donde existe solucin. Con este objetivo definiremos la funcin de coste ptima y luego a partir de ella el conjunto solucionable<sup>1</sup>.

<sup>1</sup> A continuacin utilizaremos adrede definiciones con nombres utilizados ampliamente en los problemas de control ptimo, de manera que el lector pueda notar la gran analogía que existe entre el problema finito dimensional y el problema de control ptimo propuesto en la subseccin ??.

**Definition 3.1 (Optimal Cost Function)** We call optimal cost function  $V_{\mathcal{S}} : \mathbb{R}^{N_a} \times \mathbb{R}^{N_b} \rightarrow \mathbb{R}$ , the function that take as input variables the vector targets  $\mathbf{a}_T$  and  $\mathbf{b}_T$  and return the optimal cost of the Problem 3.1.

**Definition 3.2 (Resolvable set)** We define a solutionable set  $\mathcal{R}_{\mathcal{S}}$  as:

$$\mathcal{R}_{\mathcal{S}} = \{(\mathbf{a}_T, \mathbf{b}_T) \in \mathbb{R}^{N_a+N_b} | V_{\mathcal{S}}(\mathbf{a}_T, \mathbf{b}_T) = 0\} \quad (3.1)$$

En la prctica solo son relevantes las soluciones del Problema 3.1 para los targets  $(\mathbf{a}_T, \mathbf{b}_T) \in \mathcal{R}_{\mathcal{S}}$ , de manera que es ideal obtener una funcin que contenga los ngulos de conmutacin ptimos  $\Phi_{\mathcal{S}}^*$  en todo el conjunto  $\mathcal{R}_{\mathcal{S}}$ . Definiremos esta funcin llamandola Poltica con la forma de onda  $\mathcal{S}$

**Definition 3.3 (Policy Function)** Una funcin  $\Pi_{\mathcal{S}} : \mathcal{R}_{\mathcal{S}} \rightarrow [0, \pi]^M$  tal que  $\Phi^* = \Pi_{\mathcal{S}}(\mathbf{a}_T, \mathbf{b}_T)$ , siendo  $\Phi^*$  los ngulos de conmutacin ptimos solucin del Problema 2.1 para los targets  $(\mathbf{a}_T, \mathbf{b}_T)$ .

Con el fin de reconstruir la funcin poltica  $\Pi_{\mathcal{S}}$  se calculan soluciones numricas del Problema 3.1 en pocos puntos de  $\mathbb{R}^{N_a+N_b}$ , luego se comprueba que el coste ptimo sea nulo y por ltimo se interpolar la funcin  $\Pi_{\mathcal{S}}$  en el conjunto convexo formado por los puntos obtenidos.

Let us point out some of the main difficulties and drawback arising in this approach.

### 3.1 Combinatory problem

In practice, one does not dispose of a suitable waveform  $\mathcal{S}$  which yields a solution to the Problem 2.1. A common approach to solve the SHM problem consists in fixing the number of switches  $M$ , and then solve Problem 3.1 for all the possible combinations of  $M$  elements of  $\mathcal{U}$ . However, taking into account that the number of possible  $M$ -tuples in  $\mathcal{U}$  is of the order  $(L-1)^M$ , it is evident that the complexity of the above approach increases rapidly when  $L > 1$ . This problem has been studied in [Yang et al., 2015] where, through appropriate algebraic transformations, the authors are able to convert the SHM problem into a polynomial system whose solutions' set contains all the possible waveforms  $\mathcal{S}$  of  $M$  elements in  $\mathcal{U}$ . As a drawback of this approach, the number of switches  $M$  needs to be prefixed. However, in some cases, determining the number of switches which is necessary to reach the desired Fourier coefficients is not a straightforward task.

### 3.2 Problema con el tamao de conjunto solucionable

Dado una forma de onda  $\mathcal{S}$ , el tamao de su conjunto solucionable  $\mathcal{R}_{\mathcal{S}}$  suele ser pequeno dando politicas  $\Pi_{\mathcal{S}}$  poco

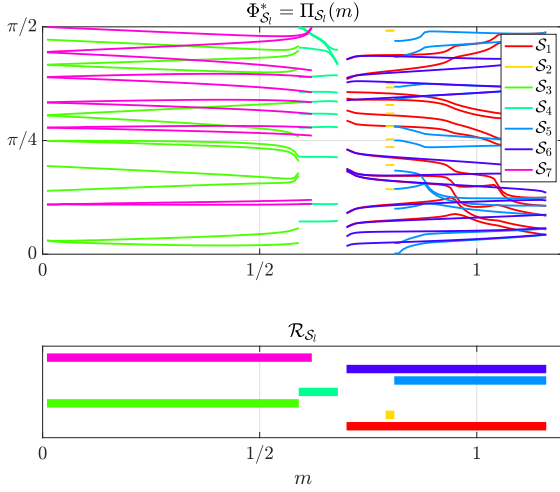


Fig. 2. En la figura superior se muestra los ngulos de conmutacin ptimos  $\Phi_{S_i}^*$  para varias formas de onda  $\{S_i\}_{i=1}^7$  para un problema SHE (vase Remark 2.1), considerando los conjuntos  $\mathcal{E}_b = \{1, 5, 7, 11, 13, 17, 19, 23, 25, 29, 31\}$  y  $\mathcal{E}_a = \{1\}$ . Se ha tomado el target  $(a_1, b_1) = (m, 0) \forall m \in [0, 1.2]$ . En la Figura inferior muestra los espacios solucionables para cada forma de onda.

tiles. Esta problemtica se aborda solucionando el Problema 3.1 para un conjunto de formas de onda  $\{S_i\}_{i=1}^r$ , obteniendo polticas  $\{\Pi_{S_i}\}_{i=1}^r$  y conjuntos solucionables  $\{\mathcal{R}_{S_i}\}_{i=1}^r$  para cada uno de ellas. Entonces mediante la unin de polticas es posible obtener un poltica en un conjunto ms mplio. Sin embargo, dado que en general las distintas forma de onda puede tener conjuntos solucionables disjuntos o solapados, esta unin es cotica, obteniendo varias soluciones para un mismo vector target  $(\mathbf{a}_T, \mathbf{b}_T)$  o incluso regiones sin solucin (vase Figura 2).

### 3.3 Problema de la poltica catica

Debido al complejidad de un poltica formada por unin de distintas formas de onda, la continuidad de los ngulos de conmutacin no se puede asegurar. Este comportamiento lo informa [Yang et al., 2017], mostrando el caos la funcin  $\Pi_S$  (vase Figura 2).

## 4 SHM problem as an Optimal Control Problem

Our main contribution in the present paper consists in formulating the SHM problem from the viewpoint of optimal control. In this formulation, the Fourier coefficients of the signal  $u(t)$  are identified with the terminal state of a controlled dynamical system of  $N_a + N_b$  components defined in the time-interval  $[0, \pi]$ . The control of the system is precisely the signal  $u(t)$ , defined as a function  $[0, \pi] \rightarrow \mathcal{U}$ , which has to steer the state from the origin to the desired values of the prescribed Fourier coefficients.

The starting point of this approach is to rewrite the

Fourier coefficients of the function  $u(t)$  as the final state of a dynamical system controlled by  $u(t)$ .

To this end, let us first note that, in view of (2.3), for all  $u \in L^\infty([0, \pi]; \mathbb{R})$  any Fourier coefficient  $a_j$  satisfies

$$a_j = y(\pi),$$

with  $y \in C([0, \pi]; \mathbb{R})$  defined as

$$y(t) = \frac{2}{\pi} \int_0^t u(\tau) \cos(j\tau) d\tau.$$

Besides, as a consequence of the fundamental theorem of calculus,  $y(\cdot)$  is the unique solution to the differential equation

$$\begin{cases} \dot{y}(t) = \frac{2}{\pi} \cos(jt) u(t), & t \in [0, \pi] \\ y(0) = 0. \end{cases} \quad (4.1)$$

Analogously, we can also write the Fourier coefficients  $b_j$ , defined in (2.3), as the solution at time  $t = \pi$  of a differential equation similar to (4.1).

Hence, for  $\mathcal{E}_a$ ,  $\mathcal{E}_b$ ,  $\mathbf{a}_T$  and  $\mathbf{b}_T$  given, the SHM Problem 2.1 can be reduced to finding a control  $u(\cdot)$  of the form (2.4), satisfying (2.5), such that the solution  $\mathbf{y} \in C([0, \pi]; \mathbb{R}^{N_a + N_b})$  to the dynamical system

$$\begin{cases} \dot{\mathbf{y}}(t) = \frac{2}{\pi} \mathcal{D}(t) u(t), & t \in [0, \pi] \\ \mathbf{y}(0) = 0. \end{cases} \quad (4.2)$$

satisfies

$$\mathbf{y}(\pi) = [\mathbf{a}_T; \mathbf{b}_T]^\top,$$

where

$$\mathcal{D}(t) = [\mathcal{D}^a(t); \mathcal{D}^b(t)]^\top, \quad (4.3)$$

with  $\mathcal{D}^a(t) \in \mathbb{R}^{N_a}$  and  $\mathcal{D}^b(t) \in \mathbb{R}^{N_b}$  given by

$$\mathcal{D}^a(t) = \begin{bmatrix} \cos(e_a^1 t) \\ \cos(e_a^2 t) \\ \vdots \\ \cos(e_a^{N_a} t) \end{bmatrix}, \quad \mathcal{D}^b(t) = \begin{bmatrix} \sin(e_b^1 t) \\ \sin(e_b^2 t) \\ \vdots \\ \sin(e_b^{N_b} t) \end{bmatrix} \quad (4.4)$$

Here,  $e_a^i$  and  $e_b^i$  denote the elements in  $\mathcal{E}_a$  and  $\mathcal{E}_b$ , i.e.

$$\mathcal{E}_a = \{e_a^1, e_a^2, e_a^3, \dots, e_a^{N_a}\}, \quad \mathcal{E}_b = \{e_b^1, e_b^2, e_b^3, \dots, e_b^{N_b}\}.$$

Moreover, for notation simplicity, we reverse the time in (4.2) using the transformation  $\mathbf{x}(t) = \mathbf{y}(\pi - t)$ . In



this way, the SHM problem converts into the following null-controllability one with initial condition  $\mathbf{x}(0) = [\mathbf{a}_T; \mathbf{b}_T]^\top$  (see also Figure 3):

**Problem 4.1 (SHM as null-controllability problem)**

Let  $\mathcal{U}$  be a given set as in (2.1), and let  $\mathcal{E}_a$  and  $\mathcal{E}_b$  be two finite sets of odd numbers of cardinality  $|\mathcal{E}_a| = N_a$  and  $|\mathcal{E}_b| = N_b$  respectively. For any two given vectors  $\mathbf{a}_T \in \mathbb{R}^{N_a}$  and  $\mathbf{b}_T \in \mathbb{R}^{N_b}$ , we look for a function  $u : [0, \pi] \rightarrow [-1, 1]$  of the form (2.4), satisfying (2.5), such that the solution to the initial-value problem

$$\begin{cases} \dot{\mathbf{x}}(t) = -\frac{2}{\pi} \mathcal{D}(t)u(t), & t \in [0, \pi] \\ \mathbf{x}(0) = \mathbf{x}_0 := [\mathbf{a}_T; \mathbf{b}_T] \end{cases} \quad (4.5)$$

satisfies  $\mathbf{x}(\pi) = 0$ , where  $\mathcal{D}$  is given by (4.3)–(4.4).

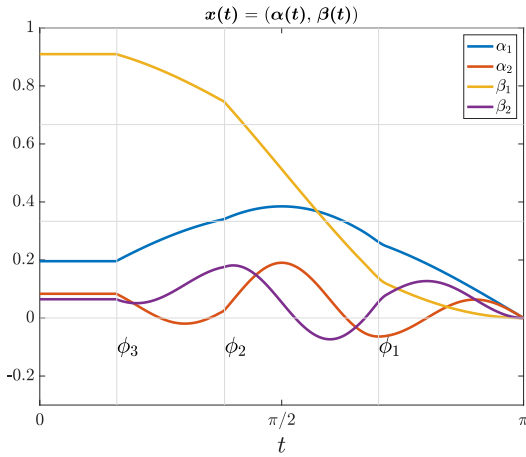


Fig. 3. Evolution of the dynamical system (4.5) with  $\mathcal{E}_a = \{1, 2\}$  and  $\mathcal{E}_b = \{1, 2\}$  corresponding to the control  $u$  in Figure 1. The positions of the switching angles  $\phi$  are displayed as well.

Gracias al Theorem 3.6.1 (Bang-bang theorem for compact polyhedra) formulado en [Schättler and Ledzewicz, 2012, Chapter 2.7], sabemos que un problema de control para un sistema Linear Time Invariant se puede construir un control constante a trozos (2.1) para cualquier puntos que este en el conjunto alcanzable. El conjunto alcanzable es el conjunto ms grande en el existe solucin al problema SHM. De esta manera, debido a la libertad de cambiar de forma de onda en el problema de control, podemos solucionar el problema de la combinatoria (Subseccin 3.2) y el Problema del tamao del conjunto solucionable.

A natural approach for controllability problems such as Problem 4.1 is to formulate them as an optimal control one, where the cost functional to be minimized is the euclidean distance from the final state  $\mathbf{x}(\pi)$  to the target, in this case, the origin. In what follows, for a given vector  $\mathbf{v} \in \mathbb{R}^d$ , we shall always denote by  $\|\mathbf{v}\|$  the euclidean norm  $\|\mathbf{v}\|_{\mathbb{R}^d}$ .

**Problem 4.2 (OCP for SHM)** Let  $\mathcal{U}$  be a given set as in (2.1) and let  $\mathcal{E}_a$  and  $\mathcal{E}_b$  be two finite sets of odd numbers of cardinality  $|\mathcal{E}_a| = N_a$  and  $|\mathcal{E}_b| = N_b$  respectively. For any two given vectors  $\mathbf{a}_T \in \mathbb{R}^{N_a}$  and  $\mathbf{b}_T \in \mathbb{R}^{N_b}$ , we look for an admissible control  $u \in \mathcal{U}_{ad}$  solution to the following optimal control problem:

$$\min_{u \in \mathcal{U}_{ad}} \frac{1}{2} \|\mathbf{x}(\pi)\|^2 \quad \text{subject to the dynamics (4.5),}$$

where the set of admissible controls  $\mathcal{U}_{ad}$  is defined as

$$\mathcal{U}_{ad} := \{u \in L^\infty([0, \pi]; [-1, 1]) \text{ of the form (2.4) satisfying (2.5)}\}.$$

**Remark 4.1** Note that, due to the coercivity of the cost functional, Problem 4.2 admits at least one solution for any target  $[\mathbf{a}_T; \mathbf{b}_T]^\top$ . However, we say that the SHM problem admits a solution if and only if the minimum in Problem 4.2 is equal to zero. Otherwise, we say that the target  $[\mathbf{a}_T; \mathbf{b}_T]^\top$  is unreachable. In this work, we will not discuss the reachable set for the control problem (4.2).

A main feature of the SHM problem is that we are looking for signal functions  $u$  of the form (2.4) satisfying (2.5). In principle, this can be guaranteed by adding directly this constraint on the set of admissible controls  $\mathcal{U}_{ad}$ . Notwithstanding that, the inclusion of such a constraint makes the optimal control problem extremely difficult, as it prevents us from applying standard arguments such as the Pontryagin maximum principle, and implementing all the computational techniques developed in the last decades to solve optimal control problems. As a matter of fact, the classical tools in optimal control theory seem not to be adapted to handle such constraints on the control. In order to bypass this difficulty, we propose a modification of Problem 4.2 by adding to the cost functional a penalization term for the control.

In what follows, if not stated differently, we will always denote

$$L^\infty := L^\infty([0, \pi]; [-1, 1]).$$

We can then give the following equivalent formulation of Problem 4.2.

**Problem 4.3 (Penalized OCP for SHM)** Fix  $\varepsilon > 0$  and a function  $\mathcal{L} \in C([-1, 1]; \mathbb{R})$ . Let  $\mathcal{E}_a$  and  $\mathcal{E}_b$  be two finite sets of odd numbers of cardinality  $|\mathcal{E}_a| = N_a$  and  $|\mathcal{E}_b| = N_b$  respectively. For any two given vectors  $\mathbf{a}_T \in \mathbb{R}^{N_a}$  and  $\mathbf{b}_T \in \mathbb{R}^{N_b}$ , we look for a control  $u \in L^\infty$  solution to the following optimal control problem:

$$\begin{aligned} \min_{u \in L^\infty} \quad & \frac{1}{2} \|\mathbf{x}(\pi)\|^2 + \varepsilon \int_0^\pi \mathcal{L}(u(t)) dt \\ \text{subject to the dynamics} \quad & (4.5). \end{aligned}$$

Observe that, in Problem 4.3, we do not impose the constraint that the control has to be the form (2.4) and satisfy the staircase property. Nevertheless, as we shall see, these properties of  $u$  will arise naturally from a suitable choice of the Lagrangian  $\mathcal{L}$ .

Another important aspect that needs to be taken into account is that the penalization term for the control might prevent the optimal trajectory from reaching the target. In other words, even if there exists a control for which the optimal trajectory satisfies  $\mathbf{x}(\pi) = 0$ , the optimal control in Problem 4.3 might not do so, and therefore, the solution to Problem 4.3 would not be a solution to the SHM problem. This issue may be prevented by a proper selection of the weighting parameter  $\varepsilon$  which allows to tune the precision of the optimal control for the perturbed problem, guaranteeing that the final state of the optimal trajectory is close enough to zero. In more detail, we have the following result.

**Proposition 4.1** *Assume that  $[\mathbf{a}_T, \mathbf{b}_T]^\top$  is such that Problem 4.1 admits a solution, and let  $u^* \in L^\infty$  be the solution to Problem 4.3. Then the associated trajectory  $\mathbf{x}^* \in C([0, \pi]; \mathbb{R})$ , solution to (4.5), satisfies*

$$\|\mathbf{x}^*(\pi)\|^2 \leq 4\varepsilon\pi\|\mathcal{L}\|_\infty,$$

where  $\|\cdot\|_\infty$  denotes the max-norm in  $C([-1, 1]; \mathbb{R})$ .

The proof of Proposition 4.1 is postponed to Section 6. Let us now describe the construction of penalization functions  $\mathcal{L}$  which guarantee that any solution to Problem 4.3 has the form (2.4) and satisfies (2.5). To this end, we will distinguish two cases, depending on the cardinality of  $\mathcal{U}$ .

#### 4.1 Bilevel SHM problem via OCP (Bang-Bang Control)

In this case, the control set  $\mathcal{U}$  defined in (2.1) has only two elements, i.e.  $\mathcal{U} = \{-1, 1\}$ . In the control theory literature, a control taking only two values is known as *bang-bang control*. In the SHM literature, this kind of solution are called *bi-level solutions*. Note that in this case, any  $u$  with the form (2.4) trivially satisfies the staircase property (2.5).

The first main result of the present paper is the following.

**Theorem 4.1** *Let  $\mathcal{U} = \{-1, 1\}$ , and consider Problem 4.3. If the function  $\mathcal{L} \in C([-1, 1]; \mathbb{R})$  is concave, then any optimal control  $u^*$ , solution to Problem 4.3, has a bang-bang structure, i.e. it has the form (2.4).*

The proof of Theorem 4.1 is postponed to Section 6, and follows from the optimality conditions given by the Pontryagin's maximum principle. The fact that the control

$u$  acts linearly in (4.5), together with the concavity of  $\mathcal{L}$ , implies that the Hamiltonian is also concave, and then, the minimum is always attained at the limits of the interval  $[-1, 1]$ .

We point out that, by choosing different penalization functions  $\mathcal{L}$ , we can obtain solutions to the SHE problem with different waveforms. Typical choices which give rise to bang-bang controls are

$$\mathcal{L}(u) = u, \quad \mathcal{L}(u) = -u, \quad \mathcal{L}(u) = -u^2.$$

See Figure 7 for an illustration.

#### 4.2 Multilevel SHM problem via OCP

Inspired by the ideas of the previous subsection, we can address the case when  $\mathcal{U}$  contains more than two elements. This is known in the electrical engineering literature as the *multilevel SHM problem*. Now, the goal is to construct a function  $\mathcal{L}$  such that the Hamiltonian associated to Problem 4.3 always attains the minimum at points in  $\mathcal{U}$ . A way to construct such a function  $\mathcal{L}$  is to interpolate a parabola in  $[-1, 1]$  by affine functions, considering the elements in  $\mathcal{U}$  as the interpolating points. Since, between any two points in  $\mathcal{U}$ , the function  $\mathcal{L}$  is a straight line, the Hamiltonian is a concave function in these intervals, and hence, the minimum is always attained at points in  $\mathcal{U}$ .

**Theorem 4.2** *Let  $\mathcal{U}$  be a given set as in (2.1). For some  $0 < a \in \mathbb{R}$  and  $b \in \mathbb{R}$ , set the function*

$$\mathcal{P}(u) = a(u - b)^2.$$

Consider the Problem 4.3 with

$$\mathcal{L}(u) = \begin{cases} \lambda_k(u) & \text{if } u \in [u_k, u_{k+1}[ \\ \mathcal{P}(1) & \text{if } u = u_L \end{cases} \quad (4.6)$$

$$\forall k \in \{1, \dots, L-1\}$$

where

$$\lambda_k(u) = \frac{(u - u_k)\mathcal{P}(u_{k+1}) + (u_{k+1} - u)\mathcal{P}(u_k)}{u_{k+1} - u_k}. \quad (4.7)$$

Then any optimal control  $u^*$ , solution to Problem 4.3, has the form (2.4) and satisfies (2.5).

**Remark 4.2 (Bang-off-bang control)** We note that when  $\mathcal{U} = \{-1, 0, 1\}$ , as penalization function we can just use the  $L^1$ -norm of the control, i.e.  $\mathcal{L}(u) = |u|$ . This yields to the so-called bang-off-bang controls, that are widely studied in the literature [Wang and Toppo, 2020]. By taking a different parabola  $\mathcal{P}$ , one can then obtain different bang-off-bang solutions to the SHM problem.

We illustrate in Figure 8 different examples of penalization functions  $\mathcal{L}$  giving rise to multilevel solutions to the SHM problem. We point out that, by varying the values of  $a$  and  $b$  in Theorem 4.2, we can obtain solutions with different waveforms.

**Remark 4.3** For completeness, we shall mention that in Theorem 4.2 the Lagrangian  $\mathcal{L}$  can actually have a more general form than (4.6), still yielding to an optimal control  $u^*$  with the form (2.4) and satisfying (2.5). In more detail, given a strictly convex function  $\mathcal{P}(u)$ , we can define  $\mathcal{L}(u)$  as in (4.6) but with a more general selection of  $\lambda_k$ . For having controls in the desired form for the SHM problem, we only have to take into account the sufficient condition that, for all  $k \in \{1, \dots, N_u - 1\}$ , the functions  $\lambda_k$  are concave.

Examples of functions which are compatible with this situation are

1. A union of concave functions (see the second plot in Figure 8):

$$\lambda_k(u) = -4u^2 + 2(u_k + u_{k+1}) - 2u_k.$$

2. Linear approximation of shifted parabola (see the third plot in Figure 8)

$$\lambda_k(u) = \frac{1}{4} \left[ (u_{k+1} + u_k)(u - u_k - 1) + u_k^2 \right].$$

## 5 Numerical simulations

In this section, we present several examples in which we solve our optimal control problem through the direct method and the non-linear constrained optimization tool CasADi [Andersson et al., 2019]. All the simulations we are going to present can be found also in [Oroya et al., 2021]. Our Experiments were conducted on a personal MacBook Pro laptop (1,4 GHz Quad-Core Intel Core i5, 8GB RAM, Intel Iris Plus Graphics 1536 MB).

### 5.1 Smooth approximation of piece-wise linear penalization

With the final aim of using an optimization software to solve our optimal control problem, we will approximate our piece-wise linear penalization with the help of the Heaviside function  $h : \mathbb{R} \rightarrow \mathbb{R}$  and its smooth approximation defined as follows:

$$h(x) = \begin{cases} 1 & \text{if } x > 0 \\ 1/2 & \text{if } x = 0 \\ 0 & \text{if } x < 0 \end{cases}, \quad \begin{cases} h^\theta(x) = \frac{1 + \tanh(\theta x)}{2} \\ \lim_{\theta \rightarrow +\infty} h^\theta = h. \end{cases}$$

Using  $h^\theta$ , for all  $k \in \{1, \dots, N_u - 1\}$  we can then define the (smooth) function  $\Pi_{u_k, u_{k+1}}^\theta : \mathbb{R} \rightarrow \mathbb{R}$  as:

$$\begin{aligned} \Pi_{u_k, u_{k+1}}^\theta(x) &= -1 + h^\theta(x - u_k) + h^\theta(-x + u_{k+1}) \\ &= \frac{\tanh[\theta(x - u_k)] + \tanh[\theta(u_{k+1} - x)]}{2}. \end{aligned}$$

and use it to build the smooth version of (4.6)

$$\mathcal{L}^\theta(u) = \sum_{k=1}^{N_u-1} \left[ (u_{k+1} + u_k)(u - u_k) + u_k^2 \right] \Pi_{u_k, u_{k+1}}^\theta(u),$$

so that  $\mathcal{L}^\theta \rightarrow \mathcal{L}$  when  $\theta \rightarrow +\infty$ .

### 5.2 Direct method for OCP-SHE

To solve the optimal control problem (4.3), we use a direct method. If we consider a partition  $\mathcal{P} = \{\tau_0, \tau_1, \dots, \tau_T\}$  of interval  $[0, T]$ , we can represent a function  $\{u(\tau) \mid \tau \in [0, T]\}$  as a vector  $\mathbf{u} \in \mathbb{R}^T$  where component  $u_t = u(\tau_t)$ . Then the optimal control problem (4.2) can be written as optimization problem with variable  $\mathbf{u} \in \mathbb{R}^T$ . This problem is a nonlinear programming, for this we use CasADi software to solve. Hence, given a partition of the interval  $[0, \pi)$ , we can formulate the problem 4.3 as the following one in discrete time

**Problem 5.1 (Numerical OCP)** Given two sets of odd numbers  $\mathcal{E}_a$  and  $\mathcal{E}_b$  with cardinalities  $|\mathcal{E}_a| = N_a$  and  $|\mathcal{E}_b| = N_b$  respectively, given the target vectors  $\mathbf{a}_T \in \mathbb{R}^{N_a}$ , so that  $\mathbf{x}_0 = [\mathbf{a}_T, \mathbf{b}_T]^\top$  and  $\mathbf{b}_T \in \mathbb{R}^{N_b}$  and a partition  $\mathcal{P}_\tau = \{\tau_0, \tau_1, \dots, \tau_T\}$  of the interval  $[0, \pi)$ , we search a vector  $\mathbf{u} \in \mathbb{R}^T$  that solves the following minimization problem:

$$\min_{\mathbf{u} \in \mathbb{R}^T} \left[ \|\mathbf{x}^T\|^2 + \varepsilon \sum_{t=0}^{T-1} \left[ \frac{\mathcal{L}^\eta(u_t) + \mathcal{L}^\eta(u_{t+1})}{2} \Delta\tau_t \right] \right]$$

subject to:

$$\forall \tau \in \mathcal{P} \begin{cases} \mathbf{x}^{t+1} = \mathbf{x}^t - \Delta\tau_t \mathbf{k}/6 \\ \mathbf{x}^0 = \mathbf{x}_0 \end{cases}$$

where

$$\begin{cases} \mathbf{k} = (k_1 + 2k_2 + 2k_3 + k_4) \\ k_1 = \\ k_2 = \\ k_3 = \\ k_4 = \end{cases} \quad (5.1)$$



### 5.3 Examples

All the examples that we are going to present will share the following common parameter:  $\varepsilon = 10^{-5}$ ,  $\eta = 10^{-5}$  and  $\mathcal{P}_t = \{0.0, 0.1, 0.2, \dots, \pi\}$ . Moreover, we will consider  $\mathcal{E}_a = \{1, 5, 7\}$  and  $\mathcal{E}_b = \{1, 5, 7\}$ , and the target vectors  $\mathbf{a}_T = (m, 0, 0)^T$  and  $\mathbf{b}_T = (m, 0, 0)^T$  for all  $m \in [-0.8, 0.8]$ . We will treat all the cases specific types of controls we mentioned before, that is, *bang-bang*, *bang-off-bang* and *multilevel*.

**Simulation 5.1 (Bang-Bang)** We consider Problem 5.1 with a set of admissible controls:

$$\mathcal{U} = \{-1, 1\}$$

In accordance with Theorem 4.1, we can see in Figure 4 how for all  $i_a \in [-0.8, 0.8]$  the optimal control only takes the values  $\{-1, 1\}$  displayed in the plot in blue and red, respectively.

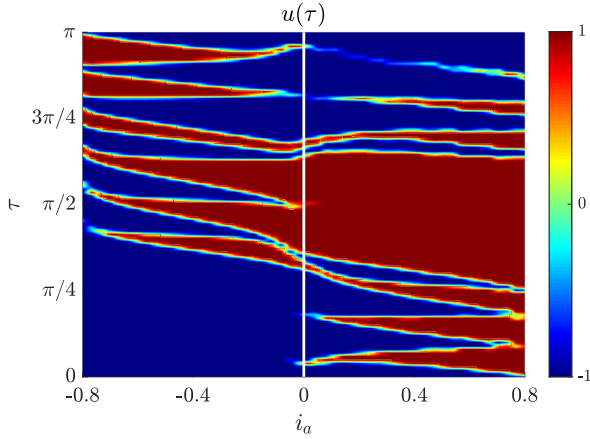


Fig. 4. Results of simulation of Simulation 5.1.

**Simulation 5.2 (Bang-off-Bang)** We consider Problem 5.1 with a set of admissible controls:

$$\mathcal{U} = \{-1, 0, 1\}$$

In the same way as before, when we consider three possible values  $\mathcal{U} = \{-1, 0, 1\}$ , we recover the *bang-off-bang* controls (Figure 5).

**Simulation 5.3 (Multi-level)** We consider Problem 5.1 with a set of admissible controls:

$$\mathcal{U} = \{-1, -1/2, 0, 1/2, 1\}$$

Finally, we can see in Figure 6 the same behavior when considering el mismo comportamiento cuando consideramos de el conjunto de controles admisibles  $\mathcal{U} = \{-1, 1\}$ .

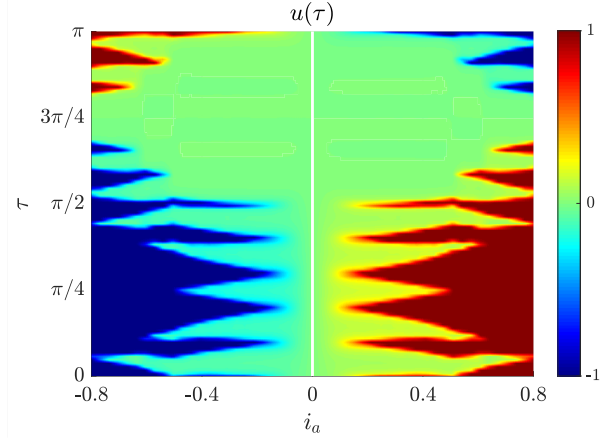


Fig. 5. Results of simulation of Simulation 5.2.

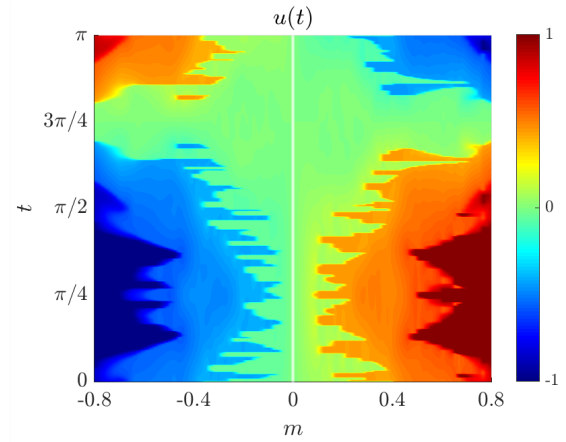


Fig. 6. Results of simulation of Simulation 5.3.

This methodology allows obtaining a  $10^{-4} - 10^{-5}$  precision in the distance to the target vector

## 6 Proofs of results in Section 4

We give here the proofs of the results presented in Section 4. We start with the proof of Proposition 4.1, which gives an upper estimate for the error in the solution to the optimal control problem with the penalization term for the control.

**Proof: (of Proposition 4.1)** Since we are supposing that Problem 4.1 has a solution, there exists a control  $\tilde{u} \in L^\infty$  such that its corresponding trajectory  $\tilde{\mathbf{x}}$ , solution to (4.5), satisfies  $\tilde{\mathbf{x}}(\pi) = 0$ .

Now, let  $u^*$  be the solution to Problem 4.3, and let  $\mathbf{x}^*$  be its corresponding trajectory. By the optimality of  $u^*$  we have

$$\frac{1}{2} \|\mathbf{x}^*(\pi)\|^2 + \varepsilon \int_0^\pi \mathcal{L}(u^*(\tau)) d\tau \leq \varepsilon \int_0^\pi \mathcal{L}(\tilde{u}(\tau)) d\tau,$$

and hence, we deduce that  $\|\mathbf{x}^*(\pi)\|^2 \leq 4\varepsilon\pi\|\mathcal{L}\|_\infty$ .  $\square$

Are we sure that here is  $4\varepsilon$  and not  $2\varepsilon$ ?

### 6.1 Proofs of Theorems 4.1 and 4.2

The proofs of Theorems 4.1 and 4.1 are based on the optimality conditions for the Optimal Control Problem 4.3, which can be deduced by means of Pontryagin's maximum principle [Bryson, 1975, Chapter 2.7].

To this end, let us first introduce the Hamiltonian function associated to the Optimal Control Problem 4.3:

$$\mathcal{H}(u, \mathbf{p}, t) = \varepsilon\mathcal{L}(u) - \frac{2}{\pi}(\mathbf{p} \cdot \mathcal{D}(t))u(t), \quad (6.1)$$

where  $\mathbf{p} \in \mathbb{R}^{N_a+N_b}$  is the so-called adjoint variable, which arises from the restriction imposed by the dynamical system (4.5). In view of the definition of  $\mathcal{D}(t)$  in (4.3)-(4.4), we will sometimes write the state and the adjoint variables using the following notation:

$$\mathbf{x}(t) = \begin{bmatrix} \mathbf{a}(t) \\ \mathbf{b}(t) \end{bmatrix} \quad \text{and} \quad \mathbf{p}(t) = \begin{bmatrix} \mathbf{p}^a(t) \\ \mathbf{p}^b(t) \end{bmatrix}.$$

Now, let us derive the optimality conditions arising from Pontryagin's principle.

1. **The adjoint system:** for any  $u^* \in L^\infty$ , solution to the Problem 4.3, there exists a unique adjoint trajectory  $\mathbf{p}^* \in C([0, \pi]; \mathbb{R}^{N_a+N_b})$  which satisfies the following terminal-value problem

$$\begin{cases} \dot{\mathbf{p}}^*(t) = -\nabla_x \mathcal{H}(u(t), \mathbf{p}^*(t), t), & t \in [0, \pi] \\ \mathbf{p}^*(\pi) = \nabla_x \Psi(\mathbf{x}^*(\pi)) \end{cases}$$

where  $\Psi$  is the terminal cost of the Optimal Control Problem 4.3. In our case, we have  $\Psi(\mathbf{x}) = \frac{1}{2}\|\mathbf{x}\|^2$ . Moreover, since the Hamiltonian does not depend on the state variable  $\mathbf{x}$ , we simply have  $\dot{\mathbf{p}}^*(t) = 0$  for all  $t \in [0, \pi]$ . We therefore deduce that the adjoint trajectory is constant, and given by

$$\mathbf{p}^*(t) = \mathbf{x}^*(\pi), \quad \forall t \in [0, \pi]. \quad (6.2)$$

2. **The Optimal Control:** now, using the optimal adjoint trajectory, we can deduce the necessary optimality condition for the control, which reads as follows:

$$u^*(t) \in \arg \min_{|u| \leq 1} \mathcal{H}(t, \mathbf{p}^*(t), u), \quad \forall t \in [0, \pi]. \quad (6.3)$$

As we will see, for penalization functions  $\mathcal{L}$  as the ones we consider in Theorems 4.1 and 4.2, this argmin is a

singleton for almost every  $t \in [0, \pi]$ . Hence, given the adjoint  $\mathbf{p}^*$ , condition (6.3) uniquely determines the control as a function in  $L^\infty$ .

In view of (6.1) and (6.2), we can write the optimality condition (6.3) as

$$\begin{aligned} u^*(t) &\in \arg \min_{|u| \leq 1} \mathcal{H}_\mu(u, \mu^*(t)) \\ &:= \arg \min_{|u| \leq 1} [\varepsilon\mathcal{L}(u) - \mu^*(t)u] \end{aligned} \quad (6.4)$$

where

$$\begin{aligned} \mu^*(t) &:= \frac{2}{\pi}(\mathbf{x}^*(\pi) \cdot \mathcal{D}(t)) \\ &= \sum_{i \in \mathcal{E}_a} a_i^*(\pi) \cos(it) + \sum_{j \in \mathcal{E}_b} b_j^*(\pi) \sin(jt). \end{aligned} \quad (6.5)$$

### 6.2 Proof of Theorem 4.1

We need to prove that, if  $\mathcal{L}$  is a concave function, then any optimal control  $u^*$  has the form (2.4) with  $\mathcal{U} = \{-1, 1\}$ . It then suffices to prove that the argmin in (6.3) is the singleton  $\{-1\}$  or  $\{1\}$  for all  $t \in [0, \pi]$  except for a finite set of  $t$ .

Since  $\mathcal{L}$  is a concave function, so is  $\mathcal{H}_\mu(u, \mu)$  as a function of  $u$ . Hence, in view of (6.4), we have

$$u^*(t) = -1 \quad \text{whenever} \quad \mathcal{H}_\mu(-1, \mu^*(t)) < \mathcal{H}_\mu(1, \mu^*(t))$$

and

$$u^*(t) = 1 \quad \text{whenever} \quad \mathcal{H}_\mu(1, \mu^*(t)) < \mathcal{H}_\mu(-1, \mu^*(t)).$$

Now, in view of the definition of  $\mathcal{H}_\mu$  in (6.4), we see that  $u^*(t)$  is uniquely determined, and belongs to  $\mathcal{U}$ , for all  $t \in [0, \pi]$  except when

$$\mu^*(t) = \frac{\varepsilon}{2} [\mathcal{L}(1) - \mathcal{L}(-1)].$$

Finally, in view of the form of  $\mu^*$  in (6.5), it is clear that the above equality can only hold a finite number of times in  $[0, \pi]$ . Indeed, the times  $t \in [0, \pi]$  such that it holds are precisely the switching angles.  $\square$

### 6.3 Proof of Theorem 4.2

In this case, we suppose that  $\mathcal{U} = \{u_i\}_{i=1}^L$  is a finite set of real numbers in  $[-1, 1]$  satisfying

$$-1 = u_1 < u_2 < \dots < u_L = 1, \quad \text{with } L > 2.$$

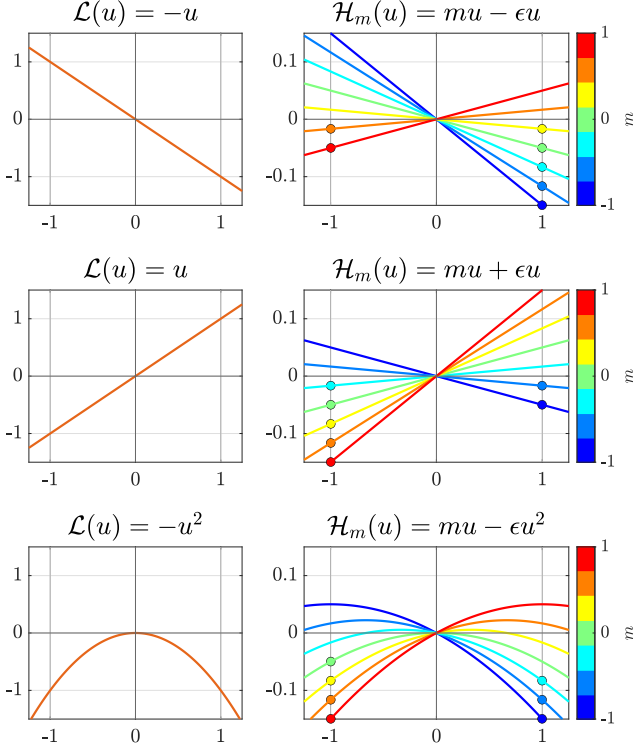


Fig. 7. Bi-level SHE: in the left column we show three types of concave penalization compatibles with Theorem 4.1. In the right columns we display the behavior of the corresponding Hamiltonian for different values of  $\mu$ .

The case  $L = 2$  is just the bi-level case. As in the previous proof, we only need to show that the argmin in (6.4) is a singleton and belongs to  $\mathcal{U}$  for every  $t \in [0, \pi]$  except for a finite number of times  $t \in [0, \pi]$ .

In this case, the study of the minimizers of  $\mathcal{H}_\mu$  is slightly more involved since the penalization function  $\mathcal{L}$  defined in (4.6)-(4.7) is not differentiable. However,  $\mathcal{L}$  is an affine interpolation of a convex function, so it is Lipschitz and convex. Hence,  $\mathcal{H}_\mu$  is also Lipschitz and convex as a function of  $u$ .

In view of this, we know that  $u$  is a minimizer of  $\mathcal{H}_\mu(u, \mu)$  if and only if

$$0 \in \partial_u \mathcal{H}_\mu(u, \mu), \quad (6.6)$$

where  $\partial_u$  denotes the subdifferential with respect to  $u$ . Let us recall the definition of subdifferential from convex analysis:

$$\begin{aligned} \partial_u \mathcal{H}_\mu(u, \mu) = \{c \in \mathbb{R} \text{ s.t.} \\ \mathcal{H}_\mu(v, \mu) - \mathcal{H}_\mu(u, \mu) \geq c(v - u) \\ \forall v \in [-1, 1]\}. \end{aligned}$$

In the case of a convex function, one may show that the subdifferential at  $u \in (-1, 1)$  is the nonempty interval

$[a, b]$ , where  $a$  and  $b$  are the one-sided derivatives

$$\begin{aligned} a &= \lim_{v \rightarrow u^-} \frac{\mathcal{H}_\mu(v, \mu) - \mathcal{H}_\mu(u, \mu)}{v - u} \\ b &= \lim_{v \rightarrow u^+} \frac{\mathcal{H}_\mu(v, \mu) - \mathcal{H}_\mu(u, \mu)}{v - u}. \end{aligned}$$

Moreover, the subdifferential at  $u = -1$  and  $u = 1$  are given by  $(-\infty, b]$  and  $[a, +\infty)$  respectively.

Using this characterization of the subdifferential, we can compute  $\partial_u \mathcal{H}_\mu(u, \mu)$  for all  $u \in [-1, 1]$  in terms of  $\mu$ . Let us define

$$p_k := \frac{\mathcal{P}(u_{k+1}) - \mathcal{P}(u_k)}{u_{k+1} - u_k}$$

for all  $k \in \{1, \dots, L-1\}$ . Now, using the definition of  $\mathcal{H}_\mu$  in (6.4) and  $\mathcal{L}$  in (4.6), we can compute

$$\begin{aligned} \partial_u \mathcal{H}_\mu(-1, \mu) &= (-\infty, \varepsilon p_1 - \mu], \\ \partial_u \mathcal{H}_\mu(1, \mu) &= [\varepsilon p_{L-1} - \mu, +\infty), \\ \partial_u \mathcal{H}_\mu(u_k, \mu) &= [\varepsilon p_{k-1} - \mu, \varepsilon p_k - \mu], \end{aligned}$$

for all  $k \in \{2, \dots, L-1\}$ , and

$$\partial_u \mathcal{H}_\mu(u, \mu) = \{\varepsilon p_k - \mu\},$$

for all  $u \in (u_k, u_{k+1})$  and all  $k \in \{1, \dots, L-1\}$ .

Now we observe that

$$\begin{aligned} 0 \in \partial_u \mathcal{H}_\mu(-1, \mu) &\quad \text{iff} \quad \mu \leq \varepsilon p_1, \\ 0 \in \partial_u \mathcal{H}_\mu(1, \mu) &\quad \text{iff} \quad \mu \geq \varepsilon p_{L-1}, \\ 0 \in \partial_u \mathcal{H}_\mu(u_k, \mu) &\quad \text{iff} \quad \varepsilon p_{k-1} \leq \mu \leq \varepsilon p_k, \end{aligned} \quad (6.7)$$

for all  $k \in \{2, \dots, L-1\}$ , and, for all  $k \in \{1, \dots, L-1\}$ ,

$$0 \in \partial_u \mathcal{H}_\mu(u, \mu) \quad \text{for all } u \in [u_k, u_{k+1}]$$

if and only if  $\mu = \varepsilon p_k$ .

Using this, along with the optimality condition (6.6), we deduce that, for almost every  $\mu \in \mathbb{R}$ , we have

$$\arg \min_{u \leq 1} \mathcal{H}_\mu(u, \mu) = \{u_k\}$$

for some  $u_k \in \mathcal{U}$ . Indeed, (6.6) does not hold if and only if  $\mu = \varepsilon p_k$ , for some  $k \in \{1, \dots, L-1\}$ .

Hence, using the optimality condition for  $u^*(t)$  in (6.4), we see that  $u^*(t)$  is uniquely determined, and belongs to

$\mathcal{U}$ , for all  $t \in [0, \pi]$  except when

$$\mu^*(t) = \varepsilon \frac{\mathcal{P}(u_{k+1}) - \mathcal{P}(u_k)}{u_{k+1} - u_k}$$

for some  $k \in \{1, \dots, L-1\}$ . In view of the form of  $\mu^*(t)$  in (6.5), it is clear that the above equality can only hold a finite number of times in the interval  $[0, \pi]$ . Indeed, we refer to the set of times when the above equality holds as the switching angles. Due to the continuity of  $\mu^*(t)$ , along with (6.7), it is clear that  $u^*(t)$  does not change value between two consecutive switching angles. The staircase property of the waveform (2.5) can also be deduced from (6.4) and (6.7), along with the continuity of the function  $\mu^*(t)$ .  $\square$

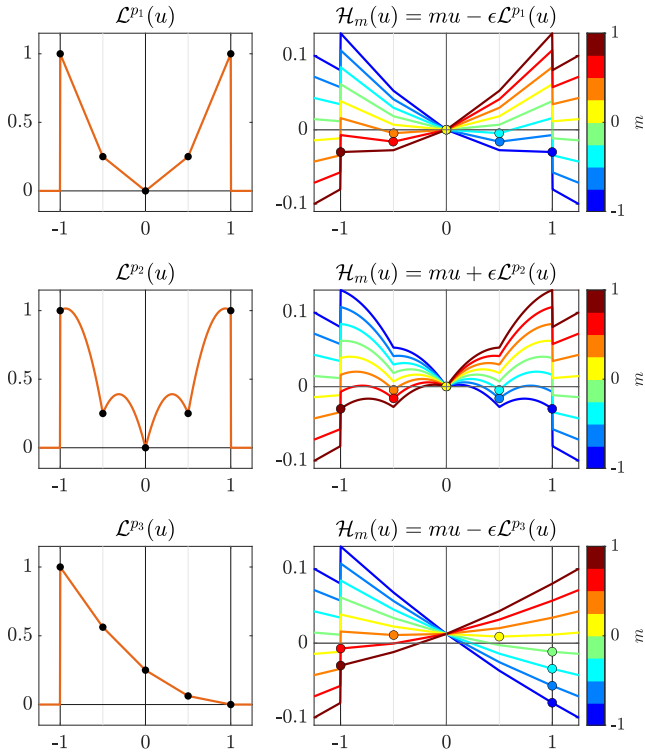


Fig. 8. Multilevel SHM: in the left column we show three types of penalizations which are compatible with our theoretical results. In the right column we show the behavior of the corresponding Hamiltonian for different values of  $\mu$ .

## 7 Conclusions

We presented the SHE problem from the point of view of control theory. Nevertheless, comparing with methodologies where the commutation number is fixed a priori, our approximation is computationally more expensive. Notwithstanding that, the optimal control provides solutions in the entire range of the modulation index, although the number of solutions or their locations change dramatically.

This methodology for the SHE problem connects control theory with harmonic elimination. In this way, the SHE problem can be solved through classical tools.

### 7.1 Quarter wave symmetry

We shall mention that, in the SHE literature [Wu, 2009], it is usual to distinguish among the half-wave symmetry problem (addressed in the present paper) and the quarter-wave symmetry one in which

$$u\left(\tau + \frac{\pi}{2}\right) = -u(\tau) \quad \text{for all } \tau \in \left[0, \frac{\pi}{2}\right).$$

In quarter-wave symmetry, the SHE problem simplifies, as the Fourier coefficients  $\{a_i\}_{i \in \mathcal{E}_a}$  turn out to be all zero. Hence, only the phases of the converter's signal can be controlled, while the half-wave SHE allows to deal with the amplitudes as well. It is worth to remark nonetheless that our optimal control formulation can be easily adapted to the quarter-wave symmetry problem by simply replacing the Fourier coefficients (2.3) with

$$a_i = 0, \quad b_j = \frac{4}{\pi} \int_0^{\frac{\pi}{4}} u(\tau) \sin(j\tau) d\tau.$$

Entonces podemos introducir el siguiente sistema dinámico:

$$\begin{cases} \dot{\beta}(\tau) = \frac{2}{\pi} \mathcal{D}^\beta(\tau) u(\tau), & \tau \in [0, \pi/2) \\ \beta(0) = \mathbf{b}_T \end{cases} \quad (7.1)$$

Además del siguiente problema de control:

**Problem 7.1** Let  $\mathcal{U}$  be defined as in (2.1) and let  $\mathcal{E}_b$  be a set of odd numbers with cardinality  $|\mathcal{E}_b| = N_b$ . Given the vector  $\mathbf{b}_T \in \mathbb{R}^{N_b}$ . We look for  $u : [0, \pi/2) \rightarrow \mathcal{U}$  such that the solution of (7.1) with initial datum  $\mathbf{x}(0) = \mathbf{x}_0$  satisfies  $\mathbf{x}(\pi) = 0$ .

Donde de la misma forma que en el problema con simetría de media onda la solución de este problema es también solución del problema SHE con simetría de cuarto de onda.

### 7.2 Generalizations

Consideramos que el tipo de penalizaciones utilizados en este problema pueden ser extendido a sistemas LTI:

$$\min_{u \in \mathcal{U}} \frac{1}{2} \|\mathbf{x}(T)\|^2 + \int_0^T \mathcal{L}(u(\tau)) d\tau \quad (7.2)$$

$$\text{subject to: } \begin{cases} \dot{\mathbf{x}}(\tau) = A(\tau)\mathbf{x}(\tau) + B(\tau)u(\tau), & \tau \in [0, T) \\ \mathbf{x}(0) = \mathbf{x}_0 \end{cases}$$

De manera que dado un trmino de penalizacin compatible con el Teorema ??, sea capaz de condicionar el control ptimo como un control digital.

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## References

- [Andersson et al., 2019] Andersson, J. A. E., Gillis, J., Horn, G., Rawlings, J. B., and Diehl, M. (2019). CasADi – A software framework for nonlinear optimization and optimal control. *Math. Program. Comput.*, 11(1):1–36.
- [Bryson, 1975] Bryson, A. E. (1975). *Applied optimal control: optimization, estimation and control*. CRC Press.
- [Duranay and Guldemir, 2017] Duranay, Z. B. and Guldemir, H. (2017). Selective harmonic eliminated v/f speed control of single-phase induction motor. *IET Power Electron.*, 11(3):477–483.
- [Janabi et al., 2020] Janabi, A., Wang, B., and Czarkowski, D. (2020). Generalized chudnovsky algorithm for real-time pwm selective harmonic elimination/modulation: two-level vsi example. *IEEE Trans. Power Electron.*, 35(5):5437–5446.
- [Konstantinou and Agelidis, 2010] Konstantinou, G. S. and Agelidis, V. G. (2010). Bipolar switching waveform: novel solution sets to the selective harmonic elimination problem. *Proc. IEEE International Conference on Industrial Technology*, pages 696–701.
- [Oroya et al., 2021] Oroya, J., Esteve, C., and Biccari, U. (2021). djoroya/she-optimal-control-paper, github repository. <https://github.com/djoroya/SHE-Optimal-Control-paper>. Accessed: 2021-02-22.
- [Rodríguez et al., 2002] Rodríguez, J., Lai, J. S., and Peng, F. Z. (2002). Multilevel inverters: a survey of topologies, controls, and applications. *IEEE Trans. Ind. Electron.*, 49(4):724–738.
- [Schättler and Ledzewicz, 2012] Schättler, H. and Ledzewicz, U. (2012). *Geometric optimal control: theory, methods and examples*, volume 38. Springer Science & Business Media.
- [Sun et al., 1996] Sun, J., Beineke, S., and Grotstollen, H. (1996). Optimal pwm based on real-time solution of harmonic elimination equations. *IEEE Trans. Power Electron.*, 11(4):612–621.
- [Wang and Topputo, 2020] Wang, Y. and Topputo, F. (2020). Robust Bang-Off-Bang Low-Thrust Guidance Using Model Predictive Static Programming. *Acta Astronautica*, 176(May):357–370.
- [Wu, 2009] Wu, B. (2009). A Generalized Formulation of Quarter-Wave Symmetry SHE-PWM Problems for Multilevel Inverters. *IEEE Transactions on Power Electronics*, 24(7):1758–1766.
- [Yang et al., 2015] Yang, K., Yuan, Z., Yuan, R., Yu, W., Yuan, J., and Wang, J. (2015). A groebner bases theory-based method for selective harmonic elimination. *IEEE Trans. Power Electron.*, 30(12):6581–6592.
- [Yang et al., 2017] Yang, K., Zhang, Q., Zhang, J., Yuan, R., Guan, Q., Yu, W., and Wang, J. (2017). Unified selective harmonic elimination for multilevel converters. *IEEE Trans. Power Electron.*, 32(2):1579–1590.
- [Zuazua, 2011] Zuazua, E. (2011). Switching control. *Journal of the European Mathematical Society*, 13(1):85–117.