Polytech Nice

Project report

 $\begin{array}{c} \text{numerical solution of the optimal orbit transfer problem in} \\ \text{the plane} \end{array}$

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1 Theory

1.1 Nomenclature

• r := satellite radial distance attracting body

• u := radial velocity

• v := tangential speed

 \bullet m := satellite mass

• $\frac{\partial m}{\partial t} = \dot{m} = -\frac{T}{g_0 \cdot Isp} :=$ dm motor mass flow

• $\phi :=$ angle of thrust direction with r

• $\mu := \text{gravitational constant of the attracting body}$

• θ := the angle formed by r with r(0)

1.2 Initial and Final Conditions:

$$r(0) = r_0, \quad u(0) = 0, \quad v(0) = \sqrt{\frac{\mu}{r_0}}$$
$$r(t_f) = r_f, \quad u(t_f) = 0, \quad v(t_f) = \sqrt{\frac{\mu}{r_f}}$$

1.3 Hamiltonian and Adjoint State Dynamics:

$$\dot{x} = \begin{bmatrix} \dot{r} \\ \dot{u} \\ \dot{v} \end{bmatrix} = f(x, t, \phi) = \begin{bmatrix} u \\ \frac{v^2}{r} - \frac{\mu}{r^2} + \frac{T}{m_0 + \dot{m} \times t} \sin(\phi) \\ -\frac{uv}{r} + \frac{T}{m_0 + \dot{m} \times t} \cos(\phi) \end{bmatrix}$$

$$\lambda^T = \begin{bmatrix} \lambda_r & \lambda_u & \lambda_v \end{bmatrix}$$

$$H = 1 + \lambda^T f = 1 + \lambda_r u + \lambda_u \left(\frac{v^2}{r} - \frac{\mu}{r^2} + \frac{T}{m_0 + \dot{m} \times t} \sin(\phi) \right) + \lambda_v \left(-\frac{uv}{r} + \frac{T}{m_0 + \dot{m} \times t} \cos(\phi) \right)$$

$$\dot{\lambda} = \begin{bmatrix} \dot{\lambda}_r \\ \dot{\lambda}_u \\ \dot{\lambda}_v \end{bmatrix} = -\frac{\partial f^T}{\partial x} \lambda = \begin{bmatrix} -\lambda_u \left(-\frac{v^2}{r^2} + \frac{2\mu}{r^3} \right) - \lambda_v \left(\frac{uv}{r^2} \right) \\ -\lambda_r + \lambda_v \frac{v}{r} \\ -\lambda_u \frac{2v}{r} + \lambda_v \frac{u}{r} \end{bmatrix}$$

1.4 Control Equation

$$H = 1 + \lambda^T f = 1 + \lambda_r u + \lambda_u \left(\frac{v^2}{r} - \frac{\mu}{r^2} + \frac{T}{m_0 + \dot{m} \times t} \sin(\phi) \right) + \lambda_v \left(-\frac{uv}{r} + \frac{T}{m_0 + \dot{m} \times t} \cos(\phi) \right)$$

$$\frac{\partial H}{\partial \phi} = \lambda_u \left(\frac{T}{m_0 + \dot{m} \times t} \cos(\phi) \right) - \lambda_v \left(\frac{T}{m_0 + \dot{m} \times t} \sin(\phi) \right) = 0$$

$$\frac{\partial H}{\partial \phi} = 0 \Leftrightarrow \lambda_u \cos(\phi) - \lambda_v \sin(\phi) = 0 \Rightarrow \tan(\phi) = \frac{\lambda_u}{\lambda_v}$$

1.5 Optimality Condition on t_f

$$\left[H = 1 + \lambda^T f = 1 + \lambda_r u + \lambda_u \left(\frac{v^2}{r} - \frac{\mu}{r^2} + \frac{T}{m_0 + \dot{m} \times t} \sin(\phi)\right) + \lambda_v \left(-\frac{uv}{r} + \frac{T}{m_0 + \dot{m} \times t} \cos(\phi)\right)\right]_{t_j} = 0$$

This condition can be replaced by the equivalent condition at $t_0 = 0$:

$$\left[\lambda^T \times \lambda\right]_{t_0} = 1$$

1.6 Two-Point Boundary Value Problem Formulation

Unknowns at $t = t_0 = 0$: $\lambda(t_0)^T = [\lambda_r \ \lambda_u \ \lambda_v]_{t_0}$ Unknown at $t = t_f$: t_f System of equations to solve:

$$g(\lambda_{r0}, \lambda_{u0}, \lambda_{v0}, t_f) = \begin{bmatrix} r(t_f) - r_f \\ u(t_f) \\ v(t_f) - \sqrt{\frac{\mu}{r_f}} \\ \left[1 + \lambda^T f\right]_{t_f} \end{bmatrix} = 0$$

2 Implementation

Algorithm 1 Problem data declaration

Data:

- Initial Conditions:
 - -AU = 149597870690
 - $-r_0 = AU$
 - $-u_0 = 0$

$$- v_0 = \sqrt{\frac{\mu_{\text{body}}}{r_0}}$$

- $-m_0 = 1000$
- Final Conditions:
 - $-r_f = 1.5 \times AU$
 - $-u_{f}=0$

$$-v_f = \sqrt{\frac{\mu_{\text{body}}}{r_f}}$$

- Power:
 - Set T to an array containing values from 0.1 to 0.6 with a step size of 0.1.
- Earth gravity
 - $-g_0 = 9.80665$
- Specific motor impulse
 - Isp = 3000
- sun's gravitational constant
 - $\mu_{body} = 1.32712440018e + 20$

2.1 Problem Data declaration

In this code, we have to normalize the data due to the problem of orders of magnitude. This allows us to make the resolution easier and more accurate for the solver.

Algorithm 2 Problem data declaration

Normalized unit:

- <u>Unit:</u>
 - $-DU = r_0;$
 - $-VU = v_0;$
 - $MU = m_0;$
 - $-TU = \frac{DU}{VU};$
 - $-FU = \frac{MU \times DU}{TU^2};$

2.2 System Dynamics Resolution function

Algorithm 3 Dynamical model of a spacecraft

Definition of the struct:

```
\begin{aligned} & \text{param.}DU = DU; \\ & \text{param.}VU = VU; \\ & \text{param.}MU = MU; \\ & \text{param.}TU = TU; \\ & \text{param.}FU = FU; \\ & \text{param.}\mu_{\text{body}} = \frac{\mu_{\text{body}}}{\text{param.}DU^3 \times \text{param.}TU^2}; \\ & \text{param.}T = \frac{T(1)}{\text{param.}FU}; \\ & \text{param.}g0_{\text{Isp}} = \frac{g0 \times Isp}{\text{param.}VU}; \\ & \text{param.}m0 = \frac{m0}{\text{param.}MU}; \\ & \text{param.}t0 = 0; \\ & \text{param.}x_{t0} = \begin{bmatrix} \frac{r0}{\text{param.}DU}} \\ \frac{v0}{\text{param.}VU} \\ \frac{v0}{\text{param.}VU} \end{bmatrix}; \\ & \text{param.}x_{tf} = \begin{bmatrix} \frac{rf}{\text{param.}DU}} \\ \frac{vf}{\text{param.}VU} \\ \frac{vf}{\text{param.}VU} \end{bmatrix}; \end{aligned}
```

Algorithm 4 Function dynpol

Input:

• t: A given time step

•
$$x = \begin{bmatrix} r \\ u \\ v \\ \lambda_r \\ \lambda_u \\ \lambda_v \end{bmatrix}$$
: the state of the dynamic system vector

function dynpol(t, x)

$$\begin{array}{l} \mu = \mathrm{param}.\mu_{\mathrm{body}} \\ T = \mathrm{param}.T; \\ g0_{\mathrm{Isp}} = \mathrm{param}.g0_{\mathrm{Isp}} \\ m0 = \mathrm{param}.m0; \\ t0 = \mathrm{param}.t0; \\ dm = -\frac{T}{g0_{\mathrm{Isp}}} \\ \phi = \mathrm{arctan}\left(\frac{x(5)}{x(6)}\right) \\ dx = x; \\ dx(1) = x(2); \\ dx(2) = \frac{x(3)^2}{x(1)} - \frac{\mu}{x(1)^2} + \frac{T}{m0 + dm \times t} \times \sin(\phi) \\ dx(3) = -\frac{x(2) \times x(3)}{x(1)} + \frac{T}{m0 + dm \times t} \times \cos(\phi) \\ dx(4) = -x(5) \times \left(-\frac{x(3)^2}{x(1)^2} + 2 \times \frac{\mu}{x(1)^3}\right) - x(6) \times \frac{x(2) \times x(3)}{x(1)^2} \\ dx(5) = -x(4) + x(6) \times \frac{x(3)}{x(1)}; \\ dx(6) = -x(5) \times (2 \times \frac{x(3)}{x(1)}) + x(6) \times \frac{x(2)}{x(1)}; \\ \mathrm{nd} \ \ \mathrm{function} \end{array}$$

end function

Algorithm 5 Function gnultmin

Input:

end function

```
• p = \begin{bmatrix} t_f \\ \lambda_{r_0} \\ \lambda_{u_0} \\ \lambda_{v_0} \end{bmatrix}: the state of the dynamic system vector
function GNMT(p)
    t0 = param.t0;
    r0 = \text{param.} x_{t0}(1);
     u0 = \text{param.} x_{t0}(2);
     v0 = \text{param.} x_{t0}(3);
     rf = \text{param.} x_{tf}(1);
     uf = \text{param.} x_{tf}(2);
     vf = \text{param.} x_{tf}(3);
     y0 = [r0 \quad u0 \quad v0 \quad p(2) \quad p(3) \quad p(4)];
    tf = p(1);
     atol = 1e - 10;
    rtol = 1e - 10;
     y = ode("rk", y0, t0, tf, dynpol, atol, rtol);
     gnmt = p;
    gnmt(1) = y(1) - rf;
    gnmt(2) = y(2) - uf;
     gnmt(3) = y(3) - \sqrt{\operatorname{param}.\mu_{\text{body}}/rf};
    gnmt(4) = p(2)^2 + p(3)^2 + p(4)^2 - 1;
```

The function gnultmin() takes p as a parameter and returns the column vector gnmt.

Absolute and relative tolerances of 10^{-10} have been chosen for solving the differential equation using the function ode(). If the order of magnitude of the relative tolerance is smaller than 10^{-10} , Scilab displays a warning message and increases this tolerance to have an order of magnitude of 10^{-10} .

We aim to achieve better accuracy than that of the solver.

Algorithm 6 Control History Calculation

```
1: for i = 1 to 6 do
        param.T \leftarrow \frac{T(i)}{param.FU}
 3:
         [x, v, info] \leftarrow \text{fsolve}(p0, gnultmin)
         disp(info)
 4:
         \operatorname{disp}(v)
 5:
         InitialCond \leftarrow [param.x t0(1); param.x t0(2); param.x t0(3); x(2); x(3); x(4)]
 6:
 7:
         t \leftarrow 0: 0.01: x(1)
 8:
         sol \leftarrow ode(InitialCond, param.t0, t, dynpol)
         timeInDays \leftarrow t \times \frac{param.TU}{86400}
 9:
         AngleInDegrees \leftarrow atan(sol(5,:),sol(6,:)) \times \frac{360}{2\pi}
10:
         figure
11:
         plot(timeInDays, AngleInDegrees)
12:
13:
         xlabel('Time(days)')
         ylabel('Controlindeg')
14:
         powerInNewton \leftarrow param.T \times param.FU
15:
         xgrid
16:
17:
         h \leftarrow \gcd()
         h.\text{background} \leftarrow \text{color}('white')
18:
         title('Controlhistory for' + string(powerInNewton) + 'N')
19:
20: end for
```

Each thrust (in Newtons) is contained in T. We proceed with increasing thrusts, starting with the lowest one (here 0.1), so the thrust duration decreases. The very first time the function gnultmin() is passed as a parameter to the function fsolve(), i.e., to calculate the trajectory with the first thrust, we use p_0 for the initialization of gnultmin(). The result of the first use of fsolve() called p_1 will serve for the second time when we pass gnultmin() as a parameter to fsolve(), hence for calculating the trajectory with the second thrust. Thus, for each thrust except the first one, we use the result of fsolve() from the previous thrust. With p_0 , we can then obtain p_i for all $i \in [1; 6]$.

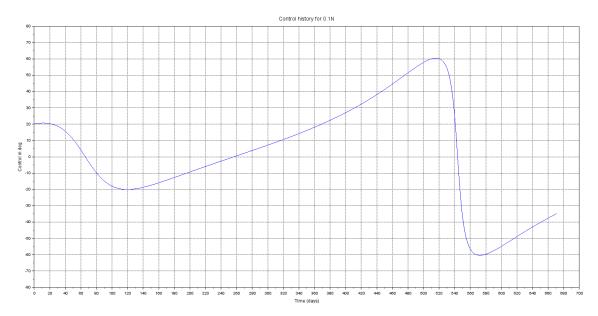


Figure 1: thrusts of 0.1N

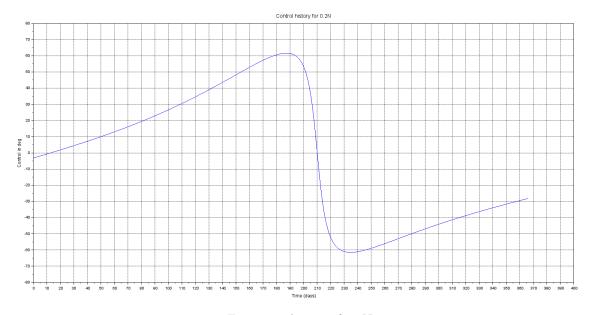


Figure 2: thrusts of 0.2N

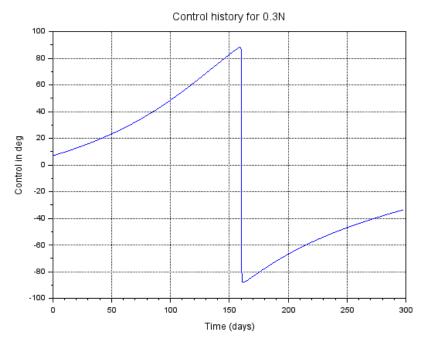


Figure 3: thrusts of 0.3N

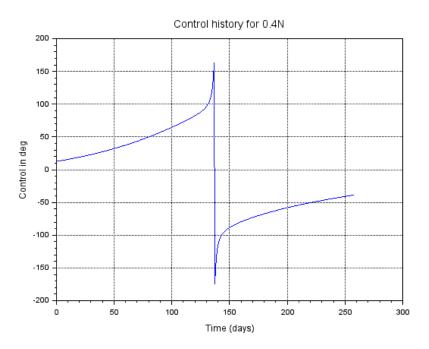


Figure 4: thrusts of 0.4N

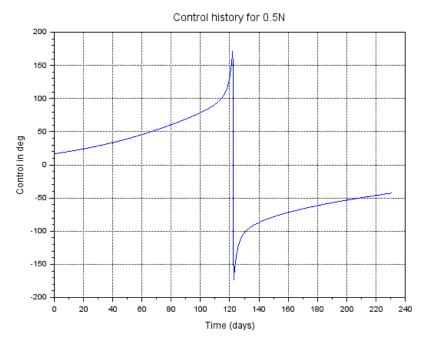


Figure 5: thrusts of 0.5N

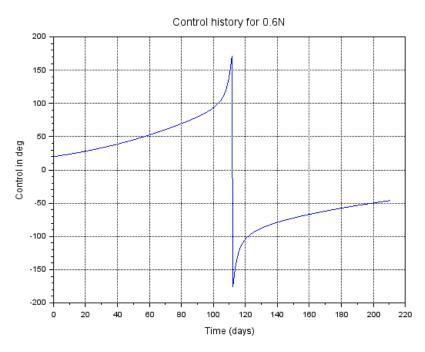


Figure 6: thrusts of 0.6N