

# Low Thrust Interplanetary Mission Trajectory Optimization using Differential Evolution

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# Interplanetary Transfers with Electrically Propelled Spacecraft

- Electric propulsion offers a good option for interplanetary transfers.
- Spacecraft in initial heliocentric orbit.
- To rendezvous with a final heliocentric orbit under heliocentric gravitational dynamics.
- Thrust vector magnitude and direction are the control variables that are to be determined to optimize:
  - Flight duration
  - Propellant consumption
- Only gradual velocity changes are possible due to low thrust.
- Coasting is to be allowed and should arise naturally out of the solution.
- Indirect approach to optimal control has been followed to solve this problem.

# Electric propulsion (EP) power-plant types considered

- Nuclear electric propulsion (NEP)
  - Constant power availability
- Solar electric propulsion (SEP)
  - Inverse square law model (first approximation to available power)
  - Williams and Coverstone-Carroll model (from experimental data)

# Equations of motion

$$\dot{x} = v_x \quad (1)$$

$$\dot{y} = v_y \quad (2)$$

$$\dot{v}_x = -\frac{\mu_s x}{r^3} + a_x \quad (3)$$

$$\dot{v}_y = -\frac{\mu_s y}{r^3} + a_y \quad (4)$$

$$\dot{m} = -\frac{m \sqrt{a_x^2 + a_y^2}}{g_0 I_{sp}} \quad (5)$$

The above equations govern both time and fuel optimal trajectories.

# Cost functions

$$J_{time} = \Phi_f + \int_{t_0}^{t_f} dt \quad \text{subject to} \quad m\sqrt{a_x^2 + a_y^2} = T_{max} \quad (6)$$

$$J_{fuel} = \Phi_f + \int_{t_0}^{t_f} \frac{m\sqrt{a_x^2 + a_y^2}}{g_0 I_{sp}} dt \quad \text{subject to} \quad m\sqrt{a_x^2 + a_y^2} \leq T_{max} \quad (7)$$

$\Phi_f$  represents the error in achieving the final desired orbit.

# Two Point Boundary Value Problem (TPBVP) Formulation

## TPBVP formulation

- Costates are introduced and the Hamiltonian is formed
- System dynamics, cost functionals and costates form the Hamiltonians.
- Pontryagin's minimum principle gives the costate dynamics and the optimal control law.
- Initial costates are unknown.
- Problem is reduced to the determination of initial costates such that the final state is achieved with maximum accuracy.

## Variables

- States -  $[x \ y \ v_x \ v_y \ m]$ , Costates -  $[\lambda_x \ \lambda_y \ \lambda_{v_x} \ \lambda_{v_y} \ \lambda_m]$
- Controls -  $[a_x \ a_y]$

# Optimal Control Law with Constraints

- Results in constrained minimization problem.
- Lagrange multipliers or the Karush-Kuhn-Tucker (KKT) conditions are utilized to obtain the control law.

$$I = \frac{\sqrt{\lambda_{v_x}^2 + \lambda_{v_y}^2}}{m} - \frac{1 - \lambda_m}{g_0 I_{sp}} \quad k = -\frac{T_{max}/m}{\sqrt{\lambda_{v_x}^2 + \lambda_{v_y}^2}} \quad (8)$$

$$\text{Fuel optimal - } \begin{cases} \text{If } I \geq 0, & a_x = k\lambda_{v_x} & a_y = k\lambda_{v_y} \\ \text{If } I < 0, & a_x = 0 & a_y = 0 \end{cases} \quad (9)$$

$$\text{Time optimal - } a_x = k\lambda_{v_x} \quad a_y = k\lambda_{v_y} \quad (10)$$

# Differential Evolution (DE)

- DE is used to determine the initial unknown costates.
- DE is an evolutionary algorithm.
- Search based global optimization method.
- Utilizes three operation: crossover, mutation and selection.

## DE parameters that influence the convergence

- Crossover ratio (CR)
  - Mutation factor (F)
  - Population size (NP)
- 
- The DE algorithm requires the selection of 3 distinct members from the population to generate a trial vector.
  - The Durstenfeld version of the Fischer-Yates shuffle is utilized.



# DE robustness and performance

**Table:** Test problem for optimization.

25 dimensional Rastrigin function			1 thread	2 threads	4 threads	
Lower bounds	Upper bounds	Generations	Time(ms)	Time(ms)	Time(ms)	Max Speedup
-1	1	1000	857.626	546.38	379.27	2.26
-10	10	1750	1460.03	934.662	621.439	2.35
-100	100	1950	1613.14	1021.72	682.483	2.36
-1000	1000	2125	1764.25	1088.09	751.529	2.35
-10000	10000	2350	1954.39	1202.36	816.597	2.39
	Solution	(0,0, . . . ,0)				

- Final cost  $< 10^{-7}$ , NP=250, CR=0.1, F=0.8, run on a 2 core machine with random seeds.
- The above table shows the robustness of DE for the standard test problem taken. (25 dimensional Rastrigin function)
- The efficiency of multi-threading has also been simultaneously demonstrated.
- This provides confidence to apply DE to solve the TPBVP formed by indirect optimal control.

# Model validation

1AU to 1.5AU time optimal transfer, 6000s  $I_{sp}$  at  $1\text{mm/s}^2$  initial acceleration level.

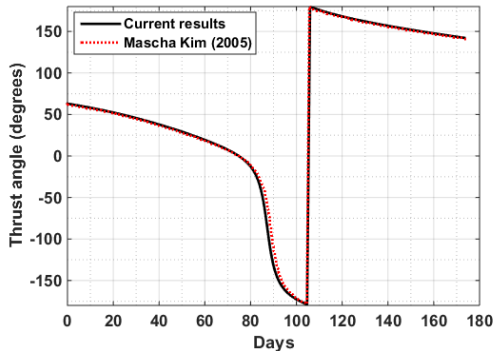
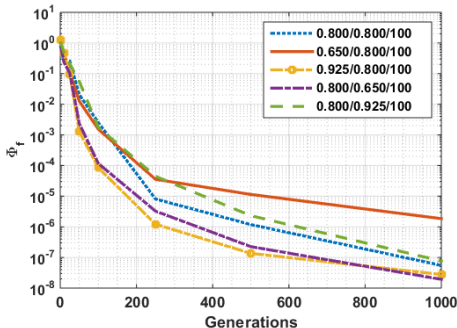


Figure: Comparison of obtained results with literature.

# DE performance for Earth-Mars fuel optimal transfers

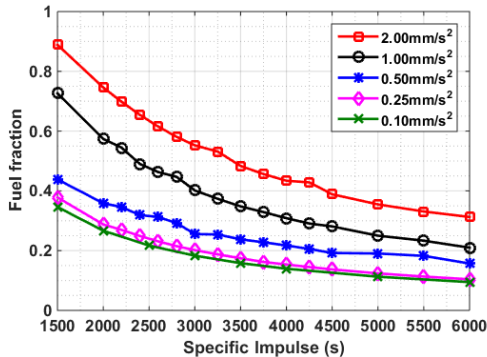
- Different CR/F/NP combinations are values. (200 days, 2000s  $I_{sp}$ ,  $1\text{mm/s}^2$ )



It is observed that CR/F ratios greater than 1 are suitable for rapid convergence. CR=0.9 and F=0.8 has been chosen for further analysis.

# Time optimal transfers

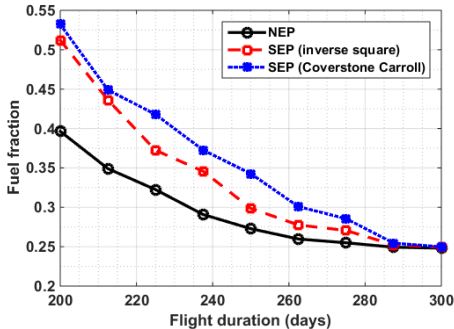
- Earth-Mars transfer with varying initial acceleration levels.



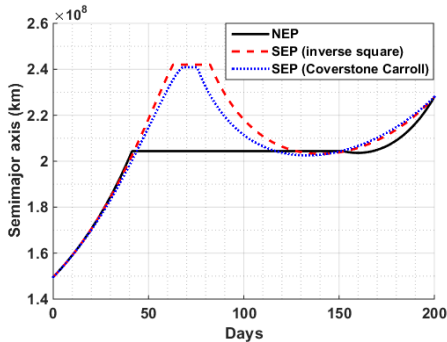
Very low acceleration levels result in spiral transfers. These transfers take similar fuel fractions.

# Earth-Mars fuel optimal transfers

- NEP and SEP power models have been used.
- For small flight durations, NEP consumes much lower fuel than SEP.
- For large flight durations, all power models converge to the same fuel fraction.



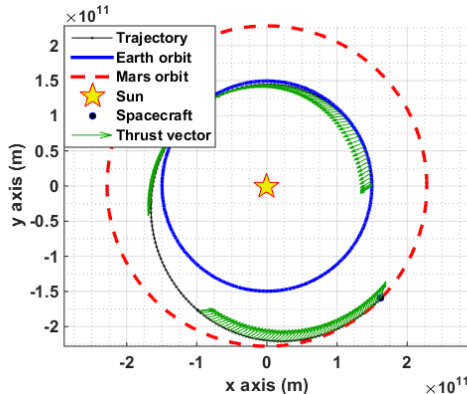
# Earth-Mars fuel optimal transfers - Power models



SEP models for a 200 day transfer show that solar power models tend to waste energy to satisfy the flight time constraint. This is due to reduced thrust at larger radii from the Sun.

# Sample trajectory - Earth to Mars - Heliocentric

- 400 day fuel optimal transfer, 2000s  $I_{sp}$ , initial mass 1000kg, thrust level 236mN. (NEXT class thruster)



# Thank You