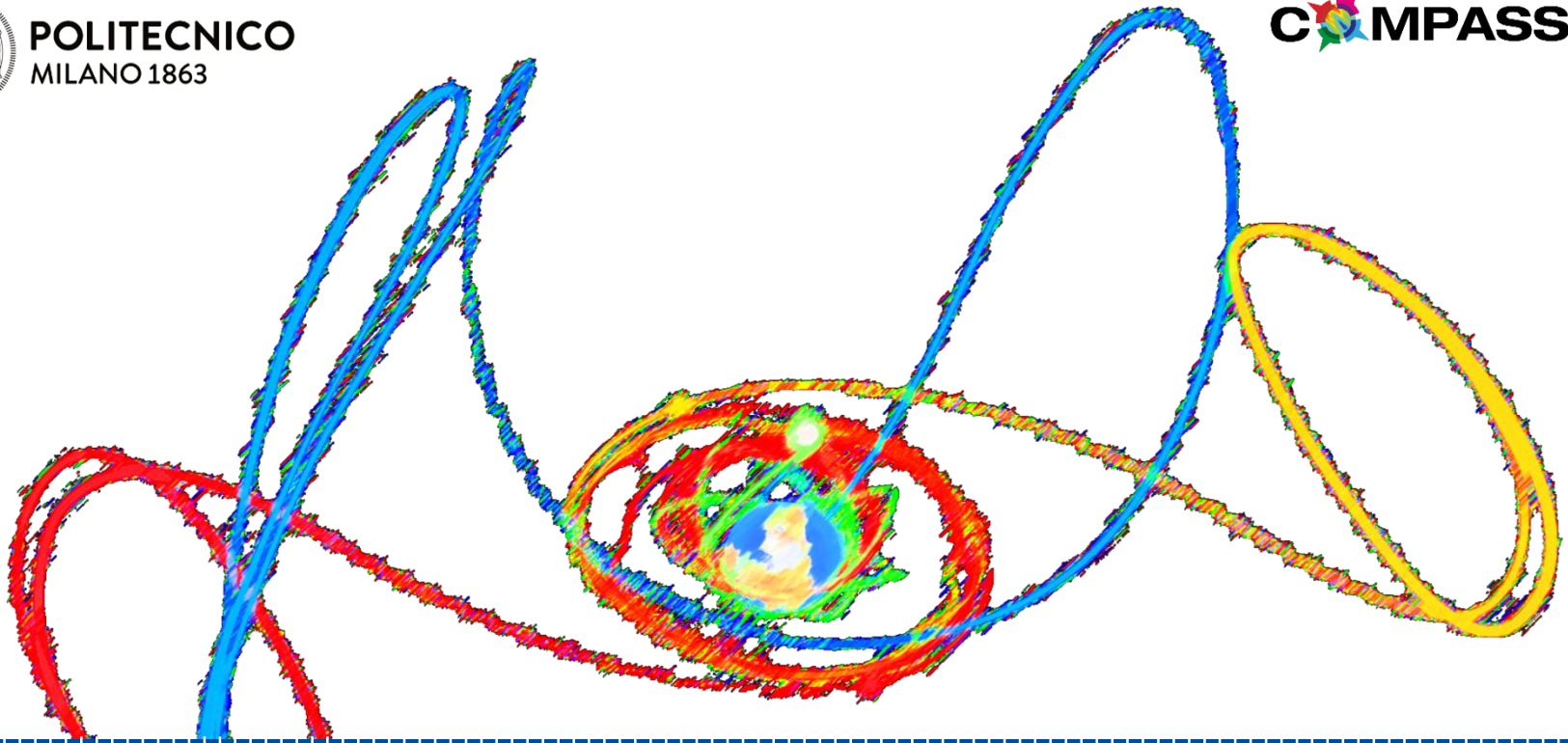




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# Planetary protection analysis for interplanetary missions

SDSM 2017 Summer School, San Martino al Cimino (VT)

Matteo Romano & Camilla Colombo

- Planetary protection
  - Motivations
  - Suite for Numerical Analysis of Planetary Protection
  - Explored ideas
- Numerical integration
  - Symplectic methods
  - Selection of methods and other techniques
  - Applications
- Sampling techniques
  - Line Sampling
  - Other ideas
  - Applications
- Conclusions and future work



# PLANETARY PROTECTION

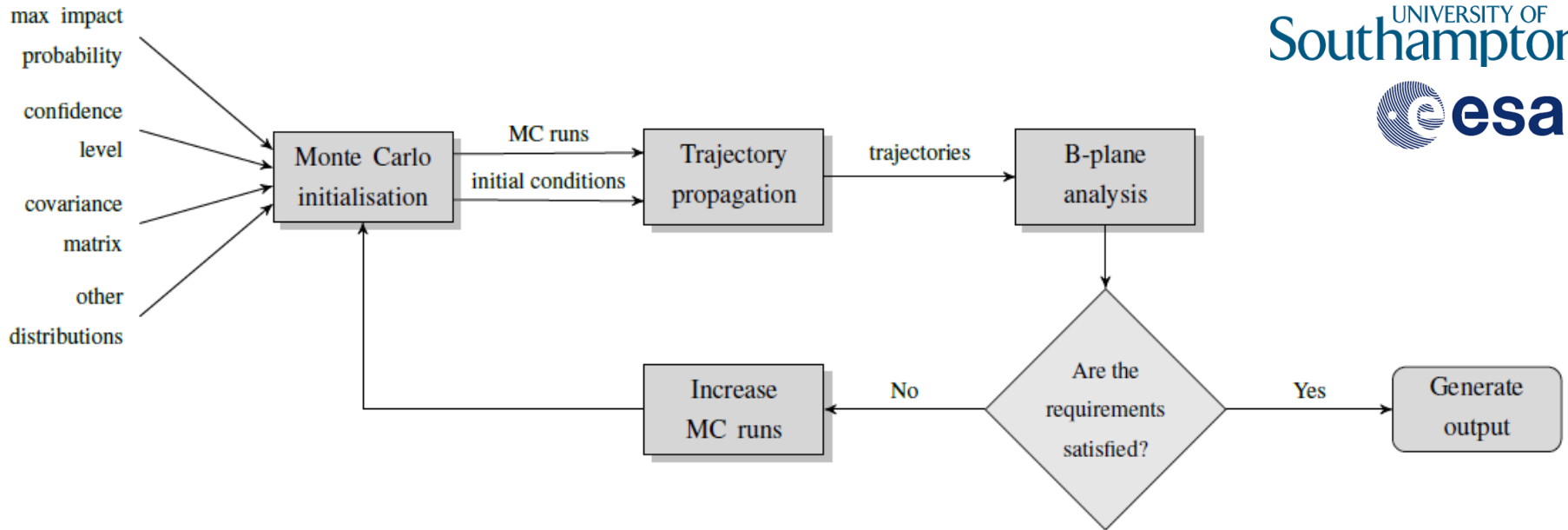
## Motivations

- During interplanetary missions, spacecraft and debris may impact with other planets over long times
  - Impacts from man-made objects can cause **biological contamination**
  - Sensible targets for scientific research (Mars, Europa, Enceladus) impose very stringent **planetary protection requirements**<sup>1</sup>
  - Space missions must satisfy these requirements during design phase
  
- Driving factors
  - Uncertainty over the initial state of the launcher injection
  - Uncertainty over spacecraft design parameters (e.g. area/mass ratio)
  - Random failures of spacecraft propulsion system

<sup>1</sup> G. Kminek, *ESA planetary protection requirements*, Technical Report ESSB-ST-U-001, European Space Agency, February 2012

## Suite for Numerical Analysis of Planetary Protection

**SNAPPshot** (Suite for the Numerical Analysis of Planetary Protection)<sup>2</sup>, developed by the University of Southampton under a study for the European Space Agency

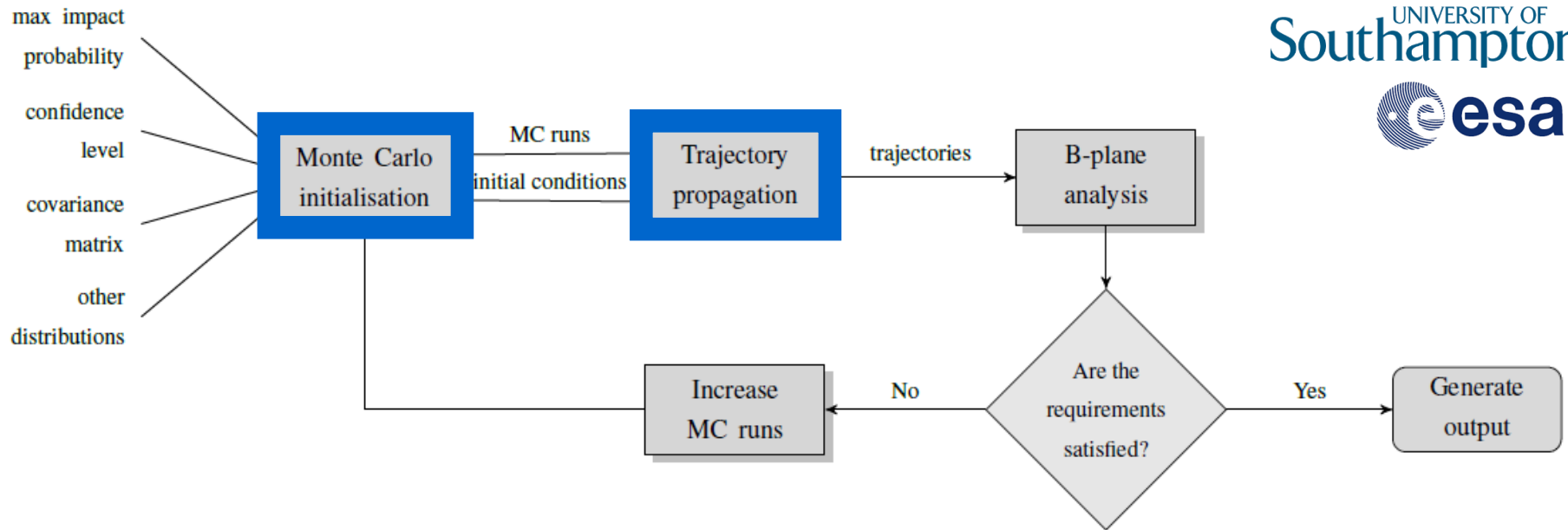


<sup>2</sup> Letizia F., Colombo C., Van den Eynde J., Armellin R., Jehn R., *SNAPPSHOT: Suite for the numerical analysis of planetary protection*, ICATT 2016

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## Our approach

The main goal is to **improve the accuracy and the efficiency** of the Planetary Protection analysis:

### ■ Numerical integration

Understand how the errors in a single propagation may affect planetary protection verification

- RK schemes (current)
- Symplectic and energy-preserving methods (in development)
- Other (future)

### ■ Sampling techniques

Efficient methods to sample the initial dispersion

- Monte Carlo (current)
- Line Sampling & Subset Simulation (in development)
- Other (future)



# NUMERICAL INTEGRATION



## Introduction

- Numerical methods accumulate errors during long-term integrations
  - **Fictitious dissipation** of total energy of the system
  - **Large errors** in the propagation introduced by fly-bys
  - Effect of numerical errors in single propagations on the overall planetary protection has to be determined
  
- Alternative numerical approaches may improve the accuracy of the orbital propagation
  - Symplectic schemes preserve constants of motion exactly or with bounded oscillations
  - Additional numerical techniques can help in maintaining the correct qualitative behaviour of the solution (no energy dissipation)

## Selection of numerical methods

Method	Explanation	Pros	Cons
RK	Generic Runge-Kutta schemes can become symplectic when applied to Hamiltonian systems under <b>conditions on their coefficients</b>	<ul style="list-style-type: none"><li>• Step adaptation is possible</li></ul>	<ul style="list-style-type: none"><li>• Symplectic schemes have to be implicit</li></ul>
GLRK	Implicit method based on <b>Gauss-Legendre quadrature points</b>	<ul style="list-style-type: none"><li>• Symplectic</li><li>• Numerically stable</li></ul>	<ul style="list-style-type: none"><li>• Implicit</li><li>• Fixed step</li></ul>
RKN	<b>Runge-Kutta-Nystrom</b> methods for <b>separable Hamiltonian</b> , use different schemes to integrate coordinates and momenta of the Hamiltonian	<ul style="list-style-type: none"><li>• Symplectic</li><li>• Can be explicit</li></ul>	<ul style="list-style-type: none"><li>• High number of evaluations</li><li>• Fixed step</li></ul>
SY	Derived from the Hamiltonian formulation, make use of successive <b>canonical transformations</b>	<ul style="list-style-type: none"><li>• Symplectic</li><li>• Explicit</li></ul>	<ul style="list-style-type: none"><li>• High number of evaluations</li><li>• Fixed step</li></ul>

# Numerical integration

## Selection of numerical methods

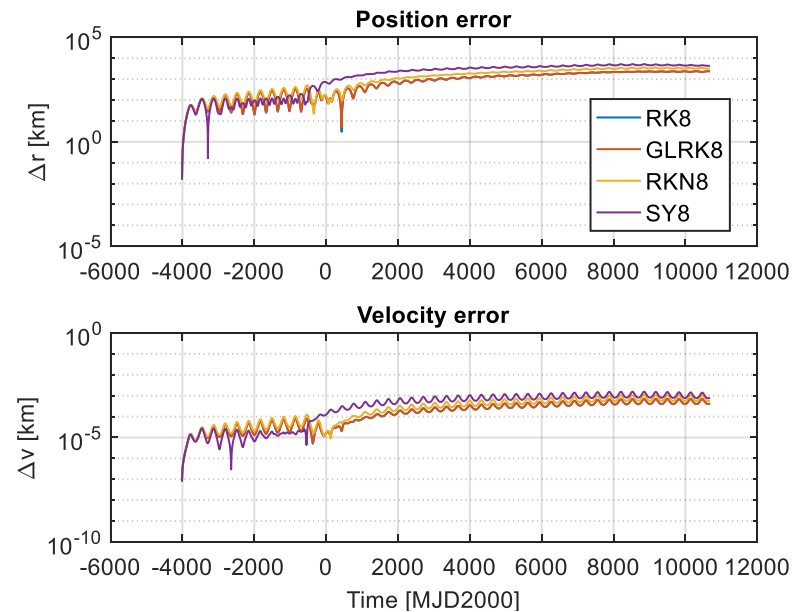
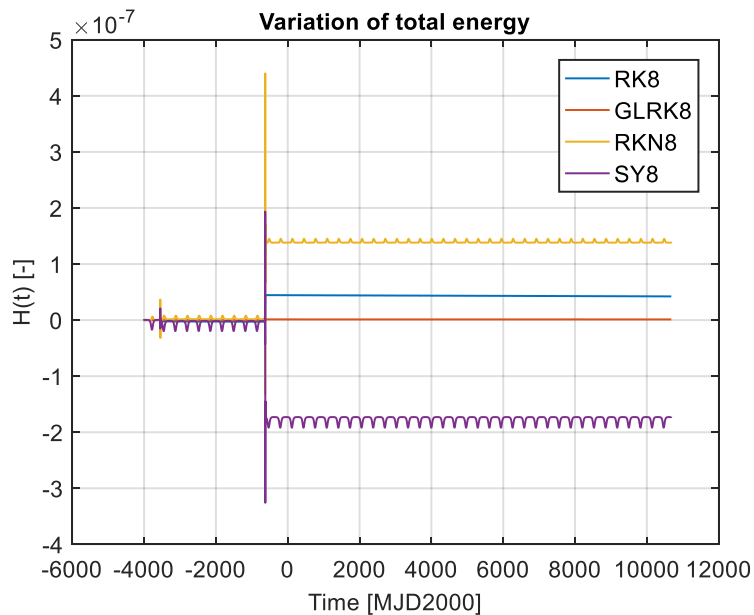
Method	(Order, Stages)	Type	Time step	Property	Reference
RK	(4,4) <b>(8,13)</b>	Explicit	Fixed step		Dormand and Prince, 1980
	(5/4,7) <b>(8/7,13)</b>	Explicit	<b>Adaptive step</b>		Prince and Dormand, 1981
GLRK	(4,2)	<b>Implicit</b>	Fixed step	<b>Symplectic</b>	Butcher, 1964
	(6,3)				Jones et al., 2012
	<b>(8,4)</b>				Aristoff et al., 2012
RKN	(6,6) <b>(8,26)</b>	Explicit	Fixed step	<b>Symplectic</b>	Dormand et al., 1987 Calvo et al., 1993
	(4,4) (6,8) <b>(8,16)</b>	Explicit	Fixed step	<b>Symplectic, canonical</b>	Yoshida et al., 1990 Neri, 1988

## Tests

Propagation of Apophis 1989-2029 (reference ephemeris from JPL SPICE)

8<sup>th</sup> order, fixed-step methods

(initial step determined according to RK8(7) with relative tolerance 1e-12)



\*Propagations performed in Matlab on processor Intel® Core™ i7-6500U CPU @ 2,50GHz

## Additional techniques

### ▪ Step regularisation

Step is rescaled during the integration according to the behaviour of the dynamics

- Maximum eigenvalue  $\Lambda$  of Jacobian matrix taken as reference value<sup>4</sup>

$$h_{n+1} = h_n \Lambda(t_n) / \Lambda(t_{n+1})$$

- More efficient tracking of dynamics, but change in time step during the integration **breaks down the conservation properties** of symplectic methods

### ▪ Projection methods

Correct the numerical solution according to the **gradient of the energy function**<sup>5</sup>

- Energy error is minimised
- **Implicit non-linear problem** has to be solved

<sup>4</sup> F. Debatin, A. Tilgner, F. Hechler, *Fast numerical integration of interplanetary orbits*. In *Second International Symposium on Spacecraft Flight Dynamics*, 1986

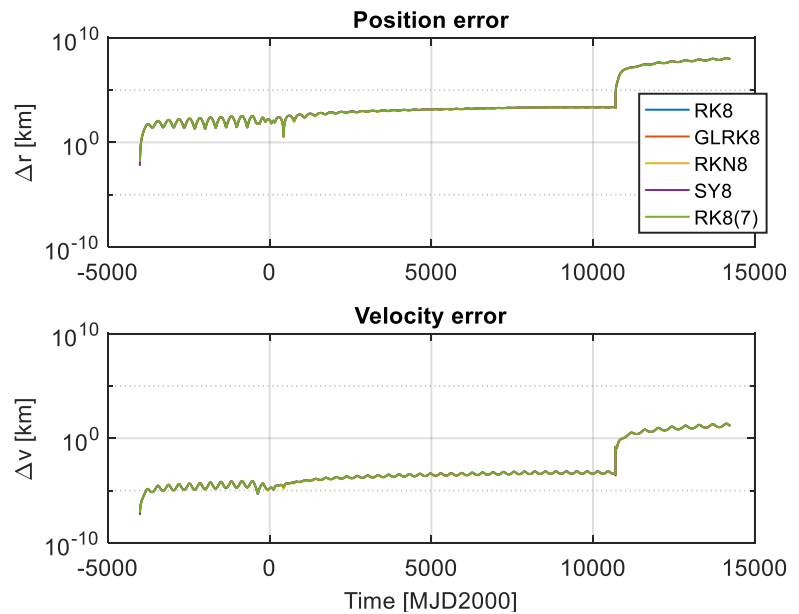
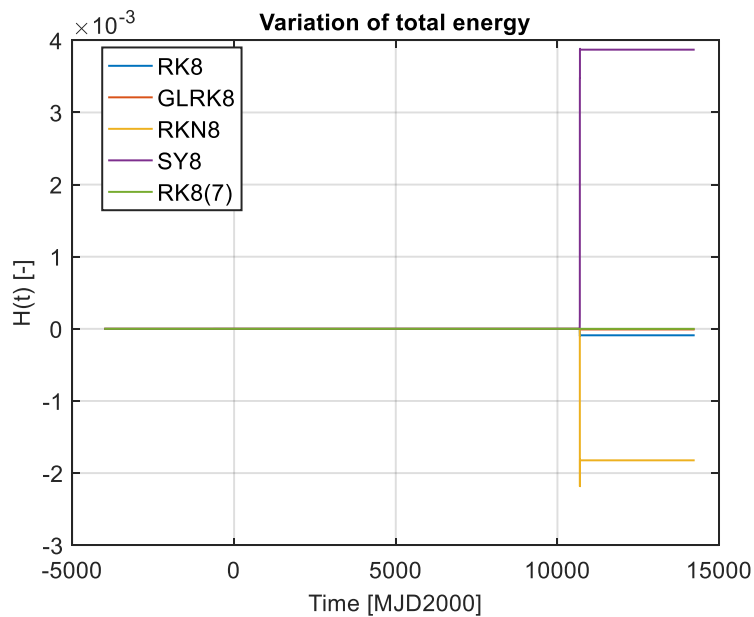
<sup>5</sup> Hairer, *Geometric Numerical Integration Structure-preserving algorithms for ordinary differential equations* (1990)

## Tests

Propagation of Apophis 1989-2039 (reference ephemeris from JPL SPICE)  
Fly-by of 2029 is included

8<sup>th</sup> order, Regularised step

(initial step determined according to RK8(7) with relative tolerance 1e-12)



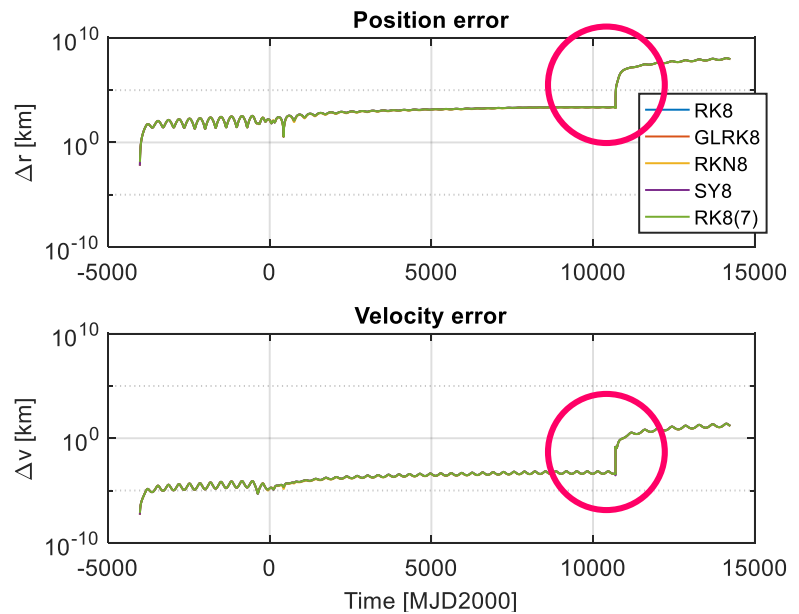
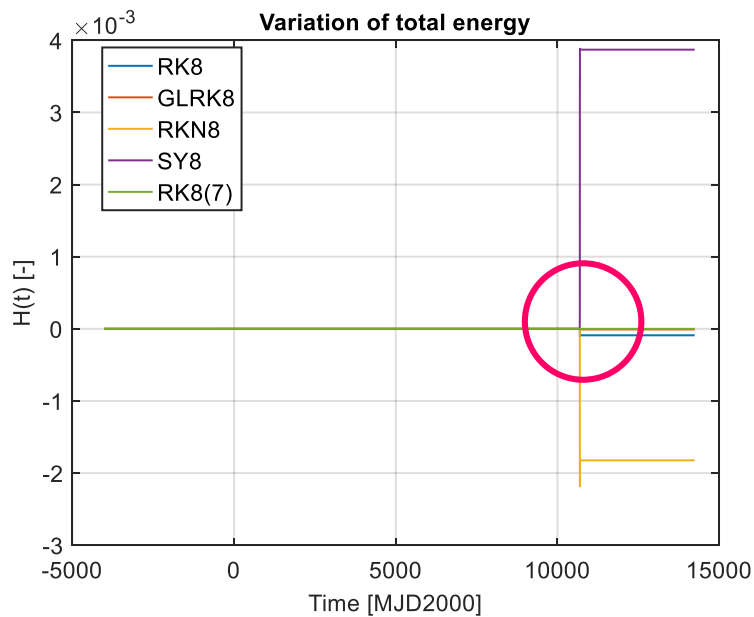
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## Additional techniques

Fly-bys heavily affect the numerical solution, **magnifying small numerical errors** due to a strong non-linearity of the dynamics

- Possible solution: apply other techniques only during the fly-by
  - Switch the centre of the propagation<sup>6</sup>
  - Add projection
  - other
- Problem: identification of the fly-by condition
  - Set distance from the planet (arbitrary or SOI definition)
  - **Alternative approach**: detection of fly-by according to the behaviour of the dynamics (Jacobian)

<sup>6</sup>D. Amato, G. Baù, C. Bombardelli, *Accurate orbit propagation in the presence of planetary close encounters*, *Monthly Notices of the Royal Astronomical Society*, Volume 470, Issue 2, 11 September 2017, Pages 2079–2099



## Fly-by detection through Jacobian

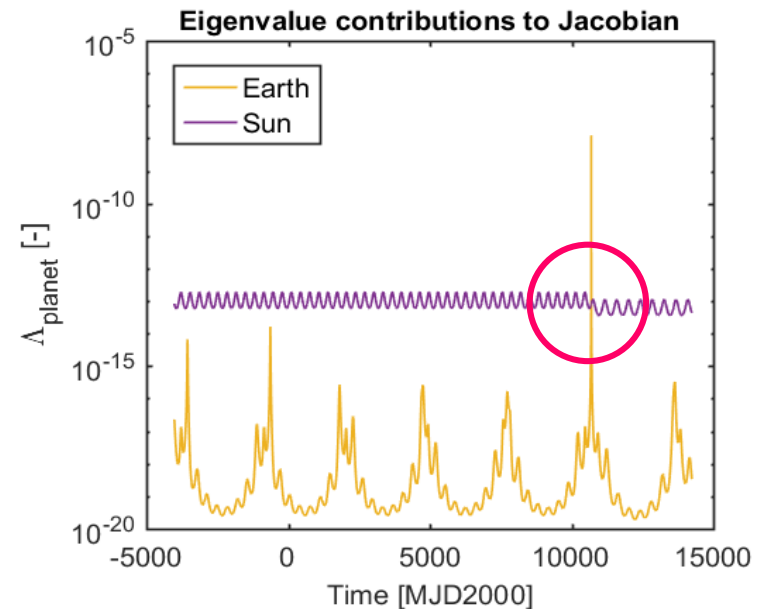
Considering the singular planet contributions to the Jacobian

- Value of planet contribution

$$\Lambda_j = \frac{2\mu_j}{\|\underline{r} - \underline{r}_j\|^3}$$

- Time variation of planet contribution

$$\dot{\Lambda}_j = -2\mu_j \frac{3(\underline{r} - \underline{r}_j)(\underline{v} - \underline{v}_j)}{\|\underline{r} - \underline{r}_j\|^5}$$



Fly-by detection criteria (approximation)

- Relative value w.r.t. main attractor:  $\Lambda_j / \Lambda_0 \geq \text{tol}$  (similar to SOI definition)
- Relative variation w.r.t. main attractor:  $\dot{\Lambda}_j / \dot{\Lambda}_0 \geq \text{tol}$

## Fly-by detection through Jacobian

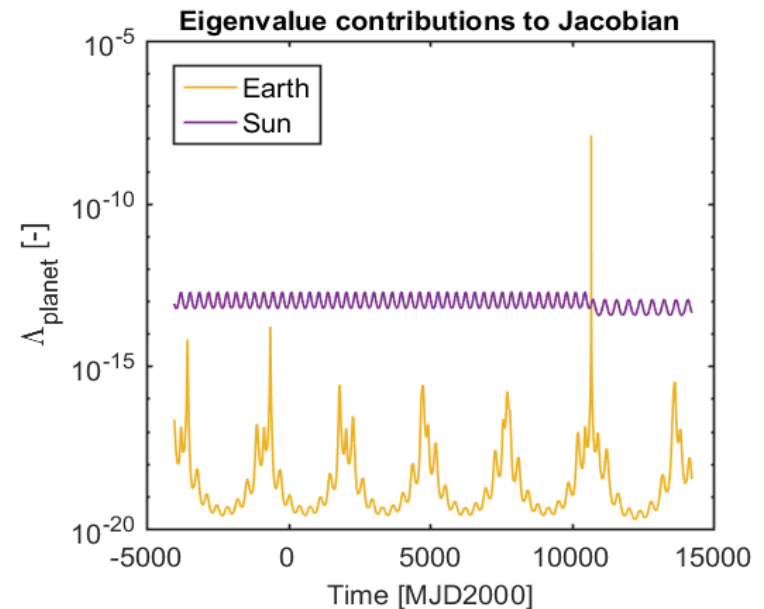
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# SAMPLING TECHNIQUES

## Explored ideas

- Monte Carlo
  - Number of runs selected to ensure the desired confidence level is respected (SNAPPShot)
- Other approaches to sampling
  - Methods compared in previous work<sup>7</sup>:
    - Line Sampling
      - Aimed to increase accuracy of probability estimation
    - Subset Simulation
      - Aimed to increase efficiency by reducing number of propagations

<sup>7</sup> M. Romano, M. Losacco, C. Colombo, P. Di Lizia, *Estimation of impact probability of asteroids and space debris through Monte Carlo Line Sampling and Subset Simulation*, KePASSA 2017 Workshop

## Line Sampling

The Line Sampling (LS) is a **Monte Carlo sampling** method that probes the uncertainty domain by using **lines** instead of random points

- The lines are used to **identify the boundaries of the impact region** inside the coordinate space
  - This can be done **independently from the initial uncertainty** and the probability estimation
  - The lines follow a reference direction pointing toward the impact subdomain
- The estimation of impact probability is reduced to a number of 1D integration problems along each line
  - **Analytical integration** results into a more accurate solution<sup>3</sup>

<sup>3</sup> Enrico Zio, Nicola Pedroni, *Subset Simulation and Line Sampling for Advanced Monte Carlo Reliability Analysis*, Proceedings of the European Safety and RELiability (ESREL) 2009 Conference, 2009, pp.687-694. <hal-007210>

## Line Sampling

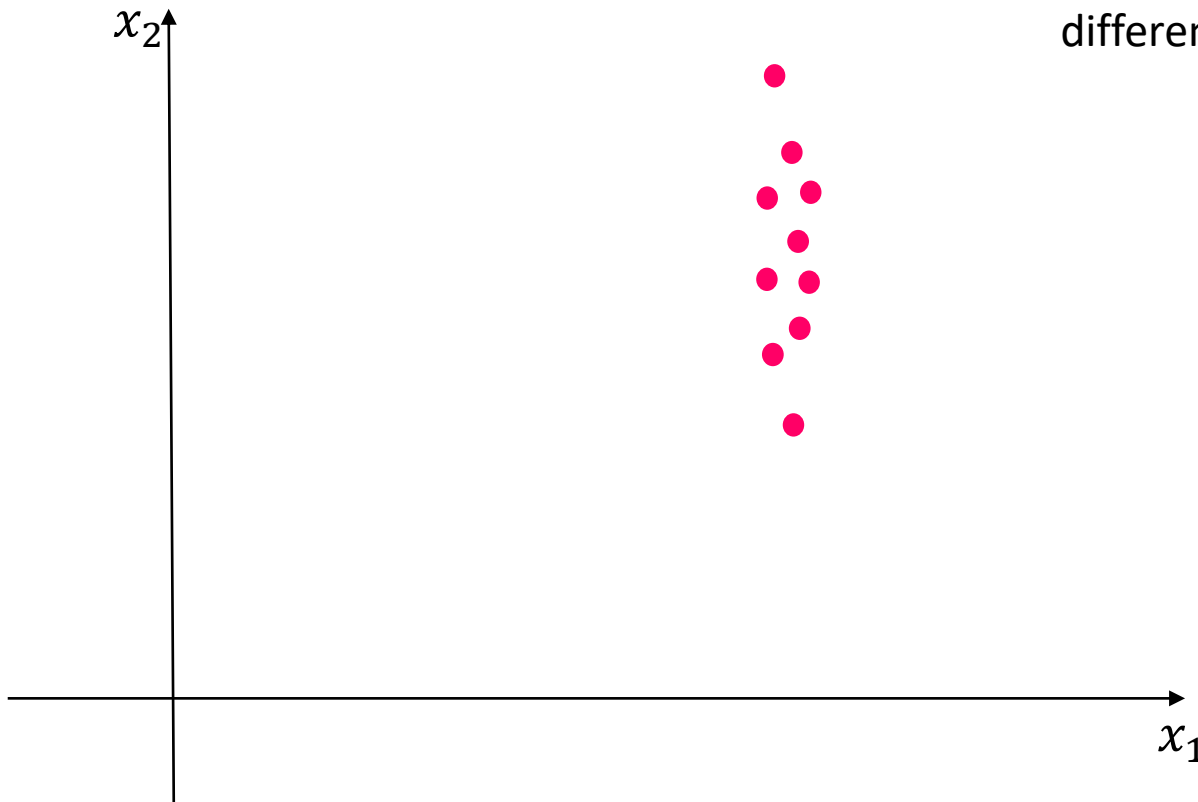
1. Determination of the “reference direction”
  - a. Impact region not known a priori



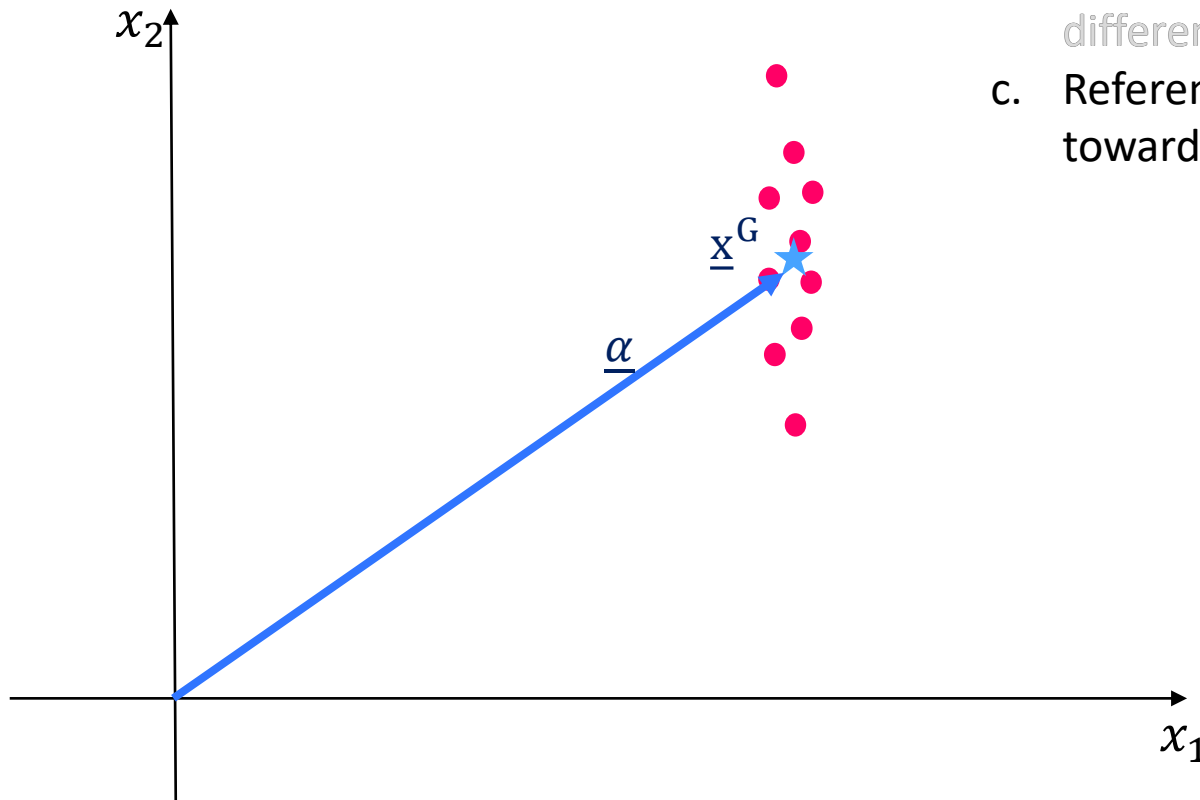
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## Line Sampling



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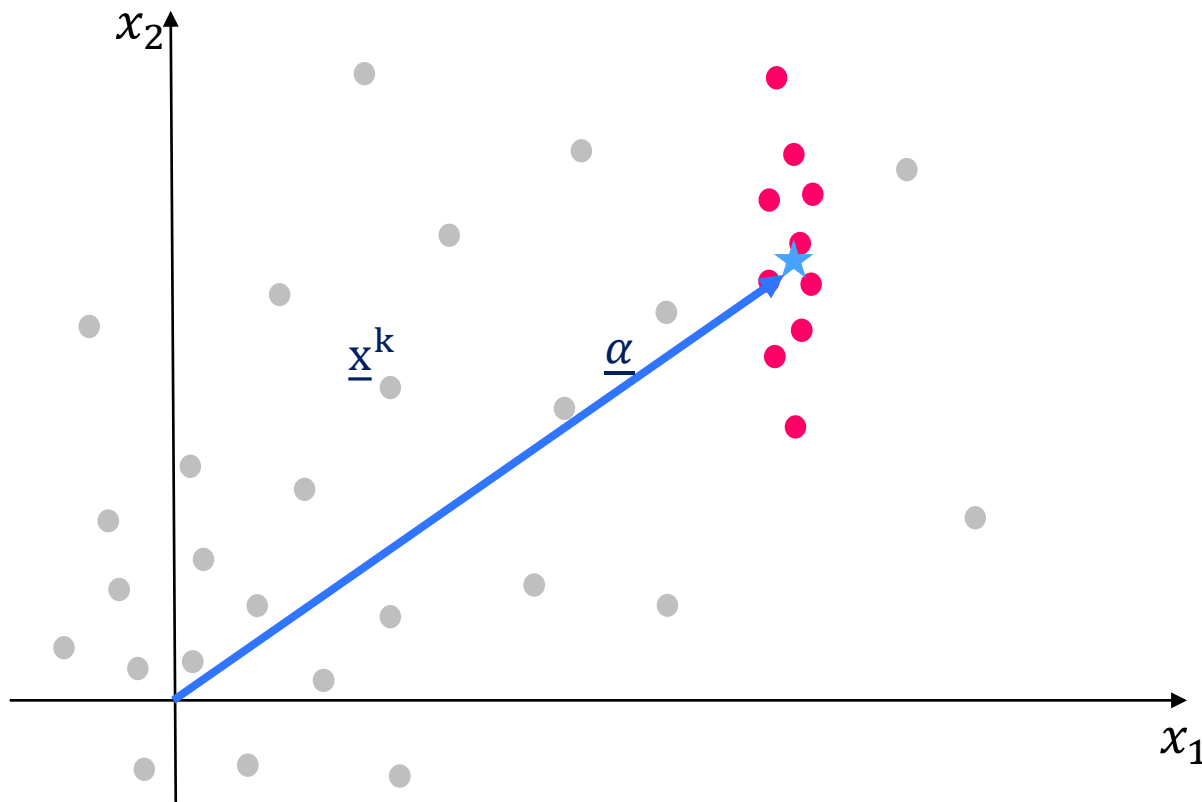
- Impact region not known a priori
- Information can be obtained in different ways
- Reference direction generally pointing toward impact region



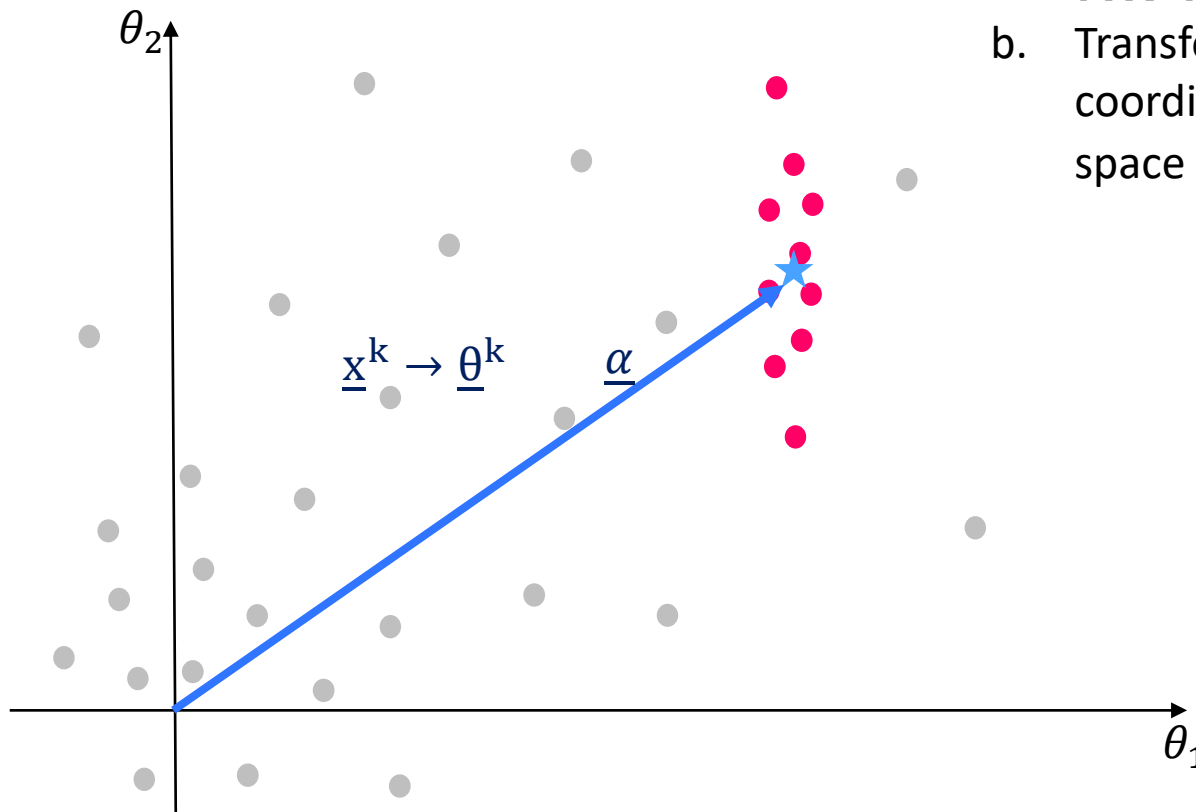
## Line Sampling

### 2. Mapping onto the standard normal space

- a. Generation of random samples according to given distribution



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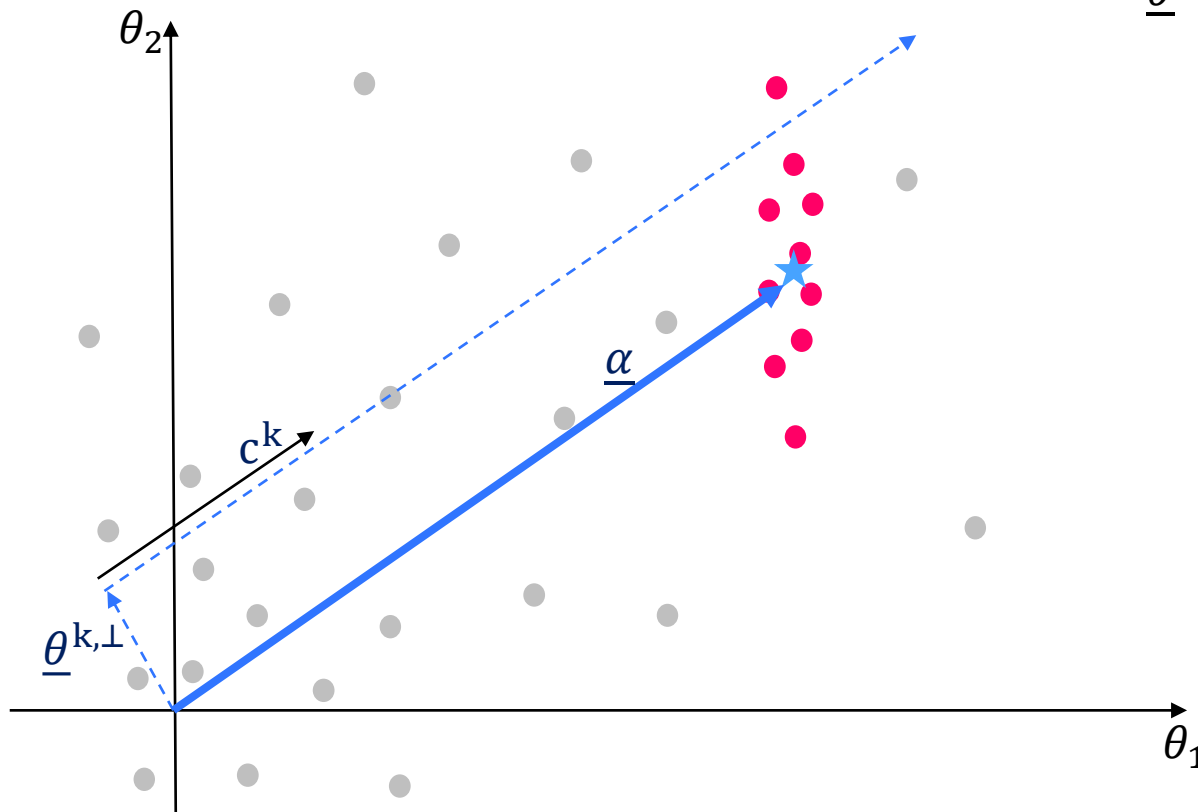
- Generation of random samples according to given distribution
- Transformation from physical coordinates into normalised standard space following  $\Phi(\underline{\theta}^k) = F(\underline{x}^k)$

## Line Sampling

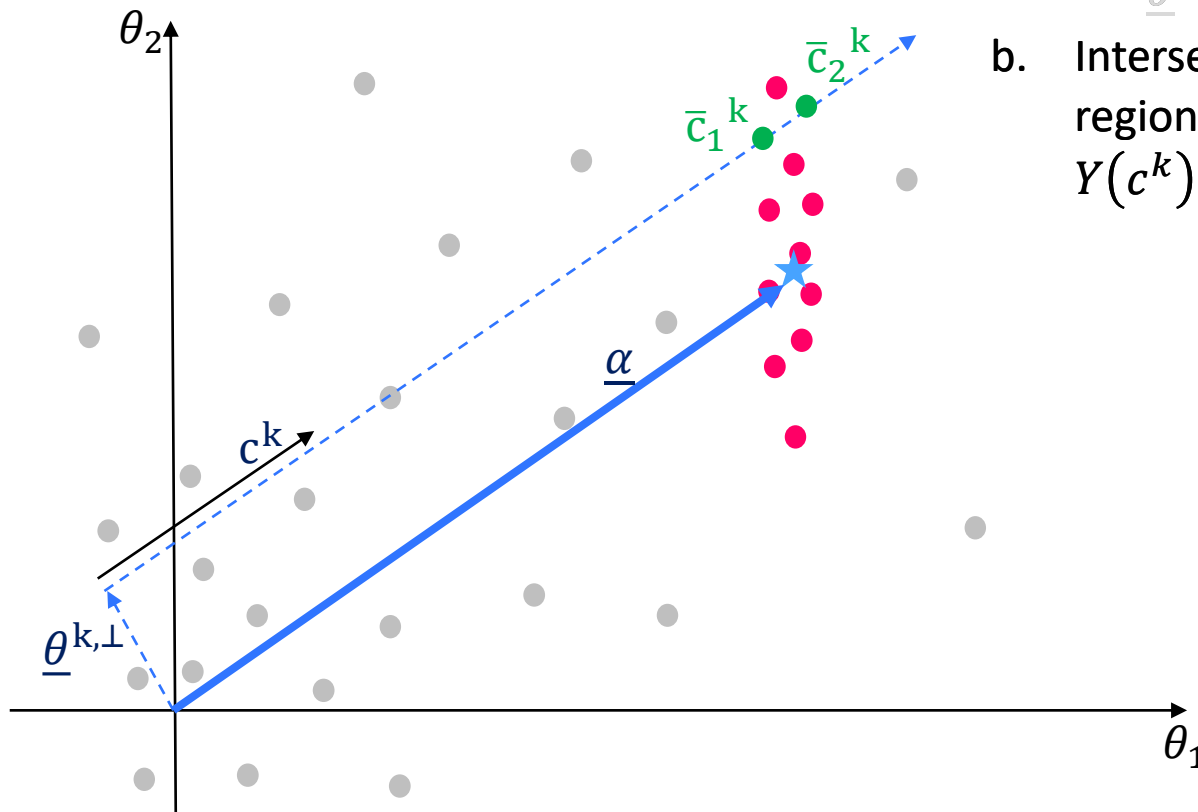
### 3. Sampling along the lines

- a. Lines defined in normalised space

$$\tilde{\theta}^k = c^k \underline{\alpha} + \underline{\theta}^{k,\perp}$$



## Line Sampling



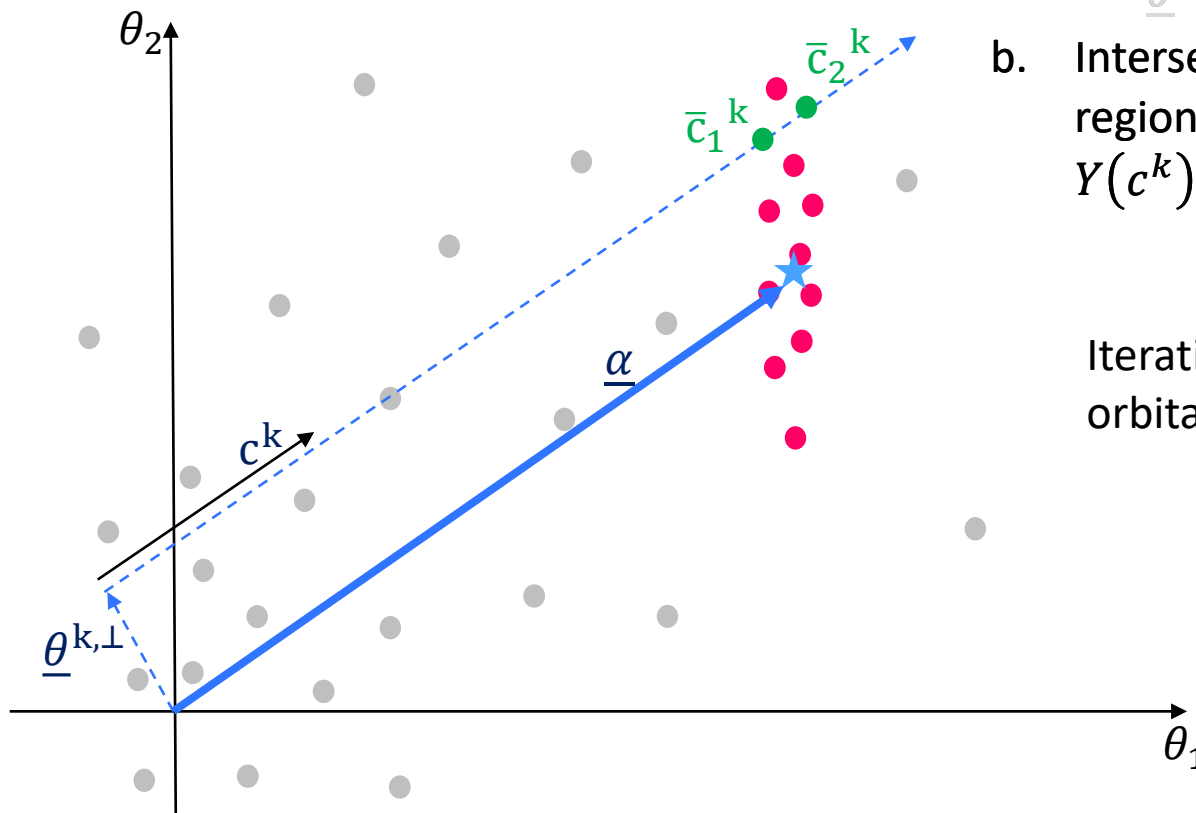
### 3. Sampling along the lines

- a. Lines defined in normalised space

$$\tilde{\theta}^k = c^k \underline{\alpha} + \underline{\theta}^{k,\perp}$$

- b. Intersections  $(\bar{c}_1^k, \bar{c}_2^k)$  with impact region found where objective function  $Y(c^k) = 0$

## Line Sampling



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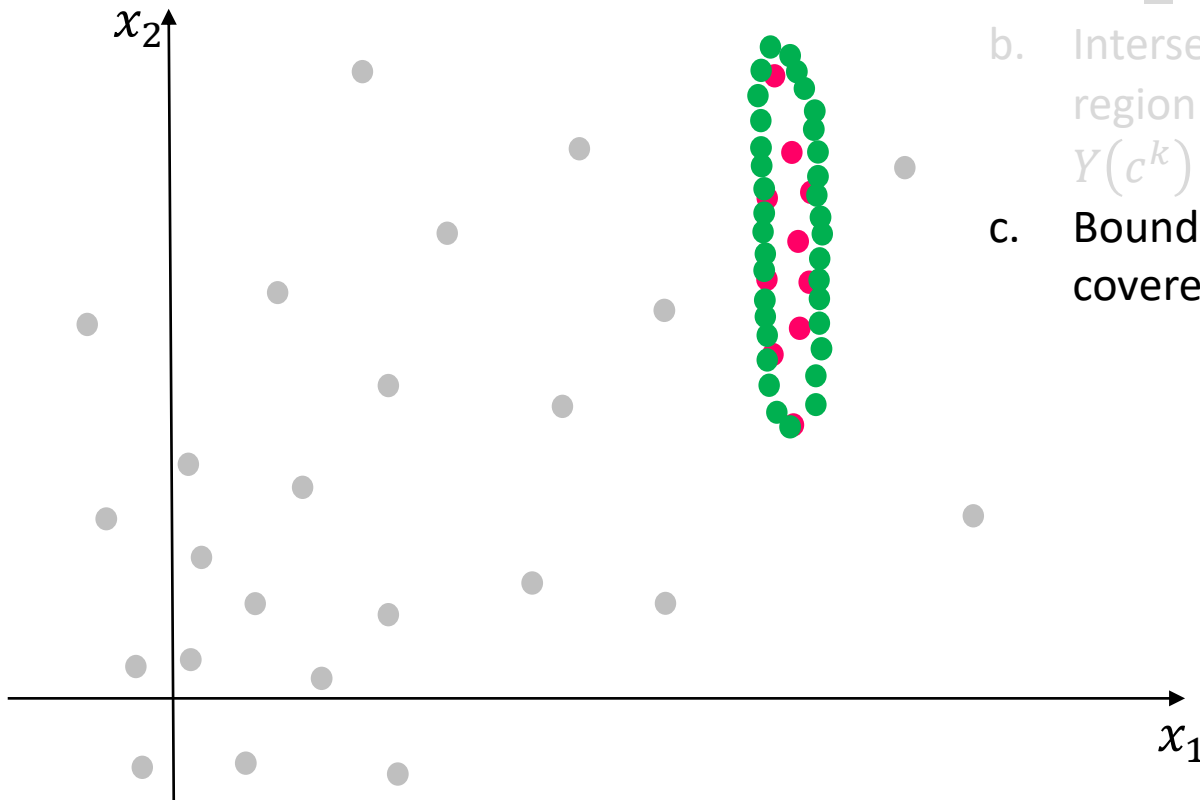
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Iterative procedure requires extra orbital propagations

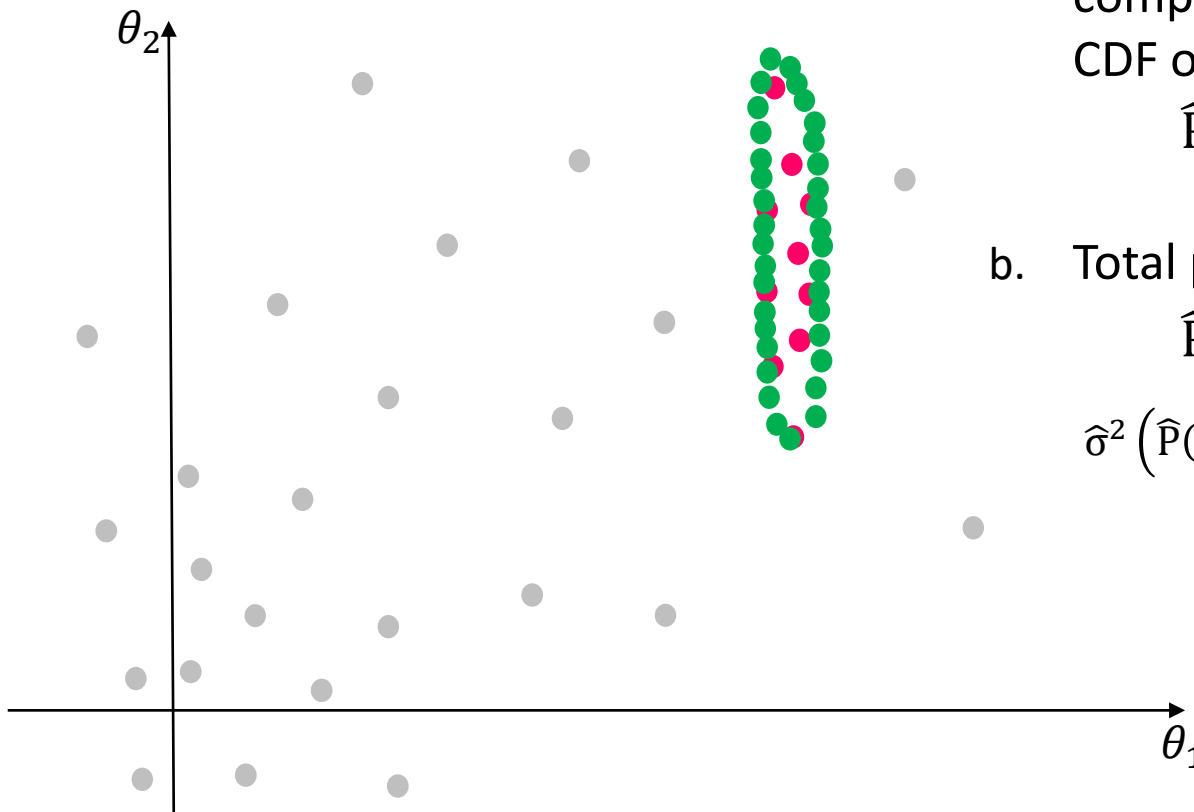
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- Intersections  $(\bar{c}_1^k, \bar{c}_2^k)$  with impact region found where objective function  $Y(c^k) = 0$
- Boundaries of the impact region are covered

## Line Sampling



### 4. Estimation of impact probability

- a. Partial probability estimates are computed along each line using the CDF of the unit gaussian:

$$\hat{P}^k(I) = \Phi(\bar{c}_2^k) - \Phi(\bar{c}_1^k)$$

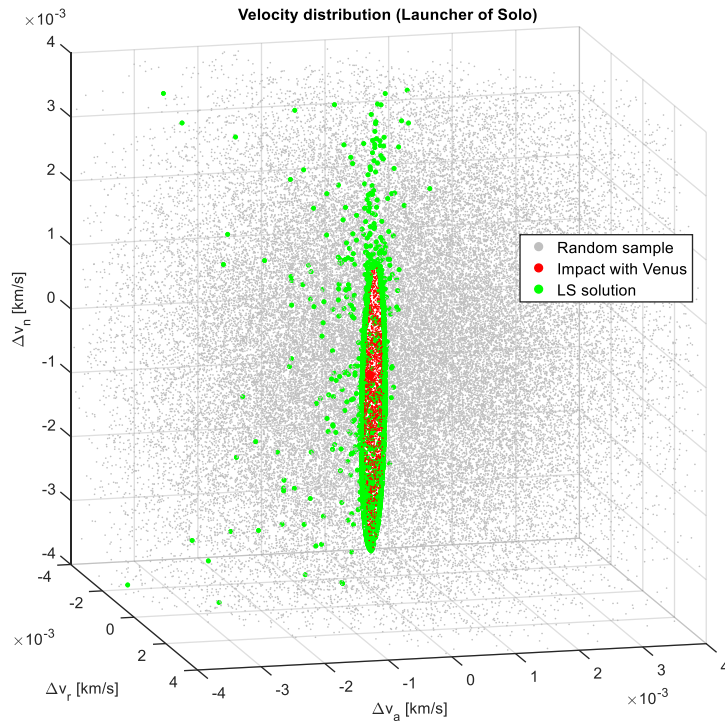
- b. Total probability and variance

$$\hat{P}(I) = \frac{1}{N_T} \sum_{k=1}^{N_T} \hat{P}^k(I)$$

$$\hat{\sigma}^2(\hat{P}(I)) = \frac{1}{N_T(N_T-1)} \sum_{k=1}^{N_T} (\hat{P}^k(I) - \hat{P}(I))^2$$

## Test: Launcher of Solo

Solar Orbiter (Solo) is a planned Sun-observing satellite, under development by ESA.  
Analysed event: fly-by of Venus



	$N_{\text{Samples}}$	$N_{\text{Prop}}$	$\hat{P}(I)$	$\hat{\sigma}$
MCS	54114	54114	4.2e-2	8.6e-4
LS	~54000	~200000	4.3e-2	5.5e-4

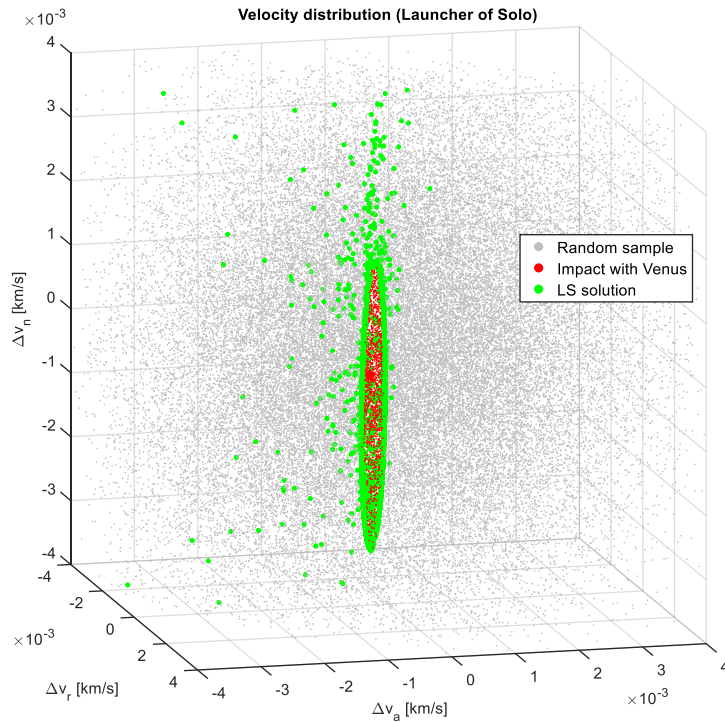
LS identifies well the boundaries of the impact region  
but

**Large expected probability** and **compact impact region**  
make the method less efficient for the same  
confidence level



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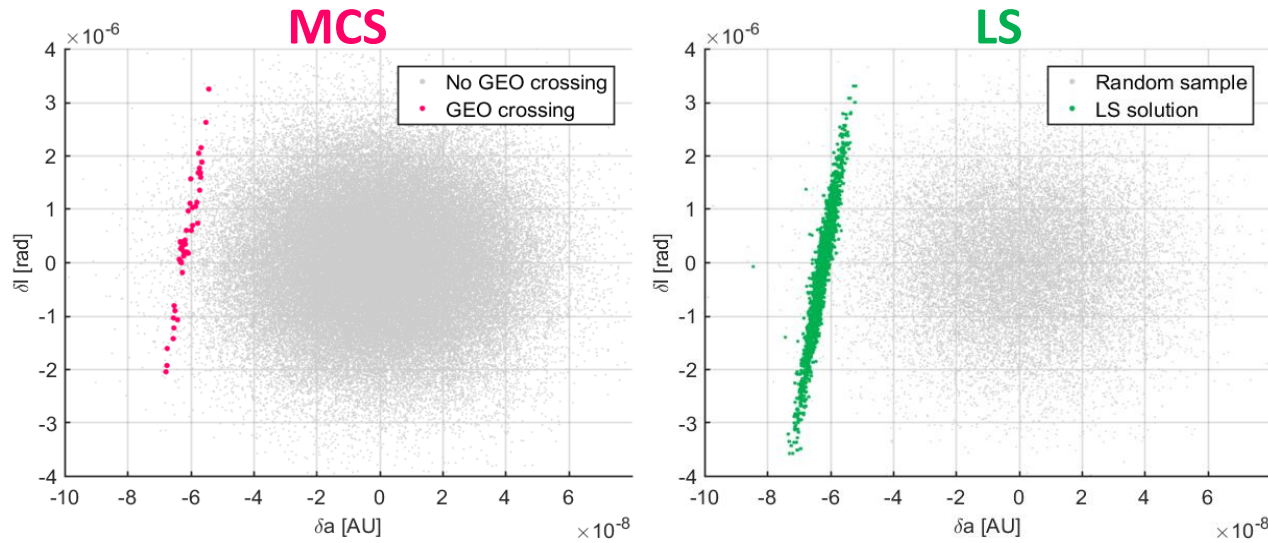
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## Test: Apophis

Analysed event: return in 2036 (according to observations in 2009)<sup>9</sup>



Small expected probability  
Distributed impact region

	$N_{\text{Samples}}$	$N_{\text{Prop}}$	$\hat{P}(I)$	$\hat{\sigma}$
<b>MCS</b>	1e6	1e6	5.00e-5	6.86e-6
<b>LS</b>	1e4	$\sim 1e5$	5.38e-5	1.18e-6
	1e5	$\sim 1e6$	5.32e-5	3.45e-7

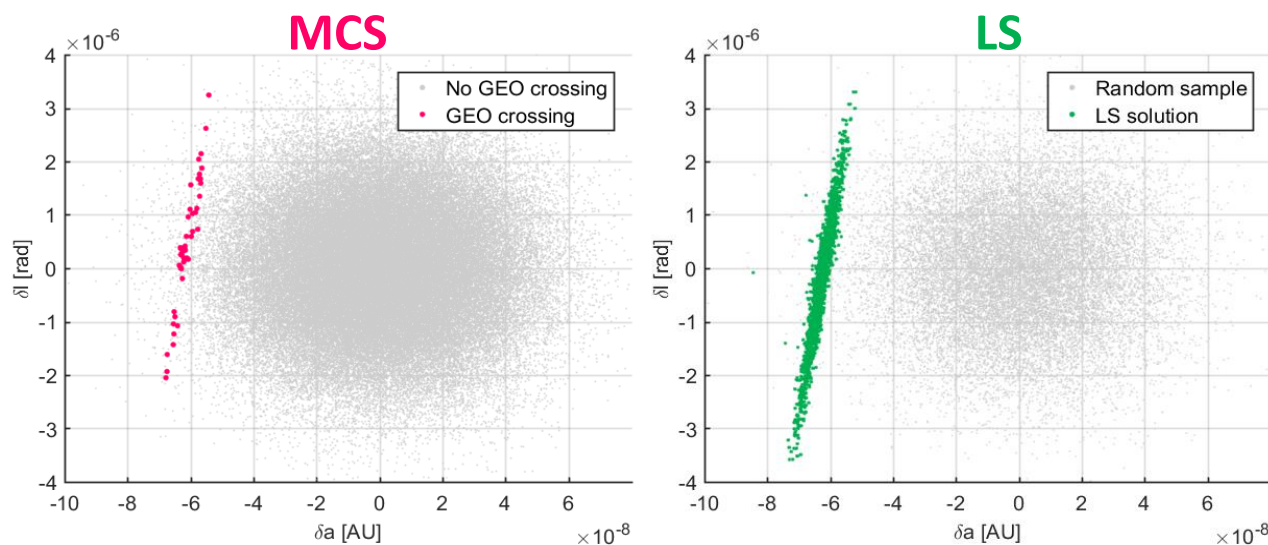
Similar confidence level as MC

Similar number of orbital  
propagations as MC

<sup>9</sup> <http://newton.dm.unipi.it/neodys>

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	1e5	~1e6	5.32e-5	3.45e-7 ↓

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# CONCLUSIONS

## Final considerations

- Integration
  - Symplectic methods show good performance in cases of regular dynamics
  - Very close **fly-bys** introduce very large numerical **errors** in the integration
  - Techniques to cancel the effect of a fly-by on the propagation exist and are being investigated
  
- Sampling
  - LS can achieve a lower variance of the solution (**higher accuracy**) with the same number of random samples, with larger efficiency as the impact probability gets lower
  - Current implementation supposes a unique impact region with a regular shape, during a given time window
  - Current implementation uses **extra evaluations** to probe each line, thus decreasing the efficiency of LS

## Future work

- Numerical integration
  - Explore alternative methods (symplectic and non)
  - Explore alternative formulations of the dynamics (Keplerian/equinoctial parameters, Delaunay parameters, universal variables, etc.)
  - Develop other numerical schemes (symplectic and non)
- Uncertainty sampling
  - Obtain an analytical expression for the confidence interval in LS
  - Improve the computation of zeros
  - Explore other techniques to improve efficiency
- **Final goal:** apply the two approaches (efficient sampling and conservative integration) to planetary protection analysis



**THANK YOU FOR YOUR ATTENTION  
ANY QUESTION?**



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