

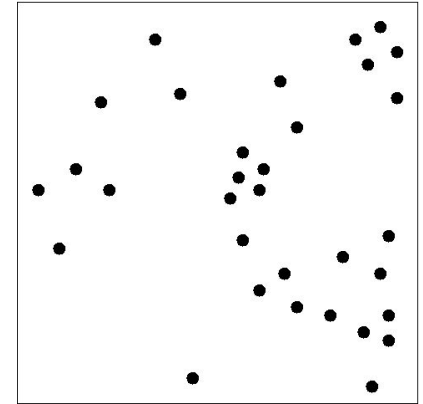
Mass-Spring Systems

Pop Worksheet!

Teams of 2. Hand in to Jeramey after we discuss.

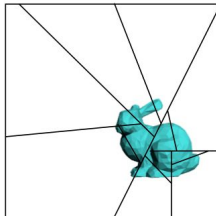
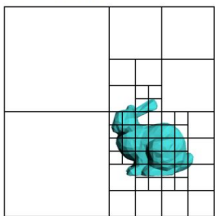
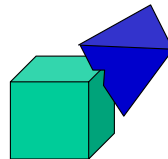
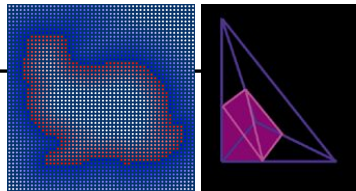
- For each adaptive grid method (quad tree, k-d tree, binary space partition) sketch the resulting grid if we split cells with > 2 elements and allow a maximum tree height of 5 (max of 4 splits from root).

*When >1 choice is available:
minimize the # of leaf nodes
and maximize the distance
from each point to the split*



Last Time?


- Implicit Surfaces & Marching Cubes/Tetras
- Collision Detection & Conservative Bounding Regions
- Spatial Acceleration Data Structures
 - Octree, k-d tree, BSF tree



Today

- **Particle Systems**
 - Equations of Motion (Physics)
 - Forces: Gravity, Spatial, Damping
 - Numerical Integration (Euler, Midpoint, etc.)
- Mass Spring System Examples
 - String, Hair, Cloth
- Stiffness
- Discretization

Types of Dynamics

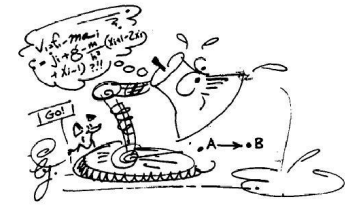
- Point 

- Rigid body 

- Deformable body
(include clothes, fluids, smoke, etc.)



Carlson, Mucha, Van Horn, & Turk 2002

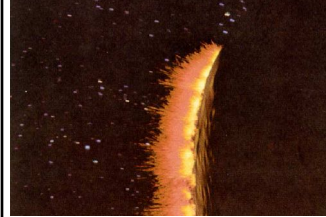


Witkin & Kass, "Spacetime Constraints", 1988.

What is a Particle System?

- Collection of many small simple particles that maintain *state* (position, velocity, color, etc.)
- Particle motion influenced by external *force fields*
- Integrate* the laws of mechanics (ODE Solvers)
- To model: sand, dust, smoke, sparks, flame, water, etc.

Star Trek, The Wrath of Kahn, 1982



Sateesh Malla, 2008,
<http://www.sateeshmalla.com/blog/2008/05/water-fountain-simulation/>

Particle Motion

- mass m , position x , velocity v
- equations of motion:

$$\frac{d}{dt} x(t) = v(t)$$

$$\frac{d}{dt} v(t) = \frac{1}{m} F(x, v, t) \quad F = ma$$

- Analytic solutions can be found for some classes of differential equations, but most can't be solved analytically
- Instead, we will numerically approximate a solution to our *initial value problem*

Higher Order ODEs

- Basic mechanics is a 2nd order ODE:

$$\frac{d^2}{dt^2} x = \frac{1}{m} F$$

- Express as 1st order ODE by defining $v(t)$:

$$\frac{d}{dt} x(t) = v(t)$$

$$\frac{d}{dt} v(t) = \frac{1}{m} F(x, v, t)$$

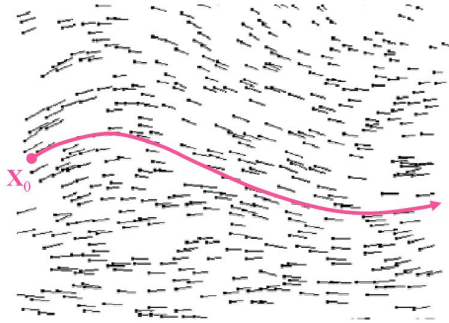
$$\mathbf{X} = \begin{pmatrix} x \\ v \end{pmatrix} \quad f(\mathbf{X}, t) = \begin{pmatrix} v \\ \frac{1}{m} F(x, v, t) \end{pmatrix}$$

\mathbf{X} is a vector storing the
current state of the particle

$f(\mathbf{X}, t)$ describes how to
update the state of the particle

Path Through a Field

- $f(\mathbf{X}, t)$ is a vector field defined everywhere
– E.g. a velocity field which may change over time



Note: In the simplest particle systems, the particles do *not* interact with each other, only with external force fields

- $X(t)$ is a path through the field

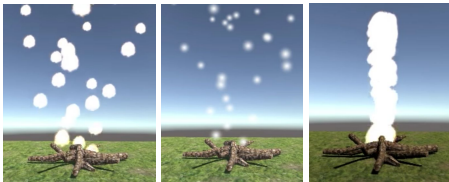
For a Collection of 3D particles...

$$\mathbf{X} = \begin{pmatrix} p_x^{(1)} \\ p_y^{(1)} \\ p_z^{(1)} \\ v_x^{(1)} \\ v_y^{(1)} \\ v_z^{(1)} \\ p_x^{(2)} \\ p_y^{(2)} \\ p_z^{(2)} \\ v_x^{(2)} \\ v_y^{(2)} \\ v_z^{(2)} \\ \vdots \end{pmatrix} \quad f(\mathbf{X}, t) = \begin{pmatrix} v_x^{(1)} \\ v_y^{(1)} \\ v_z^{(1)} \\ \frac{1}{m_1} F_x^{(1)}(\mathbf{X}, t) \\ \frac{1}{m_1} F_y^{(1)}(\mathbf{X}, t) \\ \frac{1}{m_1} F_z^{(1)}(\mathbf{X}, t) \\ v_x^{(2)} \\ v_y^{(2)} \\ v_z^{(2)} \\ \frac{1}{m_2} F_x^{(2)}(\mathbf{X}, t) \\ \frac{1}{m_2} F_y^{(2)}(\mathbf{X}, t) \\ \frac{1}{m_2} F_z^{(2)}(\mathbf{X}, t) \\ \vdots \end{pmatrix}$$

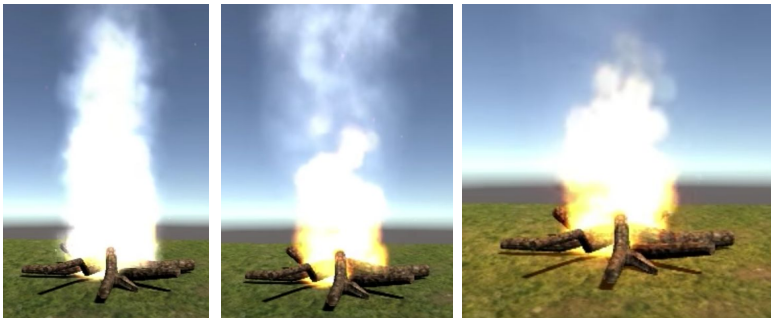
more generally, we can define X as a huge vector storing the current state of *all* particles in a system

Questions?

<https://www.youtube.com/watch?v=M-Hz9Za5mCE>
MixPixVisuals, Mikael Bellander



Note: current state X can also include color & transparency. And $f(X, t)$ can animate changes in these values over time!



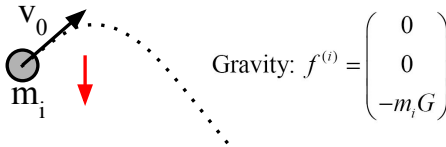
Today

- Particle Systems
 - Equations of Motion (Physics)
 - **Forces: Gravity, Spatial, Damping**
 - Numerical Integration (Euler, Midpoint, etc.)
- Mass Spring System Examples
 - String, Hair, Cloth
- Stiffness
- Discretization

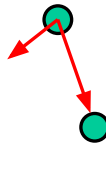
Forces: Gravity

*For smoke, flame:
make "gravity" point up!*

- Simple gravity:
depends only on
particle mass



- N-body problem:
depends on all other particles
 - Magnitude inversely
proportional to square distance
 - $F_{ij} = G m_i m_j / r^2$



*Quickly gets impractical to compute analytically,
and expensive to numerically approximate too!*

Forces: Spatial Fields

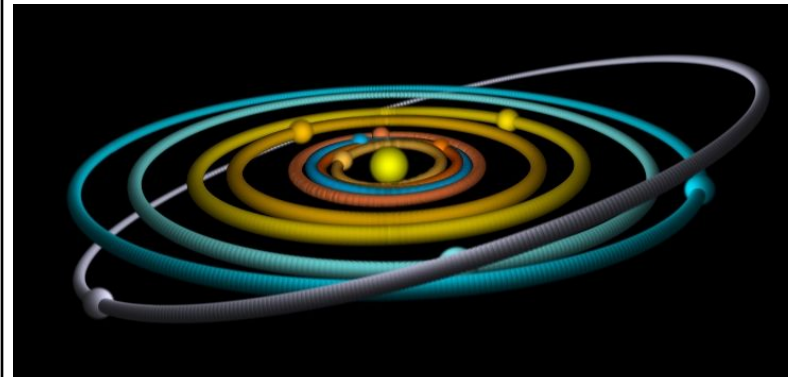
- Force on particle i depends only on position of i
 - wind
 - attractors
 - repulsers
 - vortices
- Can depend on time (e.g., wind gusts)
- Note: these forces will generally add energy to the system, and thus may need damping...

Forces: Damping

$$f^{(i)} = -dv^{(i)}$$

- Force on particle i depends only on velocity of i
- Force opposes motion
 - A hack mimicking real-world friction/drag
- Removes energy, so system can settle
- Small amount of damping can stabilize solver
- Too much damping makes motion too glue-like

Questions?



<http://www.lactamme.polytechnique.fr/Mosaic/images/NCOR.U1.2048.D/display.html>

Today

- Particle Systems
 - Equations of Motion (Physics)
 - Forces: Gravity, Spatial, Damping
 - Numerical Integration (Euler, Midpoint, etc.)
- Mass Spring System Examples
 - String, Hair, Cloth
- Stiffness
- Discretization

Euler's Method

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

$$t_1 = t_0 + h$$

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

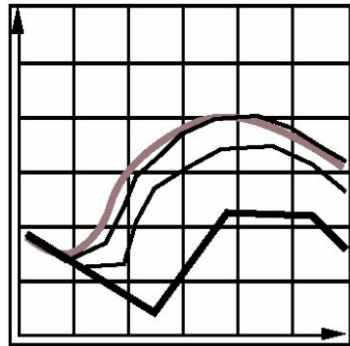
$$\mathbf{X} = \begin{pmatrix} x \\ v \end{pmatrix} \quad f(\mathbf{X}, t) = \begin{pmatrix} v \\ \frac{1}{m} F(x, v, t) \end{pmatrix}$$

update the position
by adding a
little bit of the
current velocity
&
update the velocity
by adding a little
bit of the current
acceleration

- Piecewise-linear approximation to the curve

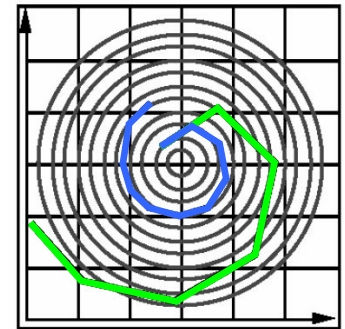
Effect of Step Size

- Step size controls accuracy
- Smaller steps more closely follow curve
- For animation, we may want to take *many* small steps per frame
 - How many frames per second for animation?
 - How many steps per frame?



Euler's Method: Inaccurate

- Simple example: particle in stable circular orbit around planet (origin)
- Current velocity is always tangent to circle
- Force is perpendicular to circle
- Euler method will spiral outward no matter how small h is



Euler's Method: Unstable

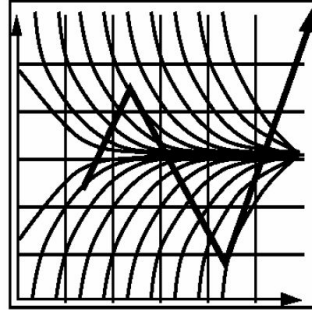
- Problem: $f(x, t) = -kx$
- Solution: $x(t) = x_0 e^{-kt}$

- Limited step size:

$$x_1 = x_0 (1 - hk)$$

$$\begin{cases} h \leq 1/k & \text{ok} \\ h > 1/k & \text{oscillates } \pm \\ h > 2/k & \text{explodes} \end{cases}$$

- If k is big, h must be small



Analysis using Taylor Series

- Expand exact solution $\mathbf{X}(t)$

$$\mathbf{X}(t_0 + h) = \mathbf{X}(t_0) + h \left(\frac{d}{dt} \mathbf{X}(t) \right) \Big|_{t_0} + \frac{h^2}{2!} \left(\frac{d^2}{dt^2} \mathbf{X}(t) \right) \Big|_{t_0} + \frac{h^3}{3!} (\dots) + \dots$$

- Euler's method:

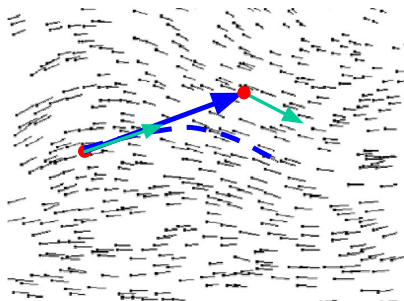
$$\mathbf{X}(t_0 + h) = \mathbf{X}_0 + h f(\mathbf{X}_0, t_0) \quad \dots + O(h^2) \text{ error}$$

$$h \rightarrow h/2 \Rightarrow \text{error} \rightarrow \text{error}/4 \text{ per step} \times \text{twice as many steps} \rightarrow \text{error}/2$$

- First-order method: Accuracy varies with h
 - To get 100x better accuracy need 100x more steps

Can we do better than Euler's Method?

- Problem: f has varied along the step
- Idea: look at f at the arrival of the step and compensate for variation



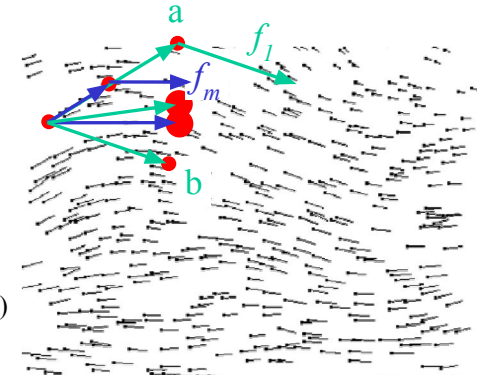
2nd-Order Methods

- Midpoint:

- $\frac{1}{2}$ Euler step
- evaluate f_m
- full step using f_m

- Trapezoid:

- Euler step (a)
- evaluate f_1
- full step using f_1 (b)
- average (a) and (b)

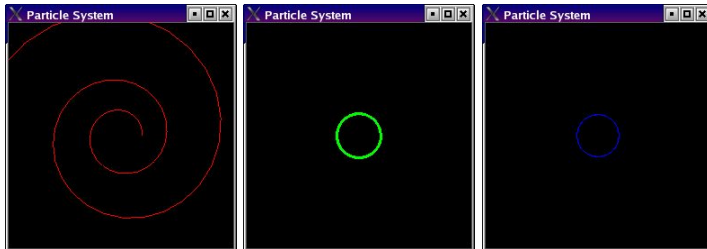


- Midpoint & trapezoid do not yield exactly the same result, but they have same order of accuracy

Comparison: Euler, Midpoint, Runge-Kutta

- *initial position:* (1,0,0)
- *initial velocity:* (0,5,0)
- *force field:* pulls particles to origin with magnitude proportional to distance from origin
- *correct answer:* circle

A 4th order method!

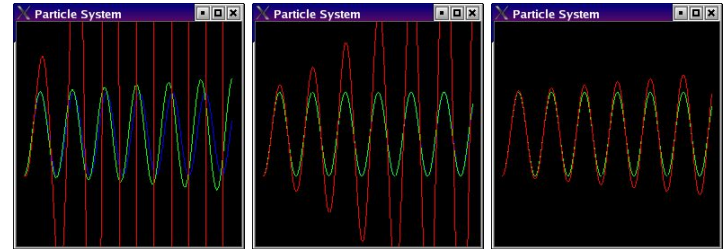


Euler will always diverge (even with small dt)

Comparison: Euler, Midpoint, Runge-Kutta

- *initial position:* (0,-2,0)
- *initial velocity:* (1,0,0)
- *force field:* pulls particles to line $y=0$ with magnitude proportional to distance from line
- *correct answer:* sine wave

A 4th order method!



Decreasing the timestep (dt) improves the accuracy

Questions?

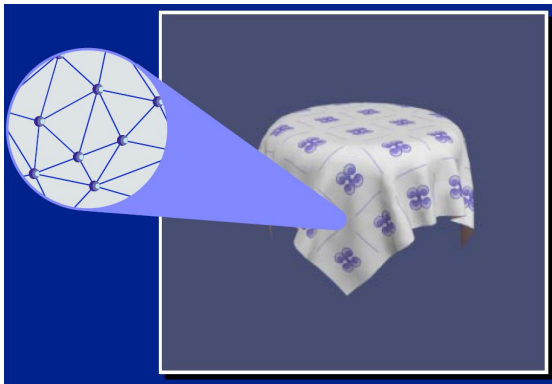


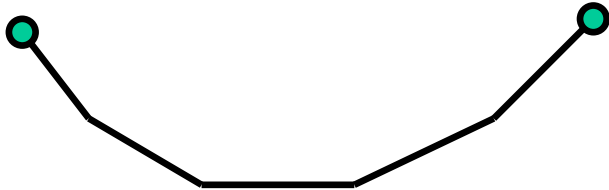
Image by
Baraff, Witkin, Kass

Today

- Particle Systems
 - Equations of Motion (Physics)
 - Numerical Integration (Euler, Midpoint, etc.)
 - Forces: Gravity, Spatial, Damping
- Mass Spring System Examples
 - String, Hair, Cloth
- Stiffness
- Discretization

How would you simulate a string?

- Each particle is linked to two particles
- Forces try to keep the distance between particles constant
- What force?

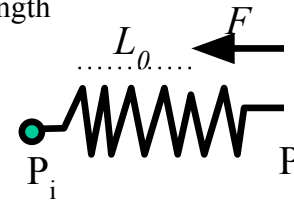


Spring Forces

- Force in the direction of the spring and proportional to difference with rest length L_0

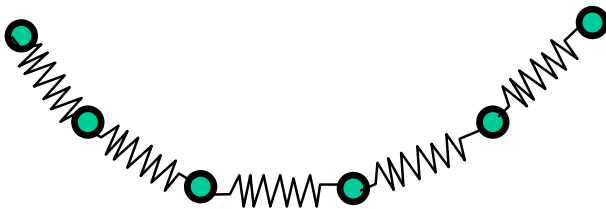
$$F(P_i, P_j) = K(L_0 - \|\vec{P_i P_j}\|) \frac{\vec{P_i P_j}}{\|\vec{P_i P_j}\|}$$

- K is the stiffness of the spring
 - When K gets bigger, the spring really wants to keep its rest length



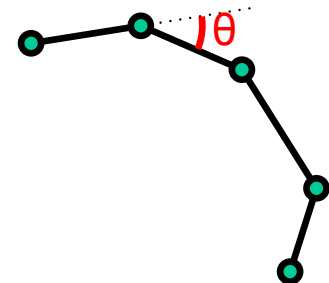
How would you simulate a string?

- Springs link the particles
- Springs try to keep their rest lengths and preserve the length of the string
- Problems?
 - Stretch, actual length will be greater than rest length
 - Numerical oscillation



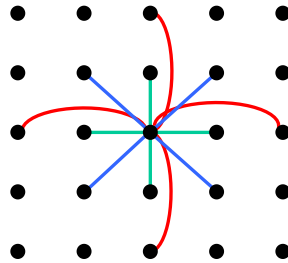
How would you simulate hair?

- Similar to string...
- Also... add deformation forces proportional to the angle between segments (hair wants to stay straight or curly)



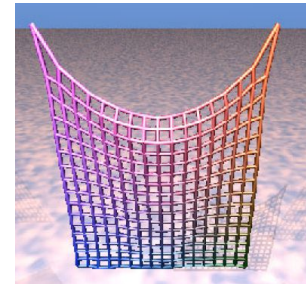
Cloth Modeled with Mass-Spring

- Network of masses and springs
- **Structural springs:**
 - link (i, j) & $(i+1, j)$
and (i, j) & $(i, j+1)$
- **Shear springs**
 - link (i, j) & $(i+1, j+1)$
and $(i+1, j)$ & $(i, j+1)$
- **Flexion (Bend) springs**
 - link (i, j) & $(i+2, j)$
and (i, j) & $(i, j+2)$
- Be careful not to index out of bounds on the cloth edges!

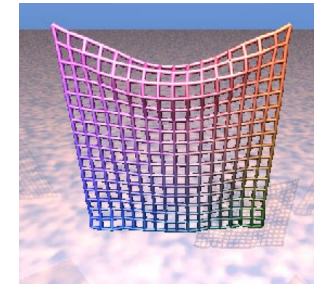


Reading for Tuesday: Everyone should read this (simple cloth model used in HW2)

- “Deformation Constraints in a Mass-Spring Model to Describe Rigid Cloth Behavior”, Provat, 1995.



Simple mass-spring system



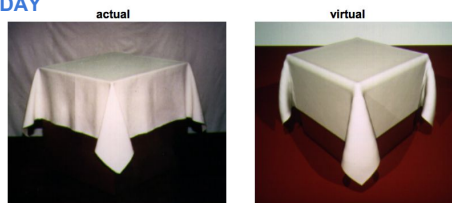
Improved solution

Post a comment/question on the LMS discussion by 10am

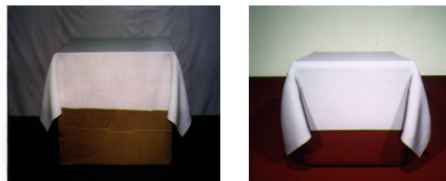
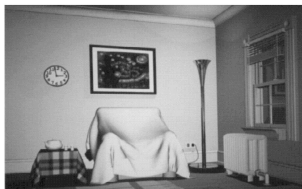
“Predicting the Drape of Woven Cloth Using Interacting Particles”

OPTIONAL READING FOR TUESDAY

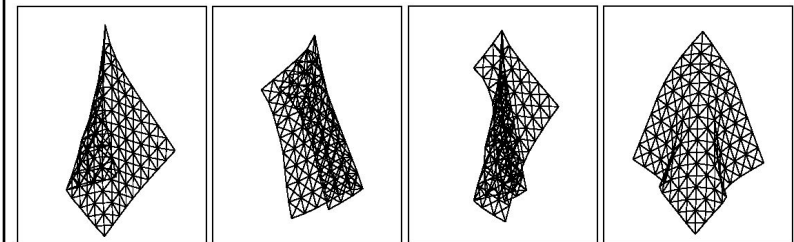
- Breen, House, and Wozny
- SIGGRAPH 1994



100% Cotton Weave



Questions?



*Interactive Animation of
Structured Deformable Objects*
Desbrun, Schröder, & Barr 1999

OPTIONAL READING FOR TUESDAY

Today

- Particle Systems
 - Equations of Motion (Physics)
 - Numerical Integration (Euler, Midpoint, etc.)
 - Forces: Gravity, Spatial, Damping
- Mass Spring System Examples
 - String, Hair, Cloth
- **Stiffness**
- **Discretization**

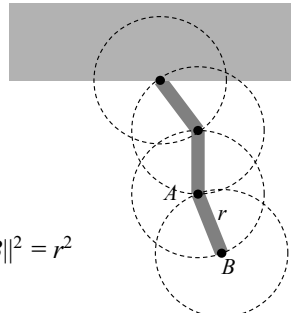
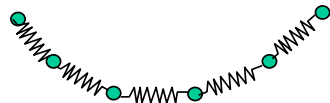
The Stiffness Issue

- What relative stiffness do we want for the different springs in the network?
- Cloth is barely elastic, shouldn't stretch so much!
- Inverse relationship between stiffness & Δt
- We really want constraints (not springs)
- Many numerical solutions
 - reduce Δt
 - use constraints
 - implicit integration
 - ...

How would you simulate a string?

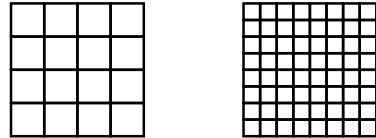
- Springs link the particles. Problems?
 - Stretch, actual length will be greater than rest length
 - Numerical oscillation
- Rigid, fixed-length bars link the particles
 - Dynamics &
 - Constraints

(must be solved simultaneously
non-trivial, even for tiny systems)



<https://www.youtube.com/watch?v=AwT0k09w-jw>

The Discretization Problem

- What happens if we discretize our cloth more finely, or with a different mesh structure?
- 
- Do we get the same behavior?
 - Usually not! It takes a lot of effort to design a scheme that does not depend on the discretization.
 - Using (explicit) Euler, how many timesteps before a force propagates across the mesh?

Explicit vs. Implicit Integration

- With an explicit/forward integration scheme:

$$\mathbf{y}_{k+1} = \mathbf{y}_k + h \mathbf{g}(\mathbf{y}_k)$$

The future state (position & velocity) of this particle is a function of the current state of the particle.

we must use a very small timestep to simulate stable, stiff cloth.

- Alternatively we can use an implicit/backwards scheme:

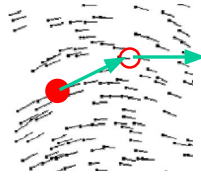
$$\mathbf{y}_{k+1} = \mathbf{y}_k + h \mathbf{g}(\mathbf{y}_{k+1})$$

$$\mathbf{y}_k = \mathbf{y}_{k+1} - h \mathbf{g}(\mathbf{y}_{k+1})$$

The future state of this particle depends on the current state AND the future state.

Solving one step is much more expensive (Newton's Method, Conjugate Gradients, ...) but overall faster than the thousands of explicit timesteps required for very stiff springs.

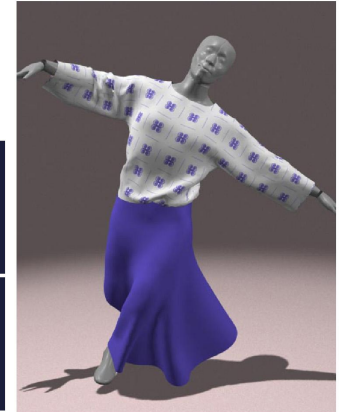
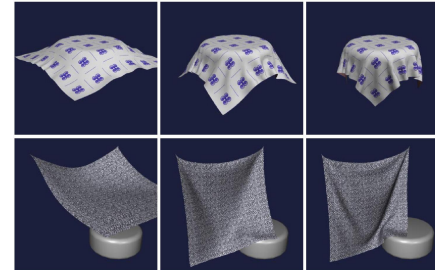
- Larger timesteps are possible with implicit methods.



Questions?

David Baraff & Andrew Witkin
Large Steps in Cloth Simulation
SIGGRAPH 1998

OPTIONAL READING FOR TUESDAY

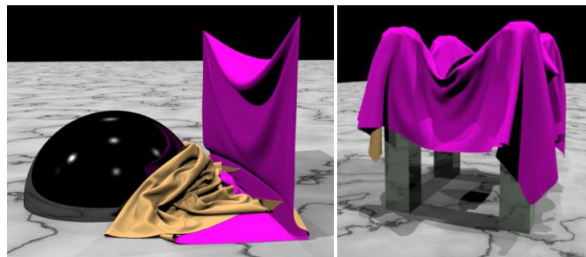
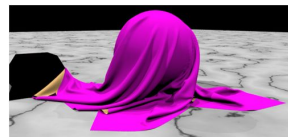


Cloth Collision

OPTIONAL READING FOR TUESDAY

Robert Bridson, Ronald Fedkiw & John Anderson
Robust Treatment of Collisions, Contact and Friction for Cloth Animation
SIGGRAPH 2002

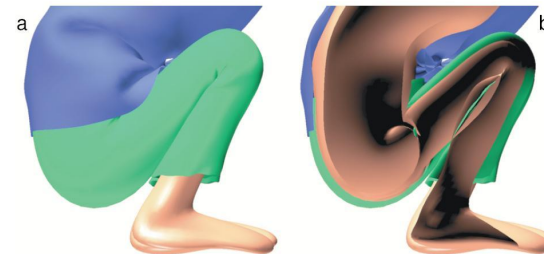
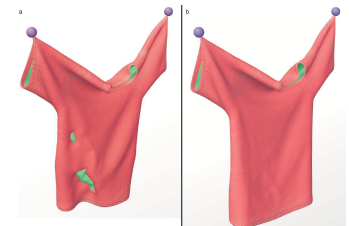
- A cloth has many points of contact
- Often stays in contact
- Requires
 - Efficient collision detection
 - Efficient numerical treatment (stability)



Cloth in Practice (w/ Animation)

OPTIONAL READING FOR TUESDAY

- Baraff, Witkin & Kass
Untangling Cloth
SIGGRAPH 2003

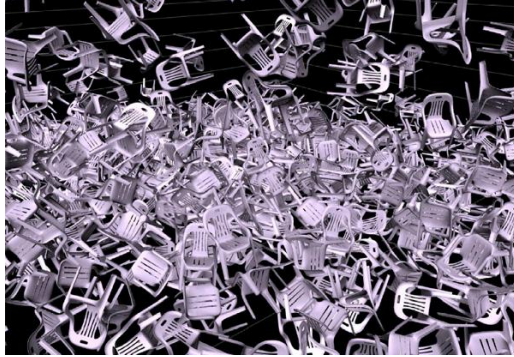


Reduced Deformation

OPTIONAL READING FOR TUESDAY

Doug L. James & Dinesh K. Pai
*BD-Tree: Output-Sensitive Collision
Detection for Reduced Deformable Models*
SIGGRAPH 2004

- Collisions are expensive
- Deformation is expensive
- This is a lot of geometry!
- Simplify the simulation model



HW2: Cloth & Fluid Simulation

