
Deformation & Skinning

形变与蒙皮

北京大学 前沿计算研究中心

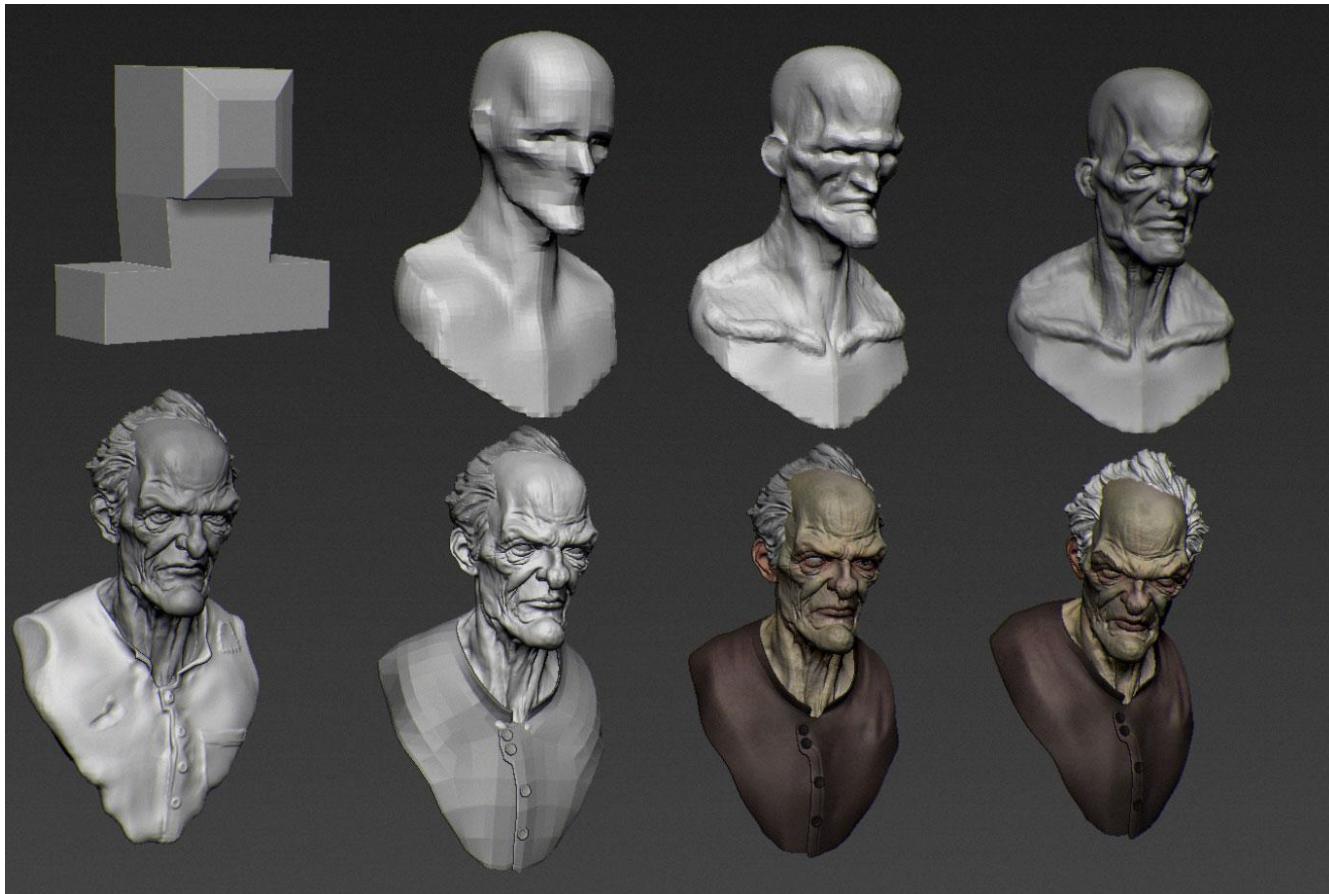
刘利斌

本节内容

- 几何形状的编辑
 - 自由形变 (Free-form Deformation)
 - 基于优化的形变
- 蒙皮形变 (Skinning Deformation)
 - 线性蒙皮
 - Linear-Blend Skinning (LBS)
 - Multi-linear Skinning
 - 非线性蒙皮
 - Dual-quaternion Skinning (DQS)
 - 基于样例的蒙皮
 - Blendshapes
 - 人脸动画

几何形变与编辑

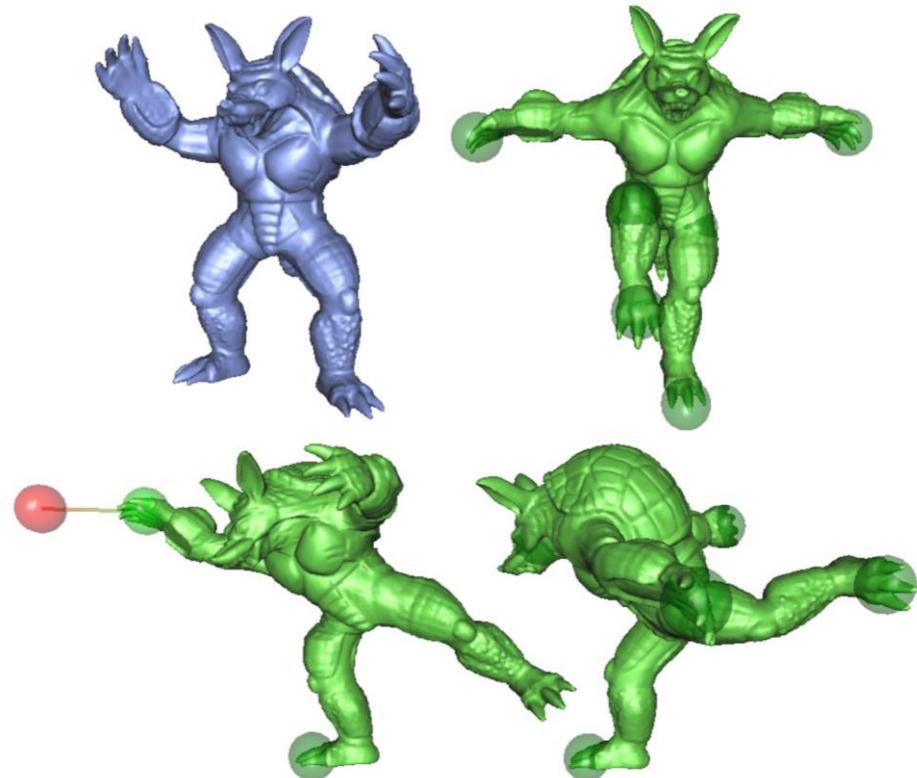
- 直接编辑几何模型顶点/控制点



<https://khanhha.github.io/portfolio/skeleton-to-mesh-conversion/>

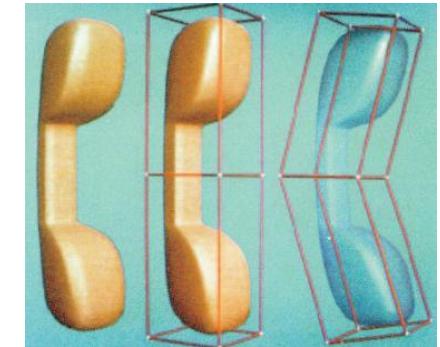
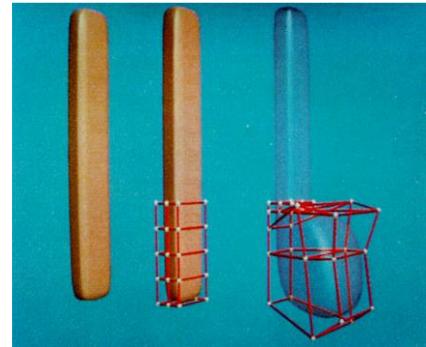
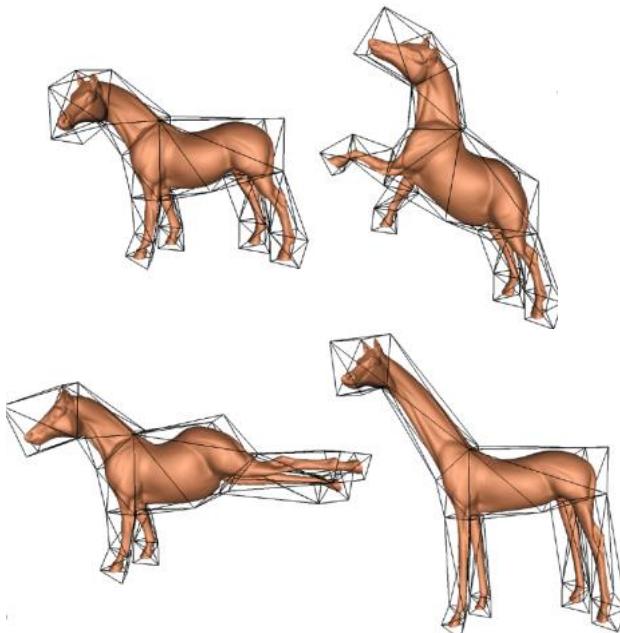
几何形变与编辑

- 编辑修改几何形状，同时保留几何模型细节
- 一般通过整体性的改变几何模型的参数来实现
 - 网格 -> 顶点
 - NURBS -> 控制点
 -



自由变形与Cage-based Deformation

- “扭曲”模型附近的空间来实现变形
 - 将几何模型嵌入到简单网格中
 - 改变简单网格的形状间接的改变模型的形状



Thomas W. Sederberg and Scott R. Parry. 1986. *Free-form deformation of solid geometric models*. In Proceedings of the 13th annual conference on Computer graphics and interactive techniques (SIGGRAPH '86).

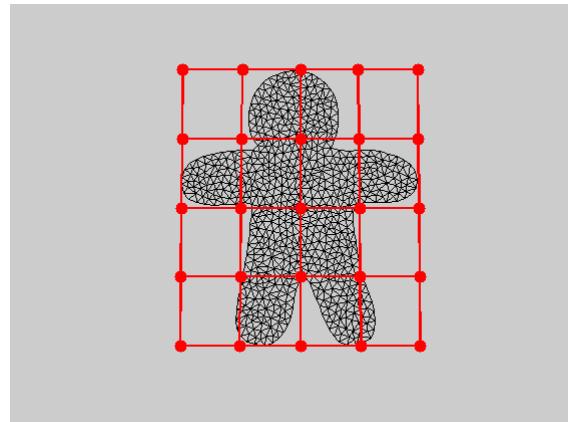
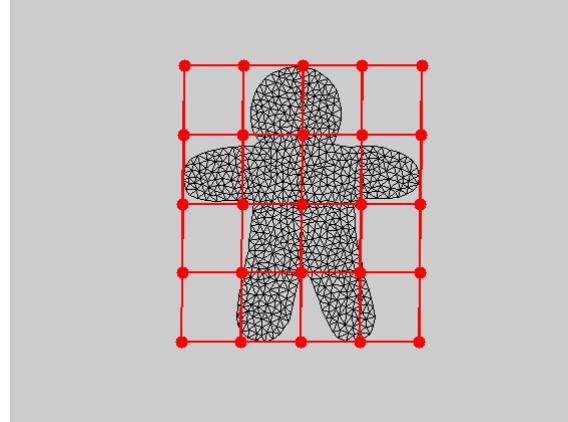
自由变形

- 如何定义嵌入关系?
 - 嵌入的精细网格 $\{x_i\}$
 - 简单网格 $\{p_j\}$

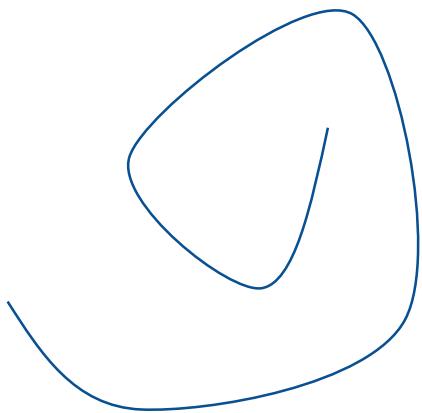
嵌入关系：

$$x_i = \sum_j \omega_{ij} p_j$$

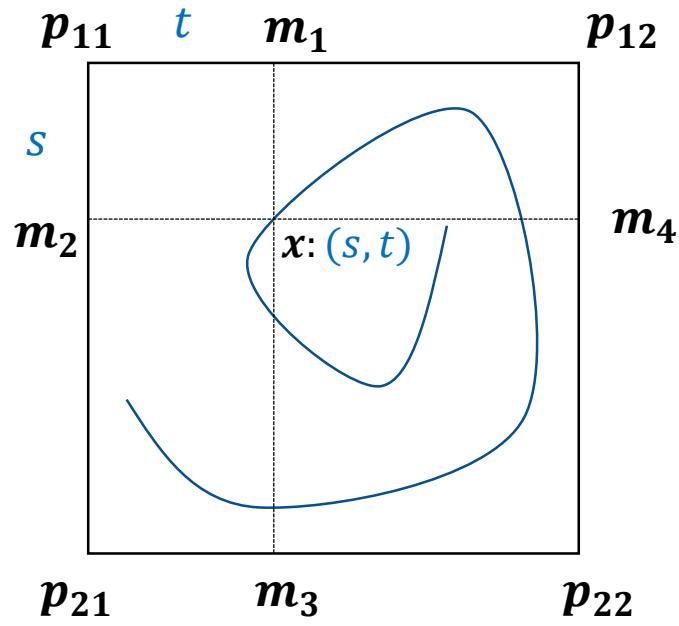
- 如何确定 ω_{ij} ??



线性插值

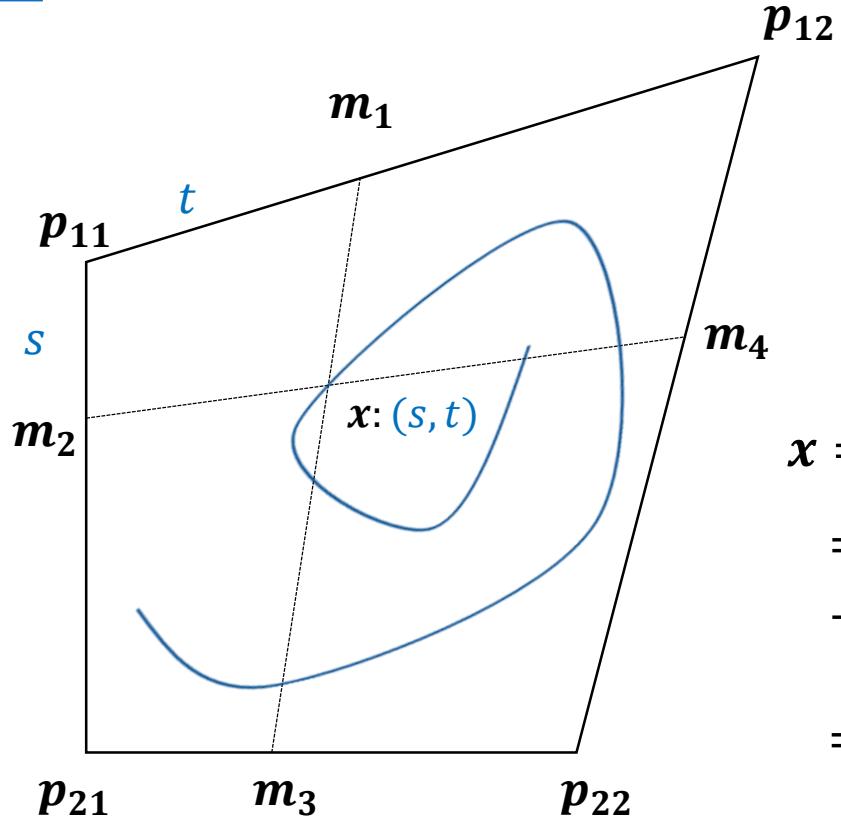


线性插值



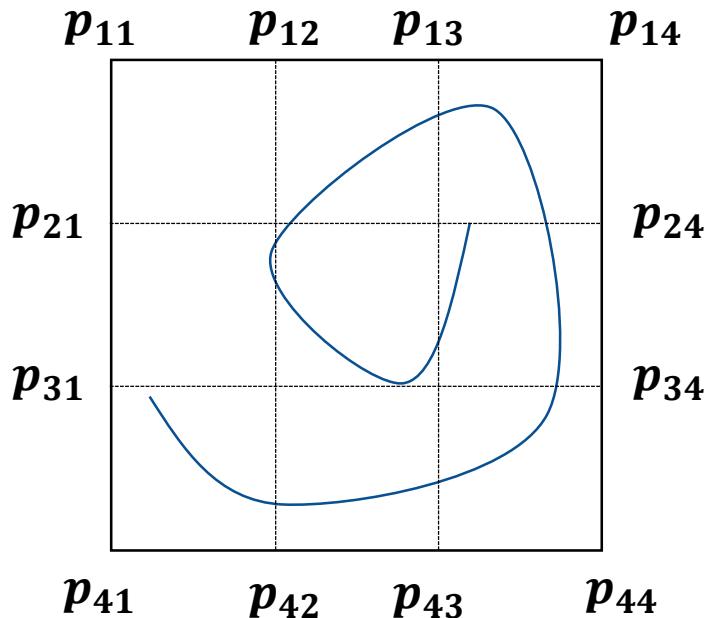
$$\begin{aligned}x &= (1 - t)m_2 + tm_4 \\&= (1 - t)((1 - s)p_{11} + sp_{21}) \\&\quad + t((1 - s)p_{12} + sp_{22}) \\&= \sum_i \sum_j B_i(s)B_j(t)\mathbf{p}_{ij}\end{aligned}$$

线性插值



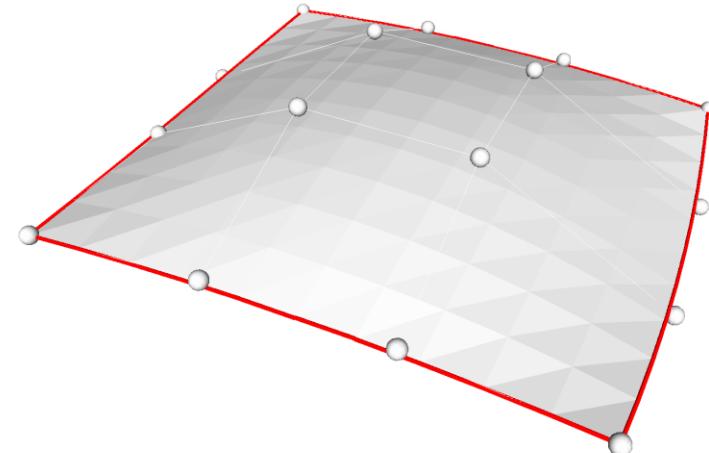
$$\begin{aligned}x &= (1 - t)m_2 + tm_4 \\&= (1 - t)((1 - s)p_{11} + sp_{21}) \\&\quad + t((1 - s)p_{12} + sp_{22}) \\&= \sum_i \sum_j B_i(s)B_j(t) \mathbf{p}_{ij}\end{aligned}$$

曲线插值



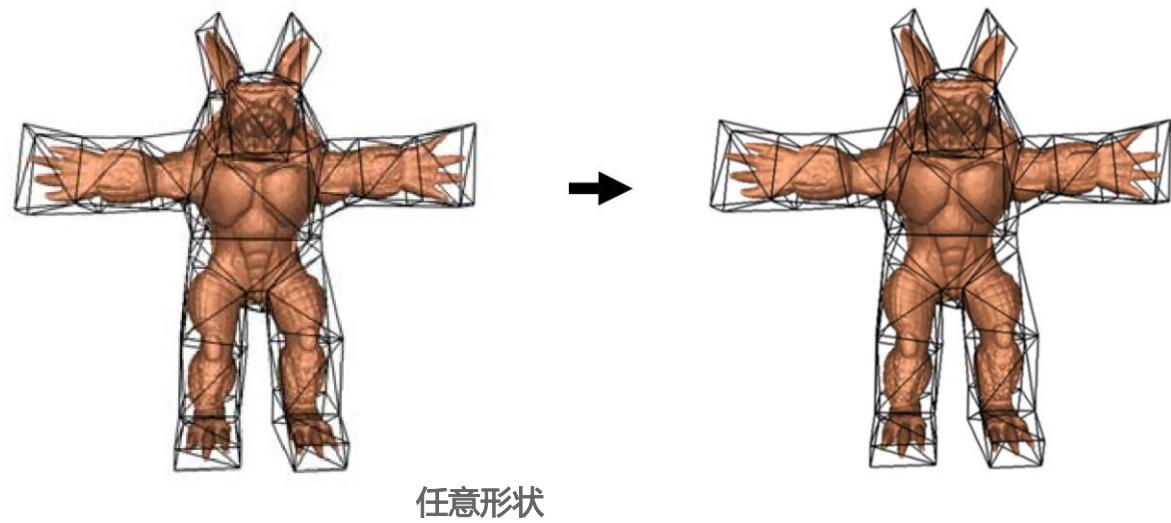
$$x = \sum_{i=0}^n \sum_{j=0}^m w_{i,j}(u, v) p_{i,j}$$

Bézier Surface, NURBS Surface...

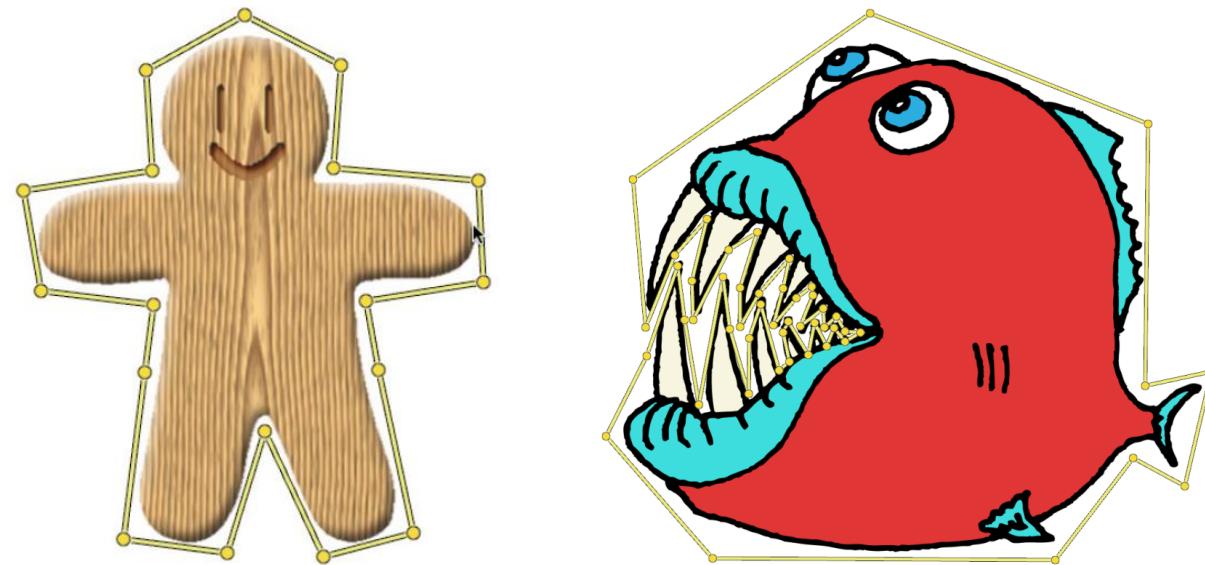


Cage-based Deformation

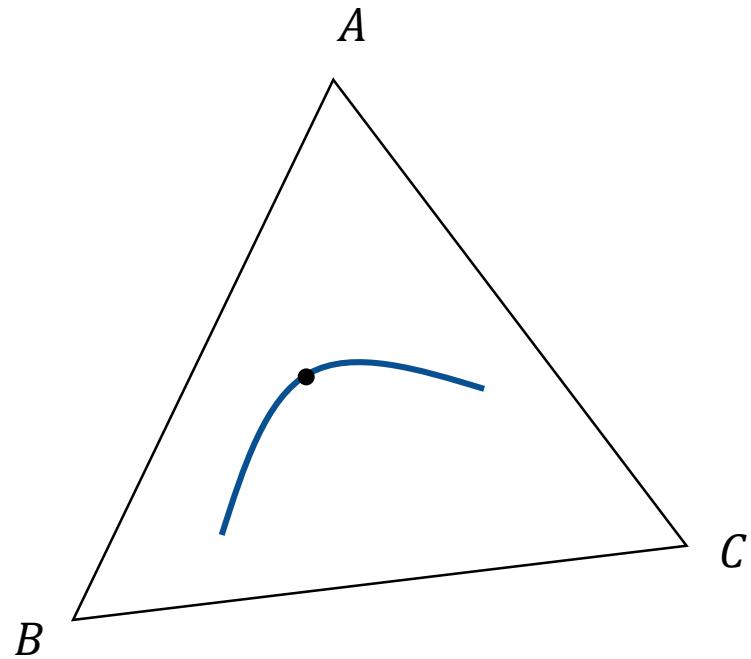
$$x_i = \sum_j \omega_{ij} p_j$$



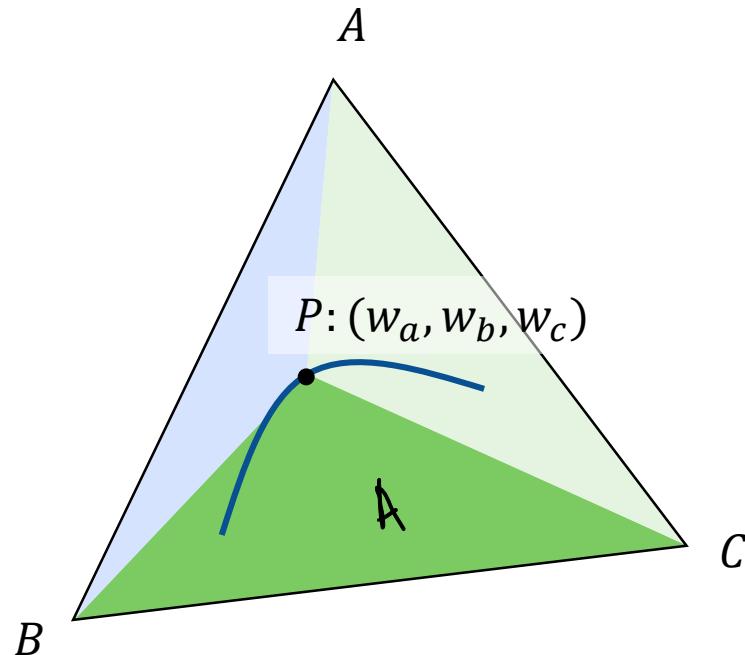
基于Cage计算权重



重心坐标



重心坐标



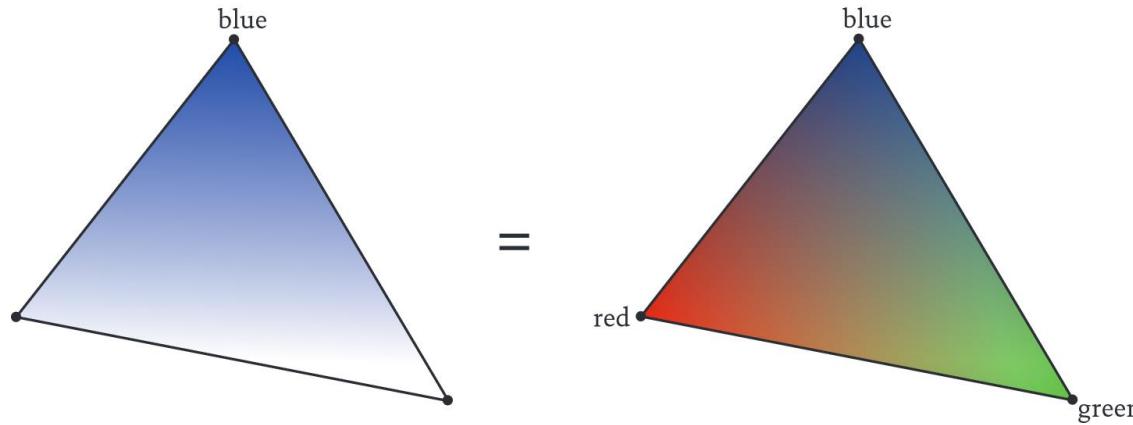
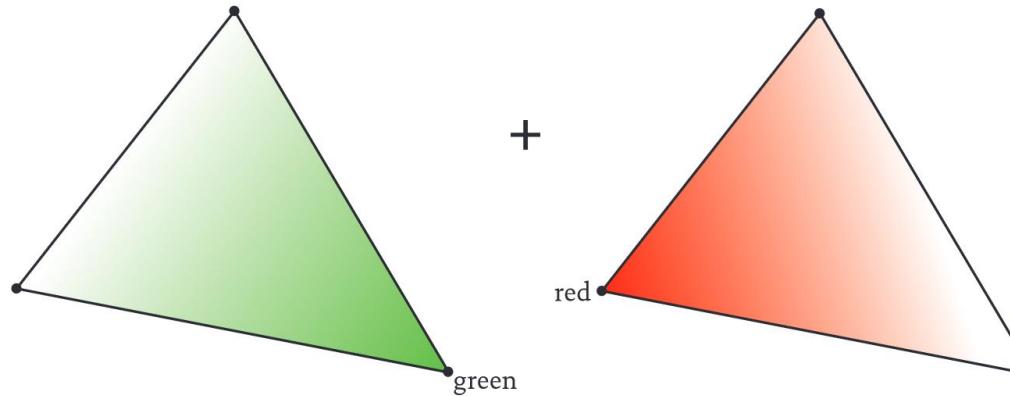
$$P = w_a A + w_b B + w_c C$$

$$w_a = \frac{\Delta PBC}{\Delta ABC}$$

$$w_b = \frac{\Delta PCA}{\Delta ABC}$$

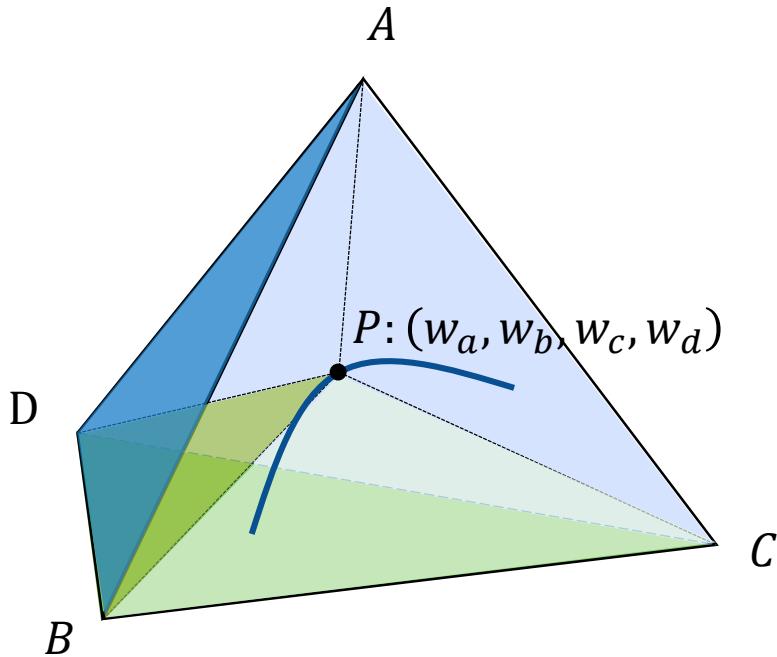
$$w_c = \frac{\Delta PAC}{\Delta ABC}$$

重心坐标



Lidberg, Petter. "Barycentric and Wachspress coordinates in two dimensions: theory and implementation for shape transformations." (2011).

重心坐标



$$P = w_a A + w_b B + w_c C + w_d D$$

$$w_a = \frac{\Delta PBCD}{\Delta ABCD}$$

$$w_b = \frac{\Delta PCDA}{\Delta ABCD}$$

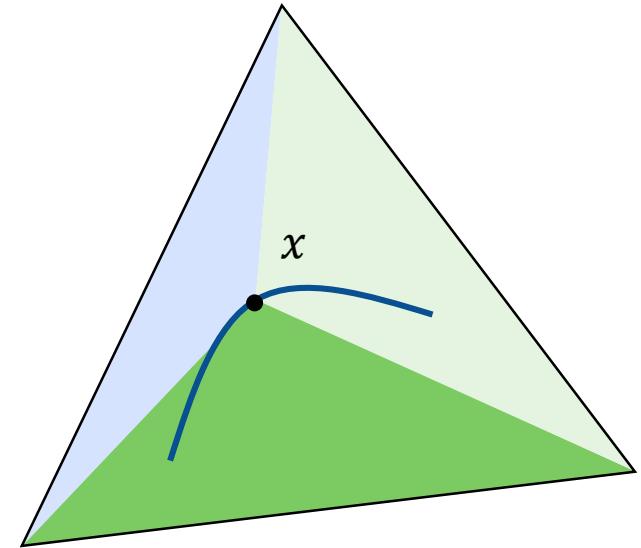
$$w_c = \frac{\Delta PDAB}{\Delta ABCD}$$

$$w_d = \frac{\Delta PABC}{\Delta ABCD}$$

三角形的重心坐标

$$x = \sum_i w_i p_i, \quad \sum_i w_i = 1, \text{且 } w_i \geq 0$$

$$x = \frac{\sum_i w'_i p_i}{\sum_i w'_i}$$



- 可写作 $x = (w_0, w_1, \dots, w_n)$
- 三角形中的点有唯一重心坐标 (why?)
 - 不同坐标组合仅差常系数
 - Homogeneous coordinates
 - 可扩展到N-维单纯形

$$\begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} (w_0, w_1, w_2) \end{bmatrix} \begin{bmatrix} Ax & Bx & 1 \\ Cx & Cy & 1 \\ Dx & Dy & 1 \end{bmatrix}^T$$

任意多边形的情况

$$x = \sum_i w_i p_i / \sum_i w_i$$

- 任意多边形重心坐标不唯一

- Wachspress (WP) coordinates

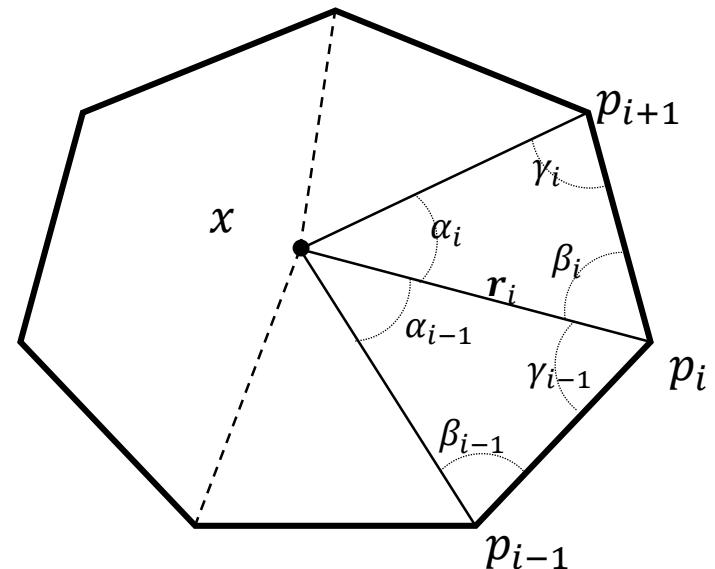
$$w_i = \frac{\cot \gamma_{i-1} + \cot \beta_i}{r_i^2}$$

- mean value (MV) coordinates

$$w_i = \frac{\tan(\alpha_{i-1}/2) + \tan(\alpha_i/2)}{r_i}$$

- discrete harmonic (DH) coordinates

$$w_i = \cot \beta_{i-1} + \cot \gamma_i$$



任意多边形的情况

$$x = \sum_i w_i p_i / \sum_i w_i$$

- 任意多边形重心坐标不唯一

- Wachspress (WP) coordinates**

$$w_i = \frac{\cot \gamma_{i-1} + \cot \beta_i}{r_i^2}$$

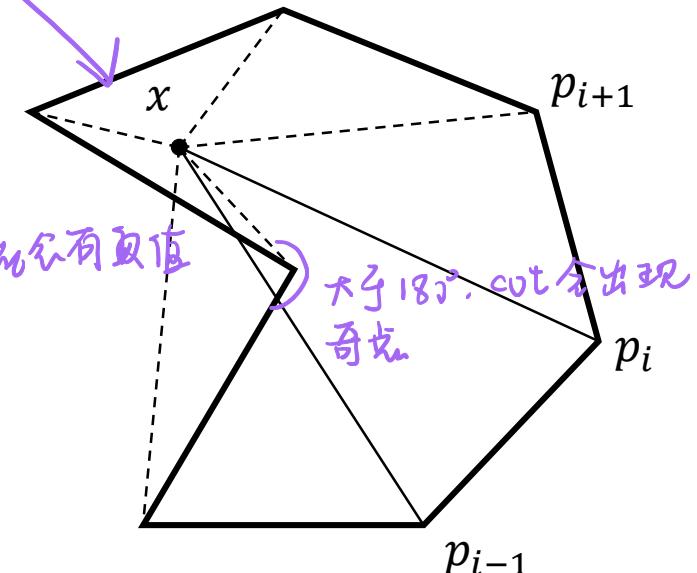
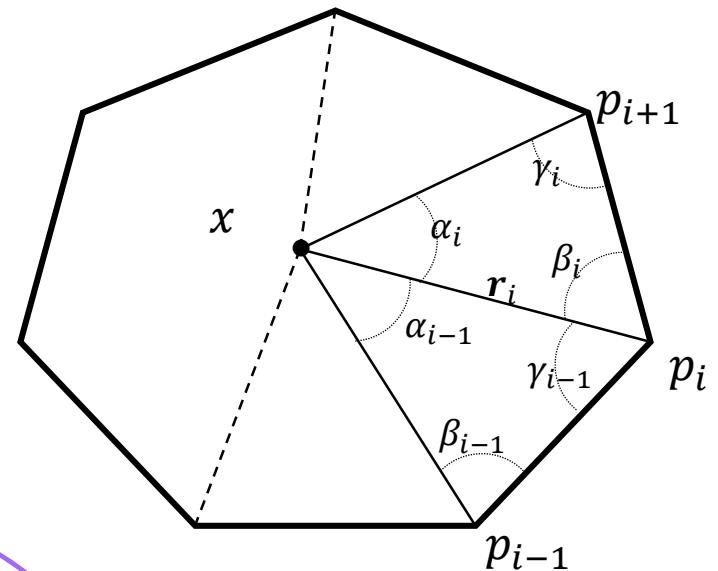
- mean value (MV) coordinates**

$$w_i = \frac{\tan(\alpha_{i-1}/2) + \tan(\alpha_i/2)}{r_i}$$

非凸的时候可能会有负值

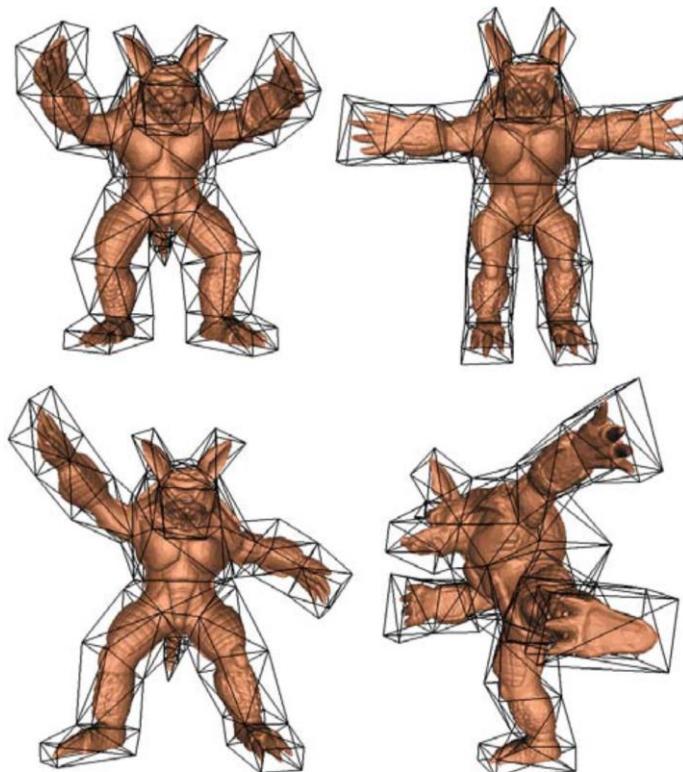
- discrete harmonic (DH) coordinates**

$$w_i = \cot \beta_{i-1} + \cot \gamma_i$$



任意多边形的情况

- Generalize to 3D Polyhedron is non-trivial
 - See <https://www.inf.usi.ch/hormann/barycentric/>



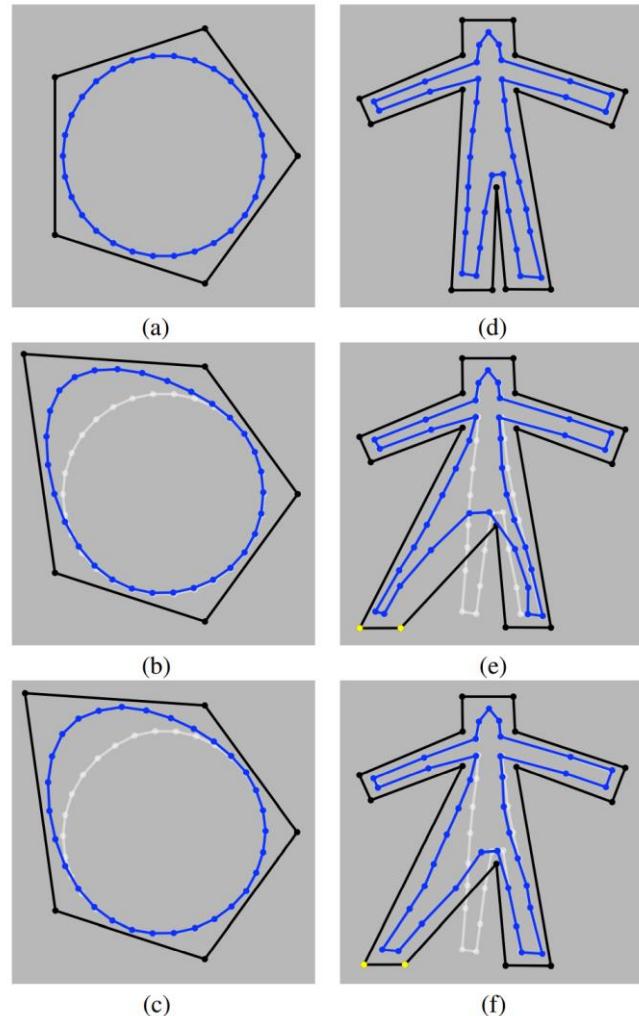
Ju, T., Schaefer, S., Warren, J.: ***Mean value coordinates for closed triangular meshes***. In: ACM SIGGRAPH 2005 Papers, SIGGRAPH '05, pp. 561–566.

更一般的情况

- Cage权重的一些性质：
 - 控制点本身的权重为1
 - 每个点的权重和为1
 - Cage内部权重连续/光滑
 - 权重非负
 - 支持线性插值

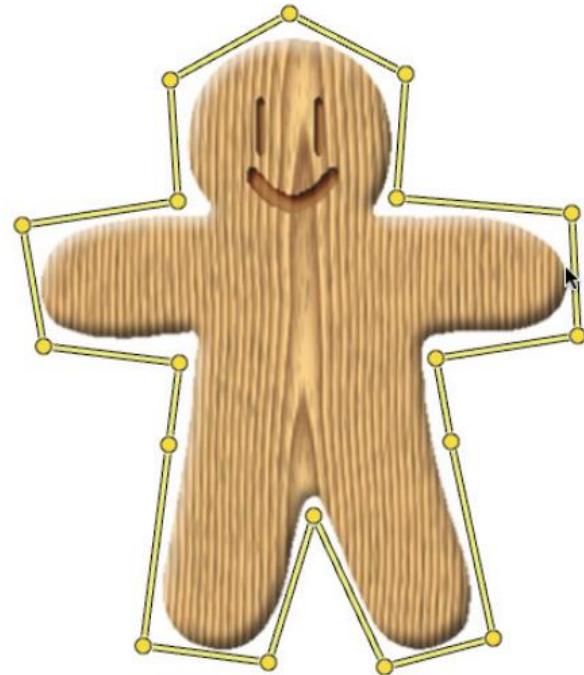
$$x_i = \sum_j \omega_{ij} p_j$$

拆三角形（三角形过大）可能导致不光滑



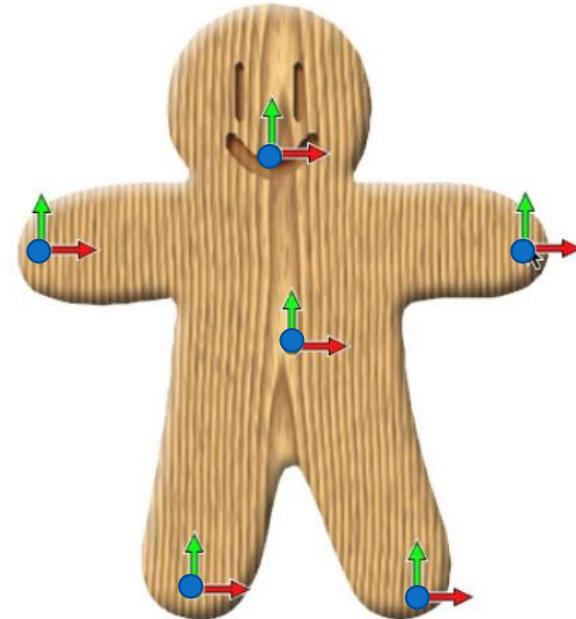
另一种形式的cage

$$x_i = \sum_j \omega_{ij} p_j$$



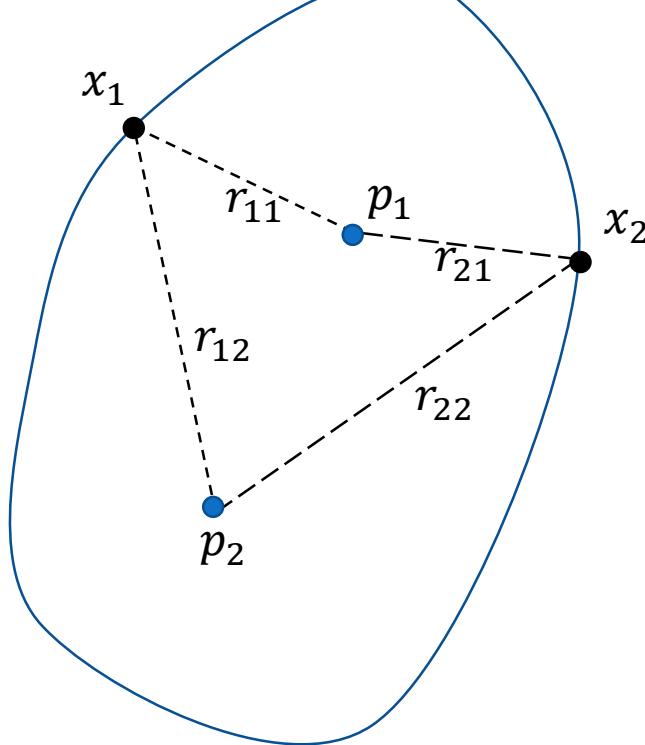
Cage

$$x_i = \sum_j \omega_{ij} p_j \quad ??$$



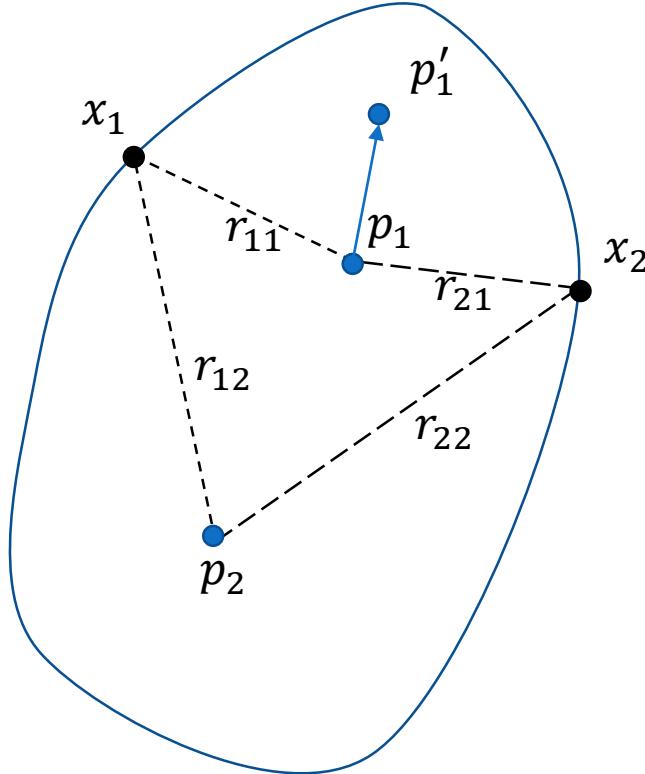
控制点

基于控制点的变形



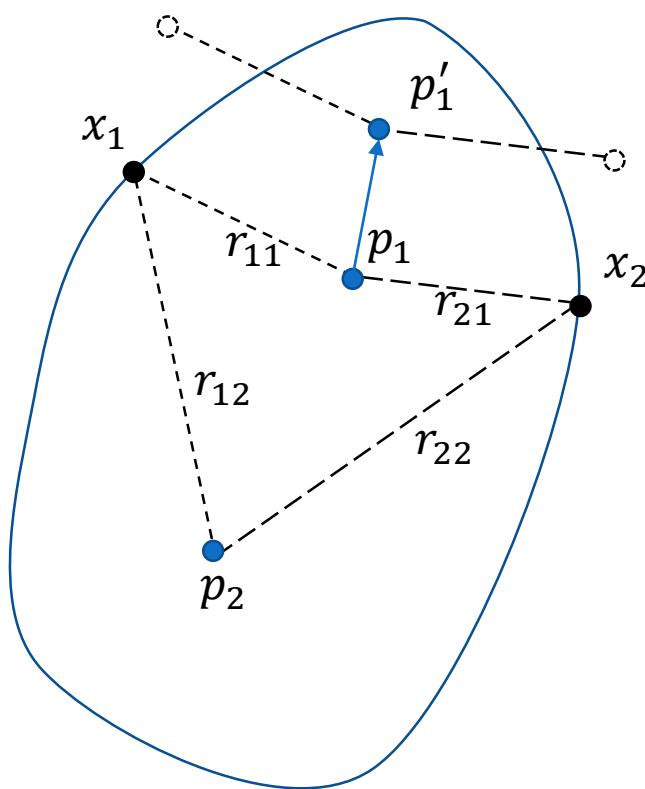
$$\begin{aligned}x_i &= \sum_j \omega_{ij} (x_i - p_j + p_j) \\&= \sum_j \omega_{ij} (r_{ij} + p_j)\end{aligned}$$

基于控制点的变形



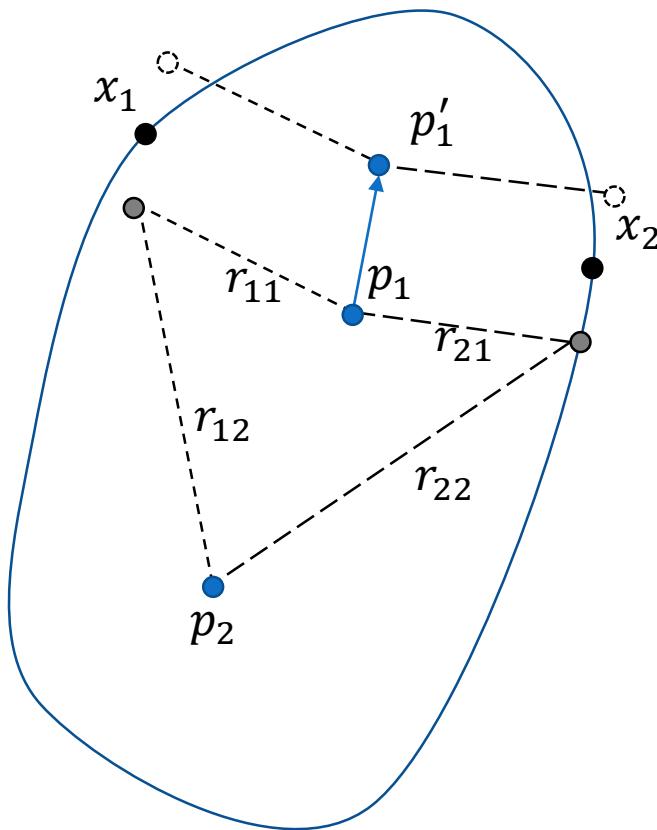
$$x'_i = \sum_j \omega_{ij} (\textcolor{red}{r}_{ij} + p'_j)$$

基于控制点的变形



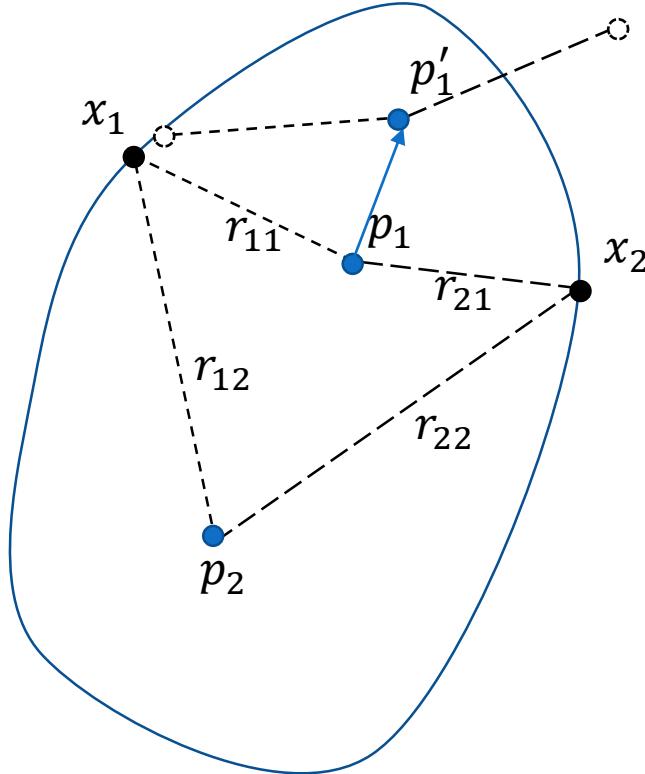
$$x'_i = \sum_j \omega_{ij} (\textcolor{red}{r}_{ij} + p'_j)$$

基于控制点的变形



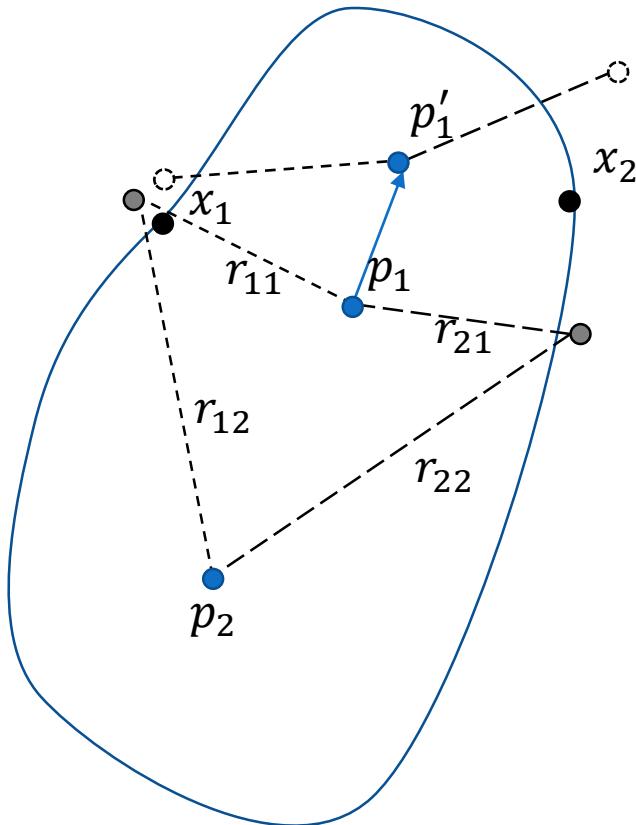
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基于控制点的变形



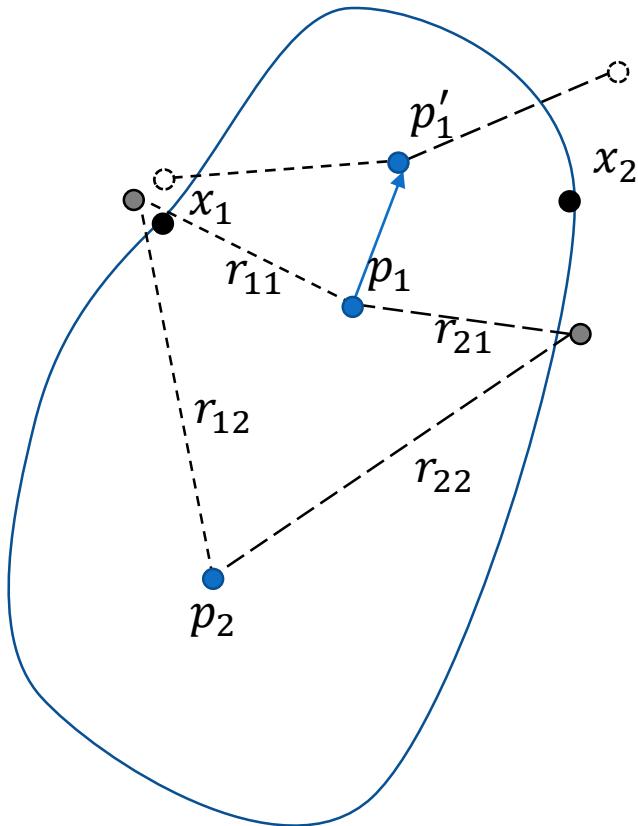
$$x'_i = \sum_j \omega_{ij} (R_j r_{ij} + p'_j)$$

基于控制点的变形



$$x'_i = \sum_j \omega_{ij} (R_j r_{ij} + p'_j)$$

基于控制点的变形



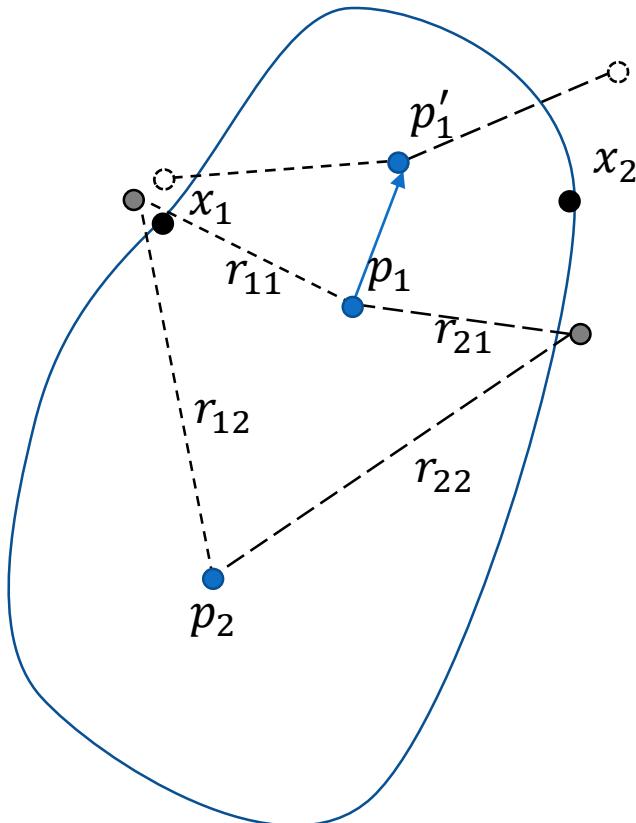
$$x'_i = \sum_j \omega_{ij} (R_j r_{ij} + p'_j)$$



$$r_{ij} = x_i - p_j$$

$$x'_i = \sum_j \omega_{ij} (R_j x_i + t_j)$$

基于控制点的变形



$$x'_i = \sum_j \omega_{ij} (R_j r_{ij} + p'_j)$$



$$r_{ij} = x_i - p_j$$

$$x'_i = \sum_j \omega_{ij} (R_j x_i + t_j)$$



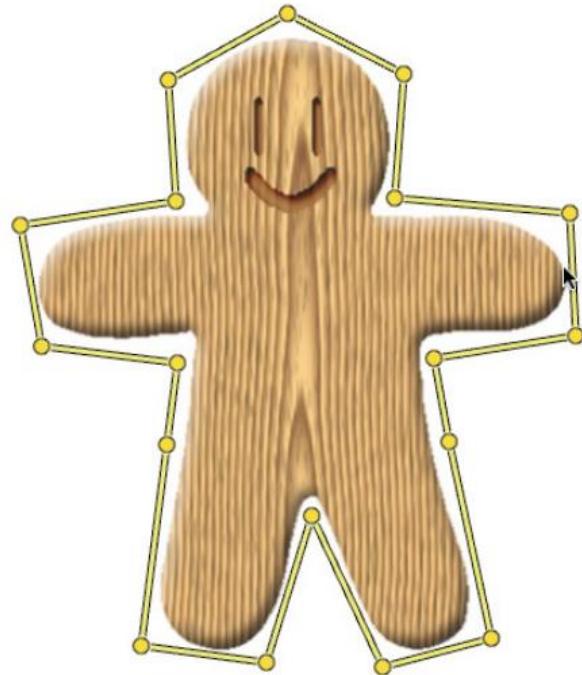
$$T = \begin{pmatrix} R & t \\ 0 & 1 \end{pmatrix}$$

$$\begin{pmatrix} x'_i \\ 1 \end{pmatrix} = \sum_j \omega_{ij} T_j \begin{pmatrix} x_i \\ 1 \end{pmatrix}$$

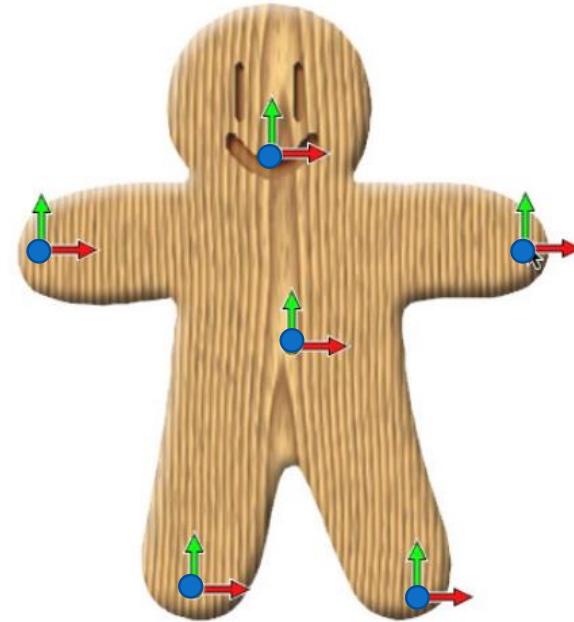
另一种形式的cage

$$x_i = \sum_j \omega_{ij} p_j$$

$$x'_i = \sum_j \omega_{ij} T_j x_i$$



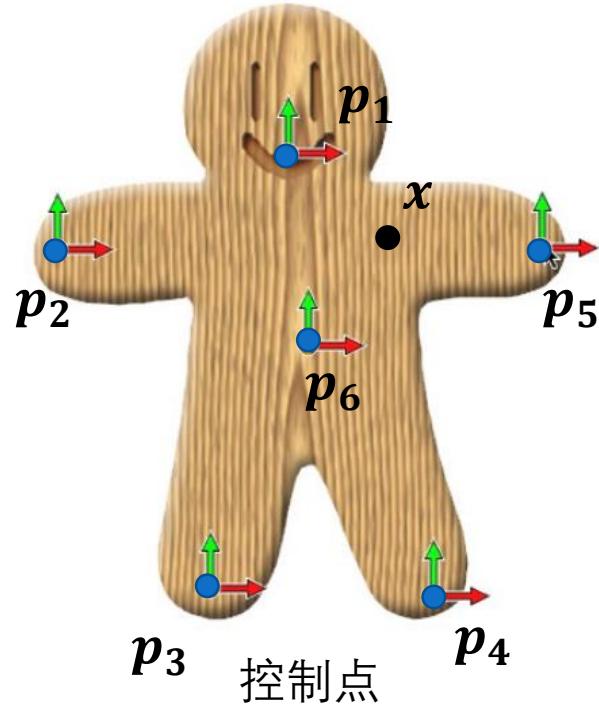
Cage



控制点

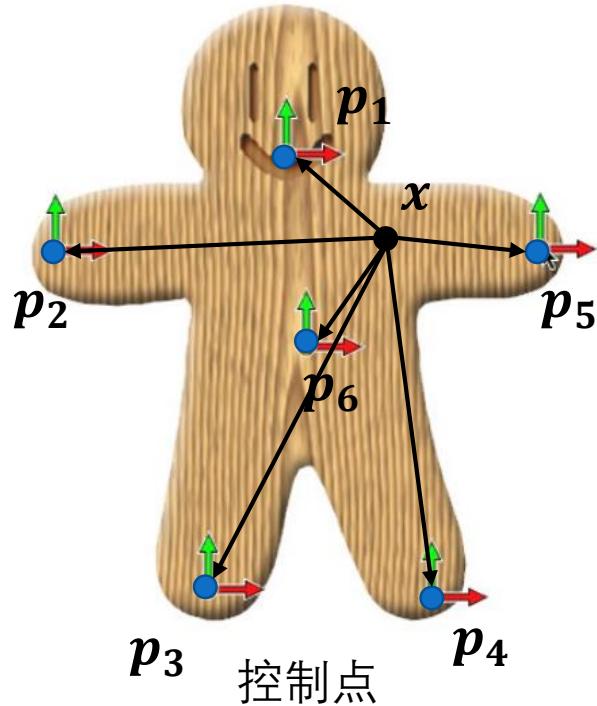
控制点权重的计算

$$x'_i = \sum_j \omega_{ij} T_j x_i$$



控制点权重的计算

$$x'_i = \sum_j \omega_{ij} T_j x_i$$



基于距离的权重

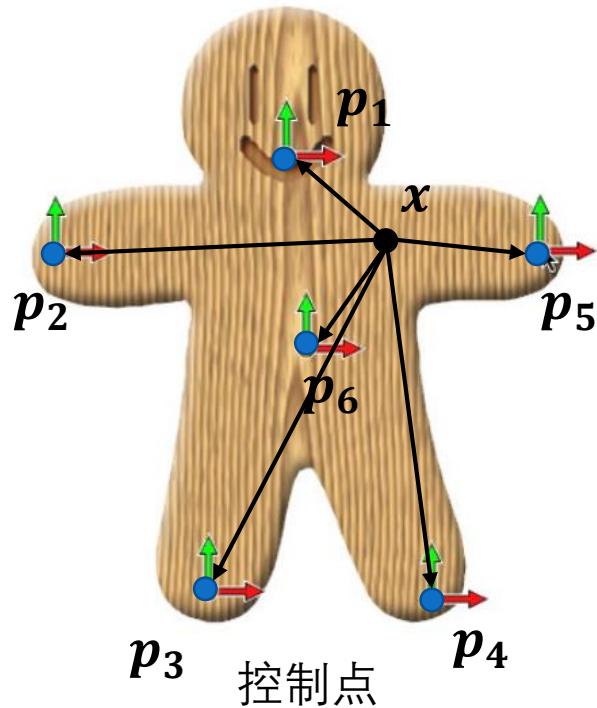
欧氏距离: $d_j = \|x - p_j\|$

权重: $w_{ij} = \begin{cases} 1, & d_j < d_k, \forall k \neq j \\ 0, & \text{otherwise} \end{cases}$

常用 但是不连续

控制点权重的计算

$$x'_i = \sum_j \omega_{ij} T_j x_i$$



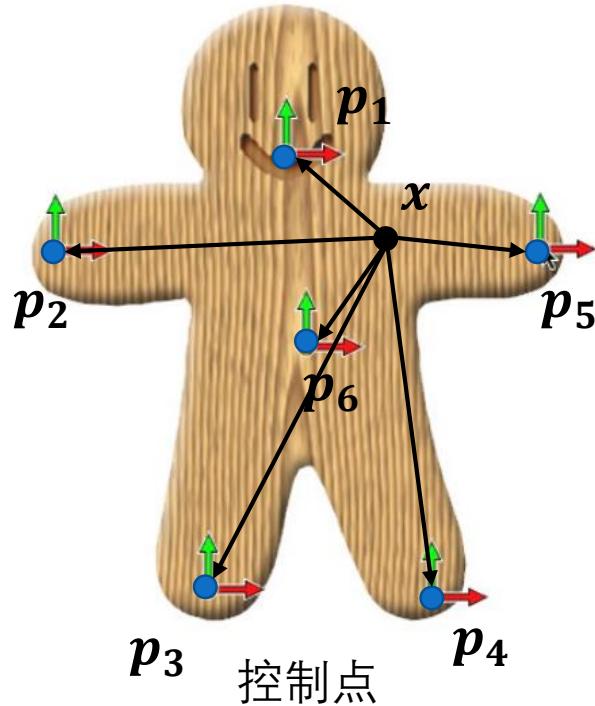
基于距离的权重

欧氏距离: $d_j = \|x - p_j\|$

权重: $w_{ij} = \frac{1}{d_j^p}$

控制点权重的计算

$$x'_i = \sum_j \omega_{ij} T_j x_i$$



基于距离的权重

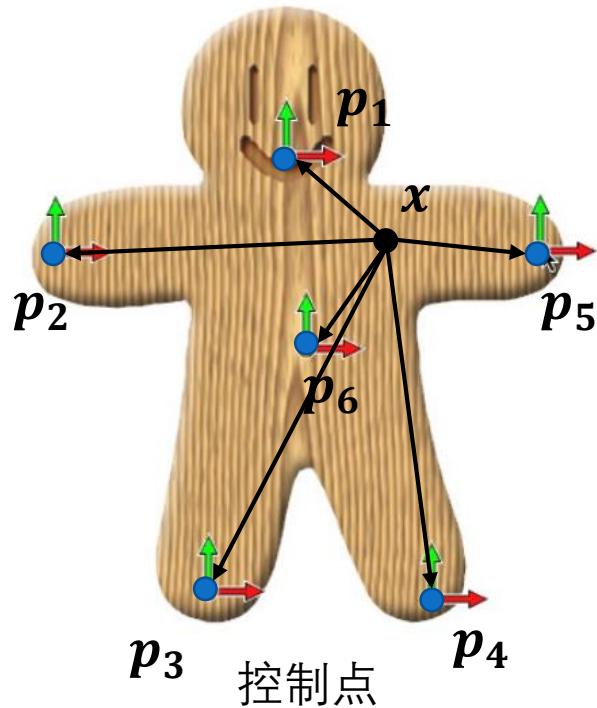
欧氏距离: $d_j = \|x - p_j\|$

权重: $\hat{w}_{ij} = \frac{1}{d_j^p}$

$$w_{ij} = \frac{\hat{w}_{ij}}{\sum_j \hat{w}_{ij}}$$

控制点权重的计算

$$x'_i = \sum_j \omega_{ij} T_j x_i$$

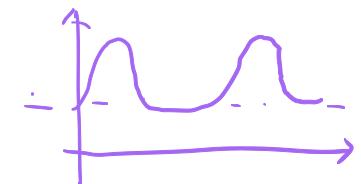


基于距离的权重

欧氏距离: $d_j = \|x - p_j\|$

权重: $\hat{w}_{ij} = \frac{1}{d_j^p + \epsilon}$

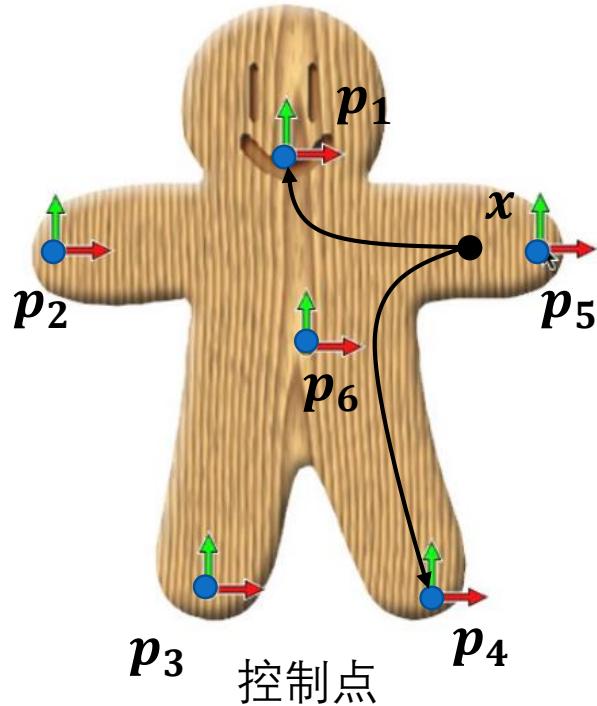
$$w_{ij} = \frac{\hat{w}_{ij}}{\sum_j \hat{w}_{ij}}$$



中越大，尖峰越陡
控制点影响范围越
小，实际上 >0 即可

控制点权重的计算

$$x'_i = \sum_j \omega_{ij} T_j x_i$$



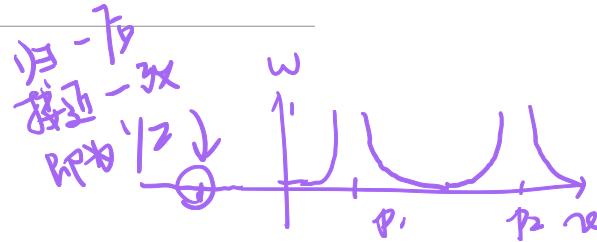
基于距离的权重

几何距离: $d_j = d(x, p_i)$

权重: $\hat{w}_{ij} = \frac{1}{d_j^p + \varepsilon}$

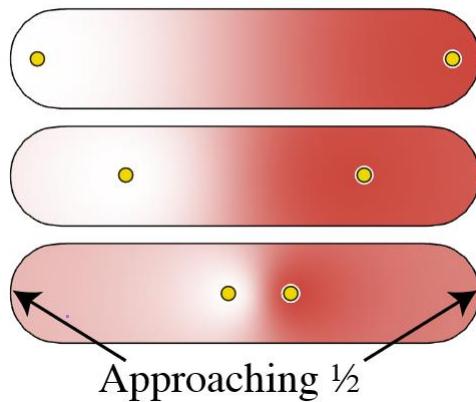
$$w_{ij} = \frac{\hat{w}_{ij}}{\sum_j \hat{w}_{ij}}$$

控制点权重的计算



例如距离

Inverse distance



Original

Inverse distance



控制点权重的计算

- 基于能量优化的控制点权重

• 权重在控制点处为1，平滑过渡到整个空间

• 平滑能量函数

$w_i(p_i) = 1, i=1, \dots, m$

mg控制点

权重分布与梯度传递相似

Common name	Least squares description	Energy	PDE
Dirichlet	f should be as constant as possible	$\int_{\Omega} \ \nabla f\ ^2 dV$	$\Delta f = 0$
Laplacian	f should be as harmonic as possible	$\int_{\Omega} (\Delta f)^2 dV$	$\Delta^2 f = 0$
Laplacian gradient	Δf should be as constant as possible	$\int_{\Omega} \ \nabla \Delta f\ ^2 dV$	$\Delta^3 f = 0$
...			

$$\begin{aligned} f(x) &= \Delta f \\ &= \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2} + \frac{\partial^2 f}{\partial z^2} \end{aligned}$$

位移变化
速度变化
加速度

控制点权重的计算

Bounded Biharmonic Weights

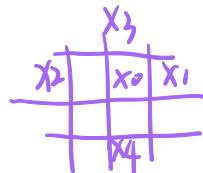
$$\operatorname{argmin}_{w_1, \dots, w_m} \sum_{j=1}^m \frac{1}{2} \int_{\Omega} (\Delta w_j(\mathbf{v}))^2 dV,$$

r3. 有愧一游

subject to $w_j(\mathbf{v}) = \varphi_j(\mathbf{v}) \quad \forall \mathbf{v} \in H,$

$$w_j(\mathbf{v}) > 0,$$

$$\sum_{j=1}^m w_j(\mathbf{v}) = 1.$$



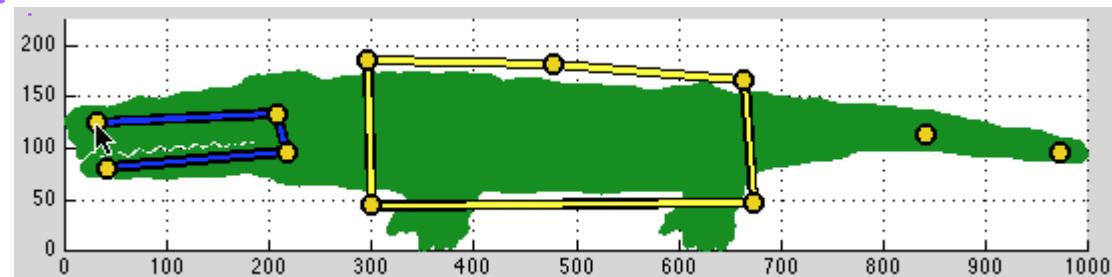
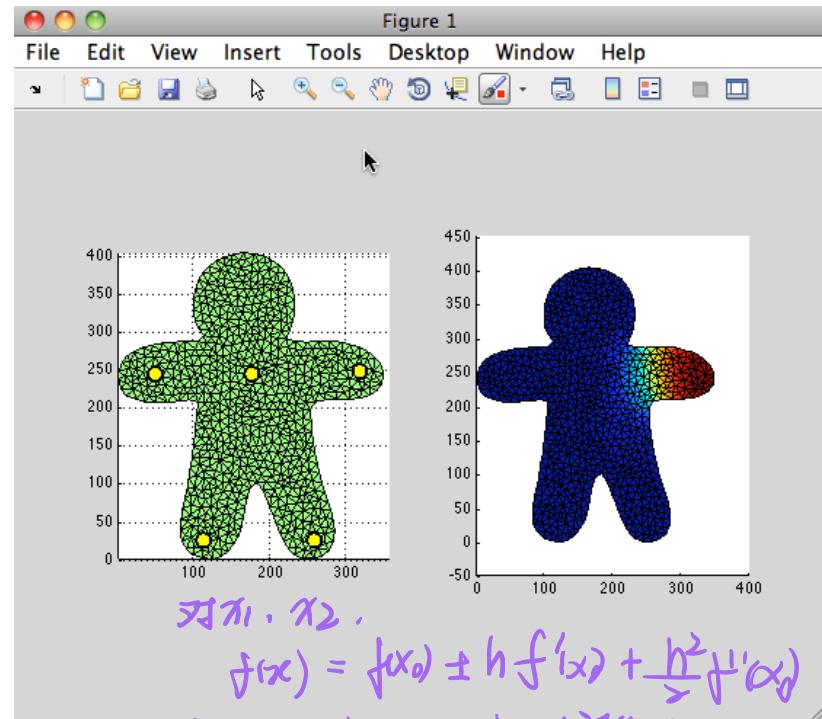
$$\frac{\partial^2 f}{\partial x^2} = \frac{x_1 + x_2 - 2x_0}{2}$$

$$\frac{\partial f}{\partial x^2} = \frac{x_3 + x_4 - 2x_0}{2}$$

$$\Delta f(x_0) = \frac{1}{4}(x_1 + x_2 + x_3 + x_4) - x_0$$

$$= (-1, \frac{1}{4}, \frac{1}{4}, \frac{1}{4}, \frac{1}{4}) \begin{bmatrix} x_0 \\ x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$$

$$\Delta f(x) = \begin{bmatrix} x_0 \\ x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$$

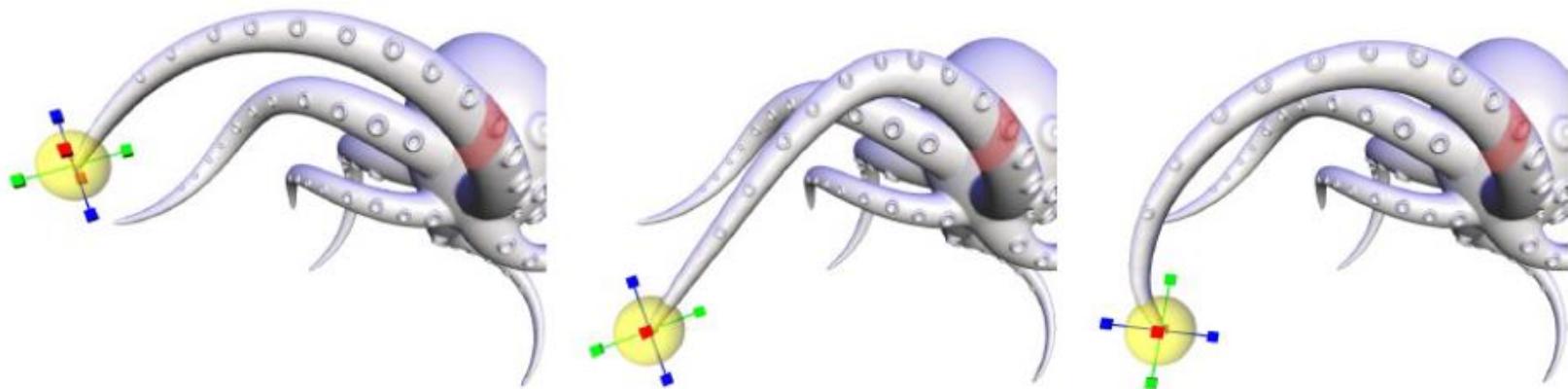


Alec Jacobson, Ilya Baran, Jovan Popović, and Olga Sorkine-Hornung. 2014. *Bounded biharmonic weights for real-time deformation*. Commun. ACM 57, 4 (April 2014)

↓ 翻译组强项

作为优化问题的几何形变

- 更加直接的形变编辑
 - 直接操作几何模型上的点
 - 将修改平滑地施加在整个几何模型上
 - 构造优化问题求解

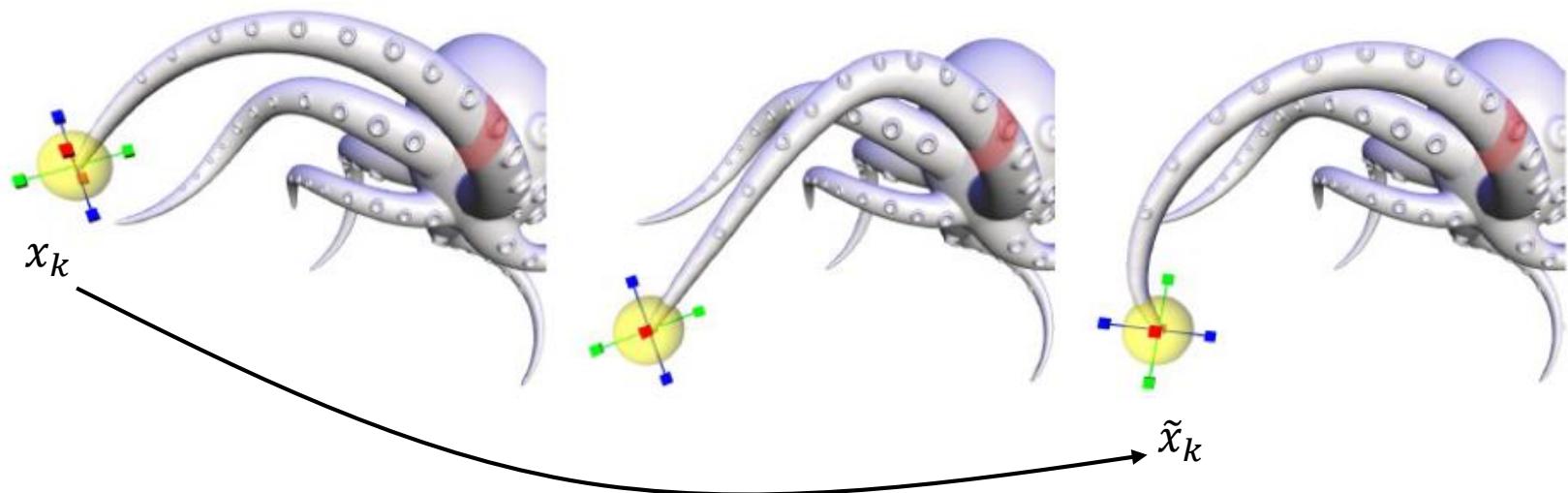


作为优化问题的几何形变

$$\min_{X'} \sum_k \|x'_k - \tilde{x}_k\|^2 + E(X, X')$$

编辑约束

平滑能量



作为优化问题的几何形变

$$\min_{X'} \sum_k \|x'_k - \tilde{x}_k\|^2 + E(X, X')$$

编辑约束

平滑能量

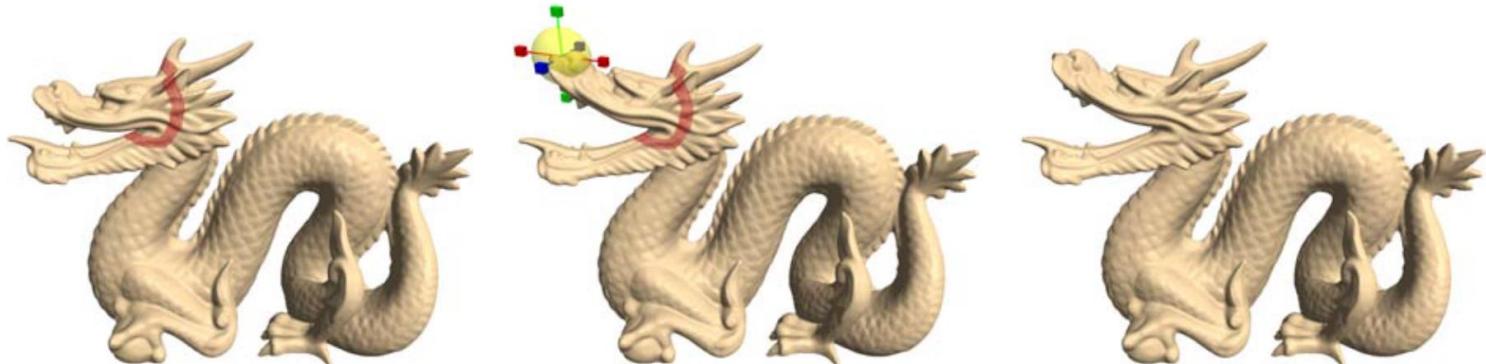
Common name	Least squares description	Energy	PDE
Dirichlet	f should be as constant as possible	$\int_{\Omega} \ \nabla f\ ^2 dV$	$\Delta f = 0$
Laplacian	f should be as harmonic as possible	$\int_{\Omega} (\Delta f)^2 dV$	$\Delta^2 f = 0$
Laplacian gradient	Δf should be as constant as possible	$\int_{\Omega} \ \nabla \Delta f\ ^2 dV$	$\Delta^3 f = 0$
...			

Laplacian Mesh Editing

$$\min_{X'} \sum_k \|x'_k - \tilde{x}_k\|^2 + \sum_i \|T_i(X')\mathcal{L}(x_i) - \mathcal{L}(x'_i)\|^2$$

编辑约束

平滑能量



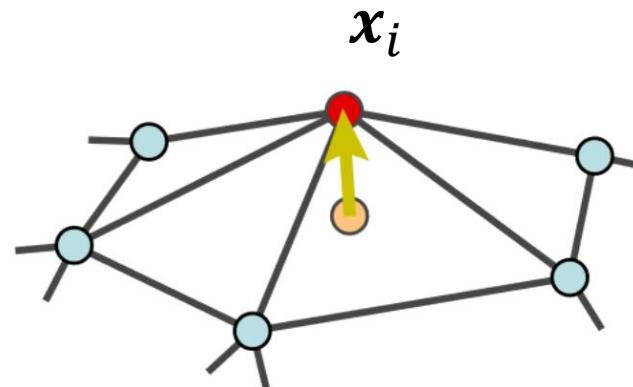
O. Sorkine, D. Cohen-Or, Y. Lipman, M. Alexa, C. Rössl, and H.-P. Seidel. 2004. *Laplacian surface editing*. In *Proceedings of the 2004 Eurographics/ACM SIGGRAPH symposium on Geometry processing (SGP '04)*, Association for Computing Machinery, New York, NY, USA, 175–184.

Laplacian坐标

$$\min_{X'} \sum_k \|x'_k - \tilde{x}_k\|^2 + \sum_i \|T_i(X')\mathcal{L}(x_i) - \mathcal{L}(x'_i)\|^2$$

$$\mathcal{L}(x) = \Delta f(x)$$

Laplace-Beltrami operator



\mathcal{N}_i : 1-neighbors

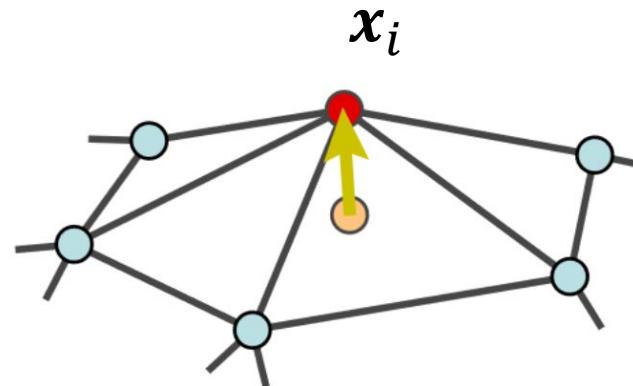
Laplacian坐标

$$\min_{X'} \sum_k \|x'_k - \tilde{x}_k\|^2 + \sum_i \|T_i(X')\mathcal{L}(x_i) - \mathcal{L}(x'_i)\|^2$$

$$\mathcal{L}(x_i) = x_i - \frac{1}{d_i} \sum_{j \in \mathcal{N}_i} x_j$$

“Umbrella” operator

可以告知某个地方是否有尖点，说明一个平面的光滑性



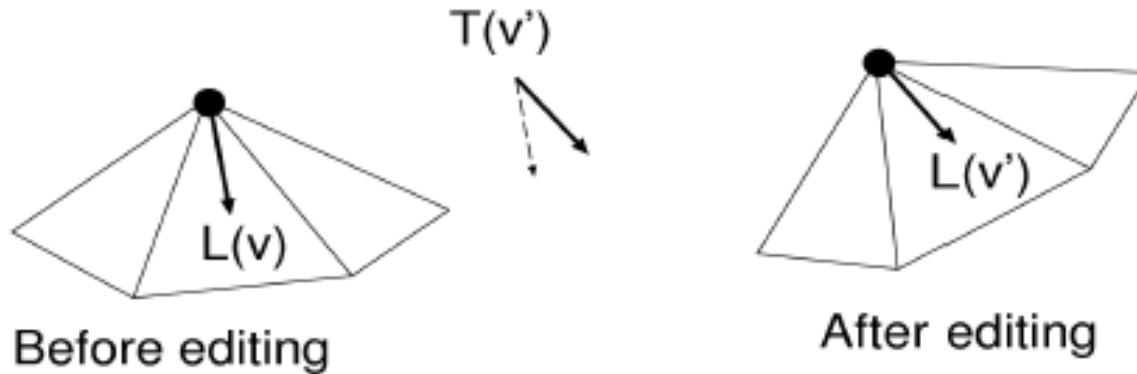
\mathcal{N}_i : 1-neighbors
一阶邻居

Laplacian坐标

$$\min_{X'} \sum_k \|x'_k - \tilde{x}_k\|^2 + \sum_i \|T_i(X')\mathcal{L}(x_i) - \mathcal{L}(x'_i)\|^2$$

- 我们希望变换过后每个点的Laplacian坐标不变

维持平滑性



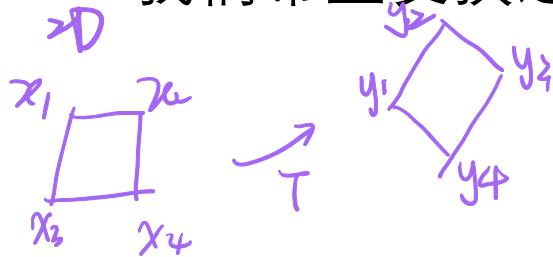
Laplacian坐标

$$\min_{x', \mathbf{T}_i} \sum_k \|x'_k - \tilde{x}_k\|^2 + \sum_i \|T_i(x') \mathcal{L}(x_i) - \mathcal{L}(x'_i)\|^2$$

T 的求解：

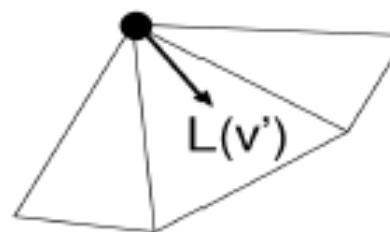
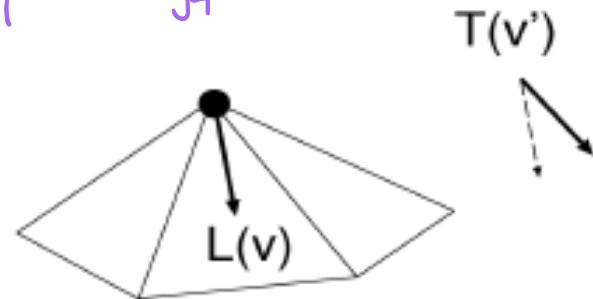
- 我们希望变换过后每个点的 Laplacian 坐标不变

用梯度下降方向，通过迭代求解。
先固定 T ，估计 \mathcal{L} ，
再对 x 求优化，再
估计 T ……



$$Tx_i \rightarrow y_i$$

$$T \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$



After editing

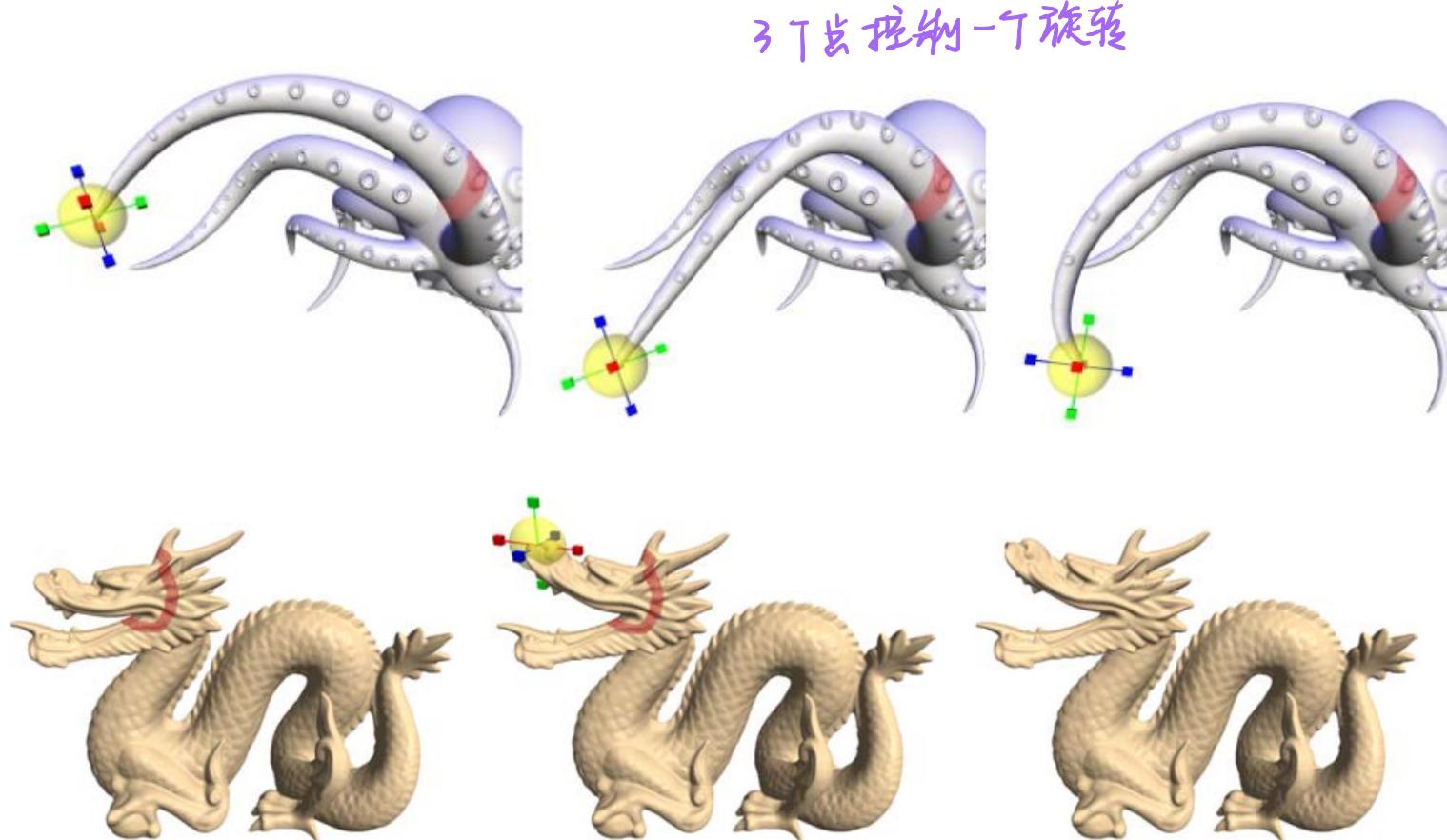
$$\begin{bmatrix} x_1 & x_2 & 0 & 1 & 0 & 0 \\ 0 & 0 & x_1 & x_1^2 & 0 & 1 \\ \vdots & & & & & \end{bmatrix} \begin{bmatrix} t_1 \\ t_2 \\ t_3 \\ t_p \\ v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

$$T = [t_1, t_2, v_1, t_3, t_p, v_2]$$

(对约束)

合在一起是一个最小二乘问题，左边为一个 8×6 矩阵

Laplacian Mesh Editing

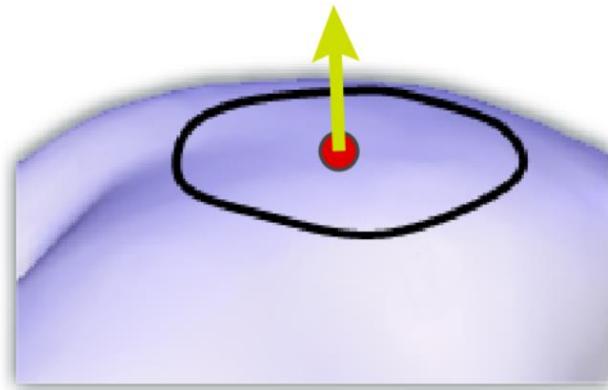
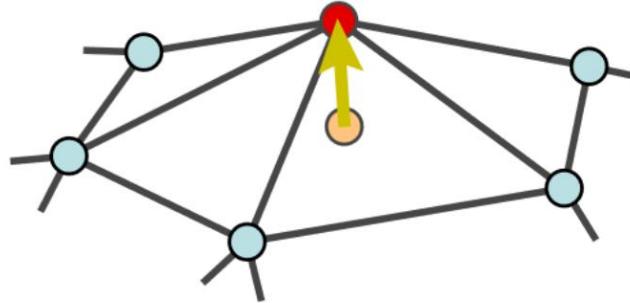


O. Sorkine, D. Cohen-Or, Y. Lipman, M. Alexa, C. Rössl, and H.-P. Seidel. 2004. *Laplacian surface editing*. In *Proceedings of the 2004 Eurographics/ACM SIGGRAPH symposium on Geometry processing (SGP '04)*, Association for Computing Machinery, New York, NY, USA, 175–184.

Laplacian坐标的含义

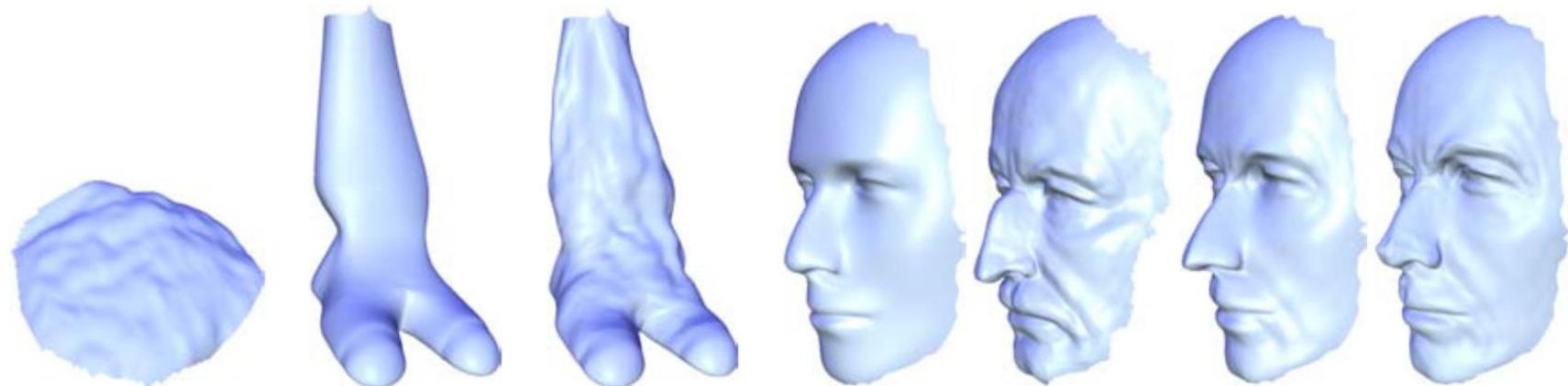
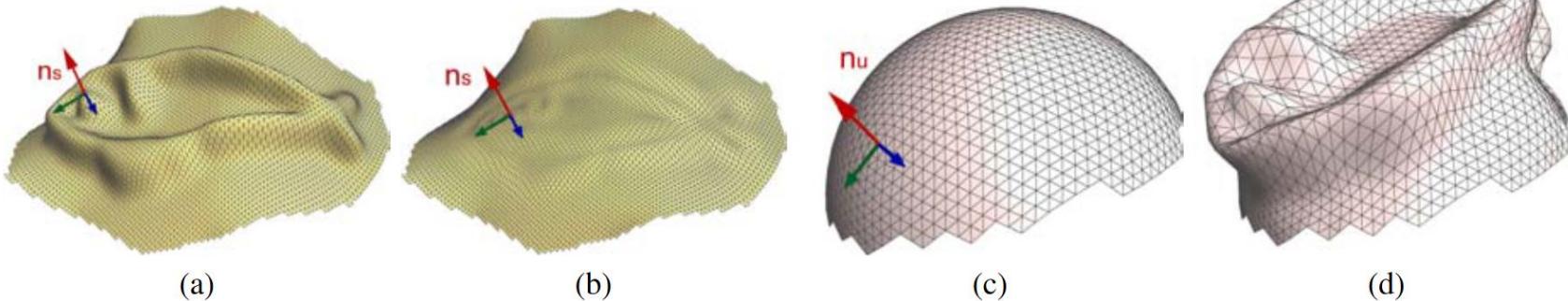
- 几何表面局部的细节

- 方向 \approx 法向
- 大小 \approx 曲率

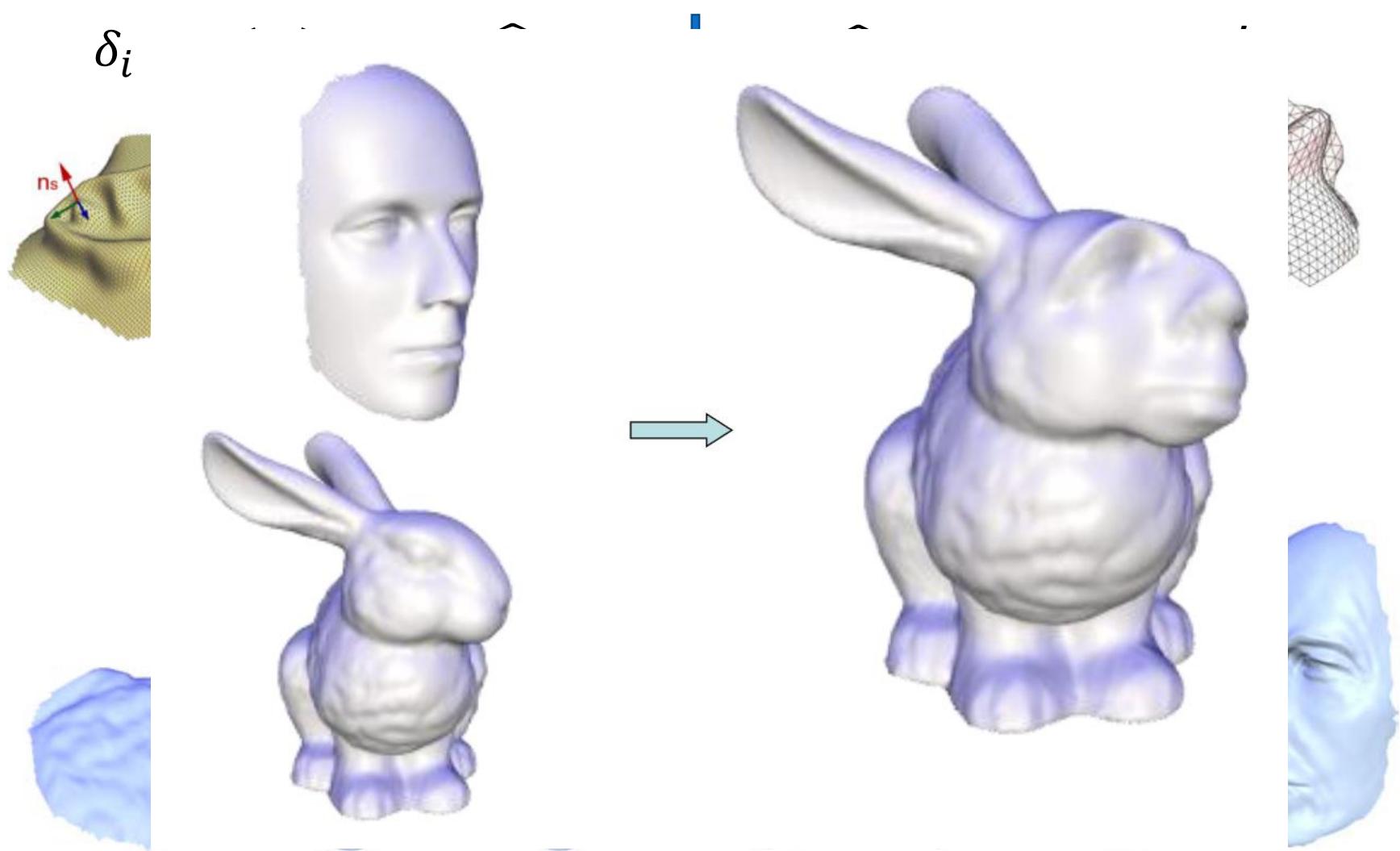


Coating Transfer

$$\delta_i = \mathcal{L}(x_i) - \widehat{\delta}_i + \widehat{\delta}'_i = \delta'_i$$



Coating Transfer



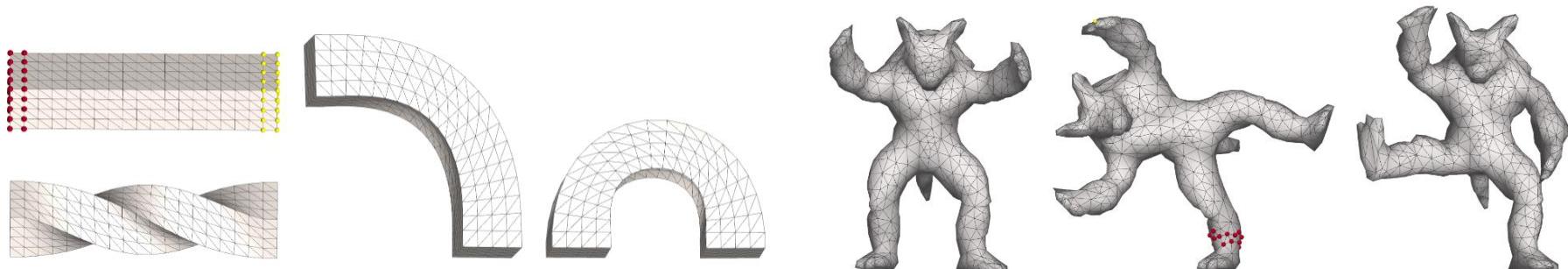
As Rigid As Possible Deformation

$$\min_{X'} \sum_k \|x'_k - \tilde{x}_k\|^2 + E(X, X')$$

编辑约束
平滑能量项

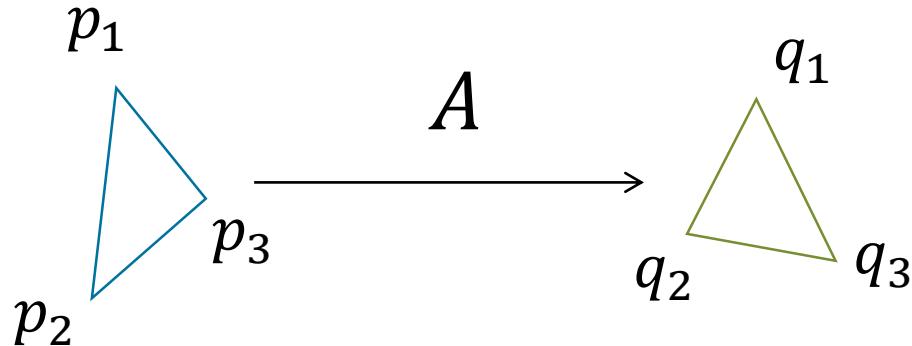
刚性很强 $T = \begin{bmatrix} P & T \\ 0 & I \end{bmatrix}$
 $\det(T) = 1$

尽可能减少非“刚性”形变 → 减少不必要的缩放



Olga Sorkine and Marc Alexa. 2007. *As-rigid-as-possible surface modeling*. In *Proceedings of the fifth Eurographics symposium on Geometry processing (SGP '07)*, Eurographics Association, Goslar, DEU, 109–116.

As Rigid As Possible Deformation



$$\mathbf{q}_i = A\mathbf{p}_i + \mathbf{t}$$

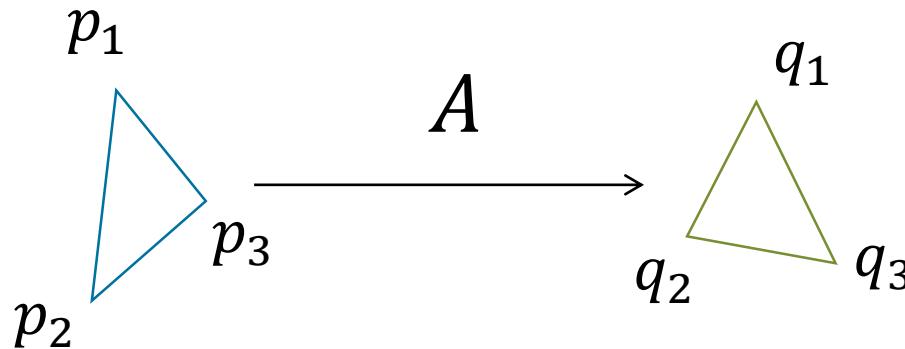
剛體變形. 平移 + 軸旋轉

$$\min_{R,t} \sum_{i=1}^3 \|R\mathbf{p}_i + \mathbf{t} - \mathbf{q}_i\|^2$$

s.t. $R^T R = 1$

$$A = U\Sigma V^T = (UV^T)(VDV^T) = RS$$

As Rigid As Possible Shape Interpolation



Interpolation between two shapes

$$\mathbf{q}_i(t) = \mathbf{A}(t)\mathbf{p}_i + \mathbf{t}$$

r, s 分别插值， s 正且对称

$$\mathbf{A}(t) = \mathbf{R}(t)((1-t)\mathbf{I} + t\mathbf{S})$$

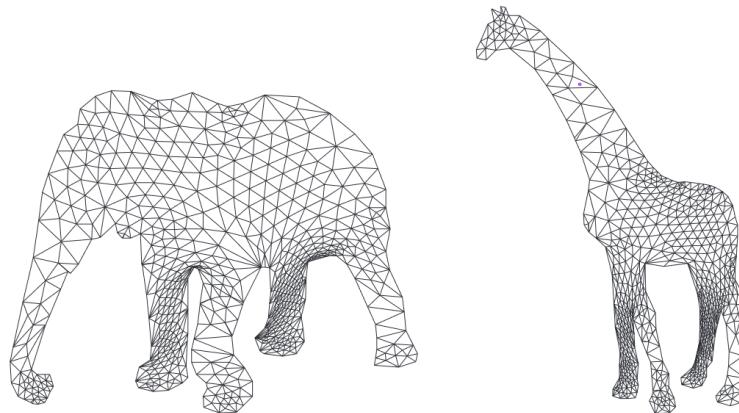
顶点线性插值



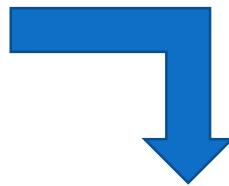
刚性变换插值



As Rigid As Possible Shape Interpolation



多个三角形通过之前的计算方式可能会分开，但是可以通过优化所有的顶点的坐标值，使得每个三角形的变换与插值得到的R、S相近，这样就可以保持一个完整的形状。这个问题实际上也是一个二次问题。



$$\min_{D(i)} = \sum_{i=1}^n \| A(x_i) - R_i S_i \|_F^2$$



Marc Alexa, Daniel Cohen-Or, and David Levin. 2000. *As-rigid-as-possible shape interpolation*. In *Proceedings of the 27th annual conference on Computer graphics and interactive techniques (SIGGRAPH '00)*

As Rigid As Possible Deformation

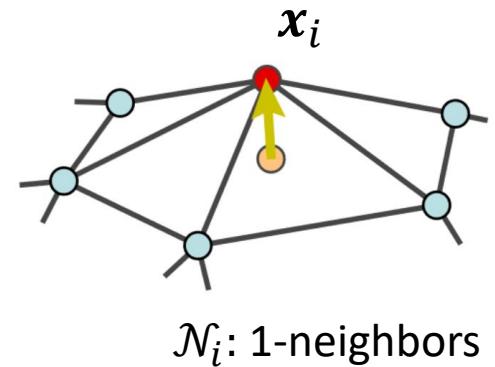
$$\min_{X'} \sum_k \|x'_k - \tilde{x}_k\|^2 + E(X, X')$$

编辑约束

平滑能量项

尽可能减少非“刚性”形变 → 减少不必要的缩放

$$E(X, X') = \sum_{j \in \mathcal{N}_i} w_{ij} \|(x'_i - x'_j) - R_i(x_i - x_j)\|^2$$



As Rigid As Possible Deformation

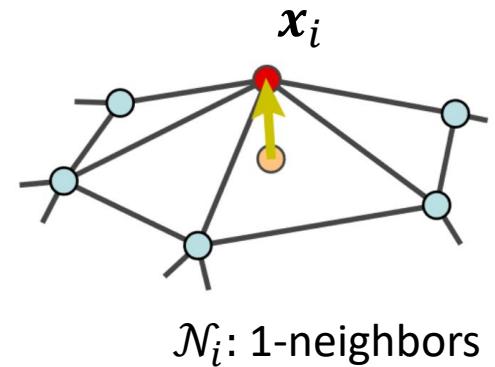
$$\min_{X', \mathbf{R}_i} \sum_k \|x'_k - \tilde{x}_k\|^2 + E(X, X')$$

编辑约束

平滑能量项

尽可能减少非“刚性”形变 → 减少不必要的缩放

$$E(X, X') = \sum_{j \in \mathcal{N}_i} w_{ij} \|(x'_i - x'_j) - R_i(x_i - x_j)\|^2$$



As Rigid As Possible Deformation



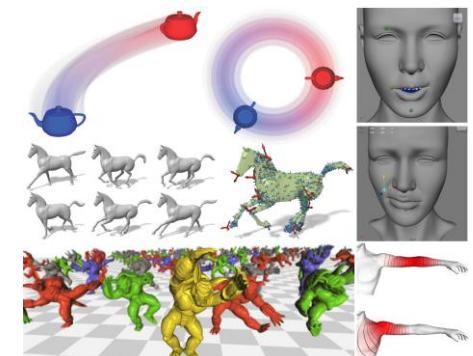
尽可能的保持了体积不变

蒙皮变形

Skinning

What is covered

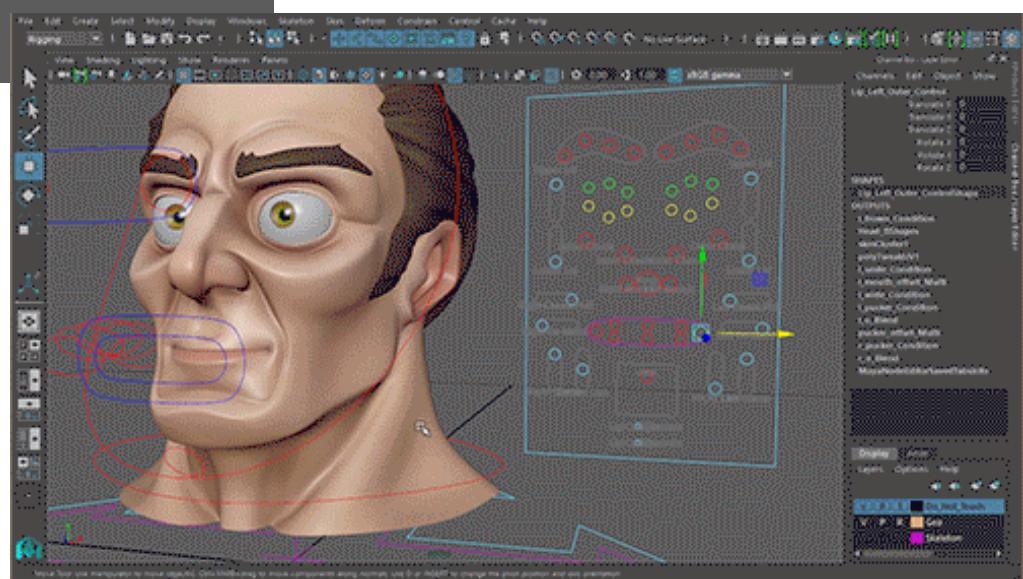
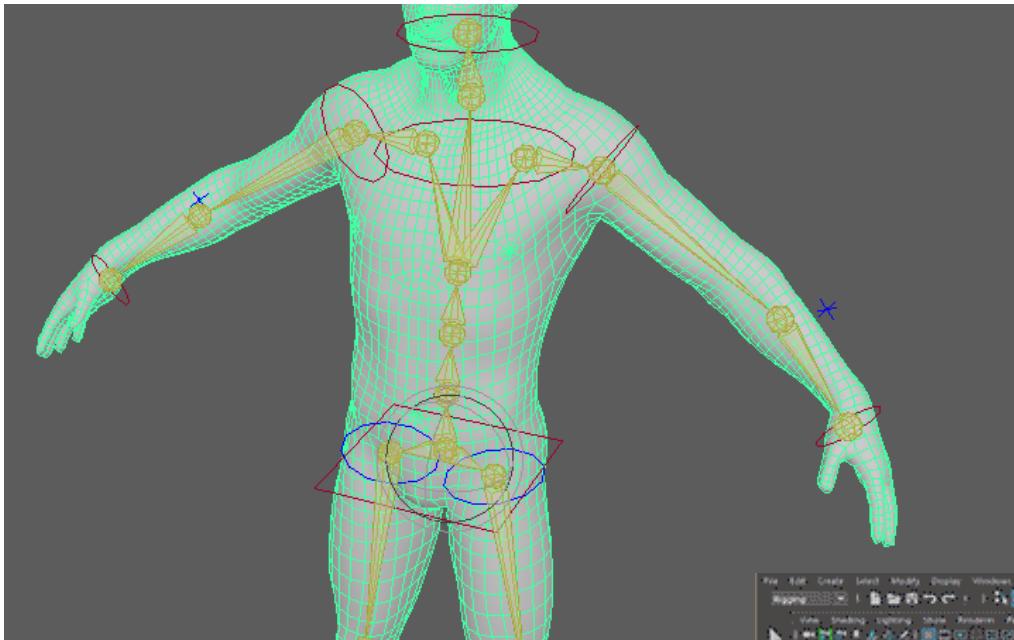
- How to deform a mesh according to skeletal animation?
 - Linear Blend Skinning (LBS)
 - Multi-linear Skinning
 - Non-linear Skinning
 - Dual-Quaternion Skinning (DQS)
- How to compute weights?
- Example-based methods?



Many images are from: <https://skinning.org/>

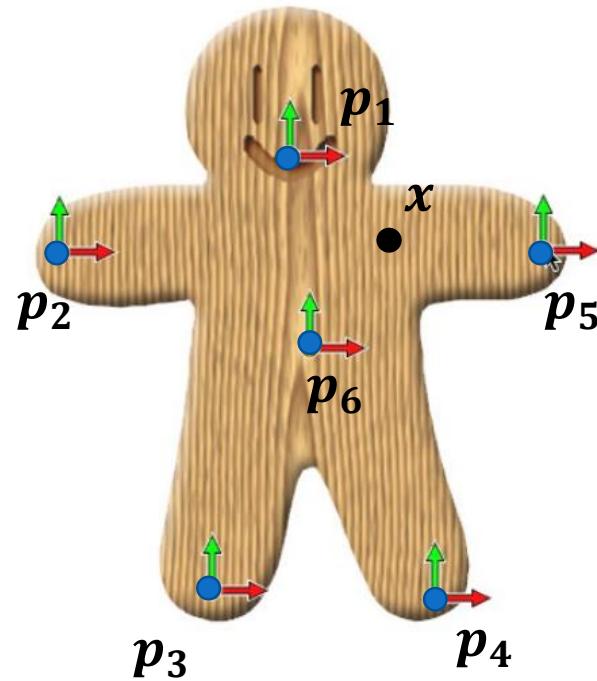
Alec Jacobson, Zhigang Deng, Ladislav Kavan, and J. P. Lewis. 2014. *Skinning: real-time shape deformation*. In ACM SIGGRAPH 2014 Courses (SIGGRAPH '14)

Skinning



Recall: deformation with control points

$$x'_i = \sum_j \omega_{ij} T_j x_i$$

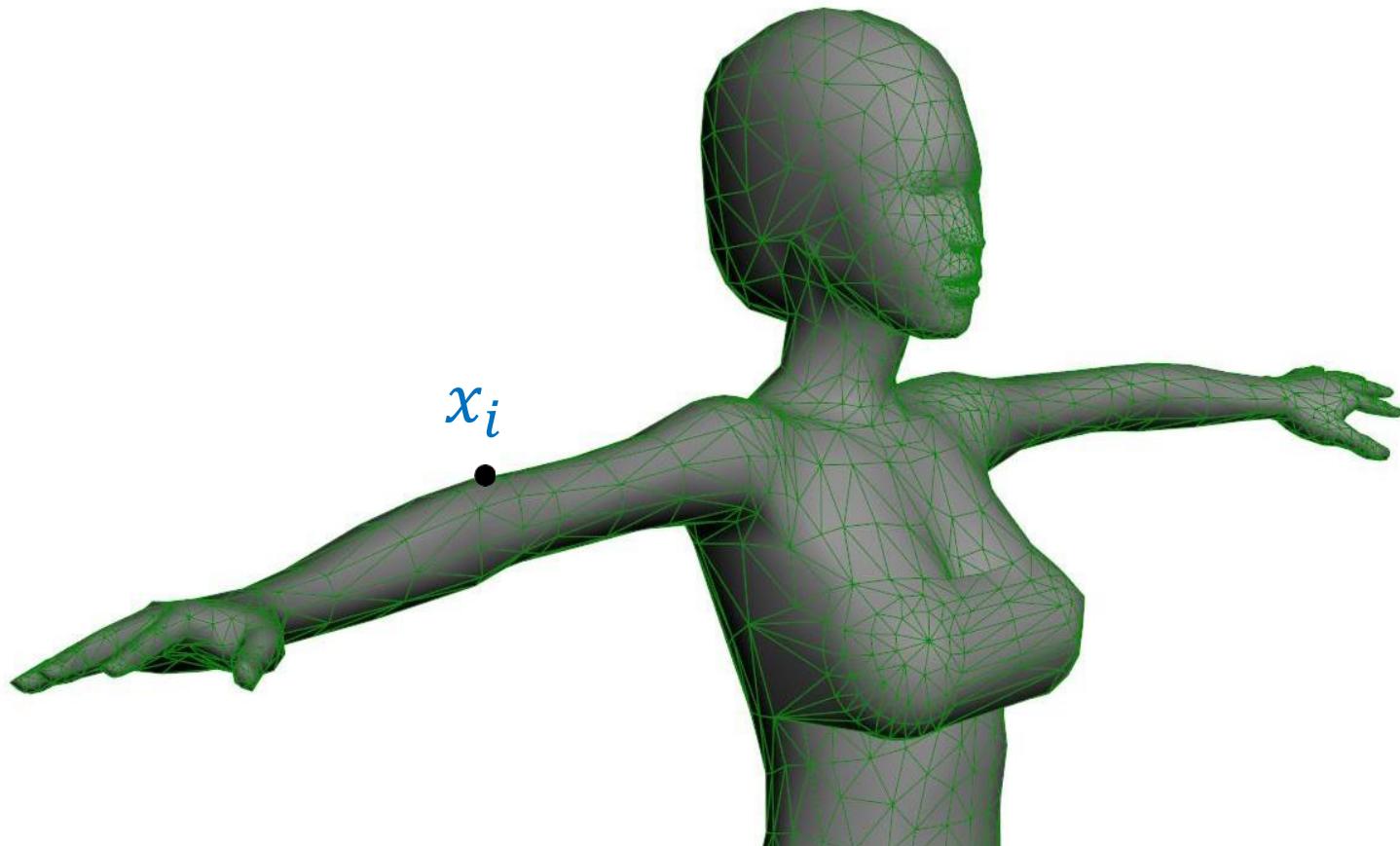


Basic Setup

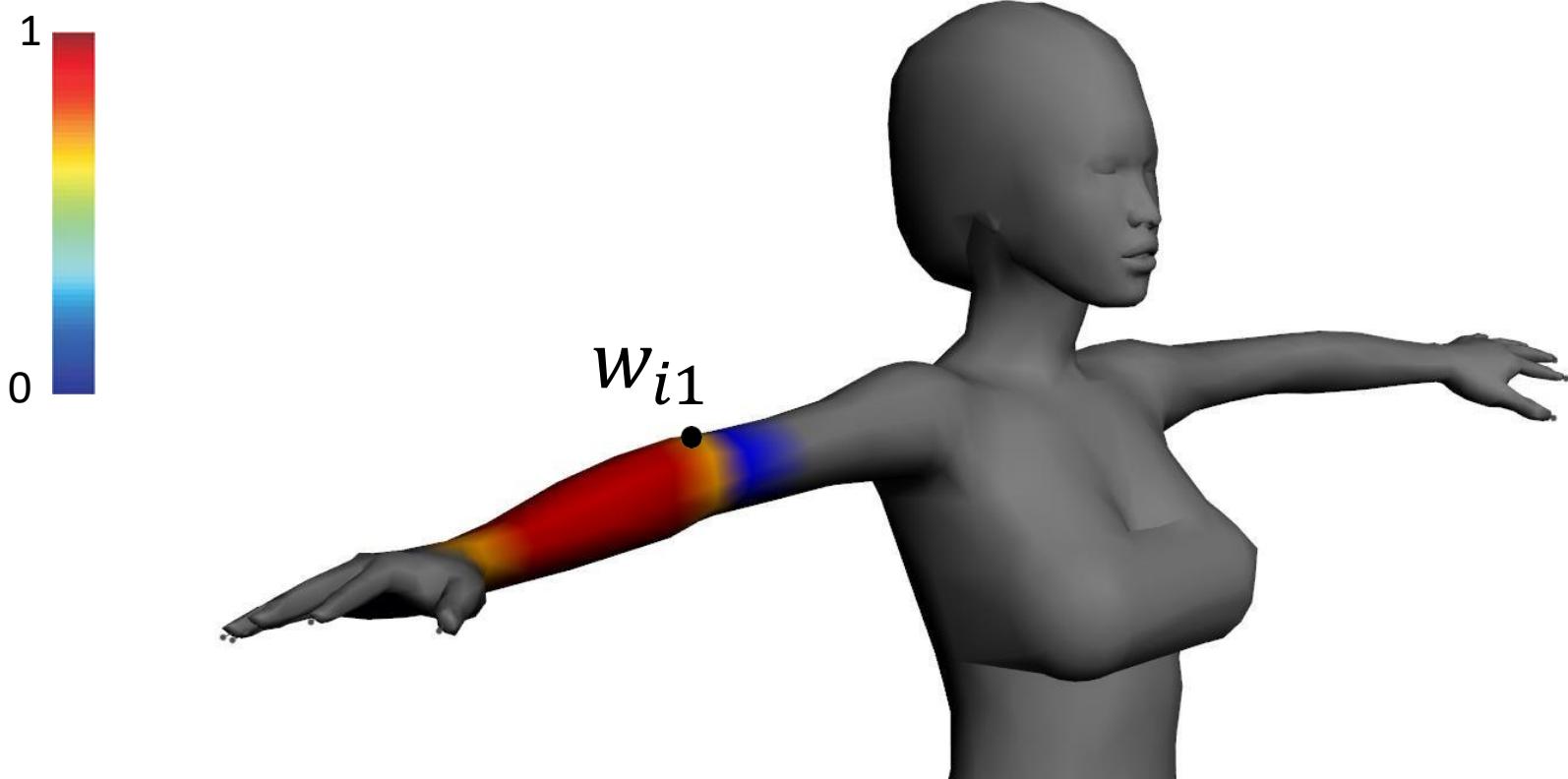
- Rest pose x_0
- Skinning weights ω
- Skinning transformation T

$$x'_i = \sum_{j=1}^m \omega_{ij} T_j x_i$$

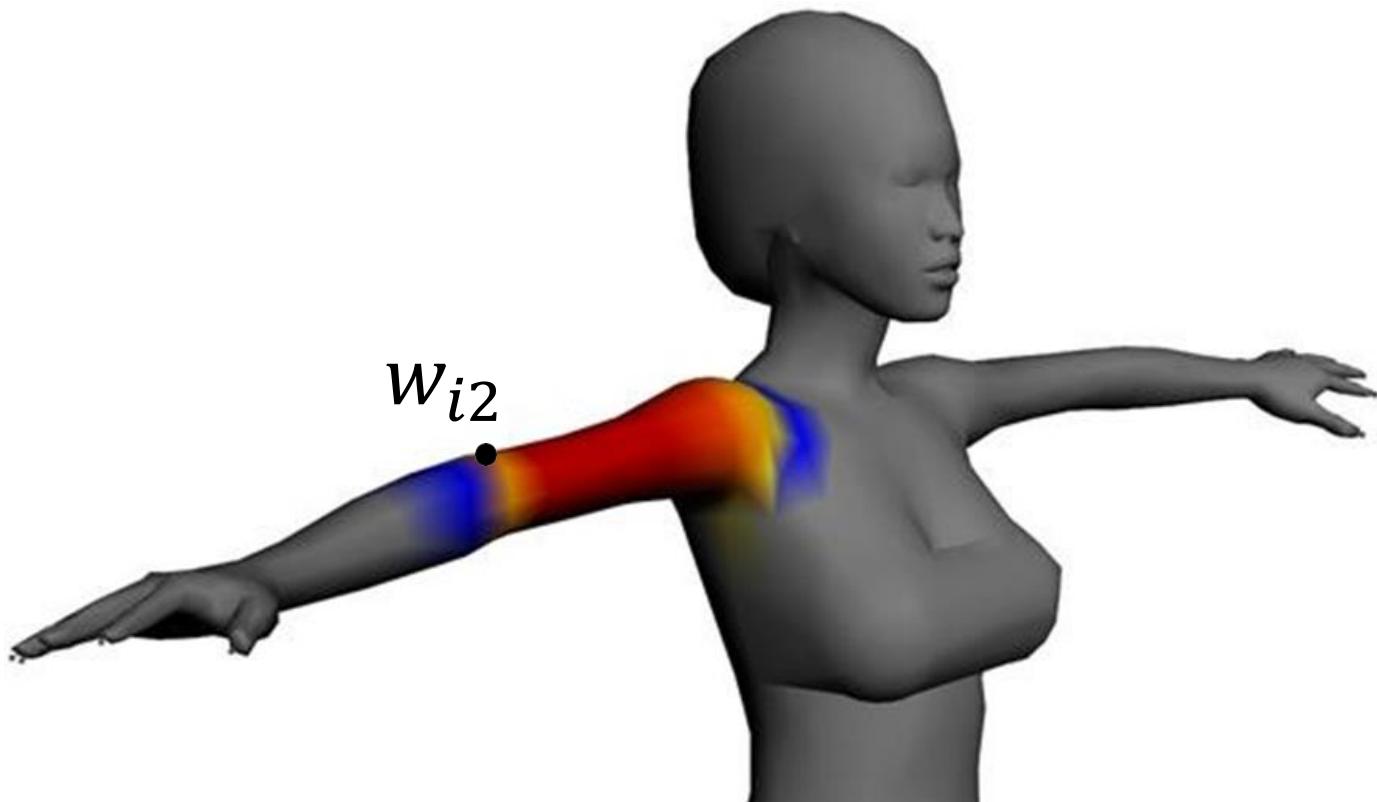
Basic Setup: Rest Pose/Bind Pose



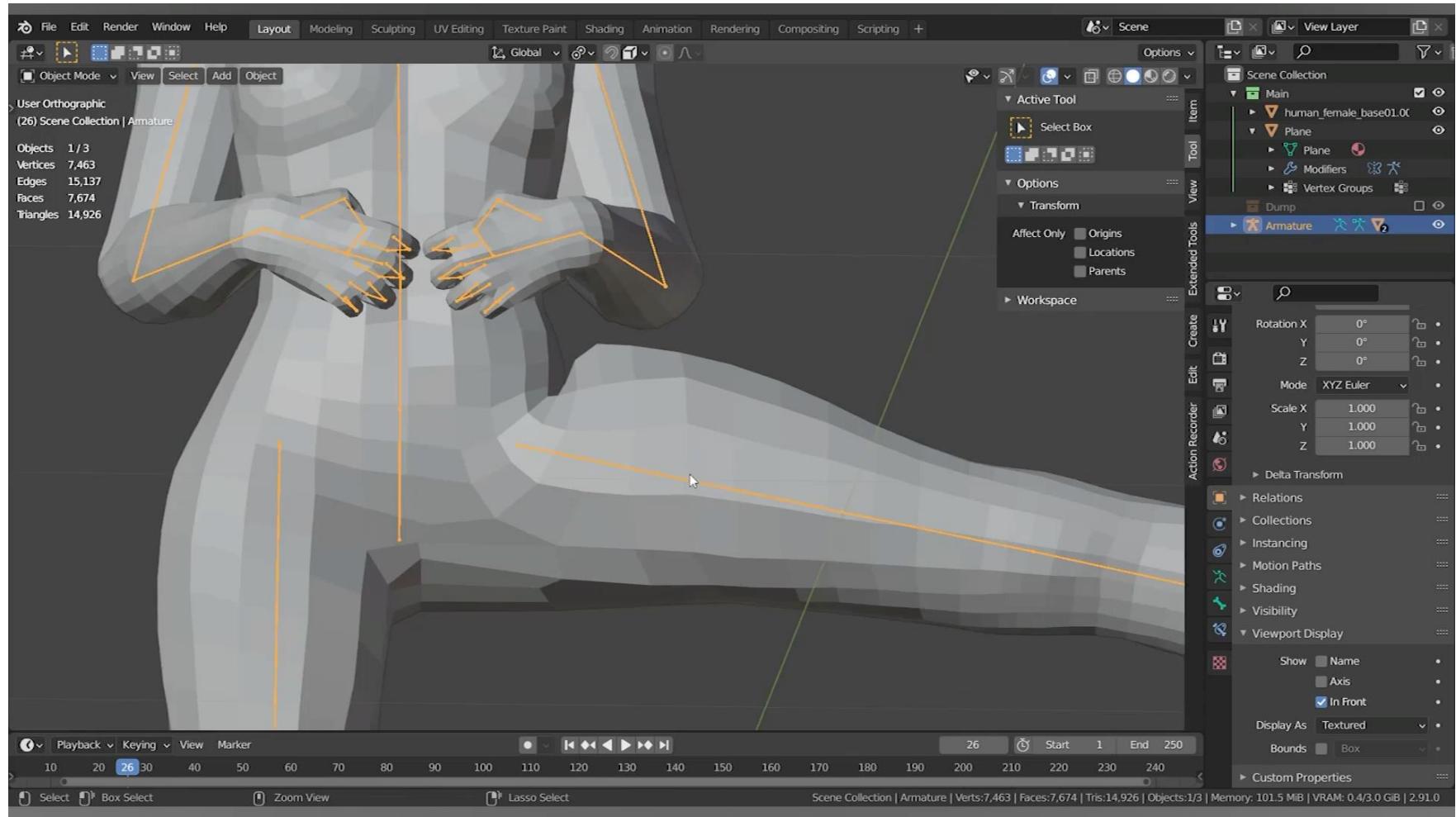
Basic Setup: Skinning Weights



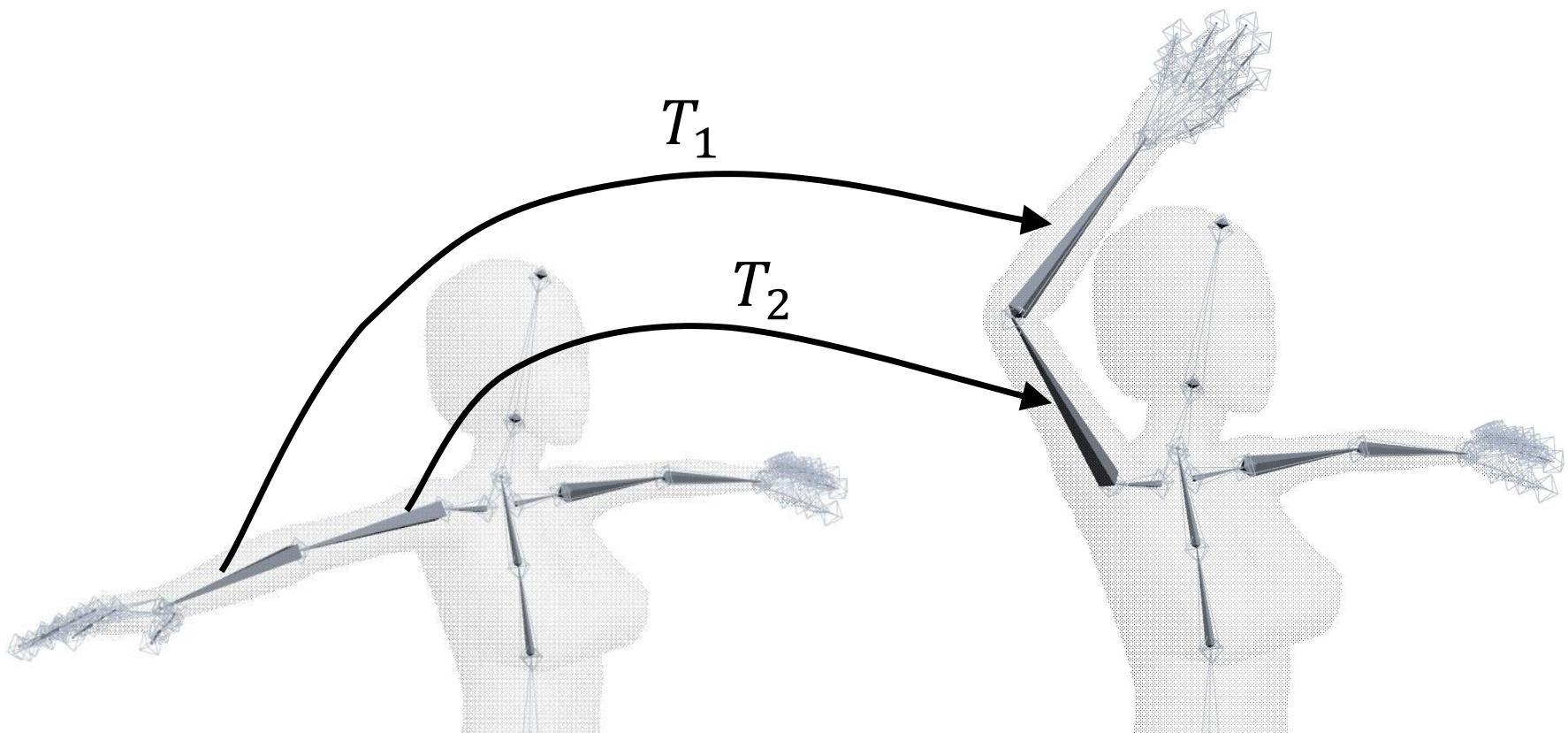
Basic Setup: Skinning Weights



Basic Setup: Skinning Weights

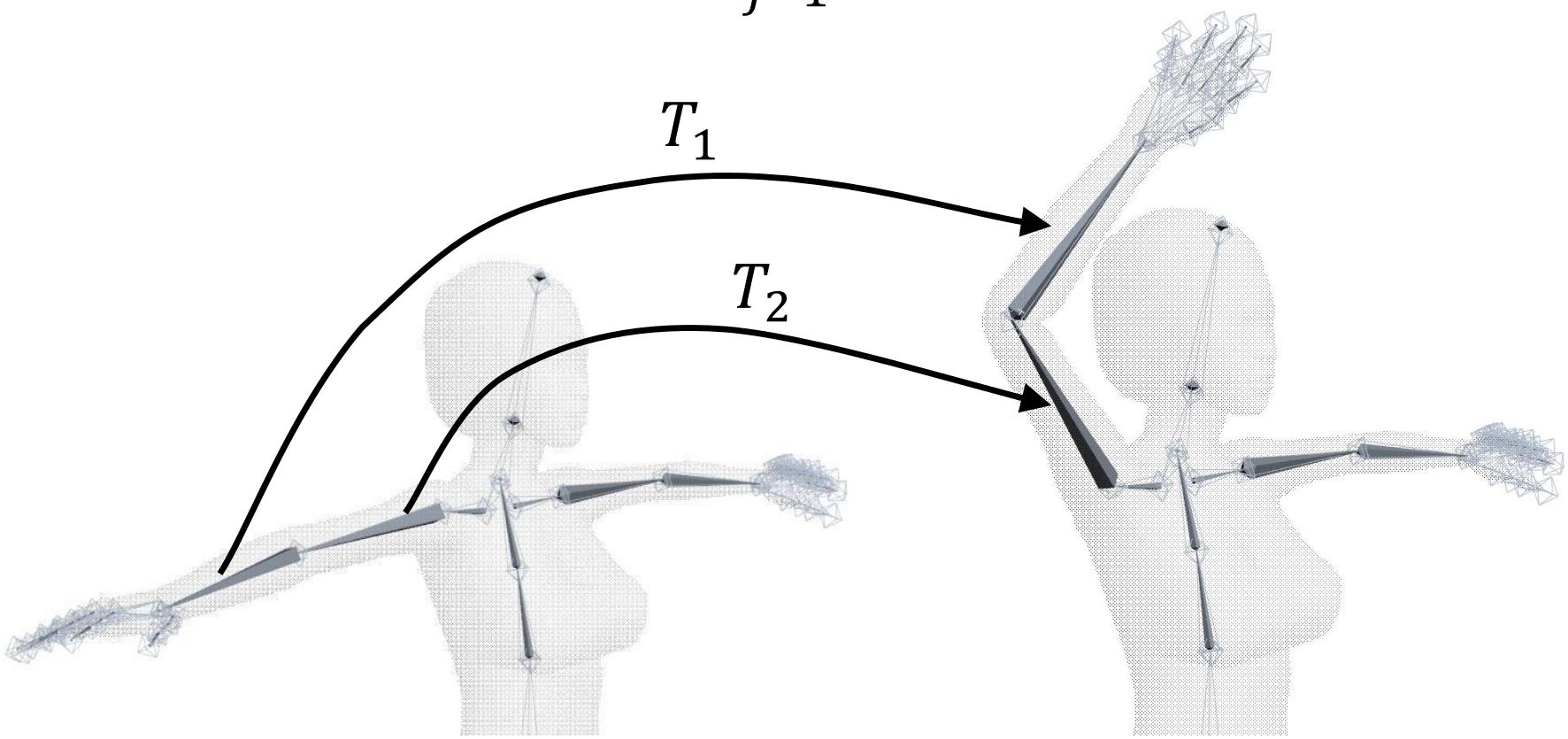


Basic Setup: Skinning Transformation



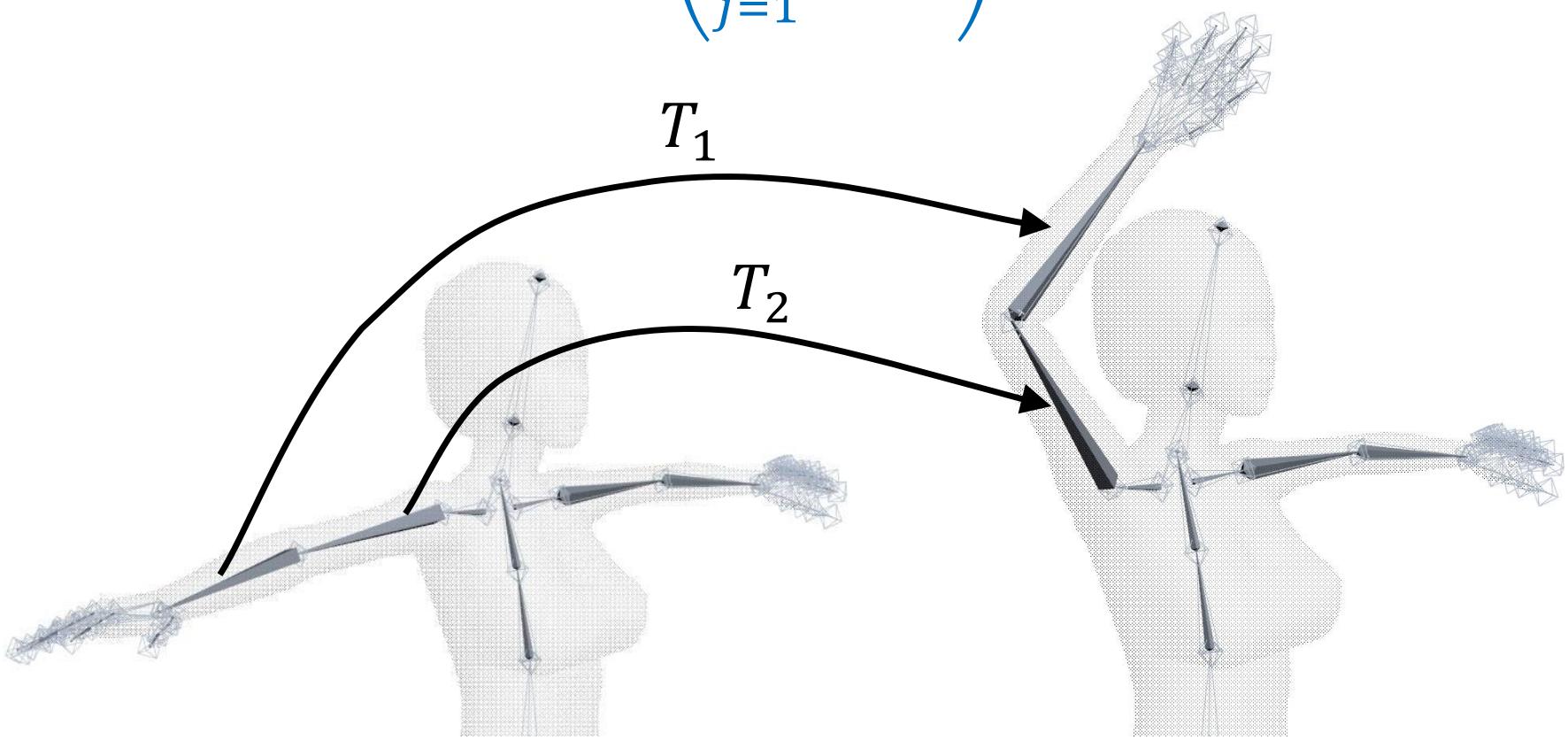
Linear Blend Skinning (LBS)

$$x'_i = \sum_{j=1}^m \omega_{ij} T_j x_i$$



Linear Blend Skinning (LBS)

$$x'_i = \left(\sum_{j=1}^m \omega_{ij} T_j \right) x_i$$



Linear Blend Skinning (LBS)

$$x'_i = \left(\sum_{j=1}^m \omega_{ij} T_j \right) x_i$$

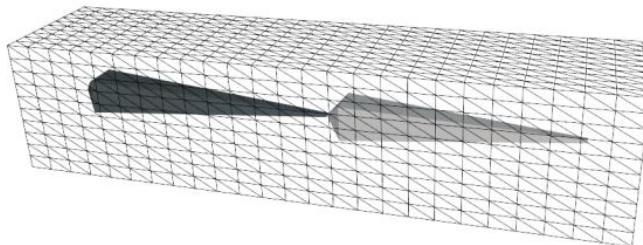
- Used widely in the industry
- Efficient and GPU-friendly
- Games like it

基于GPU约束，优化权重，减少权重的使用（即节点受多少关节的影响）

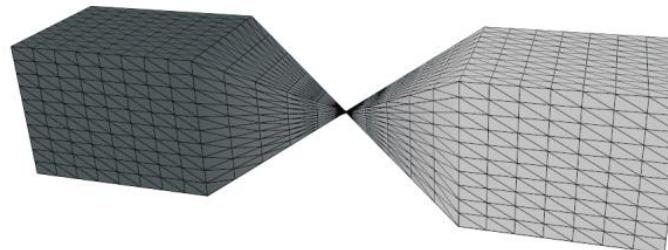


Halo 3

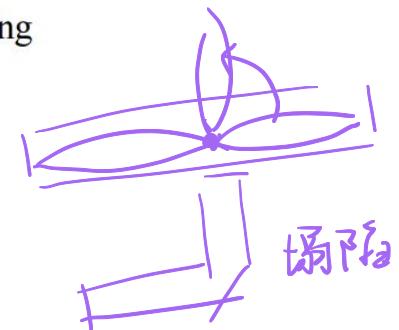
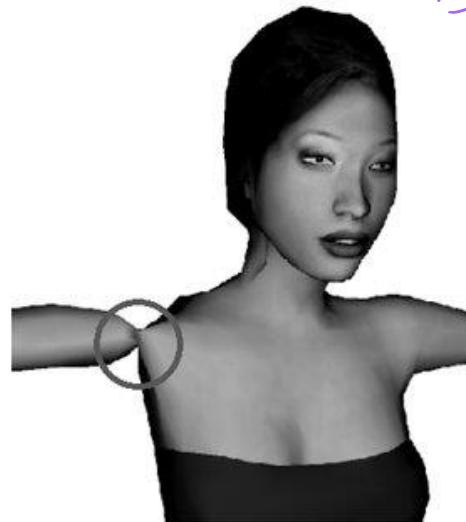
Candy-Wrapper Artifact



Rest pose



Linear blend skinning



Candy-Wrapper Artifact



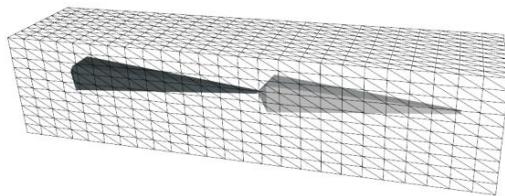
$$x'_i = \left(\sum_{j=1}^m \omega_{ij} T_j \right) x_i$$

Consider $T = (R \quad t)$ and

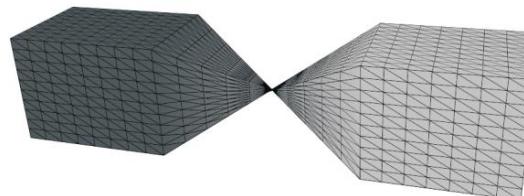
$$\mathbf{R}_1 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad \mathbf{R}_2 = \begin{bmatrix} -1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Advanced skinning methods

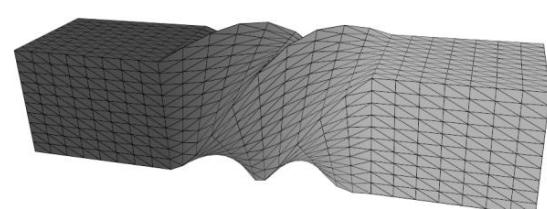
- Multi-linear Skinning
- Nonlinear Skinning
- Dual-quaternion Skinning (DQS)



Rest pose



Linear blend skinning



Dual quaternion skinning

Multi-linear Skinning

- Multi-weight enveloping [Wang and Phillips 2002]
- Animation Space [Merry et al. 2006]

Multi-linear Skinning

$$x' = Tx$$

$$T = [R \mid u] \in \mathbb{R}^{3 \times 4}$$



$$v' = Xt$$

$v' \in \mathbb{R}^{3n}$: stack of all vertices x'

$t \in \mathbb{R}^{12m}$: stack of all transformation T

$X \in \mathbb{R}^{3n \times 12m}$: skinning matrix

Multi-linear Skinning

The structure of matrix \mathbf{X}

$$\mathbf{X} \in R^{3n \times 12m} \quad \mathbf{X} = \begin{bmatrix} \mathbf{X}_{11} & \cdots & \mathbf{X}_{1m} \\ \vdots & \ddots & \vdots \\ \mathbf{X}_{n1} & \cdots & \mathbf{X}_{nm} \end{bmatrix}$$

$$\mathbf{X}_{ij} \in R^{3 \times 12}$$

Multi-linear Skinning

Matrix vectorization

$$\text{vec}(\mathbf{A}) = (a_{1,1}, a_{1,2}, a_{1,3}, \dots, a_{m,n})^T$$

Kronecker product

$$\mathbf{A} \otimes \mathbf{B} = \begin{bmatrix} a_{1,1}\mathbf{B} & \cdots & a_{1,q}\mathbf{B} \\ \vdots & \ddots & \vdots \\ a_{p,1}\mathbf{B} & \cdots & a_{p,q}\mathbf{B} \end{bmatrix} \in \mathbb{R}^{pr \times qs}$$

B r x s

$$\text{vec}(\mathbf{AB}) = (\mathbf{I}_s \otimes \mathbf{A}) \underset{m \times n}{\text{vec}(\mathbf{B})} = (\mathbf{B}^T \otimes \mathbf{I}_p) \text{vec}(\mathbf{A})$$

Multi-linear Skinning

$$x'_i = \sum_{j=1}^m \omega_{ij} T_j x_i$$



$$x'_i = \text{vec}(x'_i) = \sum_{j=1}^m \text{vec}(T_j \omega_{ij} x_i) = \sum_{j=1}^m (\omega_{ij} x_i \otimes I) \text{vec}(T_j)$$

$$= \sum_{j=1}^m X_{ij}^{LBS} t_j = X_i^{LBS} t$$

LBS in matrix form

$$X_{ij}^{LBS} = [w_{ij}x_{i1}\mathbf{I}_3 \quad w_{ij}x_{i2}\mathbf{I}_3 \quad w_{ij}x_{i3}\mathbf{I}_3 \quad w_{ij}\mathbf{I}_3]$$

$$\mathbf{I}_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad \mathbf{x}_i = \begin{bmatrix} \mathbf{x}_{i1} \\ \mathbf{x}_{i2} \\ \mathbf{x}_{i3} \end{bmatrix}$$

$$\begin{bmatrix} w_{ij}\mathbf{x}_{i,1} & 0 & 0 & w_{ij}\mathbf{x}_{i,2} & 0 & 0 & w_{ij}\mathbf{x}_{i,3} & 0 & 0 & w_{ij} & 0 & 0 \\ 0 & w_{ij}\mathbf{x}_{i,1} & 0 & 0 & w_{ij}\mathbf{x}_{i,2} & 0 & 0 & w_{ij}\mathbf{x}_{i,3} & 0 & 0 & w_{ij} & 0 \\ 0 & 0 & w_{ij}\mathbf{x}_{i,1} & 0 & 0 & w_{ij}\mathbf{x}_{i,2} & 0 & 0 & w_{ij}\mathbf{x}_{i,3} & 0 & 0 & w_{ij} \end{bmatrix}$$

LBS has one weight per vertex/bone pair

Multi-linear Skinning

$$X_{ij} = \begin{vmatrix} * & * & * & * & * & * & * & * & * & * & * & * \\ * & * & * & * & * & * & * & * & * & * & * & * \\ * & * & * & * & * & * & * & * & * & * & * & * \end{vmatrix} \in \mathbb{R}^{3 \times 12}$$

We can have 36 weights per vertex/bone pair in the most general form

Multi-weight Enveloping [Wang and Phillips 2002]

$X_{ij}^{MWE} =$

$$\begin{bmatrix} w_{i,j}^1 x_{i,1} & 0 & 0 & w_{i,j}^4 x_{i,2} & 0 & 0 & w_{i,j}^7 x_{i,3} & 0 & 0 & w_{i,j}^{10} & 0 & 0 \\ 0 & w_{i,j}^2 x_{i,1} & 0 & 0 & w_{i,j}^5 x_{i,2} & 0 & 0 & w_{i,j}^8 x_{i,3} & 0 & 0 & w_{i,j}^{11} & 0 \\ 0 & 0 & w_{i,j}^3 x_{i,1} & 0 & 0 & w_{i,j}^6 x_{i,2} & 0 & 0 & w_{i,j}^9 x_{i,3} & 0 & 0 & w_{i,j}^{12} \end{bmatrix}$$

12 weights per vertex/bone

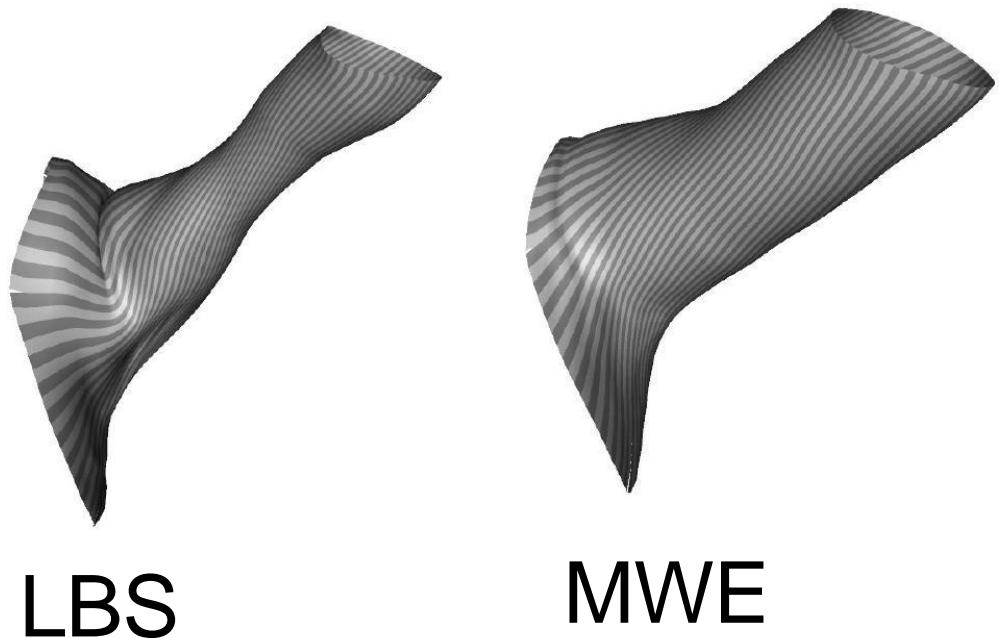
通过样例坐标值进行优化



Multi-weight Enveloping [Wang and Phillips 2002]

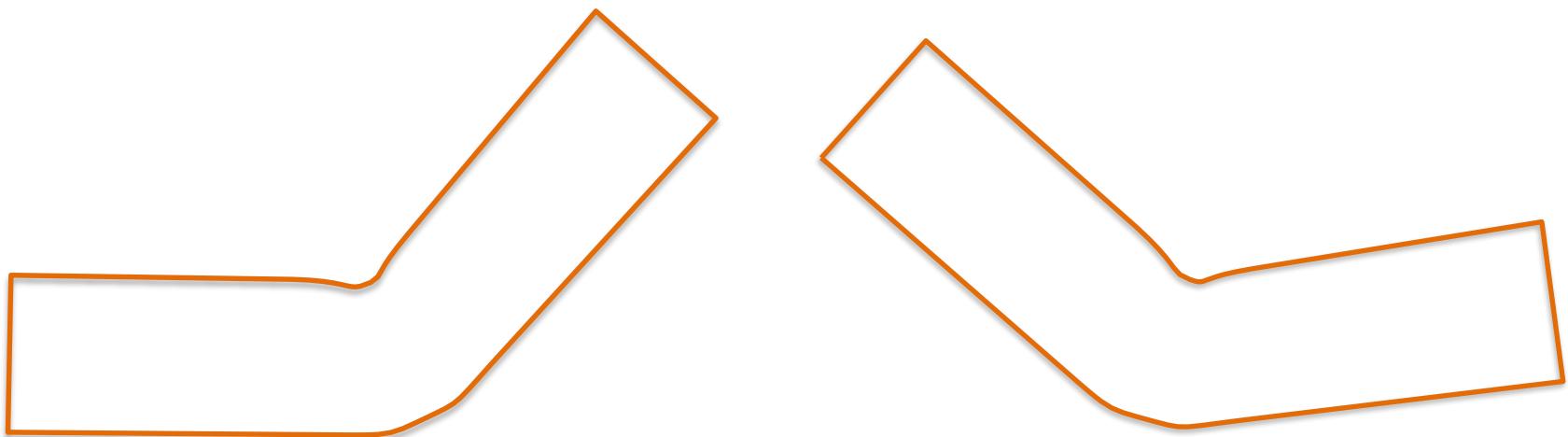
More powerful

Weights trained
from examples



Animation Space [Merry et al. 2006]

World-space rotation invariance.



不满足旋转不变性，对应的T不同，外形会发生改变

Animation Space [Merry et al. 2006]

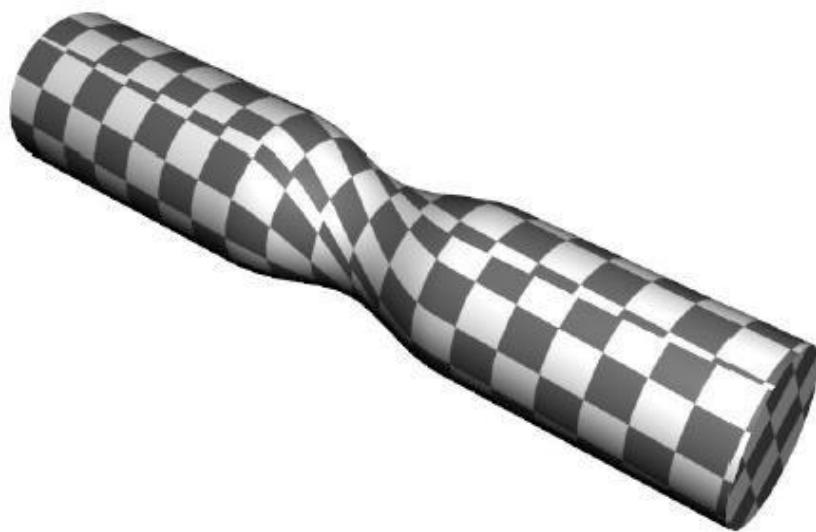
Rotation-invariant results in 4 weights per vertex/bone pair

每个block的weight相同，一共四个

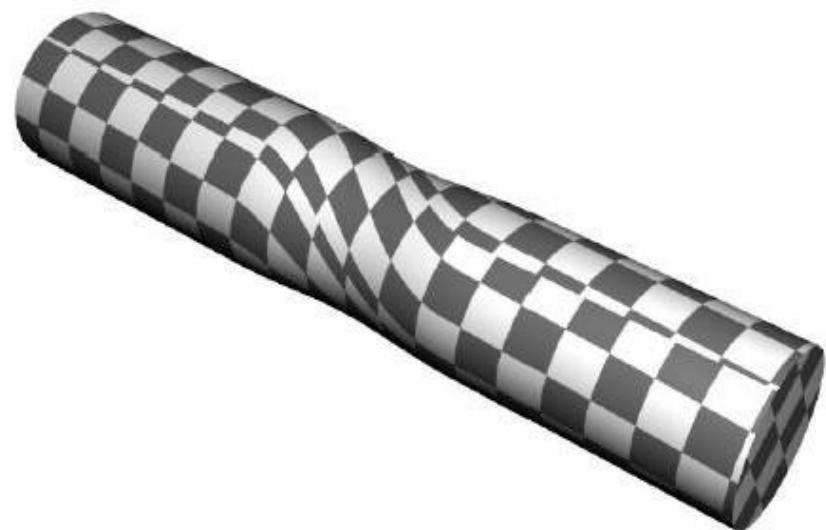
$$X_{ij}^{LBS} = [y_{ij1}\mathbf{I}_3 \quad y_{ij2}\mathbf{I}_3 \quad y_{ij3}\mathbf{I}_3 \quad y_{ij4}\mathbf{I}_3]$$

$$\sum_{j=1}^m y_{ij4} = 1$$

Animation Space [Merry et al. 2006]



LBS



Animation space

Non-linear Skinning

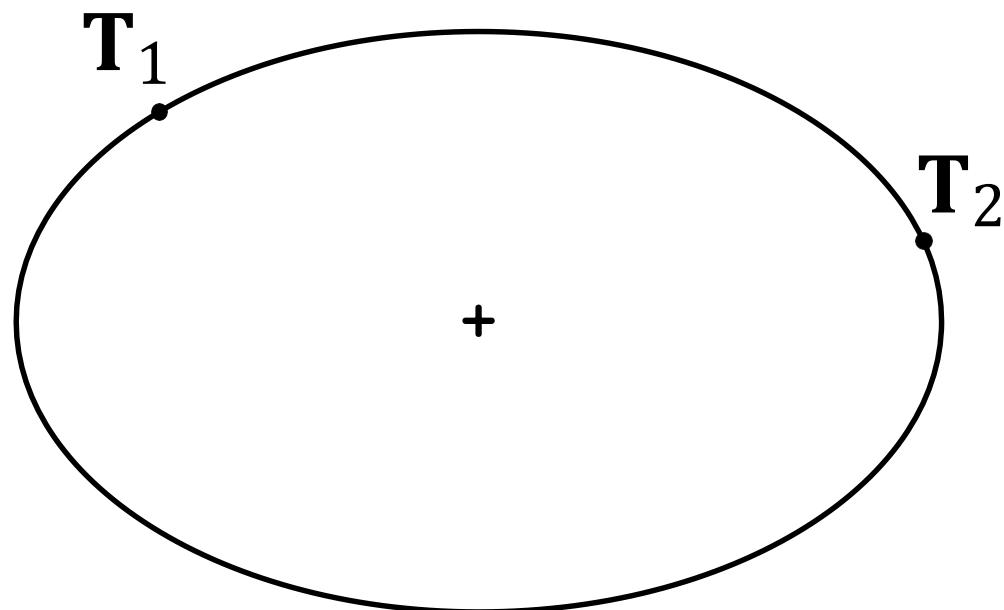
$$x'_i = \left(\sum_{j=1}^m \omega_{ij} T_j \right) x_i$$

$T_j \in SE(3)$: 刚性变换群

$R \in SO(3)$: 旋转变换群

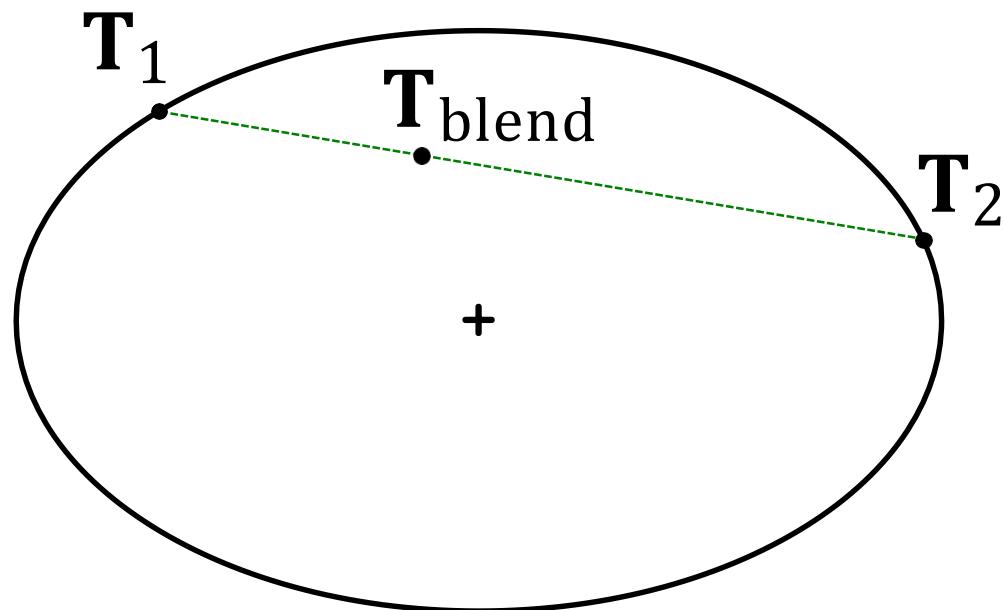
Interpolation in SE(3)

SE(3)



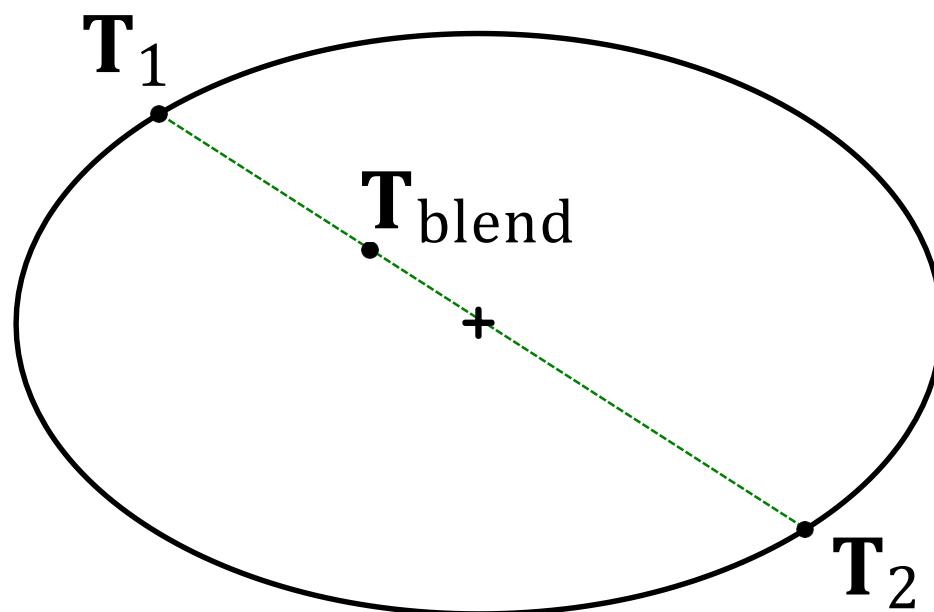
Interpolation in SE(3)

SE(3)

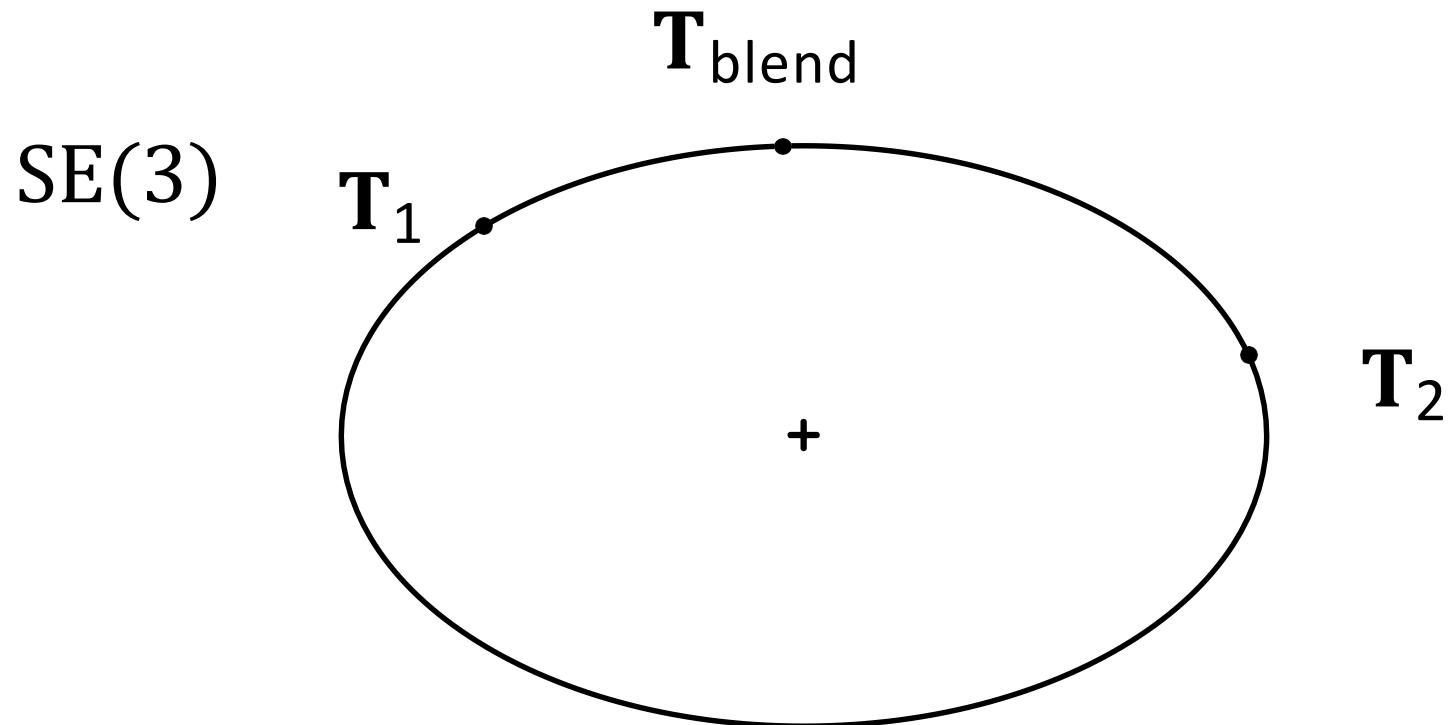


Interpolation in SE(3)

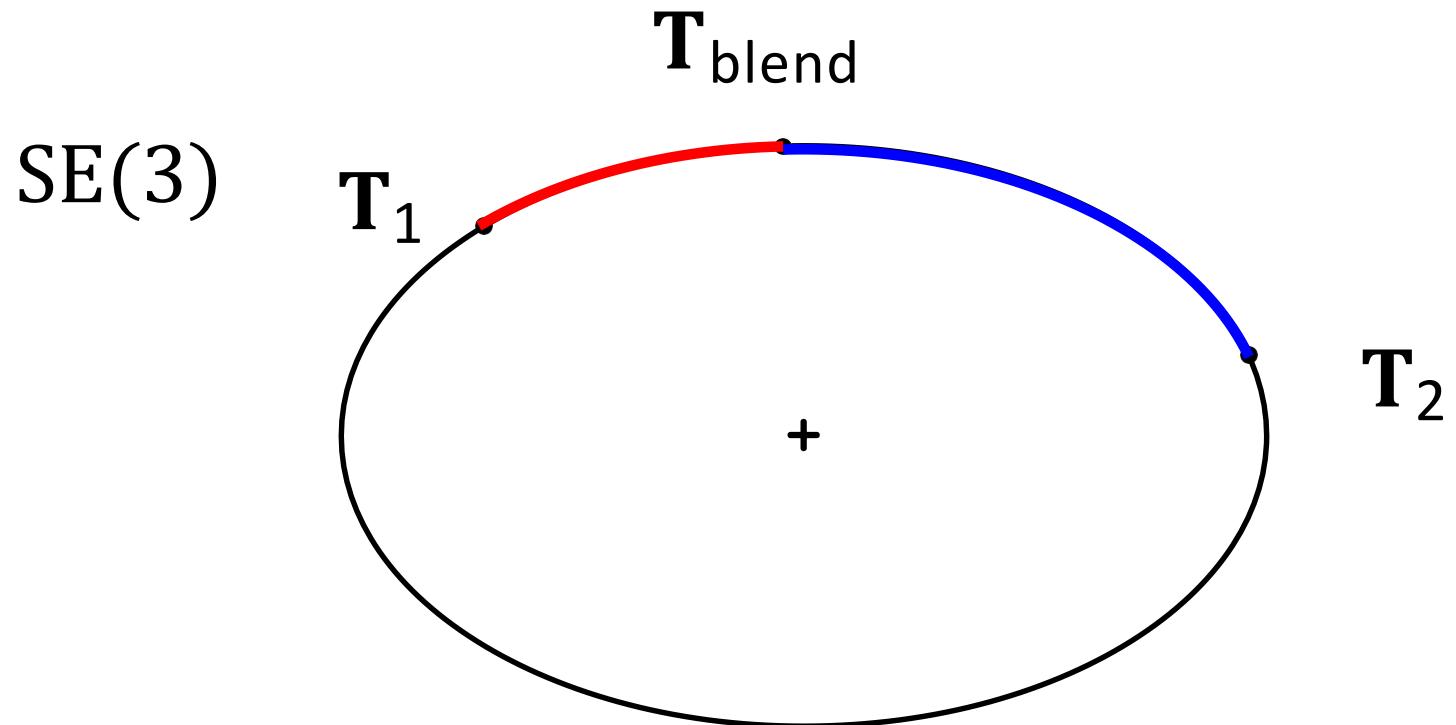
SE(3)



Intrinsic blending

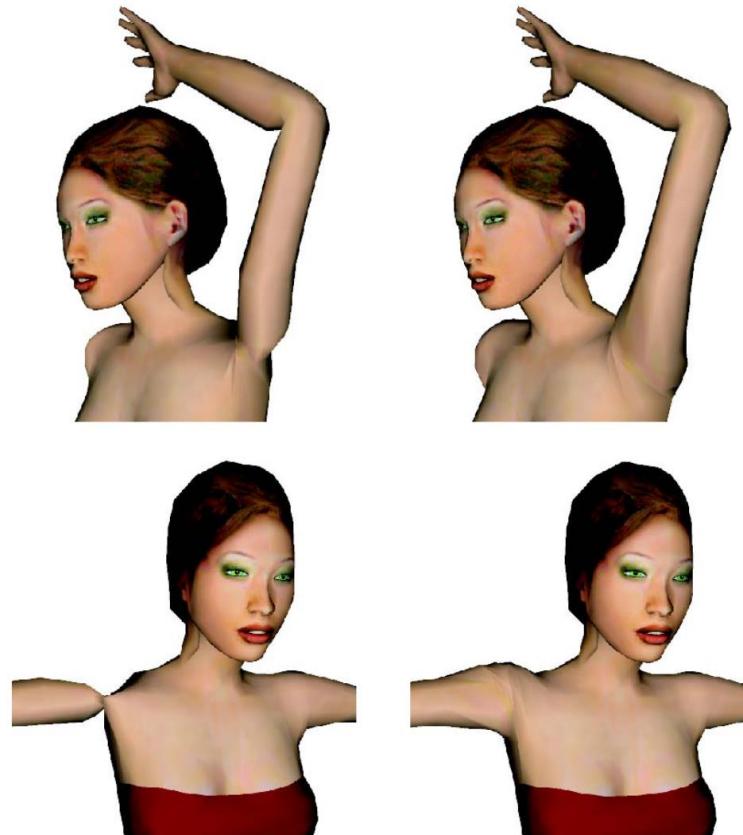


Intrinsic blending



Dual-Quaternion Skinning (DQS)

- Approximation of intrinsic averages in $SE(3)$



Ladislav Kavan, Steven Collins, Jiri Zara, Carol O'Sullivan. ***Geometric Skinning with Approximate Dual Quaternion Blending***, ACM Transaction on Graphics, 27(4), 2008.

Dual-Quaternion

- Dual number

$$x = a + b\epsilon$$

where $\epsilon^2 = 0$

$$y = c + d\epsilon$$

$$x \cdot y = ac + ad\epsilon + bc\epsilon$$

Dual-Quaternion

- Dual quaternion

$$\hat{\mathbf{q}} = \mathbf{q}_0 + \varepsilon \mathbf{q}_\varepsilon \quad \text{8个变量}$$

- Conjugation 三种不同的共轭

$$\hat{\mathbf{q}}^* = \mathbf{q}_0^* + \varepsilon \mathbf{q}_\varepsilon^*$$

① \mathbf{q}^*
② $-\varepsilon$

③ \mathbf{q}^* 同时 $-\varepsilon$

- Norm

$$\|\hat{\mathbf{q}}\| = \sqrt{\hat{\mathbf{q}}^* \hat{\mathbf{q}}} = \|\mathbf{q}_0\| + \varepsilon \frac{\langle \mathbf{q}_0, \mathbf{q}_\varepsilon \rangle}{\|\mathbf{q}_0\|}$$

$$\begin{aligned} & (\mathbf{q}_0 + \varepsilon \mathbf{q}_\varepsilon)(\mathbf{q}_0^* + \varepsilon \mathbf{q}_\varepsilon^*) \\ &= \mathbf{q}_0 \mathbf{q}_0^* + \varepsilon \mathbf{q}_0 \mathbf{q}_\varepsilon^* + \varepsilon \mathbf{q}_\varepsilon \mathbf{q}_0^* + \varepsilon^2 \mathbf{q}_\varepsilon \mathbf{q}_\varepsilon^* \end{aligned}$$

Dual-Quaternion

Scalar Multiplication

$$s\mathbf{q} = s\mathbf{q}_r + s\mathbf{q}_d\varepsilon$$

Addition

$$\mathbf{q}_1 + \mathbf{q}_2 = \mathbf{q}_{r1} + \mathbf{q}_{r2} + (\mathbf{q}_{d1} + \mathbf{q}_{d2})\varepsilon$$

Multiplication

$$\mathbf{q}_1 \mathbf{q}_2 = \mathbf{q}_{r1} \mathbf{q}_{r2} + (\mathbf{q}_{r1} \mathbf{q}_{d2} + \mathbf{q}_{d1} \mathbf{q}_{r2})\varepsilon$$

Dual-Quaternion

- Norm

$$\|\hat{\mathbf{q}}\| = \sqrt{\hat{\mathbf{q}}^* \hat{\mathbf{q}}} = \|\mathbf{q}_0\| + \varepsilon \frac{\langle \mathbf{q}_0, \mathbf{q}_\varepsilon \rangle}{\|\mathbf{q}_0\|}$$

- Unit dual quaternion: $\|\hat{\mathbf{q}}\| = 1$

- $\|\mathbf{q}_0\| = 1$
- $\mathbf{q}_0 \cdot \mathbf{q}_\varepsilon = 0$

Dual-Quaternion

- Like quaternion, any rigid transformation $T \in SE(3)$ can be converted into a dual-quaternion

$$T = [R \mid t] \rightarrow \hat{\mathbf{q}} = \mathbf{q}_0 + \varepsilon \mathbf{q}_\varepsilon$$

$$\mathbf{q}_0 = \overset{R}{\textcolor{purple}{\downarrow}} \mathbf{r}$$

$$\mathbf{q}_\varepsilon = \frac{1}{2} \mathbf{t} \mathbf{r}$$

Dual-Quaternion

- Transform a vector using unit dual quaternion

$$\hat{v}' = \hat{q} \hat{v} \hat{q}^* \rightarrow \text{是第3种方法}$$

Where

$$\hat{v} = 1 + \varepsilon(0, v)$$

$$\bar{v} = [1, 0, 0, 0, 0, 1, y, z]$$

~~~~~  
↓  
虚数部分

# Dual-Quaternion

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- Like quaternion, any rigid transformation  $T \in SE(3)$  can be converted into a dual-quaternion

$$Tx = Rx + t$$

$$T = [R \mid t] \rightarrow \hat{q} = q_0 + \varepsilon q_\varepsilon$$

$$q_0 = r$$

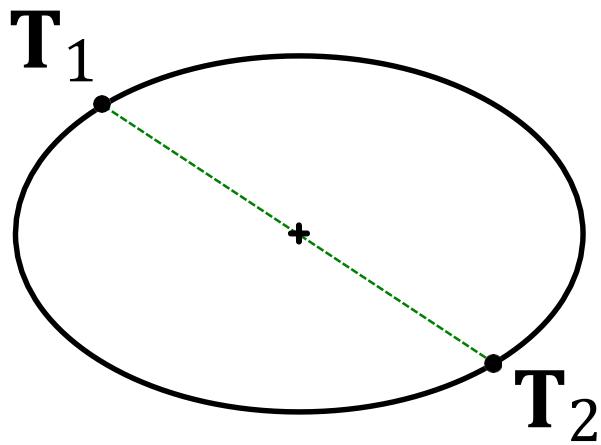
$$q_\varepsilon = \frac{1}{2} tr \quad t = (0, t)$$

$\hat{q}$  and  $-\hat{q}$  represent the same transformation

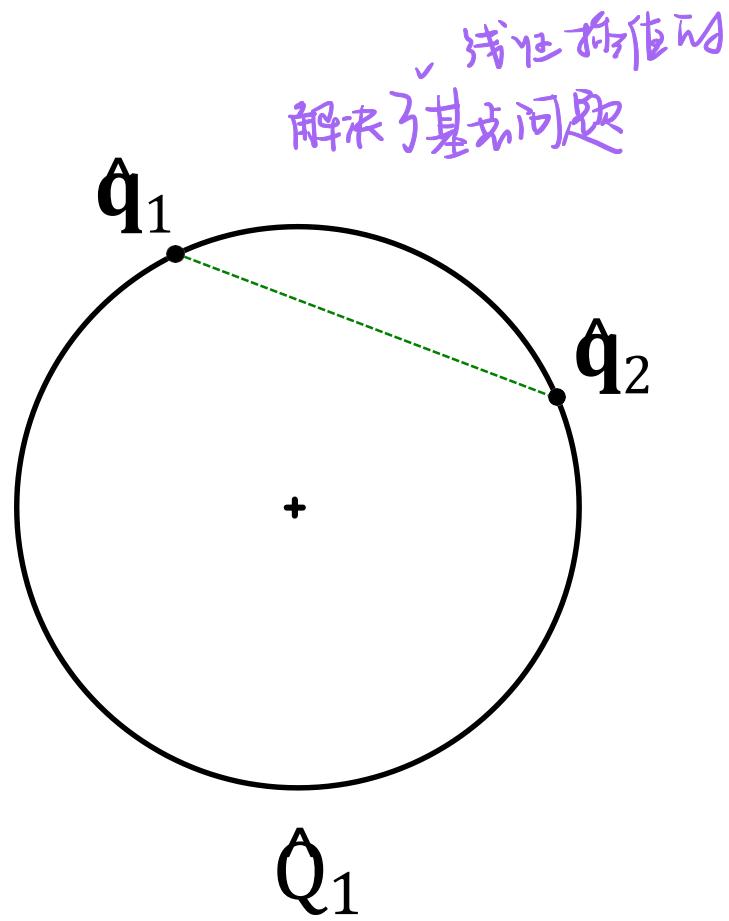
$\hat{Q}_1$  is a double cover of  $SE(3)$

# Double cover visualized

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$\text{SE}(3)$



$\hat{\mathbf{Q}}_1$

# Interpolating Dual-Quaternion

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$$\hat{q} = (1 - t)\hat{q}_0 + t\hat{q}_1$$



$$\hat{q} = \frac{(1 - t)\hat{q}_0 + t\hat{q}_1}{\|(1 - t)\hat{q}_0 + t\hat{q}_1\|}$$

# Dual-Quaternion Skinning

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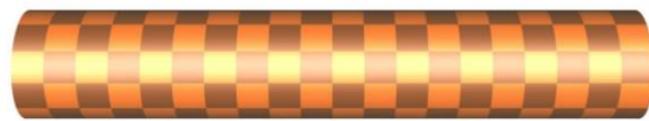
$$\hat{q} = (1 - t)\hat{q}_0 + t\hat{q}_1$$



$$\hat{q} = \frac{(1 - t)\hat{q}_0 + t\hat{q}_1}{\|(1 - t)\hat{q}_0 + t\hat{q}_1\|}$$

# Dual-Quaternion Skinning

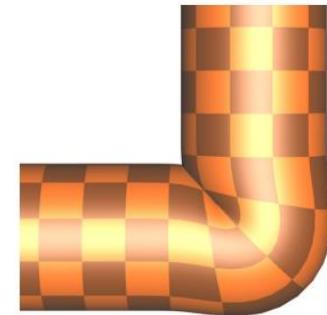
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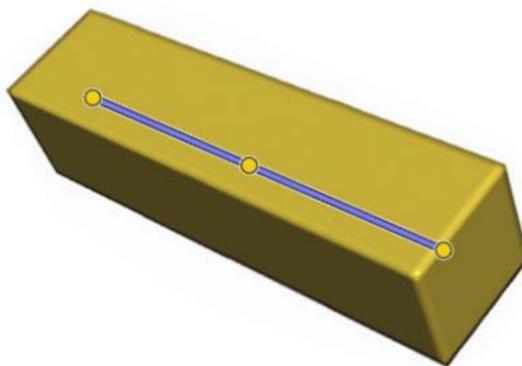
Rest pose



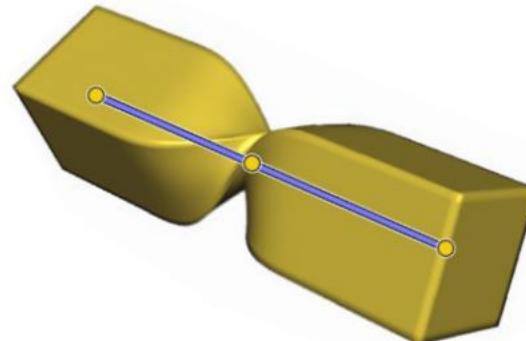
Dual quaternions: twist



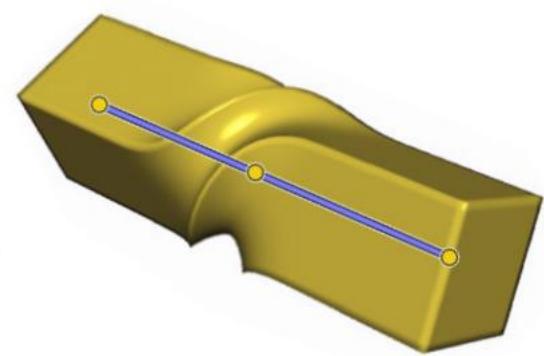
Dual quaternions: bend



Rest pose



Linear blend skinning



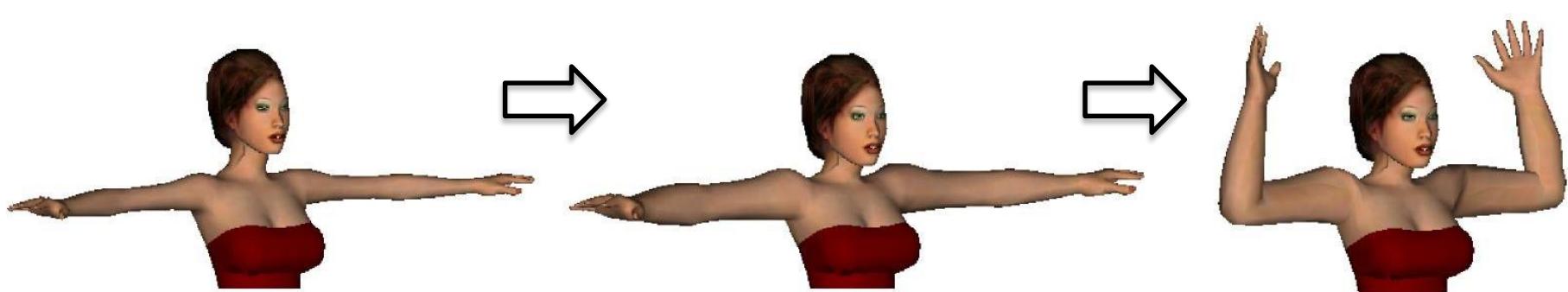
Dual quaternion skinning

# Non-rigid transformations

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1. Scaling

2. Rotations



# Two-phase skinning

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Enhanced for production use by Disney

[Lee et al. 2013]



# Uncovered Topics

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- Deformers
- Skinning weight reduction and compression
- Skeleton Extraction

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有没有问题？

Any Questions