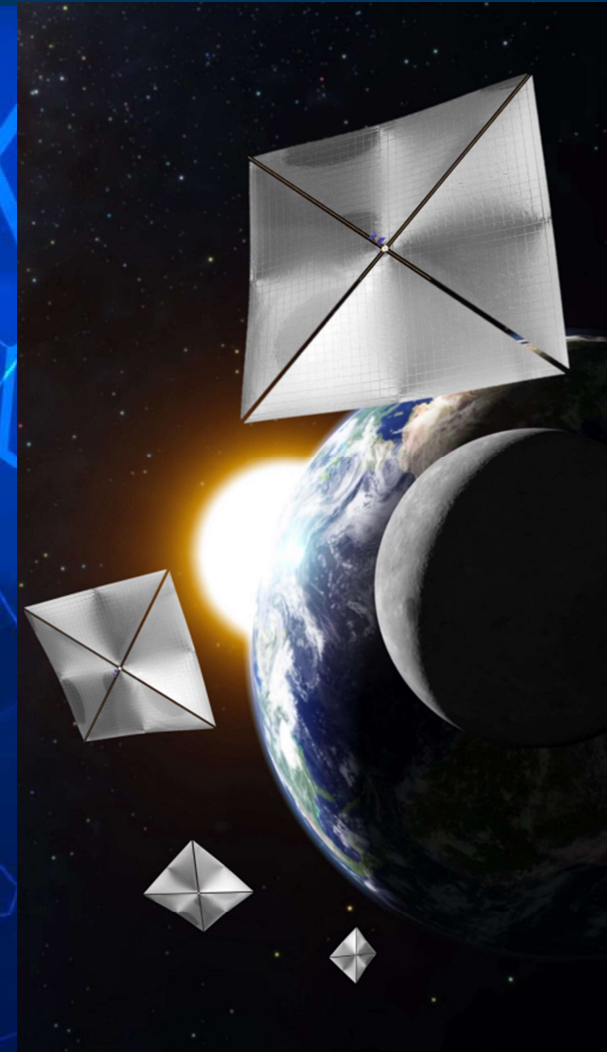


Advances in preliminary solar-sail trajectory design

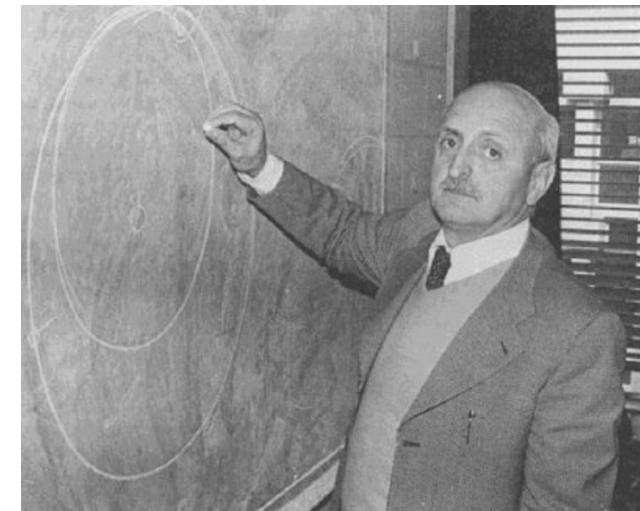
Shape-Based Methods and Artificial Neural Networks

Matteo Ceriotti, Giulia Viavattene, Andrea Caruso, Giovanni Mengali, Alessandro Quarta



Why preliminary trajectory design?

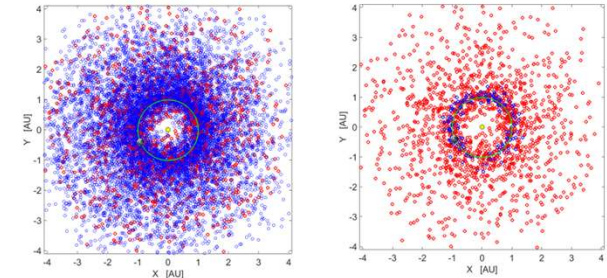
- Early mission design phase
 - Concept study, pre-phase A
- Fast preliminary trajectory design
- Interested in trajectory key figures of merit:
 - Δv , TOF, maximum thrust, propellant mass, feasibility
- Not interested in full control history (as long as feasible)
- Approximate solution
 - Design margins are high
- Fast to calculate
 - Can calculate hundreds to millions of options



Giuseppe "Bepi" Colombo. Credit: ESA

Multi-target missions

- Multi-target missions, wide launch window
 - Appealing for solar sailing: “infinite” Δv
- Multi-asteroid rendezvous mission
 - Complete database: $n = \sim 13,000$ NEAs
 - Near-Earth Object Human Space Flight Accessible Target Study (NHATS) (for a low-thrust return mission)
 Reduced database: $n = \sim 1,800$ objects



NHATS criteria: {

- total Δv required ≤ 8 km/s
- total mission duration ≤ 450 days
- stay time at the object ≥ 8 days
- launch: 2015–2040
- $H \leq 26$ magnitude
- OCC ≤ 7

- Number of unique sequences of q objects out of n :

$$P_q^n = \frac{n!}{(n - q)!}$$

Database	n	$q = 3$	$q = 4$	$q = 5$
Complete database	12,840	2.1×10^{12}	2.7×10^{16}	3.5×10^{20}
Reduced database	1,801	5.8×10^9	1.0×10^{13}	1.9×10^{16}

Peloni, A., Ceriotti, M. and Dachwald, B. (2016) Solar sail trajectory design for a multiple near-Earth asteroid rendezvous mission. Journal of Guidance, Control, and Dynamics, 39(12), pp. 2712-2724. (doi: 10.2514/1.G000470)

Solar sailing vs. low thrust (EP)

- Continuous, low acceleration

Solar sailing

- Thrust direction and magnitude related
- Dependent on sun direction

$$\mathbf{a} = a_c \left(\frac{r_{\oplus}}{r} \right)^2 \cos^2 \alpha \hat{\mathbf{N}}$$

$$a_c = 2P \frac{A}{M}$$

$$\hat{\mathbf{N}} = \begin{bmatrix} N_r \\ N_g \\ N_h \end{bmatrix} = \begin{bmatrix} \cos \alpha \\ \sin \alpha \cos \delta \\ \sin \alpha \sin \delta \end{bmatrix}$$

$$\alpha = \angle(\mathbf{r}_{sun}, \hat{\mathbf{N}})$$

$$\alpha < 90^\circ$$

- Objective:
 - Time of flight

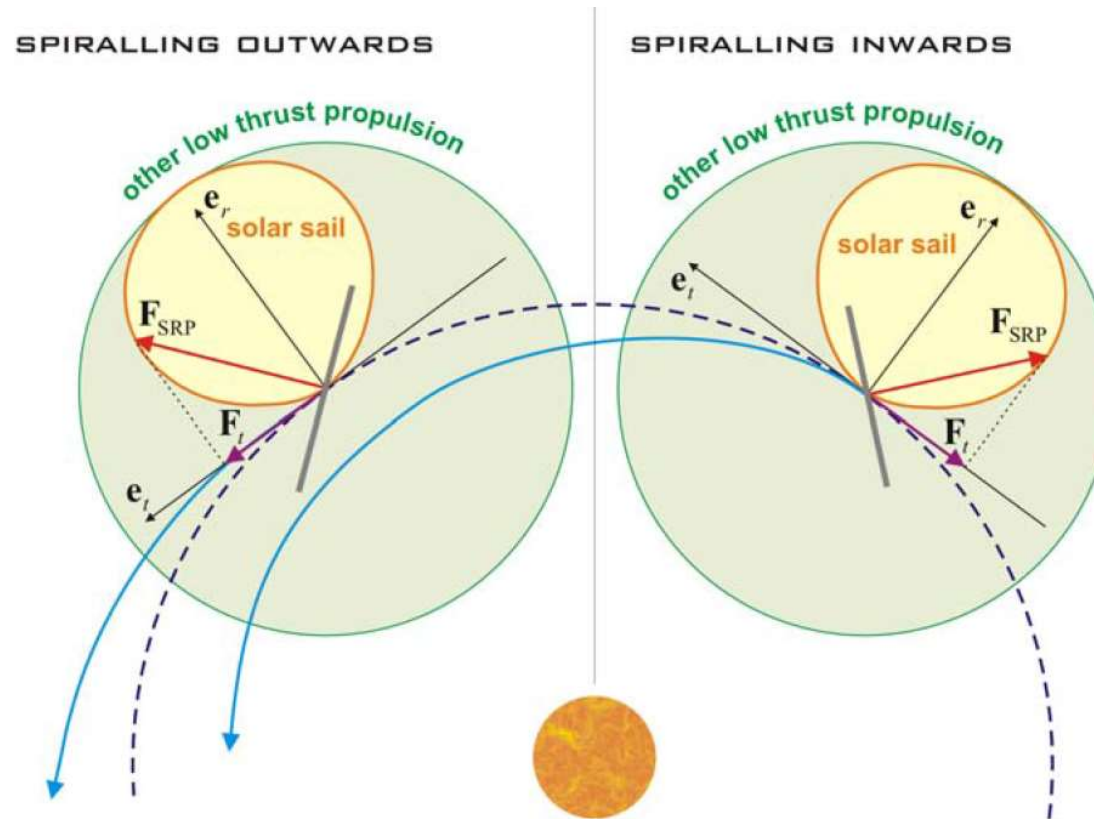
Low Thrust (EP)

- Free thrust direction
- Free magnitude (up to the max)

$$\mathbf{a} = \begin{bmatrix} a_x \\ a_y \\ a_z \end{bmatrix}, \quad |\mathbf{a}| < a_{max}$$

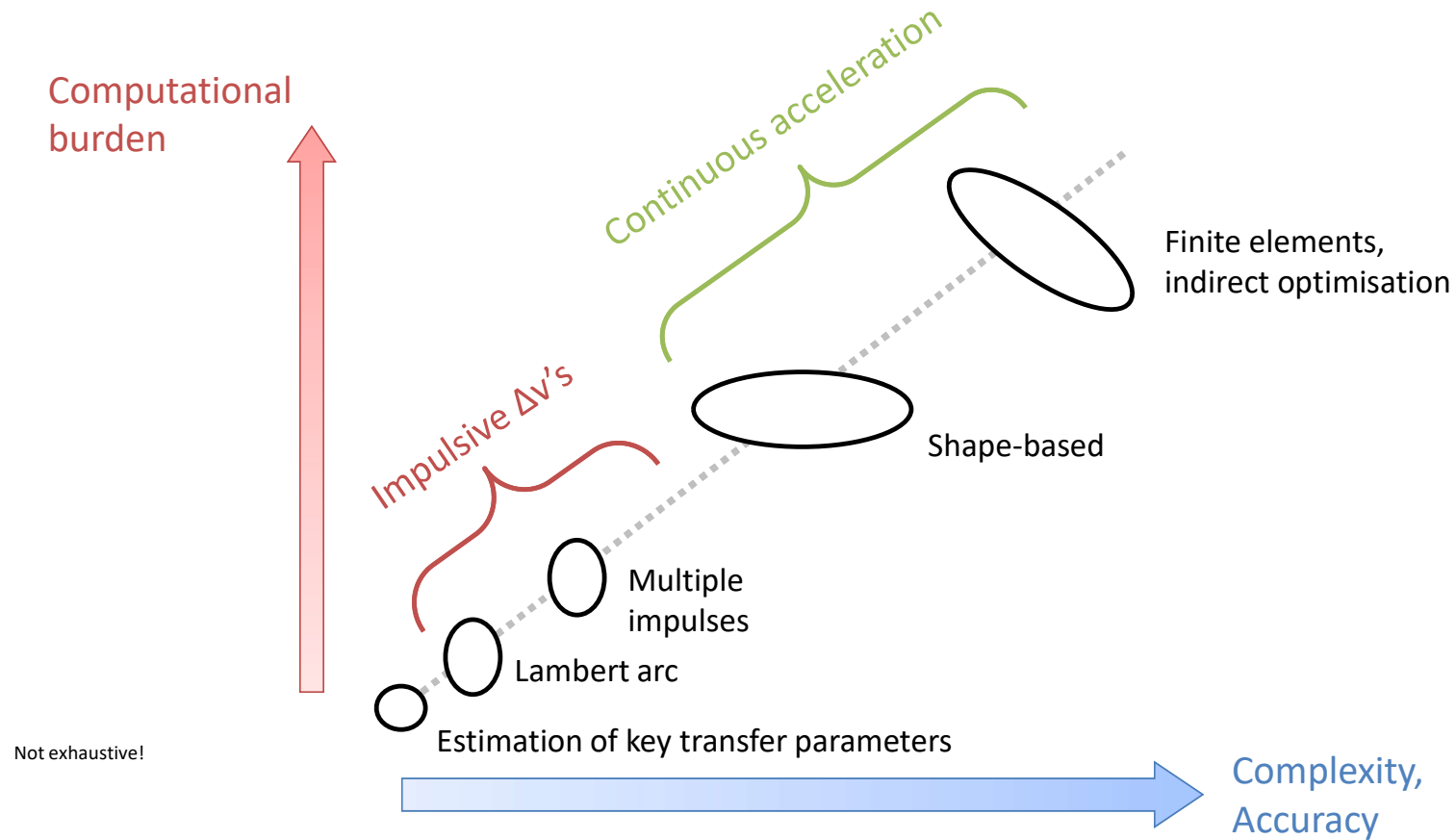
- (Multi-) Objective:
 - Propellant mass
 - Time of flight

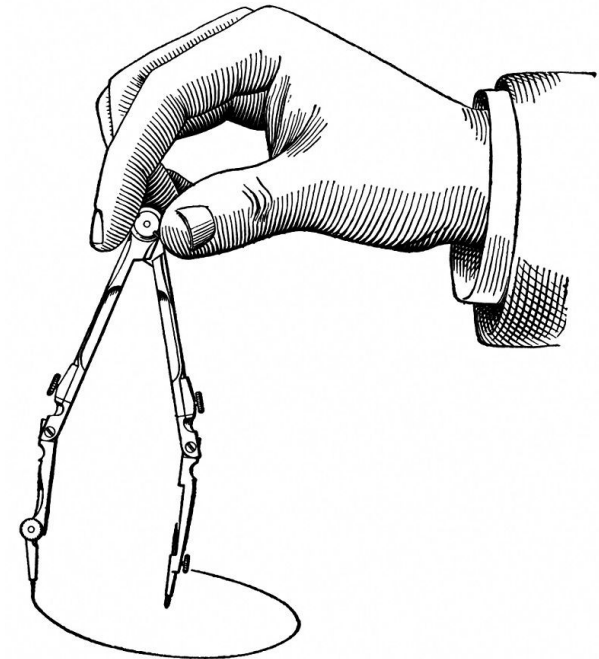
Solar sailing vs. low thrust (EP)



Dachwald, B., "Low-Thrust Trajectory Optimization and Interplanetary Mission Analysis Using Evolutionary Neurocontrol", Institut für Raumfahrttechnik, Universität der Bundeswehr München, 2004.

Trajectory models





SHAPE-BASED TRAJECTORIES FOR SOLAR SAILING

Shape-based methods

- Take the Equations of Motion: $\ddot{\mathbf{x}} = f(\mathbf{x}, t) + \mathbf{a}(t)$
- Assign a “shape”, i.e. trajectory function of time: $\mathbf{x}(t) = \mathbf{x}^*(t, \mathbf{p})$, $\dot{\mathbf{x}}$, $\ddot{\mathbf{x}}$
- Set shape parameters \mathbf{p} to meet the boundary conditions (e.g. rendezvous): $\mathbf{x}^*(t_0, \mathbf{p}) = \mathbf{x}_0$, $\mathbf{x}^*(t_f, \mathbf{p}) = \mathbf{x}_f$
- “Invert” the EoM to find the control (acceleration) profile: $\mathbf{a}(t) = \ddot{\mathbf{x}}^* - f(\mathbf{x}^*, t)$



- Usually fast to calculate
- Provides a thrust profile
 - Can be used as initial guess for OPC
- Inversion of EoM only possible for specific shapes
- Does not always satisfy the constraints!

2D shape for solar sailing

EoM Modified Equinoctial Elements

2D Shape

In-plane motion

$$\left\{ \begin{array}{l} p = a(1 - e^2) \\ f = e \cos(\Omega + \omega) \\ g = e \sin(\Omega + \omega) \end{array} \right.$$

$$p = p_I \exp[p_F(L - L_0)] + \lambda_p \sin(L + \varphi_p)$$

$$f = f_I + f_F(L - L_0) + \lambda_{fg} \sin(L + \varphi_{fg})$$

$$g = g_I + g_F(L - L_0) - \lambda_{fg} \cos(L + \varphi_{fg})$$

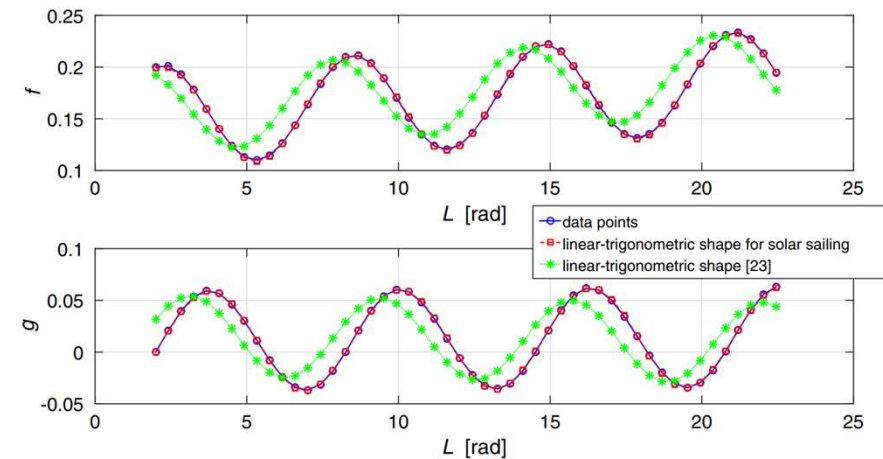
$$h = \tan(i/2) \cos(\Omega)$$

$$k = \tan(i/2) \sin(\Omega)$$

$$L = \Omega + \omega + \nu$$

Peloni, A., Ceriotti, M. and Dachwald, B. (2016) Solar sail trajectory design for a multiple near-Earth asteroid rendezvous mission. *Journal of Guidance, Control, and Dynamics*, 39(12), pp. 2712-2724. (doi: 10.2514/1.G000470)

Coplanar Earth-Mars



3D shape for solar sailing

EoM Modified Equinoctial Elements

3D Shape

In-plane motion	{	$p = a(1 - e^2)$	$p(L) = \tilde{p}_0 + \tilde{p}_f (L - L_0) + a_p (L - L_0)^2 + \lambda_p \sin(L - L_0 + \phi_p)$
		$f = e \cos(\Omega + \omega)$	$f(L) = \tilde{f}_0 + \tilde{f}_f (L - L_0) + \lambda_{fg} \sin(L - L_0 + \phi_{fg})$
		$g = e \sin(\Omega + \omega)$	$g(L) = \tilde{g}_0 + \tilde{g}_f (L - L_0) - \lambda_{fg} \cos(L - L_0 + \phi_{fg})$
Out-of-plane motion	{	$h = \tan(i/2) \cos(\Omega)$	$h(L) = \tilde{h}_0 + \tilde{h}_f (L - L_0) + a_h \exp[b_h (L - L_0)] + \lambda_{hk} \sin[2(L - L_0) + \phi_{hk}]$
		$k = \tan(i/2) \sin(\Omega)$	$k(L) = \tilde{k}_0 + \tilde{k}_f (L - L_0) + a_k \exp[b_k (L - L_0)] - \lambda_{hk} \cos[2(L - L_0) + \phi_{hk}]$
		$L = \Omega + \omega + \nu$	

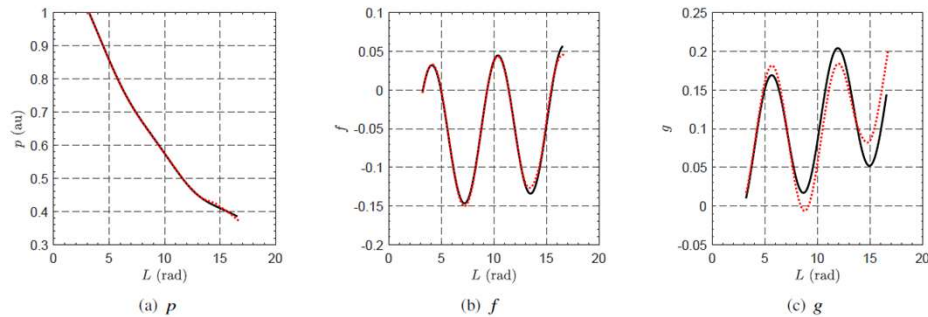


Free parameters

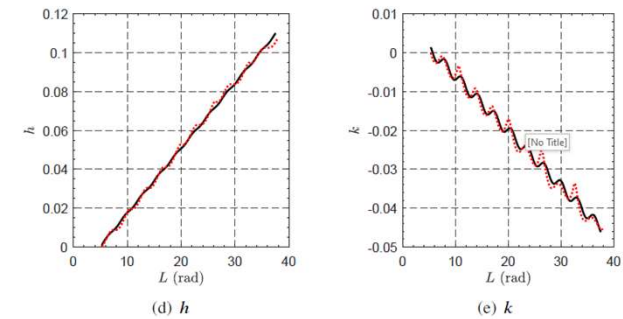
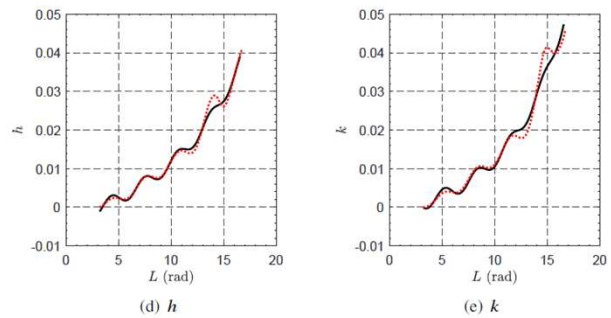
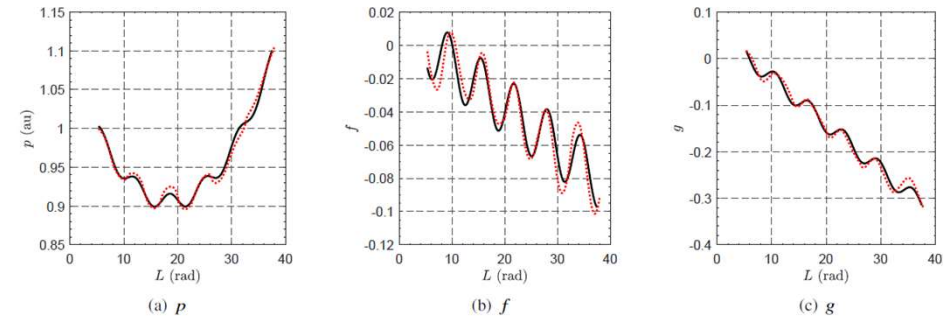
Caruso, A., Quarta, A., Mengali, G. and Ceriotti, M. (2020) Shape-based approach for solar sail trajectory optimization. Aerospace Science and Technology, 107, 106363. (doi: 10.1016/j.ast.2020.106363)

3D shape fitting

Earth-Mercury, $a_c = 0.6 \text{ mm/s}^2$



Earth-1620 Geographos ($i = 13^\circ$), $a_c = 0.2 \text{ mm/s}^2$




..... Optimal transfers computed with indirect method
 ————— Curve fitting to find free parameters

Shape-based trajectory design for solar sailing

$$\min \left\{ \text{Time of Flight} \quad T = \int_{L_0}^{L_f} \frac{dt}{dL} dL \right\}$$

$a_p, \lambda_p, \phi_p, \lambda_{fg}, \phi_{fg}, a_h, b_h, a_k, b_k, \lambda_{hk}, \phi_{hk}, L_0$ and L_f

Boundary constraints (pos, vel)  $\tilde{p}_0, \tilde{p}_f, \tilde{f}_0, \tilde{f}_f, \tilde{g}_0, \tilde{g}_f, \tilde{h}_0, \tilde{h}_f, \tilde{k}_0,$ and \tilde{k}_f

■ Solar sail constraints:

1. : $\max_t (\|\mathbf{a}\| - a_{\max}) < 0$
2. : $\min_t (a_r) > 0$
3. : $\max_t (|a_r - a_c (\frac{r_{\oplus}}{r})^2 \cos^3 \alpha|) < \epsilon$
4. : $\max_t (|a_t - a_c (\frac{r_{\oplus}}{r})^2 \cos^2 \alpha \sin \alpha|) < \epsilon$

Constraint satisfaction

Shape-based trajectory design for solar sailing

1. Shape-based trajectory:

- Genetic Algorithm + gradient-based
- 3 methods:
 1. Solar sail shape + all constraints 1-4
 2. Solar sail shape + constraints 1-2
 3. Linear-trigonometric shape (LT, Pascale, Vasile) + constraint 1

$$1. : \max_t (\|\mathbf{a}\| - a_{\max}) < 0$$

$$2. : \min_t (a_r) > 0$$

$$3. : \max_t (|a_r - a_c (\frac{r_{\oplus}}{r})^2 \cos^3 \alpha|) < \epsilon$$

$$4. : \max_t (|a_t - a_c (\frac{r_{\oplus}}{r})^2 \cos^2 \alpha \sin \alpha|) < \epsilon$$



2. Optimisation with full dynamics:

- Multiple-shooting
- Shape-based solution used as initial guess

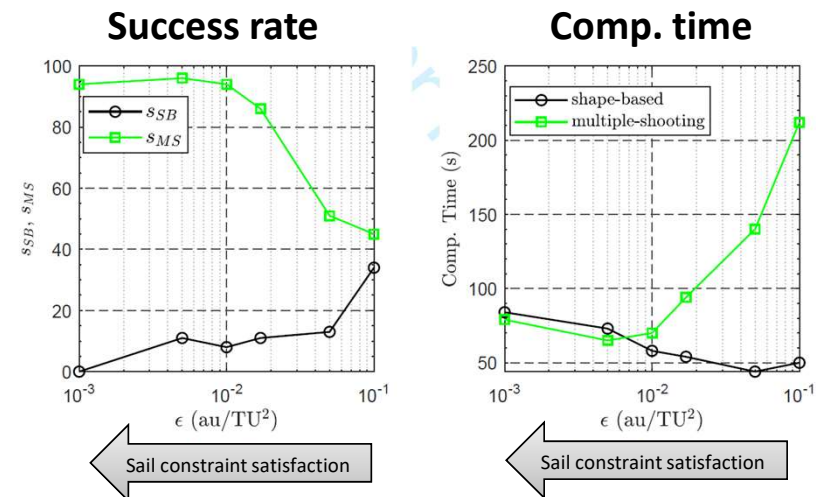
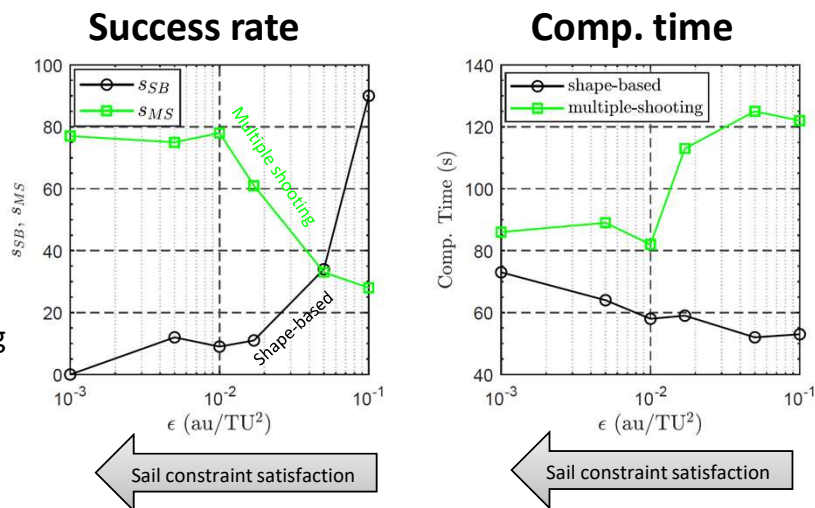
Performance evaluation

Earth - 2002 DU₃ $a_c = 0.3 \text{ mm/s}^2$

Earth - 2007 MK₁₃ $a_c = 0.3 \text{ mm/s}^2$

Method 1:

Shape-based
Multiple shooting



Method 2:

Method 3:

	Shape-based		Multiple shooting	
	Success rate	Comp. time	Success rate	Comp. time
Method 2:	93 %	58 s	38 %	104 s
Method 3:	50 %	52 s	16 %	187 s

	Shape-based		Multiple shooting	
	Success rate	Comp. time	Success rate	Comp. time
Method 2:	100 %	45 s	46 %	221 s
Method 3:	100 %	16 s	2 %	284 s

Results

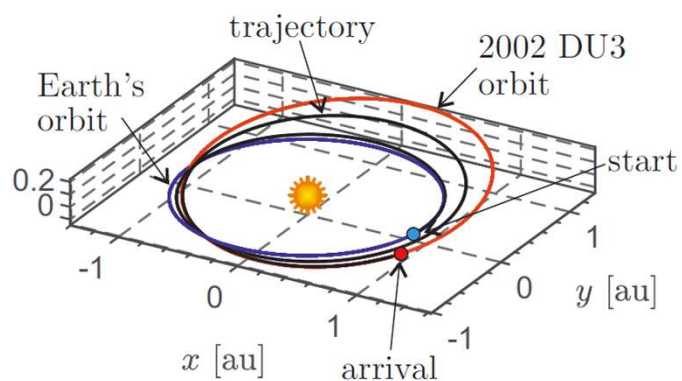
$$a_c = 0.2 \text{ mm/s}^2$$

Method 1

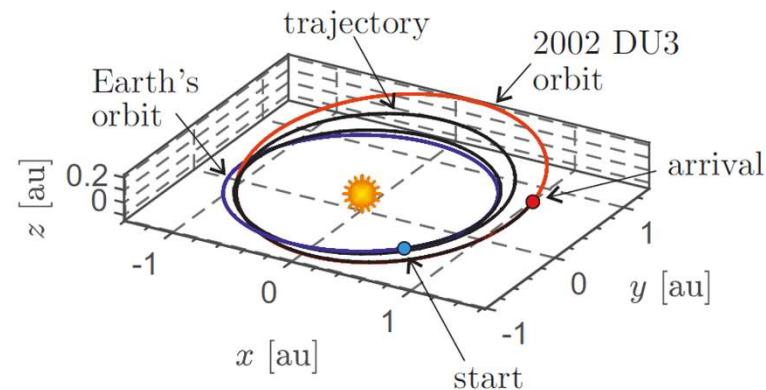
$$\varepsilon = 10^{-2}$$

Earth – 2002 DU₃

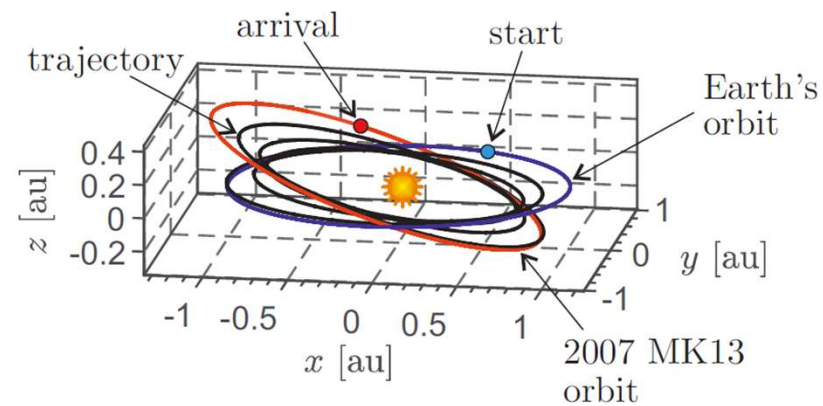
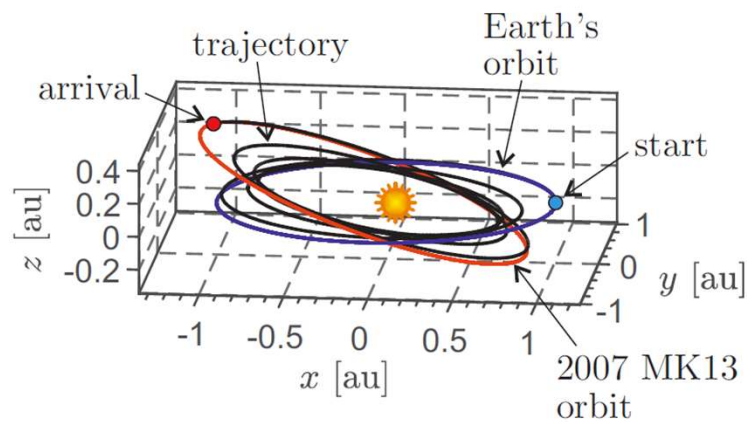
Shape-based

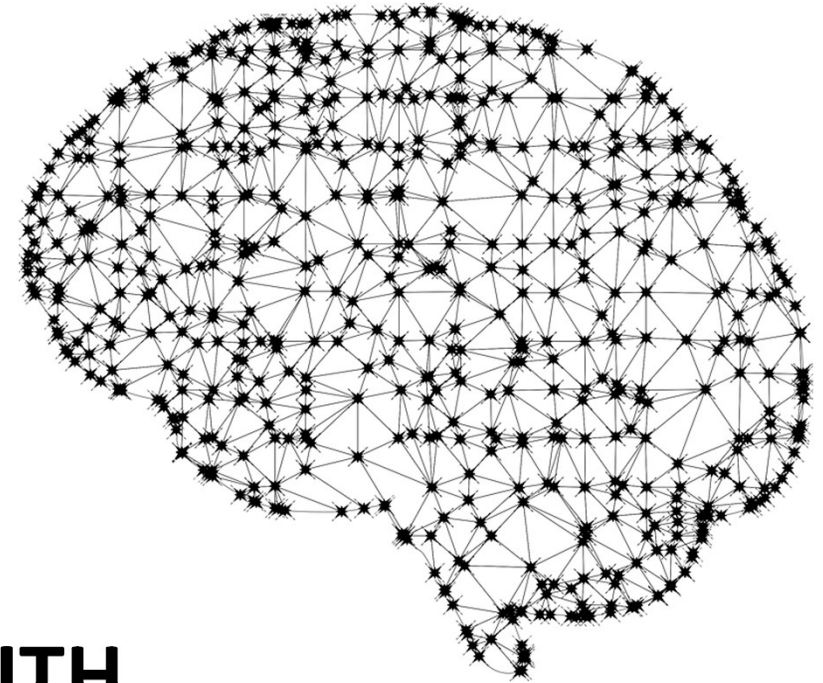


Multiple shooting (optimal)



Earth – 2007 MK₁₃

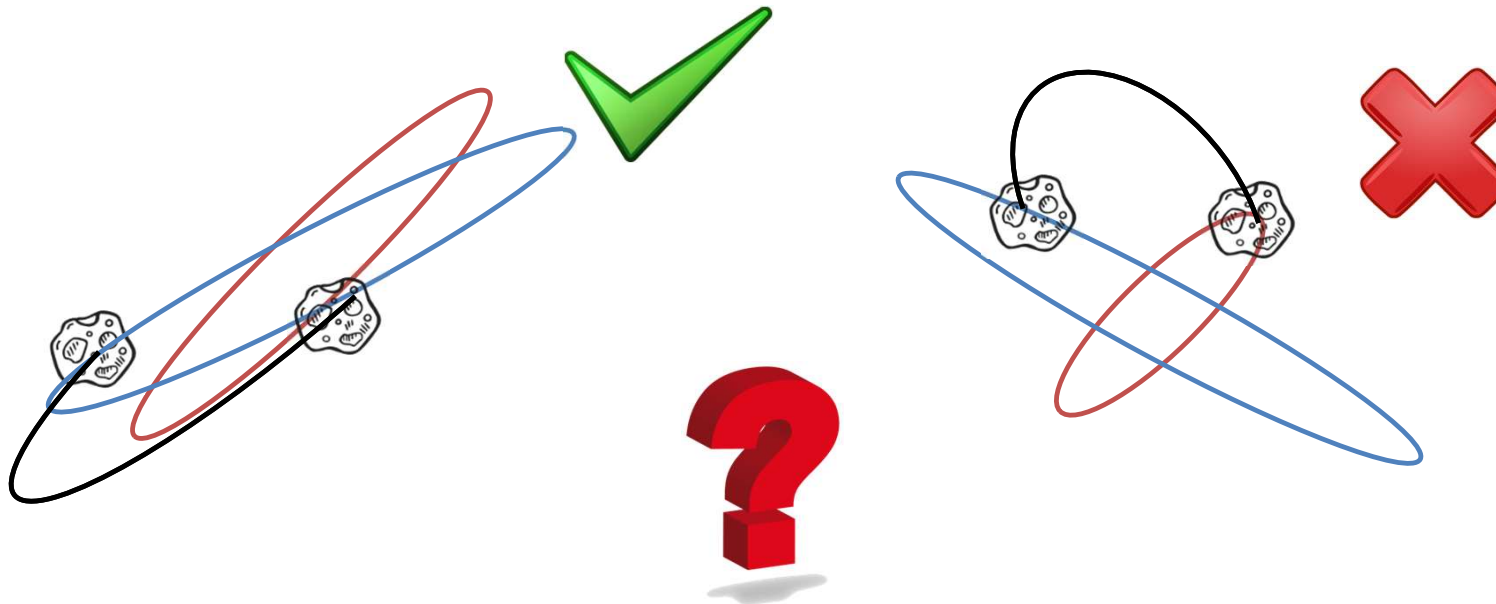




SAIL TRAJECTORY DESIGN WITH ARTIFICIAL NEURAL NETWORKS

An even faster trajectory estimation

- Can a machine to “learn” to estimate the “cost” of transfers between asteroids?

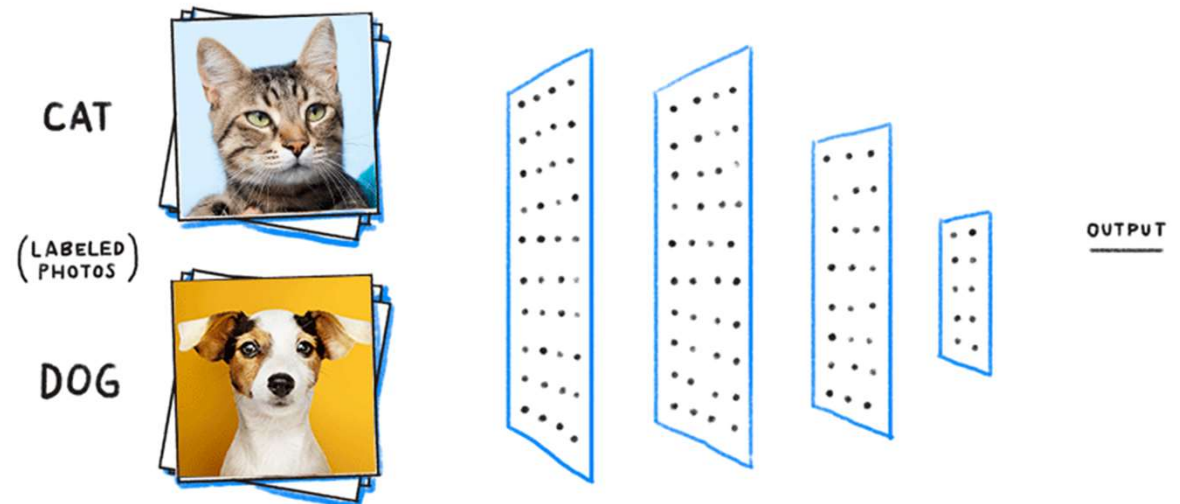


Artificial Neural Networks

Select all images with a **bus**
Click verify once there are none left.

⏪ 🔊 ⓘ **VERIFY**

HOW DOES MACHINE LEARNING WORK?



Kate Libbie, QATestLab

Neural Network Design

TRAINING DATABASE

Collection of $(\mathbf{x}, \mathbf{t})_i$ with $i \in [1, N]$

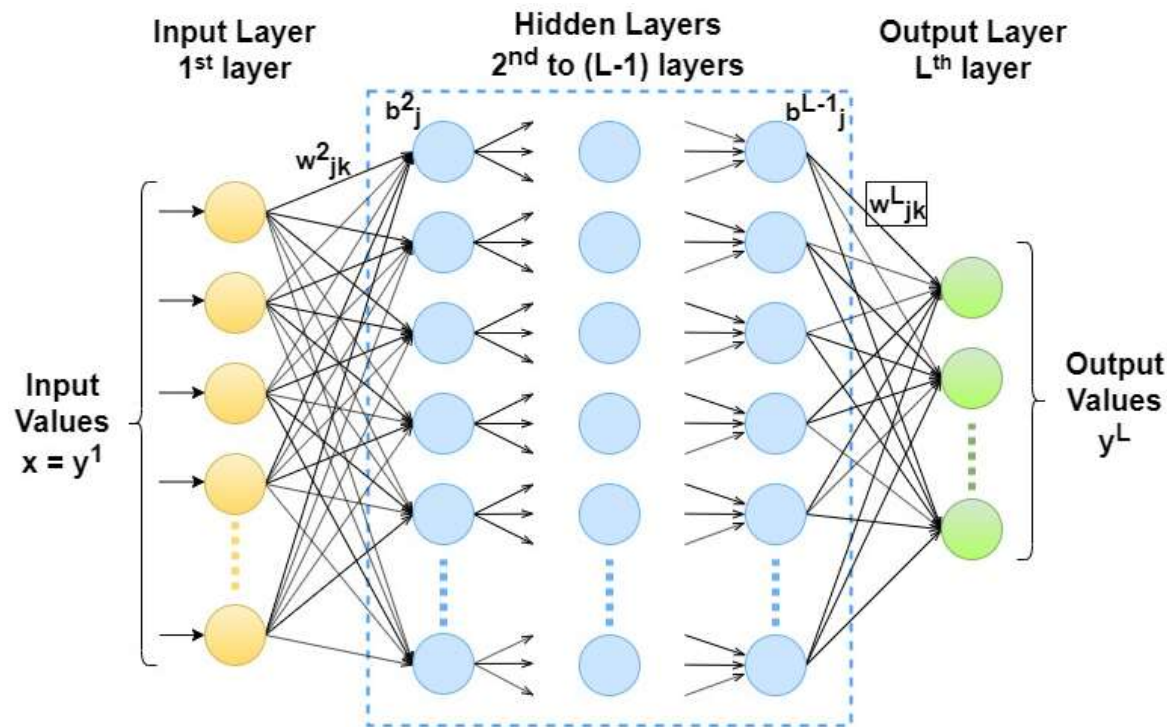


TRAINING

Find optimal w_{jk}^l and b_j^l

so that **Mean Square Error (MSE)** is minimised

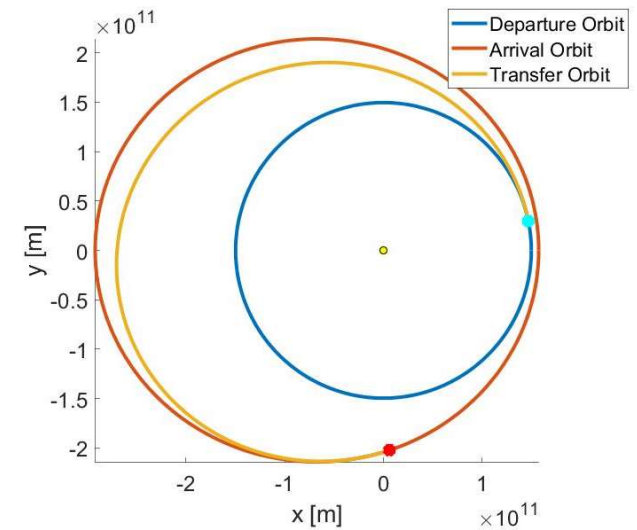
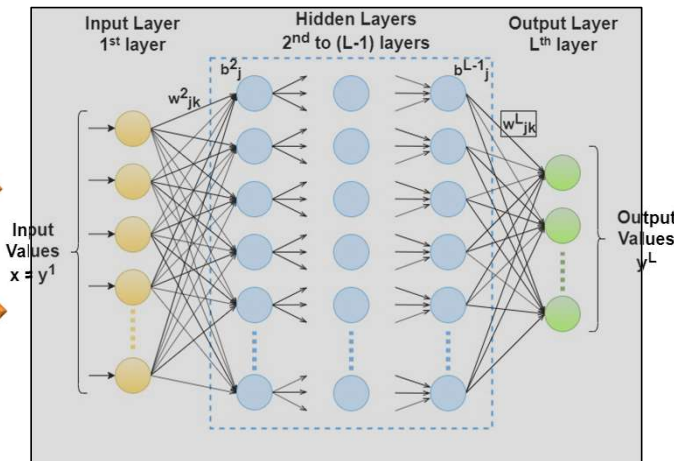
$$\text{MSE} = \frac{1}{N} \sum_{i=1}^N \|y_i - \mathbf{t}_i\|$$



Neural Network Design

Departure object orbital parameters

Arrival object orbital parameters



$$\mathbf{x} = [p_1, f_1, g_1, h_1, k_1, L_{1,0}, p_2, f_2, g_2, h_2, k_2, L_{2,0}]$$

Viavattene, G. and Ceriotti, M. (2020) Artificial Neural Network Design for Tours of Multiple Asteroids. In: 15th International Conference on Hybrid Artificial Intelligence Systems (HAIS 2020), Gijón, Spain, 11-13 Nov 2020, pp. 751-762. ISBN 9783030617042 (doi:10.1007/978-3-030-61705-9_63)

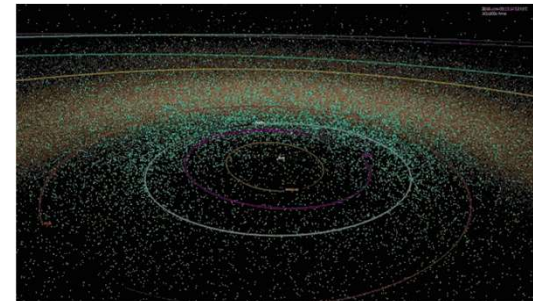
Additional challenges

- Orbit parameterisation

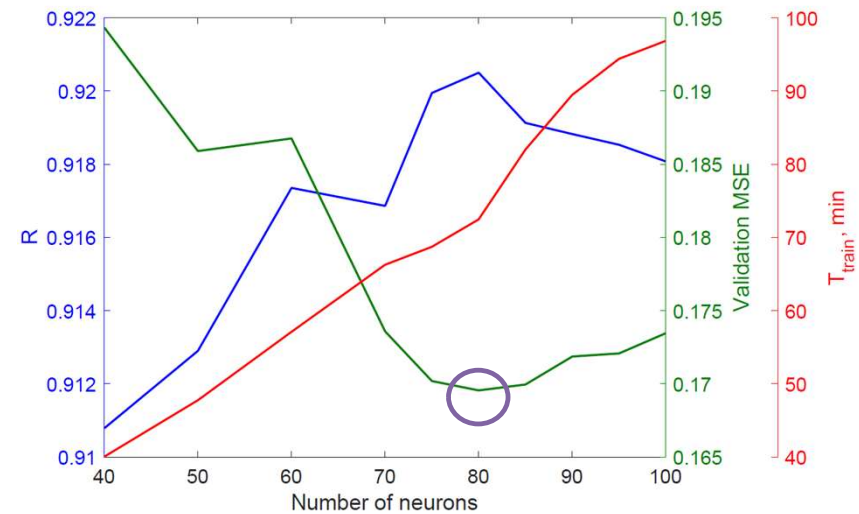
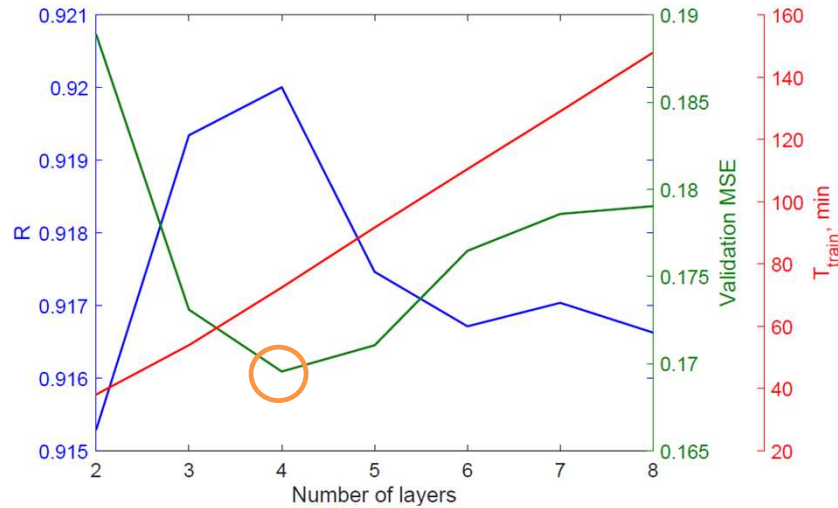
		Correlation	Validation-Set Error
Classical Orbital Elements	COE	0.855	0.530
Equinoctial Elements	EE	0.856	0.487
Modified Equinoctial Elements	MEE	0.925	0.236
Cartesian Coordinates	Cartesian	0.551	0.761
Delaunay Elements	Delaunay	0.694	0.862
Eccentricity and angular momentum vector	eH	0.908	0.221

- Dependence on time (phasing problem)

- Orbital transfer \neq Rendezvous



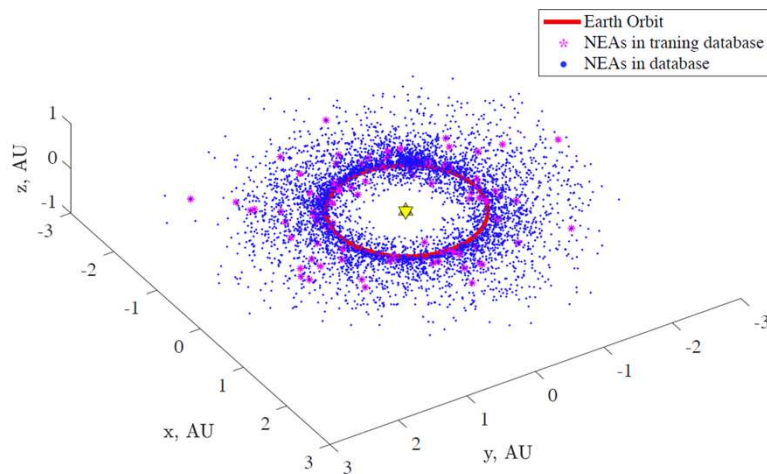
ANN architecture design



ANN parameter	Search space	Optimal value
No. hidden layers	[2, 8]	4
No. neurons	[40, 100]	80
Learning algorithm	Levenberg-Marquardt Resilient back-propagation Scaled conjugate gradient Gradient descent	Levenberg-Marquardt
Activation function	tansig, sigmoid, ReLu	Sigmoid

ANN Training

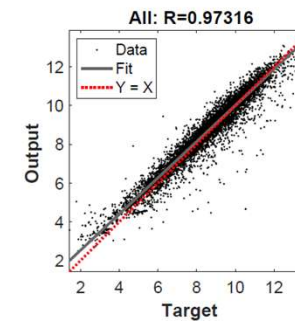
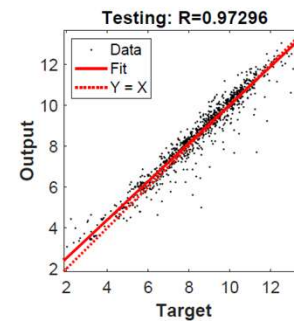
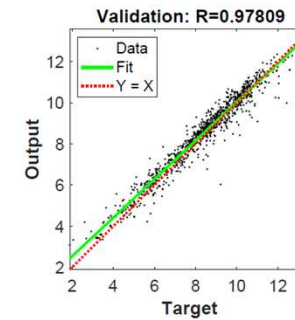
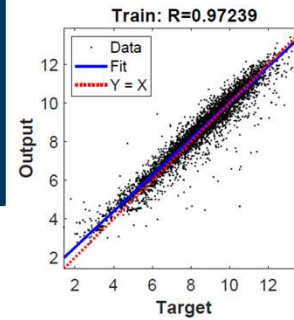
- 100 NEA in training database + Earth



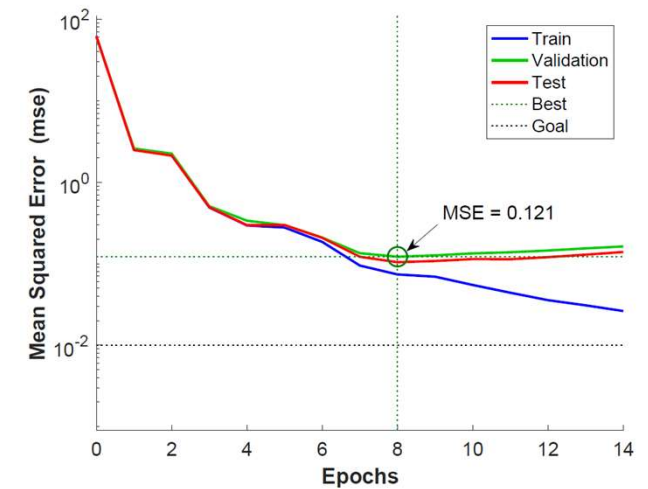
- 10,100 low-thrust transfers

- Training set (70%)
- Validation set (15%)
- Test set (15%)

- Exponential sinusoid shape with $a_{\max} = 0.1 \text{ mm/s}^2$



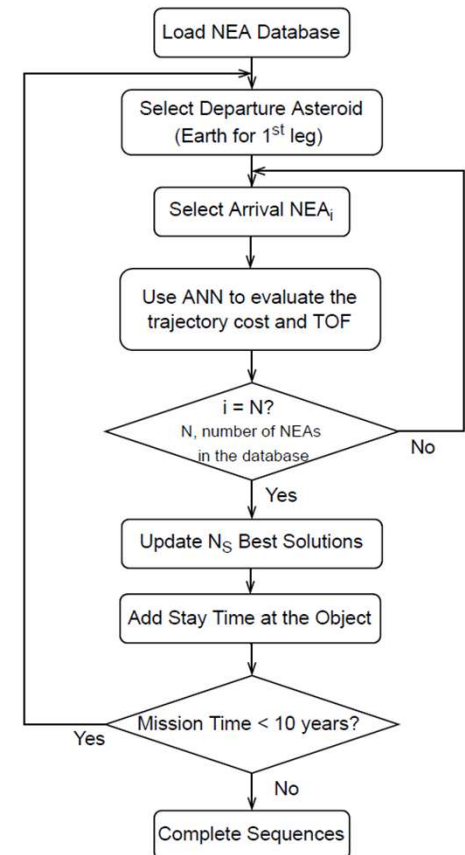
$R = 0.97$
 $MSE = 0.1211$



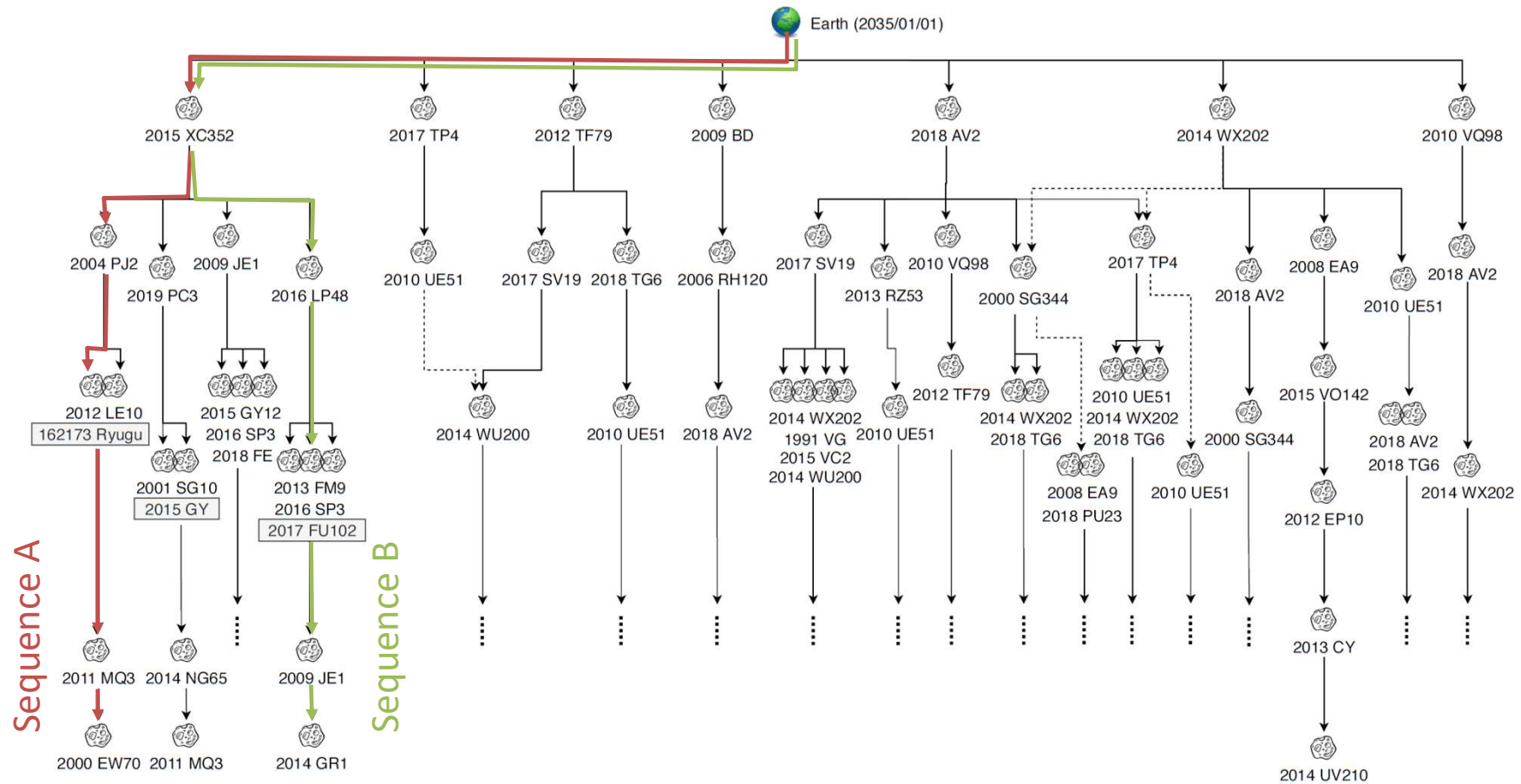
Sequence search with ANN

Multiple asteroid rendezvous

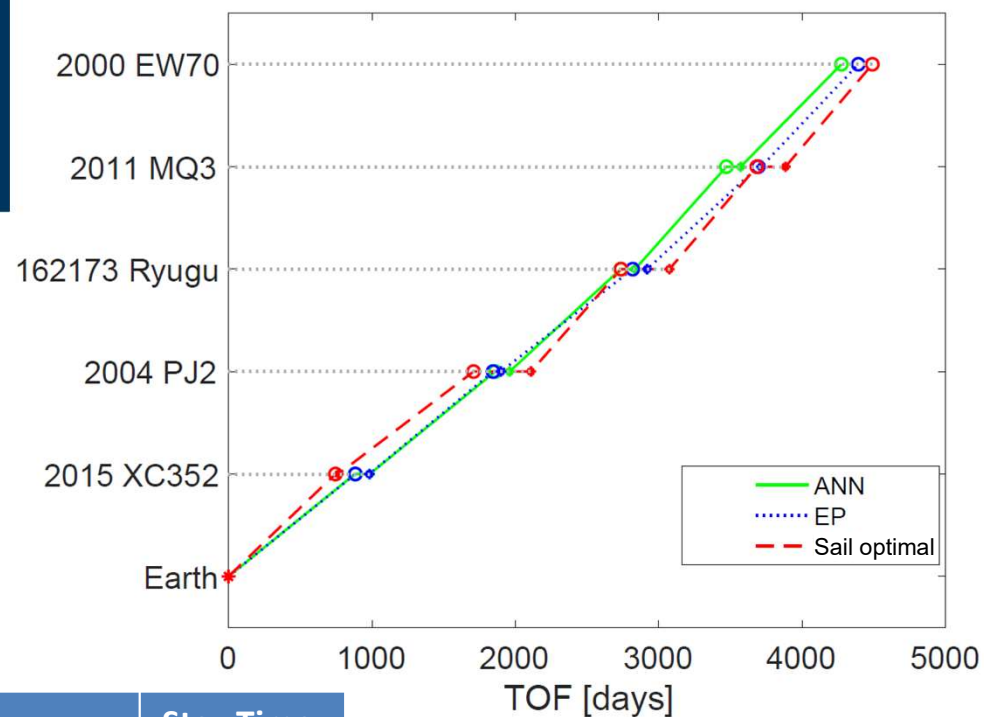
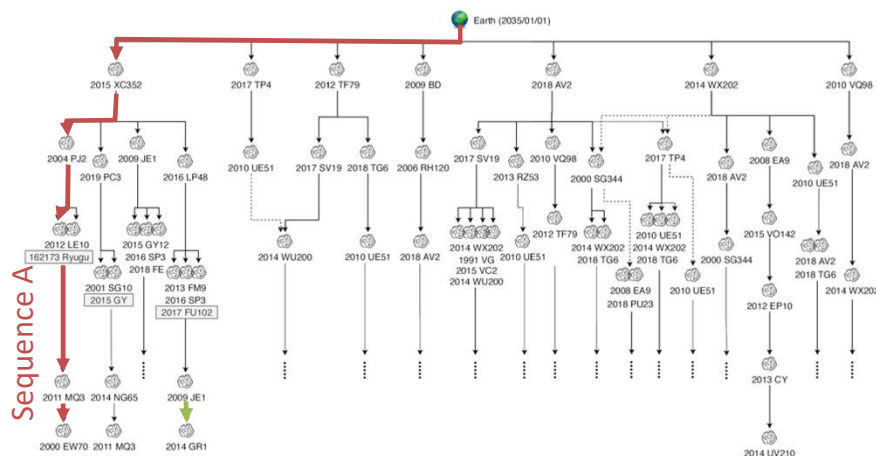
- 6,286 asteroids
 - ~300 PHAs, ~1,450 NHATS
 - Excluded highly inclined ($i \geq 20^\circ$) and eccentric ($e \geq 0.4$)
- Launch date fixed: 2035/01/01
 - Systematic scan of launch window could be done
- Stay time at asteroid: 100 days
- Only best 200 transfers with lowest ToF stored
 - Limits exponential growth
- 2 sequences selected for full optimisation
 - OPC solver GPOPS-II [Patterson, Rao]
 - Ideal solar sail with $a_c = 0.2 \text{ mm/s}^2$



Asteroid sequence tree graph

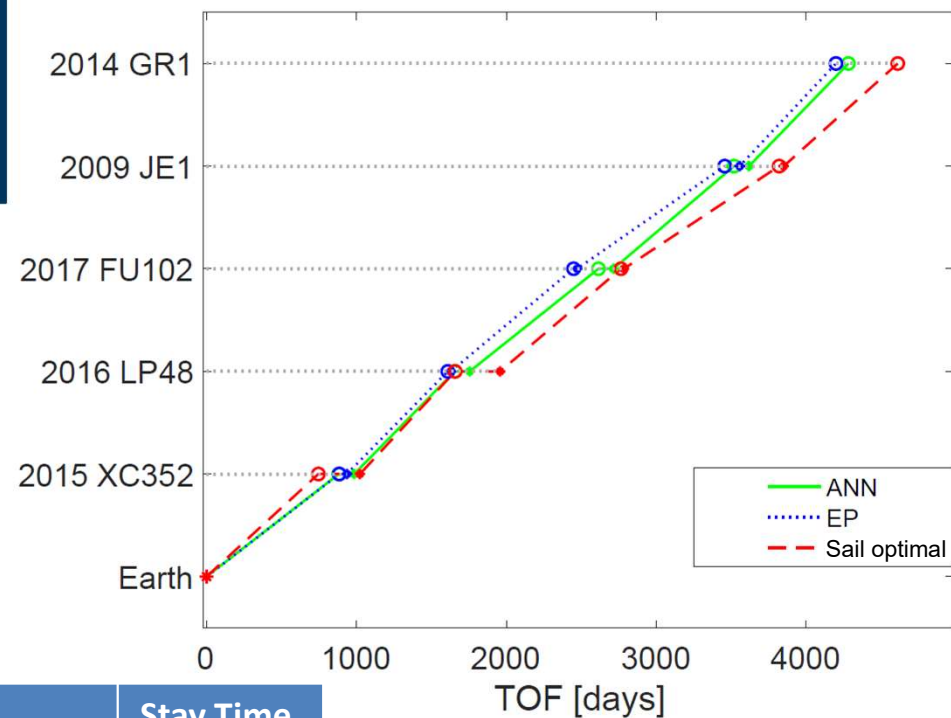
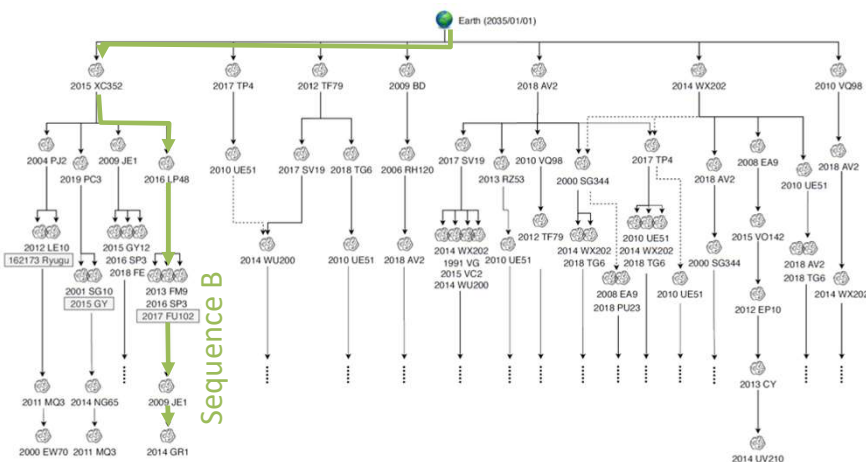


Sequence A



Leg	Departure	Arrival	TOF, days		Stay Time, days
			ANN estimate	Optimal	
Earth - 2015 XC352	2035-02-08	2037-04-04	882	746	20
2015 XC352 - 2004 PJ2	2037-04-24	2039-12-15	878	943	400
2004 PJ2 - 162173 Ryugu	2037-04-24	2042-10-08	775	628	338
162173 Ryugu - 2011MQ3	2043-09-20	2045-05-13	637	609	200
2011MQ3 - 2000 EW70	2045-11-29	2047-07-28	702	606	-

Sequence B



Leg	Departure	Arrival	TOF, days		Stay Time, days
			ANN estimate	Optimal	
Earth - 2015 XC352	2035-02-08	2037-04-04	882	746	277
2015 XC352 - 2016 LP48	2037-11-28	2039-08-26	673	635	300
2016 LP48 - 2017 FU102	2040-06-21	2042-09-05	859	806	20
2017 FU102 - 2009 JE1	2042-09-26	2045-07-27	984	1035	31
2009 JE1 - 2014 GR1	2045-08-27	2047-09-26	642	760	-

Performance analysis

Error in Time of Flight estimation

$$\mathcal{E}_{TOF} = \frac{1}{N} \sum_{i=1}^N \left(\frac{|TOF_{i,opt} - TOF_{i,ANN}|}{TOF_{i,opt}} \right) = 13.4 \%$$

ANN within sequence search algorithm

200 best sequences in less than 8 hours

~100x faster than using shape-based trajectories*

ANN **not** trained on fully-optimal solar sail trajectories

*compared to the method used in: A. Peloni, M. Ceriotti, and B. Dachwald. Solar-Sail Trajectory Design for a Multiple Near-Earth-Asteroid Rendezvous Mission. Journal of Guidance, Control, and Dynamics, 39(12):2712–2724, Sep 2016

- **Shape-based methods** can provide an approximate thrust profile
 - Still require optimisation
 - Trade-off between accuracy of solution and computational time
 - A better shape-based solution eases the follow-up optimisation, but takes longer

- **Artificial neural networks** can provide a very fast estimate of transfer “cost”
 - Require training
 - The position-time problem (phasing of bodies) is not fully solved

- **Future work**
 - Train ANN with full solar sail trajectory model

Space and Exploration Technology Group

Thank you!

Institution of
**MECHANICAL
ENGINEERS**

Thanks to the IMechE which
partially supported my
participation to ISSS 2023



University
of Glasgow

James Watt
School of
Engineering