## Computer Animation

Animation Methods
Keyframing
Interpolation
Kinematics
Inverse Kinematics

Slides courtesy of Leonard McMillan and Jovan Popovic


## Administrative

## Office hours

- Durand \& Teller by appointment
- Ngan Thursday 4-7 in W20-575

Deadline for proposal: Friday Nov 1
Meeting with faculty \& staff about proposal

- Next week
- Web page for appointment


## Animation

## 4 approaches to animation

 Pros ? Cons ?


## Computer-Assisted Animation

Keyframing

- automate the inbetweening
- good control
- less tedious
- creating a good animation still requires considerable skill and talent
Procedural animation


ACM © 1987 "Principles of traditional animation applied to 3D computer animation"

- describes the motion algorithmically
- express animation as a function
of small number of parameteres
- Example: a clock with second, minute and hour hands
- hands should rotate together
- express the clock motions in terms of a "seconds" variable
- the clock is animated by varying the seconds parameter
- Example 2: A bouncing ball
- $\operatorname{Abs}\left(\sin \left(\omega t+\theta_{0}\right)\right)^{*} e^{-k t}$



## Computer-Assisted Animation

## Physically Based Animation

- Assign physical properties to objects (masses, forces, inertial properties)
- Simulate physics by solving equations
- Realistic but difficult to control



## Motion Capture

- Captures style, subtle nuances and realism
- You must observe someone do something




## Overview

# Keyframing and interpolation Interpolation of rotations, quaternions Kimematrics, articulation 

Particles

Rigid bodies



Deformable objects, clothes, fluids


## Kinematics vs. Dynamics

## Kinematics

Describes the positions of the body parts as a function of the joint angles.

## Dynamics

Describes the positions of the body parts as a function of the applied forces.

## Now

## Dynamics



## Particle

A single particle in 2-D moving in a flow field

- Position $\mathbf{x}=\left[\begin{array}{l}x_{1} \\ x_{2}\end{array}\right]$
- Velocity $\mathbf{v}=\left[\begin{array}{l}v_{1} \\ v_{2}\end{array}\right], \mathbf{v}=\frac{d \mathbf{x}}{d t}$
- The flow field function dictates particle velocity $\mathbf{V}=\mathbf{g}(\mathbf{x}, t)$



## Vector Field

The flow field $\mathbf{g}(\mathbf{x}, t)$ is a vector field that defines a vector for any particle position $\mathbf{x}$ at any time $t$.


How would a particle move in this vector field?

## Differential Equations

The equation $\mathbf{v}=\mathbf{g}(\mathbf{x}, t)$ is a first order differential equation:

$$
\frac{d \mathbf{x}}{d t}=\mathbf{g}(\mathbf{x}, t)
$$

The position of the particle is computed by integrating the differential equation:

$$
\mathbf{x}(t)=\mathbf{x}\left(t_{0}\right)+\int_{t_{0}}^{t} \mathbf{g}(\mathbf{x}, t) d t
$$

For most interesting cases, this integral cannot be computed analytically.

## Numeric Integration

Instead we compute the particle's position by numeric integration: starting at some initial point $\mathbf{x}\left(t_{0}\right)$ we step along the vector field to compute the position at each subsequent time instant. This type of a problem is called an initial value problem.


## Euler's Method

Euler's method is the simplest solution to an initial value problem. Euler's method starts from the initial value and takes small time steps along the flow:

$$
\mathbf{x}(t+\Delta t)=\mathbf{x}(t)+\Delta t \mathbf{g}(\mathbf{x}, t)
$$

Why does this work?
Let's look at a Taylor series expansion of function $\mathbf{x}(t)$ :

$$
\mathbf{x}(t+\Delta t)=\mathbf{x}(t)+\Delta t \frac{d \mathbf{x}}{d t}+\frac{\Delta t^{2}}{2} \frac{d^{2} \mathbf{x}}{d t^{2}}+\cdots
$$

Disregarding higher-order terms and replacing the first derivative with the flow field function yields the equation for the Euler's method.

## Other Methods

Euler's method is the simplest numerical method. The error is proportional to $\Delta t^{2}$. For most cases, the Euler's method is inaccurate and unstable requiring very small steps.

Other methods:

- Midpoint (2 ${ }^{\text {nd }}$ order Runge-Kutta)

- Higher order Runge-Kutta (4 $4^{\text {th }}$ order, $6^{\text {th }}$ order)
- Adams
- Adaptive Stepsize


## Particle in a Force Field

What is a motion of a particle in a force field?
The particle moves according to Newton's Law:

$$
\frac{d^{2} \mathbf{x}}{d t^{2}}=\frac{\mathbf{f}}{m} \quad(\mathbf{f}=m a)
$$

The mass $m$ of a particle describes the particle's inertial properties: heavier particles are easier to move than lighter particles. In general, the force field $\mathbf{f}(\mathbf{x}, \mathbf{v}, t)$ may depend on the time $t$ and particle's position $\mathbf{x}$ and velocity $\mathbf{v}$.

## Second-Order Differential Equations

Newton's Law yields an ordinary differential equation of second order:

$$
\frac{d^{2} \mathbf{x}(t)}{d t^{2}}=\frac{\mathbf{f}(\mathbf{x}, \mathbf{v}, t)}{m}
$$

A clever trick allows us to reuse the same numeric differentiation solvers for first-order differential equations. If we define a new phase space vector $\mathbf{y}$, which consists of particle's position $\mathbf{x}$ and velocity $\mathbf{v}$, then we can construct a new first-order differential equation whose solution will also solve the second-order differential equation.

$$
\mathbf{y}=\left[\begin{array}{l}
\mathbf{x} \\
\mathbf{v}
\end{array}\right], \quad \frac{d \mathbf{y}}{d t}=\left[\begin{array}{l}
d \mathbf{x} / d t \\
d \mathbf{v} / d t
\end{array}\right]=\left[\begin{array}{c}
\mathbf{v} \\
\mathbf{f} / m
\end{array}\right]
$$

## Particle Animation

## AnimateParticles $\left(n, \mathbf{y}_{0}, t_{0}, t_{f}\right)$

\{

$$
\begin{aligned}
& \mathbf{y}=\mathbf{y}_{0} \\
& t=t_{0}
\end{aligned}
$$

DrawParticles $(n, \mathbf{y})$
while $\left(t!=t_{f}\right)$ \{
$\mathbf{f}=$ ComputeForces $(\mathbf{y}, t)$
dydt = AssembleDerivative $(\mathbf{y}, \mathbf{f})$
$\{\mathbf{y}, t\}=$ ODESolverStep( $6 n, \mathbf{y}, \mathrm{~d} \mathbf{y} / \mathrm{dt}$ )
DrawParticles( $n, \mathbf{y}$ )
\}
\}

## Particle Animation [Reeves et al. 1983]



## Particle Modeling [Reeves et al. 1983]



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## Rigid-Body Dynamics

We could compute the motion of a rigid-body by computing the motion of all constituent particles. However, a rigid body does not deform and position of few of its particles is sufficient to determine the state of the body in a phase space. We'll start with a special particle located at the body's center of mass.


Net Force


## Net Torque



## Rigid-Body Equation of Motion

$$
\frac{d}{d t} \mathbf{y}(t)=\frac{d}{d t}\left[\begin{array}{c}
\mathbf{x}(t) \\
\mathbf{R}(t) \\
M \mathbf{v}(t) \\
\mathbf{I}(t) \boldsymbol{\omega}(t)
\end{array}\right]=\left[\begin{array}{c}
\mathbf{v}(t) \\
\omega(t) \times \mathbf{R}(t) \\
\mathbf{f}(t) \\
\mathbf{T}(t)
\end{array}\right]
$$

$M \mathbf{v}(t) \rightarrow$ linear momentum
$\mathbf{I}(t) \boldsymbol{\omega}(t) \rightarrow$ angular momentum

## Simulations with Collisions

Simulating motions with collisions requires that we detect them (collision detection) and fix them (collision response).


Lecture 14

## Collision Response

The mechanics of collisions are complicated and many mathematical models have been developed. We'll just look at a one simple model, which assumes that when the collision occurs the two bodes exchange collision impulse instantaneously.


## Frictionless Collision Model



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## Deformable models

Shape deforms due to contact
Discretize the problem
Animation runs with smaller time steps than rendering (between $1 / 10,000$ s and $1 / 100$ s)


Images from Debunne et al. 2001


Imas from Debunne et al. 2001

## Mass-Spring system

Network of masses and springs
Express forces
Integrate
Deformation of springs simulates deformation of objects



Images from Debunne et al. 2001

## Implicit Finite Elements

Discretize the problem
Express the interrelationship
Solve a big system
More principled than mass-spring


Object
Slide from Debunne et al. 2001


Finite
Elements

## Large matricial system

## Explicit Finite Elements

Discretize the problem
Solve locally
Simpler but less stable than implicit


## Formally: Finite Elements

We are trying to solve a continuous problem

- Deformation of all points of the object
- Infinite space of functions

We project to a finite set of basis functions

- E.g. piecewise linear, piecewise constant

We project the equations governing the problem
This results in a big linear system


Object


Finite Elements Slide 33


Large matricial
system
6.837 Fall 2002

## Cloth animation

## Discretize cloth

Write physical equations
Integrate
Collision detection


Image from Meyer et al. 2001

## Fluid simulation

## Discretize volume of fluid

- Exchanges and velocity at voxel boundary Write Navier Stokes equations
- Incompressible, etc.

Numerical integration

- Finite elements, finite differences

Challenges:

- Robust integration, stability
- Speed
- Realistic surface

Figure from Fedkiw et al. 2001


Figure 1: Water being poured into a glass ( $55 \times 120 \times 55$ grid cells). Figure from Enright et al. 2002

## Other physical animation

Aging of materials

- Metallic patina, rust
- Water flow
. Stone aging
[Dorsey 1996-1999]



## How do they animate movies?

## Keyframing mostly

Articulated figures, inverse kinematics
Skinning

- Complex deformable skin
- Muscle, skin motion

Hierarchical controls

- Smile control, eye blinking, etc.

- Keyframes for these higher-level controls


Images from the Maya tutorial

A huge time is spent building the 3D models, its skeleton and its controls

Physical simulation for secondary motion

- Hair, cloths, water
- Particle systems for "fuzzy" objects


## Next time: Texture mapping



