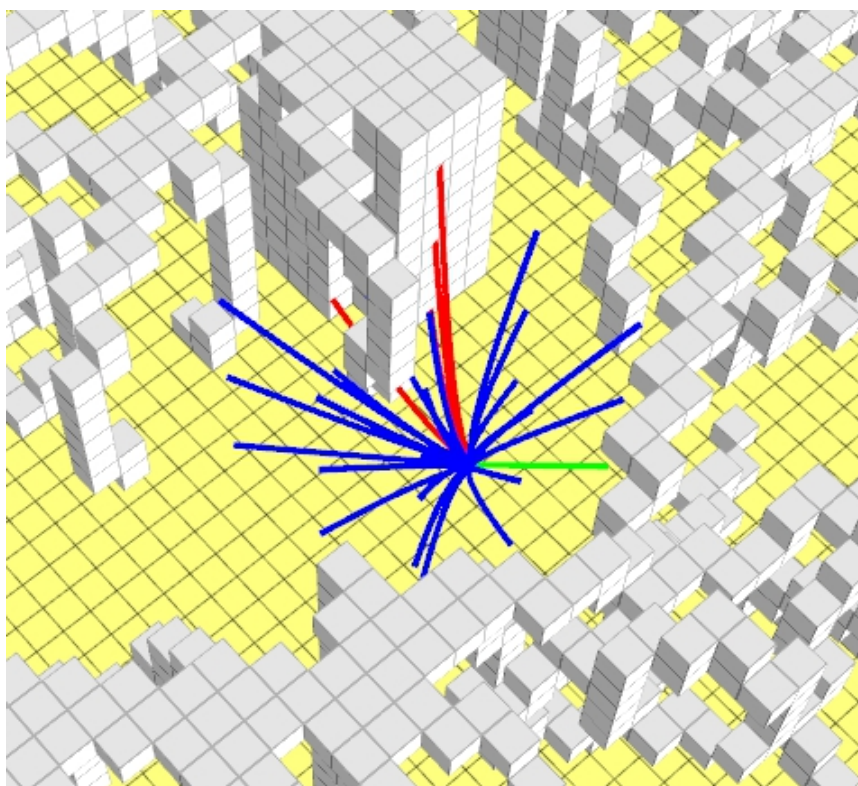


## HomeWork 4

### 一、成果展示



## 二、公式推导

$$J = \int_0^T (1 + a_x^2 + a_y^2 + a_z^2) dt$$

$$J = T + \left( \frac{1}{3} \alpha_1^2 T^3 + \alpha_1 \beta_1 T^2 + \beta_1^2 T \right) + \left( \frac{1}{3} \alpha_2^2 T^3 + \alpha_2 \beta_2 T^2 + \beta_2^2 T \right) + \left( \frac{1}{3} \alpha_3^2 T^3 + \alpha_3 \beta_3 T^2 + \beta_3^2 T \right)$$

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

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

由上式可推出：

$$\begin{aligned} \frac{12 \Delta p_x}{-T^3} + \frac{6 v_{x0}}{T^2} &= a_1 \\ \frac{12 \Delta p_y}{-T^3} + \frac{6 v_{y0}}{T^2} &= a_2 \\ \frac{12 \Delta p_z}{-T^3} + \frac{6 v_{z0}}{T^2} &= a_3 \\ \frac{6 \Delta p_x}{T^2} - \frac{8}{T} v_{x0} &= \beta_1 \\ \frac{6 \Delta p_y}{T^2} - \frac{8}{T} v_{y0} &= \beta_2 \\ \frac{6 \Delta p_z}{T^2} - \frac{8}{T} v_{z0} &= \beta_3 \end{aligned}$$

$$\begin{aligned} \Delta p_x &= p_{xf} - p_{x0} \\ \Delta p_y &= p_{yf} - p_{y0} \\ \Delta p_z &= p_{zf} - p_{z0} \end{aligned}$$

$\alpha$ 和 $\beta$ 是  $T$  的函数，将其带入  $J$  中，得出  $J$  是只与  $T$  有关的函数。为了令  $J$  最小，对  $T$

求导，让求导表达式=0，得到如下多项式

$$T^4 - 4 \left( \underbrace{V_{x_0}^2 + V_{x_0} V_{x_f} + V_{x_f}^2}_{\substack{\downarrow \\ V_{x_0}^2 + V_{x_f}^2}} + \underbrace{V_{y_0}^2 + V_{y_0} V_{y_f} + V_{y_f}^2}_{\substack{\downarrow \\ V_{y_0}^2 + V_{y_f}^2}} + \underbrace{V_{z_0}^2 + V_{z_0} V_{z_f} + V_{z_f}^2}_{\substack{\downarrow \\ V_{z_0}^2 + V_{z_f}^2}} \right) T^2 + 24 \left( (P_{x_f} - P_{x_0})(V_{x_f} - V_{x_0}) + (P_{y_f} - P_{y_0})(V_{y_f} - V_{y_0}) + (P_{z_f} - P_{z_0})(V_{z_f} - V_{z_0}) \right) T - 36 \left( (P_{x_f} - P_{x_0})^2 + (P_{y_f} - P_{y_0})^2 + (P_{z_f} - P_{z_0})^2 \right).$$

多项式求解采用伴随矩阵求特征值法：

In linear algebra, the **Frobenius companion matrix** of the **monic polynomial**

$$p(t) = c_0 + c_1 t + \cdots + c_{n-1} t^{n-1} + t^n,$$

is the **square matrix** defined as

$$C(p) = \begin{bmatrix} 0 & 0 & \cdots & 0 & -c_0 \\ 1 & 0 & \cdots & 0 & -c_1 \\ 0 & 1 & \cdots & 0 & -c_2 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \cdots & 1 & -c_{n-1} \end{bmatrix}.$$

对矩阵求解特征值，然后将结果带入 J 中。

## 三、代码分析

### 1. forward integration 前向积分

```
*/
//pos vel delta

pos(0) = pos(0)+vel(0)*delta_time+0.5*acc_input(0)*delta_time*delta_time; //  $x=x_0+vt+0.5at^2$ 
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; //  $v=v_0+at$ 
vel(1) = vel(1)+acc_input(1)*delta_time;
vel(2) = vel(2)+acc_input(2)*delta_time;
```

```
Position.push_back(pos);
Velocity.push_back(vel);
```

前向积分公式为  $V=V_0+at$  和  $X=X_0+V_0t+0.5at^2$

### 2. OBVP

```
*/
//初始的位置和速度
double p_x0=_start_position(0);
double p_y0=_start_position(1);
double p_z0=_start_position(2);
double v_x0=_start_velocity(0);
double v_y0=_start_velocity(1);
double v_z0=_start_velocity(2);
//终点的位置和速度
double p_xf=_target_position(0);
double p_yf=_target_position(1);
double p_zf=_target_position(2);
double v_xf=0 ,v_yf=0 ,v_zf=0;
```

设置初始位置和速度，终点位置和速度，终点速度指定为 0，加速度为自由量。

```
Eigen::Matrix<double,4,4> m;

double c0 = -36*((p_xf-p_x0)*(p_xf-p_x0) + (p_yf-p_y0)*(p_yf-p_y0) + (p_zf-p_z0)*(p_zf-p_z0));
double c1 = 24*((p_xf-p_x0)*(v_xf+v_x0) + (p_yf-p_y0)*(v_yf+v_y0) + (p_zf-p_z0)*(v_zf+v_z0));
double c2 = -4*(v_x0*v_x0 + v_x0*v_xf + v_xf*v_xf + v_y0*v_y0 + v_y0*v_yf + v_yf*v_yf + v_z0*v_z0 + v_z0*v_zf + v_zf*v_zf);
double c3 = 0.0;

m << 0,0,0,-c0,
      1,0,0,-c1,
      0,1,0,-c2,
      0,0,1,c3;

double J;

Eigen::Matrix<complex<double>,Eigen::Dynamic,Eigen::Dynamic> eigenValules;
eigenValules = m.eigenvalues(); //返回特征值
```

多项式求根采用伴随矩阵求特征值方法，损失函数  $J$  是关于  $T$  的函数，所以求  $J$  关于  $T$

的导数，令其为 0，得到  $T$ ，在带入  $J$  中得出  $J$  的最小值

```

for(int i=0; i<4; ++i){
    double T = std::real(eigenValules(i)); //返回实数部分
    double img = std::imag(eigenValules(i)); //返回虚数部分
    //对T进行筛选, 不要负数和虚数
    if(T < 0 || std::abs(img) >= 0){
        continue;
    }

    Eigen::Vector3d alpha,beta;

    alpha(0)=-12*(p_xf-p_x0)/T/T/T + 6*v_x0/T/T;
    alpha(1)=-12*(p_yf-p_y0)/T/T/T + 6*v_y0/T/T;
    alpha(2)=-12*(p_zf-p_z0)/T/T/T + 6*v_z0/T/T;

    beta(0)=6*(p_xf-p_x0)/T/T - 8*v_x0/T;
    beta(1)=6*(p_yf-p_y0)/T/T - 8*v_y0/T;
    beta(2)=6*(p_zf-p_z0)/T/T - 8*v_z0/T;

    J=T+(1/3*alpha(0)*alpha(0)*pow(T,3)+alpha(0)*beta(0)*T*T+beta(0)*beta(0)*T)
        +(1/3*alpha(1)*alpha(1)*pow(T,3)+alpha(1)*beta(1)*T*T+beta(1)*beta(1)*T)
        +(1/3*alpha(2)*alpha(2)*pow(T,3)+alpha(2)*beta(2)*T*T+beta(2)*beta(2)*T);

```

将 $\alpha$ 和 $\beta$ 带入求 J。