

Simulation in Computer Graphics

Particle-based Fluid Simulation

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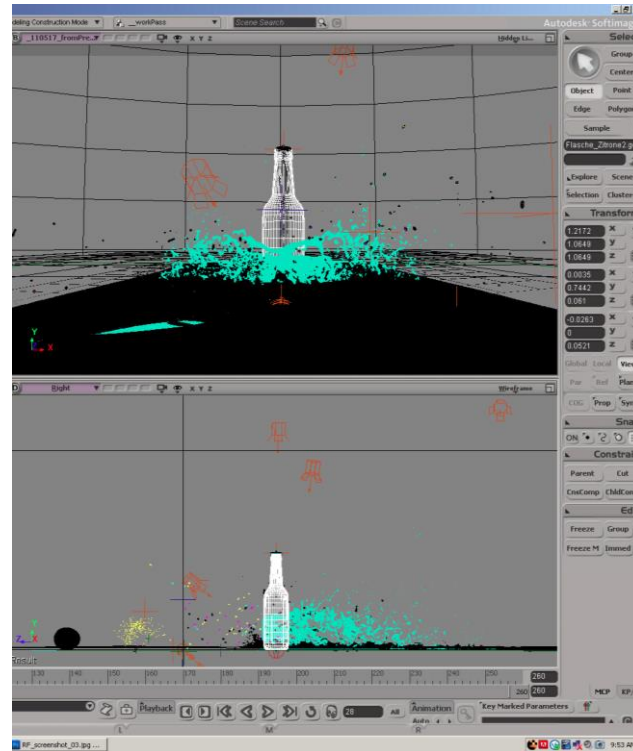
Albert-Ludwigs-Universität Freiburg

Application (with Pixar)



10 million fluid + 4 million rigid particles, 50 s simulated,
50 h computation time on a 16-core PC, www.youtube.com/cgfreiburg

Application (Commercials)



Copyright
NHB Studios,
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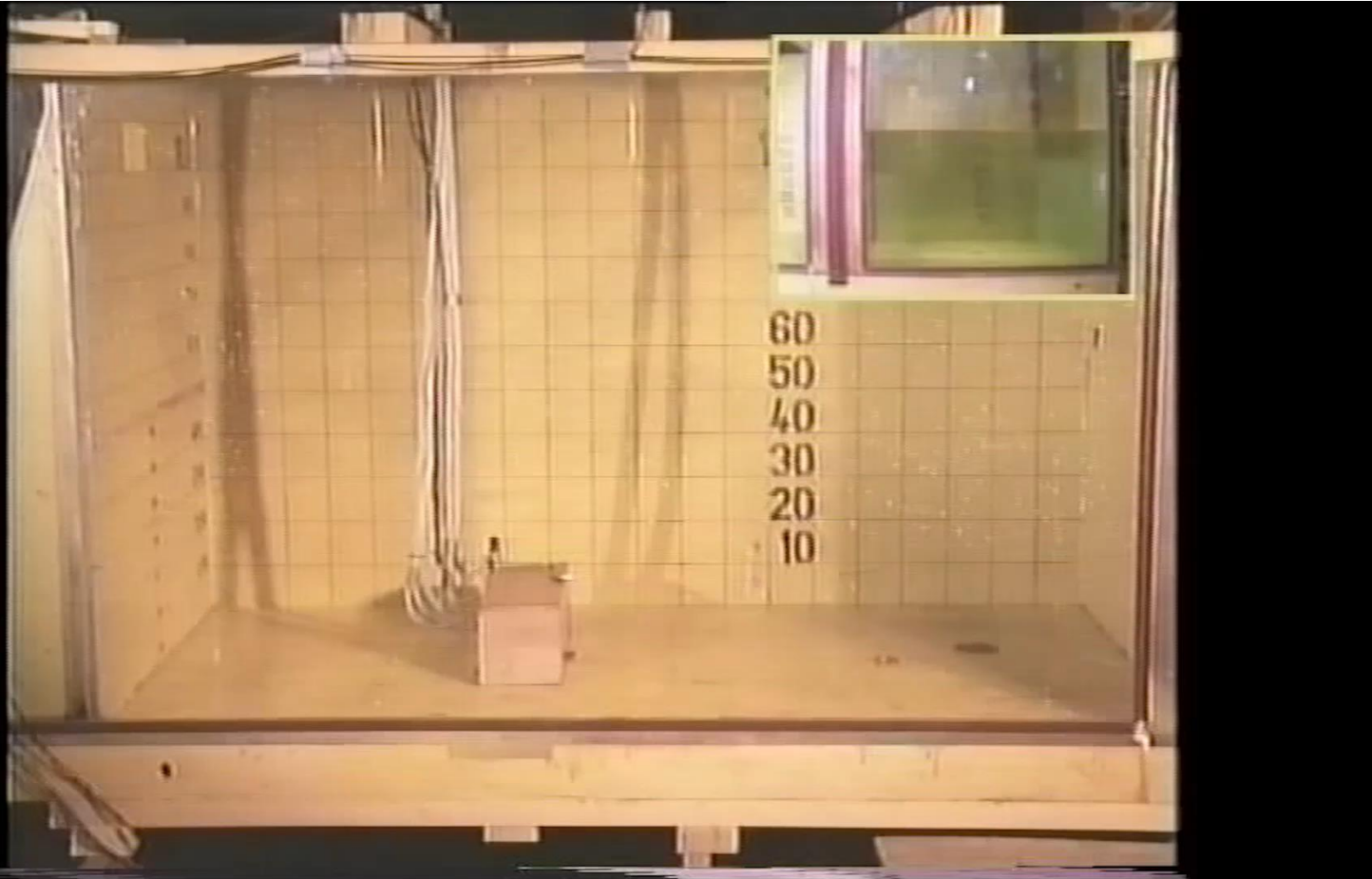
Application (with FIFTY2 Technology)

PreonLab: Drive Through



PreonLab, FIFTY2 Technology GmbH, www.youtube.com -> fifty2

Application (with FIFTY2 Technology)

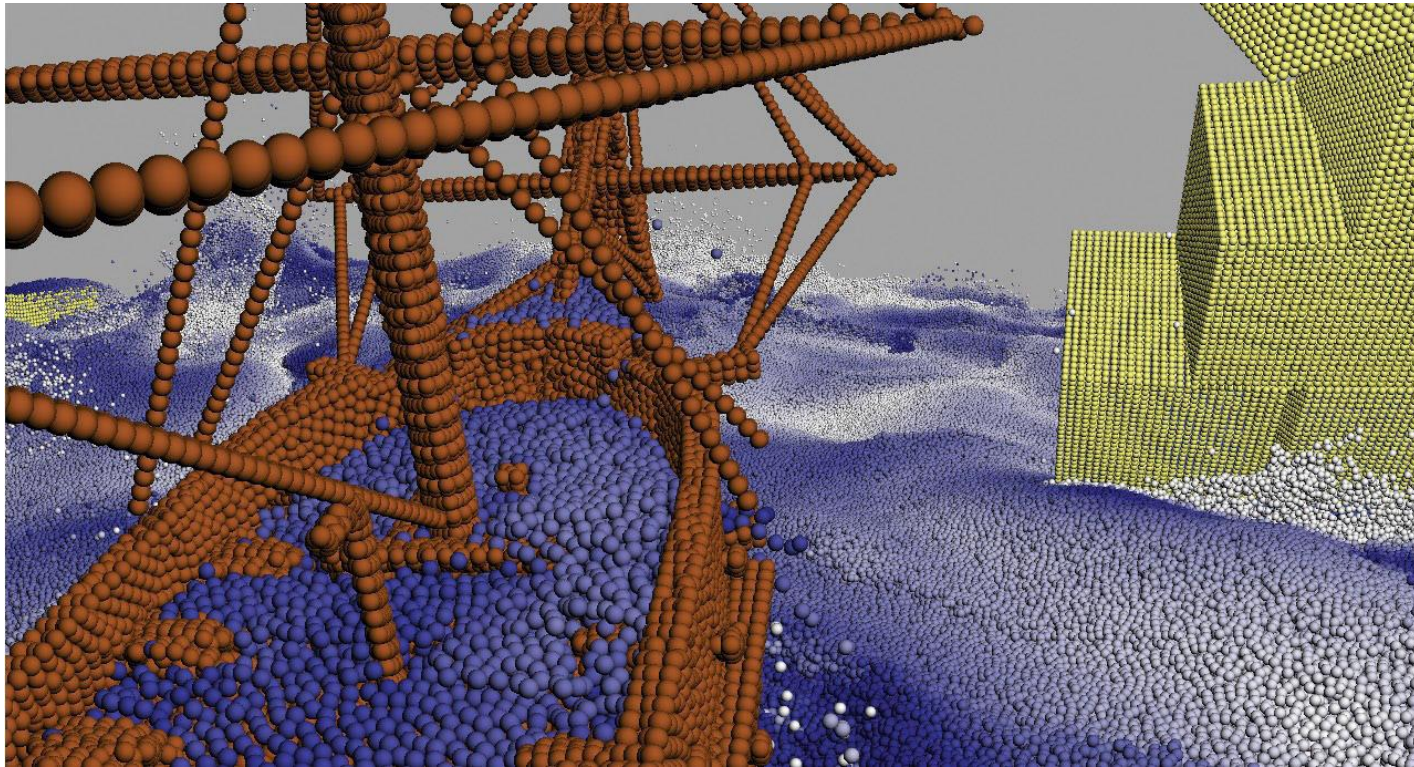


PreonLab, FIFTY2 Technology GmbH, www.youtube.com -> fifty2

Outline

- concept of an SPH fluid simulator
- momentum equation
- SPH basics
- neighborhood search
- boundary handling
- incompressibility
- surface reconstruction

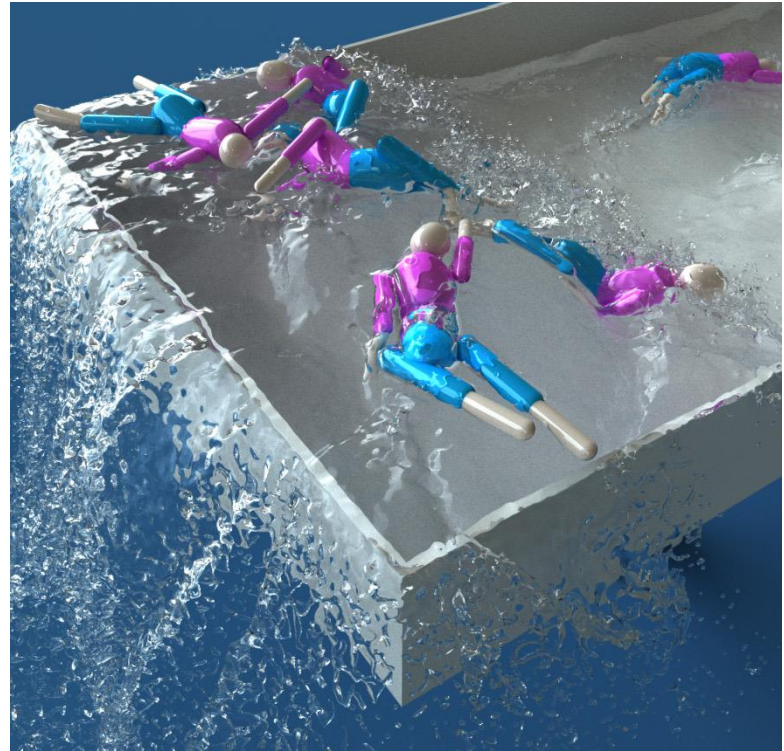
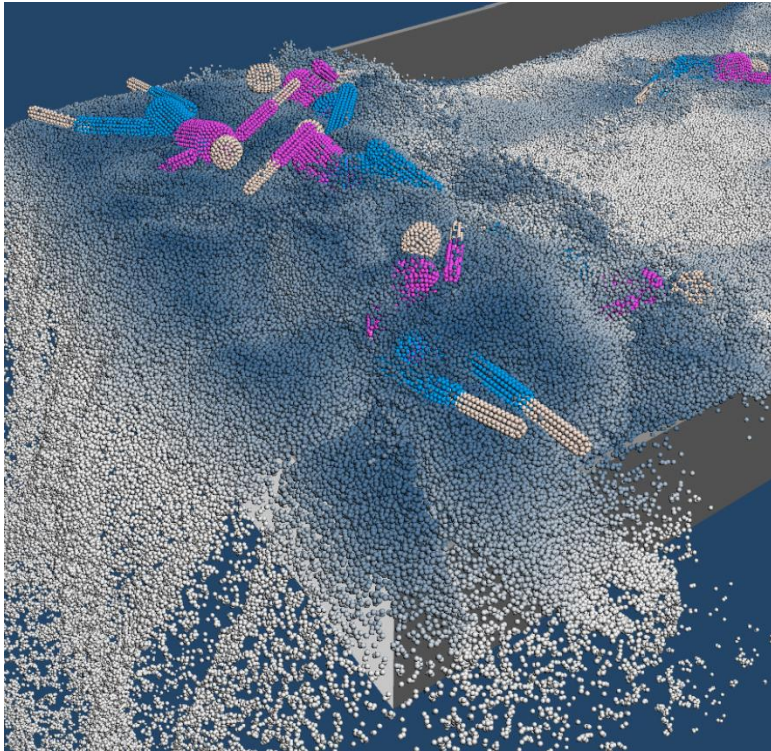
Concept



Concept

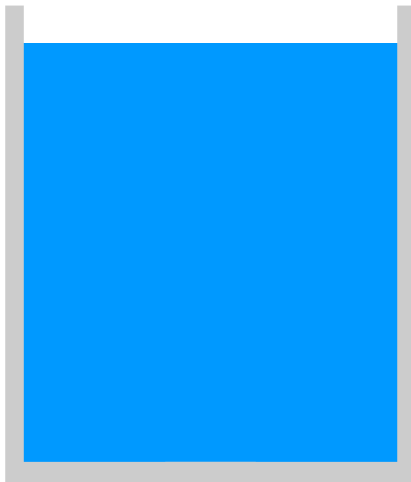


Fluid Representation

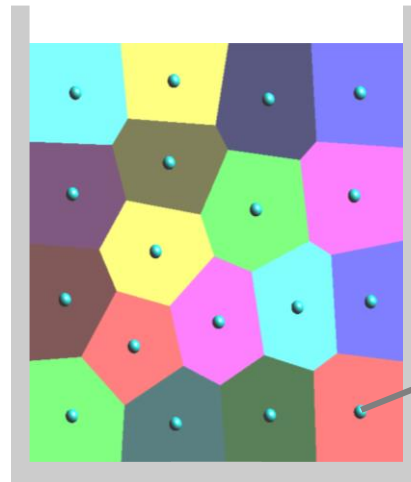


Fluid Representation

- fluid body is subdivided into small moving parcels, i.e. particles, with fluid properties



Fluid body

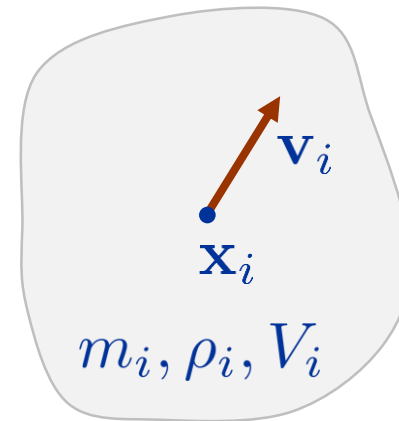


Set of fluid parcels

$\mathbf{x}, \mathbf{v}, m, V, \rho, p$

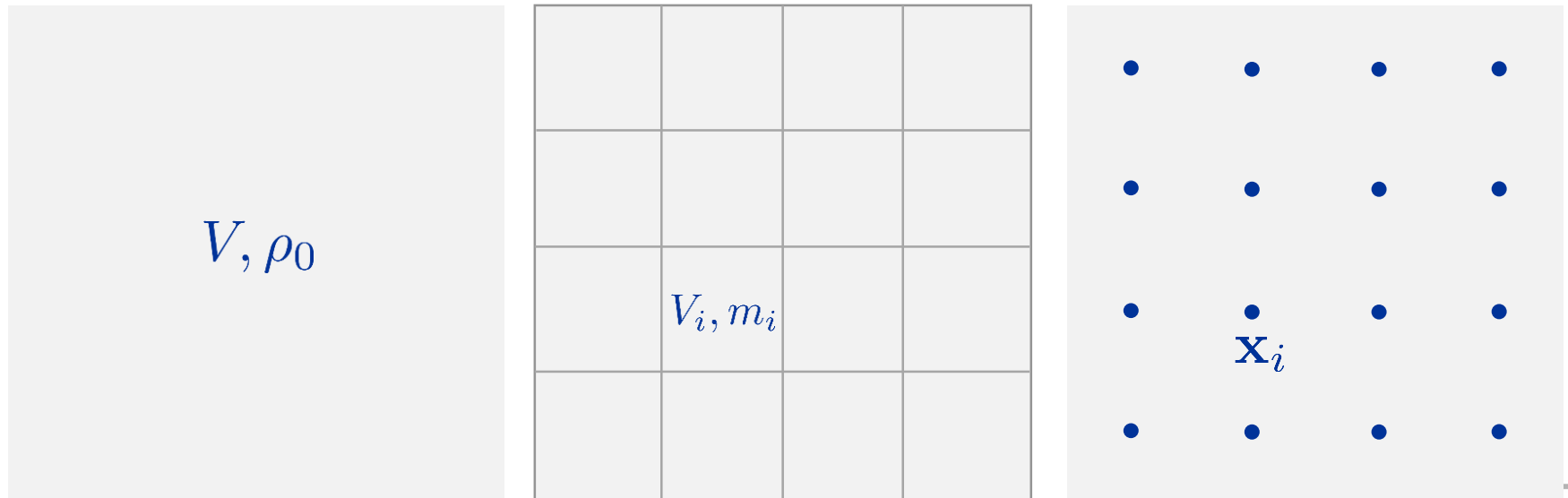
Particles / Fluid Parcels

- represent small fluid portions
- are represented by a sample position $\mathbf{x}_i$
- move with their velocity $\mathbf{v}_i$
- have a fixed mass m_i
- volume and density are related by $V_i = \frac{m_i}{\rho_i}$
 - preservation of density / volume over time is one of the challenges of a fluid simulator
- shape is not considered



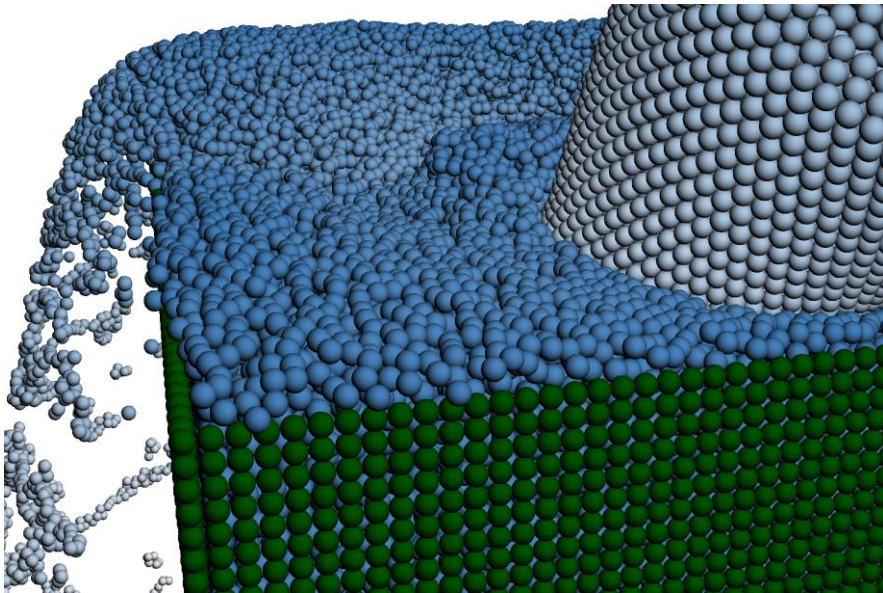
Typical Setup

- define overall fluid volume V and fluid density ρ_0
- define number n of particles
- assume particles of uniform size $V_i = \frac{V}{n}$
- compute particle mass as $m_i = \rho_0 \cdot V_i$
- sample $\mathbf{x}_i$ represents a particle in the simulation

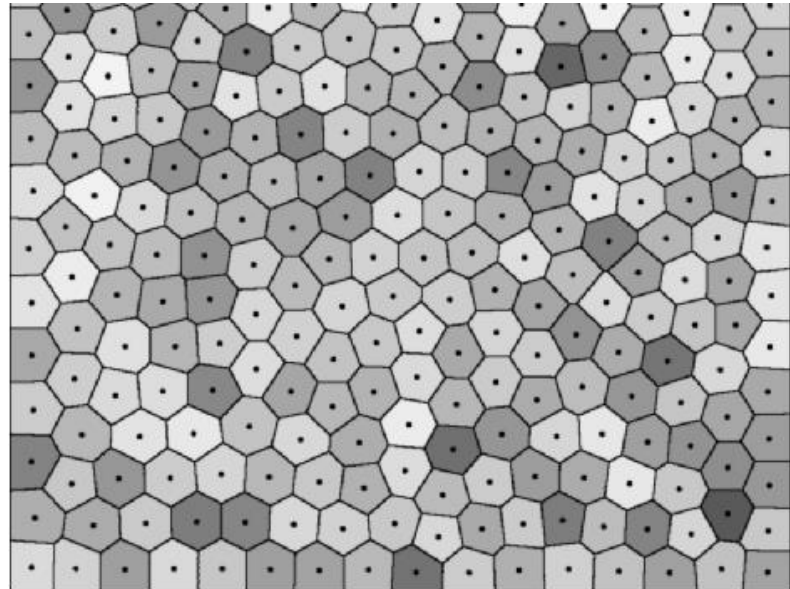


Particle Shape

- typically initialized as a cube
- typically visualized as a sphere
- implicitly handled as Voronoi cell by the simulation



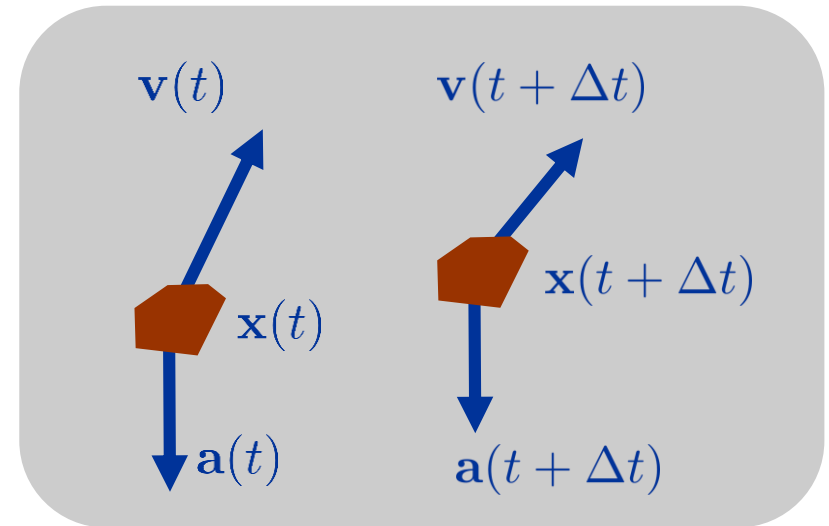
PreonLab, FIFTY2 Technology GmbH



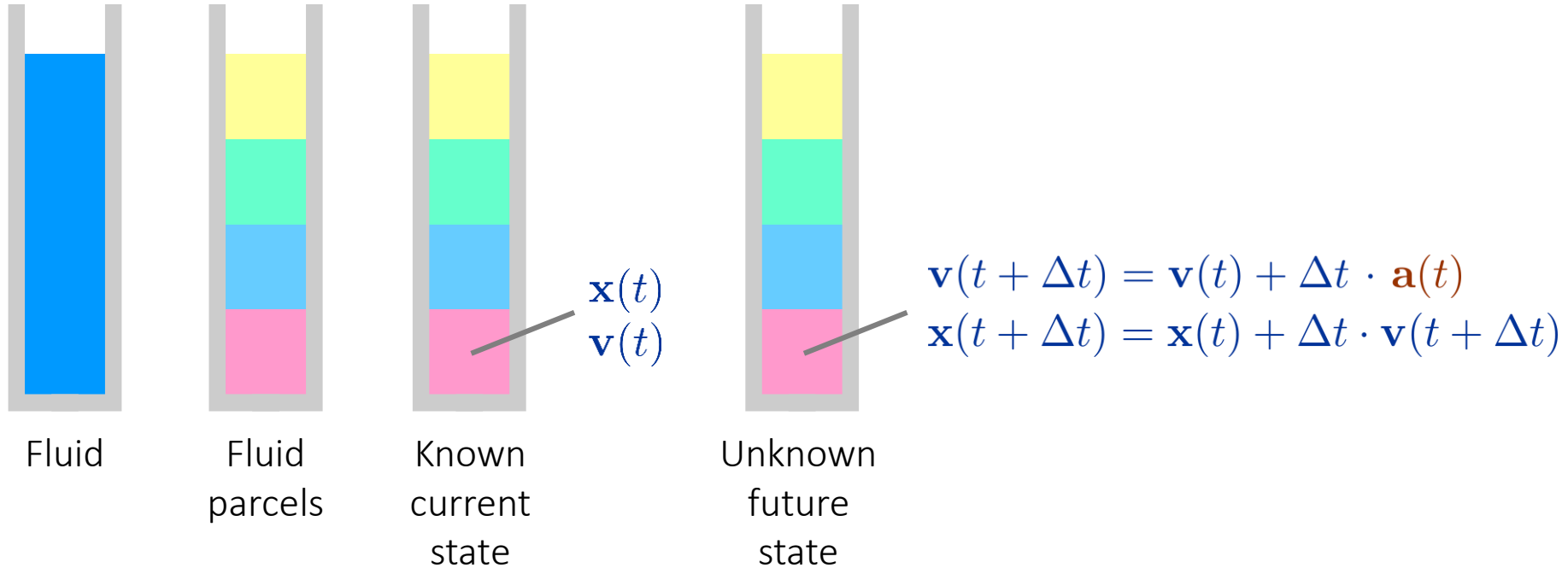
Adrian Secord: Weighted Voronoi Stippling, NPAR 2002.

Fluid Simulation

- computation of positions and velocities of fluid parcels over time
 - velocity change from current time t to subsequent time $t + \Delta t$
$$\mathbf{v}(t + \Delta t) = \mathbf{v}(t) + \Delta t \cdot \mathbf{a}(t)$$
 - position change
$$\mathbf{x}(t + \Delta t) = \mathbf{x}(t) + \Delta t \cdot \mathbf{v}(t + \Delta t)$$



Example

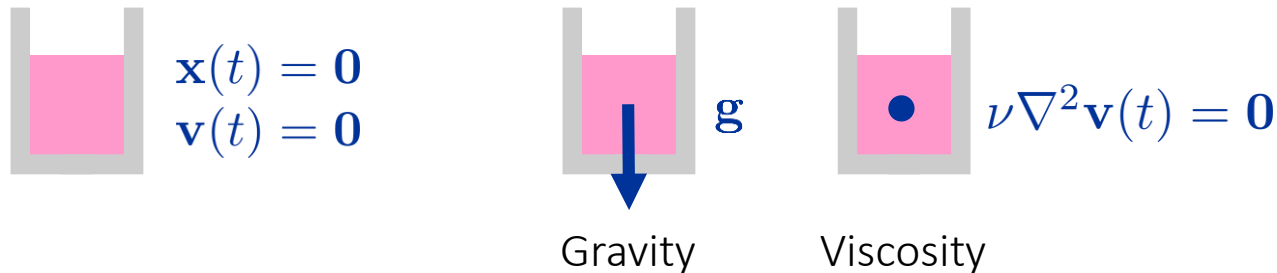


Accelerations

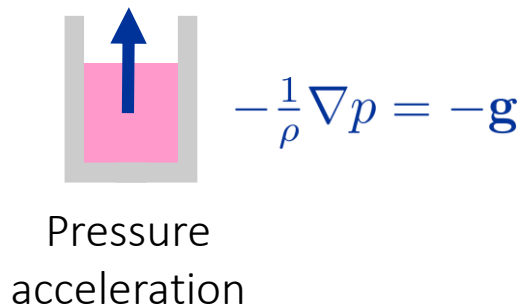
- gravity $\mathbf{g}$
- viscosity $\nu \nabla^2 \mathbf{v}$
 - friction
 - accelerate parcel towards the average velocity of adjacent fluid parcels
- pressure acceleration $-\frac{1}{\rho} \nabla p$
 - prevent fluid parcels from density / volume changes

Simulation Step - Example

- gravity and viscosity would change the parcel volume

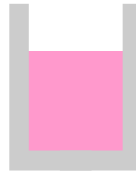


- pressure acceleration avoids the volume / density change



Simulation Step - Example

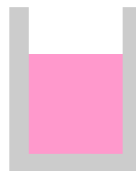
- current state


$$\begin{aligned}\mathbf{x}(t) &= \mathbf{0} \\ \mathbf{v}(t) &= \mathbf{0}\end{aligned}$$

- overall acceleration

$$\begin{aligned}\mathbf{a}(t) &= \mathbf{g} + \nu \nabla^2 \mathbf{v}(t) - \frac{1}{\rho} \nabla p \\ &= \mathbf{g} + \mathbf{0} - \mathbf{g} = \mathbf{0}\end{aligned}$$

- subsequent state


$$\begin{aligned}\mathbf{x}(t + \Delta t) &= \mathbf{x}(t) + \Delta t \cdot \mathbf{v}(t) = \mathbf{0} \\ \mathbf{v}(t + \Delta t) &= \mathbf{v}(t) + \Delta t \cdot \mathbf{a}(t) = \mathbf{0}\end{aligned}$$

Neighboring Parcels

- computations require neighboring parcels j

- density or volume

$$\rho_i = \sum_j m_j W_{ij} \quad V_i = \frac{V_i^0}{\sum_j V_j^0 W_{ij}}$$

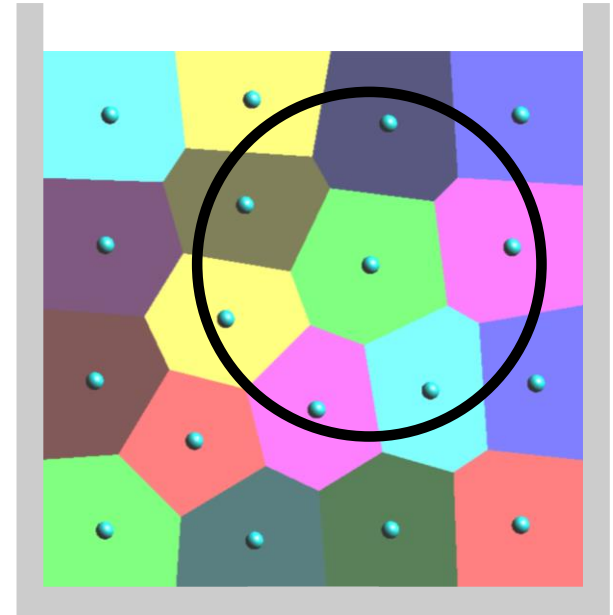
- pressure acceleration

$$-\frac{V_i}{m_i} \nabla p = -\frac{V_i}{m_i} \sum_j (p_i + p_j) V_j \nabla W_{ij}$$

$$-\frac{1}{\rho_i} \nabla p_i = -\sum_j m_j \left(\frac{p_i}{\rho_i^2} + \frac{p_j}{\rho_j^2} \right) \nabla W_{ij}$$

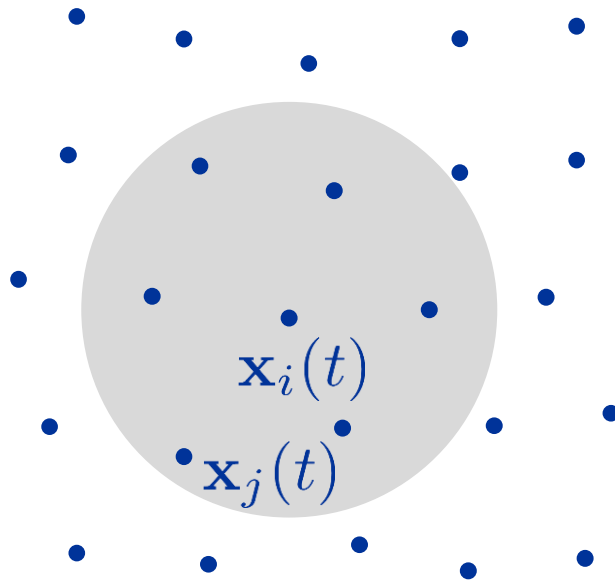
- Smoothed Particle Hydrodynamics

- Gingold and Monaghan, Lucy



Simulation Step - Implementation

- determine adjacent particles / neighbors
 $\mathbf{x}_j(t)$ of particle $\mathbf{x}_i(t)$ ($\mathbf{x}_i(t)$ is neighbor of $\mathbf{x}_i(t)$!)
- compute forces $\mathbf{F}_i(t) = \sum_j \dots$ as sums of neighbors
- advect the particles, e.g. Euler-Cromer
- determine neighbors
of particle $\mathbf{x}_i(t + \Delta t)$
- ...



Governing Equations

- particles /sample positions $\mathbf{x}_i$ and the respective attributes are advected with the local fluid velocity $\mathbf{v}_i$

$$\frac{d\mathbf{x}_i}{dt} = \mathbf{v}_i$$

- time rate of change of the velocity $\mathbf{v}_i$ is governed by the Lagrange form of the Navier-Stokes equation

$$\frac{d\mathbf{v}_i}{dt} = -\frac{1}{\rho_i} \nabla p_i + \nu \nabla^2 \mathbf{v}_i + \frac{\mathbf{F}_i^{\text{other}}}{m_i}$$

- this form of the Navier-Stokes equation requires that the particle positions are advected with the flow
- in contrast to the Eulerian form, it does not contain the convective acceleration $\mathbf{v}_i \cdot \nabla \mathbf{v}_i$, which is handled by the advection of the particles / sample positions
- in Eulerian approaches, sample positions are not necessarily advected with the flow

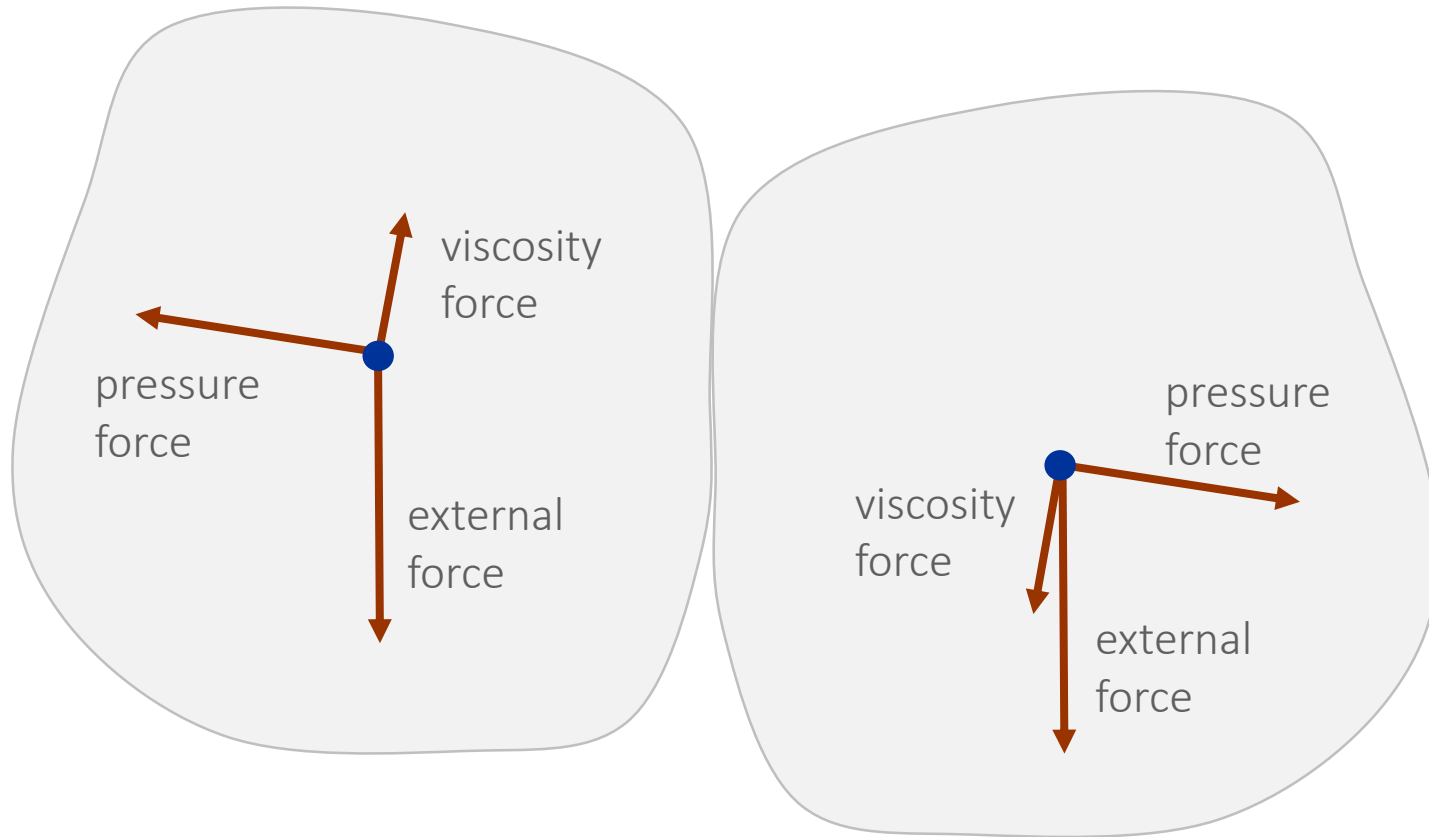
Accelerations

- $-\frac{1}{\rho_i} \nabla p_i$: acceleration due to pressure differences
 - preserves the fluid volume / density
 - pressure forces act in normal direction at the surface of the fluid element
 - small and preferably constant density deviations are important for high-quality simulation
- $\nu \nabla^2 \mathbf{v}_i$: acceleration due to friction forces between particles with different velocities
 - friction forces act in tangential (and normal) direction at the surface of the fluid element
 - kinematic viscosity $\nu \approx 10^{-6} \text{m}^2 \cdot \text{s}^{-1}$: larger friction is less realistic, but can improve the stability, dynamic viscosity $\eta = \mu = \nu \cdot \rho_0$
- $\frac{\mathbf{F}_i^{\text{other}}}{m_i}$: e.g., gravity, boundary handling

Accelerations

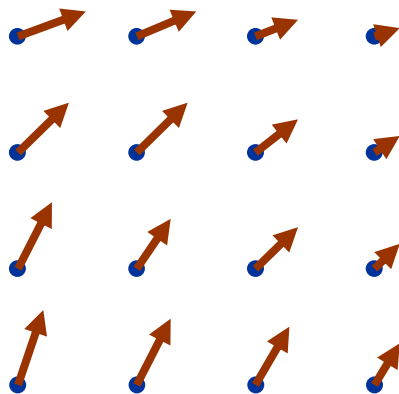
- $-\frac{1}{\rho} \nabla p = -\frac{1}{\rho} \begin{pmatrix} \frac{\partial p}{\partial x_x} \\ \frac{\partial p}{\partial x_y} \\ \frac{\partial p}{\partial x_z} \end{pmatrix} = -\frac{1}{\rho} \nabla \cdot \begin{pmatrix} p & 0 & 0 \\ 0 & p & 0 \\ 0 & 0 & p \end{pmatrix}$
- $\nu \nabla^2 \mathbf{v} = \nu \nabla \cdot (\nabla \mathbf{v}) = \nu \nabla \cdot \begin{pmatrix} \frac{\partial v_x}{\partial x_x} & \frac{\partial v_x}{\partial x_y} & \frac{\partial v_x}{\partial x_z} \\ \frac{\partial v_y}{\partial x_x} & \frac{\partial v_y}{\partial x_y} & \frac{\partial v_y}{\partial x_z} \\ \frac{\partial v_z}{\partial x_x} & \frac{\partial v_z}{\partial x_y} & \frac{\partial v_z}{\partial x_z} \end{pmatrix}$
 $= \nu \begin{pmatrix} \frac{\partial^2 v_x}{\partial x_x^2} + \frac{\partial^2 v_x}{\partial x_y^2} + \frac{\partial^2 v_x}{\partial x_z^2} \\ \frac{\partial^2 v_y}{\partial x_x^2} + \frac{\partial^2 v_y}{\partial x_y^2} + \frac{\partial^2 v_y}{\partial x_z^2} \\ \frac{\partial^2 v_z}{\partial x_x^2} + \frac{\partial^2 v_z}{\partial x_y^2} + \frac{\partial^2 v_z}{\partial x_z^2} \end{pmatrix}$

Forces



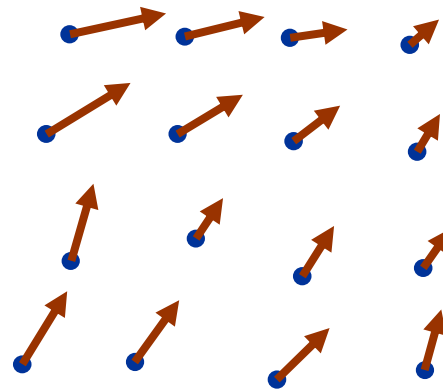
Lagrangian Fluid Simulation

- fluid simulators compute the velocity field over time
- Lagrangian approaches compute the velocities for samples $\mathbf{x}_i$ that are advected with their velocity $\mathbf{v}_i$



$$\mathbf{v}_i(x_i, y_i, z_i, t) = (u_i, v_i, w_i)$$

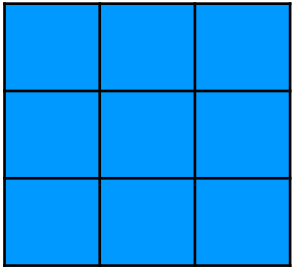
$$\mathbf{x}_i(t) = (x_i, y_i, z_i)$$



$$\mathbf{v}_i(x_i + \Delta t \cdot u_i, y_i + \Delta t \cdot v_i, z_i + \Delta t \cdot w_i, t + \Delta t)$$

$$\mathbf{x}_i(t + \Delta t) = (x_i + \Delta t \cdot u_i, y_i + \Delta t \cdot v_i, z_i + \Delta t \cdot w_i)$$

Moving Parcels vs. Static Cells



$$\frac{d\mathbf{v}}{dt} = \mathbf{g} + \nu \nabla^2 \mathbf{v} - \frac{V}{m} \nabla p$$

Acceleration of a moving
parcel

Smoothed Particle Hydrodynamics

- proposed by Gingold / Monaghan and Lucy (1977)
- SPH can be used to interpolate fluid quantities at arbitrary positions and to approximate the spatial derivatives in the Navier-Stokes equation with a finite number of samples, i.e., adjacent particles
- SPH in a Lagrangian fluid simulation
 - fluid is represented with particles
 - particle positions and velocities are governed by $\frac{d\mathbf{x}_i}{dt} = \mathbf{v}_i$ and $\frac{d\mathbf{v}_i}{dt} = -\frac{1}{\rho_i} \nabla p_i + \nu \nabla^2 \mathbf{v}_i + \frac{\mathbf{F}_i^{other}}{m_i}$
 - ρ_i , $-\frac{1}{\rho_i} \nabla p_i$ and $\nu \nabla^2 \mathbf{v}_i$ are computed with SPH
- SPH is typically used in Lagrangian, mesh-free approaches, but not limited to

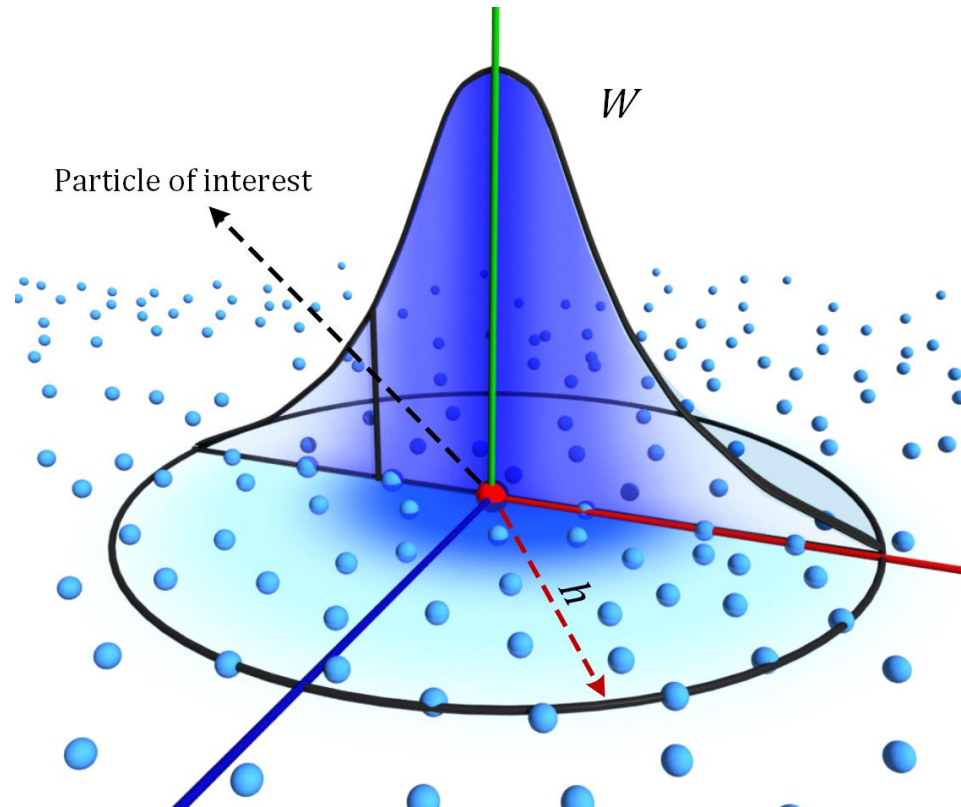
SPH Interpolation

- quantity A_i at an arbitrary position $\mathbf{x}_i$ is approximately computed with a set of known quantities A_j at sample positions $\mathbf{x}_j$

$$A_i = \sum_j V_j A_j W_{ij} = \sum_j \frac{m_j}{\rho_j} A_j W_{ij}$$

- $\mathbf{x}_i$ is not necessarily a sample position
 - if $\mathbf{x}_i$ is a sample position, it contributes to the sum
- W_{ij} is a kernel function that weights the contributions of sample positions $\mathbf{x}_j$ according to their distance to $\mathbf{x}_i$
$$W_{ij} = W\left(\frac{\|\mathbf{x}_i - \mathbf{x}_j\|}{h}\right) = W(q)$$
 - d is the dimensionality of the simulation domain
 - h is the so-called smoothing length

SPH Interpolation – 2D

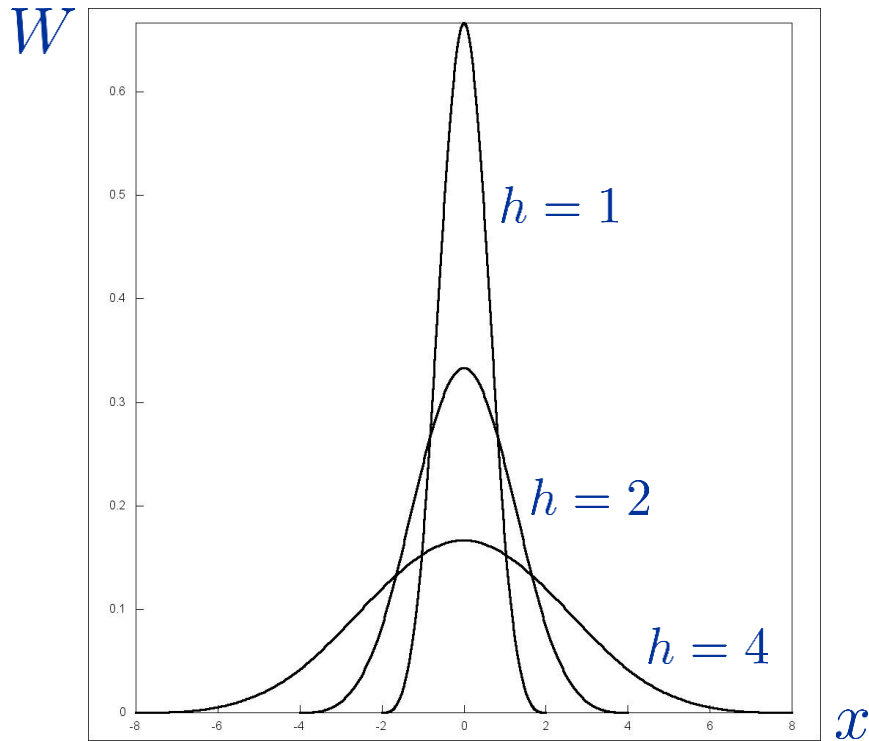


Neutrino Physics Guide

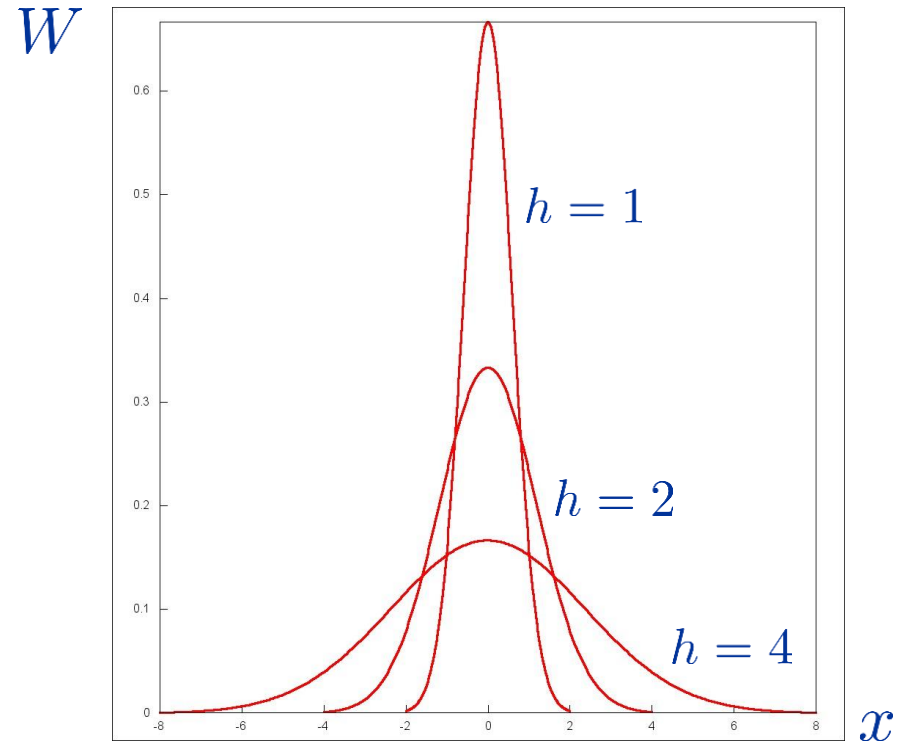
Kernel Function

- close to a Gaussian, but with compact support
 - support typically between h and $3h$
- e.g. cubic spline (1D: $\alpha = \frac{1}{6h}$ 2D: $\alpha = \frac{5}{14\pi h^2}$ 3D: $\alpha = \frac{1}{4\pi h^3}$)
$$W(q) = \alpha \begin{cases} (2-q)^3 - 4(1-q)^3 & 0 \leq q < 1 \\ (2-q)^3 & 1 \leq q < 2 \\ 0 & q \geq 2 \end{cases} \quad q = \frac{\|\mathbf{x}_i - \mathbf{x}_j\|}{h}$$
- number of particles / samples that are considered in the interpolation depends on
 - dimensionality, kernel support, particle spacing
 - e.g., 3D, cubic spline support $2h$, particle spacing h result in 30-40 neighboring particles
 - number of neighbors should not be too small to appropriately sample the kernel function

Kernel Function in 1D



$$W(x) = \frac{1}{6h} \begin{cases} (2 - \frac{|x|}{h})^3 - 4(1 - \frac{|x|}{h})^3 & 0 \leq |x| < h \\ (2 - \frac{|x|}{h})^3 & h \leq |x| < 2h \\ 0 & |x| \geq 2h \end{cases}$$



$$W(x) = \frac{2}{3h} e^{-\frac{x^2}{2 \cdot (0.59h)^2}}$$

Spatial Derivatives with SPH

- original approximations

$$\nabla A_i = \sum_j \frac{m_j}{\rho_j} A_j \nabla W_{ij}$$

$$\nabla^2 A_i = \sum_j \frac{m_j}{\rho_j} A_j \nabla^2 W_{ij}$$

- currently preferred approximations

$$\nabla A_i = \rho_i \sum_j m_j \left(\frac{A_i}{\rho_i^2} + \frac{A_j}{\rho_j^2} \right) \nabla W_{ij}$$

$$\nabla^2 A_i = 2 \sum_j \frac{m_j}{\rho_j} A_{ij} \frac{\mathbf{x}_{ij} \cdot \nabla W_{ij}}{\mathbf{x}_{ij} \cdot \mathbf{x}_{ij} + 0.01 h^2}$$

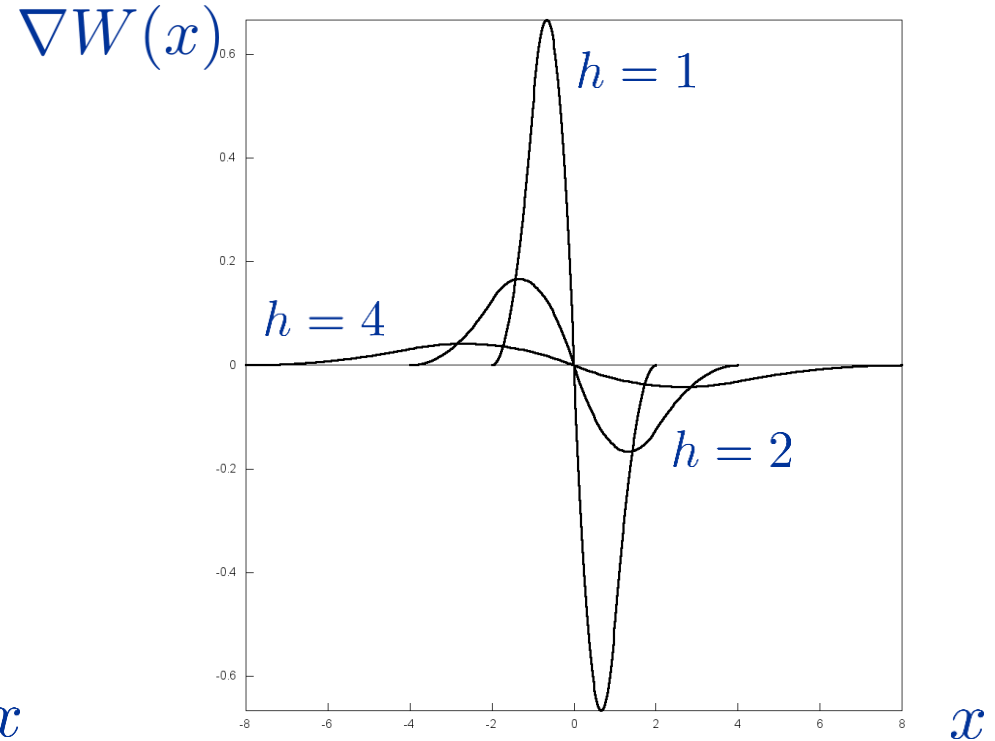
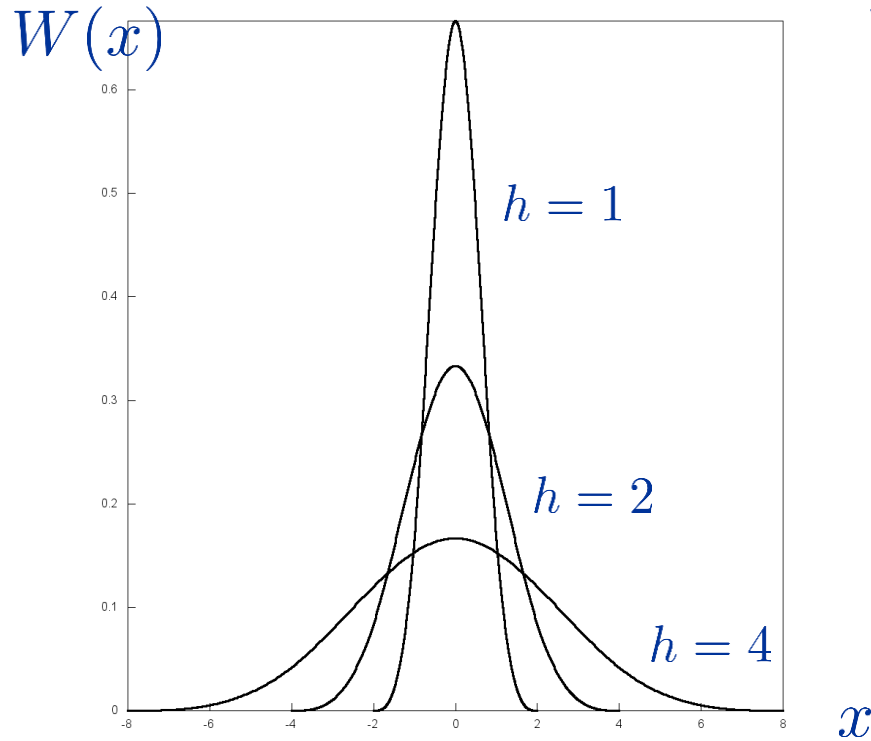
$$\nabla \cdot \mathbf{A}_i = -\frac{1}{\rho_i} \sum_j m_j \mathbf{A}_{ij} \nabla W_{ij}$$

$$A_{ij} = A_i - A_j \quad \mathbf{A}_{ij} = \mathbf{A}_i - \mathbf{A}_j \quad \mathbf{x}_{ij} = \mathbf{x}_i - \mathbf{x}_j$$

preserves linear and angular momentum, when used for pressure forces

more robust as it avoids the second derivative of W

Kernel Derivative in 1D



$$W(x) = \frac{1}{6h} \begin{cases} (2 - \frac{|x|}{h})^3 - 4(1 - \frac{|x|}{h})^3 & 0 \leq |x| < h \\ (2 - \frac{|x|}{h})^3 & h \leq |x| < 2h \\ 0 & |x| \geq 2h \end{cases} \quad \nabla W(x) = -\frac{x}{6h^2|x|} \begin{cases} -3(2 - \frac{|x|}{h})^2 + 12(1 - \frac{|x|}{h})^2 & 0 \leq |x| < h \\ -3(2 - \frac{|x|}{h})^2 & h \leq |x| < 2h \\ 0 & |x| \geq 2h \end{cases}$$

Density Computation

- explicit form

$$\rho_i = \sum_j \frac{m_j}{\rho_j} \rho_j W_{ij} = \sum_j m_j W_{ij}$$

- comparatively exact
- erroneous for incomplete neighborhood, e.g. at the free surface

- differential update

- using the continuity equation
- time rate of change of the density is related to the divergence of the velocity field $\frac{d\rho_i}{dt} = -\rho_i \nabla \cdot \mathbf{v}_i$

$$\frac{d\rho_i}{dt} = \sum_j m_j \mathbf{v}_{ij} \nabla W_{ij}$$

- no issues for incomplete neighborhoods
- drift, i.e. less accurate for large time steps

Simple SPH Fluid Solver

- find all neighbors j of particle i
 - typically accelerated with a uniform grid
 - cell size equal to kernel support , e.g. $2h$
- compute pressure p_i
 - e.g., from density ρ_i using a state equation, e.g. $p_i = k \left(\left(\frac{\rho_i}{\rho_0} \right)^7 - 1 \right)$
 - ρ_0 is the desired rest density of the fluid
 - k is a user-defined stiffness constant that scales pressure, pressure gradient, and the resulting pressure force
 - SPH with state equation is referred to as SESP
- compute pressure force, viscosity / gravitational force
- compute other forces, e.g. due to boundaries
- update velocity and position

Simple SPH Fluid Solver

- **for all particle i do**
 find neighbors j
for all particle i do
 $\rho_i = \sum_j m_j W_{ij}$
 compute p_i from ρ_i
for all particle i do
 $\mathbf{F}_i^{\text{pressure}} = -\frac{m_i}{\rho_i} \nabla p_i$
 $\mathbf{F}_i^{\text{viscosity}} = m_i \nu \nabla^2 \mathbf{v}_i$
 $\mathbf{F}_i^{\text{other}} = m_i \mathbf{g}$
 $\mathbf{F}_i(t) = \mathbf{F}_i^{\text{pressure}} + \mathbf{F}_i^{\text{viscosity}} + \mathbf{F}_i^{\text{other}}$
for all particle i do
 $\mathbf{v}_i(t + \Delta t) = \mathbf{v}_i(t) + \Delta t \mathbf{F}_i(t) / m_i$
 $\mathbf{x}_i(t + \Delta t) = \mathbf{x}_i(t) + \Delta t \mathbf{v}_i(t + \Delta t)$

Simple SPH Fluid Solver

- **for all particle i do**
 find neighbors j
for all particle i do
 $\rho_i = \sum_j m_j W_{ij}$
 compute p_i from ρ_i
for all particle i do
 $\mathbf{a}_i^{\text{pressure}} = -\frac{1}{\rho_i} \nabla p_i$
 $\mathbf{a}_i^{\text{viscosity}} = \nu \nabla^2 \mathbf{v}_i$
 $\mathbf{a}_i^{\text{other}} = \mathbf{g}$
 $\mathbf{a}_i(t) = \mathbf{a}_i^{\text{pressure}} + \mathbf{a}_i^{\text{viscosity}} + \mathbf{a}_i^{\text{other}}$
for all particle i do
 $\mathbf{v}_i(t + \Delta t) = \mathbf{v}_i(t) + \Delta t \mathbf{a}_i(t)$
 $\mathbf{x}_i(t + \Delta t) = \mathbf{x}_i(t) + \Delta t \mathbf{v}_i(t + \Delta t)$

Accelerations with SPH

- $$\mathbf{a}_i^{pressure} = - \sum_j m_j \left(\frac{p_i}{\rho_i^2} + \frac{p_j}{\rho_j^2} \right) \nabla W_{ij}$$
$$\mathbf{a}_i^{viscosity} = 2\nu \sum_j \frac{m_j}{\rho_j} \mathbf{v}_{ij} \frac{\mathbf{x}_{ij} \cdot \nabla W_{ij}}{\mathbf{x}_{ij} \cdot \mathbf{x}_{ij} + 0.01h^2}$$
$$\mathbf{a}_i^{other} = \mathbf{g}$$

Setting

- kernel has to be defined, e.g. cubic with support of $2h$
- particle mass m_i has to be specified
 - e.g., $m_i = h^3 \rho_0$ for a particle spacing of h
 - smaller spacing would result in smaller mass and more neighbors per particle
- numerical integration scheme
 - semi-implicit Euler (a.k.a. symplectic Euler or Euler-Cromer) is commonly used
- time step
 - size is governed by the Courant-Friedrich-Levy (CFL) condition
 - e.g., $\Delta t \leq \lambda \frac{h}{\|\mathbf{v}^{\max}\|}$ with $\lambda = 0.1$ and particle spacing h
 - motivation: for $\lambda \leq 1$, a particle moves less than its diameter per time step

Simulation in Computer Graphics

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Albert-Ludwigs-Universität Freiburg

Simple SPH Fluid Solver

- **for all particle i do**
 find neighbors j
for all particle i do
 $\rho_i = \sum_j m_j W_{ij}$
 compute p_i from ρ_i

$$\begin{aligned} \mathbf{a}_i^{pressure} &= - \sum_j m_j \left(\frac{p_i}{\rho_i^2} + \frac{p_j}{\rho_j^2} \right) \nabla W_{ij} \\ \mathbf{a}_i^{viscosity} &= 2\nu \sum_j \frac{m_j}{\rho_j} \mathbf{v}_{ij} \frac{\mathbf{x}_{ij} \cdot \nabla W_{ij}}{\mathbf{x}_{ij} \cdot \mathbf{x}_{ij} + 0.01h^2} \\ \mathbf{a}_i^{other} &= \mathbf{g} \end{aligned}$$

SPH approximations

- for all particle i do**

$$\begin{aligned} \mathbf{a}_i^{pressure} &= -\frac{1}{\rho_i} \nabla p_i \\ \mathbf{a}_i^{viscosity} &= \nu \nabla^2 \mathbf{v}_i \\ \mathbf{a}_i^{other} &= \mathbf{g} \\ \mathbf{a}_i(t) &= \mathbf{a}_i^{pressure} + \mathbf{a}_i^{viscosity} + \mathbf{a}_i^{other} \end{aligned}$$

Navier-Stokes
equation

- for all particle i do**

$$\begin{aligned} \mathbf{v}_i(t + \Delta t) &= \mathbf{v}_i(t) + \Delta t \mathbf{a}_i(t) \\ \mathbf{x}_i(t + \Delta t) &= \mathbf{x}_i(t) + \Delta t \mathbf{v}_i(t + \Delta t) \end{aligned}$$

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- surface reconstruction

Force Types

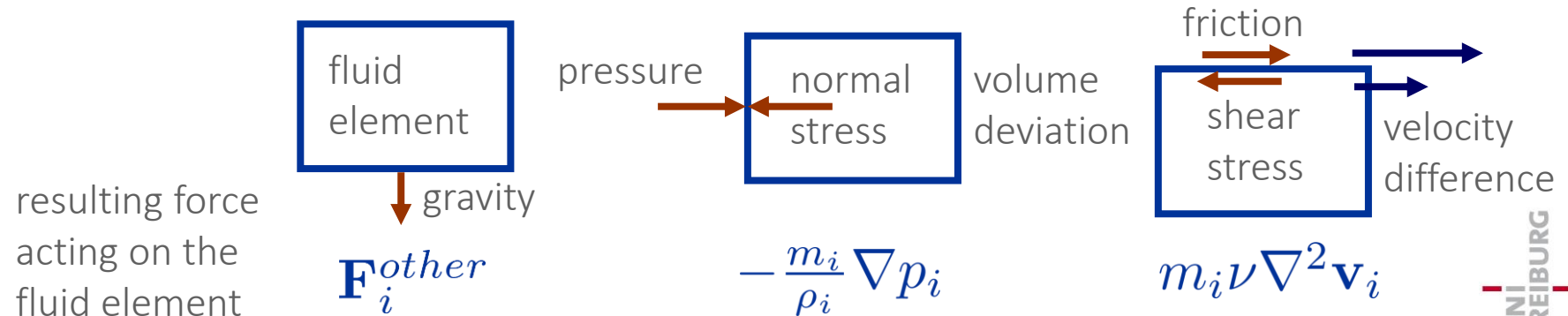
- momentum equation

$$\frac{d\mathbf{v}_i}{dt} = -\frac{1}{\rho_i} \nabla p_i + \nu \nabla^2 \mathbf{v}_i + \frac{\mathbf{F}_i^{\text{other}}}{m_i}$$

- body forces, e.g. gravity (external)

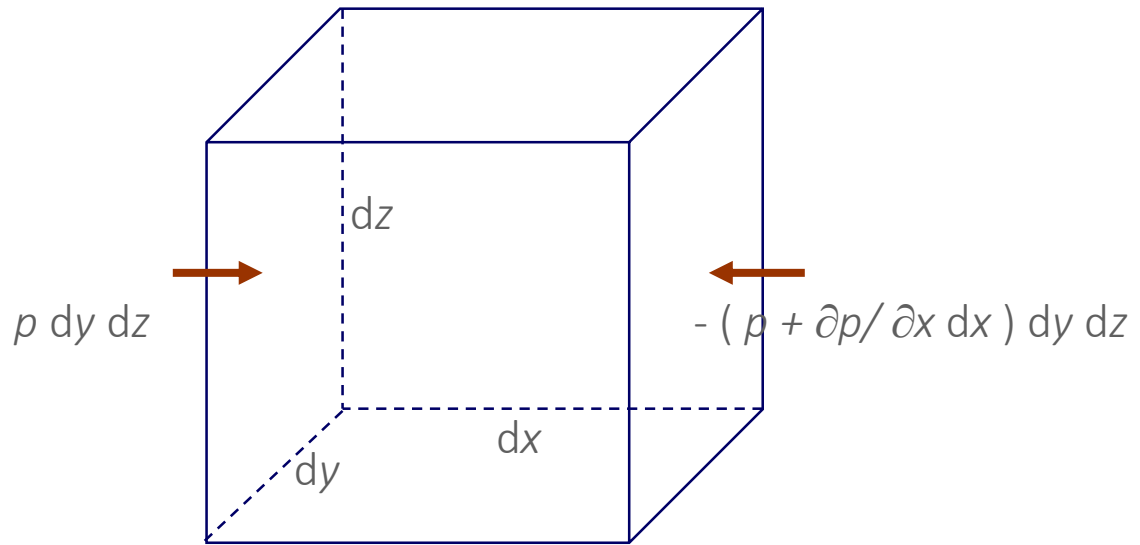
- surface forces (internal, i.e. conservative)

- based on shear and normal stress distribution on the surface due to deformation of the fluid element
- normal stress related to volume deviation
- normal and shear stress related to friction due to velocity differences



Pressure Force in x-direction

- pressure force acts orthogonal to the surface of the fluid element



- resulting pressure force

$$\left(p - \left(p + \frac{\partial p}{\partial x} dx \right) \right) dy \, dz = -\frac{\partial p}{\partial x} dx \, dy \, dz = -\frac{\partial p}{\partial x} V$$

Overall Pressure Force

- force at particle i

$$\mathbf{F}_i^{pressure} = - \begin{pmatrix} \frac{\partial p_i}{\partial x_{i,x}} \\ \frac{\partial p_i}{\partial x_{i,y}} \\ \frac{\partial p_i}{\partial x_{i,z}} \end{pmatrix} V_i = -\nabla p_i \quad V_i = -\frac{m_i}{\rho_i} \nabla p_i$$

- respective acceleration

$$\mathbf{a}_i^{pressure} = \frac{\mathbf{F}_i^{pressure}}{m_i} = -\frac{1}{\rho_i} \nabla p_i$$

Cauchy Momentum Equation

- Lagrange form

$$\frac{d\mathbf{v}}{dt} = \frac{1}{\rho} \nabla \cdot \sigma + \frac{\mathbf{F}^{other}}{m}$$

- σ is the stress tensor (a 3x3 matrix in 3D) describing the pressure distribution at the surface of a fluid element $\sigma = -p\mathbf{I}_3 + \tau$

- $\nabla \cdot \sigma$ is the resulting force per volume acting on the fluid element

- τ is the viscous stress tensor $\tau = \nu \begin{pmatrix} \frac{\partial u}{\partial x} + \frac{\partial u}{\partial x} & \frac{\partial v}{\partial x} + \frac{\partial u}{\partial y} & \frac{\partial w}{\partial x} + \frac{\partial u}{\partial z} \\ \frac{\partial u}{\partial y} + \frac{\partial v}{\partial x} & \frac{\partial v}{\partial y} + \frac{\partial v}{\partial y} & \frac{\partial w}{\partial y} + \frac{\partial v}{\partial z} \\ \frac{\partial u}{\partial z} + \frac{\partial w}{\partial x} & \frac{\partial v}{\partial z} + \frac{\partial w}{\partial y} & \frac{\partial w}{\partial z} + \frac{\partial w}{\partial z} \end{pmatrix}$

- $\nabla \cdot \tau = \nu \nabla^2 \mathbf{v}$ is the resulting viscosity force per volume

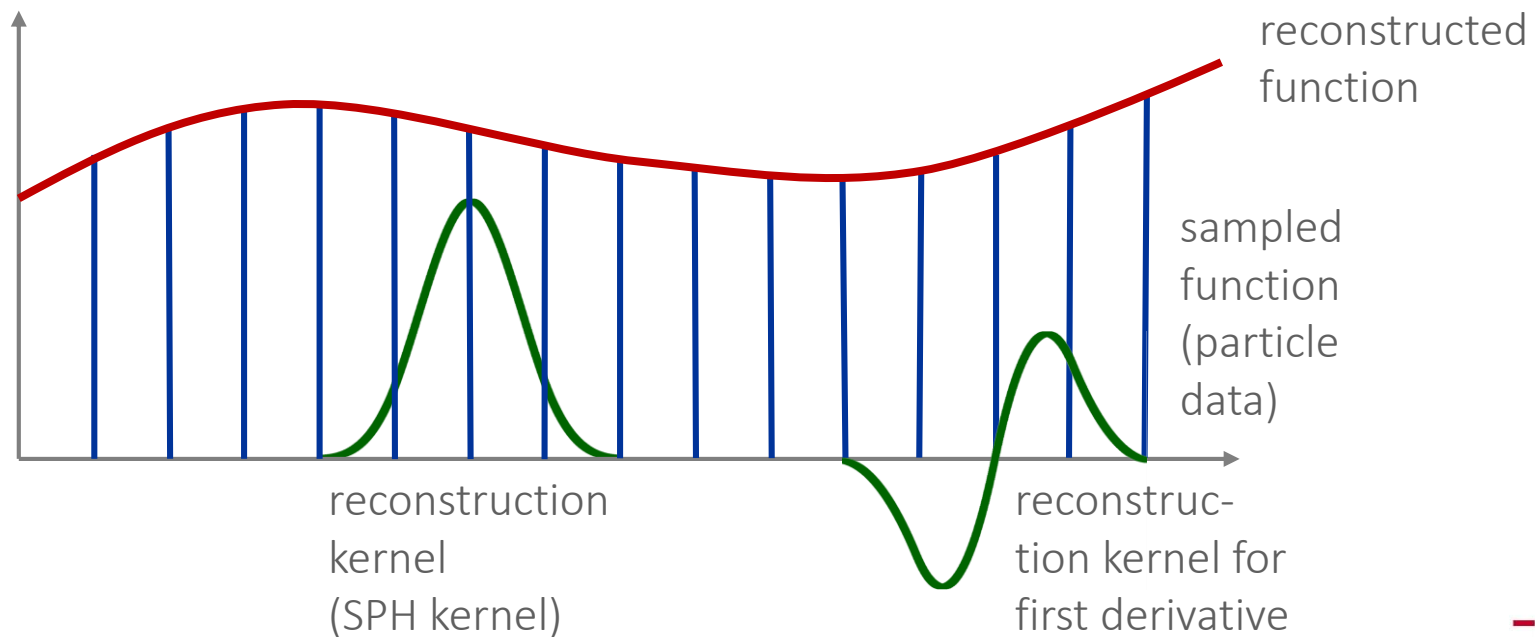
- $\frac{d\mathbf{v}_i}{dt} = -\frac{1}{\rho_i} \nabla p_i + \nu \nabla^2 \mathbf{v}_i + \frac{\mathbf{F}_i^{other}}{m_i}$

Outline

- concept of an SPH fluid simulator
- momentum equation
- SPH basics
- neighborhood search
- boundary handling
- incompressibility
- surface reconstruction

Illustration

- approximate a function and its derivatives from discrete samples, e.g. $\rho, \nabla p, \nabla^2 \mathbf{v}$
- convolution of discrete samples with reconstruction filter, e.g. cubic spline



Derivation

- quantity A at position $\mathbf{x}$ can be written as

$$A(\mathbf{x}) = \int_{\Omega} A(\mathbf{x}') \delta(\mathbf{x} - \mathbf{x}') d\mathbf{x}'$$

- dirac delta $\delta(\mathbf{x}) = \delta(x)\delta(y)\delta(z)$ and $\delta(x) = \begin{cases} \infty & x = 0 \\ 0 & x \neq 0 \end{cases}$

- $\int_{-\infty}^{+\infty} \delta(x) dx = 1$

- dirac delta is approximated with a kernel function with limited local support h

$$A(\mathbf{x}) \approx \int_{\Omega_h} A(\mathbf{x}') W(\mathbf{x} - \mathbf{x}', h) d\mathbf{x}'$$

Kernel Function

- integral should be normalized (unity condition)
$$\int_{\Omega} W(\mathbf{x} - \mathbf{x}', h) d\mathbf{x}' = 1$$
- support should be compact
$$W(\mathbf{x} - \mathbf{x}', h) = 0 \text{ for } \|\mathbf{x} - \mathbf{x}'\| > h$$
- should be symmetric
$$W(\mathbf{x} - \mathbf{x}', h) = W(\mathbf{x}' - \mathbf{x}, h)$$
- should be non-negative
$$W(\mathbf{x} - \mathbf{x}', h) \geq 0$$
- should converge to the Dirac delta for $h \rightarrow 0$
- should be differentiable

Kernel Function

- close to a Gaussian, but with compact support
 - support typically between h and $3h$
- e.g. cubic spline (1D: $\alpha = \frac{1}{6h}$ 2D: $\alpha = \frac{5}{14\pi h^2}$ 3D: $\alpha = \frac{1}{4\pi h^3}$)
$$W(q) = \alpha \begin{cases} (2-q)^3 - 4(1-q)^3 & 0 \leq q < 1 \\ (2-q)^3 & 1 \leq q < 2 \\ 0 & q \geq 2 \end{cases} \quad q = \frac{\|\mathbf{x}_i - \mathbf{x}_j\|}{h}$$
- number of particles / samples that are considered in the interpolation depends on
 - dimensionality, kernel support, particle spacing
 - e.g., 3D, cubic spline support $2h$, particle spacing h result practically in 30-40 neighbors
 - number of neighbors should not be too small to appropriately sample the kernel function

First Kernel Derivative

- $\nabla W_{ij} = \left(\frac{\partial W_{ij}}{\partial x_{i,x}}, \frac{\partial W_{ij}}{\partial x_{i,y}}, \frac{\partial W_{ij}}{\partial x_{i,z}} \right)^T$

$$\nabla W_{ij} = \frac{\partial W(q)}{\partial q} \nabla q$$

- e.g. cubic spline (1D: $\alpha = \frac{1}{6h}$ 2D: $\alpha = \frac{5}{14\pi h^2}$ 3D: $\alpha = \frac{1}{4\pi h^3}$)

$$\nabla q = \frac{\mathbf{x}_{ij}}{\|\mathbf{x}_{ij}\|h}$$

$$\frac{\partial W(q)}{\partial q} = \alpha \begin{cases} -3(2-q)^2 + 12(1-q)^2 & 0 \leq q < 1 \\ -3(2-q)^2 & 1 \leq q < 2 \\ 0 & q \geq 2 \end{cases}$$

$$\nabla W_{ij} = \alpha \frac{\mathbf{x}_{ij}}{\|\mathbf{x}_{ij}\|h} \begin{cases} -3(2-q)^2 + 12(1-q)^2 & 0 \leq q < 1 \\ -3(2-q)^2 & 1 \leq q < 2 \\ 0 & q \geq 2 \end{cases}$$

Second Kernel Derivative

- $\nabla^2 W_{ij} = \nabla \cdot (\nabla W_{ij}) = \frac{\partial^2 W_{ij}}{\partial x_{i,x}^2} + \frac{\partial^2 W_{ij}}{\partial x_{i,y}^2} + \frac{\partial^2 W_{ij}}{\partial x_{i,z}^2}$

$$\nabla^2 W_{ij} = \frac{\partial^2 W(q)}{\partial q^2} (\nabla q)^2 + \frac{\partial W(q)}{\partial q} (\nabla \cdot (\nabla q))$$

- e.g. cubic spline (1D: $\alpha = \frac{1}{6h}$ 2D: $\alpha = \frac{5}{14\pi h^2}$ 3D: $\alpha = \frac{1}{4\pi h^3}$)

$$(\nabla q)^2 = \frac{\mathbf{x}_{ij}}{\|\mathbf{x}_{ij}\|h} \cdot \frac{\mathbf{x}_{ij}}{\|\mathbf{x}_{ij}\|h} = \frac{\|\mathbf{x}_{ij}\|^2}{\|\mathbf{x}_{ij}\|^2 h^2} = \frac{1}{h^2}$$

$$\nabla \cdot (\nabla q) = \frac{2}{h\|\mathbf{x}\|}$$

$$\frac{\partial W(q)}{\partial q} = \alpha \begin{cases} -3(2-q)^2 + 12(1-q)^2 & 0 \leq q < 1 \\ -3(2-q)^2 & 1 \leq q < 2 \\ 0 & q \geq 2 \end{cases}$$

$$\frac{\partial^2 W(q)}{\partial q^2} = \alpha \begin{cases} 6(2-q) - 24(1-q) & 0 \leq q < 1 \\ 6(2-q) & 1 \leq q < 2 \\ 0 & q \geq 2 \end{cases}$$

Design of a Kernel Function 1D

- shape close to a Gaussian, e.g.

- $\alpha \tilde{W}\left(\frac{\|\mathbf{x}_i - \mathbf{x}_j\|}{h}\right) = \alpha \tilde{W}\left(\frac{x}{h}\right) = \alpha \tilde{W}(q) = W(q) = \alpha \begin{cases} (2-q)^3 - 4(1-q)^3 & 0 \leq q < 1 \\ (2-q)^3 & 1 \leq q < 2 \\ 0 & q \geq 2 \end{cases}$

- $2 \int_0^{2h} \alpha \tilde{W}(x) dx = 2 \int_0^2 \alpha \tilde{W}(q) h dq = 1$ integration by substitution

- $\alpha = \frac{1}{2 \int_0^2 \tilde{W}(q) h dq}$

- 1D: integration over a line segment $2 \int_0^2 \tilde{W}(q) h dq =$
 $2 \int_0^1 [(2-q)^3 - 4(1-q)^3] h dq + 2 \int_1^2 (2-q)^3 h dq = 2 \frac{11}{4} h + 2 \frac{1}{4} h$
 $\alpha = \frac{1}{6h}$

Design of a Kernel Function

- 2D: integration over the area of a circle

$$\begin{aligned}\int_0^{2\pi} \int_0^{2h} \tilde{W}(x) x \, dx d\phi &= \int_0^{2\pi} \int_0^2 \tilde{W}(q) h q h \, dq d\phi = \\ 2\pi \int_0^1 [q(2-q)^3 - 4q(1-q)^3] h^2 dq &+ 2\pi \int_1^2 q(2-q)^3 h^2 dq = 2\pi \frac{11}{10} h^2 + 2\pi \frac{3}{10} h^2 \\ \alpha &= \frac{5}{14\pi h^2}\end{aligned}$$

- 3D: integration over the volume of a sphere

$$\begin{aligned}\int_0^{2\pi} \int_{-\frac{\pi}{2}}^{\frac{\pi}{2}} \int_0^{2h} \tilde{W}(x) x^2 \sin\theta \, dx d\theta d\phi &= \int_0^{2\pi} \int_{-\frac{\pi}{2}}^{\frac{\pi}{2}} \int_0^2 \tilde{W}(q) (qh)^2 h \sin\theta \, dq d\theta d\phi = \\ 4\pi \int_0^1 [q^2(2-q)^3 - 4q(1-q)^3] h^3 dq &+ 4\pi \int_1^2 q^2(2-q)^3 h^3 dq = 4\pi \frac{19}{30} h^3 + 4\pi \frac{11}{30} h^3 \\ \alpha &= \frac{1}{4\pi h^3}\end{aligned}$$

Particle Approximation

- $$A(\mathbf{x}) \approx \int_{\Omega_h} A(\mathbf{x}') W(\mathbf{x} - \mathbf{x}', h) d\mathbf{x}'$$
$$= \int_{\Omega_h} \frac{A(\mathbf{x}')}{\rho(\mathbf{x}')} W(\mathbf{x} - \mathbf{x}', h) \rho(\mathbf{x}') d\mathbf{x}'$$
- consider a limited number of samples / particles $\mathbf{x}_j$ representing a mass $m(\mathbf{x}_j) = \rho(\mathbf{x}_j) V(\mathbf{x}_j)$
$$A(\mathbf{x}_i) \approx \sum_j A(\mathbf{x}_j) W(\mathbf{x}_i - \mathbf{x}_j, h) V(\mathbf{x}_j)$$
$$A(\mathbf{x}_i) \approx \sum_j \frac{A(\mathbf{x}_j)}{\rho(\mathbf{x}_j)} W(\mathbf{x}_i - \mathbf{x}_j, h) m(\mathbf{x}_j)$$
- typical notation
$$A_i = \sum_j \frac{m_j}{\rho_j} A_j W_{ij}$$

Spatial Derivatives

- $\nabla_{\mathbf{x}} A(\mathbf{x}) \approx \int_{\Omega_h} [\nabla_{\mathbf{x}'} A(\mathbf{x}')] W(\mathbf{x} - \mathbf{x}', h) d\mathbf{x}'$
- $\nabla_{\mathbf{x}'} [A(\mathbf{x}') W(\mathbf{x}' - \mathbf{x}, h)] = [\nabla_{\mathbf{x}'} A(\mathbf{x}')] W(\mathbf{x}' - \mathbf{x}, h) + A(\mathbf{x}') \nabla_{\mathbf{x}'} W(\mathbf{x}' - \mathbf{x}, h)$
W is symmetric
$$\nabla_{\mathbf{x}'} [A(\mathbf{x}') W(\mathbf{x}' - \mathbf{x}, h)] = [\nabla_{\mathbf{x}'} A(\mathbf{x}')] W(\mathbf{x} - \mathbf{x}', h) + A(\mathbf{x}') \nabla_{\mathbf{x}'} W(\mathbf{x}' - \mathbf{x}, h)$$
$$[\nabla_{\mathbf{x}'} A(\mathbf{x}')] W(\mathbf{x} - \mathbf{x}', h) = \nabla_{\mathbf{x}'} [A(\mathbf{x}') W(\mathbf{x}' - \mathbf{x}, h)] - A(\mathbf{x}') \nabla_{\mathbf{x}'} W(\mathbf{x}' - \mathbf{x}, h)$$
- $\int_{\Omega_h} \nabla_{\mathbf{x}'} [A(\mathbf{x}') W(\mathbf{x}' - \mathbf{x}, h)] d\mathbf{x}' = \int_S A(\mathbf{x}') W(\mathbf{x}' - \mathbf{x}, h) \mathbf{n} dS$ Gauss theorem
S is the surface of Ω
$$\int_S A(\mathbf{x}') W(\mathbf{x}' - \mathbf{x}, h) \mathbf{n} dS = 0 \quad W = 0 \text{ on the surface } S$$
$$\nabla_{\mathbf{x}} A(\mathbf{x}) \approx - \int_{\Omega_h} A(\mathbf{x}') \nabla_{\mathbf{x}'} W(\mathbf{x}' - \mathbf{x}, h) d\mathbf{x}' = \int_{\Omega_h} A(\mathbf{x}') \nabla_{\mathbf{x}} W(\mathbf{x} - \mathbf{x}', h) d\mathbf{x}'$$
- $\nabla_x A(\mathbf{x}_i) \approx \sum_j A(\mathbf{x}_j) \nabla W(\mathbf{x}_i - \mathbf{x}_j, h) V(\mathbf{x}_j)$
$$\nabla_x A(\mathbf{x}_i) \approx \sum_j \frac{m(\mathbf{x}_j)}{\rho(\mathbf{x}_j)} A(\mathbf{x}_j) \nabla W(\mathbf{x}_i - \mathbf{x}_j, h)$$

Spatial Derivatives

- original forms

$$\nabla A_i = \sum_j \frac{m_j}{\rho_j} A_j \nabla W_{ij}$$

$$\nabla^2 A_i = \sum_j \frac{m_j}{\rho_j} A_j \nabla^2 W_{ij}$$

- however, resulting forces do not preserve momentum
and are not necessarily
zero for constant values $A_i = A_j$

Gradient (Anti-symmetric)

- momentum-preserving form

$$\nabla \left(\frac{A_i}{\rho_i} \right) = \frac{\rho_i \nabla A_i - A_i \nabla \rho_i}{\rho_i^2} = \frac{\nabla A_i}{\rho_i} - \frac{A_i \nabla \rho_i}{\rho_i^2}$$

$$\nabla A_i = \rho_i \left(\nabla \left(\frac{A_i}{\rho_i} \right) + \frac{A_i \nabla \rho_i}{\rho_i^2} \right)$$

- SPH approximation

$$\begin{aligned} \nabla A_i &= \rho_i \left(\sum_j \frac{m_j}{\rho_j} \frac{A_j}{\rho_j} \nabla W_{ij} + A_i \sum_j \frac{m_j}{\rho_j} \frac{\rho_j}{\rho_i^2} \nabla W_{ij} \right) \\ &= \rho_i \sum_j m_j \left(\frac{A_i}{\rho_i^2} + \frac{A_j}{\rho_j^2} \right) \nabla W_{ij} \end{aligned}$$

- applied to pressure gradient, linear and angular momentum is preserved for arbitrary samplings

example with two particles i and j

$$\mathbf{a}_i = m_j \left(\frac{A_i}{\rho_i^2} + \frac{A_j}{\rho_j^2} \right) \nabla W_{ij} = -m_i \left(\frac{A_j}{\rho_j^2} + \frac{A_i}{\rho_i^2} \right) \nabla W_{ji} = -\mathbf{a}_j \quad \nabla W_{ij} = -\nabla W_{ji}$$

Gradient (Symmetric)

- term that vanishes for constant function values

$$\nabla (\rho_i A_i) = \rho_i \nabla (A_i) + A_i \nabla (\rho_i)$$

$$\nabla A_i = \frac{1}{\rho_i} (\nabla (\rho_i A_i) - A_i \nabla \rho_i)$$

- SPH approximation

$$\begin{aligned} \nabla A_i &= \frac{1}{\rho_i} \left(\sum_j \frac{m_j}{\rho_j} \frac{A_j}{\rho_j} \nabla W_{ij} - A_i \sum_j \frac{m_j}{\rho_j} \rho_j \nabla W_{ij} \right) \\ &= \frac{1}{\rho_i} \sum_j m_j (A_j - A_i) \nabla W_{ij} = \frac{1}{\rho_i} \sum_j m_j A_{ji} \nabla W_{ij} \end{aligned}$$

- applied to velocity divergence, zero divergence
for a constant velocity field is obtained for arbitrary samplings

Laplacian

- second derivative is error prone and sensitive to particle disorder
- too few samples to appropriately approximate the second kernel derivative
- therefore, the Laplacian is typically approximated with a finite difference approximation of the first derivative

$$\nabla^2 A_i = 2 \sum_j \frac{m_j}{\rho_j} A_{ij} \frac{\mathbf{x}_{ij} \cdot \nabla W_{ij}}{\mathbf{x}_{ij} \cdot \mathbf{x}_{ij} + 0.01h^2}$$

$$A_{ij} = A_i - A_j$$

$$\mathbf{x}_{ij} = \mathbf{x}_i - \mathbf{x}_j$$

Spatial Derivatives - Summary

- original approximations

$$\nabla A_i = \sum_j \frac{m_j}{\rho_j} A_j \nabla W_{ij}$$

$$\nabla^2 A_i = \sum_j \frac{m_j}{\rho_j} A_j \nabla^2 W_{ij}$$

- currently preferred approximations

- improved robustness in case of particle disorder, i.e. $\sum_j \nabla W_{ij} \neq \mathbf{0}$

- $\nabla p_i = \rho_i \sum_j m_j \left(\frac{p_i}{\rho_i^2} + \frac{p_j}{\rho_j^2} \right) \nabla W_{ij}$ preserves linear and angular momentum

- $\nabla^2 \mathbf{v}_i = 2 \sum_j \frac{m_j}{\rho_j} \mathbf{v}_{ij} \frac{\mathbf{x}_{ij} \cdot \nabla W_{ij}}{\mathbf{x}_{ij} \cdot \mathbf{x}_{ij} + 0.01 h^2}$ improved robustness as it avoids the second kernel derivative

- $\nabla \cdot \mathbf{v}_i = -\frac{1}{\rho_i} \sum_j m_j \mathbf{v}_{ij} \nabla W_{ij}$ zero for uniform velocity field

- $\mathbf{v}_{ij} = \mathbf{v}_i - \mathbf{v}_j \quad \mathbf{x}_{ij} = \mathbf{x}_i - \mathbf{x}_j$

Kernel Properties

- in case of ideal sampling
- $\rho_i = \sum_j m_j W_{ij} = m_i \sum_j W_{ij} \quad m_i = m_j$
- $m_i \sum_j W_{ij} = \rho_i = \frac{m_i}{V_i} \Rightarrow \sum_j W_{ij} = \frac{1}{V_i}$
- $\nabla W_{ij} = -\nabla W_{ji} \quad \nabla W_{ij} = \alpha \frac{\mathbf{x}_{ij}}{\|\mathbf{x}_{ij}\|^h} \dots$
- $\sum_j \nabla W_{ij} = \mathbf{0}$

Kernel Illustration

- 1D illustration

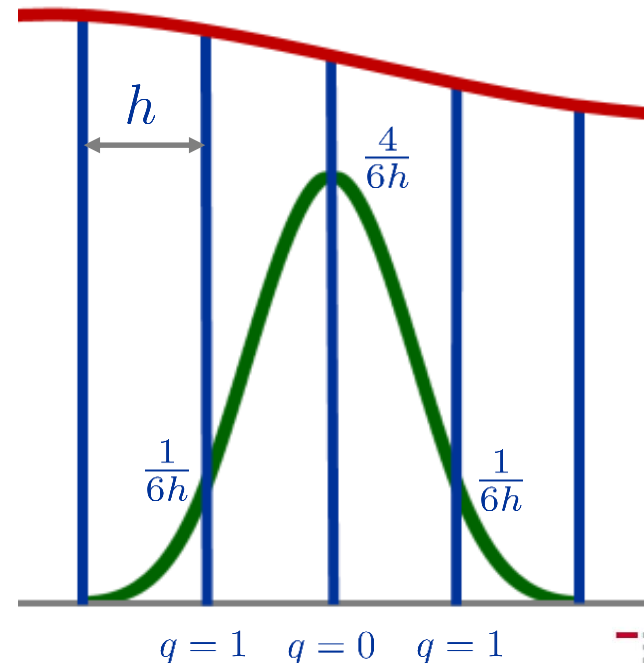
- $$W(q) = \frac{1}{6h} \begin{cases} (2-q)^3 - 4(1-q)^3 & 0 \leq q < 1 \\ (2-q)^3 & 1 \leq q < 2 \\ 0 & q \geq 2 \end{cases} \quad q = \frac{\|\mathbf{x}_i - \mathbf{x}_j\|}{h}$$

- $$W(0) = \frac{1}{6h} ((2-0)^3 - 4(1-0)^3) = \frac{4}{6h}$$

$$W(1) = \frac{1}{6h} (2-1)^3 = \frac{1}{6h}$$

$$W(2) = 0$$

- $$\sum_j W_{ij} = W(0) + 2W(1) + 2W(2) = \frac{1}{h}$$



Kernel Illustration

■ 2D illustration

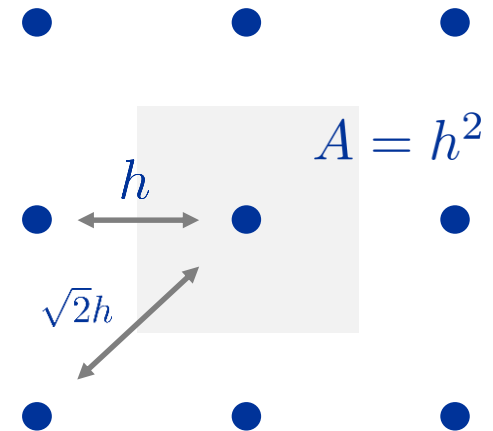
$$W(q) = \frac{5}{14\pi h^2} \begin{cases} (2-q)^3 - 4(1-q)^3 & 0 \leq q < 1 \\ (2-q)^3 & 1 \leq q < 2 \\ 0 & q \geq 2 \end{cases} \quad q = \frac{\|\mathbf{x}_i - \mathbf{x}_j\|}{h}$$

$$W(0) = \frac{5}{14\pi h^2} ((2-0)^3 - 4(1-0)^3) = \frac{20}{14\pi h^2}$$

$$W(1) = \frac{5}{14\pi h^2} (2-1)^3 = \frac{5}{14\pi h^2}$$

$$W(\sqrt{2}) = \frac{5}{14\pi h^2} (2-\sqrt{2})^3 \approx \frac{1.005}{14\pi h^2}$$

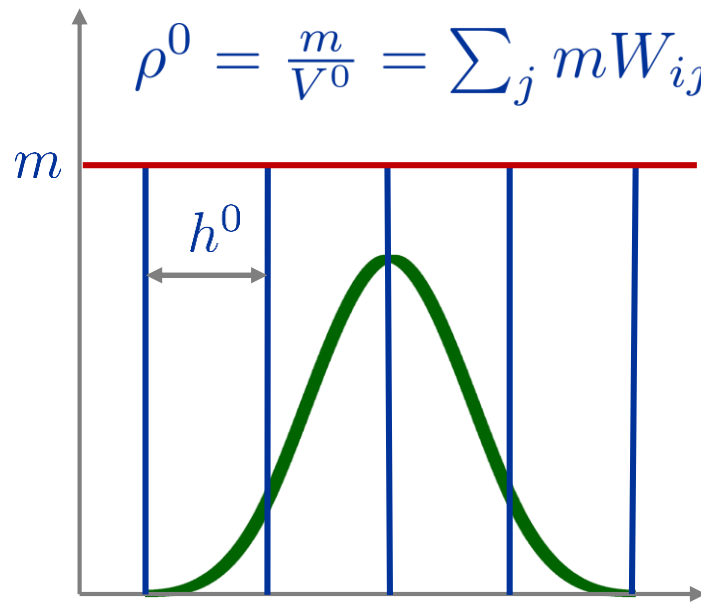
$$\sum_j W_{ij} = W(0) + 4W(1) + 4W(\sqrt{2}) \approx \frac{1.001}{h^2}$$



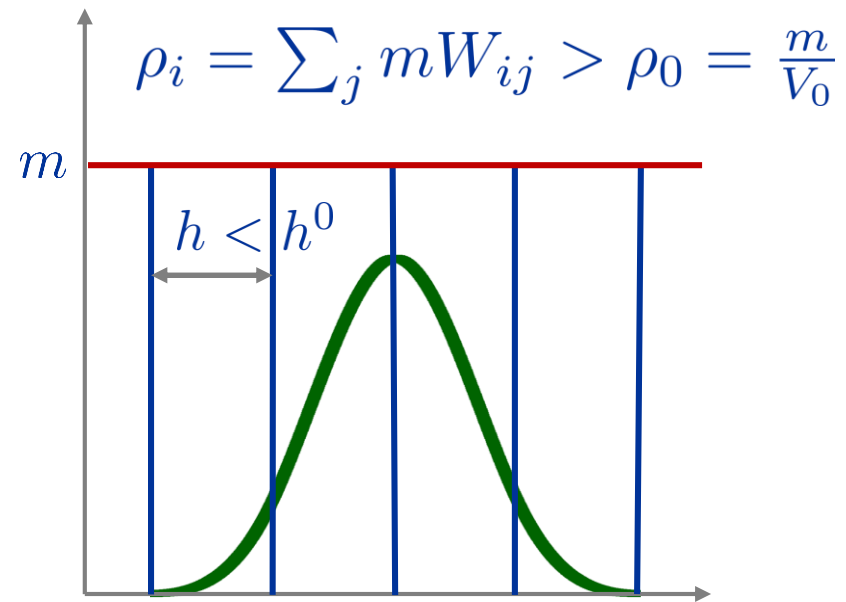
Kernel Illustration

- density computation

- is not an interpolation of the function m , but detects erroneous sampling



correct sampling



dense sampling
(kernel contributions do
not sum up to $1/V$)

Simulation in Computer Graphics

Particle-based Fluid Simulation

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Simple SPH Fluid Solver

- **for all particle i do**
 find neighbors j
for all particle i do
 $\rho_i = \sum_j m_j W_{ij}$
 compute p_i from ρ_i

$$\begin{aligned} \mathbf{a}_i^{\text{pressure}} &= - \sum_j m_j \left(\frac{p_i}{\rho_i^2} + \frac{p_j}{\rho_j^2} \right) \nabla W_{ij} \\ \mathbf{a}_i^{\text{viscosity}} &= 2\nu \sum_j \frac{m_j}{\rho_j} \mathbf{v}_{ij} \frac{\mathbf{x}_{ij} \cdot \nabla W_{ij}}{\mathbf{x}_{ij} \cdot \mathbf{x}_{ij} + 0.01h^2} \\ \mathbf{a}_i^{\text{other}} &= \mathbf{g} \end{aligned}$$

SPH approximations

- for all particle i do**

$$\begin{aligned} \mathbf{a}_i^{\text{pressure}} &= -\frac{1}{\rho_i} \nabla p_i \\ \mathbf{a}_i^{\text{viscosity}} &= \nu \nabla^2 \mathbf{v}_i \\ \mathbf{a}_i^{\text{other}} &= \mathbf{g} \\ \mathbf{a}_i(t) &= \mathbf{a}_i^{\text{pressure}} + \mathbf{a}_i^{\text{viscosity}} + \mathbf{a}_i^{\text{other}} \end{aligned}$$

Navier-Stokes
equation

- for all particle i do**

$$\begin{aligned} \mathbf{v}_i(t + \Delta t) &= \mathbf{v}_i(t) + \Delta t \mathbf{a}_i(t) \\ \mathbf{x}_i(t + \Delta t) &= \mathbf{x}_i(t) + \Delta t \mathbf{v}_i(t + \Delta t) \end{aligned}$$

Outline

- concept of an SPH fluid simulator
- momentum equation
- SPH basics
- neighborhood search
- boundary handling
- incompressibility
- surface reconstruction

SPH Simulation Step

Using a State Equation (SESPH)

- foreach particle do
 - compute density
 - compute pressure
- foreach particle do
 - compute forces
 - update velocities and positions
- density and force computation
process all neighbors of a particle

Neighbor Search

- for the computation of SPH sums in 3D, each particle needs to know at least 30-40 neighbors in each simulation step
- current scenarios
 - up to 30 million fluid particles
 - up to 1 billion neighbors
 - up to 10000 simulation steps
 - up to 10^{13} neighbors processed per simulation
- efficient construction and processing of dynamically changing neighbor sets is **essential**

Motivation



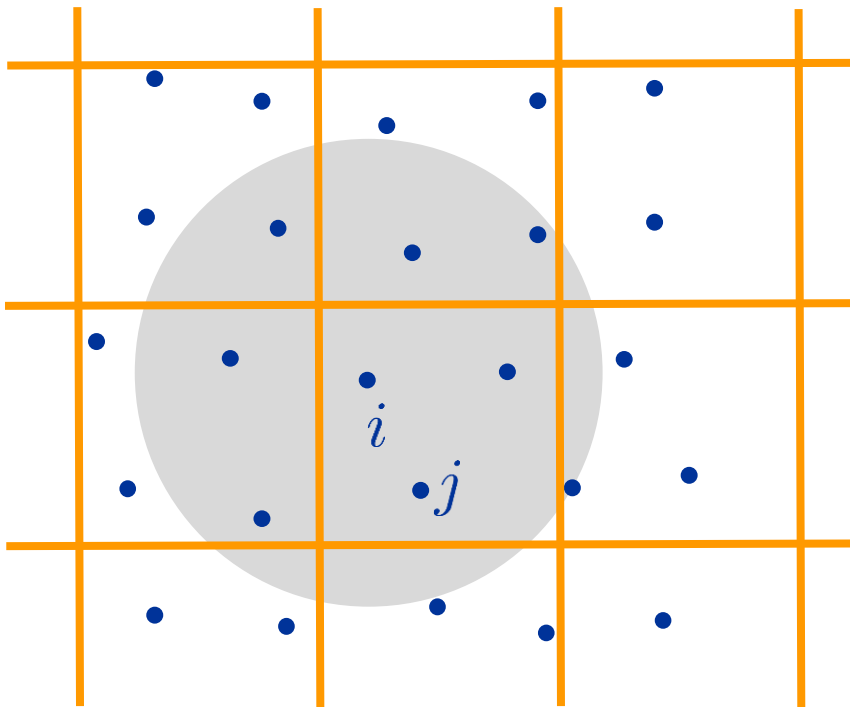
up to 30 million fluid particles, up to 1 billion neighbors,
11 s computation time for neighbor search on a 16-core PC

Characteristics

- SPH computes sums
 - dynamically changing sets of neighboring particles
 - temporal coherence
- spatial data structures accelerate the neighbor search
 - fast query
 - fast generation (at least once for each simulation step)
 - sparsely, non-uniformly filled simulation domain
- space subdivision
 - each particle is placed in a convex space cell, e.g. a cube
- similarities to collision detection and intersection tests in raytracing
 - however, cells adjacent to the cell of a particle have to be accessed

Characteristics

- hierarchical data structures are less efficient
 - construction in $O(n \log n)$, access in $O(\log n)$
- uniform grid is generally preferred
 - construction in $O(n)$, access in $O(1)$



Characteristics

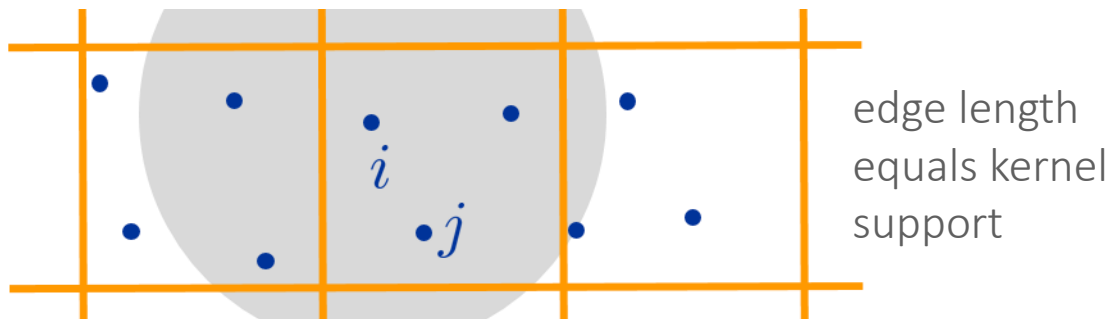
- Verlet lists
 - motivated by temporal coherence
 - potential neighbors are computed within a distance larger than the actual kernel support
 - actual neighbors are computed from the set of potential neighbors
 - potential neighbors are updated every n-th simulation step
 - memory-intensive (processes more neighbors than a standard grid)
- storing neighbors is generally expensive
 - might be avoided for, e.g., a low number of neighbor queries per simulation step or in case of very efficient computation
- data structures have to process
 - fluid particles of multiple phases, e.g. air
 - rigid particles (static or moving)
 - deformable particles

Outline

- concept of an SPH fluid simulator
- momentum equation
- SPH basics
- neighborhood search
 - uniform grid
 - index sort
 - spatial hashing
 - discussion
- boundary handling
- incompressibility
- surface reconstruction

Basic Grid

- particle is stored in a cell with coordinates (k, l, m)
- 27 cells are queried in the neighborhood search ($k \pm 1, l \pm 1, m \pm 1$)
- cell size equals the kernel support of a particle
 - larger cells increase the number of tested particles
 - smaller cells increase the number of tested cells
- parallel construction suffers from race conditions
 - insertion of particles from different threads in the same cell

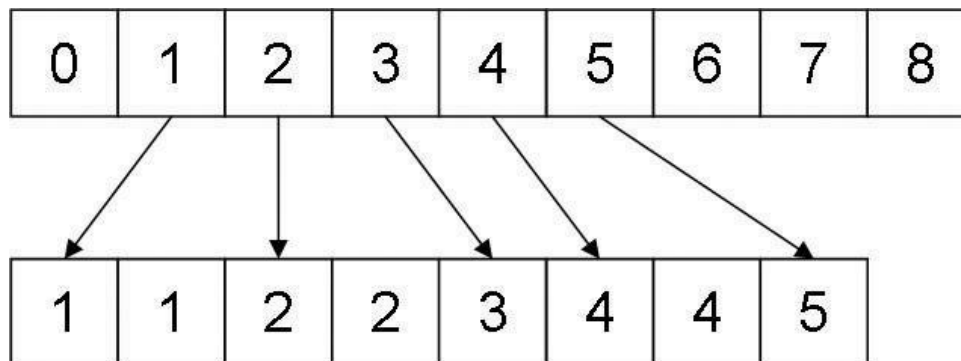


Outline

- concept of an SPH fluid simulator
- momentum equation
- SPH basics
- neighborhood search
 - uniform grid
 - index sort
 - spatial hashing
 - discussion
- incompressibility
- boundary handling
- surface reconstruction

Construction

- cell index $c = k + l \cdot K + m \cdot K \cdot L$
is computed for a particle
 - K and L denote the number of cells in x and y direction
- particles are sorted with respect to their cell index
 - e.g., radix sort, $O(n)$
- each grid cell (k, l, m) stores a reference to the first particle in the sorted list

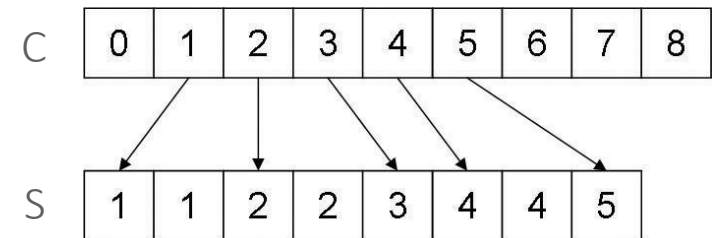


uniform grid

sorted particles with
their cell indices

Construction

- generate C
- store the number of particles in each cell of C
 - loop over all particles and increment the respective value in C
- accumulate the values in C
- generate S
- associate particle i with cell j:
 $S[\text{--}C[j]] = i$
 - stores the particles in reversed order into S
 - after insertion C contains the correct offsets



Construction

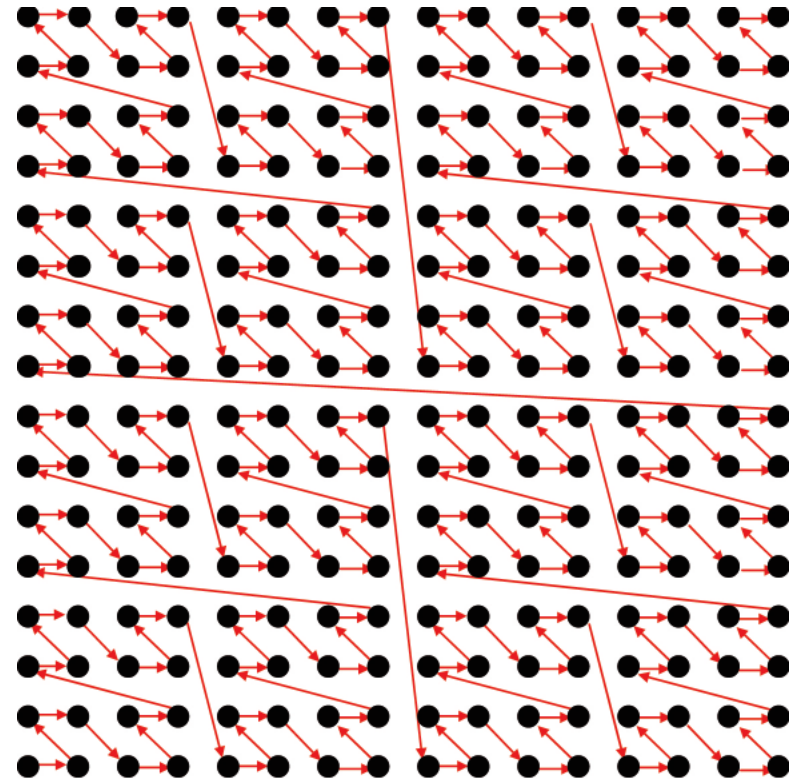
- parallelizable
- memory allocations are avoided
- constant memory consumption
- entire spatial grid has to be represented to find neighboring cells

Query

- sorted particle array is queried
 - parallelizable
- particles in the same cell are queried
- references to particles of adjacent cells are obtained from the references stored in the uniform grid
- improved cache-hit rate
 - particles in the same cell are close in memory
 - particles of neighboring cells are not necessarily close in memory

Z-Index Sort

- particles are sorted with respect to a z-curve index
- improved cache-hit rate
 - particles in adjacent cells are close in memory
- efficient computation of z-curve indices possible



z-curve

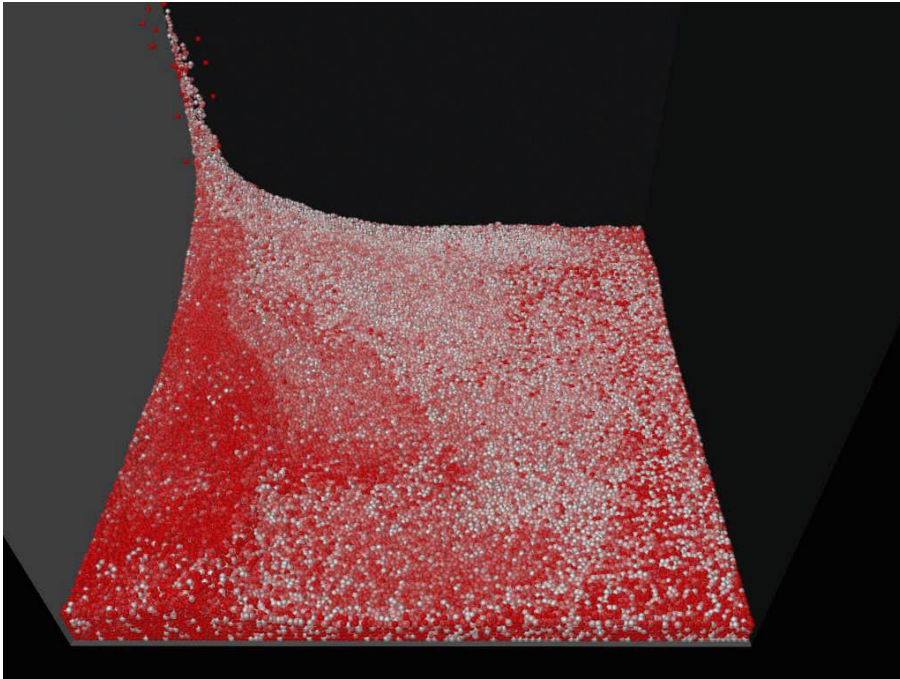
Z-Index Sort - Sorting

- particle attributes and z-curve indices can be processed separately
- handles (particle identifier, z-curve index) are sorted in each time step
 - reduces memory transfer
 - spatial locality is only marginally influenced due to temporal coherence
- attribute sets are sorted every n^{th} simulation step
 - restores spatial locality

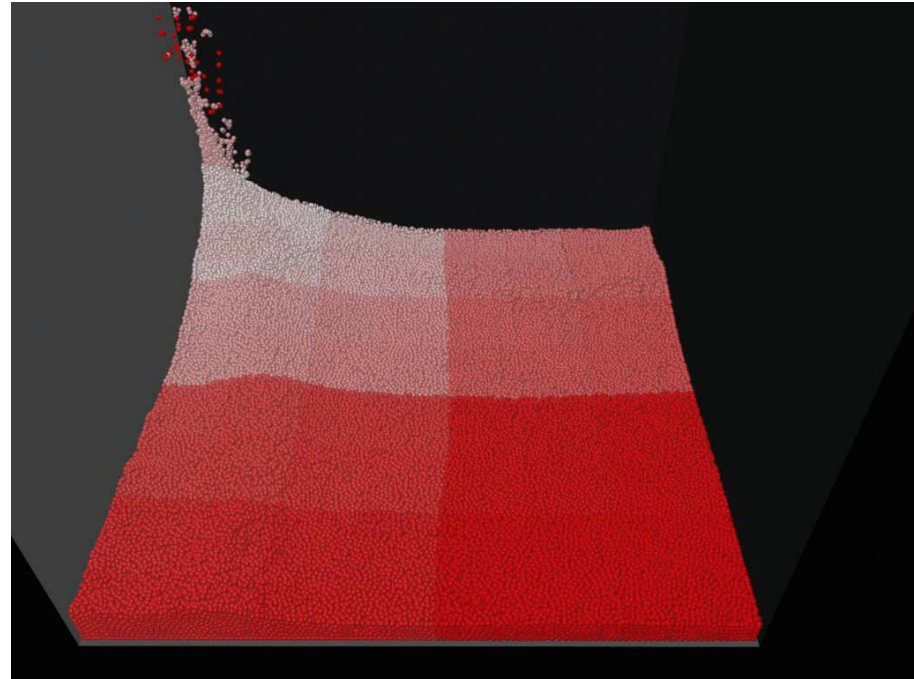
Z-Index Sort - Sorting

- instead of radix sort, insertion sort can be employed
 - $O(n)$ for almost sorted arrays
 - due to temporal coherence, only 2% of all particles change their cell, i.e. z-curve index, in each time step

Z-Index Sort - Reordering



particles colored according
to their location in memory



spatial compactness is
enforced using a z-curve

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Spatial Hashing

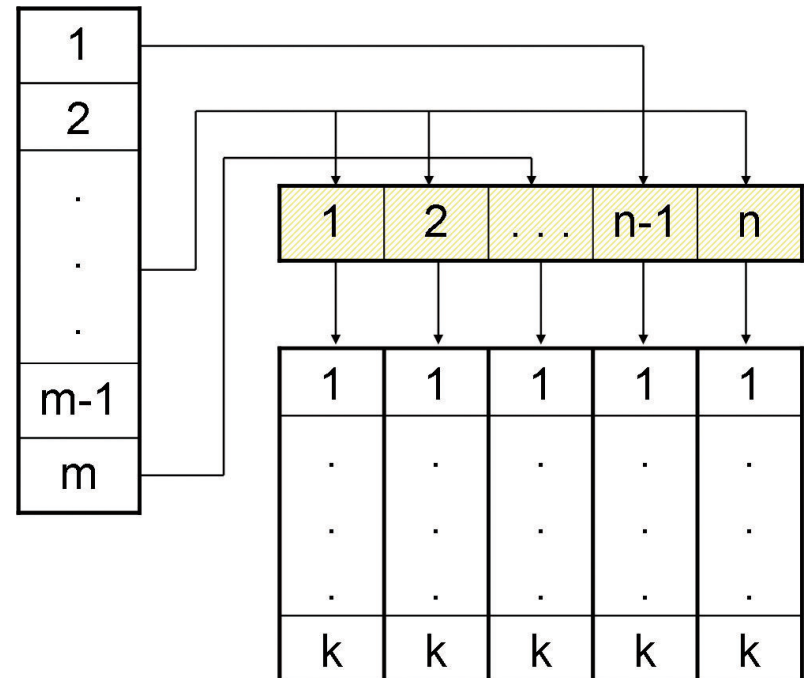
- hash function maps a grid cell to a hash cell
 - infinite 3D domain is mapped to a finite 1D list
 - in contrast to index sort, infinite domains can be handled
- implementation
 - compute a cell index c or a cell identifier (x, y, z) for a particle
 - compute a hash function $i = h(c)$ or $i = h(x, y, z)$
 - store the particle in a 1D array (hash table) at index i

Spatial Hashing

- large hash tables reduce number of hash collisions
 - hash collisions occur, if different spatial cells are mapped to the same hash cell
 - hash collisions slow down the query
- reduced memory allocations
 - memory for a certain number of entries is allocated for each hash cell
- reduced cache-hit rate
 - hash table is sparsely filled
 - filled and empty cells are alternating

Compact Hashing

- hash cells store handles to a compact list of used cells
 - k entries are pre-allocated for each element in the list of used cells
 - elements in the used-cell list are generated if a particle is placed in a new cell
 - elements are deleted, if a cell gets empty
- memory consumption is reduced from $O(m \cdot k)$ to $O(m + n \cdot k)$ with $m \gg n$
- list of used cells is queried in the neighbor search



Compact Hashing - Construction

- not parallelizable
 - particles from different threads might be inserted in the same cell
- larger hash table compared to spatial hashing to reduce hash collisions
- temporal coherence is employed
 - list of used cells is not rebuilt, but updated
 - set of particles with changed cell index is estimated (about 2% of all particles)
 - particle is removed from the old cell and added to the new cell (not parallelizable)

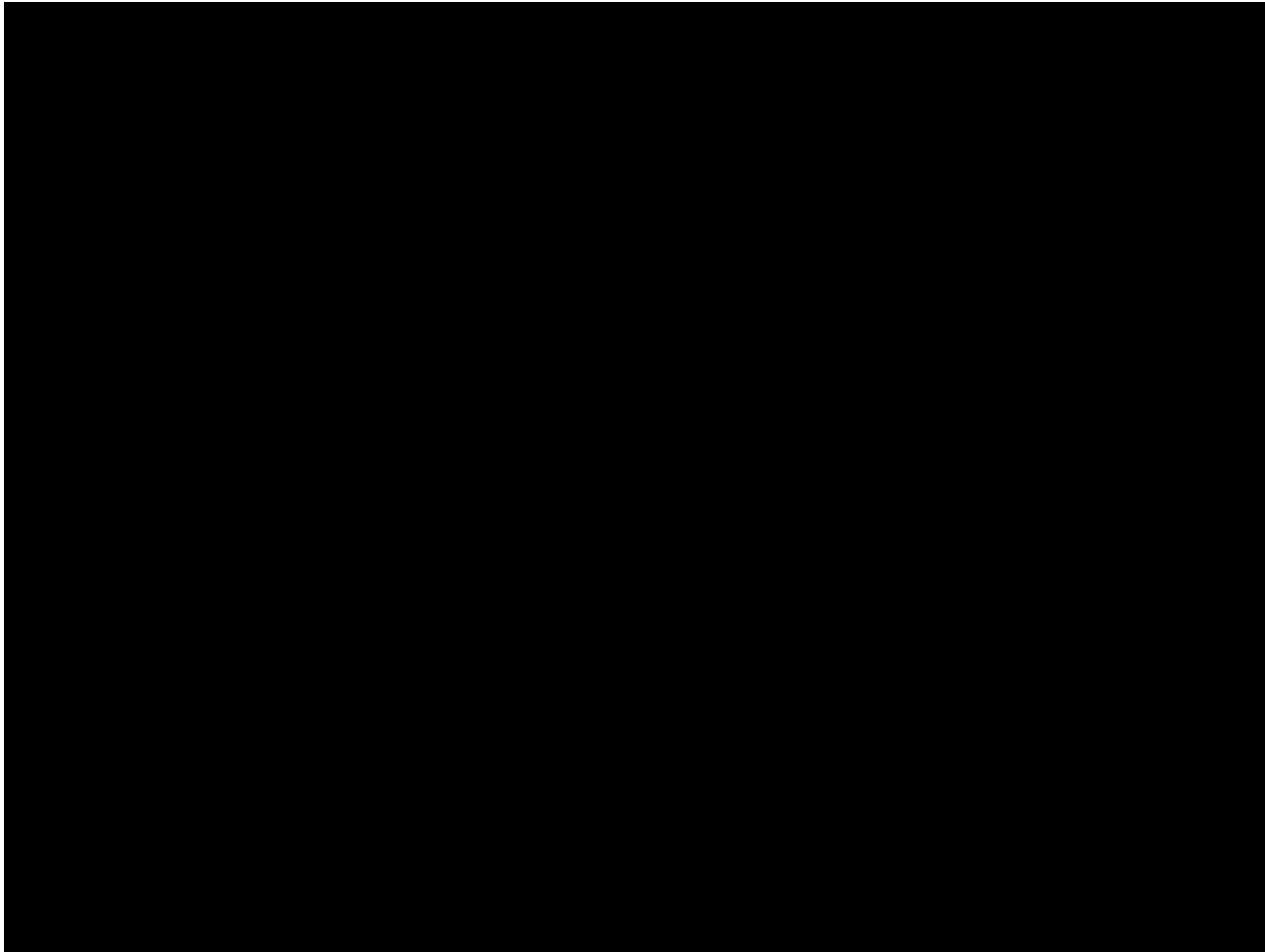
Compact Hashing - Query

- processing of used cells
 - bad spatial locality
 - used cells close in memory are not close in space
- hash-collision flag
 - if there is no hash collision in a cell, hash indices of adjacent cells have to be computed only once for all particles in this cell
 - large hash table results in 2% cells with hash collisions

Compact Hashing - Query

- particles are sorted with respect to a z-curve every n^{th} step
- after sorting, the list of used cells has to be rebuilt
- if particles are serially inserted into the list of used cells, the list is consistent with the z-curve
 - improved cache hit rate during the traversal of the list of used cells

Compact Hashing - Reordering



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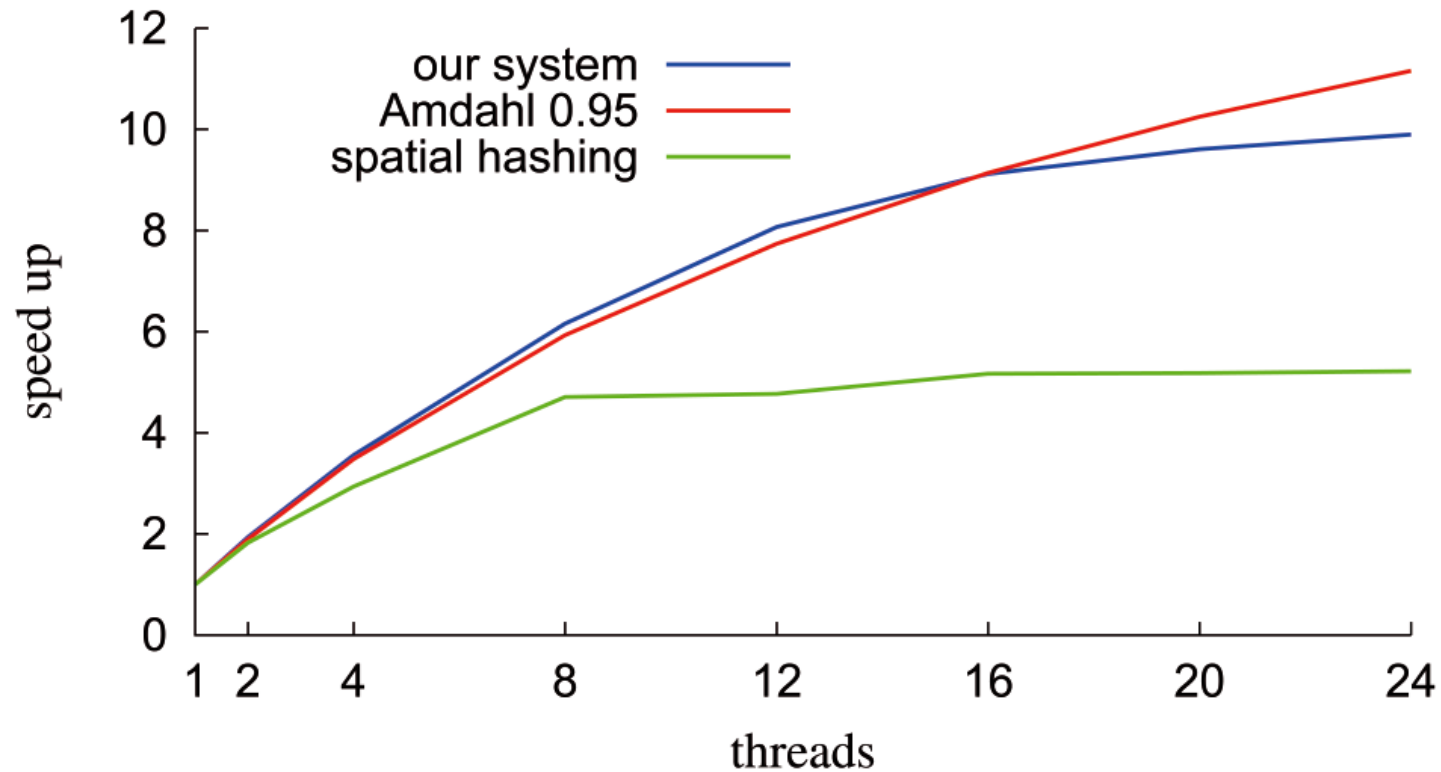
Comparison

method	construction	query	total
basic grid	26	38	64
index sort	36	29	65
z-index sort	16	27	43
spatial hashing	42	86	128
compact hashing	8	32	40

- measurements in ms for 130K particles
- ongoing research
 - currently, compact hashing is used, compact list stores references to a sorted particle list



Parallel Scaling



Discussion

- index sort
 - fast query as particles are processed in the order of cell indices
- z-index sort
 - fast construction due to radix sort or insertion sort of an almost sorted list
 - sorting with respect to the z-curve improves cache-hit rate
- spatial hashing
 - slow query due to hash collisions and due to the traversal of the sparsely filled hash table
- compact hashing
 - fast construction (update) due to temporal coherence
 - fast query due to the compact list of used cells, due to the hash-collision flag and due to z-curve

References

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 - IHMSEN M., AKINCI N., BECKER M., TESCHNER M.: A Parallel SPH Implementation on Multi-core CPUs. *Computer Graphics Forum*, 2011.
- index sort
 - PURCELL T. J., DONNER C., CAMMARANO M., JENSEN H. W., HANRAHAN P.: Photon Mapping on Programmable Graphics Hardware. *ACM SIGGRAPH/EUROGRAPHICS Conference on Graphics Hardware*, 2003.
- spatial hashing
 - TESCHNER M., HEIDELBERGER B., MÜLLER M., POMERANETS D., GROSS M.: Optimized Spatial Hashing for Collision Detection of Deformable Objects. *Vision, Modeling, Visualization*, 2003.

Simulation in Computer Graphics

Particle-based Fluid Simulation

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Albert-Ludwigs-Universität Freiburg

Simple SPH Fluid Solver

- **for all particle i do**
 find neighbors j
for all particle i do
 $\rho_i = \sum_j m_j W_{ij}$
 compute p_i from ρ_i

$$\begin{aligned}\mathbf{a}_i^{pressure} &= -\sum_j m_j \left(\frac{p_i}{\rho_i^2} + \frac{p_j}{\rho_j^2} \right) \nabla W_{ij} \\ \mathbf{a}_i^{viscosity} &= 2\nu \sum_j \frac{m_j}{\rho_j} \mathbf{v}_{ij} \frac{\mathbf{x}_{ij} \cdot \nabla W_{ij}}{\mathbf{x}_{ij} \cdot \mathbf{x}_{ij} + 0.01h^2} \\ \mathbf{a}_i^{other} &= \mathbf{g}\end{aligned}$$

SPH approximations

- for all particle i do**

$$\begin{aligned}\mathbf{a}_i^{pressure} &= -\frac{1}{\rho_i} \nabla p_i \\ \mathbf{a}_i^{viscosity} &= \nu \nabla^2 \mathbf{v}_i \\ \mathbf{a}_i^{other} &= \mathbf{g} \\ \mathbf{a}_i(t) &= \mathbf{a}_i^{pressure} + \mathbf{a}_i^{viscosity} + \mathbf{a}_i^{other}\end{aligned}$$

Navier-Stokes
equation

- for all particle i do**

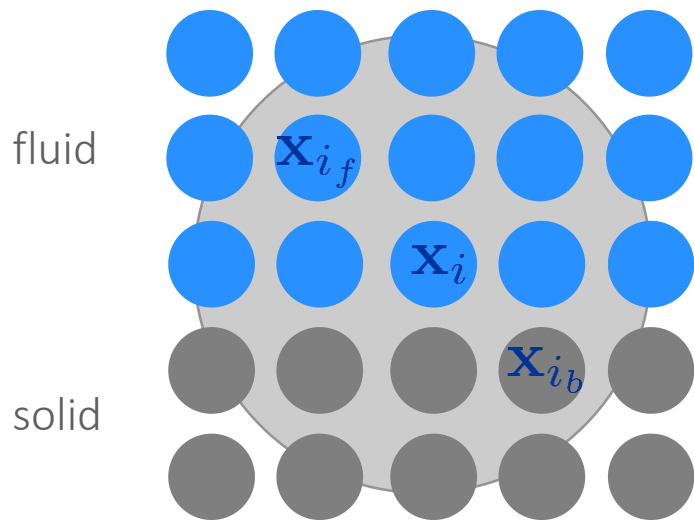
$$\begin{aligned}\mathbf{v}_i(t + \Delta t) &= \mathbf{v}_i(t) + \Delta t \mathbf{a}_i(t) \\ \mathbf{x}_i(t + \Delta t) &= \mathbf{x}_i(t) + \Delta t \mathbf{v}_i(t + \Delta t)\end{aligned}$$

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Concept

- rigids are uniformly sampled with particles



$$\rho_i = \sum_{i_f} m_{i_f} W_{ii_f} + \sum_{i_b} m_{i_b} W_{ii_b}$$

$$m_i = m_{i_f} = m_{i_b}$$

$$\rho_i \approx m_i \sum_{i_f} W_{ii_f} + m_i \sum_{i_b} W_{ii_b}$$

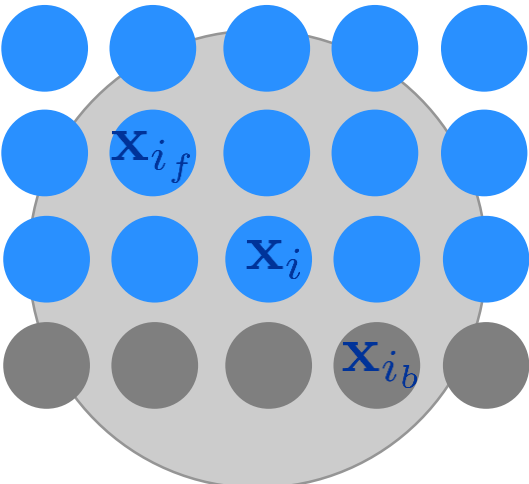
$$p_i = \dots$$

$$\mathbf{a}_i^p = - \sum_{i_f} m_{i_f} \left(\frac{p_i}{\rho_i^2} + \frac{p_{i_f}}{\rho_{i_f}^2} \right) \nabla W_{ii_f} - \sum_{i_b} m_{i_b} \left(\frac{p_i}{\rho_i^2} + \frac{p_{i_b}}{\rho_{i_b}^2} \right) \nabla W_{ii_b}$$

$$\mathbf{a}_i^p \approx -m_i \sum_{i_f} \left(\frac{p_i}{\rho_0^2} + \frac{p_{i_f}}{\rho_0^2} \right) \nabla W_{ii_f} - m_i \sum_{i_b} \left(\frac{p_i}{\rho_0^2} + \frac{p_{i_b}}{\rho_0^2} \right) \nabla W_{ii_b}$$

Missing Contributions

- rigids are uniformly sampled with particles



fluid

solid

$$\rho_i \approx m_i \sum_{i_f} W_{ii_f} + m_i \sum_{i_b} W_{ii_b} + x$$

$$\rho_i \approx m_i \sum_{i_f} W_{ii_f} + m_i \gamma_1 \sum_{i_b} W_{ii_b}$$

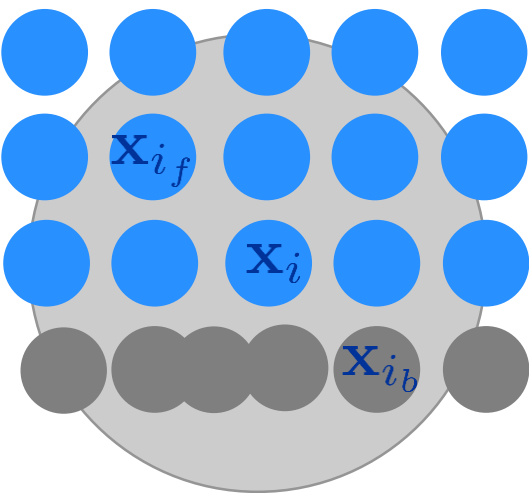
$$\sum_{i_f} W_{ii_f} + \gamma_1 \sum_{i_b} W_{ii_b} = \frac{1}{V_i}$$

$$\mathbf{a}_i^p \approx -m_i \sum_{i_f} \left(\frac{p_i}{\rho_0^2} + \frac{p_{i_f}}{\rho_0^2} \right) \nabla W_{ii_f} - m_i \gamma_2 \sum_{i_b} \left(\frac{p_i}{\rho_0^2} + \frac{p_{i_b}}{\rho_0^2} \right) \nabla W_{ii_b}$$

$$\sum_{i_f} \nabla W_{ii_f} + \gamma_2 \sum_{i_b} \nabla W_{ii_b} = \mathbf{0}$$

Non-uniform Sampling

- rigids are non-uniformly sampled with particles



The diagram illustrates a fluid-solid interface. A central gray circular region represents the solid, with a grid of blue circles representing fluid particles. The fluid particles are arranged in a grid, with some labeled $\mathbf{x}_{i_f}$ and $\mathbf{x}_i$. The solid particles are represented by gray circles at the bottom, with one labeled $\mathbf{x}_{i_b}$. The labels 'fluid' and 'solid' are on the left. To the right of the diagram are three equations:

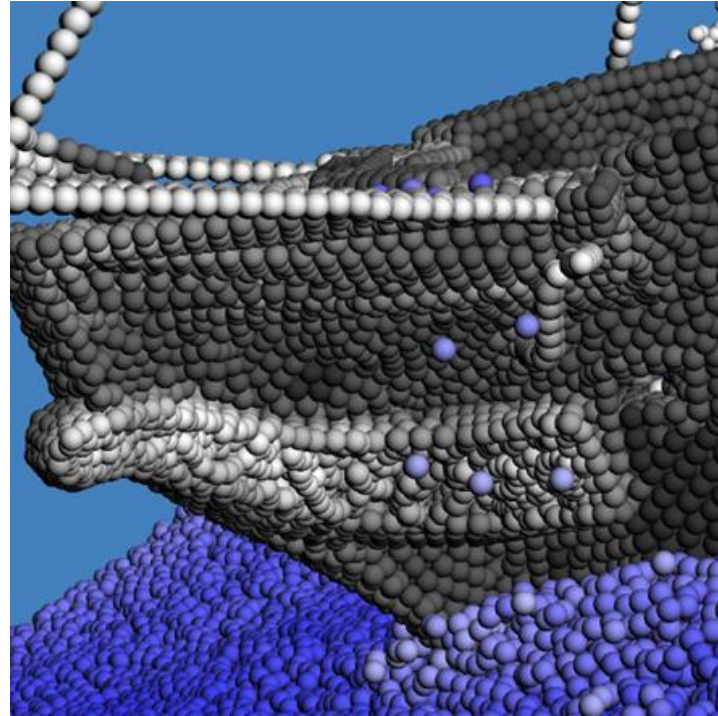
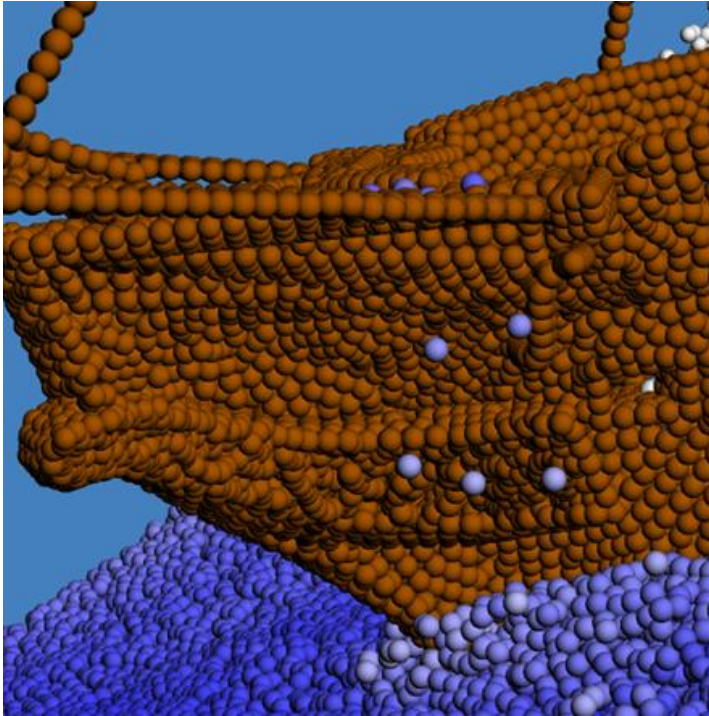
$$\rho_i \approx m_i \sum_{i_f} W_{ii_f} + \gamma_1 \sum_{i_b} m_{i_b} W_{ii_b}$$

$$\frac{1}{V_{i_b}} \approx \gamma_3 \sum_{i_{bb}} W_{i_b i_{bb}}$$

$$m_{i_b} = \frac{\rho_0}{\gamma_3 \sum_{i_{bb}} W_{i_b i_{bb}}}$$

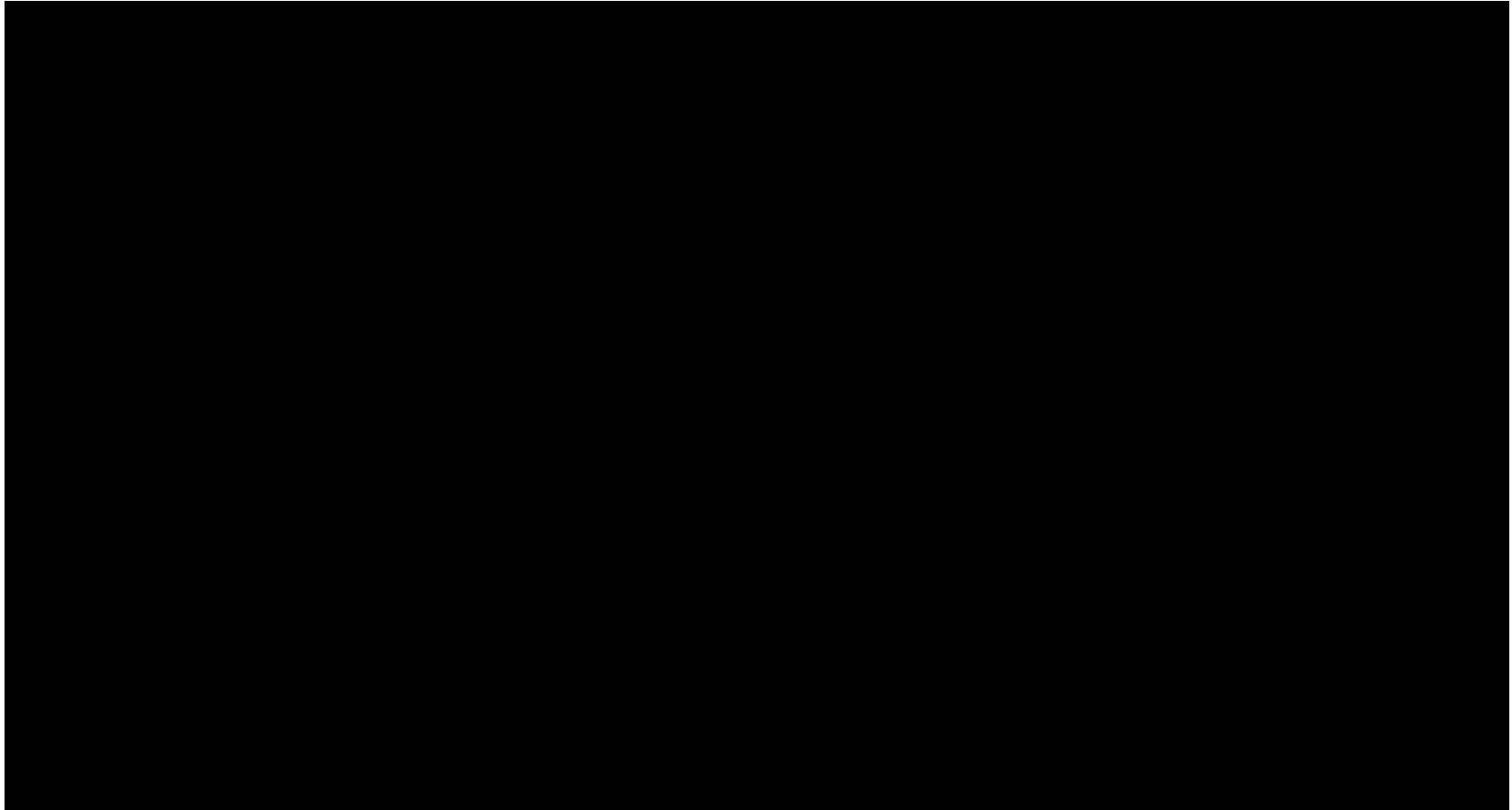
$$\mathbf{a}_i^p \approx -m_i \sum_{i_f} \left(\frac{p_i}{\rho_0^2} + \frac{p_{i_f}}{\rho_0^2} \right) \nabla W_{ii_f} - \gamma_2 \sum_{i_b} m_{i_b} \left(\frac{p_i}{\rho_0^2} + \frac{p_{i_b}}{\rho_0^2} \right) \nabla W_{ii_b}$$

Non-uniform Sampling



color-coded volume
of boundary particles

Rigid-Fluid Coupling



www.youtube.com/cgfreiburg

Rigid-Fluid Coupling



www.youtube.com/cgfreiburg

References

- rigid-fluid coupling

- IHMSEN M., AKINCI N., GISSLER M., TESCHNER M.: Boundary Handling and Adaptive Time-stepping for PCISPH. *VRIPHYS*, 2010.
- AKINCI N., IHMSEN M., AKINCI G., SOLENTHALER B., TESCHNER M.: Versatile Rigid-Fluid Coupling for Incompressible SPH. *ACM TOG (SIGGRAPH 2012)*, 2012.

- elastic-fluid coupling

- AKINCI N., CORNELIS J., AKINCI G., TESCHNER M.: Coupling Elastic Solids with Smoothed Particle Hydrodynamics Fluids. *Journal of Computer Animation and Virtual Worlds (CASA 2013)*, 2013.

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Incompressibility

- is **essential** for a realistic fluid behavior
 - less than 0.1% in typical scenarios
- inappropriate compression leads, e.g., to oscillations at the free surface
- compression is time-step dependent
 - volume changes should be imperceptible in adaptive time-stepping schemes
- is computationally expensive
 - simple computations require small time steps
 - large time steps require complex computations

State Equations (EOS, SESPH)

- pressure forces resolve compression induced by non-pressure forces (penalty approach)
 - density fluctuations in the fluid result in density gradients
 - density gradients result in pressure gradients
 - pressure gradients result in pressure force from high to low pressure
- fast computation, but small time steps
- pressure is computed from density, e.g.
 - $p_i = k(\rho_i - \rho_0)$ $p_i = k\rho_i$ $p_i = k\left(\frac{\rho_i}{\rho_0} - 1\right)$ k is user-defined
 - in graphics referred to as compressible SPH
 - $p_i = k_1\left(\left(\frac{\rho_i}{\rho_0}\right)^{k_2} - 1\right)$ k_1, k_2 are user-defined
 - in graphics referred to as weakly compressible SPH
 - compressibility is governed by the stiffness constant(s) and limits the time step

Non-iterative EOS solver (SESPH)

- for all *particle i* do
 - find neighbors *j*
- for all *particle i* do
 - $\rho_i = \sum_j m_j W_{ij}$
 - compute p_i from ρ_i with a state equation
- for all *particle i* do
 - $\mathbf{F}_i^{\text{pressure}} = -\frac{m_i}{\rho_i} \nabla p_i$
 - $\mathbf{F}_i^{\text{viscosity}} = m_i \nu \nabla^2 \mathbf{v}_i$
 - $\mathbf{F}_i^{\text{other}} = m_i \mathbf{g}$
 - $\mathbf{F}_i(t) = \mathbf{F}_i^{\text{pressure}} + \mathbf{F}_i^{\text{viscosity}} + \mathbf{F}_i^{\text{other}}$
- for all *particle i* do
 - $\mathbf{v}_i(t + \Delta t) = \mathbf{v}_i(t) + \Delta t \mathbf{F}_i(t) / m_i$
 - $\mathbf{x}_i(t + \Delta t) = \mathbf{x}_i(t) + \Delta t \mathbf{v}_i(t + \Delta t)$

SESPH with Splitting

- compute pressure after advecting the particles with non-pressure forces
- concept
 - compute all non-pressure forces $\mathbf{F}_i^{nonp}(t)$
 - compute intermediate velocity $\mathbf{v}_i^* = \mathbf{v}_i(t) + \Delta t \frac{\mathbf{F}_i^{nonp}}{m_i}$
 - compute intermediate position $\mathbf{x}_i^* = \mathbf{x}_i(t) + \Delta t \mathbf{v}_i^*$
 - compute intermediate density $\rho_i^*(\mathbf{x}_i^*)$
 - compute pressure p_i from intermediate density ρ_i^* using an EOS
 - compute final velocity $\mathbf{v}_i(t + \Delta t) = \mathbf{v}_i^* - \Delta t \frac{1}{\rho_i^*} \nabla p_i$
- motivation
 - consider competing forces
 - take (positive or negative) effects of non-pressure forces into account when computing the pressure forces

SESPH with Splitting

■ for all particle i do

find neighbors j

for all particle i do

$$\mathbf{F}_i^{viscosity} = m_i \nu \nabla^2 \mathbf{v}_i$$

$$\mathbf{F}_i^{other} = m_i \mathbf{g}$$

$$\mathbf{v}_i^* = \mathbf{v}_i(t) + \Delta t \frac{\mathbf{F}_i^{viscosity} + \mathbf{F}_i^{other}}{m_i}$$

for all particle i do

$$\rho_i^* = \sum_j m_j W_{ij} + \Delta t \sum_j m_j (\mathbf{v}_i^* - \mathbf{v}_j^*) \nabla W_{ij}$$

- follows from the continuity equation

compute p_i using ρ_i^*

- avoids neighbor search

- see next slide

for all particle i do

$$\mathbf{F}_i^{pressure} = -\frac{m_i}{\rho_i^*} \nabla p_i$$

for all particle i do

$$\mathbf{v}_i(t + \Delta t) = \mathbf{v}_i^* + \Delta t \mathbf{F}_i^{pressure} / m_i$$

$$\mathbf{x}_i(t + \Delta t) = \mathbf{x}_i(t) + \Delta t \mathbf{v}_i(t + \Delta t)$$

Differential Density Update

- continuity equation

$$\frac{D\rho_i}{Dt} = -\rho_0 \nabla \cdot \mathbf{v}_i$$

velocity divergence corresponds to an in- / outflow at a fluid element which corresponds to a density change

- time discretization

$$\frac{\rho_i^* - \rho_i(t)}{\Delta t} = -\rho_0 \nabla \cdot \mathbf{v}_i^*$$

- space discretization

$$\frac{\rho_i^* - \sum_j m_j W_{ij}}{\Delta t} = -\rho_0 \left(-\frac{1}{\rho_0} \sum_j m_j (\mathbf{v}_i^* - \mathbf{v}_j^*) \nabla W_{ij} \right)$$

- predicted density due to the divergence of $\mathbf{v}_i^*$

$$\rho_i^* = \sum_j m_j W_{ij} + \Delta t \sum_j m_j (\mathbf{v}_i^* - \mathbf{v}_j^*) \nabla W_{ij}$$

Iterative SESP with Splitting

- pressure forces are iteratively accumulated and refined
- concept
 - compute non-pressure forces, intermediate velocity and position
 - iteratively
 - compute intermediate density from intermediate position
 - compute pressure from intermediate density
 - compute pressure forces
 - update intermediate velocity and position
- motivation
 - parameterized by a desired density error, not by a stiffness constant
 - provides a fluid state with a guaranteed density error

Iterative SESP_H with Splitting

■ for all particle i do

find neighbors j

for all particle i do

$$\mathbf{F}_i^{\text{viscosity}} = m_i \nu \nabla^2 \mathbf{v}_i \quad \mathbf{F}_i^{\text{other}} = m_i \mathbf{g}$$

$$\mathbf{v}_i^* = \mathbf{v}_i(t) + \Delta t \frac{\mathbf{F}_i^{\text{viscosity}} + \mathbf{F}_i^{\text{other}}}{m_i} \quad \mathbf{x}_i^* = \mathbf{x}_i(t) + \Delta t \mathbf{v}_i^*$$

repeat

for all particle i do

compute ρ_i^* using $\mathbf{x}_i^*$

compute p_i using ρ_i^* , e.g. $p_i = k(\rho_i^* - \rho_0)$

compute ρ_{err} , e.g. average or maximum

for all particle i do

$$\mathbf{F}_i^{\text{pressure}} = -\frac{m_i}{\rho_i^*} \nabla p_i \quad \mathbf{v}_i^* = \mathbf{v}_i^* + \Delta t \frac{\mathbf{F}_i^{\text{pressure}}}{m_i} \quad \mathbf{x}_i^* = \mathbf{x}_i^* + \Delta t^2 \frac{\mathbf{F}_i^{\text{pressure}}}{m_i}$$

until $\rho_{err} < \eta$ user-defined density error

for all particle i do

$$\mathbf{v}_i(t + \Delta t) = \mathbf{v}_i^* \quad \mathbf{x}_i(t + \Delta t) = \mathbf{x}_i^*$$

Iterative SESP - Variants

- different quantities are accumulated
 - pressure forces (local Poisson SPH)
 - pressure (predictive-corrective SPH, PCISPH)
 - advantageous, if pressure is required for other computations
 - distances (position-based fluids, PBF)
 - $\Delta \mathbf{x}_i = -\frac{1}{\rho_0} \sum_j (\frac{p_i}{\beta_i} + \frac{p_j}{\beta_j}) \nabla W_{ij}$ β is a pre-computed constant
- different EOS and stiffness constants are used
 - $k = \frac{\rho_i^* r_i^2}{2\rho_0 \Delta t^2}$ in local Poisson SPH
 - $k = \frac{2\rho_0^2}{m_i^2 \cdot \Delta t^2 \sum_j \nabla W_{ij}^0 \cdot \sum_j \nabla W_{ij}^0 + \sum_j (\nabla W_{ij}^0 \cdot \nabla W_{ij}^0)}$ in PCISPH W^0 is precomputed
 - $k = 1$ in PBF ($p_i = \frac{\rho_i}{\rho_0} - 1$)

PCISPH - Motivation

- density at the next timestep should be rest density

$$\rho(t + \Delta t) = \rho_0 = \underbrace{\sum_j m_i W_{ij} + \Delta t \sum_j m_j (\mathbf{v}_i^* - \mathbf{v}_j^*) \nabla W_{ij}}_{\rho_i^*} + \Delta t \sum_j m_j (\mathbf{v}_i^p - \mathbf{v}_j^p) \nabla W_{ij}$$

- $\mathbf{v}_i^p = \Delta t \frac{\mathbf{F}_i^p}{m_i}$

$$\mathbf{F}_i^p = -m_i \sum_j \left(\frac{p_i}{\rho_i^2} - \frac{p_j}{\rho_j^2} \right) \nabla W_{ij} \approx -m_i^2 \frac{2p_i}{\rho_0^2} \sum_j \nabla W_{ij}$$

$$\rho_0 - \rho_i^* = \Delta t \cdot m_i \cdot \sum_j \left(-m_i^2 \frac{2p_i}{\rho_0^2} \sum_j \nabla W_{ij} + m_i^2 \frac{2p_j}{\rho_0^2} \sum_k \nabla W_{jk} \right) \nabla W_{ij}$$

- for particle j, do not consider all contributions, but only the contribution from i

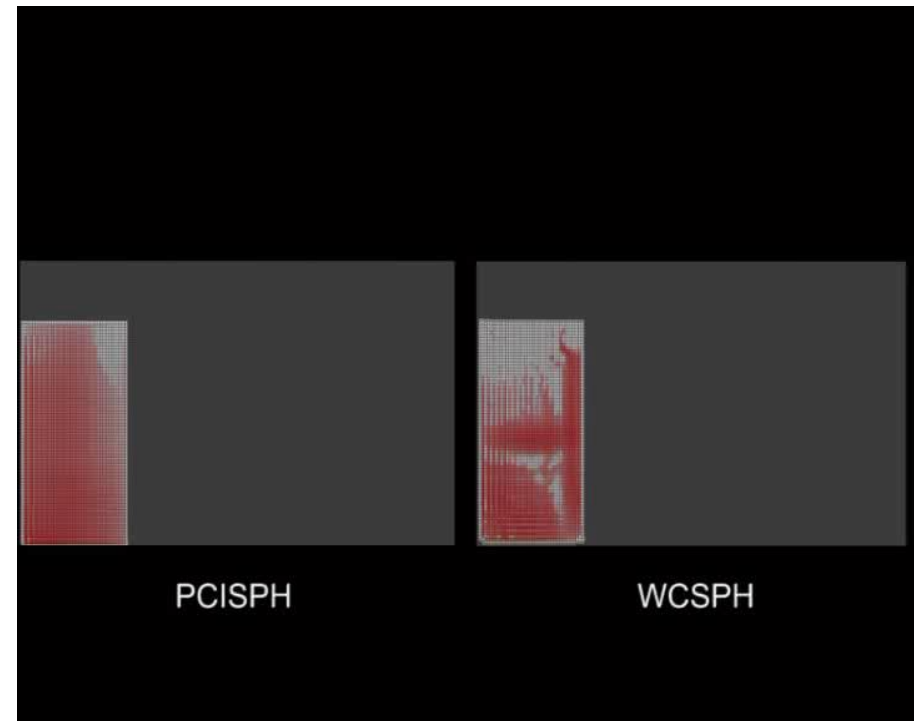
$$\mathbf{F}_{j|i}^p = -m_i \left(\frac{p_i}{\rho_i^2} - \frac{p_j}{\rho_j^2} \right) \nabla W_{ij} \approx -m_i^2 \frac{2p_i}{\rho_0^2} \sum_j \nabla W_{ij}$$

*Iterative SESP*H - Performance

- typically three to five iterations for density errors between 0.1% and 1%
- typical speed-up over non-iterative SESPH: 50
 - more computations per time step compared to SESPH
 - significantly larger time step than in SESPH
- EOS and stiffness constant influence the number of required iterations to get a desired density error
 - rarely analyzed
- non-linear relation between time step and iterations
 - largest possible time step does not necessarily lead to an optimal overall performance

Pressure Computation

- iterative SESP (PCISPH)
 - [Solenthaler 2009]
 - iterative pressure computation
 - large time step
- non-iterative SESP (WCSPH)
 - [Becker and Teschner 2007]
 - efficient to compute
 - small time step
- computation time for the PCISPH scenario is 20 times shorter than WCSPH



Projection Schemes

- compute pressure with a pressure Poisson equation
$$\Delta t \nabla^2 p_i = \rho_0 \nabla \cdot \mathbf{v}_i^* = \frac{1}{\Delta t} (\rho_0 - \rho_i^*)$$
- $\mathbf{v}_i^*$ is the predicted velocity considering all non-pressure forces
- ρ_i^* is the corresponding predicted density,
e.g. $\rho_i^* = \rho_i - \Delta t \cdot \rho_i \cdot \nabla \cdot \mathbf{v}_i^*$
- density invariance is preferred
- divergence-free schemes suffer from drift

Projection Schemes - Derivation

$$\blacksquare \quad \frac{d\mathbf{v}_i}{dt} = -\frac{1}{\rho_i} \nabla p_i + \nu \nabla^2 \mathbf{v}_i + \frac{\mathbf{F}_i^{\text{other}}}{m_i}$$

$$\mathbf{v}_i(t + \Delta t) = \mathbf{v}_i(t) + \Delta t \left(-\frac{1}{\rho_i} \nabla p_i + \nu \nabla^2 \mathbf{v}_i + \frac{\mathbf{F}_i^{\text{other}}}{m_i} \right)$$

$$\mathbf{v}_i^* = \mathbf{v}_i(t) + \Delta t \left(\nu \nabla^2 \mathbf{v}_i + \frac{\mathbf{F}_i^{\text{other}}}{m_i} \right)$$

$$\mathbf{v}_i(t + \Delta t) = \mathbf{v}_i^* - \Delta t \frac{1}{\rho_i} \nabla p_i$$

$$\frac{\mathbf{v}_i(t + \Delta t) - \mathbf{v}_i^*}{\Delta t} = -\frac{1}{\rho_i} \nabla p_i$$

$$\nabla \cdot \frac{\mathbf{v}_i(t + \Delta t) - \mathbf{v}_i^*}{\Delta t} = -\nabla \cdot \frac{1}{\rho_i} \nabla p_i$$

Projection Schemes - Derivation

- $$\nabla \cdot \frac{\mathbf{v}_i(t+\Delta t) - \mathbf{v}_i^*}{\Delta t} = -\nabla \cdot \frac{1}{\rho_i} \nabla p_i$$

$$\nabla \cdot \mathbf{v}_i(t + \Delta t) - \nabla \cdot \mathbf{v}_i^* = -\nabla \cdot \Delta t \frac{1}{\rho_i} \nabla p_i$$

$$\nabla \cdot \mathbf{v}_i(t + \Delta t) = 0$$

divergence of the velocity at the next time step should be zero.

$$\nabla \cdot \mathbf{v}_i^* = \Delta t \frac{1}{\rho_i} \nabla^2 p_i$$

Projection Schemes

- linear system with unknown pressure values
- iterative solvers
 - Conjugate Gradient
 - relaxed Jacobi
- fast computation per iteration
 - 30-40 non-zero entries in each equation
 - very few information per particle
 - matrix-free implementations
- huge time steps
- convergence tends to be an issue
 - up to 100 iterations, dependent on the formulation

Implicit Incompressible SPH

Derivation

- discretizing the continuity equation

$$\frac{D\rho_i(t+\Delta t)}{Dt} = -\rho_i(t+\Delta t)\nabla \cdot \mathbf{v}_i(t+\Delta t) \text{ to}$$

$$\frac{\rho_i(t+\Delta t) - \rho_i(t)}{\Delta t} = \sum_j m_j \mathbf{v}_{ij}(t+\Delta t) \nabla W_{ij}(t)$$

forward difference

SPH

- **unknown velocity** $\mathbf{v}_i(t+\Delta t)$ can be rewritten using known non-pressure accel. $\mathbf{a}_i^{nonp}(t)$ and **unknown pressure accel.** $\mathbf{a}_i^p(t)$

$$\mathbf{v}_i(t+\Delta t) = \underbrace{\mathbf{v}_i(t) + \Delta t \mathbf{a}_i^{nonp}(t)}_{\mathbf{v}_i^*} + \Delta t \mathbf{a}_i^p(t)$$

predicted velocity only
using non-pressure forces

Implicit Incompressible SPH

Derivation

- with $\rho_i(t + \Delta t) = \rho_0$ rest density is the desired density at $t + \Delta t$

and $\rho_i^* = \rho_i(t) + \Delta t \sum_j m_j \mathbf{v}_{ij}^* \nabla W_{ij}(t)$

predicted density, if only non-pressure forces are applied

the discretized continuity equation can be written as

$$\frac{\rho_0 - \rho_i^*}{\Delta t} = \Delta t \sum_j m_j (\mathbf{a}_i^p(t) - \mathbf{a}_j^p(t)) \nabla W_{ij}(t)$$

with unknown pressure accel. $\mathbf{a}_i^p(t)$

- unknown pressure accel. $\mathbf{a}_i^p(t)$ can be rewritten using unknown pressures $p_i(t)$

$$\mathbf{a}_i^p(t) = - \sum_j m_j \left(\frac{p_i(t)}{\rho_i^2(t)} - \frac{p_j(t)}{\rho_j^2(t)} \right) \nabla W_{ij}(t)$$

resulting in a linear system with unknown pressures

Implicit Incompressible SPH

Linear System

- one equation per particle

$$\sum_j a_{ij} p_j = s_i = \frac{\rho_0 - \rho_i^*}{\Delta t}$$

- iterative solver, e.g.,

$$p_i^{l+1} = (1 - \omega) p_i^l + \omega \frac{s_i - \sum_{j \neq i} a_{ij} p_j^l}{a_{ii}}$$

- relaxed Jacobi
- matrix-free implementation
- user-defined $\omega (= 0.5)$
- p_i^l is the computed pressure in iteration l
- p_i^0 is initialized, e.g., $p_i^0 = 0$ or $p_i^0 = p_i(t - \Delta t)$
- system is not necessarily symmetric
(a prerequisite for Conjugate Gradient)

Implicit Incompressible SPH Interpretation

- PPE $\Delta t^2 \nabla^2 p_i = \rho_0 - \rho_i^*$
 density change due to pressure accelerations predicted density error

- discretized PPE

- $\mathbf{A} \mathbf{p} = \mathbf{s}$

- $\underbrace{\Delta t^2 \sum_j m_j (\mathbf{a}_i^p - \mathbf{a}_j^p) \nabla W_{ij}}_{(\mathbf{A} \mathbf{p})_i} = \underbrace{\rho_0 - \rho_i^*}_{s_i} \quad \mathbf{a}_i^p = - \sum_j m_j \left(\frac{p_i}{\rho_i^2} - \frac{p_j}{\rho_j^2} \right) \nabla W_{ij}$

- $\Delta t \sum_j m_j (\Delta t \mathbf{a}_i^p - \Delta t \mathbf{a}_j^p) \nabla W_{ij} = \rho_0 - \rho_i^*$ pressure accel. causes a velocity change $\mathbf{v}^p$ whose divergence causes a density change
 - $\Delta t \sum_j m_j (\mathbf{v}_i^p - \mathbf{v}_j^p) \nabla W_{ij} = \rho_0 - \rho_i^*$
 - $\Delta t \cdot \rho_i \cdot \nabla \cdot \mathbf{v}_i^p = \rho_0 - \rho_i^*$

Implicit Incompressible SPH

Boundary Handling

- PPE

$$\Delta t^2 \nabla^2 p_f = \rho_0 - \rho_f^* = \rho_0 - \rho_f + \Delta t \rho_0 \nabla \cdot \mathbf{v}_f^*$$

- discretized PPE including boundary handling $\mathbf{A}\mathbf{p} = \mathbf{s}$

$$\mathbf{a}_f^p = - \sum_{f_f} m_{f_f} \left(\frac{p_f}{\rho_f^2} + \frac{p_{f_f}}{\rho_{f_f}^2} \right) \nabla W_{ff_f} - \gamma \sum_{f_b} m_{f_b} 2 \frac{p_f}{\rho_f^2} \nabla W_{ff_b}$$

$$\underbrace{\Delta t^2 \sum_{f_f} m_{f_f} \left(\mathbf{a}_f^p - \mathbf{a}_{f_f}^p \right) \nabla W_{ff_f} + \Delta t^2 \sum_{f_b} m_{f_b} \mathbf{a}_f^p \nabla W_{ff_b}}_{(\mathbf{A}\mathbf{p})_f} =$$

$$\underbrace{\rho_0 - \rho_f - \Delta t \sum_{f_f} m_{f_f} \left(\mathbf{v}_f^* - \mathbf{v}_{f_f}^* \right) \nabla W_{ff_f} - \Delta t \sum_{f_b} m_{f_b} \left(\mathbf{v}_f^* - \mathbf{v}_{f_b}(t + \Delta t) \right) \nabla W_{ff_b}}_{\mathbf{s}_f}$$

Implicit Incompressible SPH

Implementation with Boundary Handling

■ initialization

- density $\rho_f = \sum_{f_f} m_{f_f} W_{ff_f} + \sum_{f_b} m_{f_b} W_{ff_b}$
- predicted velocity $\mathbf{v}_f^* = \mathbf{v}_f + \Delta t \mathbf{a}_f^{nonp}$
- source term $s_f = \rho_0 - \rho_f - \Delta t \sum_{f_f} m_{f_f} (\mathbf{v}_f^* - \mathbf{v}_{f_f}^*) \nabla W_{ff_f} - \Delta t \sum_{f_b} m_{f_b} \dots$
- pressure $p_f^0 = 0$
- diagonal element of matrix **A**

$$\begin{aligned}
 a_{ff} = & \Delta t^2 \sum_{f_f} m_{f_f} \left(- \sum_{f_f} \frac{m_{f_f}}{\rho_0^2} \nabla W_{ff_f} - 2\gamma \sum_{f_b} \frac{m_{f_b}}{\rho_0^2} \nabla W_{ff_b} \right) \nabla W_{ff_f} \\
 & + \Delta t^2 \sum_{f_f} m_{f_f} \left(\frac{m_{f_f}}{\rho_0^2} \nabla W_{ff_f} \right) \nabla W_{ff_f} \\
 & + \Delta t^2 \sum_{f_b} m_{f_b} \left(- \sum_{f_f} \frac{m_{f_f}}{\rho_0^2} \nabla W_{ff_f} - 2\gamma \sum_{f_b} \frac{m_{f_b}}{\rho_0^2} \nabla W_{ff_b} \right) \nabla W_{ff_b}
 \end{aligned}$$

Implicit Incompressible SPH

Implementation with Boundary Handling

- iteration l

- first particle loop

- predicted pressure acceleration

$$(\mathbf{a}_f^p)^l = - \sum_{f_f} m_{f_f} \left(\frac{p_f^l}{\rho_f^2} + \frac{p_{ff}^l}{\rho_{ff}^2} \right) \nabla W_{ff_f} - \gamma \sum_{f_b} m_{f_b} 2 \frac{p_f^l}{\rho_f^2} \nabla W_{ff_b}$$

- second particle loop

- predicted density change due to pressure acceleration

$$(\mathbf{A}\mathbf{p})_f^l = \Delta t^2 \sum_{f_f} m_{f_f} \left((\mathbf{a}_f^p)^l - (\mathbf{a}_{ff}^p)^l \right) \nabla W_{ff_f} + \Delta t^2 \sum_{f_b} m_{f_b} (\mathbf{a}_f^p)^l \nabla W_{ff_b}$$

- pressure update

$$p_f^{l+1} = \max \left(p_f^l + \omega \frac{s_f - (\mathbf{A}\mathbf{p})_f^l}{a_{ff}}, 0 \right)$$

- predicted density deviation per particle

$$(\rho_f^{\text{error}})^l = (\mathbf{A}\mathbf{p})_f^l - s_f$$

Comparison with Iterative SESP

- breaking dam
 - 100k particles
 - 0.01% average density error
 - particle radius 0.025m

Δt [s]	PCISPH			IISPH			PCISPH / IISPH		
	avg. iter.	total comp. time [s]		avg. iter.	total comp. time [s]		ratio		
		pressure	overall		pressure	overall	iterations	pressure	overall
0.0005	4.3	540	1195	2.2	148	978	2.0	3.6	1.2
0.00067	7.2	647	1145	2.9	149	753	2.5	4.3	1.5
0.001	14.9	856	1187	4.9	164	576	3.0	5.2	2.1
0.0025	66.5	1495	1540	18.4	242	410	3.6	6.2	3.8
0.004	-	-	-	33.5	273	379	-	-	-
0.005	-	-	-	45.8	297	383	-	-	-

- largest possible time step does not necessarily result in the best performance

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Outline

- concept of an SPH fluid simulator
- momentum equation
- SPH basics
- neighborhood search
- boundary handling
- incompressibility
- surface reconstruction