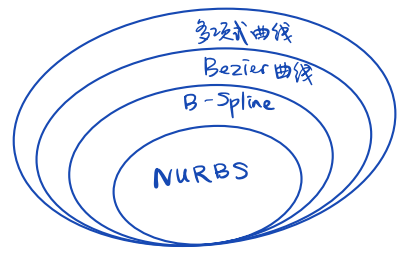


# 样条曲线的数学表达与样条拟合



1.  $C(u) = [a_0 \ a_1 \ \dots \ a_n] \begin{bmatrix} 1 \\ u \\ \vdots \\ u^n \end{bmatrix} = [a_i]^T [u^i]$  多项式  
对于交互式设计是不自然的

2. Bezier 曲线:

$C(u) = (1-u)P_0 + uP_1$  两个点 即为直线

$C(u) = (1-u)^2 P_0 + 2u(1-u)P_1 + u^2 P_2$  三个点 抛物线

$C(u) = (1-u)^3 P_0 + 3u(1-u)^2 P_1 + 3u^2(1-u)P_2 + u^3 P_3$  四个点

依此类推

Bezier 曲线具有凸包性, 即曲线包含在其控制点的凸包中

$$\begin{cases} \frac{2!}{0!2!} (1-u)^2 = (1-u)^2 \\ \frac{2!}{1!1!} u(1-u) = 2u(1-u) \\ \frac{2!}{2!0!} u^2 = u^2 \end{cases}$$

$$\begin{cases} \frac{3!}{0!3!} u^0 (1-u)^3 \\ \frac{3!}{1!2!} u (1-u)^2 \\ \frac{3!}{2!1!} u^2 (1-u) \\ \frac{3!}{3!0!} u^3 \end{cases}$$

通式

$$\frac{n!}{i!(n-i)!} u^i (1-u)^{n-i}$$

3. B-Spline (B样条曲线)

$N_{i,p}(u)$   $p$ 次 B 样条基函数  $C(u) = \sum_{i=0}^n N_{i,p}(u) P_i$  控制点

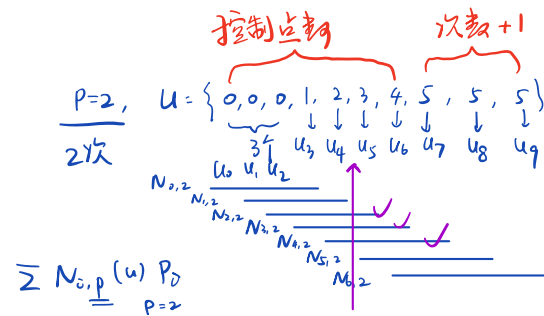
$N_{i,0}(u) = \begin{cases} 1 & u_i \leq u < u_{i+1} \\ 0 & \text{其他} \end{cases}$

$N_{i,p}(u) = \frac{u-u_i}{u_{i+p}-u_i} N_{i,p-1}(u) + \frac{u_{i+p+1}-u}{u_{i+p+1}-u_{i+1}} N_{i+1,p-1}(u)$  控制点

节点向量  $U = \{ \underbrace{a, \dots, a}_{p+1 \uparrow}, \underbrace{u_{p+1}, \dots, u_{m-p-1}}_{p+1 \uparrow}, \underbrace{b, \dots, b}_{p+1 \uparrow} \}$  控制点

例  $p=2, U = \{0, 0, 0, 1, 2, 3, 4, 5, 5, 5\}$

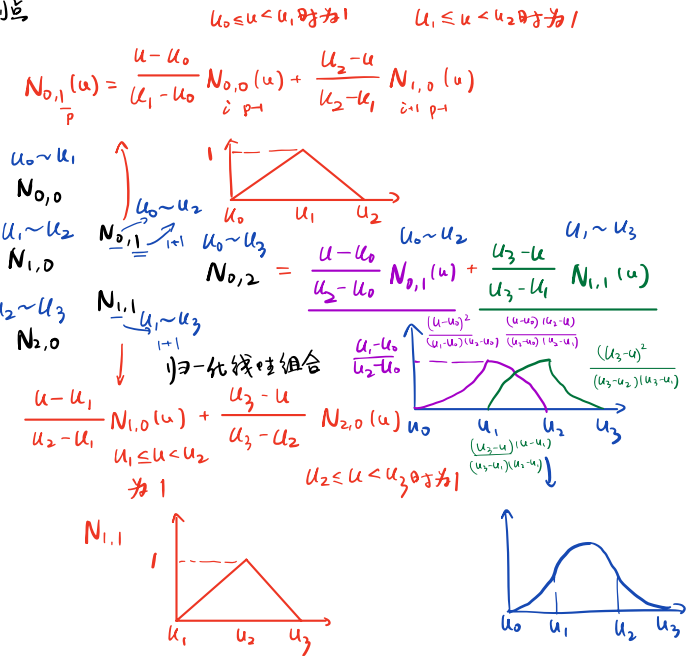
求  $C(2.5) = ?$



$C(2.5) = N_{2,2}(\frac{5}{2}) P_2 + N_{3,2}(\frac{5}{2}) P_3 + N_{4,2}(\frac{5}{2}) P_4$

4. NURBS:  $C(u) = \frac{\sum_{i=0}^n N_{i,p}(u) w_i P_i}{\sum_{i=0}^n N_{i,p}(u) w_i}$

(Non Uniform Rational B-Spline)



例: 写出基函数  $N_{0,3}$  ?

$N_{0,0}$   $N_{0,1}$   $N_{0,2}$   $N_{0,3}$   $N_{1,0}$   $N_{1,1}$   $N_{1,2}$   $N_{1,3}$   $N_{2,0}$   $N_{2,1}$   $N_{2,2}$   $N_{2,3}$   $N_{3,0}$   $N_{3,1}$   $N_{3,2}$   $N_{3,3}$   $N_{4,0}$   $N_{4,1}$   $N_{4,2}$   $N_{4,3}$

# 5. 分段三次样条

PCS

目标: 在给定拟合点时, 确定方程组并求解

$$f_i(u) = D_i + C_i(u - u_i) + B_i(u - u_i)^2 + A_i(u - u_i)^3 \quad u \in [u_i, u_{i+1}]$$

已知: 拟合点:  $(u_i, t_i)$

则有 ①  $f_i(u_i) = t_i = D_i$

②  $f_i''(u_i) = 2B_i$  且  $f_i''(u_i) = m_i$  则  $B_i = \frac{1}{2}m_i$

③  $f_i(u_{i+1}) = D_i + C_i(u_{i+1} - u_i) + B_i(u_{i+1} - u_i)^2 + A_i(u_{i+1} - u_i)^3$   
 $\underbrace{t_{i+1}}_{t_{i+1} - t_i} \quad \underbrace{t_i}_{\text{记为 } h_i} \quad \underbrace{h_i^2} \quad \underbrace{h_i^3}$

则  $\frac{t_{i+1} - t_i}{h_i} = C_i + B_i h_i + A_i h_i^2$   
 $\underbrace{\quad}_{\text{记为 } d_i}$

则  $C_i = d_i - \frac{h_i(m_{i+1} - m_i) + 3m_i h_i}{6} = d_i - \frac{h_i(m_{i+1} + 2m_i)}{6}$

④ 又有  $f_i''(u_{i+1}) = 2B_i + 6A_i(u_{i+1} - u_i)$   
 $\underbrace{m_{i+1}}_{m_i} \quad \underbrace{h_i}$   
 $\Rightarrow A_i = \frac{m_{i+1} - m_i}{6h_i}$

由  $f_i'(u_i) = C_i \xrightarrow{\text{连续}} f_{i-1}'(u_i) = C_{i-1} + 2B_{i-1}(u_i - u_{i-1}) + 3A_i(u_i - u_{i-1})^2$

$\Rightarrow d_i - \frac{h_i(m_{i+1} + 2m_i)}{6} = d_{i-1} - \frac{h_{i-1}(m_i + 2m_{i-1})}{6} + m_{i-1}h_{i-1} + \frac{1}{2}h_{i-1}(m_i - m_{i-1})$

$\Rightarrow d_i - d_{i-1} = \frac{1}{6}h_i m_{i+1} + \frac{1}{6}[2m_i h_i + 2m_{i-1} h_i] + \frac{1}{6}h_{i-1} m_{i-1}$

$\Rightarrow 6(d_i - d_{i-1}) = h_i m_{i+1} + 2h_i(m_i + m_{i-1}) + h_{i-1} m_{i-1} \quad i=1, 2, \dots, N-1$   $N-1$  个方程,  $N+1$  个未知数

增加两个边界条件:  $\begin{cases} \text{自然样条} & m_0 = m_N = 0 \\ \text{或} \\ \text{夹持样条} \end{cases}$

$f'(u_0) = C_0 = d_0 - \frac{h_0(m_1 + 2m_0)}{6}$   
 $\downarrow$   
 指定值  $m_0 = \frac{6(d_0 - f'(u_0))}{2h_0} - \frac{m_1}{2} = \frac{3}{h_0}[d_0 - f'(u_0)] - \frac{m_1}{2}$

类似地, 指定终端条件  
可约束  $m_N$

可写为线性方程组  $AM = V$ , 其中对于夹持样条,  $A$  为三对角矩阵, 可用追赶法 (Thomas 算法) 求解.



## 8. 最小二乘曲线逼近 (Least Squares Curve Approximation, LSCA)

与上面的区别: 控制点少于样拟合点数

假设设有  $n$  个控制点,  $k$  个样拟合点, 且  $n < k$ ,

步骤:

(1) 指定  $P_0, P_n$ , 使  $P_0 = Q_0, P_n = Q_k$

(2) 选取  $\bar{u}_k \quad k=0, 1, \dots, k$  规则与上页所述相同 (一般用规则 2 或 3)

(3) 确定节点向量  $U_{p+i} = \frac{1}{p} \sum_{j=\alpha}^{\alpha+p-1} u_j$  注意  $\alpha$  取为  $\text{int}\left(\frac{i \cdot k}{n}\right)$  而不像之前那样直接取为  $i$

比如  $k=9 \quad n=5 \quad p=3$  则  $U = \left\{ \underbrace{0, 0, 0, 0, u_4}_{\text{点积}}, \underbrace{1, 1, 1, 1, 1}_{p+1} \right\}$

$$u_4 = \frac{1}{3} \sum_{j=2}^{2+3-1} \bar{u}_j = \frac{1}{3} (\bar{u}_2 + \bar{u}_3 + \bar{u}_4)$$

$$\alpha = \text{int}\left(1 \times \frac{9}{5}\right) = 2$$

(4) 构建平方误差:

$$f = \sum_{k=1}^{k-1} |Q_k - C(\bar{u}_k)|^2 = \sum_{k=1}^{k-1} \left| Q_k - \sum_{i=0}^n N_{i,p}(\bar{u}_k) P_i \right|^2 = \sum_{k=1}^{k-1} \left| R_k - \sum_{i=0}^{n-1} N_{i,p}(\bar{u}_k) P_i \right|^2$$

||  
 $Q_k = N_{0,p} Q_0 + N_{n,p}(\bar{u}_k) Q_k, (P_n)$

则精确匹配时应满足  $NP = R$

$$\text{其中 } N = \begin{bmatrix} N_{0,p}(\bar{u}_1) & N_{1,p}(\bar{u}_1) & \dots & N_{n-1,p}(\bar{u}_1) \\ \vdots & \vdots & & \vdots \\ N_{0,p}(\bar{u}_{k-1}) & N_{1,p}(\bar{u}_{k-1}) & \dots & N_{n-1,p}(\bar{u}_{k-1}) \end{bmatrix}_{(k-1) \times (n-1)}, \quad P = \begin{bmatrix} P_1 \\ P_2 \\ \vdots \\ P_{n-1} \end{bmatrix}, \quad R = \begin{bmatrix} R_1 \\ \vdots \\ R_{k-1} \end{bmatrix}$$

则相应的法方程 (最小二乘时满足的方程) 应为  $N^T N P = N^T R$   $\rightarrow$  由此获得最小二乘时  $P$

(5) 检验误差

拟合方法常需迭代进行, 主要有两种策略:

1. 一开始就拟合一整段, 如果不符合要求就退回 (缩小范围), 直至拟合成功, 再从末尾点往前走;

2. 一开始只拟合一小段, 若符合就扩大范围, 直至不符合要求, 则退一步并从末尾点往前走。

But, the best results are obtained with new files made with the right CAM process in mind. You can use fitting method with old files but the results may not be as optimal.

还有变控制点数的方法 (见 ppt)