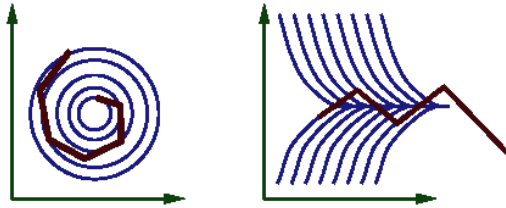


Solving the Equations of Motion



- Forward (explicit) Euler integration
 - $x(t+\Delta t) \leftarrow x(t) + \Delta t v(t)$
 - $v(t+\Delta t) \leftarrow v(t) + \Delta t f(x(t), v(t), t) / m$
- Problem:
 - Accuracy decreases as Δt gets bigger



Hodgins

Solving the Equations of Motion



Explicit Euler step

$$y_{n+1} = y_n + hf(t_n, y_n)$$

Implicit Euler step

$$y_{n+1} = y_n + hf(t_{n+1}, y_{n+1})$$

Why are these methods called like this?

- **Explicit:** all quantities are known (*given explicitly*)
- **Implicit:** y_{n+1} is unknown (*given implicitly*)

→ solve (nonlinear) equation(s)!

Solving the Equations of Motion



Explicit Euler step

$$y_{n+1} = y_n + hf(t_n, y_n)$$

Implicit Euler step

$$y_{n+1} = y_n + hf(t_{n+1}, y_{n+1})$$

Why are these methods called like this?

- **Explicit:** all quantities are known (*given explicitly*)
- **Implicit:** y_{n+1} is unknown (*given implicitly*)
- Stability conditions for Euler
 - Explicit $y_n = (1+h\lambda)^n y_0 < \infty \Leftrightarrow |1+h\lambda| < 1$
 - Implicit $y_n = (1-h\lambda)^{-n} y_0 < \infty \Leftrightarrow |1-h\lambda|^{-1} < 1$

⇒ **Implicit Euler is stable for all $h > 0$!**

More Complex Particle Systems

Particle System Forces



- Gravity
 - Force due to gravitational pull (of earth)
 - g = acceleration due to gravity (m/s^2)

$$f_g = mg$$

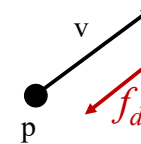
\downarrow $g = (0, -9.80665, 0)$

Particle System Forces



- Drag
 - Force due to resistance of medium
 - k_{drag} = drag coefficient (kg/s)

$$f_d = -k_{drag} v$$



- Air resistance sometimes taken as proportional to v^2

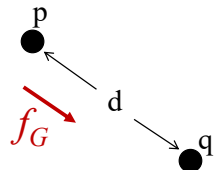
Particle System Forces



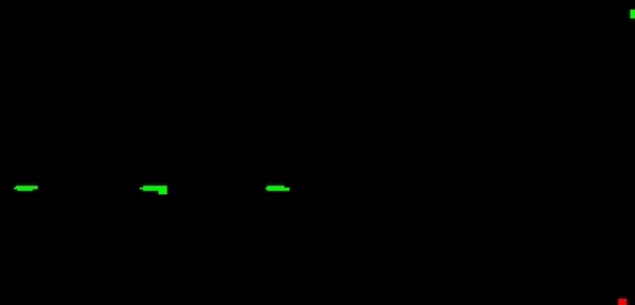
- Gravitational pull of other particles
 - Newton's universal law of gravitation

$$f_G = G \frac{m_1 \cdot m_2}{d^2}$$

$$G = 6.67428 \times 10^{-11} \text{ N m}^2 \text{ kg}^{-2}$$



Examples: Fountain, Gravity



Spring-Mass Models

Particle System Forces



- Springs
 - Hooke's law

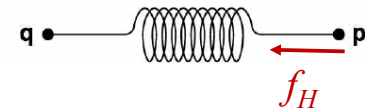
$$f_H(p) = k_s(d(p, q) - s) D$$

$$D = (q - p) / \|q - p\|$$

$$d(p, q) = \|q - p\|$$

s = resting length

k_s = spring coefficient



Particle System Forces



- Springs
 - Hooke's law with damping

$$f_H(p) = [k_s(d(p, q) - s) + k_d(v(q) - v(p)) \cdot D] D$$

$$D = (q - p) / \|q - p\|$$

$$d(p, q) = \|q - p\|$$

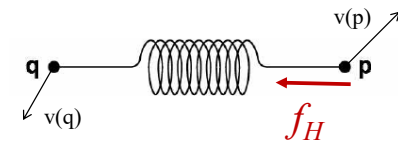
s = resting length

k_s = spring coefficient

k_d = damping coefficient

$v(p)$ = velocity of p

$v(q)$ = velocity of q



$$k_d \sim 2\sqrt{mk_s}$$

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

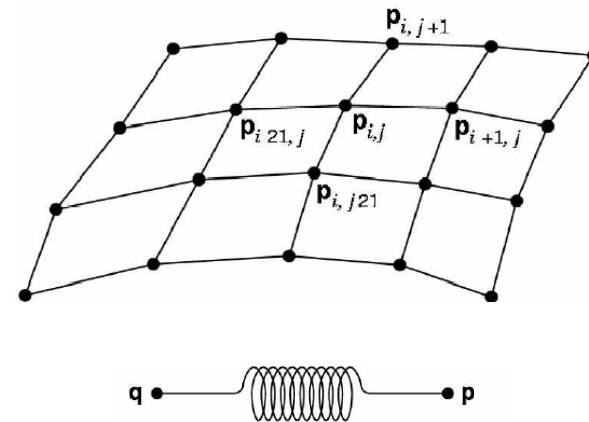
Examples: Rope

<http://www.cs.princeton.edu/courses/archive/spring14/cos426/assignment4/examples/examples.html>

Particle System Forces



- Spring-mass mesh



Hodgins

Solving Spring-Mass Systems

- Can be set up as an Euler integration
- \mathbf{f}_0 accumulates the spring forces, \mathbf{M}^{-1} is a matrix of the masses at each point
- $\Delta \mathbf{x}$, $\Delta \mathbf{v}$ describe the discretized position and velocity update

$$\begin{pmatrix} \Delta \mathbf{x} \\ \Delta \mathbf{v} \end{pmatrix} = h \begin{pmatrix} \mathbf{v}_0 \\ \mathbf{M}^{-1} \mathbf{f}_0 \end{pmatrix}$$

- Compare with implicit form:

$$\begin{pmatrix} \Delta \mathbf{x} \\ \Delta \mathbf{v} \end{pmatrix} = h \begin{pmatrix} \mathbf{v}_0 + \Delta \mathbf{v} \\ \mathbf{M}^{-1} \mathbf{f}(\mathbf{x}_0 + \Delta \mathbf{x}, \mathbf{v}_0 + \Delta \mathbf{v}) \end{pmatrix}$$

Baraff, Witkin. SIGGRAPH 1998. <https://www.cs.cmu.edu/~baraff/papers/sig98.pdf>

Examples: Cloth, Flag



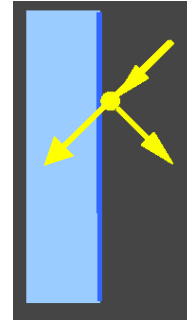
<http://www.cs.princeton.edu/courses/archive/spring14/cos426/assignment4/examples/examples.html>

Collision

Particle System Forces



- Collisions
 - Collision detection
 - Collision response

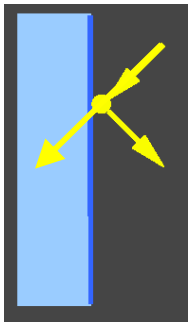


Witkin

Particle System Forces



- Collision detection
 - Intersect ray with scene
 - Compute up to Δt at time of first collision, and then continue from there

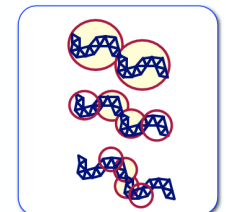
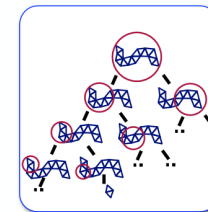
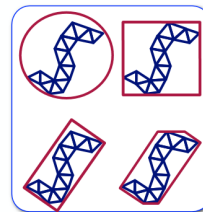


Witkin

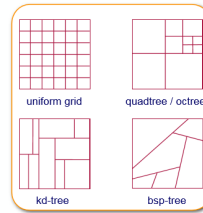
Collision Detection



Bounding Volumes

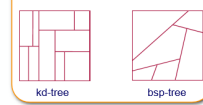


Spatial Partitioning



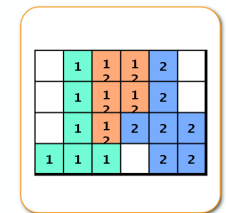
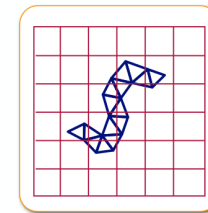
uniform grid

quadtree / octree



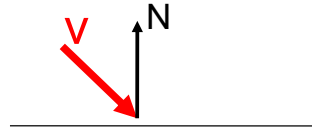
kd-tree

bsp-tree



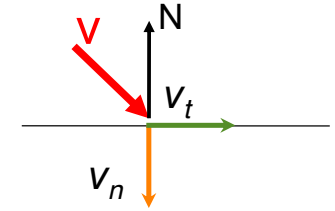
| | | | | | |
|---|---|---|---|---|---|
| | 1 | 1 | 1 | 2 | |
| | 1 | 1 | 1 | 2 | |
| | 1 | 1 | 2 | 2 | 2 |
| 1 | 1 | 1 | | 2 | 2 |

Collision Response for Particles



66

Collision Response for Particles



$$v = v_n + v_t$$

normal component
tangential component

67

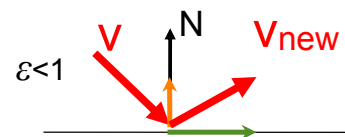
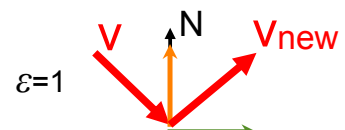
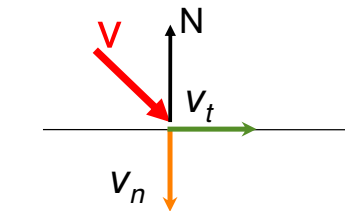
Collision Response for Particles

- Tangential velocity v_t often unchanged
- Normal velocity v_n reflects:

$$v = v_t + v_n$$

$$v \leftarrow v_t - \varepsilon v_n$$

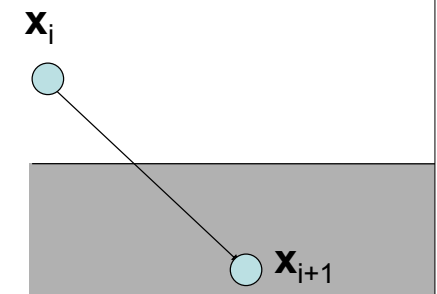
- Coefficient of restitution ε
- When $\varepsilon = 1$, mirror reflection



68

Collisions – Overshooting

- Usually, we detect collision when it is too late: we are already inside

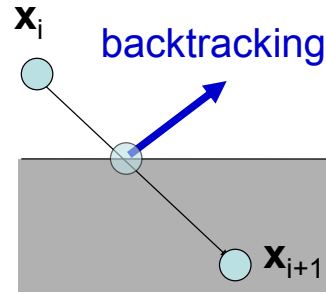


69

Collisions – Overshooting

- Usually, we detect collision when it is too late: we are already inside
- Solution: Back up

- Compute intersection point
- Ray-object intersection!
- Compute response there
- Advance for remaining fractional time step

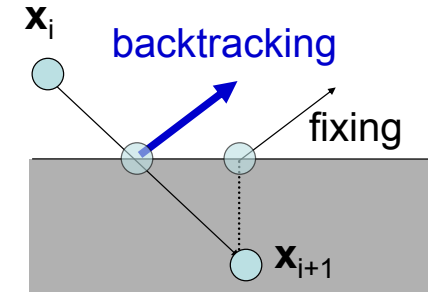


70

Collisions – Overshooting

- Usually, we detect collision when it is too late: we are already inside
- Solution: Back up

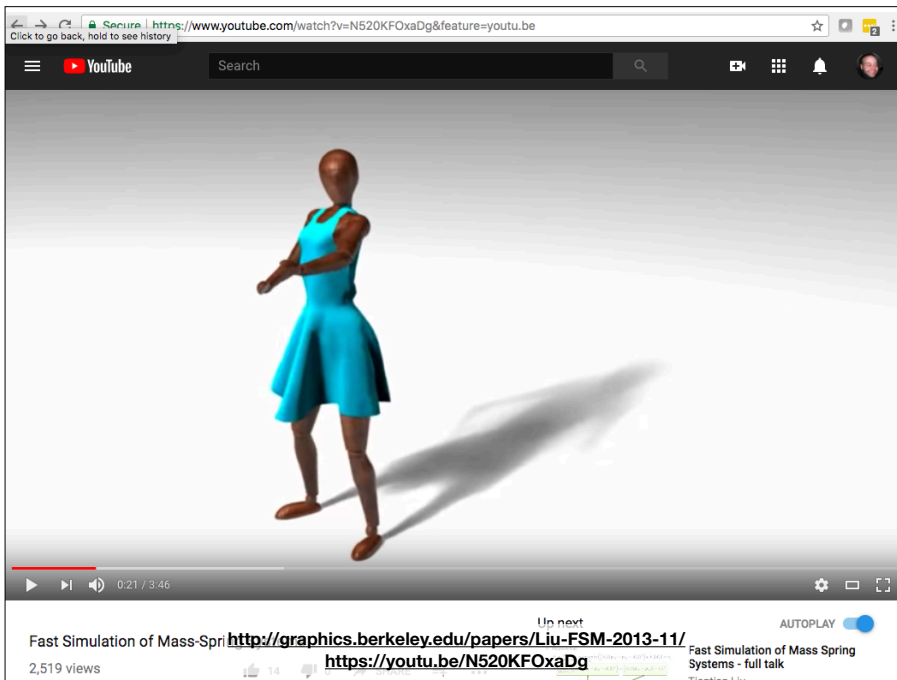
- Compute intersection point
- Ray-object intersection!
- Compute response there
- Advance for remaining fractional time step



- Other solution: Quick and dirty hack

- Just project back to object closest point

71



Animation So Far

- Interpolation based animation:
 - Interpolation of images, for example based on keyframes
 - Interpolation of motions, for example scripted character animation by designing transformation on per entity basis
- Physically-based animation techniques where we solve an ODE of some sort
 - We've started talking about this with particle-based physics
 - Today we'll talk about another scheme for it
- Next: we'll talk about non-physical animations through behaviors.

Animating Based on Behavioral Rules

Boids



- Powerful, simple model
 - No central control
 - Only simple rules for each individual
 - Complex, emergent phenomena
 - Self-organization, swarm intelligence



Reynolds

Boids



- Computer graphics motivation
 - Scripting of the path of many individual objects using traditional computer animation techniques is tedious.

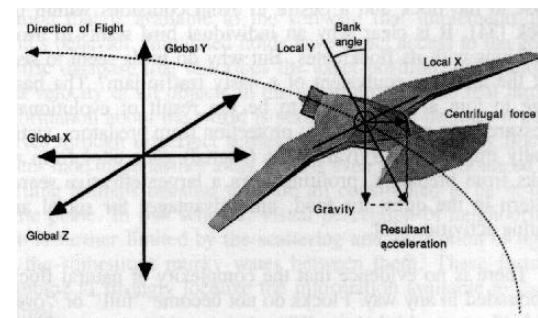


Reynolds

Boids



- Like a particle system, except ...
 - Each boid may be an entire polygonal object with a local coordinate system (rather than a point)

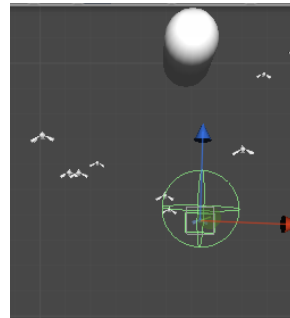
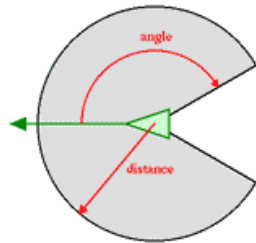


Reynolds

Boids



- Like a particle system, except ...
 - Each boid can “perceive” a local region around it, e.g., a spherical neighborhood

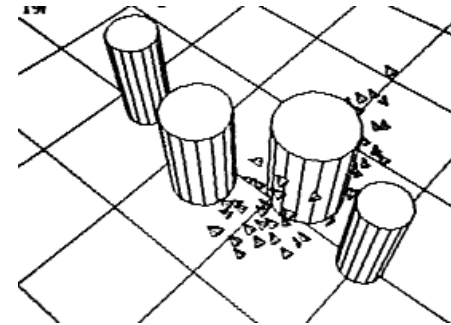


<http://www.arges-systems.com>

Boids



- Like a particle system, except ...
 - Each boid exerts “intentional forces”



Reynolds

Flocking

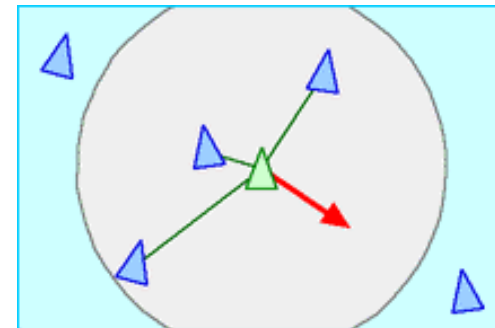


- Complex flocking behaviors can be modeled with simple “intentional forces”
 - Separation
 - Alignment
 - Cohesion

Flocking – 3 Behaviors (1)



- Separation = collision avoidance:
avoid collisions with nearby flockmates

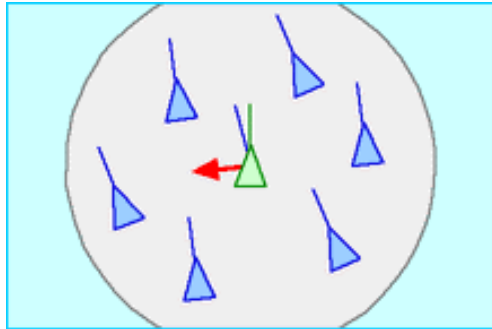


<http://www.red3d.com>

Flocking – 3 Behaviors (2)



- Alignment = velocity matching:
attempt to match velocity with nearby flockmates

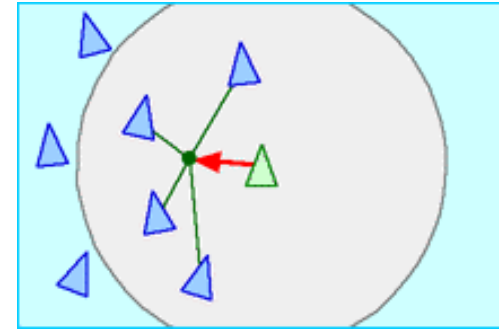


<http://www.red3d.com>

Flocking – 3 Behaviors (3)



- Cohesion = flock centering:
attempt to stay close to nearby flockmates



<http://www.red3d.com>

Other Examples (single behavior)



- Example behaviors
 - Seek
 - Flee
 - Evasion
 - Pursuit
 - Wander
 - Arrival
 - Obstacle Avoidance
 - Containment
 - Wall Following
 - Path Following

<http://www.red3d.com/cwr/steer/>

Examples

COURSE: 07
COURSE ORGANIZER: DEMETRI TERZOPOULOS

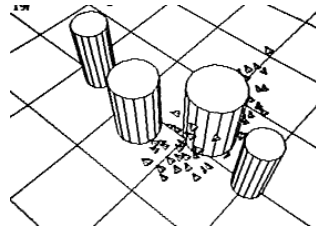
"BOIDS DEMOS"
CRAIG REYNOLDS
SILICON STUDIOS, MS 3L-980
2011 NORTH SHORELINE BLVD.
MOUNTAIN VIEW, CA 94039-7311

<http://www.red3d.com/cwr/boids/>

Obstacle Avoidance (1)



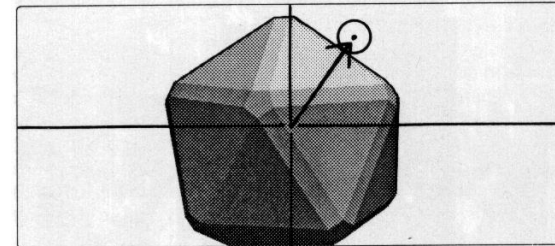
- Force field approach
 - Obstacles have a field of repulsion
 - Boids increasingly repulsed as they approach obstacle
- Drawbacks:
 - Approaching a force in exactly the opposite direction
 - Flying alongside a wall



Obstacle Avoidance (2)



- Steer-to-avoid approach
 - Boid only considers obstacles directly in front of it
 - Finds silhouette edge of obstacle closest to point of eventual impact
 - A vector is computed that will aim the boid at a point one body length beyond the silhouette edge



<http://www.red3d.com>

Arbitrating Independent Behaviors



- Navigation module of boid brain to collect relevant acceleration requests and then determine single behaviorally desired acceleration
 - Weighted average according to priority
- Emergency acceleration allocated to satisfy pressing needs first
 - Example: Centering ignored in order to maneuver around obstacles

ncase.me/improv-wip/

Improv.js [WIP]

a tool to make tools to make [explorable explanations](#)

Let's say you have a model of a complex system. And you want others to be able to play around with it, explore the system, change its rules, maybe even create their own models with it! Well, that's an incredibly specific goal to have, but hey, you're in luck.

Improv lets you write words normally like this...

```
A "boid" is like a bird, but worse.  
Let's make a flock of {NUMBER count: min=0,max=100} boids,  
and paint 'em all {CHOOSE color: black, red, blue, random colors}!
```

...which will get you something like this:

```
A "boid" is like a bird, but worse.  
Let's make a flock of 50 boids,  
and paint 'em all 
```

<http://ncase.me/improv-wip/>



