



轨迹优化-闭式求解

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1.多项式轨迹表示方法

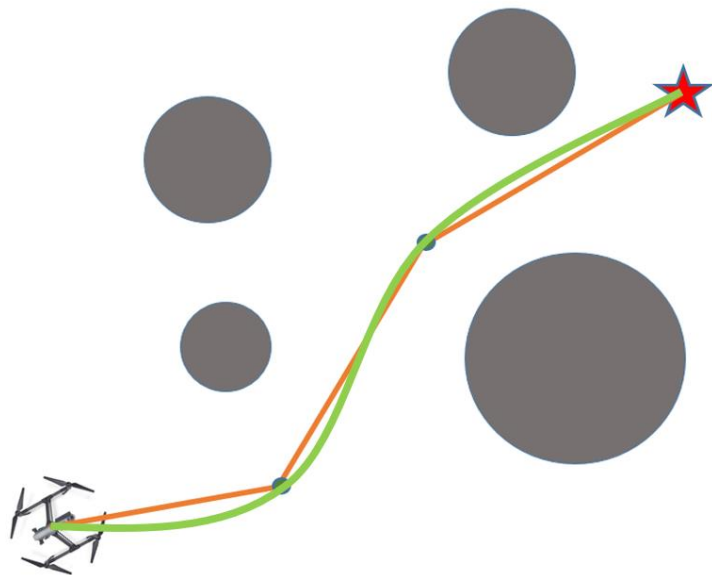
2.Minimum snap

3.闭式求解

4.软约束

5.贝塞尔曲线和硬约束

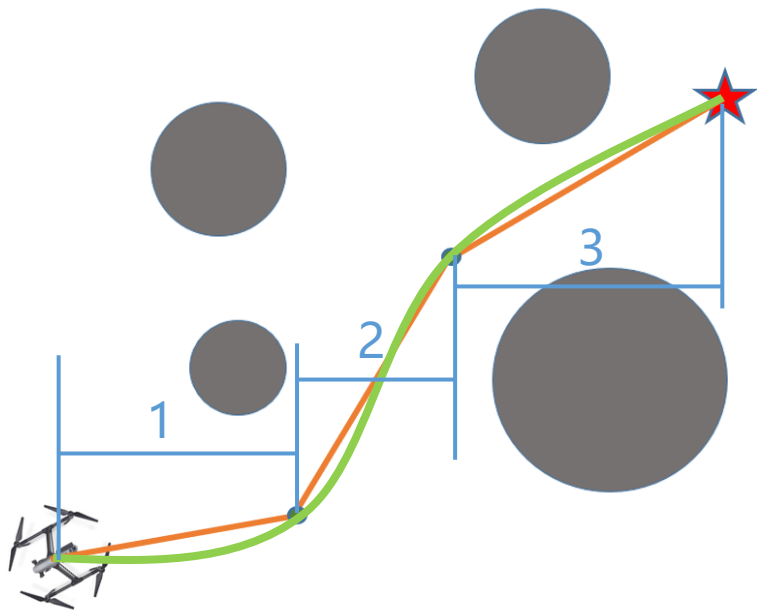
6.论文总结



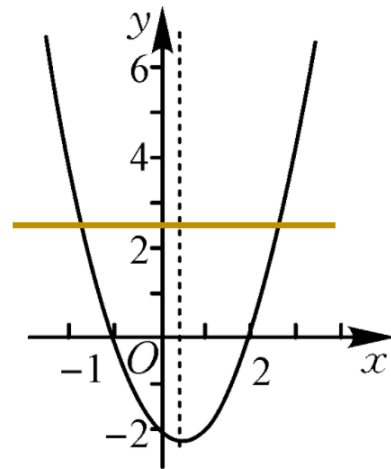


闭式求解

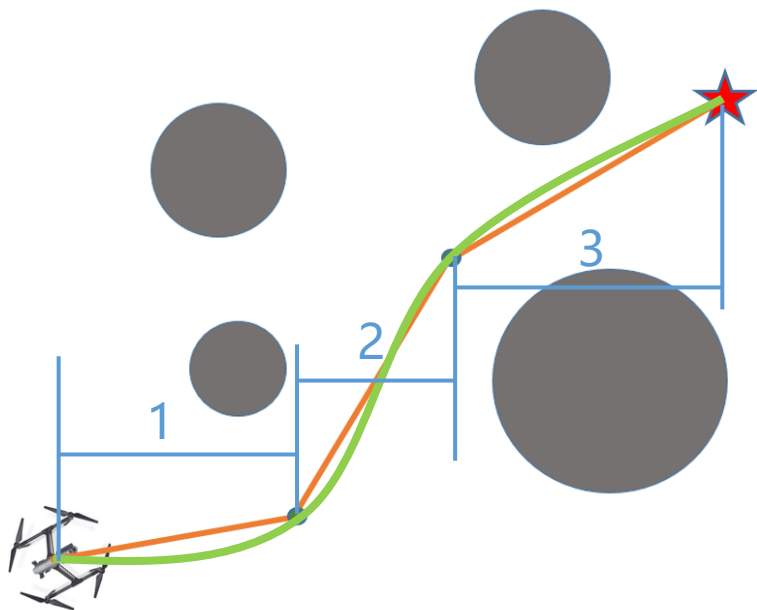
闭式求解



$$\min p^T Q p$$
$$s. t. A_{eq} p = b_{eq},$$



闭式求解



$$\min p^T Q p$$

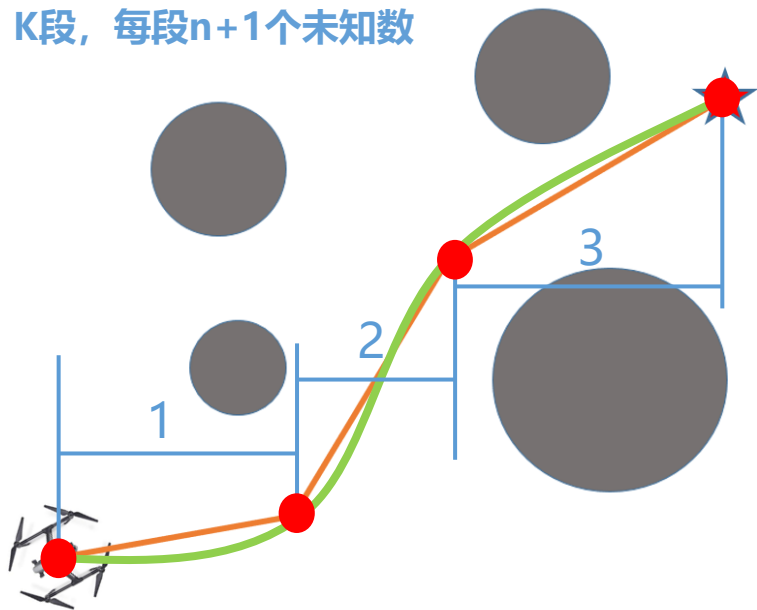
$$s. t. \quad A_{eq} p = b_{eq},$$



$$A_i p_i = d_i, \quad A_i = [A_0 \quad A_t]_i^T, \quad d_i = [d_0, d_T]_i$$

闭式求解

K段，每段n+1个未知数



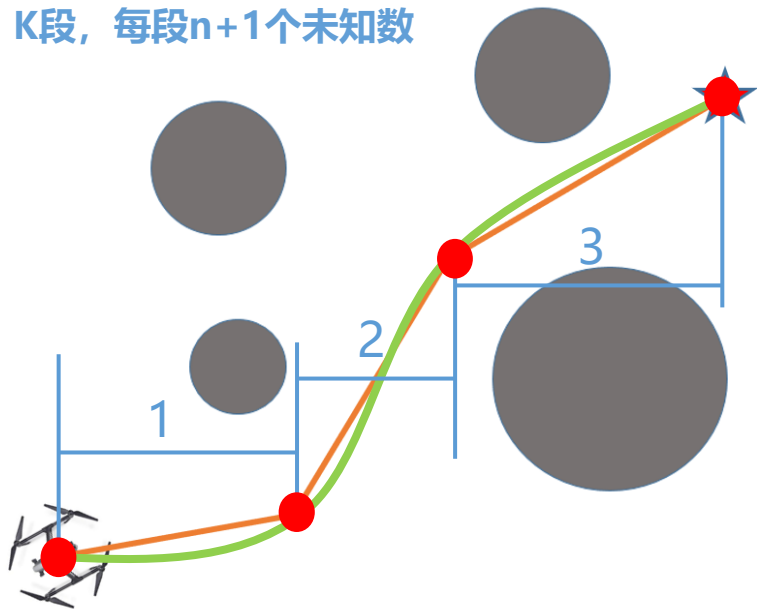
$$\underbrace{A_{total}}_{k(n+1) \times 6k} \begin{bmatrix} p_1 \\ \vdots \\ p_k \end{bmatrix} = \begin{bmatrix} d_1 \\ \vdots \\ d_k \end{bmatrix} = \begin{bmatrix} p_1(t_0) \\ v_1(t_0) \\ a_1(t_0) \\ p_1(t_1) \\ v_1(t_1) \\ a_1(t_1) \\ \vdots \\ p_k(t_{k-1}) \\ v_k(t_{k-1}) \\ a_k(t_{k-1}) \\ p_k(t_k) \\ v_k(t_k) \\ a_k(t_k) \end{bmatrix}$$

$6k \times 1$

$$p = A^{-1} \boxed{d}$$

连续性约束

K段，每段n+1个未知数



$$\underbrace{A_{total}}_{k(n+1) \times 6k} \begin{bmatrix} p_1 \\ \vdots \\ p_k \end{bmatrix} = \begin{bmatrix} d_1 \\ \vdots \\ d_k \end{bmatrix} = \underbrace{\begin{bmatrix} p_1(t_0) \\ v_1(t_0) \\ a_1(t_0) \\ p_1(t_1) \\ v_1(t_1) \\ a_1(t_1) \\ \vdots \\ p_k(t_{k-1}) \\ v_k(t_{k-1}) \\ a_k(t_{k-1}) \\ p_k(t_k) \\ v_k(t_k) \\ a_k(t_k) \end{bmatrix}}_{6k \times 1}$$

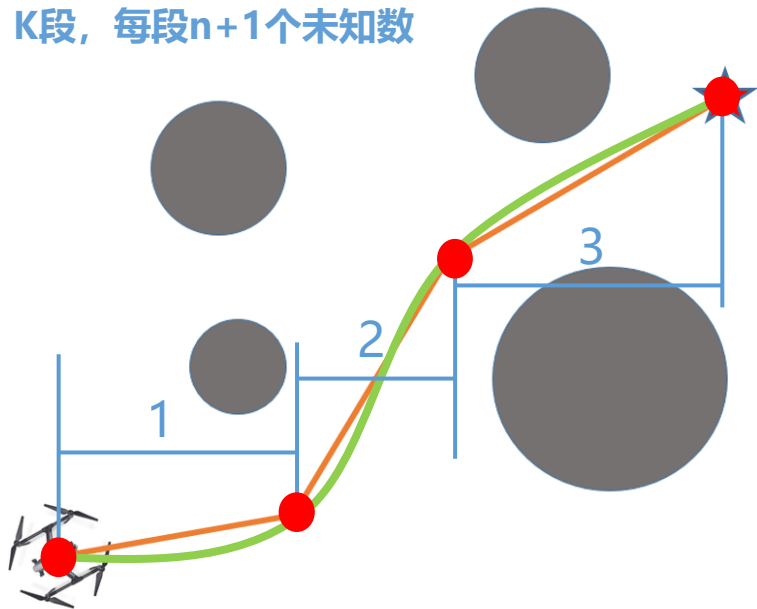
$$p_i(t_i) = p_{i+1}(t_i), \quad v_i(t_i) = v_{i+1}(t_i), \quad a_i(t_i) = a_{i+1}(t_i)$$

消除重复变量

$$\begin{bmatrix} a \\ a \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} a$$

连续性约束

K段，每段n+1个未知数



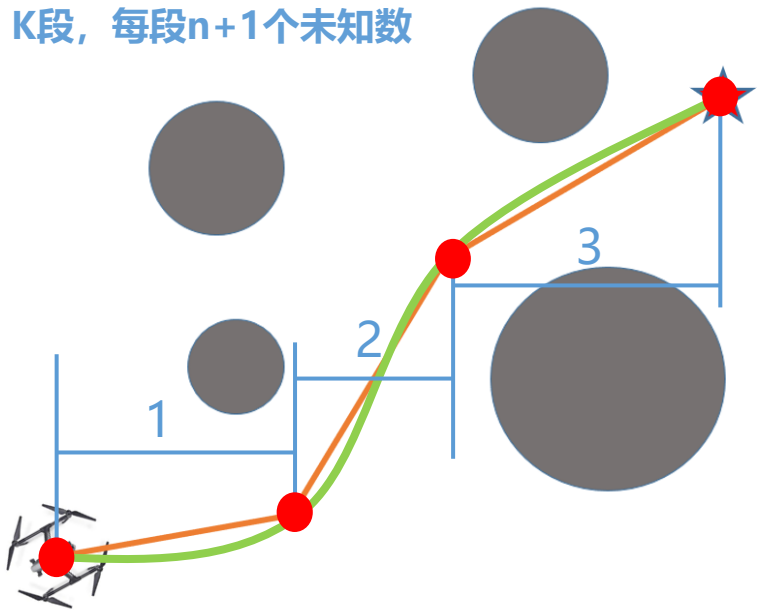
消除重复变量

$$\underbrace{\begin{bmatrix} d_1 \\ \vdots \\ d_k \end{bmatrix}}_{6k \times 1} = \underbrace{\begin{bmatrix} 1 & & & & & & & & \\ & 1 & & & & & & & \\ & & 1 & & & & & & \\ & & & 1 & & & & & \\ & & & & 1 & & & & \\ & & & & & 1 & & & \\ & & & & & & 1 & & \\ & & & & & & & 1 & \\ & & & & & & & & 1 \\ & & & & & & & & & \ddots \end{bmatrix}}_M \underbrace{\begin{bmatrix} p(t_0) \\ v(t_0) \\ a(t_0) \\ p(t_1) \\ v(t_1) \\ a(t_1) \\ p(t_2) \\ v(t_2) \\ a(t_2) \\ \vdots \\ p(t_k) \\ v(t_k) \\ a(t_k) \end{bmatrix}}_{3(k+1) \times 1}$$

$$d = Md'$$

元素位置置换

K段，每段n+1个未知数



$$d' = C \begin{bmatrix} d_F \\ d_P \end{bmatrix}$$

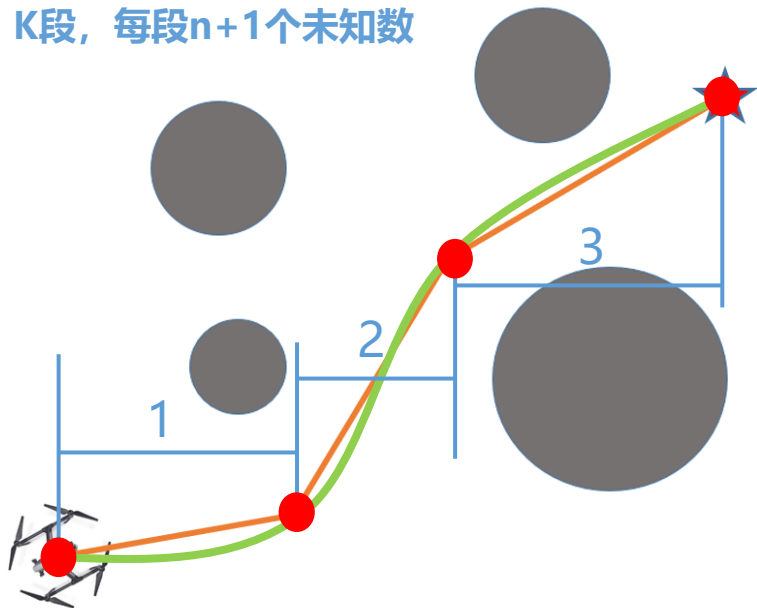
例子:

$$\begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} = \underbrace{\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}}_C \begin{bmatrix} a \\ c \\ d \\ b \end{bmatrix}$$

无约束优化问题



K段，每段n+1个未知数



$$d = MC \begin{bmatrix} d_F \\ d_P \end{bmatrix}$$

$$p = A^{-1}d = \underbrace{A^{-1}MC}_K \begin{bmatrix} d_F \\ d_P \end{bmatrix} = K \begin{bmatrix} d_F \\ d_P \end{bmatrix}$$

$$\min J = p^T Q p$$

$$J = \begin{bmatrix} d_F \\ d_P \end{bmatrix}^T \underbrace{K^T Q K}_R \begin{bmatrix} d_F \\ d_P \end{bmatrix}$$

$$= \begin{bmatrix} d_F \\ d_P \end{bmatrix}^T \begin{bmatrix} R_{FF} & R_{FP} \\ R_{PF} & R_{PP} \end{bmatrix} \begin{bmatrix} d_F \\ d_P \end{bmatrix}$$

$$= d_F^T R_{FF} d_F + d_F^T R_{FP} d_P + d_P^T R_{PF} d_F + d_P^T R_{PP} d_P$$

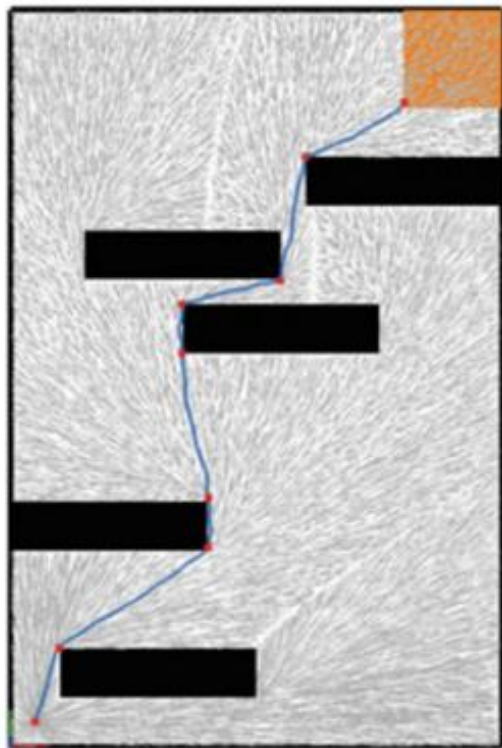
$$Q \text{ 对称} \Rightarrow R \text{ 对称} \Rightarrow = d_F^T R_{FF} d_F + 2d_F^T R_{FP} d_P + d_P^T R_{PP} d_P$$

$$\Rightarrow 2R_{FP}^T d_F + 2R_{PP} d_P = 0 \quad (\text{注意 } R_{PP}^T = R_{PP})$$

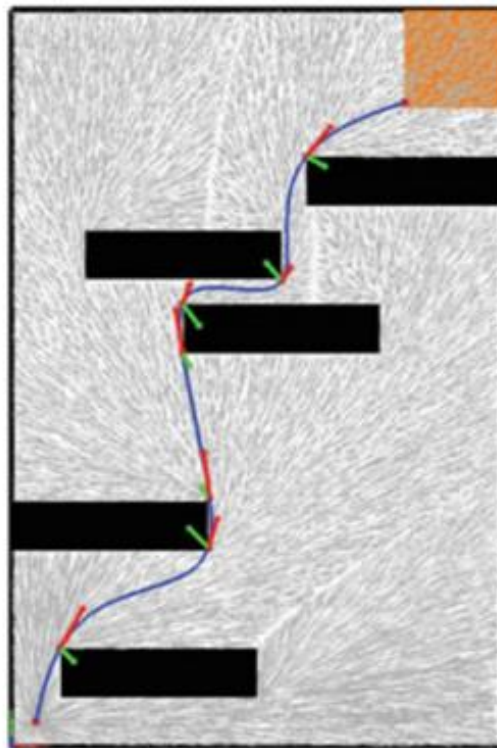
$$\Rightarrow d_P = -R_{PP}^{-1} R_{FP}^T d_F$$

$$p = K \begin{bmatrix} d_F \\ d_P \end{bmatrix}$$

闭式求解

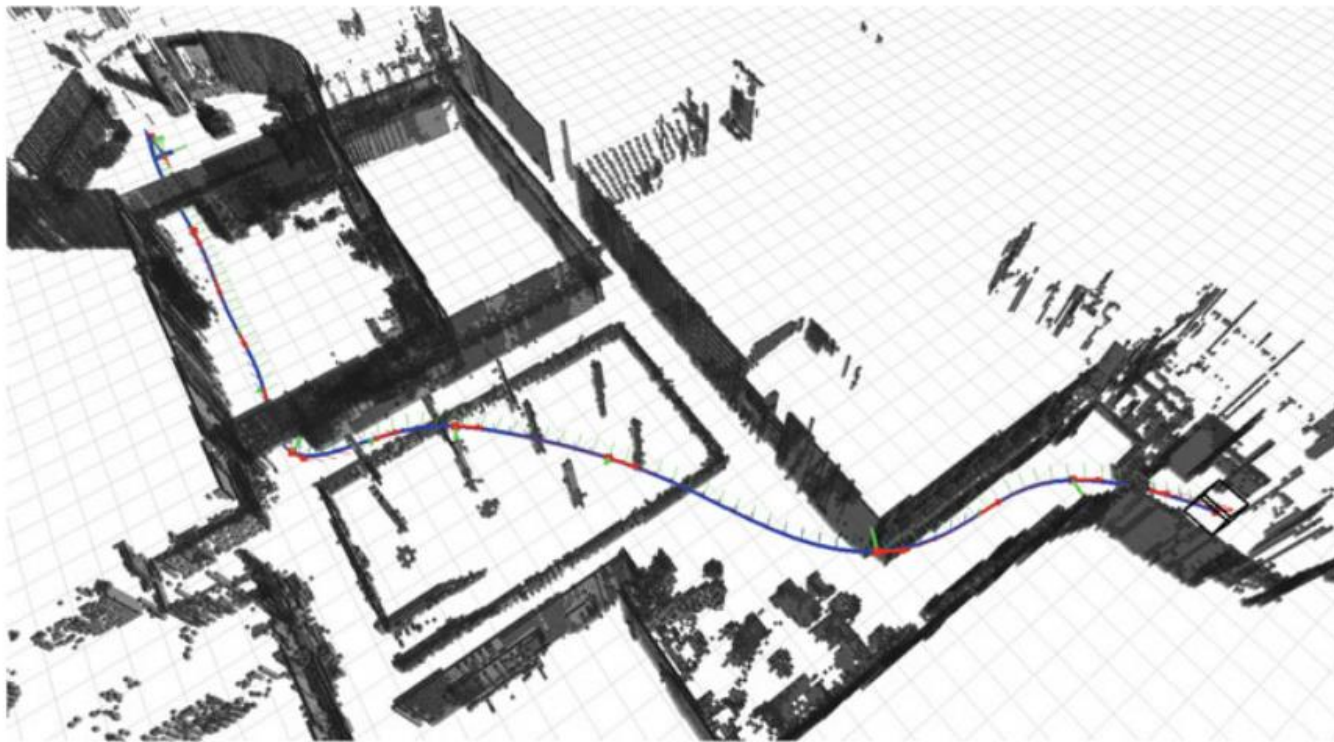


Pruned waypoints from straight-line RRT* become waypoints in 6c.



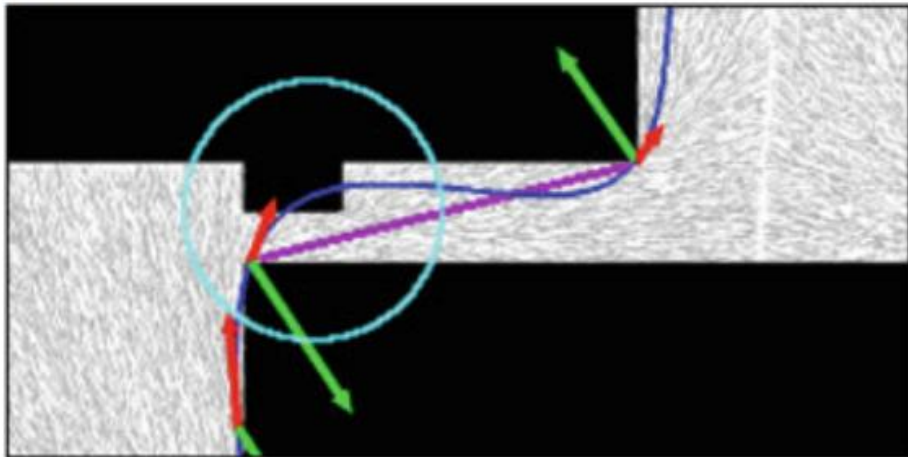
Solution by our algorithm after 3s running time, finds much lower cost than 6a.

闭式求解

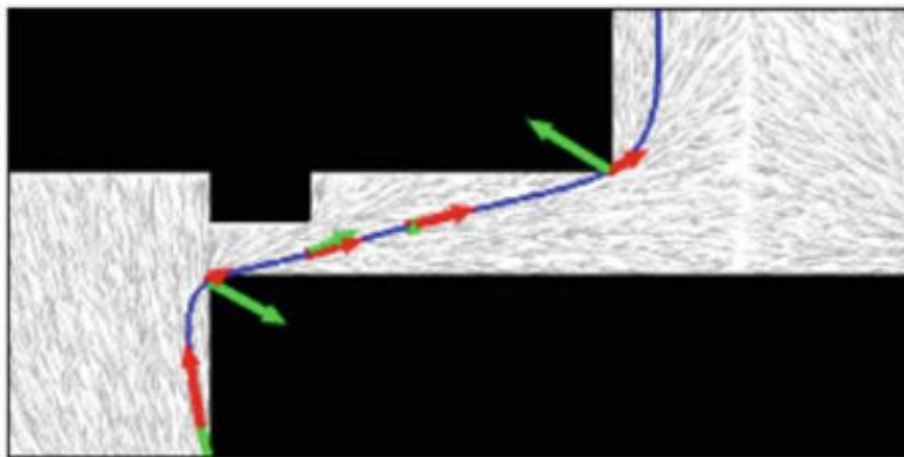


闭式求解

(a)



(b)



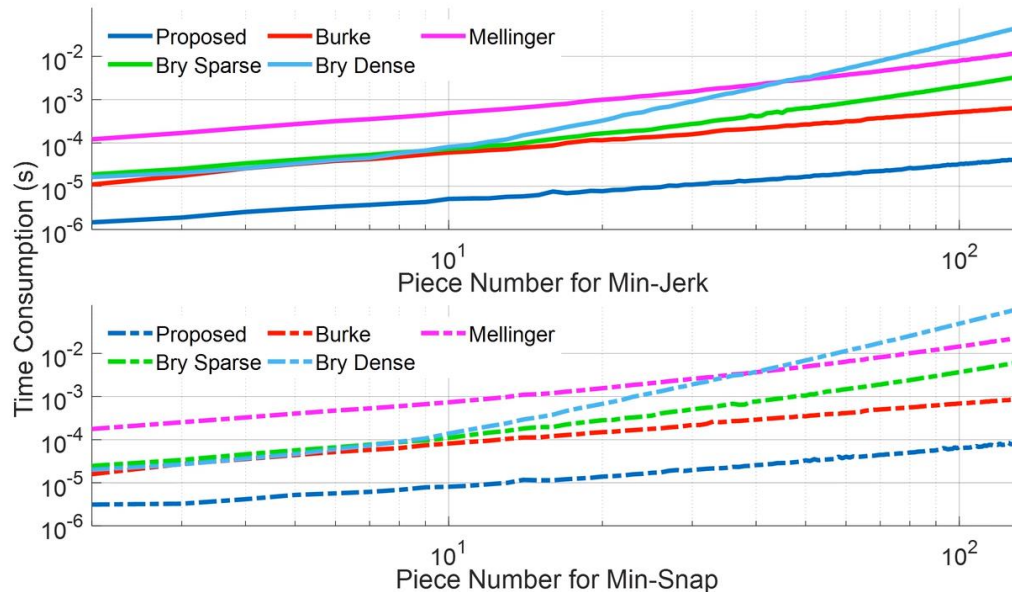


闭式求解代码讲解

总结一下整个闭式法的步骤：

1. 先确定轨迹阶数（比如5阶），再确定 d 向量中的约束量（pva），进而根据各段的时间分配求得 A_{total} 。
2. 根据连续性约束构造映射矩阵 M ，并确定 d 向量中哪些量是Fix(比如起点终点pva，中间点的p等)，哪些量是Free，进而构造置换矩阵 C ，并求得 $K = A^{-1}MC$ 。
3. 计算QP目标函数中的 Q ($\min Jerk/Snap$) 并计算 $R = K^T Q K$ ，根据fix变量的长度将 R 拆分成 $R_{FF}, R_{FP}, R_{PF}, R_{PP}$ 四块。
4. 填入已知变量得到 d_F ，并根据 $d_p = -R_{PP}^{-1}R_{FP}^T d_F$ 计算得到 d_P 。
5. 根据公式 $p = K \begin{bmatrix} d_F \\ d_P \end{bmatrix}$ 计算得到轨迹参数 p 。

Speeding Up Traditional Min-Jerk/Snap Trajectory Generation



D. Mellinger and V. Kumar, "Minimum snap trajectory generation and control for quadrotors," ICRA 2011.

A. Bry, C. Richter, A. Bachrach, and N. Roy, "Aggressive flight of fixed-wing and quadrotor aircraft in dense indoor environments," IJRR 2015.

D. Burke, A. Chapman, and I. Shames, "Generating minimum-snap quadrotor trajectories really fast," ArXiv, vol. abs/2008.00595, IROS 2020.

THANKS

线性代数-求导

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$$\frac{dx^T}{dx} = I \quad \frac{dx}{dx^T} = I \quad \frac{dx^T A}{dx} = A$$

A 为 $n \times n$ 的矩阵, x 为 $n \times 1$ 的列向量

$$\frac{dAx}{dx^T} = A \quad \frac{dAx}{dx} = A^T \quad \frac{dxA}{dx} = A^T$$

$$\frac{\partial u}{\partial x^T} = \left(\frac{\partial u^T}{\partial x} \right)^T \quad \frac{\partial u^T v}{\partial x} = \frac{\partial u^T}{\partial x} v + \frac{\partial v^T}{\partial x} u^T \quad \frac{\partial uv^T}{\partial x} = \frac{\partial u}{\partial x} v^T + u \frac{\partial v^T}{\partial x}$$

$$\frac{dx^T x}{dx} = 2x \quad \frac{dx^T Ax}{dx} = (A + A^T)x \quad \frac{dX^T AX}{dX} = 2AX \quad (\text{如果} A \text{为对称阵})$$

$$\frac{\partial AB}{\partial x} = \frac{\partial A}{\partial x} B + A \frac{\partial B}{\partial x} \quad \frac{\partial u^T X v}{\partial X} = uv^T$$

$$\frac{\partial u^T X^T X u}{\partial X} = 2Xuu^T \quad \frac{\partial [(Xu-v)^T (Xu-v)]}{\partial X} = 2(Xu-v)u^T$$