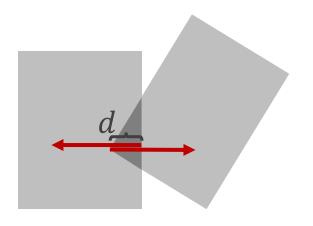
XPBD Extended Position Based Dynamics

Matthias Müller, Ten Minute Physics

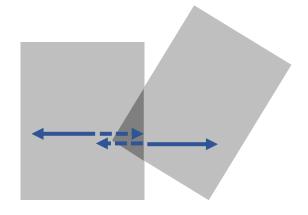
matthiasmueller.info/tenMinutePhysics

Motivation

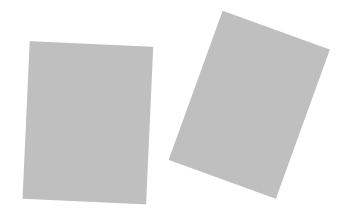
Force Based Simulation



penetration causes force f = kd



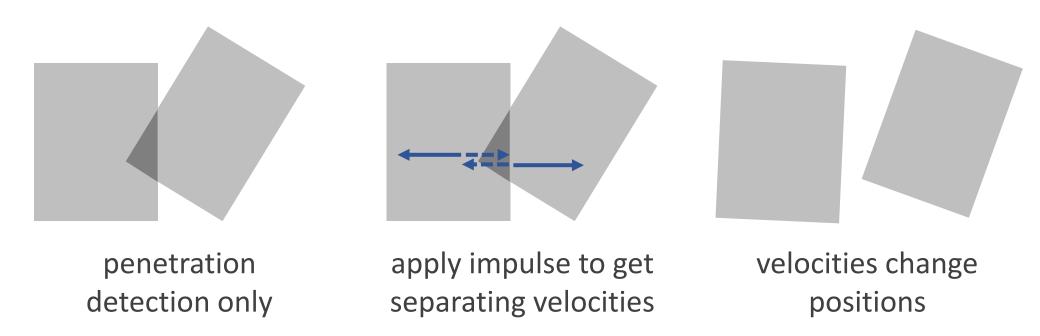
forces change velocities



velocities change positions

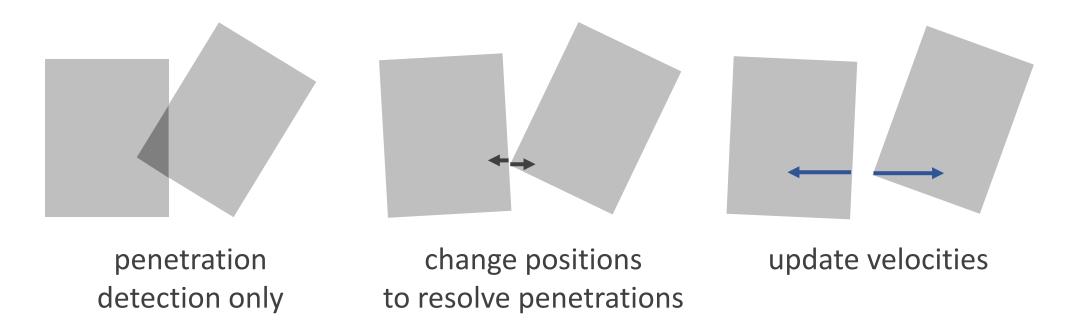
- Overlap needed
- Reaction lag
- Large stiffness $k \rightarrow$ stability problems, overshooting
- Small stiffness → squishy system
- How to choose k for a hard constraint?

Impulse Based Simulation



- More stable
- Controlled velocity update, no overshooting
- Drift: consistent velocities do not guarantee consistent positions!

Position Based Simulation



- Controlled position change, unconditionally stable
- No drift

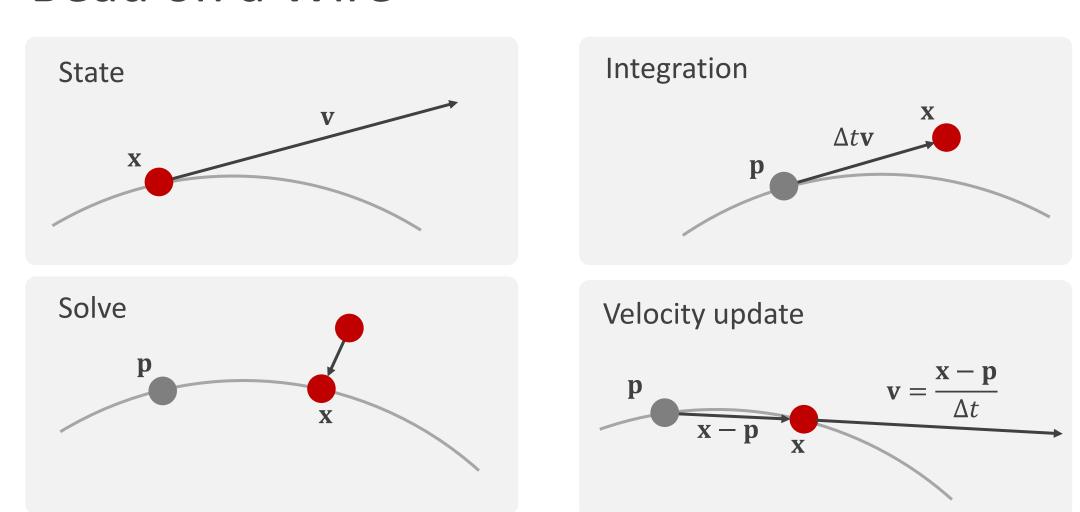
Physical? Accurate?

- We found that PBD is very closely related to implicit Euler integration
- To be precise, it corresponds to...

... the first iteration of the Newton minimization of a backward Euler integration step in variational position-based form, using the non-linear Gauss-Seidel method, where the Newton solution is initialized with the unconstrained / predicted / inertial position using external forces.

- Original PBD is unphysical only in the way it handles softness
- Problem fixed with XPBD

Bead on a Wire



PBD = integrator and solver!

PBD Algorithm

```
while simulating
for all particles i
\mathbf{v}_i \leftarrow \mathbf{v}_i + \Delta t \mathbf{g}
\mathbf{p}_i \leftarrow \mathbf{x}_i
\mathbf{x}_i \leftarrow \mathbf{x}_i + \Delta t \mathbf{v}_i
```

for all constraints C solve(C, Δt)

for all particles i $\mathbf{v}_i \leftarrow (\mathbf{x}_i - \mathbf{p}_i)/\Delta t$

```
solve(C, \Delta t):

for all particles i of C

compute \Delta \mathbf{x}_i
```

 $\mathbf{x}_i \leftarrow \mathbf{x}_i + \Delta \mathbf{x}_i$

Iterations vs. Sub-steps

```
\begin{array}{l} \textbf{while} \ \text{simulating} \\ \textbf{for all} \ \text{particles} \ i \\ \textbf{v}_i \leftarrow \textbf{v}_i + \Delta t \ \textbf{g} \\ \textbf{p}_i \leftarrow \textbf{x}_i \\ \textbf{x}_i \leftarrow \textbf{x}_i + \Delta t \ \textbf{v}_i \\ \textbf{for} \ n \ \text{iterations} \\ \textbf{for all} \ \text{constraints} \ \mathcal{C} \\ \text{solve}(\mathcal{C}, \Delta t) \\ \textbf{for all} \ \text{particles} \ i \\ \textbf{v}_i \leftarrow (\textbf{x}_i - \textbf{p}_i)/\Delta t \end{array}
```

```
\Delta t_{\rm S} \leftarrow \Delta t/n
while simulating
            for n substeps
                     for all particles i
                              \mathbf{v}_i \leftarrow \mathbf{v}_i + \Delta t_s \mathbf{g}
                               \mathbf{p}_i \leftarrow \mathbf{x}_i
                              \mathbf{x}_i \leftarrow \mathbf{x}_i + \Delta t_s \mathbf{v}_i
                     for all constraints C
                               solve(C, \Delta t_s)
                     for all particles i
                              \mathbf{v}_i \leftarrow (\mathbf{x}_i - \mathbf{p}_i)/\Delta t_s
```

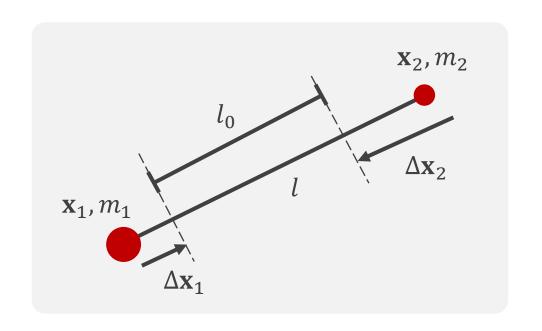
- Spending fixed time budget with sub-steps is much is more effective than with iterations!
- XPBD much simpler (no λ tracking)

Distance Constraint

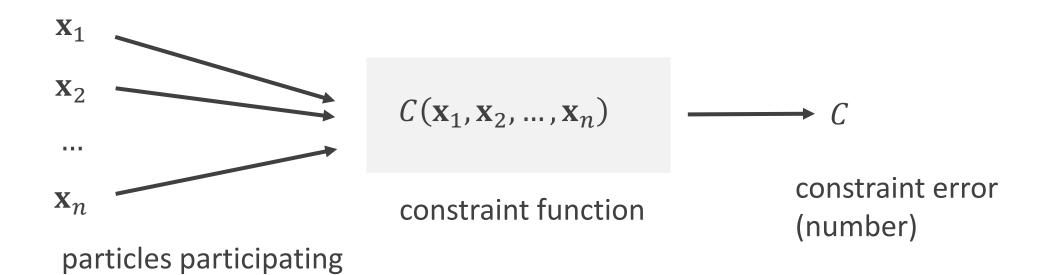
- Rest distance l_0
- Current distance l
- Masses m_i
- Inverse masses $w_i = 1/m_1$

$$\Delta \mathbf{x}_1 = \frac{w_1}{w_1 + w_2} (l - l_0) \frac{\mathbf{x}_2 - \mathbf{x}_1}{|\mathbf{x}_2 - \mathbf{x}_1|}$$

$$\Delta \mathbf{x}_2 = -\frac{w_2}{w_1 + w_2} (l - l_0) \frac{\mathbf{x}_2 - \mathbf{x}_1}{|\mathbf{x}_2 - \mathbf{x}_1|}$$



General Constraint



For the distance constraint:

in constraint

$$C_{\text{dist}}(\mathbf{x}_1, \mathbf{x}_2) = |\mathbf{x}_2 - \mathbf{x}_1| - l_0$$

Constraint Gradient ∇*C*

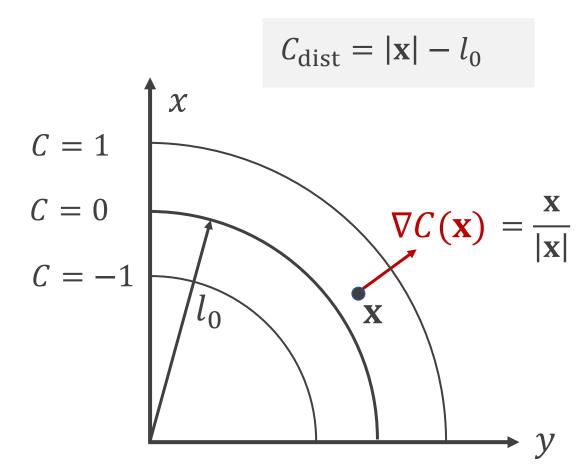
 $\nabla C(\mathbf{x})$ is a vector

Vector direction:

direction in which
 C increases the most

Vector length:

how much C changes
 when moving x by one unit



Solving a General Constraint (PBD)

Compute the scalar value λ (same for all participating particles):

$$\lambda = \frac{-C}{w_1 |\nabla C_1|^2 + w_2 |\nabla C_2|^2 + \dots + w_n |\nabla C_n|^2}$$

 ∇C_i : How to move \mathbf{x}_i for a maximal increase of C

 $|\nabla C_i|^2$ the squared length of ∇C_i

Compute correction for point x_i as:

$$\Delta \mathbf{x}_i = \lambda w_i \; \nabla C_i$$

Making the Constraint Soft

PBD:

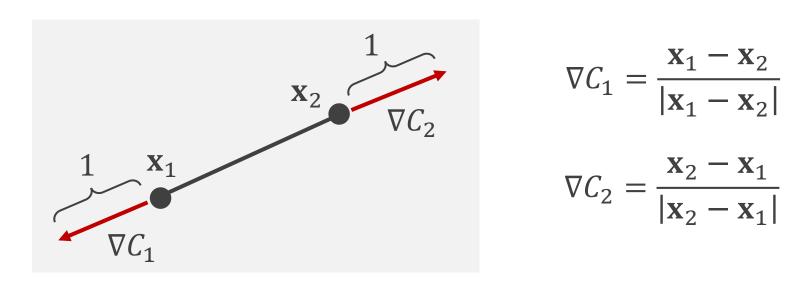
- Scale the correction as $\Delta \mathbf{x}_i = \mathbf{k} \lambda w_i \nabla C_i$
- Stiffness $k \in [0,1]$
- Easy to tune!
- Dependent on time step (stiffer for smaller time steps)

XPBD:

$$\lambda = \frac{-C}{w_1 |\nabla C_1|^2 + w_2 |\nabla C_2|^2 + \cdots + w_n |\nabla C_n|^2 + \frac{\alpha}{\Delta t^2}}$$

- Compliance α is the inverse of physical stiffness
- Infinitely stiff (hard) when $\alpha = 0$

Example: Distance Constraint



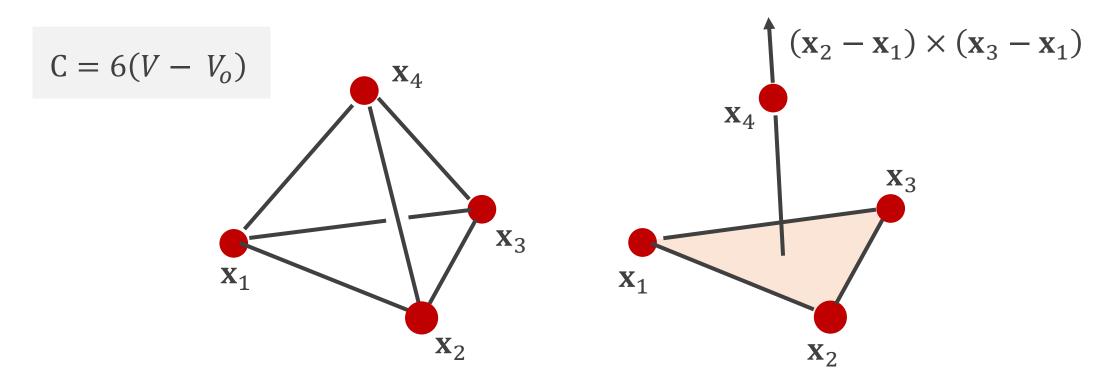
$$\nabla C_1 = \frac{\mathbf{x}_1 - \mathbf{x}_2}{|\mathbf{x}_1 - \mathbf{x}_2|}$$

$$\nabla C_2 = \frac{\mathbf{x}_2 - \mathbf{x}_1}{|\mathbf{x}_2 - \mathbf{x}_1|}$$

$$\lambda = \frac{-C}{w_1 |\nabla C_1|^2 + w_2 |\nabla C_2|^2 + \dots + w_n |\nabla C_n|^2} = \frac{-(l - l_0)}{w_1 \cdot 1 + w_2 \cdot 1}$$

$$\Delta \mathbf{x}_1 = \lambda w_1 \nabla C_1 = -\frac{w_1}{w_1 + w_2} (l - l_0) \frac{\mathbf{x}_2 - \mathbf{x}_1}{|\mathbf{x}_2 - \mathbf{x}_1|}$$

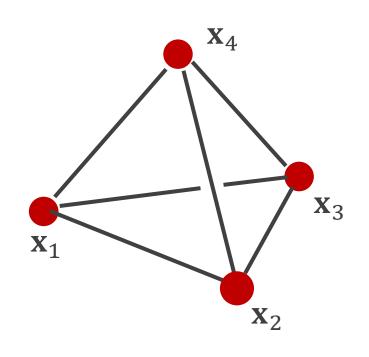
Volume Conservation Constraint



$$C = 6(V - V_o) = [(\mathbf{x}_2 - \mathbf{x}_1) \times (\mathbf{x}_3 - \mathbf{x}_1)] \cdot (\mathbf{x}_4 - \mathbf{x}_1) - 6V_o$$

$$\nabla_4 C = (\mathbf{x}_2 - \mathbf{x}_1) \times (\mathbf{x}_3 - \mathbf{x}_1)$$

Solve



Right hand rule

$$\nabla_1 \mathbf{C} = (\mathbf{x}_4 - \mathbf{x}_2) \times (\mathbf{x}_3 - \mathbf{x}_2)$$

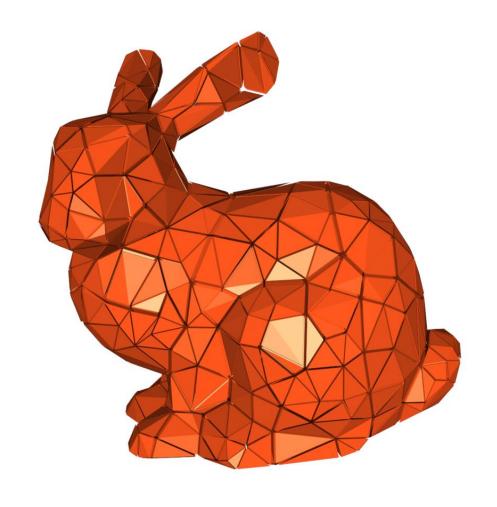
$$\nabla_2 C = (\mathbf{x}_3 - \mathbf{x}_1) \times (\mathbf{x}_4 - \mathbf{x}_1)$$

$$\nabla_3 C = (\mathbf{x}_4 - \mathbf{x}_1) \times (\mathbf{x}_2 - \mathbf{x}_1)$$

$$\nabla_4 C = (\mathbf{x}_2 - \mathbf{x}_1) \times (\mathbf{x}_3 - \mathbf{x}_1)$$

$$\lambda = \frac{-6(V - V_o)}{w_1 |\nabla C_1|^2 + w_2 |\nabla C_2|^2 + w_3 |\nabla C_3|^2 + w_4 |\nabla C_4|^2}$$

$$\Delta \mathbf{x}_i = \lambda \, w_i \, \nabla C_i$$



→ See soft body tutorial for action!