

Introduction to Computer Graphics

– Animation (2) –

May 20, 2021

Kenshi Takayama

Physics-based animation of deforming objects

- Classic: faithful simulation of physical phenomena
 - Mass-spring system
 - Finite Element Method (FEM)

- CG-specific: plausible & robust, but not faithful simulation
 - Shape Matching (Position-Based Dynamics)

Simple example: single mass & spring in 1D

- Mass m , position x , spring coefficient k , rest length l , gravity g :

$$\boxed{\text{Equation of motion}} \quad m \frac{d^2 x}{dt^2} = -k(x - l) + m g \\ = f_{\text{int}}(x) + f_{\text{ext}}$$

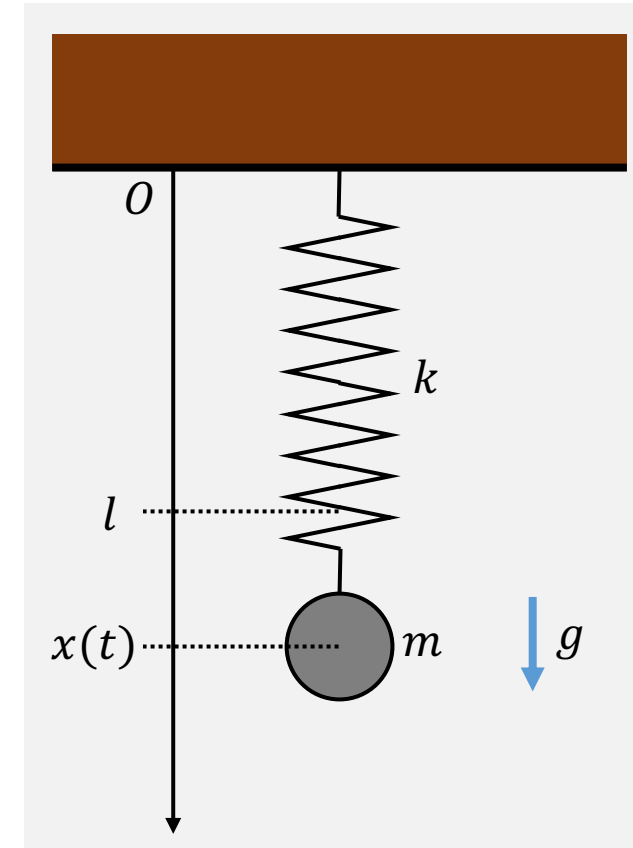
- f_{ext} : External force (gravity, collision, user interaction)
- $f_{\text{int}}(x)$: Internal force (pulling the system back to original)

- Spring's internal energy (potential):

$$E(x) := \frac{k}{2} (x - l)^2$$

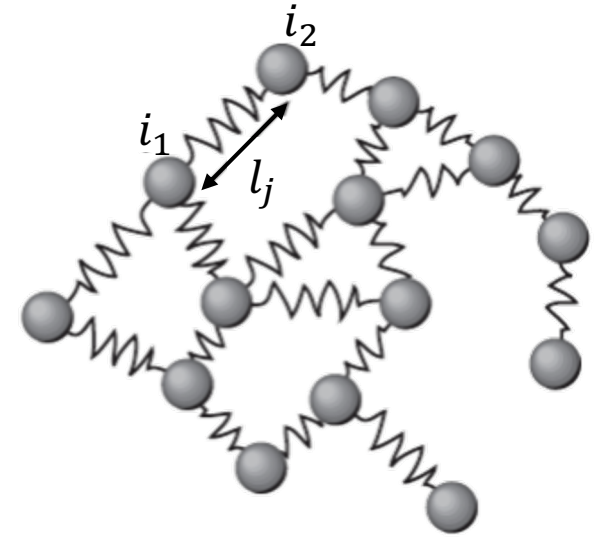
- Internal force is the opposite of potential gradient:

$$f_{\text{int}}(x) := -\frac{dE}{dx} = -k(x - l)$$



Mass-spring system in 3D

- N masses: i -th mass m_i , position $x_i \in \mathbb{R}^3$
- M springs: j -th spring $e_j = (i_1, i_2)$
 - Coefficient k_j , rest length l_j



- System's potential energy for a state $\mathbf{x} = (x_1, \dots, x_N) \in \mathbb{R}^{3N}$:

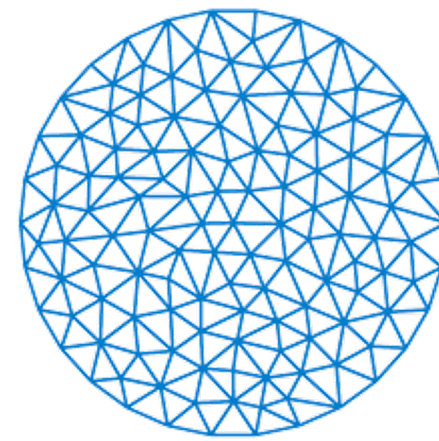
$$E(\mathbf{x}) := \sum_{e_j=(i_1, i_2)} \frac{k_j}{2} (\|x_{i_1} - x_{i_2}\| - l_j)^2$$

- Equation of motion:

$$\mathbf{M} \frac{d^2 \mathbf{x}}{dt^2} = -\nabla E(\mathbf{x}) + \mathbf{f}_{\text{ext}}$$

- $\mathbf{M} \in \mathbb{R}^{3N \times 3N}$: Diagonal matrix made of m_i (mass matrix)

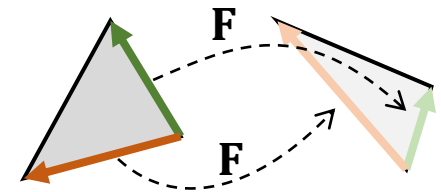
Continuous elastic model in 2D (Finite Element Method)



Tessellate the domain into triangular mesh

- N vertices: i -th position $x_i \in \mathbb{R}^2$
- M triangles: j -th triangle $t_j = (i_1, i_2, i_3)$
- Undeformed state: $\mathbf{X} = (X_1, \dots, X_N) \in \mathbb{R}^{2N}$
- Deformed state: $\mathbf{x} = (x_1, \dots, x_N) \in \mathbb{R}^{2N}$
- Deformation gradient:

$$\mathbf{F}_j(\mathbf{x}) := \begin{pmatrix} x_{i_1} & x_{i_2} \\ x_{i_1} & x_{i_3} \end{pmatrix} \begin{pmatrix} X_{i_1} & X_{i_2} \\ X_{i_1} & X_{i_3} \end{pmatrix}^{-1} \in \mathbb{R}^{2 \times 2}$$



Linear transformation which maps edges

- System's potential:

$$E(\mathbf{x}) := \sum_{t_j=(i_1, i_2, i_3)} \frac{A_j}{2} \left\| \mathbf{F}_j(\mathbf{x})^\top \mathbf{F}_j(\mathbf{x}) - \mathbf{I} \right\|_{\mathcal{F}}^2$$

Area of t_j

Green's strain energy

- Equation of motion:

$$\mathbf{M} \frac{d^2 \mathbf{x}}{dt^2} = -\nabla E(\mathbf{x}) + \mathbf{f}_{\text{ext}}$$

- $\mathbf{M} \in \mathbb{R}^{2N \times 2N}$: Diagonal matrix made of vertices' Voronoi areas

Computing dynamics

- Problem: Given initial value of position $\mathbf{x}(t)$ and velocity $\mathbf{v}(t) := \frac{d\mathbf{x}}{dt}$ as

$$\mathbf{x}(0) = \mathbf{x}_0 \quad \text{and} \quad \mathbf{v}(0) = \mathbf{v}_0 ,$$

compute $\mathbf{x}(t)$ and $\mathbf{v}(t)$ for $t > 0$. (Initial Value Problem)

- Simple case of single mass & spring:

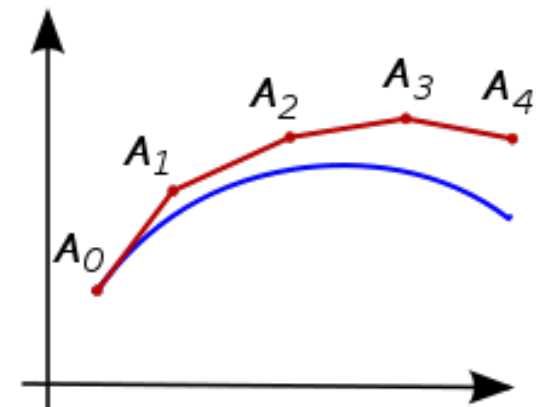
$$m \frac{d^2 x}{dt^2} = -k(x - l) + m g$$

→ analytic solution exists (sine curve)

- General problems don't have analytic solution

→ From state $(\mathbf{x}_n, \mathbf{v}_n)$ at time t , compute next state $(\mathbf{x}_{n+1}, \mathbf{v}_{n+1})$ at time $t + h$. (time integration)

- h : time step



Explicit method

Discretize acceleration using finite difference:

$$\mathbf{M} \frac{\mathbf{v}_{n+1} - \mathbf{v}_n}{h} = \mathbf{f}_{\text{int}}(\mathbf{x}_n) + \mathbf{f}_{\text{ext}}$$

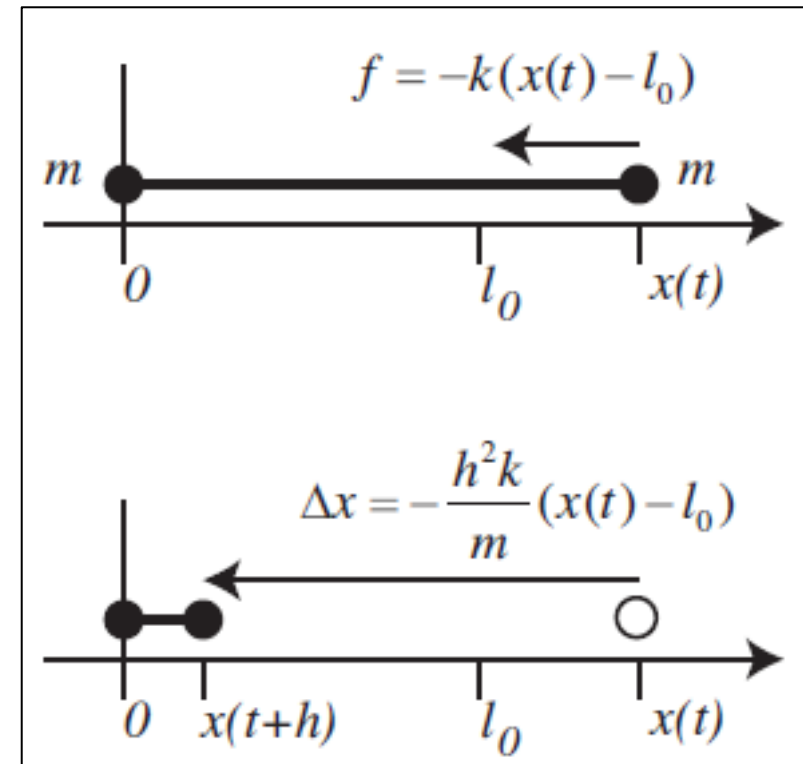
Forward Euler

$$\begin{aligned}\mathbf{x}_{n+1} &= \mathbf{x}_n + h \mathbf{v}_n \\ \mathbf{v}_{n+1} &= \mathbf{v}_n + h \mathbf{M}^{-1} (\mathbf{f}_{\text{int}}(\mathbf{x}_n) + \mathbf{f}_{\text{ext}})\end{aligned}$$

Symplectic Euler

$$\begin{aligned}\mathbf{x}_{n+1} &= \mathbf{x}_n + h \mathbf{v}_{n+1} \\ \mathbf{v}_{n+1} &= \mathbf{v}_n + h \mathbf{M}^{-1} (\mathbf{f}_{\text{int}}(\mathbf{x}_n) + \mathbf{f}_{\text{ext}})\end{aligned}$$

- Pro: easy to compute
- Con: overshooting
 - With larger time steps, mass can easily go beyond the initial amplitude
 - System energy explodes over time



Implicit method: backward Euler

$$\begin{aligned}\mathbf{x}_{n+1} &= \mathbf{x}_n + h \mathbf{v}_{n+1} \\ \mathbf{v}_{n+1} &= \mathbf{v}_n + h \mathbf{M}^{-1} (\mathbf{f}_{\text{int}}(\mathbf{x}_{n+1}) + \mathbf{f}_{\text{ext}})\end{aligned}$$

- Represent \mathbf{v}_{n+1} using unknown position \mathbf{x}_{n+1}
- Pros: can avoid overshoot
- Cons:
 - Expensive to compute (i.e. solve equation)
 - Numerical damping (energy dissipation)

Inside of backward Euler

$$\mathbf{x}_{n+1} = \mathbf{x}_n + h \mathbf{v}_{n+1}$$

$$\mathbf{v}_{n+1} = \mathbf{v}_n + h \mathbf{M}^{-1} (\mathbf{f}_{\text{int}}(\mathbf{x}_{n+1}) + \mathbf{f}_{\text{ext}})$$

$$= \mathbf{x}_n + h \mathbf{v}_n + h^2 \mathbf{M}^{-1} (\mathbf{f}_{\text{int}}(\mathbf{x}_{n+1}) + \mathbf{f}_{\text{ext}})$$

$$= \mathbf{x}_n + h \mathbf{v}_n + h^2 \mathbf{M}^{-1} (-\nabla E(\mathbf{x}_{n+1}) + \mathbf{f}_{\text{ext}})$$

Denote unknown \mathbf{x}_{n+1} as \mathbf{y}

$$h^2 \nabla E(\mathbf{y}) + \mathbf{M} \mathbf{y} - \mathbf{M}(\mathbf{x}_n + h \mathbf{v}_n) - h^2 \mathbf{f}_{\text{ext}} = \mathbf{0}$$

$\mathbf{F}(\mathbf{y})$

- Reduce to root-finding problem of function $\mathbf{F}: \mathbb{R}^{3N} \mapsto \mathbb{R}^{3N}$

→ Newton's method:

$$\mathbf{y}^{(i+1)} \leftarrow \mathbf{y}^{(i)} - \left(\frac{d\mathbf{F}}{d\mathbf{y}} \right)^{-1} \mathbf{F}(\mathbf{y}^{(i)})$$

$$= \mathbf{y}^{(i)} - \left(h^2 \mathcal{H}_E(\mathbf{y}^{(i)}) + \mathbf{M} \right)^{-1} \mathbf{F}(\mathbf{y}^{(i)})$$

2nd derivative of potential E (Hessian matrix)

- Coefficient matrix of large linear system changes at every iteration

→ high computational cost!

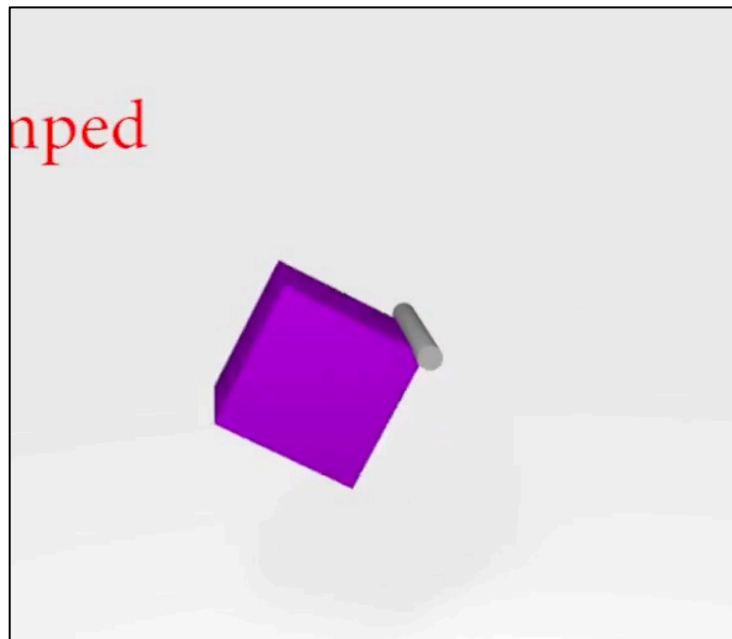
Recent paper: Energy conservation through combination of forward Euler & backward Euler

Implicit Midpoint

$$\mathbf{x}_{n+1} = \mathbf{x}_n + h \frac{\mathbf{v}_n + \mathbf{v}_{n+1}}{2}$$
$$\mathbf{v}_{n+1} = \mathbf{v}_n + h \mathbf{M}^{-1} \mathbf{f} \left(\frac{\mathbf{x}_n + \mathbf{x}_{n+1}}{2} \right)$$

Backward Euler

Proposed method



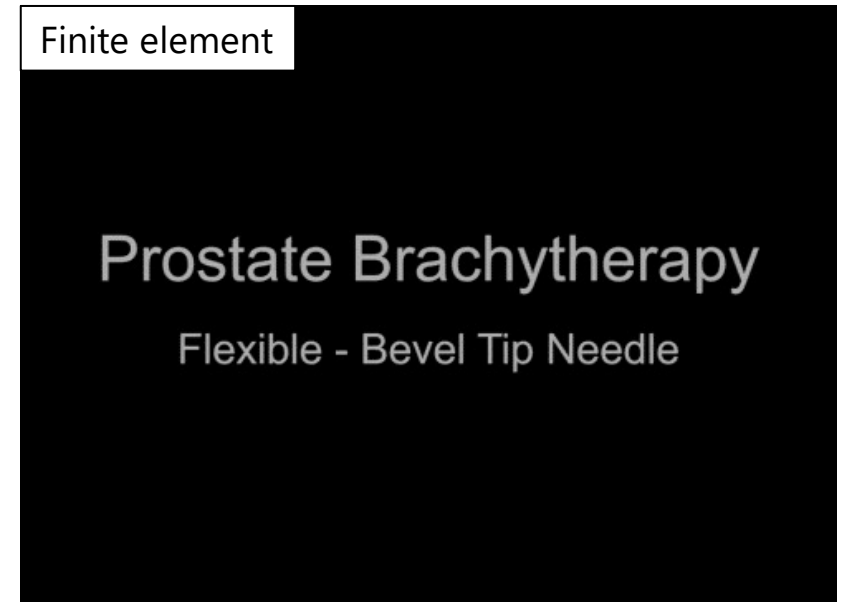
☹ Energy increase

☹ Energy decrease

☺ Energy conservation

Mass spring vs. Finite elements

- Both define potential as sum of deformations of small elements
 - Differ only in the measure of deformations



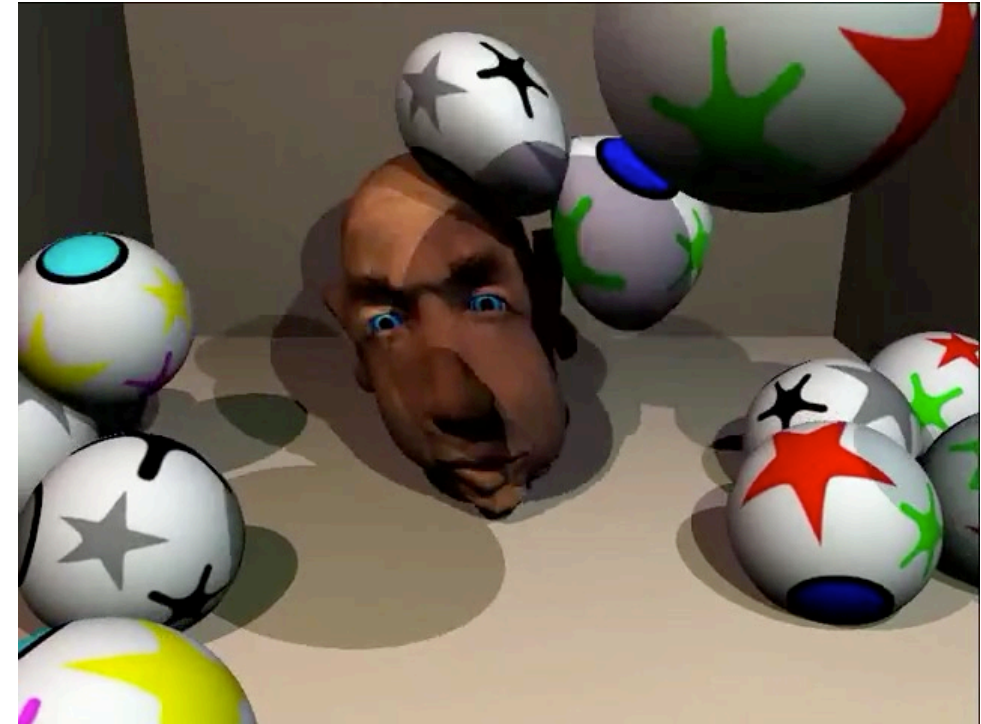
Physical accuracy	△	○
Compute/impl cost	○	△

Physics-based animation of deforming objects

- Classic: faithful simulation of physical phenomena
 - Mass-spring system
 - Finite Element Method (FEM)
- CG-specific: plausible & robust, but not faithful simulation
 - Shape Matching (Position-Based Dynamics)

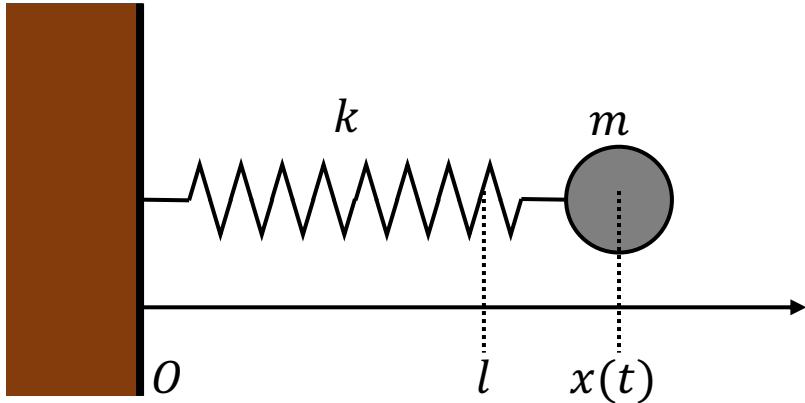
Shape Matching: physics animation method tailored for CG

- Meshless deformations based on shape matching
[Müller, Heidelberger, Techner, Gross, SIGGRAPH 2005]
- Features
 - Unconditionally stable
 - Goes back to rest state no matter what
 - Easy to compute/implement
- Caveat: cannot be interpreted as *real* physics



➔ Perfectly suitable for CG

Case of single mass & spring (no ext. force)



Symplectic Euler

$$v_{n+1} \leftarrow v_n + \frac{h k}{m} (l - x_n)$$
$$x_{n+1} \leftarrow x_n + h v_{n+1}$$

Shape Matching

$$v_{n+1} \leftarrow v_n + \frac{\alpha}{h} (l - x_n)$$
$$x_{n+1} \leftarrow x_n + h v_{n+1}$$

- $0 \leq \alpha \leq 1$ is "stiffness" parameter unique to PBD
 - $\alpha = 0 \rightarrow$ No update of velocity (spring is infinitely soft)
 - $\alpha = 1 \rightarrow$ Spring is infinitely stiff (?)
 - \rightarrow Energy never explodes in any case 😊
- Note: Unit of α/h is $(\text{time})^{-1} \rightarrow \alpha$ has no physical meaning!
 - Reason why PBD is called non physics-based but geometry-based

Case of general deforming shape (no ext. force)

Explicit Euler

$$\mathbf{v}_{n+1} \leftarrow \mathbf{v}_n - h \mathbf{M}^{-1} \nabla E(\mathbf{x}_n)$$

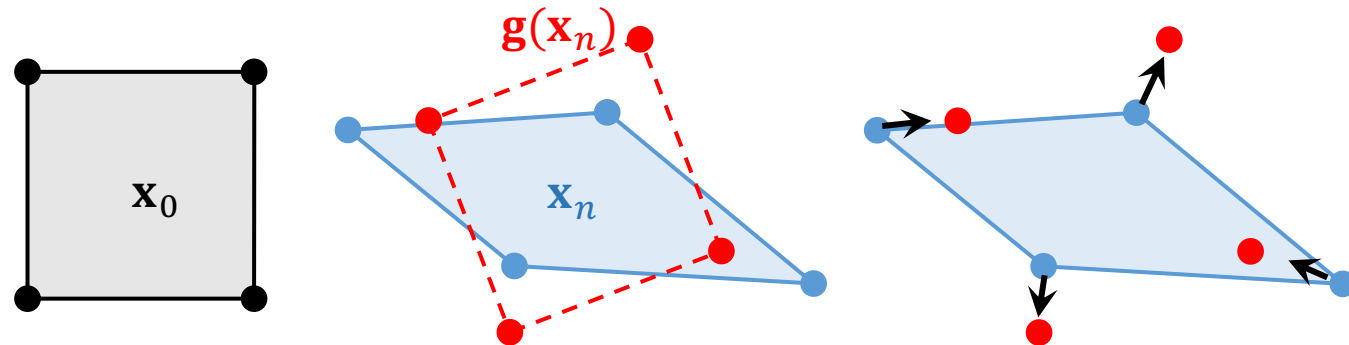
$$\mathbf{x}_{n+1} \leftarrow \mathbf{x}_n + h \mathbf{v}_{n+1}$$

Shape Matching

$$\mathbf{v}_{n+1} \leftarrow \mathbf{v}_n + \frac{\alpha}{h} (\mathbf{g}(\mathbf{x}_n) - \mathbf{x}_n)$$

$$\mathbf{x}_{n+1} \leftarrow \mathbf{x}_n + h \mathbf{v}_{n+1}$$

- Goal position \mathbf{g}
 - Rest shape rigidly transformed such that it best matches the current deformed state
 - (SVD of moment matrix)



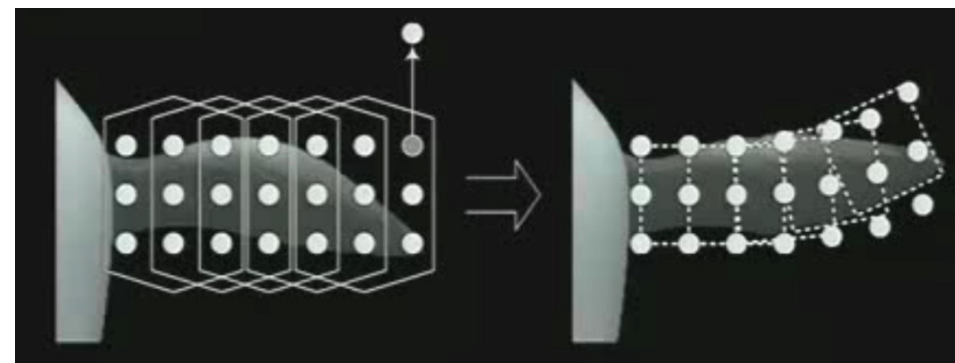
- Called "Shape Matching"
 - One technique within the PBD framework
 - Connectivity info (spring/mesh) not needed \rightarrow *meshless*

Shape Matching per local regions

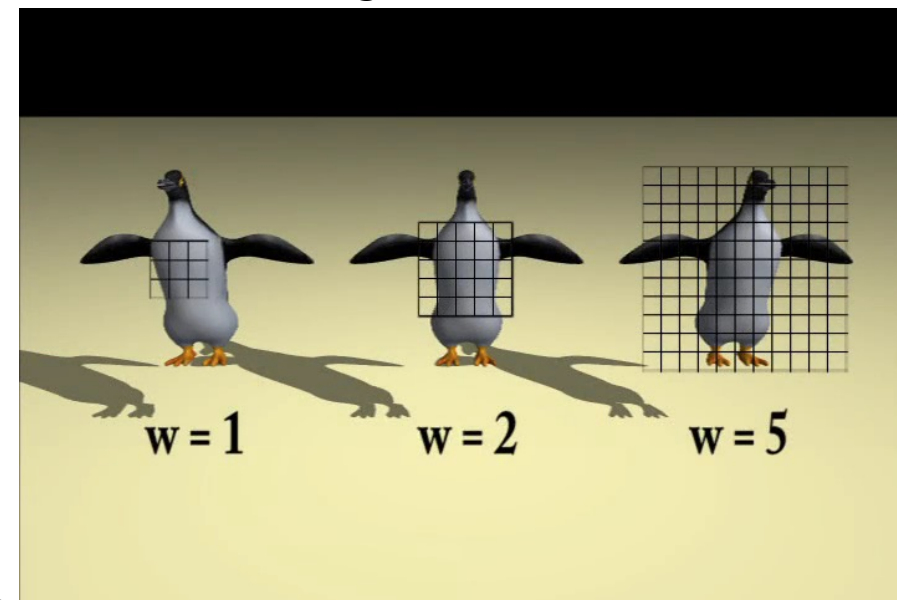
- To allow more complex deformations



Represent shape as overlapped local regions

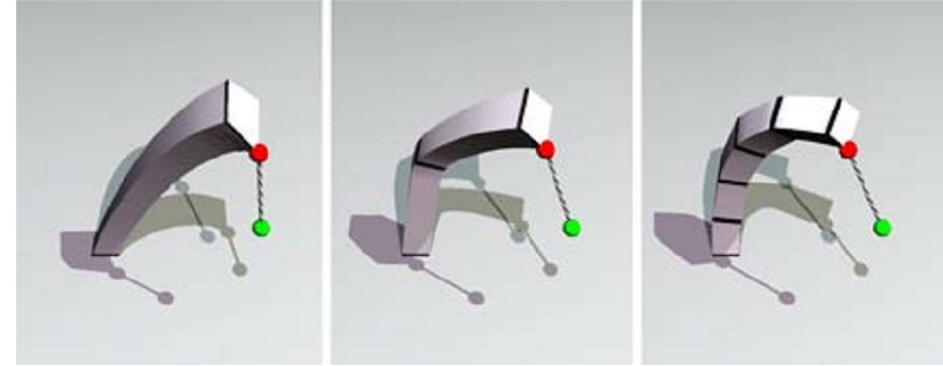


Size of local region determines stiffness

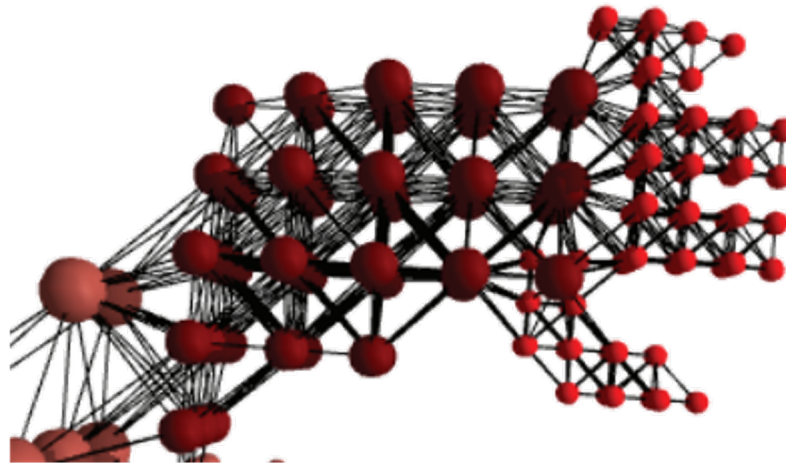


Shape Matching per (overlapping) local region

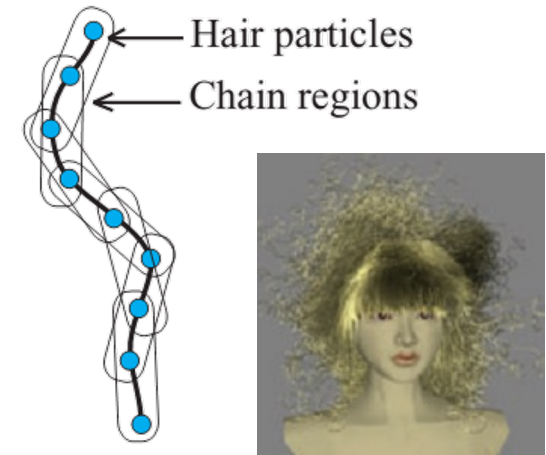
- More complex deformations
- Acceleration techniques



Local regions from voxel lattice



Local regions from octree



Animating hair using
1D chain structure



FastLSM; fast lattice shape matching for robust real-time deformation [Rivers SIGGRAPH07]

Fast adaptive shape matching deformations [Steinemann SCA08]

Chain Shape Matching for Simulating Complex Hairstyles [Rungjiratananon CGF10]

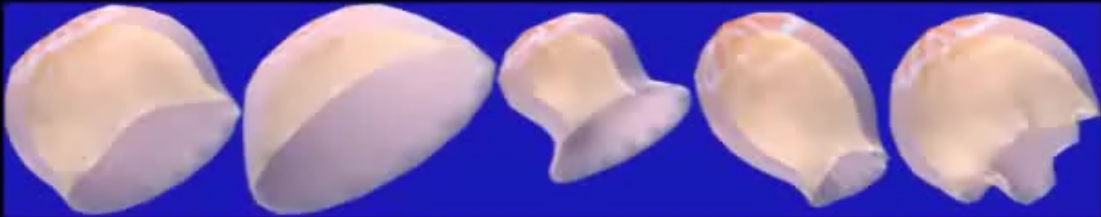
Examples of applying this idea

Self-actuated soft objects



ProcDef: Local-to-global Deformation for Skeleton-free Character Animation

Takashi Ijiri^{1,2}, Kenshi Takayama², Hideo Yokota¹, Takeo Igarashi^{2,3}
¹Riken, ²The University of Tokyo, ³JST ERATO



Soft objects whose deformation behavior can be specified via examples

SCA 2012

Real-Time Example-Based Elastic Deformation

Yuki Koyama¹
Kenshi Takayama^{1, 2}
Nobuyuki Umetani¹
Takeo Igarashi^{1, 3}

¹The University of Tokyo
²ETH Zürich
³JST ERATO

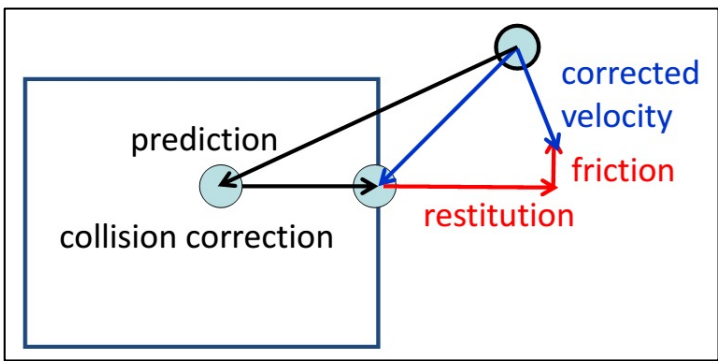
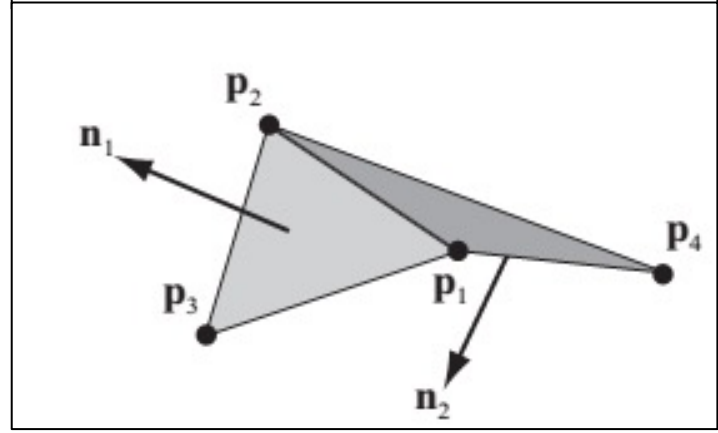
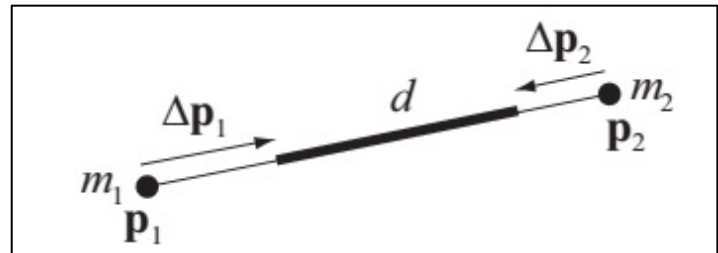


Position-Based Dynamics (PBD)

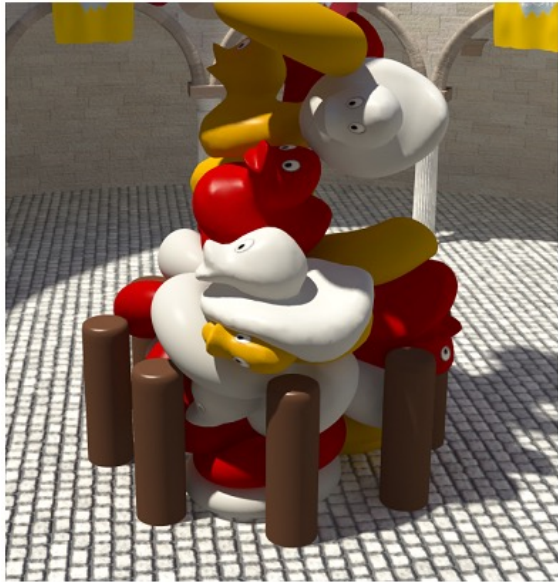
- General framework including Shape Matching
- Input: initial position \mathbf{x}_0 & velocity \mathbf{v}_0
- At every frame:

\mathbf{p}	$= \mathbf{x}_n + h \mathbf{v}_n$	prediction
\mathbf{x}_{n+1}	$= \text{modify}(\mathbf{p})$	position correction
\mathbf{u}	$= (\mathbf{x}_{n+1} - \mathbf{x}_n)/h$	velocity update
\mathbf{v}_{n+1}	$= \text{modify}(\mathbf{u})$	velocity correction

(My understanding is still weak)



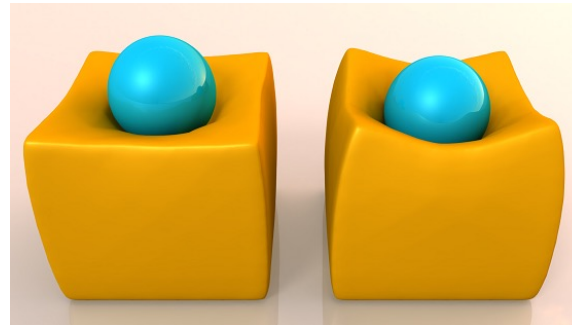
Various geometric constraints available in PBD (other than Shape Matching)



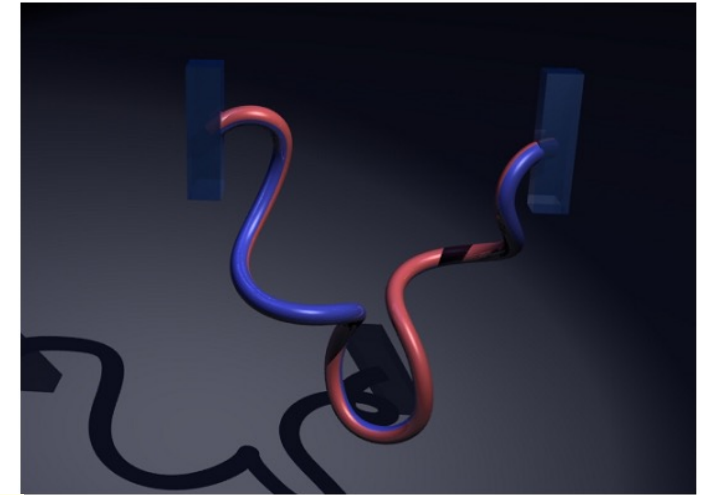
Volume constraint



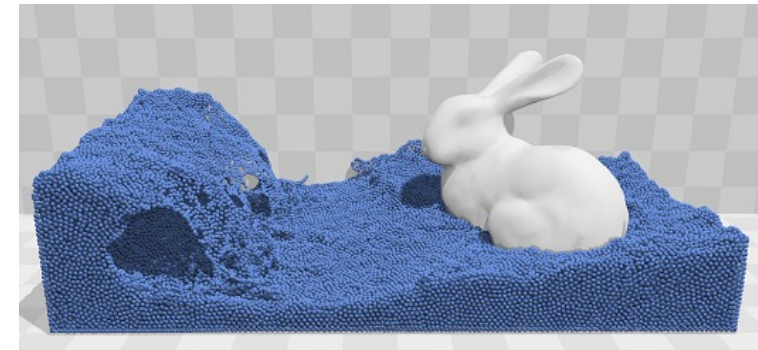
Stretch constraint



Strain constraint



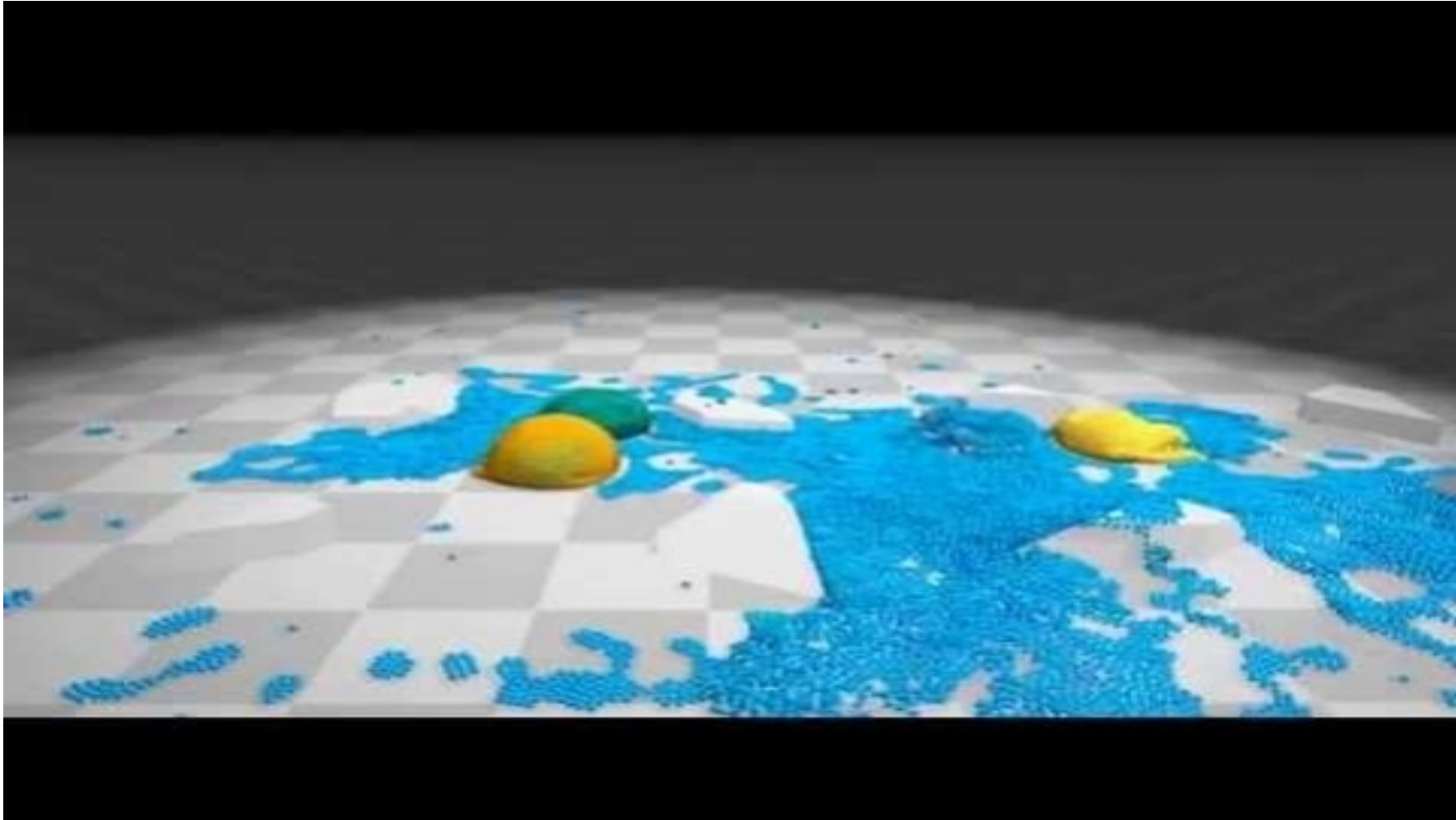
Twist constraint



Density constraint

Robust Real-Time Deformation of Incompressible Surface Meshes [Dziol SCA11]
Long Range Attachments - A Method to Simulate Inextensible Clothing in Computer Games [Kim SCA12]
Position Based Fluids [Macklin SIGGRAPH13]
Position-based Elastic Rods [Umetani SCA14]
Position-Based Simulation of Continuous Materials [Bender Comput&Graph14]

Putting everything together: FLEX in PhysX

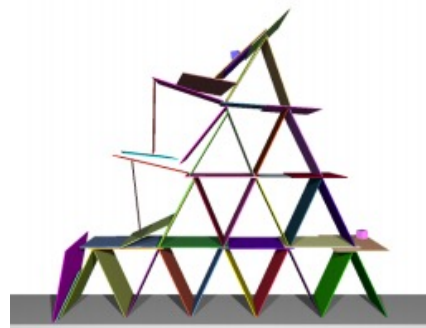


- SDK released by NVIDIA!

<https://www.youtube.com/watch?v=z6dAahLUbZg>

Collisions

- Another tricky issue
- Popular methods in PBD:
 - For each voxel grid, record which particles it contains
 - Test collisions only among nearby particles
- Recent method specialized for PBD



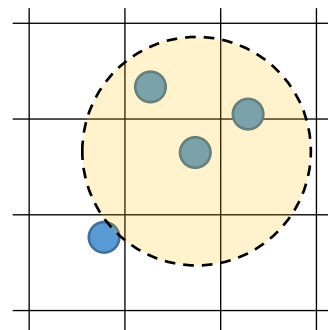
[Kaufman08]



[Harmon09]



[Zheng12]



[Muller15]

Collision detection for deformable objects [Teschner CGF05]
Staggered Projections for Frictional Contact in Multibody Systems [Kaufman SIGGRAPHAsia08]
Asynchronous Contact Mechanics [Harmon SIGGRAPH09]
Energy-based Self-Collision Culling for Arbitrary Mesh Deformations [Zheng SIGGRAPH12]
Air Meshes for Robust Collision Handling [Muller SIGGRAPH15]

Pointers

- Surveys, tutorials
 - A Survey on Position-Based Simulation Methods in Computer Graphics [Bender CGF14]
 - http://www.csee.umbc.edu/csee/research/vangogh/I3D2015/matthias_muller_slides.pdf
 - Position-Based Simulation Methods in Computer Graphics [Bender EG15Tutorial]
 - http://www.tkim.graphics/DYNAMIC_DEFORMABLES/ (recommended)
- Libraries, implementations
 - <https://code.google.com/p/opencloth/>
 - <http://shapeop.org/>
 - <http://matthias-mueller-fischer.ch/demos/matching2dSource.zip>
 - <https://bitbucket.org/yukikoyama>
 - <https://developer.nvidia.com/physx-flex>
 - <https://github.com/janbender/PositionBasedDynamics>
 - <https://github.com/InteractiveComputerGraphics/PositionBasedDynamics>