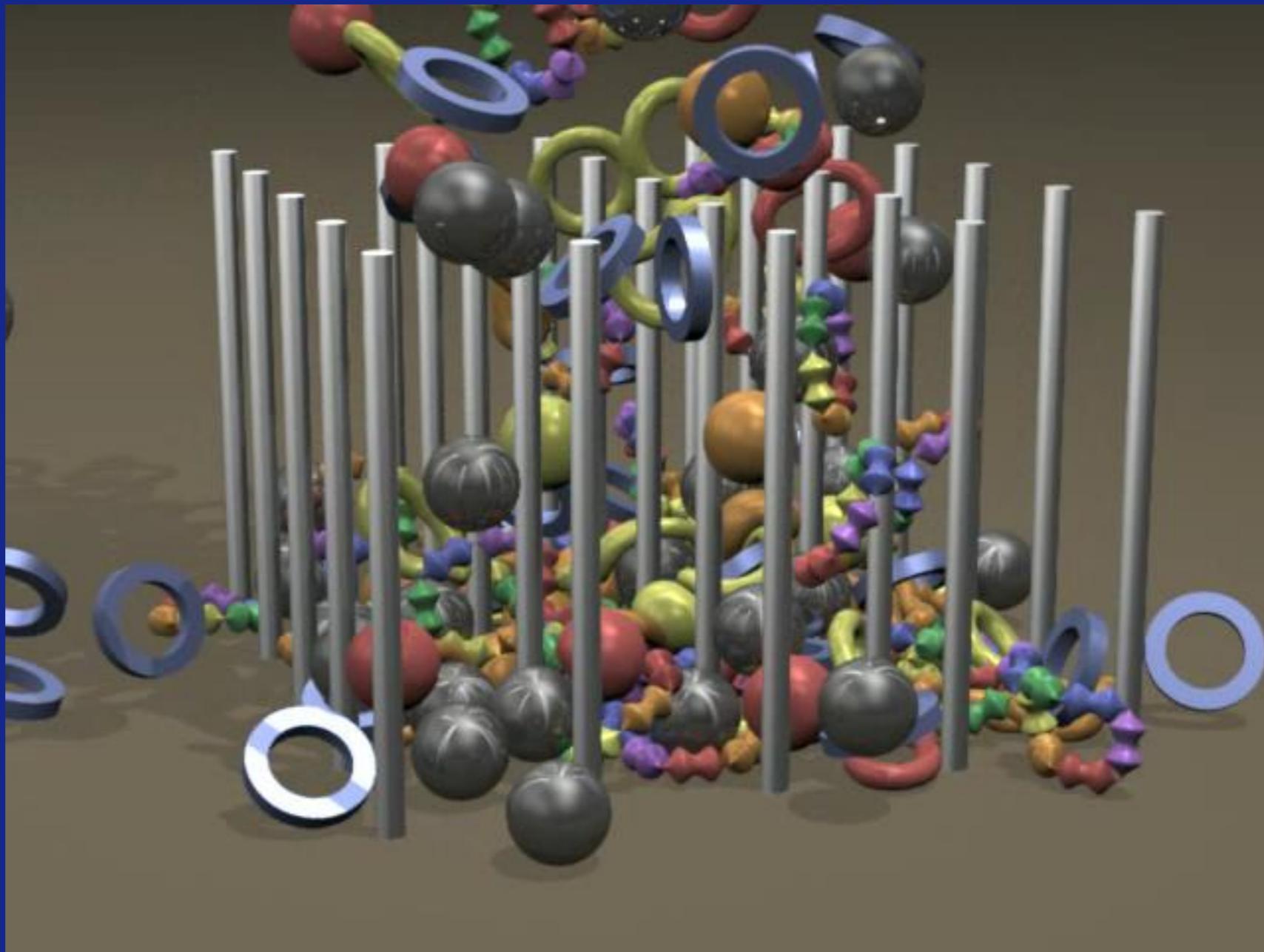


Differential Equations & Particle Systems

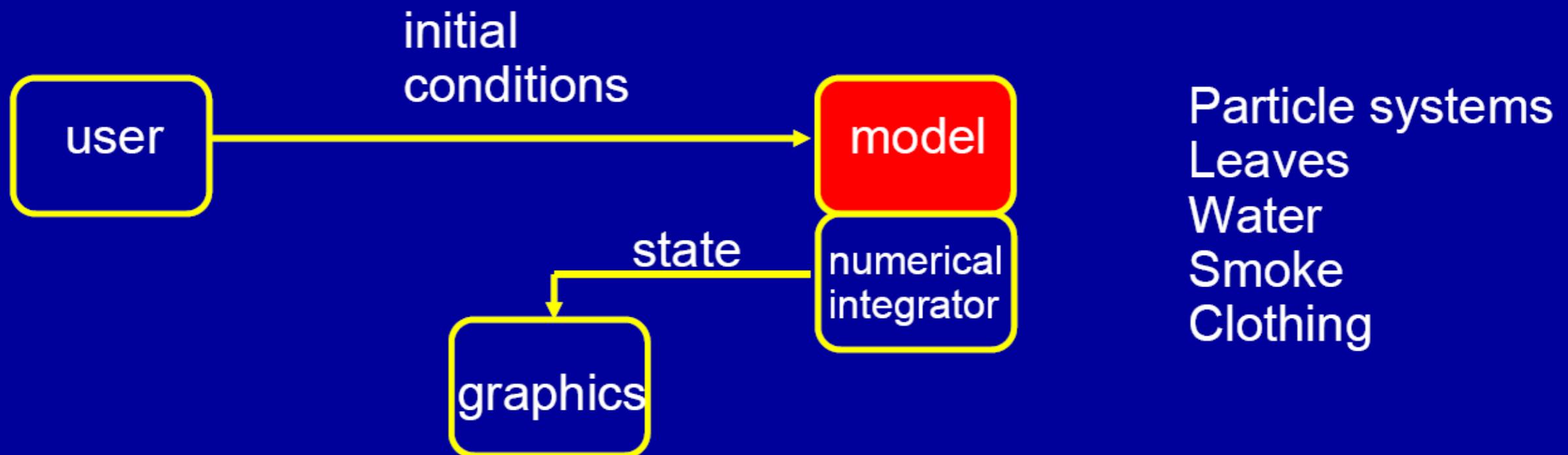
Thanks to Trueille, Popovic, Baraff, Witkin

Physics-based Animation

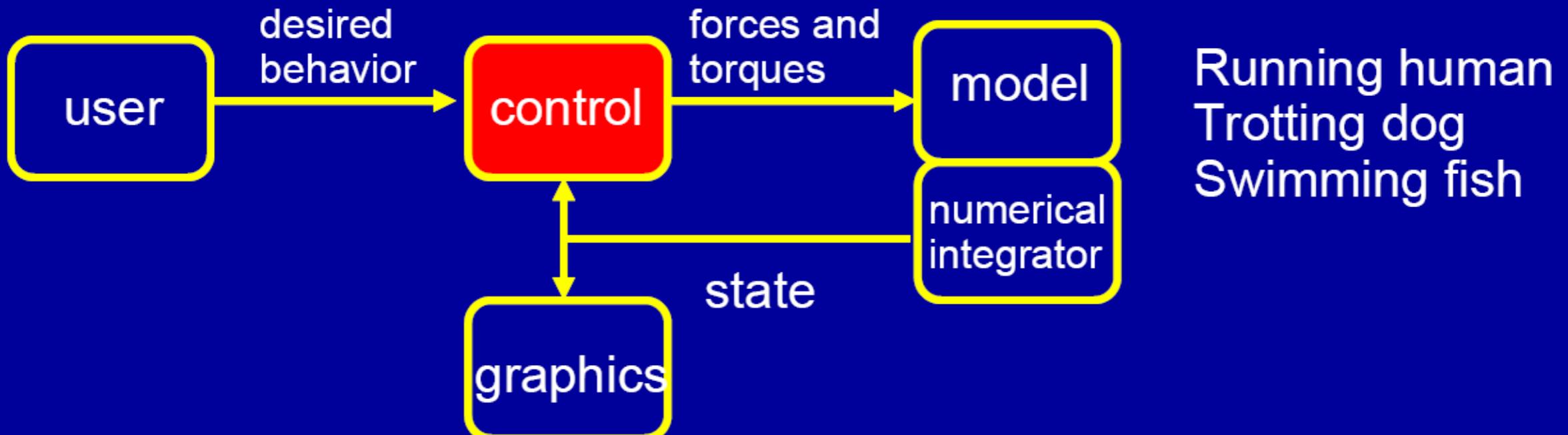


http://physbam.stanford.edu/~fedkiw/animations/large_pile.avi

Passive—no muscles or motors



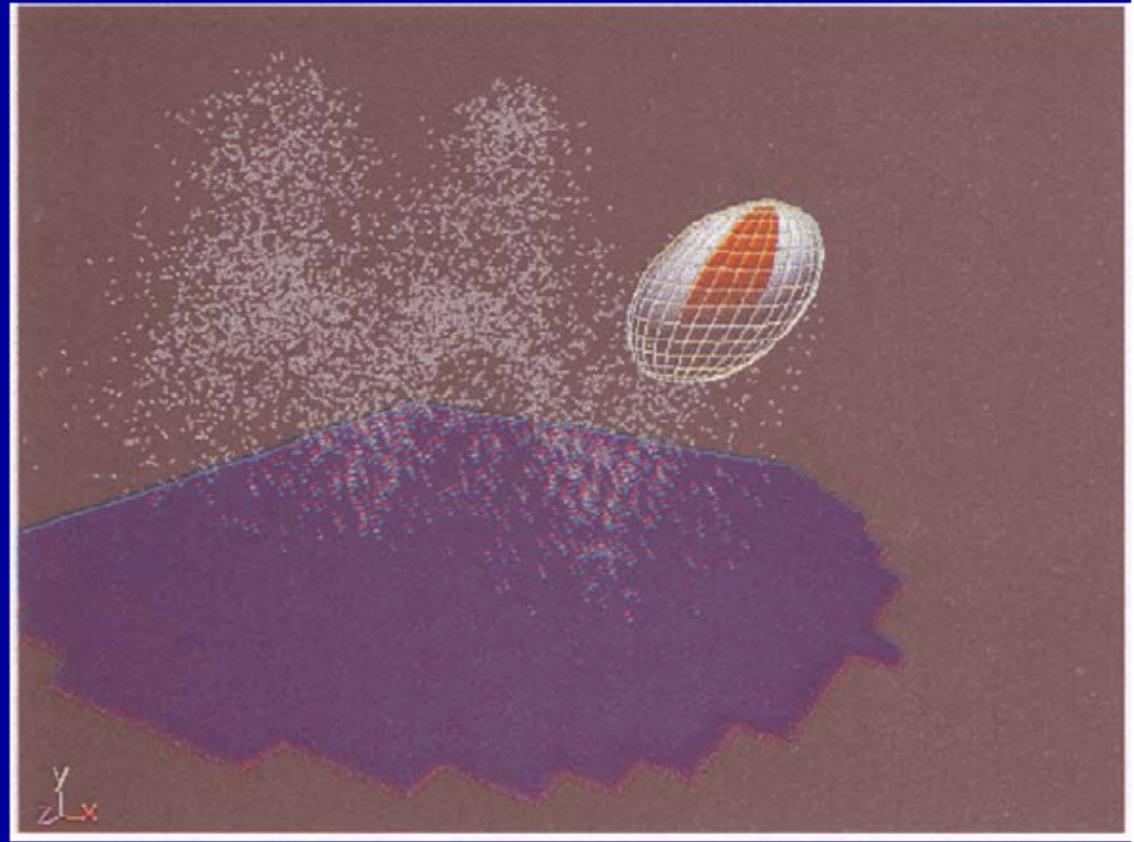
Active—internal sources of energy



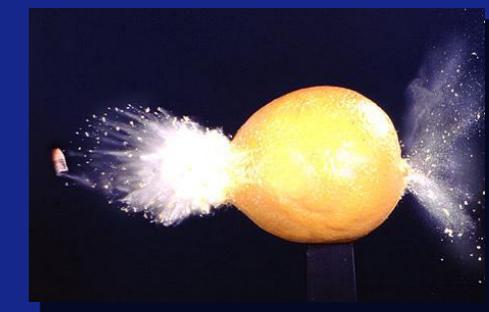
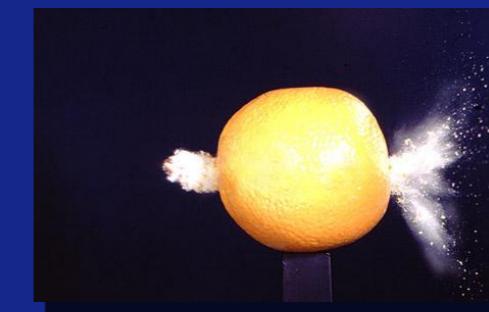
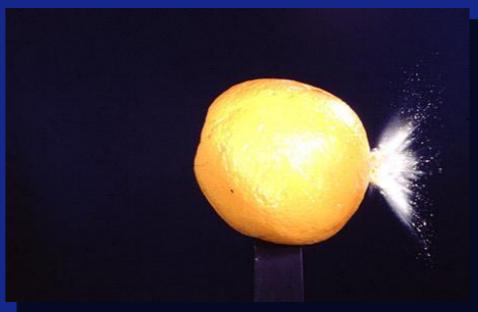
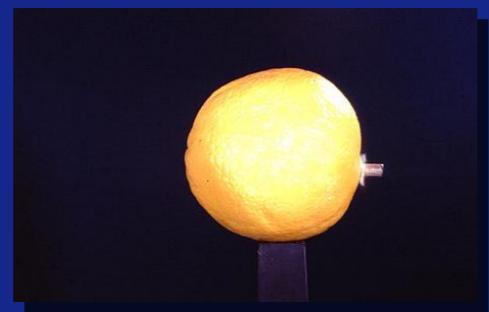
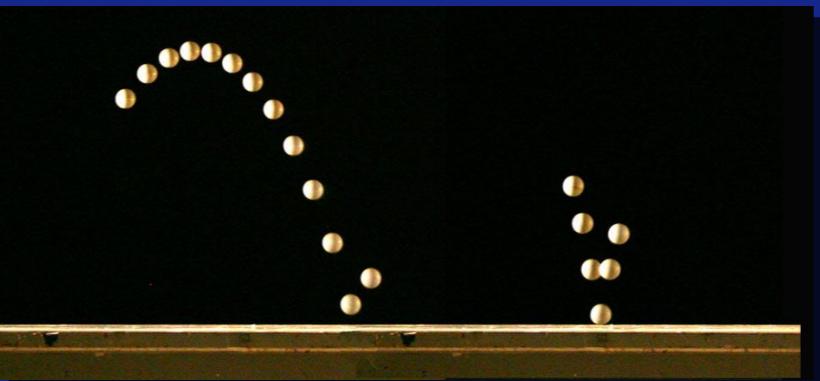
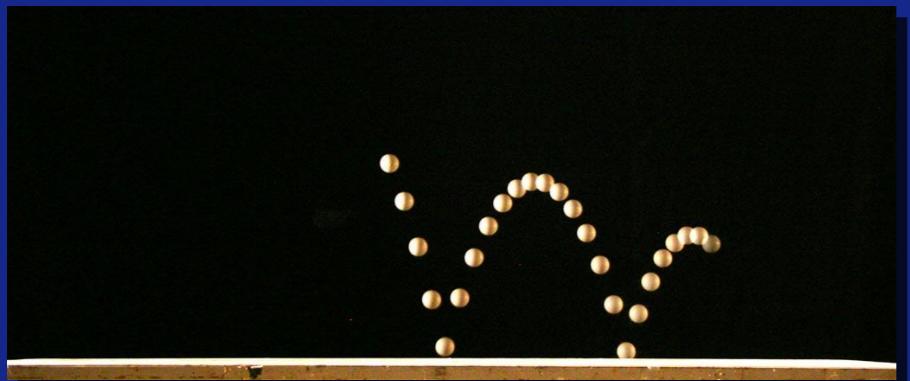
Dynamics

- Generate motion by specifying mass and force, apply physical laws (e.g., Newton's laws)
 - particles
 - soft objects
 - rigid bodies
- Simulates physical phenomena
 - gravity
 - momentum (inertia)
 - collisions
 - friction
 - fluid flow (drag, turbulence, ...)
 - solidity, flexibility, elasticity
 - fracture

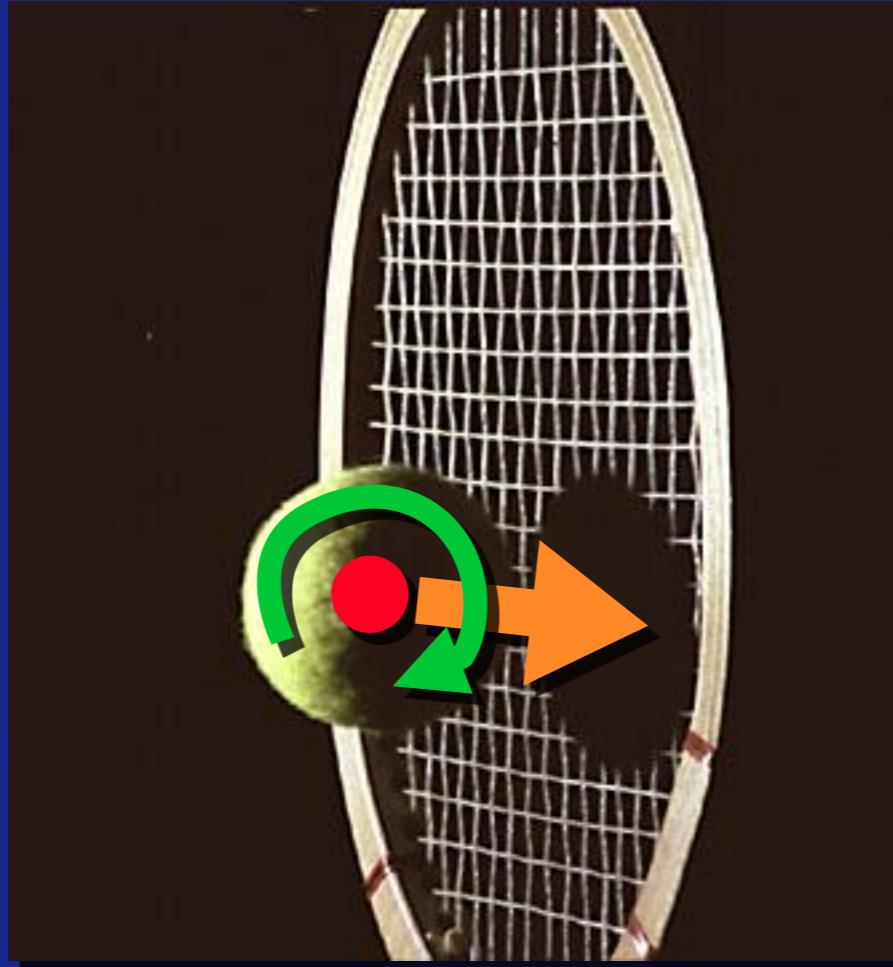
Maya Dynamics



Describing Physics



What variables do we need?



<http://people.rit.edu/andpph/exhibit-8.html>

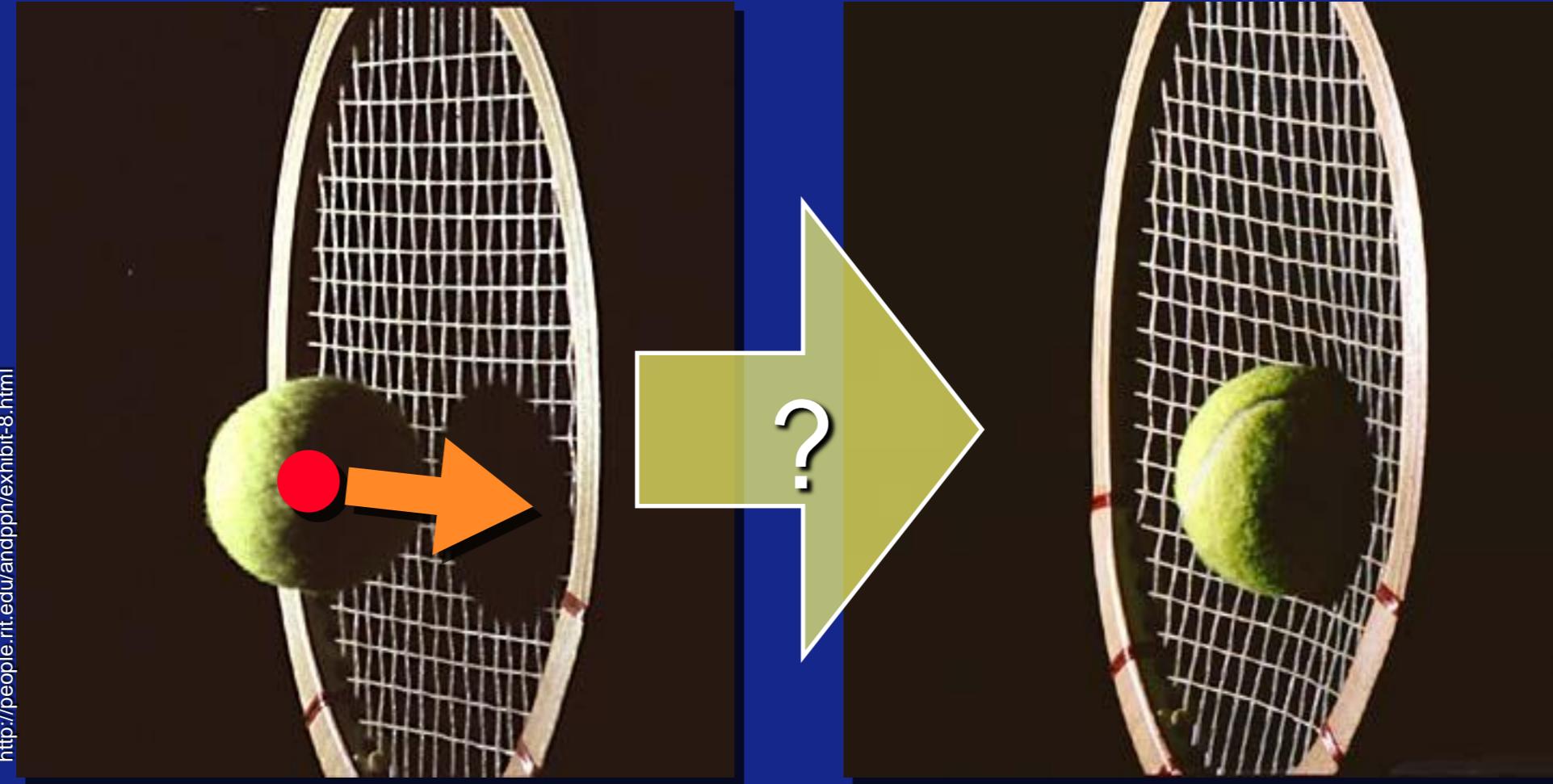
Static

- Radius
- Mass
- Racquet Info

Dynamic

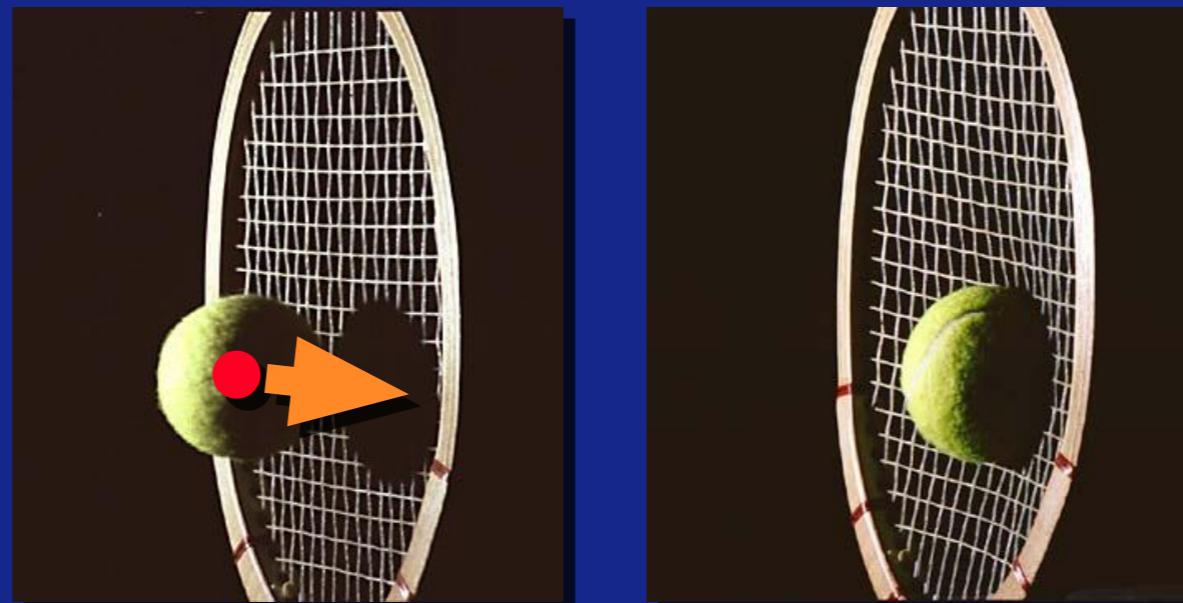
- Position
- Velocity
- Rotation?

What Happens Next?



- Position
 - Velocity
- } $\mathbf{x} = \begin{bmatrix} x \\ y \\ z \\ \dot{x} \\ \dot{y} \\ \dot{z} \end{bmatrix}$
- Discrete Time: $\mathbf{x}_{t+1} = \mathbf{f}(\mathbf{x}_t)$
- Continuous Time: $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x})$

Differential Equations



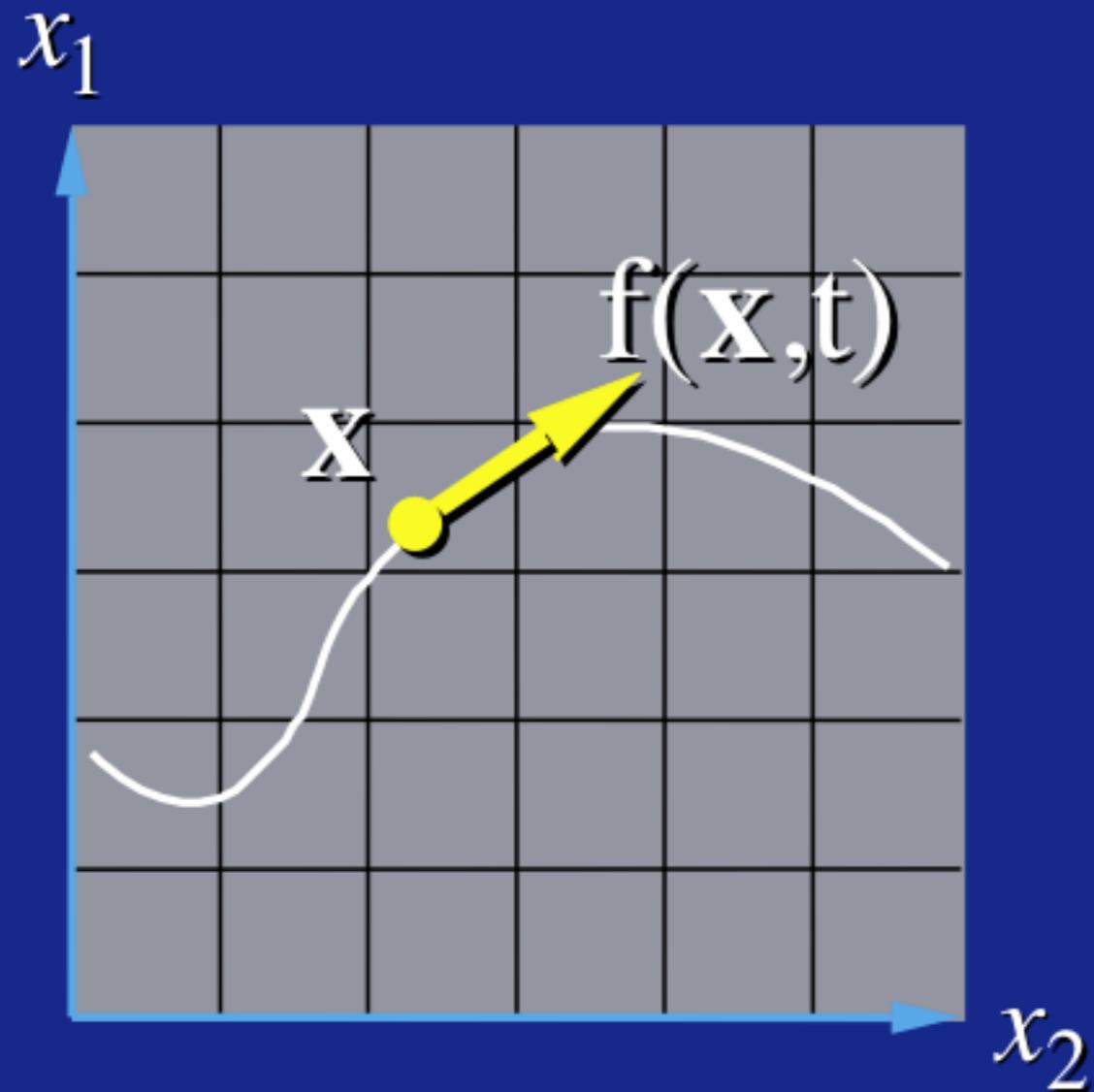
$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x})$$

Differential Equation Basics

Andrew Witkin



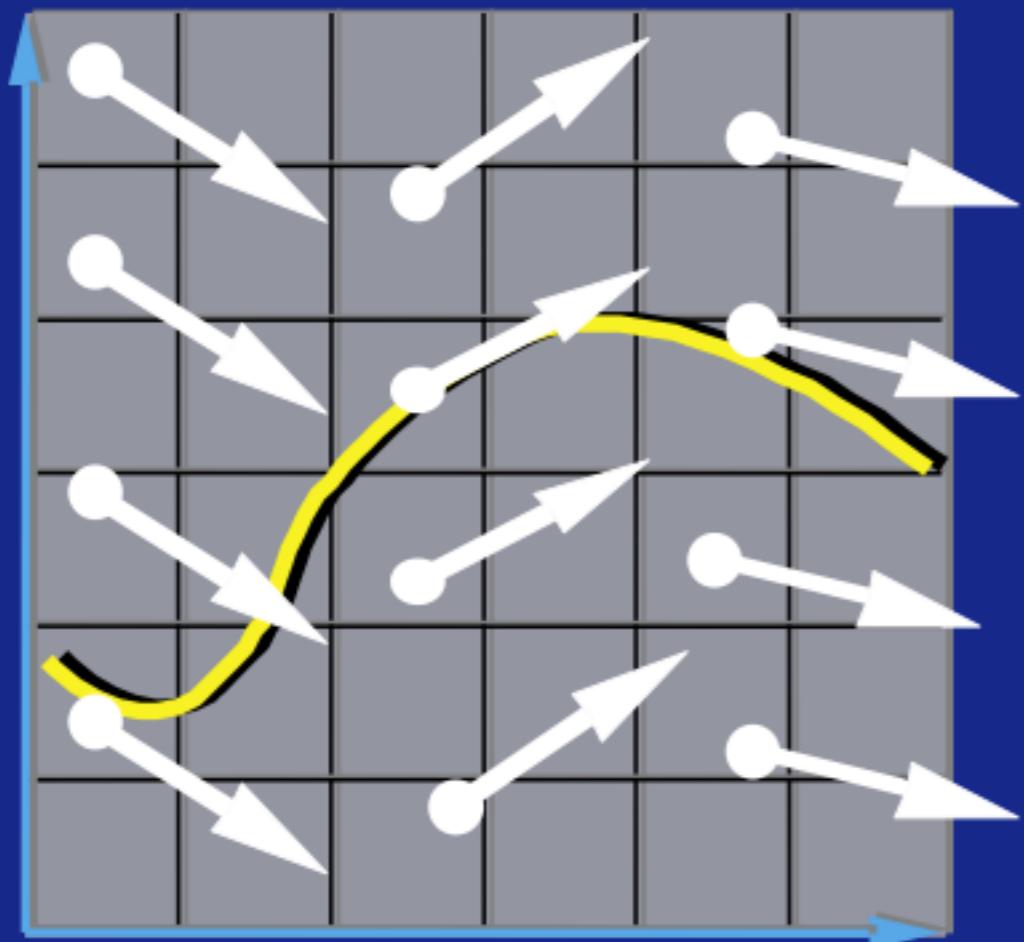
A Canonical Differential Equation



$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, t)$$

- $\mathbf{x}(t)$: a moving point.
- $\mathbf{f}(\mathbf{x}, t)$: \mathbf{x} 's velocity.

Vector Field

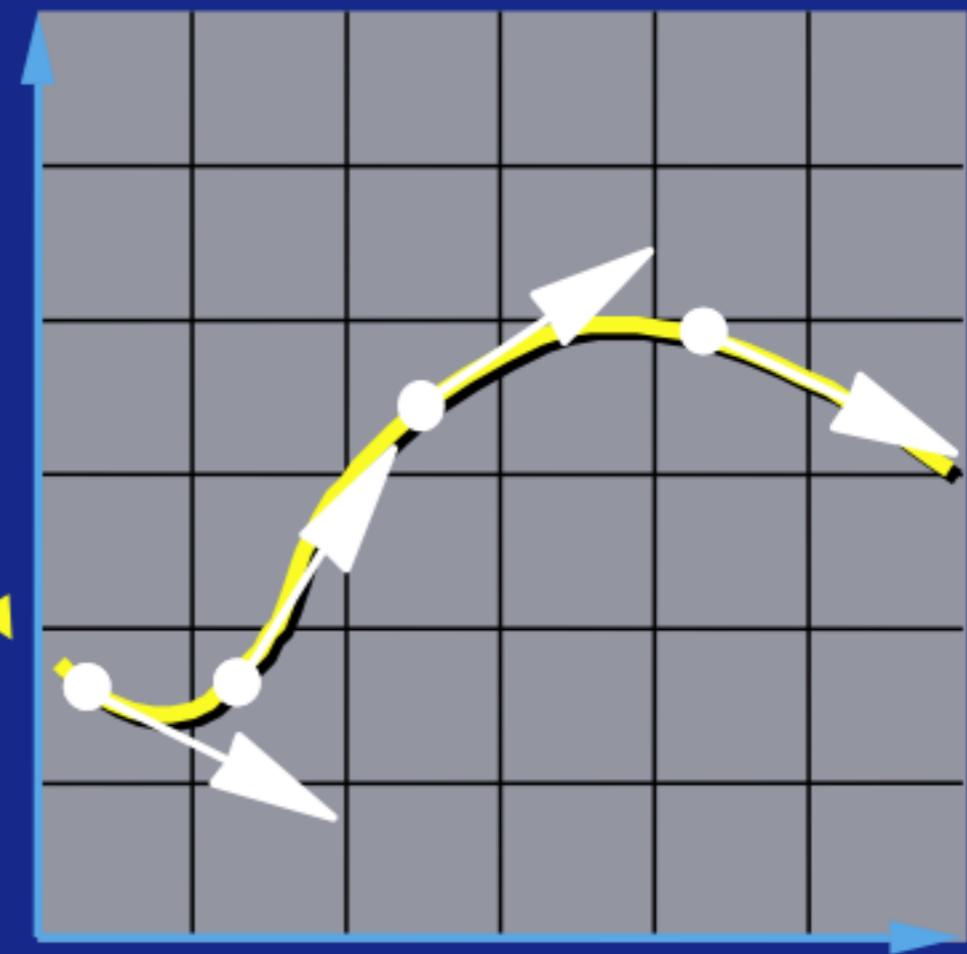


The differential
equation
 $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, t)$
defines a vector
field over \mathbf{x} .

Integral Curves

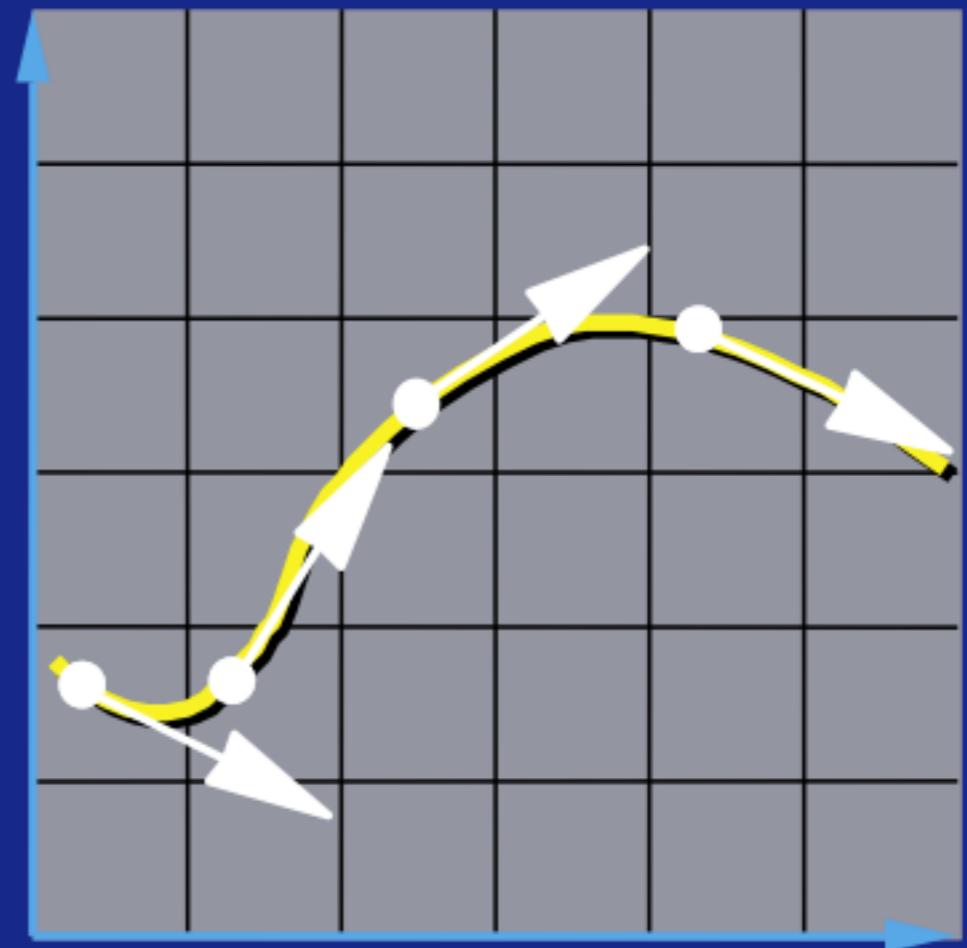
Start Here

Pick any starting point,
and follow the vectors.

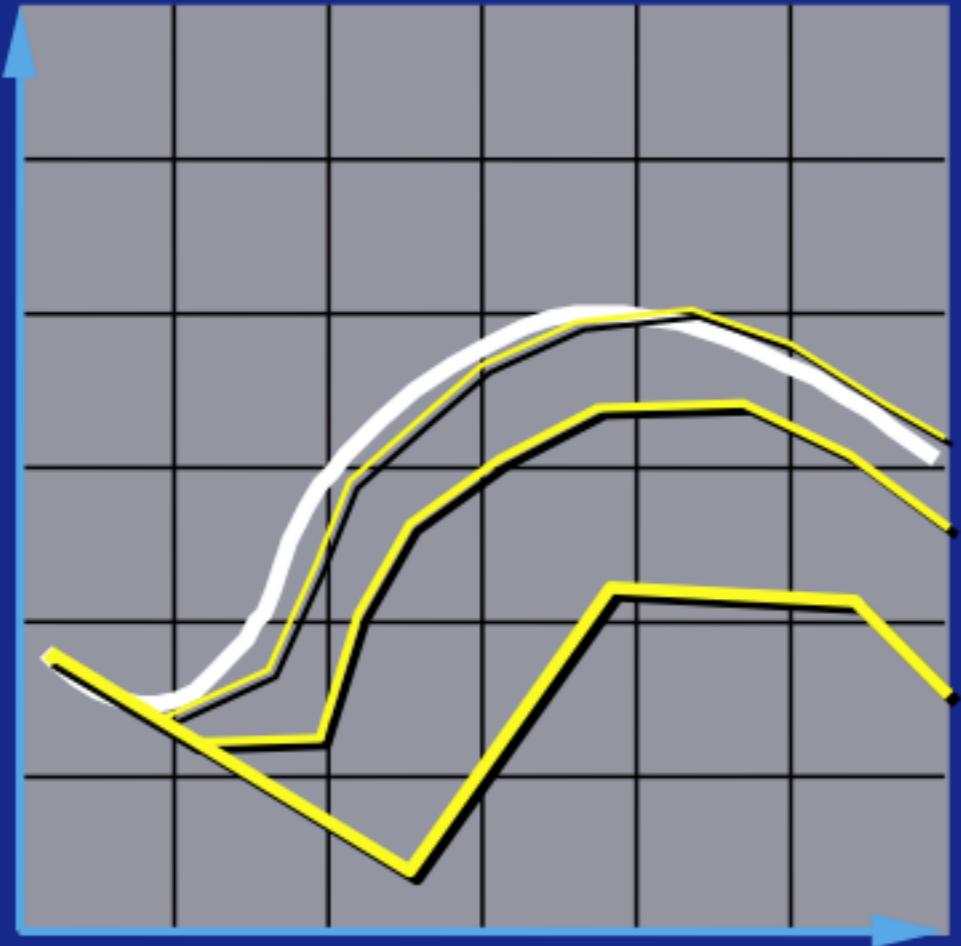


Initial Value Problems

Given the starting point,
follow the integral curve.



Euler's Method



- Simplest numerical solution method
- Discrete time steps
- Bigger steps, bigger errors.

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

Two Problems

- Accuracy
- Instability

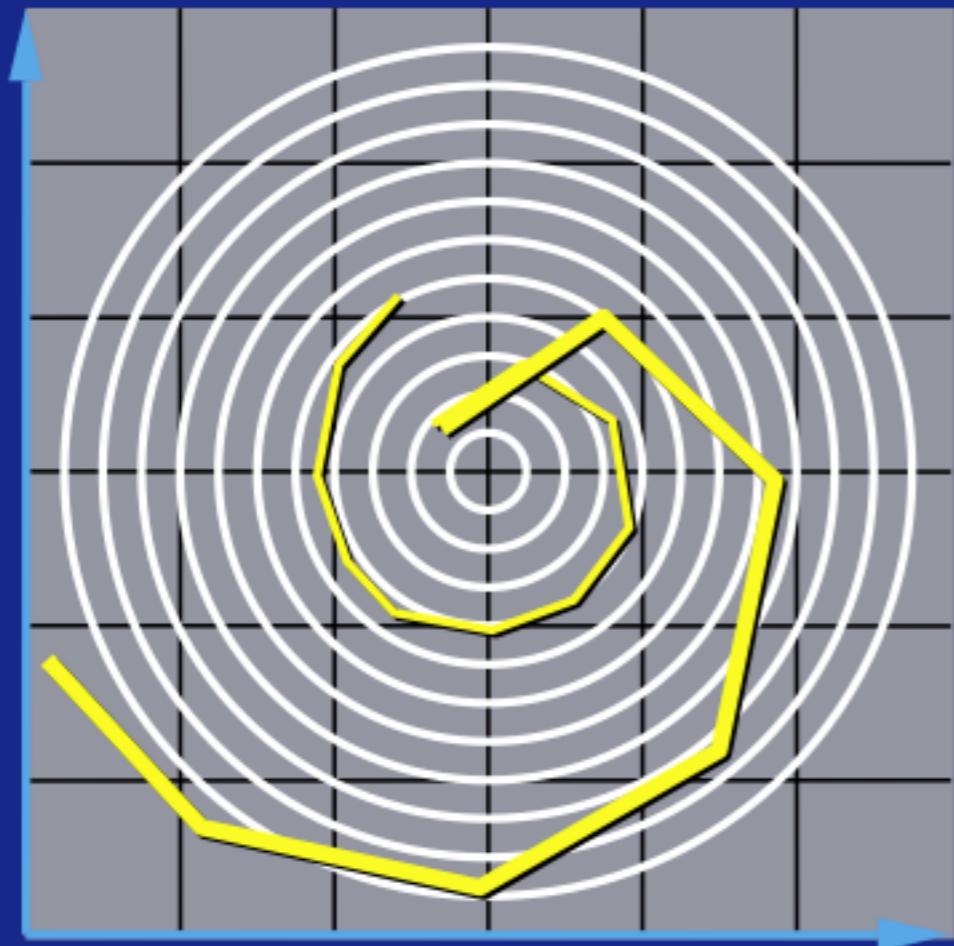
Accuracy

Consider the equation:

$$\dot{\mathbf{x}} = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} \mathbf{x}$$

What do the integral curves look like?

Problem I: Inaccuracy



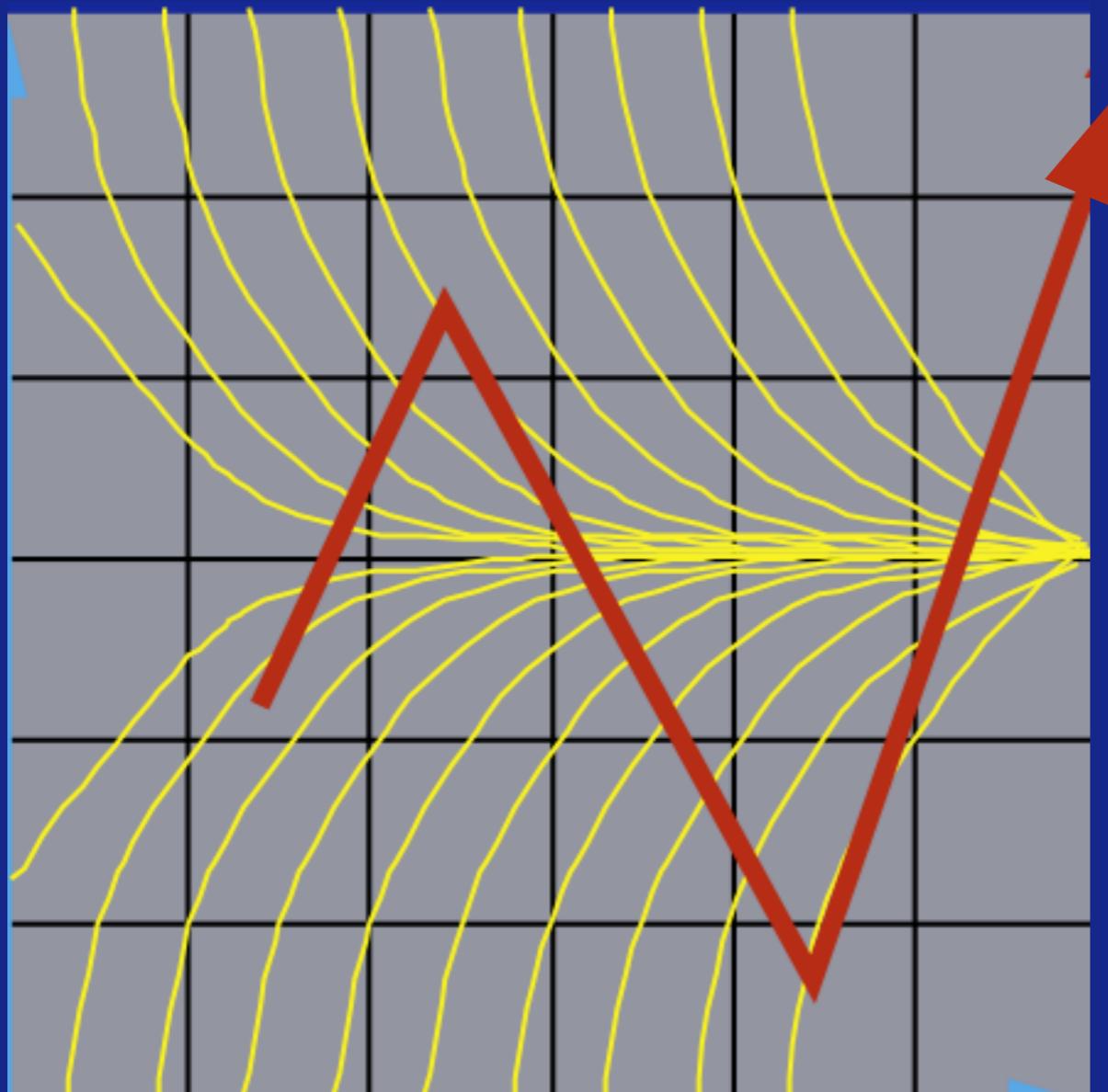
Error turns $x(t)$ from a circle into the spiral of your choice.

Problem 2: Instability

- Consider the following system:

$$\begin{cases} \dot{x} = -x \\ x(0) = 1 \end{cases}$$

Problem 2: Instability



to Neptune!

Accuracy of Euler Method

$$\dot{x} = f(x)$$

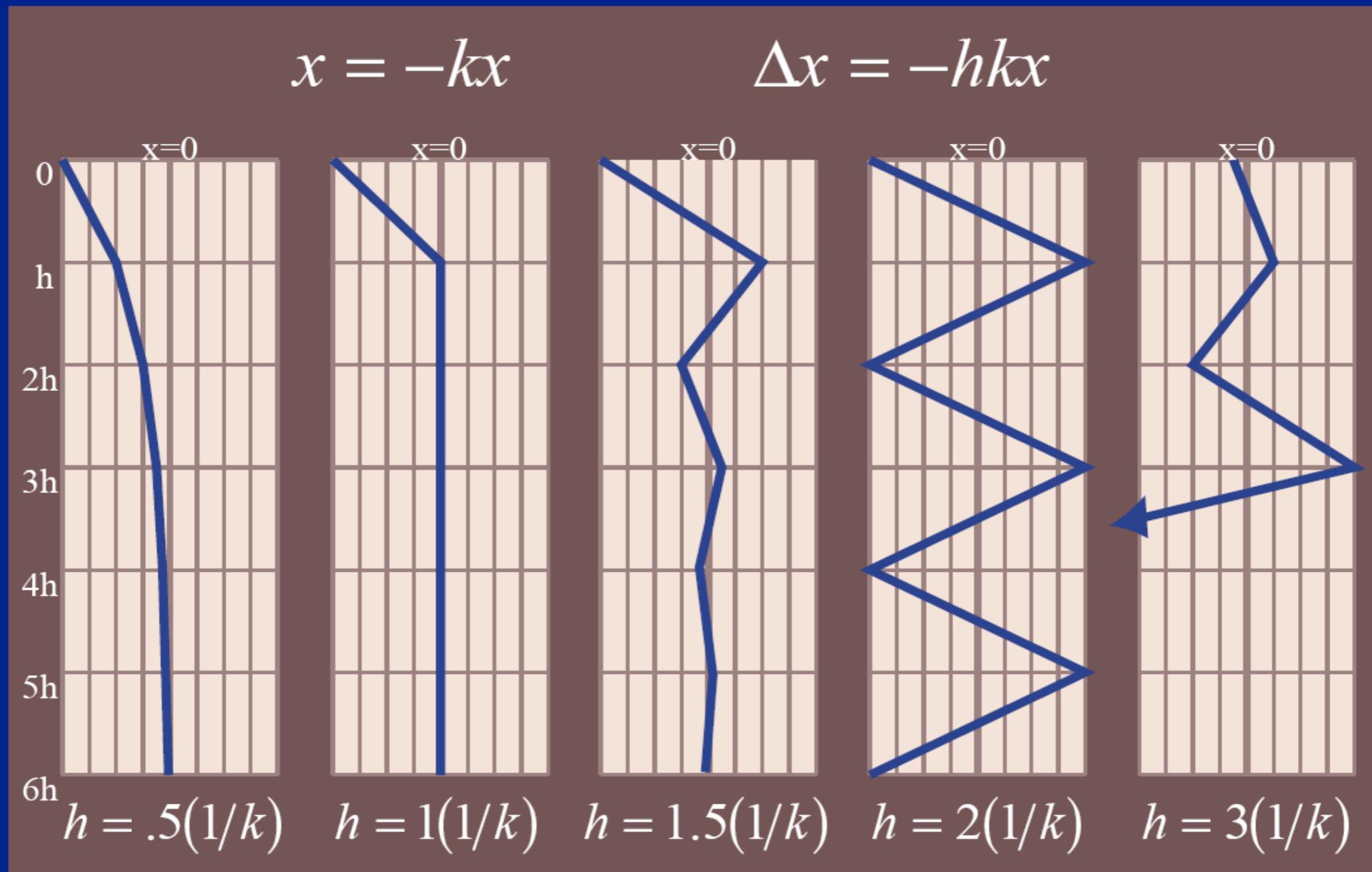
Consider Taylor Expansion about $x(t)$...

$$x(t+h) \equiv \frac{\text{Euler step}}{\underline{x(t)} + \underline{h f(x(t))}} + \frac{\text{error}}{\underline{\text{constant}} \quad \underline{\text{linear}} \quad \underline{\text{everything else}}}$$

Euler's method has error $O(h^2)$... first order.

How can we get to $O(h^3)$ error?

Euler's method has a speed limit



$h > 1/k$: oscillate.

$h > 2/k$: explode!

The Midpoint Method

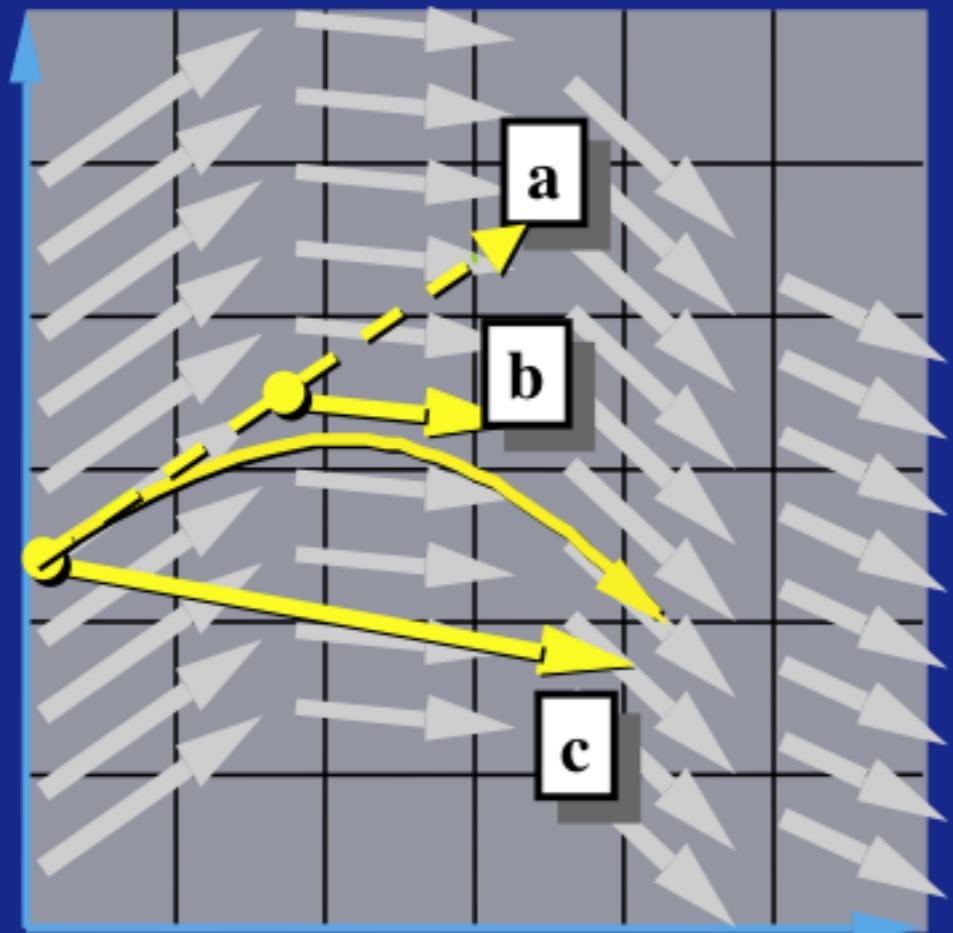
- Also known as second order Runge-Kutta:

$$k_1 = h(f(x_0, t_0))$$

$$k_2 = h f\left(x_0 + \frac{k_1}{2}, t_0 + \frac{h}{2}\right)$$

$$x(t_0 + h) \equiv x_0 + k_2 + O(h^3)$$

The Midpoint Method



a. Compute an Euler step

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

b. Evaluate \mathbf{f} at the midpoint

$$\mathbf{f}_{\text{mid}} = \mathbf{f}\left(\frac{\mathbf{x} + \Delta \mathbf{x}}{2}, \frac{t + \Delta t}{2}\right)$$

c. Take a step using the midpoint value

$$\mathbf{x}(t + \Delta t) = \mathbf{x}(t) + \Delta t \mathbf{f}_{\text{mid}}$$

4th-Order Runge-Kutta

$$k_1 \equiv h f(x_0, t_0)$$

$$k_2 \equiv h f\left(x_0 + \frac{k_1}{2}, t_0 + \frac{h}{2}\right)$$

Very popular

$$k_3 \equiv h f\left(x_0 + \frac{k_2}{2}, t_0 + \frac{h}{2}\right)$$

$$k_4 \equiv h f(x_0 + k_3, t_0 + h)$$

$$x(t_0 + h) = x_0 + \frac{1}{6}k_1 + \frac{1}{3}k_2 + \frac{1}{3}k_3 + \frac{1}{6}k_4 + O(h^5)$$

q-Stage Runge-Kutta

General Form:

$$x(t_0 + h) \equiv x_0 + h \sum_{i=1}^q w_i k_i$$

where:

$$k_i = f \left(x_0 + h \sum_{j=1}^{i-1} \beta_{ij} k_j \right)$$

Find the constant that ensure accuracy $O(h^n)$.

stability is all stability is all stability is all

- If your step size is too big, your simulation blows up. It isn't pretty.
- Sometimes you have to make the step size so small that you never get anywhere.
- Nasty cases: cloth, constrained systems.

Implicit Euler Method

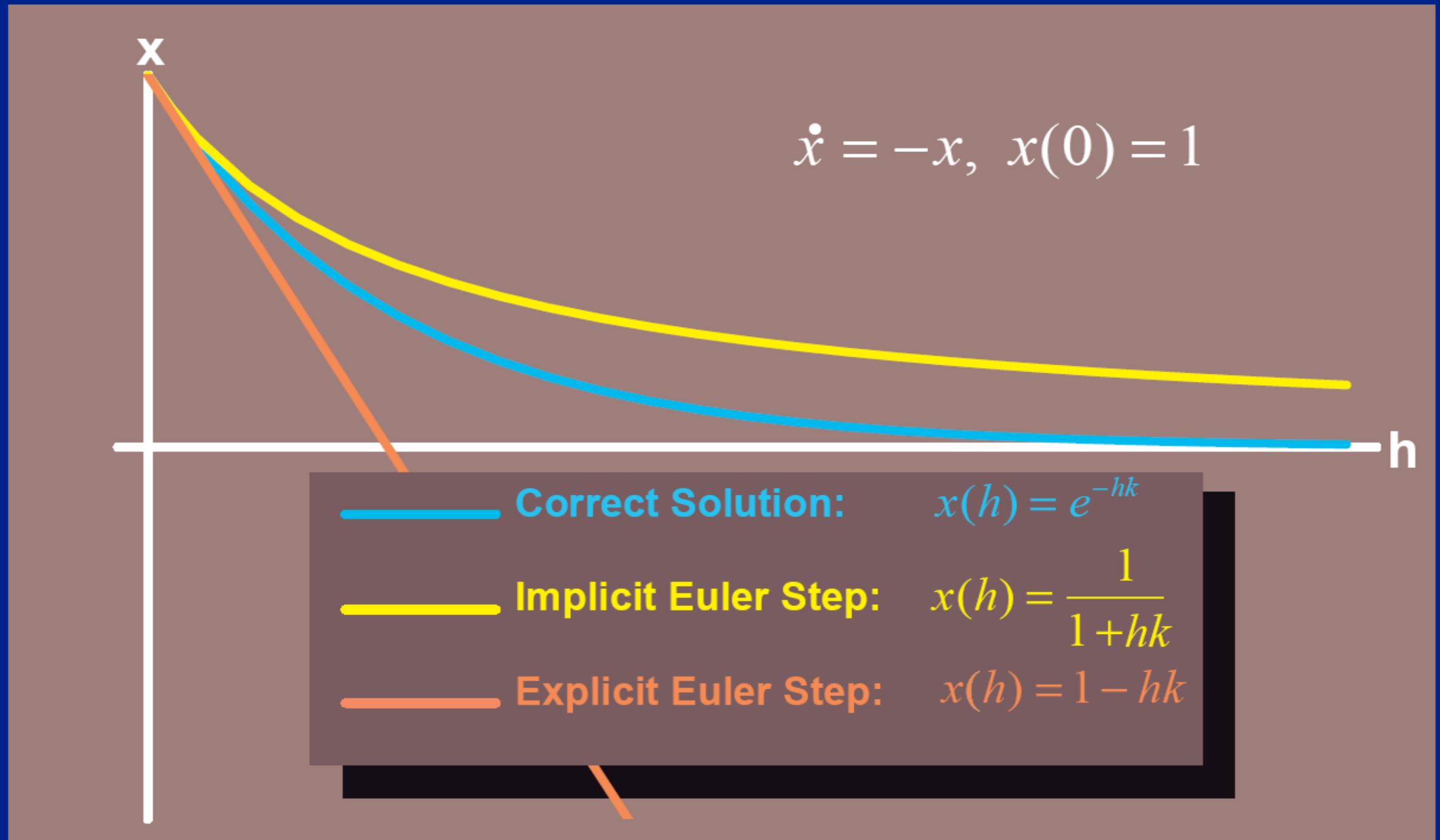
$$x(t_0 + h) = x(t_0) + h \dot{x}(t_0)$$

$$x(t_0 + h) = x(t_0) + h \dot{x}(t_0 + \Delta t)$$

Implicit Euler for $\dot{x} = -kx$

$$\begin{aligned}x(t+h) &= x(t) + h \dot{x}(t+h) \\&= x(t) - h k x(t+h) \\&= \frac{x(t)}{1 + hk}\end{aligned}$$

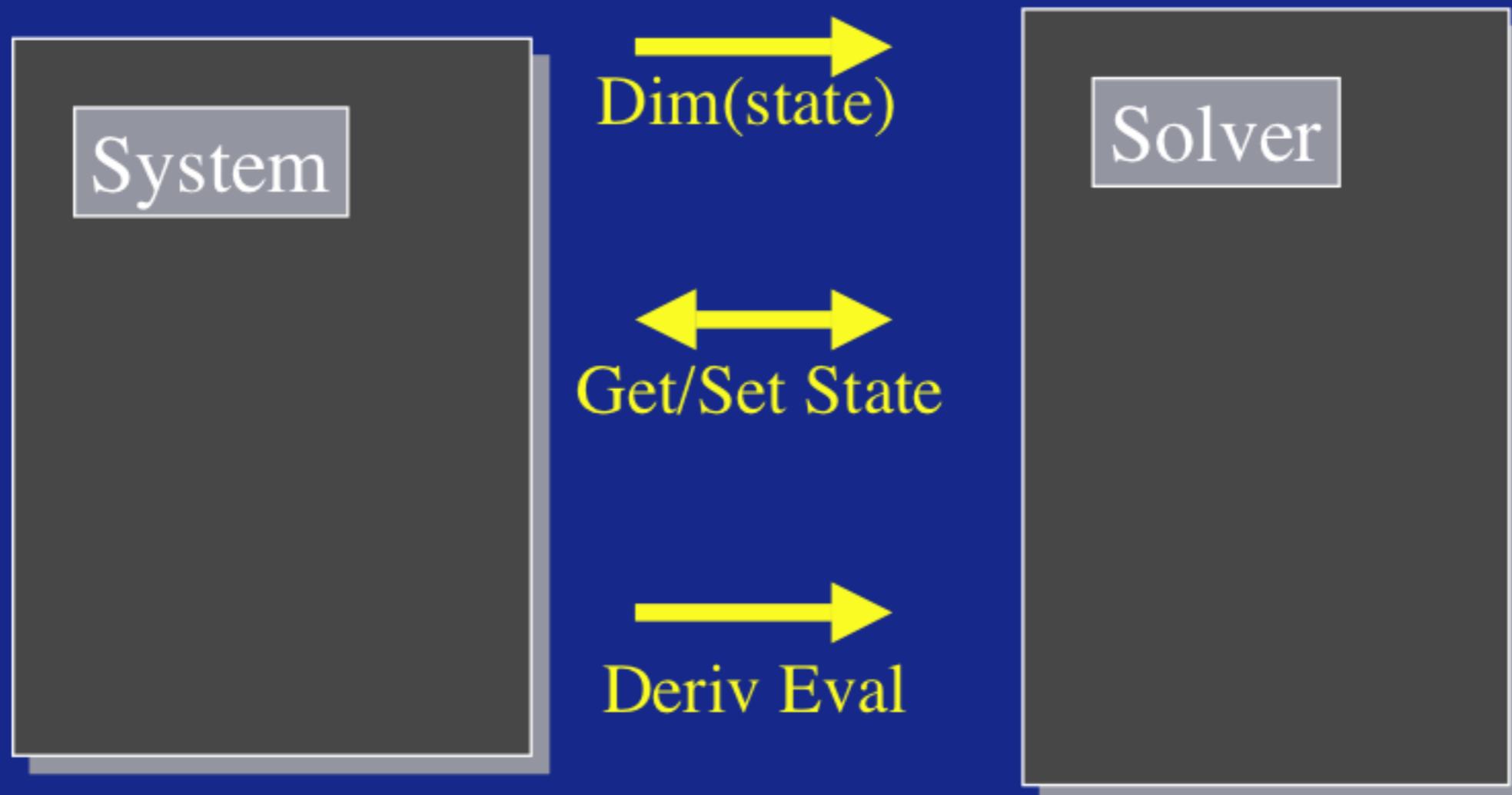
One Step: Implicit vs. Explicit



Modular Implementation

- Generic operations:
 - Get $\text{dim}(x)$
 - Get/set x and t
 - Deriv Eval at current (x,t)
- Write solvers in terms of these.
 - Re-usable solver code.
 - Simplifies model implementation.

Solver Interface



A Code Fragment

```
void eulerStep(Sys sys, float h) {  
    float t = getTime(sys);  
    vector<float> x0, deltaX;  
  
    t = getTime(sys);  
    x0 = getState(sys);  
    deltaX = derivEval(sys, x0, t);  
    setState(sys, x0 + h*deltaX, t+h);  
}
```

Particle Systems

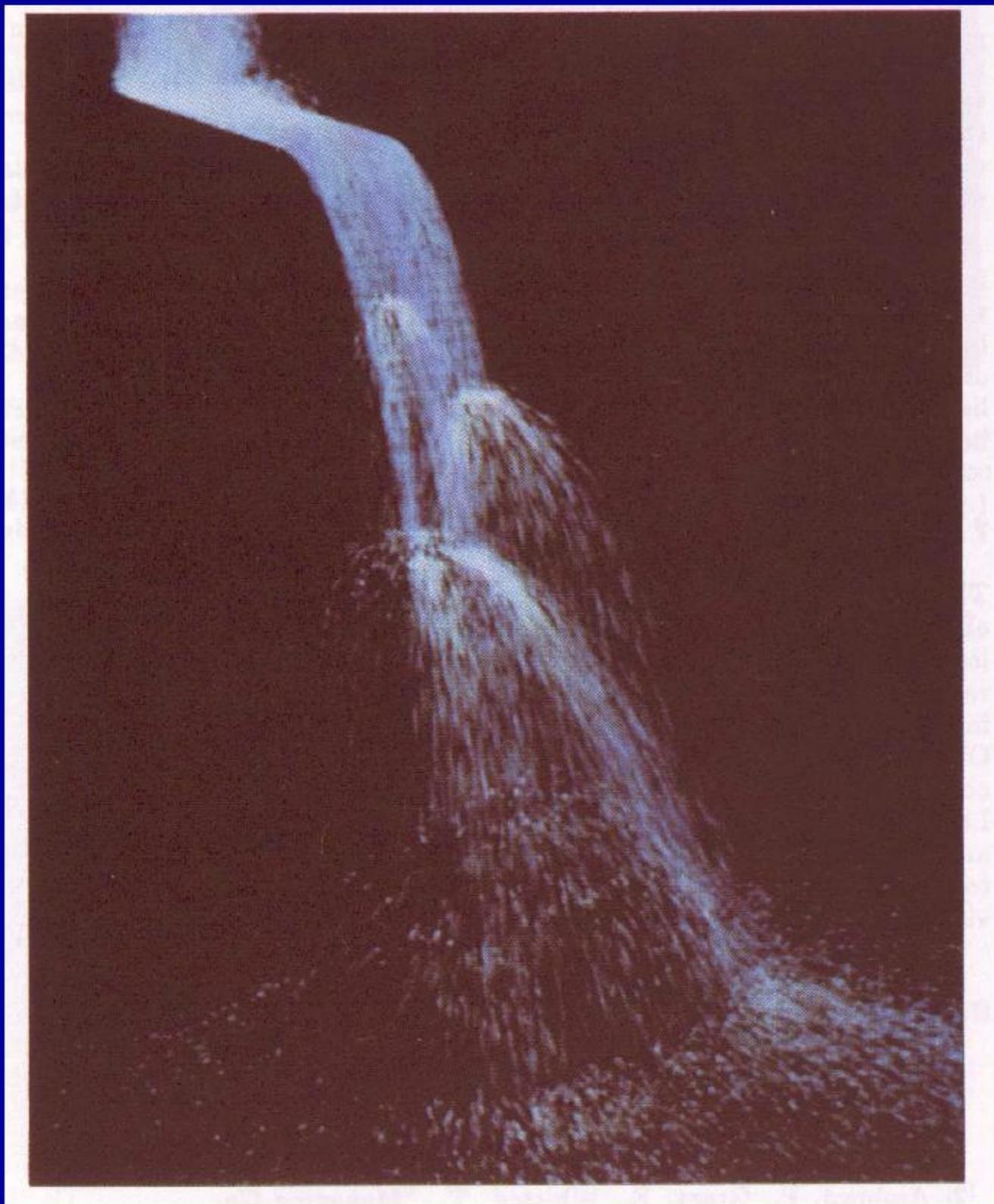
Particle Systems

Clouds
Smoke
Fire
Waterfalls
Fireworks



Reeves '83, the Wrath of Khan
Batman Returns, using Reynold's flocking algorithms

Karl Sims, Particle Dreams



What are particle systems?

A **particle system** is a collection of point masses that obeys some physical laws (e.g, gravity or spring behaviors).

Particle systems can be used to simulate all sorts of physical phenomena:

- Smoke
- Snow
- Fireworks
- Hair
- Cloth
- Snakes
- Fish

Overview

1. One lousy particle
2. Particle systems
3. Forces: gravity, springs
4. Implementation

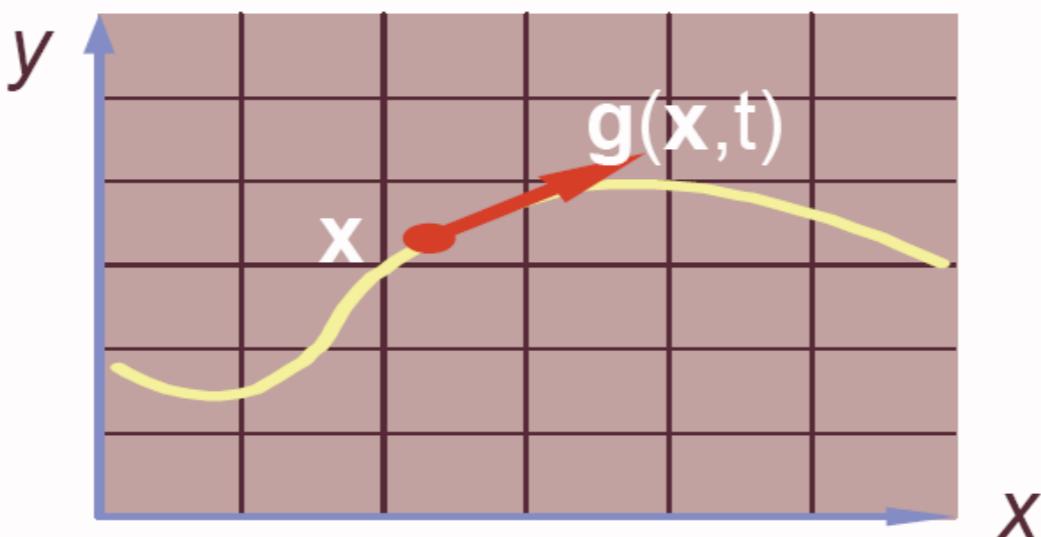
Particle in a flow field

We begin with a single particle with:

- Position, $\mathbf{x} = \begin{bmatrix} x \\ y \end{bmatrix}$
- Velocity, $\mathbf{v} \equiv \dot{\mathbf{x}} = \frac{d\mathbf{x}}{dt} = \begin{bmatrix} dx/dt \\ dy/dt \end{bmatrix}$

Suppose the velocity is dictated by some driving function \mathbf{g} :

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



Particle in a force field

- Now consider a particle in a force field \mathbf{f} .
- In this case, the particle has:
 - Mass, m
 - Acceleration, $\mathbf{a} \equiv \ddot{\mathbf{x}} = \frac{d\mathbf{v}}{dt} = \frac{d^2\mathbf{x}}{dt^2}$
- The particle obeys Newton's law: $\mathbf{f} = m\mathbf{a} = m\ddot{\mathbf{x}}$
- The force field \mathbf{f} can in general depend on the position and velocity of the particle as well as time.
- Thus, with some rearrangement, we end up with:

$$\ddot{\mathbf{x}} = \frac{\mathbf{f}(\mathbf{x}, \dot{\mathbf{x}}, t)}{m}$$

Second order equations

This equation: $\ddot{\mathbf{x}} = \frac{\mathbf{f}(\mathbf{x}, \mathbf{v}, t)}{m}$

is a **second order differential equation**.

Our solution method, though, worked on first order differential equations.

We can rewrite this as:

$$\begin{bmatrix} \dot{\mathbf{x}} = \mathbf{v} \\ \dot{\mathbf{v}} = \frac{\mathbf{f}(\mathbf{x}, \mathbf{v}, t)}{m} \end{bmatrix}$$

where we have added a new variable \mathbf{v} to get a pair of **coupled first order equations**.

Phase space

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{v} \end{bmatrix}$$

Concatenate \mathbf{x} and \mathbf{v} to make a 6-vector: position
in **phase space**.

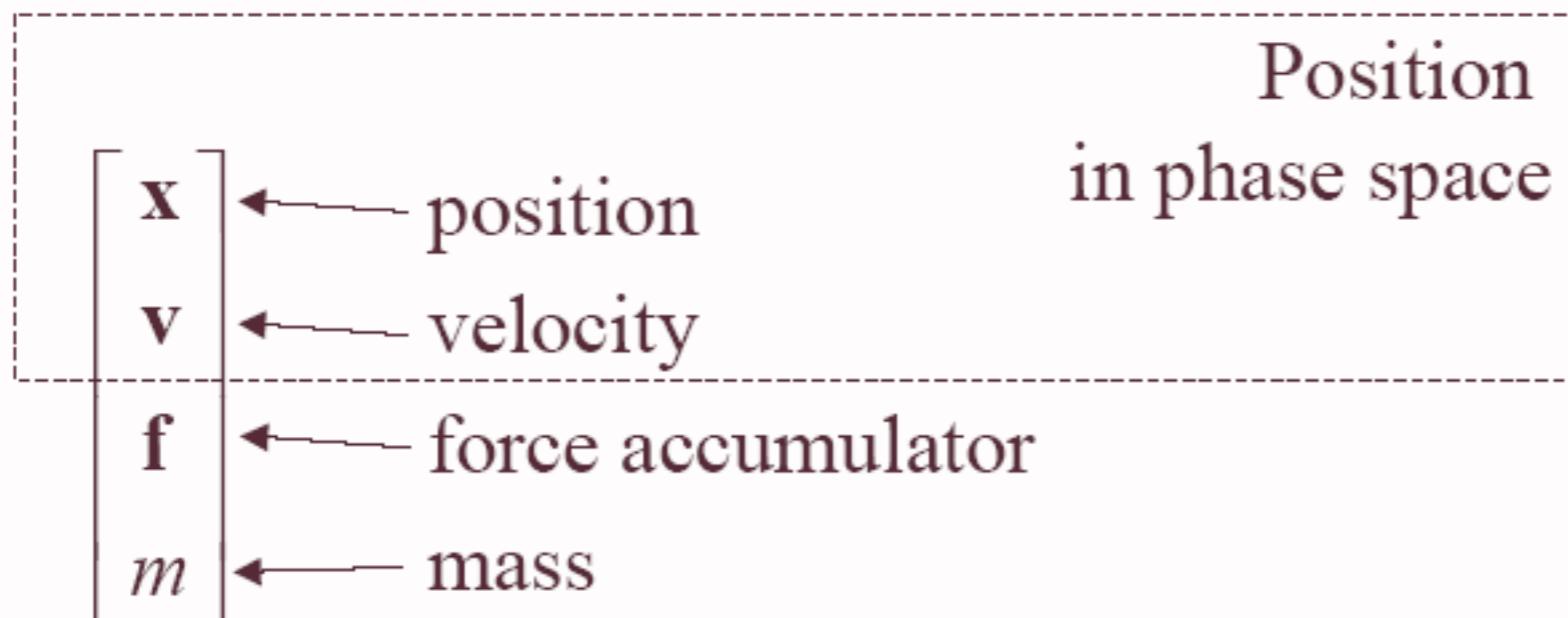
$$\begin{bmatrix} \dot{\mathbf{x}} \\ \dot{\mathbf{v}} \end{bmatrix}$$

Taking the time derivative: another 6-vector.

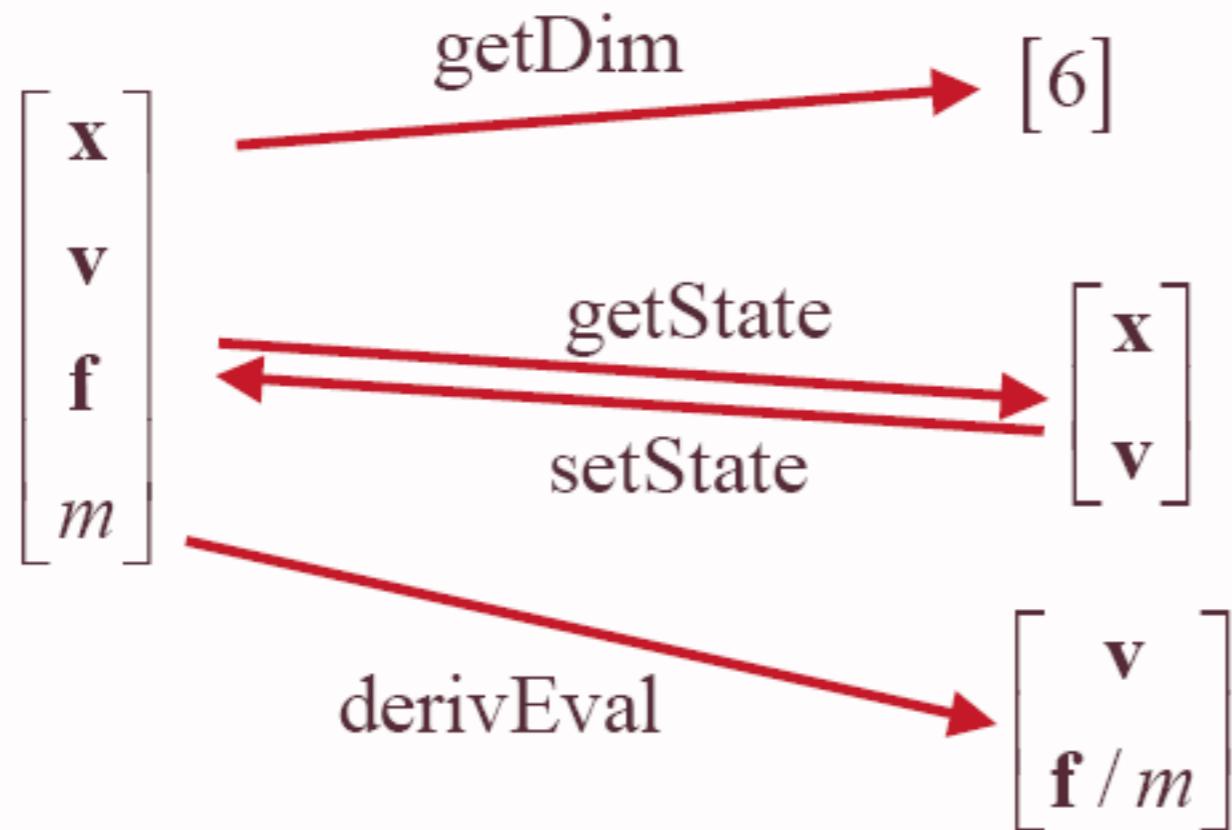
$$\begin{bmatrix} \dot{\mathbf{x}} \\ \dot{\mathbf{v}} \end{bmatrix} = \begin{bmatrix} \mathbf{v} \\ \mathbf{f}/m \end{bmatrix}$$

A vanilla 1st-order differential equation.

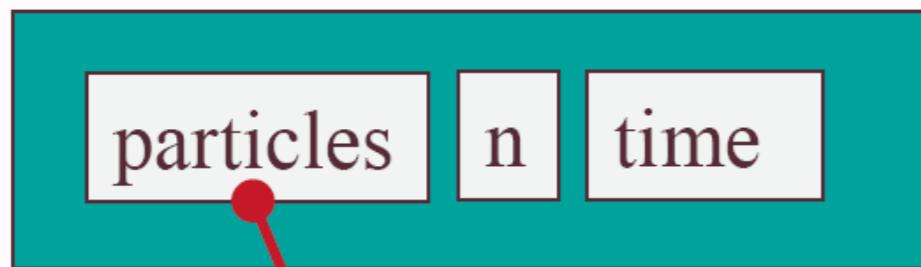
Particle structure



Solver interface

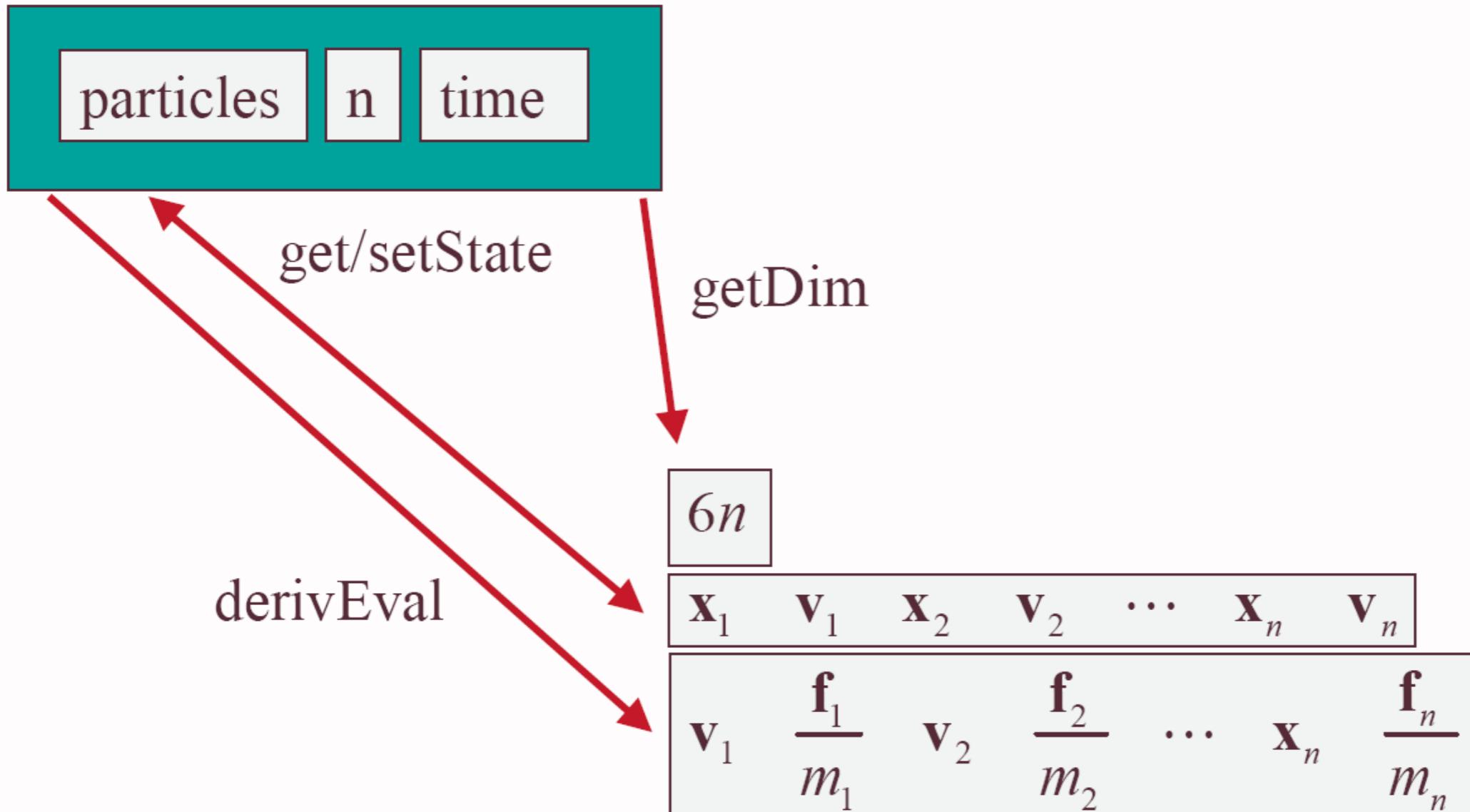


Particle systems



$$\begin{bmatrix} \mathbf{x}_1 \\ \mathbf{v}_1 \\ \mathbf{f}_1 \\ m_1 \end{bmatrix} \begin{bmatrix} \mathbf{x}_2 \\ \mathbf{v}_2 \\ \mathbf{f}_2 \\ m_2 \end{bmatrix} \begin{bmatrix} \mathbf{x}_3 \\ \mathbf{v}_3 \\ \mathbf{f}_3 \\ m_3 \end{bmatrix} \dots \begin{bmatrix} \mathbf{x}_n \\ \mathbf{v}_n \\ \mathbf{f}_n \\ m_n \end{bmatrix}$$

Solver interface



Forces

- Constant (gravity)
- Position/time dependent (force fields)
- Velocity-dependent (drag)
- N-ary (springs)

Gravity

Force law:

$$\mathbf{f}_{grav} = m\mathbf{G}$$

$$\mathbf{p} \rightarrow \mathbf{f} += \mathbf{p} \rightarrow \mathbf{m} * \mathbf{F} \rightarrow \mathbf{G}$$

Viscous drag

Force law:

$$\mathbf{f}_{drag} = -k_{drag} \mathbf{v}$$

$$\mathbf{p} \rightarrow \mathbf{f} = \mathbf{F} \rightarrow \mathbf{k} * \mathbf{p} \rightarrow \mathbf{v}$$

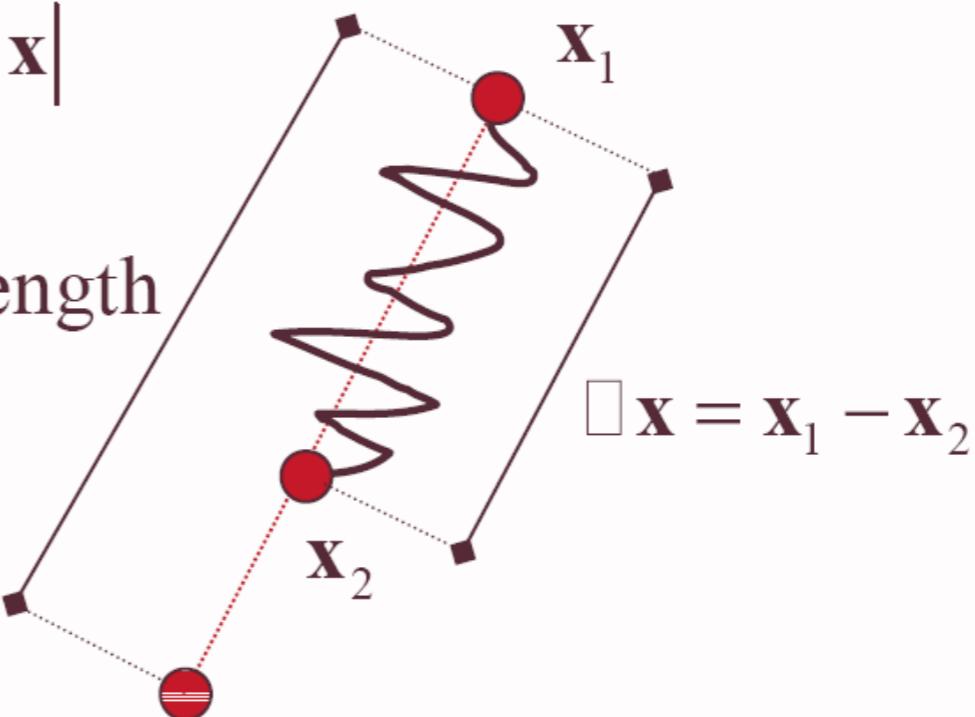
Damped spring

Force law:

$$\mathbf{f}_1 = - \left[k_s (|\mathbf{x}| - \mathbf{r}) + k_d \left(\frac{\mathbf{v} \cdot \mathbf{x}}{|\mathbf{x}|} \right) \right] \frac{\mathbf{x}}{|\mathbf{x}|}$$

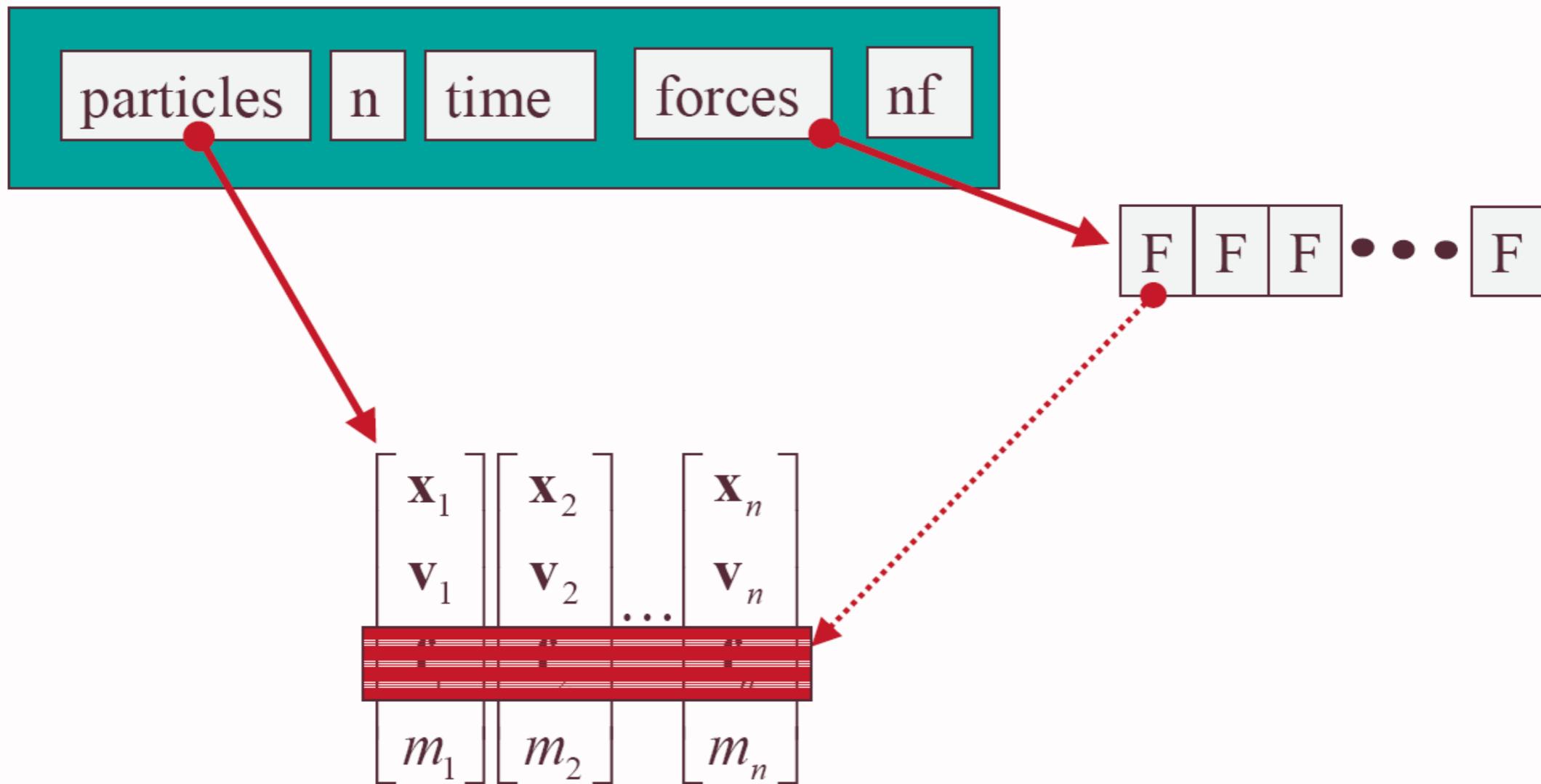
$$\mathbf{f}_2 = -\mathbf{f}_1$$

\mathbf{r} = rest length



$$\Delta \mathbf{v} = \mathbf{v}_1 - \mathbf{v}_2$$

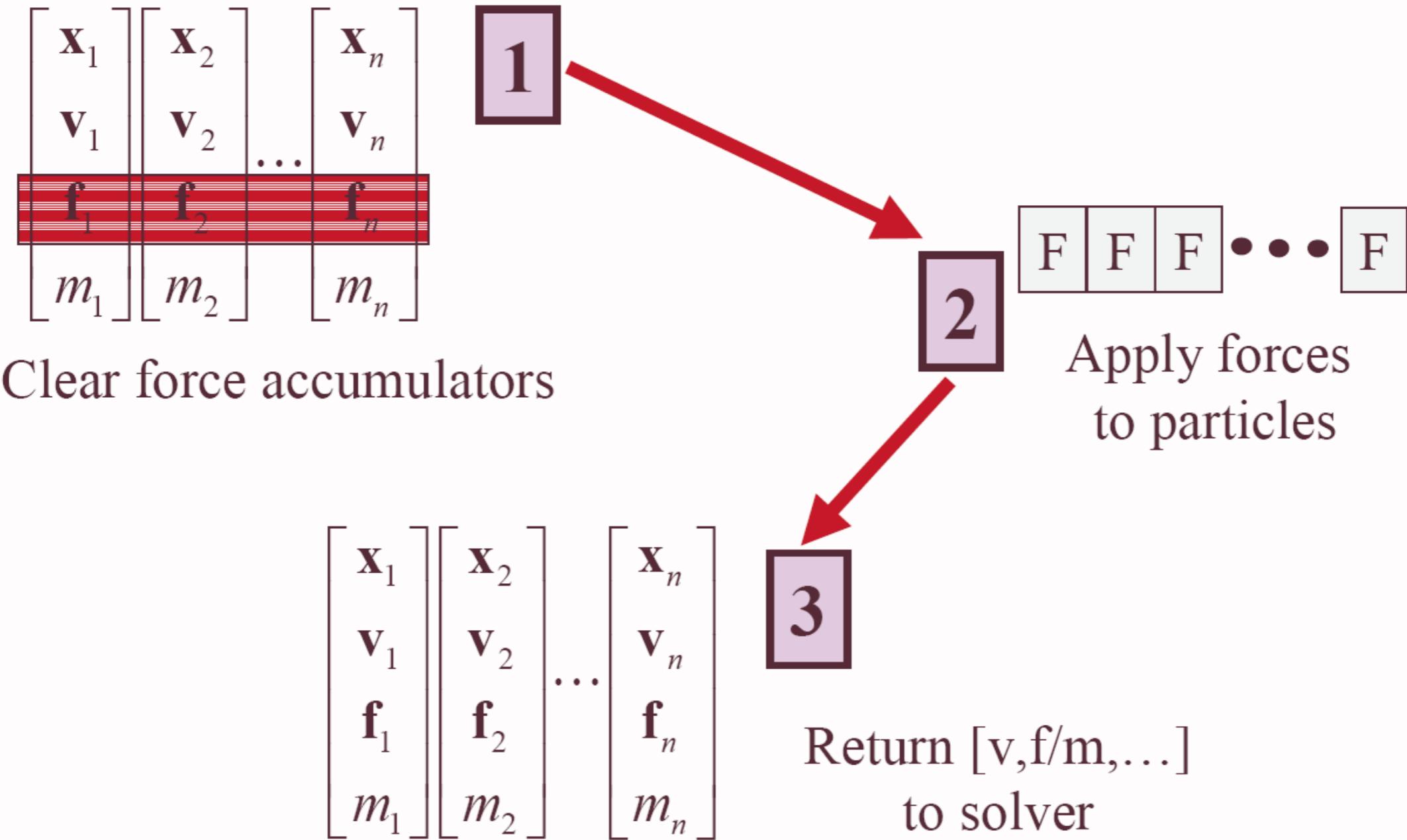
Particle systems with forces



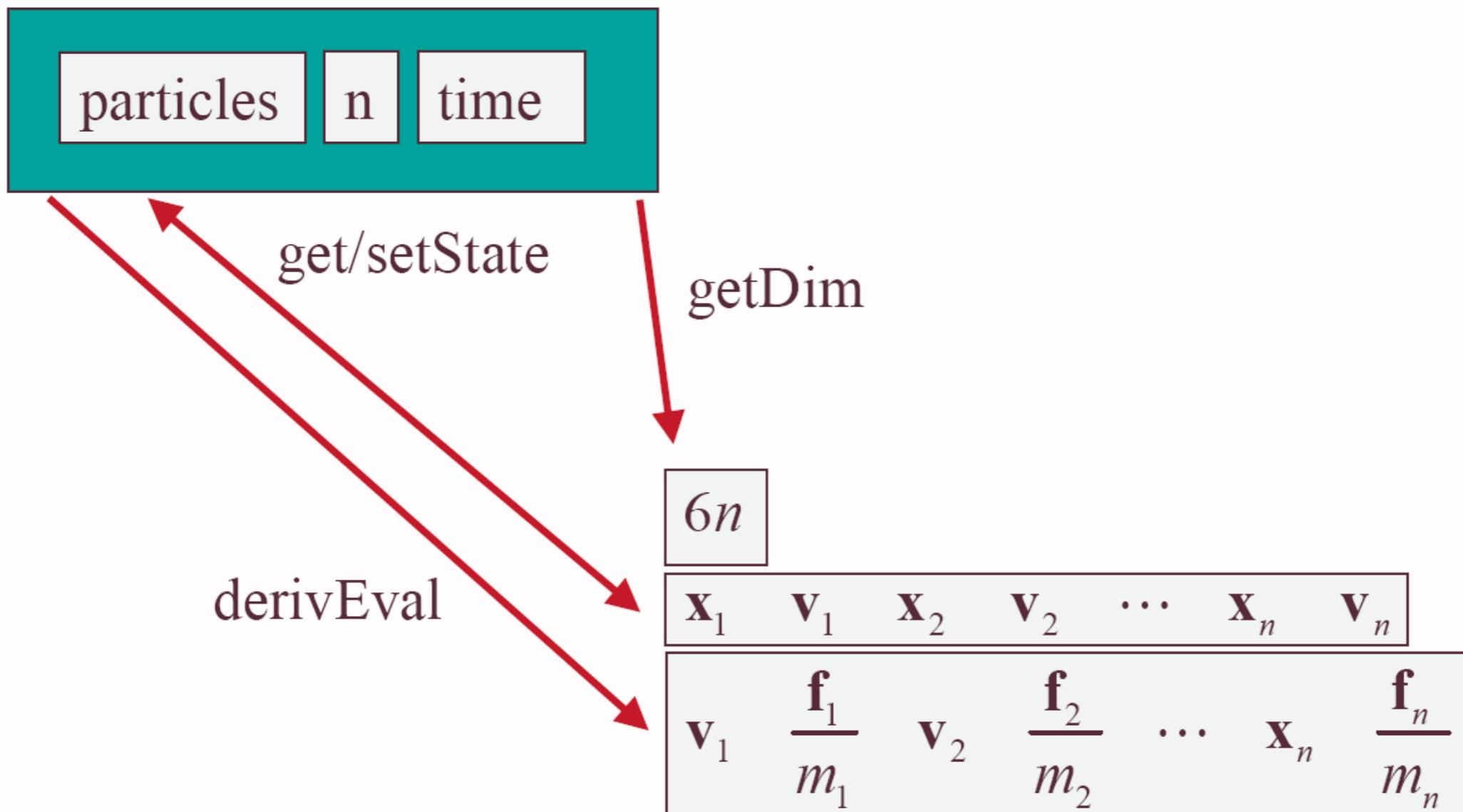
derivEval loop

1. Clear forces
 - Loop over particles, zero force accumulators
2. Calculate forces
 - Sum all forces into accumulators
3. Gather
 - Loop over particles, copying v and f/m into destination array

derivEval Loop



Solver interface



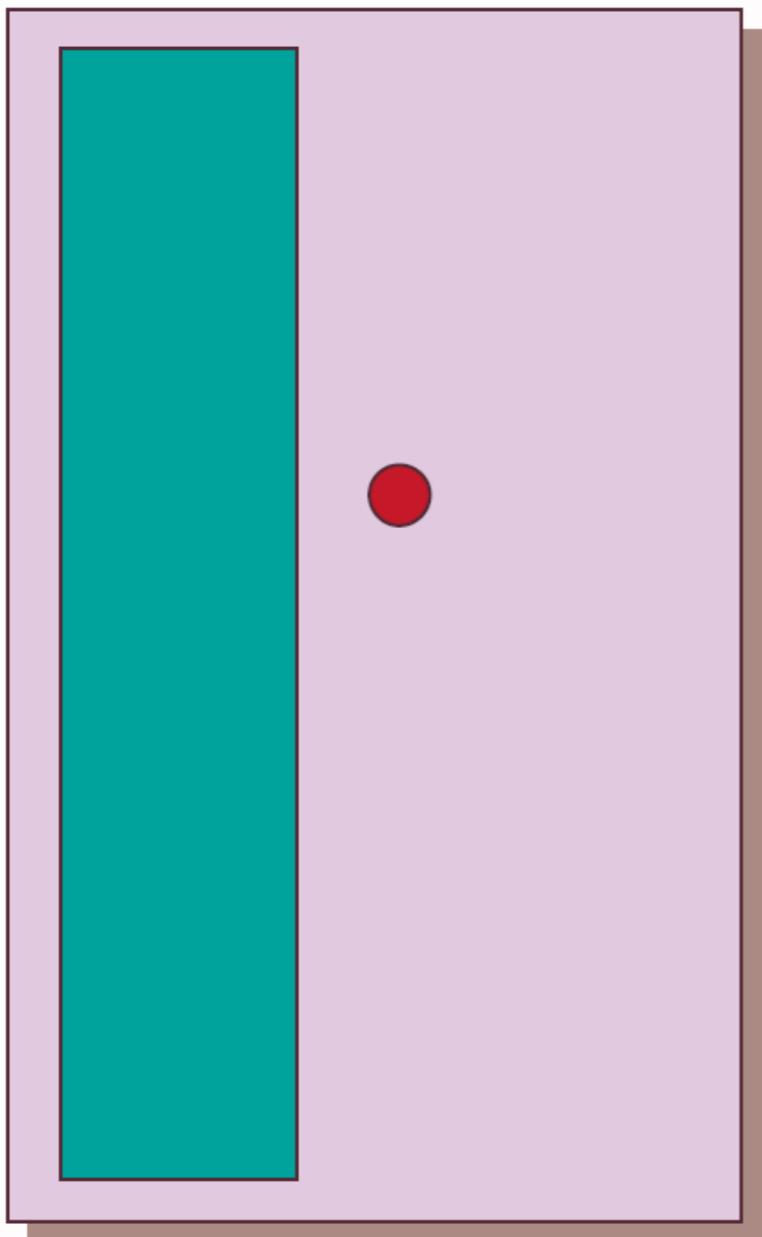
Differential equation solver

$$\begin{bmatrix} \dot{\mathbf{x}} \\ \dot{\mathbf{v}} \end{bmatrix} = \begin{bmatrix} \mathbf{v} \\ \mathbf{f} / m \end{bmatrix}$$

Euler method:

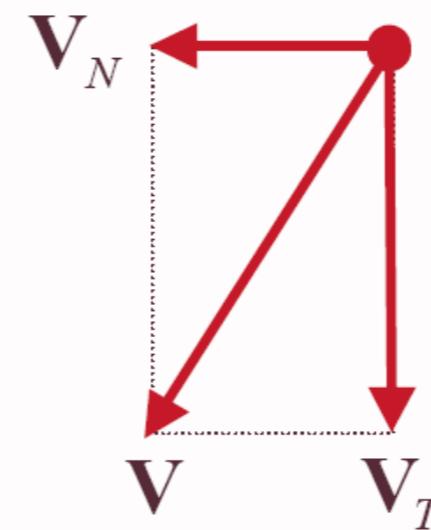
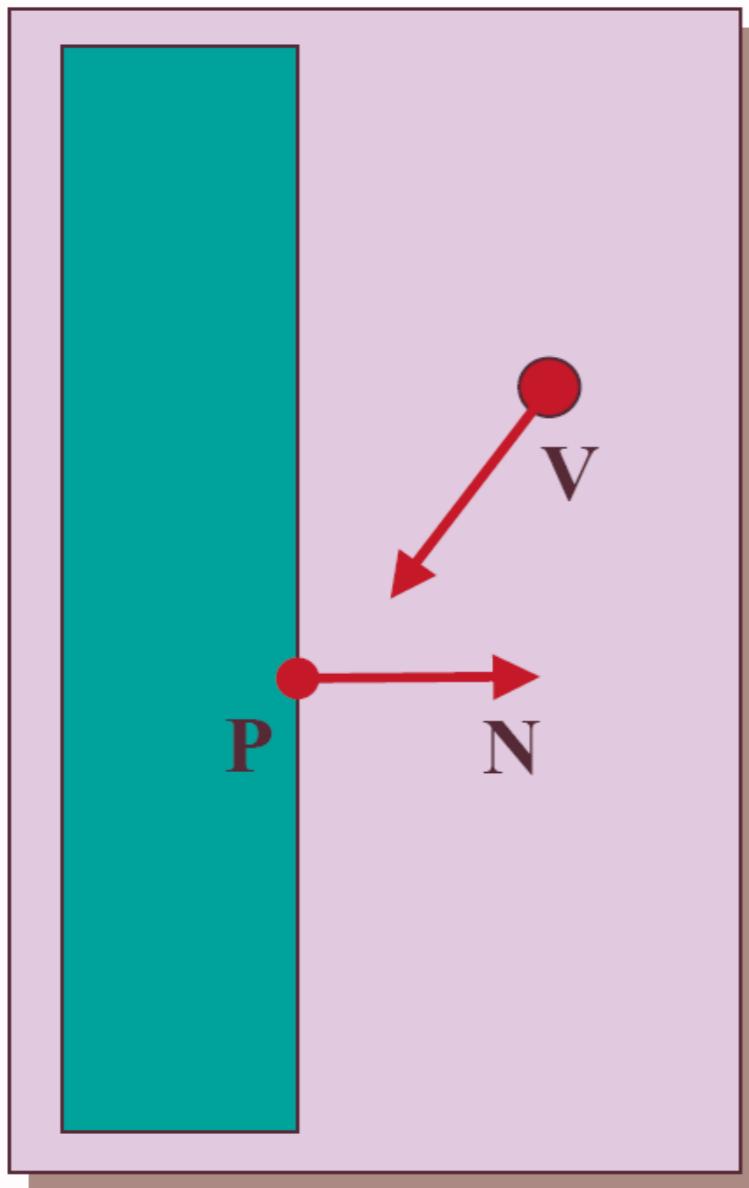
$$\begin{bmatrix} \mathbf{x}_1^{i+1} \\ \mathbf{v}_1^{i+1} \\ \vdots \\ \mathbf{x}_n^{i+1} \\ \mathbf{v}_n^{i+1} \end{bmatrix} = \begin{bmatrix} \mathbf{x}_1^i \\ \mathbf{v}_1^i \\ \vdots \\ \mathbf{x}_n^i \\ \mathbf{v}_n^i \end{bmatrix} + \square t \begin{bmatrix} \mathbf{v}_1^i \\ \mathbf{f}_1^i / m_1 \\ \vdots \\ \mathbf{v}_n^i \\ \mathbf{f}_n^i / m_n \end{bmatrix}$$

Bouncing off the walls



- Add-on for a particle simulator
- For now, just simple point-plane collisions

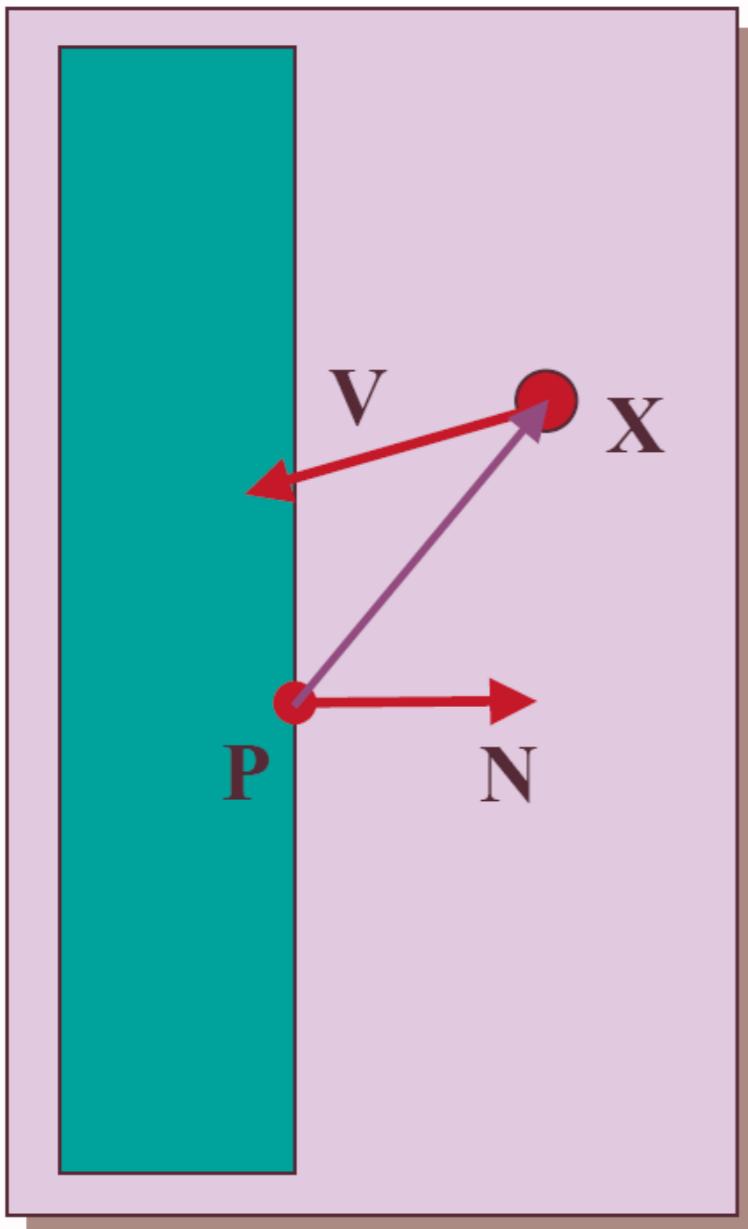
Normal and tangential components



$$\mathbf{V}_N = (\mathbf{N} \cdot \mathbf{V})\mathbf{N}$$

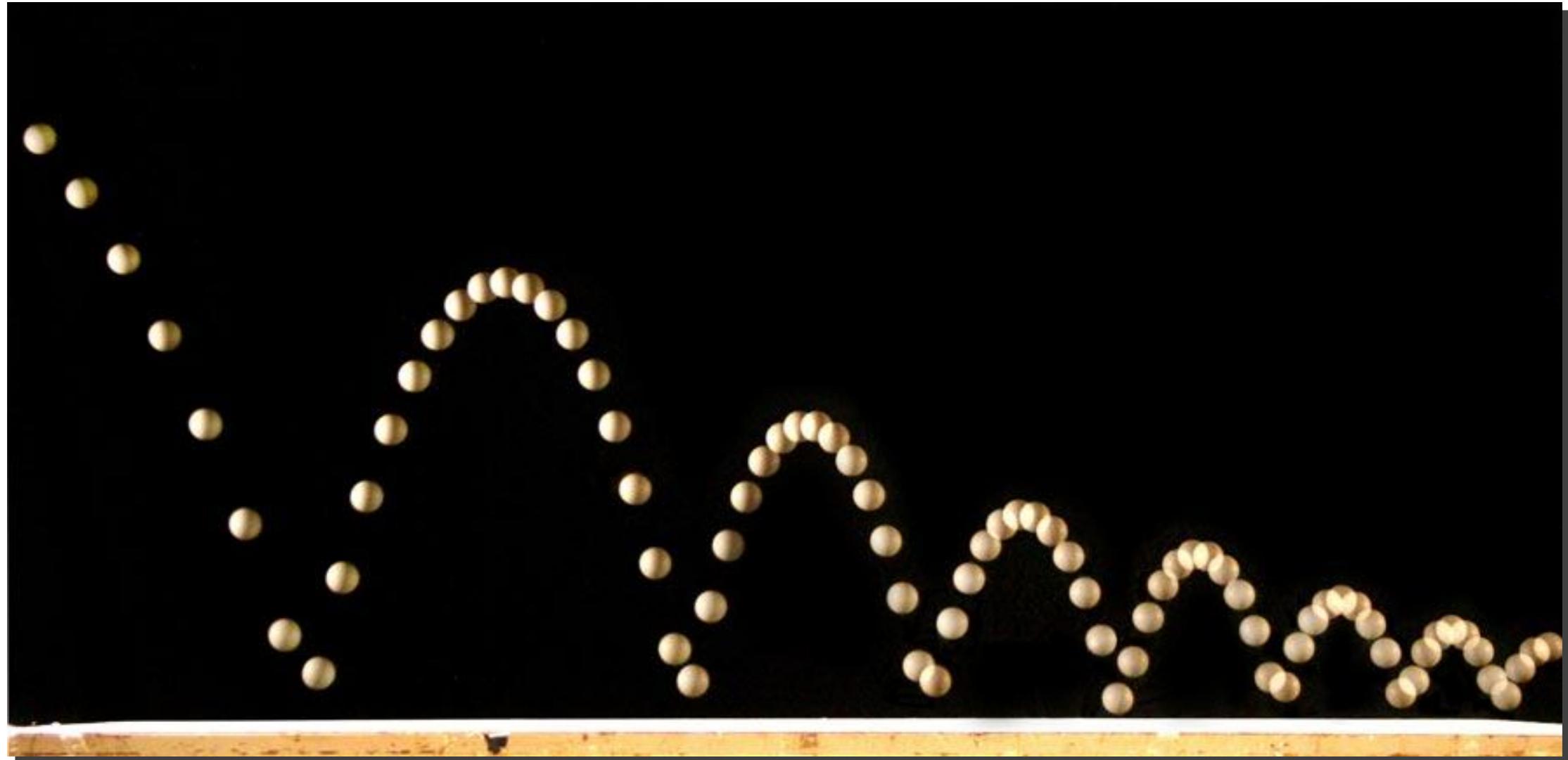
$$\mathbf{V}_T = \mathbf{V} - \mathbf{V}_N$$

Collision Detection



$(\mathbf{X} - \mathbf{P}) \cdot \mathbf{N} < \varepsilon$ Within ε of the wall
 $\mathbf{N} \cdot \mathbf{V} < 0$ Heading in

Collision Response



$$\mathbf{V}' = \mathbf{V}_T - k_r \mathbf{V}_N$$

Summary

- Physics of a particle system
- Various forces acting on a particle
- Combining particles into a particle system
- Euler method for solving differential equations

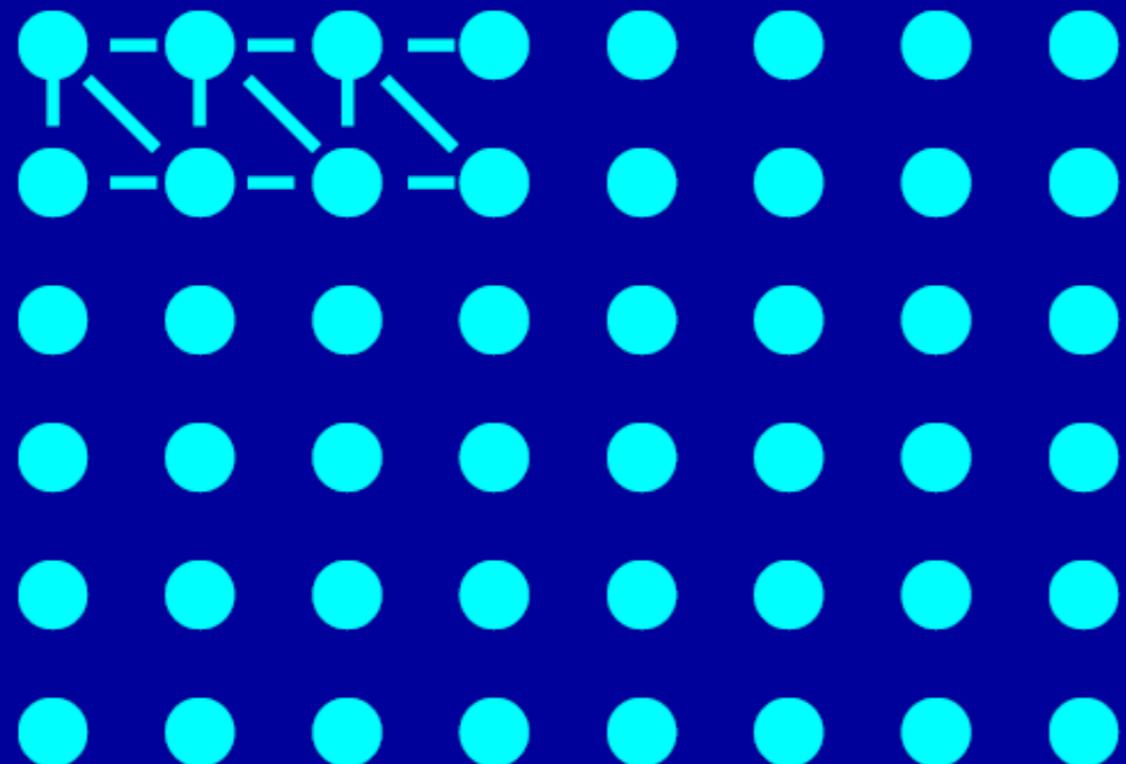
Example



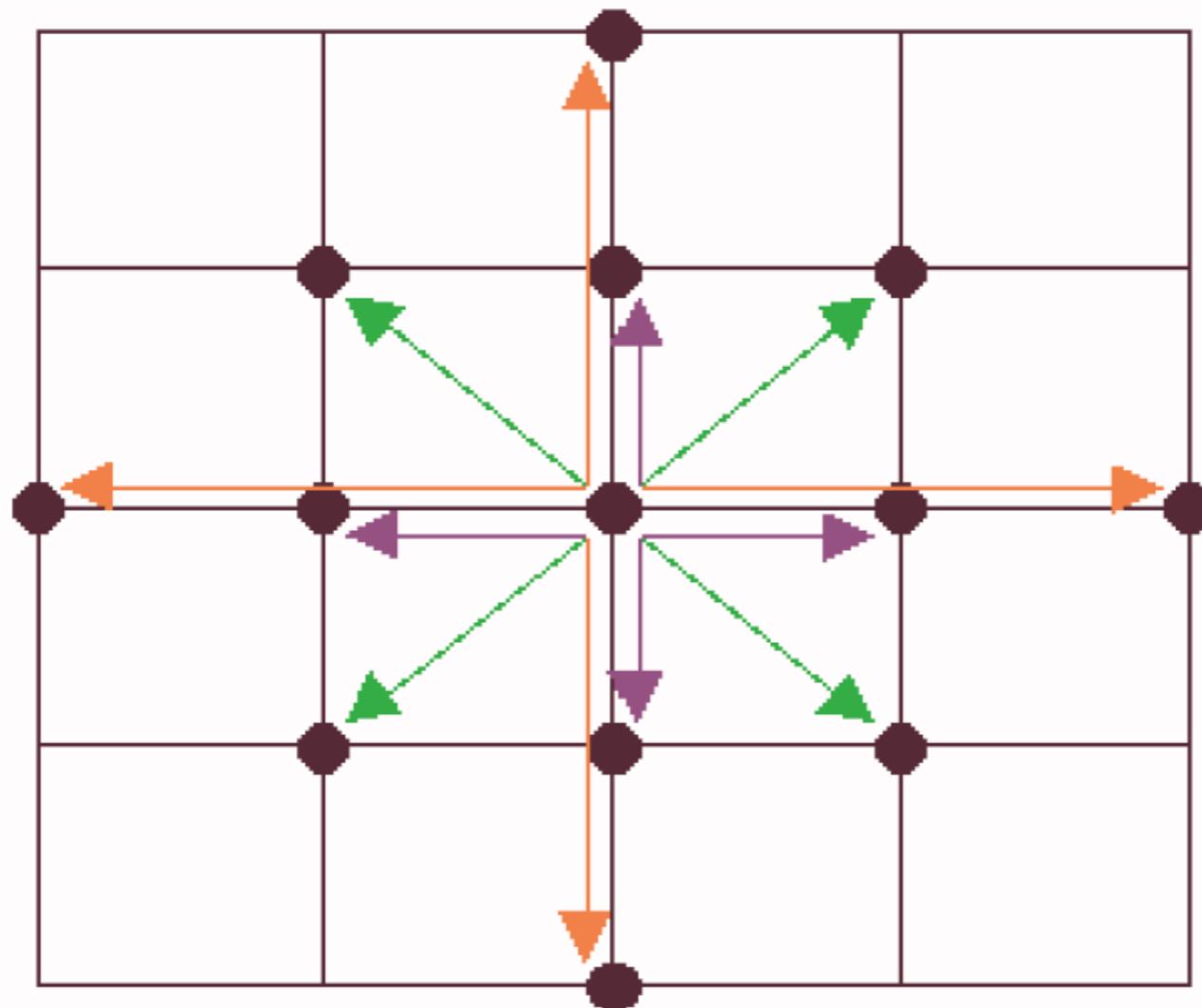
http://www.youtube.com/watch?v=3_fLO4xjTqg

Spring-Mass Systems

Cloth in 2D
Jello in 3D



Cloth Simulation



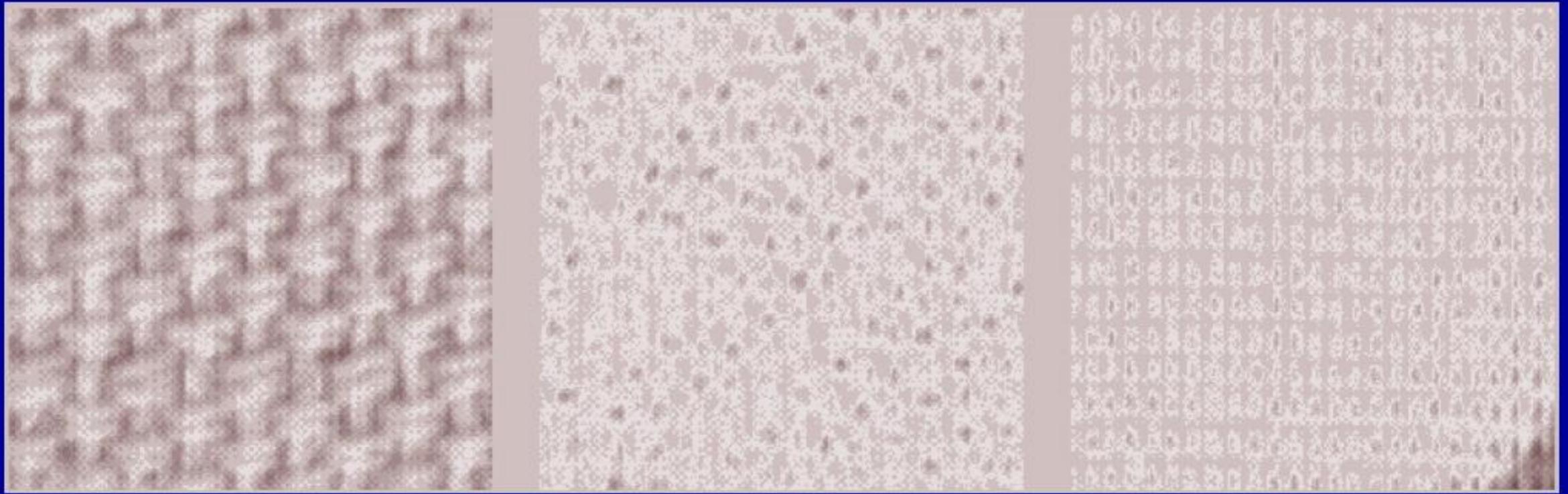
Cloth forces:

Blue (short horizontal & vertical) = stretch springs

Green (diagonal) = shear springs

Red (long horizontal & vertical) = bend springs

Cloth



Many types of cloth
Very different properties
Not a simple elastic surface
Woven fabrics tend to be very stiff
Anisotropic

Breen '95

Artificial Fish

