

### Mini-course:

# Optimal Control of Space Trajectories using GEKKO

Lecture 3: Reentry gliding

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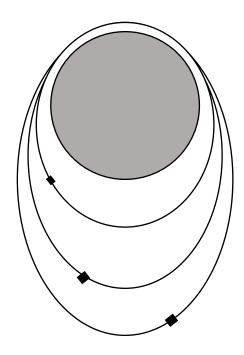
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CBDO - 2024

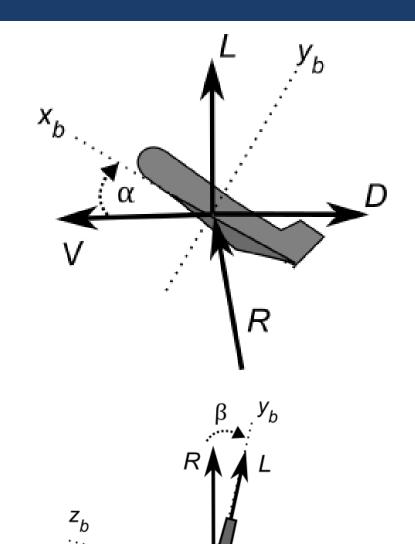
06/12/2024

# Summary

- I. Case of study Reentry gliding.
- II. Final remarks.



# I. Space shuttle gliding reentry [2, 3].



$$\dot{R} = V \sin \gamma \tag{1}$$

$$\dot{\theta} = \frac{V \cos \gamma \cos A}{R \cos \varphi} \tag{2}$$

$$\dot{\varphi} = \frac{V \cos \gamma \sin A}{R} \tag{3}$$

$$\dot{V} = \frac{-D(\alpha)}{m} - g \sin \gamma \tag{4}$$

$$\dot{\gamma} = \frac{L_{(\alpha)} \cos \beta}{mV} - \frac{g \cos \gamma}{V} + \frac{V \cos \gamma}{R} \tag{5}$$

$$\dot{A} = \frac{L_{(\alpha)} \sin \beta}{mV \cos \gamma} - \frac{V \tan \varphi \cos A \cos \gamma}{R} \tag{6}$$

$$U\{\alpha(t),\beta(t)\}\tag{7}$$

## The Optimal Control Problem - Constraints

$$t_0 < t \le t_f$$

 $1000 s \le t_f \le 3000 s$ 

$$24384 \ km \le h(t) \le 79248km$$

$$V(t) \le 7802.88$$

$$0 \ deg \le \theta(t) \le 180 \ deg$$

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

$$-89 \ deg \le \gamma(t) \le 89 \ deg$$

$$0 \deg \le A(t) \le 180.0 \deg$$

$$7.4 \ deg \le \alpha(t) \le 18.0 \ deg$$

$$-90 deg \le \beta(t) \le 1.0 deg$$

#### **Additional assumptions**

- Non-rotational spherical Earth.
- Static isothermal exponential atmosphere.
- Space shuttle aerodynamic model.

# Verification and validation, V&V The reentry problem.

#### • Initial conditions:

- Altitude = 79248 m
- Velocity = 7802.88 m/s
- FPA = -1 deg
- $AZI = 90 \deg$
- LON & LAT = 0.0 deg

#### • Final conditions:

- Altitude: 24384 m
- Velocity: 762 m/s
- FPA: -5 deg

Objective: Maximize cross-range or:

$$\boldsymbol{J} = \max_{t_f}(\varphi)$$

```
from gekko import GEKKO
import numpy as np
import matplotlib.pyplot as plt
import math
# GEKKO Initialization -
m = GEKKO()
# Time parameters --
nt = 301
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 #m.Const if only they are applied in the Variables, without previous calculations

mu = 3.986031954093051e14  # earth gravitational param (m^3/s^2)

Re = 6371203.92  # mean radius of the earth (m)
g0 = 9.8  # mean gravity at the SML (m/s^2)

# Atm info
rho0 = 1.225570827014494  # msl atmospheric density (kg/m^3)
H = 7254.24  # atm scale height (m)
```

```
# Vehicle info
Surf = 249.9091776
                         # Surface area m**2
                          # spacecraft mass (kg)
mass = 92079.2525560557
A2m = Surf/mass
#propmass = 0
# Aerodynamic info
b0 = 0.07854
                                   # Cd base, cd0
   = -0.3529
                                   # Cd 1 (1/rad)
                                   # Cd polar (1/rad^2)
   = 2.0400
   = -0.20704
                                   # Cl base
a1 = 1.6756
                                   # cl rate (1/rad)
```

```
# Initial boundary conditions at (t0)
   = 79248
                                    # initial altitude (m)
hØ
     = Re+h0
RØ
LONGO = 0
LATO = 0
     = 7802.88
                                    # inital velocity in m/s
VØ
FPA0 = -1.0*deg2rad
AZIO = pi2
msp0 = mass
AOAO = 0
BA0 = 0
```

```
# Final boundary conditions at (tf)
hf = 24384  # final desired altitude (m)
Rf = Re+hf
Vf = 762  # final velocity in m/s
FPAf = -5.0*deg2rad  # final FPA
mspf = mass-propmass
```

```
# Variables constraints --
# State vector
Ru = R0
R1 = Rf
LONG1 = 0
LONGu = pi
LAT1 = -pi2
LATu = pi2
V1 = Vf
Vu = V0
FPAl = -89.0*deg2rad
FPAu = -FPA1
AZI1 = 0
AZIu = pi
```

```
#Manipulated variables ------
aoau = 18*deg2rad
aoal = 7.4*deg2rad

BAl = -pi2
BAu = 1*deg2rad #Betts p. 248

#Time guess ------
Timel = 1000.0 #(s)
Timeu = 3000.0 #(s)
```

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

# Manipulated variables
BA = m.MV (lb=BAl, ub=BAu)
BA.STATUS = 1

AOA = m.MV (lb=aoal, ub=aoau)
AOA.STATUS = 1
```

```
= m.Intermediate(mu/r**2) # local gravity (m/s^2)
alt = m.Intermediate(r-Re) # Local altitude (m)
# Atm -----
rho = m.Intermediate(rho@*m.exp(-alt/H)) # local density (kg/m^3)
pdyn = m.Intermediate(0.5*rho*v**2)
                               # dynamic pressure/mass
# Aero -
cl
   = m.Intermediate(a0+a1*A0A)
    = m.Intermediate(b0+b1*A0A+b2*A0A**2)
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 = 0.0
                               # Thrust to mass ratio (m/s^2)
AT = 0.0
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))
```

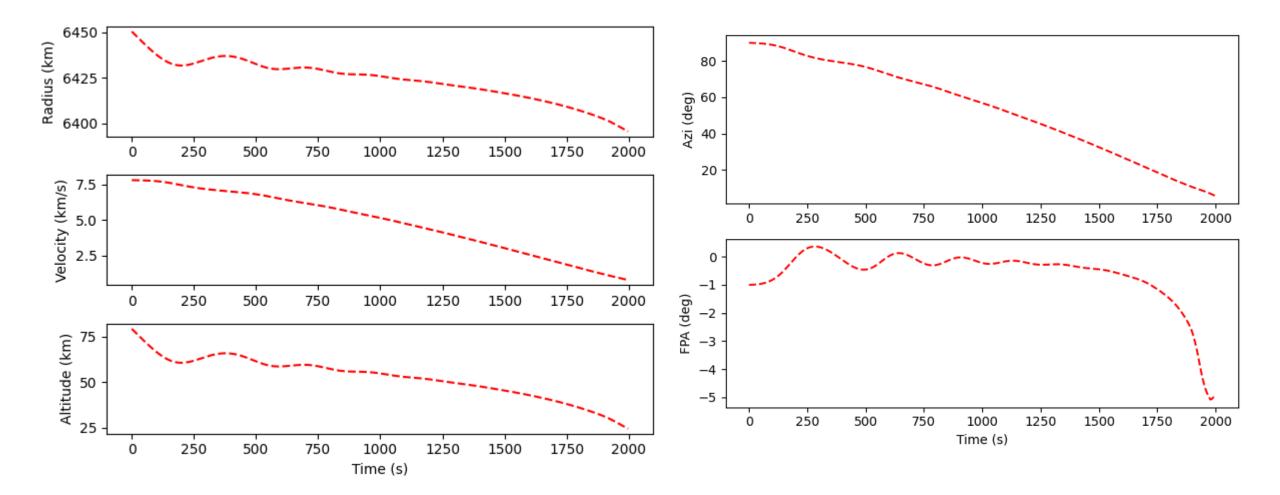
```
# Objetive funtion -
m.Maximize(lat*final)
# Setup solution
m.options.IMODE
m.options.MAX_ITER = 2000
m.options.NODES = 1
##m.options.OTOL = 1e-3
##m.options.RTOL = 1e-3
\#m.options.SOLVER = 3
m.solve(disp=True)
# get additional solution information ---
import json
with open(m.path+'//results.json') as f:
   results = json.load(f)
```

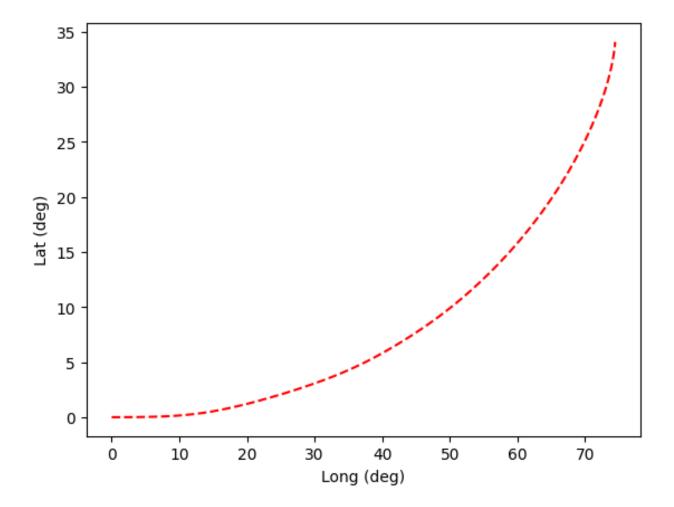
```
EXIT: Optimal Solution Found.

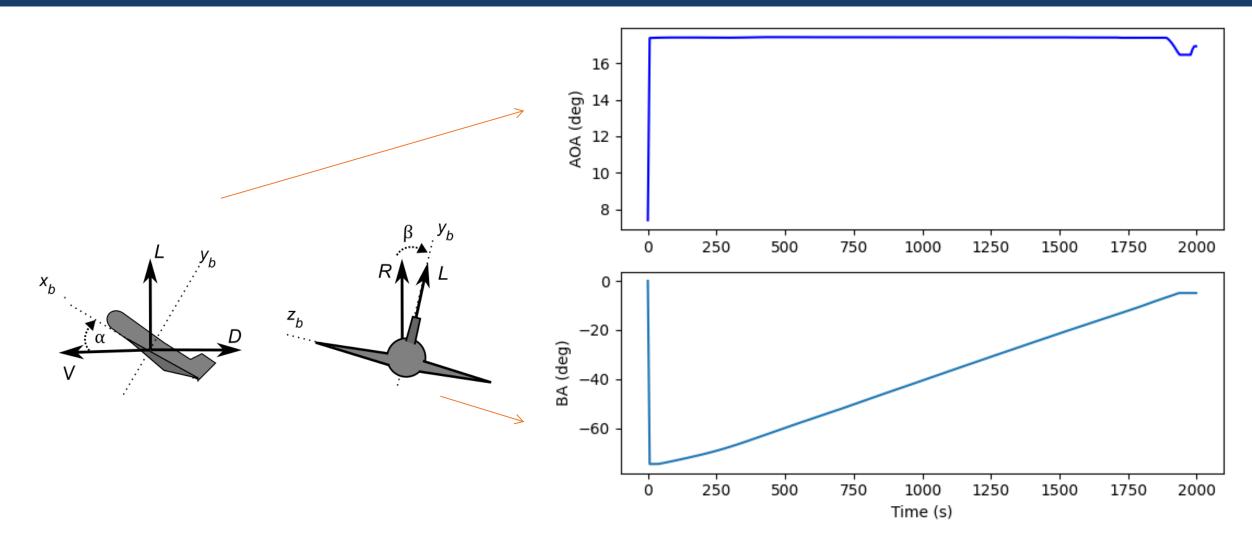
The solution was found.

The final value of the objective function is -0.594489548598784

Solver: IPOPT (v3.12)
Solution time: 6.52729999998701 sec
Objective: -0.594489548598784
Successful solution
```







### II. Final remarks

I invite you to follow the work of the researchers:

• PhD. Omkar Mulekar – Optimal control for landers based on ML.

Johnson Space Center - NASA

https://scholar.google.com/citations?hl=en&user=5HTQrk4AAAAJ

- PhD(C). Emanuela Gaglio ML based optimal control for aeromaneuvers.
- Optimal drag-based collision avoidance: Balancing miss distance and orbital decay. Acta Astronautica, 2024.

Scuola Superiore Meridionale - Italy

https://scholar.google.com/citations?hl=en&user=5HTQrk4AAAAJ

• PhD(C). Luis Mendoza Zambrano – Optimal Control on solar sailing and cislunar trajectories.

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## Thank you and have fun!

Questions?