

# *Simulation in Computer Graphics*

# *Particle-based Fluid Simulation*

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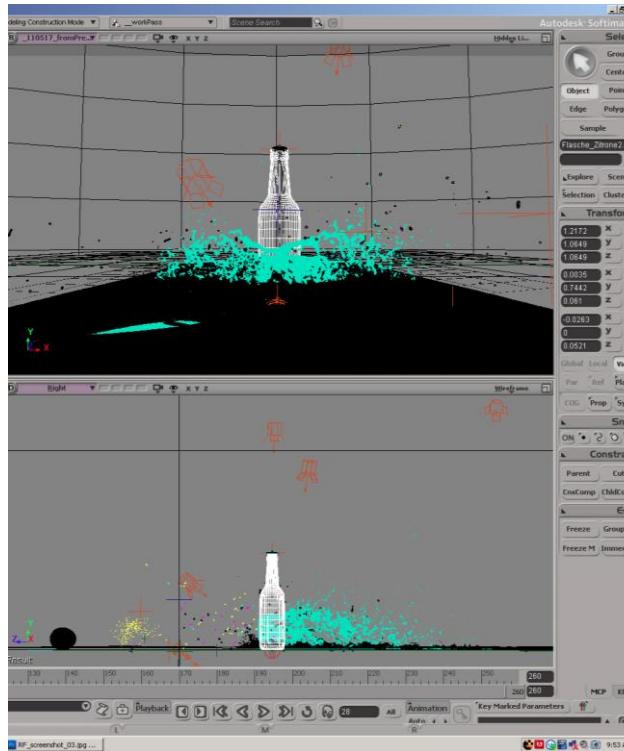
# *Application (with Pixar)*

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

# *Application (Commercials)*



Copyright  
NHB Studios,  
Berlin,  
Hamburg,  
Dusseldorf

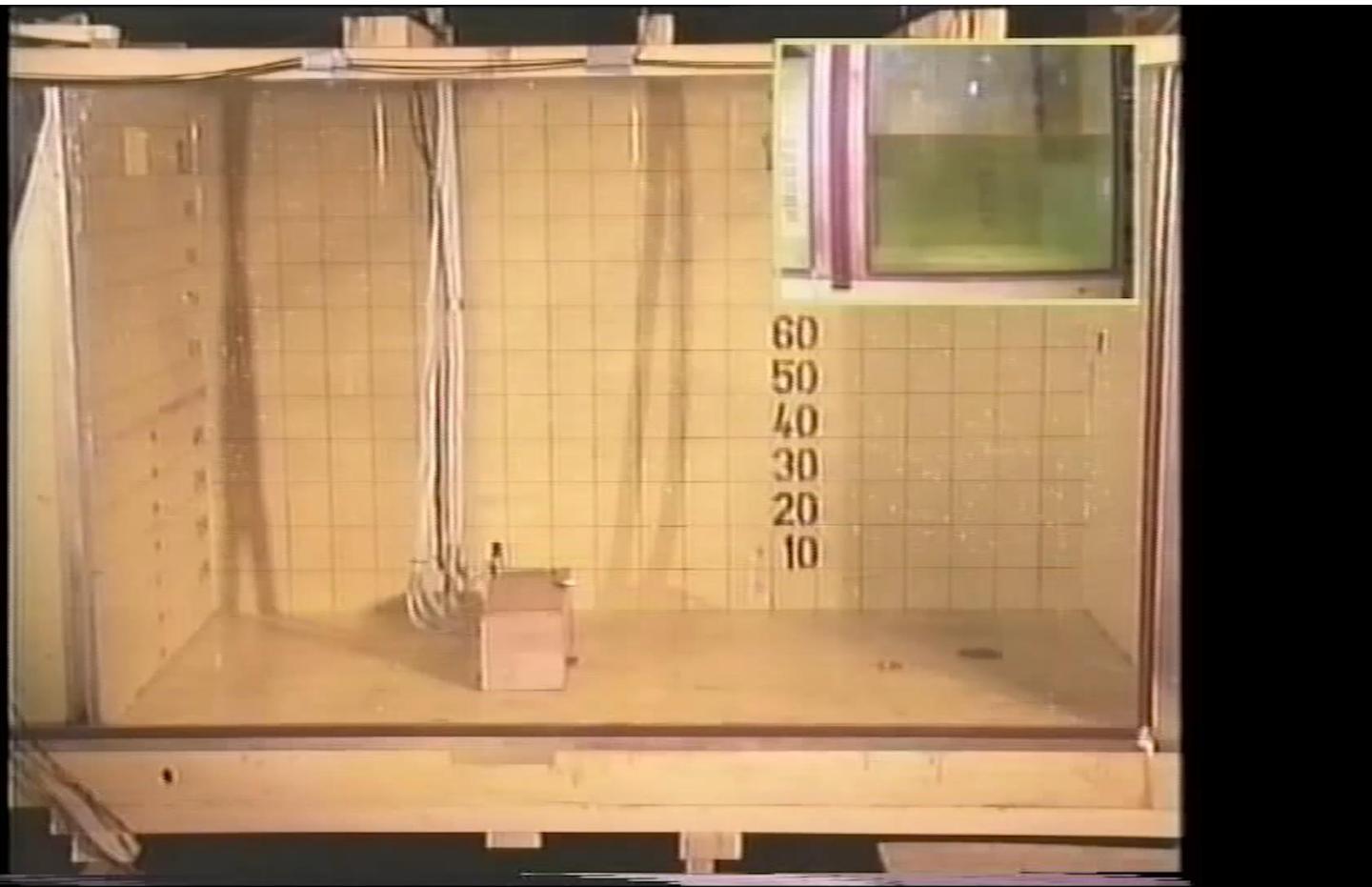
# *Application (with FIFTY2 Technology)*

PreonLab: Drive Through



PreonLab, FIFTY2 Technology GmbH, [www.youtube.com -> fifty2](http://www.youtube.com->fifty2)

# *Application (with FIFTY2 Technology)*



PreonLab, FIFTY2 Technology GmbH, [www.youtube.com -> fifty2](http://www.youtube.com->fifty2)

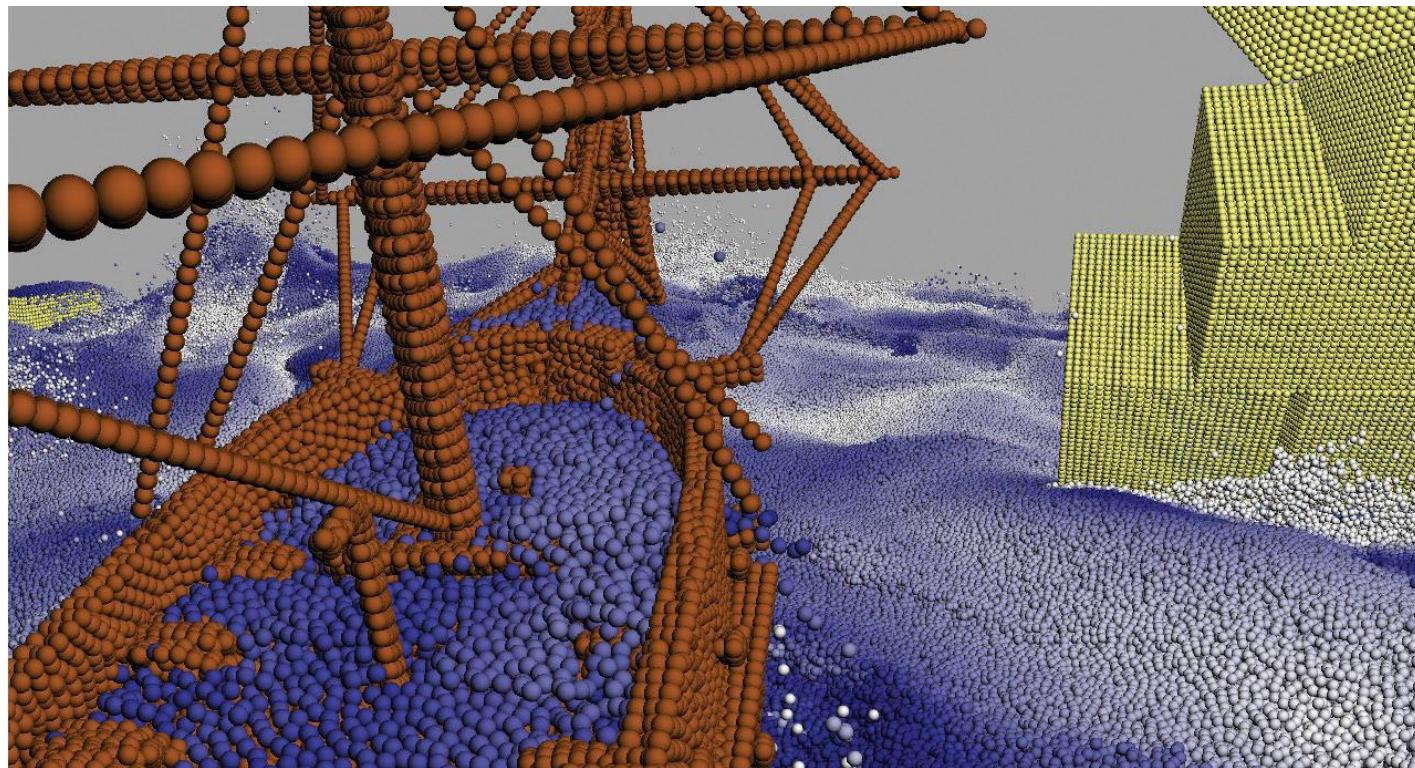
# *Outline*

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

# *Concept*

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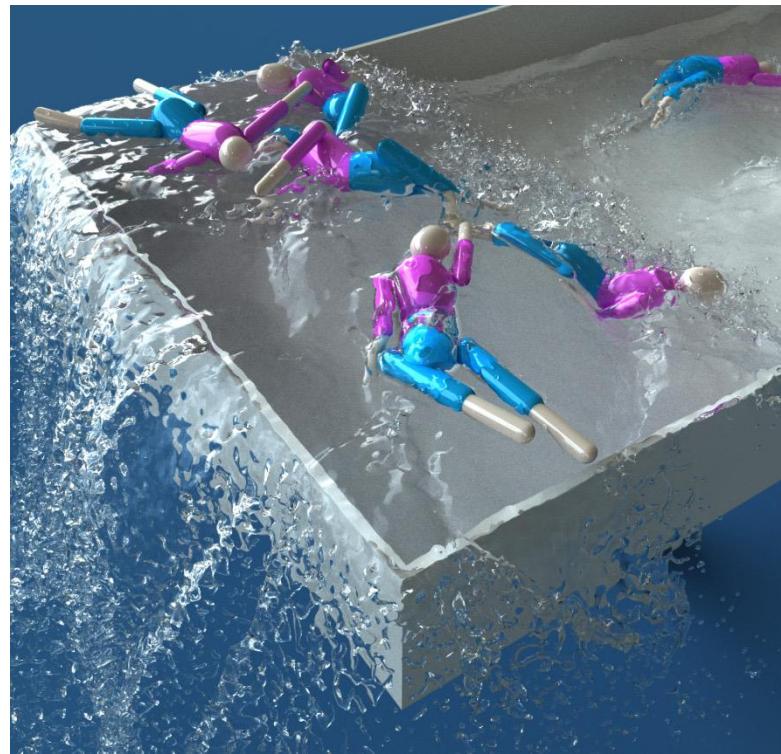
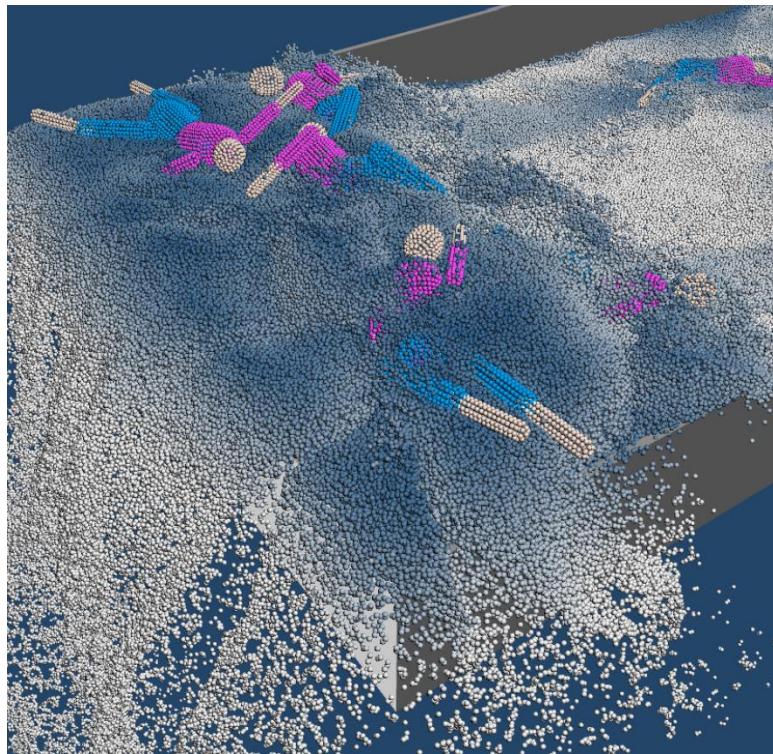
# *Concept*

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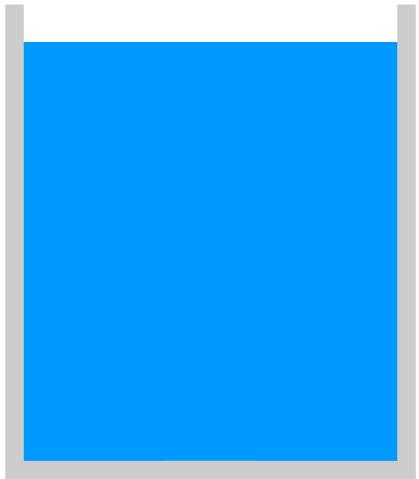
# *Fluid Representation*

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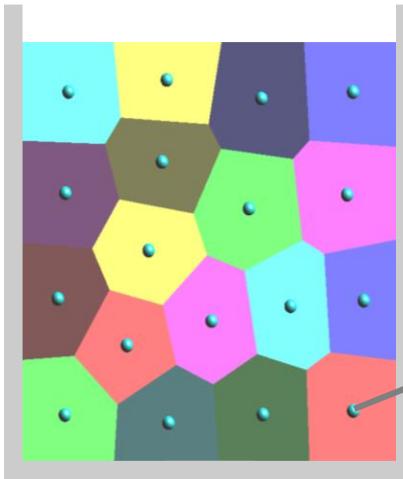


# Fluid Representation

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



Fluid body



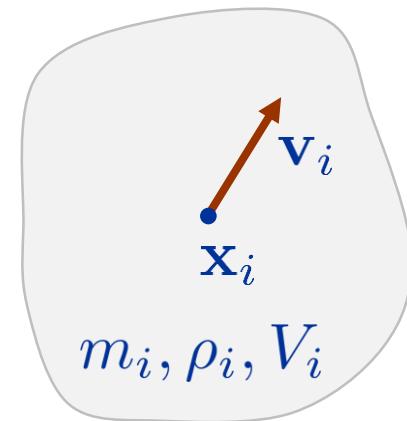
Set of fluid  
parcels

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

# Particles / Fluid Parcels

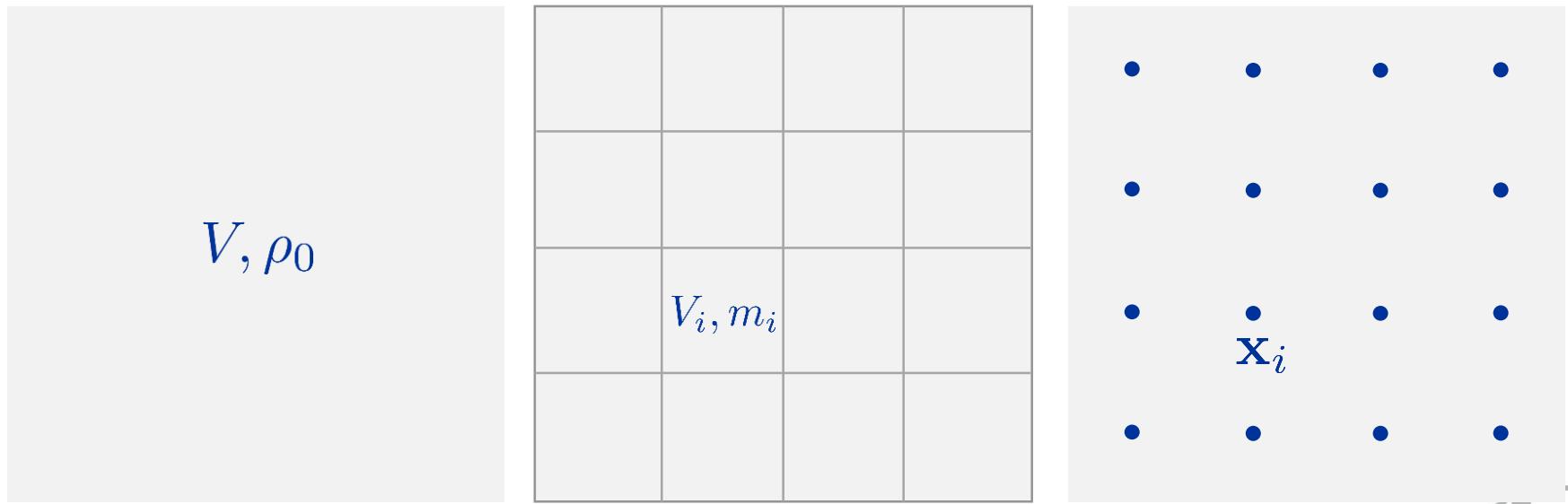
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- represent small fluid portions
- are represented by a sample position  $\mathbf{x}_i$
- move with their velocity  $\mathbf{v}_i$
- have a fixed mass  $m_i$
- volume and density are related by  $V_i = \frac{m_i}{\rho_i}$ 
  - preservation of density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



# Typical Setup

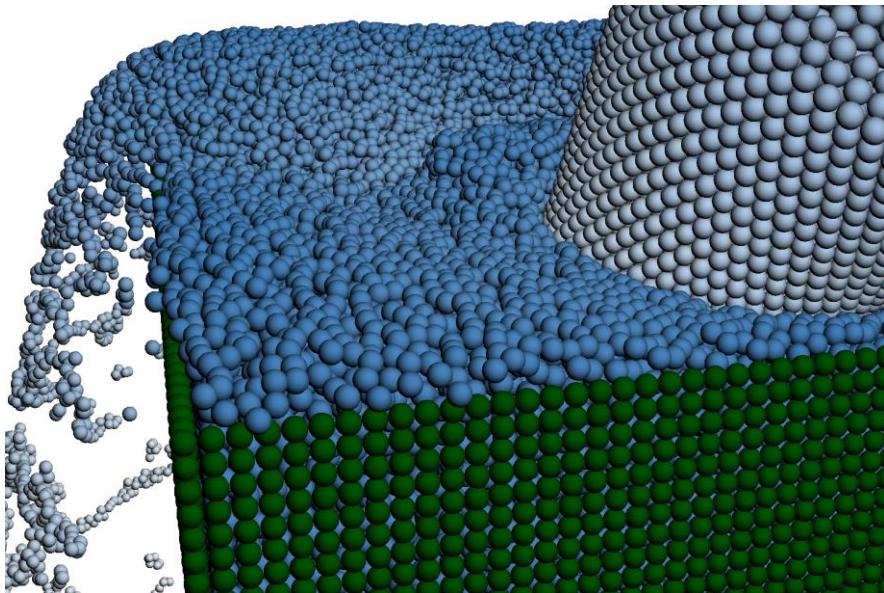
- define overall fluid volume  $V$  and fluid density  $\rho_0$
- define number  $n$  of particles
- assume particles of uniform size  $V_i = \frac{V}{n}$
- compute particle mass as  $m_i = \rho_0 \cdot V_i$
- sample  $\mathbf{x}_i$  represents a particle in the simulation



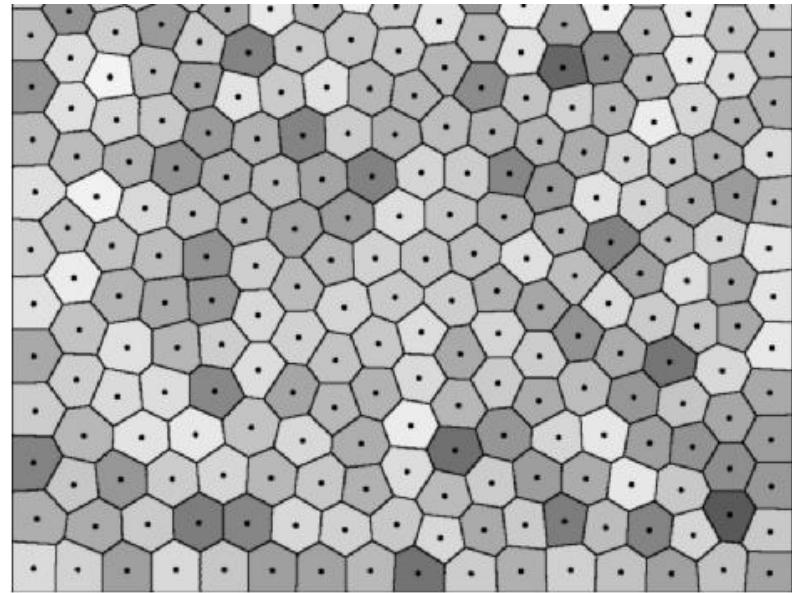
# *Particle Shape*

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- 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

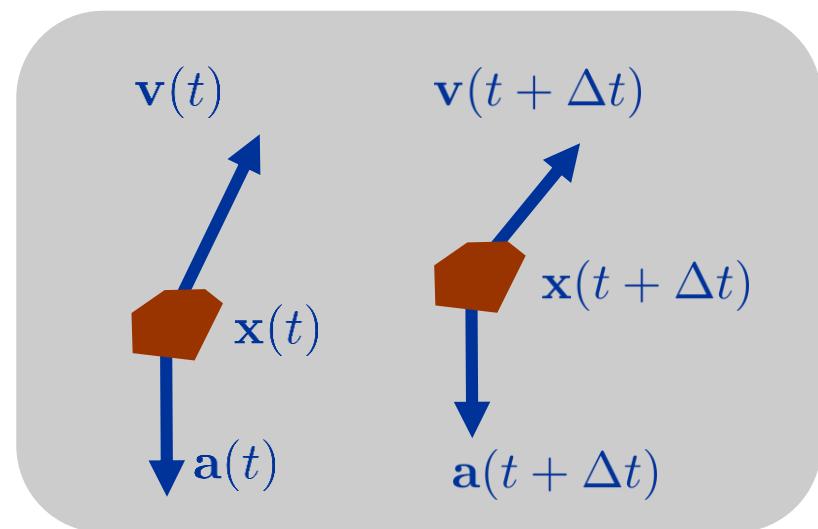
- computation of positions and velocities of fluid parcels over time

- velocity change from current time  $t$  to subsequent time  $t + \Delta t$

$$\mathbf{v}(t + \Delta t) = \mathbf{v}(t) + \Delta t \cdot \mathbf{a}(t)$$

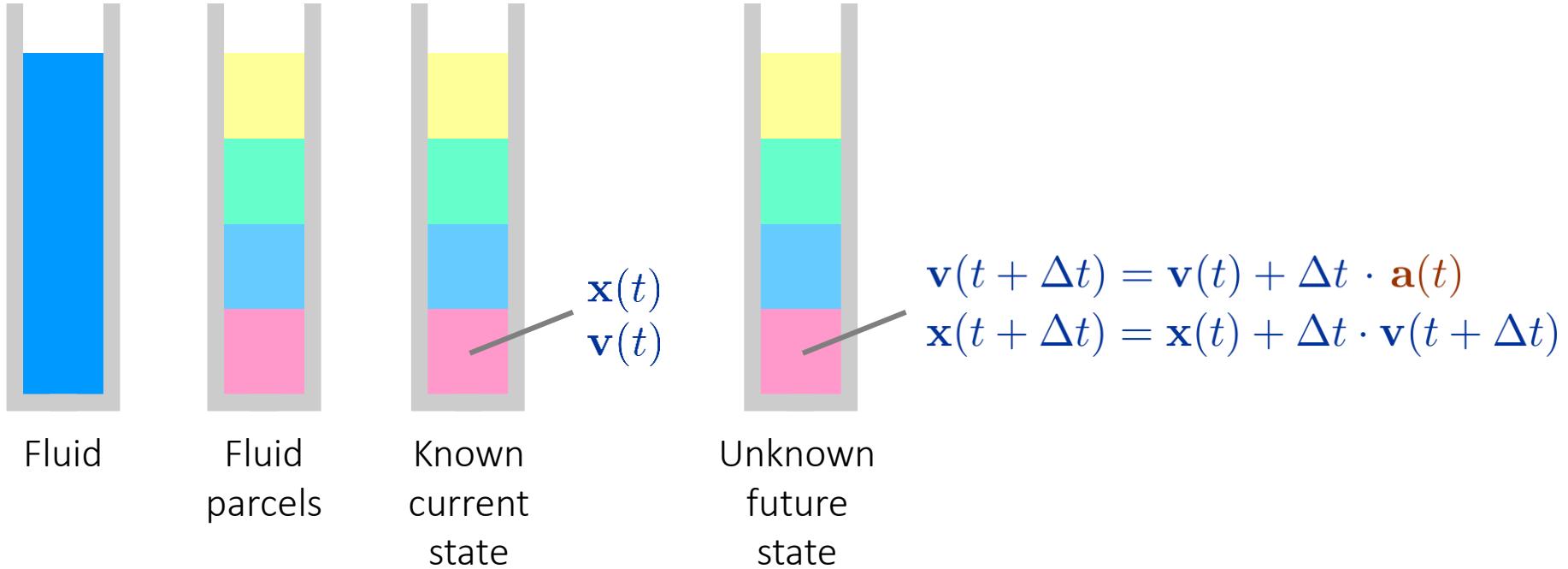
- position change

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



# Example

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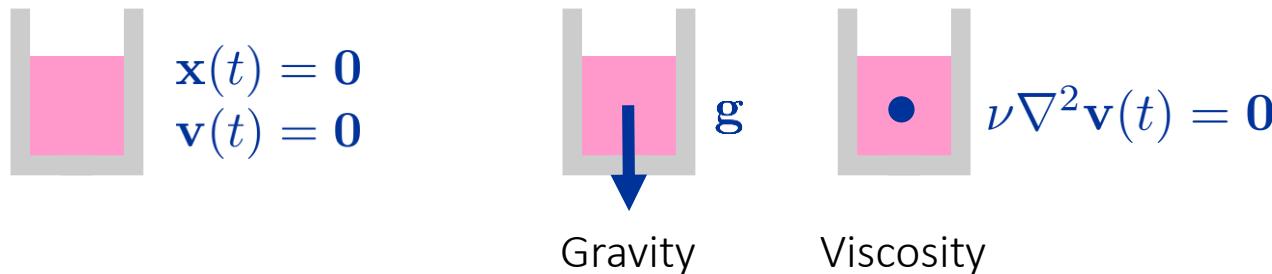
# *Accelerations*

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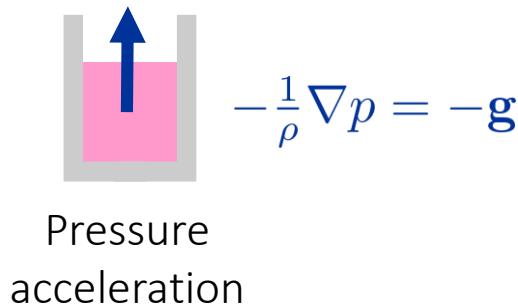
- gravity  $\mathbf{g}$
- viscosity  $\nu \nabla^2 \mathbf{v}$ 
  - friction
  - accelerate parcel towards the average velocity of adjacent fluid parcels
- pressure acceleration  $-\frac{1}{\rho} \nabla p$ 
  - prevent fluid parcels from density / volume changes

# Simulation Step - Example

- gravity and viscosity would change the parcel volume



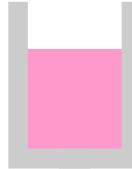
- pressure acceleration avoids the volume / density change



# Simulation Step - Example

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- current state

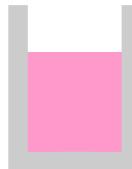


$$\mathbf{x}(t) = \mathbf{0}$$
$$\mathbf{v}(t) = \mathbf{0}$$

- overall acceleration

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

- subsequent state



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

# Neighboring Parcels

- computations require neighboring parcels  $j$

- density or volume

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

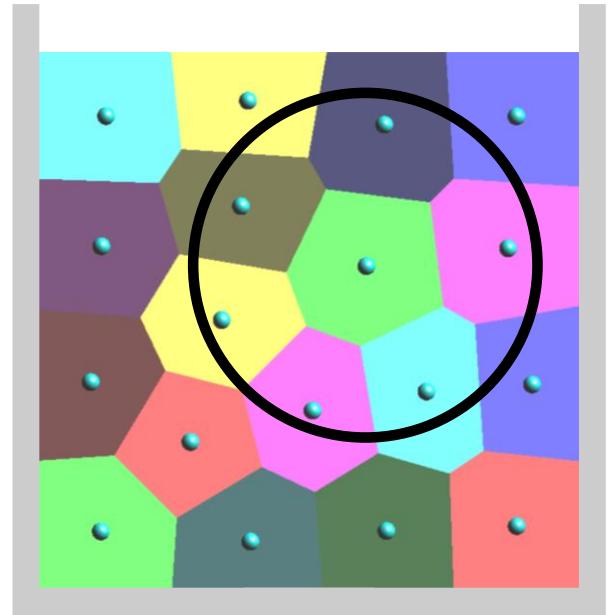
- pressure acceleration

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

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

- Smoothed Particle Hydrodynamics

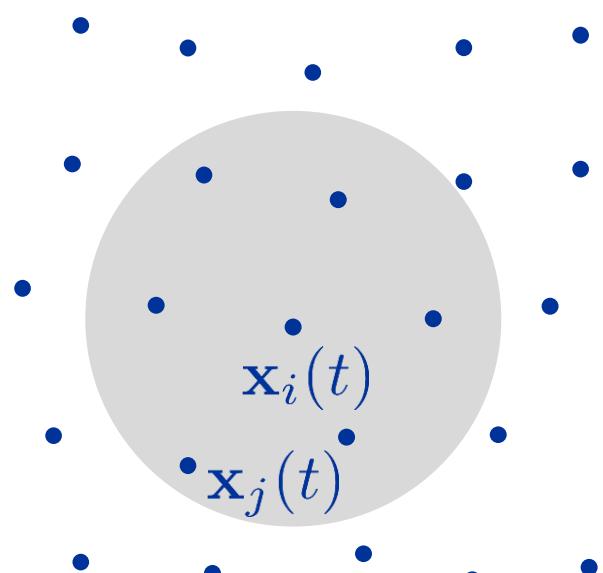
- Gingold and Monaghan, Lucy



# Simulation Step - Implementation

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- 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

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- 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

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

- 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  $\eta = \mu = \nu \cdot \rho_0$
- $\frac{\mathbf{F}_i^{other}}{m_i}$  : e.g., gravity, boundary handling

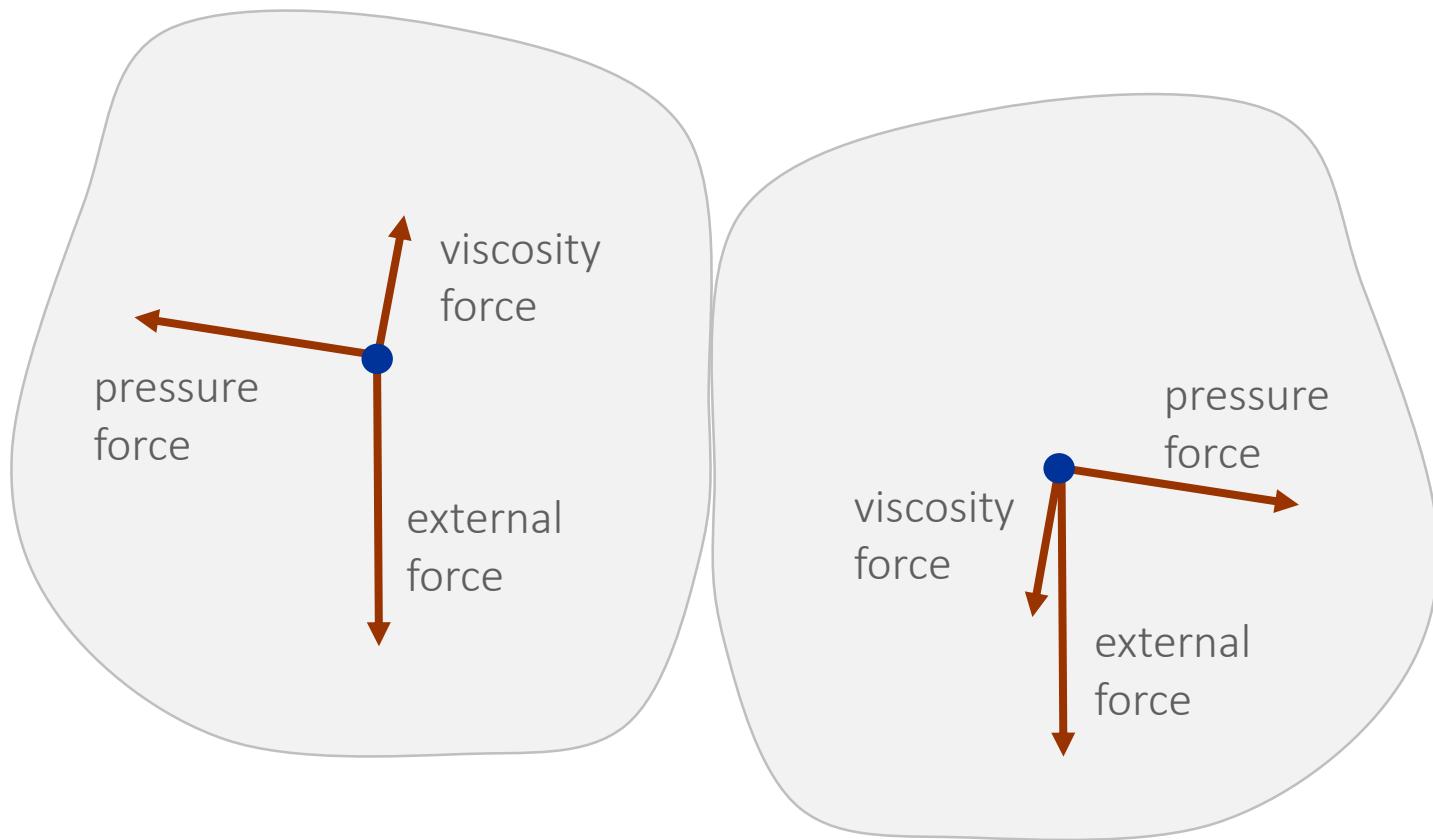
# Accelerations

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

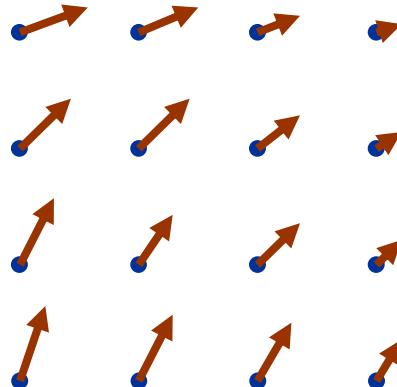
# Forces

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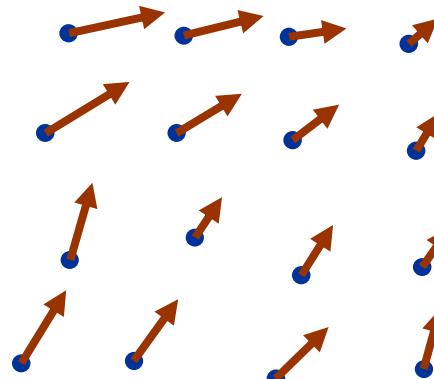


# Lagrangian Fluid Simulation

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



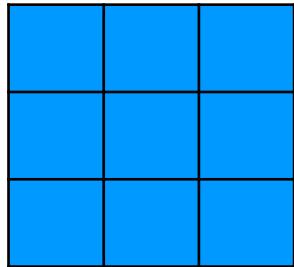
$$\mathbf{v}_i(x_i, y_i, z_i, t) = (u_i, v_i, w_i)$$
$$\mathbf{x}_i(t) = (x_i, y_i, z_i)$$



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

# *Moving Parcels vs. Static Cells*

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

Acceleration of a moving  
parcel

# *Smoothed Particle Hydrodynamics*

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

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

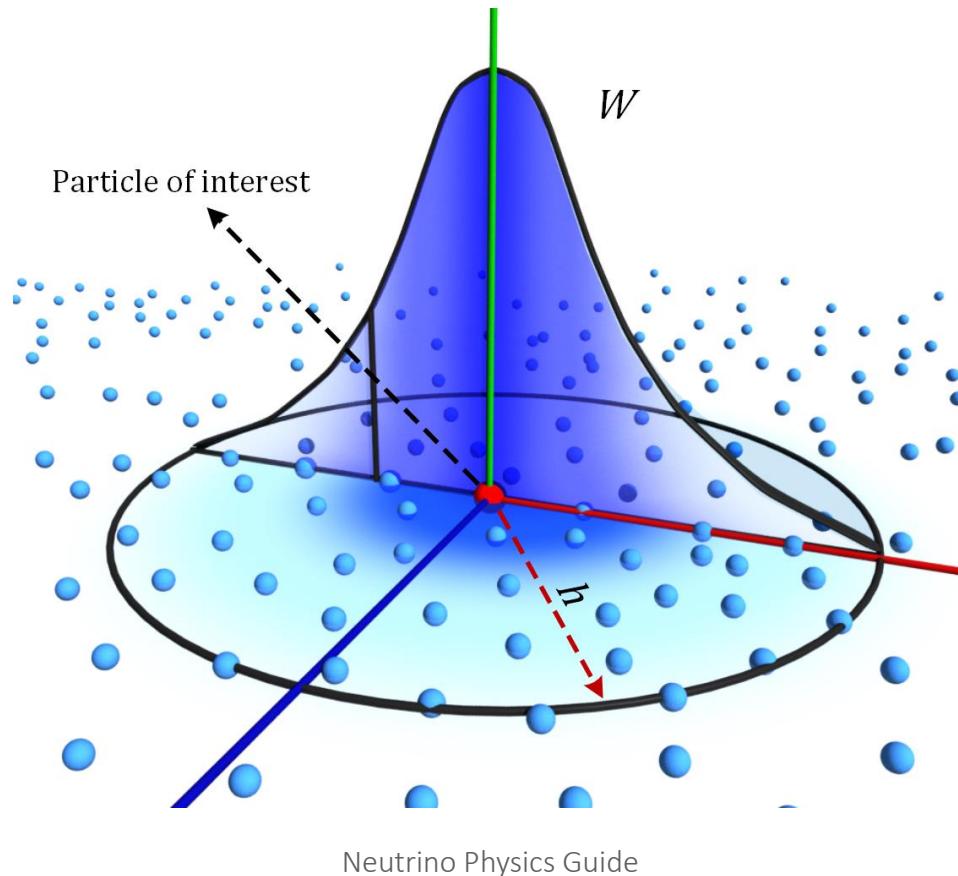
$$A_i = \sum_j V_j A_j W_{ij} = \sum_j \frac{m_j}{\rho_j} A_j W_{ij}$$

- $\mathbf{x}_i$  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*

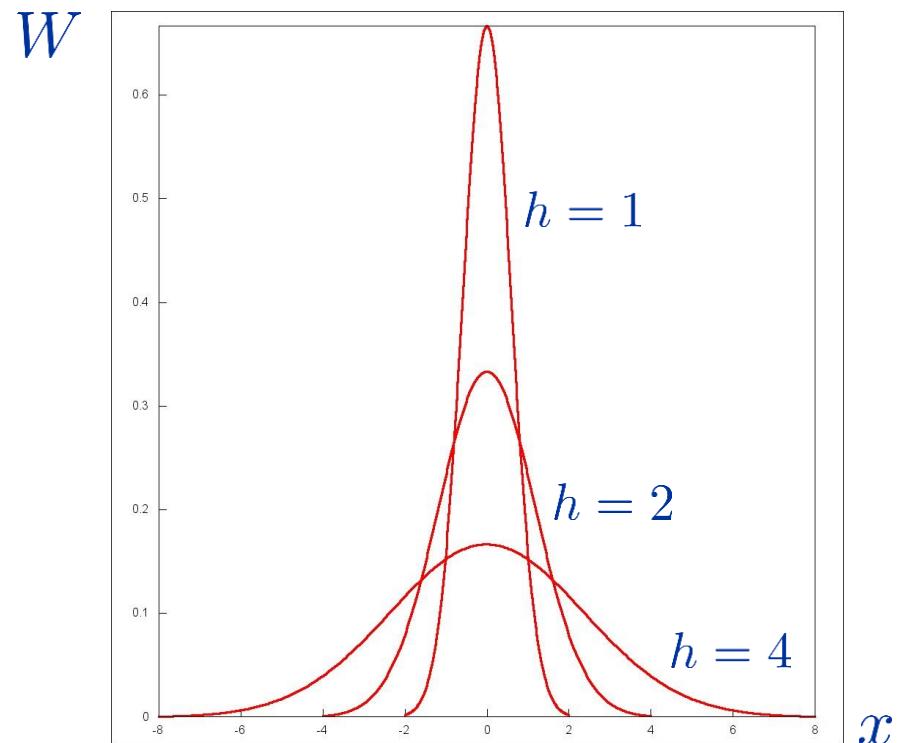
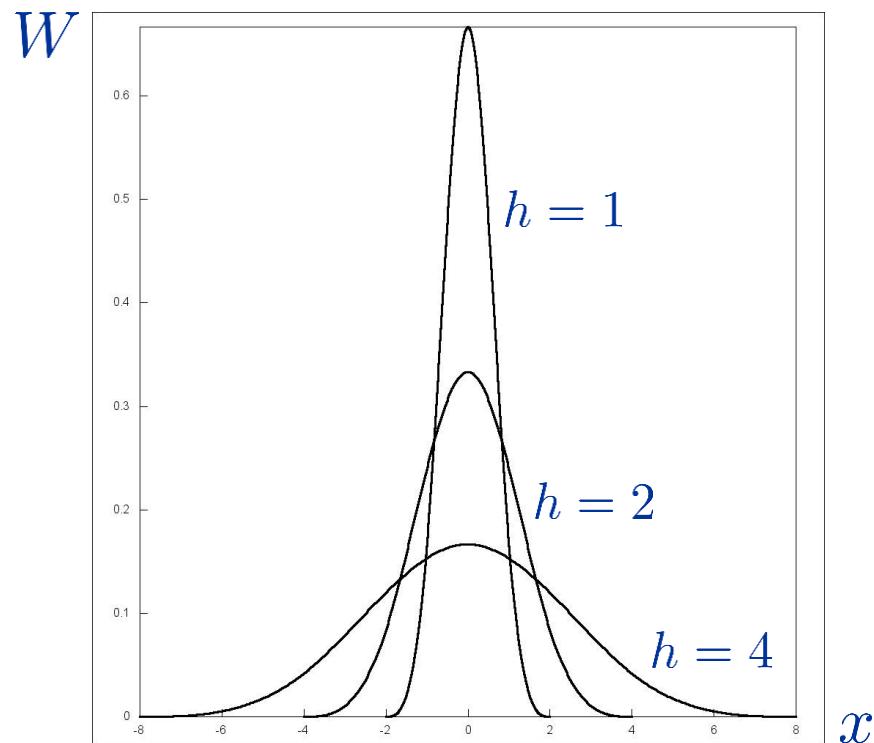


# Kernel Function

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- 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

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

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

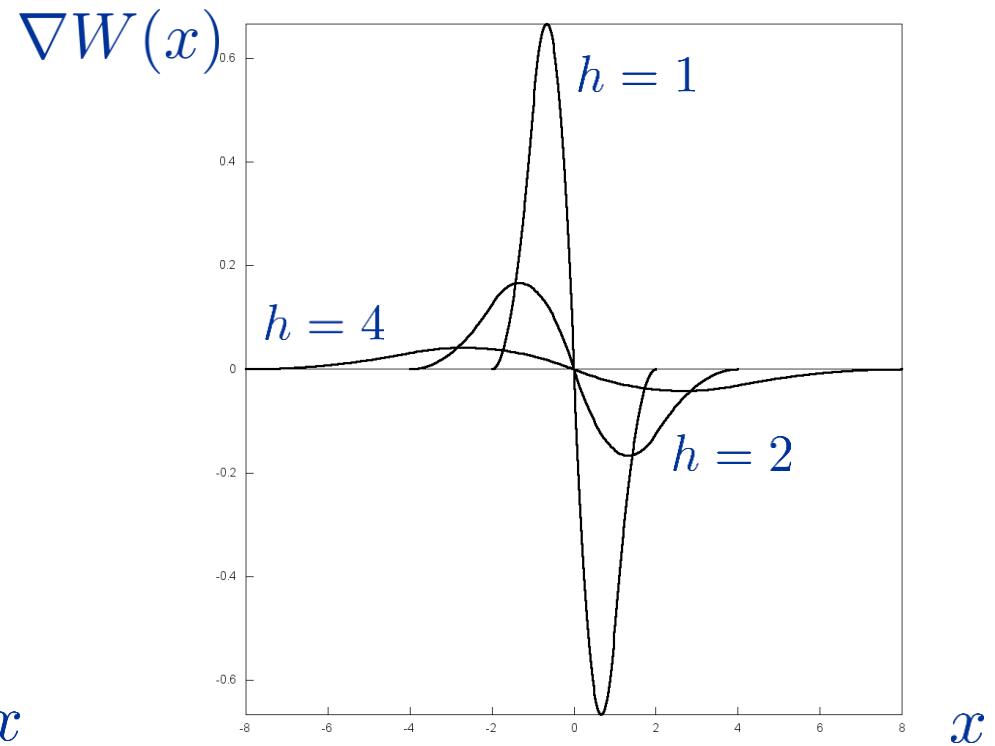
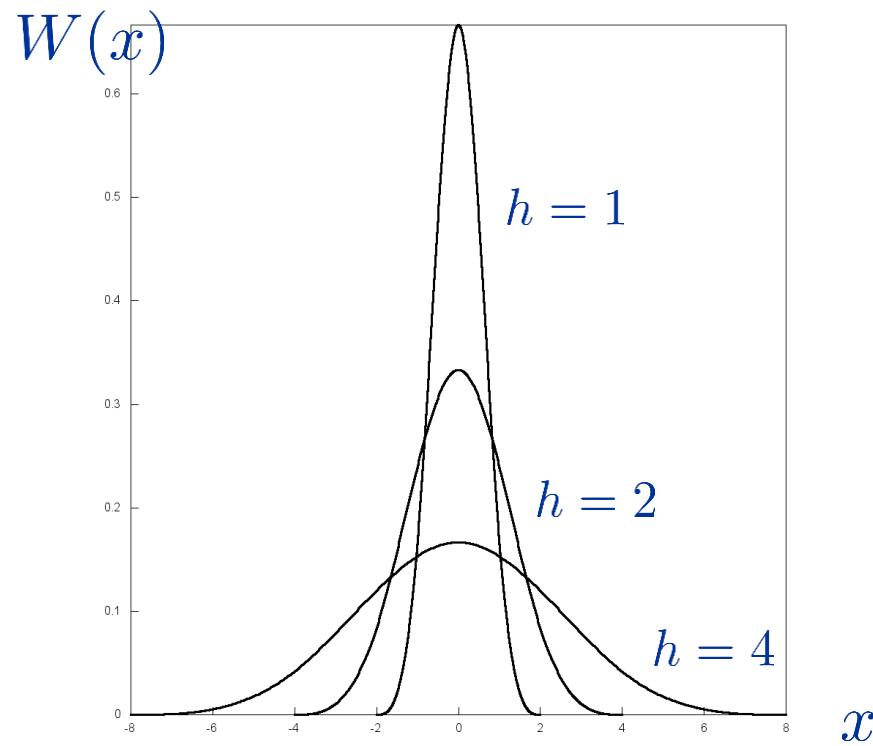
$$\nabla \cdot \mathbf{A}_i = -\frac{1}{\rho_i} \sum_j m_j \mathbf{A}_{ij} \nabla W_{ij}$$

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

preserves linear and angular momentum, when used for pressure forces

more robust as it avoids the second derivative of W

# Kernel Derivative in 1D



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

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

# 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  $\rho_i$  using a state equation, e.g.  $p_i = k \left( \left( \frac{\rho_i}{\rho_0} \right)^7 - 1 \right)$
  - $\rho_0$  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 SESPH
- 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  $\rho_i$
  - for all particle  $i$  do**
    - $\mathbf{F}_i^{pressure} = -\frac{m_i}{\rho_i} \nabla p_i$
    - $\mathbf{F}_i^{viscosity} = m_i \nu \nabla^2 \mathbf{v}_i$
    - $\mathbf{F}_i^{other} = m_i \mathbf{g}$
    - $\mathbf{F}_i(t) = \mathbf{F}_i^{pressure} + \mathbf{F}_i^{viscosity} + \mathbf{F}_i^{other}$
  - 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  $\rho_i$
  - for all particle  $i$  do**
    - $\mathbf{a}_i^{pressure} = -\frac{1}{\rho_i} \nabla p_i$
    - $\mathbf{a}_i^{viscosity} = \nu \nabla^2 \mathbf{v}_i$
    - $\mathbf{a}_i^{other} = \mathbf{g}$
    - $\mathbf{a}_i(t) = \mathbf{a}_i^{pressure} + \mathbf{a}_i^{viscosity} + \mathbf{a}_i^{other}$
  - 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

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

# 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*

# *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  $\rho_i$

**for all particle  $i$  do**

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

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

$$\mathbf{a}_i^{other} = \mathbf{g}$$

$$\mathbf{a}_i(t) = \mathbf{a}_i^{pressure} + \mathbf{a}_i^{viscosity} + \mathbf{a}_i^{other}$$

**for all particle  $i$  do**

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

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

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

SPH approximations

Navier-Stokes  
equation

# *Outline*

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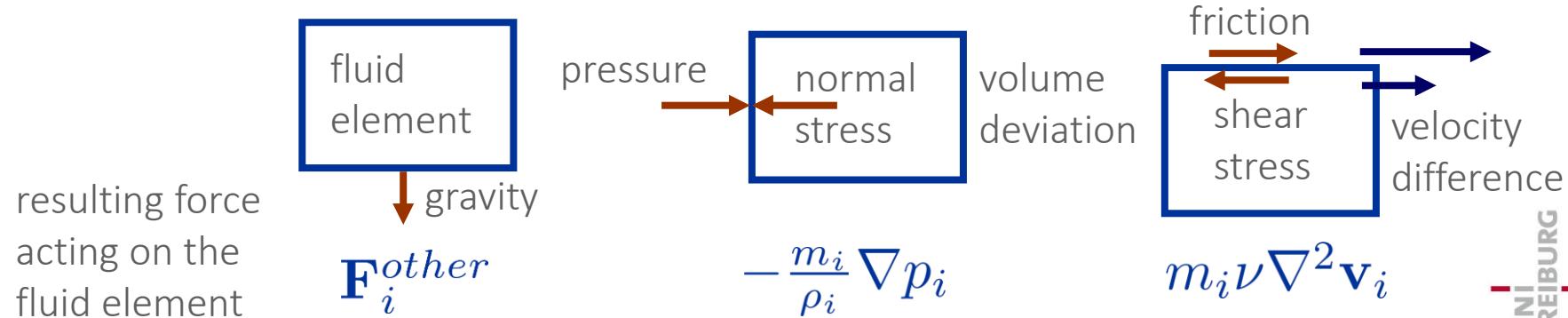
- concept of an SPH fluid simulator
- momentum equation
- SPH basics
- neighborhood search
- boundary handling
- incompressibility
- surface reconstruction

# Force Types

- momentum equation

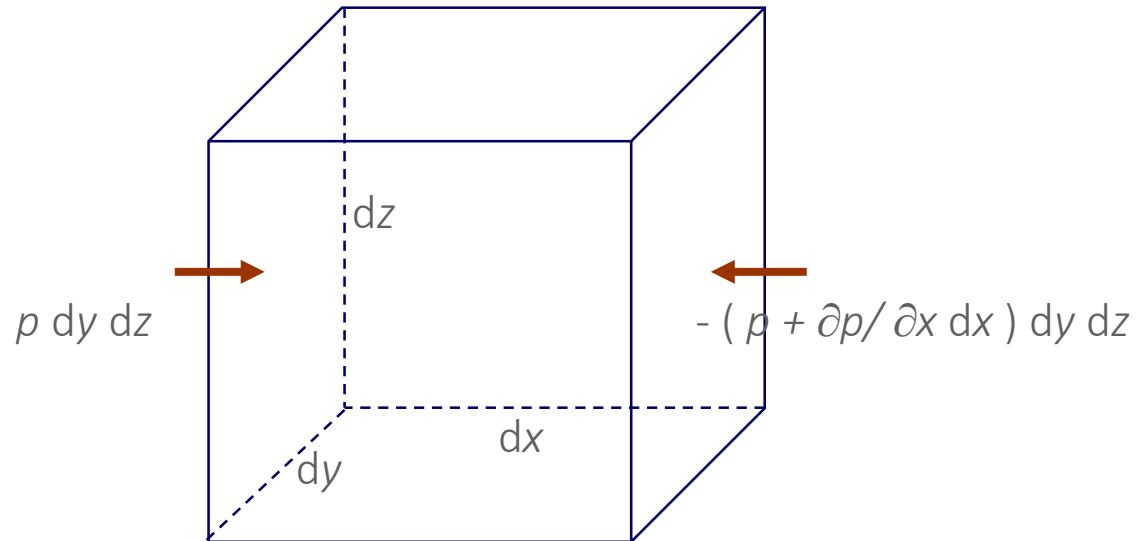
$$\frac{d\mathbf{v}_i}{dt} = -\frac{1}{\rho_i} \nabla p_i + \nu \nabla^2 \mathbf{v}_i + \frac{\mathbf{F}_i^{other}}{m_i}$$

- 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  $i$

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

- 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

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

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

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

- $\boldsymbol{\tau}$  is the viscous stress tensor

$$\boldsymbol{\tau} = \nu \begin{pmatrix} \frac{\partial u}{\partial x} + \frac{\partial u}{\partial x} & \frac{\partial v}{\partial x} + \frac{\partial u}{\partial y} & \frac{\partial w}{\partial x} + \frac{\partial u}{\partial z} \\ \frac{\partial u}{\partial y} + \frac{\partial v}{\partial x} & \frac{\partial v}{\partial y} + \frac{\partial v}{\partial y} & \frac{\partial w}{\partial y} + \frac{\partial v}{\partial z} \\ \frac{\partial u}{\partial z} + \frac{\partial w}{\partial x} & \frac{\partial v}{\partial z} + \frac{\partial w}{\partial y} & \frac{\partial w}{\partial z} + \frac{\partial w}{\partial z} \end{pmatrix}$$

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

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

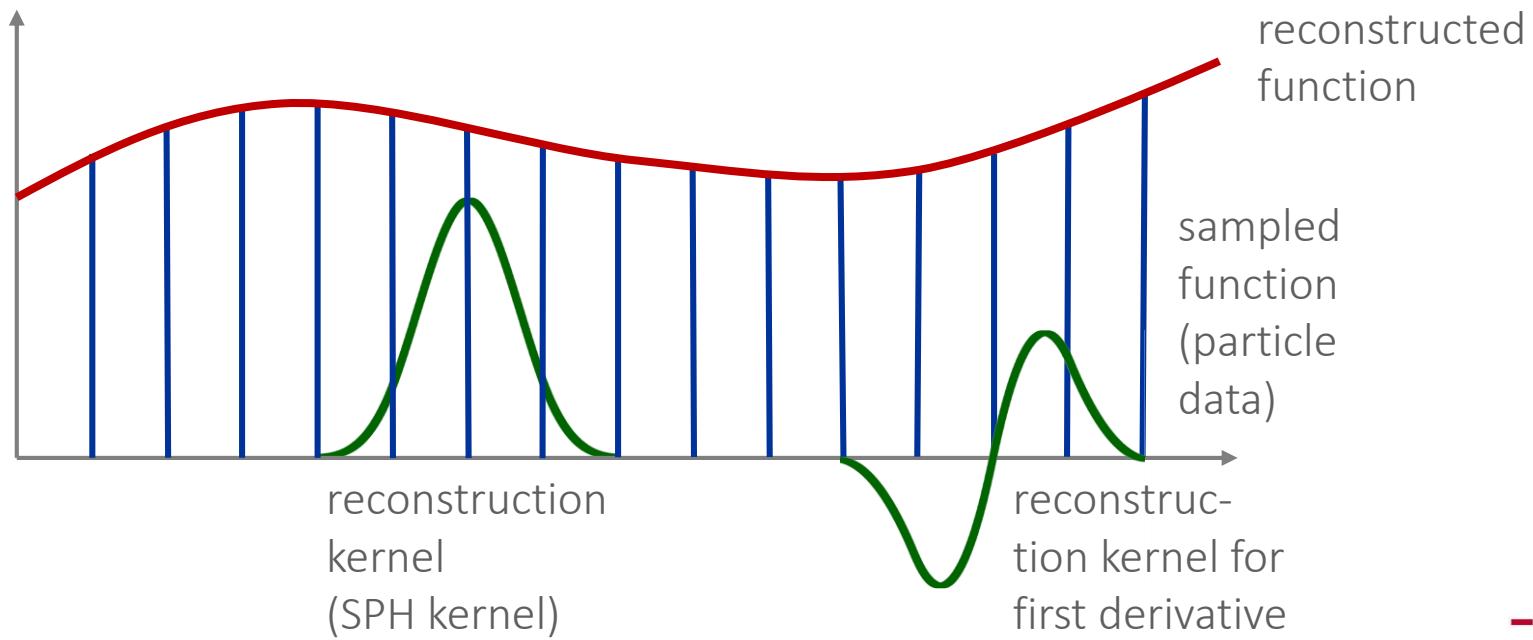
# *Outline*

---

- 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

$$\int_0^{2\pi} \int_0^{2h} \tilde{W}(x)x \, dx d\phi = \int_0^{2\pi} \int_0^2 \tilde{W}(q)hq \, dq d\phi =$$

$$2\pi \int_0^1 [q(2-q)^3 - 4q(1-q)^3] h^2 dq + 2\pi \int_1^2 q(2-q)^3 h^2 dq = 2\pi \frac{11}{10} h^2 + 2\pi \frac{3}{10} h^2$$

$$\alpha = \frac{5}{14\pi h^2}$$

- 3D: integration over the volume of a sphere

$$\int_0^{2\pi} \int_{-\frac{\pi}{2}}^{\frac{\pi}{2}} \int_0^{2h} \tilde{W}(x)x^2 \sin\theta \, dx d\theta d\phi = \int_0^{2\pi} \int_{-\frac{\pi}{2}}^{\frac{\pi}{2}} \int_0^2 \tilde{W}(q)(qh)^2 h \sin\theta \, dq d\theta d\phi =$$

$$4\pi \int_0^1 [q^2(2-q)^3 - 4q(1-q)^3] h^3 dq + 4\pi \int_1^2 q^2(2-q)^3 h^3 dq = 4\pi \frac{19}{30} h^3 + 4\pi \frac{11}{30} h^3$$

$$\alpha = \frac{1}{4\pi h^3}$$

# *Particle Approximation*

---

- $$\begin{aligned} A(\mathbf{x}) &\approx \int_{\Omega_h} A(\mathbf{x}') W(\mathbf{x} - \mathbf{x}', h) d\mathbf{x}' \\ &= \int_{\Omega_h} \frac{A(\mathbf{x}')}{\rho(\mathbf{x}')} W(\mathbf{x} - \mathbf{x}', h) \rho(\mathbf{x}') d\mathbf{x}' \end{aligned}$$
- 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  
$$\int_S A(\mathbf{x}') W(\mathbf{x}' - \mathbf{x}, h) \mathbf{n} dS = 0 \quad W = 0 \text{ on the surface } S$$
$$\nabla_{\mathbf{x}} A(\mathbf{x}) \approx - \int_{\Omega_h} A(\mathbf{x}') \nabla_{\mathbf{x}'} W(\mathbf{x}' - \mathbf{x}, h) d\mathbf{x}' = \int_{\Omega_h} A(\mathbf{x}') \nabla_{\mathbf{x}} W(\mathbf{x} - \mathbf{x}', h) d\mathbf{x}'$$
- $\nabla_x A(\mathbf{x}_i) \approx \sum_j A(\mathbf{x}_j) \nabla W(\mathbf{x}_i - \mathbf{x}_j, h) V(\mathbf{x}_j)$ 
$$\nabla_x A(\mathbf{x}_i) \approx \sum_j \frac{m(\mathbf{x}_j)}{\rho(\mathbf{x}_j)} A(\mathbf{x}_j) \nabla W(\mathbf{x}_i - \mathbf{x}_j, h)$$

# *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

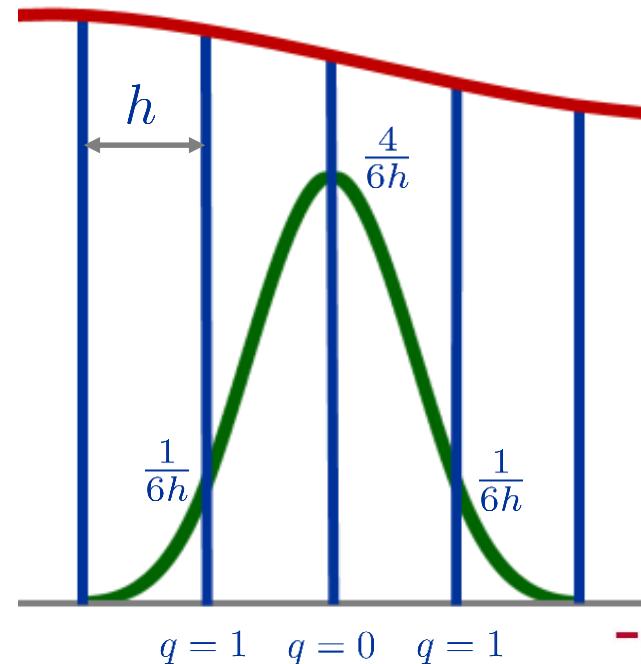
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- in case of ideal sampling
- $\rho_i = \sum_j m_j W_{ij} = m_i \sum_j W_{ij} \quad m_i = m_j$
- $m_i \sum_j W_{ij} = \rho_i = \frac{m_i}{V_i} \Rightarrow \sum_j W_{ij} = \frac{1}{V_i}$
- $\nabla W_{ij} = -\nabla W_{ji} \quad \nabla W_{ij} = \alpha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

# 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

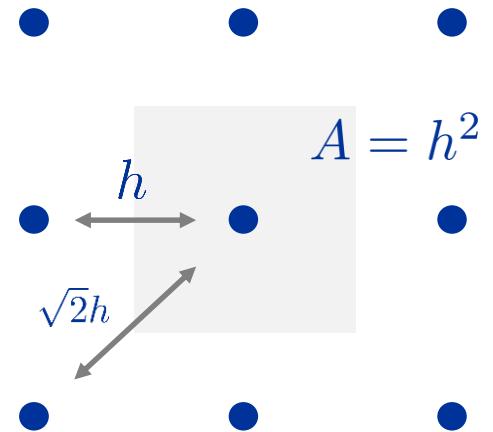
$$W(q) = \frac{5}{14\pi h^2} \begin{cases} (2-q)^3 - 4(1-q)^3 & 0 \leq q < 1 \\ (2-q)^3 & 1 \leq q < 2 \\ 0 & q \geq 2 \end{cases} \quad q = \frac{\|\mathbf{x}_i - \mathbf{x}_j\|}{h}$$

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

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

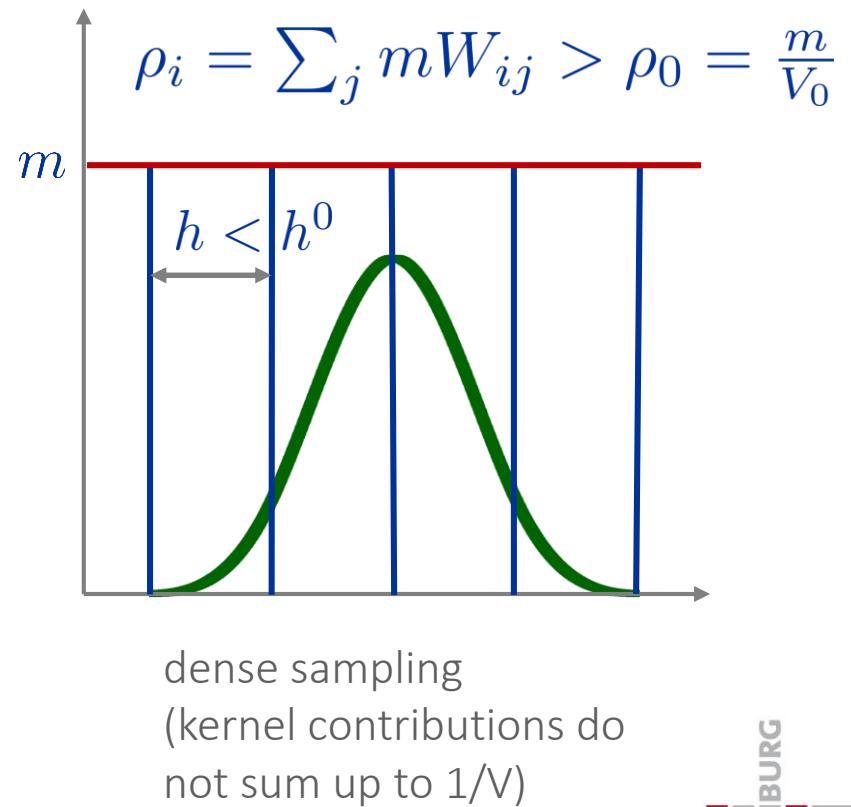
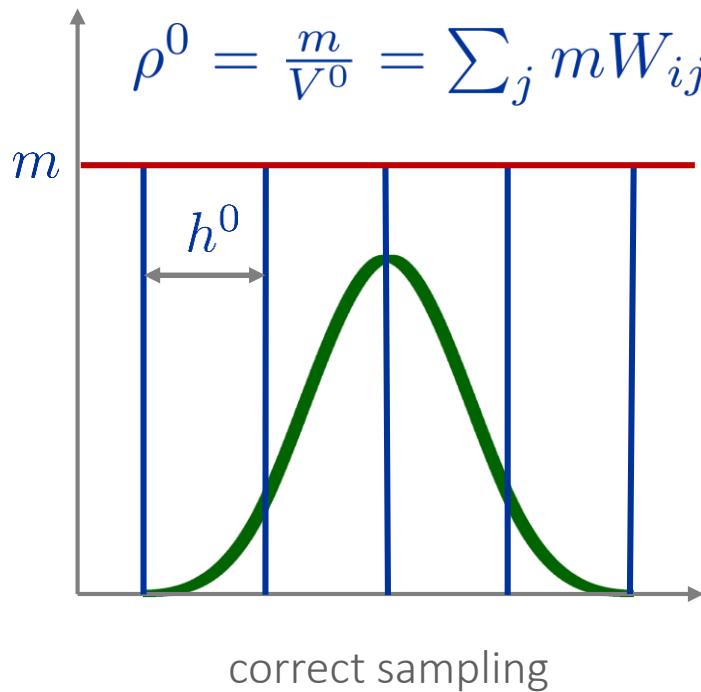
$$W(\sqrt{2}) = \frac{5}{14\pi h^2} (2-\sqrt{2})^3 \approx \frac{1.005}{14\pi h^2}$$

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



# Kernel Illustration

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



# *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  $\rho_i$

**for all particle  $i$  do**

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

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

$$\mathbf{a}_i^{other} = \mathbf{g}$$

$$\mathbf{a}_i(t) = \mathbf{a}_i^{pressure} + \mathbf{a}_i^{viscosity} + \mathbf{a}_i^{other}$$

**for all particle  $i$  do**

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

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

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

SPH approximations

Navier-Stokes  
equation

# *Outline*

---

- concept of an SPH fluid simulator
- momentum equation
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- 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*

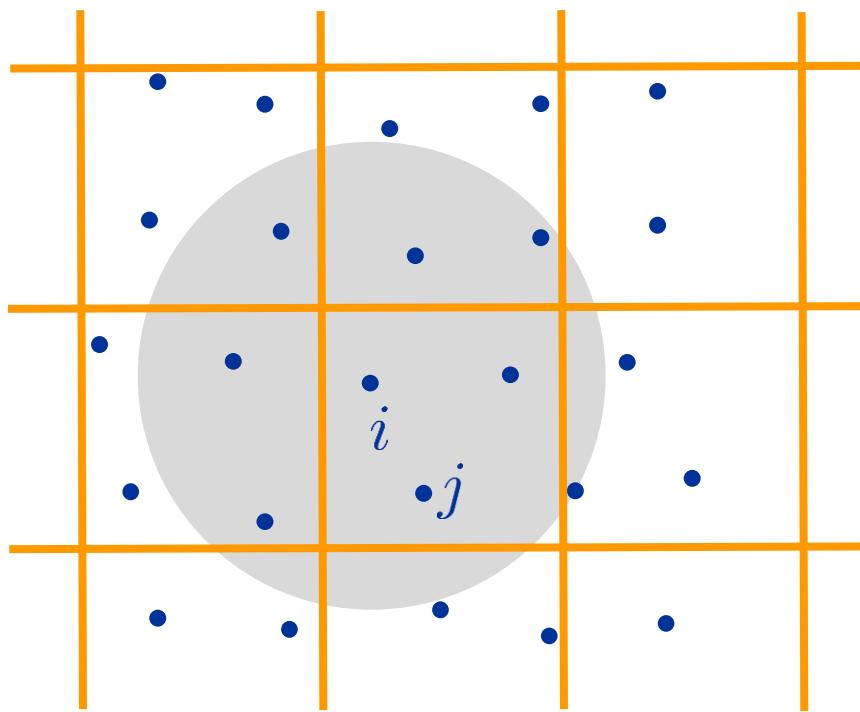
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- SPH computes sums
  - dynamically changing sets of neighboring particles
  - temporal coherence
- spatial data structures accelerate the neighbor search
  - fast query
  - fast generation (at least once for each simulation step)
  - sparsely, non-uniformly 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

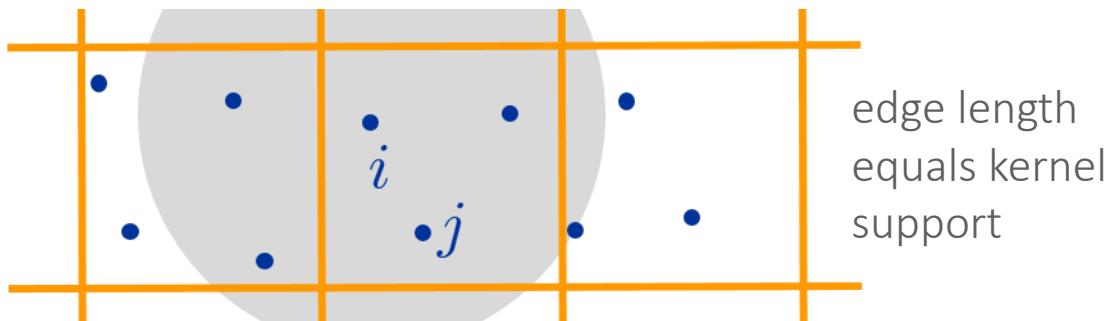
# *Outline*

---

- 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



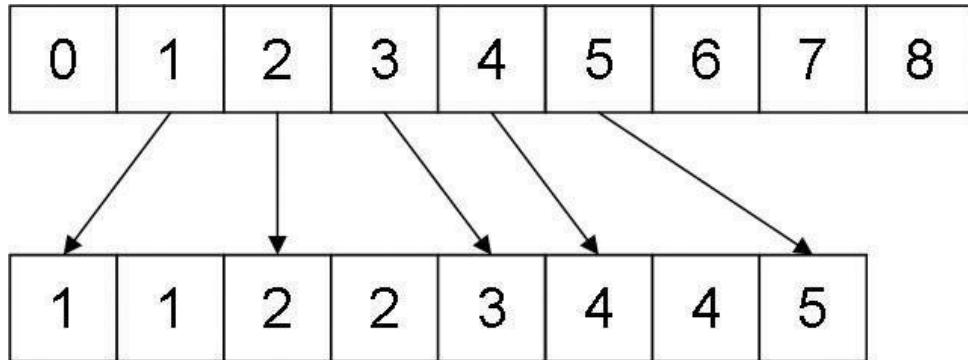
# *Outline*

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# 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



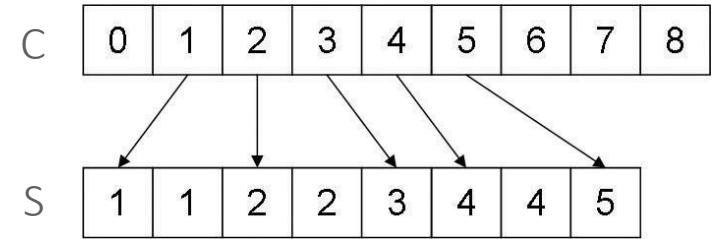
uniform grid

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:  
 $S [ --C [ j ] ] = i$ 
  - 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

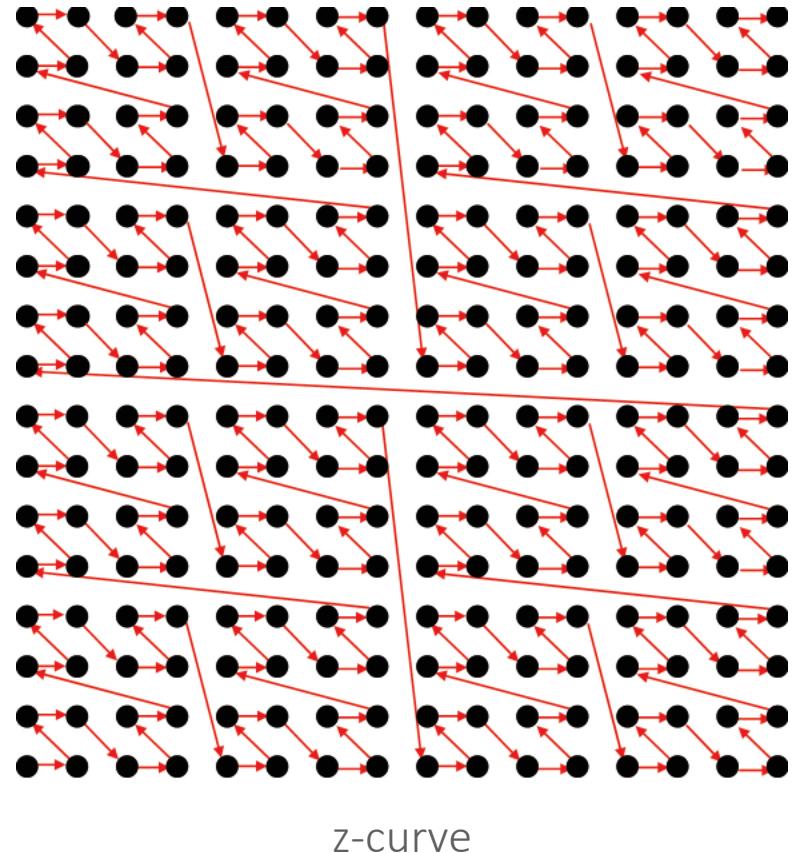
# *Query*

---

- 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-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  $n^{\text{th}}$  simulation step
  - restores spatial locality

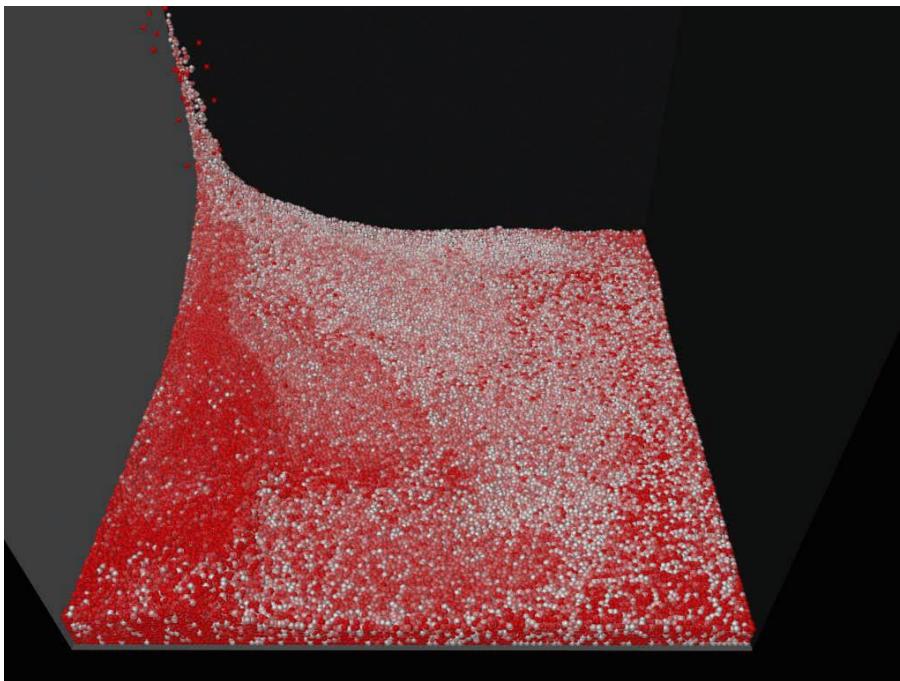
# *Z-Index Sort - Sorting*

---

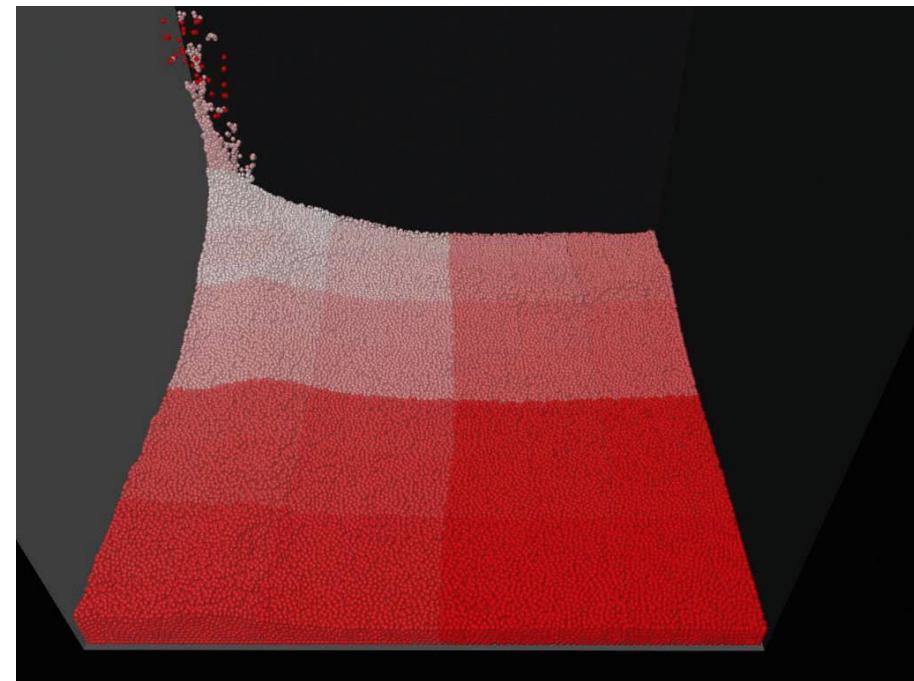
- instead of radix sort, insertion sort can be employed
  - $O(n)$  for almost sorted arrays
  - due to temporal coherence, only 2% of all particles change their cell, i.e. z-curve index, in each time step

# *Z-Index Sort - Reordering*

---



particles colored according  
to their location in memory



spatial compactness is  
enforced using a z-curve

# *Outline*

---

- concept of an SPH fluid simulator
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  - uniform grid
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  - spatial hashing
  - discussion
- boundary handling
- incompressibility
- surface reconstruction

# *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  $i$

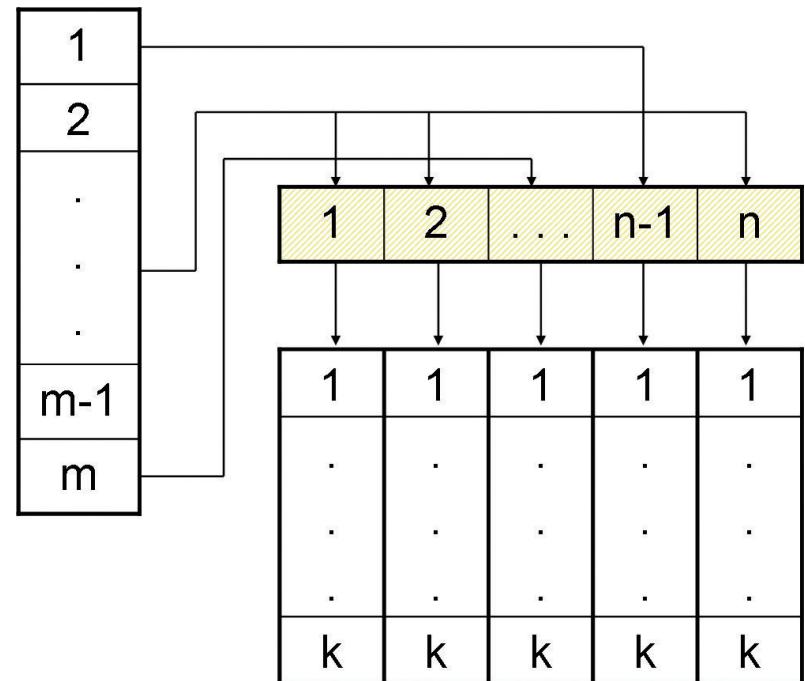
# *Spatial Hashing*

---

- 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

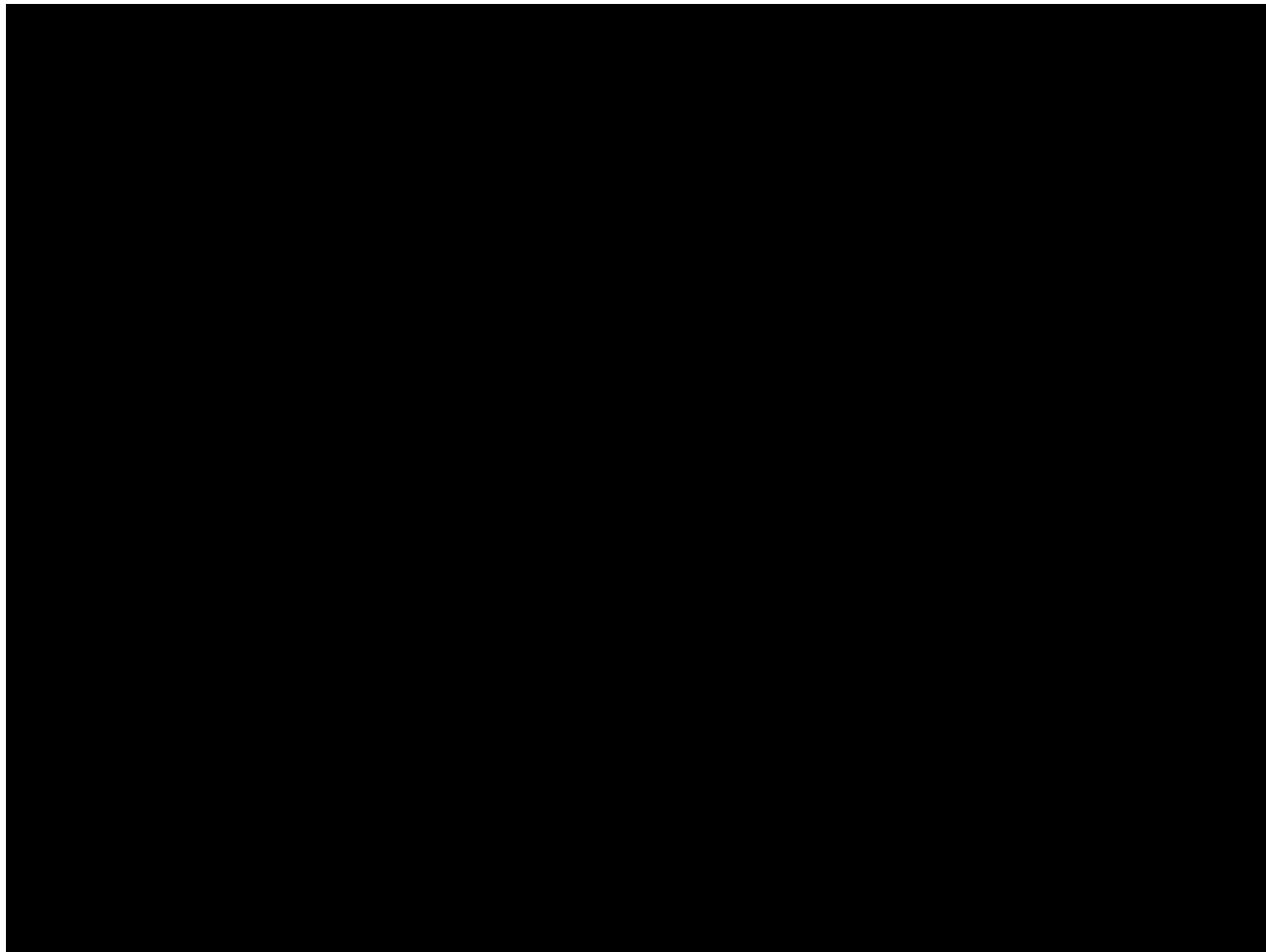
# *Compact Hashing - Query*

---

- particles are sorted with respect to a z-curve every  $n^{\text{th}}$  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*

---



# *Outline*

---

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  - index sort
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# Comparison

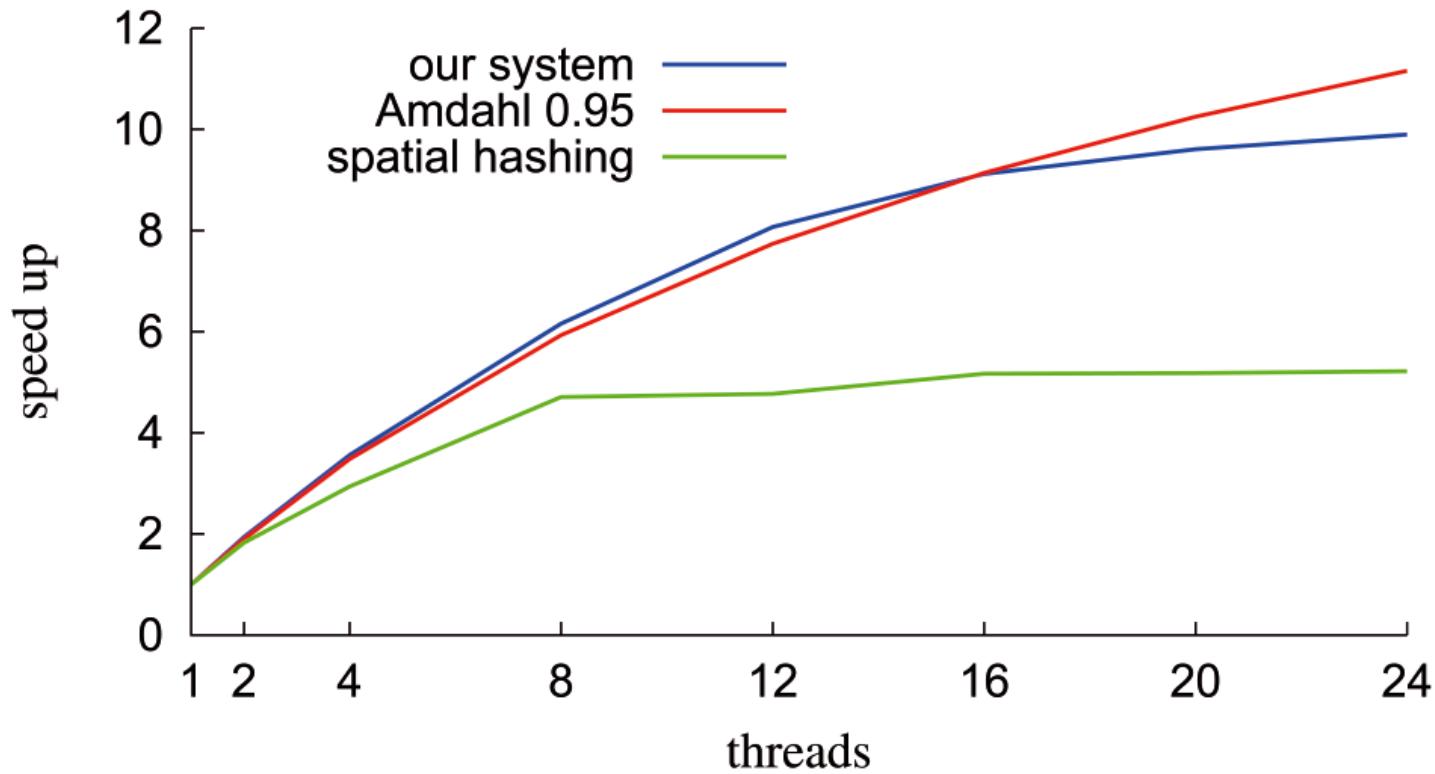
method	construction	query	total
basic grid	26	38	64
index sort	36	29	65
z-index sort	16	27	43
spatial hashing	42	86	128
compact hashing	8	32	40

- 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

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# *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  $\rho_i$

**for all particle  $i$  do**

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

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

$$\mathbf{a}_i^{other} = \mathbf{g}$$

$$\mathbf{a}_i(t) = \mathbf{a}_i^{pressure} + \mathbf{a}_i^{viscosity} + \mathbf{a}_i^{other}$$

**for all particle  $i$  do**

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

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

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

SPH approximations

Navier-Stokes  
equation

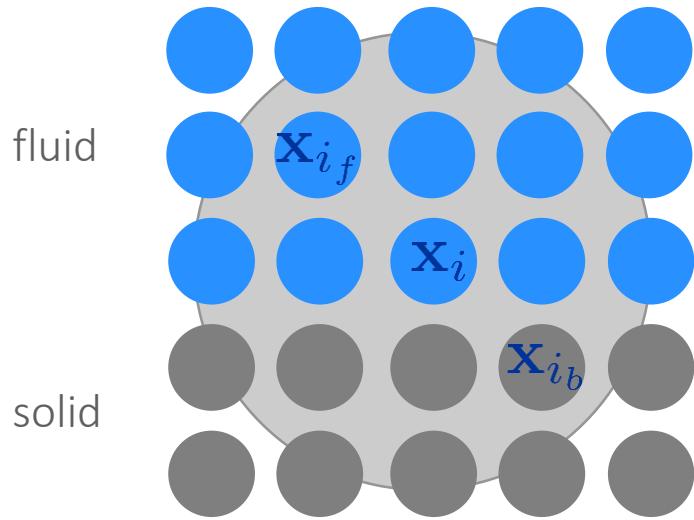
# *Outline*

---

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# Concept

- rigids are uniformly sampled with particles



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

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

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

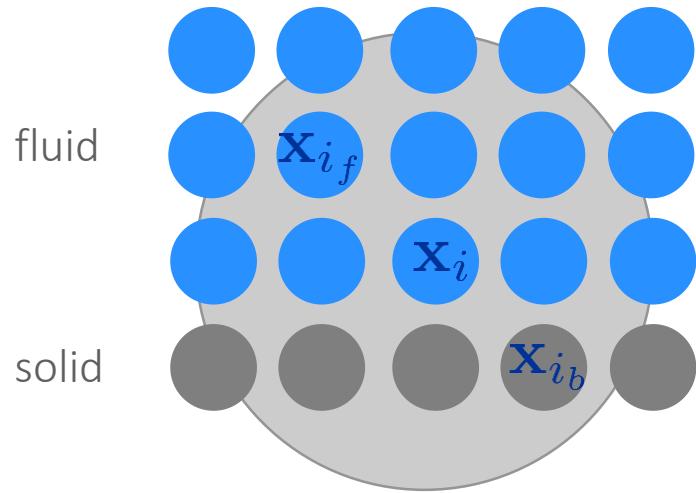
$$p_i = \dots$$

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

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

# Missing Contributions

- rigids are uniformly sampled with particles



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

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

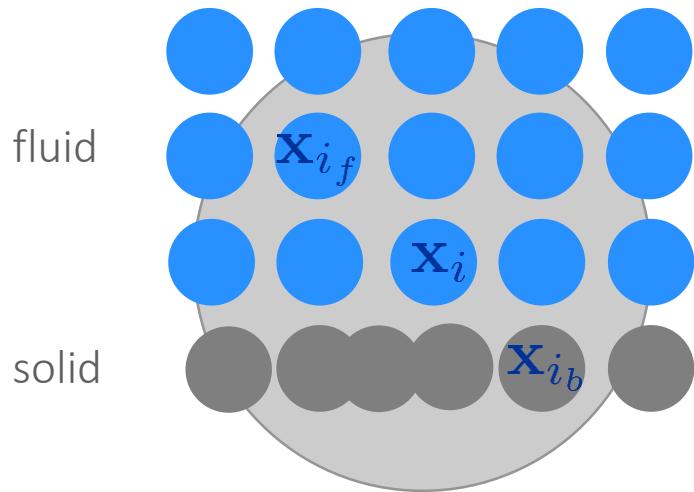
$$\sum_{i_f} W_{ii_f} + \gamma_1 \sum_{i_b} W_{ii_b} = \frac{1}{V_i}$$

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

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

# Non-uniform Sampling

- rigids are non-uniformly sampled with particles



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

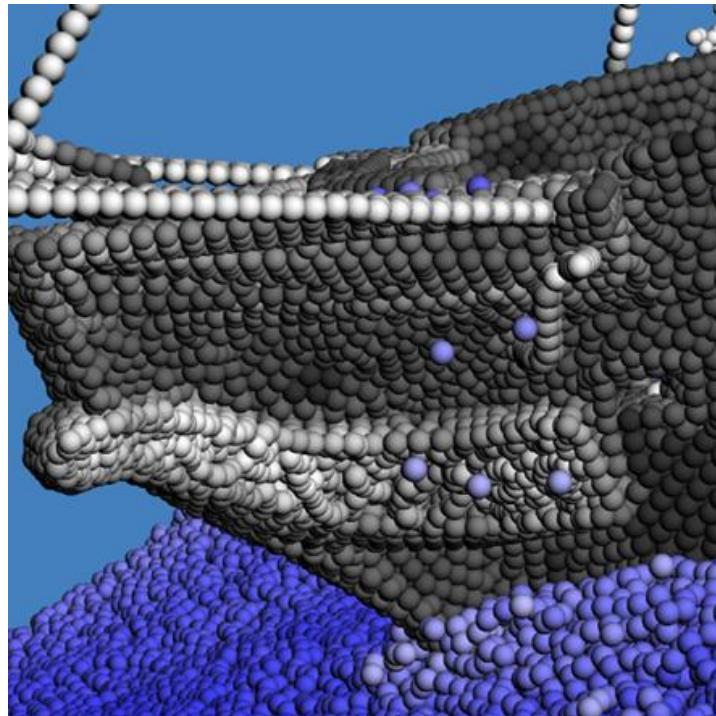
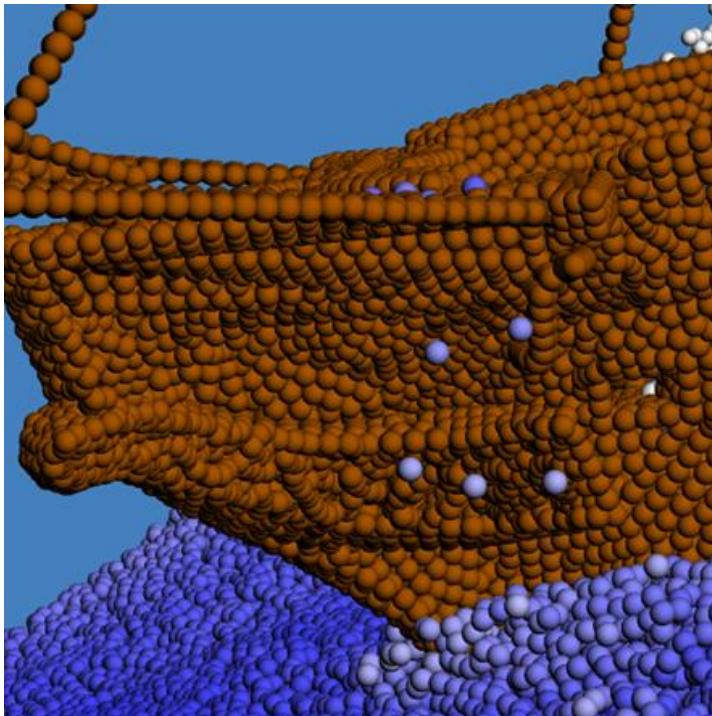
$$\frac{1}{V_{i_b}} \approx \gamma_3 \sum_{i_{b_b}} W_{i_b i_{b_b}}$$

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

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

# *Non-uniform Sampling*

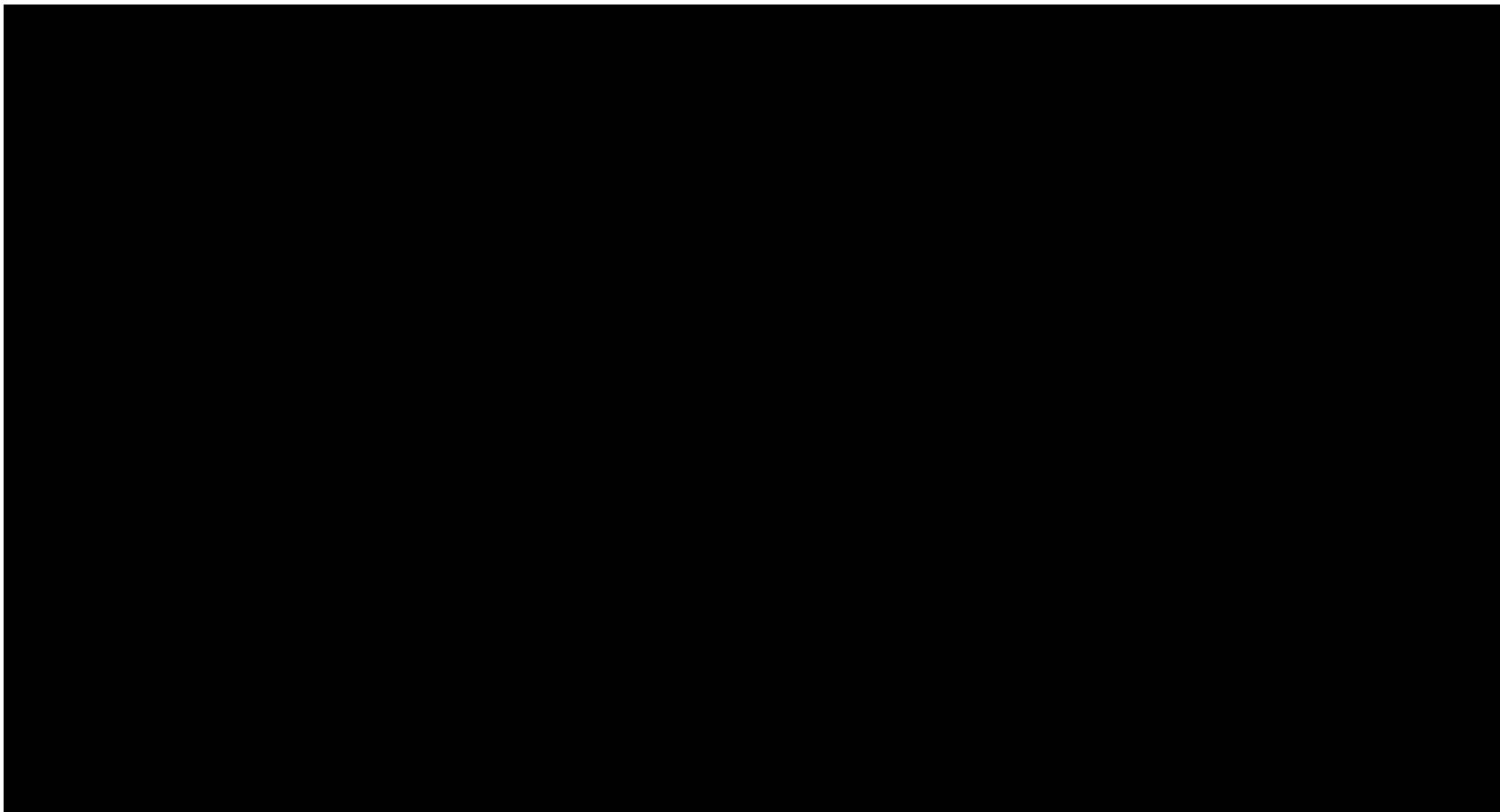
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color-coded volume  
of boundary particles

# *Rigid-Fluid Coupling*

---



[www.youtube.com/cgfreiburg](http://www.youtube.com/cgfreiburg)

# *Rigid-Fluid Coupling*

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[www.youtube.com/cgfreiburg](http://www.youtube.com/cgfreiburg)

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  - IHMSEN M., AKINCI N., GISSLER M., TESCHNER M.: Boundary Handling and Adaptive Time-stepping for PCISPH. *VRIPHYS*, 2010.
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# *Outline*

---

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

# *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<sub>1</sub>, k<sub>2</sub> 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  $\rho_i$  with a state equation
  - for all particle  $i$  do**
    - $\mathbf{F}_i^{pressure} = -\frac{m_i}{\rho_i} \nabla p_i$
    - $\mathbf{F}_i^{viscosity} = m_i \nu \nabla^2 \mathbf{v}_i$
    - $\mathbf{F}_i^{other} = m_i \mathbf{g}$
    - $\mathbf{F}_i(t) = \mathbf{F}_i^{pressure} + \mathbf{F}_i^{viscosity} + \mathbf{F}_i^{other}$
  - 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  $\rho_i^*$  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  $\rho_i^*$  - 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

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

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

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

# *Iterative SESPH 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 SESPH with Splitting

---

- **for all particle  $i$  do**
  - find neighbors  $j$
  - for all particle  $i$  do**
    - $\mathbf{F}_i^{viscosity} = m_i \nu \nabla^2 \mathbf{v}_i$      $\mathbf{F}_i^{other} = m_i \mathbf{g}$
    - $\mathbf{v}_i^* = \mathbf{v}_i(t) + \Delta t \frac{\mathbf{F}_i^{viscosity} + \mathbf{F}_i^{other}}{m_i}$      $\mathbf{x}_i^* = \mathbf{x}_i(t) + \Delta t \mathbf{v}_i^*$
  - repeat**
    - for all particle  $i$  do**
      - compute  $\rho_i^*$  using  $\mathbf{x}_i^*$
      - compute  $p_i$  using  $\rho_i^*$ , e.g.  $p_i = k(\rho_i^* - \rho_0)$
    - compute  $\rho_{err}$ , e.g. average or maximum
    - for all particle  $i$  do**
      - $\mathbf{F}_i^{pressure} = -\frac{m_i}{\rho_i^*} \nabla p_i$      $\mathbf{v}_i^* = \mathbf{v}_i^* + \Delta t \frac{\mathbf{F}_i^{pressure}}{m_i}$      $\mathbf{x}_i^* = \mathbf{x}_i^* + \Delta t^2 \frac{\mathbf{F}_i^{pressure}}{m_i}$
  - until**  $\rho_{err} < \eta$     user-defined density error
  - for all particle  $i$  do**
    - $\mathbf{v}_i(t + \Delta t) = \mathbf{v}_i^*$      $\mathbf{x}_i(t + \Delta t) = \mathbf{x}_i^*$

# *Iterative SESPH - 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)
    - $\Delta\mathbf{x}_i = -\frac{1}{\rho_0} \sum_j (\frac{p_i}{\beta_i} + \frac{p_j}{\beta_j}) \nabla W_{ij}$     $\beta$  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

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

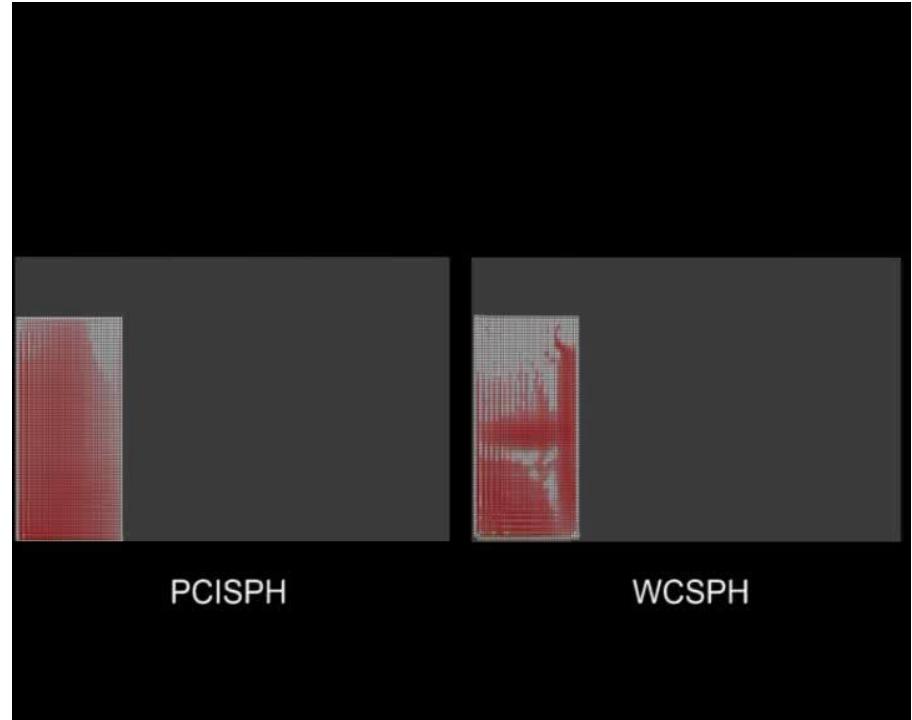
# *Iterative SESPH - Performance*

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- 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 SESPH (PCISPH)
  - [Solenthaler 2009]
  - iterative pressure computation
  - large time step
- non-iterative SESPH (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*

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- 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
- $\rho_i^*$  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

---

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

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

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

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

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

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

# *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

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- 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      predicted  
due to pressure      density error  
accelerations
- discretized PPE
  - $\mathbf{Ap} = \mathbf{s}$
  - $\underbrace{\Delta t^2 \sum_j m_j (\mathbf{a}_i^p - \mathbf{a}_j^p) \nabla W_{ij}}_{(\mathbf{Ap})_i} = \underbrace{\rho_0 - \rho_i^*}_{s_i} \quad \mathbf{a}_i^p = - \sum_j m_j \left( \frac{p_i}{\rho_i^2} - \frac{p_j}{\rho_j^2} \right) \nabla W_{ij}$
  - $\Delta t \sum_j m_j (\Delta t \mathbf{a}_i^p - \Delta t \mathbf{a}_j^p) \nabla W_{ij} = \rho_0 - \rho_i^*$       pressure accel. causes a  
 $\Delta t \sum_j m_j (\mathbf{v}_i^p - \mathbf{v}_j^p) \nabla W_{ij} = \rho_0 - \rho_i^*$       velocity change  $\mathbf{v}^p$  whose  
 $\Delta t \cdot \rho_i \cdot \nabla \cdot \mathbf{v}_i^p = \rho_0 - \rho_i^*$       divergence causes a  
density change

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

$$a_{ff} = \Delta t^2 \sum_{ff} m_{ff} \left( - \sum_{ff} \frac{m_{ff}}{\rho_0^2} \nabla W_{ff} - 2\gamma \sum_{fb} \frac{m_{fb}}{\rho_0^2} \nabla W_{ff} \right) \nabla W_{ff}$$
$$+ \Delta t^2 \sum_{ff} m_{ff} \left( \frac{m_f}{\rho_0^2} \nabla W_{ff} \right) \nabla W_{ff}$$
$$+ \Delta t^2 \sum_{fb} m_{fb} \left( - \sum_{ff} \frac{m_{ff}}{\rho_0^2} \nabla W_{ff} - 2\gamma \sum_{fb} \frac{m_{fb}}{\rho_0^2} \nabla W_{ff} \right) \nabla W_{ff}$$

# *Implicit Incompressible SPH*

## *Implementation with Boundary Handling*

---

- iteration  $l$

- first particle loop

- predicted pressure acceleration

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

- second particle loop

- predicted density change due to pressure acceleration

$$(\mathbf{Ap})_f^l = \Delta t^2 \sum_{ff} m_{ff} \left( (\mathbf{a}_f^p)^l - (\mathbf{a}_{ff}^p)^l \right) \nabla W_{ff} + \Delta t^2 \sum_{fb} m_{fb} (\mathbf{a}_f^p)^l \nabla W_{fb}$$

- pressure update

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

- predicted density deviation per particle

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

# *Comparison with Iterative SESPH*

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

$\Delta t$ [s]	PCISPH				IISPH				PCISPH / IISPH		
	avg.	iter.	pressure	overall	avg.	iter.	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	<b>1145</b>	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	<b>6.2</b>	3.8
0.004	-		-	-	33.5		273	<b>379</b>	-	-	-
0.005	-		-	-	45.8		297	383	-	-	-

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

# References

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# *Outline*

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