

Asteroid Rendezvous Missions

An Application of Optimal Control

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Launch	Mission	Time (years)	Δv (m/s)
1978	<i>ISEE-3</i>	7.08	430
1989	<i>Galileo</i>	2.03	1300
1996	<i>NEAR</i>	1.36	1450
1997	<i>Cassini</i>	2.21	500
1998	<i>Deep Space 1</i>	2.91	1300
1999	<i>Stardust</i>	5.34	230
2003	<i>Hayabusa</i>	2.32	1400
2004	<i>Rosetta</i>	4.51	2200
2005	<i>DIXI</i>	5.63	190
2007	<i>Dawn</i>	3.76	10000
2014	<i>Hayabusa 2</i>	4.00*	
2016*	<i>Osiris-Rex</i>	2.00*	
2020*	<i>ARM</i>	4.00*	

NASA'S Plan to Haul an Asteroid

An ambitious plan included in NASA's 2014 budget proposal would send a robotic spacecraft out to capture an asteroid and haul it back to an orbit around the moon for study. One of NASA's stated goals is to send an asteroid by the year 2025.

How to Bag a Space Rock

A 2012 feasibility study was based on asteroid 2008 TC₂₆ and Plan A, NASA's preferred option of intercepting an asteroid at 0.28 AU (around 28% of the way from Earth to the Sun) and about the same distance from the Sun in space by firing the rocket engines.

The spacecraft's main propulsion needs to be controlled by fuel-efficient thrusters. This is a goal of the program to reduce the fuel expense. A counterweight to an electric field can capture propellant molecules that can be used for long runs to orbit an asteroid.

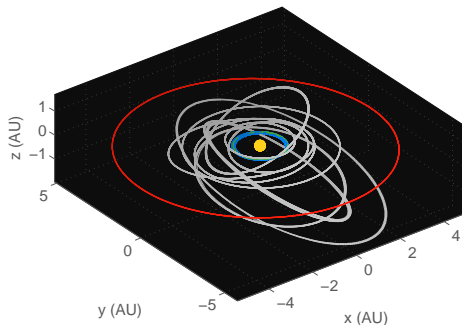
An asteroid would be moved from its original orbit by a slow tugboat from the moon, cutting its velocity by the amount of the intended orbital velocity.

Mission: Asteroid

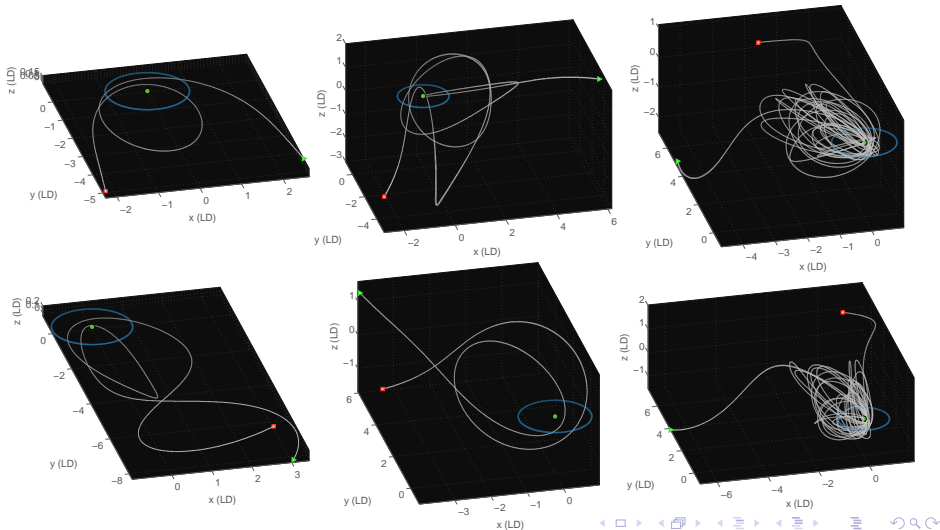
- The asteroid orbit spacecraft would be launched from Earth on an Asteroid Redirect Mission, shortly after the end of the space race.
- The spacecraft arrives for 12 years orbit to monitor the target asteroid.
- Operations on the asteroid take about 90 days. The capture bag is deployed, and once secured, the asteroid is captured by towing.
- The cabin back to the vicinity of the Earth takes less than 10 years.
- After another 90-day mission around the moon, the asteroid is placed in a stable orbit, where it can be studied by the nearest direct-view satellite for study.

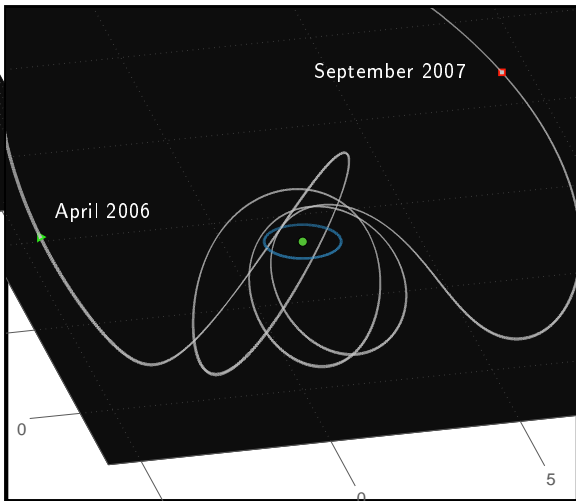
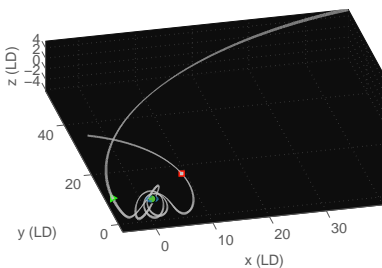
ILLUSTRATION BY NASA/JPL/AMOS

Asteroid	Perigee (LD)	Diameter
Ida	680.3	31 km
Mathilde	369.7	53 km
Gaspra	326.1	18 km
Borrelly	188.1	8 km
Halley	164.7	11 km
Churyumov	153.4	4 km
Braille	123.3	2 km
Eros	59.0	17 km
Itokawa	13.6	0.54 km
2006RH ₁₂₀	0.7	0.003 km



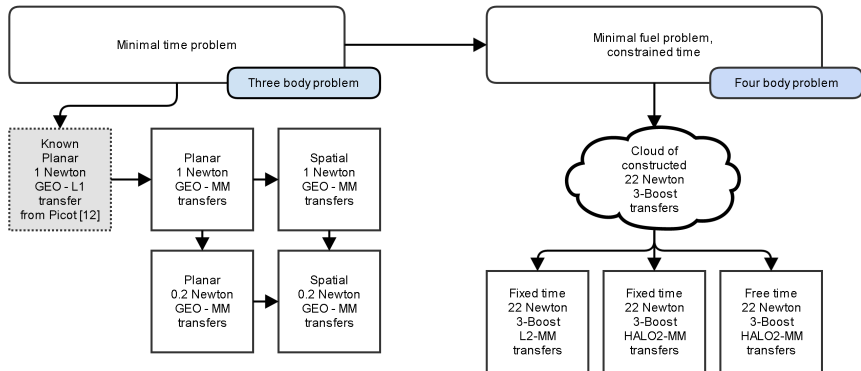
- ▶ Database of 16,923 simulated *minimoons*
- ▶ At least one in orbit at any time (1-meter diameter)





Defn. *temporary capture* :

1. the geocentric Keplerian energy $E < 0$
2. the geocentric distance is less than three Hill radii ≈ 0.03 AU.
3. at least one full revolution around Earth in co-rotating frame



Consider a *general* optimal control problem

$$\left\{ \begin{array}{l} \dot{x}(t) = f(t, x(t), u(t)) \\ \min_{u(\cdot) \in U} \int_{t_0}^{t_f} f_0(t, x(t), u(t)) dt + g(t_f, x_f) \\ x(t_0) = x_0 \in M_0 \subset M, \quad x(t_f) = x_f \in M_1 \subset M \end{array} \right.$$

In 1958, the Russian mathematician Lev Pontryagin stated a fundamental *necessary first order* optimality condition, known as the *Pontryagin's Maximum Principle*.

This result generalizes the *Euler-Lagrange* equations from the theory of calculus of variations.

If u is optimal on $[0, t_f]$ then there exists $p \in T_x^* M$ and $p^0 \in \mathbb{R}^-$ such that $(p^0, p) \neq (0, 0)$ and almost everywhere in $[t_0, t_f]$ there holds

- ▶ $z(t) = (x(t), p(t)) \in T^* M$ is *solution* to the *pseudo-Hamiltonian system*

$$\dot{x}(t) = \frac{\partial H}{\partial p}(x(t), p^0, p(t), u(t)), \quad \dot{p}(t) = -\frac{\partial H}{\partial q}(x(t), p^0, p(t), u(t))$$

where

$$H(x, p^0, p, u) = p^0 f^0(x, u) + \langle p, f(x, u) \rangle ;$$

- ▶ *maximization condition*

$$H(x(t), p^0, p(t), u(t)) = \max_{v \in U \subseteq N} H(x(t), p^0, p(t), v) ;$$

- ▶ *transversality condition*

$$p(0) \perp T_{x(0)} M_0 \text{ and } p(t_f) - p^0 \frac{\partial g}{\partial t}(t_f, x(t_f)) \perp T_{x(t_f)} M_1.$$

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- ▶ A triplet (x, p, u) satisfying these 3 conditions is called an *extremal solution*.

Locally $U = \mathbb{R}^n \Rightarrow \frac{\partial H}{\partial u} = 0$.

Assumption : The quadratic form $\frac{\partial^2 H}{\partial u^2}$ is **negative definite** along the extremal $(x(t), p(t), u(t))$.

Implicit function theorem

\Rightarrow In a neighborhood of u , extremal controls are *feedback controls* i.e. smooth functions

$$u_r(t) = u_r(x(t), p(t))$$

\Rightarrow extremal curves are pairs $(x(t), p(t))$ solutions of the *true Hamiltonian system*

$$\dot{x}(t) = \frac{\partial H_r}{\partial p}(x(t), p(t)), \quad \dot{p}(t) = -\frac{\partial H_r}{\partial x}(x(t), p(t))$$

where H_r is the *true Hamiltonian function* defined by

$$H_r(x, p) = H(x, p, u_r(x, p)).$$

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The variational equation

$$\dot{\delta z}(t) = d\vec{H}_r(z) \cdot \delta z(t)$$

is called the *Jacobi equation* along z . One calls a *Jacobi field* a nontrivial solution $J(t)$ of the Jacobi equation along z . t is said to be *vertical* at time t if

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A time t_c is said to be *geometrically conjugate* if there exists a Jacobi field vertical at 0 and t_c . In which case, $x(t_c)$, is said to be *conjugate* to $x(0)$.

Denote $\exp_t(\vec{H}_r)$ the *flow* of the Hamiltonian vectorfield \vec{H}_r . One defines the *exponential mapping* by

$$\exp_{x_0,t}(p(0)) \longrightarrow \Pi_x(z(t, z_0)) = x(t, q_0, p_0)$$

where $z(t, z_0)$, with $z(0) = (x(0), p(0))$ is the trajectory of \vec{H} such that $z(0, z_0) = z_0$.

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Proposition : Let $x_0 \in M$, $L_0 = T_{x_0}^*M$ and $L_t = \exp_t(\vec{H}_r)(L_0)$. Then L_t is a Lagrangian submanifold of T^*M whose tangent space is *spanned* by Jacobi fields starting from L_0 . Moreover $q(t_c)$ is *geometrically conjugate* to x_0 if and only if \exp_{x_0,t_c} is *not a immersion* at p_0 .

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\Rightarrow calculating a conjugate point is *equivalent* to verifying a *rank condition*.

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Theorem : Let t_c^1 be the *first* conjugate time along z . The trajectory $q(\cdot)$ is *locally optimal* on $[0, t_c^1)$ in L^∞ topology ; if $t > t_c^1$ then $x(\cdot)$ is *not locally optimal* on $[0, t]$.

Objective : use optimal control theory to compute *optimal space transfers* in the *Earth-Moon* system

- ▶ *time*-minimal space transfers
- ▶ *energy*-minimal space transfers

First question : How to *model* the *motion* of a spacecraft in the Earth-Moon system ?

- ▶ *Neglect the influences of other planets*
- ▶ The spacecraft *does not* affect the motion of the Earth and the Moon
- ▶ Eccentricity of orbit of the Moon is *very small* (≈ 0.05)

The Earth-Moon-spacecraft system

The motion of the spacecraft in the Earth-Moon system can be modelled by *the planar restricted 3 body problem*.

Description :

- ▶ The Earth (mass M_1) and Moon (mass M_2) are *circularly* revolving around their *center of mass* G .
- ▶ The spacecraft is *negligeable point mass* M involves in the plane defined by the Earth and the Moon.
- ▶ Normalization of the masses : $M_1 + M_2 = 1$
- ▶ Normalization of the distance : $d(M_1, M_2) = 1$.

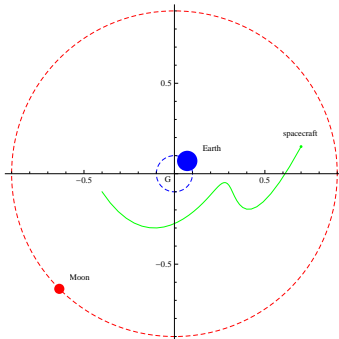


Figure : The circular restricted 3-body problem. The blue dashed line is the orbit of the Earth and the red one is the orbit of the Moon. The trajectory of spacecraft lies in the plan defined by these two orbits.

The Rotating Frame

Idea : Instead of considering a fixed frame $\{G, X, Y\}$, we consider a *dynamic rotating* frame $\{G, x, y\}$ which rotates with the same angular velocity as the Earth and the Moon.

→ *rotation* of angle t

→ *substitution*

$$\begin{pmatrix} X \\ Y \end{pmatrix} = \begin{pmatrix} \cos(t)x + \sin(t)y \\ -\sin(t)x + \cos(t)y \end{pmatrix}$$

→ *simplifies* the equations of the model

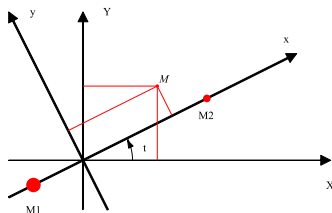


Figure : Comparison between the fixed frame $\{G, X, Y\}$ and the rotating frame $\{G, x, y\}$.

In the rotating frame

- ▶ define the *mass ratio* $\mu = \frac{M_2}{M_1+M_2}$
- ▶ the *Earth* has mass $1 - \mu$ and is located at $(-\mu, 0)$;
- ▶ the *Moon* has mass μ and is located at $(1 - \mu, 0)$;
- ▶ *Equations of motion*

$$\begin{cases} \ddot{x} - 2\dot{y} - x = \frac{\partial V}{\partial x} \\ \ddot{y} + 2\dot{x} - y = \frac{\partial V}{\partial y} \end{cases}$$

where

$-V$: is the *mechanical potential*

$$V = \frac{1 - \mu}{\varrho_1^3} + \frac{\mu}{\varrho_2^3}$$

ϱ_1 : *distance* between the *spacecraft* and the *Earth*

$$\varrho_1 = \sqrt{(x + \mu)^2 + y^2}$$

ϱ_2 : *distance* between the *spacecraft* and the *Moon*

$$\varrho_2 = \sqrt{(x - 1 + \mu)^2 + y^2}.$$

Hill Regions

There are 5 *possible* regions of motion, known as the *Hill regions*. Each region is defined by the value of the *total energy* of the system.

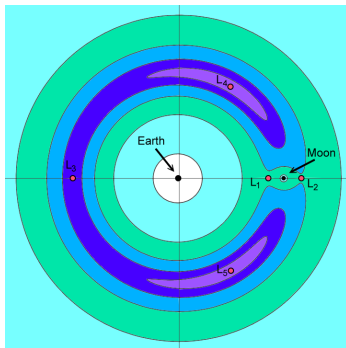


Figure: The Hill regions of the planar restricted 3-body problem

Topology/Shape of the regions is determined with respect to the total energy at the *equilibrium points* of the system

Equilibrium points

Critical points of the mechanical potential

→ Points (x, y) where $\frac{\partial V}{\partial x} = \frac{\partial V}{\partial y} = 0$

- ▶ *Euler points* : colinear points L_1, L_2, L_3 located on the axis $y = 0$, with $x_1 \simeq 1.1557, x_2 \simeq 0.8369, x_3 \simeq -1.0051$.
- ▶ *Lagrange points* : L_4, L_5 which form equilateral triangles with the primaries.

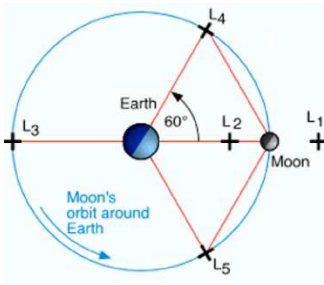


Figure: Equilibrium points of the planar restricted 3-body problem

The controlled restricted 3-Body problem

Control on the motion of the spacecraft?

→ Thrust/Propulsion provided by the engines of the spacecraft

→ control term $u = (u_1, u_2)$ must be added to the equations of motion

→ *controlled dynamics* of the spacecraft

$$\begin{cases} \ddot{x} - 2\dot{y} - x = \frac{\partial V}{\partial x} + u_1 \\ \ddot{y} + 2\dot{x} - y = \frac{\partial V}{\partial y} + u_2. \end{cases}$$

Setting $q = (x, y, \dot{x}, \dot{y})$

→ *bi-input system*

$$\dot{q} = F_0(q) + F_1(q)u_1 + F_2(q)u_2$$

where

$$F_0(q) = \begin{pmatrix} q_3 \\ q_4 \\ 2q_4 + q_1 - (1 - \mu) \frac{q_1 + \mu}{((q_1 + \mu)^2 + q_2^2)^{\frac{3}{2}}} - \mu \frac{q_1 - 1 + \mu}{((q_1 - 1 + \mu)^2 + q_2^2)^{\frac{3}{2}}} \\ -2q_3 + q_2 - (1 - \mu) \frac{q_2}{((q_1 + \mu)^2 + q_2^2)^{\frac{3}{2}}} - \mu \frac{q_2}{((q_1 - 1 + \mu)^2 + q_2^2)^{\frac{3}{2}}} \end{pmatrix},$$

$$F_1(q) = \frac{\partial}{\partial q_3}, \quad F_2(q) = \frac{\partial}{\partial q_4}$$

Objective : minimizing transfer time between *geostationary* orbit \mathcal{O}_G and a *circular parking* orbit \mathcal{O}_L around the Moon when *low-thrust* is applied.

→ solve the the *optimal control problem*

$$\left\{ \begin{array}{l} \dot{q} = F_0(q) + \epsilon(F_1(q)u_1 + F_2(q)u_2), \epsilon > 0 \\ \min_{u(\cdot) \in B_{\mathbb{R}^2}(0,1)} \int_{t_0}^{t_f} dt \\ q(0) \in \mathcal{O}_G, \quad q(t_f) \in \mathcal{O}_L. \end{array} \right.$$

where $\epsilon =$ *bound* on the control = *maximum* thrust allowed

This optimal control problem *can not* be solved analytically

- ▶ Highly non-linear
- ▶ Singularities

→ *Approximate* low-thrust optimal solutions

- ▶ Apply *Pontryagin's Maximum Principle* (necessary conditions)
- ▶ Turn the *pseudo*-Hamiltonian system into a *true* Hamiltonian system (Implicit function Theorem)
- ▶ Use *shooting method* to compute *extremal curves* of the problem
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→ compute the *first conjugate time* along each extremal and verify that it is *greater* than the *transfer time*

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⇒ one can use a *Newton* type algorithm.

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2. One chooses a *discretization* $0 = \lambda_0, \lambda_1, \dots, \lambda_N = 1$ such that the shooting function is solved *iteratively* at λ_{i+1} from λ_i .

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1. Setting $\lambda = 0$, one computes the *extremal* $z(t)$ on $[0, t_f]$ starting from $z(0) = (q_0, p_0^0)$ using a simple *shooting* method.
2. One chooses a *discretization* $0 = \lambda_0, \lambda_1, \dots, \lambda_N = 1$ such that the shooting function is solved *iteratively* at λ_{i+1} from λ_i .
3. One builds up a *sequence* p_0^0, \dots, p_0^N with zeros of shooting functions $E_{\lambda_0}, \dots, E_{\lambda_N}$.

Problem : To ensure the convergence of the Newton method, we need a *precise guess* for the initial condition p_0 we are searching.

Idea : Consider H_r as the element H_1 of a family $(H_\lambda)_{\lambda \in [0,1]}$ of *smooth Hamiltonians*.

\Rightarrow build a *one-parameter family* $(E_\lambda)_{\lambda \in [0,1]}$ of shooting functions such that the shooting function associated with E_0 is *easy* to solve.

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4. p_0^N is the zero that we wanted to determine.

Theorem : For each λ , the exponential mapping \exp_{x_0, t_f}^λ is of textitmaximal rank if and only if the point $x_1 = \exp_{x_0, t_f}^\lambda(p(0))$ is *non-conjugate* to x_0 . Moreover, solutions of the parametrized shooting equation contain a *smooth curve* which can be parametrized by λ and the derivative E'_λ can be computed integrating the Jacobi equation.

\Rightarrow *convergence* of the smooth continuation is *guaranteed* if there is *no conjugate point* along any extremal curve z_{λ_f} .

Time-minimal problem $\min_{u \in B_{\mathbb{R}^2}(0, \epsilon)} \int_{t_0}^{t_f} dt$

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Maximization condition

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\Rightarrow if $(H_1, H_2) \neq 0$, we have

$$u_i = \frac{H_i}{\sqrt{H_1^2 + H_2^2}}, \text{ with } H_i(q, p) = \langle p, F_i(q) \rangle$$

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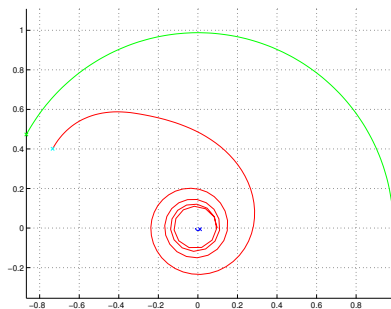
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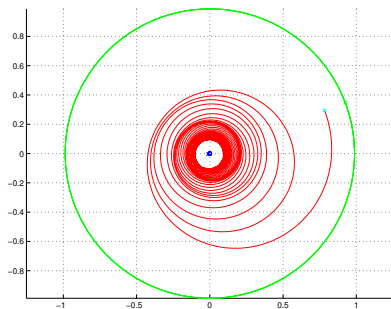
Continuation on ϵ (maximal thrust)

Compute Earth- L_1 trajectories prior to Earth-*Moon* trajectories



(a) Fixed frame

Figure: Thrust=1N. Red curve : time-minimal transfer to L_1 . Green curve : orbit of the Moon.



(a) Fixed frame

Figure: Thrust=0.08N. Red curve : time-minimal transfer to L_1 . Green curve : orbit of the Moon.

Earth-Moon time-minimal trajectories

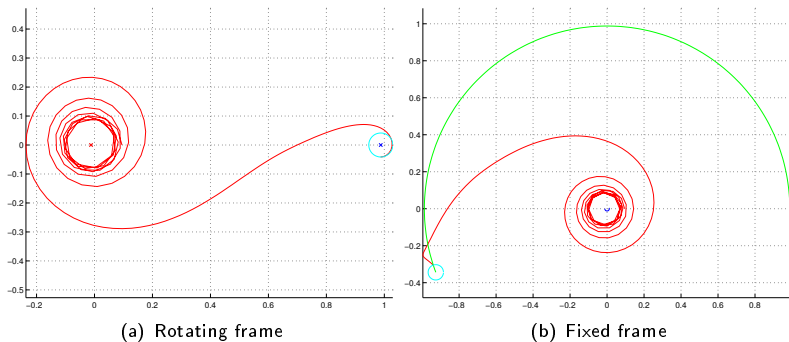


Figure: Thrust=1. Red curve : time-minimal transfer to a circular orbit around the Moon. Green curve : orbit of the Moon. Blue curve : circular parking orbit around the Moon

Earth-Moon time-minimal trajectories

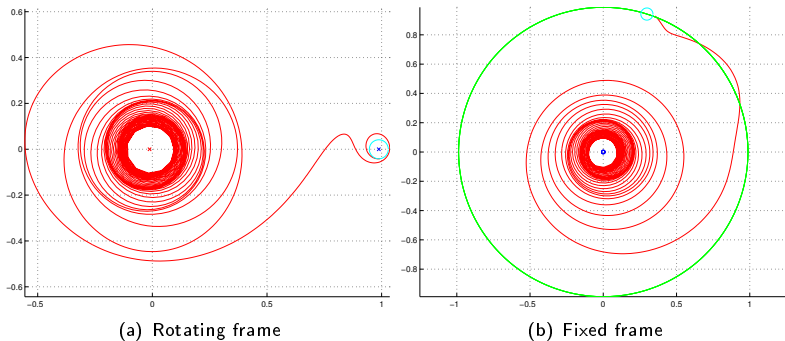


Figure: Thrust=0.08. Red curve : time-minimal transfer to a circular orbit around the Moon. Green curve : orbit of the Moon. Blue curve : circular parking orbit around the Moon

Objective : minimizing *energy-cost* of a transfer between *geostationary* orbit \mathcal{O}_G and a *circular parking* orbit \mathcal{O}_L around the Moon.

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→ minimizing the *L^2 -cost* of the control u along the transfer

→ solve the the *optimal control problem*

$$\left\{ \begin{array}{l} \dot{q} = F_0(q) + F_1(q)u_1 + F_2(q)u_2 \\ \min_{u(\cdot) \in \mathbb{R}^2} \int_{t_0}^{t_f} u_1^2 + u_2^2 dt \\ q(0) \in \mathcal{O}_G, \quad q(t_f) \in \mathcal{O}_L. \end{array} \right.$$

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Continuation on μ (mass ratio)

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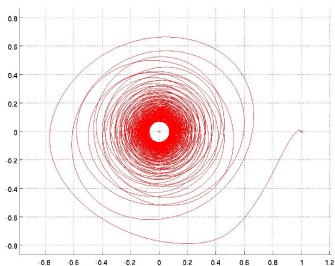
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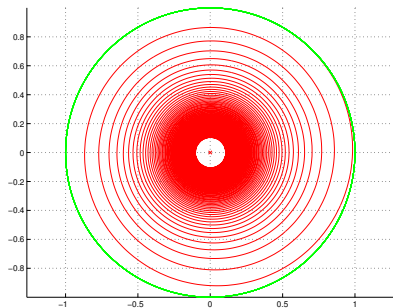
Continuation on μ (mass ratio)

Compute Earth- L_1 trajectories prior to Earth-*Moon* trajectories

Earth- L_1 energy minimal trajectories



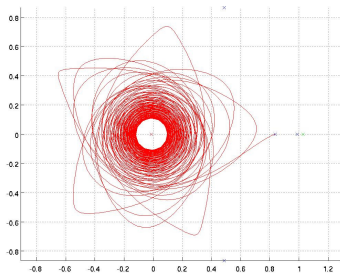
(a) Rotating frame



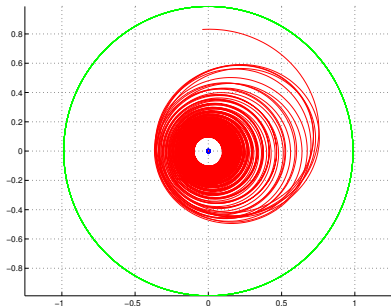
(b) Fixed frame

Figure: $\mu=0$. Red curve : energy-minimal transfer to L_1 . Green curve : orbit of the Moon.

Earth- L_1 energy minimal trajectories

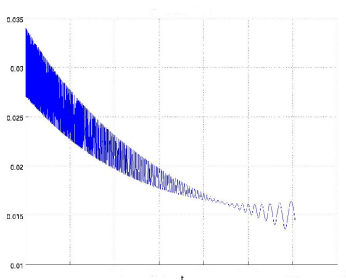


(a) Rotating frame

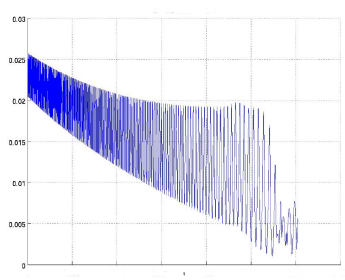


(b) Fixed frame

Figure: $\mu = 1.2153e - 2$. Red curve : energy-minimal transfer to L_1 . Green curve : orbit of the Moon.



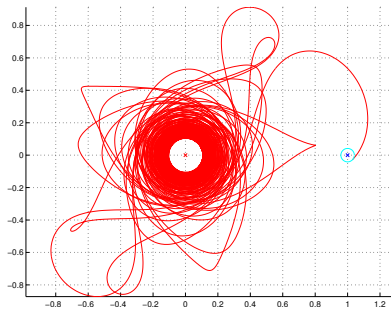
(a) $\mu=0$



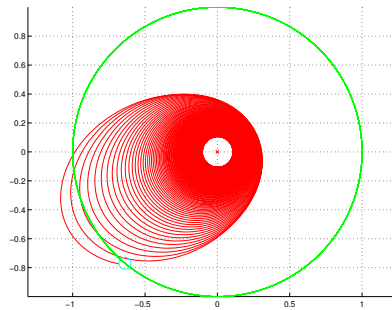
(b) $\mu = 1.2153e - 2$

Figure: Control norms

Earth-Moon energy minimal trajectories



(a) Rotating frame



(b) Fixed frame

Figure: $\mu=0$. Red curve : energy-minimal transfer to a circular orbit around the Moon. Green curve : orbit of the Moon. Blue curve : circular parking orbit around the Moon

Earth-Moon energy minimal trajectories

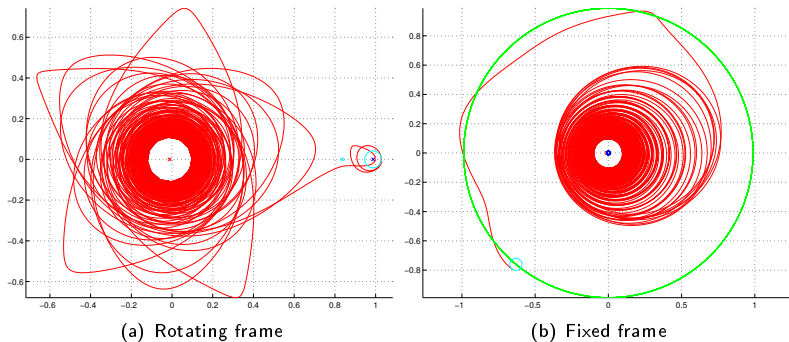
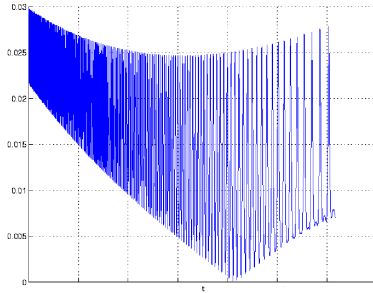
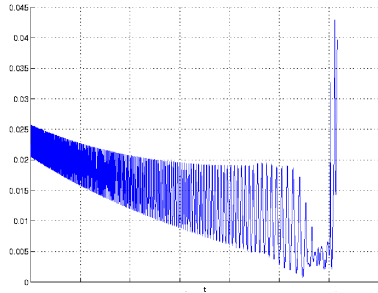


Figure: $\mu = 1.2153e - 2$. Red curve : energy-minimal transfer to a circular orbit around the Moon. Green curve : orbit of the Moon. Blue curve : circular parking orbit around the Moon



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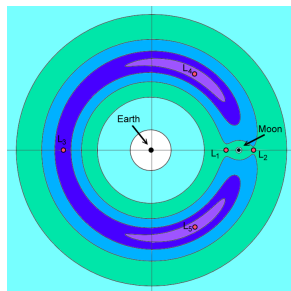


(b) $\mu = 1.2153e - 2$

Figure: Control norms

Szebehely [1967] *Circular restricted three-body problem*

$$\begin{cases} \ddot{x} = 2\dot{y} - \frac{\partial V}{\partial x} \\ \ddot{y} = -2\dot{x} - \frac{\partial V}{\partial y} \\ \ddot{z} = -\frac{\partial V}{\partial z} \end{cases}$$



where V is the potential energy function :

$$-V = \frac{x^2 + y^2}{2} + \frac{1 - \mu}{\rho_1} + \frac{\mu}{\rho_2} + \frac{\mu(1 - \mu)}{2}$$

and

$$\rho_1 = \sqrt{(x + \mu)^2 + y^2 + z^2}$$

$$\rho_2 = \sqrt{(x - 1 + \mu)^2 + y^2 + z^2}$$

Simó *et al.* [1995] *Circular restricted four-body problem*

Sun at $(r_s \cos \theta, r_s \sin \theta)$ with mass μ_s

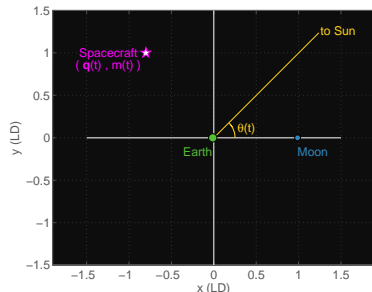
$$\begin{cases} \ddot{x} = 2\dot{y} - \frac{\partial V_4}{\partial x} \\ \ddot{y} = -2\dot{x} - \frac{\partial V_4}{\partial y} \\ \ddot{z} = -\frac{\partial V_4}{\partial z} \end{cases}$$

where

$$V_4 = V - V_s, \quad V_s(t) = \frac{\mu_s}{\rho_s} - \frac{\mu_s}{r_s^2} (x \cos \theta + y \sin \theta)$$

$$\rho_s = \sqrt{(x - r_s \cos \theta)^2 + (y - r_s \sin \theta)^2 + z^2}, \quad \mu_s = 329012.5, \quad r_s = 389.2,$$

$$\dot{\theta} = \omega_s = -0.925 \quad \rightarrow \quad \theta(t) = \omega_s t + \theta_0$$



State variables :

$$\mathbf{X}(t) = \begin{pmatrix} \mathbf{q}(t) \\ m(t) \end{pmatrix} = \begin{pmatrix} \mathbf{s}(t) \\ \mathbf{v}(t) \\ m(t) \end{pmatrix}$$

Uncontrolled system :

$$\dot{\mathbf{X}} = \mathbf{F}_0(\mathbf{X}) = \begin{pmatrix} \dot{x} \\ \dot{y} \\ \dot{z} \\ 2\dot{y} + x - (1 - \mu) \frac{x + \mu}{\rho_1^3} - \mu \frac{x - 1 + \mu}{\rho_2^3} - \mu_s \frac{x - r_s \cos \theta}{\rho_s^3} - \frac{\mu_s}{r_s^2} \cos \theta \\ -2\dot{x} + y - (1 - \mu) \frac{y}{\rho_1^3} - \mu \frac{y}{\rho_2^3} - \mu_s \frac{y - r_s \sin \theta}{\rho_s^3} - \frac{\mu_s}{r_s^2} \sin \theta \\ -(1 - \mu) \frac{z}{\rho_1^3} - \mu \frac{z}{\rho_2^3} - \mu_s \frac{z}{\rho_s^3} \\ 0 \end{pmatrix}$$

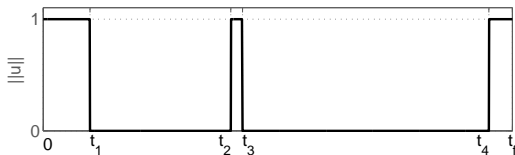
Control variables :

$$\mathbf{u}(t) = \begin{pmatrix} u_1(t) \\ u_2(t) \\ u_3(t) \end{pmatrix}, \quad \|\mathbf{u}\| \leq 1, \quad \mathbf{u} \in \bar{\mathcal{U}}$$

Set of admissible controls :

$$\bar{\mathcal{U}} = \left\{ \mathbf{u} : \mathbb{R} \rightarrow B_{\mathbb{R}^3}(0, 1) : \begin{array}{l} \mathbf{u} \text{ measurable, } \|\mathbf{u}(t)\| = \bar{\alpha}(t) \text{ for a sequence} \\ \text{of times } 0 \leq t_1 \leq t_2 \leq t_3 \leq t_4 \leq t_f \end{array} \right\}$$

$$\bar{\alpha}(t) = \left\{ \begin{array}{ll} 1, & t \in [0, t_1] \cup [t_2, t_3] \cup [t_4, t_f] \\ 0, & \text{otherwise} \end{array} \right\}$$



$t_i, i = 1..4$ are called *switching times*

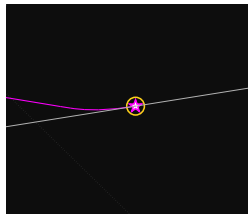
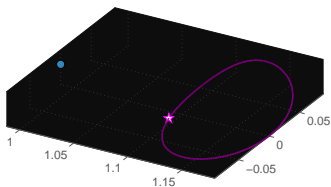
Minimal fuel problem

We can now state the minimal cost problem :

$$(\star)_D \begin{cases} \min_{\mathbf{u} \in \bar{\mathcal{U}}} \int_0^{t_f} \|\mathbf{u}(t)\| dt \\ \dot{\mathbf{X}} = \mathbf{F}_0(\mathbf{X}) + \frac{T_{max}}{m} \sum_{i=1}^3 u_i \mathbf{F}_i - T_{max} \beta \|\mathbf{u}\| \mathbf{F}_4 \\ \mathbf{q}(0) = \mathbf{q}_{dprt}, \quad m(0) = m_0 \\ \mathbf{q}(t_f) = \mathbf{q}_{rdvz}, \quad \theta(t_f) = \theta_{rdvz} \end{cases}$$

D is the *problem data*

$$D = (t_f, \mathbf{q}_{dprt}, \mathbf{q}_{rdvz}, \theta_{rdvz})$$



(Pontryagin [1962]) Let (\mathbf{X}, \mathbf{u}) be an optimal pair for $(\star)_D$. Then there exists an absolutely continuous function $\mathbf{P}(\cdot)$, called the costate vector, and a constant $p_0 \leq 0$, $(\mathbf{P}, p_0) \neq (\mathbf{0}, 0)$, such that for a.e. t we have :

Pseudo-Hamiltonian Equations :

$$\dot{\mathbf{P}} = -\frac{\partial H}{\partial \mathbf{X}}, \quad \dot{\mathbf{X}} = \frac{\partial H}{\partial \mathbf{P}}$$

where $H(t, \mathbf{P}, \mathbf{X}, \mathbf{u}) = p_0 \|\mathbf{u}\| + \langle \mathbf{P}, \dot{\mathbf{X}} \rangle$.

Maximization Condition :

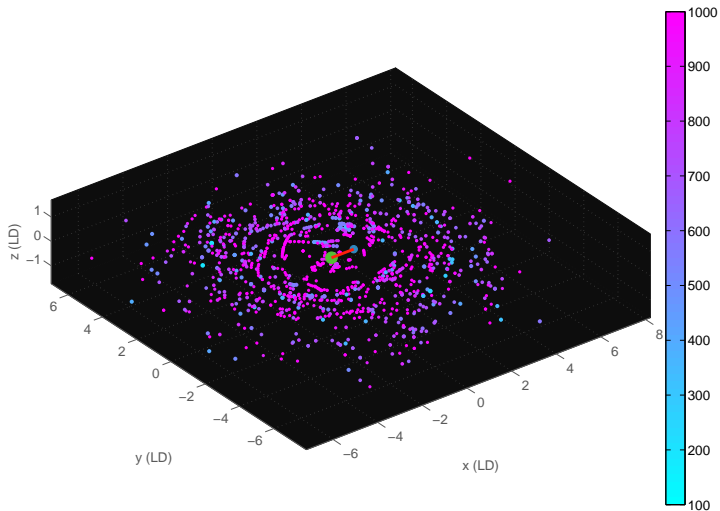
$$H(t, \mathbf{P}(t), \mathbf{X}(t), \mathbf{u}(t)) = \max_{\mathbf{v} \in B_{\mathbb{R}^3}(0,1)} H(t, \mathbf{P}(t), \mathbf{X}(t), \mathbf{v})$$

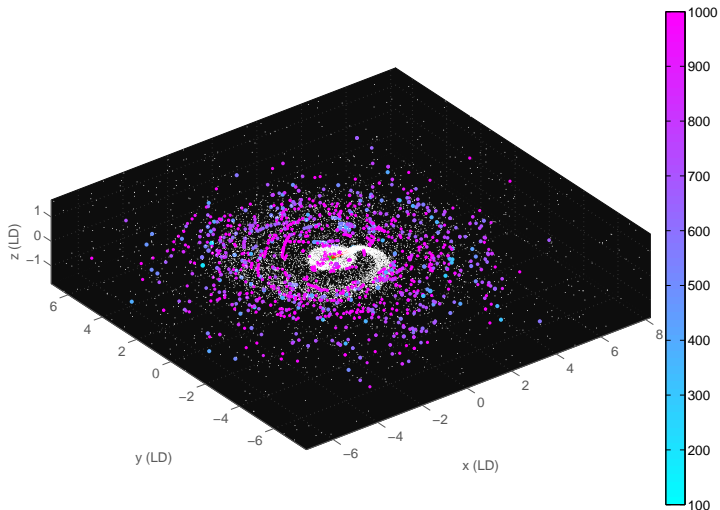
Transversality Condition :

$$\mathbf{P}(0) \perp T_{\mathbf{X}(0)}M_0, \quad \mathbf{P}(t_f) \perp T_{\mathbf{X}(t_f)}M_f$$

A solution $(\mathbf{P}, \mathbf{X}, \mathbf{u})$ is called an *extremal*.

ΔV at rendezvous (m/s)

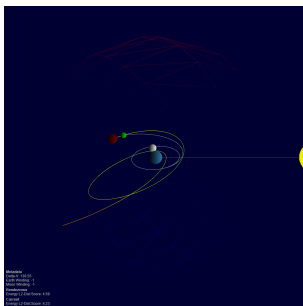




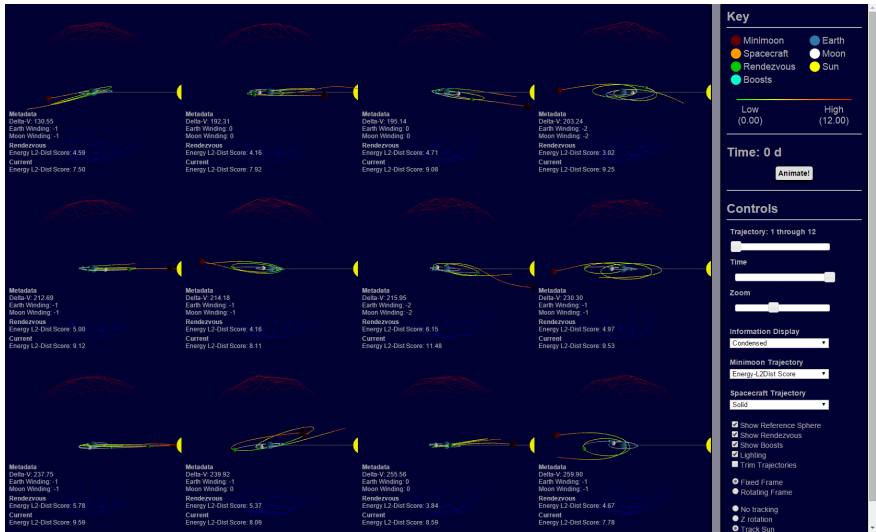
Further questions :

- ▶ Why were some transfers better than others?
- ▶ What characterizes a good candidate asteroid?
- ▶ What some general features that allow low cost transfers?

This prompted exploratory work using visualization and statistical analysis.



Visualization demo



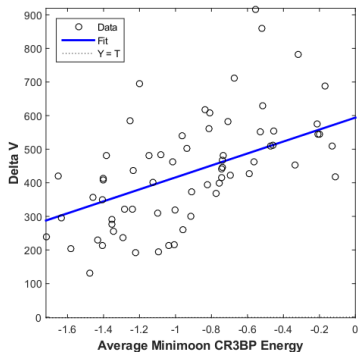
Potential predictors based on visualization and talks with industry :

- ▶ Average energy
- ▶ Average velocity
- ▶ Planarity
- ▶ Lunar planarity
- ▶ Z-displacement
- ▶ Distance from barycenter
- ▶ Eccentricity
- ▶ Average curvature

The most important factors were :

- ▶ Average energy
- ▶ Lunar planarity
- ▶ Variance of distance from barycenter

[include other graphs here]



The following are fair predictors of good transfers :

- ▶ Low average energy relative to the departure energy (from earth-moon L_2)
- ▶ Minimoons trajectories that lie mostly in the lunar plane
- ▶ Roughly circular geocentric orbits

For mission design, this implies that parking an array of spacecraft at various energy levels could be a viable solution to maximize low cost interceptions. Furthermore, prioritizing minimoons with roughly moon-like orbits could help reduce fuel costs. This has implications for missions like NASA's ARM as well.