

# An improved method for asteroid impact probability due to swarm intelligence algorithms

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# Introduction

- This research explores two fundamental questions regarding the orbital dynamics of potentially hazardous objects (PHOs):
  - (1) given the best orbit determination solution, i.e. a state and uncertainty at a specific time, what is the closest that asteroid ever gets to the Earth when propagated forwards in time?
  - (2) how can the computationally expensive process of running a Monte Carlo simulation, with potentially thousands or millions of samples, be improved?

# Method

- pose the PHO close approach problem as a trajectory optimization problem
- objective function set equal to perigee radius

$$J = r_p = \left[ r_x^2(t_p) + r_y^2(t_p) + r_z^2(t_p) \right]^{1/2}$$

- decision vector defined as an orbital state with constraints defined as the +/- 3 sigma state uncertainties for each element

$$\mathbf{x} = [a \quad e \quad i \quad \Omega \quad \omega \quad M]$$

$$\begin{aligned} \mathbf{x}_{\text{ub}} &= 3 [\sigma_a \quad \sigma_e \quad \sigma_i \quad \sigma_\Omega \quad \sigma_\omega \quad \sigma_M] \\ \mathbf{x}_{\text{lb}} &= -3 [\sigma_a \quad \sigma_e \quad \sigma_i \quad \sigma_\Omega \quad \sigma_\omega \quad \sigma_M] \end{aligned}$$

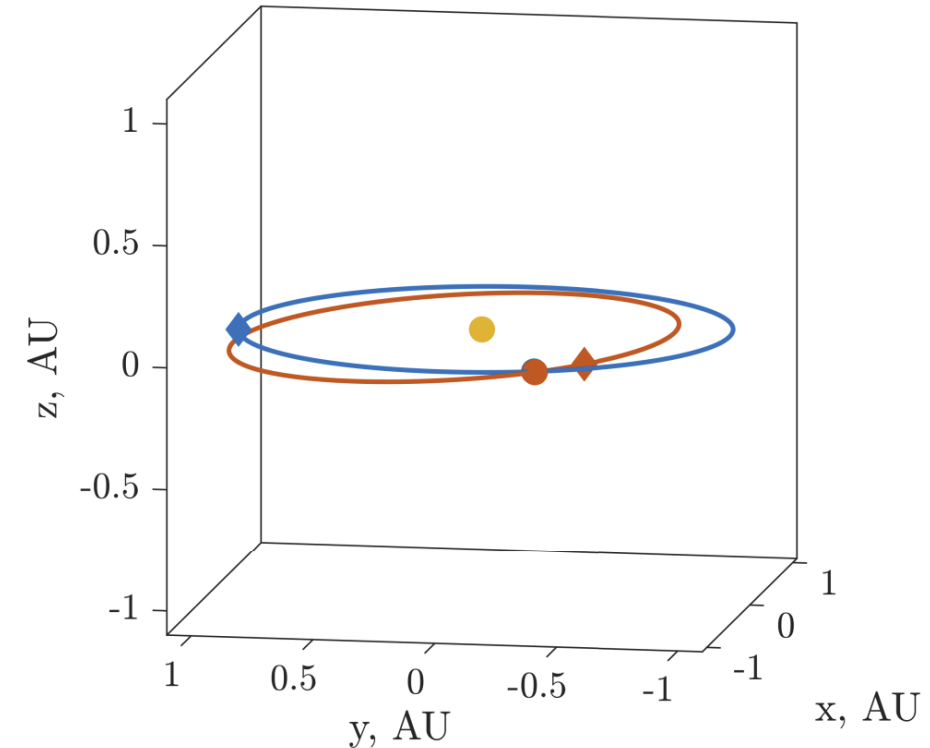
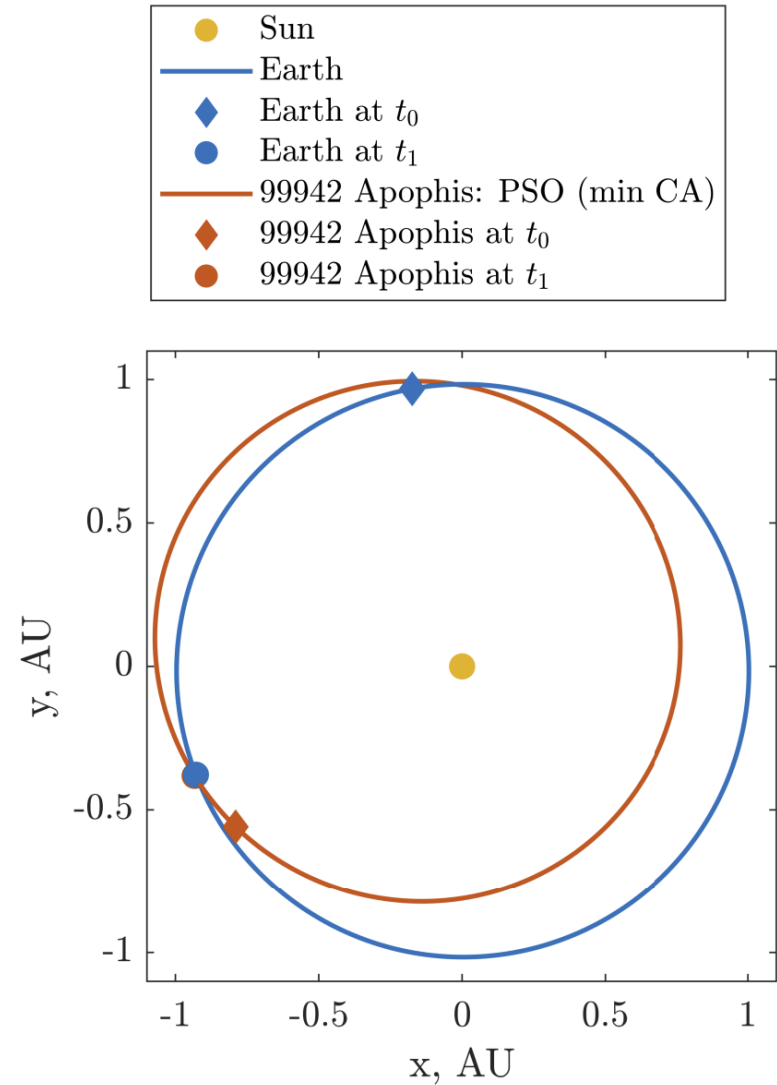
# Method

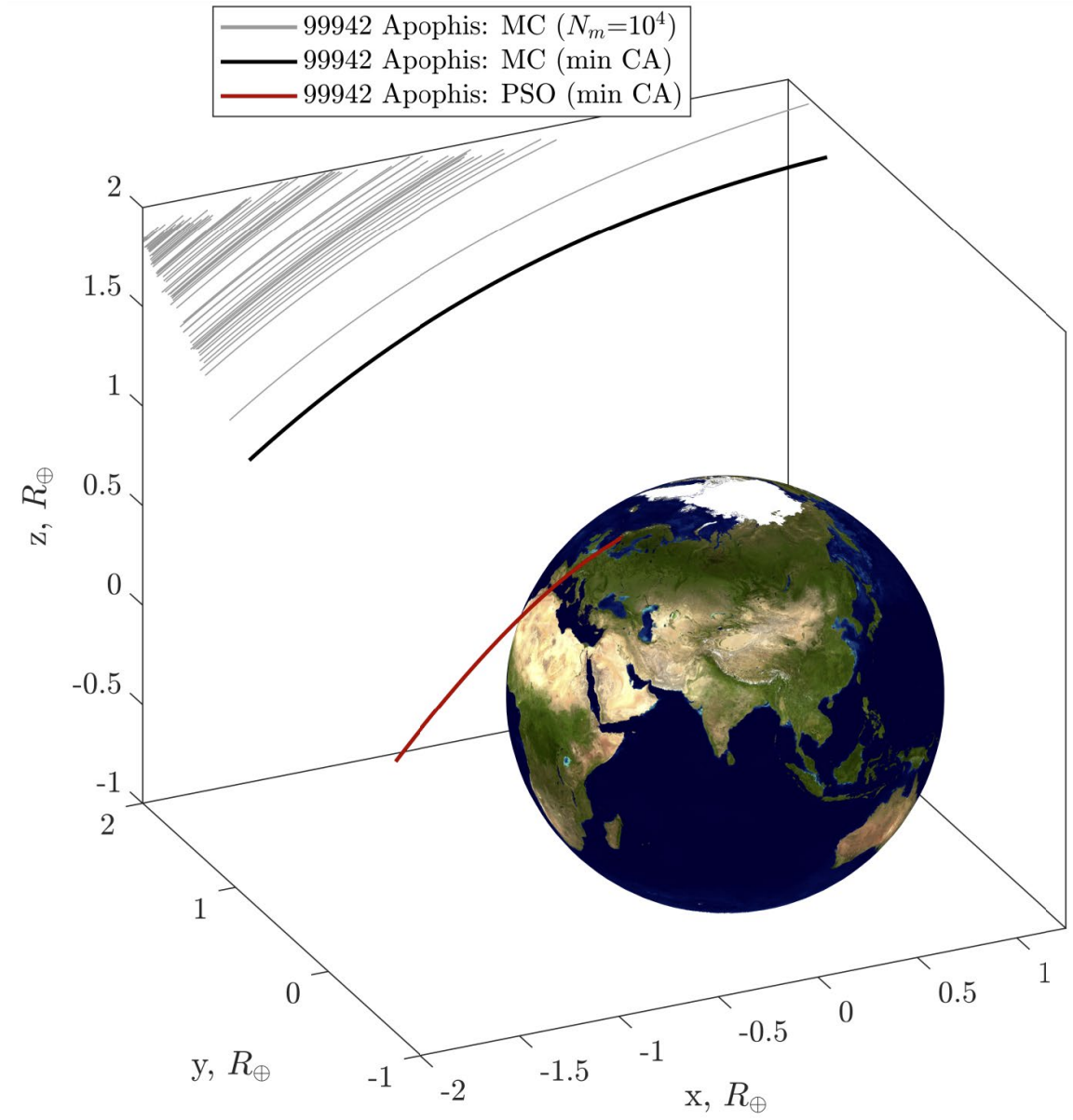
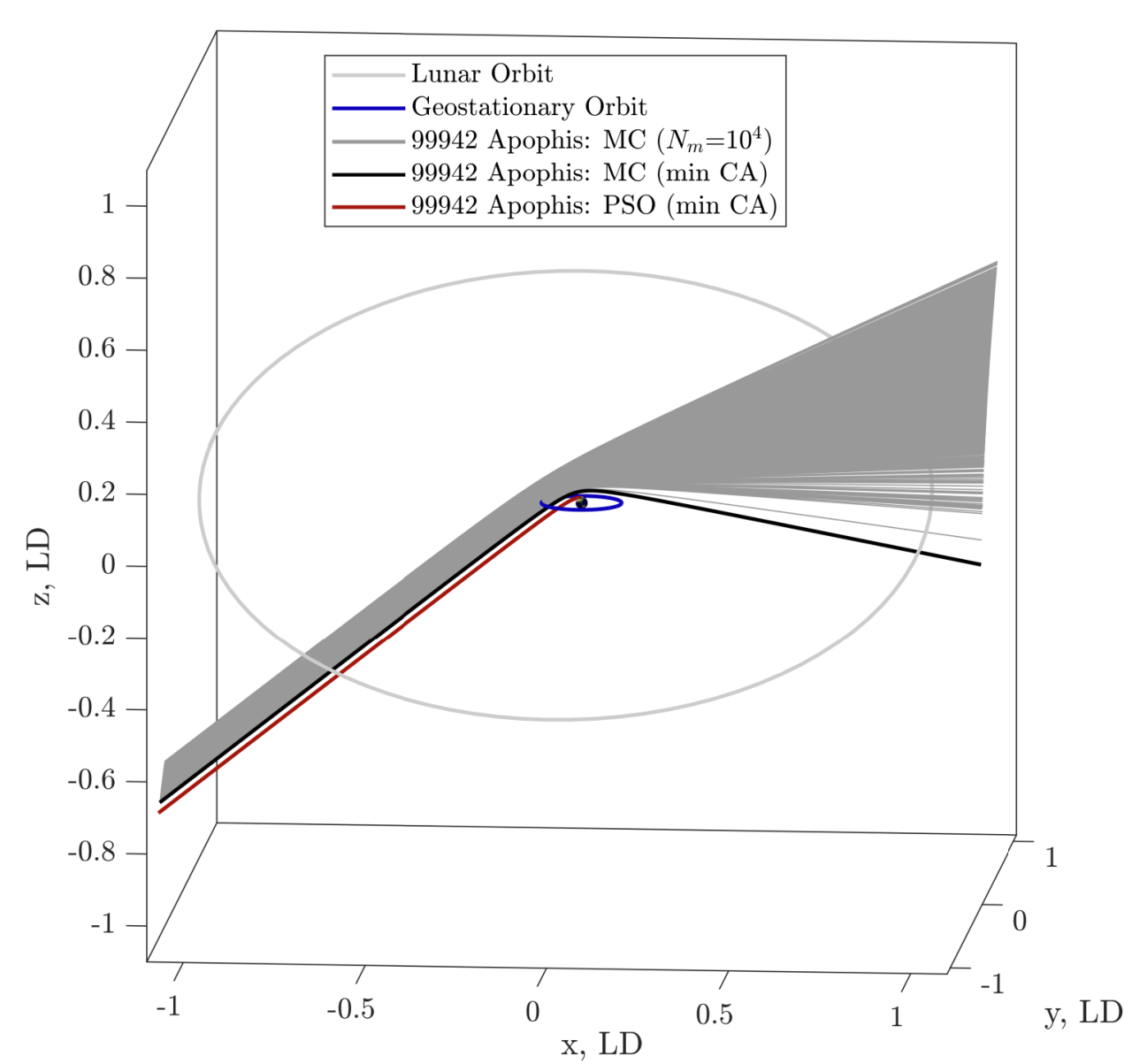
- Particle Swarm Optimization (PSO) – a swarm intelligence algorithm
- Generate random initial population of particles in n-dimensional search box – just like a Monte Carlo
  - $n=6$  since the search space is a 6-element orbit state
- Particles have a position vector for its location in the feasible n-dimensional search box and an associated score based on some cost function
  - Cost function is the perigee radius when that state is propagated until Earth close approach

# Method

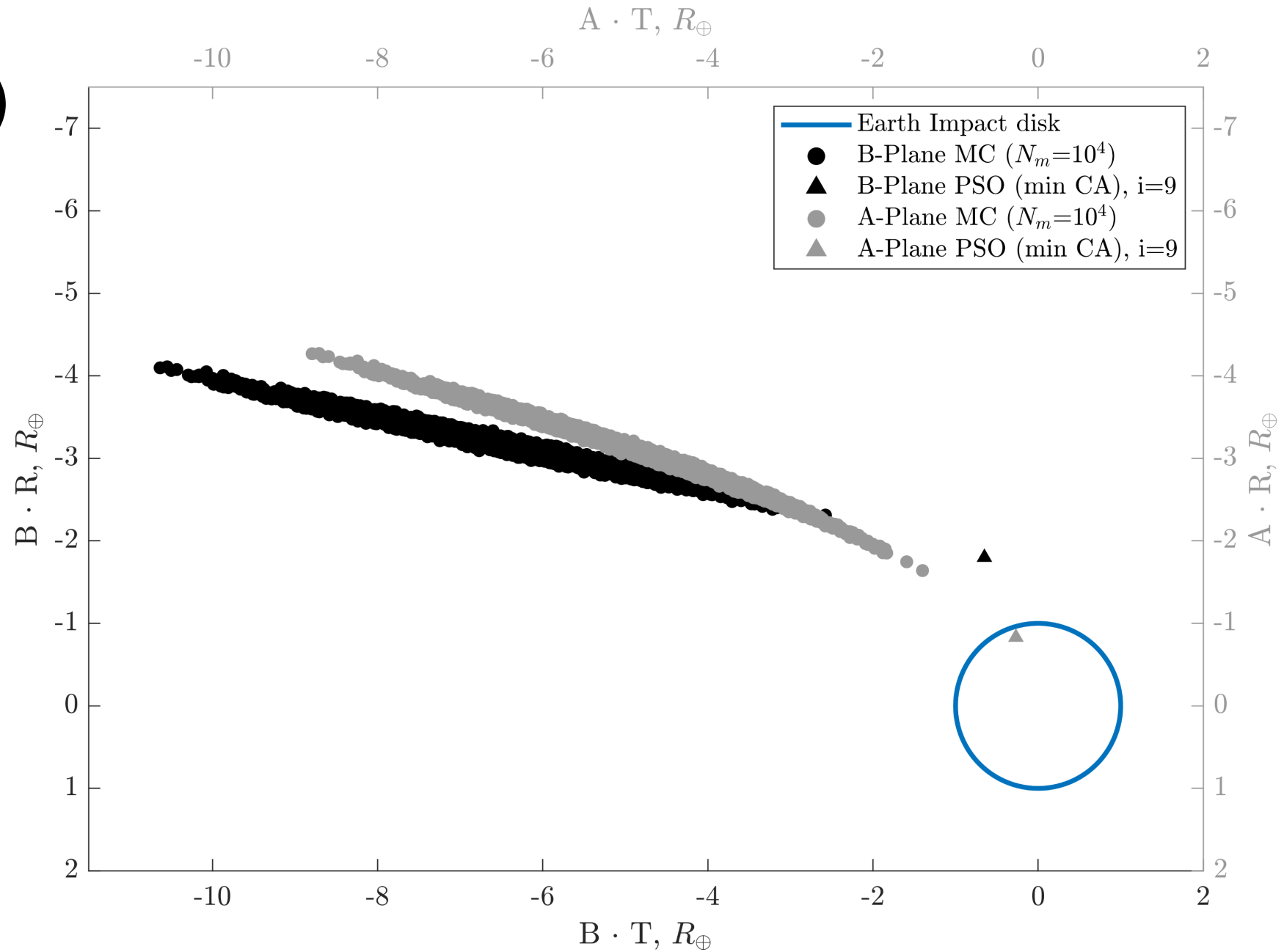
- PSO iterates, updating the position vector of each particle based upon three components with associated weighting coefficients:
  - *Inertial*: biases each particle to continue moving in the direction it was moving in during the previous iteration
  - *Cognitive*: biases the particle to move towards the best position that it has experienced (its personal best fitness)
  - *Social*: biases the velocity towards the global best fitness found by any particle in the population
  
- The classic Monte Carlo method can be reframed as zeroth order PSO

# Case Study 1) Apophis



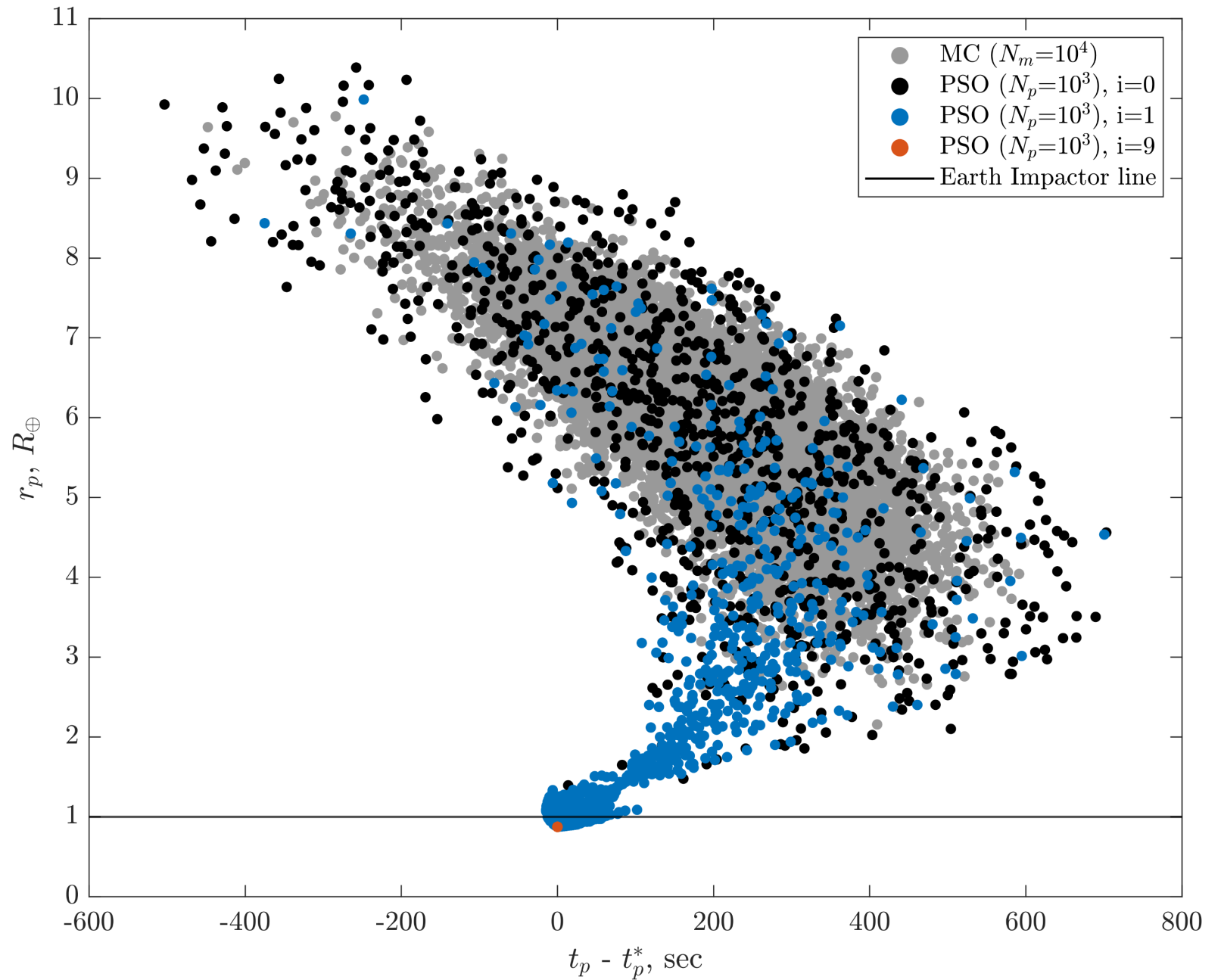


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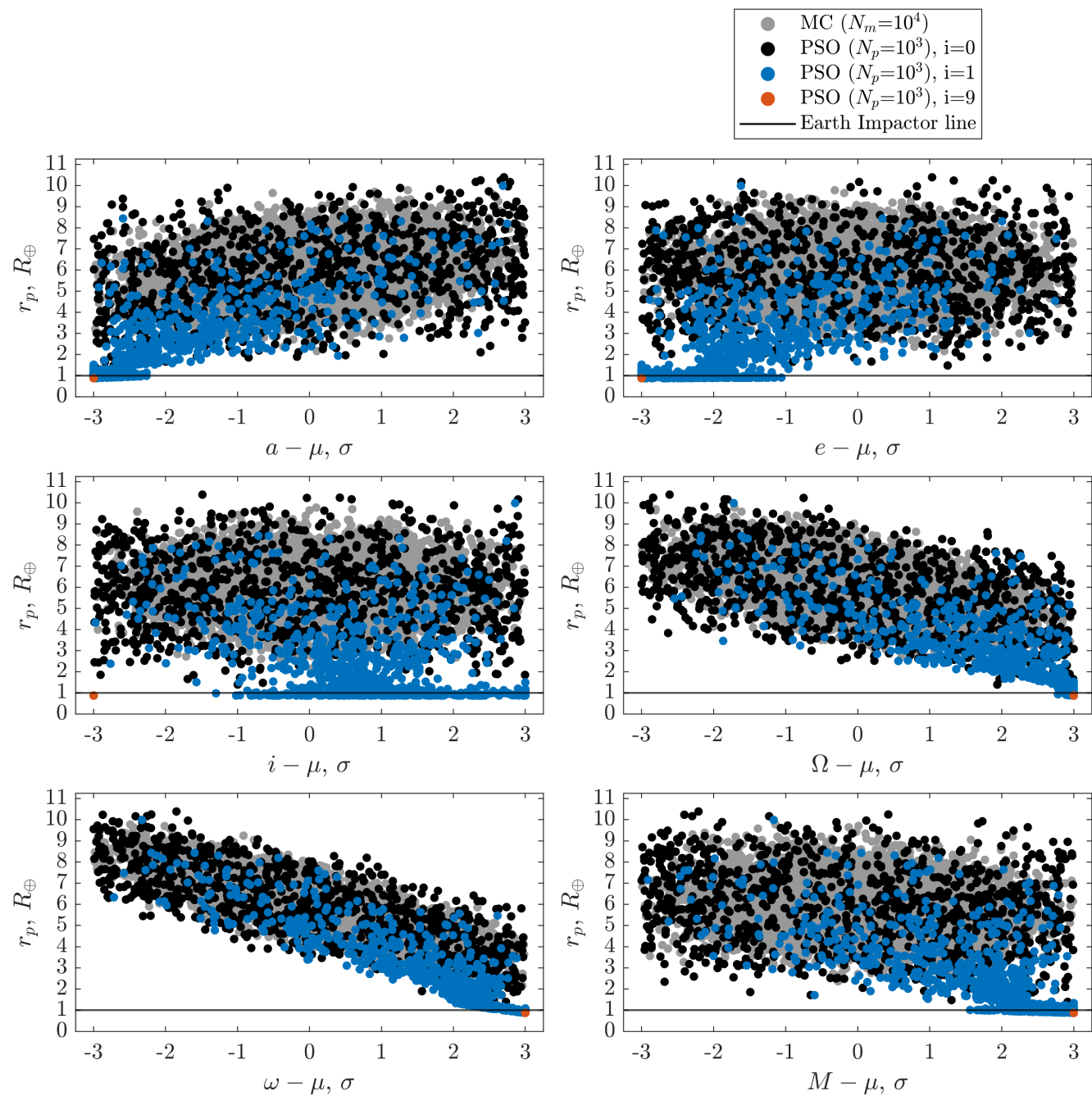




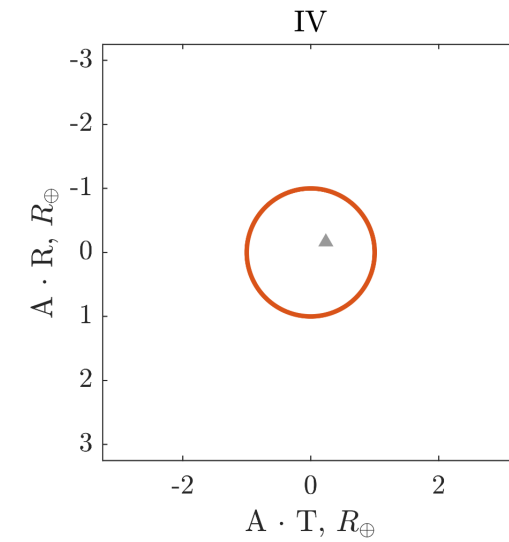
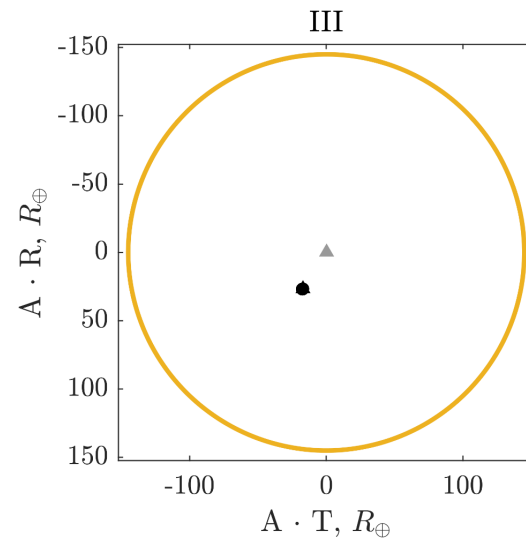
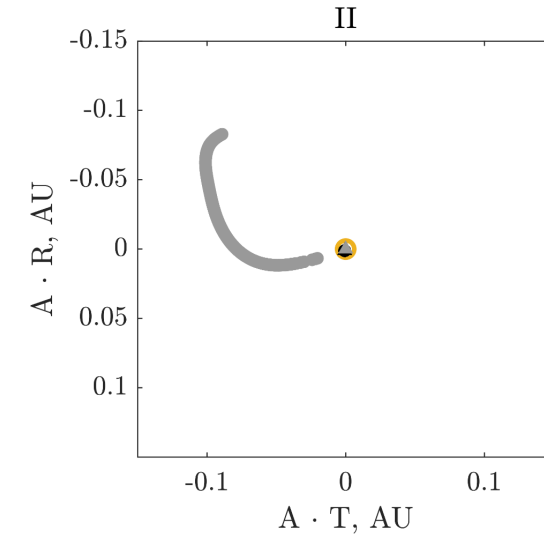
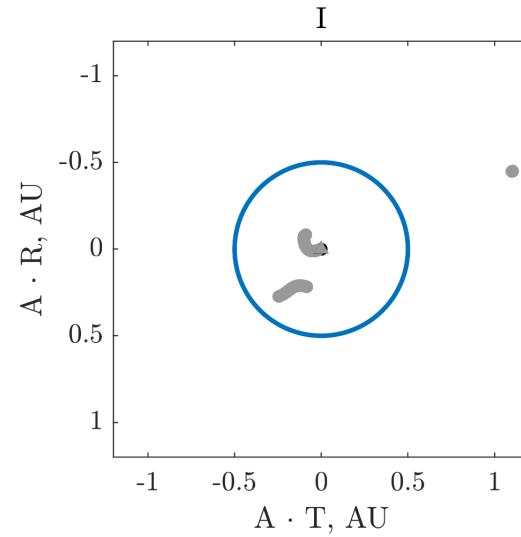
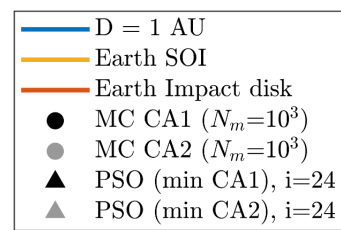
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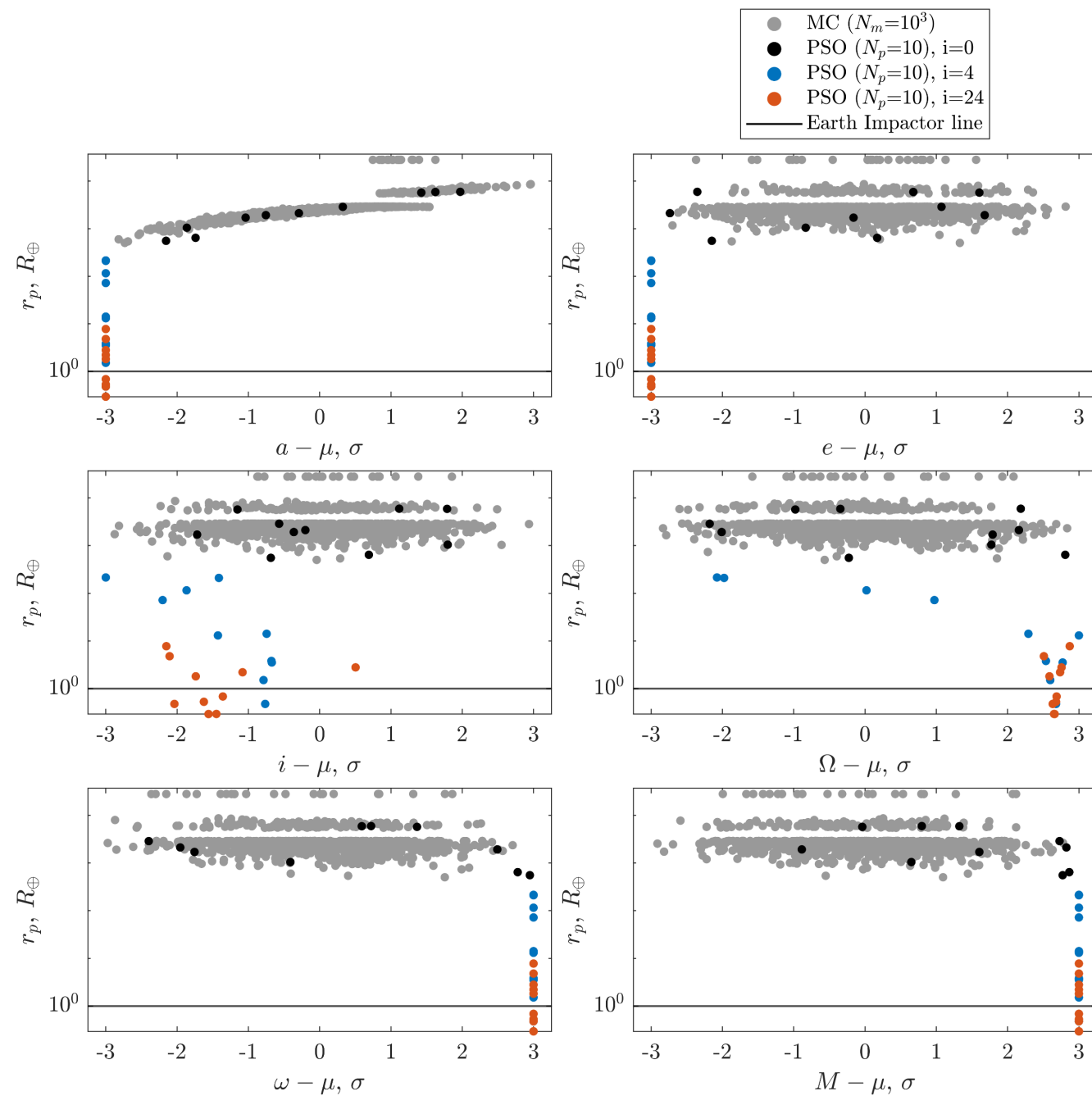
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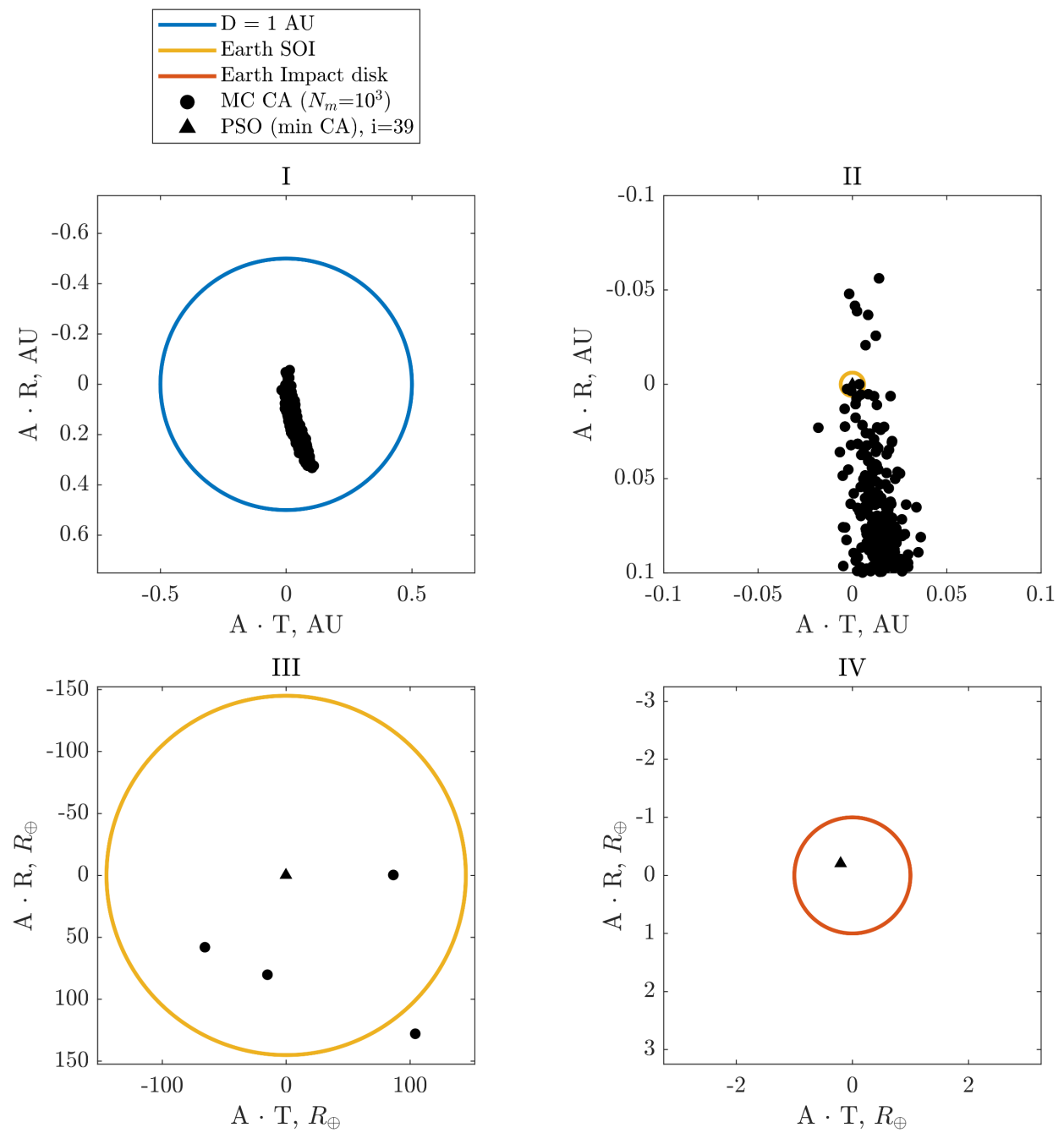
# Case Study 2) Bennu



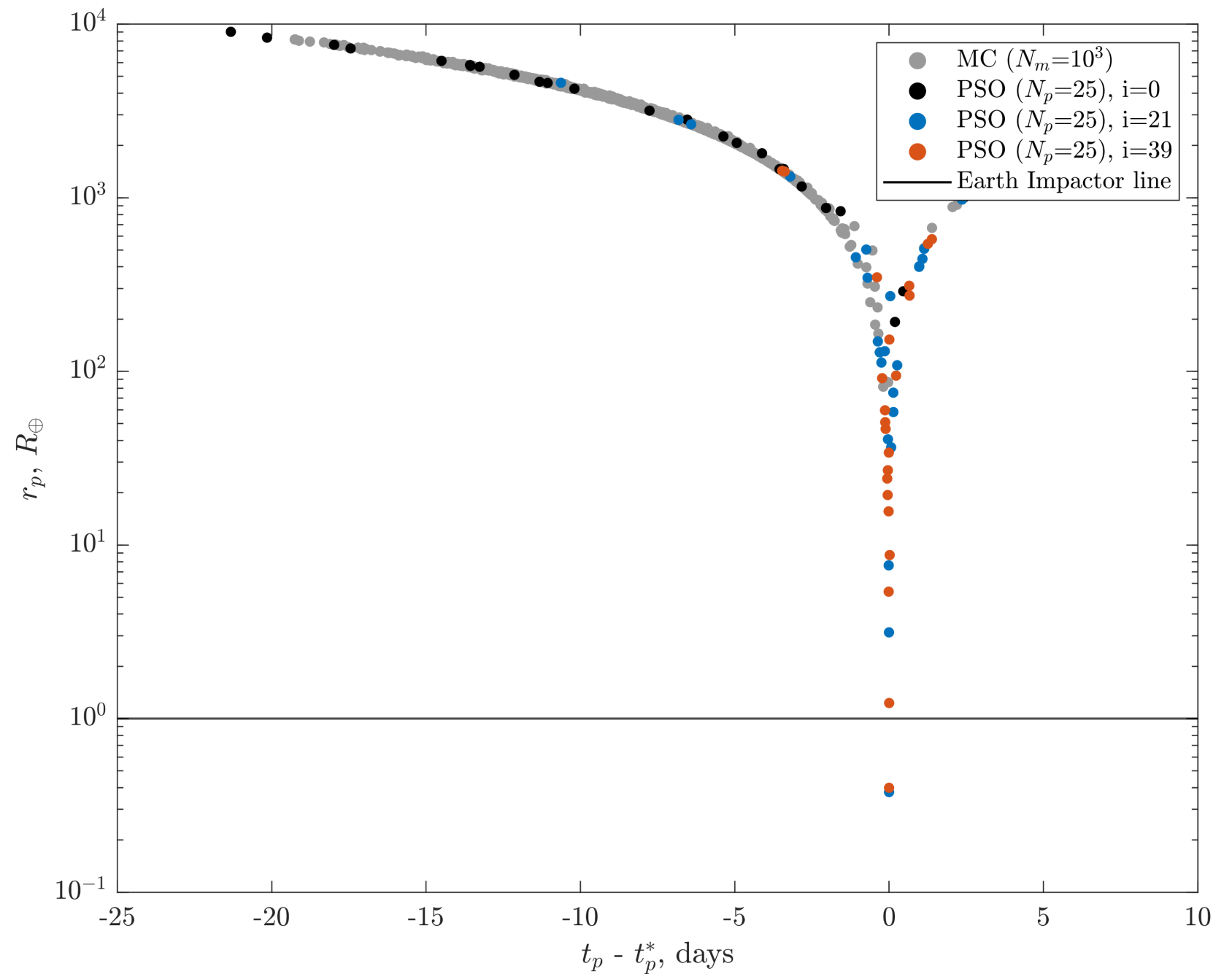
# Case Study 2) Bennu



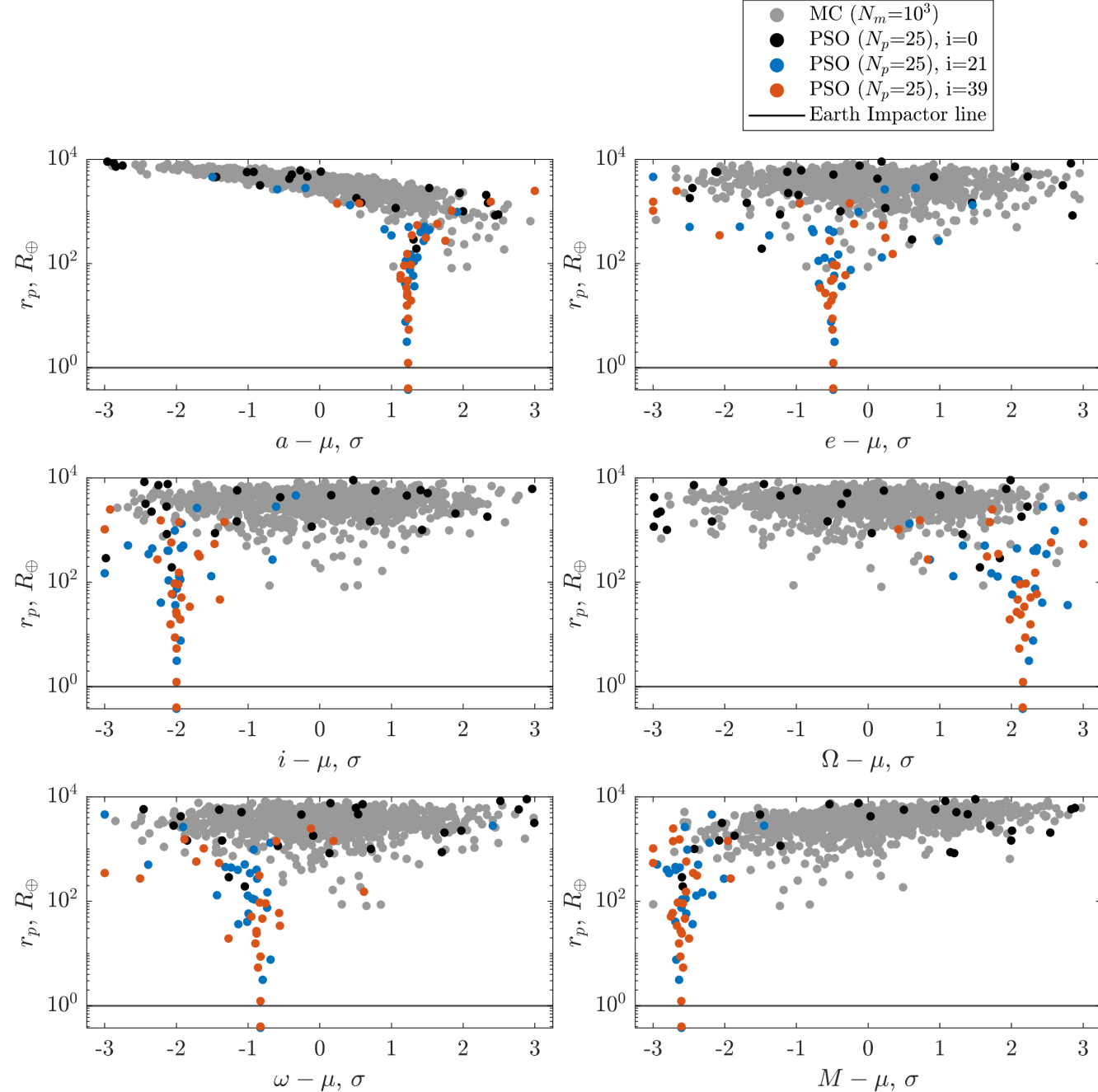
# Case Study 3) Swift-Tuttle



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# Conclusions

- The introduced method is an improvement in computational efficiency and effectiveness at finding virtual impactors that are not detected in a Monte Carlo simulation.
- The PSO PHO close approach method accurately and efficiently predicts the “worst-case scenario” for a given potentially hazardous object and its orbit determination solution.