Post-Print

This is the accepted version of:

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Low-Thrust Minimum-Fuel Optimization in the Circular Restricted Three-Body Problem
doi:10.2514/1.G001080

The final publication is available at http://dx.doi.org/10.2514/1.G001080

Access to the published version may require subscription.

When citing this work, cite the original published paper.
I. Introduction

The challenging missions envisioned for the upcoming decades involve innovative spacecraft trajectory concepts. For instance, installing outposts at Earth–Moon Lagrange points requires novel transfers. These orbits shall, at minimum, be able to combine the benefits inherent to multi-body dynamics with those enabled by advanced propulsion technologies. The three-body dynamics already suffice to produce free transport mechanisms, whereas the high specific impulse typical of low-thrust propulsion allows considerable propellant savings. The challenge is then to design optimal low-thrust, three-body transfers.

This subject has recently gained the attention of the scientific community. In [1] design synthesis is achieved by combining invariant manifolds and attainable sets, which are sets containing ballistic and low-thrust orbits, respectively. In [2] transfers to distant periodic orbits have been designed by targeting their invariant manifolds with low-thrust propulsion. Using invariant manifolds as a first guess solution in high-fidelity, low-thrust optimization was examined in [3]. Low-energy, low-thrust
transfers to the Moon have been formulated in [4, 5]. Low-thrust optimization in a three-body model was studied also in [6–10] without relying on invariant manifolds. These concepts were demonstrated by SMART-1 [11], where the pioneering ideas in [12] were applied.

In this field, the case of low-thrust transfers to Earth–Moon Lagrange point orbits is of interest. As the latter are chosen as ideal target for a number of future applications [13–15], it is crucial to assess their accessibility in terms of cost and time. In doing so, it is worth considering departure from a geostationary transfer orbit (GTO). This orbit is chosen because launch vehicles are optimized for GTO, and the case of piggy back spacecraft may also be accounted for. Nevertheless, finding an optimal low-thrust trajectory from GTO to Lagrange point orbits in the Earth–Moon system is particularly challenging because of: 1) the long-duration nature of the solutions, 2) the high number of revolutions, 3) the strong nonlinearities in the three-body vector field, 4) the high number of bang-bang structures in the optimal guidance law. These features have prevented finding end-to-end optimal transfers, and some sort of approximations, leading to suboptimal solutions, have been favored over optimality. In [16], the transfer orbit is defined by patching together a tangential thrust spiral with a leg targeting a portion of the stable manifold associated to the final orbit. As optimization is carried out in the second phase only, overall optimality is lost. The same idea is replicated in a number of later works [17–20]. Alternatively, much higher initial orbits are chosen [10, 21–24], yet not matching practical applications.

In this Note, the complete optimal low-thrust GTO-to-Halo transfer is solved for the first time. This result is achieved with an indirect approach and constant specific impulse engine. Thrust-to-mass ratios in agreement with currently available technology are considered. Some effective techniques are applied to cope with problem complexity. These methods involve solving the minimum-fuel, minimum-energy, and minimum-time problems [10, 25], implementing energy-to-fuel homotopy [10, 26–28], continuing the maximum thrust magnitude [10], computing the analytic Jacobians [9], and accurately detecting the switching points through a combination of Newton and bisection methods [29]. In the present approach the solution structure is not prescribed (as in [16–18, 20, 30, 31]), but rather it is found a posteriori, with low-thrust propulsion which is free to act everywhere; the final coast arc, recalling a stable manifold branch, is found automatically. The techniques presented
in this Note are useful in practical cases where very low-thrust accelerations are used for long times in highly nonlinear vector fields.

The Note is organized as follows. In Section II, the dynamical model is presented, and the minimum-fuel, minimum-energy, and minimum-time problems are stated. In Section III, the analytical Jacobians for the minimum-energy and minimum-fuel problems are derived. In Section IV, the original, hybrid switching detection technique is presented, and details on practical implementation are given. The case study is solved in Section V and critical analysis is conducted.

II. Statement of the Problem

A. Controlled Restricted Three-Body Problem

The circular restricted three-body problem studies the motion of a massless spacecraft, $P_3$, under the gravitational field generated by two primaries $P_1, P_2$ of masses $m_1, m_2$, respectively, which move in circular motion due to their mutual interaction. The dynamics are written in a rotating frame with non-dimensional units: the angular velocity of $P_1, P_2$, their distance, and the sum of their masses are all set to unity. The Earth–Moon model is considered in this work, where $\mu = m_2/(m_1 + m_2) \approx 0.012$. The Earth, of mass $(1 - \mu)$, is located at $(-\mu, 0, 0)$, whereas the Moon, of mass $\mu$, is located at $(1 - \mu, 0, 0)$; see [32] for details. When low-thrust propulsion is considered, it is convenient to write the equations of motions as [9]

$$
\begin{align*}
\dot{x} &= f(x, \alpha, u) \\
\dot{v} &= \begin{bmatrix} v \\ g(r) + h(v) + u T_{\text{max}} \alpha/m \\ -u T_{\text{max}}/c \end{bmatrix},
\end{align*}
$$

where $r = [x, y, z]^T$ and $v = [v_x, v_y, v_z]^T$ are the spacecraft position and velocity vectors, respectively, $m$ denotes the spacecraft mass, $T_{\text{max}}$ is the maximum thrust magnitude, $c = I_{\text{sp}} g_0$ represents the exhaust velocity ($I_{\text{sp}}$ is the thruster specific impulse and $g_0$ is the gravitational acceleration at the sea level). The control variables are the throttle factor, $u \in [0, 1]$, and the thrust direction unit.
vector, \( \alpha \). The functions \( g(r) \) and \( h(v) \) are defined as [1, 18]

\[
\begin{bmatrix}
x - (1 - \mu)(x + \mu)/r_1^2 - \mu(x + \mu - 1)/r_2^2 \\
y - (1 - \mu)y/r_1^2 - \mu y/r_2^2 \\
-(1 - \mu)z/r_1^2 - \mu z/r_2^2
\end{bmatrix},
\begin{bmatrix}
2v_y \\
-2v_x \\
0
\end{bmatrix},
\] (2)

where \( r_1 \) and \( r_2 \) represent the distance of \( P_3 \) to the Earth and Moon, respectively; i.e.,

\[
\begin{align*}
r_1 &= [(x + \mu)^2 + y^2 + z^2]^{1/2}, \\
r_2 &= [(x + \mu - 1)^2 + y^2 + z^2]^{1/2}.
\end{align*}
\] (3)

**B. Construction of the Shooting Function**

In minimum-fuel problems, a solution of (1) minimizes

\[
J_f = \frac{T_{\text{max}}}{c} \int_{t_i}^{t_f} u \, dt,
\] (4)

where \( t_i, t_f \) denote the initial, final time, respectively. In these problems \( u \) is either zero or one [33]. This dichotomy causes discontinuities in the shooting function [28], and poses severe restrictions on numerical methods. A smoothing technique, or homotopic approach, was introduced in [26] to relax the control profile and to enforce discontinuity in a gradual way [34]. It consists in using the objective function

\[
J = \frac{T_{\text{max}}}{c} \int_{t_i}^{t_f} [u - \varepsilon u(1 - u)] \, dt, \quad \varepsilon \in [0, 1].
\] (5)

The minimum-energy problem (\( \varepsilon = 1 \)) is solved first; the solution is then continued by decreasing \( \varepsilon \) until (4) is reached (\( \varepsilon = 0 \)) [28]. In this work, optimal transfers between two given states (a point on the GTO, \( (r_i, v_i) \), and a point on the \( L_1 \) halo, \( (r_f, v_f) \)) are designed, and therefore the boundary conditions read

\[
\begin{align*}
r(t_i) - r_i &= 0, & v(t_i) - v_i &= 0, & m(t_i) - 1 &= 0, \\
r(t_f) - r_f &= 0, & v(t_f) - v_f &= 0,
\end{align*}
\] (6)

where the initial mass is scaled to one, whereas the final mass is obviously unconstrained. The fixed terminal state formulation in (6), as opposed to that in [16, 18, 22–24], has been chosen to favor convergence. To vary the arrival point on the Halo orbit, an outer loop can be implemented, or a problem accounting for free terminal conditions can be stated.
The Hamiltonian of the problem is [35]

\[ H = \lambda_r \cdot v + \lambda_v \cdot \left[ g(r) + h(v) + \frac{u T_{\text{max}}}{m} \alpha \right] - \lambda_m \frac{u T_{\text{max}}}{c} \left[ u - \varepsilon u(1 - u) \right] , \]

where \( \lambda = [\lambda_r, \lambda_v, \lambda_m]^T \) is the vector of costates, whose dynamics are

\[
\dot{\lambda} = -\frac{\partial H}{\partial x} \Rightarrow \begin{bmatrix} \dot{\lambda}_r \\ \dot{\lambda}_v \\ \dot{\lambda}_m \end{bmatrix} = \begin{bmatrix} -G^T \lambda_v \\ -\lambda_r - H^T \lambda_v \\ \frac{u T_{\text{max}}}{m^2} \lambda_v \cdot \alpha \end{bmatrix},
\]

with \( G = \partial g(r)/\partial r \) and \( H = \partial h(v)/\partial v \) (only nonzero elements listed).

\[
G_{1,1} = 1 - (1 - \mu)/r_1^3 + 3(1 - \mu)(x + \mu)^2/r_1^5 - \mu/r_2^2 + 3\mu(x + \mu - 1)/r_2^5,
\]

\[
G_{2,2} = 1 - (1 - \mu)/r_1^3 + 3(1 - \mu)g^2/r_1^5 - \mu/r_2^2 + 3\mu g^2/r_2^5,
\]

\[
G_{3,3} = -(1 - \mu)/r_1^3 + 3(1 - \mu)z^2/r_1^5 - \mu/r_2^2 + 3\mu z^2/r_2^5,
\]

\[
G_{1,2} = G_{2,1} = 3(1 - \mu)(x + \mu)g/r_1^5 + 3\mu(x + \mu - 1)g/r_2^5,
\]

\[
G_{1,3} = G_{3,1} = 3(1 - \mu)(x + \mu)z/r_1^5 + 3\mu(x + \mu - 1)z/r_2^5,
\]

\[
G_{2,3} = G_{3,2} = 3(1 - \mu)yz/r_1^5 + 3\mu yz/r_2^5, \quad H_{1,2} = -H_{2,1} = 2.
\]

Due to (6), the boundary conditions for the costates are all unknown except for \( \lambda_m \). As the final mass is free, its associated costate must be zero at \( t_f \) [35], i.e.,

\[
\lambda_m(t_f) = 0.
\]

Conditions on the control variables \( u, \alpha \) are derived by applying the Pontryagin Maximum Principle (PMP) [36], which states that the Hamiltonian is minimized along an optimal trajectory. Since \( u T_{\text{max}}/m \geq 0 \) in (7), then the optimal thrust direction is [37]

\[
\alpha^* = -\frac{\lambda_v}{\lambda_v}.
\]

Substituting Eq. (10) into Eq. (7) yields

\[
H = \lambda_r \cdot v + \lambda_v \cdot \left[ g(r) + h(v) \right] + \frac{u T_{\text{max}}}{c} (S - \varepsilon + \varepsilon u),
\]

where a switching function, \( S \), is defined as

\[
S = -\lambda_v \frac{c}{m} - \lambda_m + 1,
\]
along with its first derivative

\[
\dot{S} = \frac{c}{m} \left( -\lambda_r - H^\top \lambda_v \right) \cdot \lambda_v.
\]  

(13)

The throttle factor \( u \) is the only decision variable in Eq. (11). Its value can be related to \( x, \lambda \) by applying again the PMP. The optimal value, \( u^* \), can be stated in terms of \( S \) through

\[
u^* = 0 \quad \text{if } S > \varepsilon \\
u^* = (\varepsilon - S)/2\varepsilon \quad \text{if } -\varepsilon \leq S \leq \varepsilon \\
u^* = 1 \quad \text{if } S < -\varepsilon.
\]

(14)

Note that in minimum-fuel problems (\( \varepsilon = 0 \)) the magnitude of \( u^* \) depends only upon the sign of \( S \), and a bang-bang profile is generated.

Once the optimal control variables \( \alpha^*, u^* \) are determined as functions of the states and costates, through Eqs. (10) and (14), the motion can be integrated implicitly with dynamics

\[
\dot{y} = F(y) \quad \Rightarrow \quad \begin{bmatrix}
\dot{r} \\
\dot{v} \\
\dot{m} \\
\dot{\lambda}_r \\
\dot{\lambda}_v \\
\dot{\lambda}_m
\end{bmatrix} = \begin{bmatrix}
v \\
g(r) + h(v) - (\lambda_v / \lambda_c) u T_{max} / m \\
-u T_{max} / c \\
-G^\top \lambda_v \\
-\lambda_r - H^\top \lambda_v \\
-\lambda_c u T_{max} / m^2
\end{bmatrix}.
\]

(15)

where \( y = [x, \lambda]^\top \) is a 14-dimensional canonical variable [9]. A two-point boundary value problem (TPBVP) is defined by (15) together with the boundary conditions (6) and (9). If the initial costate vector \( \lambda_i \) was given, one could integrate (15) (with \( u \) as in (14)) and check if the final conditions are verified. In the likely case in which these are not met, \( \lambda_i \) can be adjusted based on first-order information. This scheme is the essence of the shooting procedure.

**Remark 1** Let \([x(t), \lambda(t)]^\top = \varphi([x_1, \lambda_1]^\top, t_i, t)\) be the solution of (15) integrated from \([x_i, \lambda_i]^\top\), initial time \( t_i \) to a generic time \( t \). The optimization problem is stated as follows.

Find \( \lambda_i \) such that \([x(t_f), \lambda(t_f)]^\top = \varphi([x_i, \lambda_i]^\top, t_i, t_f)\) satisfies

\[
\begin{align*}
r(t_f) - r_f &= 0 \\
v(t_f) - v_f &= 0 \\
\lambda_m(t_f) &= 0.
\end{align*}
\]

(16)
More synthetically, the problem is to find $\lambda_i$ such that $Z(\lambda_i) = 0$, where $Z(\lambda)$, or the shooting function, is the seven-dimensional vector valued function given by the final boundary conditions (16). As finding the zeros of $Z(\lambda)$ is not trivial, a number of techniques have to be implemented.

C. Minimum-Time Problem

A minimum-time step has to be solved to infer the minimum possible transfer time, $t_{f_{\min}}$, for each $T_{\text{max}}$; see [10]. The minimum-fuel problem (16) has to be solved then for $t_f \geq t_{f_{\min}}$. The performance index of minimum-time problems is $J_{t} = \int_{t_i}^{t_f} dt$, and therefore the Hamiltonian reads

$$H_t = \lambda_r \cdot v + \lambda_v \cdot \left[ g(r) + h(v) + \frac{u T_{\text{max}}}{m} \alpha \right] - \lambda_m \frac{u T_{\text{max}}}{c} + 1. \quad (17)$$

Applying the PMP to (17) yields

$$u^* = 0 \quad \text{if } S_t > 0$$

$$u^* \in [0, 1] \quad \text{if } S_t = 0$$

$$u^* = 1 \quad \text{if } S_t < 0,$$

where the minimum-time switching function is

$$S_t = -\lambda_v \frac{c}{m} - \lambda_m. \quad (19)$$

When the final time is free, the transversality condition sets $H_t(t_f)$ to zero [25].

Remark 2 The minimum-time problem is stated as follows.

$$\begin{align*}
\text{Find } (\lambda_i, t_f) \text{ such that } [x(t_f), \lambda(t_f)]^\top &= \varphi([x_i, \lambda_i]^\top, t_i, t_f) \text{ satisfies}
\end{align*}$$

$$\begin{align*}
\begin{cases}
\begin{align*}
& r(t_f) - r_f = 0 \\
& v(t_f) - v_f = 0 \\
& \lambda_m(t_f) = 0 \\
& H_t(t_f) = 0.
\end{align*}
\end{cases}
\end{align*} \quad (20)$$

III. Analytic Derivatives

To increase accuracy and robustness of the shooting procedure, analytic derivatives are provided. In particular, the state transition matrix (STM) of (15), $\Phi(t_i, t) = d\varphi(y, t_i)/dy$, is derived. The STM maps small variations in the initial condition, $\delta y_i$, over $t_i \to t$, i.e., $\delta y(t) = \Phi(t_i, t)\delta y(t_i)$. 

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The STM is subject to the variational equation

$$\dot{\Phi}(t_i, t) = D_y F \Phi(t_i, t), \quad \Phi(t_i, t_j) = I_{14 \times 14}, \quad \text{(21)}$$

where $D_y F$, the Jacobian of $F(y)$ in (15), has two different expressions: A) for $u^* = 0$ or $u^* = 1$ and B) for $u^* = (\varepsilon - S)/2$ (see Eqs. (12), (14), and (15)), i.e.

$$D_y F^{(A)} = \begin{bmatrix}
0 & I & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
G & H & \lambda_v & u_{T_{\text{max}}} & 0 & -u_{T_{\text{max}}} & \left( \frac{I}{\lambda_v} - \frac{\lambda_v \lambda_v^\top}{\lambda_v^2} \right) & 0
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
\frac{\partial (G^\top \lambda_v)}{\partial \rho} & 0 & 0 & 0 & -G^\top & 0
0 & 0 & 0 & -I & -H^\top & 0
0 & 2 \lambda_v u_{T_{\text{max}}} & 0 & -\lambda_v^\top u_{T_{\text{max}}} & 0
\end{bmatrix} \quad \text{(22)}$$

$$D_y F^{(B)} = \begin{bmatrix}
0 & I & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
G & H & \Omega_1 & 0 & \Omega_2 & \Omega_3 & 0 & 0 & 0 & 0 & -G^\top & 0
0 & 0 & \Omega_4 & 0 & \Omega_5 & \Omega_6 & 0 & 0 & 0 & 0 & -I & -H^\top & 0
\frac{\partial (G^\top \lambda_v)}{\partial \rho} & 0 & 0 & 0 & -G^\top & 0
0 & 0 & 0 & -I & -H^\top & 0
0 & 0 & \Omega_7 & 0 & \Omega_8 & \Omega_9 & 0
\end{bmatrix} \quad \text{(23)}$$

where

$$\begin{align*}
\Omega_1 &= \frac{\lambda_v u_{T_{\text{max}}}}{\lambda_v^2} + \frac{\lambda_v c_{T_{\text{max}}}}{2 \varepsilon m^2} \\
\Omega_2 &= -\frac{\lambda_v \lambda_v^\top c_{T_{\text{max}}}}{2 \varepsilon m^2} - \frac{u_{T_{\text{max}}}}{\lambda_v} \left( \frac{I}{\lambda_v} - \frac{\lambda_v \lambda_v^\top}{\lambda_v^2} \right) \\
\Omega_3 &= -\frac{\lambda_v T_{\text{max}}}{\lambda_v^2} \quad \text{and} \quad \Omega_4 = -\frac{\lambda_v T_{\text{max}}}{\lambda_v^2} \\
\Omega_5 &= \frac{\lambda_v T_{\text{max}}}{2 \varepsilon m^2} \\
\Omega_6 &= -\frac{\lambda_v^2 u_{T_{\text{max}}}}{\lambda_v^2} - \frac{\lambda_v c_{T_{\text{max}}}}{2 \varepsilon m^2} \\
\Omega_7 &= -\frac{\lambda_v^2 u_{T_{\text{max}}}}{2 \varepsilon m^2} \\
\Omega_8 &= -\frac{\lambda_v^2 c_{T_{\text{max}}}}{2 \varepsilon m^2} \\
\Omega_9 &= -\frac{\lambda_v^2 T_{\text{max}}}{2 \varepsilon m^2}.
\end{align*}$$

Note that Eq. (21) is equivalent to 196 first-order differential equations for the elements of $\Phi(t, t)$; it requires $D_y F$ to be evaluated along $y(t)$, and therefore Eqs. (15) and (21) have to be integrated simultaneously. Let $z$ be a vector containing $y$ and columns of $\Phi$; its first-order variation is

$$\dot{z} = F(z) \quad \Rightarrow \quad \begin{bmatrix}
\dot{y} \\
\text{vec}(\Phi)
\end{bmatrix} = \begin{bmatrix}
F(y) \\
\text{vec}(D_y F \Phi)
\end{bmatrix}, \quad \text{(24)}$$

where ‘vec’ is an operator which converts the matrix into a column vector. Eqs. (24) yields 210 nonlinear differential equations. It is worth mentioning that $\Phi(t_i, t)$ only maps states along a continuous orbit. If at time $t_j$ there is a ‘bang-bang’ switching point, a discontinuity arises, and the STM across such discontinuity, $\Psi(t_j)$, has to be determined. This STM can be computed as [9]

$$\Psi(t_j) = \left( \frac{\partial y(t_j^+)}{\partial y(t_j^-)} \right) = I_{14 \times 14} + \left( y(t_j^+) - y(t_j^-) \right) \left( \frac{\partial S}{\partial y} \right) \bigg|_{t_j^-}, \quad \text{(25)}$$

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where $t_j^-$, $t_j^+$ represent the times immediately before, after the discontinuity. In the likely case in which there are $N$ ‘bang-bang’ points at times $t_1, \ldots, t_N$, the composite STM is given by [9]

$$\Phi(t_f, t_i) = \Phi(t_f, t_N^-) \Psi(t_N^-) \Phi(t_N^-, t_{N-1}) \Psi(t_{N-1}) \ldots \Phi(t_2^-, t_1^-) \Psi(t_1^-) \Phi(t_1^-, t_i) = \frac{\partial y(t_f)}{\partial y(t_i)}.$$ (26)

### IV. Switching Detection Techniques

It has been mentioned that the right-hand side of Eq. (15) is discontinuous. When numerical integration is performed, discontinuities cause the integration error to accumulate about the switching points if these are not explicitly determined, which in turn threatens the entire shooting process. A hybrid switching detection technique combined with a variable step integration has been implemented to precisely detect the switching times. Let $t_k$, $t_{k+1}$ be two consecutive integration steps and let $y_k = y(t_k)$, $y_{k+1} = y(t_{k+1})$. These states are linked through $y_{k+1} = \varphi_{\text{RK}}(y_k, t_k, t_{k+1})$, where $\varphi_{\text{RK}}$ is the numerical integration map. Let also $t_{\text{sw}}$ be the switching time, and $S_k = S(y(t_k))$, $S_{k+1} = S(y(t_{k+1}))$ as per the switching function (12). With reference to Eq. (14), suppose that $S_k > \varepsilon$ and $0 < S_{k+1} < \varepsilon$; extending the analysis to the other cases in Eq. (14) is straightforward.

The focus is to determine the switching time $t_{\text{sw}}$, $t_{\text{sw}} \in (t_k, t_{k+1})$, which zeroes $f(t_{\text{sw}}) = S(t_{\text{sw}}) - \varepsilon$.

The hybrid switching detection procedure developed is made of two steps. First, the Newton method is implemented because of its high efficiency; in general, only 4-5 iterations are needed to reach machine precision (i.e., $10^{-15}$). The first guess is set to $t_k$ and the analytic derivative in (13) is used. In case the Newton method fails or converges to solutions outside of the searching interval $(t_k, t_{k+1})$, a bisection procedure is executed. The latter is slower but more robust.

**Algorithm 1** The switching time, $t_{\text{sw}}$, is determined with the following procedure (Newton method).

i) Initialize $t_{\text{sw}} = t_k$

ii) Compute $t_{\text{sw}} = t_{\text{sw}} - f(t_{\text{sw}})/\dot{S}(t_{\text{sw}})$

iii) Perform $y(t_{\text{sw}}) = \varphi_{\text{RK}}(y_k, t_k, t_{\text{sw}})$

iv) Evaluate $f(t_{\text{sw}})$ and $\dot{S}(t_{\text{sw}})$ with $y(t_{\text{sw}})$

v) Repeat steps ii)–iv) until $f(t_{\text{sw}}) < \text{tol}$
In case \( f(t_{sw}) > \text{tol} \) or \( t_{sw} < t_k \) or \( t_{sw} > t_{k+1} \) execute a Bisection method within \((t_k, t_{k+1})\).

A. Practical Implementation

The low-thrust trajectory optimization problem has been implemented in a numerical framework. The 7th/8th-order Runge–Kutta scheme, with relative and absolute tolerances set to \(10^{-14}\), has been used to integrate Eqs. (15) and (21) (or equivalently Eqs. (24)), with the implicit optimal thrusting strategy in (14). The implementation of a generic integration step is outlined in Fig. 1.
The input required to execute an integration step are: 1) $t_k$, the initial integration time, 2) $h_p$, the step-size predicted at previous integration step (a value is guessed at first step), 3) $z_k$, the 210-dimensional composite state at $t_k$ (state and costate, $y_k$, and the elements of $\Phi(t_i, t_k)$, see Eqs. (24), 4) $u_{\text{type}}$, the logical thrust type at previous integration step, i.e., ‘On’, ‘Medium’, or ‘Off’ according to (14) (only ‘On’ or ‘Off’ in minimum-fuel problems). According to $u_{\text{type}}$, the flow is redirected to one of the three integration blocks in Fig. 1 where a prediction on $z_{k+1}$ is made, $z_{k+1} = \varphi_{\text{RK}}(z_k, t_k, t_k + h_p)$. Note that $z_{k+1}$ is the composite state at time $t_{k+1} = t_k + h_f$, where $h_f$ is the step-size corrected during the Runge–Kutta integration, which also updates the value of $h_p$ for the subsequent integration. A correction is then made according to the value of the switching function $S_{k+1}$. The branch tagged with ‘Thrust On’ is analyzed for brevity. According to the relations (14), if $S_{k+1} < -\varepsilon$ the thrusting strategy is not changed over $[t_k, t_{k+1}]$, and therefore $t_{k+1}$, $z_{k+1}$ are saved. When $-\varepsilon < S_{k+1} < \varepsilon$ a switching (from ‘On’ to ‘Medium’) occurs in $[t_k, t_{k+1}]$. In this case, the switching time, $t_{sw}$, as well as the state at the switching point, $z_{sw}$, are detected with Algorithm 1, and $u_{\text{type}}$ is updated (see ‘Block 1’ at the bottom of Fig. 1). In the case $S_{k+1} > \varepsilon$ two options are possible: a) in the minimum-energy problem ($\varepsilon \neq 0$) jumping from thrust ‘On’ to ‘Off’ is not possible, and therefore the step size has to be reduced and the process must be repeated; b) in the minimum-fuel problem ($\varepsilon = 0$) the switching time is detected, along with the state at the switching point, and the control type is updated. In the latter case, the STM is updated by considering the control discontinuity as in (25).

V. Simulations and Results

The case study is a low-thrust transfer from a GTO to a $L_1$ halo orbit in the Earth–Moon system. The GTO has periapsis and apoapsis altitudes of $h_p = 400$ km and $h_a = 35,864$ km, respectively. The specific impulse is 3000 s, and the initial satellite mass is 1500 kg. The transfer begins at periapsis. From this point on, the low-thrust propulsion is used to increase the energy until a particular point on the halo orbit is targeted. Numerical experiments show that varying the target point has negligible impact over the total cost of the transfer. This orbit has an out-of-plane amplitude of 8000 km, and is consistent with [16–18, 20]. Table 1 summarizes the parameters used
in this work. The basis for distance, time, velocity, and mass units are the Earth–Moon distance, the inverse of their orbital angular velocity, the lunar circular speed, and the initial spacecraft mass, respectively. Table 2 reports the initial and final conditions.

<table>
<thead>
<tr>
<th>Table 1 Physical constants.</th>
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</thead>
<tbody>
<tr>
<td>Physical constant</td>
</tr>
<tr>
<td>Mass parameter, $\mu$</td>
</tr>
<tr>
<td>Gravitational field, $g_0$</td>
</tr>
<tr>
<td>Length unit, LU</td>
</tr>
<tr>
<td>Time unit, TU</td>
</tr>
<tr>
<td>Speed unit, VU</td>
</tr>
<tr>
<td>Mass unit, MU</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Table 2 Boundary conditions.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Boundary condition</td>
</tr>
<tr>
<td>Initial position vector</td>
</tr>
<tr>
<td>Initial velocity vector</td>
</tr>
<tr>
<td>Final position vector</td>
</tr>
<tr>
<td>Final velocity vector</td>
</tr>
</tbody>
</table>

A. Minimum-Time Solutions

A minimum-time step is necessary to infer the minimum time required to accomplish the transfer, $t_{f_{\text{min}}}$, for a given maximum thrust, $T_{\text{max}}$. Once $t_{f_{\text{min}}}$ is known, the subsequent energy-to-fuel optimization is formulated with a fixed final time $t_f$ satisfying $t_f \geq t_{f_{\text{min}}}$ at each $T_{\text{max}}$. A continuation on $T_{\text{max}}$ is performed. That is, at iteration $k$ problem (20) is solved for a given value of thrust, and the pair $(\lambda^{[k]}, t_{f_{\text{min}}}^{[k]})$ is used as first guess solution in iteration $k + 1$, with decreased $T_{\text{max}}$ (see Section II C). Solutions with $T_{\text{max}}$ ranging from 10 N down to 0.3 N have been achieved, with transfer times that increase accordingly from 7.8 days to 171.6 days. The results, in terms of $t_{f_{\text{min}}}$, $m_f$ are reported in Table 3 for different $T_{\text{max}}$, and their trend versus the initial thrust-to-mass ratio, $T_{\text{max}}/m_i$, is drawn in Fig. 2. The transfer trajectories corresponding to $T_{\text{max}}$ of 10 N, 2 N, 1
Table 3 Minimum-time solutions computed.

<table>
<thead>
<tr>
<th>$T_{\text{max}}$ (N)</th>
<th>$T_{\text{max}}/m_i$ (m/s$^2$)</th>
<th>$t_{\text{fmin}}$ (days)</th>
<th>$m_f$</th>
<th>$\Delta v$ (km/s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>10 \times 10^{-3}</td>
<td>10.6667</td>
<td>7.8549</td>
<td>0.8462</td>
<td>4.8015</td>
</tr>
<tr>
<td>9 \times 10^{-3}</td>
<td>9.0000</td>
<td>8.6861</td>
<td>0.8469</td>
<td>4.7777</td>
</tr>
<tr>
<td>8 \times 10^{-3}</td>
<td>5.3333</td>
<td>9.6522</td>
<td>0.8488</td>
<td>4.7133</td>
</tr>
<tr>
<td>7 \times 10^{-3}</td>
<td>4.6667</td>
<td>10.8133</td>
<td>0.8518</td>
<td>4.6119</td>
</tr>
<tr>
<td>6 \times 10^{-3}</td>
<td>4.0000</td>
<td>12.6278</td>
<td>0.8516</td>
<td>4.6186</td>
</tr>
<tr>
<td>5 \times 10^{-3}</td>
<td>3.3333</td>
<td>12.9634</td>
<td>0.8730</td>
<td>3.9050</td>
</tr>
<tr>
<td>4 \times 10^{-3}</td>
<td>2.6667</td>
<td>16.0510</td>
<td>0.8742</td>
<td>3.8655</td>
</tr>
<tr>
<td>3 \times 10^{-3}</td>
<td>2.0000</td>
<td>21.1363</td>
<td>0.8758</td>
<td>3.8130</td>
</tr>
<tr>
<td>2 \times 10^{-3}</td>
<td>1.3333</td>
<td>29.1512</td>
<td>0.8858</td>
<td>3.4865</td>
</tr>
<tr>
<td>1 \times 10^{-4}</td>
<td>6.6667 \times 10^{-4}</td>
<td>56.2458</td>
<td>0.8898</td>
<td>3.3570</td>
</tr>
<tr>
<td>0.9 \times 10^{-4}</td>
<td>6.0000 \times 10^{-4}</td>
<td>59.8376</td>
<td>0.8945</td>
<td>3.2055</td>
</tr>
<tr>
<td>0.8 \times 10^{-4}</td>
<td>5.3333 \times 10^{-4}</td>
<td>64.6165</td>
<td>0.8987</td>
<td>3.0708</td>
</tr>
<tr>
<td>0.7 \times 10^{-4}</td>
<td>4.6667 \times 10^{-4}</td>
<td>80.2242</td>
<td>0.8900</td>
<td>3.3505</td>
</tr>
<tr>
<td>0.6 \times 10^{-4}</td>
<td>4.0000 \times 10^{-4}</td>
<td>87.6667</td>
<td>0.8970</td>
<td>3.1253</td>
</tr>
<tr>
<td>0.5 \times 10^{-4}</td>
<td>3.3333 \times 10^{-4}</td>
<td>112.0327</td>
<td>0.8903</td>
<td>3.3408</td>
</tr>
<tr>
<td>0.4 \times 10^{-4}</td>
<td>2.6667 \times 10^{-4}</td>
<td>138.4519</td>
<td>0.8915</td>
<td>3.3021</td>
</tr>
<tr>
<td>0.3 \times 10^{-4}</td>
<td>2.0000 \times 10^{-4}</td>
<td>171.6254</td>
<td>0.8991</td>
<td>3.0580</td>
</tr>
</tbody>
</table>

N, and 0.6 N are reported in Fig. 3 in the Earth-centered frame. The throttle factor and switching function trends are instead reported in Fig. 4. In these solutions the thrust is always on; i.e., $S_i$ in (19) is always negative, which triggers the last condition (18).

B. Minimum-Fuel Solutions

Minimum-fuel solutions are computed with a two-step continuation: 1) the minimum-energy problem ($\varepsilon = 1$) is solved iteratively with decreasing $T_{\text{max}}$; 2) once a low-level of thrust is reached, $\varepsilon$ is decreased until the minimum-fuel problem ($\varepsilon = 0$) is reached. Fig. 5 reports the thrust profile and switching function for the 10 steps needed to switch from minimum-energy to minimum-fuel, $T_{\text{max}} = 10$ N. In this example, the continuation law is $\varepsilon_j = (j^2 - 1)/(N^2 - 1)$, with $j = 10, 9, \ldots, 2, 1$ and $N = 10$. The ability of the process in smoothing the thrust profile gradually is evident. For
Fig. 2 Minimum transfer time, $t_{f\min}$, and final mass ratio, $m_f$, versus thrust acceleration, $\frac{T_{max}}{m_i}$, in minimum-time solutions.

For each $T_{max}$, the transfer time is chosen such that $t_f = c_{t_f} \times t_{f\min}$, where $c_{t_f}$ is a coefficient selected manually. In step 2) above, varying $c_{t_f}$ is a practical method to overcome convergence problems.

The sensitivity of the final mass with respect to $c_{t_f}$ is not considered.

For the sake of completeness, both minimum-energy and minimum-fuel solutions are reported for different values of $T_{max}$ in Table 4. As expected, for each $T_{max}$, minimum-fuel solutions use less propellant than minimum-energy ones. The solution with $T_{max} = 0.6$ N has a thrust-to-mass ratio of $4 \times 10^{-4}$ m/s$^2$, which is consistent with the current state-of-the-art, and agrees with reference works [16–25]. This solution takes 140 days and needs just 8.5% of propellant mass fraction. Compared to the analogous minimum-time solution, this means saving 27 kg of propellant at the cost of spending additional 53 days of flight time.

Four sample minimum-fuel transfer trajectories are shown in Fig. 6 (rotating frame) and Fig. 7 (inertial frame). In both figures, the thick and thin lines indicate thrust and coast arcs, respectively. It can be seen that in the minimum-fuel solutions the thruster is on duty across the periapses (see Fig. 7). The algorithm is also capable of exploiting the stable structure associated to the halo orbit: all solutions foresee a final coast arc, which injects the spacecraft toward $L_1$ (the very final short thrust arc is the orbit injection maneuver, see Fig. 6). The long-duration, multi-spiral features of the low-thrust solutions can be appreciated in Figs. 6(d) and 7(d).

The trends for $u$ and $S$ corresponding to the solutions in Figs. 6 and 7 are reported in Fig. 8. The remarkable features of the low-thrust solution are proven in Figs. 6(d), 7(d) and 8(d): in
Fig. 3 Minimum-time trajectories for different values of $T_{\text{max}}$; Earth-centered inertial frame.

In this work, the formulation of low-thrust minimum-fuel, minimum-energy, and minimum-time trajectories in the restricted three-body problem is discussed, with applications to transfers from a geostationary transfer orbit to a halo orbit about $L_1$ in the Earth–Moon model. As these problems
pose several challenges, a number of issues allowing an effective and efficient implementation of a solution method are discussed. These techniques involve solving the minimum-time problem, introducing energy-to-fuel homotopy, formulating analytical derivatives, and implementing a hybrid method for accurate switching point detection.

A fixed point of the halo is targeted without exploiting its stable manifold. The algorithm is left free to place thrust arcs everywhere, and the existence of free coast arcs, mimicking the stable manifold, is found a posteriori. By the best of the authors’ knowledge, these techniques have enabled solving, for the first time, the geostationary transfer orbit to halo low-thrust transfer entirely, so
Table 4 Solutions to minimum-energy ($\varepsilon = 1$) and minimum-fuel ($\varepsilon = 0$) cases, for different $T_{\text{max}}$.

<table>
<thead>
<tr>
<th>$T_{\text{max}}$ (N)</th>
<th>$T_{\text{max}}/m_i$ (m/s$^2$)</th>
<th>$c_t$</th>
<th>$t_f$ (days)</th>
<th>$m_f$ ($\varepsilon = 1$)</th>
<th>$m_f$ ($\varepsilon = 0$)</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>$6.6667 \times 10^{-3}$</td>
<td>1.1</td>
<td>8.6404</td>
<td>0.9034</td>
<td>0.9105</td>
</tr>
<tr>
<td>9</td>
<td>$6.0000 \times 10^{-3}$</td>
<td>1.1</td>
<td>9.5548</td>
<td>0.9016</td>
<td>0.9103</td>
</tr>
<tr>
<td>8</td>
<td>$5.3333 \times 10^{-3}$</td>
<td>1.1</td>
<td>10.6174</td>
<td>0.9003</td>
<td>0.9090</td>
</tr>
<tr>
<td>7</td>
<td>$4.6667 \times 10^{-3}$</td>
<td>1.1</td>
<td>11.8946</td>
<td>0.8988</td>
<td>0.9072</td>
</tr>
<tr>
<td>6</td>
<td>$4.0000 \times 10^{-3}$</td>
<td>1.2</td>
<td>15.1533</td>
<td>0.9009</td>
<td>0.9092</td>
</tr>
<tr>
<td>5</td>
<td>$3.3333 \times 10^{-3}$</td>
<td>1.3</td>
<td>16.8524</td>
<td>0.8964</td>
<td>0.9053</td>
</tr>
<tr>
<td>4</td>
<td>$2.6667 \times 10^{-3}$</td>
<td>1.3</td>
<td>20.8663</td>
<td>0.8936</td>
<td>0.9015</td>
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<tr>
<td>3</td>
<td>$2.0000 \times 10^{-3}$</td>
<td>1.3</td>
<td>27.4773</td>
<td>0.8992</td>
<td>0.9088</td>
</tr>
<tr>
<td>2</td>
<td>$1.3333 \times 10^{-3}$</td>
<td>1.3</td>
<td>37.8965</td>
<td>0.9091</td>
<td>0.9146</td>
</tr>
<tr>
<td>1</td>
<td>$6.6667 \times 10^{-4}$</td>
<td>1.5</td>
<td>84.3688</td>
<td>0.9038</td>
<td>0.9131</td>
</tr>
<tr>
<td>0.6</td>
<td>$4.0000 \times 10^{-4}$</td>
<td>1.6</td>
<td>140.2678</td>
<td>0.9069</td>
<td>0.9150</td>
</tr>
</tbody>
</table>

Table 5 Initial costate, $\lambda_v$, and final time, $t_f$, for the four minimum-fuel solutions presented.

<table>
<thead>
<tr>
<th>$T_{\text{max}}$</th>
<th>$\lambda_{v,i}$</th>
<th>$\lambda_{e,i}$</th>
<th>$\lambda_{m,i}$</th>
<th>$t_f$ (days)</th>
</tr>
</thead>
<tbody>
<tr>
<td>10 N</td>
<td>$[15.616017, 32.875896, -0.094522]^T$</td>
<td>$[-0.101606, 0.044791, -0.000150]^T$</td>
<td>0.133266</td>
<td>8.6</td>
</tr>
<tr>
<td>2 N</td>
<td>$[6.476119, 13.589566, 0.133954]^T$</td>
<td>$[-0.041598, 0.018240, -0.000162]^T$</td>
<td>0.126720</td>
<td>37.8</td>
</tr>
<tr>
<td>1 N</td>
<td>$[4.229762, 10.873196, -0.420268]^T$</td>
<td>$[-0.030411, 0.016232, 0.000094]^T$</td>
<td>0.123321</td>
<td>84.3</td>
</tr>
<tr>
<td>0.6 N</td>
<td>$[3.728251, 9.861519, -0.412042]^T$</td>
<td>$[-0.027239, 0.014900, -0.000007]^T$</td>
<td>0.122115</td>
<td>140.2</td>
</tr>
</tbody>
</table>

avoiding splitting up the transfers into phases, which involves suboptimality.

Acknowledgments

C.Z. would like to acknowledge the support provided by the China Scholarship Council. This work was supported by the National Natural Science Foundation of China (grant No. 11102007).

References

Fig. 6 Minimum-fuel trajectories for different values of $T_{\text{max}}$, rotating frame.


Fig. 7 Minimum-fuel trajectories for different values of $T_{\text{max}}$; Earth-centered inertial frame.


Fig. 8 Throttle factor, $u$, and switching function, $S$, profiles in minimum-fuel solutions.

doi:10.2514/2.4210.


