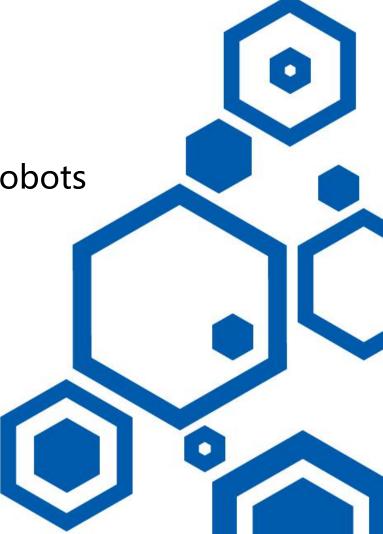


Motion Planning for Mobile Robots 第四章作业讲评





第一题



For the OBVP problem stated in slides p.25-p.29, please get the optimal solution (control, state, and time) for partially free final state case. Suppose the position is fixed, **velocity** and **acceleration** are free here.

State: $\mathbf{s}_k = [p_k, v_k, a_k]^T$ Input: $u_k = j_k$

System model: $\dot{\mathbf{s}} = f_s(\mathbf{s},u) = [v,a,j]^T, \quad \mathbf{s}(0) = [p_0,v_0,a_0], \quad t \in [0,\mathrm{T}]$

Objective function: $J=h\left(\mathbf{s}(\mathrm{T})\right)+\int_{0}^{\mathrm{T}}g\left(\mathbf{s}(t),u(t)\right)dt,\quad g(\mathbf{s}(t),u(t))=rac{1}{\mathrm{T}}j(t)^{2}$

第一题



根据题意,这里只给定最终时刻的位置约束 p_f ,因此定义 $h(\mathbf{s}(\mathbf{T}))=0$ 。最终的代价函数为:

$$J = h\left(\mathbf{s}(\mathrm{T})\right) + \int_0^{\mathrm{T}} g\left(\mathbf{s}(t), u(t)\right) dt = \int_0^{\mathrm{T}} rac{1}{\mathrm{T}} j(t)^2 dt$$
 (1)

定义协态 $\lambda = [\lambda_1, \lambda_2, \lambda_3]^T$,则哈密尔顿函数为:

$$H(\mathbf{s}, u, \boldsymbol{\lambda}) = g(\mathbf{s}, u) + \boldsymbol{\lambda}^T f(\mathbf{s}, u) = \frac{1}{\mathrm{T}} j^2 + \lambda_1 v + \lambda_2 a + \lambda_3 j$$
 (2)

应用极小值原理,有:

$$\dot{\boldsymbol{\lambda}} = -\nabla \mathbf{s} H(\mathbf{s}(*, u) *, \boldsymbol{\lambda}) = [0, -\lambda_1, -\lambda_2]^T$$
(3)

● 边界条件:

$$\lambda_{2,3}(T) = \nabla_{\mathbf{s}_{2,3}} h(\mathbf{s}^*(T)) = [0,0]^T$$
 (4)

其中 $\lambda_{2,3}$ 表示 λ 的第2和3个变量。联立公式 (3) 和 (4),可以求得 $\lambda(t)$ 的表达式:

$$oldsymbol{\lambda}(t) = rac{1}{\mathrm{T}} egin{bmatrix} -2lpha \ 2lpha(t-\mathrm{T}) \ -lpha(t-\mathrm{T})^2 \end{bmatrix}$$
 (5)

For (partially)-free final state problem:

given
$$s_i(T)$$
, $i \in I$

We have boundary condition for other costate:

$$\lambda_{j}(T) = \frac{\partial h(s^{*}(T))}{\partial s_{j}}, \text{ for } j \neq i$$

第一题



其中, v_f^*, a_f^* 表示最优的末状态。令 $\frac{\partial H(\mathbf{s}^*(t), j(t), \boldsymbol{\lambda}(t))}{\partial j(t)} = 0$,求解最优控制:

$$egin{aligned} u^*(t) &= rg\min_{j(t)} H(\mathbf{s}^*(t), j(t), oldsymbol{\lambda}(t)) = -rac{\lambda_3 \mathrm{T}}{2} \ &= rac{1}{2} lpha(t-\mathrm{T})^2 \end{aligned}$$

根据最优控制和初始状态,则最优状态为:

$$\mathbf{s}^*(t) = egin{bmatrix} rac{lpha}{120} (t-\mathrm{T})^5 + rac{(a_0 + rac{lpha}{6}\mathrm{T}^3)}{2} t^2 + (v_0 - rac{lpha}{24}\mathrm{T}^4) t + (p_0 + rac{lpha}{120}\mathrm{T}^5) \ rac{lpha}{24} (t-\mathrm{T})^4 + (a_0 + rac{lpha}{6}\mathrm{T}^3) t + (v_0 - rac{lpha}{24}\mathrm{T}^4) \ rac{lpha}{6} (t-\mathrm{T})^3 + (a_0 + rac{lpha}{6}\mathrm{T}^3) \end{bmatrix}$$

● 由末状态的位置约束, 有:

$$p_f = rac{1}{2}(a_0 \mathrm{T}^2 + rac{lpha}{6} \mathrm{T}^5) + (v_0 \mathrm{T} - rac{lpha}{24} \mathrm{T}^5) + (p_0 + rac{lpha}{120} \mathrm{T}^5)$$

则可以求解 α :

$$lpha = rac{20 \Delta p}{{
m T}^5}, \quad \Delta p = p_f - p_0 - rac{1}{2} a_0 {
m T}^2 - v_0 {
m T}$$

For (partially)-free final state problem:

given
$$s_i(T)$$
, $i \in I$

We have boundary condition for other costate:

$$\lambda_{j}(T) = \frac{\partial h(s^{*}(T))}{\partial s_{j}}, \text{ for } j \neq i$$

(9) 最终J化简为:
$$J = \int_0^{\mathrm{T}} \frac{1}{\mathrm{T}} j^*(t)^2 dt = \frac{20\Delta P^2}{\mathrm{T}^6}$$

(8)



The costate is solved as:

$$\lambda(t) = \frac{1}{T} \begin{bmatrix} \frac{-2\alpha}{2\alpha t + 2\beta} \\ -\alpha t^2 - 2\beta t - 2\gamma \end{bmatrix} = \nabla h(s^*(T)) \qquad \begin{cases} \begin{bmatrix} 2\alpha T + 2\beta \\ -\alpha T^2 - 2\beta T - 2\gamma \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \bullet \\ \frac{\alpha}{120} T^5 + \frac{\beta}{24} T^4 + \frac{\gamma}{6} T^3 = \Delta p \bullet \end{cases}$$

 α, β, γ Is solved as:

$$\begin{bmatrix} \frac{1}{120}T^5 & \frac{1}{24}T^4 & \frac{1}{6}T^3 \\ \frac{1}{24}T^4 & \frac{1}{6}T^3 & \frac{1}{2}T^2 \\ \frac{1}{6}T^3 & \frac{1}{2}T^2 & T \end{bmatrix} \begin{bmatrix} \alpha \\ \beta \\ \gamma \end{bmatrix} = \begin{bmatrix} \Delta p \\ \Delta v \\ \Delta \alpha \end{bmatrix}$$

$$\begin{cases} \begin{bmatrix} 2\alpha T + 2\beta \\ -\alpha T^2 - 2\beta T - 2\gamma \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \bullet \\ \frac{\alpha}{120} T^5 + \frac{\beta}{24} T^4 + \frac{\gamma}{6} T^3 = \Delta p \bullet \end{cases}$$

$$\begin{bmatrix} \alpha \\ \beta \\ \gamma \end{bmatrix} = \frac{1}{T^5} \begin{bmatrix} 20 \\ -20T \\ 10T^2 \end{bmatrix} \Delta \mathbf{p}$$

For (partially)-free final state problem:

given
$$s_i(T)$$
, $i \in I$

We have boundary condition for other costate:

$$\lambda_{j}(T) = \frac{\partial h(s^{*}(T))}{\partial s_{j}}, \text{ for } j \neq i$$

第二题



Build an ego-graph of the linear modeled robot. Select the best trajectory closest to the planning target.

$$J=\int_{ au}^{\mathrm{T}}\left(1+a_{x}^{2}+a_{y}^{2}+a_{z}^{2}
ight)dt$$

$$J(T) = T + 12(\delta_{px}^2 + \delta_{py}^2 + \delta_{pz}^2)T^{-3} - 12(\delta_{px}\delta_{vx1} + \delta_{py}\delta_{vx1} + \delta_{pz}\delta_{vx1})T^{-2} + 4(\delta_{vx2} + \delta_{vy2} + \delta_{vz2})T^{-1}$$

其中, δ_i 定义如下:

$$\left\{egin{aligned} \delta_{pi} &= p_{Ti} - p_{ au i} \ \delta_{vi1} &= v_{Ti} + v_{ au i} \ \delta_{vi2} &= v_{Ti}^2 + v_{Ti} v_{ au i} + v_{ au i}^2 \end{aligned}
ight.$$

要求取目标函数 J 的极小值,仅需要对其求导,令其导数为 0 即可求得最优的时间 T。

$$rac{\partial J(T)}{\partial T} = 1 - 36(\delta_{px}^2 + \delta_{py}^2 + \delta_{pz}^2)T^{-4} + 24(\delta_{px}\delta_{vx1} + \delta_{py}\delta_{vx1} + \delta_{pz}\delta_{vx1})T^{-3} - 4(\delta_{vx2} + \delta_{vy2} + \delta_{vz2})T^{-2}$$

第二题



Roots of Polynomials

- 解析解(degree <= 4)
- 多项式伴随矩阵求根(Eigen手写/PolynomialSolver)

$$\frac{\partial J(T)}{\partial T} = 1 - 36(\delta_{px}^2 + \delta_{py}^2 + \delta_{pz}^2)T^{-4} + 24(\delta_{px}\delta_{vx1} + \delta_{py}\delta_{vx1} + \delta_{pz}\delta_{vx1})T^{-3} - 4(\delta_{vx2} + \delta_{vy2} + \delta_{vz2})T^{-2}$$

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In linear algebra, the Frobenius companion matrix of the monic polynomial p(t)=c_0+c_1t+\cdots+c_{n-1}t^{n-1}+t^n\;, is the square matrix defined as C(p)=\begin{bmatrix}0&0&\dots&0&-c_0\\1&0&\dots&0&-c_1\\0&1&\dots&0&-c_2\\\vdots&\vdots&\ddots&\vdots&\vdots\\0&0&\dots&1&-c_{n-1}\end{bmatrix}. 正根
```

正根 J(T)最小

```
PolynomialSolver<double, Eigen::Dynamic> solver;
VectorXd coeff(5);
coeff[0] = -36.0f * param1;
coeff[1] = 24.0f * param2;
coeff[2] = -4.0f * param3;
coeff[3] = 0.0f;
coeff[4] = 1.0f;
solver.compute(coeff);
const Eigen::PolynomialSolver<double, Eigen::Dynamic>::RootsType& root = solver.roots();
```

$$J(T) = T + 12(\delta_{px}^2 + \delta_{py}^2 + \delta_{pz}^2)T^{-3} - 12(\delta_{px}^*\delta_{vx1} + \delta_{py}\delta_{vx1} + \delta_{pz}\delta_{vx1})T^{-2} + 4(\delta_{vx2} + \delta_{vy2} + \delta_{vz2})T^{-1}$$

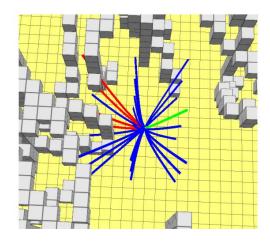
https://pdfs.semanticscholar.org/2b69/8d3602fe6ce5ec2b29f70614aa15e36fd397.pdf http://web.mit.edu/18.06/www/Spring17/Eigenvalue-Polynomials.pdf

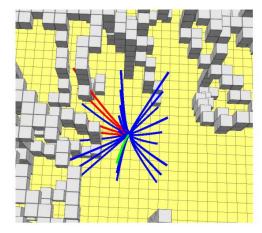
第二题



Roots of Polynomials

- Everyone learns the quadratic formula to find roots of a quadratic (degree-2) polynomial.
- There is a (horrible) cubic formula to find the roots of any cubic (degree-3) polynomial.
- There is a (terrifying) quartic formula to find the roots of any quartic (degree-4) polynomial.
- There is no formula (in terms of a *finite number* of $\pm, \times, \div, {}^{n}\sqrt{\ }$) for the roots of an **arbitrary quintic** polynomial or **any degree** \geq **5**. This is the Abel–Ruffini theorem, proved in the 19th century.







感谢各位聆听

Thanks for Listening



