

Simulation in Computer Graphics

Particle Motion 1

Matthias Teschner

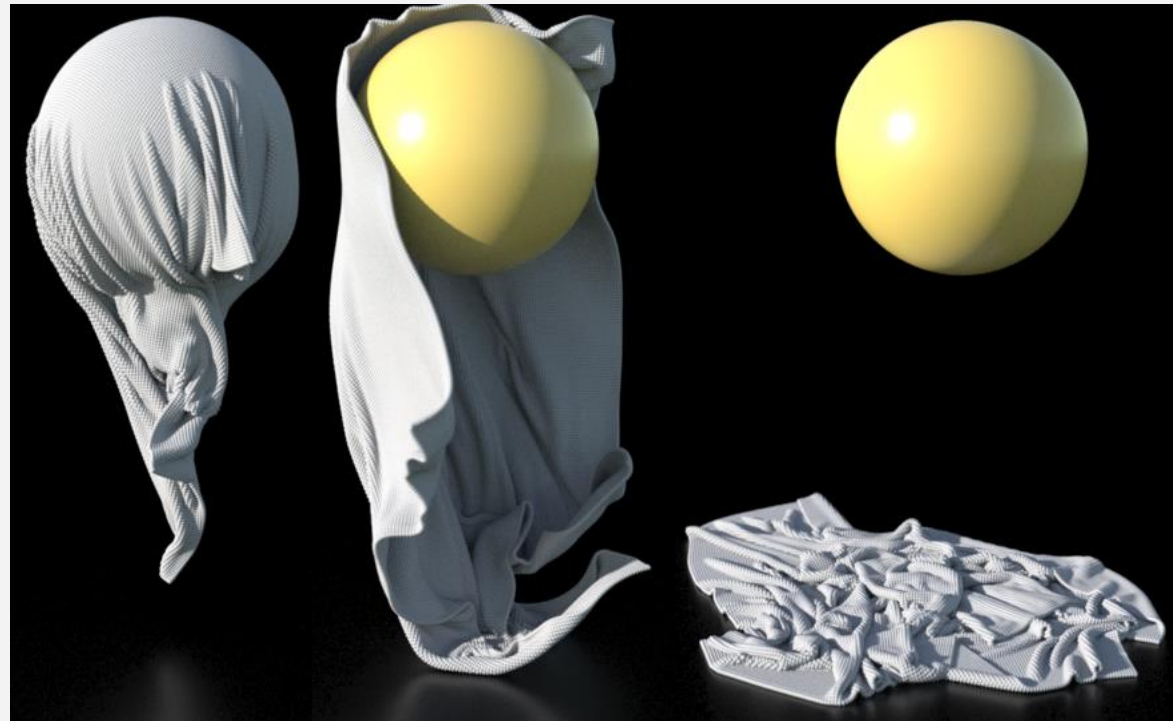


Outline

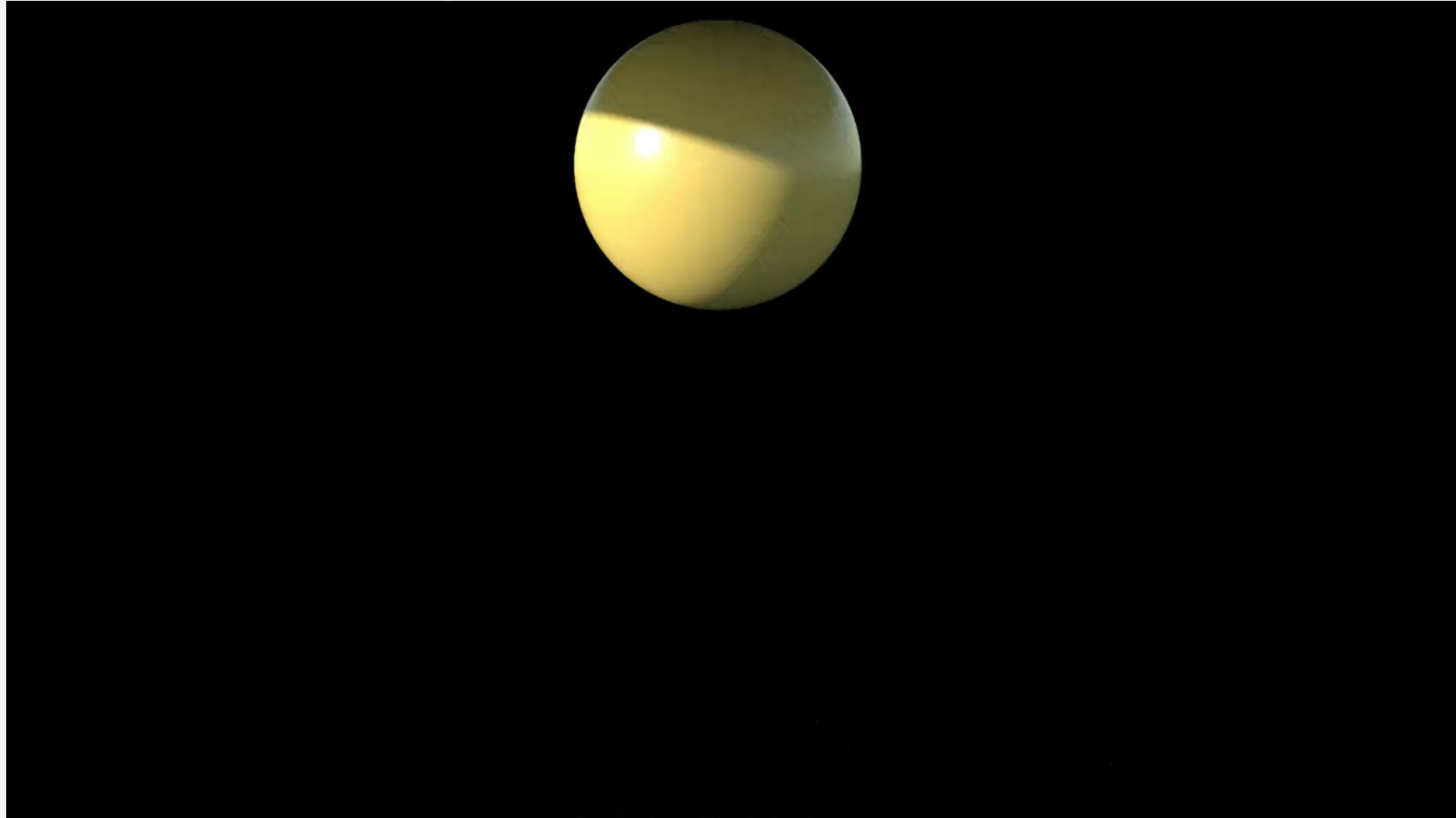
- Introduction
- Particle motion
- Finite differences
- System of first-order ODEs
- Second-order ODE
- Performance
- Discussion

Goal

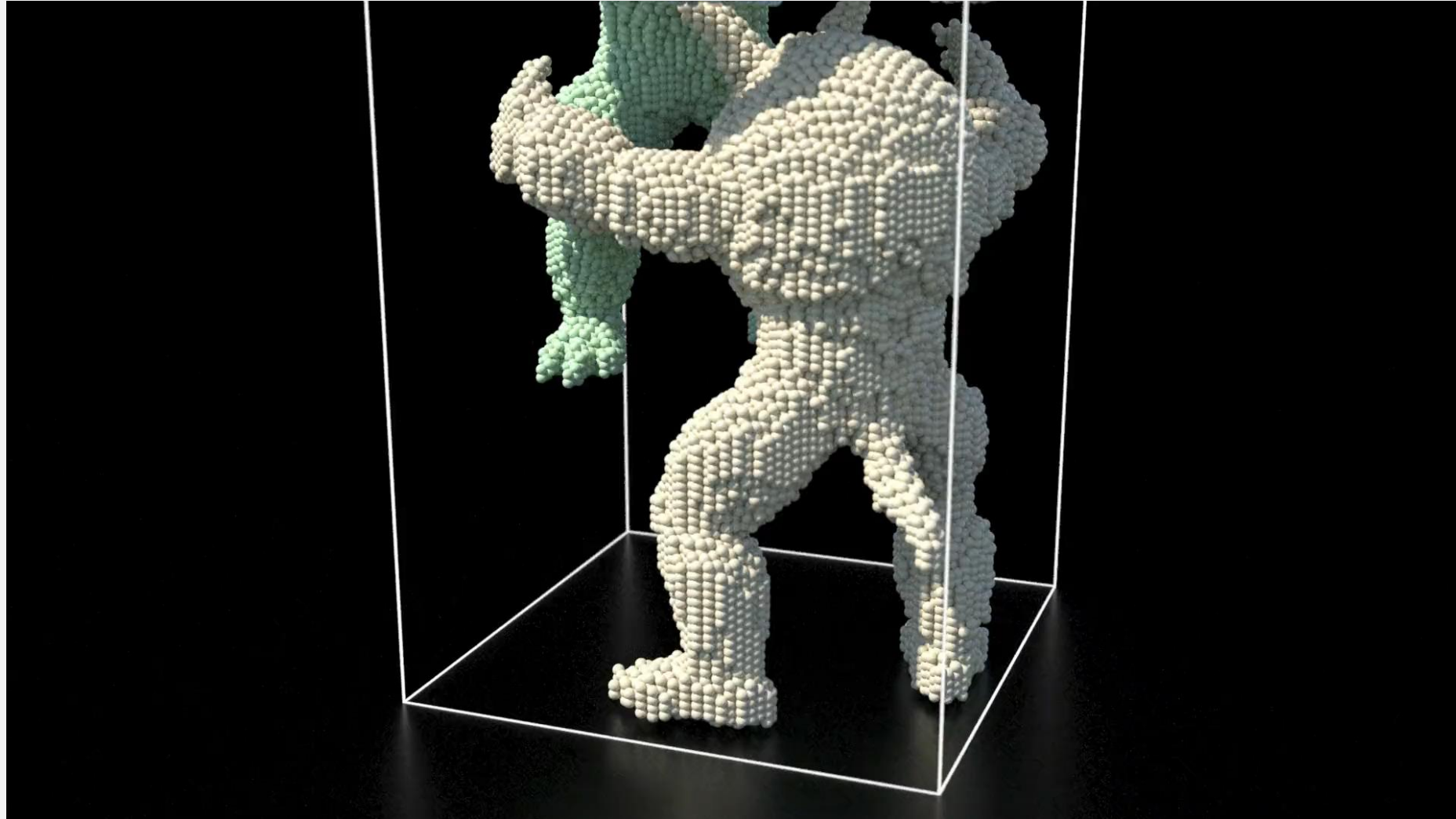
- Dynamic simulation of
 - Rigid bodies
 - Deformable objects
 - Fluids



Goal



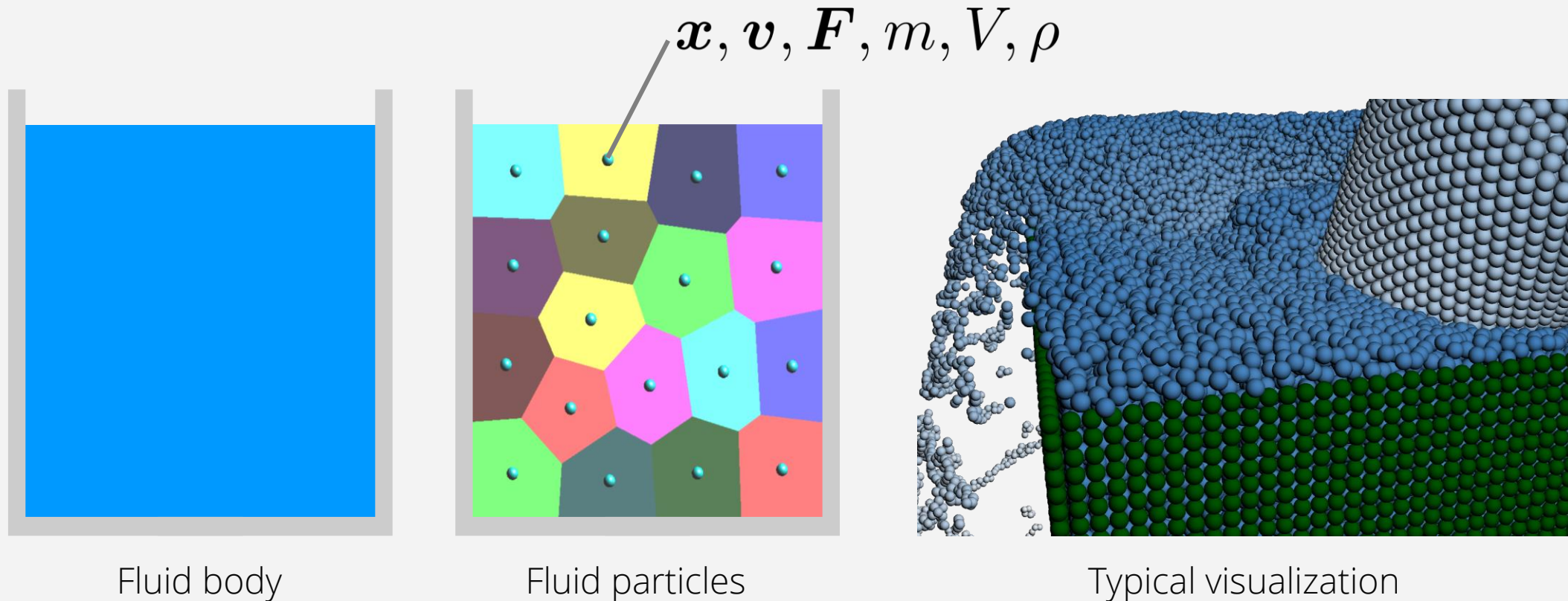
Goal



Representation

- Subdivision of objects into small parts, i.e. particles
- Particles have properties
 - Mass m , volume V , density ρ
 - Position \mathbf{x} , velocity \mathbf{v} , force \mathbf{F}
- Particles are of arbitrary shape

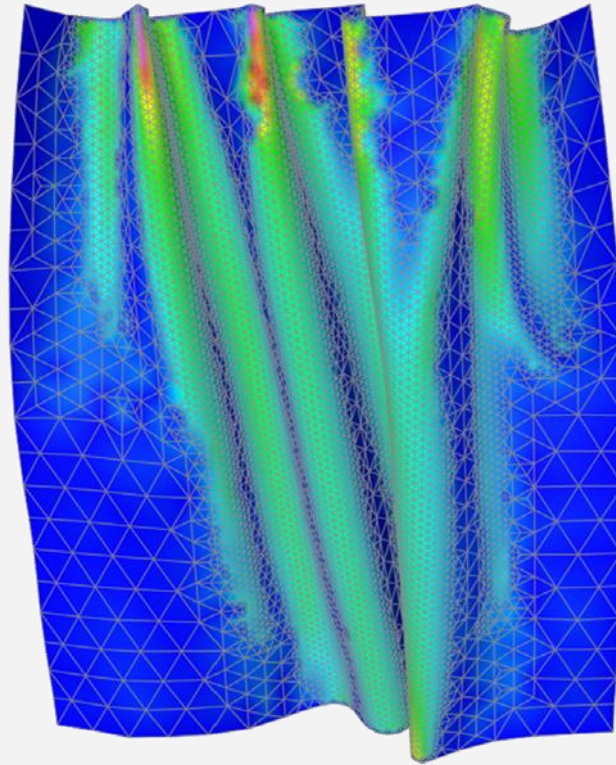
Fluid Particles



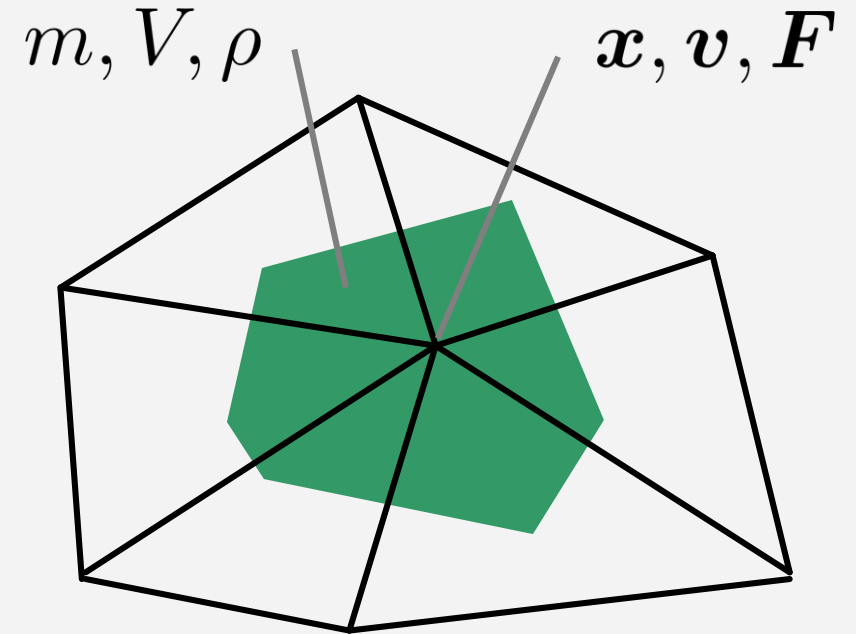
Cloth Particles



Cloth

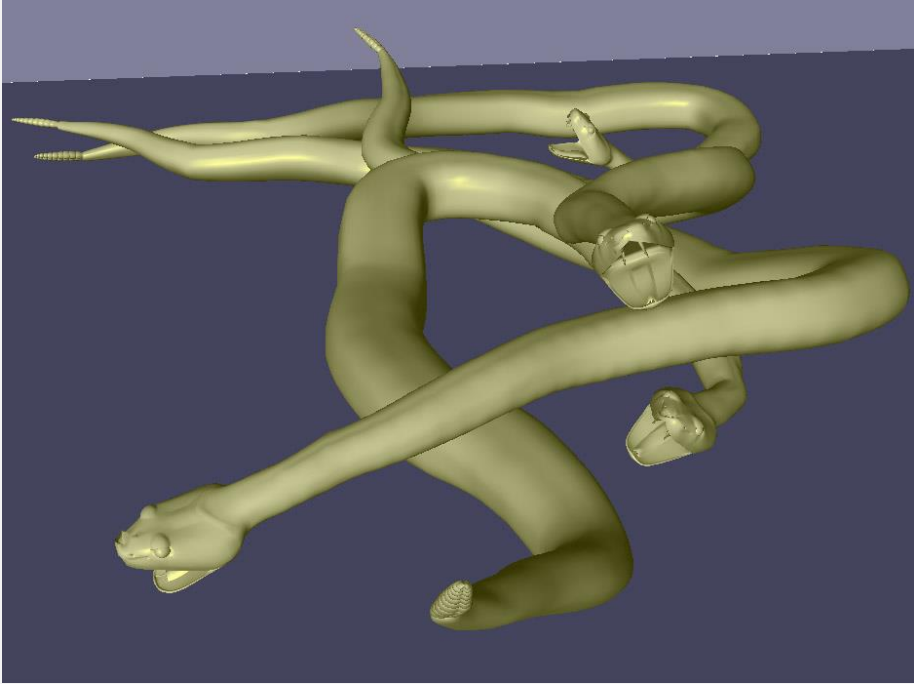


Cloth particles

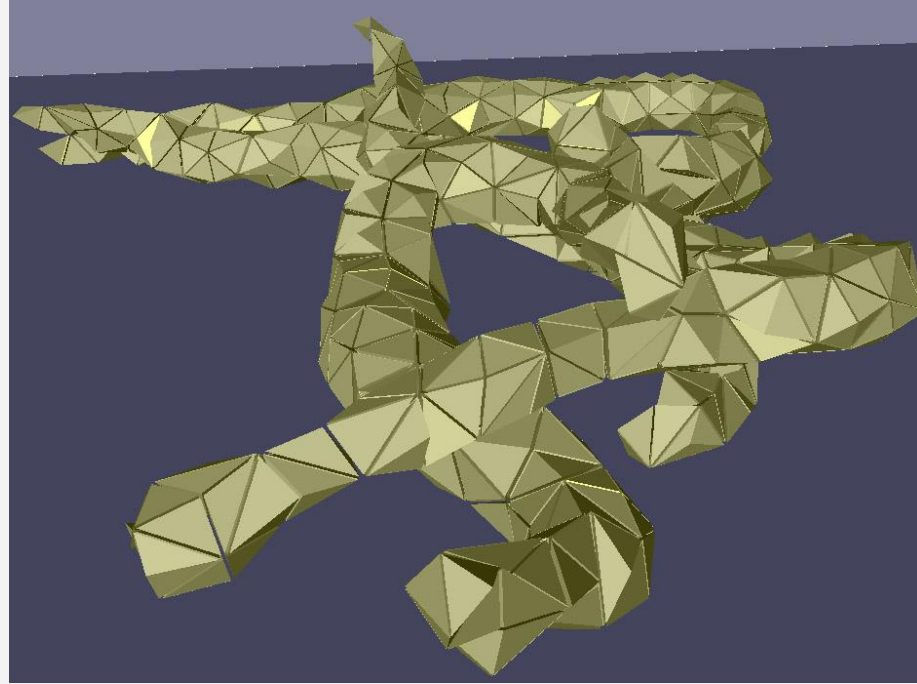


Illustration

Deformable 3D Particles



Deformable 3D object



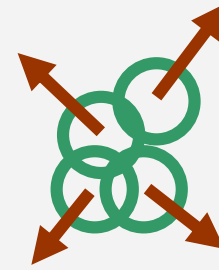
Approximate tetrahedral mesh

Particle Forces

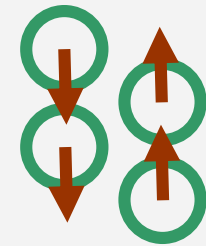
- Result from
 - Distortions, e.g. volume or shape change
 - Gravity
 - Friction, viscosity
 - Contact
 - ...



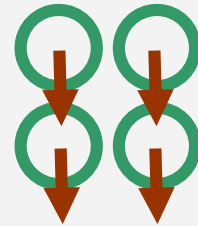
Rest state



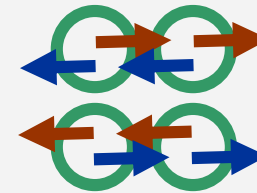
Compression



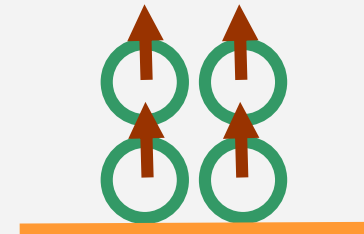
Shear



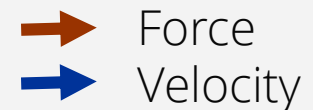
Gravity



Viscosity



Contact



Particle Motion

– Particles change position \mathbf{x} with velocity \mathbf{v} $\mathbf{v} = \frac{d\mathbf{x}}{dt}$

– Velocity governed by Newton's Second Law

– Force at a particle equals the time rate of change of its momentum

$$\mathbf{F} = \frac{d}{dt}(m\mathbf{v}) = \frac{dm}{dt}\mathbf{v} + \frac{d\mathbf{v}}{dt}m$$

– Two governing equations for two unknown functions \mathbf{x}, \mathbf{v}

$$\mathbf{F} = m\frac{d\mathbf{v}}{dt} \quad \mathbf{v} = \frac{d\mathbf{x}}{dt} \quad \text{Coupled system of first order ODEs}$$

– Can also be written as

$$\mathbf{F} = m\frac{d^2\mathbf{x}}{dt^2} \quad \text{Second order ODE}$$

Particle-based Simulation

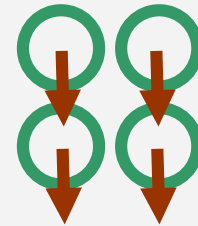
- Object subdivision into particles (spatial discretization)
- Force modeling
- Particle motion
 - Transport / advection



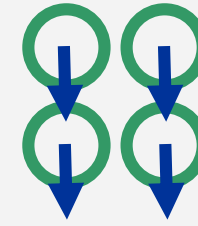
Object



Particles



Acceleration



Velocity change



Position change

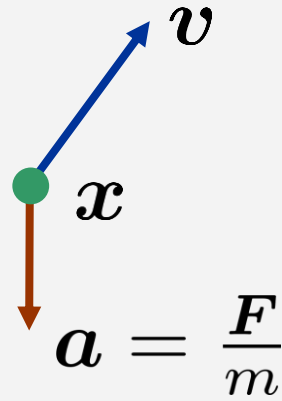
$$\frac{\mathbf{F}}{m} = \frac{d\mathbf{v}}{dt} = \frac{d^2\mathbf{x}}{dt^2}$$

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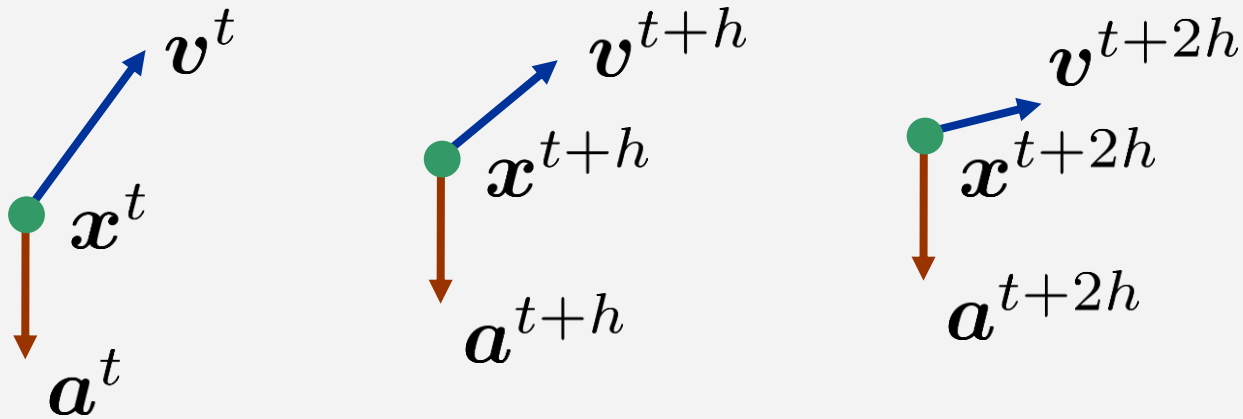
Particle Quantities

- Mass $m \in \mathbb{R}$
- Position $\mathbf{x} \in \mathbb{R}^3$
- Velocity $\mathbf{v} \in \mathbb{R}^3$
- Force $\mathbf{F} \in \mathbb{R}^3$
- Acceleration $\mathbf{a} = \frac{\mathbf{F}}{m} \in \mathbb{R}^3$



Time Discretization

- Quantities are considered at discrete time points



h is the so-called time step.

- Particle simulations are concerned with the computation of unknown future particle quantities $\mathbf{x}^{t+h}, \mathbf{v}^{t+h}$ from known current information $\mathbf{x}^t, \mathbf{v}^t, \mathbf{a}^t$

Governing Equations

- Newton's Second Law, motion equation

$$\mathbf{a}^t = \frac{d\mathbf{v}^t}{dt} = \frac{d^2\mathbf{x}^t}{dt^2}$$

- Ordinary differential equations ODEs
- Describe the behavior of \mathbf{x}^t and \mathbf{v}^t in terms of their derivatives with respect to time
- Numerical integration is employed to approximatively solve the ODE , i.e. to approximate the unknown functions \mathbf{x}^t and \mathbf{v}^t

Governing Equations

- Initial value problem of second order

$$\frac{d^2 \mathbf{x}^t}{dt^2} = \mathbf{a}^t \quad \mathbf{x}^{t_0} = \mathbf{x}^{\text{init}} \quad \frac{d\mathbf{x}^{t_0}}{dt} = \mathbf{v}^{\text{init}}$$

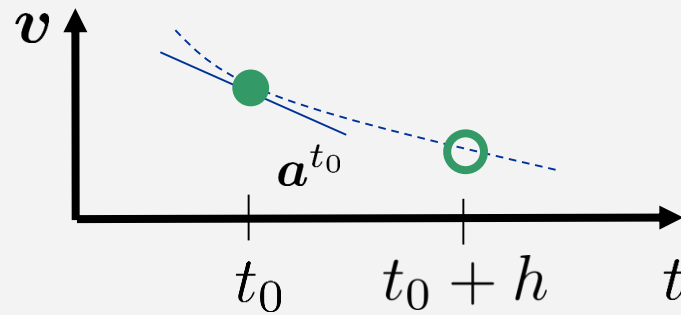
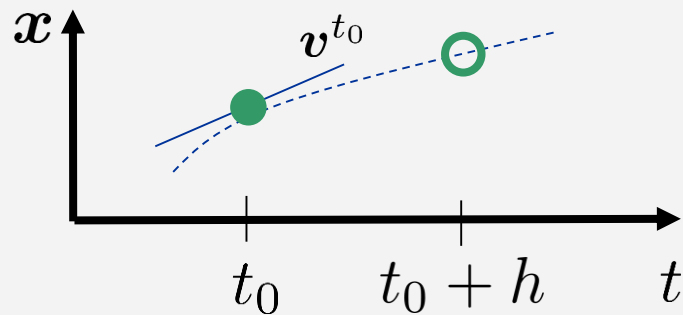
- Second-order ODEs can be rewritten as a system of two coupled equations of first order
- Initial value problems of first order

$$\frac{d\mathbf{x}^t}{dt} = \mathbf{v}^t \quad \mathbf{x}^{t_0} = \mathbf{x}^{\text{init}}$$

$$\frac{d\mathbf{v}^t}{dt} = \mathbf{a}^t \quad \mathbf{v}^{t_0} = \mathbf{v}^{\text{init}}$$

Initial Value Problem of First Order

- Functions \mathbf{x}^t and \mathbf{v}^t represent the particle motion
- Initial values \mathbf{x}^{t_0} and \mathbf{v}^{t_0} are given
- First-order differential equations are given
$$\frac{d\mathbf{x}^t}{dt} = \mathbf{v}^t \quad \frac{d\mathbf{v}^t}{dt} = \mathbf{a}^t$$
- How to estimate \mathbf{x}^{t_0+h} and \mathbf{v}^{t_0+h} ?



Particle Accelerations

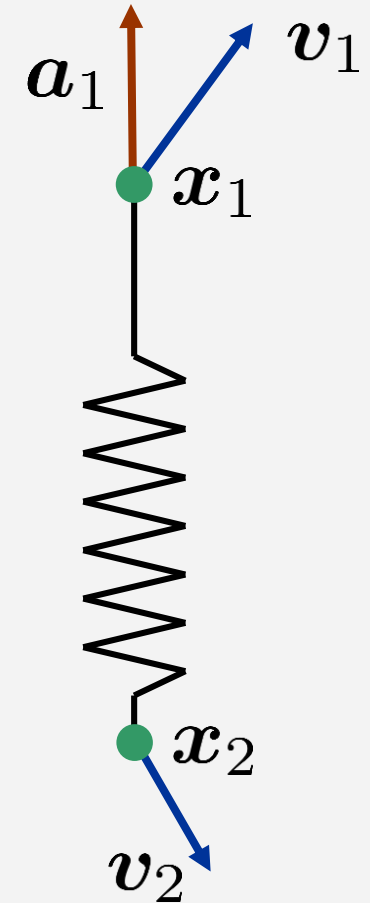
- Depend on sets of positions and velocities
- E.g., damped spring $\mathbf{a}_1(\mathbf{x}_1, \mathbf{x}_2, \mathbf{v}_1, \mathbf{v}_2)$

$$= \frac{1}{m_1} \frac{k_{12}}{L_{12}} \left(\|\mathbf{x}_2 - \mathbf{x}_1\| - L_{12} \right) \frac{\mathbf{x}_2 - \mathbf{x}_1}{\|\mathbf{x}_2 - \mathbf{x}_1\|} \quad \text{Elastic acceleration}$$

Particle mass	Spring stiffness	Actual spring length	Rest spring length	Normalized direction
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$$+ \frac{1}{m_1} \gamma \left((\mathbf{v}_2 - \mathbf{v}_1) \frac{\mathbf{x}_2 - \mathbf{x}_1}{\|\mathbf{x}_2 - \mathbf{x}_1\|} \right) \frac{\mathbf{x}_2 - \mathbf{x}_1}{\|\mathbf{x}_2 - \mathbf{x}_1\|} \quad \text{Damping acceleration}$$

Damping parameter	Relative velocity projected onto spring	Normalized direction
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Particle Accelerations

- Are typically expensive to compute
 - E.g., sums over adjacent particles
- Might need additional effort
 - E.g., contact handling forces require collision detection

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Finite Differences

- Taylor-series approximation

$$\mathbf{x}^{t+h} = \mathbf{x}^t + \frac{d\mathbf{x}^t}{dt}h + O(h^2) \quad O(h^2) - \text{order of the truncation / discretization error}$$

$$\frac{d\mathbf{x}^t}{dt} = \frac{\mathbf{x}^{t+h} - \mathbf{x}^t}{h} + O(h) \quad O(h) - \text{error order of, e.g., a scheme that employs such approximation}$$

- Continuous ODEs are replaced with discrete finite-difference equations FDEs

$$\begin{aligned} \frac{d\mathbf{x}^t}{dt} = \mathbf{v}^t &\Rightarrow \frac{\mathbf{x}^{t+h} - \mathbf{x}^t}{h} = \mathbf{v}^t &\Rightarrow \mathbf{x}^{t+h} = \mathbf{x}^t + h\mathbf{v}^t \\ \frac{d\mathbf{v}^t}{dt} = \mathbf{a}^t &\Rightarrow \frac{\mathbf{v}^{t+h} - \mathbf{v}^t}{h} = \mathbf{a}^t &\Rightarrow \mathbf{v}^{t+h} = \mathbf{v}^t + h\mathbf{a}^t \end{aligned}$$

ODE

FDE

The first approximate solution of our problem

Finite Differences

- Line fitting (assuming $\frac{d\mathbf{x}^t}{dt} = \text{const}$ near \mathbf{x}^t)

$$\mathbf{x}^t = \mathbf{b}t + \mathbf{c}$$

$$\Rightarrow \frac{d\mathbf{x}^t}{dt} = \mathbf{b} \Rightarrow \mathbf{c} = \mathbf{x}^t - \frac{d\mathbf{x}^t}{dt}t$$

$$\mathbf{x}^{t+h} = \frac{d\mathbf{x}^t}{dt}(t+h) + \mathbf{x}^t - \frac{d\mathbf{x}^t}{dt}t$$

- Resulting in

$$\frac{d\mathbf{x}^t}{dt} = \frac{\mathbf{x}^{t+h} - \mathbf{x}^t}{h} + O(h)$$

Outline

- Introduction
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- Finite differences
- System of first-order ODEs
 - Explicit schemes
 - Predictor-corrector schemes
 - Implicit schemes
- ...

Explicit Euler

- Governing equations

$$\frac{d\mathbf{x}^t}{dt} = \mathbf{v}^t \quad \frac{d\mathbf{v}^t}{dt} = \mathbf{a}^t$$

- Initialization $\mathbf{x}^{t_0} = \mathbf{x}^{\text{init}}, \mathbf{v}^{t_0} = \mathbf{v}^{\text{init}}, \mathbf{a}^{t_0}, h$

- Explicit Euler update

$$\mathbf{x}^{t_0+h} = \mathbf{x}^{t_0} + h \frac{d\mathbf{x}^{t_0}}{dt} + O(h^2) = \mathbf{x}^{t_0} + h\mathbf{v}^{t_0} + O(h^2)$$

$$\mathbf{v}^{t_0+h} = \mathbf{v}^{t_0} + h \frac{d\mathbf{v}^{t_0}}{dt} + O(h^2) = \mathbf{v}^{t_0} + h\mathbf{a}^{t_0} + O(h^2)$$

Coupled Equations

- Position update depends on velocity
- Velocity update depends on position

$$\mathbf{x}^{t_0+h} = \mathbf{x}^{t_0} + h\mathbf{v}^{t_0}$$

$$\mathbf{v}^{t_0+h} = \mathbf{v}^{t_0} + h\mathbf{a}^{t_0}(\mathbf{x}^{t_0}, \mathbf{v}^{t_0})$$

$$\mathbf{x}^{t_0+2h} = \mathbf{x}^{t_0+h} + h\mathbf{v}^{t_0+h}$$

$$\mathbf{v}^{t_0+2h} = \mathbf{v}^{t_0+h} + h\mathbf{a}^{t_0+h}(\mathbf{x}^{t_0+h}, \mathbf{v}^{t_0+h})$$

Accuracy and Stability

- **Discretization error** is the difference between the solution of the ODE and the solution of the FDE
- The FDE is **consistent**, if the discretization error vanishes if the time step h approaches zero
- The FDE is **stable**, if previously introduced errors do not grow within a simulation step
- The FDE is **convergent**, if the solution of the FDE approaches the solution of the ODE

Accuracy and Stability

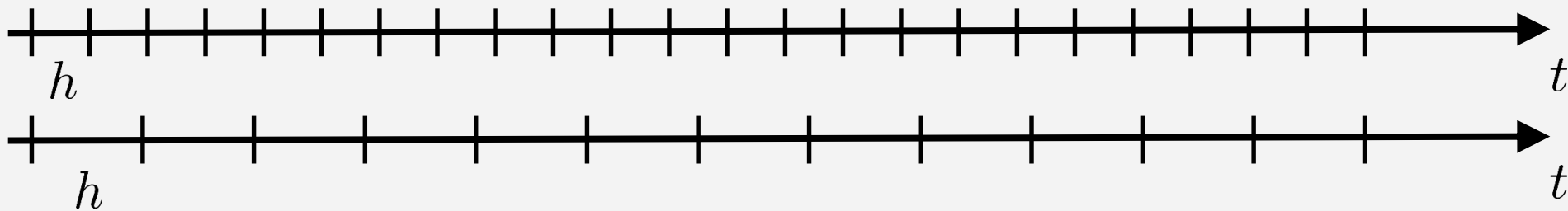
- Although the discretization error is diminished by smaller time steps in consistent schemes, the discretization error is introduced in each step of the FD scheme
- If previously introduced discretization errors are not amplified by the FD scheme, then it is stable
- Consistent and stable schemes are convergent

Stability

- If stability is influenced by the time step, the FD scheme is **conditionally stable**
- If the FD scheme is stable or unstable for arbitrary time steps, it is **unconditionally stable** or unstable
- ODE, FDE and the parameters influence the stability of a system
- Schemes with improved stability work with larger time steps \Rightarrow reduced overall computation time

Time Step

- Larger time steps result in less simulation steps and speed-up the overall computation time of a simulation



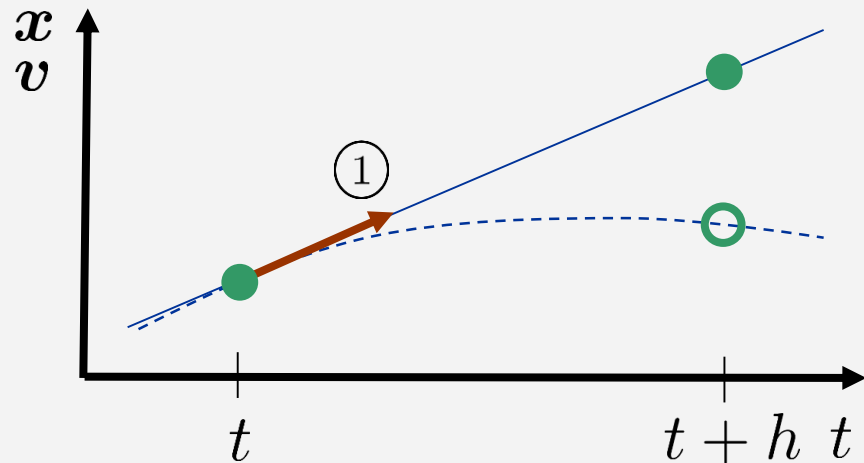
- Different FD schemes allow for different time steps
 - E.g. due to different error orders
 - Computing complexity also differs

Goal

- Stable scheme with maximized ratio between time step and computing complexity per simulation step

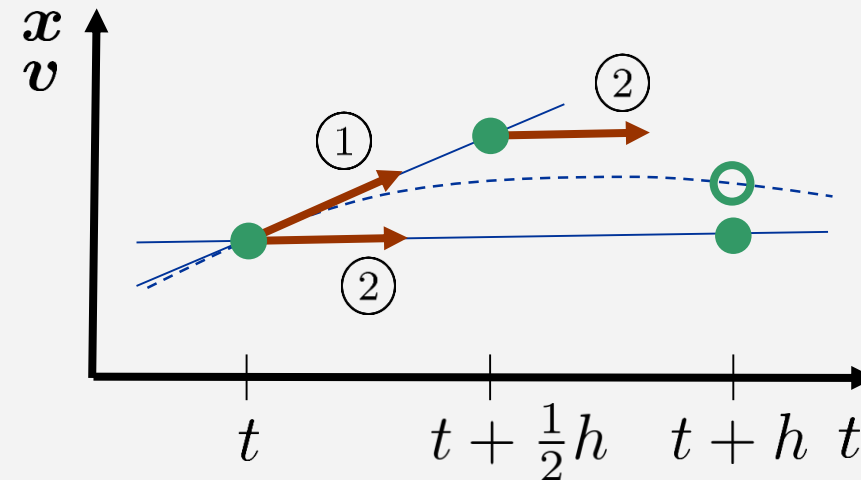
Second-Order Runge Kutta - Midpoint Method

Euler



- One derivative computation ①
- Discretization error $O(h^2)$

Midpoint



- Two derivative computations ① ②
- Requires intermediate positions and velocities
- Discretization error $O(h^3)$

Midpoint Implementation - Spring

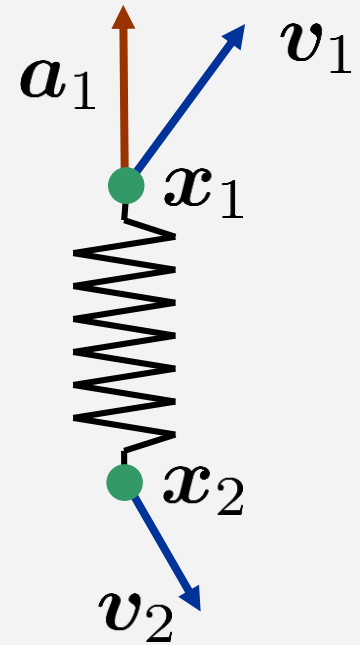
- Acceleration at time t : $\mathbf{a}_1^t(\mathbf{x}_1^t, \mathbf{x}_2^t, \mathbf{v}_1^t, \mathbf{v}_2^t)$
- Intermediate position and velocity at time $t + \frac{h}{2}$:

$$\mathbf{x}_1^* = \mathbf{x}_1^t + \frac{h}{2}\mathbf{v}_1^t \quad \mathbf{v}_1^* = \mathbf{v}_1^t + \frac{h}{2}\mathbf{a}_1^t \quad \mathbf{x}_2^* = \dots \quad \mathbf{v}_2^* = \dots$$

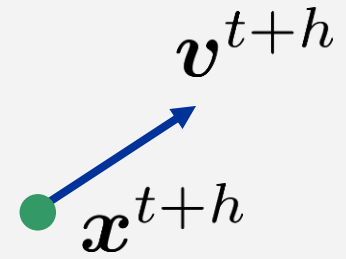
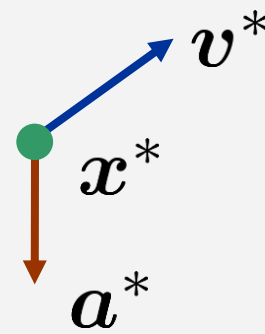
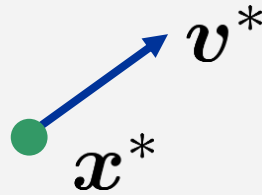
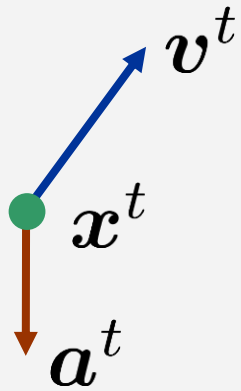
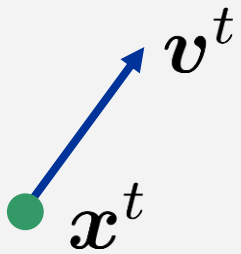
- Intermediate acceleration at time $t + \frac{h}{2}$ using intermediate positions and velocities: $\mathbf{a}_1^*(\mathbf{x}_1^*, \mathbf{x}_2^*, \mathbf{v}_1^*, \mathbf{v}_2^*)$

- Final position and velocity at time $t + h$

$$\mathbf{x}_1^{t+h} = \mathbf{x}_1^t + h\mathbf{v}_1^* \quad \mathbf{v}_1^{t+h} = \mathbf{v}_1^t + h\mathbf{a}_1^*$$



Midpoint Implementation



Current
state

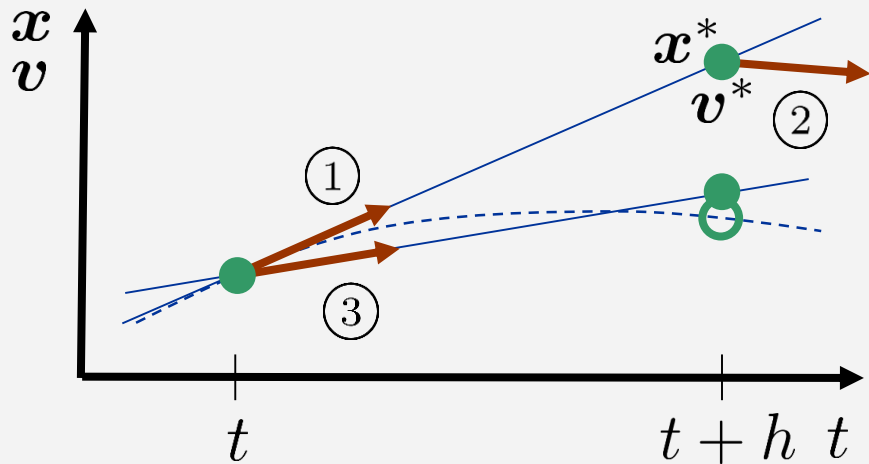
Compute **all**
accelerations

Compute **all**
predicted
pos. and vel.

Compute **all**
predicted
accelerations

Compute
all final
pos. and vel.

Second-Order Runge Kutta - Heun



① $\mathbf{v}^t, \mathbf{a}^t$

② $\mathbf{v}^*, \mathbf{a}^*$

③ $\frac{\mathbf{v}^t + \mathbf{v}^*}{2}, \frac{\mathbf{a}^t + \mathbf{a}^*}{2}$

– $\mathbf{x}^t, \mathbf{v}^t$

– $\mathbf{a}^t(\mathbf{x}^t, \mathbf{v}^t)$ ①

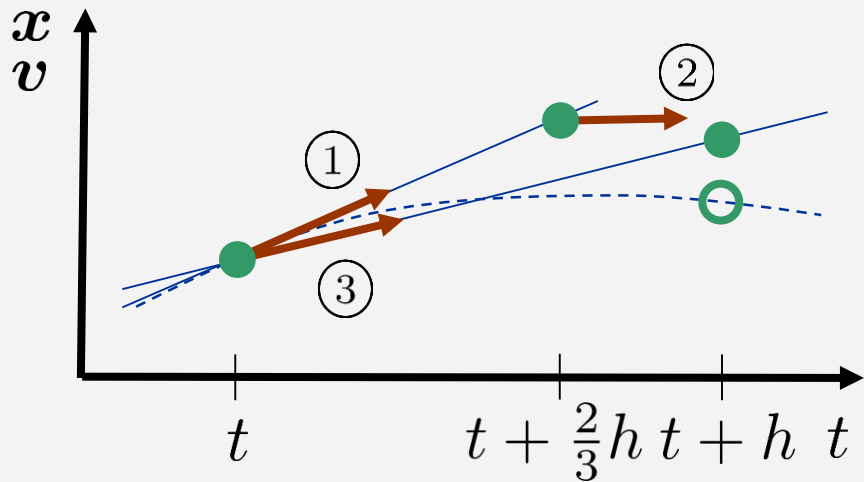
– $\mathbf{x}^* = \mathbf{x}^t + h\mathbf{v}^t \quad \mathbf{v}^* = \mathbf{v}^t + h\mathbf{a}^t$

– $\mathbf{a}^*(\mathbf{x}^*, \mathbf{v}^*)$ ②

– $\mathbf{x}^{t+h} = \mathbf{x}^t + h \frac{\mathbf{v}^t + \mathbf{v}^*}{2}$

$\mathbf{v}^{t+h} = \mathbf{v}^t + h \frac{\mathbf{a}^t + \mathbf{a}^*}{2}$ ③

Second-Order Runge Kutta - Ralston



– $\mathbf{x}^t, \mathbf{v}^t$

– $\mathbf{a}^t(\mathbf{x}^t, \mathbf{v}^t)$ ①

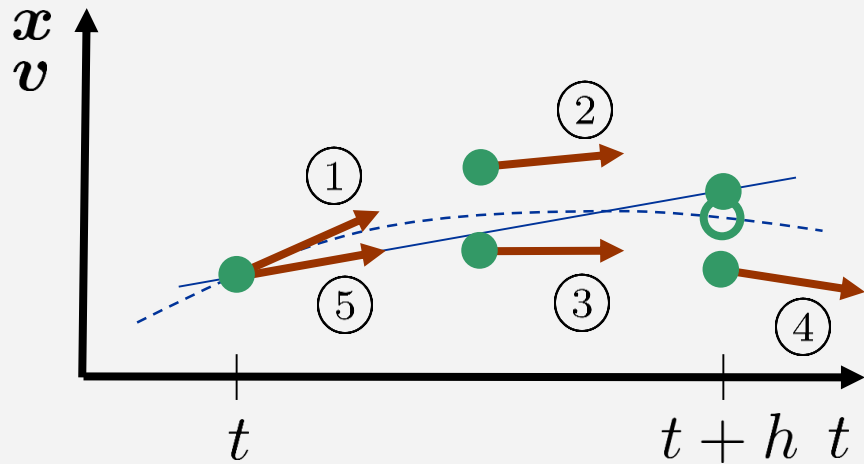
– $\mathbf{x}^* = \mathbf{x}^t + \frac{2}{3}h\mathbf{v}^t$ $\mathbf{v}^* = \mathbf{v}^t + \frac{2}{3}h\mathbf{a}^t$

– $\mathbf{a}^*(\mathbf{x}^*, \mathbf{v}^*)$ ②

– $\mathbf{x}^{t+h} = \mathbf{x}^t + \frac{1}{4}h\mathbf{v}^t + \frac{3}{4}h\mathbf{v}^*$

$\mathbf{v}^{t+h} = \mathbf{v}^t + \frac{1}{4}h\mathbf{a}^t + \frac{3}{4}h\mathbf{a}^*$ ③

Fourth-Order Runge Kutta - Classic

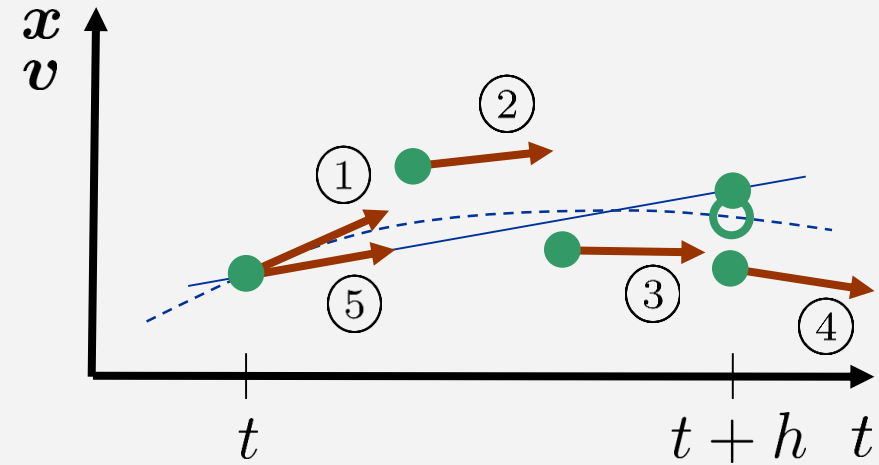


- Four derivative computations
- Discretization error $O(h^5)$

- $\mathbf{x}^t, \mathbf{v}^t$
- $\mathbf{a}^t(\mathbf{x}^t, \mathbf{v}^t)$ ①
- $\mathbf{x}^* = \mathbf{x}^t + \frac{h}{2}\mathbf{v}^t$ $\mathbf{v}^* = \mathbf{v}^t + \frac{h}{2}\mathbf{a}^t$
- $\mathbf{a}^*(\mathbf{x}^*, \mathbf{v}^*)$ ②
- $\mathbf{x}^{**} = \mathbf{x}^t + \frac{h}{2}\mathbf{v}^*$ $\mathbf{v}^{**} = \mathbf{v}^t + \frac{h}{2}\mathbf{a}^*$
- $\mathbf{a}^{**}(\mathbf{x}^{**}, \mathbf{v}^{**})$ ③
- $\mathbf{x}^{***} = \mathbf{x}^t + h\mathbf{v}^{**}$ $\mathbf{v}^{***} = \mathbf{v}^t + h\mathbf{a}^{**}$
- $\mathbf{a}^{***}(\mathbf{x}^{***}, \mathbf{v}^{***})$ ④
- $\mathbf{x}^{t+h} = \mathbf{x}^t + h \frac{\mathbf{v}^t + 2\mathbf{v}^* + 2\mathbf{v}^{**} + \mathbf{v}^{***}}{6}$
- $\mathbf{v}^{t+h} = \mathbf{v}^t + h \frac{\mathbf{a}^t + 2\mathbf{a}^* + 2\mathbf{a}^{**} + \mathbf{a}^{***}}{6}$ ⑤

Fourth-Order Runge Kutta - 3/8 Rule

- $\mathbf{x}^t, \mathbf{v}^t$
- $\mathbf{a}^t(\mathbf{x}^t, \mathbf{v}^t)$ ①
- $\mathbf{x}^* = \mathbf{x}^t + \frac{1}{3}h\mathbf{v}^t$ $\mathbf{v}^* = \mathbf{v}^t + \frac{1}{3}h\mathbf{a}^t$
- $\mathbf{a}^*(\mathbf{x}^*, \mathbf{v}^*)$ ②
- $\mathbf{x}^{**} = \mathbf{x}^t + \frac{2}{3}h(-\frac{1}{2}\mathbf{v}^t + \frac{3}{2}\mathbf{v}^*)$
- $\mathbf{v}^{**} = \mathbf{v}^t + \frac{2}{3}h(-\frac{1}{2}\mathbf{a}^t + \frac{3}{2}\mathbf{a}^*)$
- $\mathbf{a}^{**}(\mathbf{x}^{**}, \mathbf{v}^{**})$ ③
- $\mathbf{x}^{***} = \mathbf{x}^t + h(\mathbf{v}^t - \mathbf{v}^* + \mathbf{v}^{**})$
- $\mathbf{v}^{***} = \mathbf{v}^t + h(\mathbf{a}^t - \mathbf{a}^* + \mathbf{a}^{**})$
- $\mathbf{a}^{***}(\mathbf{x}^{***}, \mathbf{v}^{***})$ ④



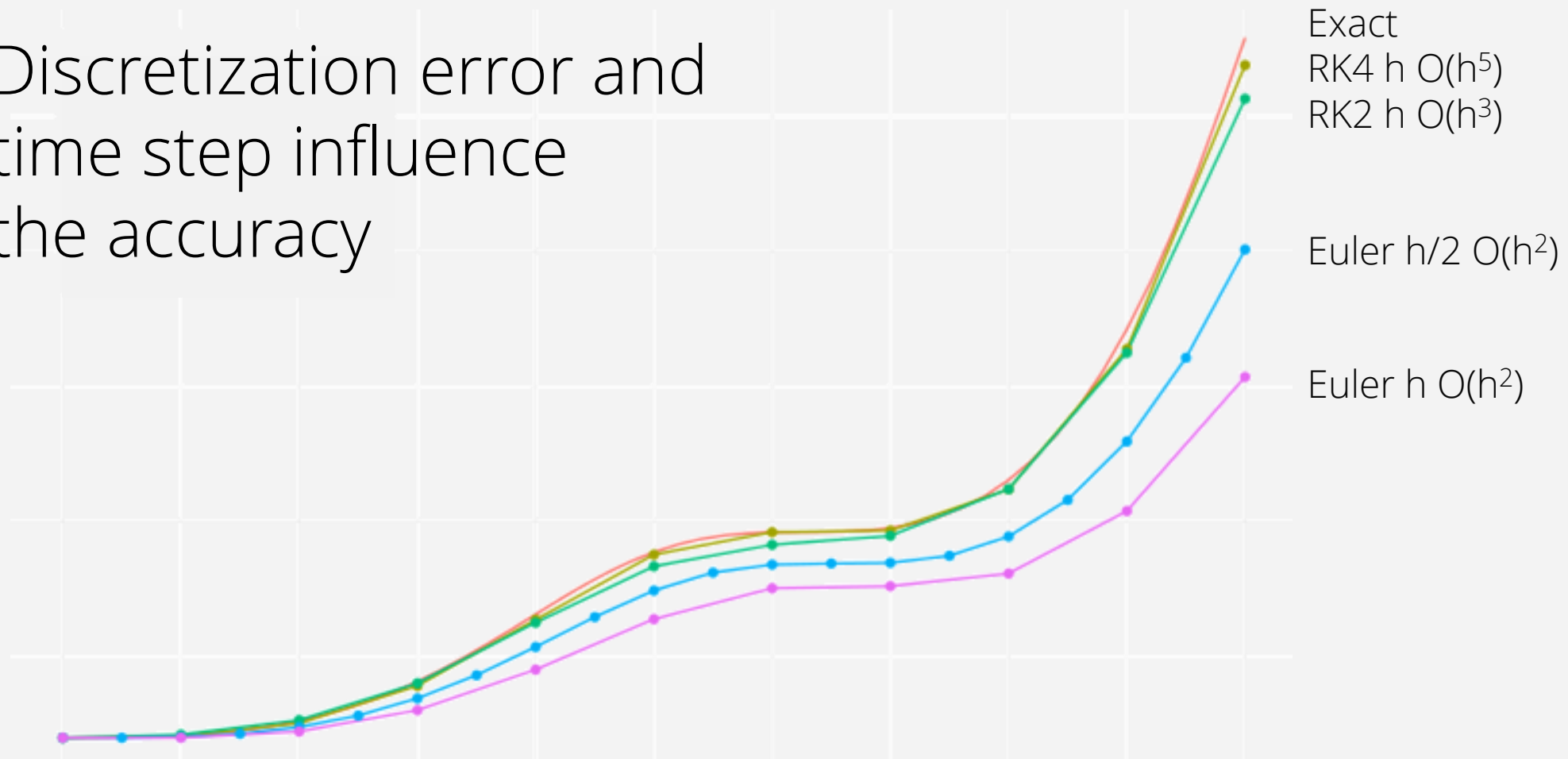
$$\mathbf{x}^{t+h} = \mathbf{x}^t + h \frac{\mathbf{v}^t + 3\mathbf{v}^* + 3\mathbf{v}^{**} + \mathbf{v}^{***}}{8}$$

$$\mathbf{v}^{t+h} = \mathbf{v}^t + h \frac{\mathbf{a}^t + 3\mathbf{a}^* + 3\mathbf{a}^{**} + \mathbf{a}^{***}}{8}$$

⑤

Accuracy

- Discretization error and time step influence the accuracy



Wikipedia: Runge-Kutta-Verfahren

Performance

- Computation dominated by derivatives, actually only by the accelerations $\mathbf{a}^t, \mathbf{a}^*, \mathbf{a}^{**}, \mathbf{a}^{***}$
- RK4 is four times as expensive as Euler
- RK2 is two times as expensive as Euler
- RK4 is more accurate than RK2 which is more accurate than Euler. Error: $O(h^5) < O(h^3) < O(h^2)$
- RK4 allows larger time steps than RK2 which allows larger times steps than Euler

Performance

- If, e.g., RK4 runs with a time step four times larger than Euler, the overall computation time is the same
 - Comparison: RK4 : Euler
 - Time per simulation step: 4 : 1
 - Simulation steps: 1 : 4
 - Overall computation time: 1 : 1

Accelerations

- $\mathbf{a}^t, \mathbf{a}^*, \mathbf{a}^{**}, \mathbf{a}^{***}$ can be very expensive to compute
- E.g., if the accelerations consider contact forces, collision detection has to be performed four times for different sets of positions $\mathbf{x}^t, \mathbf{x}^*, \mathbf{x}^{**}, \mathbf{x}^{***}$

Simulation in Computer Graphics

Particle Motion 2

Matthias Teschner



Outline

- Introduction
- Particle motion
- Finite differences
- System of first-order ODEs
 - Explicit schemes
 - Predictor-corrector schemes
 - Implicit schemes
- ...

Explicit Adams-Bashforth

- Current and previous accelerations (multistep)
 - Two acceleration computations per step
 - Previous accelerations have to be stored

$$\mathbf{v}^* = \mathbf{v}^t + \frac{h}{2}(3\mathbf{a}^t - \mathbf{a}^{t-h}) + O(h^3)$$

$$\mathbf{v}^* = \mathbf{v}^t + \frac{h}{12}(23\mathbf{a}^t - 16\mathbf{a}^{t-h} + 5\mathbf{a}^{t-2h}) + O(h^4)$$

$$\mathbf{v}^* = \mathbf{v}^t + \frac{h}{24}(55\mathbf{a}^t - 59\mathbf{a}^{t-h} + 37\mathbf{a}^{t-2h} - 9\mathbf{a}^{t-3h}) + O(h^5)$$

$$\mathbf{v}^* = \mathbf{v}^t + \frac{h}{720}(1901\mathbf{a}^t - 2774\mathbf{a}^{t-h} + 2616\mathbf{a}^{t-2h} - 1274\mathbf{a}^{t-3h} + 251\mathbf{a}^{t-4h}) + O(h^6)$$

$$\mathbf{x}^* = \mathbf{x}^t + \frac{h}{2}(3\mathbf{v}^t - \mathbf{v}^{t-h}) + O(h^3)$$

...

Implicit Adams-Moulton

- Next, current and previous accelerations (multistep)

$$\mathbf{v}^{t+h} = \mathbf{v}^t + \frac{h}{2}(\mathbf{a}^* + \mathbf{a}^t) + O(h^3)$$

$$\mathbf{v}^{t+h} = \mathbf{v}^t + \frac{h}{12}(5\mathbf{a}^* + 8\mathbf{a}^t - \mathbf{a}^{t-h}) + O(h^4)$$

$$\mathbf{v}^{t+h} = \mathbf{v}^t + \frac{h}{24}(9\mathbf{a}^* + 19\mathbf{a}^t - 5\mathbf{a}^{t-h} + \mathbf{a}^{t-2h}) + O(h^5)$$

$$\mathbf{v}^{t+h} = \mathbf{v}^t + \frac{h}{720}(251\mathbf{a}^* + 646\mathbf{a}^t - 264\mathbf{a}^{t-h} + 106\mathbf{a}^{t-2h} - 19\mathbf{a}^{t-3h}) + O(h^6)$$

$$\mathbf{x}^{t+h} = \mathbf{x}^t + \frac{h}{2}(\mathbf{v}^* + \mathbf{v}^t) + O(h^3)$$

...

A Predictor-Corrector Example

- Initialization

$$\mathbf{x}^t, \mathbf{v}^t, \mathbf{a}^t \quad \mathbf{x}^{t-h} = \mathbf{x} - h\mathbf{v}^t, \mathbf{v}^{t-h} = \mathbf{v} - h\mathbf{a}^t, \mathbf{a}^{t-h}$$

- Prediction

$$\mathbf{x}^* = \mathbf{x}^t + \frac{h}{2}(3\mathbf{v}^t - \mathbf{v}^{t-h}) \quad \mathbf{v}^* = \mathbf{v}^t + \frac{h}{2}(3\mathbf{a}^t - \mathbf{a}^{t-h}) \quad \mathbf{a}^*$$

Accelerations at predicted positions using predicted velocities

- Correction

$$\mathbf{x}^{t+h} = \mathbf{x}^t + \frac{h}{2}(\mathbf{v}^* + \mathbf{v}^t) \quad \mathbf{v}^{t+h} = \mathbf{v}^t + \frac{h}{2}(\mathbf{a}^* + \mathbf{a}^t) \quad \mathbf{a}^{t+h}$$

Discussion

- Two accelerations
- Improved accuracy, larger time steps
 - Not necessarily true for discontinuous functions, e.g., in case of contact handling
- Initialization of previous steps
- Iterative correction steps possible

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Explicit vs. Implicit Schemes

- Explicit Euler

$$\mathbf{x}^{t+h} = \mathbf{x}^t + h\mathbf{v}^t$$

$$\mathbf{v}^{t+h} = \mathbf{v}^t + h\mathbf{a}^t$$

- One unknown per equation
- Direct computation of unknowns
- Non-linear equations do not affect the approach
- Non-analytical, procedural forces can be handled

- Implicit Euler

$$\mathbf{x}^{t+h} = \mathbf{x}^t + h\mathbf{v}^{t+h}$$

$$\mathbf{v}^{t+h} = \mathbf{v}^t + h\mathbf{a}^{t+h}$$

- System of algebraic equations
- Simultaneous computation of unknowns
- Solution of a **linear system**
- **Linearization** of non-linear equations

Implicit Schemes

- Challenge
 - Solving a linear system
 - Implementation
- Benefit
 - Largely improved stability
- Issue
 - Reduced accuracy
 - Discretization error plus linearization error plus approximate solution of a linear system

Implicit Schemes – Example Overview

- Linearization of accelerations

$$\mathbf{v}^{t+h} = \mathbf{v}^t + h\mathbf{a}^{t+h}(\mathbf{x}^{t+h})$$

Here, accelerations depend only on positions.

$$\mathbf{v}^{t+h} = \mathbf{v}^t + h\mathbf{a}^{t+h}(\mathbf{x}^t + h\mathbf{v}^{t+h})$$

$$\mathbf{v}^{t+h} = \mathbf{v}^t + h(\mathbf{a}^t(\mathbf{x}^t) + \mathbf{J}^t h\mathbf{v}^{t+h})$$

\mathbf{J} is a 3x3 Jacobi matrix. $h\cdot\mathbf{v}$ is a small displacement. $\mathbf{a}(\mathbf{x}) + \mathbf{J}\cdot h\cdot\mathbf{v}$ is an approximation of the acceleration at position $\mathbf{x} + h\cdot\mathbf{v}$.

- Linear system with unknown velocities

$$(\mathbf{I} - h^2\mathbf{J}^t)\mathbf{v}^{t+h} = \mathbf{v}^t + h\mathbf{a}^t$$

- Position update

$$\mathbf{x}^{t+h} = \mathbf{x}^t + h\mathbf{v}^{t+h}$$

Linearization

– $f_{x+\Delta x} = f_x + \frac{\partial f_x}{\partial x} \Delta x + O((\Delta x)^2)$ f: 1D field of scalar values

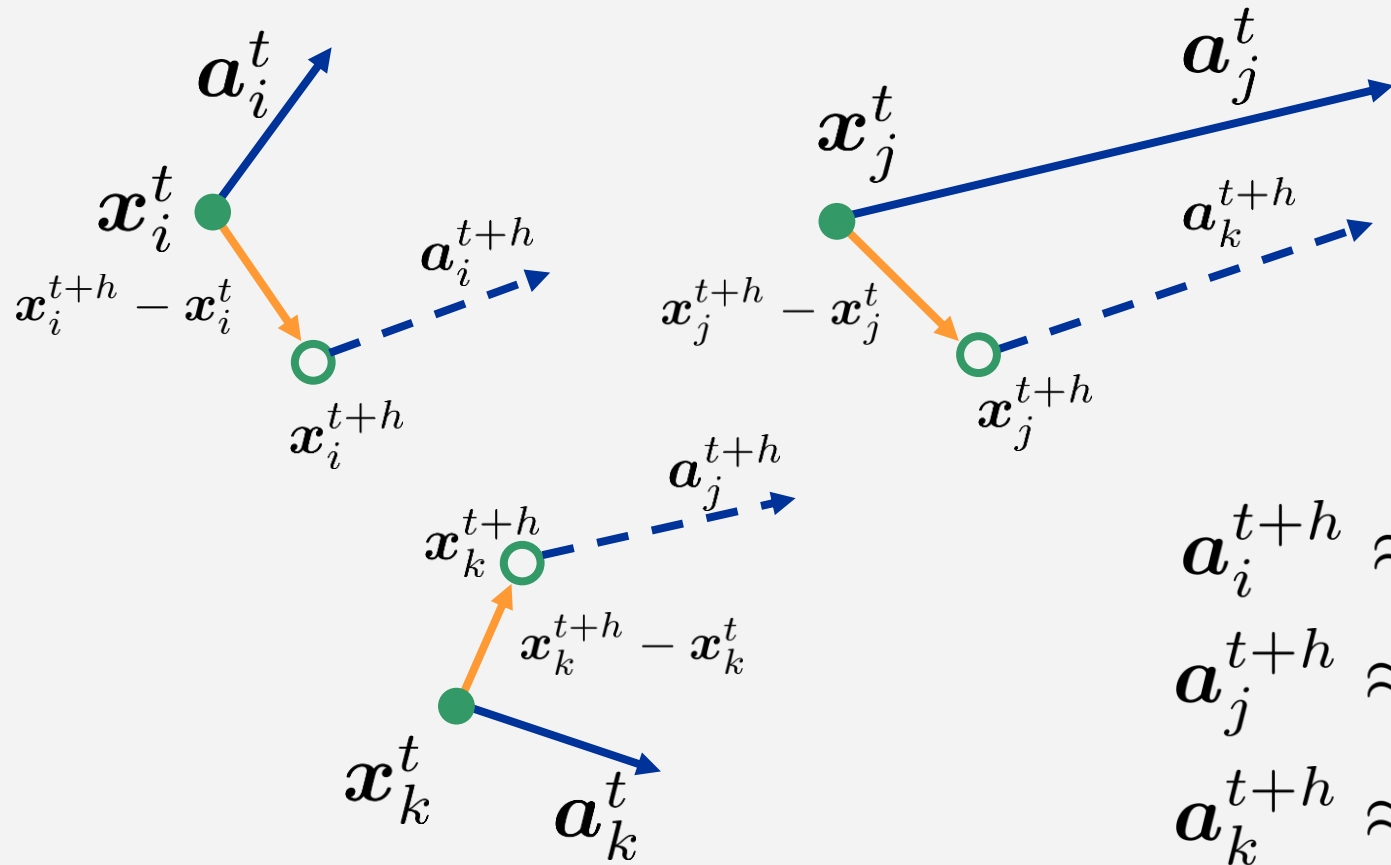
– $f_{\mathbf{x}+\Delta \mathbf{x}} = f_{\mathbf{x}} + \nabla f_{\mathbf{x}} \cdot \Delta \mathbf{x} + O(\|\Delta \mathbf{x}\|^2)$ f: 3D field of scalar values

$$\nabla f_{\mathbf{x}} = \left(\frac{\partial f_{\mathbf{x}}}{\partial x_1}, \frac{\partial f_{\mathbf{x}}}{\partial x_2}, \dots, \frac{\partial f_{\mathbf{x}}}{\partial x_n} \right)^T \quad \text{Gradient}$$

– $\mathbf{a}_{\mathbf{x}+\Delta \mathbf{x}} = \mathbf{a}_{\mathbf{x}} + \mathbf{J}_{\mathbf{x}} \Delta \mathbf{x} + O(\|\Delta \mathbf{x}\|^2)$ a: 3D field of 3D values

$$\mathbf{J}_{\mathbf{x}} = \begin{pmatrix} \frac{\partial a_{x_x}}{\partial x_x} & \frac{\partial a_{x_x}}{\partial x_y} & \frac{\partial a_{x_x}}{\partial x_z} \\ \frac{\partial a_{x_y}}{\partial x_x} & \frac{\partial a_{x_y}}{\partial x_y} & \frac{\partial a_{x_y}}{\partial x_z} \\ \frac{\partial a_{x_z}}{\partial x_x} & \frac{\partial a_{x_z}}{\partial x_y} & \frac{\partial a_{x_z}}{\partial x_z} \end{pmatrix} \quad \text{Jacobi matrix}$$
$$\mathbf{a}_{\mathbf{x}} = \begin{pmatrix} a_{x_x} \\ a_{x_y} \\ a_{x_z} \end{pmatrix} \quad \mathbf{x} = \begin{pmatrix} x_x \\ x_y \\ x_z \end{pmatrix}$$

Jacobi Matrix - Application



$$\mathbf{a}_i^{t+h} \approx \mathbf{a}_i^t + \mathbf{J}_{\mathbf{x}_i^t} (\mathbf{x}_i^{t+h} - \mathbf{x}_i^t)$$

$$\mathbf{a}_j^{t+h} \approx \mathbf{a}_j^t + \mathbf{J}_{\mathbf{x}_j^t} (\mathbf{x}_j^{t+h} - \mathbf{x}_j^t)$$

$$\mathbf{a}_k^{t+h} \approx \mathbf{a}_k^t + \mathbf{J}_{\mathbf{x}_k^t} (\mathbf{x}_k^{t+h} - \mathbf{x}_k^t)$$

Linearization

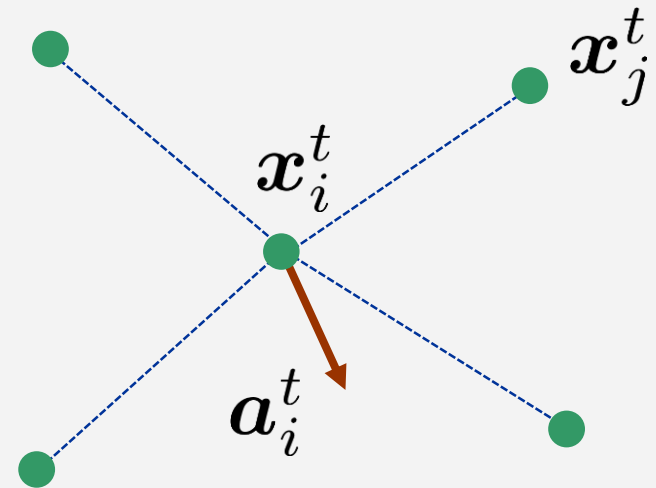
- Approximation of the acceleration at position \mathbf{x}^{t+h} using the acceleration at position \mathbf{x}^t , the Jacobi matrix \mathbf{J}^t of the acceleration at position \mathbf{x}^t and the small displacement $\mathbf{x}^{t+h} - \mathbf{x}^t = h\mathbf{v}^{t+h}$:

$$\mathbf{a}^{t+h}(\mathbf{x}^{t+h}) = \mathbf{a}^{t+h}(\mathbf{x}^t + h\mathbf{v}^{t+h}) \approx \mathbf{a}^t(\mathbf{x}^t) + \mathbf{J}^t h\mathbf{v}^{t+h}$$

- Equation $\mathbf{v}^{t+h} = \mathbf{v}^t + h\mathbf{a}^{t+h}(\mathbf{x}^{t+h})$ with unknown velocities and positions can be rewritten with unknown velocities only: $(\mathbf{I} - h^2\mathbf{J}^t)\mathbf{v}^{t+h} = \mathbf{v}^t + h\mathbf{a}^t$

Particle System

- Set of particles with, e.g., interconnecting springs
- Force at a particle depends on particle and its neighbors
- E.g. $\mathbf{a}_i^t = \frac{1}{m_i} \sum_j k_{ij} \frac{|\mathbf{x}_j^t - \mathbf{x}_i^t| - L_{ij}}{L_{ij}} \frac{\mathbf{x}_j^t - \mathbf{x}_i^t}{|\mathbf{x}_j^t - \mathbf{x}_i^t|}$
 - Position \mathbf{x}_i^t , acc. \mathbf{a}_i^t and mass m_i of particle i at time t
 - Rest distance L_{ij} and stiffness k_{ij} between particles i and j



Notation

$$\mathbf{x}^t = \begin{pmatrix} x_{1,x}^t \\ x_{1,y}^t \\ x_{1,z}^t \\ x_{2,x}^t \\ x_{2,y}^t \\ x_{2,z}^t \\ \vdots \\ x_{n,x}^t \\ x_{n,y}^t \\ x_{n,z}^t \end{pmatrix}$$

$$\mathbf{v}^t = \begin{pmatrix} v_{1,x}^t \\ v_{1,y}^t \\ v_{1,z}^t \\ v_{2,x}^t \\ v_{2,y}^t \\ v_{2,z}^t \\ \vdots \\ v_{n,x}^t \\ v_{n,y}^t \\ v_{n,z}^t \end{pmatrix}$$

$$\mathbf{a}^t = \begin{pmatrix} a_{1,x}^t \\ a_{1,y}^t \\ a_{1,z}^t \\ a_{2,x}^t \\ a_{2,y}^t \\ a_{2,z}^t \\ \vdots \\ a_{n,x}^t \\ a_{n,y}^t \\ a_{n,z}^t \end{pmatrix}$$

Linear System – Implicit Euler

- Linear system for n particles
 - $(\mathbf{I}_{3n \times 3n} - h^2 \mathbf{J}^t) \mathbf{v}^{t+h} = \mathbf{v}^t + h \mathbf{a}^t$
- Jacobian
 - $\mathbf{J}^t \in \mathbb{R}^{3n \times 3n}$
 - Spatial derivatives of all accelerations with respect to all positions

Jacobian - Example

– E.g., $\mathbf{a}_i^t = \frac{1}{m_i} \frac{k_{ij}}{L_{ij}} (|\mathbf{x}_j^t - \mathbf{x}_i^t| - L_{ij}) \frac{\mathbf{x}_j^t - \mathbf{x}_i^t}{|\mathbf{x}_j^t - \mathbf{x}_i^t|} = \frac{1}{m_i} \frac{k_{ij}}{L_{ij}} \left(\mathbf{x}_j^t - \mathbf{x}_i^t - L_{ij} \frac{\mathbf{x}_j^t - \mathbf{x}_i^t}{|\mathbf{x}_j^t - \mathbf{x}_i^t|} \right)$
depends on two positions \mathbf{x}_i^t and \mathbf{x}_j^t

$$\mathbf{J}_{i,i}^t = \frac{\partial \mathbf{a}_i^t}{\partial \mathbf{x}_i^t} = \begin{pmatrix} \frac{\partial a_{i,x}}{\partial x_{i,x}} & \frac{\partial a_{i,x}}{\partial x_{i,y}} & \frac{\partial a_{i,x}}{\partial x_{i,z}} \\ \frac{\partial a_{i,y}}{\partial x_{i,x}} & \frac{\partial a_{i,y}}{\partial x_{i,y}} & \frac{\partial a_{i,y}}{\partial x_{i,z}} \\ \frac{\partial a_{i,z}}{\partial x_{i,x}} & \frac{\partial a_{i,z}}{\partial x_{i,y}} & \frac{\partial a_{i,z}}{\partial x_{i,z}} \end{pmatrix}$$

$$= \frac{\partial}{\partial \mathbf{x}_i^t} \frac{1}{m_i} \frac{k_{ij}}{L_{ij}} \left(\mathbf{x}_j^t - \mathbf{x}_i^t - L_{ij} \frac{\mathbf{x}_j^t - \mathbf{x}_i^t}{|\mathbf{x}_j^t - \mathbf{x}_i^t|} \right)$$

$$= \frac{1}{m_i} \frac{k_{ij}}{L_{ij}} \left(-\mathbf{I} + \frac{L_{ij}}{|\mathbf{x}_j^t - \mathbf{x}_i^t|} \left(\mathbf{I} - \frac{1}{|\mathbf{x}_j^t - \mathbf{x}_i^t|^2} (\mathbf{x}_j^t - \mathbf{x}_i^t)(\mathbf{x}_j^t - \mathbf{x}_i^t)^T \right) \right)$$

$$\mathbf{J}_{i,j}^t = -\frac{m_i}{m_j} \mathbf{J}_{i,i}^t$$

Mueller et al., Real-time Physics. SIGGRAPH 2008.

Jacobian

- $\mathbf{J}^t \in \mathbb{R}^{3n \times 3n}$ is built from 3x3 matrices $\mathbf{J}_{i,j}^t \in \mathbb{R}^{3 \times 3}$
- If position \mathbf{x}_j^t influences acceleration \mathbf{a}_i^t , then $\mathbf{J}_{i,j}^t \neq \mathbf{0}$
- Otherwise, $\mathbf{J}_{i,j}^t = \mathbf{0}$

$$\mathbf{J}^t = \begin{pmatrix} \mathbf{J}_{i,i}^t & & & \mathbf{J}_{i,j}^t & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \end{pmatrix}$$

Solver

$$- \underbrace{(\mathbf{I}_{3n \times 3n} - h^2 \mathbf{J}^t)}_A \mathbf{v}^{t+h} = \underbrace{\mathbf{v}^t + h\mathbf{a}^t}_s$$

– Iterative. Start with a guess, e.g. $\mathbf{v}^0 = \mathbf{v}^t$

– Iterative updates $\mathbf{v}^0 \rightarrow \mathbf{v}^1 \rightarrow \dots \rightarrow \mathbf{v}^l$

– Result $\mathbf{v}^{t+h} = \mathbf{v}^l$

Here, superscript indicates the iteration.

Solver – Conjugate Gradient

$$l = 0$$

$$\mathbf{d}^l = \mathbf{r}^l = \mathbf{s} - \mathbf{A}\mathbf{v}^l$$

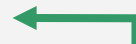
$$\alpha^l = \frac{\mathbf{r}^l \cdot \mathbf{r}^l}{\mathbf{d}^l \cdot (\mathbf{A}\mathbf{d}^l)}$$

$$\mathbf{v}^{l+1} = \mathbf{v}^l + \alpha^l \mathbf{d}^l$$

$$\mathbf{r}^{l+1} = \mathbf{r}^l - \alpha^l \mathbf{A}\mathbf{d}^l$$

$$\mathbf{d}^{l+1} = \mathbf{r}^{l+1} + \frac{\mathbf{r}^{l+1} \cdot \mathbf{r}^{l+1}}{\mathbf{r}^l \cdot \mathbf{r}^l} \mathbf{d}^l$$

$$l = l + 1$$



Scaling factor for the solution update.

Update of the solution with a scaled direction.

Residual. Exit loop, when sufficiently small.

Direction for the solution update.

Iteration count.

Solver – Conjugate Gradient

- Works (converges) for symmetric, positive-definite matrices
- Exact solution of an $n \times n$ system in n steps
- Frequently used for deformable objects
- Typically used with a fixed iteration count, e.g. 3-5

Solver - Jacobi

- $\mathbf{v}^{l+1} = \mathbf{v}^l + \omega \mathbf{D}^{-1}(\mathbf{s} - \mathbf{A}\mathbf{v}^l)$
- \mathbf{D} diagonal elements of \mathbf{A}
- ω determines convergence and convergence rate
- $0 \leq \omega \leq 2$, in practical settings typically $\omega = 0.5$
- Per-component update

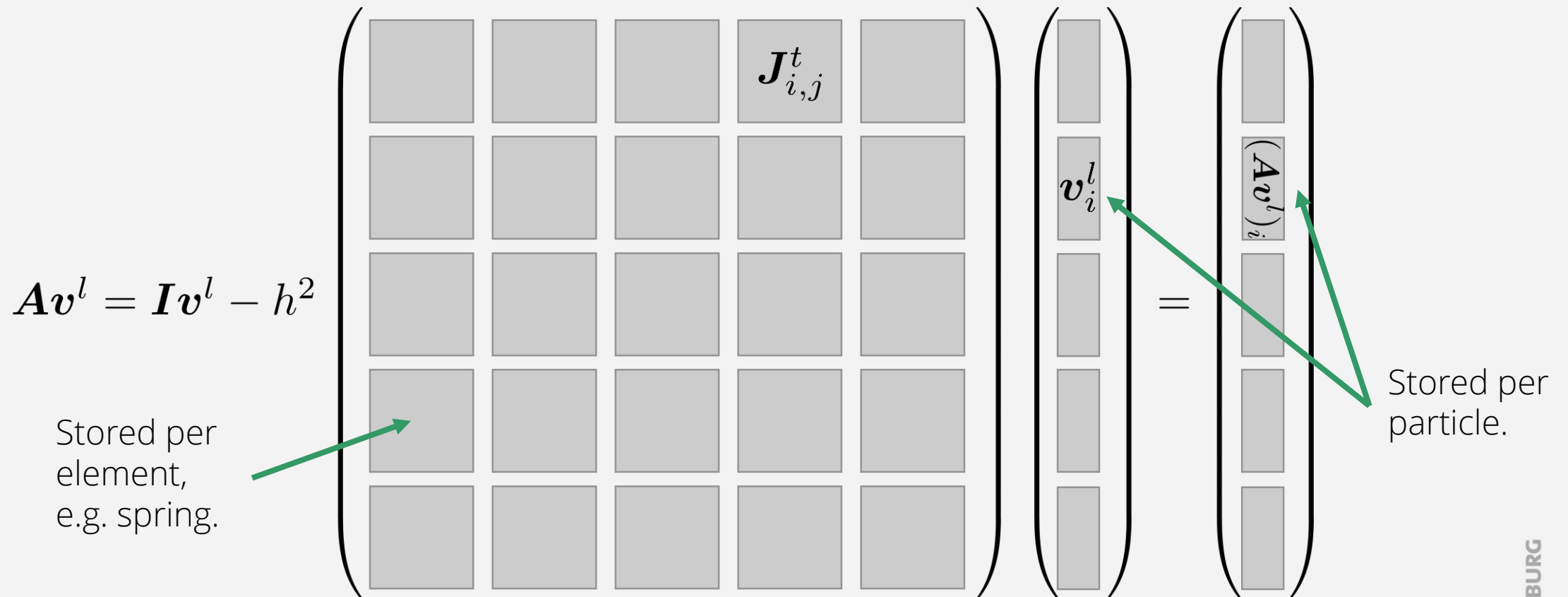
$$\begin{aligned}v_i^{l+1} &= (1 - \omega)v_i^l + \frac{\omega}{A_{ii}}(s_i - \sum_{j \neq i} A_{ij}v_j^l) \\ &= (1 - \omega)v_i^l + \frac{\omega}{A_{ii}}(s_i - (\mathbf{A}\mathbf{v}^l)_i + A_{ii}v_i^l) \\ &= v_i^l + \frac{\omega}{A_{ii}}(s_i - (\mathbf{A}\mathbf{v}^l)_i)\end{aligned}$$

Solver - Implementation

- $\mathbf{A} = \mathbf{I}_{3n \times 3n} - h^2 \mathbf{J}^t$ is not explicitly built or stored
- $\mathbf{s} = \mathbf{v}^t + h \mathbf{a}^t$ is not explicitly built or stored
- Instead
 - Per-particle information is stored at particles, e.g. \mathbf{s}_i
 - Per-element information is stored at elements, e.g. $\mathbf{J}_{i,i}^t$ for an elastic spring between i and j ,
 $\mathbf{J}_{i,j}^t = \mathbf{J}_{j,i}^t = -\frac{m_i}{m_j} \mathbf{J}_{i,i}^t$ and $\mathbf{J}_{j,j}^t = \frac{m_i}{m_j} \mathbf{J}_{i,i}^t$ can be reconstructed
 - Matrix-free implementation of solver steps

Solver - Implementation

- $\mathbf{A}v^l$ is computed and stored per particle



Solver - Implementation

- $\mathbf{A}\mathbf{v}^l$ is computed by iterating over elements
- E.g., spring connects particles j and k

For each particle:

$$(\mathbf{A}\mathbf{v}^l)_i = \mathbf{v}_i^l$$

For each spring:

$$(\mathbf{A}\mathbf{v}^l)_{j+} = -h^2 \mathbf{J}_{j,j}^t \mathbf{v}_j^l + h^2 \underbrace{\frac{m_j}{m_k} \mathbf{J}_{j,j}^t}_{-\mathbf{J}_{j,k}^t} \mathbf{v}_k^l$$

$$(\mathbf{A}\mathbf{v}^l)_{k+} = -h^2 \underbrace{\frac{m_j}{m_k} \mathbf{J}_{j,j}^t}_{\mathbf{J}_{k,k}^t} \mathbf{v}_k^l + h^2 \underbrace{\frac{m_j}{m_k} \mathbf{J}_{j,j}^t}_{-\mathbf{J}_{k,j}^t} \mathbf{v}_j^l$$

$$\mathbf{A}\mathbf{v}^l = \mathbf{I}\mathbf{v}^l - h^2$$

Solver - Discussion

- Jacobi vs. Conjugate Gradient CG:
 - CG converges faster
 - Jacobi is good-natured, e.g. in case of clamping intermediate solutions to implement constraints
- Implementation, e.g., in a particle-spring model
 - Matrix-free
 - All solver information is stored at particles and springs
 - All solver steps are realized by iterating over particles and springs

Implicit Schemes – Summary

- Implicit Euler

$$\mathbf{v}^{t+h} = \mathbf{v}^t + h\mathbf{a}^{t+h}(\mathbf{x}^{t+h})$$

- Linearization

$$\mathbf{v}^{t+h} = \mathbf{v}^t + h(\mathbf{a}^t(\mathbf{x}^t) + \mathbf{J}^t h\mathbf{v}^{t+h})$$

- Solve a linear system for velocities

$$(\mathbf{I} - h^2\mathbf{J}^t)\mathbf{v}^{t+h} = \mathbf{v}^t + h\mathbf{a}^t$$

- Update positions according to implicit Euler

$$\mathbf{x}^{t+h} = \mathbf{x}^t + h\mathbf{v}^{t+h}$$

Semi-implicit Euler (Euler-Cromer)

- Explicit Euler for the velocity update

$$\mathbf{v}^{t+h} = \mathbf{v}^t + h\mathbf{a}^t$$

- Implicit Euler for the position update

$$\mathbf{x}^{t+h} = \mathbf{x}^t + h\mathbf{v}^{t+h}$$

- No linear system

Simulation in Computer Graphics

Particle Motion 3

Matthias Teschner

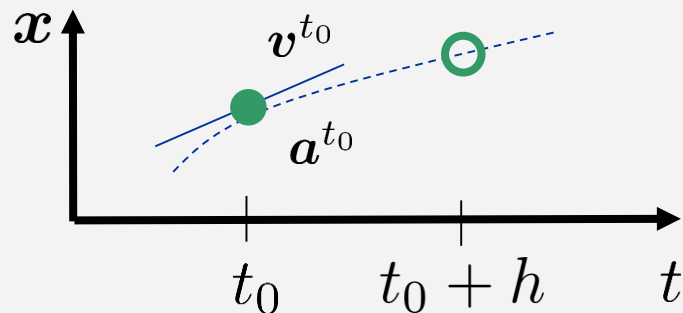


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- Introduction
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- Second-order ODE
- Performance
- Discussion

Initial Value Problem of Second Order

- Function \mathbf{x}^t represents the particle motion
- Second-order differential equation is given
$$\frac{d^2 \mathbf{x}^t}{dt^2} = \mathbf{a}^t$$
- Initial values \mathbf{x}^{t_0} and \mathbf{v}^{t_0} are given
- How to estimate \mathbf{x}^{t_0+h} ?



Motivation

- Schemes for coupled first-order ODEs update \mathbf{x} and \mathbf{v} simultaneously
- Schemes for second-order ODEs update \mathbf{x} , but not necessarily \mathbf{v}

Verlet

- Taylor approximations of \mathbf{x}^{t+h} and \mathbf{x}^{t-h}

$$\mathbf{x}^{t+h} = \mathbf{x}^t + h\mathbf{v}^t + \frac{h^2}{2}\mathbf{a}^t + \frac{h^3}{6}\frac{d^3\mathbf{x}^t}{dt^3} + O(h^4)$$

$$\mathbf{x}^{t-h} = \mathbf{x}^t - h\mathbf{v}^t + \frac{h^2}{2}\mathbf{a}^t - \frac{h^3}{6}\frac{d^3\mathbf{x}^t}{dt^3} + O(h^4)$$

- Adding both approximations

$$\mathbf{x}^{t+h} = 2\mathbf{x}^t - \mathbf{x}^{t-h} + h^2\mathbf{a}^t + O(h^4)$$

$$\mathbf{x}^{t+h} = \mathbf{x}^t + h\frac{\mathbf{x}^t - \mathbf{x}^{t-h}}{h} + h^2\mathbf{a}^t + O(h^4)$$

Verlet - Discussion

- One acceleration computation per step
 - Same computation cost as explicit Euler
- Discretization error of order 4
 - More accurate than explicit Euler
- Larger time step and improved performance compared to explicit Euler

Verlet - Discussion

- Velocity representation not necessarily required
- But:
 - Velocity typically used for collision handling and damping
 - E.g. $\mathbf{v}^{t+h} = \frac{\mathbf{x}^{t+h} - \mathbf{x}^t}{h} + O(h)$

Leap-Frog

$$\mathbf{x}^{t+h} = \mathbf{x}^t + h\mathbf{v}^{t+\frac{h}{2}}$$

$$\mathbf{v}^{t+\frac{3h}{2}} = \mathbf{v}^{t+\frac{h}{2}} + h\mathbf{a}^{t+h}$$

– Implementation, e.g.

$$\mathbf{v}^{t+\frac{h}{2}} = \frac{\mathbf{x}^{t+h} - \mathbf{x}^t}{h} \quad \mathbf{v}^{t-\frac{h}{2}} = \frac{\mathbf{x}^t - \mathbf{x}^{t-h}}{h}$$

$$\Rightarrow (\mathbf{v}^{t+\frac{h}{2}} - \mathbf{v}^{t-\frac{h}{2}})h = \mathbf{x}^{t+h} - \mathbf{x}^t - \mathbf{x}^t + \mathbf{x}^{t-h}$$

$$\Rightarrow \mathbf{a}^t h^2 = \mathbf{x}^{t+h} - \mathbf{x}^t - \mathbf{x}^t + \mathbf{x}^{t-h}$$

$$\Rightarrow \mathbf{x}^{t+h} = 2\mathbf{x}^t - \mathbf{x}^{t-h} + \mathbf{a}^t h^2 \quad \text{Verlet}$$

Velocity Verlet

- Same accuracy for position and velocity

$$\mathbf{x}^{t+h} = \mathbf{x}^t + h\mathbf{v}^t + \frac{h^2}{2}\mathbf{a}^t + O(h^3)$$

$$\mathbf{v}^{t+h} = \mathbf{v}^t + \frac{h}{2}(\mathbf{a}^t + \mathbf{a}^{t+h}) + O(h^3)$$

- One acceleration computation per step

Beeman

$$\mathbf{x}^{t+h} = \mathbf{x}^t + h\mathbf{v}^t + h^2 \left(\frac{2}{3}\mathbf{a}^t - \frac{1}{6}\mathbf{a}^{t-h} \right) + O(h^4)$$

$$\mathbf{v}^{t+h} = \mathbf{v}^t + h \left(\frac{5}{12}\mathbf{a}^{t+h} + \frac{2}{3}\mathbf{a}^t - \frac{1}{12}\mathbf{a}^{t-h} \right) + O(h^4)$$

- One acceleration computation per step
- Improved accuracy compared to Velocity Verlet
- Possibly larger time step

Gear

- Taylor approximation

$$\mathbf{x}^{t+h} = \mathbf{x}^t + \frac{d\mathbf{x}^t}{dt} \frac{h}{1!} + \frac{d^2\mathbf{x}^t}{dt^2} \frac{h^2}{2!} + \frac{d^3\mathbf{x}^t}{dt^3} \frac{h^3}{3!} + \frac{d^4\mathbf{x}^t}{dt^4} \frac{h^4}{4!} + \frac{d^5\mathbf{x}^t}{dt^5} \frac{h^5}{5!} + \dots$$

- Notation

$$\mathbf{r}_k^t = \frac{d\mathbf{x}^k}{dt^k} \frac{h^k}{k!}$$

$$\mathbf{r}_0^{t+h} = \mathbf{r}_0^t + \mathbf{r}_1^t + \mathbf{r}_2^t + \mathbf{r}_3^t + \mathbf{r}_4^t + \mathbf{r}_5^t + \dots$$

Gear

$$- \mathbf{x}^{t+h} = \mathbf{x}^t + \frac{d\mathbf{x}^t}{dt} \frac{h}{1!} + \frac{d^2\mathbf{x}^t}{dt^2} \frac{h^2}{2!} + \frac{d^3\mathbf{x}^t}{dt^3} \frac{h^3}{3!} + \frac{d^4\mathbf{x}^t}{dt^4} \frac{h^4}{4!} + \frac{d^5\mathbf{x}^t}{dt^5} \frac{h^5}{5!} + \dots$$

$$\mathbf{r}_0^{t+h} = \mathbf{r}_0^t + \mathbf{r}_1^t + \mathbf{r}_2^t + \mathbf{r}_3^t + \mathbf{r}_4^t + \mathbf{r}_5^t + \dots$$

$$- h \frac{d\mathbf{x}^{t+h}}{dt} = h \frac{d\mathbf{x}^t}{dt} + h \frac{d^2\mathbf{x}^t}{dt^2} \frac{h}{1!} + h \frac{d^3\mathbf{x}^t}{dt^3} \frac{h^2}{2!} + h \frac{d^4\mathbf{x}^t}{dt^4} \frac{h^3}{3!} + h \frac{d^5\mathbf{x}^t}{dt^5} \frac{h^4}{4!} + \dots$$

$$\mathbf{r}_1^{t+h} = \mathbf{r}_1^t + 2\mathbf{r}_2^t + 3\mathbf{r}_3^t + 4\mathbf{r}_4^t + 5\mathbf{r}_5^t + \dots$$

$$- \frac{h^2}{2} \frac{d^2\mathbf{x}^{t+h}}{dt^2} = \frac{h^2}{2} \frac{d^2\mathbf{x}^t}{dt^2} + \frac{h^2}{2} \frac{d^3\mathbf{x}^t}{dt^3} \frac{h}{1!} + \frac{h^2}{2} \frac{d^4\mathbf{x}^t}{dt^4} \frac{h^2}{2!} + \frac{h^2}{2} \frac{d^5\mathbf{x}^t}{dt^5} \frac{h^3}{3!} + \dots$$

$$\mathbf{r}_2^{t+h} = \mathbf{r}_2^t + 3\mathbf{r}_3^t + 6\mathbf{r}_4^t + 10\mathbf{r}_5^t + \dots$$

Gear - Prediction

$$\mathbf{x}^{t+h} = \mathbf{r}_0^{t+h} = \mathbf{r}_0^t + \mathbf{r}_1^t + \mathbf{r}_2^t + \mathbf{r}_3^t + \mathbf{r}_4^t + \mathbf{r}_5^t$$

$$h\mathbf{v}^{t+h} = \mathbf{r}_1^{t+h} = \mathbf{r}_1^t + 2\mathbf{r}_2^t + 3\mathbf{r}_3^t + 4\mathbf{r}_4^t + 5\mathbf{r}_5^t$$

$$\frac{h^2}{2}\mathbf{a}^{t+h} = \mathbf{r}_2^{t+h} = \mathbf{r}_2^t + 3\mathbf{r}_3^t + 6\mathbf{r}_4^t + 10\mathbf{r}_5^t$$

$$\mathbf{r}_3^{t+h} = \mathbf{r}_3^t + 4\mathbf{r}_4^t + 10\mathbf{r}_5^t$$

$$\mathbf{r}_4^{t+h} = \mathbf{r}_4^t + 5\mathbf{r}_5^t$$

$$\mathbf{r}_5^{t+h} = \mathbf{r}_5^t$$

Gear - Correction

- Error / inconsistency between the predicted acceleration $\frac{2}{h^2} \mathbf{r}_2^{t+h}$ at time $t + h$ and the acceleration $\mathbf{a}^{t+h}(\mathbf{r}_0^{t+h}, \frac{1}{h} \mathbf{r}_1^{t+h})$ at predicted positions \mathbf{r}_0^{t+h} and velocities $\frac{1}{h} \mathbf{r}_1^{t+h}$:

$$\boldsymbol{\epsilon}^{t+h} = \mathbf{r}_2^{t+h} - \frac{h^2}{2} \mathbf{a}^{t+h}$$

- Correction:

$$\mathbf{r}_k^{t+h} = \mathbf{r}_k^{t+h} - c_k \boldsymbol{\epsilon}^{t+h}$$

with coefficients

$$c_0 = \frac{3}{20}, c_1 = \frac{251}{360}, c_2 = 1, c_3 = \frac{11}{18}, c_4 = \frac{1}{6}, c_5 = \frac{1}{60}$$

Gear - Implementation

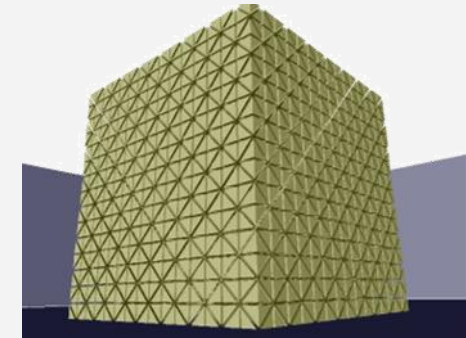
- Initialization: $\mathbf{r}_0^{t_0} = \mathbf{x}^{t_0}$ $\mathbf{r}_1^{t_0} = h\mathbf{v}^{t_0}$ $\mathbf{r}_2^{t_0} = \frac{h^2}{2}\mathbf{a}^{t_0}$
 $\mathbf{r}_3^{t_0} = \mathbf{r}_4^{t_0} = \mathbf{r}_5^{t_0} = 0$
- Prediction: $\mathbf{r}_0^{t+h} = \mathbf{r}_0^t + \dots + \mathbf{r}_5^t$
 $\mathbf{r}_k^{t+h} = \dots$
- Error: $\boldsymbol{\epsilon}^{t+h} = \mathbf{r}_2^{t+h} - \frac{h^2}{2}\mathbf{a}^{t+h}$
- Correction: $\mathbf{r}_k^{t+h} = \mathbf{r}_k^{t+h} - c_k\boldsymbol{\epsilon}^{t+h}$

Outline

- Introduction
- Particle motion
- Finite differences
- System of first-order ODEs
- Second-order ODE
- Performance
- Discussion

Comparison

- Deformable cube on a plane (4k particles, 17k tetrahedra, 22k edges), spring forces, volume preservation, gravity, contact



Scheme	Error order	Time step [ms]	Computation time [ms]	Ratio
Explicit Euler	1	0.5	9.5	0.05
RK 2	2	3.8	18.9	0.20
Implicit Euler	1	49.0	172.0	0.28
RK 4	4	17.0	50.0	0.34
Verlet	3	11.5	9.5	1.21

Time Step

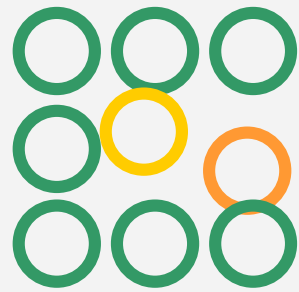
- Larger time steps are generally advantageous for the performance
- However, the time step size is limited: $h \leq \frac{d}{|\mathbf{v}|}$
- A particle should not move farther than its size in one simulation step, e.g. its diameter d : $h|\mathbf{v}| \leq d$

Time Step

- Critical states that can be avoided by a time step limit
 - Inverted elements
 - Unresolvable contacts



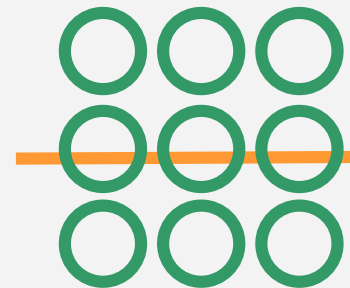
Rest state



Inverted elements



State at t :
No contact



State at $t+h$:
Contact

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Explicit Schemes

- Error order determines accuracy
- Improved accuracy may correspond to an improved stability for larger time steps
- Improved accuracy may correspond to higher costs
- Time steps are comparatively small
- Stability is generally an issue

Implicit Schemes

- Generally stable and robust
- Handle larger time steps
- Less accurate (scheme, linearization, solver)
 - Typically artificial damping / viscosity
- Decreasing accuracy for larger time steps
 - Same as for explicit schemes, but explicit schemes get unstable, while implicit schemes stay stable