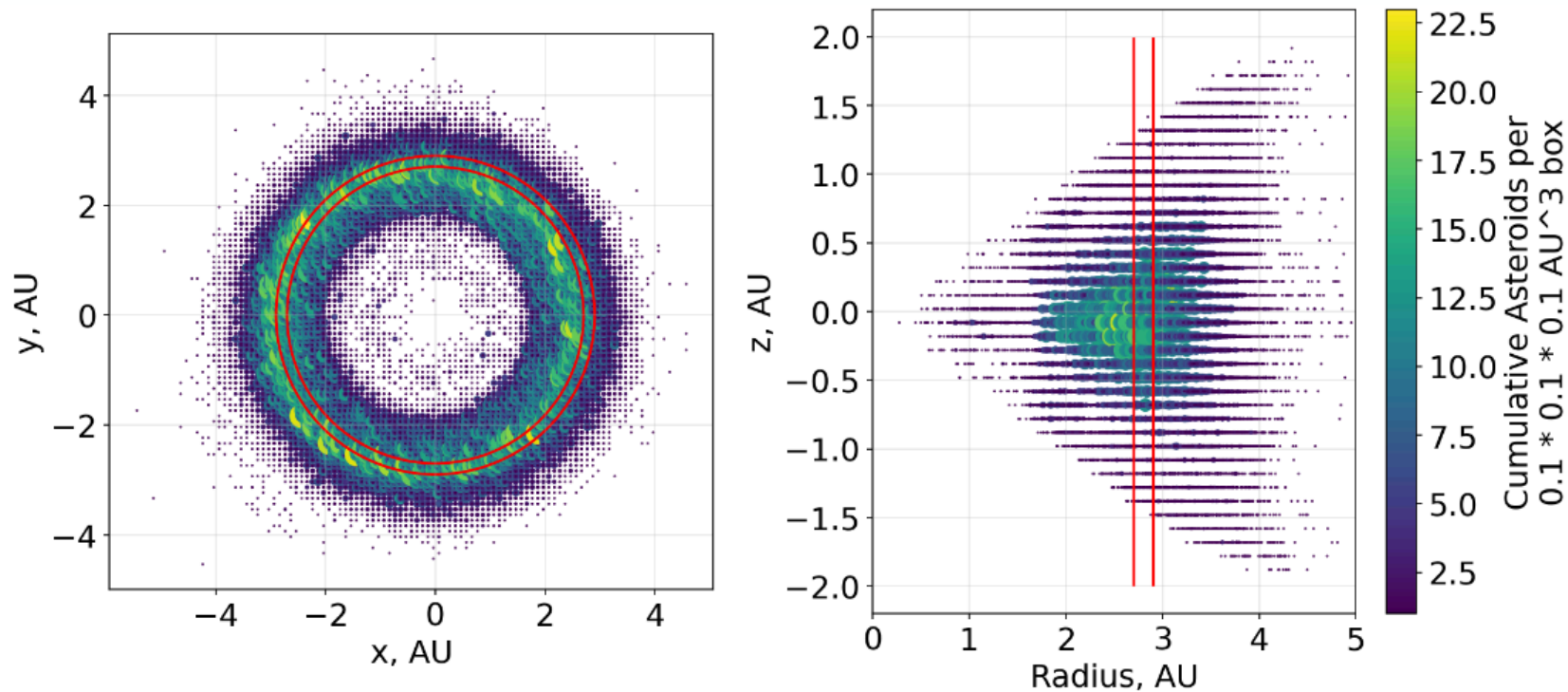


Asteroid Mining: ACT&Friends' Results for the GTOC12 Problem

The Team



- **Leaving & Returning to Earth** – potentially using Mars & Venus fly-bys
- **Deploy and Mine** – this involves extensive search for low-thrust rendezvous transfers between asteroids



Highly-dense region in the [2.7,2.9] AU range

Leaving & Returning to Earth

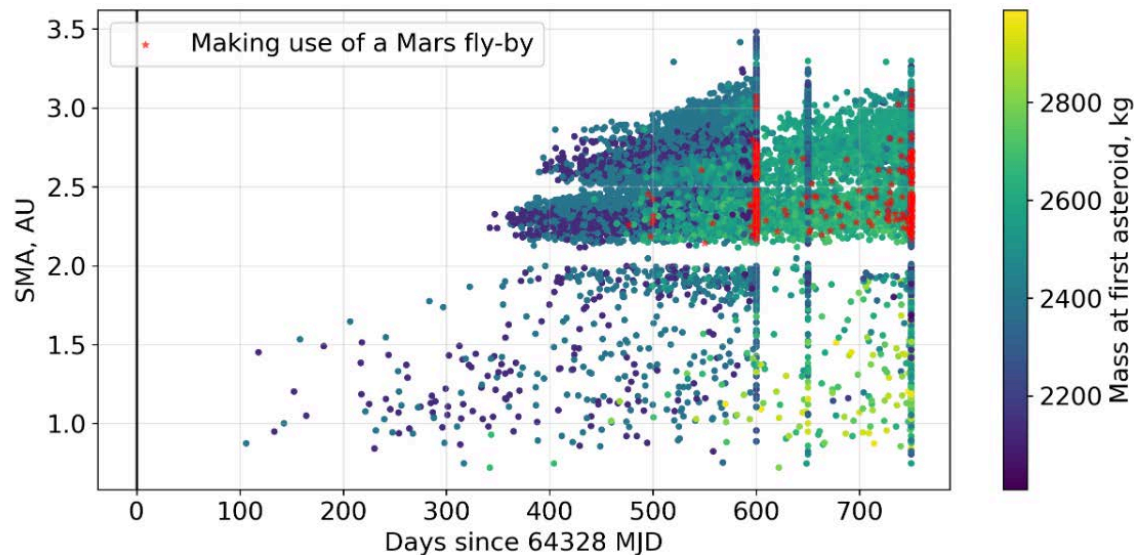


- For each asteroid, we introduce and solve **two** distinct **OCPs** (**min** arrival **time**, **max** arrival **mass**):

$$\mathcal{P}_1 : \begin{cases} \text{min: } t_f \\ \text{subject to: } m_f \geq m_{f \min} \\ t_s \geq 64328 \text{ MJD} \\ \dots \end{cases}$$

$$\mathcal{P}_2 : \begin{cases} \text{max: } m_f \\ \text{subject to: } t_f \leq t_{f \max} \\ t_s \geq 64328 \text{ MJD} \\ \dots \end{cases}$$

- Venus & Mars flybys: initial pruning based on **single DSM** model

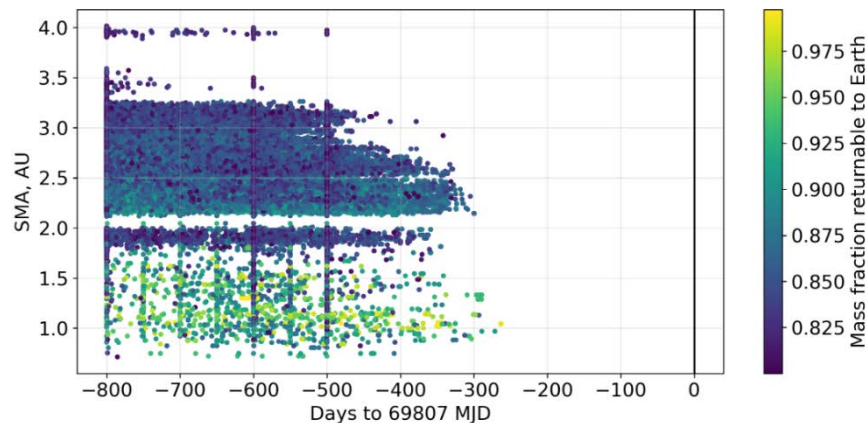


- we reduced to 18,000 out of the 60,000
- 665 were **Mars** flyby, and none **Venus** flyby
- all results were stored in a **database** of departure legs
- the database consists of **asteroid ID**, **time of arrival**, and **remaining mass**

- For each asteroid, we solve an OCP (maximum final mass aka):

$$\mathcal{P}_3 : \begin{cases} \text{max: } m_f \\ \text{subject to: } t_s \geq 69807 - \Delta t_{\max} \\ t_f \leq 69807 \\ \dots \end{cases}$$

- Effectively, this means **maximizing** the amount of **material** that can be returned from each asteroid



- the **OCP** was solved for different **maximum time of flights** (ranging from 500 to 800)
- **starting mass** is fixed at 1,200 kg
- all results were stored in a database of **return legs**
- **no flyby** seemed to help

Deploy and Mine

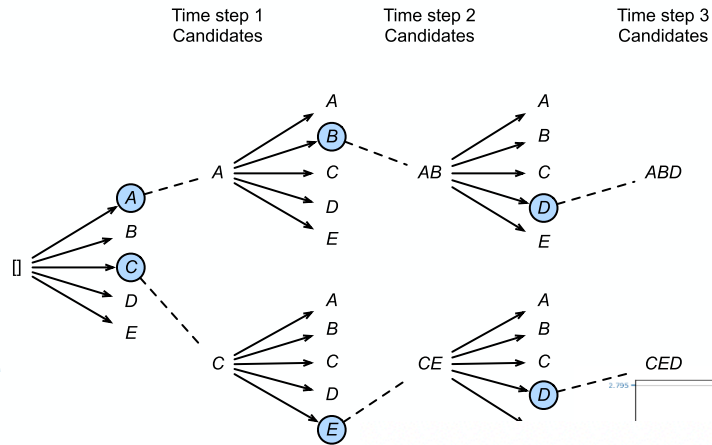
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Deploy and Mine – Part 1

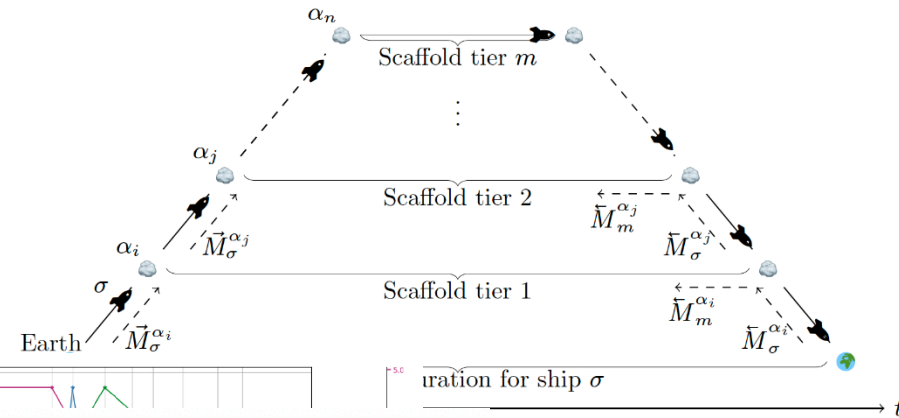
Design of Self-Sufficient Ships

- We used two strategies to approach the **combinatorial** part of the problem:

Beam search with lookahead



Trajectory scaffolding



```

1  algorithm future-score( $\alpha_i, t_i, m_i$ )
2     $h \leftarrow$  the hop that lead the beam search to  $(\alpha_i, t_i, m_i)$ 
3     $q_1 \leftarrow f(h)$ 
4     $q_2 \leftarrow \min\{f(\text{rev}(h) \oplus T \cdot 365.25) \mid T \in \{3, \dots, 9\}\}$ 
5    return  $q_1 + q_2$ 

```

Deploy and Mine – Part 2

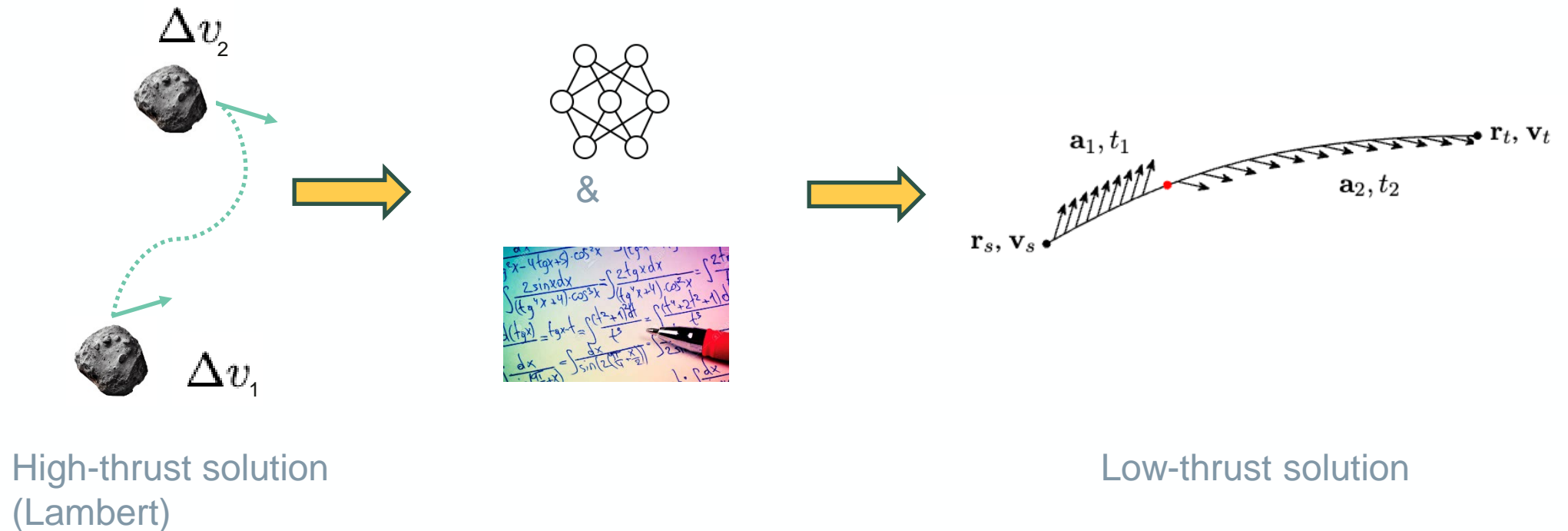
Computing Low-Thrust Hops

- Ideally, **every time** we assess a combination:
 - we should solve the associated low-thrust OCP **for different time of flights**
 - this is not feasible → hundreds of ms each!! (needs to be done billion times)
 - The control problems to solve are:
 - First find the **minimum** possible **time** for the jump (i.e., time-optimal control problem)
 - Then solve the **minimum propellant** associated control problems, starting with that time-optimal time of flight

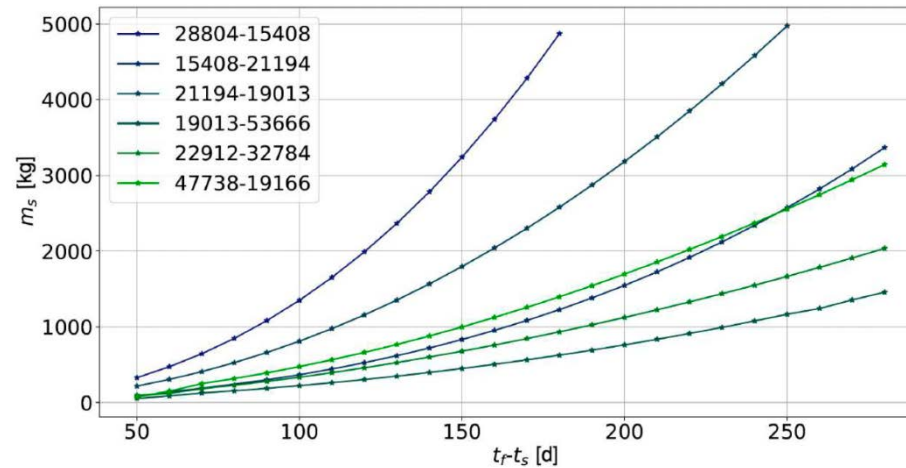
$$\mathcal{P}_{\text{MIM}} : \left\{ \begin{array}{l} \text{given: } \alpha_s, \alpha_f, t_s, t_f \\ \text{find: } \mathbf{u}(t) \\ \text{to maximize: } m_s \\ \text{subject to: } \mathbf{x}(t_s) - \mathbf{x}_{\alpha_s}(t_s) = \mathbf{0} , \\ \mathbf{x}(t_f) - \mathbf{x}_{\alpha_f}(t_f) = \mathbf{0} \\ \dot{\mathbf{y}} = \mathbf{f}(\mathbf{y}, \mathbf{u}) \\ |\mathbf{u}| \leq T_{\max} \end{array} \right. \quad \mathcal{P}_{\text{MINT}} : \left\{ \begin{array}{l} \text{given: } \alpha_s, \alpha_f, t_s, m_s \\ \text{find: } \mathbf{u}(t) \\ \text{to minimize: } t_f \\ \text{subject to: } \mathbf{x}(t_s) - \mathbf{x}_{\alpha_s}(t_s) = \mathbf{0} , \\ \mathbf{x}(t_f) - \mathbf{x}_{\alpha_f}(t_f) = \mathbf{0} \\ \dot{\mathbf{y}} = \mathbf{f}(\mathbf{y}, \mathbf{u}) \\ |\mathbf{u}| \leq T_{\max} \end{array} \right.$$

Low-Thrust Approximations

- We spent quite some time to figure out ways to compute the low-thrust solution without solving **OCP** (for both time and fuel optimal ones)
- We used for this:
 - **Astrodynamics manipulations**
 - **Machine learning**



- Given a fixed-time transfer between two asteroids (a hop), two important questions:
 - What is the **maximum initial mass** (i.e., is the hop possible)?
 - What is the **minimum propellant cost** of the hop given a time of flight?



- Maximum initial mass approximation (**MIMA**) could be used to estimate the maximum mass (GTOC7):

$$m_{\text{MIMA}}^* = 2 \frac{T_{\text{max}}}{a_D} \left(1 + \exp \left(\frac{-a_D T}{I_{sp} g_0} \right) \right)^{-1}$$

- Can we do **better**? ... it turns out we can

$$\dot{\mathbf{x}} = \mathbf{F}(\mathbf{x}) + \tilde{\mathbf{a}}(t)$$



We expand the dynamics at first order

$$\delta\dot{\mathbf{x}} = \mathbf{F}'(\mathbf{x}_K)\delta\mathbf{x} + \tilde{\mathbf{a}}(t)$$



$$\delta\mathbf{x}(t) = \mathbf{M}(t)\delta\mathbf{x}_0$$

The Keplerian state transition matrix appears



Introducing $\mathbf{z}(t) = \delta\mathbf{x}_0(t)$
(variation of parameters)

$$\dot{\mathbf{z}} = \mathbf{M}^{-1}(t)\tilde{\mathbf{a}}(t)$$



$$\delta\mathbf{z}_T = \delta\mathbf{z}_0 + \int_0^T \mathbf{M}^{-1}(s)\tilde{\mathbf{a}}(s) ds$$

The final solution can be expressed in this form

- Then, similarly to the MIMA, we impose that the thrust profile is made of **two arcs**, with constant thrusts, of the **same magnitude**
- By substituting the boundary conditions leads to the following solution (e.g. via **Simpson** rule):

$$\delta \mathbf{x}_T - \mathbf{M}_T \delta \mathbf{x}_0 = \frac{\mathbf{M}_T}{6} \left(\left(\mathbf{M}_0^{-1} + 4\mathbf{M}_{t_1/2}^{-1} + \mathbf{M}_{t_1}^{-1} \right) \Delta \mathbf{v}_1^* + \left(\mathbf{M}_{T-t_2}^{-1} + 4\mathbf{M}_{T-t_2/2}^{-1} + \mathbf{M}_T^{-1} \right) \Delta \mathbf{v}_2^* \right)$$

with: $\Delta \mathbf{v}_1^* = [\mathbf{0}, \mathbf{a}_1 t_1]^\top$ & $\Delta \mathbf{v}_2^* = [\mathbf{0}, \mathbf{a}_2 (T - t_1)]^\top$

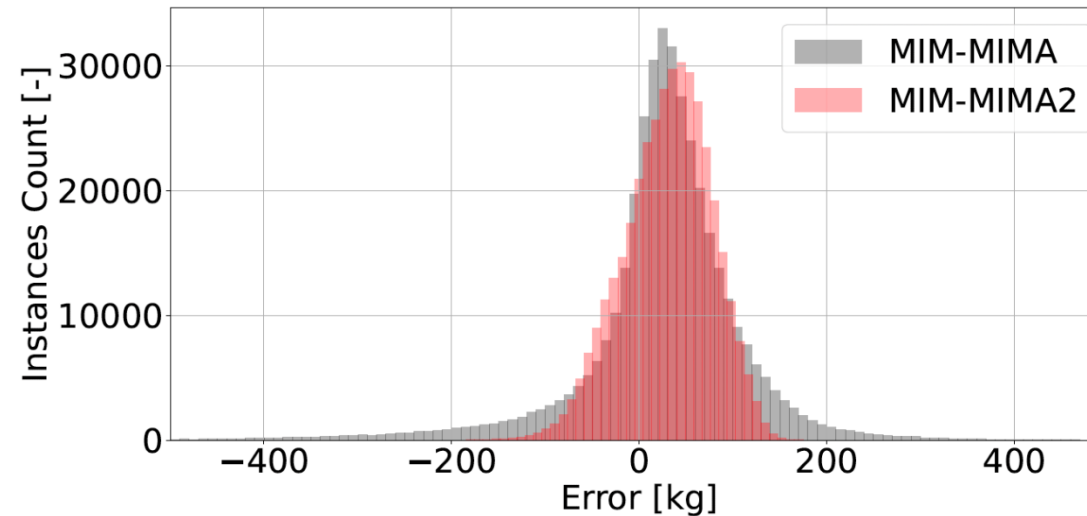
- This is a system of 6 equations in 7 unknowns (the 2 DVs, and the switching time). By imposing **equal magnitude** of the two accelerations, we add an extra equation and solve the system

$$(T - t_1) |\Delta \mathbf{v}_1^*| = t_1 |\Delta \mathbf{v}_2^*|$$

- Et voilà! We have obtained an **approximation** of the maximum initial mass, only at the cost of computing the STM for a few points on the **ballistic transfer!**

$$m_{\text{MIMA2}}^* = \frac{T_{\text{max}}}{a}$$

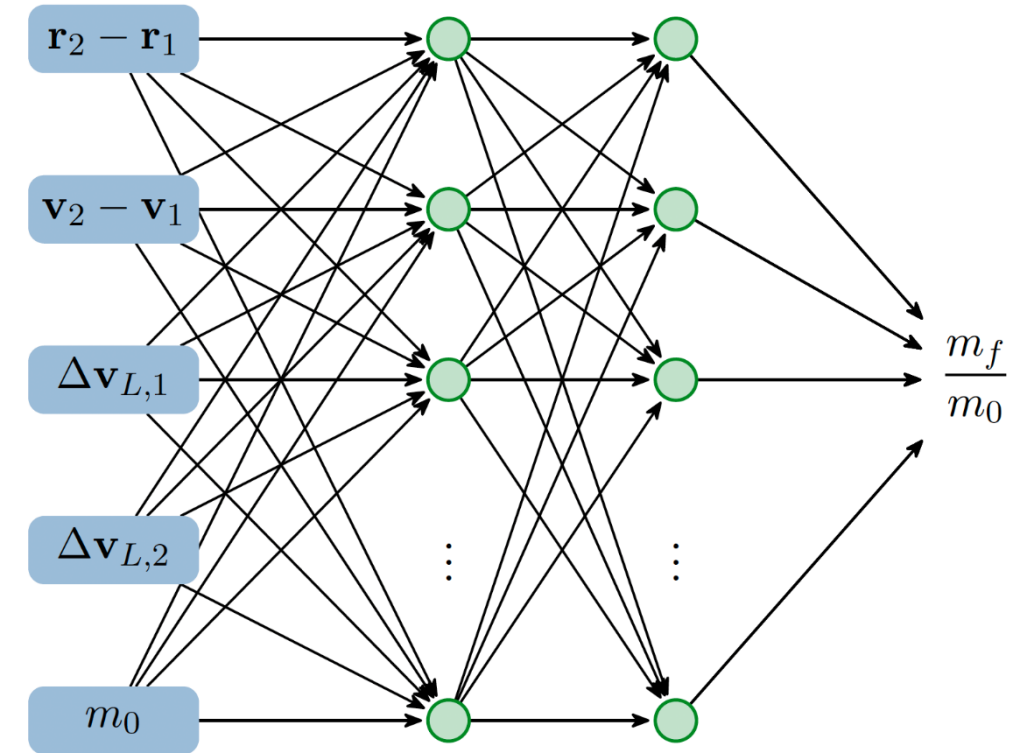
- How does **MIMA2** perform vs **MIMA**?



- With similar reasoning, we can also approximate the DV cost of a hop with mass below the MIM - this time the initial mass is known. but the switching time are both unknown (since there is intermediate **coasting**). So we add:

$$a = \frac{T_{\max}}{m_s} \quad \text{which leads to:} \quad |\Delta \mathbf{v}_1^*| = t_1 a, \quad |\Delta \mathbf{v}_2^*| = t_2 a$$
- Finally, from MIMA2 we can also extract the **MINTA2** approximation (minimum initial time): inverting the curve (e.g. via root solver)

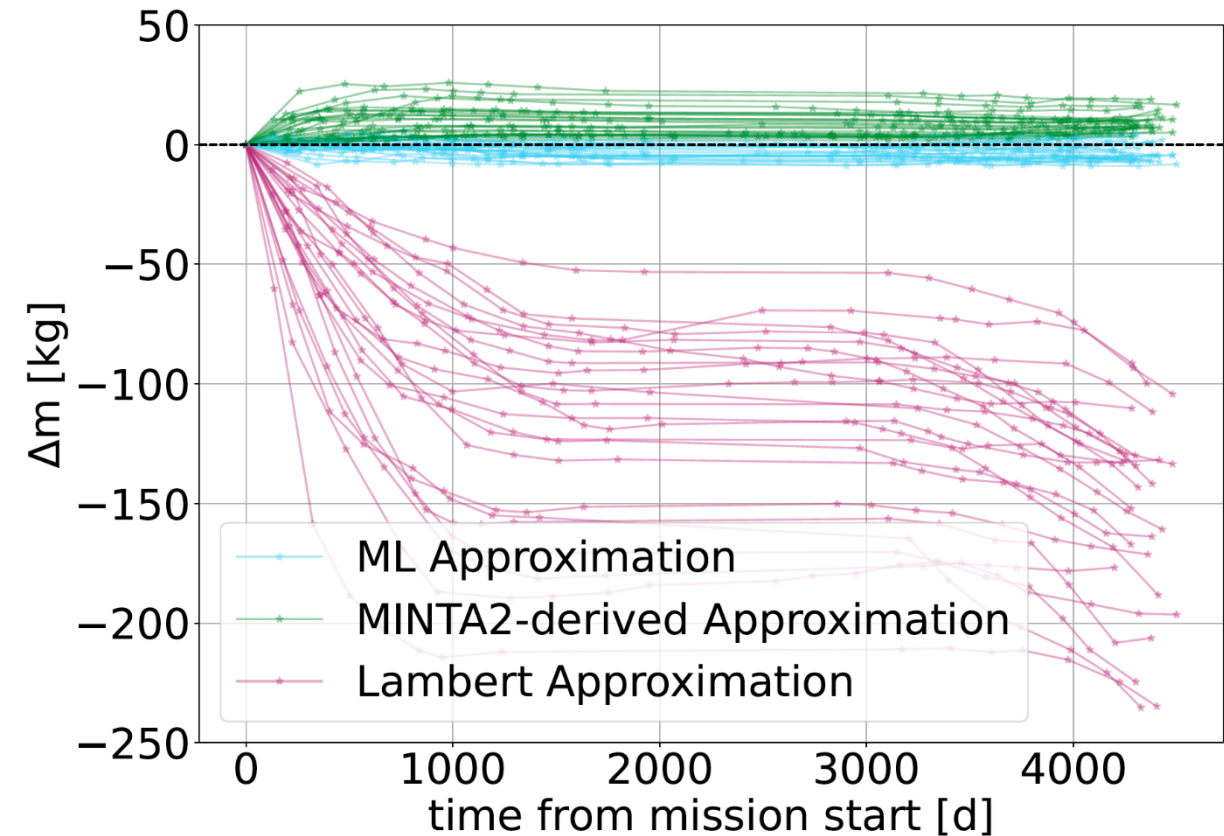
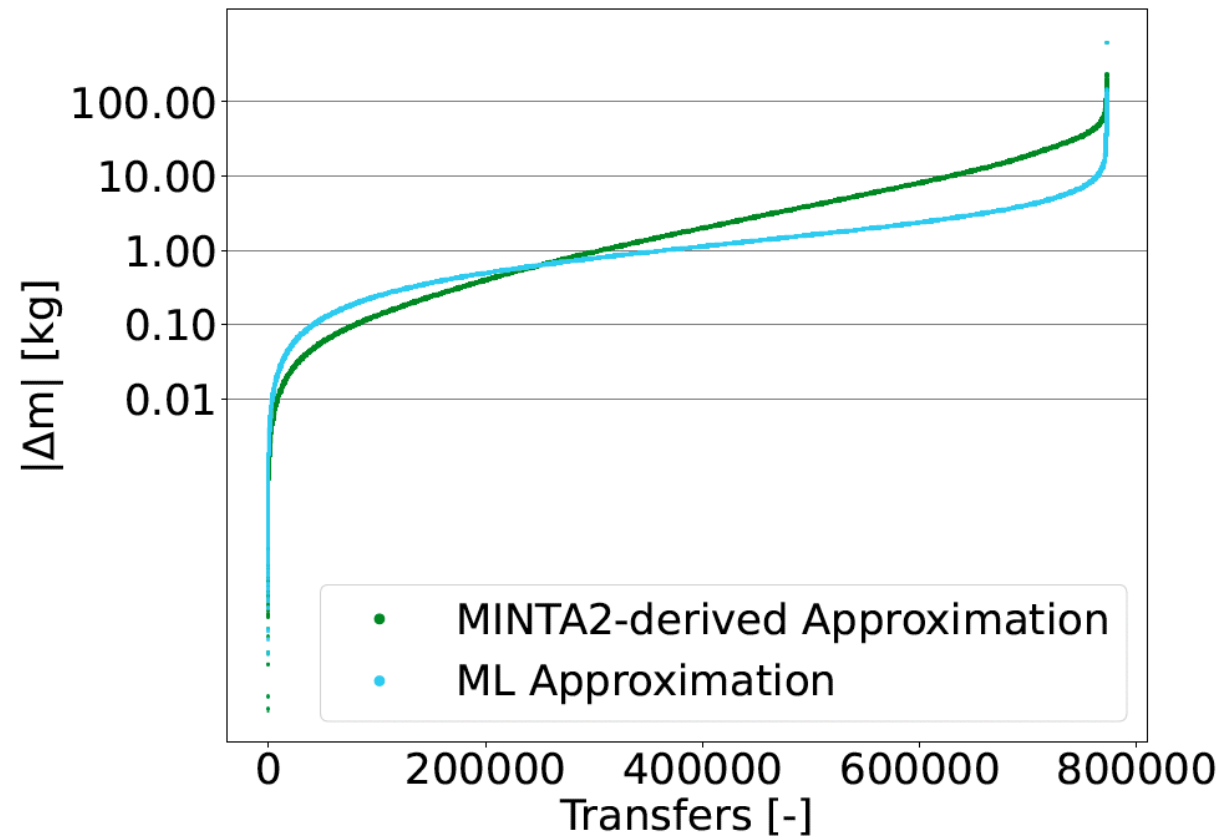
- In the same spirit, we sought to solve the problem with **ML**
- We used the **Lambert** transfer DVs as **attributes**, together with the **relative position & velocity** of the two asteroids, and the **initial mass**
- We made sure that the NN treats in the same way hops with the same geometry and distance from the Sun → hence why we used their **relative** positions as attributes
- We trained on **800,000** hops computed solving the OCP



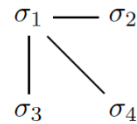
- We now obtained **analytical** & **ML** approximations
- The timings are good.... More than **100,000x** computational gain from the OCP solution
- The ML approximation is **3x** faster
- And in terms of **accuracy**? Do machines have an edge vs humans?



- Errors remain for most cases **below 1kg!**
- They even maintain error ranges below 5/10 kg across 20 and more consecutive hops!



- Having a **pool** of promising ships, we had to select a **subset** to make our final solution
- The objective was to **maximize** the cumulative **collected mass**, accounting for **penalties**, while adhering to the crucial **non-overlap condition**
- We mapped this as an **integer linear programming (ILP)** problem:

$$\mathcal{S} = \{$$
$$\sigma_1 = (300, 1, \{a, b, c\}),$$
$$\sigma_2 = (75, 75, \{a, d\}),$$
$$\sigma_3 = (75, 75, \{b, e\}),$$
$$\sigma_4 = (100, 100, \{c, f\})$$
$$\}$$


Maximize

$$1x_1 + 75x_2 + 75x_3 + 100x_4$$

Subject To

$$x_1 + x_2 \leq 1$$
$$x_1 + x_3 \leq 1$$
$$x_1 + x_4 \leq 1$$
$$x_1, x_2, x_3, x_4 \in \{0, 1\}$$

- We solved this problem using a two-pass algorithm utilizing **0-1 linear programming**
- We also added the possibility to automatically **remove** an **asteroid**
- In this way we constructed our **final solution**

