

# Optimal control of Reusable Launch Vehicles: an indirect shooting approach

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THE FRENCH AEROSPACE LAB

# Outline

## 1 Introduction

Optimal Control & Applications at ONERA  
Real-time Optimal Control

## 2 Return trajectory of Reusable Launch Vehicles (CNES-ONERA project)

RLV Model  
Pontryagin's Maximum Principle  
Homotopy method  
Numerical results: Toss-back model  
Optimal Guidance

## 3 Glider model for RLVs (CNES-ONERA Post-Doctoral Fellowship)

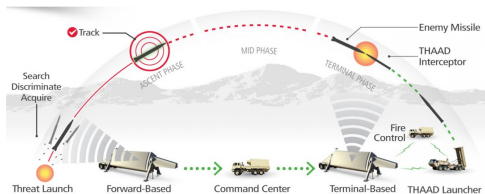
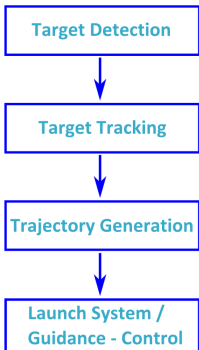
From toss-back recovery to glider model  
Numerical results: Glider model

## 4 Conclusion



# Intercept Problem (Riccardo BONALI's Ph.D. thesis)

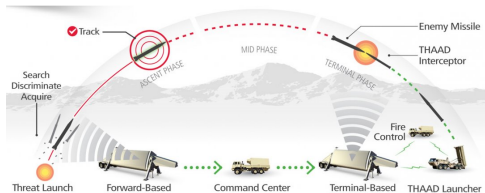
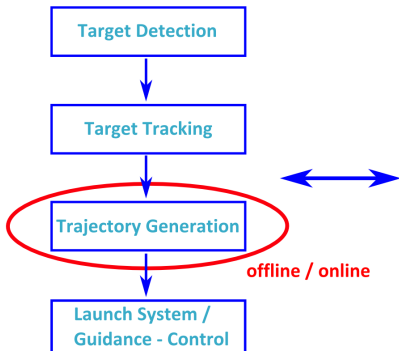
- **Objective** : Intercept the target, need high precision and high terminal velocity.
- **Difficulty** : Missiles can fly at high altitude (20-30 km), difficult to control with aerodynamic actuators due to altitude dependence of air density.



THAAD Interceptor [<https://www.mda.mil>]

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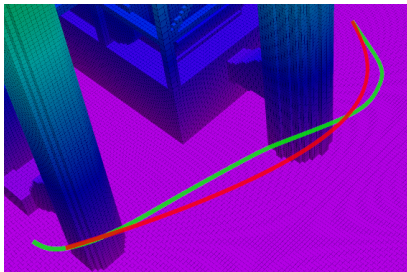
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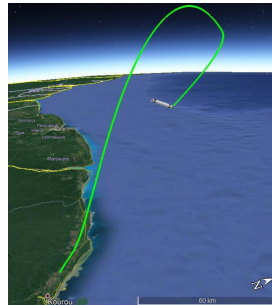
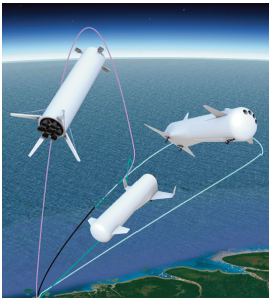
# Motion Planning for Unmanned Aerial Vehicles

- **Objective** : Motion planning of Aerial Robots in cluttered environments.
- **Difficulty** : Dynamic environments (moving obstacles, etc.), obstacle modeling.



# Reusable Launch Vehicles

- **Objective** : Return and landing of the first stage of space launchers.
- **Difficulty** : Highly constrained problem (aerodynamic and safety constraints), may need re-ignition of rocket engines.



# Real-time Optimal Control

## Challenge

Compute optimal trajectories in **real time**, by using **embedded computers**, to make the vehicle adapt its trajectory to **changes of the scenario**.

- Global approaches (e.g. HJB):

### Pros:

- Global optimum
- Do not require any initial guess

### Cons:

- Time consuming for problems of high dimension
- Cannot be used for real time computations

- Local approaches



# Real-time Optimal Control

## Challenge

Compute optimal trajectories in **real time**, by using **embedded computers**, to make the vehicle adapt its trajectory to **changes of the scenario**.

### ① Explicit Feedback Laws and Direct Methods (e.g. SQP):

#### Pros:

- Easy to implement
- Robustness

#### Cons:

- Lack of precision
- Expensive or sub-optimal

### ② Indirect Methods:

#### Pros:

- High precision
- **Fast convergence**

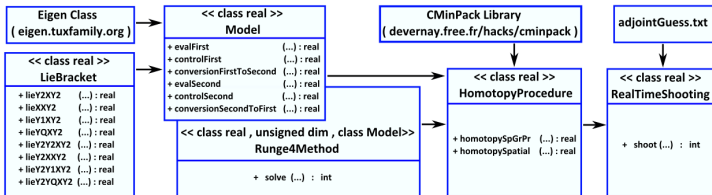
#### Cons:

- Complex analysis
- **Hard to initialize**

# A library for solving real-time optimal control problems

## SOCP (Shooting for Optimal Control Problems)

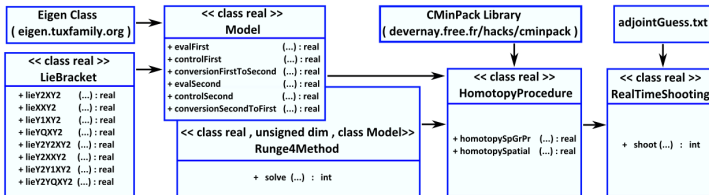
- **Indirect methods** for high precision.



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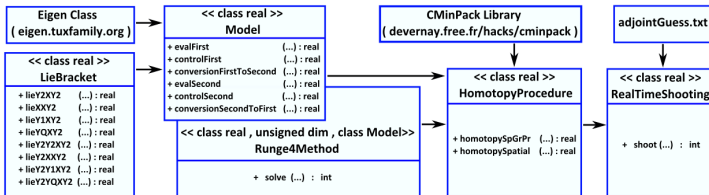
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- **Multiple shooting** for numerical robustness.



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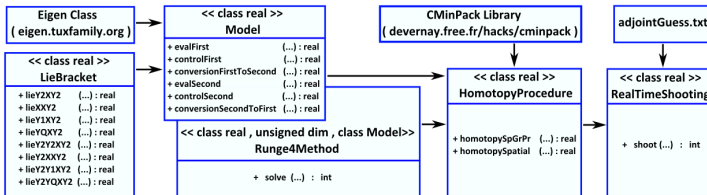
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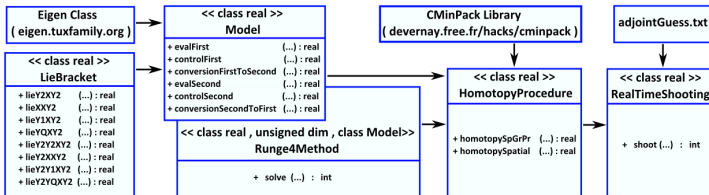
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## SOCP (Shooting for Optimal Control Problems)

- **Indirect methods** for high precision.
- **Multiple shooting** for numerical robustness.
- **Homotopy methods** for initialization problems.
- **Parallel computing** for computation time.
- **C++ library** for best performance and embedded solutions.



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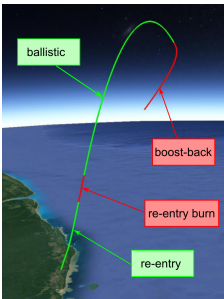


Figure: Toss-back model

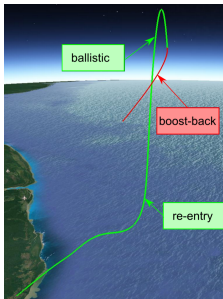


Figure: Glider model



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## RLV model

$$\left\{ \begin{array}{l}
 \min \quad -m(t_f) \\
 \dot{\mathbf{r}} = \mathbf{v} \quad , \quad (\mathbf{r}, \mathbf{v}, \mathbf{u}) \in \mathbb{R}^9 \quad , \quad \|\mathbf{r}\| = 1 \\
 \dot{\mathbf{v}} = \frac{L}{m} \mathbf{k}_a - \frac{D}{m} \frac{\mathbf{v}_a}{\|\mathbf{v}_a\|} - \mathbf{g} + \frac{T_m}{m} \gamma \mathbf{u} - 2\boldsymbol{\Omega} \times \mathbf{v} - \boldsymbol{\Omega} \times (\boldsymbol{\Omega} \times \mathbf{r}) \\
 \dot{m} = -q_m \gamma \quad , \quad 0 \leq \gamma \leq 1 \\
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# Optimal Control with pure state constraints and mixed constraints

## Optimal Control Problem

$$\left\{ \begin{array}{l} \min \int_0^{t_f} f^0(X, U, t) dt + g(X(t_f), t_f) \\ \dot{X} = f(X, U, t) \\ X(x) = X_0, \mathbf{r}(t_f) = \mathbf{r}_f \\ t_f, m(t_f) \text{ and } \mathbf{v}(t_f) \text{ are free,} \\ \forall t, c_p(X(t)) \leq 0, \\ \forall t, c_m(X(t), U(t)) \leq 0. \end{array} \right.$$

where  $X = (\mathbf{r}, \mathbf{v}, m)$  and  $U = (u, \gamma)$

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$$\forall t, c_p(X(t)) \leq 0,$$

↔ The dynamic pressure and the thermal flux

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↔ The angle-of-attack and the load factor

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# Pontryagin's Maximum Principle

$p = (p_r, p_v, p_m) \in \mathbb{R}^3 \times \mathbb{R}^3 \times \mathbb{R}$  denotes the co-state vector

Hamiltonians

The extended Hamiltonian including the constraints

$$\tilde{H} = H + \alpha_p c_p(X) + \alpha_m c_m(X, U)$$

where the Hamiltonian  $H = p \cdot f(X, U, t) + p^0 r^0(X, u, t)$



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$$\dot{p} = -\frac{\partial \tilde{H}}{\partial X}$$

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$$H(t, X, p, p^0, U) = \min_{V \in \mathcal{U}} H(t, X, p, p^0, V)$$

- If  $f$  and  $f^0$  do not explicitly depend on  $t$ ,  $H$  is constant.

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Since  $m(t_f)$  and  $v(t_f)$  are free,

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## Continuity / Jump conditions

$c_m$  is assumed regular (**MF-CQ**).

- $\alpha_p$  and  $\alpha_m$  are continuous and  $\forall t, \alpha_p c_p(X) = \alpha_m c_m(X, U) = 0$

At every contact point  $t_i$  with the boundary of the pure state constraint,

- $\tilde{H}(t_i^+) = \tilde{H}(t_i^-)$  and  $\exists \nu_i, p(t_i^+) = p(t_i^-) - \nu_i \frac{\partial c_p}{\partial X}(X(t_i))$ .

# Structure of the control

$J(\gamma, u)$  denotes the expression in the Hamiltonian that depends on the control:

$$H = J(\gamma, u) + \dots$$

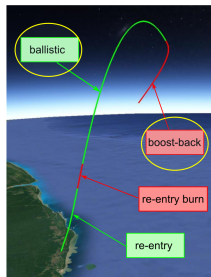
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$$\implies J(\gamma, u) = \gamma \underbrace{\left( 1 - q_m p_m + \frac{T_m}{m} \langle p_v, u \rangle \right)}_{\psi}$$



**Figure:** Boost-Back and ballistic phases

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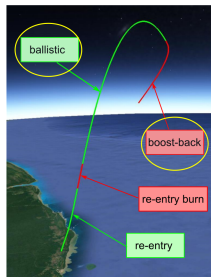
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**Minimization condition**

$$\implies u = - \frac{p_v}{\|p_v\|}$$

- When the switching function  $\Psi < 0$ ,  $\gamma = 1$ .
- When the switching function  $\Psi > 0$ ,  $\gamma = 0$ .



**Figure:** Boost-Back and ballistic phases

## Re-entry burn

The previous hypothesis is irrelevant during the re-entry burn. It may complicate the explicit computation of the optimal control input.

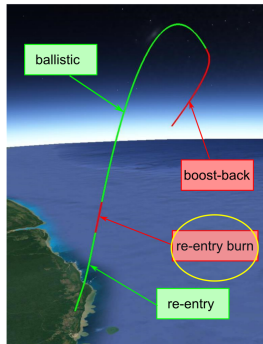


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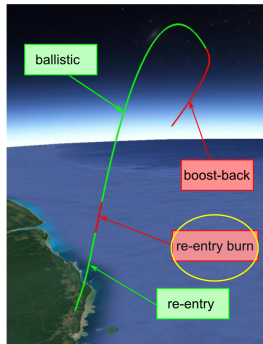


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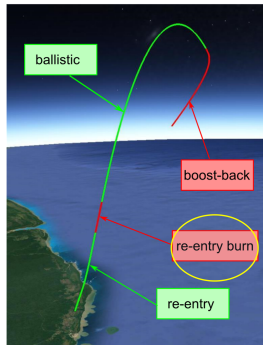


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### Solution

- A constant approximation is taken for the value of  $\gamma$  along this phase (e.g.  $\gamma = 1/3$ ).

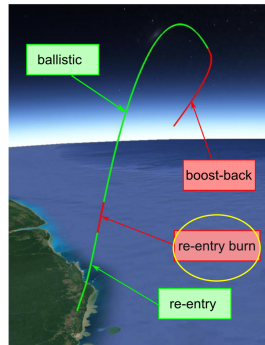


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- Ballistic phase, i.e.  $\gamma = 0$
- The aerodynamic forces are considered.
- The expression depending on the control reads

$$J = \frac{L}{m} \langle \mathbf{u}, \mathbf{v} \rangle \langle \mathbf{u}, \mathbf{p}_v \rangle$$

- $\mathbf{u}$  which minimizes  $J$  can be explicitly computed.

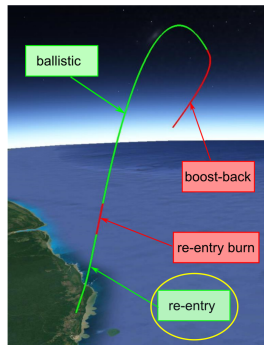


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$$J = \frac{L}{m} \langle \mathbf{u}, \mathbf{v} \rangle \langle \mathbf{u}, \mathbf{p}_v \rangle$$

- $\mathbf{u}$  which minimizes  $J$  can be explicitly computed.

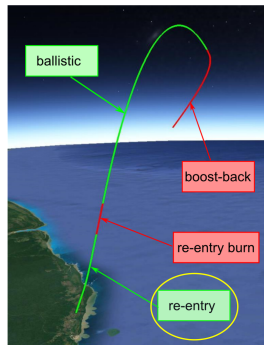


Figure: Atmospheric re-entry phase

## More details:

E. Brendel, B. Hérisse and E. Bourgeois, "Optimal guidance for Toss Back concepts of Reusable Launch Vehicles", 8th European Conference for Aeronautics AND Aerospace Sciences (EUCASS), 2019.

# Homotopy method

- Solve a simplified problem for which an explicit solution can be found.

## RLV model

$$\left\{ \begin{array}{l}
 \min \quad - \quad m(t_f) \\
 \dot{\mathbf{r}} = \mathbf{v} \quad , \quad (\mathbf{r}, \mathbf{v}, \mathbf{u}) \in \mathbb{R}^9, \quad \|\mathbf{r}\| = 1 \\
 \dot{\mathbf{v}} = \frac{L}{m} \mathbf{k}_a - \frac{D}{m} \frac{\mathbf{v}_a}{\|\mathbf{v}_a\|} - \mathbf{g} + \frac{T_m}{m} \gamma \mathbf{u} - 2\boldsymbol{\Omega} \times \mathbf{v} - \boldsymbol{\Omega} \times (\boldsymbol{\Omega} \times \mathbf{r}) \\
 \dot{m} = - \quad q_m \gamma \quad , \quad 0 \leq \gamma \leq 1 \\
 (\mathbf{r}, \mathbf{v}, m)(0) = (\mathbf{r}_0, \mathbf{v}_0, m_0) \quad , \quad (\mathbf{r}, \mathbf{v})(t_f) = (\mathbf{r}_f, \mathbf{v}_f) \\
 \alpha \leq \alpha_{\max} \quad \text{if} \quad \|\mathbf{r}\| \leq r_{\text{ref}} \quad , \quad Q \leq Q_{\max} \quad , \quad \Phi \leq \Phi_{\max} \quad , \quad \eta_x \leq \eta_{x,\max} \quad , \quad \eta_z \leq \eta_{z,\max}
 \end{array} \right.$$

# Homotopy method

- Solve a simplified problem for which an explicit solution can be found.

RLV model (homotopy start with  $\mu_x = 0$  except  $\mu_{L_2} = 1$ )

$$\left\{ \begin{array}{l} \min \quad -\mu_{L_1} m(t_f) + \mu_{L_2} \int_0^{t_f} \gamma^2 dt \\ \dot{\mathbf{r}} = \mathbf{v} \quad , \quad (\mathbf{r}, \mathbf{v}, \mathbf{u}) \in \mathbb{R}^9, \quad \|\mathbf{r}\| = 1 \\ \dot{\mathbf{v}} = \mu_L \frac{L}{m} \mathbf{k}_a - \mu_D \frac{D}{m} \frac{\mathbf{v}_a}{\|\mathbf{v}_a\|} - \mu_g \mathbf{g} + \frac{T_m}{m} \gamma \mathbf{u} - 2\boldsymbol{\Omega} \times \mathbf{v} - \boldsymbol{\Omega} \times (\boldsymbol{\Omega} \times \mathbf{r}) \\ \dot{m} = -\mu_q q_m \gamma \quad , \quad 0 \leq \gamma \leq 1 \\ (\mathbf{r}, \mathbf{v}, m)(0) = (\mathbf{r}_0, \mathbf{v}_0, m_0) \quad , \quad (\mathbf{r}, \mathbf{v})(t_f) = (\mathbf{r}_f, \mathbf{v}_f) \\ \mu_\alpha \alpha \leq \alpha_{\max} \quad \text{if} \quad \|\mathbf{r}\| \leq r_{\text{ref}} \quad , \quad \mu_Q Q \leq Q_{\max} \quad , \quad \mu_\Phi \Phi \leq \Phi_{\max} \quad , \quad \mu_\eta^x \eta_x \leq \eta_{x,\max} \quad , \quad \mu_\eta^z \eta_z \leq \eta_{z,\max} \end{array} \right.$$

# Homotopy method

- Solve a simplified problem for which an explicit solution can be found.  
 ↖ Here, a double integrator!

RLV model (homotopy start with  $\mu_x = 0$  except  $\mu_{L_2} = 1$ )

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# Homotopy method

- Solve a simplified problem for which an explicit solution can be found.
  - ↖ Here, a double integrator!
- Initialize the co-state vector of a multiple shooting algorithm.

RLV model (homotopy start with  $\mu_x = 0$  except  $\mu_{L_2} = 1$ )

$$\left\{ \begin{array}{l} \min \quad -\mu_{L_1} m(t_f) + \mu_{L_2} \int_0^{t_f} \gamma^2 dt \\ \dot{\mathbf{r}} = \mathbf{v} \quad , \quad (\mathbf{r}, \mathbf{v}, \mathbf{u}) \in \mathbb{R}^9, \quad \|\mathbf{r}\| = 1 \\ \dot{\mathbf{v}} = \mu_L \frac{L}{m} \mathbf{k}_a - \mu_D \frac{D}{m} \frac{\mathbf{v}_a}{\|\mathbf{v}_a\|} - \mu_g \mathbf{g} + \frac{T_m}{m} \gamma \mathbf{u} - 2\boldsymbol{\Omega} \times \mathbf{v} - \boldsymbol{\Omega} \times (\boldsymbol{\Omega} \times \mathbf{r}) \\ \dot{m} = -\mu_q q_m \gamma \quad , \quad 0 \leq \gamma \leq 1 \\ (\mathbf{r}, \mathbf{v}, m)(0) = (\mathbf{r}_0, \mathbf{v}_0, m_0) \quad , \quad (\mathbf{r}, \mathbf{v})(t_f) = (\mathbf{r}_f, \mathbf{v}_f) \\ \mu_\alpha \alpha \leq \alpha_{\max} \quad \text{if} \quad \|\mathbf{r}\| \leq r_{\text{ref}} \quad , \quad \mu_Q Q \leq Q_{\max} \quad , \quad \mu_\Phi \Phi \leq \Phi_{\max} \quad , \quad \mu_\eta^x \eta_x \leq \eta_{x,\max} \quad , \quad \mu_\eta^z \eta_z \leq \eta_{z,\max} \end{array} \right.$$

# Homotopy method

- Solve a simplified problem for which an explicit solution can be found.
  - ↖ Here, a double integrator!
- Initialize the co-state vector of a multiple shooting algorithm.
- Then, every neglected term is gradually added until the problem being solved is the complete problem.

RLV model (homotopy start with  $\mu_x = 0$  except  $\mu_{L_2} = 1$ )

$$\left\{ \begin{array}{l} \min \quad -\mu_{L_1} m(t_f) + \mu_{L_2} \int_0^{t_f} \gamma^2 dt \\ \dot{\mathbf{r}} = \mathbf{v} \quad , \quad (\mathbf{r}, \mathbf{v}, \mathbf{u}) \in \mathbb{R}^9, \quad \|\mathbf{r}\| = 1 \\ \dot{\mathbf{v}} = \mu_L \frac{L}{m} \mathbf{k}_a - \mu_D \frac{D}{m} \frac{\mathbf{v}_a}{\|\mathbf{v}_a\|} - \mu_g \mathbf{g} + \frac{T_m}{m} \gamma \mathbf{u} - 2\boldsymbol{\Omega} \times \mathbf{v} - \boldsymbol{\Omega} \times (\boldsymbol{\Omega} \times \mathbf{r}) \\ \dot{m} = -\mu_q q_m \gamma \quad , \quad 0 \leq \gamma \leq 1 \\ (\mathbf{r}, \mathbf{v}, m)(0) = (\mathbf{r}_0, \mathbf{v}_0, m_0) \quad , \quad (\mathbf{r}, \mathbf{v})(t_f) = (\mathbf{r}_f, \mathbf{v}_f) \\ \mu_\alpha \alpha \leq \alpha_{\max} \quad \text{if} \quad \|\mathbf{r}\| \leq r_{\text{ref}} \quad , \quad \mu_Q Q \leq Q_{\max} \quad , \quad \mu_\Phi \Phi \leq \Phi_{\max} \quad , \quad \mu_\eta^x \eta_x \leq \eta_{x,\max} \quad , \quad \mu_\eta^z \eta_z \leq \eta_{z,\max} \end{array} \right.$$

# Numerical results: Toss-back model

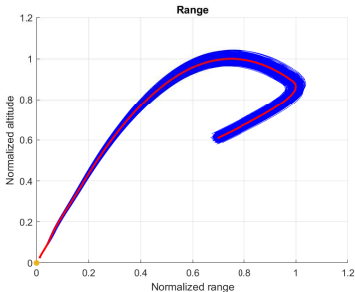
# Numerical results: Toss-back model

# Optimal Guidance algorithm

- The **offline** reference trajectory is used to initialize the guidance algorithm
- Recompute the optimal control **online** with an homotopy on the initial state
- **Evaluation**: with randomly scattered parameters, e.g. **initial state dispersion**, aerodynamic coefficients, maximal thrust, maximal mass flow rate...

# Optimal Guidance algorithm

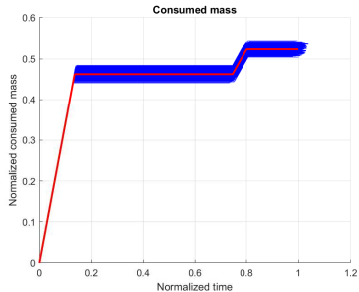
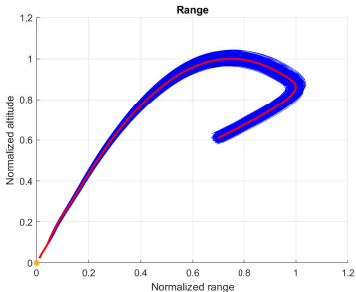
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**Satisfactory precision of the guidance algorithm:** The margin of error on final position is below 100m (at 3km of the landing platform)

# Optimal Guidance algorithm

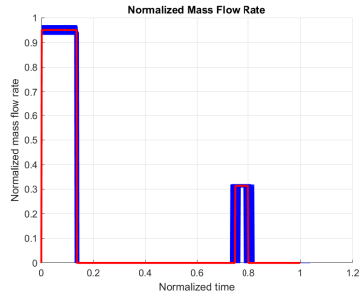
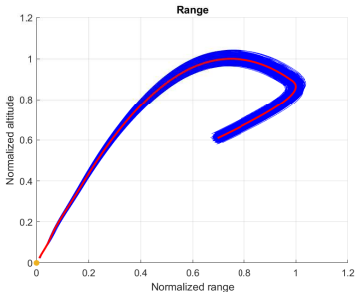
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# Optimal Guidance algorithm

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Optimal Control & Applications at ONERA  
Real-time Optimal Control

## 2 Return trajectory of Reusable Launch Vehicles (CNES-ONERA project)

RLV Model  
Pontryagin's Maximum Principle  
Homotopy method  
Numerical results: Toss-back model  
Optimal Guidance

## 3 Glider model for RLVs (CNES-ONERA Post-Doctoral Fellowship)

From toss-back recovery to glider model  
Numerical results: Glider model

## 4 Conclusion

# Toss-back vs Glider

- Similar models
- Vertical Takeoff Horizontal Landing
- The glider model uses a **larger L/D ratio**.

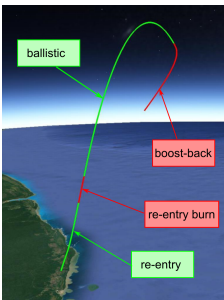


Figure: Tossback recovery

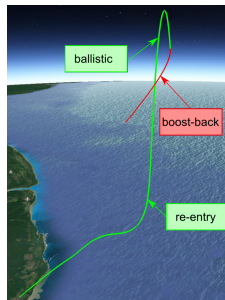


Figure: Glider model

# Numerical results: Glider model

- Homotopy on the L/D ratio

# Numerical results: Glider model

- Homotopy on the the dynamic pressure

# Numerical results: Glider model

- Saturation of the angle-of-attack

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# Conclusion

## Conclusion

- An efficient **C++ library** implementing shooting methods.
- A general homotopy scheme for solving RLVs problems.
- The algorithm can be adapted for on-line recalculations with a more realistic atmospheric model.

## Perspectives

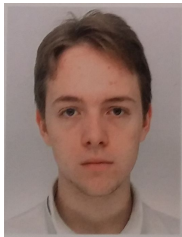
- Complete the study of the glider model (the load factor constraint, the optimal guidance, etc.)
- Compare with direct methods.
- Assess the flight-back approach.

# The SOCP Team

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