

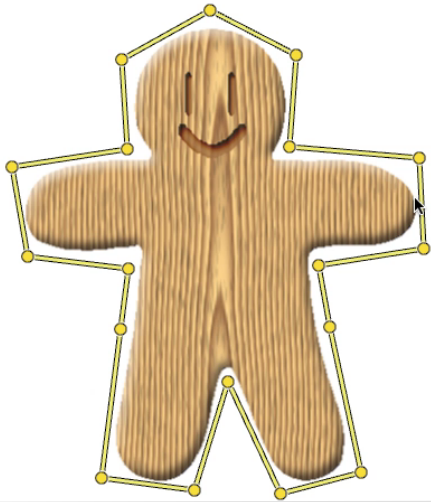
High-quality weight functions via constrained optimization

Alec Jacobson

ETH Zurich

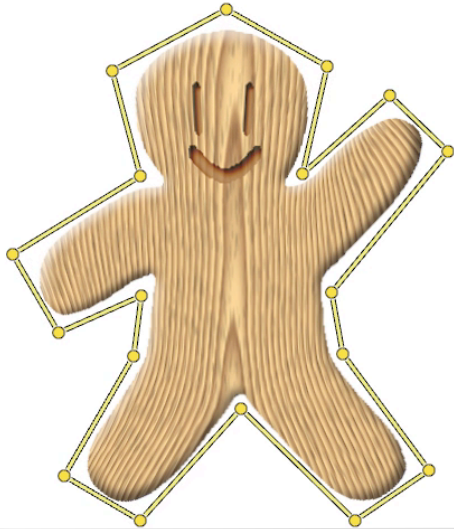
Coordinate-based deformation is a special case of Linear Blend Skinning

$$\mathbf{x}' = \sum_{j \in H} w_j(\mathbf{x}) \mathbf{h}'_j$$



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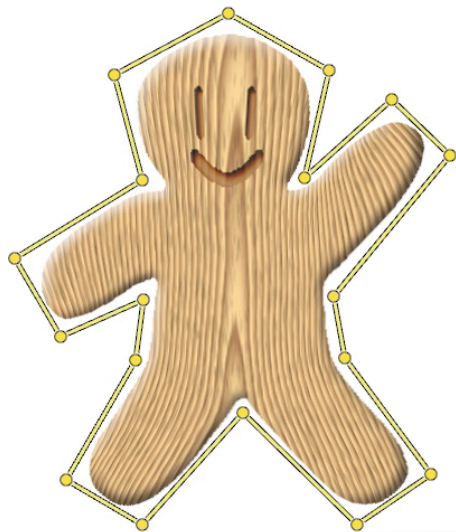
linear precision (reproduction)

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Coordinate-based deformation is a special case of Linear Blend Skinning

$$\mathbf{x}' = \sum_{j \in H} w_j(\mathbf{x}) T_j \mathbf{x}$$

restricted to translation



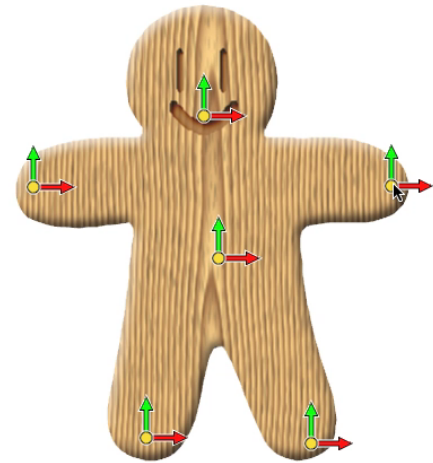
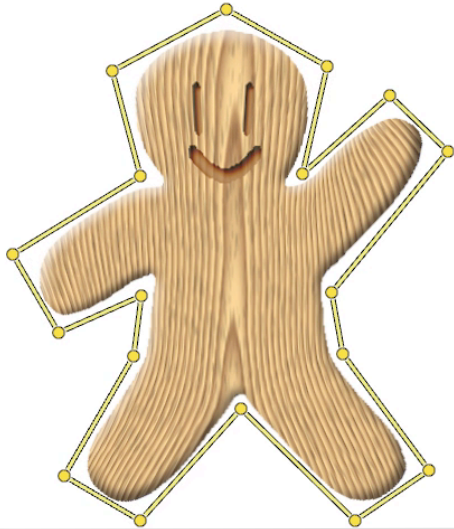
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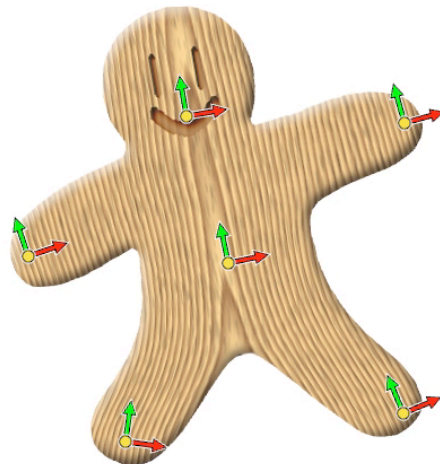
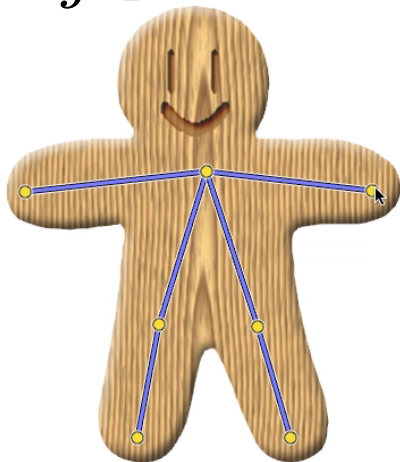
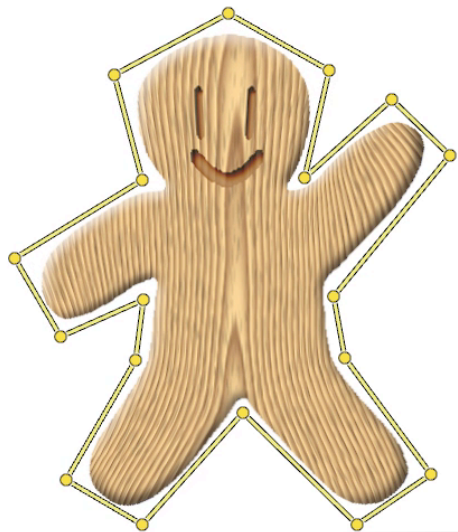
$$\mathbf{x}' = \sum_{j \in H} w_j(\mathbf{x}) T_j \mathbf{x}$$

any affine



Coordinate-based deformation is a special case of Linear Blend Skinning

$$\mathbf{x}' = \sum_{j \in H} w_j(\mathbf{x}) T_j \mathbf{x}$$



Bounded Biharmonic Weights for Real-Time Deformation

SIGGRAPH 2011

Alec Jacobson^{1,2}

Ilya Baran³

Jovan Popović⁴

Olga Sorkine^{1,2}

¹New York University, ²ETH Zurich

³Disney Research Zurich

⁴Adobe Systems



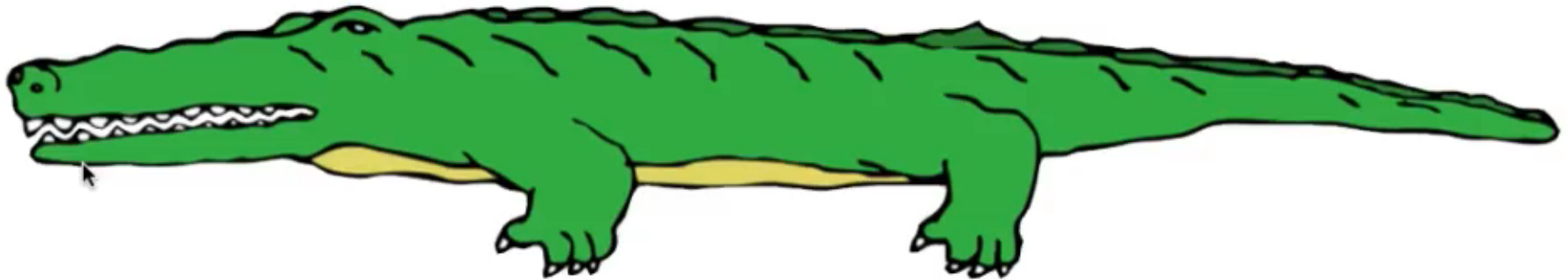
INTERACTIVE GEOMETRY LAB

August 1, 2012

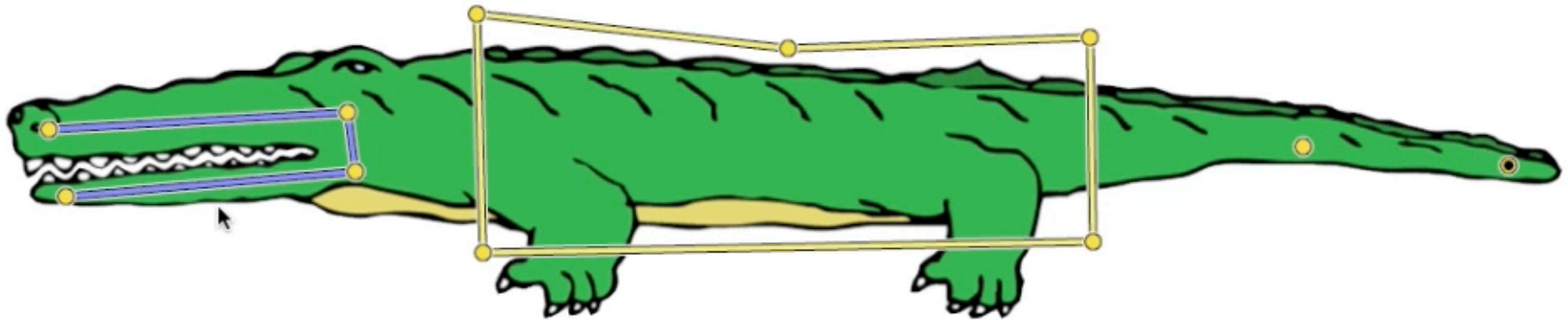
ETH

Eidgenössische Technische Hochschule Zürich
Swiss Federal Institute of Technology Zurich

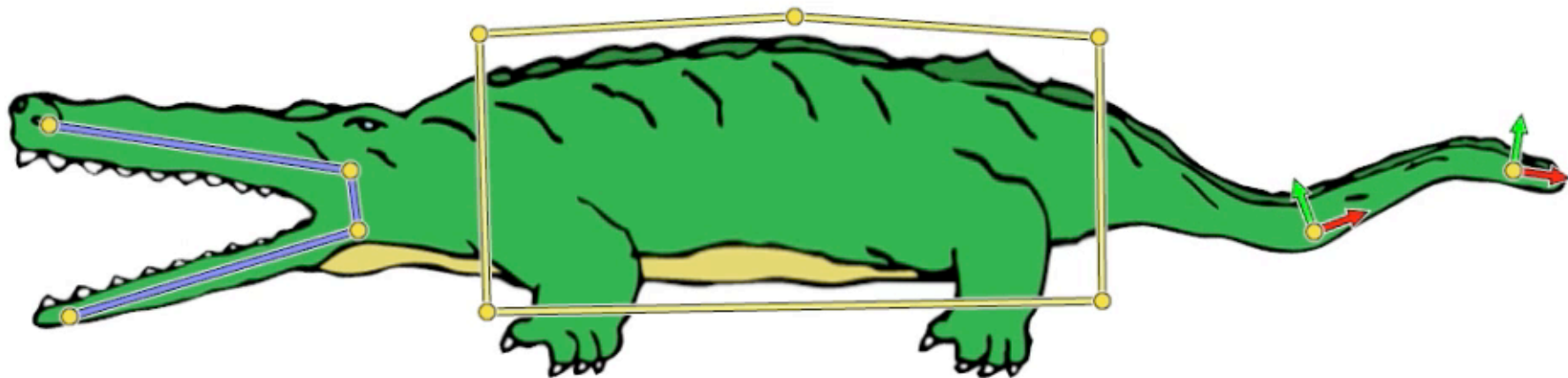
Each handle type has a specific task,
more than just *different modeling metaphor*



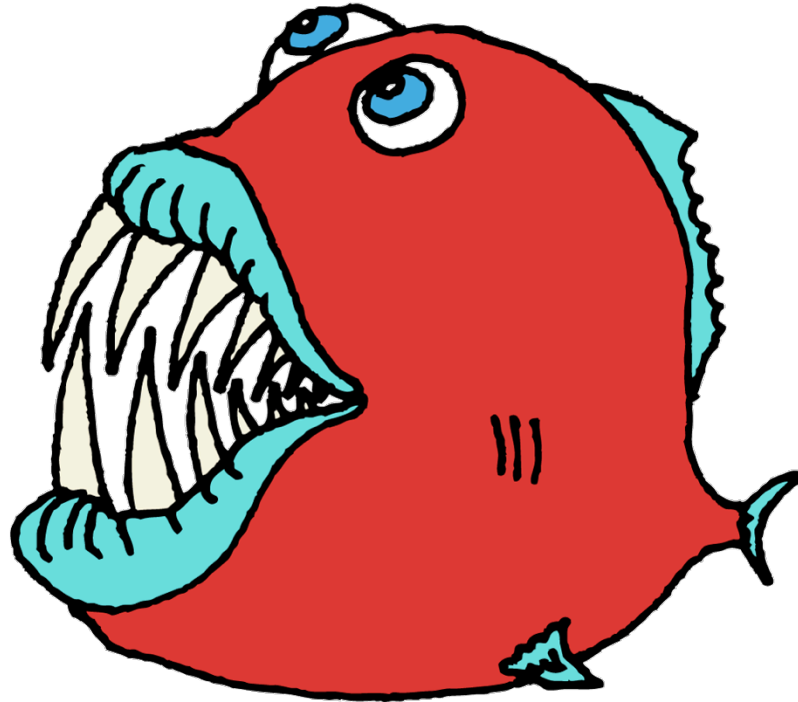
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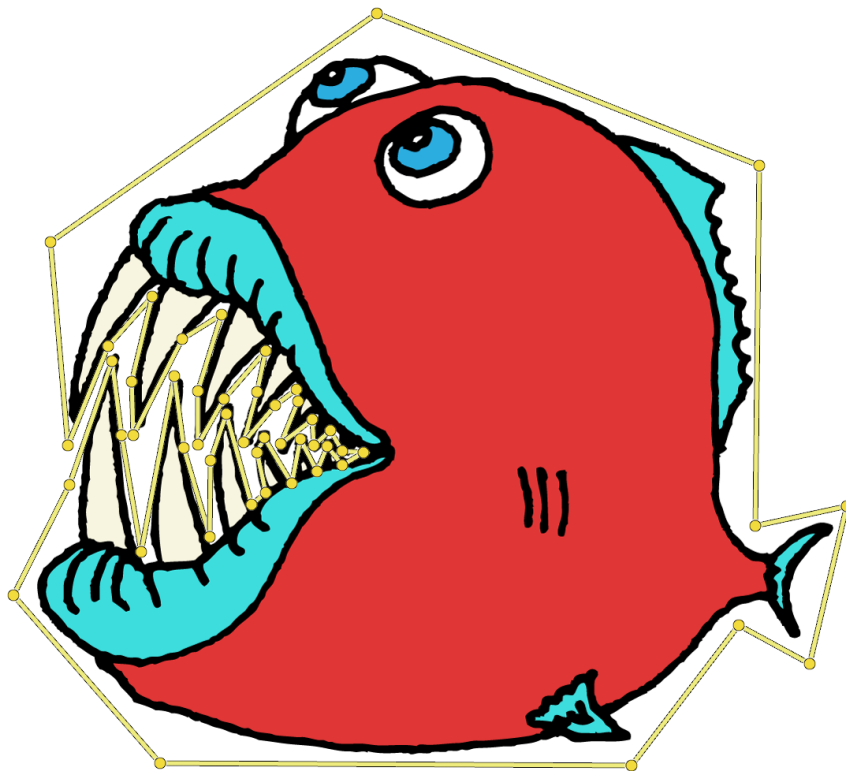
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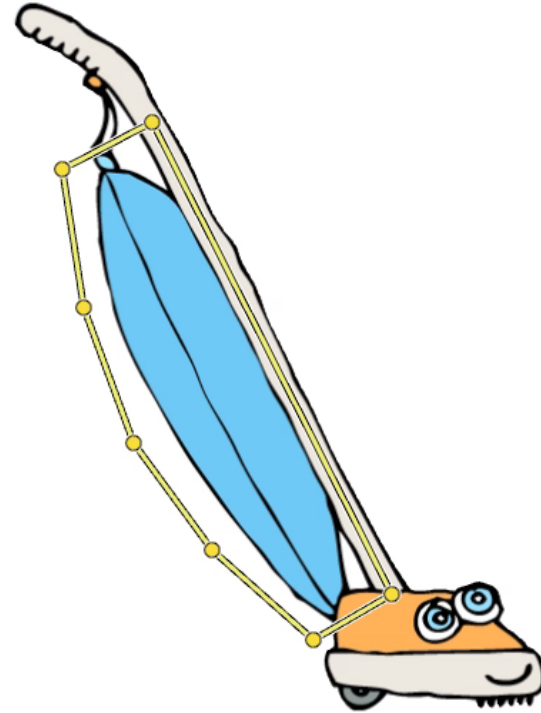
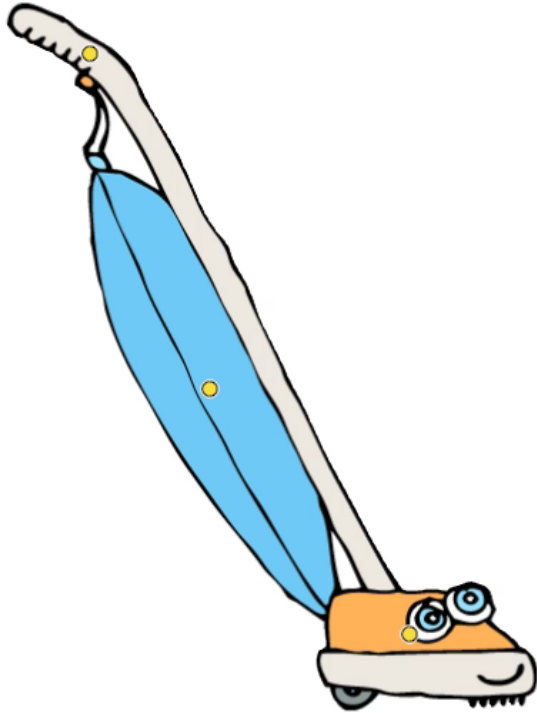
Cages can often be tedious to build and control



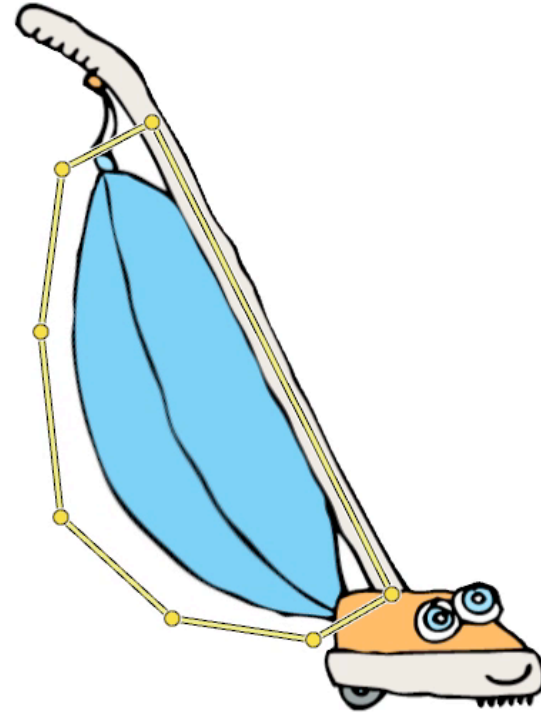
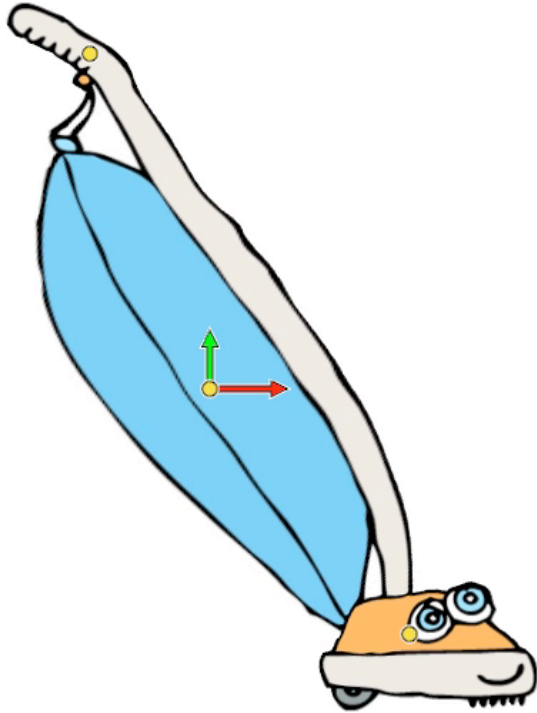
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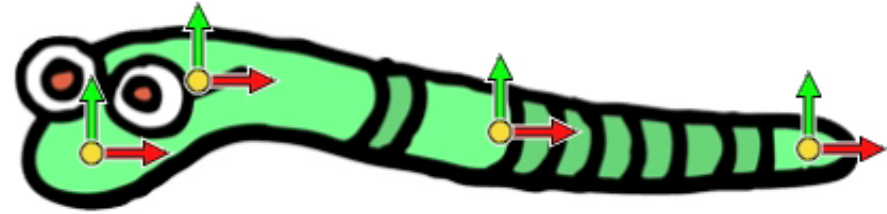
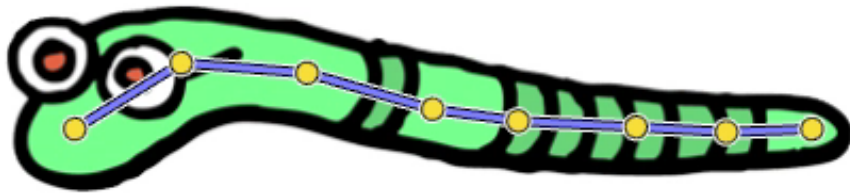
Points can only provide crude scaling



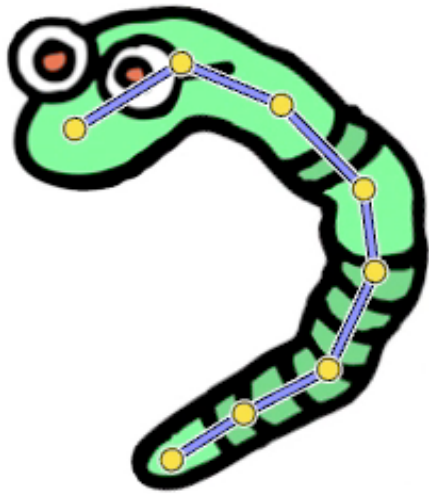
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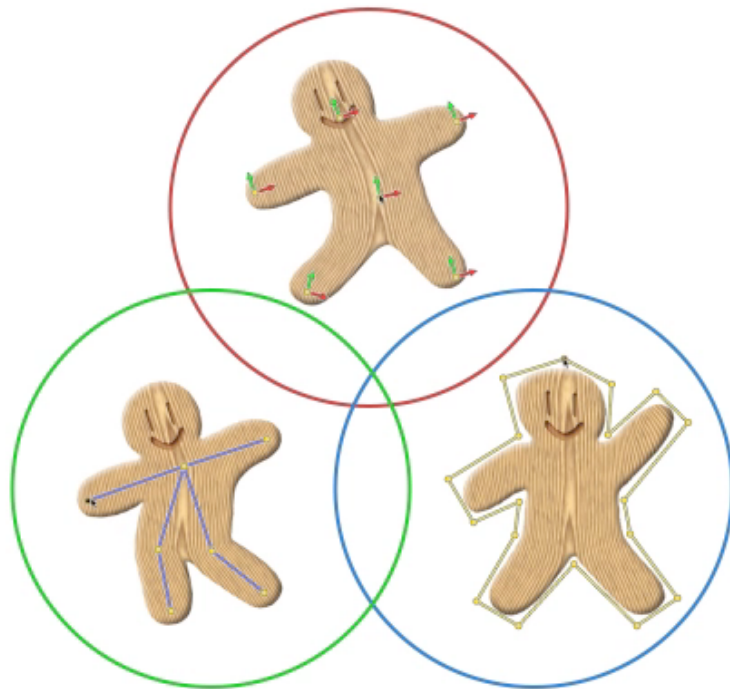
Skeletons may be too rigid or too cumbersome



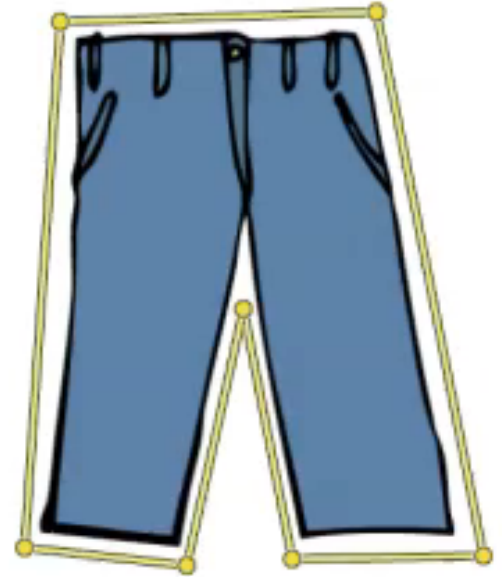
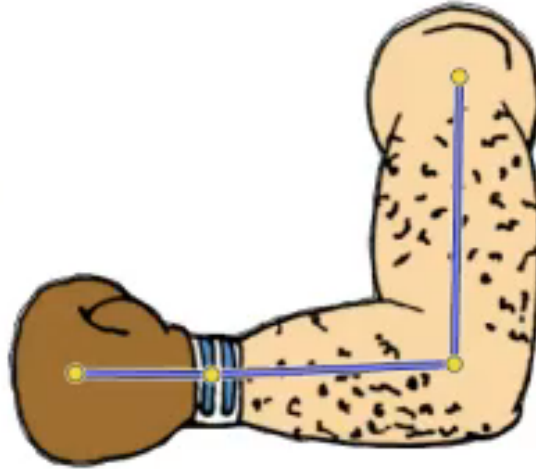
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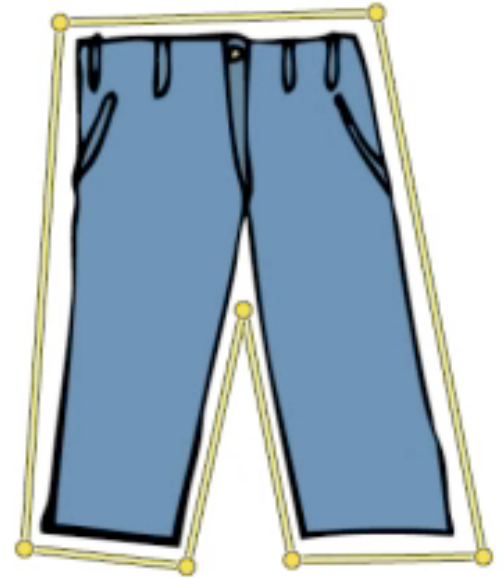
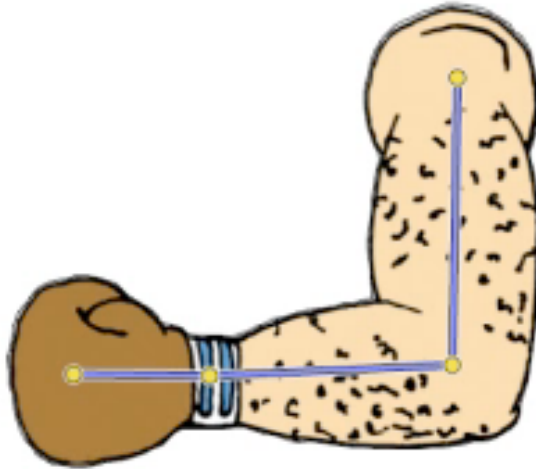
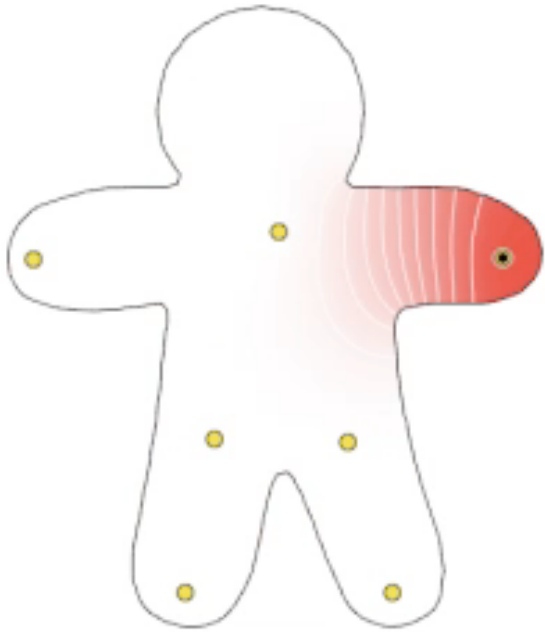
We want to compute weights that unify points, skeletons and cages



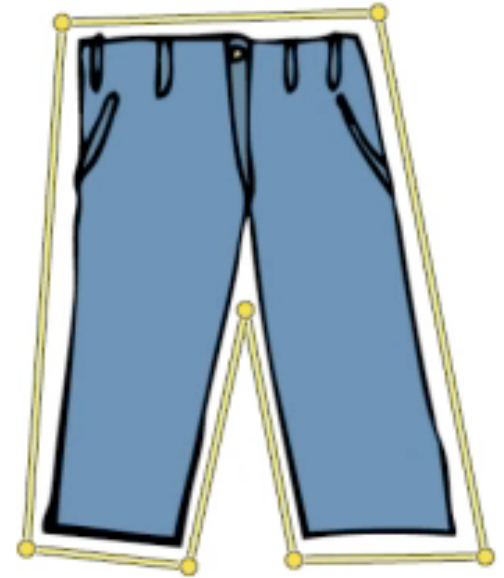
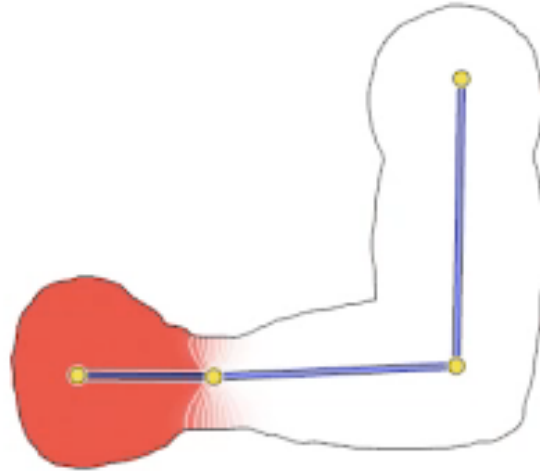
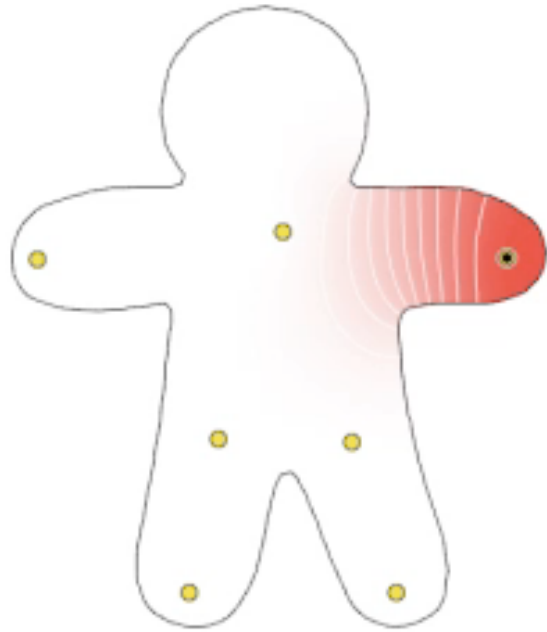
Weights should be smooth, shape-aware, positive and *intuitive*



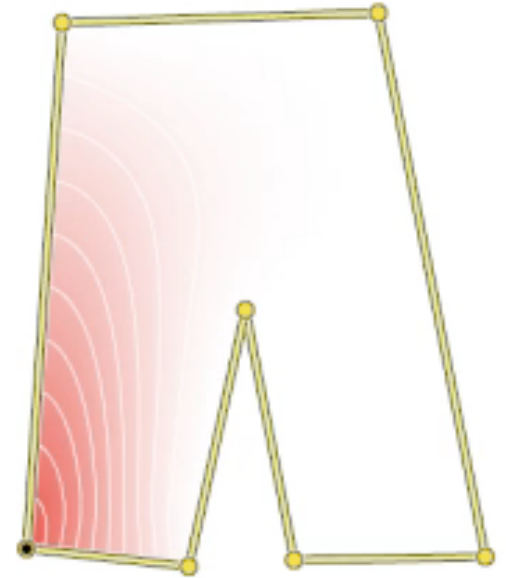
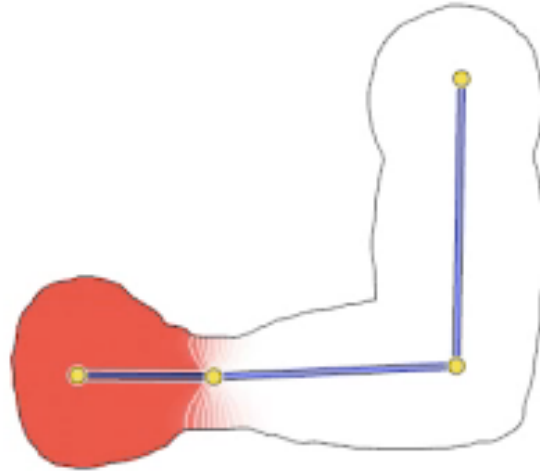
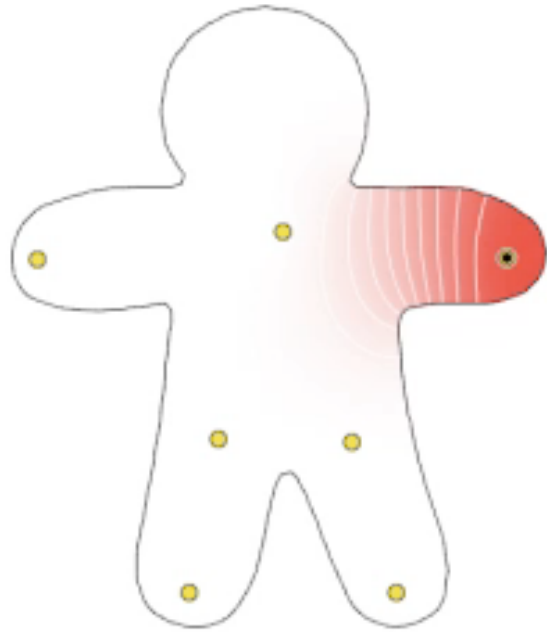
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Weights should be smooth, shape-aware, positive and *intuitive*



Weights must be smooth everywhere, *especially* at handles



Our method



Extension of Harmonic Coordinates
[Joshi et al. 2005]

Weights must be smooth everywhere, *especially* at handles

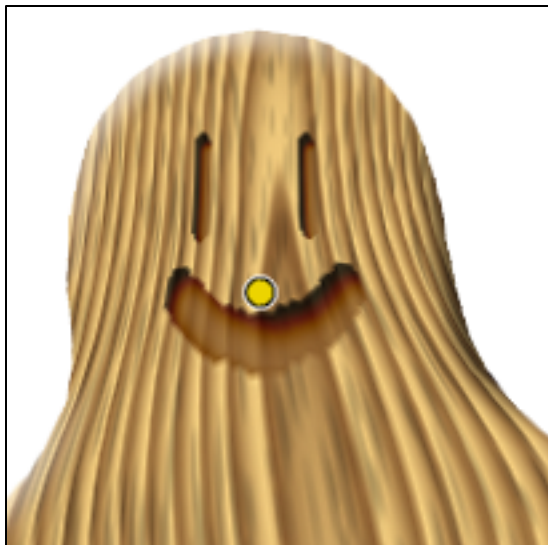


Our method

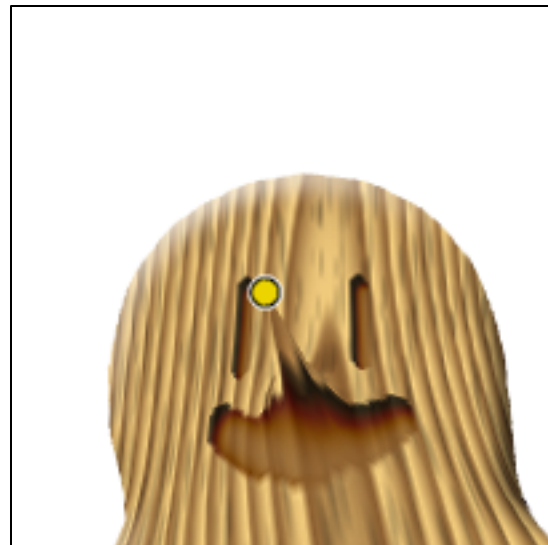


Extension of Harmonic Coordinates
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Extension of Harmonic Coordinates
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Shape-awareness ensures respect of domain's features

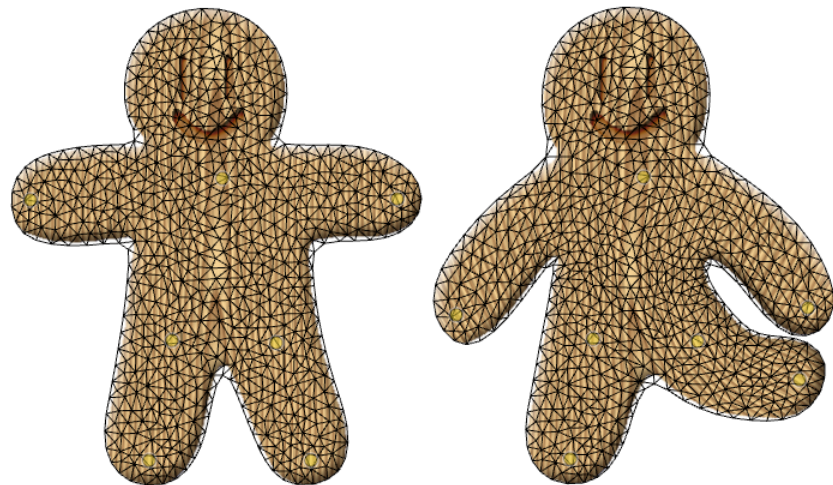


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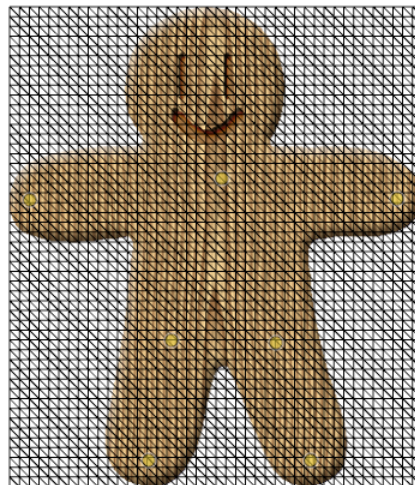


Non-shape-aware methods
e.g. [Schaefer et al. 2006]

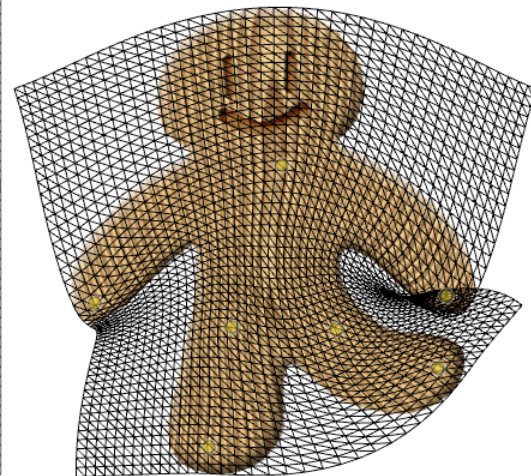
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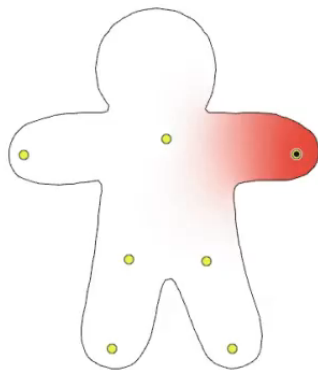


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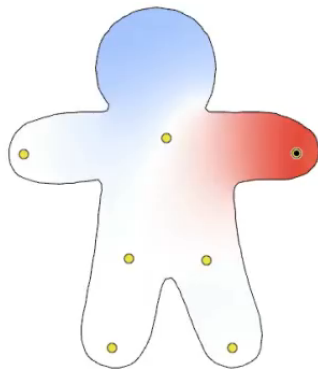


Non-negative weights are mandatory

Our method

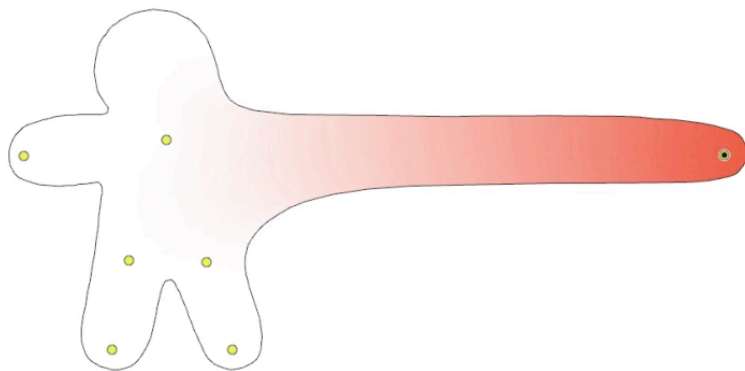


Unconstrained biharmonic
[Botsch and Kobbelt 2004]

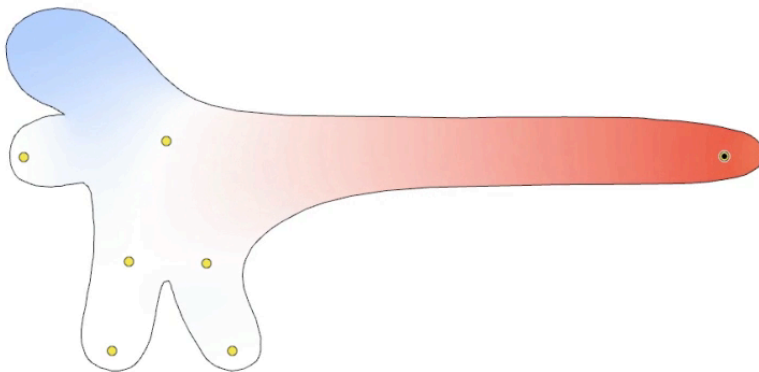


Non-negative weights are mandatory

Our method



Unconstrained biharmonic
[Botsch and Kobbelt 2004]



Weights must maintain other simple, but important properties

$$\sum_{j \in H} w_j(\mathbf{x}) = 1$$

Partition of unity

$$w_j \Big|_{H_k} = \delta_{jk}$$

w_j is linear along cage faces

Interpolation of handles

Weights must maintain other simple, but important properties

$$\sum_{j \in H} w_j(\mathbf{x}) = 1$$

Partition of unity

Set of handles, aka mesh
vertices under handles

$$w_j \Big|_{H_k} = \delta_{jk}$$

Kronecker's delta

w_j is linear along cage faces

Interpolation of handles

Previous methods only partially satisfactory

	Harmonic Coordinates [Joshi et al. 2005]	Unconstrained biharmonic [Botsch and Kobbelt 2004]	Shepard interpolation [Shepard 1968]	Natural neighbors [Sibson 1981]
Smoothness	-	✓	✓	-
Non-negativity	✓	-	✓	✓
Shape-aware	✓	✓	-	-
Locality, sparsity	-	-	-	✓
No local extrema	✓	-	-	✓

$$\Delta w_j = 0$$

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	Harmonic Coordinates [Joshi et al. 2005]	Unconstrained biharmonic [Botsch and Kobbelt 2004]	Shepard interpolation [Shepard 1968]	Natural neighbors [Sibson 1981]
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$$\Delta^2 w_j = 0$$

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Shape-aware	✓	✓	-	-
Locality, sparsity	-	-	-	✓
No local extrema	✓	-	-	✓

Inverse distance,
weighted least-squares

Inverse distance methods inherently suffer from *fall-off effect*



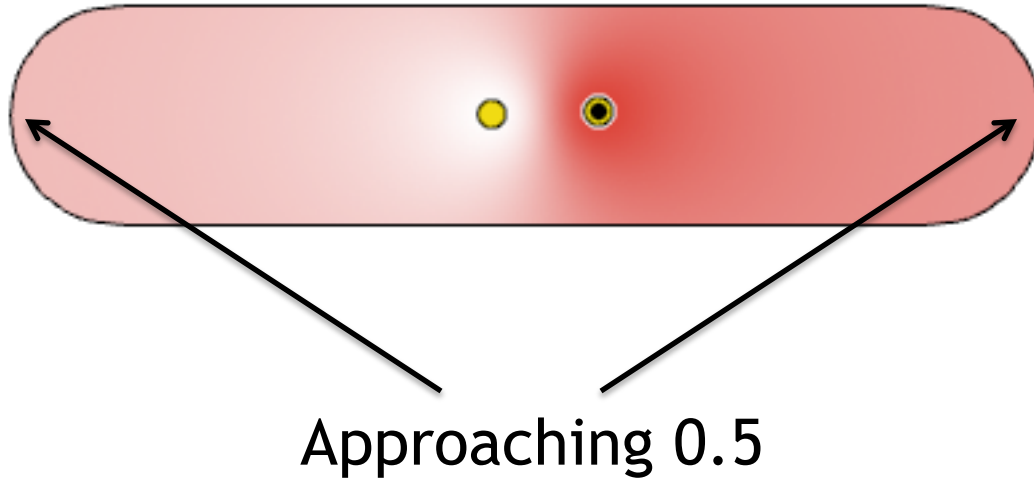
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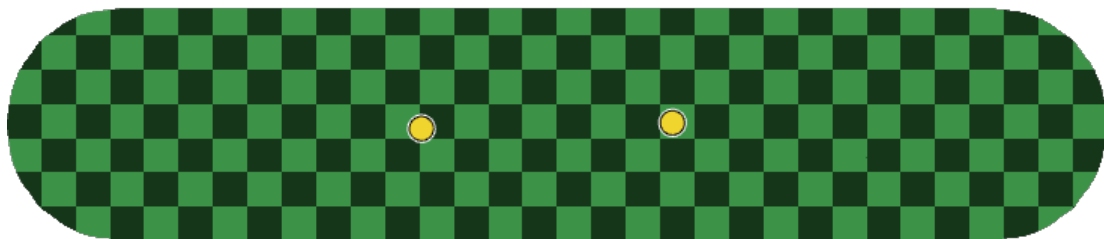


Inverse distance methods inherently suffer from *fall-off effect*

Shepard



Our method



Inverse distance methods inherently suffer from *fall-off effect*

Shepard



Our method



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Non-negativity	✓	-	✓	✓
Shape-aware	✓	✓	-	-
Locality, sparsity	-	-	-	✓
No local extrema	✓	-	-	✓

Support bones and cages?
Shape-aware?

Previous methods only partially satisfactory

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Smoothness	-	✓	✓	-
Non-negativity	✓	-	✓	✓
Shape-aware	✓	✓	-	-
Locality, sparsity	-	-	-	✓
No local extrema	✓	-	-	✓

$$\Delta^2 w_j = 0$$

Bounded biharmonic weights enforce properties as constraints to minimization

$$\arg \min_{w_j} \frac{1}{2} \int_{\Omega} \|\Delta w_j\|^2 dV$$

$$w_j \Big|_{H_k} = \delta_{jk}$$

w_j is linear along cage faces

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Constant inequality constraints

$$0 \leq w_j(\mathbf{x}) \leq 1$$

Partition of unity

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w_j is linear along cage faces

Constant inequality constraints

$$0 \leq w_j(\mathbf{x}) \leq 1$$

Solve independently, normalize

$$w_j(\mathbf{x}) = \frac{w_j(\mathbf{x})}{\sum_{i \in H_k} w_i(\mathbf{x})}$$

Weights optimized as precomputation at bind-time

$$\arg \min_{w_j} \frac{1}{2} \int_{\Omega} \|\Delta w_j\|^2 dV$$
$$w_j|_{H_k} = \delta_{jk}$$

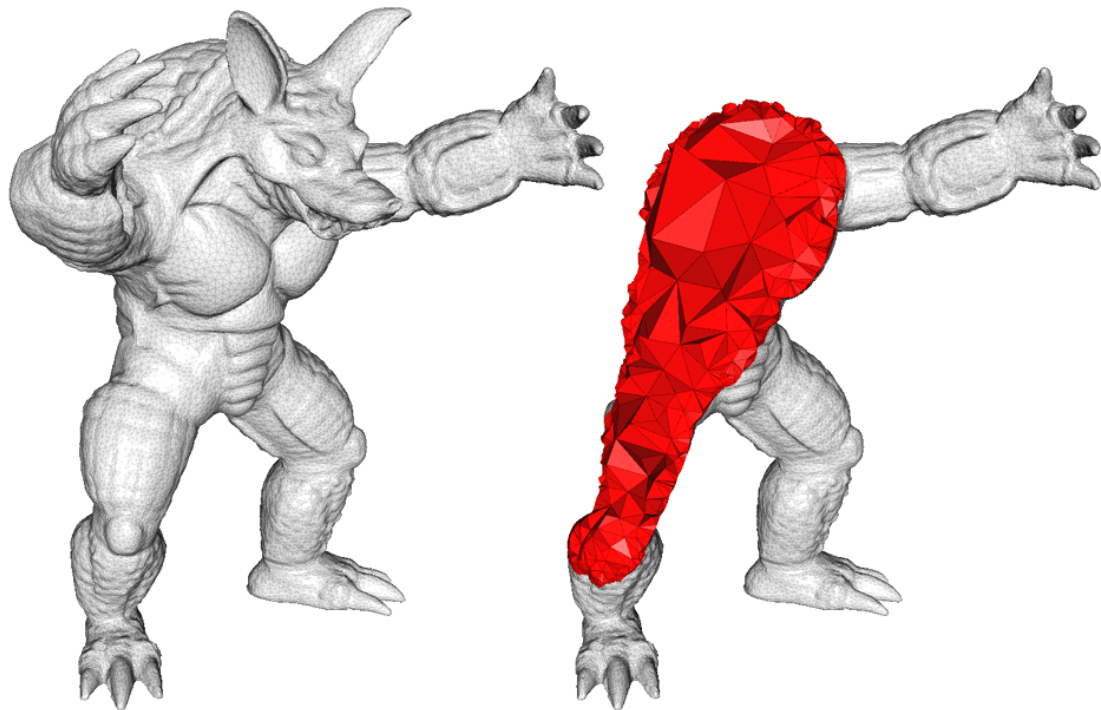
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$$0 \leq w_j(\mathbf{x}) \leq 1$$

FEM discretization

2D \rightarrow Triangle mesh

3D \rightarrow Tet mesh

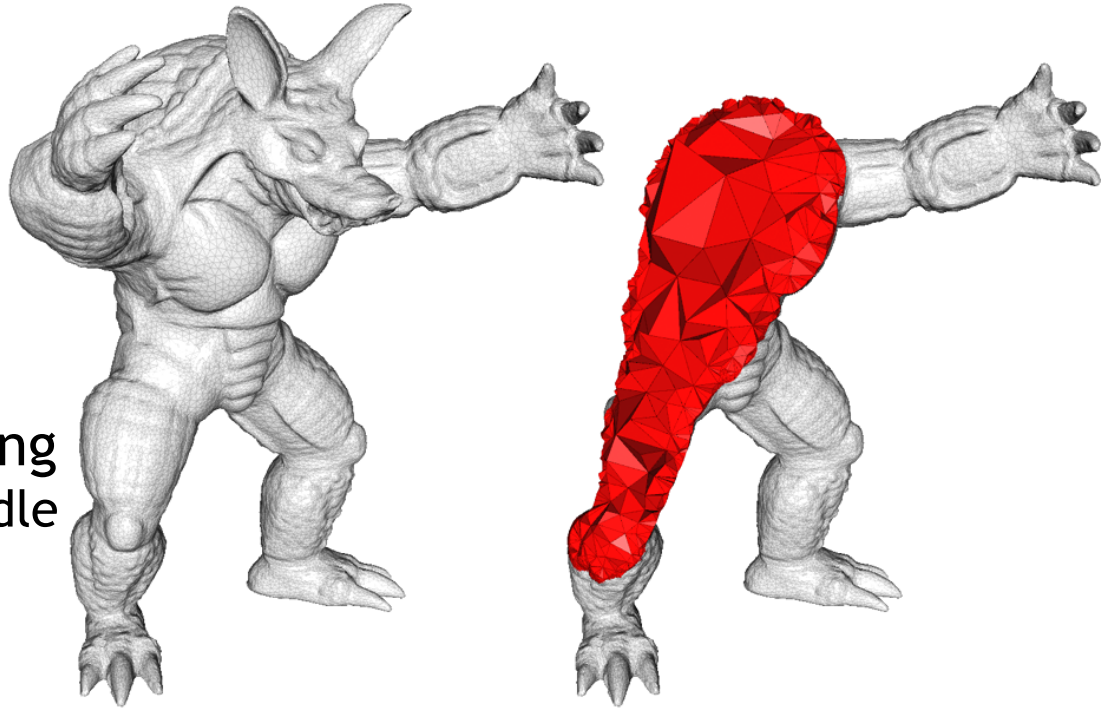


Weights optimized as precomputation at bind-time

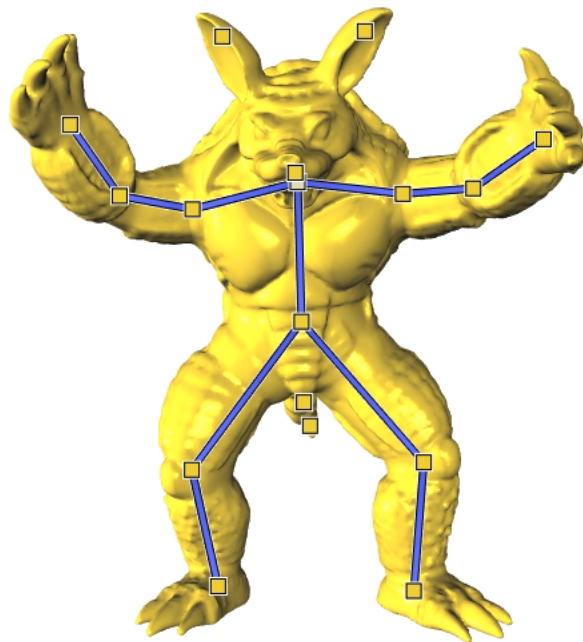
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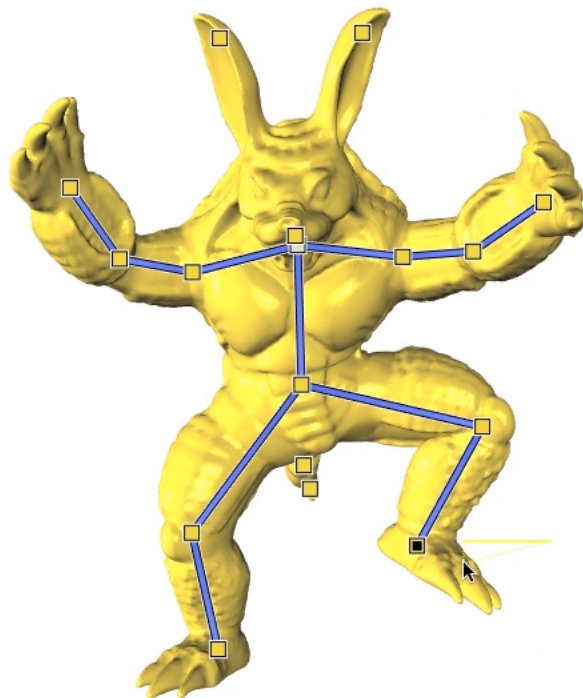
Sparse quadratic programming
2D ~O(milliseconds) per handle
3D ~O(seconds) per handle



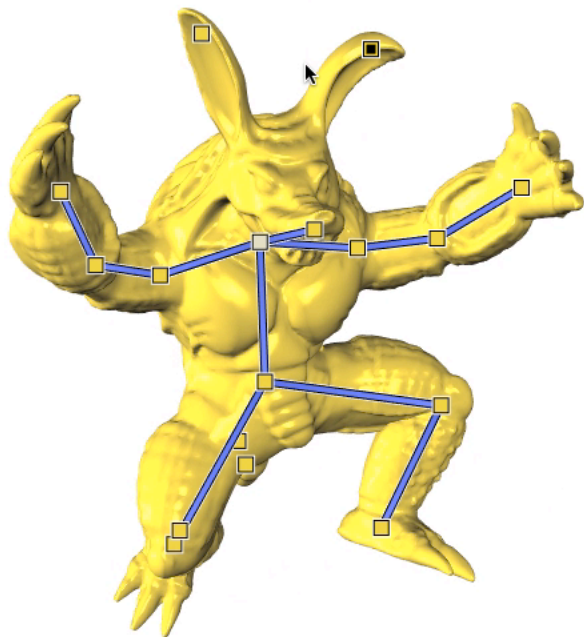
Weights in 3D also retain nice properties



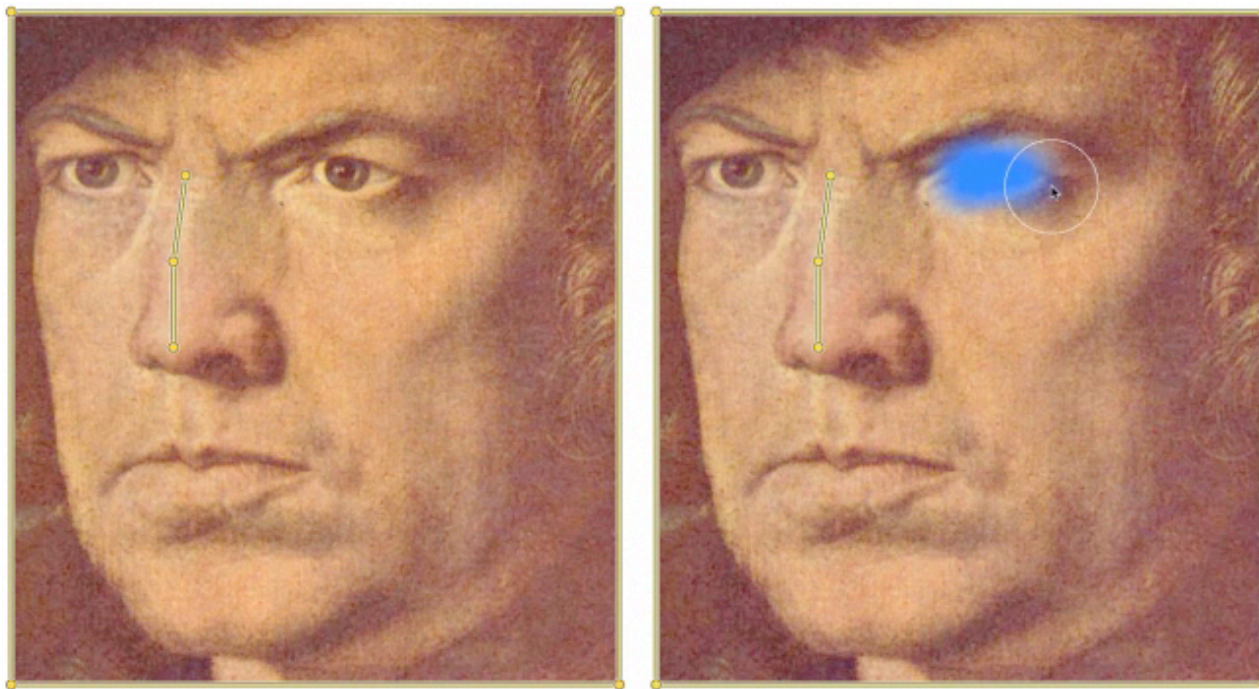
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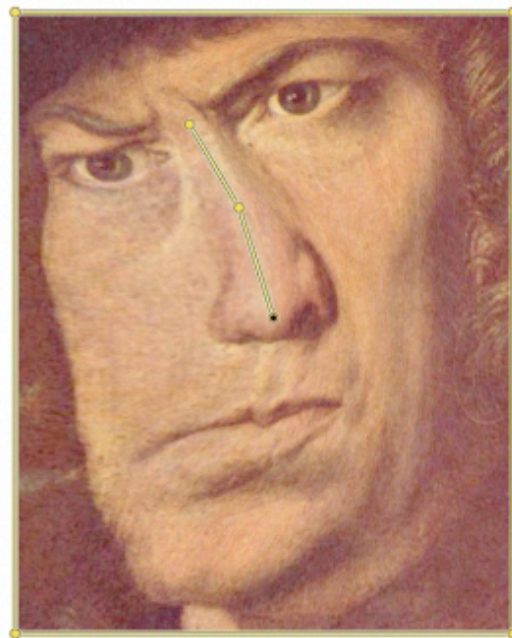
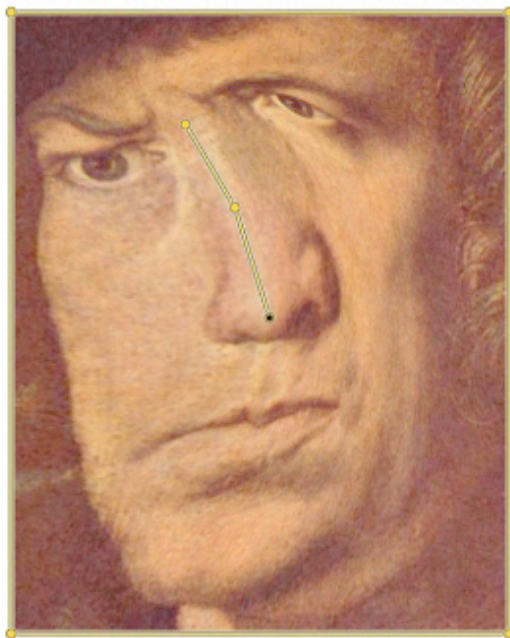
Weights in 3D also retain nice properties



Variational formulation allows additional, problem-specific constraints



Variational formulation allows additional, problem-specific constraints



Previous methods only partially satisfactory

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Smoothness	-	✓	✓	-
Non-negativity	✓	-	✓	✓
Shape-aware	✓	✓	-	-
Locality, sparsity	-	-	-	✓
No local extrema	✓	-	-	✓

$$\Delta^2 w_j = 0$$

Our weights obtain all properties...

	Harmonic Coordinates [Joshi et al. 2005]	Our Bounded Biharmonic Weights [Jacobson et al. 2011]	Shepard interpolation [Shepard 1968]	Natural neighbors [Sibson 1981]
Smoothness	-	✓	✓	-
Non-negativity	✓	✓ (Explicitly)	✓	✓
Shape-aware	✓	✓	-	-
Locality, sparsity	-	✓*	-	✓
No local extrema	✓	✓*	-	✓

*Empirically confirmed

... or so we thought

	Harmonic Coordinates [Joshi et al. 2005]	Our Bounded Biharmonic Weights [Jacobson et al. 2011]	Shepard interpolation [Shepard 1968]	Natural neighbors [Sibson 1981]
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Shape-aware	✓	✓	-	-
Locality, sparsity	-	✓*	-	✓
No local extrema	✓	-	-	✓

Smooth Shape-Aware Functions with Controlled Extrema

SGP 2012

Alec Jacobson¹

Tino Weinkauff²

Olga Sorkine¹

¹ETH Zurich

²MPI Saarbrücken



INTERACTIVE GEOMETRY LAB

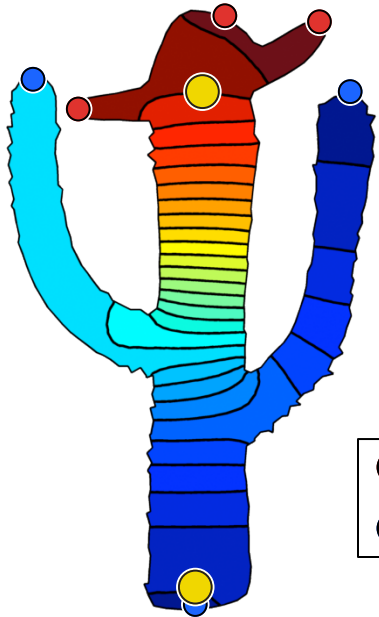
August 1, 2012

ETH

Eidgenössische Technische Hochschule Zürich
Swiss Federal Institute of Technology Zurich

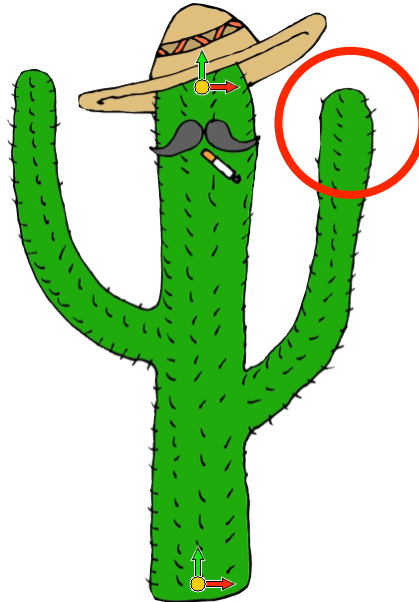
Spurious extrema cause distracting artifacts

unconstrained Δ^2
[Botsch & Kobbelt 2004]



● local max
● local min

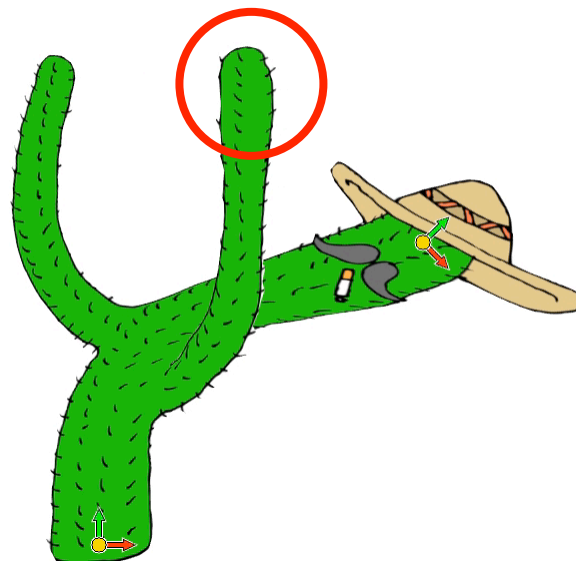
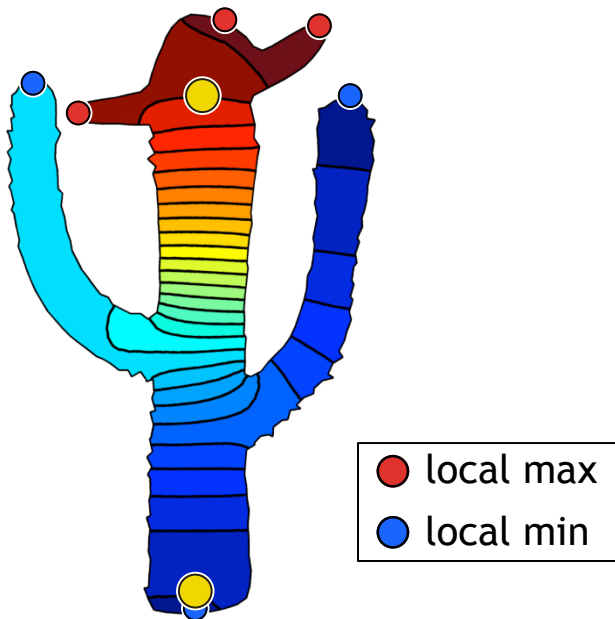
$$\mathbf{x}'_i = \sum_{j=1}^H f_j(\mathbf{x}_i) T_j \mathbf{x}_i$$



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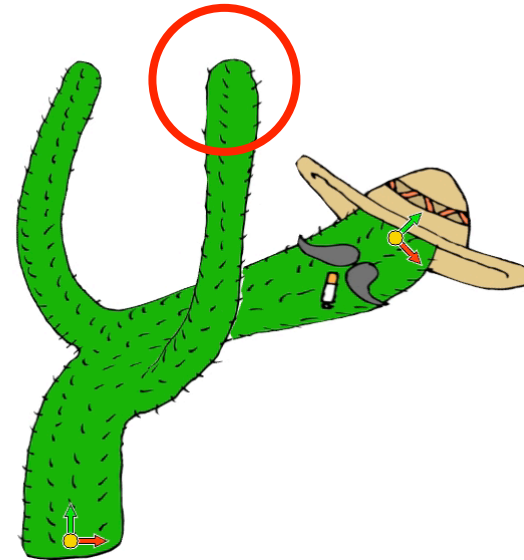
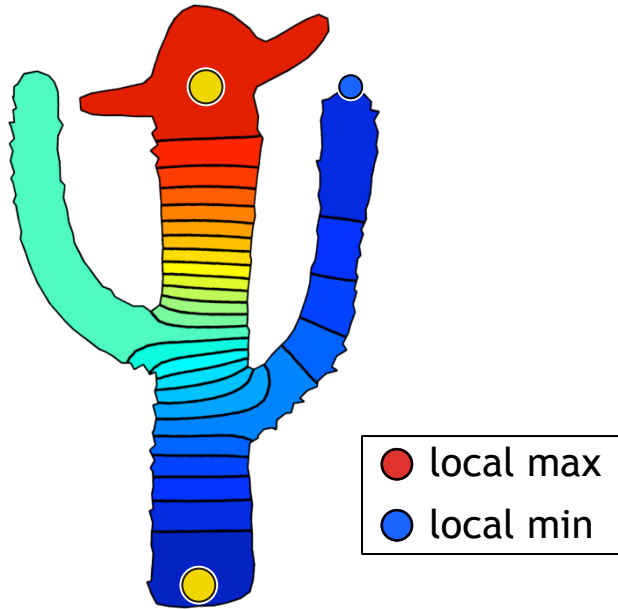
$$\mathbf{x}'_i = \sum_{j=1}^H f_j(\mathbf{x}_i) T_j \mathbf{x}_i$$



Bounds help, but don't solve problem

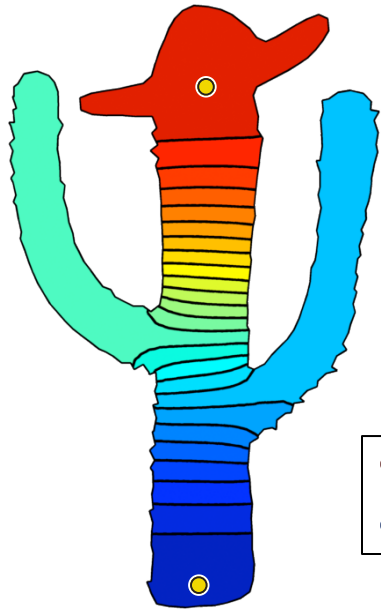
bounded Δ^2
[Jacobson et al. 2011]

$$\mathbf{x}'_i = \sum_{j=1}^H f_j(\mathbf{x}_i) T_j \mathbf{x}_i$$

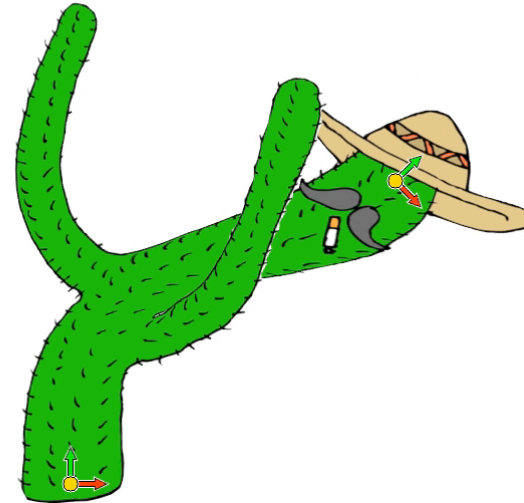


We explicitly prohibit spurious extrema

our Δ^2



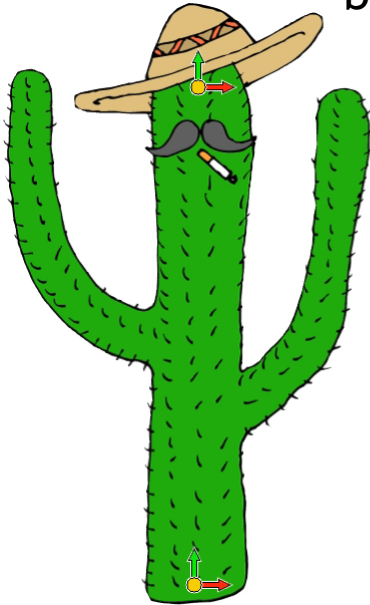
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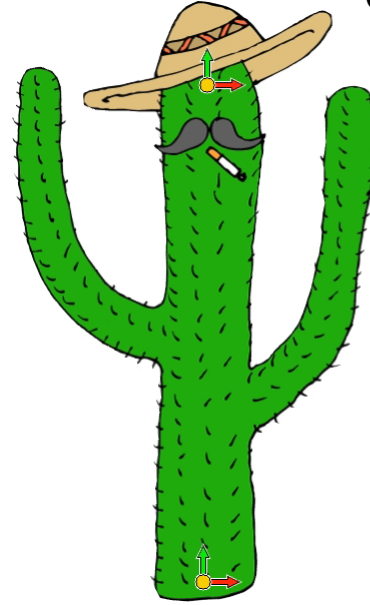
We *explicitly* prohibit spurious extrema

$$\mathbf{x}'_i = \sum_{j=1}^H f_j(\mathbf{x}_i) T_j \mathbf{x}_i$$

bounded Δ^4

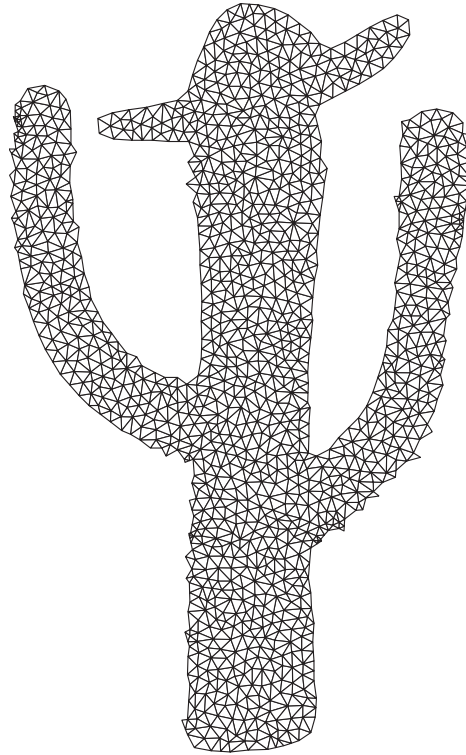


our Δ^4



Ideal discrete problem is intractable

$$\arg \min_f E(f)$$

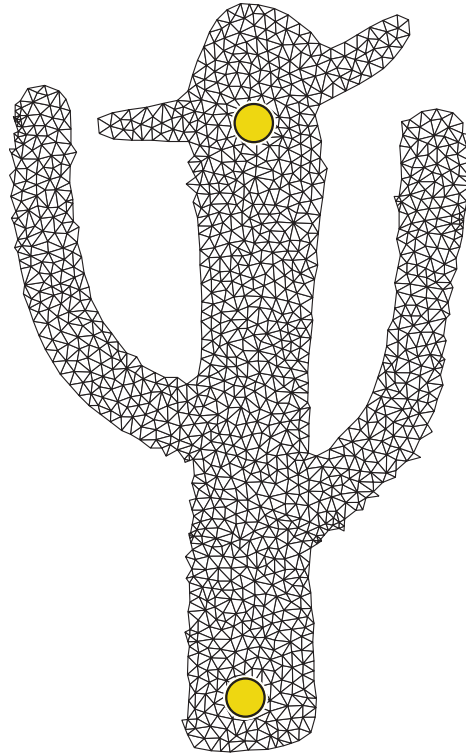


Ideal discrete problem is intractable

$$\arg \min_f E(f)$$

$$\text{s.t. } f_{\max} = \textit{known}$$

$$f_{\min} = \textit{known}$$



Ideal discrete problem is intractable

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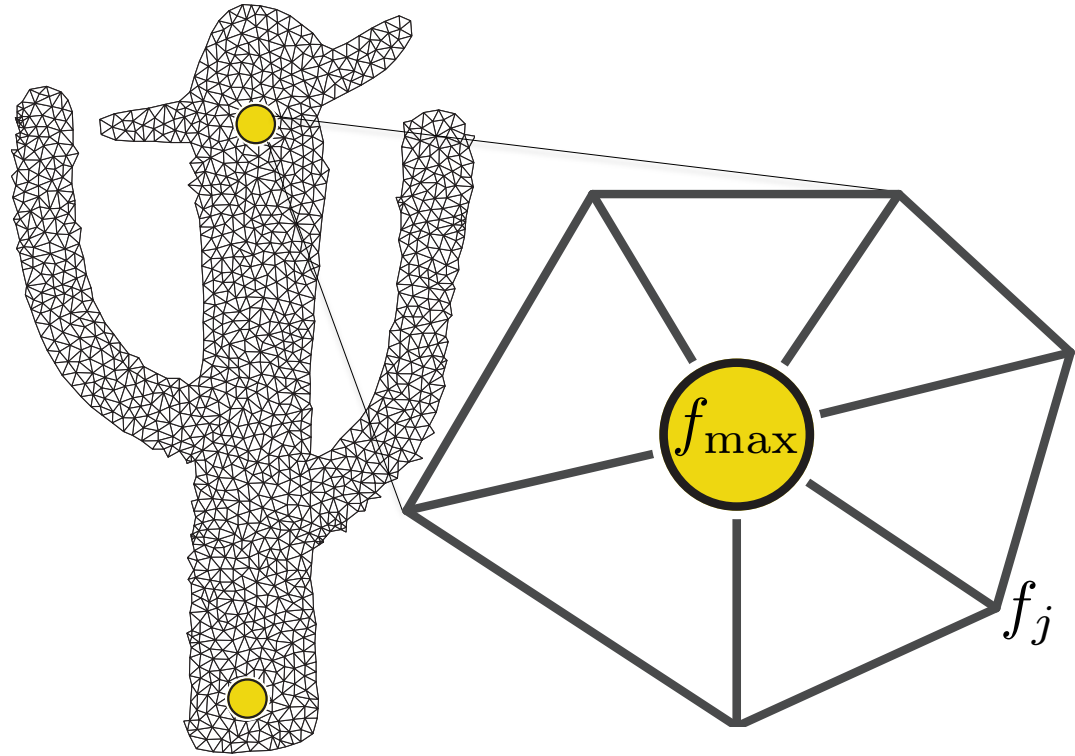
$$\text{s.t. } f_{\max} = \textit{known}$$

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linear

$$f_j < f_{\max}$$

$$f_j > f_{\min}$$



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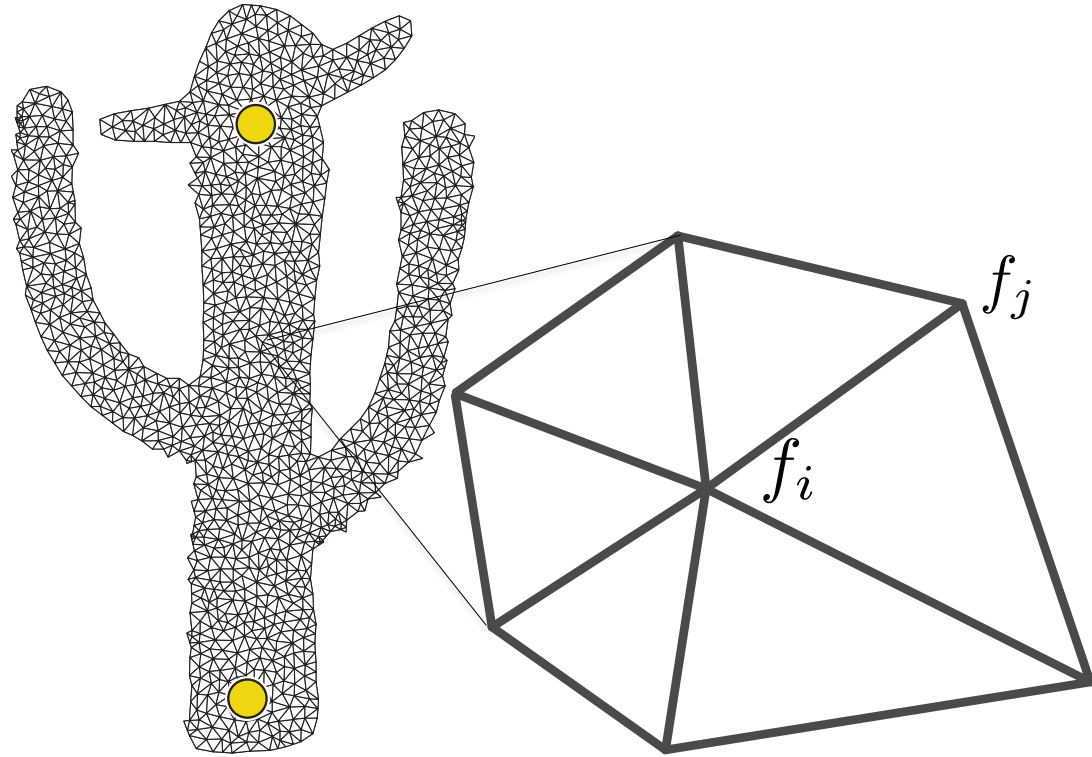
$$f_j < f_{\max}$$

$$f_j > f_{\min}$$

nonlinear

$$f_i > \min_{j \in \mathcal{N}(i)} f_j$$

$$f_i < \max_{j \in \mathcal{N}(i)} f_j$$



Assume we have a feasible solution

$$\arg \min_f E(f)$$

$$\text{s.t. } f_{\max} = \textit{known}$$

$$f_{\min} = \textit{known}$$

linear

$$f_j < f_{\max}$$

$$f_j > f_{\min}$$

nonlinear

$$f_i > \min_{j \in \mathcal{N}(i)} f_j$$

$$f_i < \max_{j \in \mathcal{N}(i)} f_j$$

“Representative function” u

$$u_j < u_{\max}$$

$$u_j > u_{\min}$$

handles

$$u_i > \min_{j \in \mathcal{N}(i)} u_j$$

$$u_i < \max_{j \in \mathcal{N}(i)} u_j$$

interior

Copy “monotonicity” of representative

$$\arg \min_f E(f)$$

$$\text{s.t. } f_{\max} = \textit{known}$$

$$f_{\min} = \textit{known}$$

$$(f_i - f_j)(u_i - u_j) > 0 \quad \text{linear} \quad \forall (i, j) \in \mathcal{E}$$

At least one edge in either
direction per vertex

Rewrite as conic optimization

QP

$$\begin{aligned} & \underset{\mathbf{f}}{\text{minimize}} && \frac{1}{2} \|\mathbf{F}\mathbf{f}\|^2 + \mathbf{c}^\top \mathbf{f} + \text{const} \\ & \text{subject to} && \mathbf{A}_{leq}^\top \mathbf{f} \leq \mathbf{b}_{leq}, \\ & && \mathbf{f} \leq \mathbf{u}_f, \quad \mathbf{f} \geq \mathbf{l}_f \end{aligned}$$



Conic

$$\begin{aligned} & \underset{\begin{bmatrix} \mathbf{f} \\ \mathbf{t} \\ v \end{bmatrix}}{\text{minimize}} && \begin{bmatrix} \mathbf{c}^\top & 0 & 1 \end{bmatrix} \begin{bmatrix} \mathbf{f} \\ \mathbf{t} \\ v \end{bmatrix} + \text{const} \\ & \text{subject to} && \begin{bmatrix} \mathbf{F} & -\mathbf{I} & 0 \\ \mathbf{A}_{leq}^\top & 0 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{f} \\ \mathbf{t} \\ v \end{bmatrix} \geq \begin{bmatrix} 0 \\ -\infty \end{bmatrix} \\ & && \begin{bmatrix} \mathbf{F} & -\mathbf{I} & 0 \\ \mathbf{A}_{leq}^\top & 0 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{f} \\ \mathbf{t} \\ v \end{bmatrix} \leq \begin{bmatrix} 0 \\ \mathbf{b}_{leq} \end{bmatrix} \\ & && \begin{bmatrix} \mathbf{f} \\ \mathbf{t} \\ v \end{bmatrix} \leq \begin{bmatrix} \mathbf{u}_f \\ \infty \\ \infty \end{bmatrix} \\ & && \begin{bmatrix} \mathbf{f} \\ \mathbf{t} \\ v \end{bmatrix} \geq \begin{bmatrix} \mathbf{l}_f \\ -\infty \\ 0 \end{bmatrix} \\ & && 2v \geq \sum_i t_i^2 \end{aligned}$$

Optimize with MOSEK

Harmonic functions obey maximum principle

	Harmonic Coordinates [Joshi et al. 2005]	Unconstrained biharmonic [Botsch and Kobbelt 2004]	Shepard interpolation [Shepard 1968]	Natural neighbors [Sibson 1981]
Smoothness	-	✓	✓	-
Non-negativity	✓	-	✓	✓
Shape-aware	✓	✓	-	-
Locality, sparsity	-	-	-	✓
No local extrema	✓	-	-	✓

$$\Delta u = 0$$

Final algorithm is simple and efficient

- Harmonic representative
 - Linear solve $\sim O(\text{milliseconds})$

Final algorithm is simple and efficient

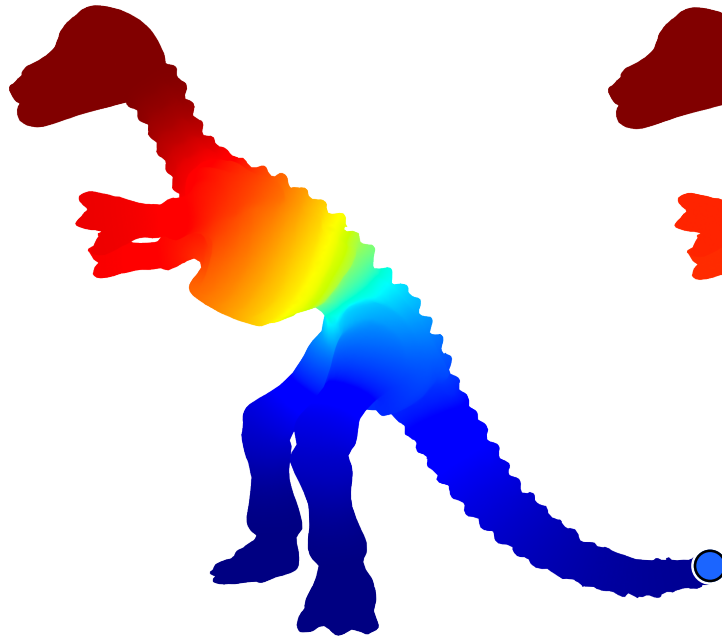
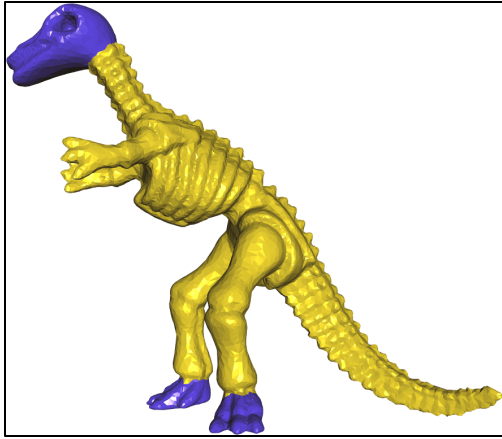
- Harmonic representative
 - Linear solve $\sim O(\text{milliseconds})$
- Conic optimization
 - 2D $\sim O(\text{milliseconds})$, 3D $\sim O(\text{seconds})$

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Again, functions are precomputed

Our weights attach appendages to body

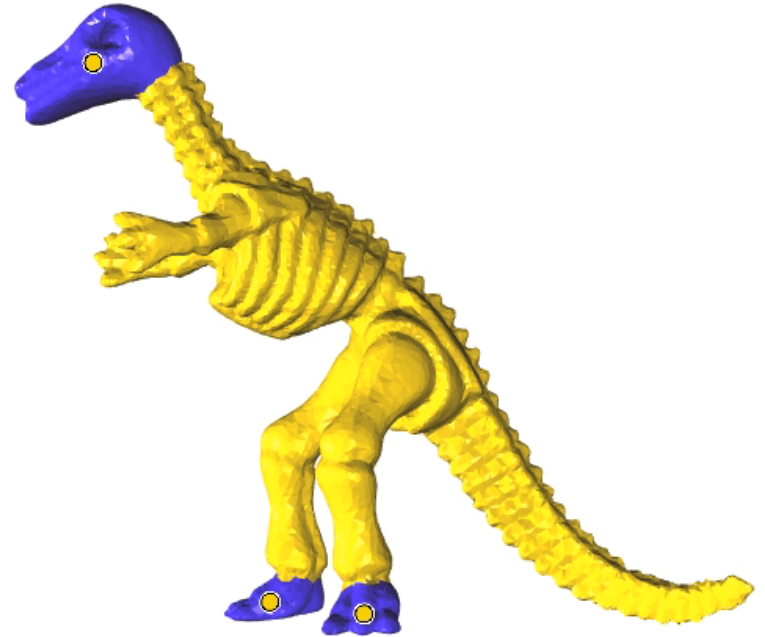
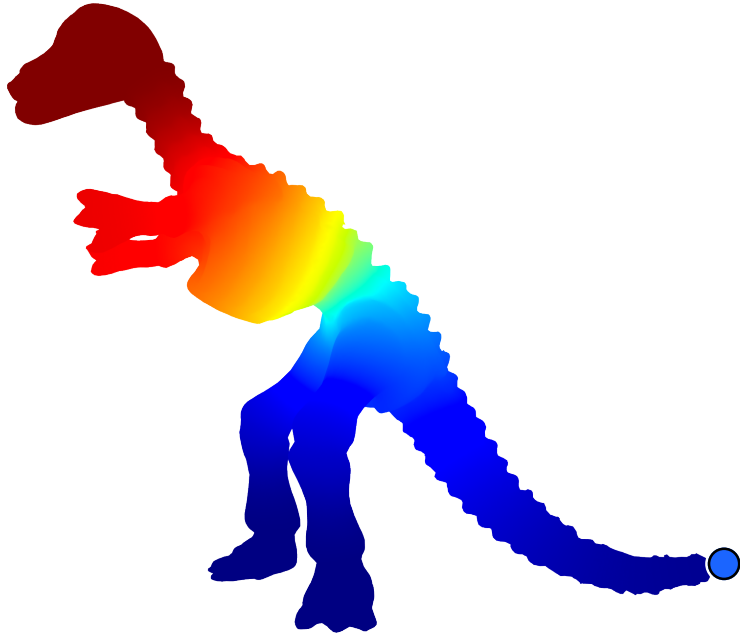


[Botsch & Kobbelt 2004,
Jacobson et al. 2011]



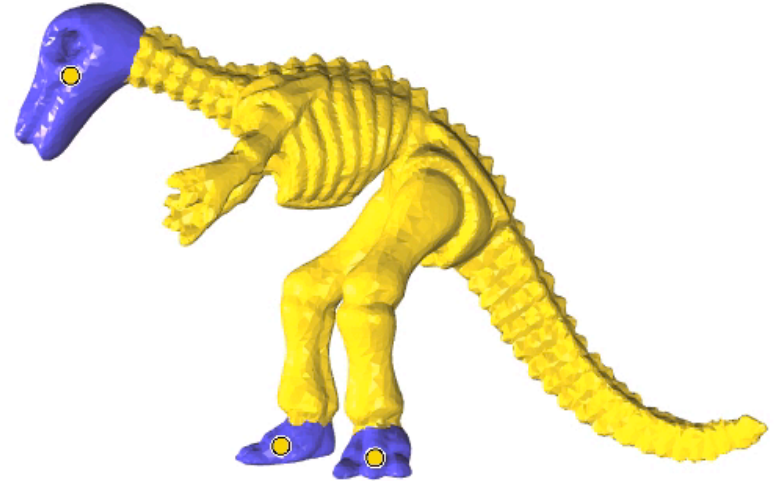
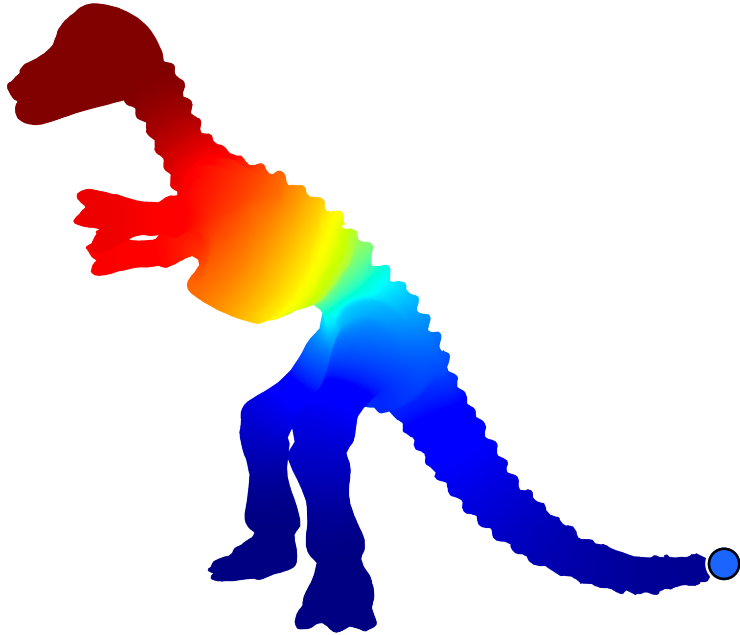
Our method

Extrema glue appendages to far-away handles



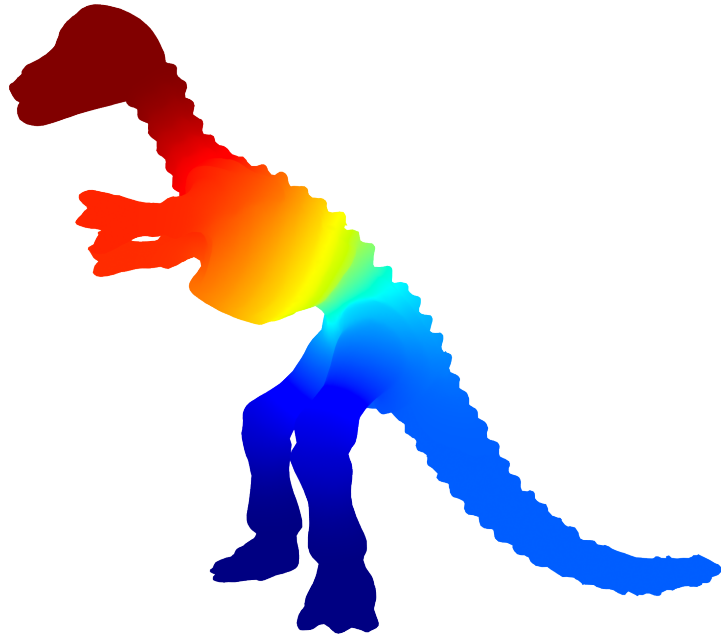
[Botsch & Kobbelt 2004, Jacobson et al. 2011]

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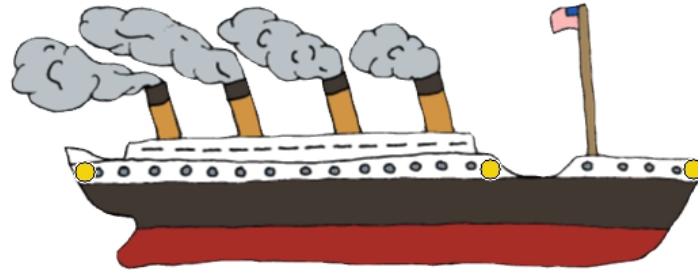
Our method

Extrema distort small features

Bounded Δ^2 [Jacobson et al. 2011]

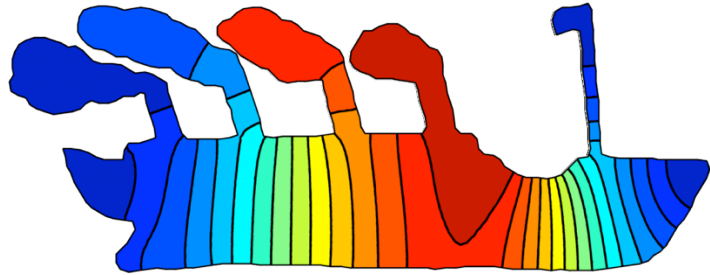


weight of middle point

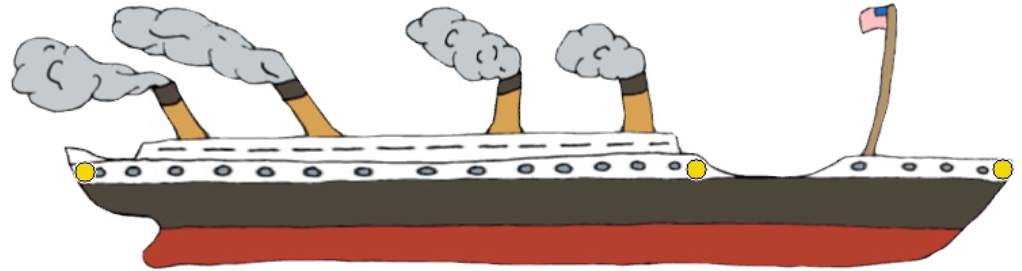


Extrema distort small features

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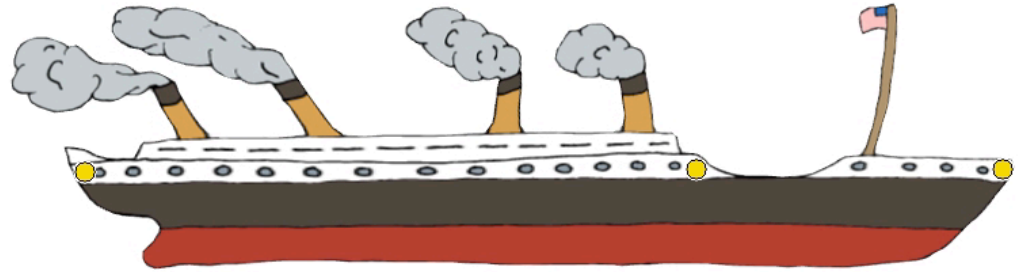
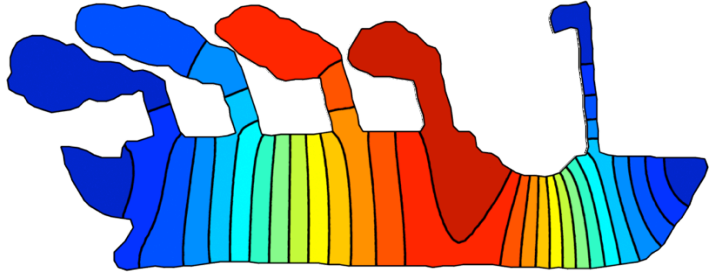


weight of middle point

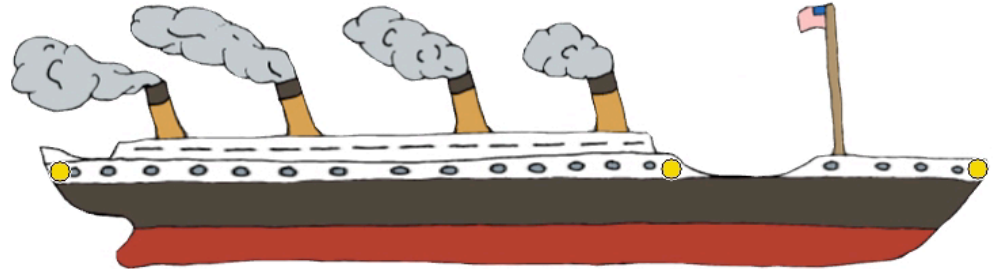
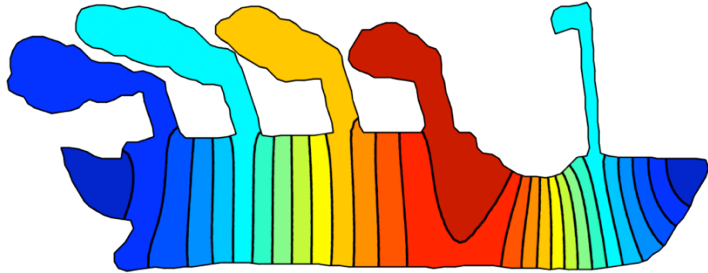


“Monotonicity” helps preserve small features

Bounded Δ^2 [Jacobson et al. 2011]



Our Δ^2



Conclusion: variational framework allows explicit control over desired properties

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 - *Implicit* locality, sparsity
- Explicit monotonicity

Future work and discussion

- Continuous formulation of monotonicity?

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- Explicit sparsity? Linear precision?

Acknowledgements

We thank Kenshi Takayama, Yang Song, Jaakko Lehtinen, Bob Sumner and Denis Zorin.

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High-quality weight functions via constrained optimization

Alec Jacobson jacobson@inf.ethz.ch

MATLAB Demos and more:

<http://igl.ethz.ch/projects/bbw/>

<http://igl.ethz.ch/projects/monotonic/>

