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ARTICLE TYPE

Optimal control of a solar sail

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Summary

Optimal control of a solar sail orbiting around a fixed center of mass is considered. The sail is modelled as a plane surface with two sides having similar optical properties. The control is assumed to be the attitude of the sail and is so represented as an element of the projective plane. Mapping this plane to compute the actual force generated by a given attitude results in a generally non-convex set of admissible control values. A suitable convex relaxation is introduced to study the optimality system associated with maximising the change of the sail orbital parameters in a given direction along one orbit. Building on previous works^{1,2,3}, a refined analysis of the control structure is given. In order to compute effective solutions, a multiple shooting approach is retained that is well suited for systems with a sequence of switchings between various control subarcs. Three issues are addressed: initialisation of shooting by means a tailored SDP relaxation that takes advantage of good convergence properties of modern convex optimisation algorithms; changes in the control structure that are accommodated by coupling shooting with differential continuation; implicit character of the Hamiltonian maximisation when using Pontrjagin maximum principle that is taken care of by incorporating the associated equation for the dynamical feedback into the shooting procedure. The method is illustrated on an example from JPL for which the sail inclination is optimally changed over one orbital period.

KEYWORDS:

solar sailing, non-convex constraints, SDP relaxation, multiple shooting, differential continuation

1 | INTRODUCTION

1.1 | Solar radiation pressure

Solar sails are satellites that use solar radiation pressure (SRP) as propulsion for orbital maneuvers. SRP is caused by the interaction between photons and the surface of the sail, and its magnitude depends on the distance between the Sun and the sail, r . Specifically, denote by $\Phi_{SR} = 1367 \text{ W m}^{-2}$ the solar flux at $r_{\oplus} = 1 \text{ AU}$ distance (*i.e.*, the average Sun-Earth distance), and by c the speed of light; then, a simple model for the SRP is given by^{4, Chap. 3}:

$$P_{SR} = \frac{\Phi_{SR}}{c} \left(\frac{r_{\oplus}}{r} \right)^2. \quad (1)$$

In this paper, similarly to⁵, we consider a flat sail with surface A and mass m . Different optical and geometrical properties have impact on the resulting SRP force, which has components of the incoming, reflected, and thermal radiations, namely f_a , f_r , and

f_e . Moreover, the reflected radiation has specular and diffuse contributions, f_{rs} and f_{ru} , respectively. Each force component has different magnitude and direction, that can be identified through the Sun-sail direction, denoted as \hat{s} and the unit vector normal to the sail having a positive component along \hat{s} , \hat{n} . In this paper we assume that both sides of the sail have the same optical properties, so only the (non-oriented) direction of the normal to the plane representing the plane will describe its attitude. We also assume that it is possible to control the attitude, and that the actual control will be the force generated by this attitude (see (6)). In this model, \hat{n} belongs the projective plane \mathbf{RP}^2 that one can describe as the union of one open hemisphere (whose axis is \hat{s}) with a circle whose antipodal points are identified (subset of directions perpendicular to \hat{s}). Fixing some basis \hat{e}_1, \hat{e}_2 of $\{\hat{s}\}^\perp$ in order that $(\hat{s}, \hat{e}_1, \hat{e}_2)$ is a direct orthogonal frame, one defines coordinates $(\beta, \delta) \in (-\pi/2, \pi/2) \times \mathbf{R}$ for \hat{n} in the open hemisphere part setting as usual (see Figure 1b)

$$\hat{n} = \sin \beta (\cos \delta \hat{e}_1 + \sin \delta \hat{e}_2) + \cos \beta \hat{s}.$$

This is not a chart¹ as no δ , even restricted to $\mathbf{R}/\pi\mathbf{Z}$, can be uniquely associated with the direction \hat{s} . (See also Remark 1.) The angle β is the so-called solar-sail *cone angle*. As shown in Fig. 1a, let us introduce the direction of specular reflection given by $\hat{\xi}$, and the unit tangent vector \hat{t} lying in the plane generated by \hat{s} and \hat{n} . The force due to the incoming radiation, f_a , is directed along \hat{s} . The force provided by the specularly reflected radiation, f_{rs} , points along $\hat{\xi}$ and is caused by photons that are reflected symmetrically with respect to the normal of the sail, thus yielding an exchange of momentum. Diffuse reflection stems from the sail surface roughness, which causes photons to be uniformly reflected in all directions, yielding a component of the force toward the direction normal to the sail, \hat{n} . Finally, as the absorbed photons are re-radiated in all directions, the force f_e is generated, which is orthogonal to the sail surface and points again along \hat{n} . We follow^{1, Chap. 2} and express the unit vectors \hat{s} and $\hat{\xi}$ in terms of \hat{n} and \hat{t} ,

$$\hat{s} = \cos \beta \hat{n} + \sin \beta \hat{t}, \quad \hat{\xi} = \cos \beta \hat{n} - \sin \beta \hat{t}, \quad (2)$$

so the above-presented forces can be expressed as⁶:

$$\begin{aligned} f_a &= \epsilon \cos \beta \hat{s} = \epsilon \cos \beta (\cos \beta \hat{n} + \sin \beta \hat{t}), \\ f_{rs} &= \epsilon \rho s \cos \beta \hat{\xi} = \epsilon \rho s \cos \beta (\cos \beta \hat{n} - \sin \beta \hat{t}), \\ f_{ru} &= \epsilon B_f \rho (1 - s) \cos \beta \hat{n}, \\ f_e &= \epsilon (1 - \rho) \frac{\epsilon_f B_f - \epsilon_b B_b}{\epsilon_b + \epsilon_f} \cos \beta \hat{n}. \end{aligned} \quad (3)$$

In (3), ϵ is equal to $AP_{SR} m^{-1}$, which combines optical and physical parameters of the sail, has small magnitude, $\rho \in [0, 1]$ is the fraction of reflected radiation to total amount of radiation illuminating the sail, $s \in [0, 1]$ is the fraction of specularly reflected radiation to total reflected radiation, ϵ_b and ϵ_f are the back and front surface emissivity coefficients, respectively, and B_b and B_f are back and front non-Lambertian coefficients, respectively. The SRP force is found as:

$$f_{SRP} = f_a + f_{rs} + f_{ru} + f_e. \quad (4)$$

Let us use the notations used in²:

$$b_1 = 1 - \rho s, \quad b_2 = 2\rho s, \quad b_3 = B_f \rho (1 - s) + (1 - \rho) \frac{\epsilon_f B_f - \epsilon_b B_b}{\epsilon_f + \epsilon_b}.$$

With these, in S the SRP force reads (note that because of the radial symmetry, its norm is independent of δ)

$$f_{SRP} = \epsilon \cos \beta \begin{pmatrix} (b_1 + b_2) \cos^2 \beta + b_3 \cos \beta + b_1 \sin^2 \beta \\ b_2 \sin \beta \cos \beta \sin \delta + b_3 \sin \beta \sin \delta \\ b_2 \sin \beta \cos \beta \cos \delta + b_3 \sin \beta \cos \delta \end{pmatrix}. \quad (5)$$

Remark 1. In our modeling, the magnitude of the SRP is continuous with respect to \hat{n} , going to zero when the Sun direction is contained in the sail plane (orthogonality of \hat{s} and \hat{n}), but its direction is not: when going through $\hat{s} \perp \hat{n}$, the illuminated side of the sail (a thickless 2D object embedded into 3D space) is changed and the orientation of \hat{n} is changed to opposite ($\beta = \pm\pi/2$ being changed to $-\beta$, still defining the same direction—a perpendicular to \hat{s} —in the projective plane). The resulting force, going to zero in such cases, is continuous but not smooth. This singularity is inherent to the modeling and would be removed in a more realistic approach describing the sail as a genuine 3D object. This lack of smoothness is nonetheless not crucial here since, as will be clear from the optimality analysis in Section 2, an optimal force will have discontinuities, being either zero or with

¹One actually retrieves the universal cover of the pointed open hemisphere by restricting to (β, δ) in $(0, \pi/2) \times \mathbf{R}$.

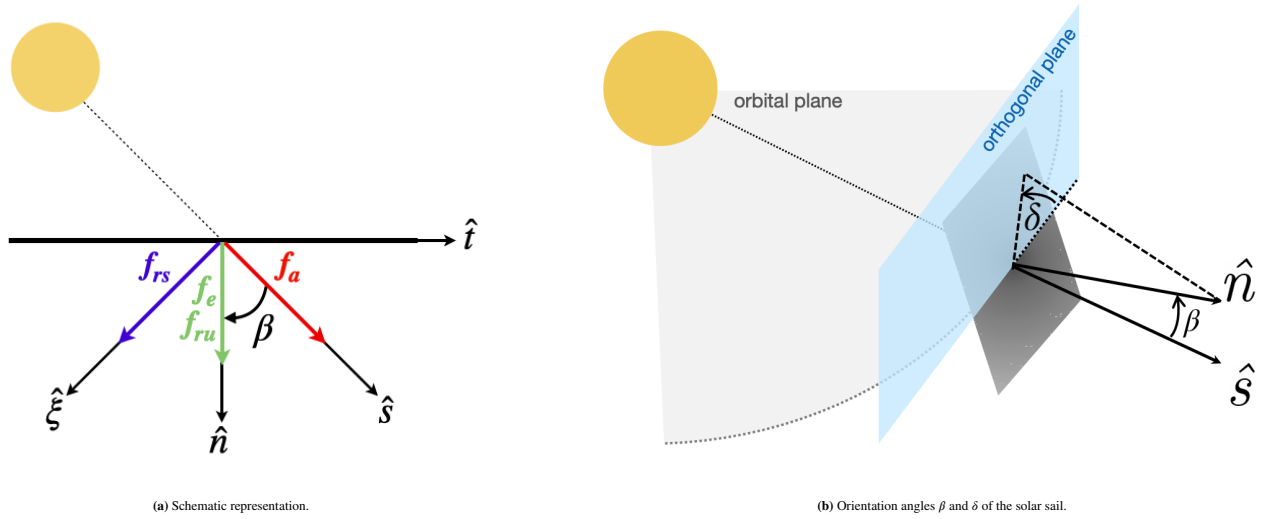


Figure 1 Components of the SRP force and orientation of th.

$\beta \in (-\beta^*, \beta^*)$ and $0 < \beta^* < \pi/2$ (if we exclude the ideal case for which $\beta^* = \pi/2$). So flips of illuminated side will not be encountered.

1.2 | Parametrisation of the control set

Controlling the sail attitude, *i.e.* the normal vector \hat{n} , allows to change the direction and magnitude of the resulting SRP. A reliable inference of optical coefficients is indeed mandatory to accurately estimate the mapping f_{SRP} . To carry out our analysis, solar sail dynamics is conveniently modeled as a nonlinear control-affine system (see Section 1.3), where the control variable is homogeneous to the renormalized force. The control set $U \subset \mathbf{R}^3$ is so given by:

$$U = \left\{ u = \frac{f_{SRP}(\hat{n})}{\epsilon} \in \mathbf{R}^3, \hat{n} \in \mathbf{RP}^2 \right\}. \quad (6)$$

It is a closed surface of revolution with axis \hat{s} that bounds a domain of \mathbf{R}^3 . Up to the scaling factor ϵ , this surface is the image of the projective plane by the mapping f_{SRP} . As a result, the surface minus the origin is parametrised by (β, δ) in $(-\pi/2, \pi/2) \times \mathbf{R}$, while the origin corresponds to the image of the circle of directions orthogonal to \hat{s} . Figure 2 shows the intersection of U with the plane generated by \hat{n} and \hat{s} for various optical properties. The resulting curve now bounds a two-dimensional domain, non-convex unless the sail is ideal ($\rho = s = 1$, so the sail is perfectly reflective, a first extreme case). In addition to ideal sails, another extreme case can be identified: contrary to perfectly reflective sails, perfectly absorptive surfaces are the worst-case scenario ($\rho = 0$, f_e neglected) because SRP is systematically parallel to \hat{s} , as shown in red in Figure 2a. Although sails are designed to be as close to ideal as possible, partial absorption of the energy is unavoidable in real-life applications and optical properties exhibit degradation in time. Hence, the fraction of reflected radiation decreases with the lifetime of the satellite as discussed in^{7,8}. Another significant optical coefficient that plays a crucial role in determining the control set is the specular reflectivity, denoted by s , as illustrated in Figure 1a. The red set represents a sail with a rough surface (corresponding to $s = 0$), where light is diffusely reflected in all directions. This results in a smaller control set, even for a high reflectivity value of $\rho = 0.8$. In Figures 2c and 2d, the impact of other optical coefficients, namely emissivity and non-Lambertian coefficients, is shown. Although these coefficients have a small but noticeable effect on the shape and convexity of the control set. (The ideal solar sail is the only exception, exhibiting a vertical tangency at its vertex.) While most control sets have a horizontal tangency at their vertices, a more comprehensive analysis is required to fully understand the dependence of the control set geometry on different optical coefficients of the sail surface.

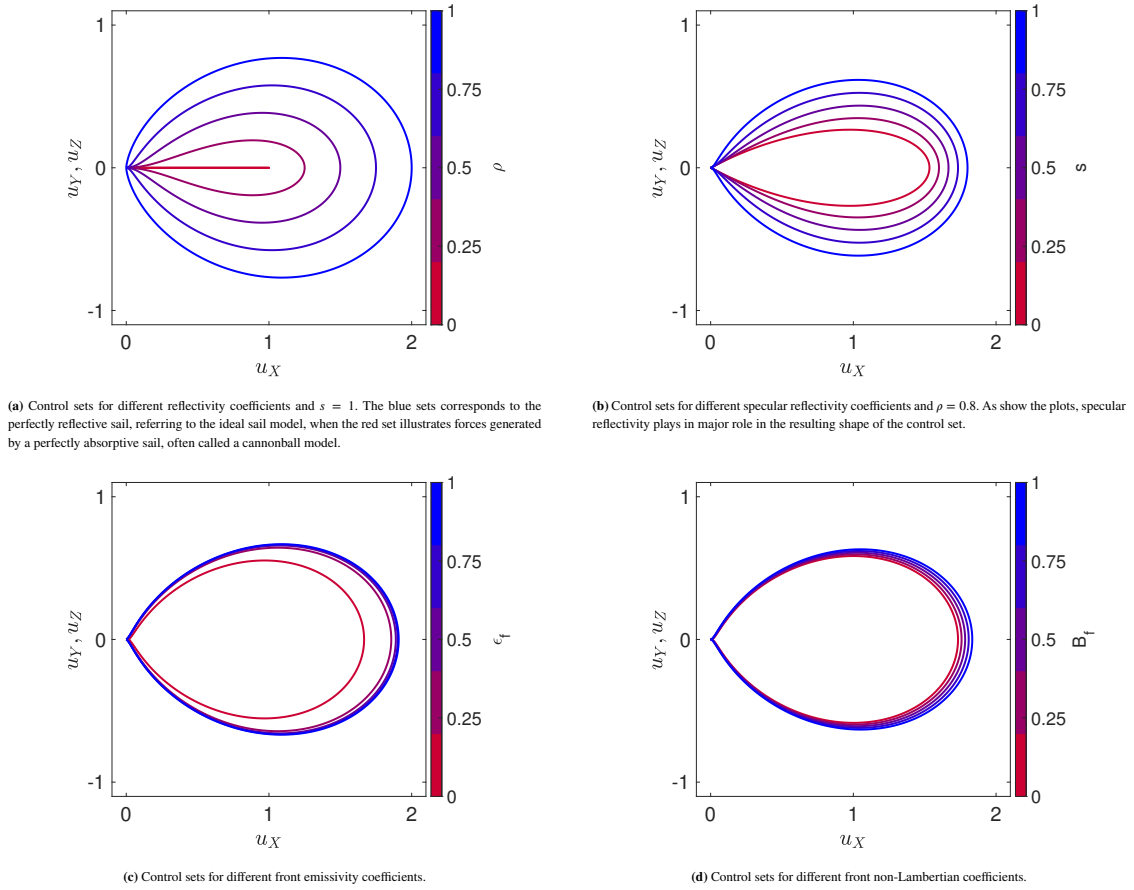


Figure 2 Deformation of the control set with respect to the various optical properties of the sail.

1.3 | Equations of motion

The following assumptions are introduced:

1. Orbital period of the sail is much smaller than the one of the heliocentric orbit of the attractor, so that variations of the Sun direction \hat{s} over a single orbit of the sail are neglected.
2. Solar eclipses are neglected. Introducing solar eclipses can be a way of improving the proposed algorithm.
3. Re-emitted radiation is neglected. In fact, this component of SRP can be reasonably regarded as a disturbance for control purposes.

Equations of motion are written in a set of Keplerian-like orbital elements, which leverages on the axial symmetry of the problem with respect to the Sun's direction. Namely, consider a reference frame \mathcal{S} with origin at the center of the planet, \hat{X} axis towards \hat{s} , \hat{Y} lies in the plane of the planet's orbit around the Sun and is orthogonal to \hat{X} , and \hat{Z} is chosen to form a right-hand frame. Because this study focuses on short-time controllability (characteristic time is of the order of one orbital period), motion of this frame is neglected by virtue of the first assumption above. Figure 3 represents the vectors \mathbf{h} , \mathbf{e} and \hat{N} , which denote the angular momentum, eccentricity and ascending node vectors, respectively. Let $\gamma_1, \gamma_2, \gamma_3$ be Euler angles orienting the eccentricity vector according to a X - Y - X rotation as depicted in Fig. 3, so that γ_2 is the angle between the angular momentum of the orbit and the Sun direction, and a , e , and f be semi-major axis, eccentricity and true anomaly, respectively. The motion of slow elements, $I = (\gamma_1, \gamma_2, \gamma_3, a, e) \in \mathcal{M}$, where \mathcal{M} is the configuration manifold, and fast angle f is governed by

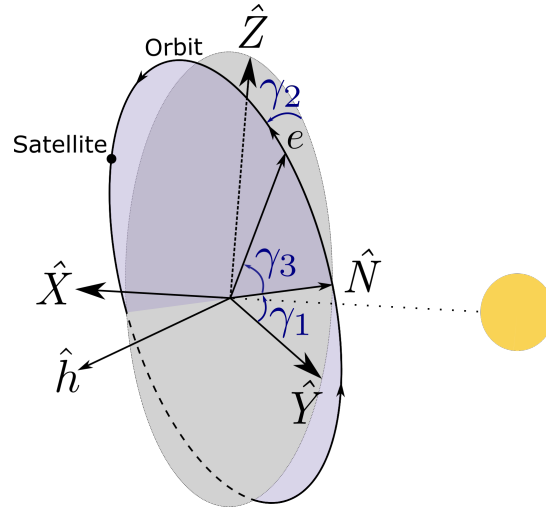


Figure 3 Euler angles γ_i orienting the orbit according to a \hat{X} - \hat{Y} - \hat{X} rotation with respect to the reference frame S . Here, h and e denote the angular momentum and eccentricity vectors.

$$\begin{aligned} \frac{dI}{dt} &= \varepsilon \sqrt{\frac{a(1-e^2)}{\mu}} G_0(I, f) R(I, f) u, \\ \frac{df}{dt} &= \omega(I, f) + \varepsilon F(I, f) R(I, f) u, \end{aligned} \quad (7)$$

where components of u are in the reference frame S , $R(I, f) = R_X(\gamma_3 + f)R_Y(\gamma_2)R_X(\gamma_1)$ is the rotation matrix from reference to local-vertical local-horizontal frames²,

$$\omega(I, f) = \sqrt{\frac{\mu}{a(1-e^2)^3}} (1 + e \cos f)^2. \quad (8)$$

Both $F(I, f)$ and $G_0(I, f)$ can be deduced from Gauss variational equations (GVE) of classical elements, where

$$G_0(I, f) = \begin{pmatrix} 0 & 0 & \frac{\sin(\gamma_3 + f)}{\sin \gamma_2 (1 + e \cos f)} \\ 0 & 0 & \frac{\cos(\gamma_3 + f)}{1 + e \cos f} \\ -\frac{\cos f}{e} & \frac{2 + e \cos f}{1 + e \cos f} \frac{\sin f}{e} & \frac{\cos(\gamma_3 + f)}{1 + e \cos f} \\ \frac{2ae}{1-e^2} \sin f & \frac{2ae}{1-e^2} (1 + e \cos f) & 0 \\ \sin f & \frac{e \cos^2 f + 2 \cos f + e}{1 + e \cos f} & 0 \end{pmatrix}. \quad (9)$$

In Section 2 we introduce the optimal control problem of interest in this paper: it consists in maximising the displacement of the sail in a given direction over one orbit. A characteristic feature of this problem is the non-convexity of its control set; we define a suitable convex relaxation and review existence and necessary conditions of the original problem, giving precise bounds of the number of switchings of the control. A tailored method of resolution that combines convex optimisation and shooting is presented in Section 3. While a bounded polyhedral cone allows to approximate the set of admissible controls, a semi-infinite formulation of the control problem is proposed. Using a sum-of-squares approach *à la Nesterov*, this problem is recast as a semi-definite program on the cone of SDP matrices. For such a program, there are efficient convex optimisation algorithms with guaranteed convergence properties. The resulting solution is expected to provide an initial guess for shooting. In particular, we rely on the SDP step to capture the switching structure of the optimal control on a bounded cone. Then the solution on this cone is connected to the solution on the original non convex control set by means of differential continuation, coupled with multiple

²Here, $R_A(f)$ denotes the rotation matrix of angle f about the axis \hat{A} .

shooting. As the structure of the control may change during continuation (appearance or disappearance of control subarcs), a callback procedure is used to monitor the process and restart it with an updated shooting function. An additional difficulty comes from the fact that there is no explicit expression of the control that maximises the Hamiltonian coming from Pontrjagin maximum principle applied to the problem. This issue is accommodated by incorporating the equation implicitly defining the control into the shooting procedure. The last section is devoted to the numerical treatment by the described approach of an example coming from the Jet Propulsion Lab: a non-ideal solar sail is considered, and the target is to increase the inclination of the initial orbit. Convex optimisation turns to provide a very precise solution to initialise the continuation. A change of structure is nonetheless observed during the early steps of homotopy as the number of arcs drops from three 5 to 3, and the callback is used to eventually obtain the zero-bang-zero optimal control of the sail.

2 | CONTROL OVER ONE ORBITAL PERIOD

2.1 | Optimal control formulation

We are interested in moving solar sail in the desired direction after one orbital period. Therefore, it is interesting to rewrite System (7) in terms of displacement of the slow state elements, denoted δI . Therefore, its dynamics is given by

$$\delta I' = \varepsilon G(I, f)u \quad (10)$$

where $' := d/df$ and with

$$G(I, f) := \frac{a(1 - e^2)^2}{\mu(1 + e \cos f)^3} G_0(I, f) R(I, f).$$

As mentioned earlier, SRP has a very small magnitude, this is why it is usually considered as a perturbation. Thus, changes on slow variables are very small over one orbital period so that I will be assumed *constant* in the rest of the paper. For the same reason, we also neglect the order one term in ε in the dynamics of the fast variable in (7). The goal is to maximise the size of the displacement in a given direction fixed by a unit vector, d_I , so that the final value of δI is parallel to d_I . This problem can be written in Mayer form as follows (note the simple form of the dynamics, given by an explicit integral, as the right-hand side does not depend on δI in our approximation):

$$\max_{u(f) \in U} (\delta I(2\pi) | d_I) \quad \text{subject to} \quad \delta I' = \varepsilon \sum_{i=1}^3 u_i G_i(I, f), \quad \delta I(0) = 0, \quad \delta I(2\pi) \text{ parallel to } d_I. \quad (11)$$

Building upon results in^{9,10}, we have access to an effective test (related to the convex SDP approximation discussed in Section 3) to check that it is indeed possible to move in the direction d_I after one revolution. So we assume in the sequel that the problem is controllable.

2.2 | Existence and necessary conditions for optimality

We first consider the relaxation of (11) obtained by replacing the control set U by its convex hull: $u(f) \in \text{conv}(U)$ (see Figure 4). As the control set is now compact and convex, and since we have assumed controllability using controls valued in $U \subset \text{conv}(U)$, Filippov theorem entails that

Proposition 1. The relaxed problem has a solution.

Let K_α be the convex cone generated both by U and by its convex hull, α denoting the half-angle at the cone vertex. To formulate the necessary optimality conditions for the problem on $\text{conv}(U)$ we introduce the costate $p_{\delta I}$ of δI , a covector of dimension 5. The Hamiltonian associated with the dynamics is

$$H(I, f, p_{\delta I}, u) = \varepsilon p_{\delta I} G(I, f)u. \quad (12)$$

Remember that I is a constant, and note that the Hamiltonian does not depend on the state δI because of the very simple form of the dynamics. (The ODE defines a mere quadrature, here.) Clearly, $p_{\delta I}$ is constant and transversality conditions write

$$(p_{\delta I} | d_I) = -p^0 \|d_I\|^2 = -p^0 \quad (13)$$

where p^0 is the nonpositive multiplier associated with the cost. In particular, $p_{\delta I}$ is not zero, since otherwise both p^0 and $p_{\delta I}$ would vanish. By homogeneity in $(p^0, p_{\delta I})$ there are two cases: (i) the abnormal case ($p^0 = 0$) when $(p_{\delta I} | d_I) = 0$ and where

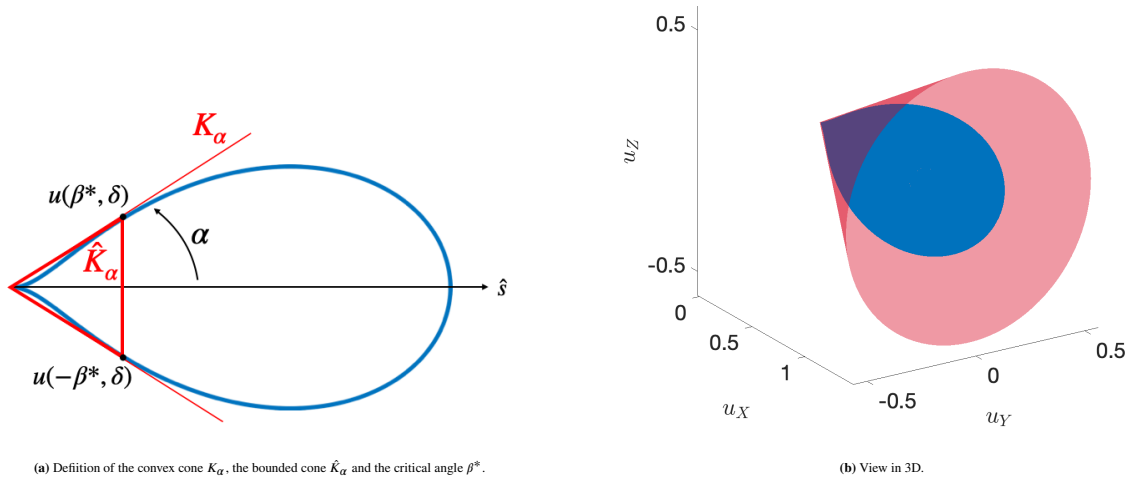


Figure 4 Convexification of the control set (in blue) by a convex cone (in red).

one can normalise setting $\|p_{\delta I}\| = 1$; (ii) the normal case ($p^0 < 0$) when $(p_{\delta I}|d_I) > 0$ and where one can normalise setting $(p_{\delta I}|d_I) = 1$. Let us set $\psi := p_{\delta I}G(I, f)$.

Lemma 1. For any I , the matrix formed by $G(I, f)$ and $\partial G(I, f)/\partial f$ has maximum rank for all $f \in [0, 2\pi]$.

Proof. This computation is actually equivalent to the rank condition that can be verified in terms of Lie brackets (and, e.g., Cartesian coordinates) in ¹¹ (check Lemma 1). \square

As a result, the zeros of the dimension three covector ψ (as a function of the true anomaly f) are isolated on $[0, 2\pi]$. Indeed, the previous lemma implies that ψ and $d\psi/df$ cannot vanish simultaneously as then, $p_{\delta I}$ would be orthogonal to all columns of $G(I, f)$ and of its derivative, so $p_{\delta I}$ would be zero (a contradiction). So there are only finitely many such zeros on $[0, 2\pi]$, defining a locus of codimension greater than one in the (I, f) space. For the sake of simplicity, we assume in the sequel that ψ actually never vanishes. For a detailed discussion on the associated singularities of the dynamics, see ¹².

The polar cone K_α^0 is the set of directions having a nonpositive scalar product with those in K_α . The drop-shaped curve obtained when intersecting the control set with a plane is parametrised by the angle β alone, and we denote $\beta^* \in (0, \pi/2)$ the parameter associated with the tangency point of this curve with its conical hull (see Figure 4a). Thus, angle α of the convex cone can be retrieved from the critical angle β^* : In the sequel, we recall and complete the analysis from ², providing precise bounds on the number of switchings on the control.

Lemma 2. ² The critical angle β^* is solution of

$$\cos \beta^* = \frac{-b_1 b_3 - 2b_2 b_3 + \sqrt{b_1^2 b_3^2 - 4b_1 b_2 b_3^2 + 8b_1^2 b_2^2 + 4b_1 b_2^3}}{4b_1 b_2 + 2b_2^2},$$

and is related to the cone angle α by the following relation:

$$\tan \alpha = \frac{(b_3 + b_2 \cos \beta^*) \sin \beta^*}{b_1 + b_2 \cos^2 \beta^* + b_3 \cos \beta^*}.$$

Proposition 2. An optimal control u verifies the following: (i) when ψ belongs to the interior of K_α^0 , u is zero; (ii) when ψ does not belong to K_α^0 , the coordinates (β, δ) of the control verify the following relations:

$$\psi_1 \sin \beta (b_1 + 3b_2 \cos^2 \beta + 2b_3 \cos \beta) = \sqrt{\psi_2^2 + \psi_3^2} (\cos^2 \beta (b_2 \cos \beta + b_3) - \sin^2 \beta (2b_2 \cos \beta + b_3)), \quad \beta \in (-\beta^*, \beta^*), \quad (14)$$

and

$$\delta = \pi/2 - \arg(\psi_2 + i\psi_3) \bmod \pi. \quad (15)$$

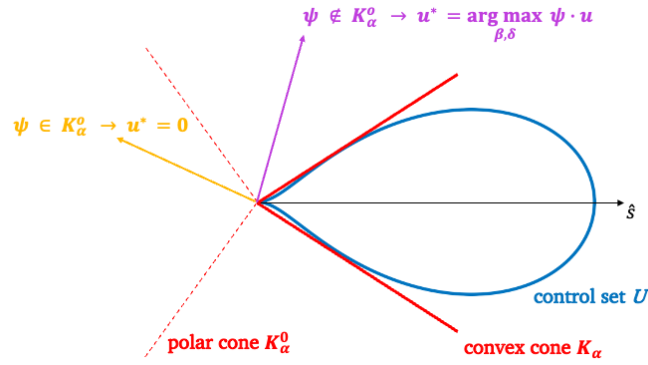


Figure 5 Geometrical interpretation of the PMP.

Moreover, any optimal control is made of finitely many subarcs corresponding to case (i) or (ii), and has at most 8 switchings (transverse contacts with ∂K_α^0) over one period.

Proof. According to Pontrjagin's maximum principle (PMP) and to the expression (12) of the Hamiltonian, for almost all true anomaly f an optimal control must be a maximiser of the scalar product $(\psi|u)$ for u in $\text{conv}(U)$. Clearly, when ψ belongs to the interior of the polar cone of K_α , this scalar product is negative for any nonzero u , so $u = 0$ is the only maximiser. Conversely, when ψ belong to the open complement of K_α^0 , maximisers must annihilate the gradient of the Hamiltonian with respect to the chosen coordinates of the control,

$$\frac{\partial H}{\partial \beta} = 0, \quad \frac{\partial H}{\partial \delta} = 0,$$

which gives the expressions in alternative (ii) of the statement. See Figure 5 for the geometrical interpretation. Moreover, ψ belongs to the boundary of K_α^0 if and only if $\psi_1 \cos \alpha + \sqrt{\psi_2^2 + \psi_3^2} \sin \alpha = 0$, implying that

$$\psi_1^2 \cos^2 \alpha - (\psi_2^2 + \psi_3^2) \sin^2 \alpha = 0. \quad (16)$$

Every component of ψ is trigonometric in f , and this (nontrivial) equation results in a trigonometric polynomial of degree 4. As it has isolated zeros, there are finitely many zeros (at most eight, see Remark 3) defining isolated contacts with ∂K_α^0 . \square

Remark 2. Equation (14) defines the optimal solution β as an implicit function of ψ and optical parameters of the sail, notably, b_1, b_2, b_3 . This relation becomes explicit when considering an ideal solar sail model, what is widely used in the litterature for different preliminarly analysis design. In this case, $b_1 = b_3 = 0$, and $b_2 = 2$. Therefore, Equation (14) becomes:

$$6\psi_1 \sin \beta \cos^2 \beta = \sqrt{\psi_2^2 + \psi_3^2} (2 \cos^2 \beta \cos \beta - 4 \sin^2 \beta \cos \beta), \quad \beta \in (-\beta^*, \beta^*), \quad (17)$$

leading to the classical well-known near-optimal steering law for an ideal sail¹:

$$\beta = \tan^{-1} \left(\frac{-3 + \sqrt{9 + \frac{\psi_2^2 + \psi_3^2}{\psi_1^2}}}{4 \frac{\psi_2^2 + \psi_3^2}{\psi_1^2}} \right) \quad (18)$$

Remark 3. Roots of a trigonometric polynomial can be found using companion-matrix methods¹³. Consider the degree 4 polynomial

$$\mathcal{T}(f) = \sum_{j=0}^4 a_j \cos(jf) + \sum_{j=1}^4 b_j \sin(jf).$$

Fourier-Frobenius companion matrix elements are

$$B_{jk} = \begin{cases} \delta_{j,k-1}, & j = 1, \dots, 7, k = 1, \dots, 8, \\ (-1) \frac{h_{k-1}}{a_4 - ib_4}, & j = 8, k = 1, \dots, 8, \end{cases} \quad (19)$$

where δ_{jk} are the Kronecker functions such that $\delta_{jk} = 0$ if $j \neq k$ and $\delta_{jj} = 1$, and h_k are

$$h_k = \begin{cases} a_{4-k} + ib_{4-k}, & k = 0, \dots, 3, \\ 2a_0, & k = 4, \\ a_{k-4} - ib_{k-4}, & k = 5, \dots, 8. \end{cases} \quad (20)$$

The roots of $\mathcal{T}(f)$ are obtained from eigenvalues z_k of the matrix defined in Eq. (19) as

$$f_{k,m} = \arg(z_k) - i \log(|z_k|) \bmod (2\pi), \quad k = 1, \dots, 8.$$

Real-valued roots of $\mathcal{T}(f)$ are such that $|z_k| = 1$. Therefore, this technique allows to find roots of the switch function and, thus, find out the structure of the solution for a given costate. It is important to stress that the trigonometric polynomial is of degree 4, which means that the switching function can have at most 8 roots.

Corollary 1. The original optimal control problem (11) has a solution.

Proof. The relaxed problem has at least one solution (Proposition 1), and any control solution actually belongs to U by virtue of Proposition 2. Such controls must be optimal for the original problem, whence existence. \square

3 | SOLUTION USING CONVEX OPTIMISATION AND CONTINUATION

3.1 | Convex approximation for a reliable initial guess

In order to use indirect shooting methods for solving optimal control problem, we need first a reliable initial guess for the costate $p_{\delta I}$. We propose an approximation by a convex mathematical program similar to the one used in⁹ for controllability check purposes. To this end, define the bounded cone \hat{K}_α obtained by truncating the K_α at its tangency points with U (check Figure 6a). This cone is bounded by a disk denoted D_α . This new control set is a subset of the convex hull of U , in order that any solution of

$$\max_{u(f) \in \hat{K}_\alpha} (\delta I(2\pi) | d_I) \quad \text{subject to} \quad \delta I' = \varepsilon \sum_{i=1}^3 u_i G_i(I, f), \quad \delta I(0) = 0, \quad \delta I(2\pi) \text{ parallel to } d_I, \quad (21)$$

will define an admissible control for the convex relaxation of the original control problem. Note that existence holds for this new problem (Filippov again, as \hat{K}_α is convex and bounded) and that any solution will also have a bang-bang structure. A similar analysis to the one of Section 2.2 on $\text{conv}(U)$ indeed allows to prove that

Proposition 3. An optimal control u of problem (21) on \hat{K}_α verifies the following: (i) when ψ belongs to the interior of K_α^0 , u is zero; when ψ does not belong to K_α^0 , (ii-a) the control is uniquely determined and belongs to the circle ∂D_α , unless (ii-b) ψ is colinear to the axis \hat{s} of the cone K_α in which case the control still belongs to the ∂D_α but is not uniquely determined, as shown in Figure 6b. Moreover, any optimal control is made of finitely many subarcs corresponding to case (i) or (ii-a) over one period.

Proof. As \hat{K}_α and K_α have the same polar cone, (i) is clear. Conversely, when ψ belongs to the open complement of K_α^0 , the colinearity condition $\psi \wedge \hat{s} = 0$ boils down to checking a polynomial condition in f and has only isolated zeros corresponding to case (ii-b). When ψ is not colinear to \hat{s} , the unique maximiser of $(\psi | u)$ for u in \hat{K}_α indeed belongs to the circle ∂D_α , which is case (ii-a). \square

This structure being analogous to that of solutions of the original problem, one hopes to retrieve a reasonable approximation to be used to initiate a differential continuation (see Section 3.2). In particular, we note that the original problem (11) on U and problem (21) on \hat{K}_α share the same switching function associated with contacts with ∂K_α^0 and given by (16).

Consider the following discretisation of (21): the control set \hat{K}_α is approximated by a polyhedral cone $\hat{K}_\alpha^g \subset \hat{K}_\alpha$ generated as the convex hull of g vertices V_1, \dots, V_g chosen in $\partial \hat{K}_\alpha$, as shown on Figure 7. (Note that the 3D cone \hat{K}_α is not finitely generated.) Any control in \hat{K}_α^g is given by a bounded conical combination

$$u(f) = \sum_{j=1}^g v_j(f) V_j, \quad v_j(f) \geq 0, \quad \sum_{j=1}^g v_j(f) \leq 1, \quad f \in \mathbf{S}^1, \quad j = 1, \dots, g. \quad (22)$$

The functions v_j are modeled using an N -dimensional basis of trigonometric polynomials, $\Phi(f) = (1, e^{if}, e^{2if}, \dots, e^{(N-1)if})$:

$$v_j(f) = (\Phi(f) | c_j)_H \quad (23)$$

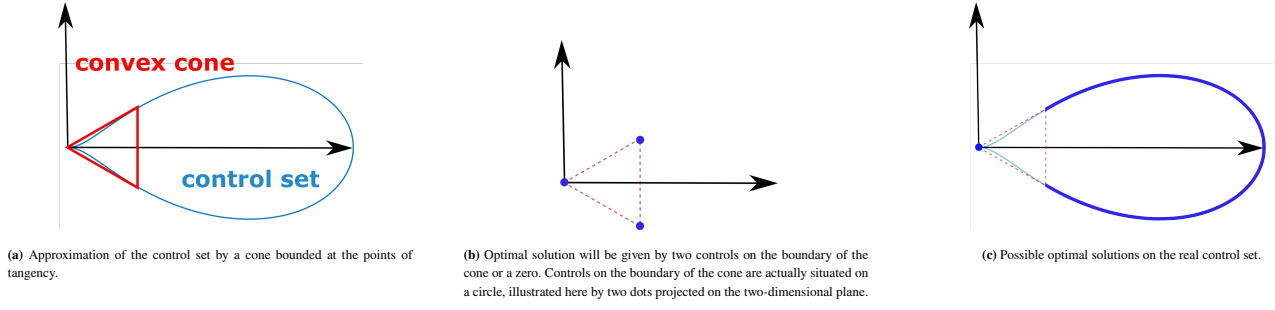


Figure 6 Approximation of the control set by a convex cone.

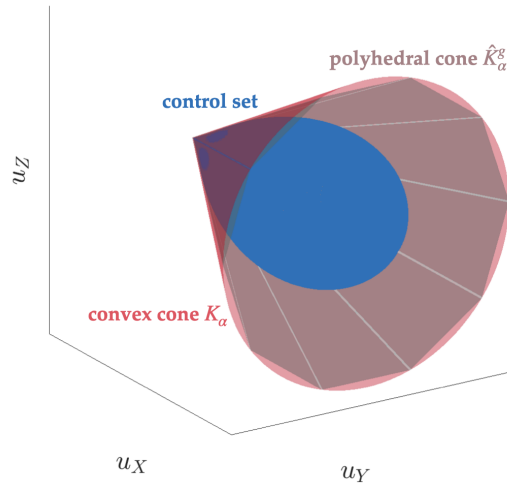


Figure 7 Approximation of the bounded convex cone \hat{K}_α by the polyhedral cone \hat{K}_α^g .

with $c_j \in \mathbb{C}^N$ complex-valued coordinates of v_j in $\Phi(f)$, and $(\cdot|\cdot)_H$ stands for the Hermitian product on \mathbb{C}^N . To enforce the positivity constraint, we leverage on the formalism of squared functional systems outlined in¹⁴. It allows to recast continuous positivity constraints into linear matrix inequalities (LMI), that can be solved using convex optimisation. Actually, $\Phi(f)$ has a corresponding squared functional system given by $S^2(f) = \Phi(f)\Phi^H(f)$, with $\Phi^H(f)$ being the conjugate transpose of $\Phi(f)$. According to¹⁴, let us define a linear operator $\Lambda_H : \mathbb{C}^N \rightarrow \mathbb{C}^{N \times N}$ that maps coefficients of a polynomial in $\Phi(f)$ to its squared base, and its adjoint operator $\Lambda_H^* : \mathbb{C}^{N \times N} \rightarrow \mathbb{C}^N$ such that

$$(Y|\Lambda_H(c))_H = (\Lambda_H^*(Y)|c)_H, \quad Y \in \mathbb{C}^{N \times N}. \quad (24)$$

The theory of squared functional systems states that, for the trigonometric polynomial $(\Phi(f)|c)$ to be non-negative, it is necessary and sufficient that there exists a Hermitian semidefinite positive matrix $Y \geq 0$, such that

$$\Lambda_H^*(Y) = c.$$

The fact that there is no gap between nonnegativeness of the polynomials and reduction to a sum of squares (that can be expressed in SDP terms) is remarkable and peculiar to the univariate case, without restriction on the degree (actually, it remains true for bivariate trigonometric polynomials, see¹⁵; this could allow us to include the second angle, δ , instead of using a polyhedral approximation). All in all, this approach permits to translate the semi-infinite constraint (24) into an SDP one that can be very efficiently treated by modern convex optimisation solvers. The operator Λ_H^* is defined by means of Toeplitz matrices:

$$\Lambda_H^*(Y) = \begin{bmatrix} \text{tr}(Y|T_0) \\ \vdots \\ \text{tr}(Y|T_{N-1}) \end{bmatrix} \quad (25)$$

with kl -coefficients of T_j such that

$$\begin{aligned} T_0 &= I, \\ T_j^{(k,l)} &= \begin{cases} 2 & \text{if } k-l = j \\ 0 & \text{otherwise} \end{cases} \quad j = 1, \dots, N-1. \end{aligned} \quad (26)$$

For an admissible control u valued in \hat{K}_α^g , one has

$$\int_0^{2\pi} \sum_{i=1}^3 u_i(f) G_i(I, f) df = \sum_{j=1}^g (L_j c_j + \bar{L}_j \bar{c}_j) \quad (27)$$

with $L_j(I)$ in $\mathbb{C}^{5 \times N}$ defined by

$$L_j(I) = \frac{1}{2} \sum_{i=1}^3 \int_{S^1} V_{ij} G_i(I, f) \Phi^H(f) df, \quad (28)$$

where $V_j = (V_{ij})_{i=1,\dots,3}$. We note that the components of $L_j(I)$ are Fourier coefficients of the function $\sum_{i=1}^3 V_{ij} G_i(I, f)$. $L_j(I)$ are approximated using the discrete Fourier transform (DFT). Since vector fields G_i are smooth, truncation of the series is justified by the fast decrease of the coefficients. Finally, for a control u valued in \hat{K}_α^g , coefficients v_j are truncated Fourier series of order $N-1$. As a result, for a given vector d_I , the SDP approximation is

$$\begin{aligned} \max_{c_j \in \mathbb{C}^N, Y_b, Y_j \in \mathbb{C}^{N \times N}} (\delta I | d_I) \quad \text{subject to} \quad & \delta I = \varepsilon \sum_{j=1}^g (L_j c_j + \bar{L}_j \bar{c}_j) \text{ parallel to } d_I \\ & Y_j \geq 0, \quad \Lambda^*(Y_j) = c_j, \quad j = 1, \dots, g, \\ & Y_b \geq 0, \quad \Lambda^*(Y_b) = (1, 0, \dots, 0) - \sum_{j=1}^g c_j. \end{aligned} \quad (29)$$

The Lagrange variable of the discretisation of the equality constraint that δI is parallel to d_I from the convex program is expected to be a fair approximation of the costate $p_{\delta I}$ of (21). More importantly, it is hoped that the bang-bang control structure associated with this $p_{\delta I}$ is indeed the same as for the solution of the problem defined on \hat{K}_α .

3.2 | Multiple shooting, differential continuation and callback

Homotopy, *aka.* continuation, allows to solve a complex problem by connecting it continuously to a simpler problem. The idea is then to follow the path (assumed to be regular enough) of solutions from the simpler problem towards the targeted one. See, *e.g.*,^{16,17} for applications in optimal control. In our case, a parameter λ defined between 0 and 1 allows to connect the problem with control set the bounded convex cone \hat{K}_α at $\lambda = 0$, to the original problem with the non-convex drop-like control set U at $\lambda = 1$. In order to be able to solve the problem for $\lambda = 0$, we rely on the solution of the convex program on \hat{K}_α^g to provide an admissible solution. This solution is used not only to compute an educated guess for the initial costate but also to devise the appropriate multiple shooting function. To do so, we use the control structure corresponding to the approximation of $p_{\delta I}$ provided by the convex optimisation and described at Proposition 3. This proposition tells us that, when ψ (a function of $p_{\delta I}$ and f) belongs to the open complement of the polar cone K_α^0 , the control must be equal to the *dynamical feedback* described in case (ii-a) (apart for some isolated points that correspond to case (ii-b) that we can neglect); we denote $u_b^0(f, p_{\delta I})$ this control. Similarly, for such values of ψ , Proposition 2 for the problem on $\text{conv}(U)$ —and actually U , check Corollary 1—, implies that the control must be a solution of (14)-(15). (While these equations provide an explicit solution for the coordinate δ of the control, β is only implicitly defined and we discuss its actual computation in Section 3.3.) We assume that this solution is unique and denote it $u_b^1(f, p_{\delta I})$. Then, for λ in $[0, 1]$ and ψ outside the polar cone, we define

$$u_b(f, p_{\delta I}, \lambda) := (1 - \lambda) u_b^0(f, p_{\delta I}) + \lambda u_b^1(f, p_{\delta I})$$

as the convex combination of the dynamical feedbacks for $\lambda = 0$ and $\lambda = 1$. Conversely, for any λ in $[0, 1]$ and ψ in the interior of the polar cone, the control is set to zero.

For a given λ , one has a finite sequence of arcs with either $u = u_b$ (bang arcs), or $u = 0$ (zero arcs). Contacts with ∂K_α^0 are characterized by (16) whose left-hand side defines the switching function, denoted $\varphi(f, p_{\delta I})$ (not depending on λ in our particular setting). To this finite sequence of arcs is associated a multiple shooting function in a standard fashion. Assume for

instance that the structure is bang-zero-bang. Then the shooting function has three arguments: the (constant) value of the costate, $p_{\delta I}$, and the two switchings times (true anomalies) bounding the central zero arc, f_1 and f_2 . (So that $(p_{\delta I}, f_1, f_2)$ belong to \mathbf{R}^7 .) Plugging $u = u_b(f, p_{\delta I}, \lambda)$ into the dynamics of δI and integrating on $[0, f_1]$ from $\delta I(0) = 0$ allows to compute $\delta I_1 := \delta I(f_1)$. As the control is zero on $[f_1, f_2]$, δI remains constant on the coast arc and we set $\delta I_2 := \delta I_1$. The control $u = u_b(f, p_{\delta I}, \lambda)$ is eventually plugged again on $[f_2, 2\pi]$ to compute $\delta I_f := \delta I(2\pi)$, starting from δI_2 . The associated value of the shooting function is obtained by concatenating the left-hand side of the four equations below, forming a vector of dimension $4 + 1 + 2 = 7$ (note that the first colinearity equation indeed has dimension $5 - 1 = 4$):

$$\delta I_f \wedge d_I = 0, \quad (30)$$

$$(p_{\delta I} | d_I) - 1 = 0, \quad (31)$$

$$\varphi(f_1, p_{\delta I}) = 0, \quad (32)$$

$$\varphi(f_2, p_{\delta I}) = 0. \quad (33)$$

This defines a shooting function $S(\xi, \lambda)$ with, for this bang-zero-bang structure, $\xi := (p_{\delta I}, f_1, f_2)$. Once the first solution for $\lambda = 0$ is obtained, the path of zeros is followed by differential continuation, typically using a parametrisation by its curvilinear abscissa:

$$s \mapsto (\lambda(s), \xi(s)) \text{ with } S(\xi(s), \lambda(s)) = 0.$$

We refer, *e.g.*, to¹⁸ for the assumptions needed to do so. Note that, according to (32), we look for normal extremals (compare with (13)).

One important issue in practice is that it might not be possible to reach $\lambda = 1$ because, at some $\lambda(\bar{s})$ in $(0, 1)$, the structure of the solution changes; for instance because one subarcs disappears. It is crucial to be able to detect such a change during homotopy since then, the shooting function has to be redefined according to the new structure. This is achieved using a standard callback mechanism along with differential continuation. On the previous bang-zero-bang example, the continuation is monitored and, at each step of the path following procedure, a simple test is performed: if the exit time of the zero arc, f_2 , becomes inferior to the entry time f_1 (this is detected by a sign change on $f_2 - f_1$, as going forward in time makes sense mathematically but is not allowed to obtain admissible trajectories), the continuation is stopped. And restarted at $\lambda(\bar{s})$ with a new shooting function (in this case, a single shooting one, as only one bang arc would be left), using $\xi(\bar{s})$ as initial guess. More elaborated tests can be constructed to detect a new arc appearing, *etc.* In our case, a callback is used to detect a structure change from 5 subarcs to 3 (see Section 4). This methodology can readily be extended to any other structure with a finite number of arcs.

3.3 | Implicit treatment of the Hamiltonian maximisation

Regarding the computation of $u_b^1(f, p_{\delta I})$, we know after Proposition 2 that the control is either zero, either solution of (14-15). The first equation for the coordinate β of u has no closed form solution. There is a preliminary numerical discussion of the number of solutions in² (we actually look for a global maximiser of the Hamiltonian over U , which may allow to eliminate some strictly local minimiser that also verify (14)) for a particular set of values of the sail parameters. More generally, while maximisation of the Hamiltonian often yields an explicit expression of the control as a dynamics feedback function of the state and the costate, it is not always the case. In such a situation, we advocate an implicit treatment of this maximisation, incorporating the stationarity equation of the Hamiltonian into the shooting procedure. We sketch below a simple way to do so in a general setting.

Assume that, after applying Pontrjagin maximum principle, one has to integrate the following system (x denoting the state, p the costate):

$$\dot{x}(t) = \nabla_p H(x(t), p(t), u(t)), \quad \dot{p}(t) = -\nabla_x H(x(t), p(t), u(t)), \quad (34)$$

where, at each time t , the m -dimensional control $u(t)$ verifies

$$\nabla_u H(x(t), p(t), u(t)) = 0. \quad (35)$$

The last stationarity equation corresponds to an unconstrained situation—whereas a Lagrangian, plus an additional finite dimensional multiplier, should be considered in the presence of constraints—and defines a semi-explicit differential-algebraic equation. Assume that the strong Legendre-Clebsch condition holds in an open neighbourhood of the reference extremal times the open control set, $\nabla_{uu}^2 H \leq -c I_m$ for some positive constant c . Then the Hamiltonian has a unique maximiser, that satisfies $\nabla_u H = 0$, and the previous differential-algebraic equation (DAE) is of index 1 (differentiating once (35) allows to solve for \dot{u}).

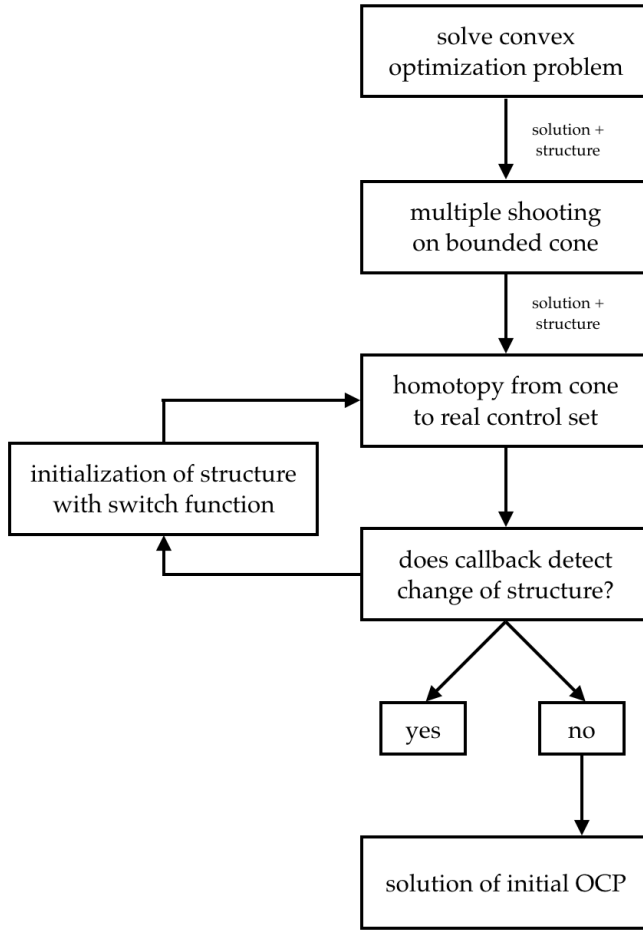


Figure 8 Algorithm for solving optimal control problem (OCP).

In particular, one can extend the Hamiltonian system (34) by adding the equation

$$\dot{u} = -\nabla_{uu}^2 H^{-1}(\nabla_{ux} H \cdot \nabla_p H - \nabla_{up} H \cdot \nabla_x H)(x, p, u) := g(x, p, u),$$

with initial condition $\nabla_u H(x(0), p(0), u(0)) = 0$. The new system remains Hamiltonian as is clear setting $\hat{x} := (x, u)$, $\hat{p} := (p, p_u)$ and

$$\hat{H}(x, u, p, p_u) := H(x, p, u) + (p_u | g(x, p, u))$$

with $p_u(0) = 0$. (One can obviously eliminate the trivial equation on p_u , which is an extra but identically zero costate.) In the case of a shooting approach, the value of $u(0)$ is an additional shooting variable. Keeping the system in Hamiltonian form is convenient in the algorithmic framework described in Section 4, but other approaches for DAE such as predictor-corrector ones can of course be considered. In our case, we use this approach with $x = \delta I$, $p = p_{\delta I}$ to deal with the implicit equation (14) on β (while we use (15) to solve it explicitly for δ). The combination of this implicit approach with multiple shooting, homotopy and callback is described in the last section.

4 | NUMERICAL EXAMPLE: THE JPL SQUARE SAIL

This section presents the results for a specific example of a trajectory around a planet, where a change in the solution's structure occurs during the process of differential continuation. It is important to note that, in our simulations, such a change in structure is more of an exception rather than the norm. The majority of the simulated trajectories exhibited the same structure for the control set, which was represented by a bounded cone and a drop-like set. However, we present this particular example to highlight the

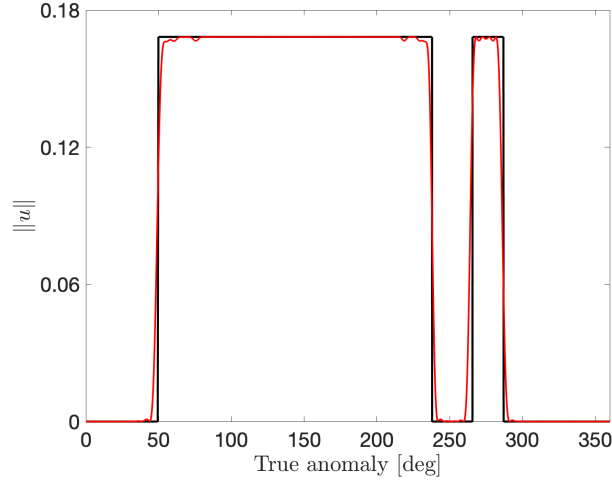


Figure 9 Results from convex optimisation (in red) on \hat{K}_α^g compared to multiple shooting on \hat{K}_α with $\lambda = 0$ (in black).

situation where the structure differs, necessitating the use of a callback and a switch function to reinitialise the multiple shooting algorithm.

Consider a solar sail whose optical properties determining shape of U are taken from JPL Square Sail defined in ¹, Table 2.1: $\rho = 0.88$, $s = 0.94$, $\varepsilon_b = 0.55$, $\varepsilon_f = 0.05$, $B_b = 0.55$, $B_f = 0.79$. An initial orbit is determined by $I = (10^\circ, 50^\circ, 30^\circ, 1, 0.1)$. The desired maneuver to be performed is increase of the inclination γ_2 , what gives $d_I = (0, 1, 0, 0, 0)$. Initially, an optimisation problem is solved to obtain the optimal solution for the control on the bounded polyhedral cone, \hat{K}_α^g . This result serves as a reliable initial guess for the optimal control problem (OCP) being considered. The solution consists of five arcs, with the first arc being zero. The costate is

$$p_{\delta I}^{conv} = (-0.0837, 1, -0.0052, 0.0398, 0.0852).$$

Switches between zeros and bangs occurs at $f_1 = 49.4^\circ$, $f_2 = 237.9^\circ$, $f_3 = 265.6^\circ$, $f_4 = 286.9^\circ$. We substitute the values $(p_{\delta I}, f_1, f_2, f_3, f_4)$ into the multiple shooting algorithm to solve the problem on the bounded cone \hat{K}_α , with the continuation parameter $\lambda = 0$. The resulting controls are depicted in Figure 9, illustrating the norm of the control u . The red line represents the controls obtained through convex optimisation on the polyhedral cone, which leads to some oscillations due to the linear combinations of a finite number of generators on the cone. In contrast, the black curve corresponds to the controls derived from the multiple shooting approach on the bounded cone. Since no discretisation is involved, the line appears smooth. Both solutions are remarkably similar, providing evidence that convex optimisation serves as a highly accurate approximation for the initial guess with $\lambda = 0$.

During the process of differential continuation using the previous solution, as expected, the callback function detects the disappearance of an arc, resulting in $f_4 < f_3$, which is physically impossible. This occurrence takes place at approximately $\lambda = 0.0256$, a relatively small value. This indicates that by "inflating" the control set, a better solution quickly emerges, leading to the removal of the second arc. Utilising the switch function, the algorithm reconfigures the solution structure by eliminating the second arc, resulting in a zero-bang-zero structure. The algorithm continues the homotopy process, ultimately arriving at the final solution on the real control set with $\lambda = 1$.

Figure 10 presents the solutions throughout the process. In Figure 10a, the solutions are shown starting from $\lambda = 0$ in black and progressing until $\lambda = 0.0256$ in red, which is when the second arc disappears. The grey curves illustrate the deformation of the solution as differential continuation is performed. At the point when $\lambda = 0.0256$, the structure of the solution is reinitialised, as depicted in Figure 10b. The subsequent continuation process leads to the final result at $\lambda = 1$, shown in blue. The final solution corresponds to the costate equal to

$$p_{\delta I}^{sol} = (-0.1637, 1, -0.0972, 0.0712, 1.6037).$$

Figure 11 shows the solution on the real control set in terms of control angles, defining attitude of the solar sail. When $\beta = \frac{\pi}{2}$, it means that the sail is aligned with the solar rays, so that no surface is exposed to the solar light, what corresponds to $\|u\| = 0$. In this case δ angle is not defined.

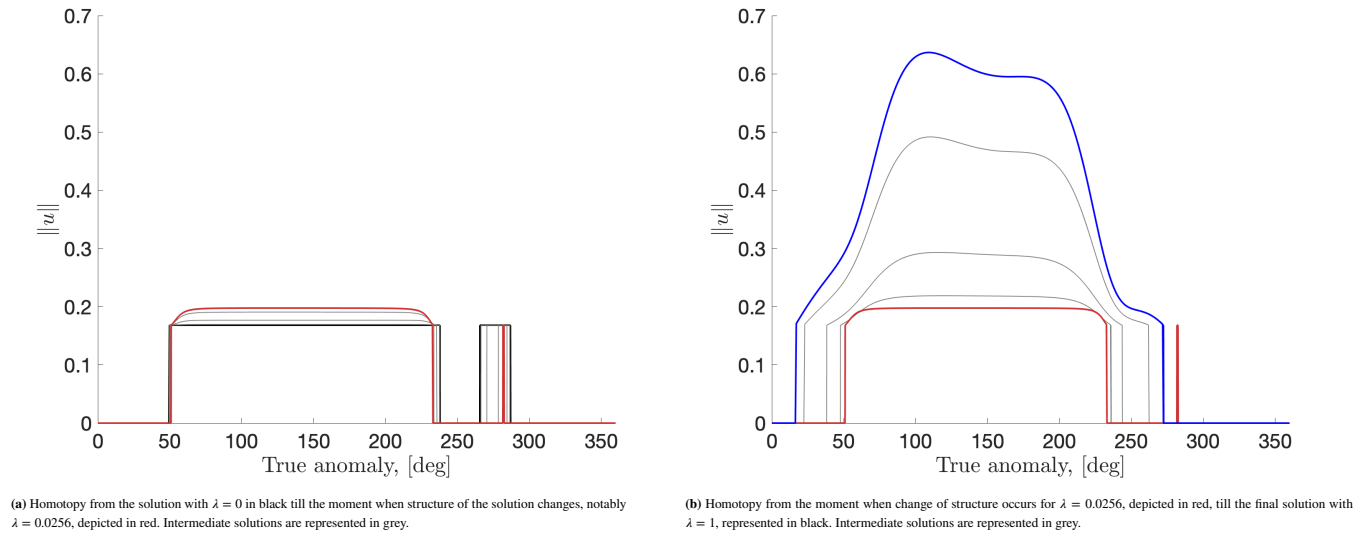


Figure 10 Deformation of the solution along the differential continuation.

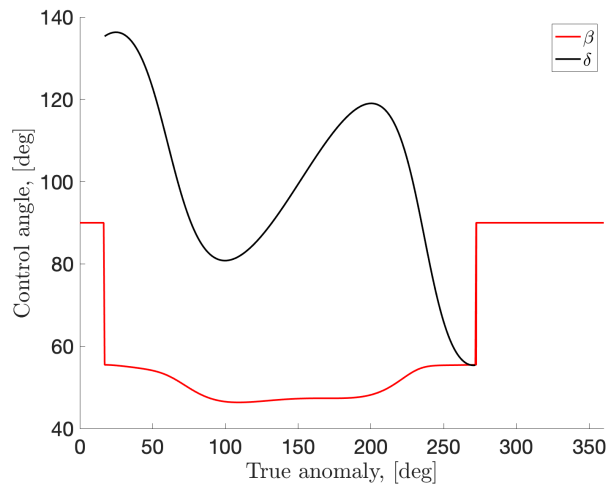


Figure 11 Controls in terms of orientation angles of the sail.

Convex optimisation is done using CVX package in Matlab^{19,20}. Multiple shooting with differential continuation is performed using *control toolbox (CT)* developed in python. Regarding the implicit computation of the maximising control discussed in Section 3.3, we monitor the computation of the Hessian of the Hamiltonian wrt. β on bang arcs along the continuation to ensure that at least of strict local solution is obtained. As we start from an initial guess provided by convex optimisation, we actually expect that a global maximiser is preserved along the entire continuation, but this analysis would require further analysis. An example of the code which is executable online is available.³

To verify exactness of the solution, trajectory of the initial system is integrated using the control given by our methodology. The resulting trajectory shows that, indeed, all orbital parameters stay constant, except for the inclination that keeps increasing. Note that our algorithm assumes that the state is fixed for one orbital period, but the final integration was performed for the real dynamics.

³<https://ct.gitlabpages.inria.fr/gallery/solarsail/solarsail-simple-version-implicit.html>

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