



深蓝学院
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第四章作业分享

主讲人 tkalpha



- 第一部分：计算推导
- 第二部分：ROS作业
- 第三部分：问题与解答

● 求解过程：基本与课件中的步骤一致

• 优化的目标函数

$$J = \int_0^T g(x, u) dt = \int_0^T (1 + u^T R u) dt = \int_0^T (1 + a_x^2 + a_y^2 + a_z^2) dt$$

• 状态转移方程

$$x = \begin{bmatrix} p_x \\ p_y \\ p_z \\ v_x \\ v_y \\ v_z \end{bmatrix}, u = \begin{bmatrix} a_x \\ a_y \\ a_z \end{bmatrix}, \dot{x} = f(x, u) = \begin{bmatrix} v_x \\ v_y \\ v_z \\ a_x \\ a_y \\ a_z \end{bmatrix}$$

• 求解

• 协状态

$$\lambda = [\lambda_1 \quad \lambda_2 \quad \lambda_3 \quad \lambda_4 \quad \lambda_5 \quad \lambda_6]^T$$

• 哈密顿方程

$$H(x, u, \lambda) = g(x, u) + \lambda^T f(x, u)$$

$$H(x, u, \lambda) = (1 + a_x^2 + a_y^2 + a_z^2) + \lambda^T f(x, u)$$

● 求解过程2

• 求解 λ

$$\dot{\lambda} = -\nabla H(x^*, u^*, \lambda) = [0 \quad 0 \quad 0 \quad -\lambda_1 \quad -\lambda_2 \quad -\lambda_3]^T$$
$$\lambda = \begin{bmatrix} 2\alpha_1 \\ 2\alpha_2 \\ 2\alpha_3 \\ -2\alpha_1 t - 2\beta_1 \\ -2\alpha_2 t - 2\beta_2 \\ -2\alpha_3 t - 2\beta_3 \end{bmatrix}$$

• 理想输入

$$u^* = \arg \min_u H(x^*, u, \lambda) = \begin{bmatrix} \alpha_1 t + \beta_1 \\ \alpha_2 t + \beta_2 \\ \alpha_3 t + \beta_3 \end{bmatrix}$$

• 理想状态

$$\dot{x}^* = f(x^*, u^*)$$
$$x^* = \begin{bmatrix} \frac{1}{6}\alpha_1 t^3 + \frac{1}{2}\beta_1 t^2 + v_{x0}t + p_{x0}t \\ \frac{1}{6}\alpha_2 t^3 + \frac{1}{2}\beta_2 t^2 + v_{y0}t + p_{y0}t \\ \frac{1}{6}\alpha_3 t^3 + \frac{1}{2}\beta_3 t^2 + v_{z0}t + p_{z0}t \\ \frac{1}{2}\alpha_1 t^2 + \beta_1 t + v_{x0} \\ \frac{1}{2}\alpha_2 t^2 + \beta_2 t + v_{y0} \\ \frac{1}{2}\alpha_3 t^2 + \beta_3 t + v_{z0} \end{bmatrix}$$
$$x^*(0) = x(0) = [p_{x0} \quad p_{y0} \quad p_{z0} \quad v_{x0} \quad v_{y0} \quad v_{z0}]^T$$

● 求解过程3

- 根据末尾状态求解 α 和 β

$$x^*(T) = \begin{bmatrix} p_{xf} & p_{yf} & p_{zf} & v_{xf} & v_{yf} & v_{zf} \end{bmatrix}^T$$

记

$$\begin{bmatrix} \Delta p_x \\ \Delta p_y \\ \Delta p_z \\ \Delta v_x \\ \Delta v_y \\ \Delta v_z \end{bmatrix} = \begin{bmatrix} p_{xf} - v_{x0}T - p_{x0} \\ p_{yf} - v_{y0}T - p_{y0} \\ p_{zf} - v_{z0}T - p_{z0} \\ v_{xf} - v_{x0} \\ v_{yf} - v_{y0} \\ v_{zf} - v_{z0} \end{bmatrix}$$

α 和 β 完全可以表示为初始状态、结束状态与 T 的函数

$$\begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \\ \beta_1 \\ \beta_2 \\ \beta_3 \end{bmatrix} = \begin{bmatrix} -\frac{12}{T^3} & 0 & 0 & \frac{6}{T^2} & 0 & 0 \\ 0 & -\frac{12}{T^3} & 0 & 0 & \frac{6}{T^2} & 0 \\ 0 & 0 & -\frac{12}{T^3} & 0 & 0 & \frac{6}{T^2} \\ \frac{6}{T^2} & 0 & 0 & -\frac{2}{T} & 0 & 0 \\ 0 & \frac{6}{T^2} & 0 & 0 & -\frac{2}{T} & 0 \\ 0 & 0 & \frac{6}{T^2} & 0 & 0 & -\frac{2}{T} \end{bmatrix} \begin{bmatrix} \Delta p_x \\ \Delta p_y \\ \Delta p_z \\ \Delta v_x \\ \Delta v_y \\ \Delta v_z \end{bmatrix}$$

将 α 和 β 作为方程组的自变量

$$\begin{bmatrix} \frac{1}{6}T^3 & 0 & 0 & \frac{1}{2}T^2 & 0 & 0 \\ 0 & \frac{1}{6}T^3 & 0 & 0 & \frac{1}{2}T^2 & 0 \\ 0 & 0 & \frac{1}{6}T^3 & 0 & 0 & \frac{1}{2}T^2 \\ \frac{1}{2}T^2 & 0 & 0 & T & 0 & 0 \\ 0 & \frac{1}{2}T^2 & 0 & 0 & T & 0 \\ 0 & 0 & \frac{1}{2}T^2 & 0 & 0 & T \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \\ \beta_1 \\ \beta_2 \\ \beta_3 \end{bmatrix} = \begin{bmatrix} \Delta p_x \\ \Delta p_y \\ \Delta p_z \\ \Delta v_x \\ \Delta v_y \\ \Delta v_z \end{bmatrix}$$

● 求解过程4

- 基于目标函数求得最优的T

$$J = \int_0^T (1 + a_x^2 + a_y^2 + a_z^2) dt$$
$$= T + (\frac{1}{3}\alpha_1^2 T^3 + \alpha_1 \beta_1 T^2 + \beta_1^2 T) + (\frac{1}{3}\alpha_2^2 T^3 + \alpha_2 \beta_2 T^2 + \beta_2^2 T) + (\frac{1}{3}\alpha_3^2 T^3 + \alpha_3 \beta_3 T^2 + \beta_3^2 T)$$

代入 α 和 β 的T表示方法可得

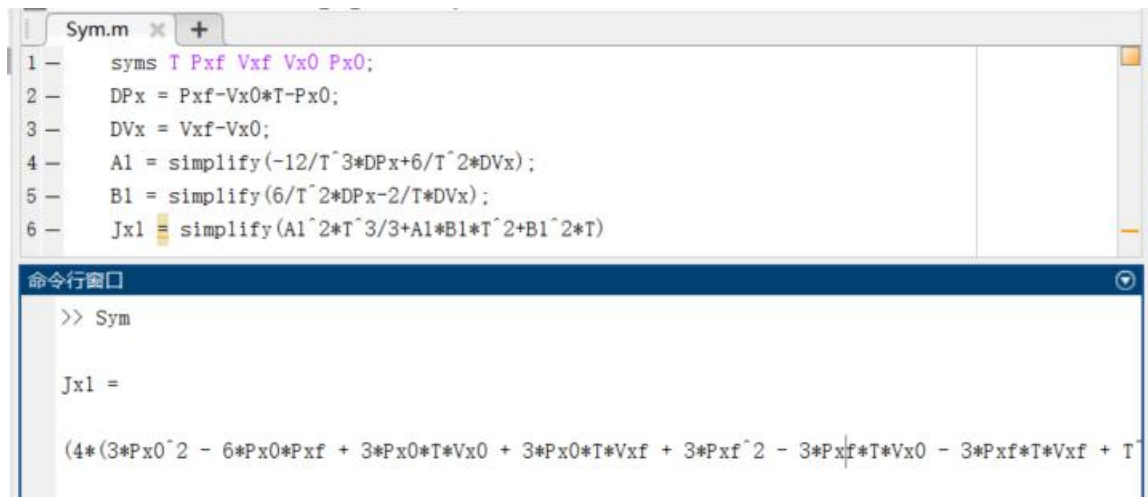
$$J = T + \frac{4}{T}((v_{xf}^2 + v_{xf}v_{x0} + v_{x0}^2) + (v_{yf}^2 + v_{yf}v_{y0} + v_{y0}^2) + (v_{zf}^2 + v_{zf}v_{z0} + v_{z0}^2)) - \frac{12}{T^2}((p_{xf} - p_{x0})(v_{xf} + v_{x0}) + (p_{yf} - p_{y0})(v_{yf} + v_{y0}) + (p_{zf} - p_{z0})(v_{zf} + v_{z0})) + \frac{12}{T^3}((p_{xf} - p_{x0})^2 + (p_{yf} - p_{y0})^2 + (p_{zf} - p_{z0})^2)$$

然后求 $\frac{dJ}{dT} = 0$ 时T符合要求的根

$$T^4 - 4T^2((v_{xf}^2 + v_{xf}v_{x0} + v_{x0}^2) + (v_{yf}^2 + v_{yf}v_{y0} + v_{y0}^2) + (v_{zf}^2 + v_{zf}v_{z0} + v_{z0}^2)) + 24T((p_{xf} - p_{x0})(v_{xf} + v_{x0}) + (p_{yf} - p_{y0})(v_{yf} + v_{y0}) + (p_{zf} - p_{z0})(v_{zf} + v_{z0})) - 36((p_{xf} - p_{x0})^2 + (p_{yf} - p_{y0})^2 + (p_{zf} - p_{z0})^2) = 0$$

● 求解过程技巧

复杂的推导不需要手算，可以借助MATLAB或者Mathematica的数值计算功能完成。



```
Sym.m  x +
1 - syms T Pxf Vxf Vx0 Px0;
2 - DPx = Pxf-Vx0*T-Px0;
3 - DVx = Vxf-Vx0;
4 - A1 = simplify(-12/T^3*DPx+6/T^2*DVx);
5 - B1 = simplify(6/T^2*DPx-2/T*DVx);
6 - Jx1 = simplify(A1^2*T^3/3+A1*B1*T^2+B1^2*T)

命令行窗口
>> Sym

Jx1 =

(4*(3*Px0^2 - 6*Px0*Pxf + 3*Px0*T*Vx0 + 3*Px0*T*Vxf + 3*Pxf^2 - 3*Pxf*T*Vx0 - 3*Pxf*T*Vxf + T
```

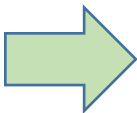
- 第一部分：计算推导
- **第二部分：ROS作业**
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● [PART I] 轨迹库的生成

直接基于恒加速运动的性质，或者基于数值积分方法（改进欧拉法/龙格库塔法）与状态方程可以推导得到

• 状态转移方程

$$x = \begin{bmatrix} p_x \\ p_y \\ p_z \\ v_x \\ v_y \\ v_z \end{bmatrix}, u = \begin{bmatrix} a_x \\ a_y \\ a_z \end{bmatrix}, \dot{x} = f(x, u) = \begin{bmatrix} v_x \\ v_y \\ v_z \\ a_x \\ a_y \\ a_z \end{bmatrix}$$



$$v_1 = v_0 + a_0 t$$

$$p_1 = p_0 + v_0 t + 0.5 a_0 t^2$$

```
pos(0) = pos(0) + vel(0)*delta_time + 0.5*acc_input(0)*delta_time*delta_time;  
pos(1) = pos(1) + vel(1)*delta_time + 0.5*acc_input(1)*delta_time*delta_time;  
pos(2) = pos(2) + vel(2)*delta_time + 0.5*acc_input(2)*delta_time*delta_time;  
  
vel(0) = vel(0) + acc_input(0)*delta_time;  
vel(1) = vel(1) + acc_input(1)*delta_time;  
vel(2) = vel(2) + acc_input(2)*delta_time;
```

● [PART II] OBVP求解

1. 核心是求解多项式。作业里使用伴随矩阵法实现。

- 对于多项式 $P(x) = x^n + c_{n-1}x^{n-1} + c_{n-2}x^{n-2} + \cdots + c_1x + c_0$

- 其伴随矩阵为

$$M_x = \begin{bmatrix} 0 & 0 & \cdots & 0 & -c_0 \\ 1 & 0 & \cdots & 0 & -c_1 \\ 0 & 1 & \cdots & 0 & -c_2 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \cdots & 1 & -c_{n-1} \end{bmatrix}$$

- 伴随矩阵的特征值就是多项式的根

2. 除此之外还可以使用费拉里法显式求解，参考[代码](#)

3. 助教作业思路的课件里面还有提到Eigen::PolynomialSolver, ZJU-FAST-Lab/Root-Finder等求解库

```
double Homeworktool::OptimalBVP(Eigen::Vector3d _start_position,Eigen::Vector3d _start_velocity,Eigen::Vector3d _target_position)
{
    double optimal_cost = 100000;
    // coefficient calculation
    double j_a1 = 4.0*(pow(_start_velocity(0),2)+pow(_start_velocity(1),2)+pow(_start_velocity(2),2));
    double j_a2 = -12.0*((_target_position(0)-_start_position(0))*_start_velocity(0)
        +(_target_position(1)-_start_position(1))*_start_velocity(1)
        +(_target_position(2)-_start_position(2))*_start_velocity(2));
    double j_a3 = 12.0*(pow((_target_position(0)-_start_position(0)),2)
        +pow((_target_position(1)-_start_position(1)),2)
        +pow((_target_position(2)-_start_position(2)),2));

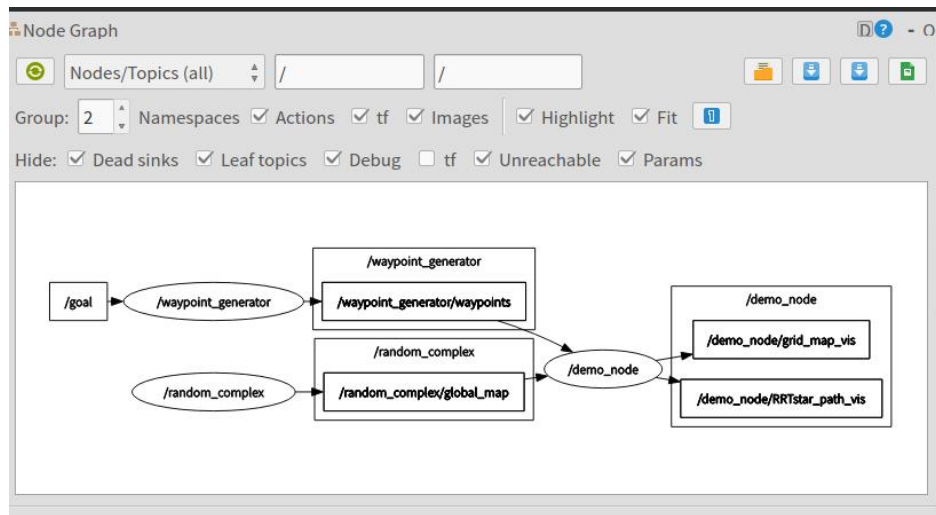
    Matrix4d t_comp = Matrix4d::Zero(4,4); // matrix generation
    t_comp(1,0) = 1.0; t_comp(2,1) = 1.0; t_comp(3,2) = 1.0;
    t_comp(2,3) = j_a1;
    t_comp(1,3) = 2.0*j_a2;
    t_comp(0,3) = 3.0*j_a3;

    Matrix<complex<double>,Dynamic,Dynamic> t_comp_eig = t_comp.eigenvalues();

    int t_num=t_comp_eig.rows(), kk=0, flag=0; // find a positive real root
    for(; kk<t_num; kk++){
        if(fabs(t_comp_eig(kk,0).imag()-0.0)<1e-8 && t_comp_eig(kk,0).real()>0){
            flag = 1;
            break;
        }
    }
    if(flag==1){ // optimal T and cost calculation
        double t = t_comp_eig(kk,0).real();
        double j = t + j_a1/t + j_a2/pow(t,2) + j_a3/pow(t,3);
        optimal_cost = j;
    }
    return optimal_cost;
}
```

代码实现

- 1.生成伴随矩阵
- 2.选择合适的根
- 3.求最优的T,cost

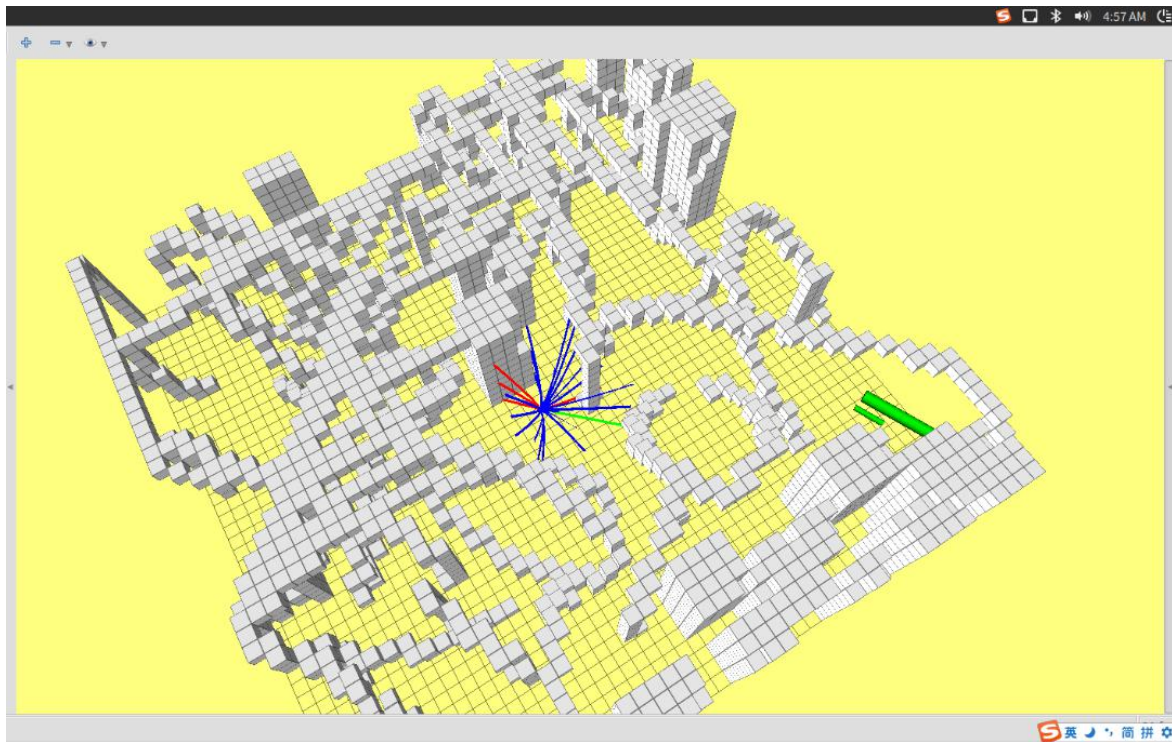


ROS rqt_graph

```
process[rviz-5]: started with pid [3064]
[ INFO] [1629533994.671714908]: [node] receive the planning target
[ INFO] [1629534020.121975636]: [node] receive the planning target
[ INFO] [1629534024.461538702]: [node] receive the planning target
[ INFO] [1629534202.101361454]: [node] receive the planning target
```

ROS_INFO

ROS实现



最终效果

ROS rviz

在线问答 - 其他

● 群里有关于边界条件符号的讨论

1. 公式推导

Define the Hamiltonian function:
$$H(s,u,\lambda) = \frac{1}{2} \dot{s}^2 + \lambda^T f(s,u)$$
$$= \frac{1}{2} \dot{s}^2 + \lambda_1 v + \lambda_2 a + \lambda_3 j$$

- 由Pontryagin's minimum principle:
$$\dot{\lambda}(t) = -\nabla_s H(s^*(t), u^*(t), \lambda(t))$$
$$\lambda(t) = \frac{1}{T} \begin{bmatrix} -2\alpha \\ 2\alpha t + 2\beta \\ -\alpha t^2 - 2\beta t - 2\gamma \end{bmatrix}$$
- 使用边界条件:
$$\lambda(T) = \nabla h(s^*(T))$$
- 由于末状态 v 和 a 并不相关
- 因此:
$$\lambda_2(T) = 0, \lambda_3(T) = 0$$
- 因此表达式可以化简为:
$$\begin{cases} \beta = -\alpha T \\ \gamma = \frac{\alpha}{2} T^2 \end{cases} \quad \lambda(t) = \frac{1}{T} \begin{bmatrix} -2\alpha \\ 2\alpha(t-T) \\ -\alpha t^2 + 2\alpha T t - \alpha T^2 \end{bmatrix}$$

Optimal Boundary Value Problem (OBVP)

- Previous slides are about fixed final state problem.
- How about the final state is (partially)-free?
 - Did you notice where is the boundary condition?

For fixed final state problem:

$$h(s(T)) = \begin{cases} 0, & \text{if } s = s(T) \\ \infty, & \text{otherwise} \end{cases} \quad \text{Not differentiable}$$

So we discard this condition, and use given $s(T)$ to directly solve for unknown variables

$$s'(t) = \begin{bmatrix} \frac{\alpha}{120} + \frac{\beta}{24} + \frac{\gamma}{6} t^2 + \frac{a_0}{2} t + v_0^T + p_0 \\ \frac{\alpha}{24} + \frac{\beta}{6} + \frac{\gamma}{2} t + a_0 \\ \frac{\gamma}{6} + \frac{\beta}{2} + \gamma t + a_0 \end{bmatrix}$$

For (partially)-free final state problem:

$$\text{given } s_j(T), i \in I$$

We have boundary condition for other costate:

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

Then we solve this problem again.

群友
@mengchaocheng

根据相关参考文献，
左下角的课件PPT才是
对的。不过不影响作
业的结果。

$pr = \nabla p_0$.

We will assume that the constraint sets E_k are convex, so that we can apply the optimality condition

$$\sum_{k=1}^{N-1} \nabla_{u_k} J(u_0^*, \dots, u_{N-1}^*)^T (u_k - u_k^*) \geq 0,$$

for all feasible (u_0, \dots, u_{N-1}) (see Appendix B). This condition can be decomposed into the N condition

$$\nabla_{u_k} J(u_0^*, \dots, u_{N-1}^*)^T (u_k - u_k^*) \geq 0, \quad \text{for all } u_k \in E_k, k = 0, \dots, N-1. \quad (3.42)$$

We then obtain:

3.4. EXTENSIONS OF THE MINIMUM PRINCIPLE

We now consider some variations of the continuous-time optimal control problem and derive corresponding variations of the Minimum Principle.

3.4.1 Fixed Terminal State

Suppose that in addition to the initial state $x(0)$, the final state $x(T)$ is given. Then the preceding adjoint derivations still hold except that the terminal condition $\lambda^T(T, x) = h(x)$ is not true anymore. In effect, here we have

$$\lambda^T(T, x) = \begin{cases} 0 & \text{if } x = x(T), \\ \infty & \text{otherwise.} \end{cases}$$

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Thus $\lambda^T(T, x)$ cannot be differentiated with respect to x , and the terminal boundary condition $\lambda^T(T) = \nabla_x h(x(T))$ for the adjoint equation does not hold. However, as compensation, we have the extra condition

$$x(T) \text{ given,}$$

thus maintaining the balance between boundary conditions and unknowns. If only some of the terminal states are fixed, that is,

$$x_i(T) \text{ given, for all } i \in I,$$

where I is some index set, we have the partial boundary condition

$$p_j(T) = \frac{\partial h(x^*(T))}{\partial x_j}, \quad \text{for all } j \notin I,$$

for the adjoint equation.

Example 3-4-1

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Figure 3-4.1 Optimal solution of the problem of connecting the two points $(0, a)$ and (T, b) with a minimum length curve (cf. Example 3-4.2).

Example 3-4.2 (The Brachistochrone Problem)

In 1696 Johann Bernoulli challenged the mathematical world with the following problem:

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$$\dot{x} = f(x, u, t) \quad (n \text{ differential equations}), \quad (2.5.4)$$

$$\dot{\lambda} = -\left(\frac{\partial f}{\partial x}\right)^T \lambda - \left(\frac{\partial L}{\partial x}\right)^T \quad (n \text{ differential equations}), \quad (2.5.5)$$

$$\left(\frac{\partial H}{\partial u}\right)^T = \left(\frac{\partial f}{\partial u}\right)^T \lambda + \left(\frac{\partial L}{\partial u}\right)^T = 0 \quad (m \text{ algebraic equations}), \quad (2.5.6)$$

$$x_k(t_0) \text{ given or } \lambda_k(t_0) = 0 \quad k = 1, \dots, n \quad (n \text{ boundary conditions}), \quad (2.5.7)$$

$$\lambda^T(t_f) = \left(\frac{\partial \phi}{\partial x} + \nu^T \frac{\partial \psi}{\partial x}\right)_{t=t_f} \quad (n \text{ boundary conditions}), \quad (2.5.8)$$

$$\psi[x(t_f), t_f] = 0 \quad (q \text{ side conditions}). \quad (2.5.9)$$

The stationarity conditions (2.5.6) determine the m -vector $u(t)$. The $2n$ differential equations (2.5.4) and (2.5.5), with the $2n$ boundary conditions (2.5.7) and (2.5.8), form a two-point boundary-value problem with q parameters ν to be found in (2.5.8) so that the q side conditions (2.5.9) are satisfied.



感谢各位聆听 !
Thanks for Listening

