

CSC 433/533

Computer Graphics

Alon Efrat
Credit: Joshua Levine

Lecture 24

Animation 2

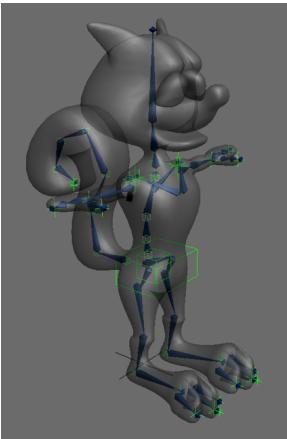
Nov. 25, 2019

Today's Agenda

- Reminders:
 - A06 questions?
- Goals for today: wrap up character animation and then introduce concepts in particle-based animation

Character Animation

Rigged character



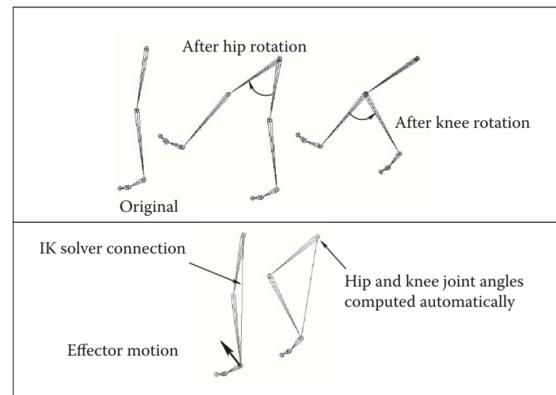
[CIS 565 staff]

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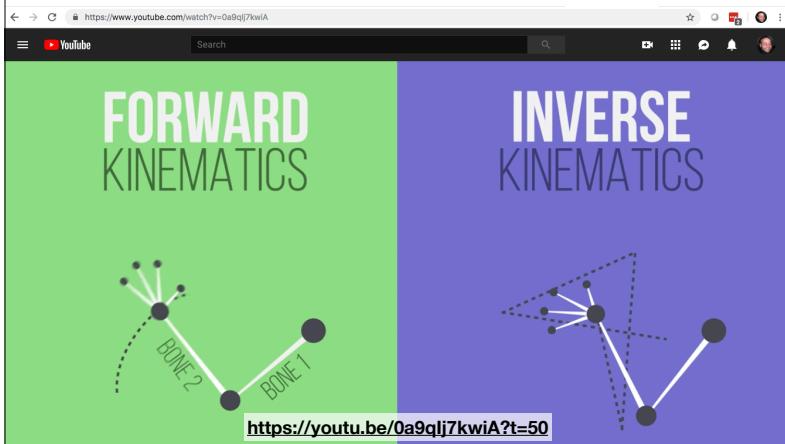
- Surface is deformed by a set of *bones*
- Bones are in turn controlled by a smaller set of *controls*
- The controls are useful, intuitive DOFs for an animator to use

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Forward vs. Inverse Kinematics

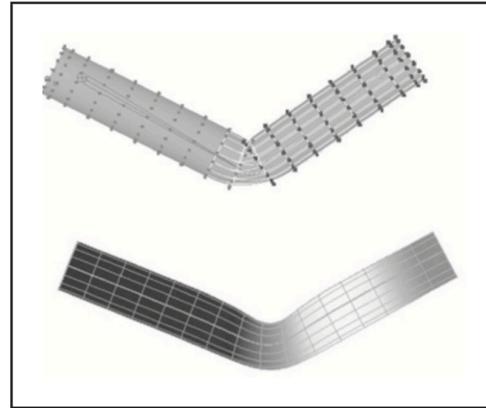


Inverse Kinematics Solves for all Intermediate Constraints



Skinning

- After solving for the skeleton, one still needs to update and deform the surface



Mesh skinning math: setup

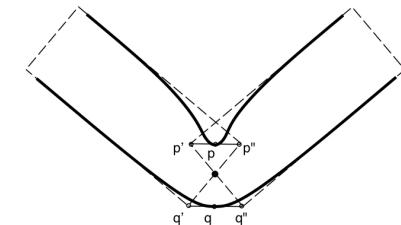
- Surface has control points \mathbf{p}_i
 - Triangle vertices, spline control points, subdiv base vertices
- Each bone has a transformation matrix M_j
 - Normally a rigid motion
- Every point–bone pair has a weight w_{ij}
 - In practice only nonzero for small # of nearby bones
 - The weights are provided by the user

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Mesh skinning math

- Deformed position of a point is a weighted sum
 - of the positions determined by each bone's transform alone
 - weighted by that vertex's weight for that bone



[Lewis et al., SIGGRAPH 2000]

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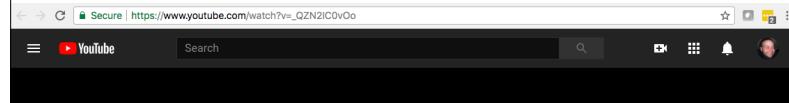
Skinning Mesh Animations

Doug L. James
Christopher D. Twigg

Carnegie Mellon University

<http://graphics.cs.cmu.edu/projects/sma/>, 2005

Motion Capture Can Be Used for Data-Driven Methods



BEYOND
TWO SOULS.TM

https://youtu.be/_QZN2IC0vOo

Physics-Based Animation

Animation vs. Simulation

- Animation methods use scripted actions to make objects change
- Simulation: simulate physical laws by associating physical properties to objects
- Solve for physics to achieve (predict) realistic effects

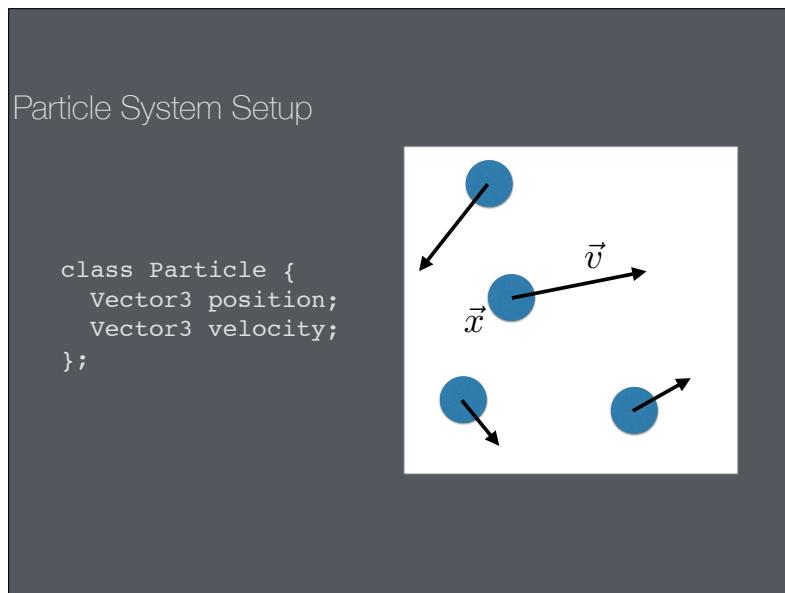
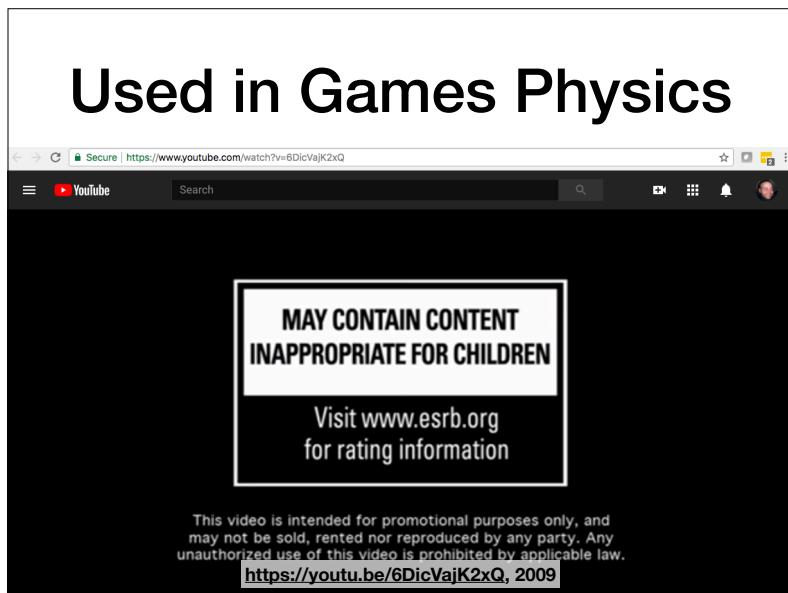
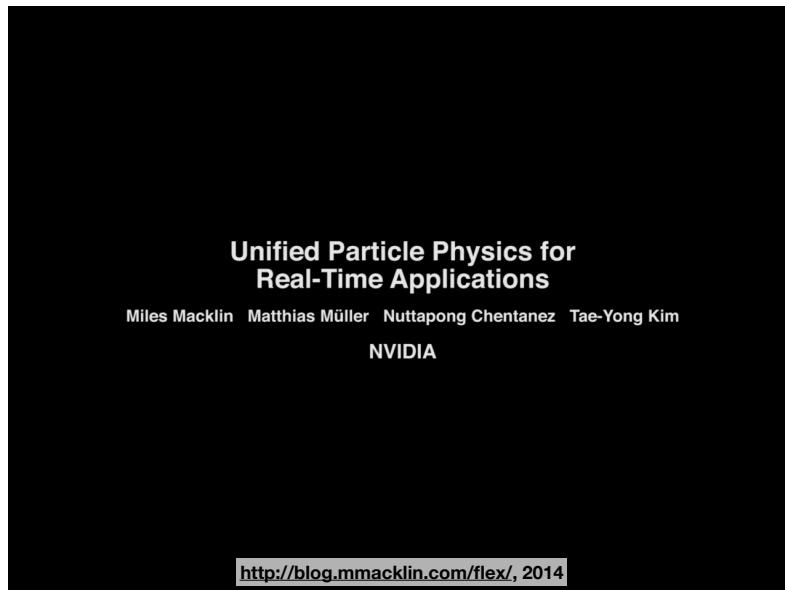
Using Particle Systems

- Idea: Represent the physics on the simplest possible entity: particles
- Used for effects like smoke, fire, water, sparks, and more
- Plenty of other approaches, this is just one family

PARTICLE DREAMS

Karl Sims
Optomystic

<http://www.karlsims.com/particle-dreams.html>, 1988



Moving Particles

Position is a function of time

- i.e., $\vec{x} \equiv \vec{x}(t)$
- Note that $\vec{v}(t) \equiv \frac{\partial \vec{x}}{\partial t}$

Use a function to control the particle's velocity

- $\vec{v}(t) = f(\vec{x}(t))$

This is an Ordinary Differential Equation (ODE)

Solve this ODE at every frame

- i.e., solve for $\vec{x}(t_0), \vec{x}(t_1), \vec{x}(t_2), \dots$
- Then we can draw each of these positions to the screen

A Simple Example

Let \vec{v} be constant

- e.g., $\vec{v} = f(\vec{x}) = (0, 0, 1)^\top$

Then we can solve for the position at any time:

- $\vec{x}(t) = \vec{x}(0) + t\vec{v}$

Not always so easy

- $f(\vec{x})$ can be anything!
- Might be unknown until runtime (e.g., user interaction)
- Often times, not solved exactly

Moving Particles, Revisited

Now, acceleration is in the mix

- $\vec{a}(t) \equiv \frac{\partial \vec{v}}{\partial t} \equiv \frac{\partial^2 \vec{x}}{\partial t^2}$

Use a function to control the particle's acceleration

- $\vec{a}(t) = f(\vec{x}(t))$

This is a Second Order ODE

Solve this ODE at every frame, same as before

- Can sometimes be reduced to a first order ODE
- Calculate position and velocity together

Physically-based Motion

Acceleration based on Newton's laws

- $\vec{f}(t) = m\vec{a}(t)$...or, equivalently... $\vec{a}(t) = \vec{f}(t)/m$
- i.e., force is mass times acceleration

Forces are known beforehand

- e.g., gravity, springs, others....
- Multiple forces sum together
- These often depend on the position, i.e., $\vec{f}(t) \equiv \vec{f}(\vec{x}(t))$
- Sometimes velocity, too

If we know the values of the forces, we can solve for particle's state

Unary Forces

Constant

- Gravity

Position/Time-Dependent

- Force fields, e.g. wind

Velocity-Dependent

- Drag

Ordinary Differential Equations

$$\frac{d \mathbf{X}(t)}{dt} = f(\mathbf{X}(t), t)$$

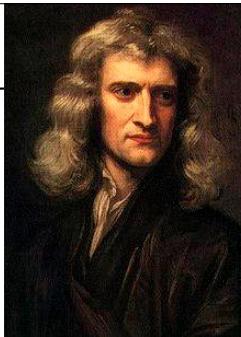
- Given a function $f(\mathbf{X}, t)$ compute $\mathbf{X}(t)$
- Typically, *initial value problems*:
 - Given values $\mathbf{X}(t_0) = \mathbf{X}_0$
 - Find values $\mathbf{X}(t)$ for $t > t_0$
- We can use lots of standard tools

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Newtonian Mechanics

- Point mass: 2nd order ODE

$$\vec{F} = m\vec{a} \quad \text{or} \quad \vec{F} = m \frac{d^2 \vec{x}}{dt^2}$$



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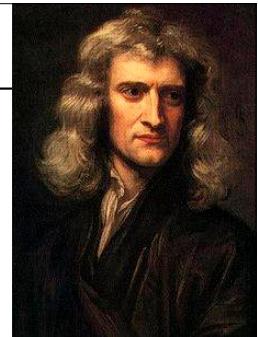
- Position \mathbf{x} and force \mathbf{F} are vector quantities
 - We know \mathbf{F} and m , want to solve for \mathbf{x}
- You have all seen this a million times before

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Reduction to 1st Order

- Point mass: 2nd order ODE

$$\vec{F} = m\vec{a} \quad \text{or} \quad \vec{F} = m \frac{d^2 \vec{x}}{dt^2}$$



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Source: Wikimedia Commons.

- Corresponds to system of first order ODEs

$$\begin{cases} \frac{d}{dt} \vec{x} = \vec{v} \\ \frac{d}{dt} \vec{v} = \vec{F}/m \end{cases}$$

2 unknowns (\mathbf{x}, \mathbf{v}) instead of just \mathbf{x}

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Reduction to 1st Order

$$\begin{cases} \frac{d}{dt} \vec{x} = \vec{v} \\ \frac{d}{dt} \vec{v} = \vec{F}/m \end{cases}$$

2 variables (\mathbf{x}, \mathbf{v}) instead of just one

- Why reduce?

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Reduction to 1st Order

$$\begin{cases} \frac{d}{dt} \vec{x} = \vec{v} \\ \frac{d}{dt} \vec{v} = \vec{F}/m \end{cases}$$

2 variables (\mathbf{x}, \mathbf{v}) instead of just one

- Why reduce?
 - Numerical solvers grow more complicated with increasing order, can just write one 1st order solver and use it
 - Note that this doesn't mean it would always be easy :-)

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Notation

- Let's stack the pair (\mathbf{x}, \mathbf{v}) into a bigger state vector \mathbf{X}

$$\mathbf{X} = \begin{pmatrix} \vec{x} \\ \vec{v} \end{pmatrix}$$

For a particle in 3D, state vector \mathbf{X} has 6 numbers

$$\frac{d}{dt} \mathbf{X} = f(\mathbf{X}, t) = \begin{pmatrix} \vec{v} \\ \vec{F}(x, v)/m \end{pmatrix}$$

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Now, Many Particles

- We have N point masses
 - Let's just stack all xs and vs in a big vector of length 6N

$$\mathbf{X} = \begin{pmatrix} \mathbf{x}_1 \\ \mathbf{v}_1 \\ \vdots \\ \mathbf{x}_N \\ \mathbf{v}_N \end{pmatrix} \quad f(\mathbf{X}, t) = \begin{pmatrix} \mathbf{v}_1 \\ \mathbf{F}^1(\mathbf{X}, t) \\ \vdots \\ \mathbf{v}_N \\ \mathbf{F}^N(\mathbf{X}, t) \end{pmatrix}$$

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Now, Many Particles

- We have N point masses
 - Let's just stack all xs and vs in a big vector of length 6N
 - \mathbf{F}^i denotes the force on particle i
 - When particles don't interact, \mathbf{F}^i only depends on \mathbf{x}_i and \mathbf{v}_i .

$$\mathbf{X} = \begin{pmatrix} \mathbf{x}_1 \\ \mathbf{v}_1 \\ \vdots \\ \mathbf{x}_N \\ \mathbf{v}_N \end{pmatrix} \quad f(\mathbf{X}, t) = \begin{pmatrix} \mathbf{v}_1 \\ \mathbf{F}^1(\mathbf{X}, t) \\ \vdots \\ \mathbf{v}_N \\ \mathbf{F}^N(\mathbf{X}, t) \end{pmatrix}$$

↑
f gives $d/dt \mathbf{X}$, remember!

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Integration Algorithm 1

Calculating Particle State from Forces: First attempt

- Use forces to update velocity: $\vec{v}(t+h) = \vec{v}(t) + \frac{h}{m} \vec{f}(t)$
- Use old velocity to update position: $\vec{x}(t+h) = \vec{x}(t) + h\vec{v}(t)$

Issues

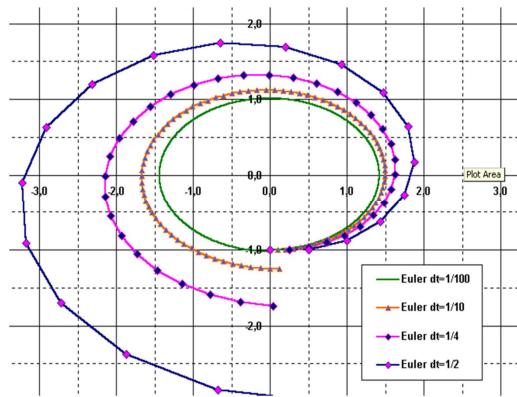
- Unstable in certain cases!
- Reducing time step can help, but this becomes computationally expensive
- Error is $O(h^2)$ per step (and accumulates!). Error is $O(h)$ globally.

This technique is called **Forward (Explicit) Euler Integration**

Example: circle

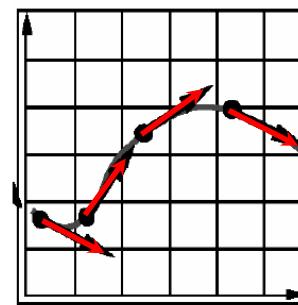
Comparison Euler, Step Sizes

Euler quality is proportional to dt



Intuitive Solution: Take Steps

- Current state \mathbf{X}
- Examine $f(\mathbf{X}, t)$ at (or near) current state
- Take a step to new value of \mathbf{X}



$$\frac{d}{dt} \mathbf{X} = f(\mathbf{X}, t)$$

$$\Rightarrow "d\mathbf{X} = dt f(\mathbf{X}, t)"$$

f = $d/dt \mathbf{X}$ is a vector that sits at each point in phase space, pointing the direction.

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Euler's Method

- Simplest and most intuitive
- Pick a **step size h**
- Given $\mathbf{X}_0 = \mathbf{X}(t_0)$, take step:

$$t_1 = t_0 + h$$

$$\mathbf{X}_1 = \mathbf{X}_0 + h f(\mathbf{X}_0, t_0)$$

- Piecewise-linear approximation to the path
- **Basically, just replace dt by a small but finite number**

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Euler, Visually

$$\frac{d}{dt} \mathbf{X} = f(\mathbf{X}, t)$$

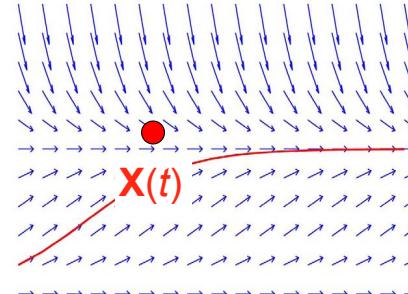


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Euler, Visually

$$\frac{d}{dt} \mathbf{X} = f(\mathbf{X}, t)$$

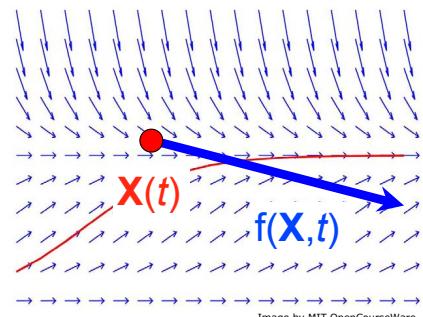


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Euler, Visually

$$\frac{d}{dt} \mathbf{X} = f(\mathbf{X}, t)$$

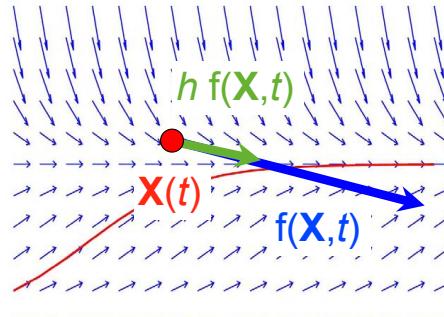
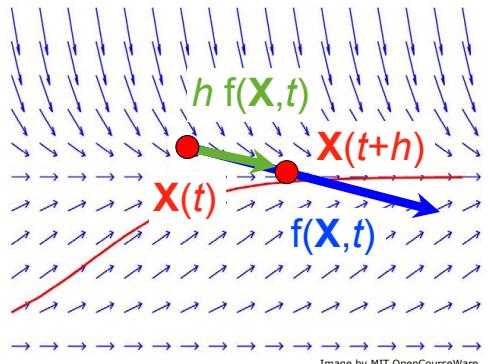


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Euler, Visually

$$\frac{d}{dt} \mathbf{X} = f(\mathbf{X}, t)$$



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Integration Algorithm 2

Another attempt

- Update velocity with forces at next time step: $\vec{v}(t+h) = \vec{v}(t) + \frac{h}{m} \vec{f}(t+h)$
- Use new velocity to update position: $\vec{x}(t+h) = \vec{x}(t) + h\vec{v}(t+h)$

Benefits

- Unconditionally stable if the system is linear!

Issues

- Solving for $\vec{f}(t+h)$ is often expensive
- Can introduce artificial viscous damping
- Error is still $O(h^2)$ per step

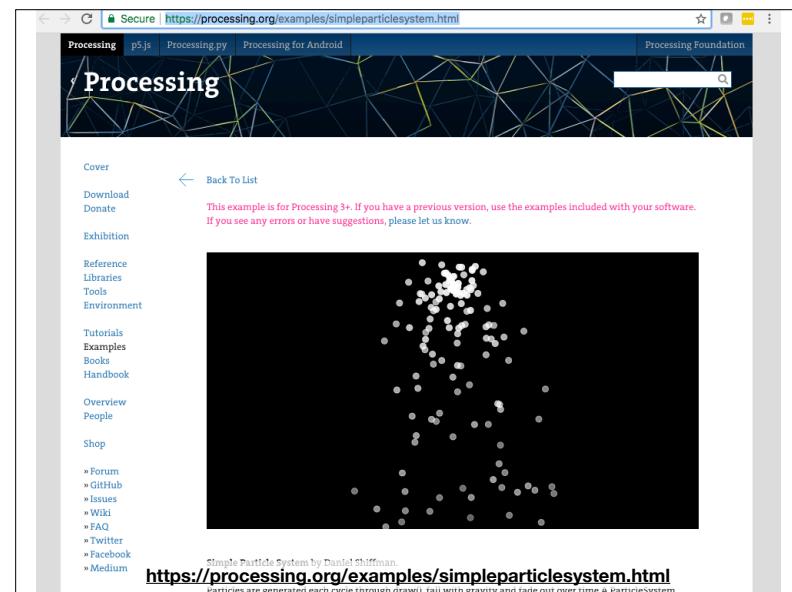
This technique is called Backward (Implicit) Euler Integration

Another Simple Example: Sprinkler

```
list<Particle> PL;
spread = 0.1; //how random the velocity is

//add k particles to the list
for (int i=0; i<k; i++) {
    Particle p;
    p->position = Vec3(0,0,0);
    p->velocity = Vec3(0,0,1) + spread*Vec3(rand(), rand(), rand());
    PL->add(p);
}

for (each time step) {
    for (each particle p in PL) {
        p->position += p->velocity*dt; //dt: time step
        p->velocity -= g*dt; //g: gravitation constant
    }
}
```



Binary, n -ary Forces

Much more interesting behaviors to be had from particles that interact

Simplest: binary forces, e.g. springs

$$\vec{f}_i(\vec{x}_i, \vec{x}_j) = -k_s(\|\vec{x}_i - \vec{x}_j\| - r_{ij}) \frac{\vec{x}_i - \vec{x}_j}{\|\vec{x}_i - \vec{x}_j\|}$$

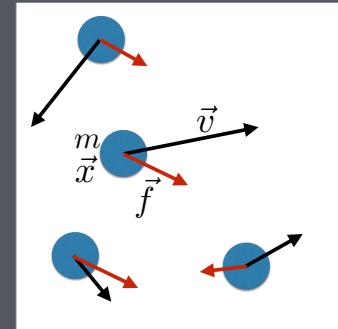
Nice example project with mass-spring systems:

- <https://vimeo.com/73188339>

More sophisticated models for deformable things use forces relating 3 or more particles

Particle System Setup, Revisited

```
class Particle {  
    float mass;  
    Vector3 position;  
    Vector3 velocity;  
    Vector3 force;  
};
```



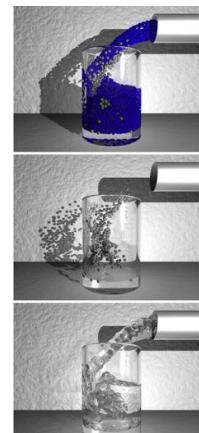
Basic Algorithm

- 1) Clear forces from previous calculations
- 2) Calculate/accumulate forces for each particle
- 3) Solve for particle's state (position, velocity) for the next time step h

Generalizations

[Müller et al. 2005](#)

- It's not all hacks:
Smoothed Particle Hydrodynamics (SPH)
 - A family of “real” particle-based fluid simulation techniques.
 - Fluid flow is described by the **Navier-Stokes Equations**, a nonlinear partial differential equation (PDE)
 - SPH discretizes the fluid as small packets (particles!), and evaluates pressures and forces based on them.



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Lec25 Required Reading

- FOCG, Ch. 16.6-16.7
- Check out paper on Boids.

Reminder: Assignment 06

Assigned: Wednesday, Nov. 13
Written Due: Monday, Nov. 25, 4:59:59 pm
Program Due: Wednesday, Nov. 27, 4:59:59 pm