



# CBDO 2024

XXII Brazilian Colloquium on Orbital Dynamics

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*Mini-course:*

## Optimal Control of Space Trajectories using GEKKO

*Lecture 2*

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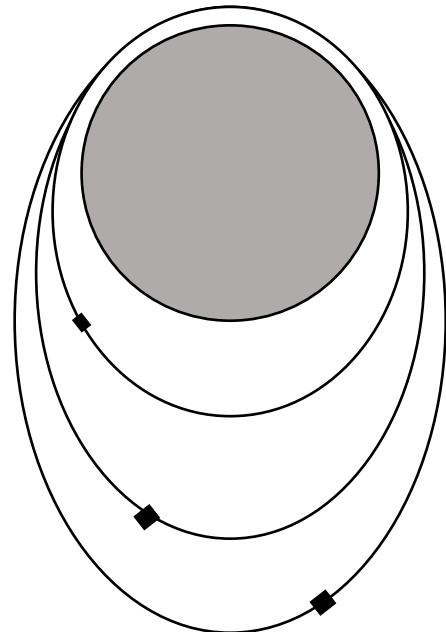
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CBDO - 2024

05/12/2024

# Summary

- I. Example – rocket launch.
- II. Description of the script.
- III. Low thrust maneuver.



# I. Example to verification – rocket launch

- Source:

<https://apmonitor.com/wiki/index.php/Apps/RocketLaunch>

# I. Example to verification – rocket launch

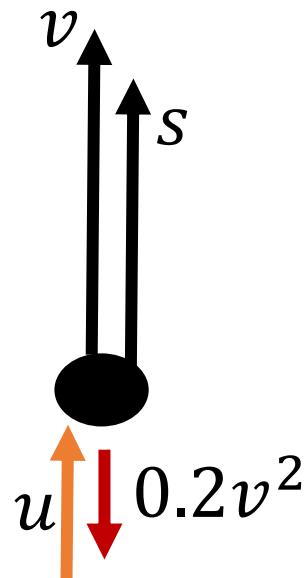
$$\min(t_f)$$

Subject to:

$$\dot{s} = v$$

$$\dot{v} = \frac{u - 0.2v^2}{m}$$

$$\dot{m} = -0.01u^2$$



$$0.0 \leq v(t) \leq 1.7$$

$$-1.1 \leq u(t) \leq 1.1$$

$$0.2 \leq m(t) \leq 1.0$$

$$s(t_0) = 0.0$$

$$v(t_0) = 0.0$$

$$m(t_0) = 1.0$$

$$s(t_f) = 10.0$$

$$v(t_f) = 0.0$$

## II. Structure of the script

```
import numpy as np
from gekko import GEKKO

# Call GEKKO
m = GEKKO()

# 1 unit of time in 101 points
m.time = np.linspace(0,1,101)

# optimal final time problem, initial value 1, lower 0.1, upper 100
tf = m.FV(value=1.0,lb=0.1,ub=100)
tf.STATUS = 1 #controlled
```

```
# optimal final time problem, initial value 1, lower 0.1, upper 100
tf = m.FV(value=1.0,lb=0.1,ub=100)
tf.STATUS = 1 #controlled

# control variable
u = m.MV(value=0,lb=-1.1,ub=1.1)
u.STATUS = 1 #controlled by the program
u.DCOST = 1e-5 #Delta penalty control mov.

# variables
s = m.Var(value=0) # position
v = m.Var(value=0,lb=0,ub=1.7) # velocity
mass = m.Var(value=1,lb=0.2) # mass
```

```
# differential equations
m.Equation(s.dt()==tf*v)
m.Equation(mass*v.dt()==tf*(u-0.2*v**2))
m.Equation(mass.dt()==tf*(-0.01*u**2))

# final conditions
m.fix(s, pos=len(m.time)-1, val=10.0)
m.fix(v, pos=len(m.time)-1, val=0.0)

# cost function
m.Obj(tf)
```

```
| # setup options for solver
| m.options.NODES = 6 # Collocations points for poly
| m.options.SOLVER = 3 #IPOPT
| m.options.IMODE = 6 #OCP DYN OPT
| m.options.MAX_ITER = 500
| m.options.MV_TYPE = 0 #INTERPOLATION BETW END POINTS

# Solving
m.solve()
```

```
apm 193.19.205.166_gk_model0 <br><pre> -----
APMonitor, Version 1.0.3
APMonitor Optimization Suite
-----
----- APM Model Size -----
Each time step contains
    Objects      :          0
    Constants    :          0
    Variables    :          5
    Intermediates:          0
    Connections   :          4
    Equations     :          4
    Residuals    :          4
```

Number of state variables:	3897
Number of total equations: -	3800
Number of slack variables: -	0
-----	
Degrees of freedom :	97

\*\*\*\*\*

### Dynamic Control with Interior Point Solver

\*\*\*\*\*

...

Objective : 3747.13119010218

Successful solution

-----

```
    print('Minimun final time: ' + str(tf.value[0]))
```

```
✓ 0.0s
```

```
Minimun final time: 7.4942617694
```

```
import matplotlib.pyplot as plt

# scaled time
ts = m.time * tf.value[0]

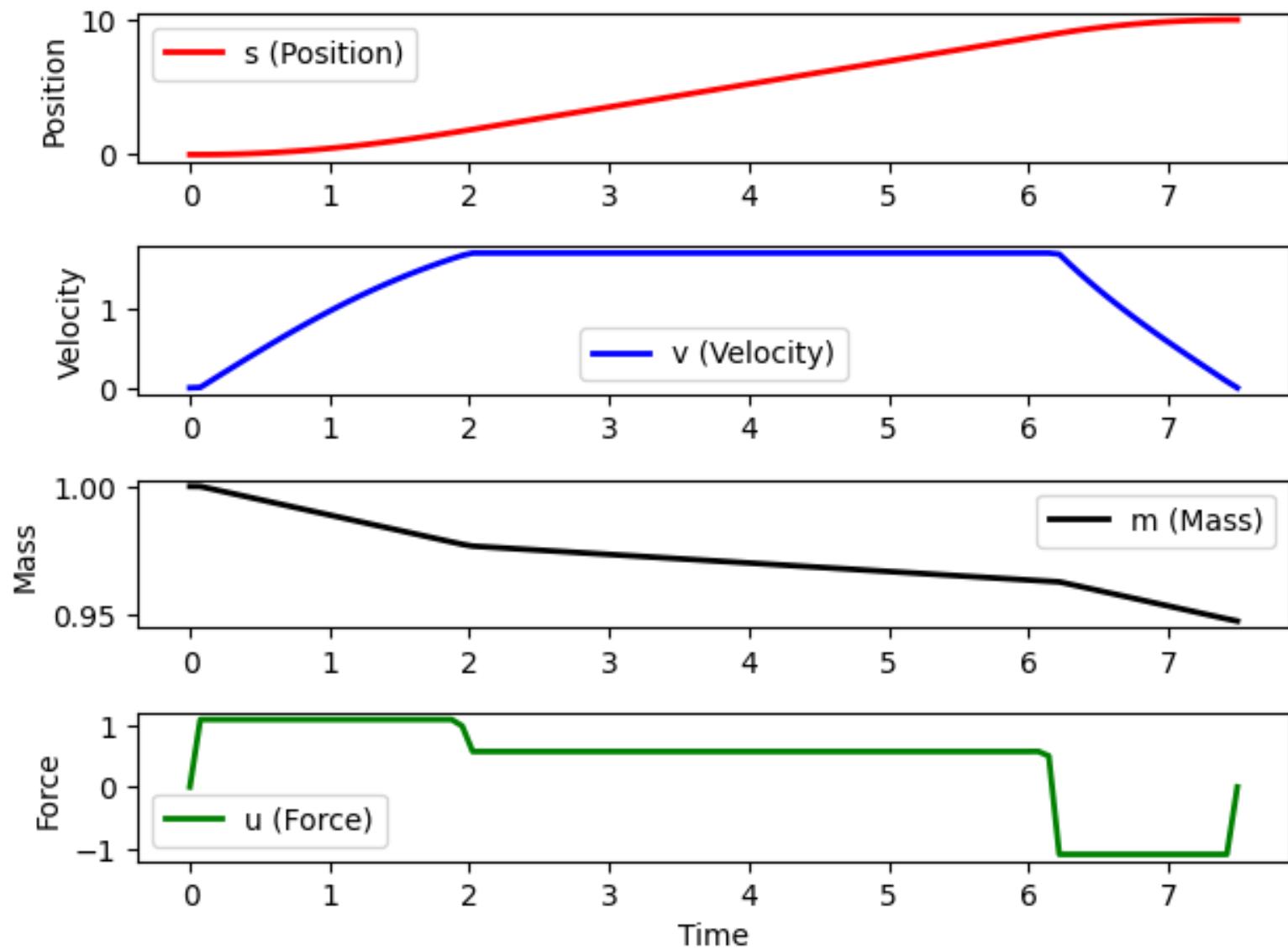
# plot results
plt.figure(1)
plt.subplot(4,1,1)
plt.plot(ts,s.value,'r-',linewidth=2)
plt.ylabel('Position')
plt.legend(['s (Position)'])
```

```
plt.subplot(4,1,2)
plt.plot(ts,v.value,'b-',linewidth=2)
plt.ylabel('Velocity')
plt.legend(['v (Velocity)'])

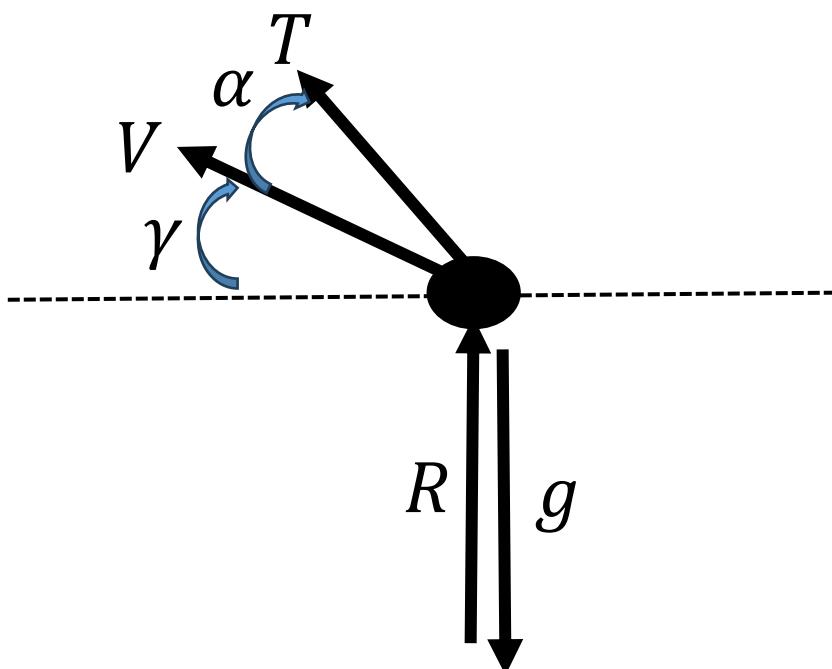
plt.subplot(4,1,3)
plt.plot(ts,mass.value,'k-',linewidth=2)
plt.ylabel('Mass')
plt.legend(['m (Mass)'])

plt.subplot(4,1,4)
plt.plot(ts,u.value,'g-',linewidth=2)
plt.ylabel('Force')
plt.legend(['u (Force)'])

plt.xlabel('Time')
plt.tight_layout()
plt.show()
```



### III. Low thrust orbital transfer



$$\dot{R} = V \sin \gamma \quad (1)$$

$$\dot{\theta} = \frac{V \cos \gamma \cos A}{R \cos \varphi} \quad (2)$$

$$\dot{\phi} = \frac{V \cos \gamma \sin A}{R} \quad (3)$$

$$\dot{V} = \frac{T \cos \alpha}{m} - g \sin \gamma \quad (4)$$

$$\dot{\gamma} = \frac{T \sin \alpha}{mV} - \frac{g \cos \gamma}{V} + \frac{V \cos \gamma}{R} \quad (5)$$

$$\dot{A} = - \frac{V \tan \varphi \cos A \cos \gamma}{R} \quad (6)$$

$$\dot{m} = - \frac{T}{V_{sp}} \quad (7)$$

$$U\{T(t), \alpha(t)\} \quad (8)$$

# The Optimal Control Problem - Constraints

$$t_0 < t \leq t_f$$

$$-20 \deg \leq \gamma(t) \leq 20 \deg$$

$$500 \text{ s} \leq t_f \leq 2000 \text{ s}$$

$$0 \deg \leq A(t) \leq 359.9 \deg$$

$$400 \text{ km} \leq h(t) \leq 402 \text{ km}$$

$$m_{min} \leq m(t) \leq m_{max}$$

$$V_{cir} \leq V(t_f)$$

$$-180 \deg \leq \alpha(t) \leq 180 \deg$$

$$-180 \deg \leq \theta(t) \leq 180 \deg$$

$$0 \text{ N} \leq T(t) \leq 0.5 \text{ N}$$

$$-90 \deg \leq \varphi(t) \leq 90 \deg$$

- Initial conditions:
  - Altitude: 400 km
  - Velocity: 7802.88 m/s
  - FPA: 0°
- Final conditions:
  - Altitude: 402 km

Objective: Final time minimization.

$$J = \min_{t_f}(t)$$

# Script

```
from gekko import GEKKO
import numpy as np
import matplotlib.pyplot as plt
import math

# GEKKO Initialization -----
m = GEKKO()

# Time parameters -----
nt = 151
tm = np.linspace(0,1,nt)
m.time = tm

p = np.zeros(nt)
p[-1] = 1.0
final = m.Param(value=p)
```

```
# Parameters and Const -----
pi   = math.pi
pi2  = pi/2.0
deg2rad = pi/180.0

# Planet info
mu   = 3.986031954093051e14          # earth gravitational param (m^3/s^2)
Re   = 6378135.                         # mean radius of the earth (m)
g0   = 9.8                               # mean gravity at the SML (m/s^2)

# Atm info
rho0 = 1.217    # msl atmospheric density (kg/m^3)
H    = 8500     # atm scale height (m)

# Vehicle info
Surf = 332.13      # Surface area m**2
mass = 180.0 #21800.0      # spacecraft mass (kg)
A2m  = Surf/mass
propmass = 20.0
```

```
# Propulsion info
isp = 230                                # Isp (s)
Ve = isp*g0                               # Rocket motor especific velocity
mdot = 0.0

# Initial boundary conditions at (t0) -----
h0    = 400000                            # initial altitude (m)
R0    = Re+h0
LONG0 = 0
LAT0  = 0
V0    = 7668.585794677108                 # initial velocity in m/s
FPA0  = 0.0*deg2rad
AZI0  = pi2
msp0  = mass
```

```
# Final boundary conditions at (tf) -----
hf    = 402000                            # final desired altitude (m)
Rf    = Re+hf
Vf    = 7667.454673925565                # final velocity in m/s
FPAf  = 0.0*deg2rad                      # final FPA
mspf  = mass-propmass
```

```

# Variables constraints -----
# State vector
Ru = Rf
Rl = Re+75000

LONGl = 0
LONGu = pi

LATl = -pi2
LATu = pi2

Vl = Vf
Vu = V0

FPAl = -20*deg2rad
FP Au = -FPAl

AZIl = 0
AZIu = pi

msp u = msp0
msp l = mspf

#Manipulated variables -----
Thrul = 0
Thruu = 0.5 #N

```

```

#Time guess -----
Timel = 500.0 #(s) 50
Timeu = 2000.0 #(s) #800

# Initialization and path constraints -----
r = m.Var (value=R0, lb=Rl, ub=Ru)          # Radio or altitude (m)
long = m.Var (value=LONG0, lb=LONGl, ub=LONGu) # Longitude angle (rad)
lat = m.Var (value=LAT0, lb=LATl, ub=LATu)    # Latitude (rad)
v = m.Var (value=V0, lb=6000)                  # Velocity (m/s)
fpa = m.Var (value=FPA0, lb=FPAl, ub=FP Au)   # Flight Path Angle
azi = m.Var (value=AZI0, lb=AZIl, ub=AZIu)      # Azimuth
msp = m.Var (value=msp0, lb=msp l, ub=msp u)

```

```

# Final time
Tf = m.FV(lb=Timel,ub=Timeu); Tf.STATUS = 1

# Manipulated variables -----
Thru = m.MV (lb=Thrul, ub=Thruu)
Thru.STATUS = 1

AT = m.MV (value=0.0, lb=-pi, ub=pi)
AT.STATUS = 1

# Additional variables *****
# Use m.Intermediate() to save and NOT USE for control and/or MV
A2m = m.Intermediate(Surf/msp)
# Gravity -----
g = m.Intermediate(mu/r**2) # local gravity (m/s^2)
alt = m.Intermediate(r-Re) # Local altitude (m)

# Atm -----
rho = m.Intermediate(rho0*m.exp(-alt/H)) # local density (kg/m^3)
pdyn = m.Intermediate(0.5*rho*v**2) # dynamic pressure/mass

# Aero -----
cl = 0.0
cd = 0.0

L2m = m.Intermediate(cl*pdyn*A2m) # lift acceleration (m/s^2)
D2m = m.Intermediate(cd*pdyn*A2m) # drag acceleration (m/s^2)

# T to mass -----
T2m = m.Intermediate(Thru/msp) # Thrust to mass ratio (m/s^2)

gf = m.Intermediate(L2m/g) #n Load or g forces
Qhc = m.Intermediate(1.83e-8*m.sqrt(rho/0.5)*v**3)
a_qr = 1.072e6*(v**-1.88)*(rho**-0.325)

Qhr = m.Intermediate(4.736e4*(0.5**a_qr)*(rho**1.22)*359)
c_acc = m.Intermediate(v**2/r) #centrifugal acc
v_esc = m.Intermediate(m.sqrt(2*g*r))
BA = 0.0

```

```

# Process model / EDOs/ DAEs ****
m.Equation(r.dt()/Tf == v*m.sin(fpa))
m.Equation(r*m.cos(lat)*long.dt()/Tf == v*m.cos(fpa)*m.sin(azi))
m.Equation(r*lat.dt()/Tf == v*m.cos(fpa)*m.cos(azi))
m.Equation(v.dt()/Tf == T2m*m.cos(AT)-D2m-g*m.sin(fpa))
m.Equation(v*fpa.dt()/Tf == T2m*m.sin(AT)+L2m*m.cos(BA)-(g-v**2/r)*m.cos(fpa))
m.Equation(v*r*m.cos(fpa)*azi.dt()/Tf == r*L2m*m.sin(BA) + (v*m.cos(fpa))**2*m.sin(azi)*m.tan(lat))
m.Equation(Ve*msp.dt()/Tf == -Thru)

# Objetive funtion -----
m.Minimize(final)

# Setup solution -----
m.options.IMODE    = 6      # 4 for SIM and 6 for Optimal Control
m.options.MAX_ITER = 2000   # Control of MAX ITER
m.options.NODES    = 2      # WITH 2 IS OK Collocation nodes
##m.options.OTOL    = 1e-3  # Iter Tolerance obje, same as GPOPS
##m.options.RTOL    = 1e-3  # Real tol eq
##m.options.SOLVER  = 3      # solver (IPOPT)

m.solve(disp=True) #True

# get additional solution information -----
import json
with open(m.path+'//results.json') as f:
    results = json.load(f)

```

```
apm 185.153.176.175_gk_model0 <br><pre> -----
APMonitor, Version 1.0.3
APMonitor Optimization Suite
-----
----- APM Model Size -----
Each time step contains
  Objects      :      0
  Constants    :      0
  Variables    :     11
  Intermediates:     13
  Connections   :      0
  Equations     :     21
  Residuals    :      8

Number of state variables:      3601
Number of total equations: -      3300
```

Number of Iterations....: 53

	(scaled)	(unscaled)
Objective.....:	1.0000592927938146e+00	1.0000592927938146e+00
Dual infeasibility.....:	2.8295922815122290e-11	2.8295922815122290e-11
Constraint violation....:	8.9391858100737346e-10	1.2118201642152826e-07
Complementarity.....:	4.7367067492974097e-08	4.7367067492974097e-08
Overall NLP error.....:	4.7367067492974097e-08	1.2118201642152826e-07

```
Number of inequality constraint evaluations      = 54
Number of equality constraint Jacobian evaluations = 54
Number of inequality constraint Jacobian evaluations = 54
Number of Lagrangian Hessian evaluations        = 53
Total CPU secs in IPOPT (w/o function evaluations) =      0.723
Total CPU secs in NLP function evaluations       =      2.362

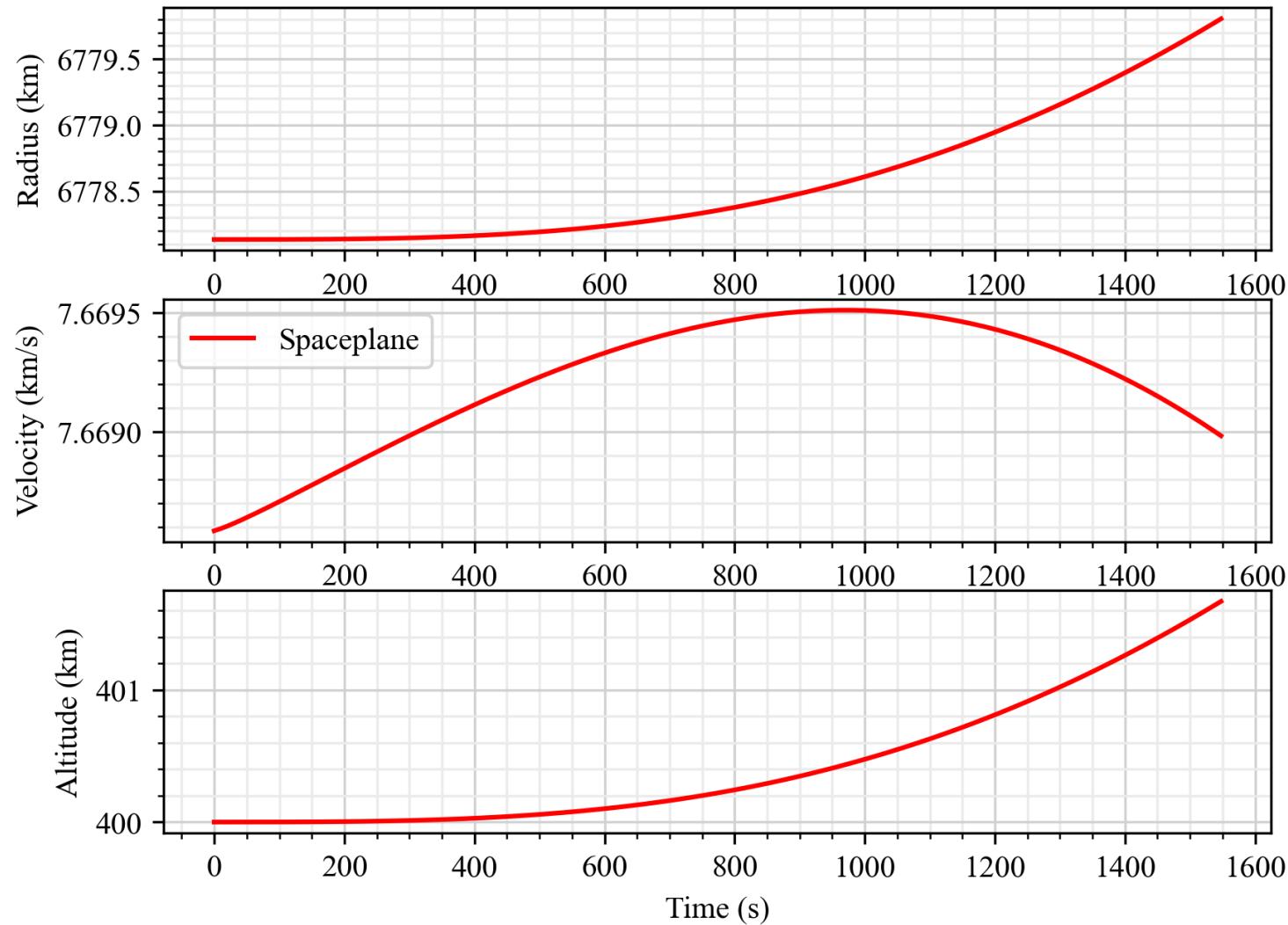
EXIT: Optimal Solution Found.

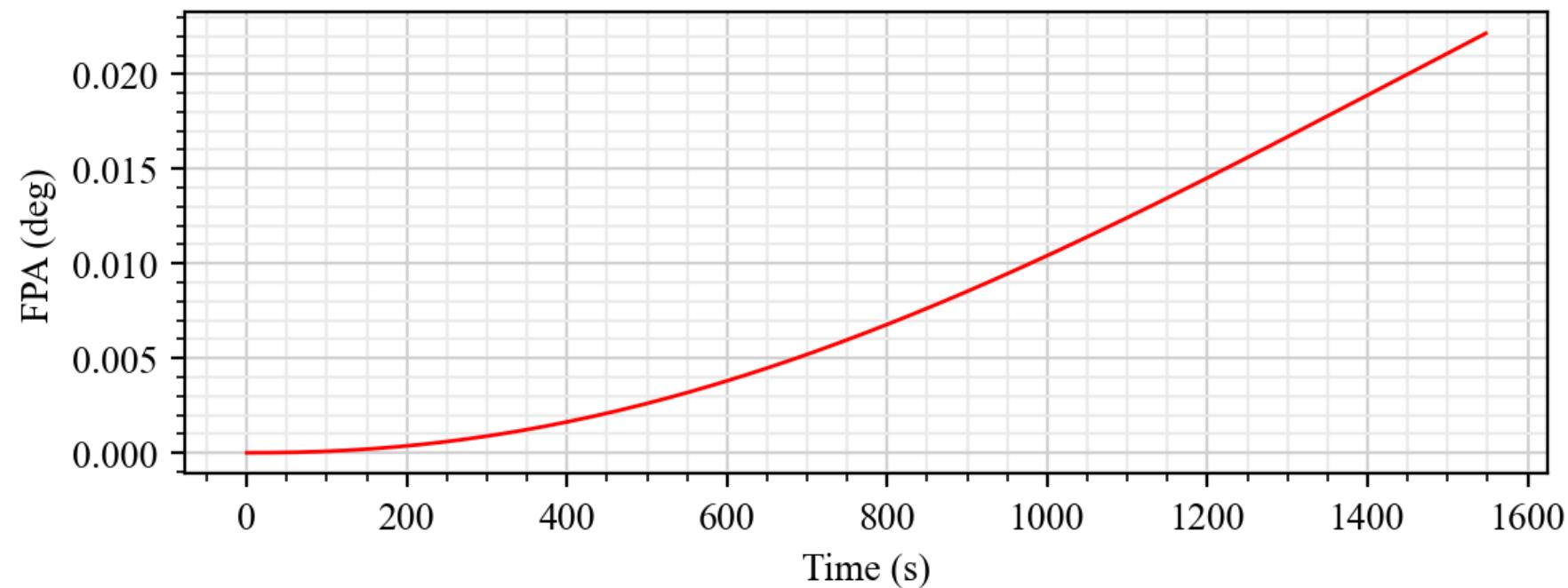
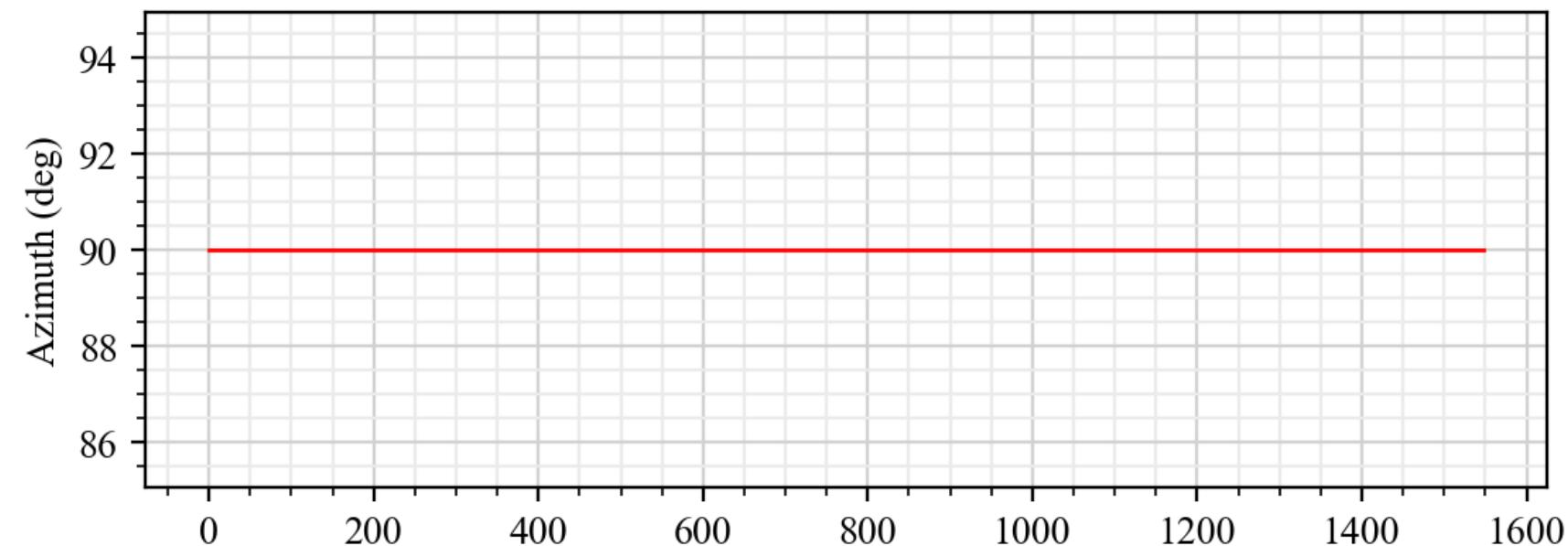
The solution was found.

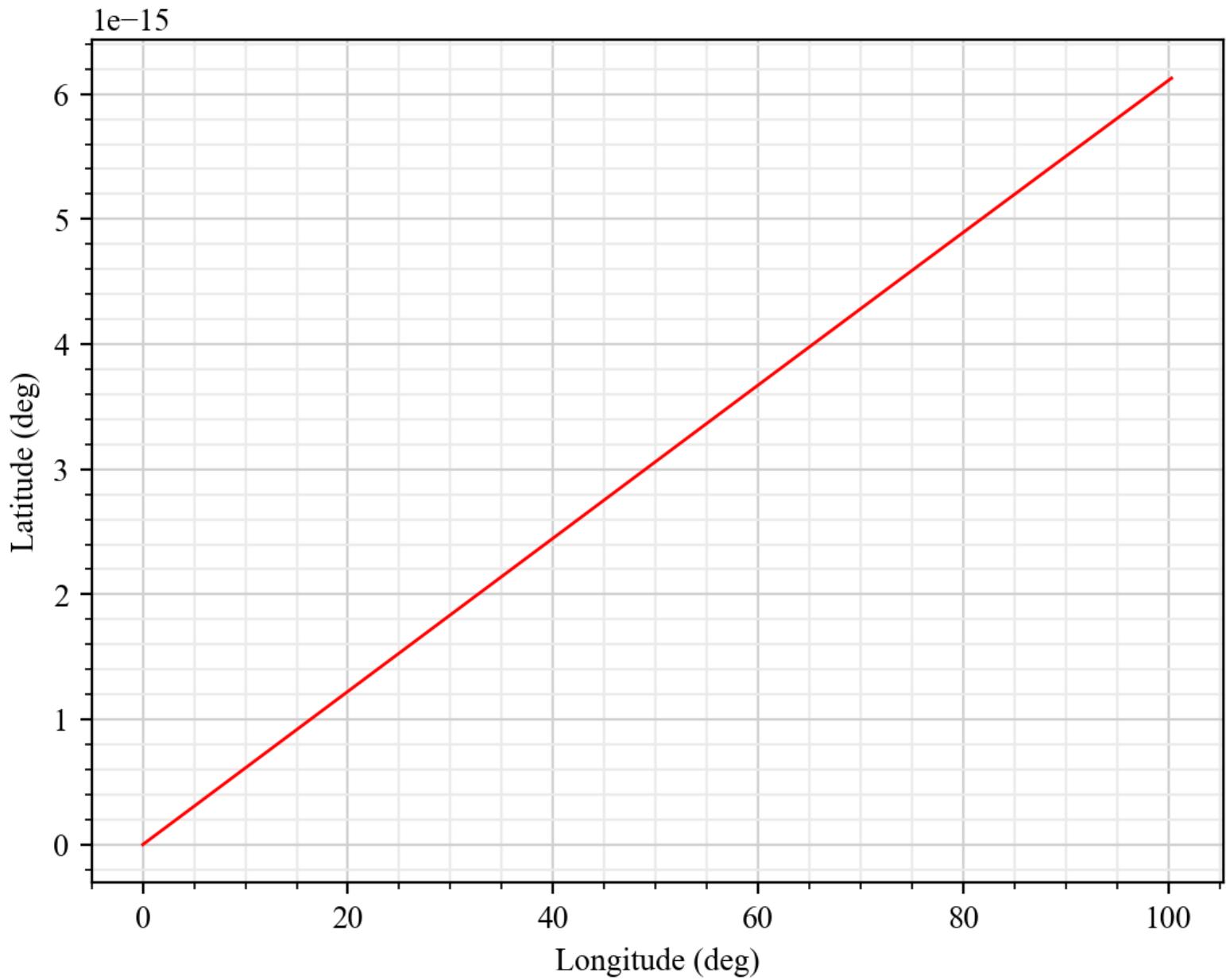
The final value of the objective function is    1.00005929279381

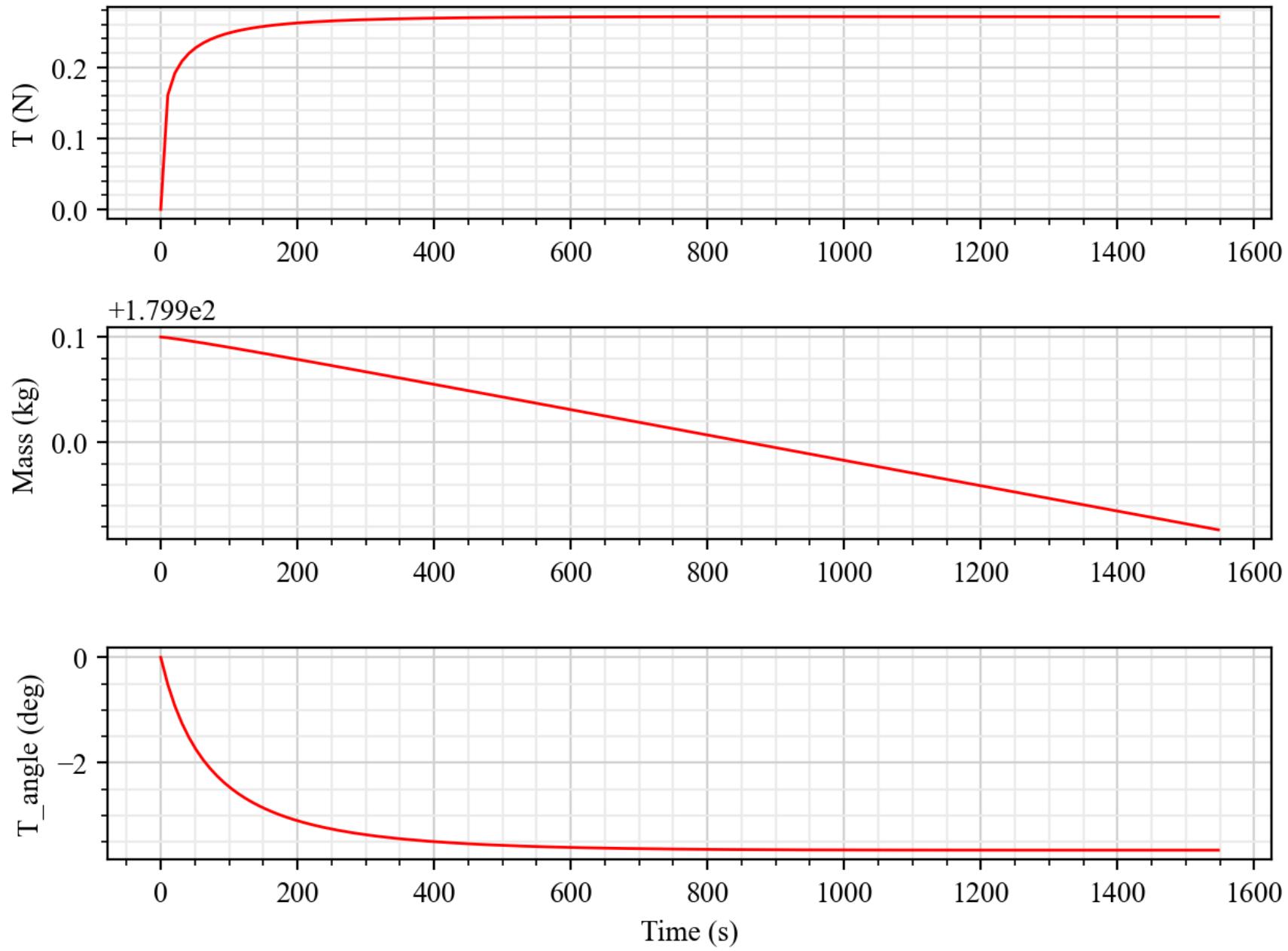
-----
Solver      : IPOPT (v3.12)
Solution time : 3.19769999998971      sec
Objective   : 1.00005929279381
Successful solution
```

# Results









# Fixing final conditions

- Final conditions:

- Altitude: 401 km
- Velocity: 7667.45 m/s
- FPA: 0°

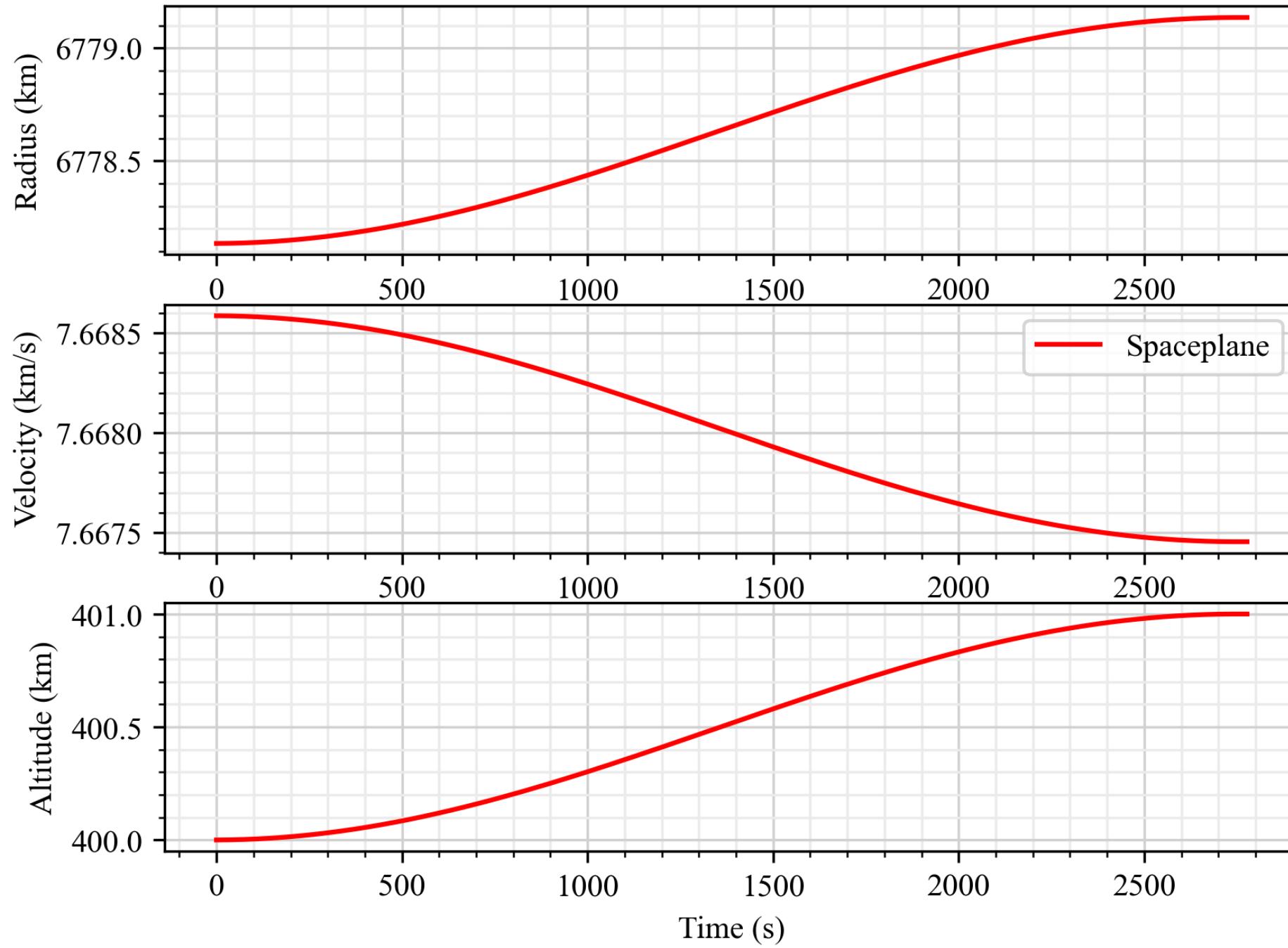
Objective: Final time minimization.

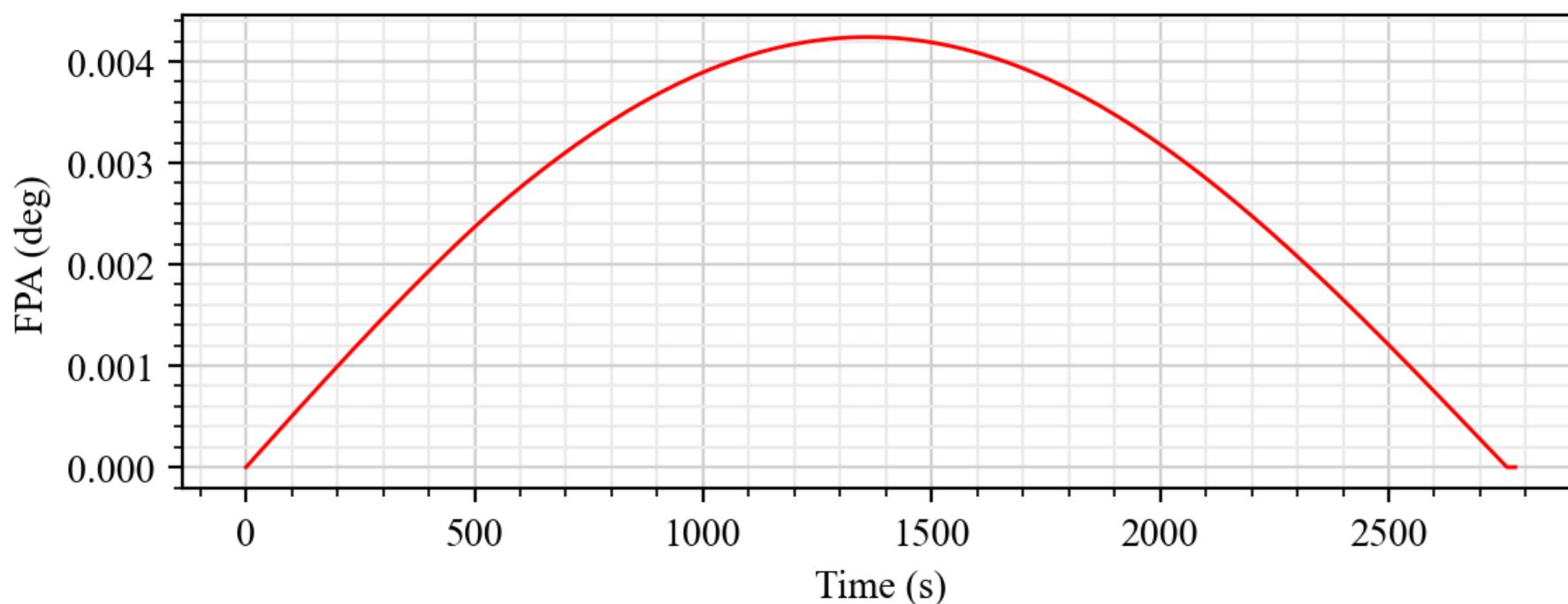
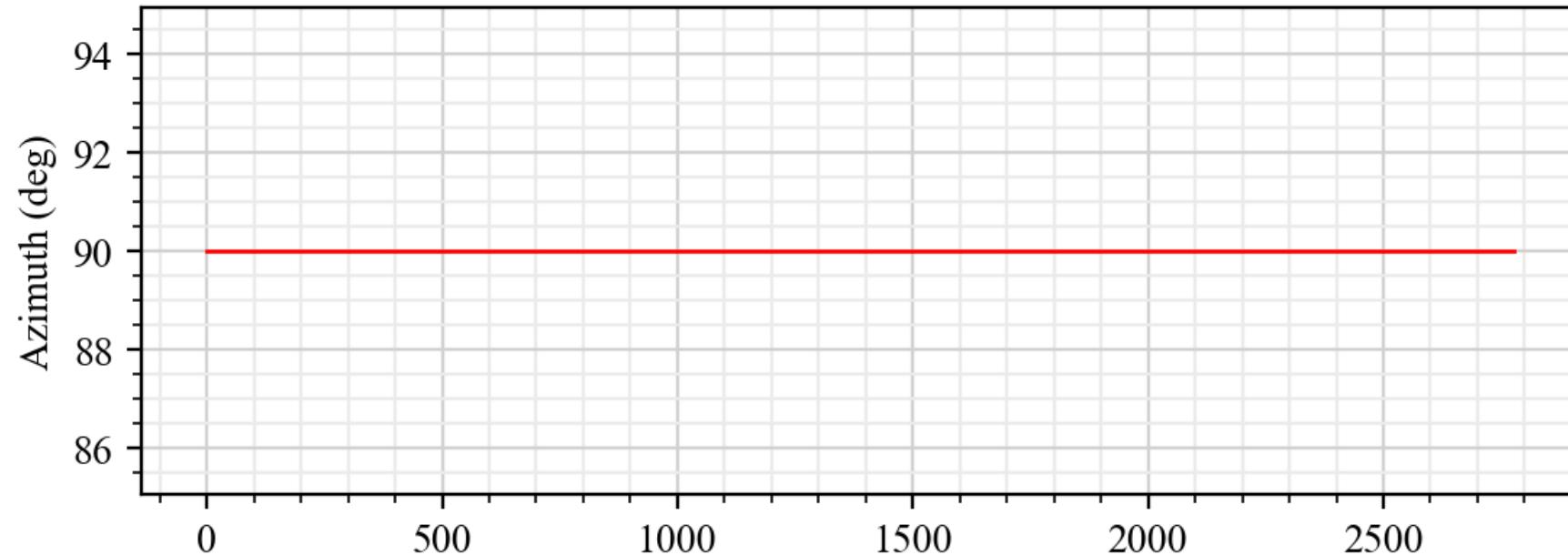
$$J = \min_{t_f}(t)$$

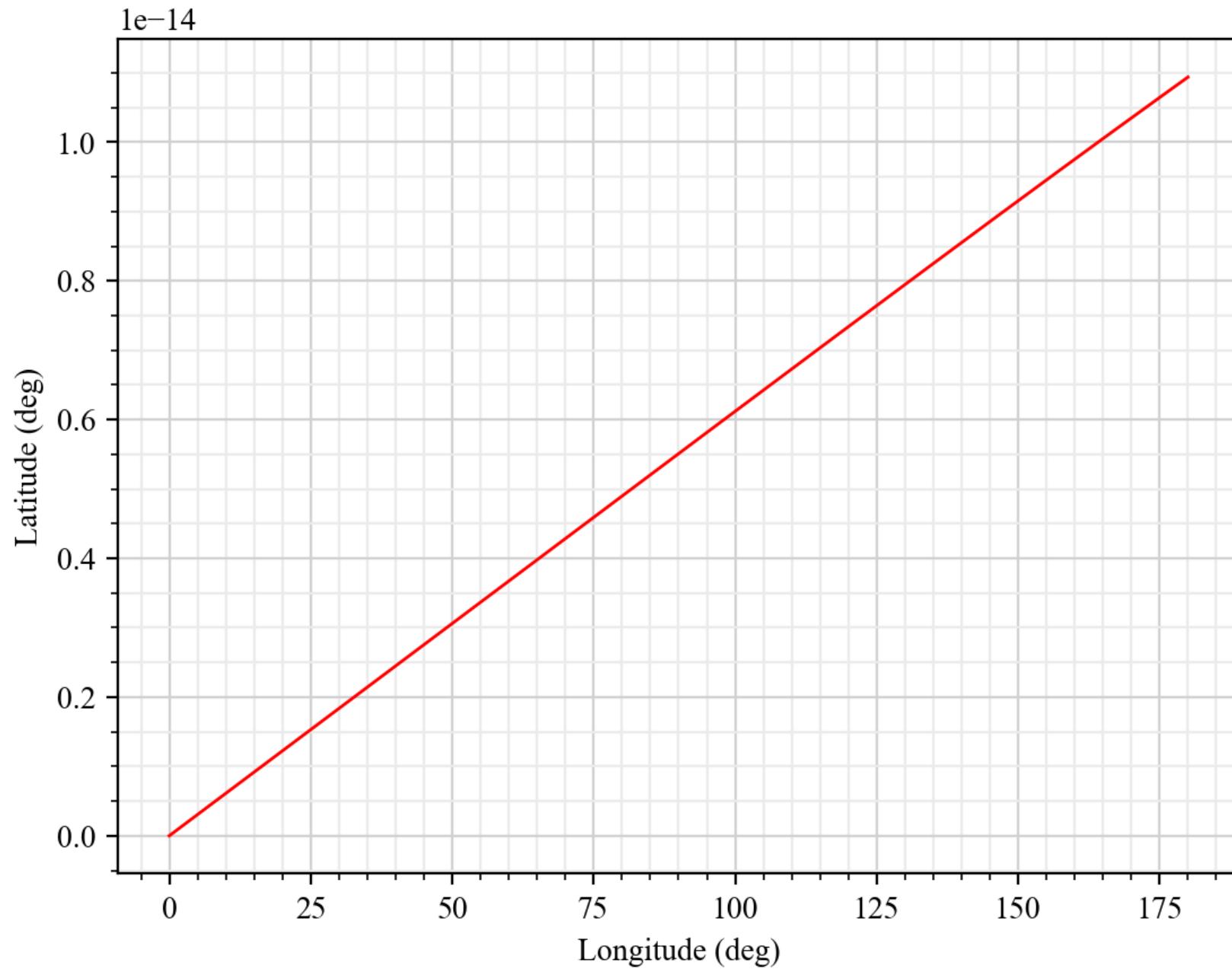
$$t_0 < t \leq t_f$$

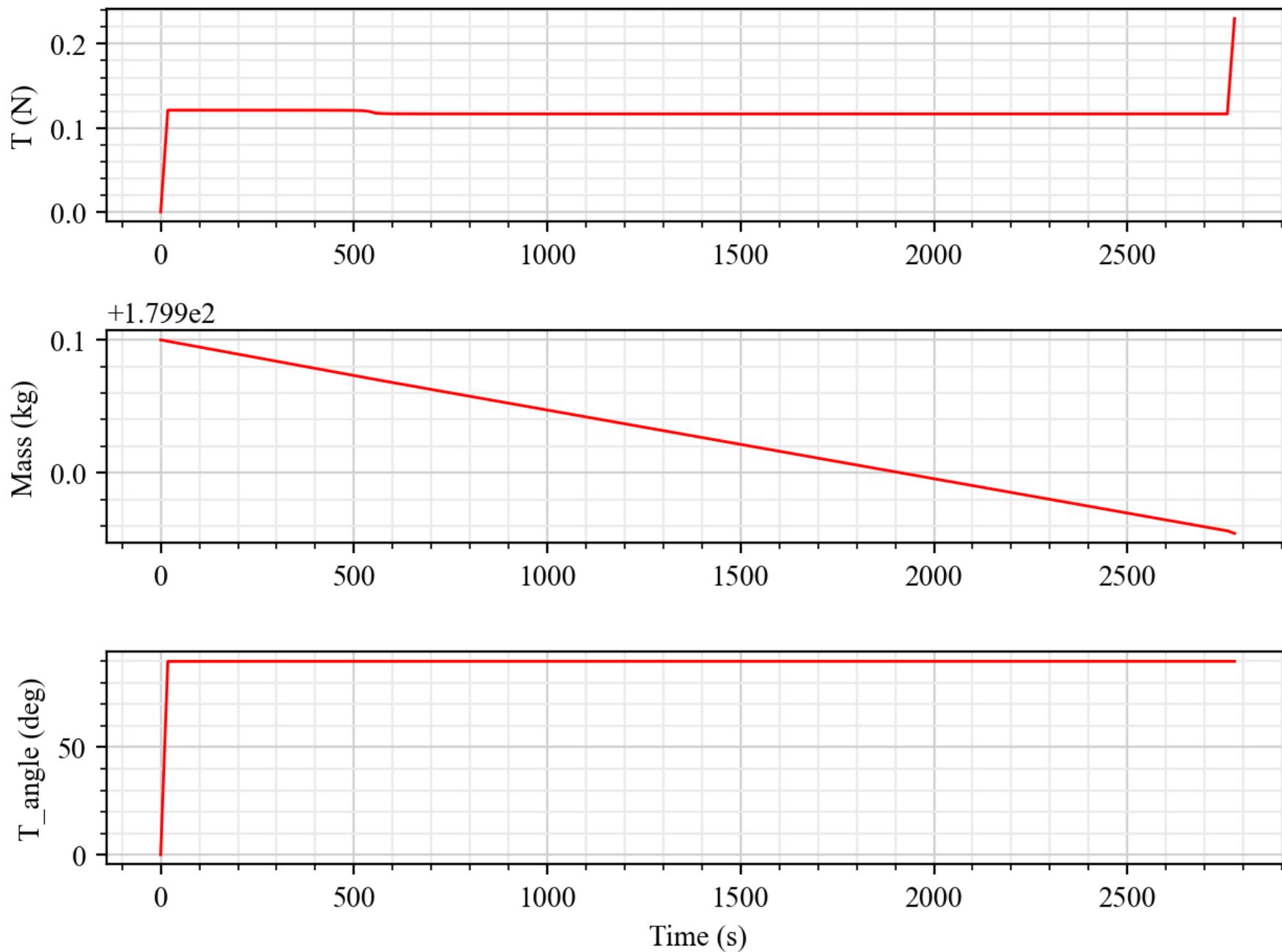
$$2000 \text{ s} \leq t_f \leq 4000 \text{ s}$$

```
# Fixed variables at tf -----
m.fix_final(r, Rf)
m.fix_final(v, Vf)
m.fix_final(fpa, FPAf)
```









# References

- [1] Armellin, R., Lavagna, M., and Ercoli-Finzi, A.,(2006) Aero-gravity assist maneuvers: controlled dynamics modeling and optimization, *Celestial Mechanics and Dynamical Astronomy*, Vol. 95,, pp. 391–405. doi: 10.1007/s10569-006-9024-y
- [2] Betts J. (2010) Practical Methods for Optimal Control Using Nonlinear Programming. 3 Ed..
- [3] Patterson, M., Rao, A. (2016) GPOPS II Manual. V 2.3. Available at: <https://gpop2.com/resources/gpop2UsersGuide.pdf>
- [4] Murcia-Piñeros, J., Prado, A. F., Dimino, I., & de Moraes, R. V. (2024). Optimal gliding trajectories for descent on Mars. *Applied Sciences*, 14(17), 7786. Doi: 10.3390/app14177786
- [5] Piñeros, J. O. M., Bevilacqua, R., Prado, A. F., & de Moraes, R. V. (2024). Optimizing aerogravity-assisted maneuvers at high atmospheric altitude above Venus, Earth, and Mars to control heliocentric orbits. *Acta Astronautica*, 215, 333-347. Doi: 10.1016/j.actaastro.2023.12.017
- [6] Murcia-Piñeros, J., Bevilacqua, R., Gaglio, E., Prado, A. B., & De Moraes, R. V. (2024). Optimization of Aero-gravity assisted maneuvers for spaceplanes at high atmospheric flight on Earth. In *AIAA SCITECH 2024 Forum* (p. 1459). Doi: 10.2514/6.2024-1459.
- [7] Beal, L.D.R., Hill, D., Martin, R.A., and Hedengren, J. D., “GEKKO Optimization Suite”, *Processes*, Vol. 6, No. 8, 2018. doi: 10.3390/pr6080106.
- [8] Hedengren, J. D., Asgharzadeh Shishavan, R., Powell, K.M., and Edgar, T.F., “Nonlinear Modeling, Estimation and Predictive Control in APMonitor”, *Computers and Chemical Engineering*, Vol. 70, 2014, pp. 133–148. doi: 10.1016/j.compchemeng.2014.04.013.

# Thank you!

## Questions?