

# Simulation & Animation

## 5. Physically-based Animation

# Physically-based animation

**Goal:** compute the motion of objects based on the underlying laws of physics.

## Motivation:

### 1. Realistic environments

A realistic virtual environment requires not only accurate rendering, but also realistic physically-based animations

### 2. Interactive environments

Arbitrary interactions require dynamic real-time simulations (we cannot precompute all possible outcomes)

### 3. More productive animation pipeline

Artists only have to specify high-level characteristics (mass, forces, initial conditions)

# Overview

The key components of a physically-based animation system are:

- 1. ODE\* Solvers**
- 2. Particle dynamics**
- 3. Rigid-body dynamics**
- 4. Collision detection and response**

**\*Ordinary Differential Equation**

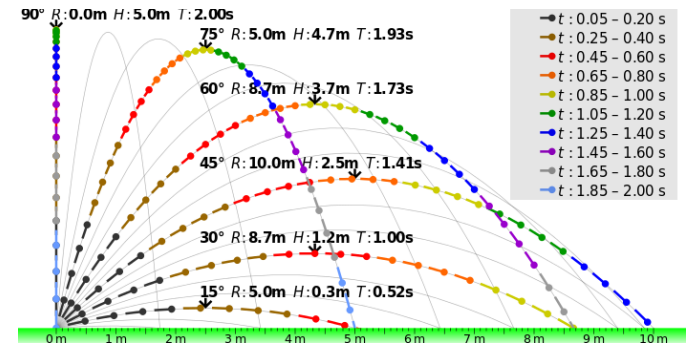
# 1. Ordinary Differential Equations

# Position Functions

- So far: position functions defining the motion of objects along smooth parametric curves (polynomials, splines)

- Simple physics-based example: parabolic motion of ballistic body

$$y = y_0 + x \tan \theta - \frac{gx^2}{2(v \cos \theta)^2}$$



- In many physical systems, an object's path is influenced based on complex interactions with the environment  
→ cannot be expressed by an analytic position function  $x(t)$ .
- However, we can define the **change of position  $x'(t)$**  depending on these interactions.  
→ also depends on the **current position  $x$**  itself

# Ordinary Differential Equations

**General form:**

$$\frac{d}{dt}x(t) = f(x, t) \quad \text{OR} \quad \dot{x}(t) = f(x, t)$$

Where:

- *$x$  is the state of a system, typically a vector that changes over time*
- *$f$  is a known function that we can evaluate*

In ***an initial value problem*** we are given the state of the system at beginning:  $\mathbf{x}(t_0) = \mathbf{x}_0$

# Equations of Motion

If  $\vec{x}(t)$  represents the position of an object, then we define:

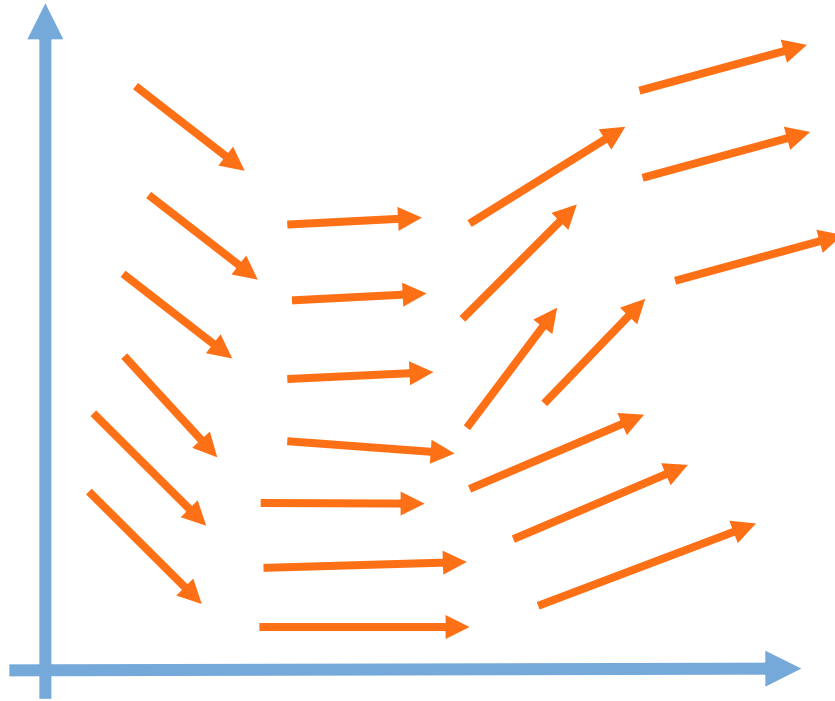
**Velocity:**  $\vec{v}(t) = \frac{d\vec{x}(t)}{dt}$  1<sup>st</sup> order ODE

**Acceleration:**  $\vec{a}(t) = \frac{d\vec{v}(t)}{dt} = \frac{d^2\vec{x}(t)}{dt^2}$  2<sup>nd</sup> order ODE

Our discussion here regards 1<sup>st</sup> order ODEs. We will see later how we can handle the 2<sup>nd</sup> order equation for the acceleration.

# Ordinary Differential Equations

$f(x, t)$  defines a vector field corresponding to the velocity of a moving point  $\mathbf{p}$  at every possible position  $\mathbf{x}$  and time  $\mathbf{t}$ .

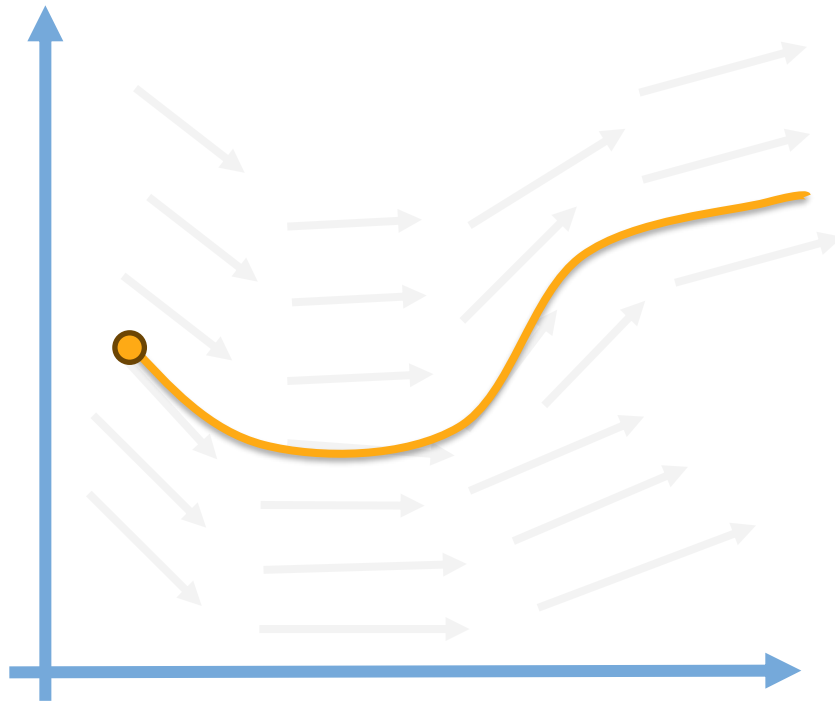


$$\frac{d}{dt}x(t) = f(x, t)$$



# Ordinary Differential Equations

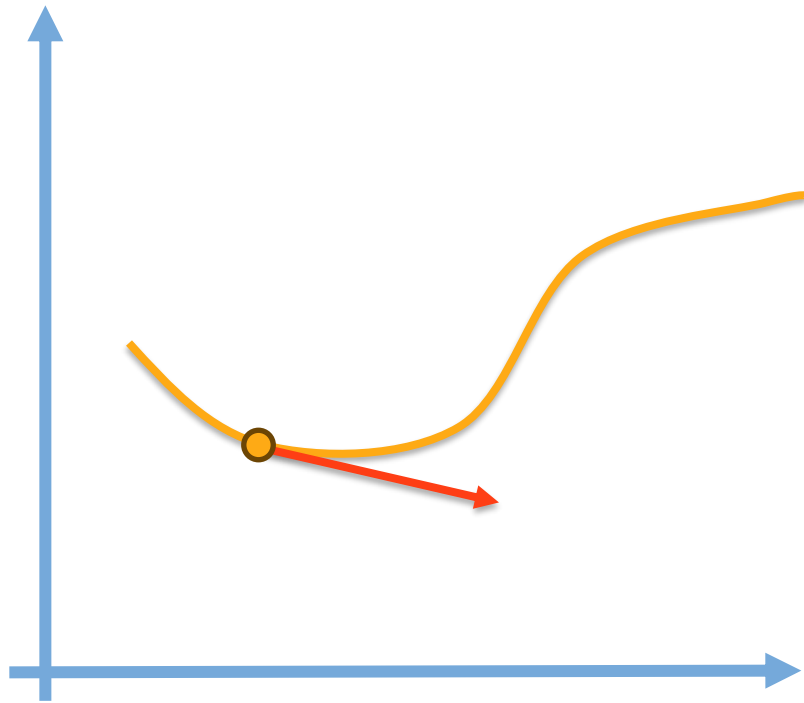
Starting at a point  $x_o$ , integrating  $x(t)$  sweeps out a curve that describes the motion of a point **p** in the plane.



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

# Ordinary Differential Equations

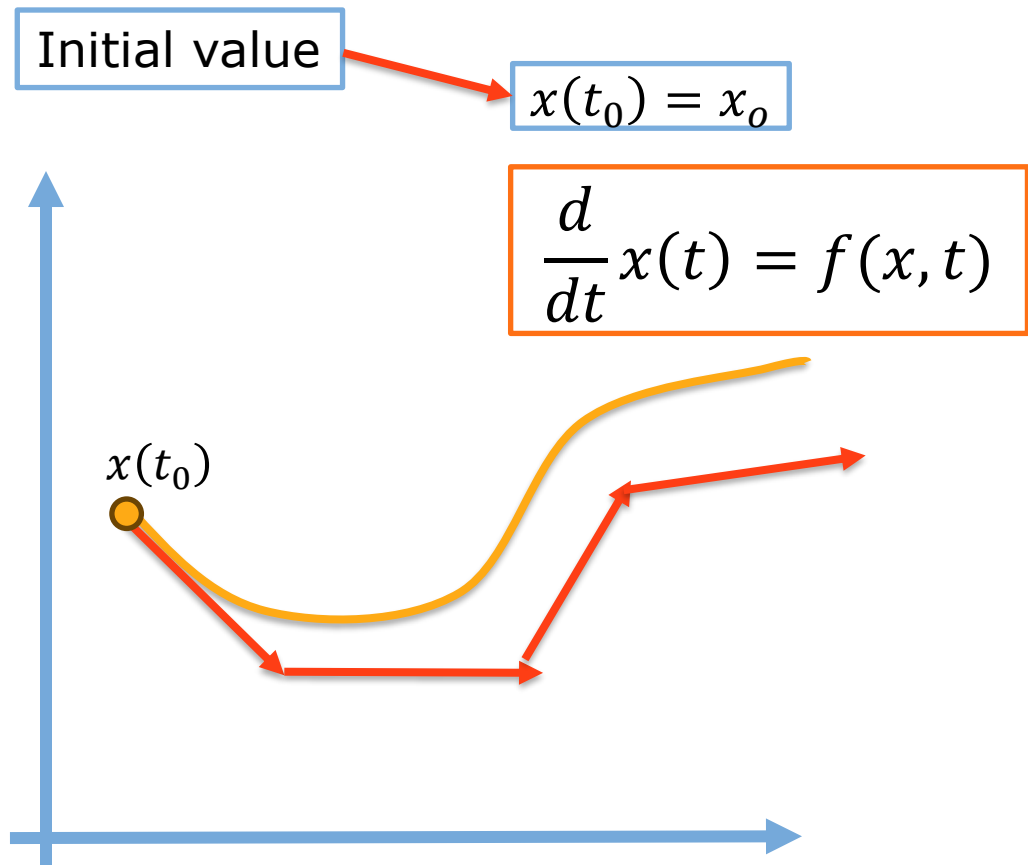
For a single position  $\mathbf{x}$  and time  $\mathbf{t}$ ,  $f(\mathbf{x}, \mathbf{t})$  defines the velocity of of point  $\mathbf{p}$  at that time, which is **tangent** to this curve.



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

# Initial value problem

Given a starting point, follow the trajectory by doing multiple **steps**.



# Taylor Series

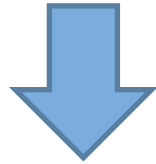
Assuming  $\mathbf{x}$  is smooth, we can express its value at the end of the step as an infinite sum:

$$\mathbf{x}(t_0 + h) = \mathbf{x}(t_0) + h\dot{\mathbf{x}}(t_0) + \frac{h^2}{2!}\ddot{\mathbf{x}}(t_0) + \frac{h^3}{3!}\dddot{\mathbf{x}}(t_0) + \dots + \frac{h^n}{n!}\frac{\partial^n \mathbf{x}}{\partial t^n} + \dots$$

# Taylor Series

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$$\mathbf{x}(t_0 + h) = \mathbf{x}(t_0) + h\dot{\mathbf{x}}(t_0) + \frac{h^2}{2!}\ddot{\mathbf{x}}(t_0) + \frac{h^3}{3!}\ddot{\mathbf{x}}(t_0) + \dots + \frac{h^n}{n!}\frac{\partial^n \mathbf{x}}{\partial t^n} + \dots$$



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

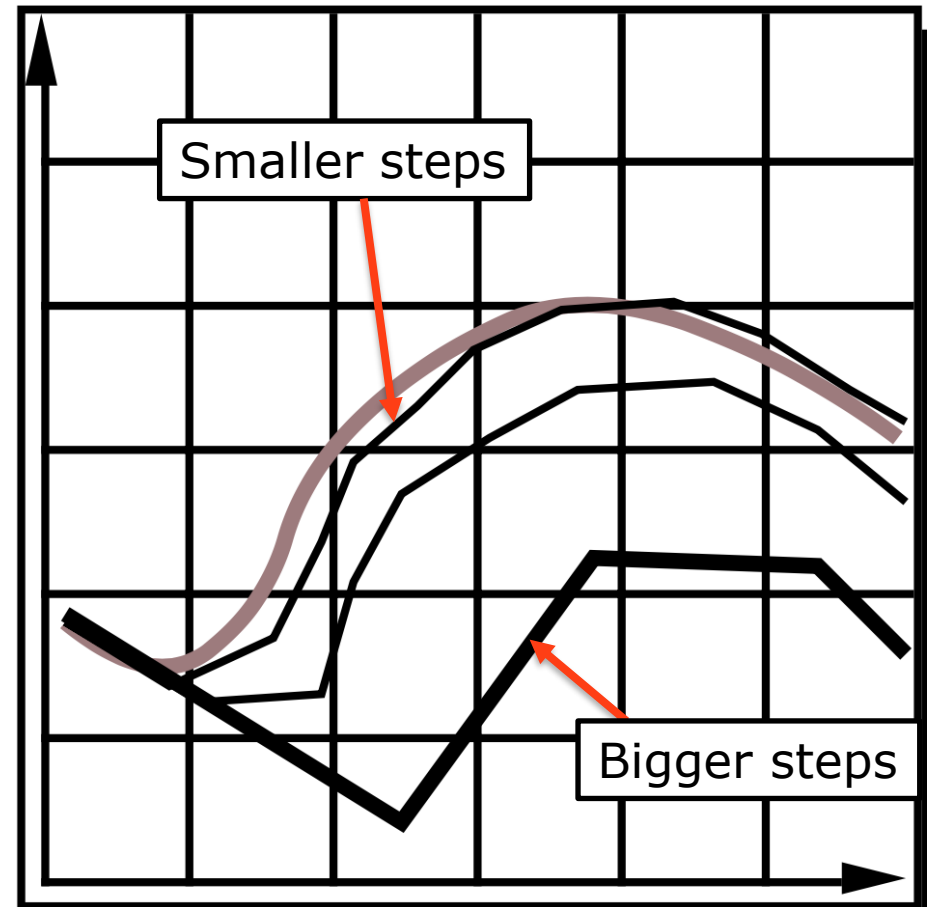
If we **truncate** the Taylor series by assuming that all derivatives except the first one are zero, we get **Euler's method**.

# Euler's Method

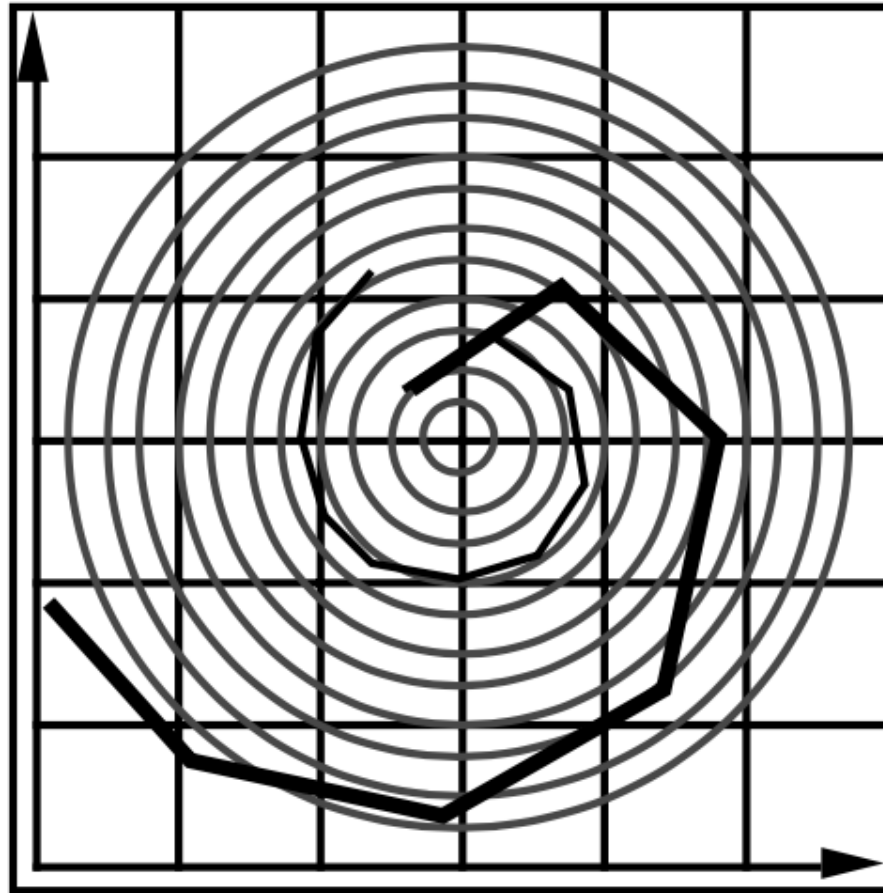
$$\mathbf{x}(t_0 + h) = \mathbf{x}(t_0) + h\dot{\mathbf{x}}(t_0)$$

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

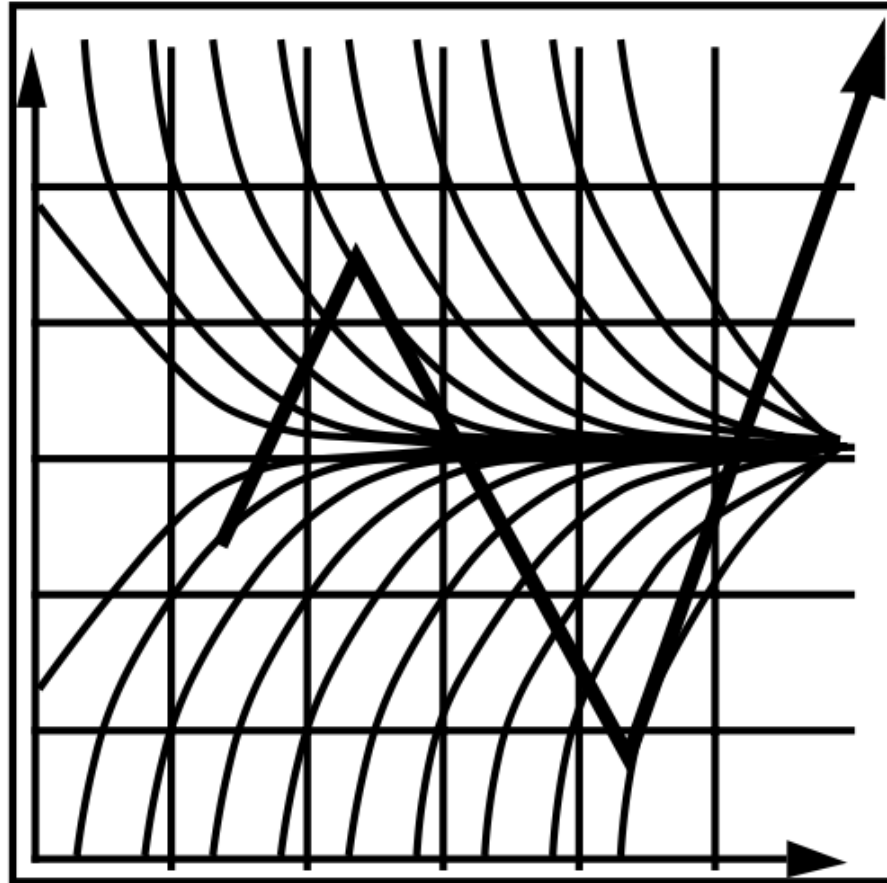
**Only correct when  $\mathbf{x}$  is linear**, otherwise error is introduced!



# Problems: Inaccuracy

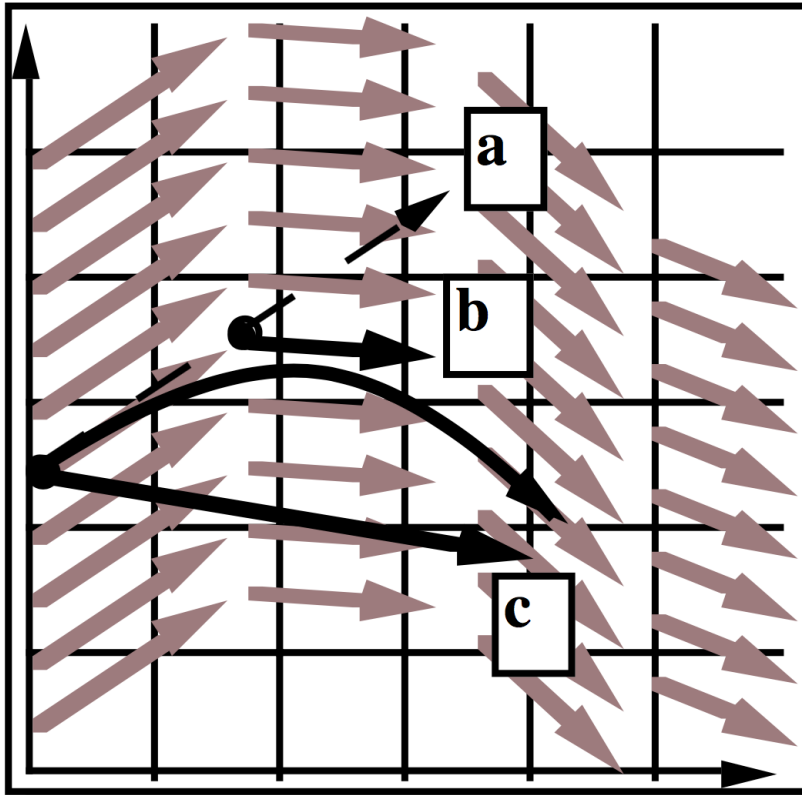


# Problems: Instability





# Midpoint Method



a. Evaluate  $f$  at the initial point:

$$f(\mathbf{x}_0)$$

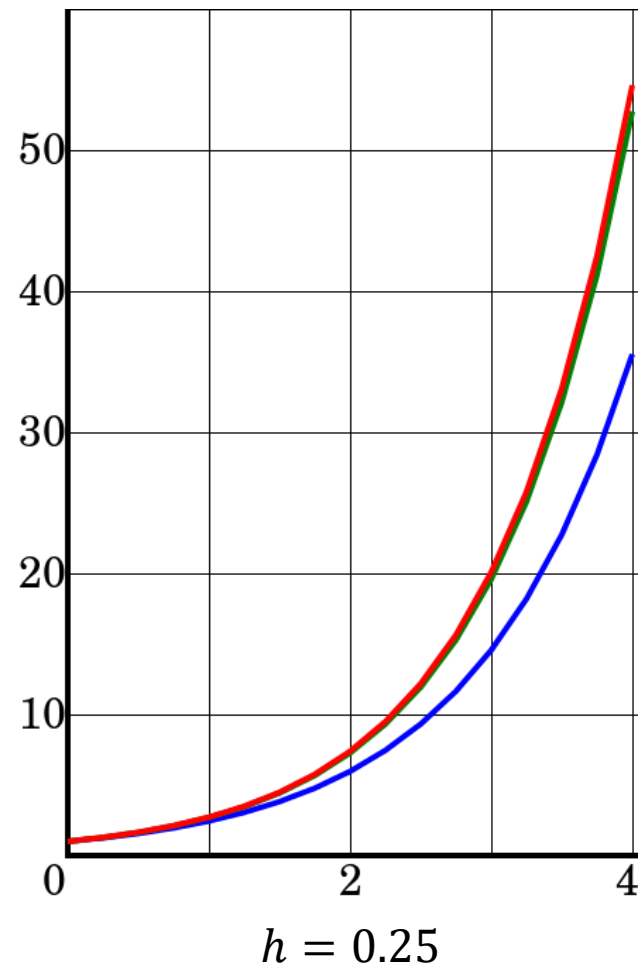
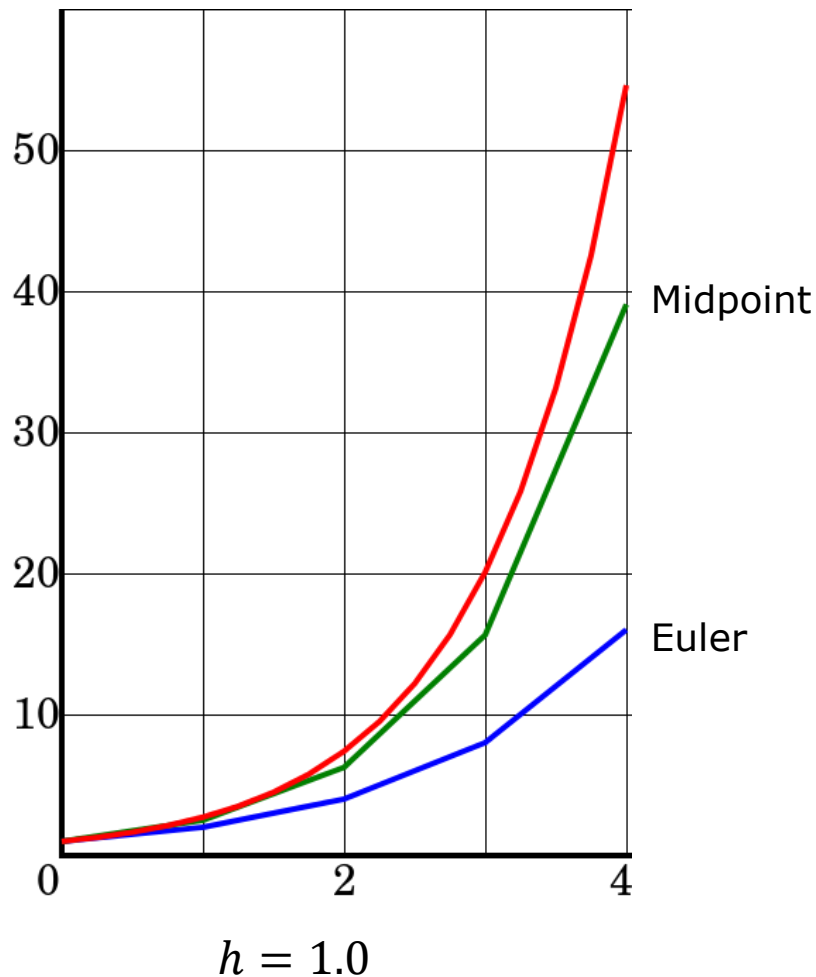
b. Evaluate  $f$  at the midpoint:

$$f\left(\mathbf{x}_0 + \frac{h}{2}f(\mathbf{x}_0)\right)$$

c. Take a step using the midpoint value:

$$\mathbf{x}(t_0 + h) = \mathbf{x}(t_0) + h\left(f(\mathbf{x}_0) + \frac{h}{2}f(\mathbf{x}_0)\right).$$

# Euler vs. Midpoint



# Higher-order methods

- Euler's method performs one function evaluation
- Midpoint performs two evaluations
- 4<sup>th</sup>-order **Runge-Kutta** performs four function evaluations:

$$k_1 = hf(\mathbf{x}_0, t_0)$$

$$k_2 = hf\left(\mathbf{x}_0 + \frac{k_1}{2}, t_0 + \frac{h}{2}\right)$$

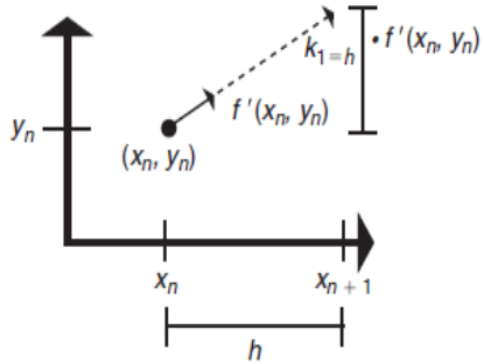
$$k_3 = hf\left(\mathbf{x}_0 + \frac{k_2}{2}, t_0 + \frac{h}{2}\right)$$

$$k_4 = hf(\mathbf{x}_0 + k_3, t_0 + h)$$

$$\mathbf{x}(t_0 + h) = \mathbf{x}_0 + \frac{1}{6}k_1 + \frac{1}{3}k_2 + \frac{1}{3}k_3 + \frac{1}{6}k_4.$$

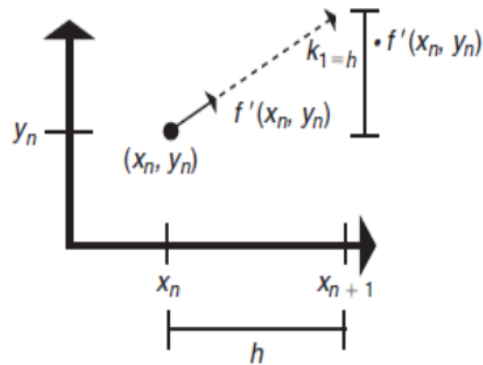
Higher accuracy, but also higher complexity

# 4<sup>th</sup>-order RK steps

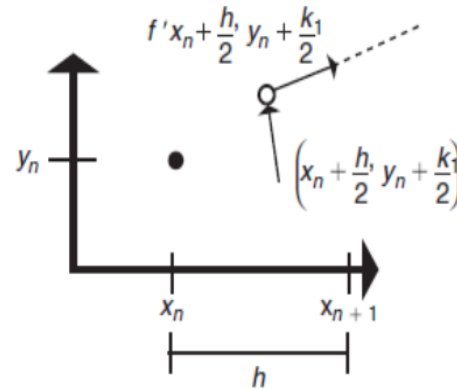


- A Compute the derivative at the beginning of the interval

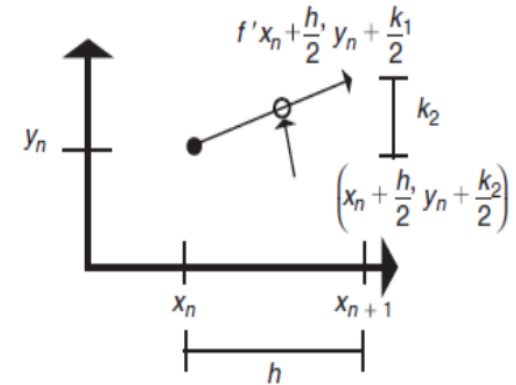
# 4<sup>th</sup>-order RK steps



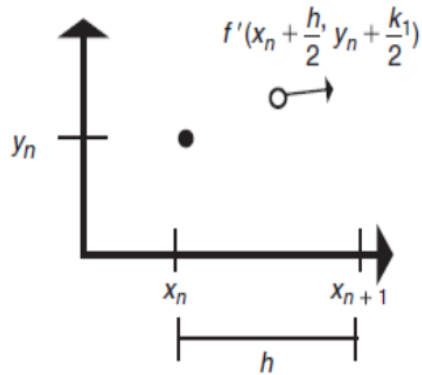
A Compute the derivative at the beginning of the interval



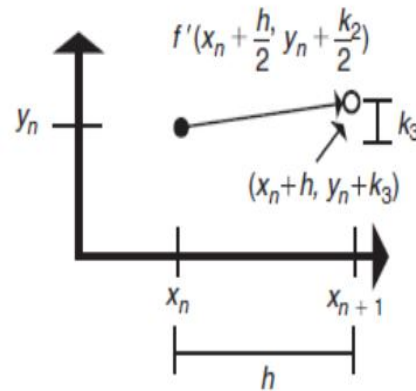
B Step to midpoint (using derivative previously computed) and compute derivative



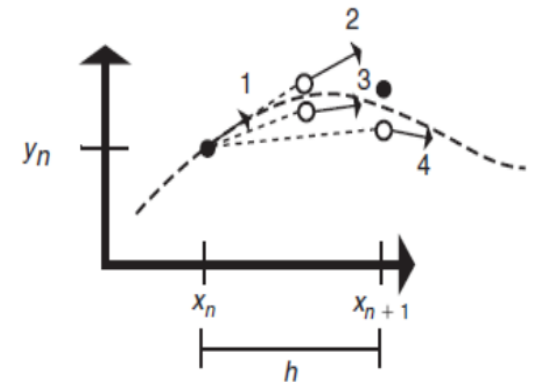
C Step to new midpoint from initial point using midpoint's derivative just computed



D Compute the derivative at the new midpoint



E Use new midpoint's derivative and step from initial point to end of interval



F Compute derivative at end of interval and average with 3 previous derivatives to step from initial point to next function value

# Adaptive step size

- Large step sizes: better performance but lower accuracy
- Small step sizes: Better accuracy but lower performance



Ideally we want the largest possible step size that does not introduce an unreasonable amount of error.

Determining such a good step size can be a problem, no matter which underlying ODE solver we are using.

Adaptive methods vary the step size over the course of solving an ODE: smaller steps are used in non-linear (curvy) segments.

# Adaptive Euler method

- Compute two estimates for  $x(t_0 + h)$
- For the first estimate  $x_a$ , use one step of size  $h$
- For the 2nd estimate  $x_b$ , use two steps of size  $h/2$
- The current step size  $h$  is adjusted based on the error value:

$$e = |x_a - x_b|$$

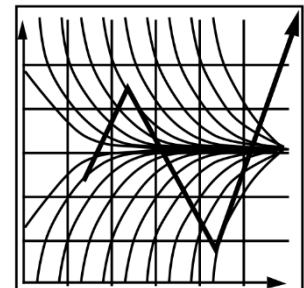
For linear segments,  $e$  will be close to zero, and larger step sizes can be used.

# Step Sizes in Interactive Simulations

- Interactive simulations, such as video games, render the world at a specific frame rate
  - Typically locked to the monitor refresh rate (vsync)
  - But can often be lower, due to limited performance.
- A naïve approach is to perform the physics simulation at the same rate as the visual refresh
  - The step size of the simulation is the interval between two subsequent frames displayed on the screen.



**Problem:** Sudden drops in the framerate can result in unstable physics simulation!!!!





# Step sizes in interactive simulations

- The update rate of the rendering and physics simulation should be decoupled
- Video games often sample the input and update the physics at a fixed rate, which is higher than the display refresh rate.

# General Solver Interface

In a C-like language, an ODE solver will typically have this interface:

```
typedef void (*dydt_func)(double t, double y[], double ydot[]);  
  
void      ode(double y0[], double yend[], int len, double t0,  
              double t1, dydt_func dydt);
```

Abstracts the underlying implementation, could be Euler's method, midpoint or RK.

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```

**Input:** Initial state at time t0

**Output:** End state at time t1

Abstracts the underlying implementation, could be Euler's method, midpoint or RK.

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```

Helper function to compute the derivatives of the state  
(*function pointer in ANSI C terms*)

Abstracts the underlying implementation, could be Euler's method, midpoint or RK.

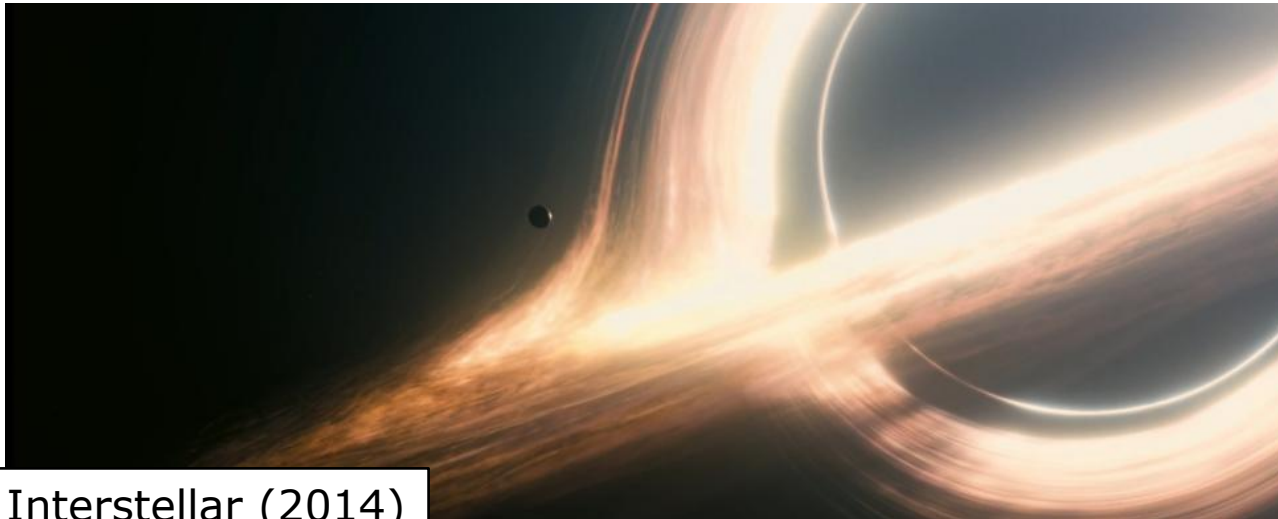
## 2. Particle Dynamics

# Newtonian Physics

- **First law:** an object either remains at rest or continues to move at a constant velocity
- **Second law:** the sum of the forces  $F$  on an object is equal to its mass  $m$  multiplied by the acceleration  $a$  of the object:  **$F = m \cdot a$**   
(accurate for a **particle of mass**, integration is required for arbitrary mass distributions/solid objects)
- **Third law:** When one body exerts a force on a second body, the second body simultaneously exerts a force equal in magnitude and opposite in direction on the first body.

# Newtonian Physics

- Empirical laws, based on observation
- Not accurate for objects that move very fast or have a very large mass
- Good enough for most simulations (unless the simulation involves black holes...)



Interstellar (2014)

# Newtonian Particle

- The motion of a **particle of mass** is governed by this differential equation:

$$\frac{d^2x(t)}{dt^2} = \frac{F(x, t)}{m}$$

where the total force  $F(x, t)$  can change depending on the position of the particle and the time.

- This equation has a second-order derivative and differs from the equations that we have seen in the previous slides about ODEs.




# Phase Space

- To handle the second-order ODE, we convert it to a first-order one by introducing extra variables

$$\frac{dx(t)}{dt} = v$$

$$\frac{dv(t)}{dt} = F/m$$



Coupled first-order ODEs  
(our solvers work for these!)

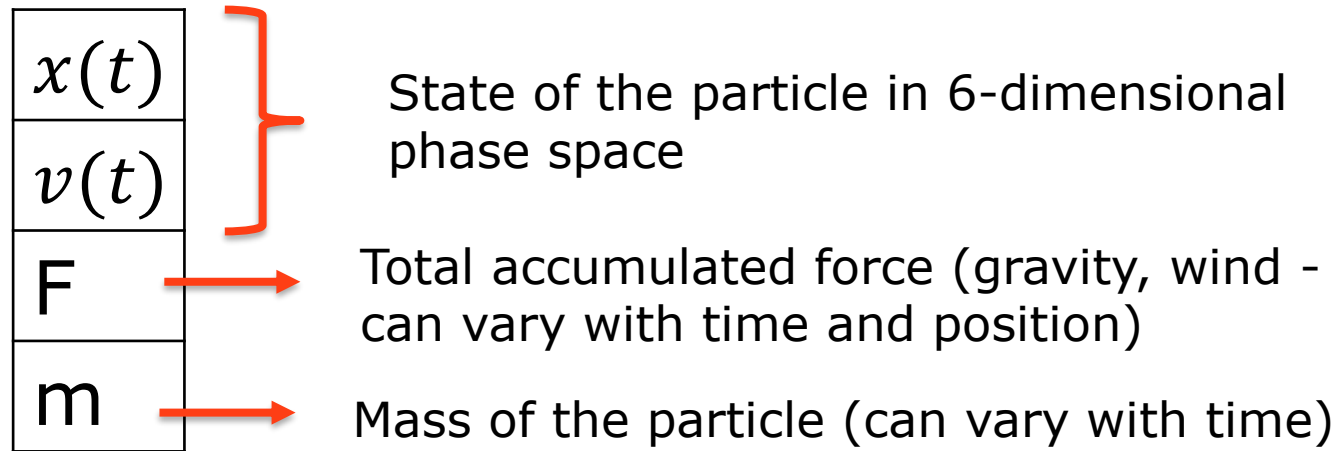
# Phase Space

We concatenate the 3D position and 3D velocity vectors to make a new 6D vector that denotes the state of the particle in ***phase-space***.

$$\begin{bmatrix} x(t) \\ v(t) \end{bmatrix} \quad \text{State in 6D phase space}$$

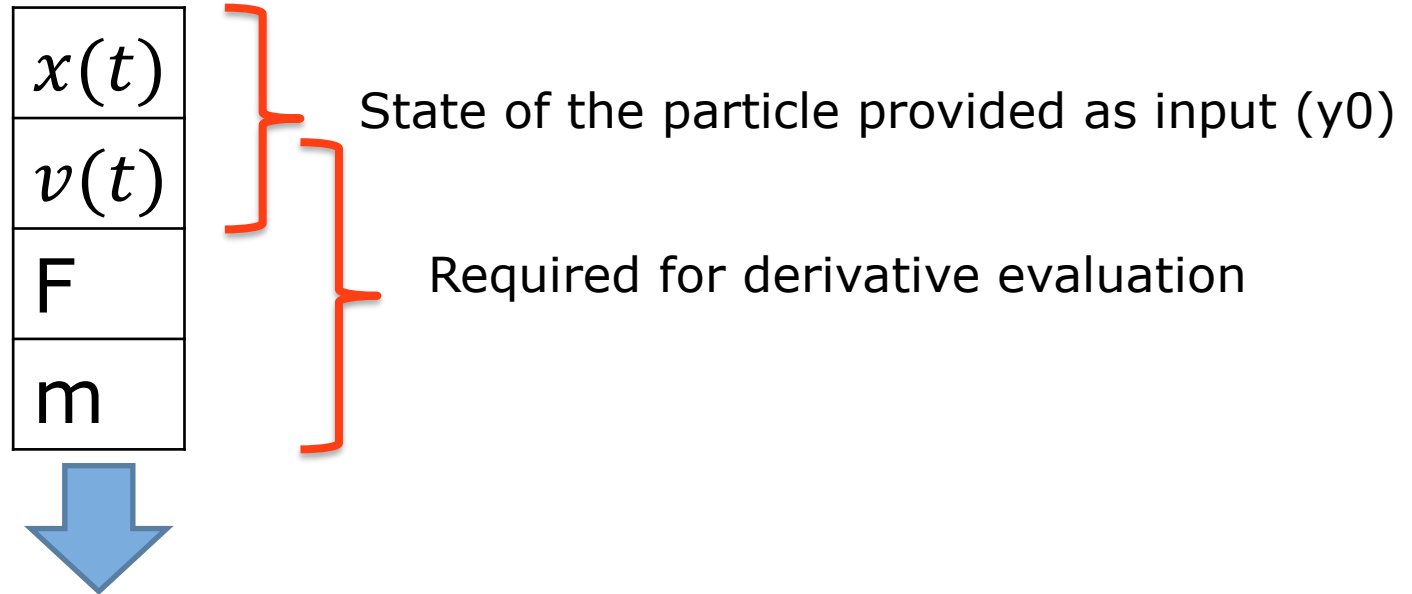
$$\frac{d}{dt} \begin{bmatrix} x(t) \\ v(t) \end{bmatrix} = \begin{bmatrix} v(t) \\ F/m \end{bmatrix} \quad \text{A standard 1<sup>st</sup> order ODE}$$

# Particle Data Structure



# Particle Dynamics

Given the state of a particle at time  $t_0$ , the ODE solver will compute the state at time  $t_1$ .



```
void ode(double y0[], double yend[], int len, double t0,  
        double t1, dydt_func dydt);
```

For a single particle,  $\text{len} = 6$

# Particle Systems

Given the state of  $N$  particles at time  $t_0$ , the ODE solver will compute the state at time  $t_1$ .

$x_0(t)$	$x_1(t)$	...	$x_N(t)$
$v_0(t)$	$v_1(t)$	...	$v_N(t)$
$F_0$	$F_1$	...	$F_N$
$m_0$	$m_1$	...	$m_N$

State of the particle provided as input ( $y_0$ )

Required for derivative evaluation



```
void ode(double y0[], double yend[], int len, double t0,  
        double t1, dydt_func dydt);
```

For  $N$  particles, **len = 6N**

# Derivative Evaluation

## 1. Zero forces

- Loop over all particles and zero the accumulators

## 2. Accumulate forces

- For each particle sum all forces

## 3. Construct the derivative vector

- For each particle copy velocity  $\mathbf{v}$  and  $\mathbf{F}/\mathbf{m}$  into the derivative vector

# Forces

- **Constant:** the thrust of a rocket engine
- **Position dependent:** gravity, magnetic fields, other force fields
- **Velocity dependent:** drag
- **n-ary:** particles connected with springs

The forces should be recomputed and accumulated on each solver step.

# Viscous Drag

Viscous force opposing the direction of motion of an object in a medium, with strength proportional to the speed (velocity magnitude):

$$F_{drag} = -K_{drag} v$$

where  $K_{drag}$  depends on density of the medium (zero for vacuum, higher for viscous fluids)



# Gravity

The magnitude of the gravitational force from earth on an object with mass  $m$  at height  $r$  is given by:

$$F_{earth} = mg, \quad g = -G \frac{m_{earth}}{r^2}$$

If the height of the object does not vary a lot during the motion, then  $g$  can be considered a constant.

# Special Cases – Free Falling Particle

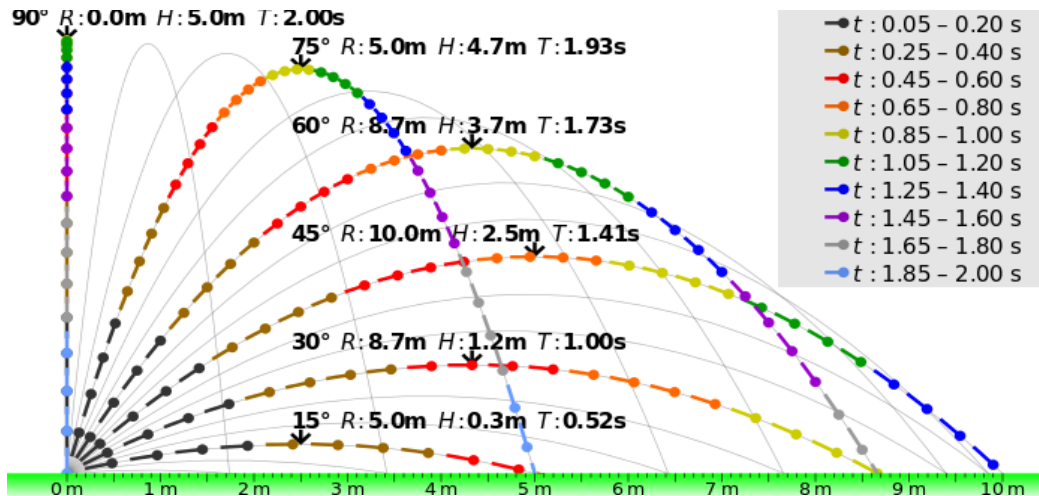
- If we assume only a constant gravitational force  $F = g \cdot m$  acts on the particle, then the position  $x(t)$  is given by

$$x(t) = x(t_o) + v_o t + 0.5 g t^2$$

Where  $g$  is the acceleration from the gravitational force, assumed here constant.

Closed-form solution, but not 100% accurate, as the gravitational force depends on the height of the particle.  
For accurate results the ODE should be solved.

# Special Cases – Ballistics Trajectories



Closed-form formula:

$$y = y_0 + x \tan \theta - \frac{gx^2}{2(v \cos \theta)^2}.$$

Trajectories of projectiles launched at different elevation angles at the same initial speed in a **vacuum (no drag)** and **uniform downward gravity field**.

In the general case the gravitational field is not constant, various additional time-varying forces act upon the projectile and the distribution of mass in the projectile is not symmetric.

For accurate results in such cases, we need to properly solve the ODE.

# Spring Forces

- Applied in n-ary “mass-spring” systems
- Guided by Hooke’s Law:

$$F_{spring} = -k \left( \|x_i - x_j\| - r_{ij} \right) \frac{x_i - x_j}{\|x_i - x_j\|}$$

Restoring Force vector of a spring

- Total potential energy:
  - $x_i$  ..... Position of the i-th mass vertex
  - $r_{ij}$  ..... Rest length of spring
  - $k$  ... .. Stiffness factor

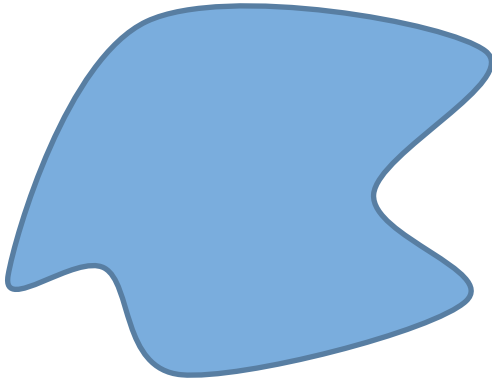
# Velocity Verlet Integration

- Specific flavour of class Verlet methods
- 2nd-order approximation of the Taylor series
- Store acceleration  $a(t)$  in the state vector
- Current state:  $x(t)$ ,  $v(t)$ ,  $a(t)$
- 1.  $\vec{x}(t + \Delta t) = \vec{x}(t) + \vec{v}(t) \Delta t + \frac{1}{2} \vec{a}(t) \Delta t^2$
- 2. Compute  $\vec{a}(t + \Delta t)$  from forces at  $\vec{x}(t + \Delta t)$
- 3.  $\vec{v}(t + \Delta t) = \vec{v}(t) + \frac{1}{2} (\vec{a}(t) + \vec{a}(t + \Delta t)) \Delta t$
- New state:  $x(t+\Delta t)$ ,  $v(t+\Delta t)$ ,  $a(t+\Delta t)$

# 3. Rigid Body Dynamics

# Rigid Bodies

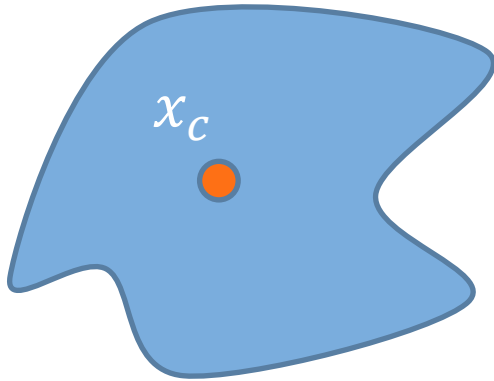
In the general case, an object (body of mass) has a non-uniform mass distribution.



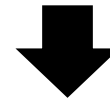
In our physically-based simulations we will assume objects are **rigid**: they can be only rotated and translated – cannot be deformed.

# Center of mass

If our object has a mass distribution with density  $\rho(x)$  within a solid  $Q$ , then we define as center of mass the point  $x_c$  that satisfies the following equation:



$$\frac{1}{M} \int_{x \text{ in } Q} \rho(x) (x - x_c) dV = 0$$



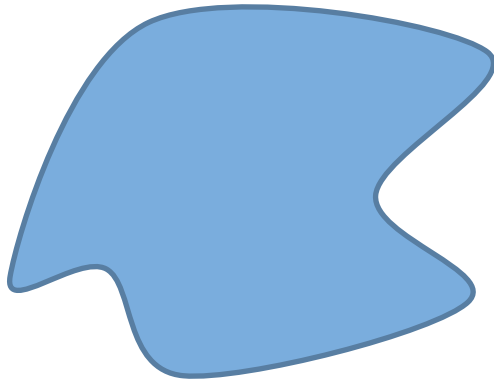
$$x_c = \frac{1}{M} \int_{x \text{ in } Q} \rho(x) x dV$$

When we refer to the position  $\mathbf{x}(\mathbf{t})$  of a rigid object, we will refer to the position (coordinates in world space) of its **center of mass**.



# Rigid Bodies

A rigid body, aside from **position**, also has **orientation**.



$q_1(t)$



$q_2(t)$

The orientation of an object is represented by a **quaternion**  $q(t)$ .  
While other representations are possible, quaternions are preferable. → see L2

# Animation state

**Particle:**

$$\begin{bmatrix} x(t) \\ v(t) \end{bmatrix}$$

**Rigid body:**

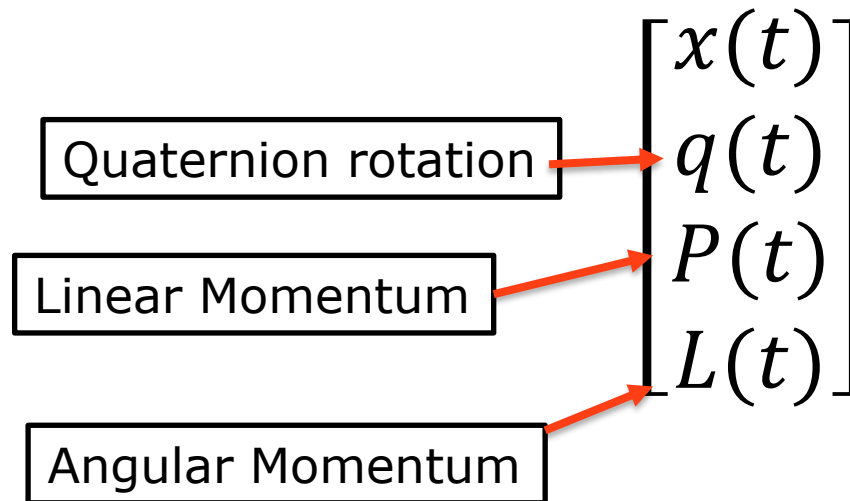
$$\begin{bmatrix} x(t) \\ q(t) \\ ? \\ ? \end{bmatrix}$$

# Animation state

**Particle:**

$$\begin{bmatrix} x(t) \\ v(t) \end{bmatrix}$$

**Rigid body:**



# Linear Momentum

**Formula:**  $P(t) = m v(t)$

**Derivative:**  $\frac{d}{dt} P(t) = F(t)$  (from Newton's second law)

If a closed system is not affected by external forces, its total linear momentum cannot change.

## Example:

A heavy truck moving rapidly has a large momentum, and it takes a large or prolonged force to get the truck up to this speed, and would take a similarly large or prolonged force to bring it to a stop.

# Torque

Just as a linear force pushes or pulls objects, torque can be thought of as a force twisting/spinning objects.

## Net torque formula:

$$\tau(t) = \sum (p_i - x_c(t)) \times f_i$$

### Remark 2:

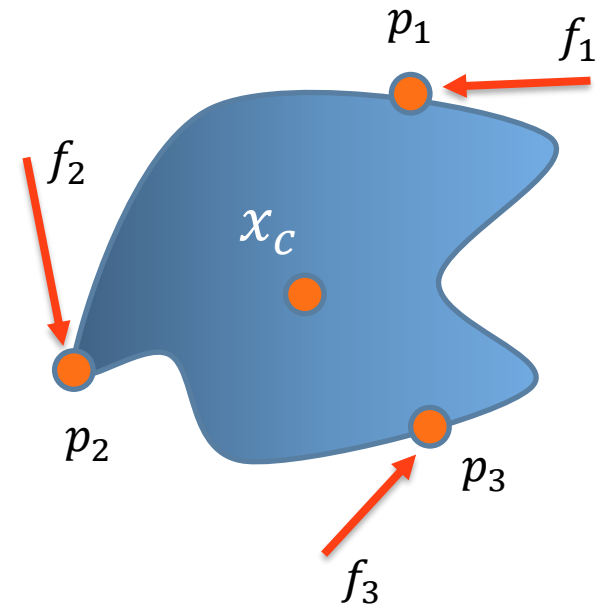
It's easier to spin objects when applying the force at a larger distance to the center of mass (larger lever)

### Remark 3:

It's easier to spin objects when applying the force orthogonally

### Remark 1:

Larger force magnitude results in larger torque



Torque will result in a rotational motion, so we need to define the speed of rotation...

# Angular Velocity

In 2D, the **scalar** rate of change of angular position of a rotating body.

**Formula:**  $\omega(t) = \frac{d\varphi(t)}{dt}$  (in 2 dimensions)

In 3D, we use a **vector**  $\vec{\omega}(t)$ , encoding both the (unit) rotation axis  $\vec{u}$  and the speed of the spin (rate of change of angular position  $\varphi(t)$  in the plane defined by  $\vec{u}$ ). *See axis-angle representation → L2*

**Formula:**  $\vec{\omega}(t) = \frac{d\varphi(t)}{dt} \vec{u}$  (in 3 dimensions)

# Angular Momentum

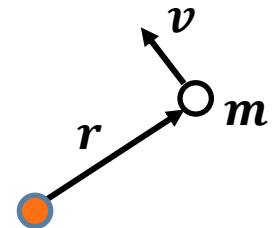
For simple **particles of mass** (ball attached to a string, satellite orbiting the earth):

**Formula:**  $\vec{L}(t) = \vec{r} \times \vec{P}(t)$

radius vec  $\times$  linear momentum

$$= \vec{r} \times \vec{v}(t) m$$

linear velocity in t



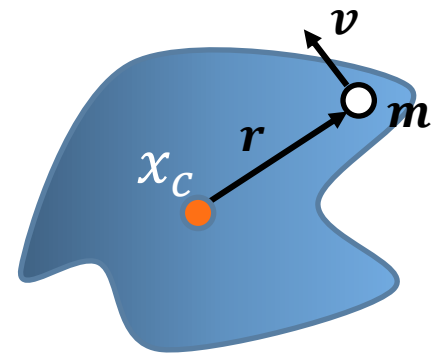
# Angular Momentum

For simple **particles of mass** (ball attached to a string, satellite orbiting the earth):

**Formula:**  $\vec{L}(t) = \vec{r} \times \vec{P}(t)$  radius vec × linear momentum!

$= \vec{r} \times \vec{v}(t) m$  linear velocity in t

For general rigid bodies, we need to consider a **non-singular volume** and **non-uniform mass distribution** ...





# Angular Momentum

For simple **particles of mass** (ball attached to a string, satellite orbiting the earth):

$$\vec{L}(t) = \vec{r} \times \vec{v}(t) m$$

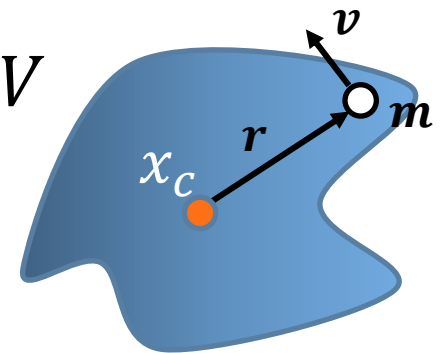
Relationship  
Linear  $\leftrightarrow$  Angular Velocity

$$\vec{v} = \vec{\omega} \times \vec{r}$$

For **general rigid objects**:

$$\begin{aligned} \vec{L}(t) &= \int_{x \in V} \vec{r}(x) \times (\vec{\omega}(t) \times \vec{r}(x)) \rho(x) dV \\ &= I(t) \vec{\omega}(t) \end{aligned}$$

Infinitesimal mass  $\rightarrow$  density



**Inertia tensor**, encodes the mass distribution of the object.

# Inertia Tensor

$$I(t) = \begin{bmatrix} I_{xx} & I_{xy} & I_{xz} \\ I_{yx} & I_{yy} & I_{yz} \\ I_{zx} & I_{zy} & I_{zz} \end{bmatrix}$$

The inertia tensor is the only thing we need to describe how an object with an arbitrary distribution of mass responds to forces!

Diagonal terms:

$$I_{xx} = \int_V \rho(x, y, z) (y^2 + z^2) dV$$

Non-diagonal terms:

$$I_{xy} = - \int_V \rho(x, y, z) x y dV$$

Note: Integration variables  $x, y, z$  defined relative to the rotation center. (Center of mass for free moving objects. Hinge point for objects rotating around a fixed hinge.)

# Inertia Tensor

$$I(t) = \begin{bmatrix} I_{xx} & I_{xy} & I_{xz} \\ I_{yx} & I_{yy} & I_{yz} \\ I_{zx} & I_{zy} & I_{zz} \end{bmatrix}$$

For objects with **uniform mass distribution** and total mass  **$M$**  :

Diagonal terms:

$$I_{xx} = M \int_V (y^2 + z^2) dV$$

Non-diagonal terms:

$$I_{xy} = -M \int_V x y dV$$

# Inertia Tensor

- In general,  $I(t)$  depends on the body rotation at time  $t$ .
- Has to be updated whenever its orientation changes.
- For rigid bodies we can simply **precompute** the integrals of its elements (MC sampling or discretization) in object space:

→ **body-space** tensor  $I_{body}$

- Gives current tensor  $I(t)$  by applying current rotation:

$$I(t) = R(t) I_{body} R(t)^T$$

where  $R(t)$  is the current rotation matrix (can be derived from  $q(t)$ )

# Angular Momentum

For **general rigid objects**:

**Formula:**  $\vec{L}(t) = I(t) \vec{\omega}(t) \iff \vec{\omega}(t) = I^{-1}(t) \vec{L}(t)$

**Derivative:**  $\frac{d}{dt} \vec{L}(t) = \vec{\tau}(t)$

Analogous to linear momentum, but for rotational motion: a heavy object that rotates fast requires large prolonged force to get it up to this speed, and an equally large force in order to stop it.

# Rigid-body Motion ODE

$$\frac{d}{dt} \begin{bmatrix} x(t) \\ q(t) \\ P(t) \\ L(t) \end{bmatrix} = \begin{bmatrix} v(t) \\ 0.5 \omega(t) q(t) \\ F(t) \\ \tau(t) \end{bmatrix}$$

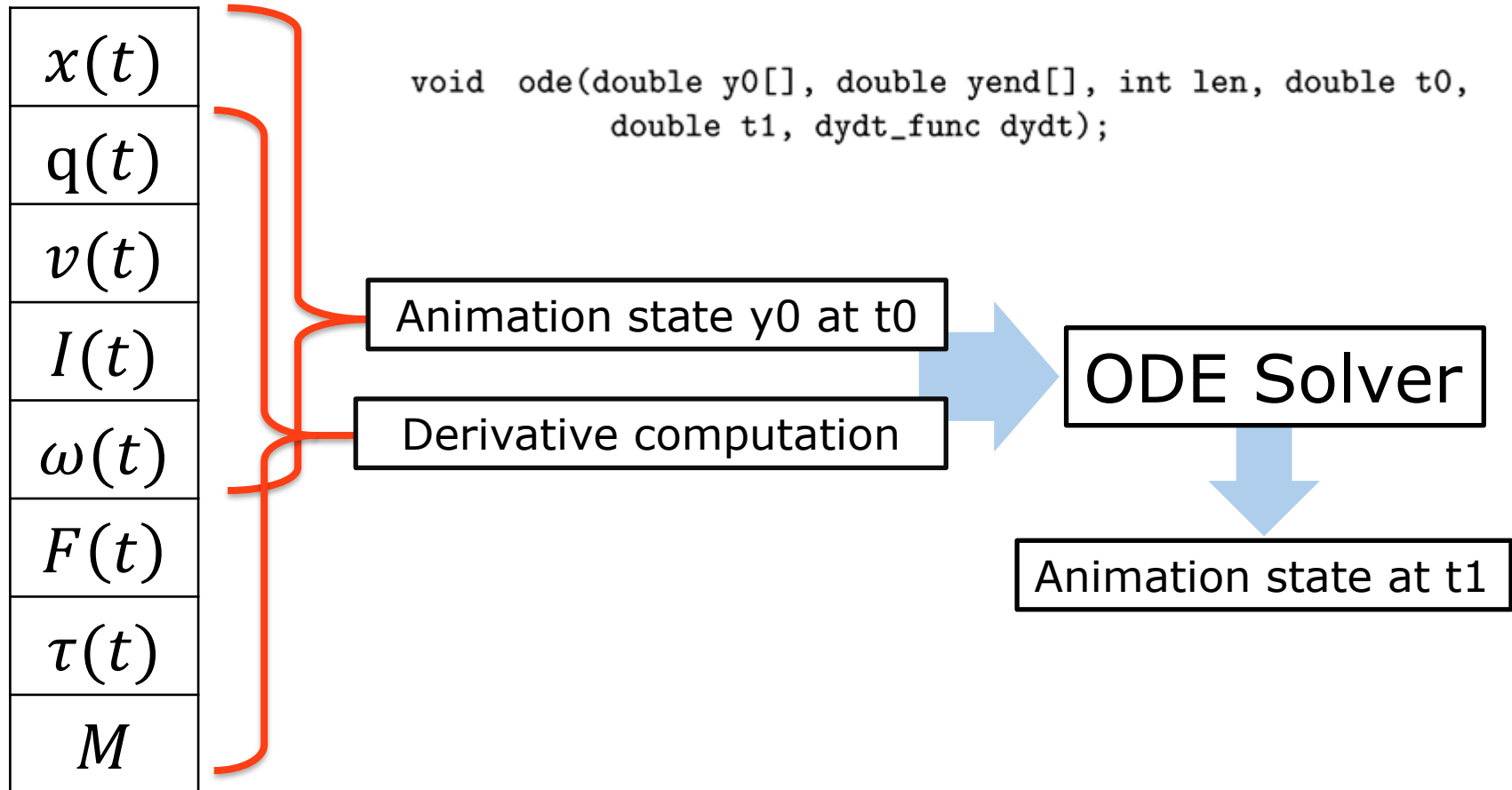
$\omega(t)$  as quaternion  $(0, \omega(t))$

Proof in the lecture notes

Sum of all forces

Total torque

# Rigid-body Representation



x N times for simulations with N objects

# Quaternions vs. Rotation Matrices

- Instead of quaternions, a 3x3 rotation matrix can be used to represent orientation
  - 9 vs. 4 variables to represent the 3DoF of rotation
  - The numerical ODE solver introduces drift
  - Less variables → less drift**(quaternions are more robust to numerical errors)**
  - Drift in the case of a rotation matrix will result in a non-orthogonal matrix that will cause a skewing effect
  - Drift in the case of a quaternion will result in a quaternion that is not unit length



**Solution:** Normalize quaternion after every solver step to obtain unit quaternion again.



# Angular Momentum in 2D

- Object extent only in xy-plane
- Perpendicular axis theorem:  $I_{zz} = I_{xx} + I_{yy}$

$$\rightarrow I = \begin{bmatrix} I_{xx} & I_{xy} & 0 \\ I_{yx} & I_{yy} & 0 \\ 0 & 0 & I_{zz} \end{bmatrix} = \begin{bmatrix} I_{xx} & I_{xy} & 0 \\ I_{yx} & I_{yy} & 0 \\ 0 & 0 & I_{xx} + I_{yy} \end{bmatrix}$$

- Axis of rotation is the z-axis:  $\vec{\omega}(t) = (0, 0, \omega_{xy}(t))$

angular velocity  
in the xy-plane

- Angular momentum  $\vec{L}(t) = I \vec{\omega}(t) = (0, 0, I_{zz} \omega_{xy}(t))$
- simplifies to scalar product  $L_{xy}(t) = I_{zz} \omega_{xy}(t)$

Single scalar, all we need  
to store for a 2D object

# **4. Collision Detection and Response**

# Collision detection types

- **Discrete** (a posteriori)

The simulation proceeds in steps. After each step, a list of colliding objects are detected, and their position is “fixed”

→ collisions are detected *after* the collision event.

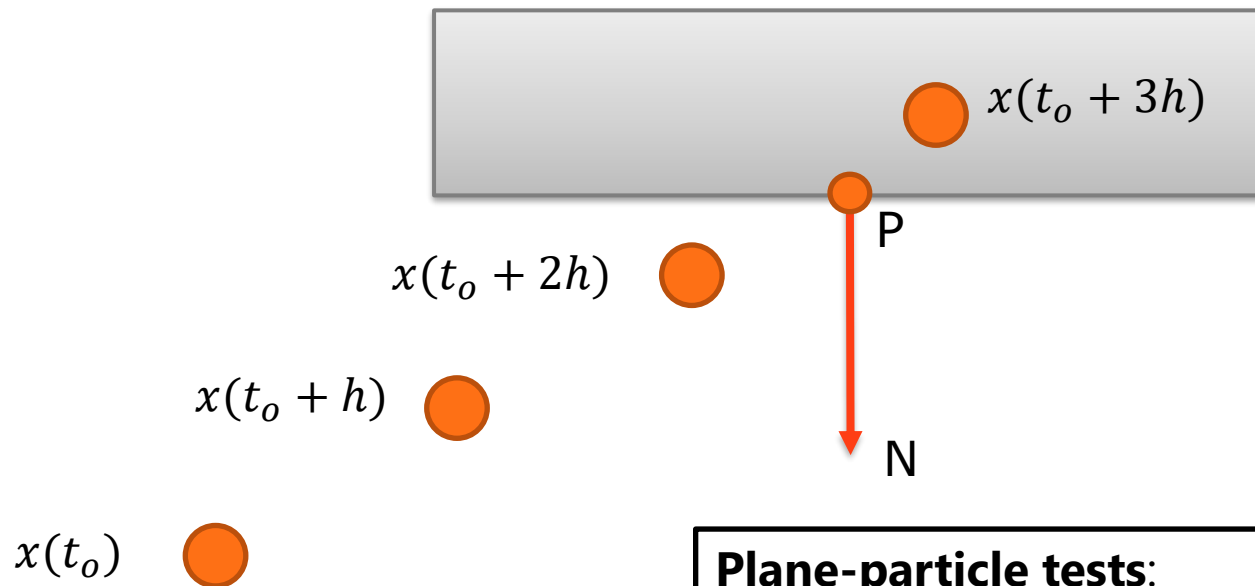
- **Continuous** (a priori)

The collision detection method is able to predict very precisely the time and place a collision happens and the physical bodies never actually interpenetrate

→ collisions are detected *before* the collision event.

# Particle collisions

Avoid interpenetrations between solid/rigid objects at collision.

**Plane-particle tests:**

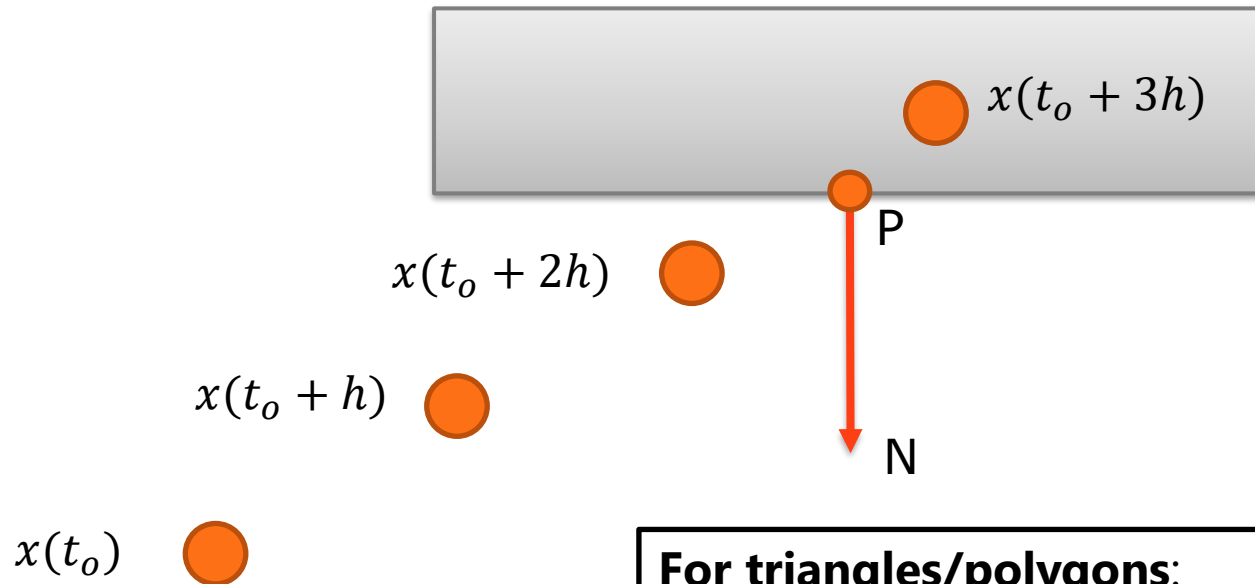
$(x - P) \cdot N > 0 \rightarrow$  in front of the plane

$(x - P) \cdot N < 0 \rightarrow$  behind the plane

$(x - P) \cdot N < \varepsilon \rightarrow$  very close, heading in

# Particle collisions

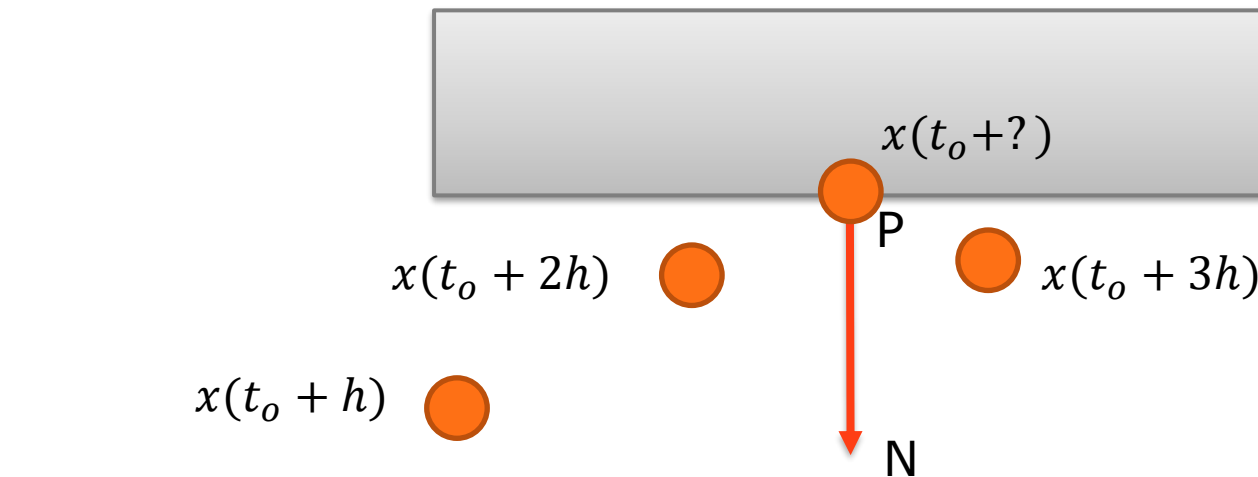
Avoid interpenetrations between solid/rigid objects at collision.

**For triangles/polygons:**

If the particle crosses the plane of the polygon, compute the ray-plane intersection and test if the intersection point lies within the polygon.

# Particle collisions

Avoid interpenetrations between solid/rigid objects at collision.



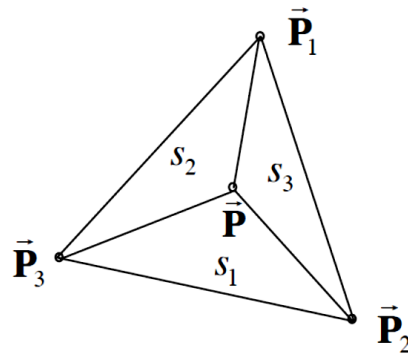
## Collision response:

1. The **position** of the particle is moved to a non-penetrating position
2. The **velocity** of the object is adjusted
3. The correct position is computed based on the new position, velocity and residual time

# Particle collisions

For linear trajectories, perform a ray-object intersection.

## Ray-triangle intersection:



### Barycentric coordinates

$$\vec{P} = s_1 \vec{P}_1 + s_2 \vec{P}_2 + s_3 \vec{P}_3$$

### Inside triangle criteria

$$s_1 = \text{area}(\triangle PP_2P_3) / \text{area}(\triangle P_1P_2P_3)$$

$$s_2 = \text{area}(\triangle P_1PP_3) / \text{area}(\triangle P_1P_2P_3)$$

$$s_3 = \text{area}(\triangle P_1P_2P) / \text{area}(\triangle P_1P_2P_3)$$

$$0 \leq s_1 \leq 1$$

$$0 \leq s_2 \leq 1$$

$$0 \leq s_3 \leq 1$$

$$s_1 + s_2 + s_3 = 1$$

### Equation for the intersection point:

$$O + tD = (1 - s_1 - s_2)P_3 + s_1P_1 + s_2P_2$$

$O$  = ray origin

$D$  = ray direction

# N-body collision

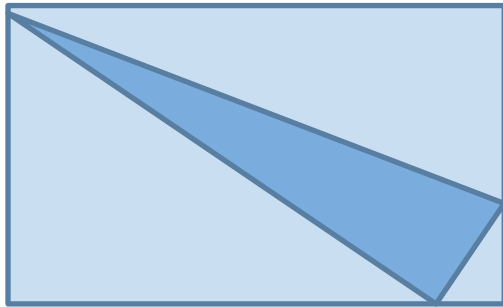
- Trivial approach: test all possible pairs
  - For  $N$  objects this will result in  $O(N^2)$  tests!
  - **Prohibitive cost** for real-time applications with high number of objects.



- Prune as-fast-as-possible collision tests of object-pairs that are not inter-penetrating.
- Only spend time for accurate collision tests on object pairs that are potentially penetrating.

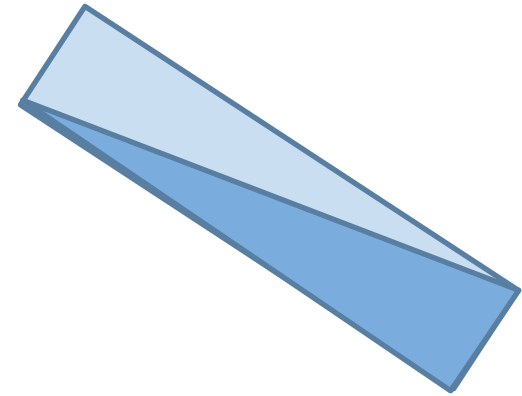


# Bounding boxes



**Axis Aligned BB**

1. Fast to compute
2. Very fast tests against points, other AABB, and polygons
3. Less tight bounds  
→ BB test will rule out less non-intersecting objects

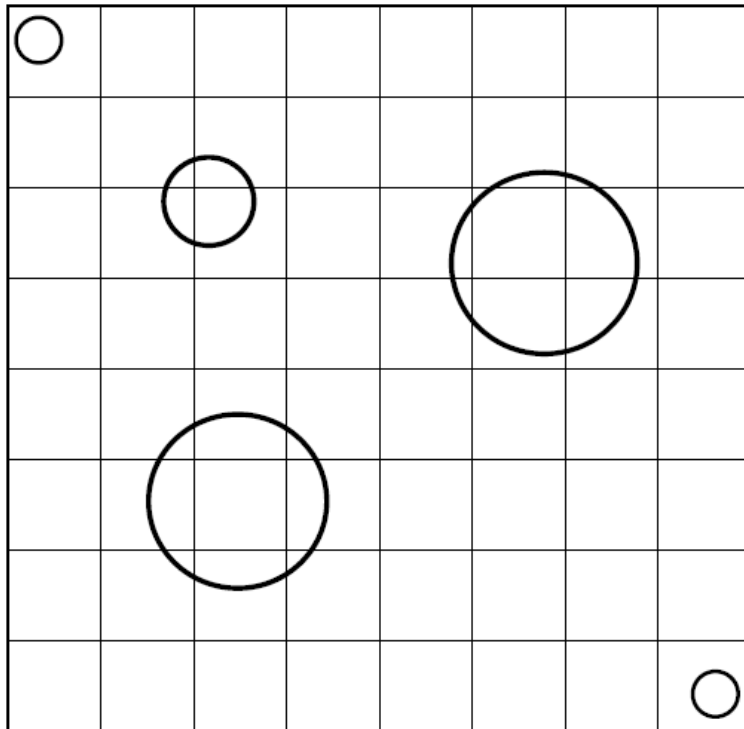


**Object-Oriented BB**

1. Tight Bounds
2. Slower collision tests
3. Slower build times

# Uniform Grids

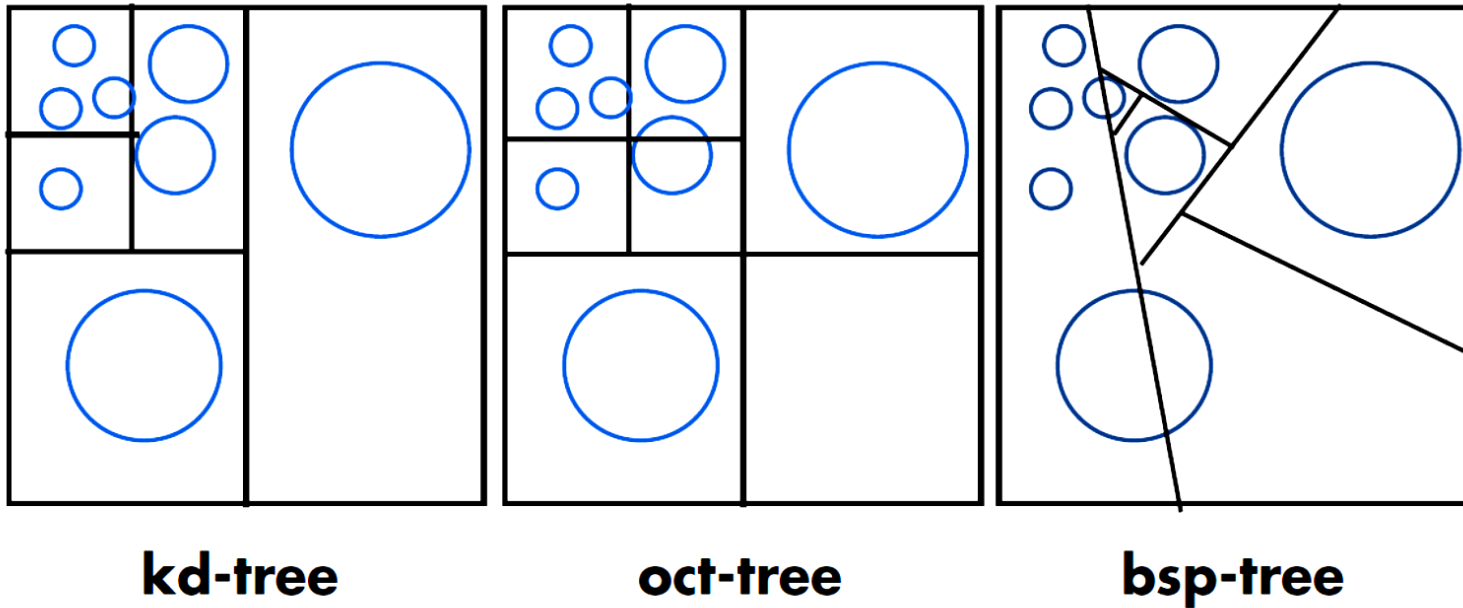
- Subdivide the space in uniform cells
- Only test objects in the same cells



## Problems:

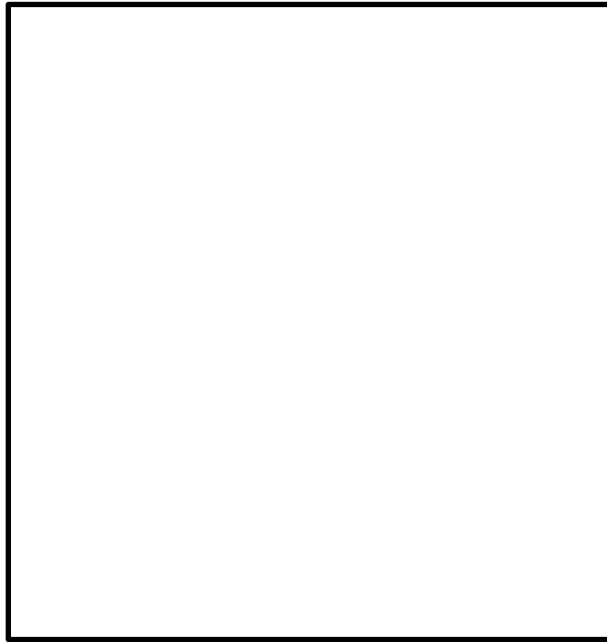
1. Does not work very well when the distribution of objects is non-uniform
2. What is the optimal cell size?

# Spatial Hierarchies

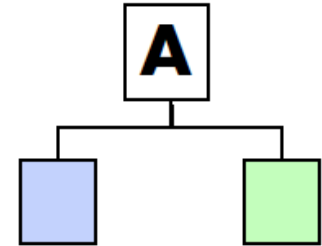
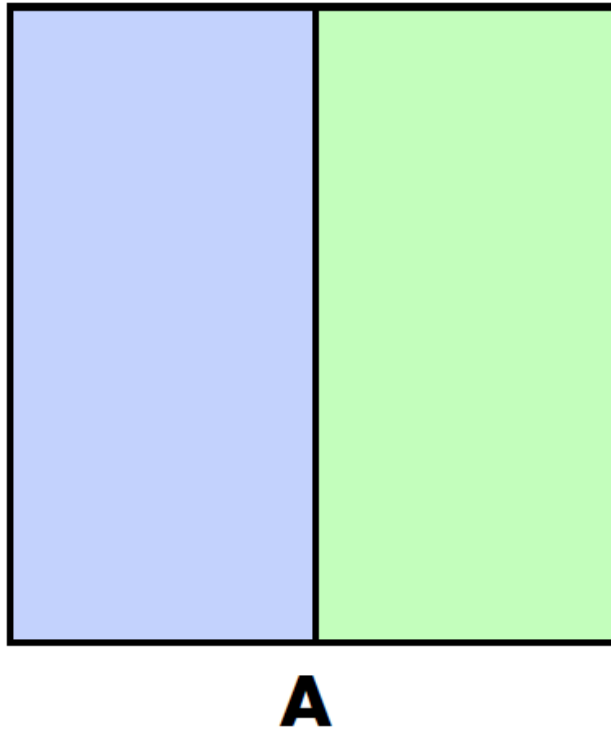


1. The space is subdivided in convex cells.
2. Only perform collision tests with objects in the same convex-cell.

# Spatial Hierarchies

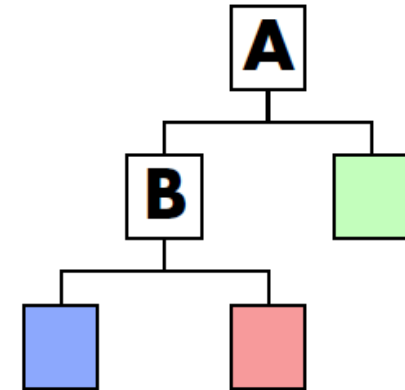
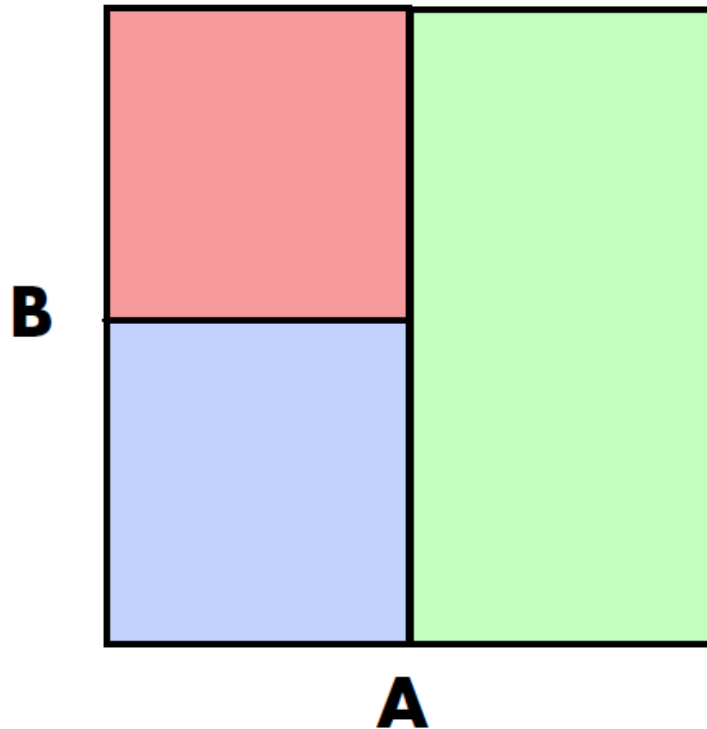


# Spatial Hierarchies



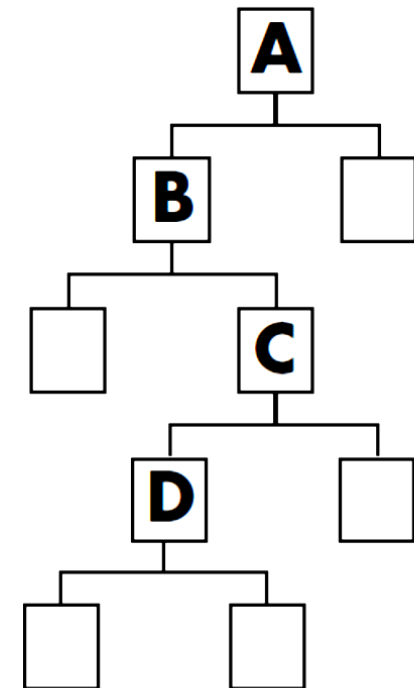
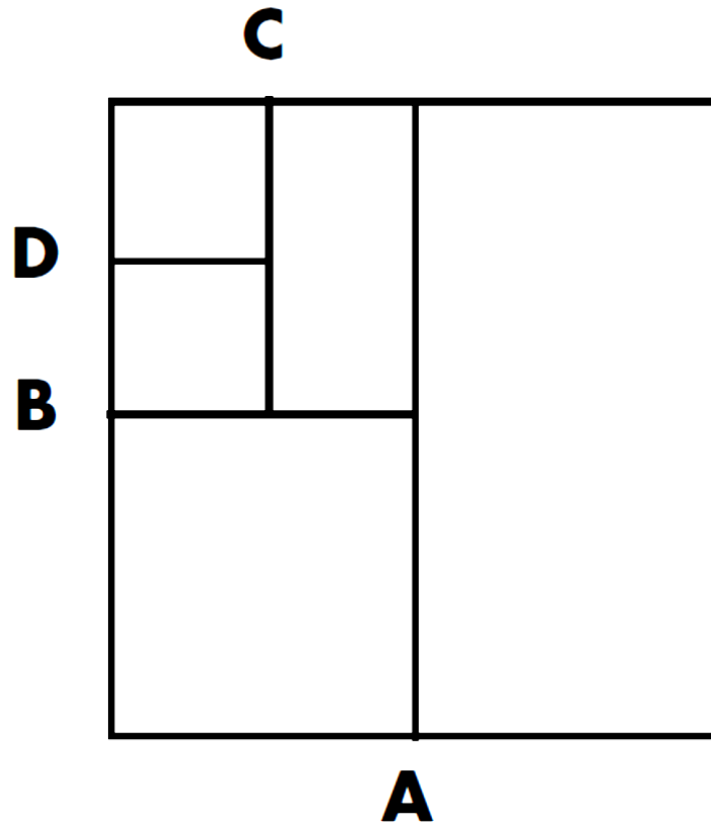
Letters correspond to planes (A)

# Spatial Hierarchies



Letters correspond to planes (A,B)

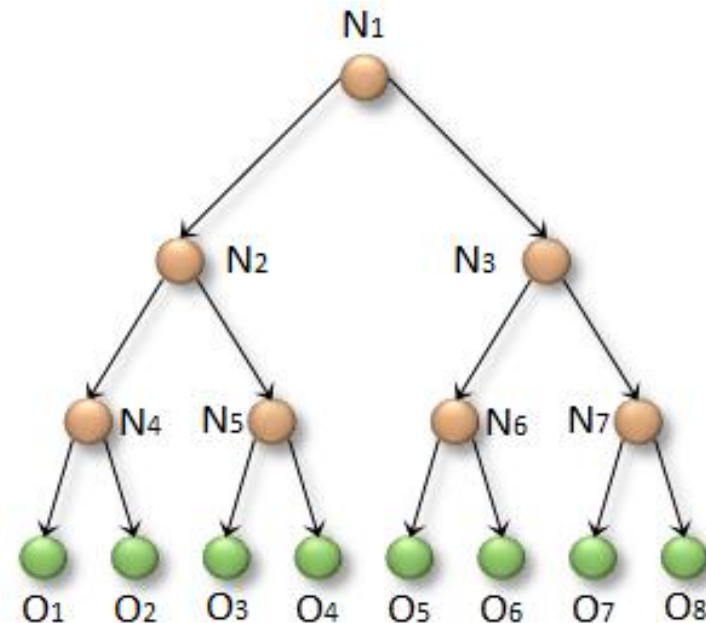
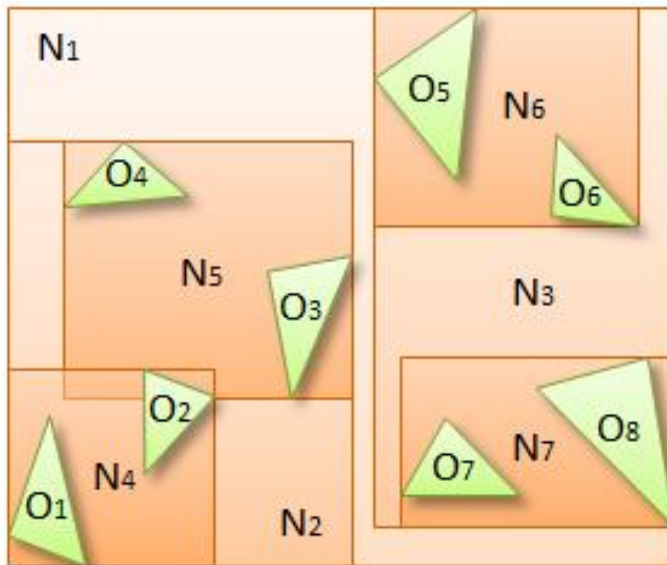
# Spatial Hierarchies



Letters correspond to planes (A,B,C,D)

# Bounding Volume Hierarchies

- Bottom-up hierarchical grouping of primitives  
Note: bounding volumes might overlap!
- Fast build times, more suitable for dynamic objects

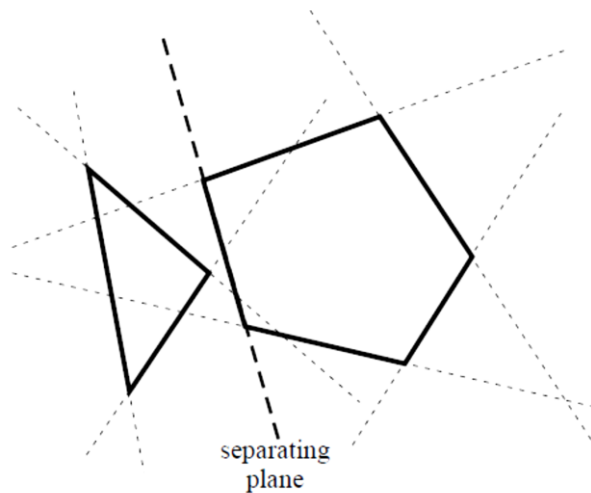




# Object-object collisions

- Convert both objects to BVHs or other spatial hierarchies
- Test the convex subspaces for collisions
- **Separating-Axis Theorem (SAT):**

*Two arbitrary convex regions do not interpenetrate if a separating axis (or plane in 3D) exists:*



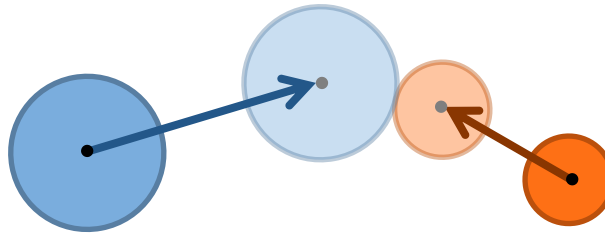
In convex triangle-meshes, finding the separating axis/plane minimizes the amount of analytical triangle-triangle intersection tests!

# Collision Response

- Objects that collide should respond according to

## **Newton's third law of motion:**

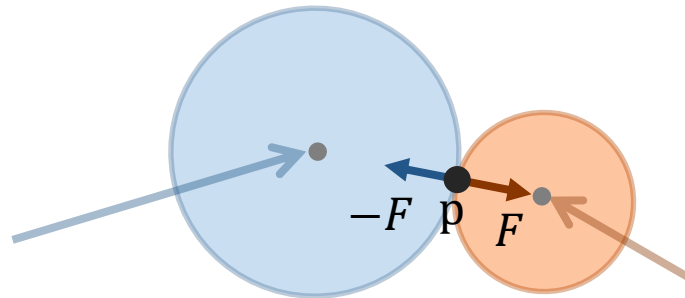
*When one body exerts a force on a second body, the second body simultaneously exerts a force equal in magnitude and opposite in direction on the first body*



- Requires updating the linear and angular momentum of both objects. (Analogously: update linear and angular velocities)

# Momentum Update

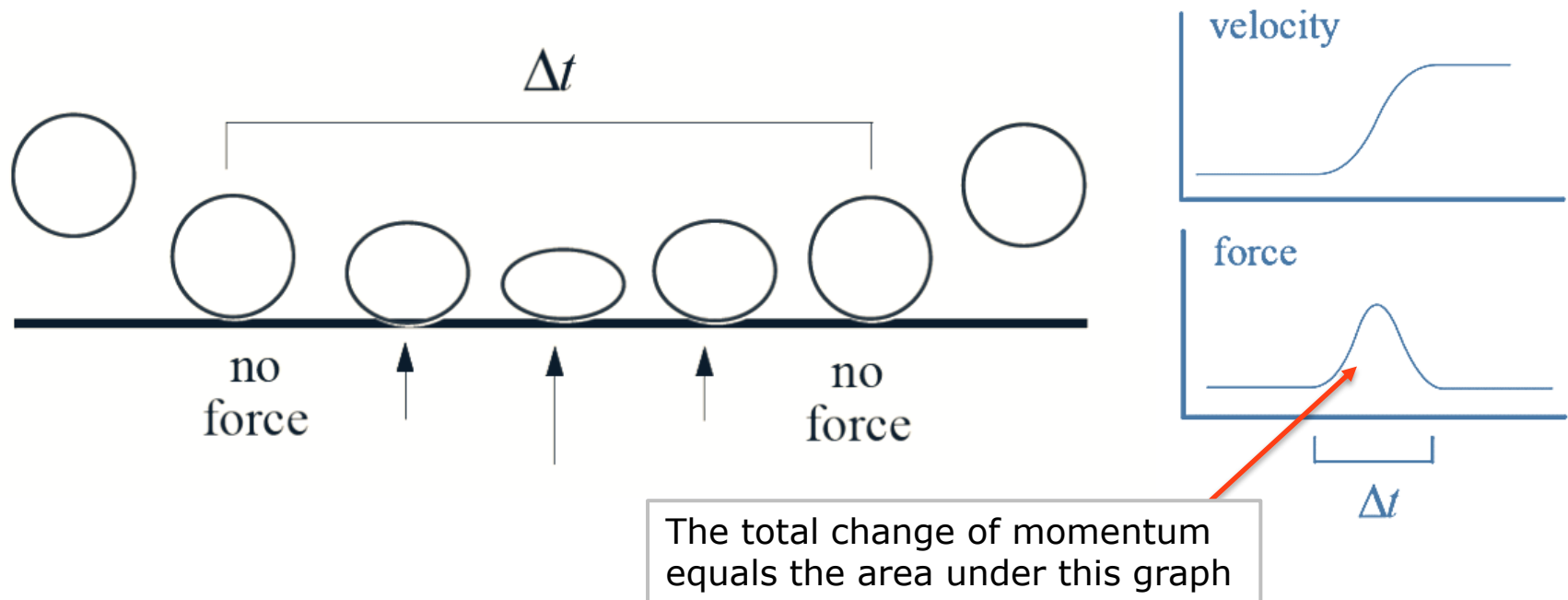
- Momentums are updated based on exerted (opposing) forces acting **at the contact point p**.



- In frictionless bodies, forces act only in direction orthogonal to the contact surface (surface normal direction).
- Change in linear momentum equals force  $F$  times duration  $d$  of exertion. 
$$\Delta P = \int_0^d F(t) dt$$
- Rigid Bodies: How large is  $F$ ? ... and  $d$ ?

# Collision Process

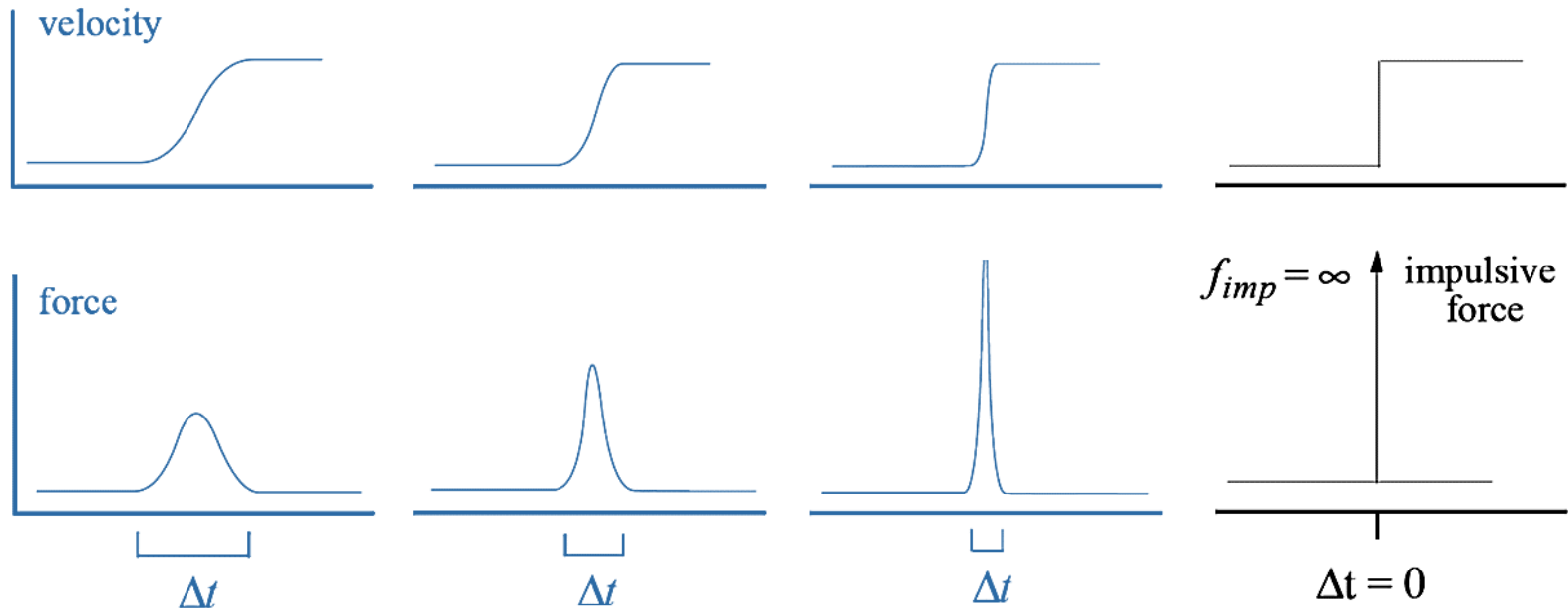
- General (non-singular) collision and bounce process:



- In practice** we will assume instantaneous rigid collisions (no deformations)  $\rightarrow \Delta t = 0$

# Collision Process

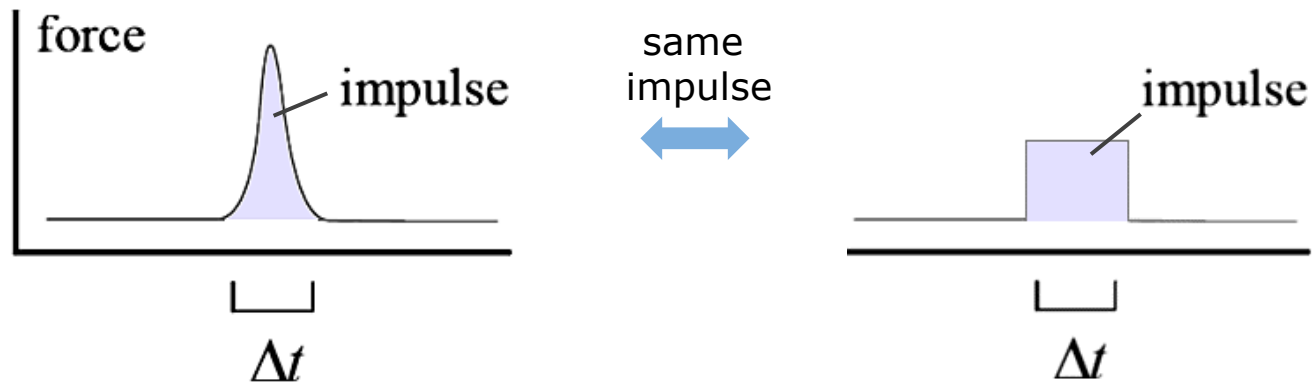
- Vanishing the contact duration/deformation time:  $\Delta t \rightarrow 0$



- Instantaneous collisions ( $\Delta t=0$ ) would require an infinite force to produce the same change in momentum  $\Delta P$ .
- In practice we will instead work with a finite **impulse**.

# Impulse (J)

- Defines the integral force  $F$  over a certain time interval  $\Delta t$ :



- Impulse  $J = F \cdot \Delta t = \Delta P \rightarrow$  defines change in momentum:
  - linear:  $dP = F dt = J$
  - angular:  $dL = \tau dt = (p - c) \times F dt = (p - c) \times J$
- If one colliding body experiences an impulse  $J$ , the other experiences an impulse  $-J$ .

# Impulse at collision point

- On frictionless colliding bodies, impulse acts only along the contact surface normal direction  $\hat{n}$ .
- The first body experiences an impulse at **collision point  $p$**  based on its **relative velocity at  $p$**  along  $\hat{n}$ :

$$J = j \hat{n}$$

$$j = \frac{-(1+\epsilon) (\dot{p}_1 - \dot{p}_2) \cdot \hat{n}}{m_1^{-1} + m_2^{-1} + \left[ \left( I_1^{-1} (r_1 \times \hat{n}) \right) \times r_1 + \left( I_2^{-1} (r_2 \times \hat{n}) \right) \times r_2 \right] \cdot \hat{n}}$$

$m_i, v_i$

$r_i = p - x_i$

$\dot{p}_i = v_i + \omega_i \times r_i$

$\epsilon \in [0,1]$

at time of contact

mass and linear velocity of body  $i$

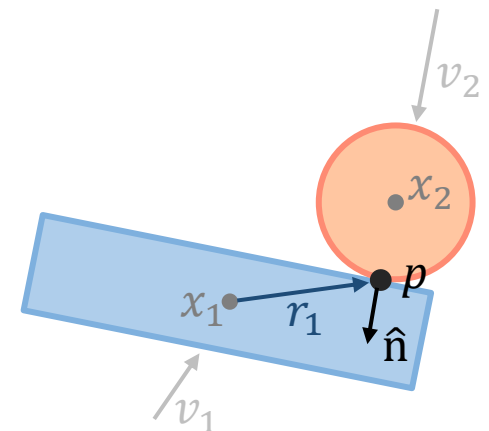
radius vector of  $p$  in body  $i$

linear velocity of  $p$  in body  $i$

restitution coefficient (bounciness)

$\epsilon=1$ : perfect elastic collision

$\epsilon<1$ : loss of kinetic energy

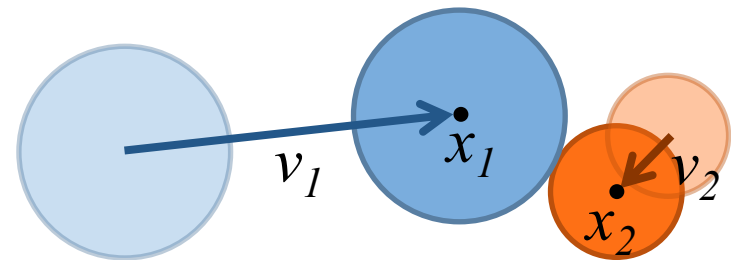


By convention,  $\hat{n}$  points towards body 1.

# Example: Elastic sphere collision

$$j = \frac{-(1+\varepsilon) (\dot{p}_1 - \dot{p}_2) \cdot \hat{n}}{m_1^{-1} + m_2^{-1} + \left[ \left( I_1^{-1} (r_1 \times \hat{n}) \right) \times r_1 + \left( I_2^{-1} (r_2 \times \hat{n}) \right) \times r_2 \right] \cdot \hat{n}}$$

- Perfect elastic collision:  $\varepsilon=1$



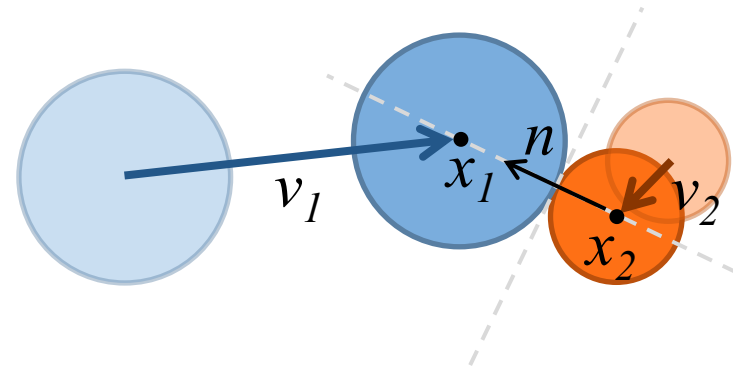


# Example: Elastic sphere collision

$$j = \frac{-2 (\dot{p}_1 - \dot{p}_2) \cdot \hat{n}}{m_1^{-1} + m_2^{-1} + \left[ \left( I_1^{-1} (\cancel{r_1 \times \hat{n}}) \right) \times \cancel{r_1} + \left( I_2^{-1} (\cancel{r_2 \times \hat{n}}) \right) \times \cancel{r_2} \right] \cdot \hat{n}}$$

- Perfect elastic collision:  $\epsilon=1$
- Contact point on spheres:

$$\hat{n} = \frac{x_1 - x_2}{\|x_1 - x_2\|} \parallel r_i$$

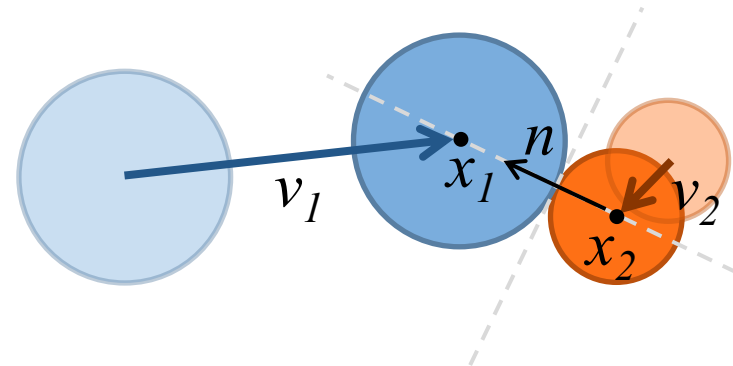


# Example: Elastic sphere collision

$$j = \frac{-2 (\dot{p}_1 - \dot{p}_2) \cdot \hat{n}}{m_1^{-1} + m_2^{-1}}$$

- Perfect elastic collision:  $\epsilon=1$
- Contact point on spheres:

$$\hat{n} = \frac{x_1 - x_2}{\|x_1 - x_2\|} \parallel r_i$$



# Example: Elastic sphere collision

$$\mathbf{j} = \frac{-2 (\mathbf{v}_1 - \mathbf{v}_2) \cdot \hat{\mathbf{n}}}{m_1^{-1} + m_2^{-1}}$$

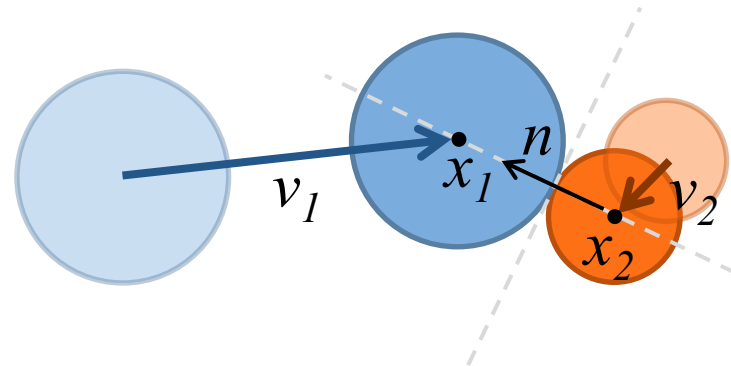
- Perfect elastic collision:  $\epsilon=1$
- Contact point on spheres:

$$\hat{\mathbf{n}} = \frac{\mathbf{x}_1 - \mathbf{x}_2}{\|\mathbf{x}_1 - \mathbf{x}_2\|} \parallel \mathbf{r}_i$$

- Assume no spin:  $\dot{\mathbf{p}}_i = \mathbf{v}_i$
- Velocity changes:

$$d\mathbf{v}_1 = \frac{d\mathbf{P}_1}{m_1} = +\mathbf{J}/m_1 = +\mathbf{j} \hat{\mathbf{n}}/m_1$$

$$d\mathbf{v}_2 = \frac{d\mathbf{P}_2}{m_2} = -\mathbf{J}/m_2 = -\mathbf{j} \hat{\mathbf{n}}/m_2$$



# Example: Elastic sphere collision

$$j = \frac{-2 (v_1 - v_2) \cdot \hat{n}}{m_1^{-1} + m_2^{-1}}$$

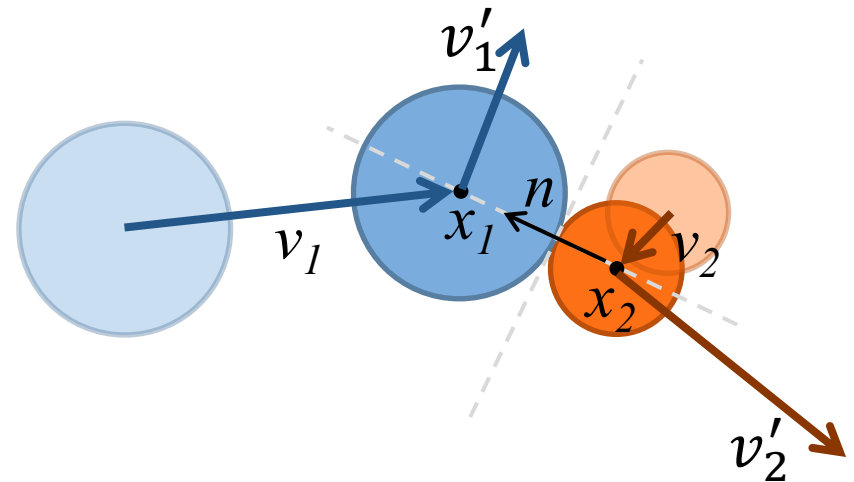
- Perfect elastic collision:  $\epsilon=1$
- Contact point on spheres:

$$\hat{n} = \frac{x_1 - x_2}{\|x_1 - x_2\|} \parallel r_i$$

- Assume no spin:  $\dot{p}_i = v_i$
- New Velocities

$$v'_1 = v_1 - \frac{2m_2}{m_1 + m_2} (v_1 - v_2) \cdot n * n$$

$$v'_2 = v_2 + \frac{2m_1}{m_1 + m_2} (v_1 - v_2) \cdot n * n$$



Many real-world collisions are inelastic, i.e., some kinetic energy is transformed to deformation energy or heat.

# Lecture Notes

- Additional details, examples and code samples in lecture nodes → [TeachCenter](#):
  - **Ordinary differential equations**  
[L6\\_ODE\\_basics.pdf](#)
  - **Rigid body dynamics (Siggraph course notes)**  
[L6\\_rigid\\_bodies.pdf](#)

## Acknowledgements:

*Andrew Witkin and David Baraff,*

**Physically Based Modeling: Principles and Practice**

# Conclusion

- Overview of the principles behind physically-based animation.
  - Particles
  - Rigid bodies
- First order ODEs and solvers
  - Euler's method, midpoint, etc...
- Equations of motion and related physical quantities
  - Inertia tensor, linear momentum, angular momentum
- Collision detection and response
  - Impulse, elastic collision