

Meshless Deformations Based on Shape Matching

M. Müller, B. Heidelberger, M. Teschner, M. Gross

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The Goal

To simulate deformable objects in a computationally efficient and stable way



Oriented towards the gaming and animation industries

Existing methods

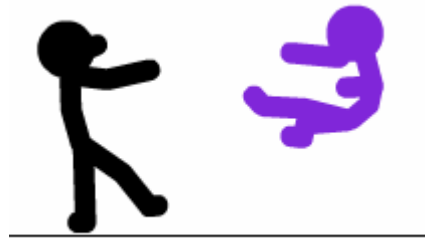
Not applicable because they are either:

- Difficult to code (FEM, BEM)
- Too slow (FEM)
- Not stable (explicit mass spring)
- Memory consuming
(Eigenmodes, stiffness matrices,
Green's functions, motion databases)

Mickey Mouse physics

To simplify the problem, the authors:

- Relax the requirement of realism
- Do not model material properties
- Avoid using the connectivity information



Simple Model

- Solid represented by points with mass
 - Simple to generate
 - Animated as a particle system
- No connectivity needed
 - Volumetric meshes not available in games
 - Each point path is calculated independently
- Explicit integration
 - Simple to code
 - Efficient to compute
 - Unconditionally stable

Basic Algorithm

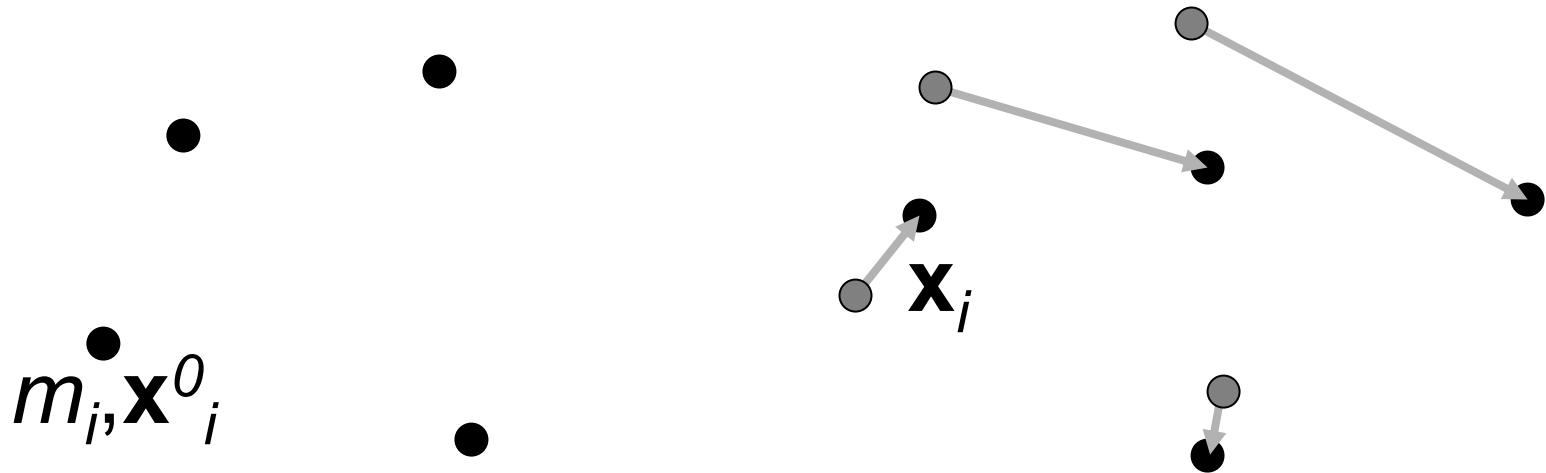
- Given n points with masses m_i and original positions \mathbf{x}^0_i

m_i, \mathbf{x}^0_i



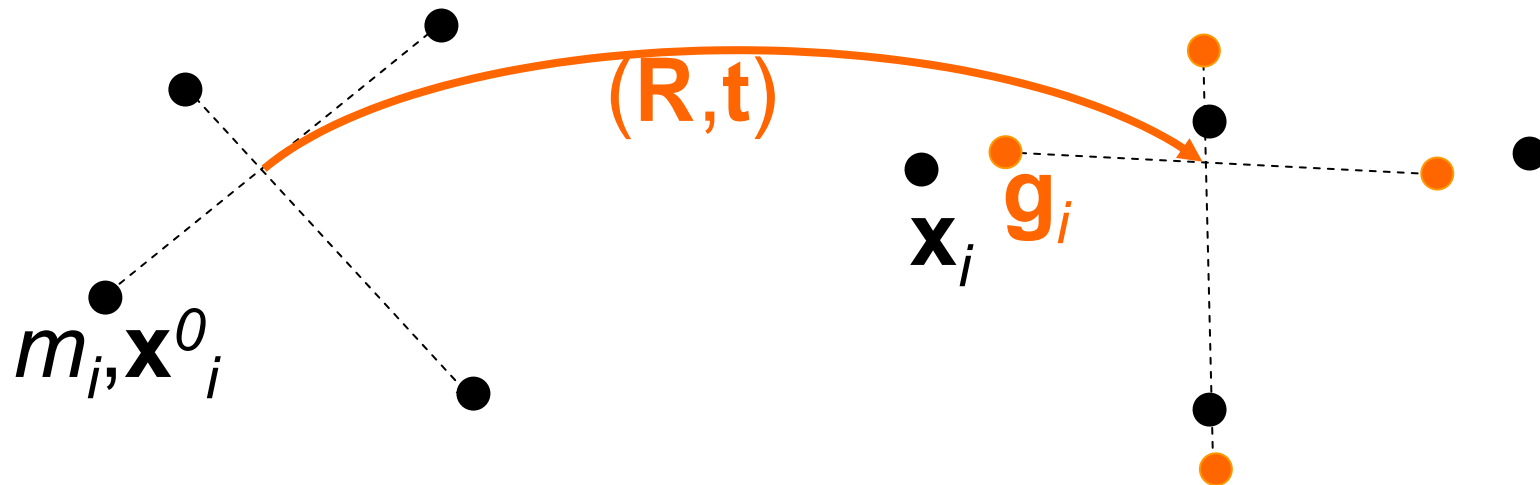
Basic Algorithm

- Animate the points as unconnected particle system considering only external forces



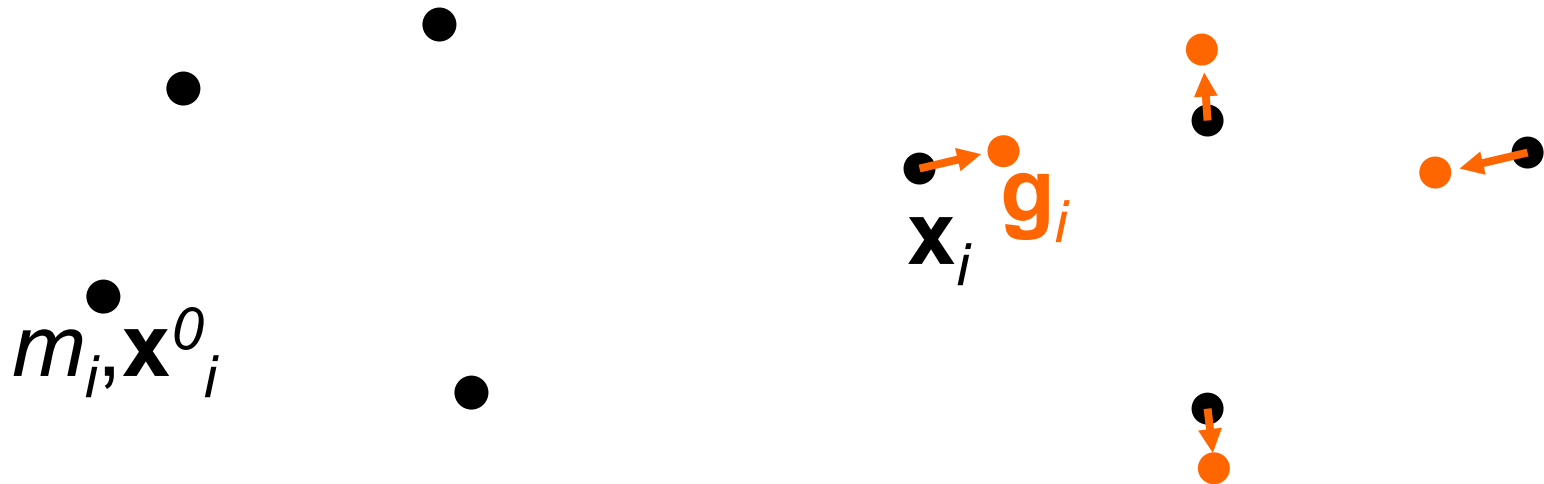
Basic Algorithm

- At every time step, match the original configuration \mathbf{x}^0_i to the actual configuration \mathbf{x}_i yielding goals \mathbf{g}_i



Basic Algorithm

- Pull the actual points towards the goal positions \mathbf{g}_i



Algorithm

Has two main components:

1. Finding the optimal transformation
(shape matching to find the goal points)
2. Moving the particles towards those points
(modeling acceleration and velocity)

Trade Off:

Subside to external forces vs. Retain the original configuration

Shape Matching

- Given \mathbf{x}_i^0 , \mathbf{x}_i and weights w_i
- Find rigid body transform \mathbf{t}_0 , \mathbf{t} , \mathbf{R} minimizing

$$\sum_i w_i \left(\mathbf{R}(\mathbf{x}_i^0 - \mathbf{t}_0) + \mathbf{t} - \mathbf{x}_i \right)^2$$

- Select $w_i = m_i$
- Optimal \mathbf{t}_0 is center of mass of the \mathbf{x}_i^0
- Optimal \mathbf{t} is center of mass of the \mathbf{x}_i

[Kanatani 1994], [Umeyama 1991] and [Lorusso et al.1995]

Optimal Rotation

- Define $\mathbf{p}_i = \mathbf{x}_i - \mathbf{x}_{cm}$ and $\mathbf{q}_i = \mathbf{x}_i^0 - \mathbf{x}_{cm}^0$
- Minimize $\sum_i m_i (\mathbf{A}\mathbf{q}_i - \mathbf{p}_i)^2$
- With optimal linear transform

$$\mathbf{A} = \left(\sum_i m_i \mathbf{p}_i \mathbf{q}_i^T \right) \left(\sum_i m_i \mathbf{q}_i \mathbf{q}_i^T \right)^{-1} = \mathbf{A}_{pq} \mathbf{A}_{qq}$$

- Optimal \mathbf{R} from polar decomposition of \mathbf{A}_{pq}

Goal positions

$$\mathbf{A} = \left(\sum_i m_i \mathbf{p}_i \mathbf{q}_i^T \right) \left(\sum_i m_i \mathbf{q}_i \mathbf{q}_i^T \right)^{-1} = \mathbf{A}_{pq} \mathbf{A}_{qq}$$

$$\mathbf{R} = \mathbf{A}_{pq} \mathbf{S}^{-1} \text{ where } \mathbf{S} = \sqrt{\mathbf{A}_{pq}^T \mathbf{A}_{pq}}$$

$$\mathbf{g}_i = \mathbf{R} (\mathbf{x}_i^0 - \mathbf{x}_{cm}^0) + \mathbf{x}_{cm}$$

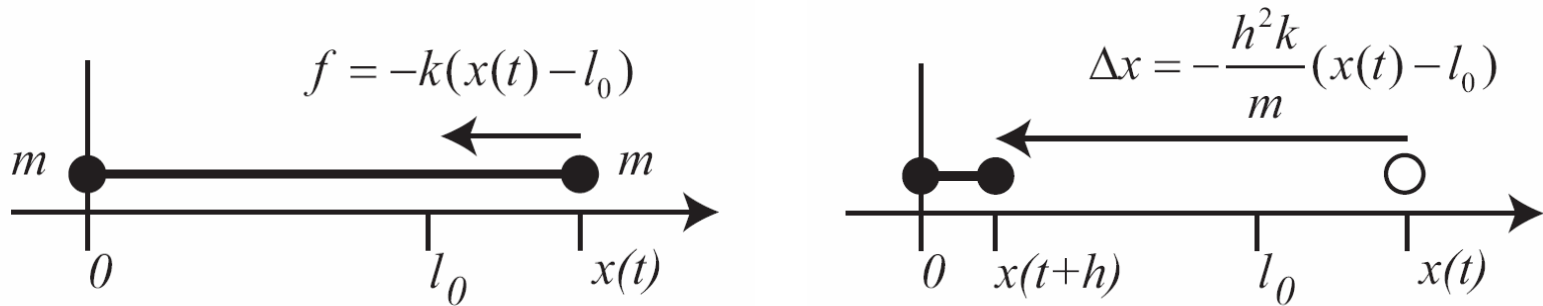
Stable Explicit Integration

$$\mathbf{v}_i(t+h) = \mathbf{v}_i(t) + \alpha \frac{\mathbf{g}_i(t) - \mathbf{x}_i(t)}{h} + \frac{h}{m_i} \mathbf{f}_{\text{ext}}(t)$$

$$\mathbf{x}_i(t+h) = \mathbf{x}_i(t) + h \mathbf{v}_i(t+h)$$

- Unconditionally stable
- Controls stiffness (rigid for $\alpha = 1$)
- Pulled out of a hat

Compare to Euler Integration



$$v(t+h) = v(t) + h \frac{-k(x(t) - l_0)}{m}$$

$$x(t+h) = x(t) + hv(t+h),$$

$$A = \begin{bmatrix} 1 & -\frac{kh}{m} \\ h & 1 - \frac{h^2 k}{m} \end{bmatrix}$$

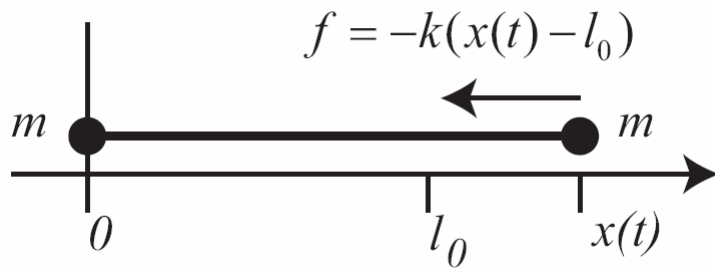
$$e_0 = 1 - \frac{1}{2m}(h^2 k - \sqrt{-4mh^2 k + h^4 k^2})$$

$$e_1 = 1 - \frac{1}{2m}(h^2 k + \sqrt{-4mh^2 k + h^4 k^2})$$

$$\clubsuit e_0 \clubsuit < 1 \text{ for } h^2 k \rightarrow$$

$$\infty \clubsuit e_1 \clubsuit < 1 \text{ iff } h < 2\sqrt{\frac{m}{k}}$$

Compare to Euler Integration



$$\mathbf{v}_i(t+h) = \mathbf{v}_i(t) + \alpha \frac{\mathbf{g}_i(t) - \mathbf{x}_i(t)}{h} + hf_{\text{ext}}(t)/m_i$$

$$\mathbf{x}_i(t+h) = \mathbf{x}_i(t) + h\mathbf{v}_i(t+h),$$

$$\begin{bmatrix} v(t+h) \\ x(t+h) \end{bmatrix} = \begin{bmatrix} 1 & -\alpha/h \\ h & 1-\alpha \end{bmatrix} \begin{bmatrix} v(t) \\ x(t) \end{bmatrix} + \begin{bmatrix} \alpha l_0/h \\ \alpha l_0 \end{bmatrix}$$

$$e_{0,1} = (1 - \alpha/2) \pm i\sqrt{4\alpha - \alpha^2}/2$$

$$\clubsuit e_{0,1} \clubsuit /$$

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Extensions

(shape matching cont'd)

1. Linear Deformations
2. Quadratic Deformations
3. Cluster-based Deformations
4. Plasticity

Linear Deformations

Recall that \mathbf{R} is the polar decomposition of \mathbf{A}_{pq} ($\mathbf{A}_{pq} = \mathbf{R}\mathbf{S}$, where $\mathbf{S} = \sqrt{\mathbf{A}_{pq}^T \mathbf{A}_{pq}}$)

Now, instead of using only \mathbf{R} to compute the goal points, also use the matrix \mathbf{A}

$$\mathbf{g}_i = (\beta \mathbf{A} + (1 - \beta) \mathbf{R})(\mathbf{x}_i^0 - \mathbf{x}_{cm}^0) + \mathbf{x}_{cm}$$

Objects are allowed to shear and stretch

Quadratic Deformations

Further modify the matrix \mathbf{A} by setting

$$\tilde{\mathbf{A}} = \begin{bmatrix} \mathbf{A} & \mathbf{Q} & \mathbf{M} \end{bmatrix}$$

and minimizing $\sum_i m_i (\tilde{\mathbf{A}} \tilde{\mathbf{q}}_i - \mathbf{p}_i)^2$ where

$$\tilde{\mathbf{q}} = [q_x, q_y, q_z, q_x^2, q_y^2, q_z^2, q_x q_y, q_x q_z, q_y q_z]$$

$$\mathbf{g}_i = (\beta \tilde{\mathbf{A}} + (1 - \beta) \tilde{\mathbf{R}}) (\mathbf{x}_i^0 - \mathbf{x}_{cm}^0) + \mathbf{x}_{cm}$$

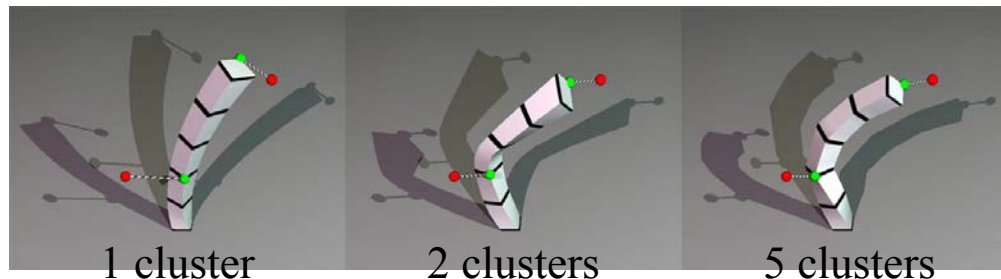
Objects are allowed to twist and bend

Cluster Based Deformations

The set of particles is divided into overlapping clusters (cubes in space)

Run shape matching for each cluster

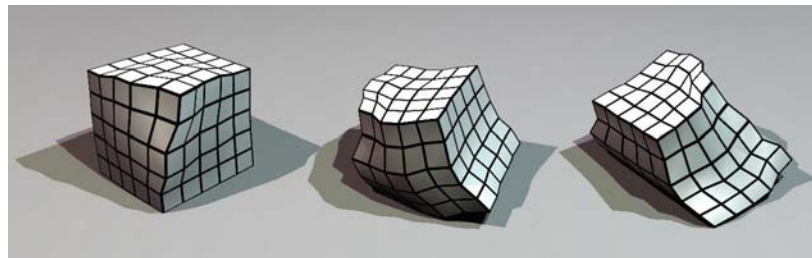
Average goal positions for each point



Plasticity

Recall $\mathbf{A} = \mathbf{R}\mathbf{S}$ where \mathbf{R} is the rotational part of the transformation

In each cluster, if the unrotated deformation $\|\mathbf{S} - \mathbf{I}\|_2$ exceeds a threshold c_{yield} , incorporate it into the original shape



Conclusions

- Pros
 - Straight forward to implement
 - Stable and scalable
 - Time and memory efficient
- Cons
 - Geometric - **not** physically derived
 - Tuned via non-physical (geometric) parameters