

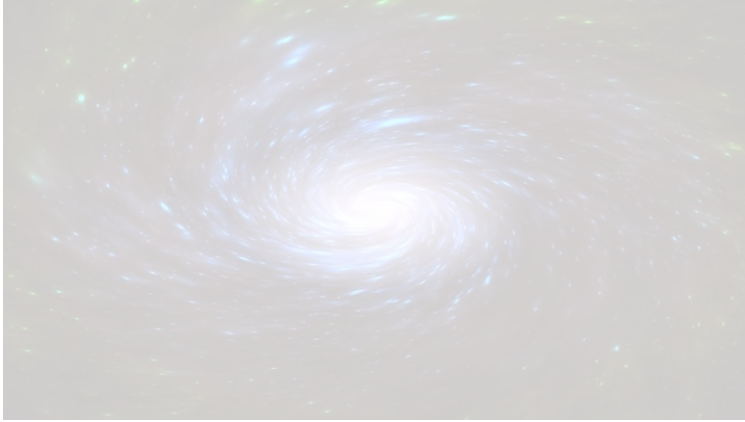


太极图形课

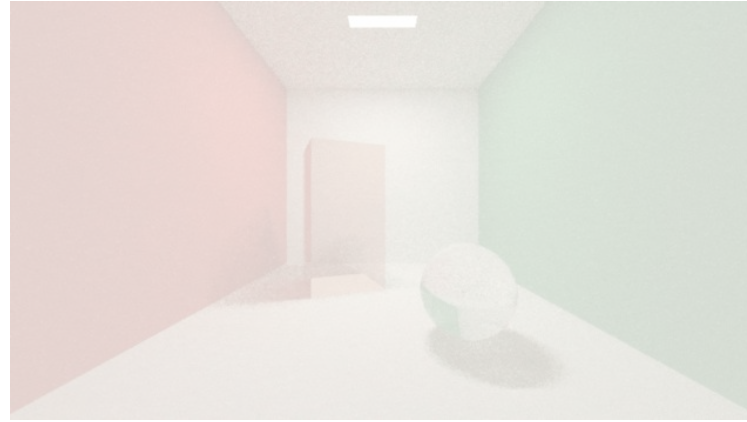
第09讲 Fluid Simulation 01: The Particle-based Methods



Where are we?



Procedural Animation



Rendering

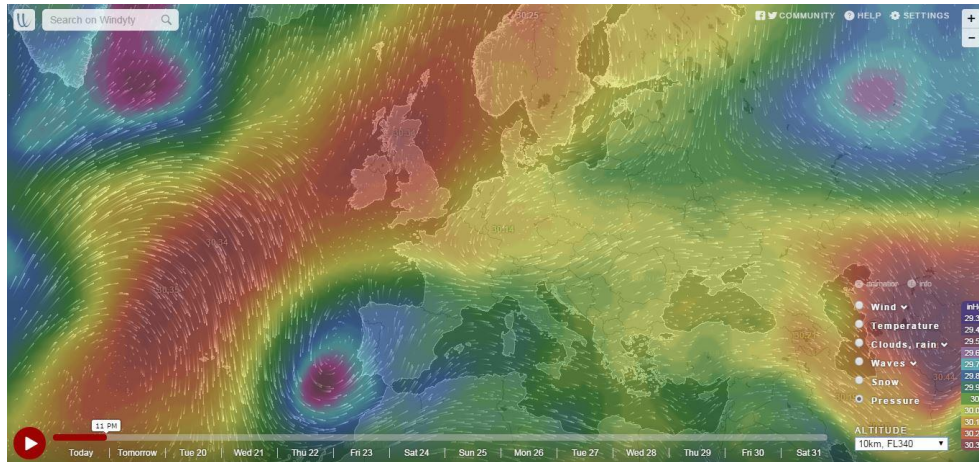


Deformable Simulation



Fluid Simulation

Fluid simulation



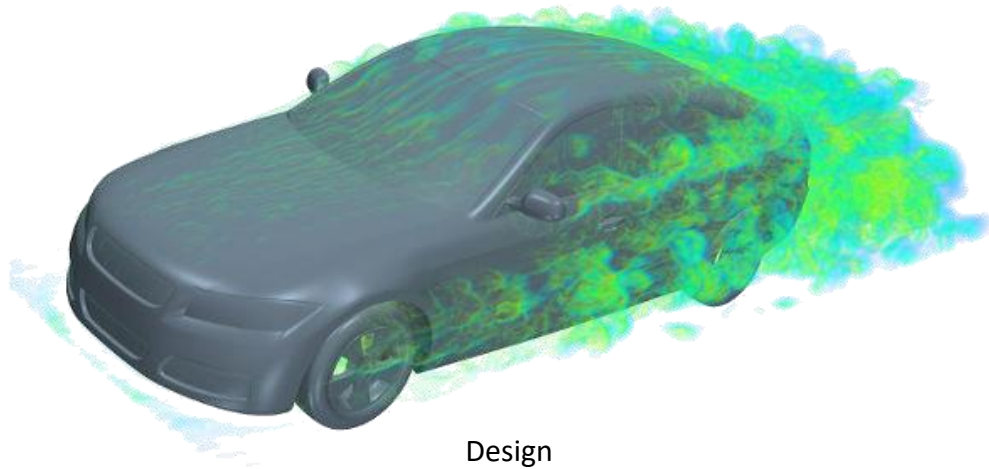
Forecast



VFX



Game



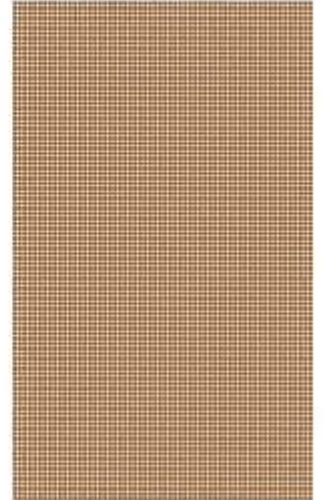
Design



Animation

Code of the day

- Code: https://github.com/taichiCourse01/taichi_sph
- Code courtesy of [@erizmr](#)

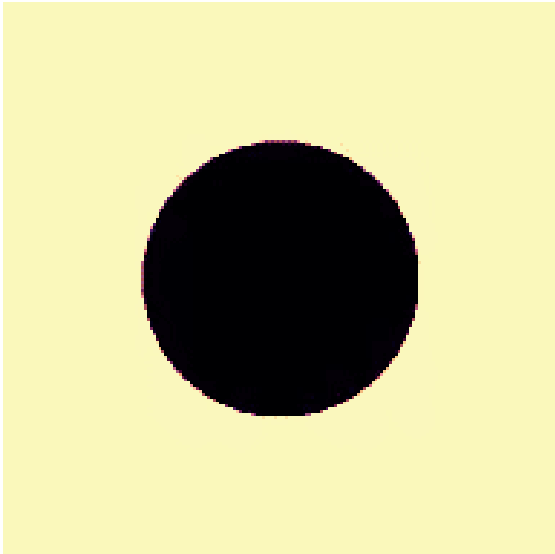


Outline today

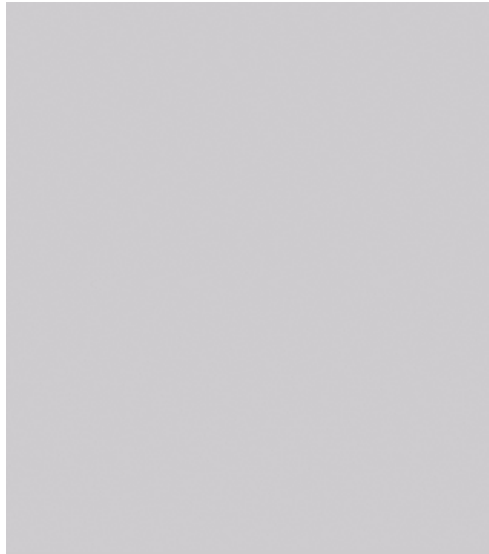
- Incompressible fluid dynamics
 - Incompressible Navier–Stokes equations
- Time discretization
 - Operator splitting
 - Integration with the weakly compressible (WC) assumption
- Spatial discretization
 - Smoothed particle hydrodynamics (SPH)
- Implementation details (WCSPH)
 - Simulation Pipeline
 - Boundary conditions
 - Neighbor search

Incompressible fluid dynamics

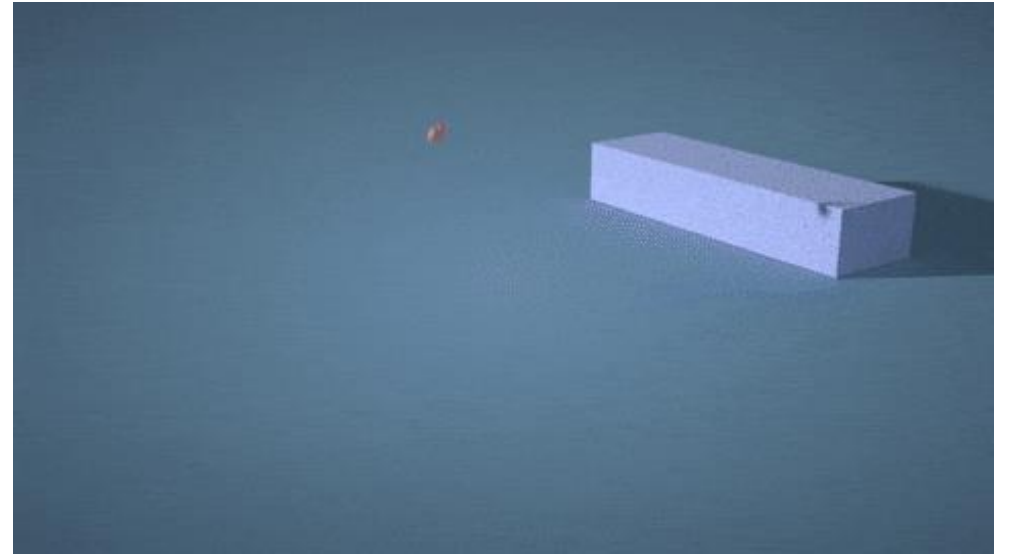
Compressible and incompressible fluids



Compressible
shock wave

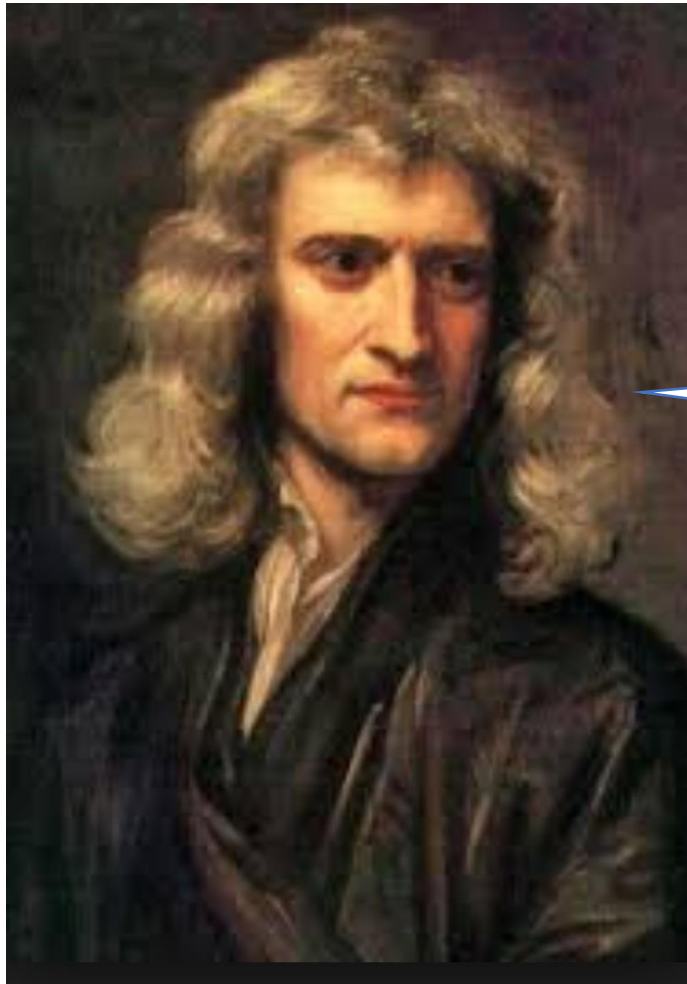


Incompressible
smoke



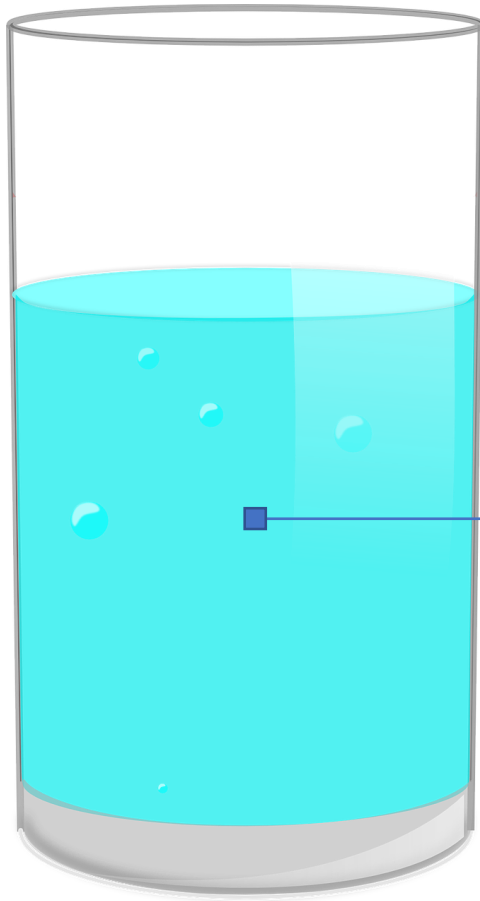
Incompressible
liquid

Laws of physics for incompressible fluids



$$\textcircled{f} = Ma$$

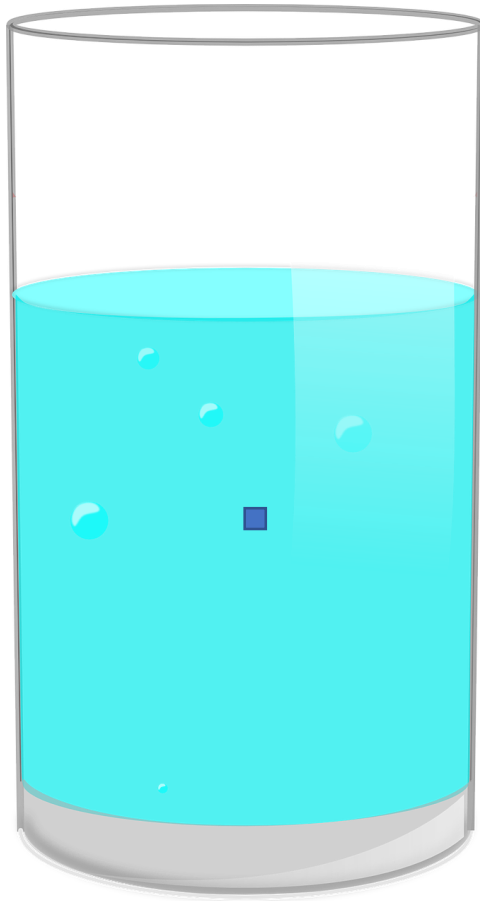
Forces for incompressible fluids



$$f = ?$$



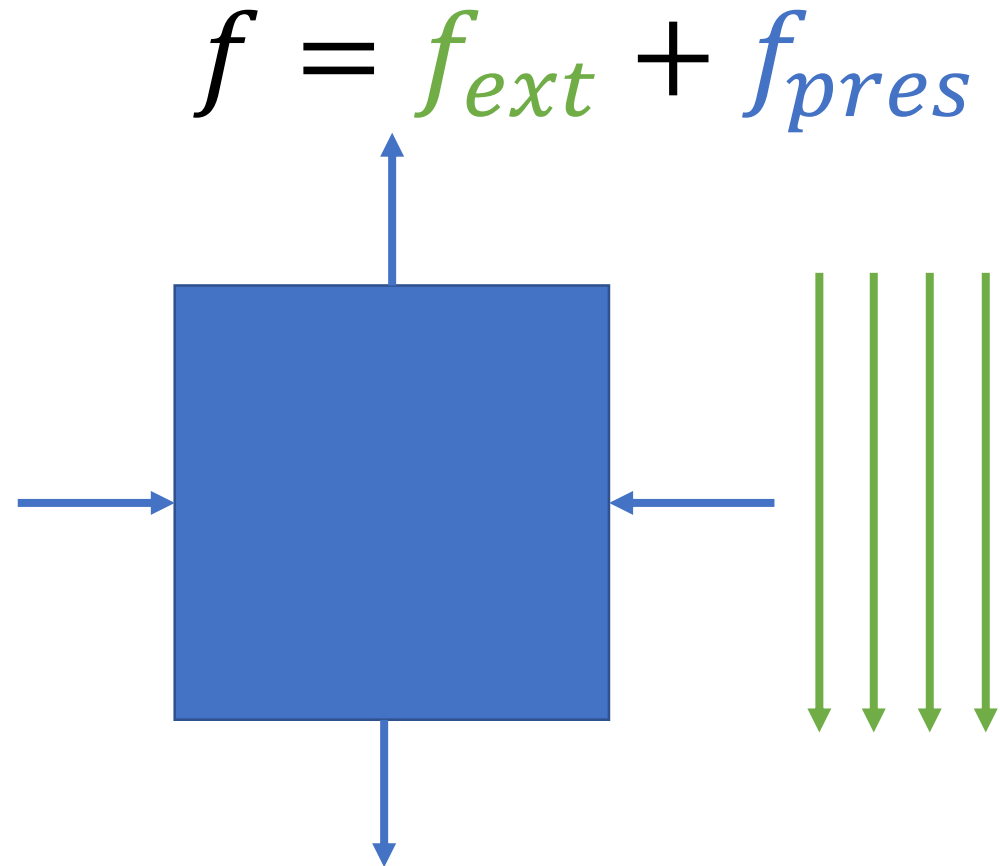
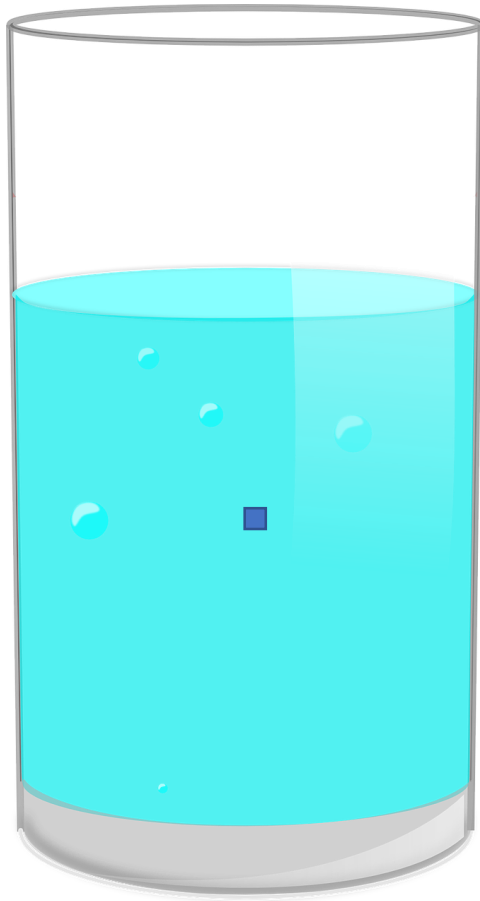
Forces for incompressible fluids



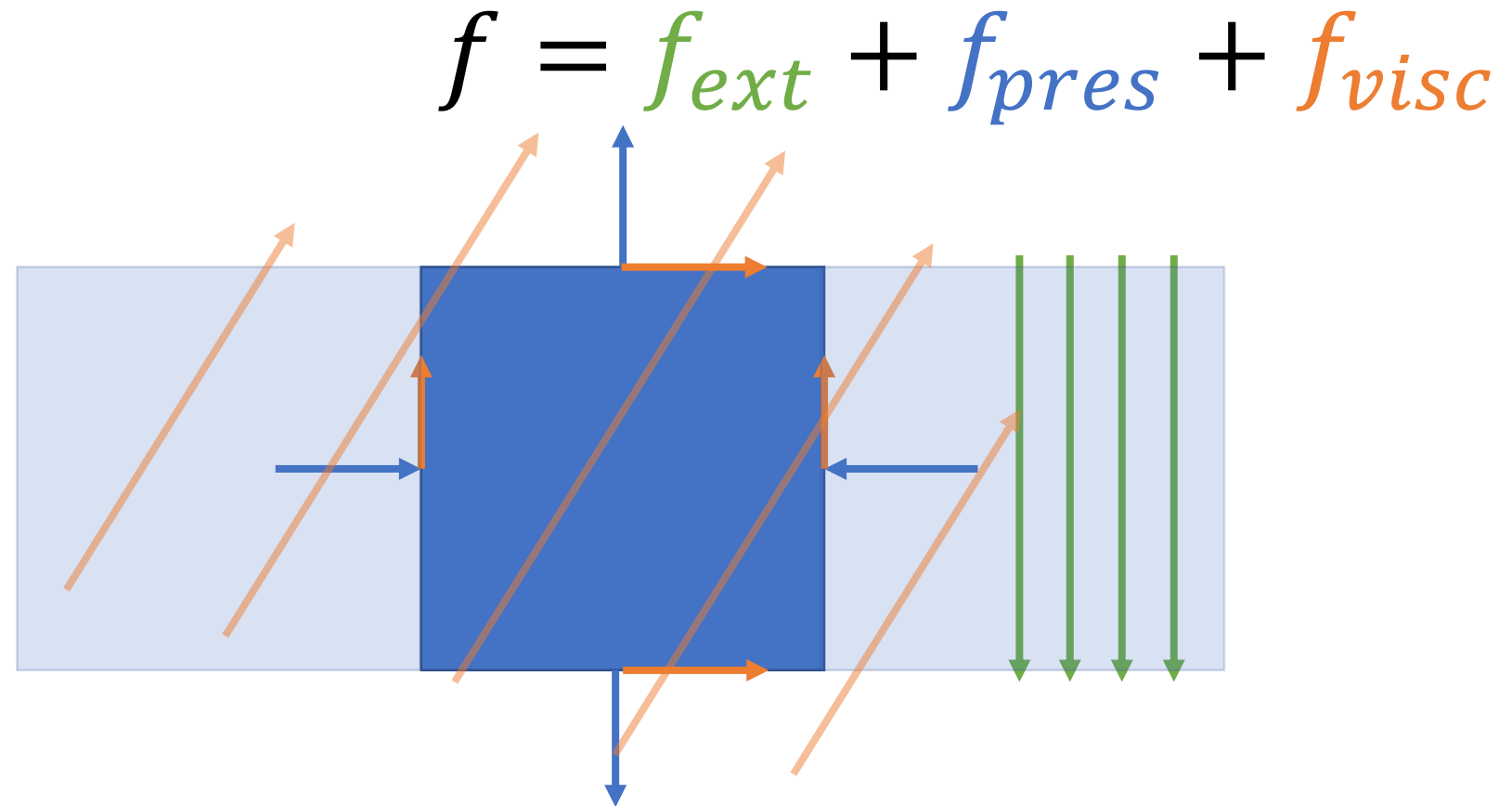
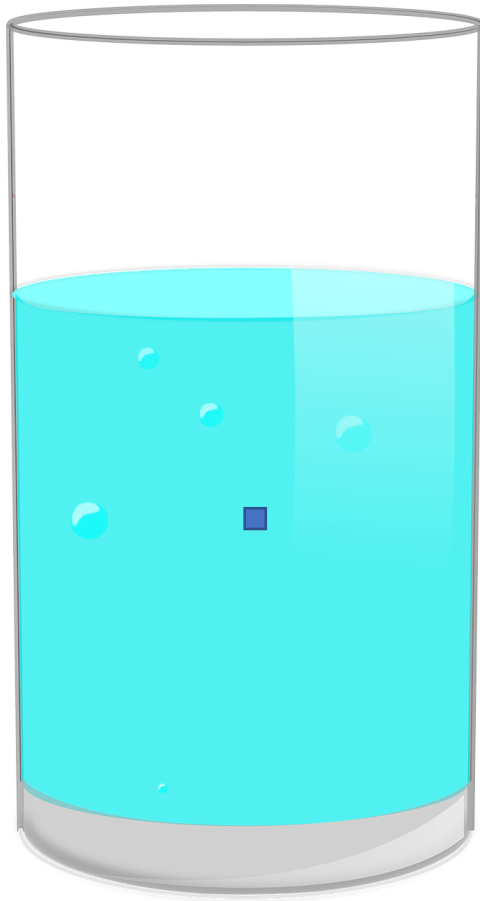
$$f = f_{ext}$$



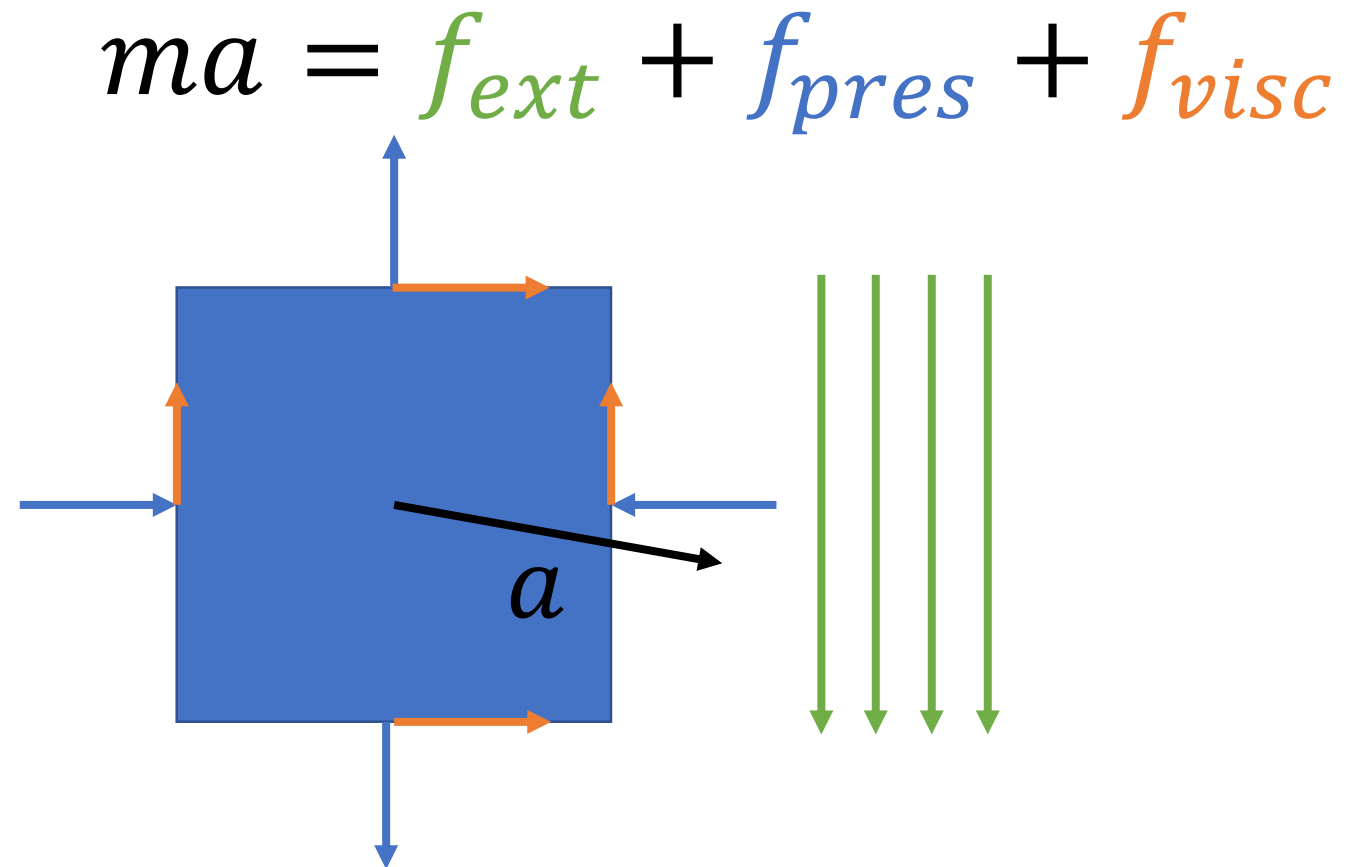
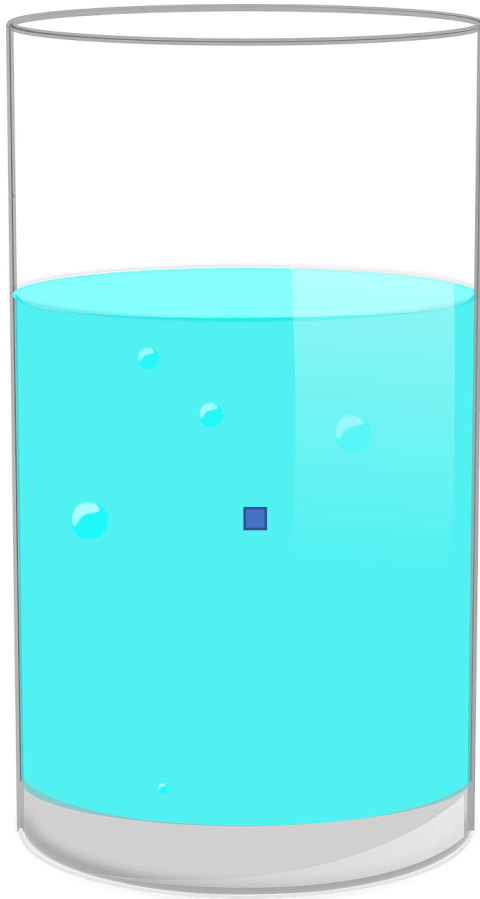
Forces for incompressible fluids



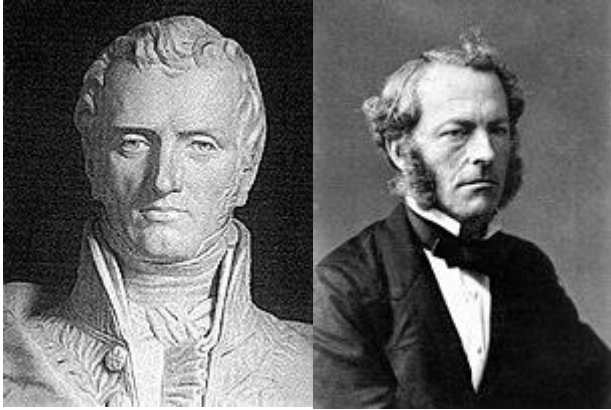
Forces for incompressible fluids



Laws of motion for incompressible fluids

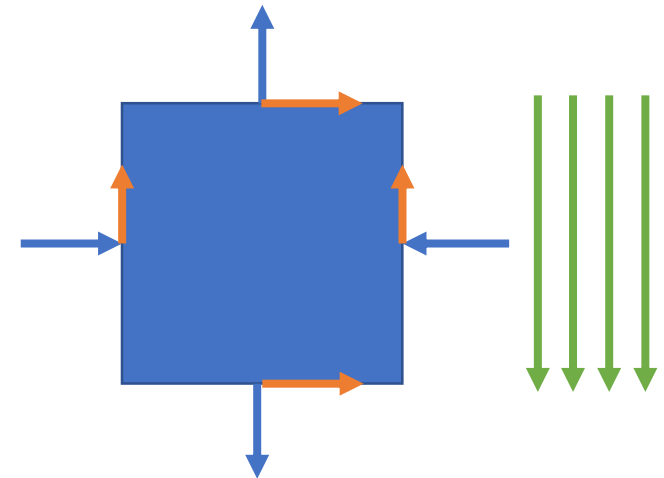


Incompressible Navier-Stokes equation



$$ma = f_{ext} + f_{pres} + f_{visc}$$

$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$
$$\nabla \cdot v = 0$$



The spatial derivative operators in 3D

$$\nabla = \begin{bmatrix} \frac{\partial}{\partial x} \\ \frac{\partial}{\partial y} \\ \frac{\partial}{\partial z} \end{bmatrix}$$

- Gradient $\nabla: \mathbb{R}^1 \rightarrow \mathbb{R}^3$
 - $\text{grad } s = \nabla s = \left[\frac{\partial s}{\partial x}, \frac{\partial s}{\partial y}, \frac{\partial s}{\partial z} \right]^T$
- Divergence $\nabla \cdot: \mathbb{R}^3 \rightarrow \mathbb{R}^1$
 - $\text{div } v = \nabla \cdot v = \frac{\partial v_x}{\partial x} + \frac{\partial v_y}{\partial y} + \frac{\partial v_z}{\partial z}$
- Curl $\nabla \times: \mathbb{R}^3 \rightarrow \mathbb{R}^3$
 - $\text{curl } v = \nabla \times v = \left[\frac{\partial v_z}{\partial y} - \frac{\partial v_y}{\partial z}, \frac{\partial v_z}{\partial x} - \frac{\partial v_x}{\partial z}, \frac{\partial v_y}{\partial x} - \frac{\partial v_x}{\partial y} \right]^T$
- Laplace $\Delta = \nabla^2 = \nabla \cdot \nabla: \mathbb{R}^n \rightarrow \mathbb{R}^n$
 - $\text{laplace } s = \text{div } (\text{grad } s) = \nabla^2 s = \frac{\partial^2 s}{\partial x^2} + \frac{\partial^2 s}{\partial y^2} + \frac{\partial^2 s}{\partial z^2}$

Incompressible Navier-Stokes equation

$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$

$$\nabla \cdot v = 0$$

ρ : mass-density

$\frac{D(\cdot)}{Dt}$: material derivative

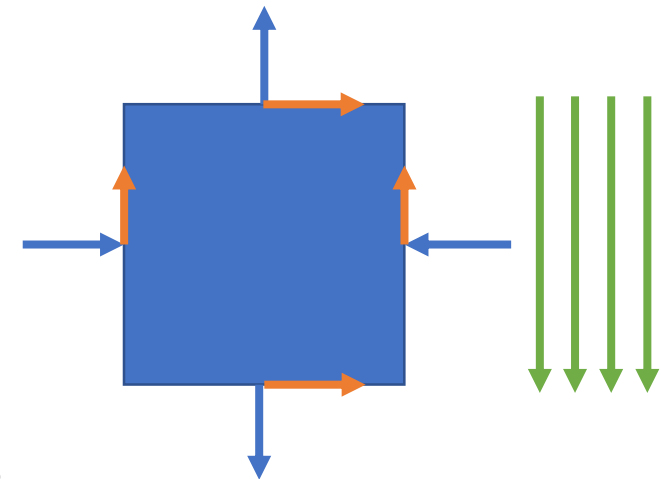
g : gravity ($g = [0, -9.8, 0]^T$ in 3D)

p : pressure, $p = k(\rho - \rho_0)$

μ : shear modulus (dynamics visc.), $\nu = \frac{\mu}{\rho_0}$: kinematic viscosity

∇ : vector differential operator ($\nabla = \left[\frac{\partial}{\partial x}, \frac{\partial}{\partial y}, \frac{\partial}{\partial z} \right]^T$ in 3D)

$\nabla^2 = \Delta$: Laplace operator, $\Delta f = \nabla \cdot \nabla f$ ($\Delta f = \frac{\partial^2}{\partial x^2} f + \frac{\partial^2}{\partial y^2} f + \frac{\partial^2}{\partial z^2} f$ in 3D)



The ma in $f = ma$

$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$

$$\nabla \cdot v = 0$$

This is simply “mass” times “acceleration” divided by “volume”

External force term

$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$
$$\nabla \cdot v = 0$$

Gravitational force divided by “volume”

Viscosity term

$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$

$$\nabla \cdot v = 0$$

$$\nabla = \begin{bmatrix} \frac{\partial}{\partial x} \\ \frac{\partial}{\partial y} \\ \frac{\partial}{\partial z} \end{bmatrix}$$

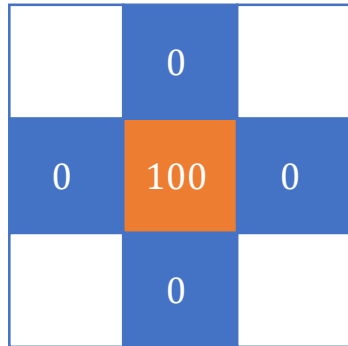
$\nabla^2 = \Delta$: Laplace operator (or diffusion operator)

∇^2 : takes a scalar/vector, returns a scalar/vector

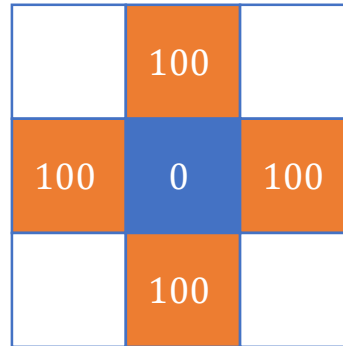
$$\nabla^2 f = \frac{\partial^2}{\partial x^2} f + \frac{\partial^2}{\partial y^2} f + \frac{\partial^2}{\partial z^2} f \text{ in 3D}$$

Still remember the diffusion problem?

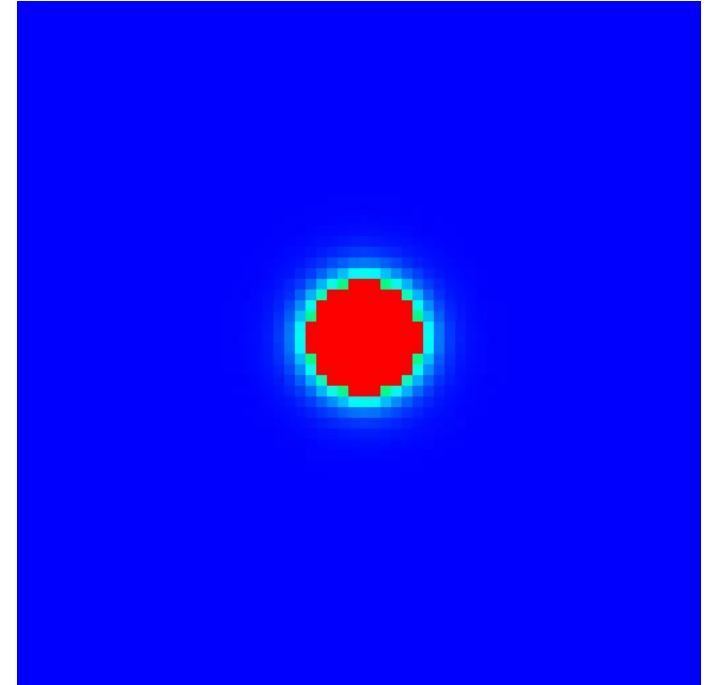
- $\frac{\partial T}{\partial t} = \kappa \nabla^2 T$



$T \downarrow$



$T \uparrow$

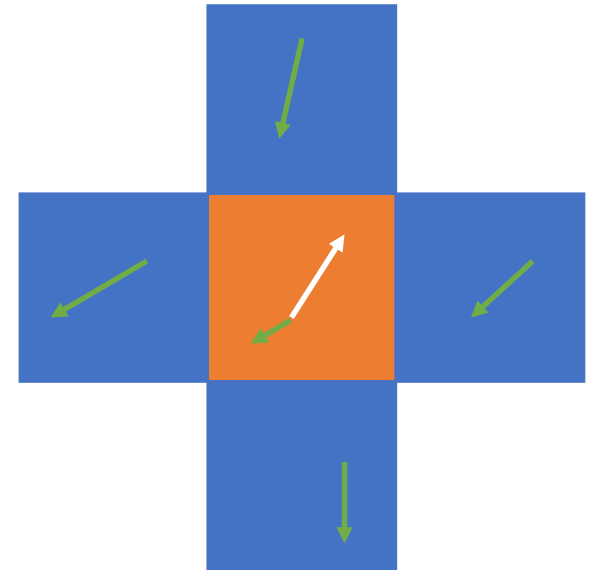


Viscosity term

$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$

$$\nabla \cdot v = 0$$

Viscosity term: how fluids want to move together



Viscosity term

$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$

$$\nabla \cdot v = 0$$

μ :



>



>



μ : some fluids are more viscous than others

Pressure term

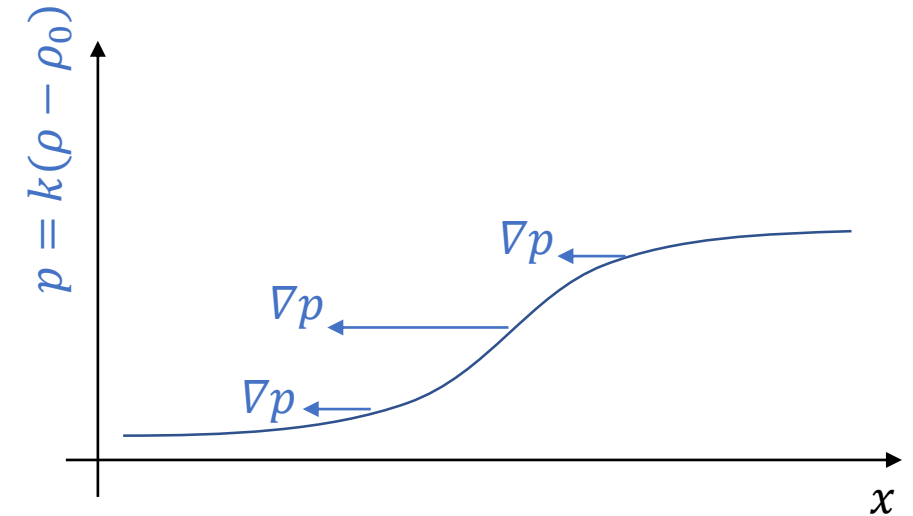
$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$

$$\nabla \cdot v = 0$$

∇ : gradient operator

∇ : takes a scalar, returns a vector

$$\nabla = \begin{bmatrix} \frac{\partial}{\partial x} \\ \frac{\partial}{\partial y} \\ \frac{\partial}{\partial z} \end{bmatrix}$$



Pressure term: fluids do not want to change volume

Pressure term

$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$

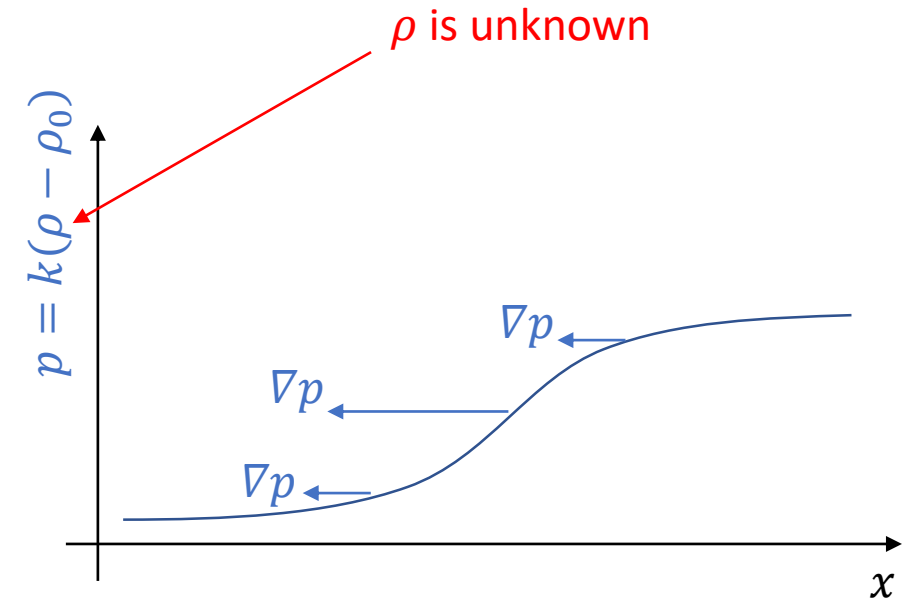
$$\nabla \cdot v = 0$$

∇ : gradient operator

∇ : takes a scalar, returns a vector

$$\nabla = \begin{bmatrix} \frac{\partial}{\partial x} \\ \frac{\partial}{\partial y} \\ \frac{\partial}{\partial z} \end{bmatrix}$$

Pressure term: fluids do not want to change volume



Divergence free condition

$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$

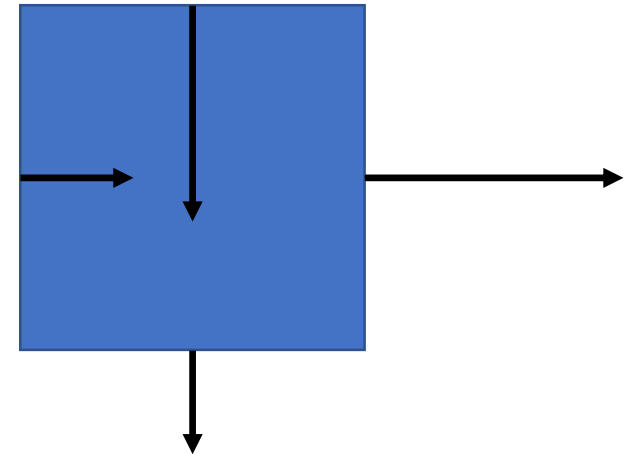
$$\nabla \cdot v = 0$$

$$\nabla = \begin{bmatrix} \frac{\partial}{\partial x} \\ \frac{\partial}{\partial y} \\ \frac{\partial}{\partial z} \end{bmatrix}$$

$\nabla \cdot$: divergence operator

$\nabla \cdot$: takes a vector, returns a scalar

$$\nabla \cdot v = \frac{\partial v_x}{\partial x} + \frac{\partial v_y}{\partial y} + \frac{\partial v_z}{\partial z} = 0$$

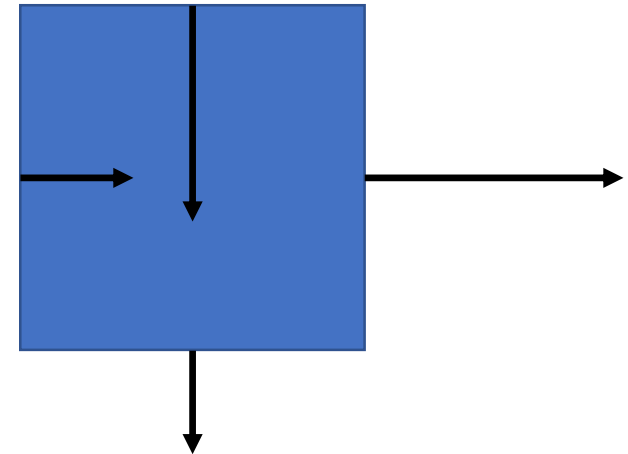


Divergence free condition

$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$

$$\nabla \cdot v = 0 \Leftrightarrow \frac{D\rho}{Dt} = (v \cdot \nabla)\rho = 0$$

Divergence ($\nabla \cdot v$) free: outbound flow equals to inbound flow
The mass conserving condition

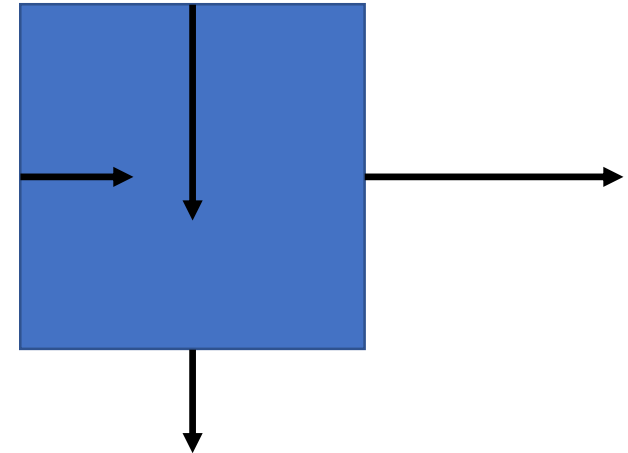


Divergence free condition

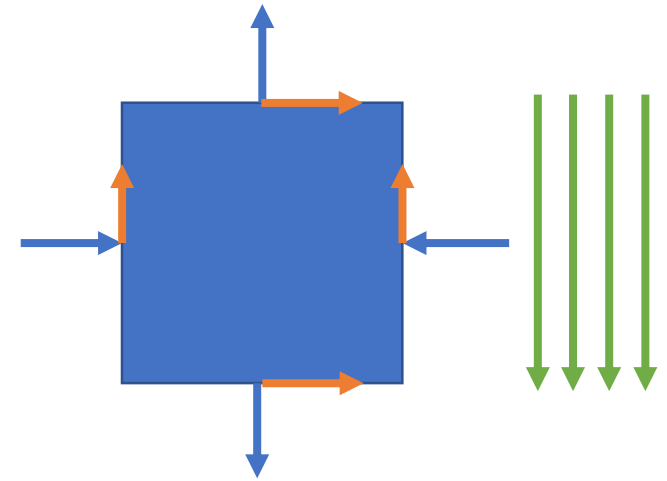
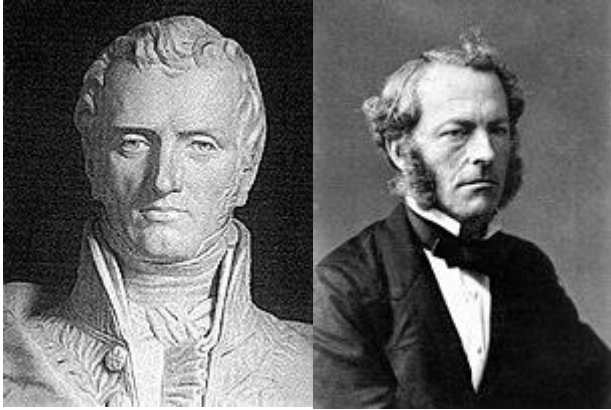
$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$

$$\nabla \cdot v = 0 \quad \leftarrow \text{The incompressibility assumption goes here!}$$

Divergence ($\nabla \cdot v$) free: outbound flow equals to inbound flow
The mass conserving condition



Incompressible Navier-Stokes equation

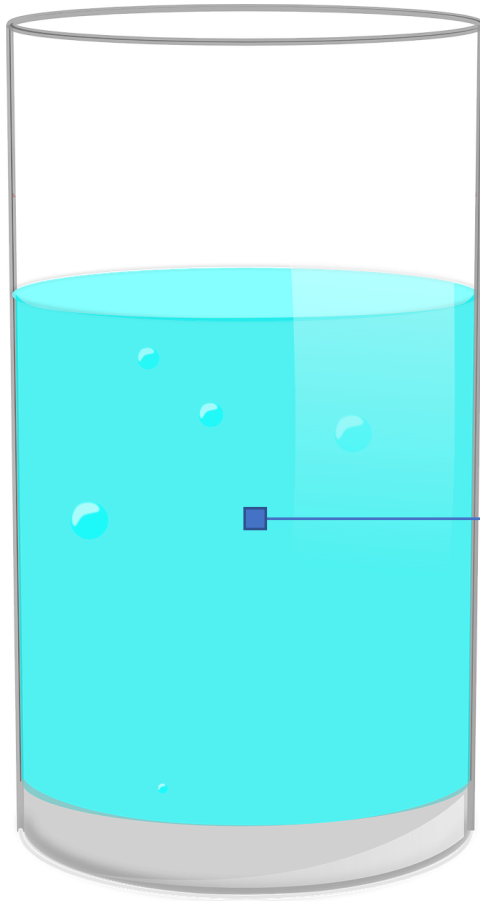


$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v \quad \leftarrow \text{The momentum equation}$$

$$\nabla \cdot v = 0$$

\leftarrow The mass conserving condition

Integrate incompressible Navier-Stokes equation?

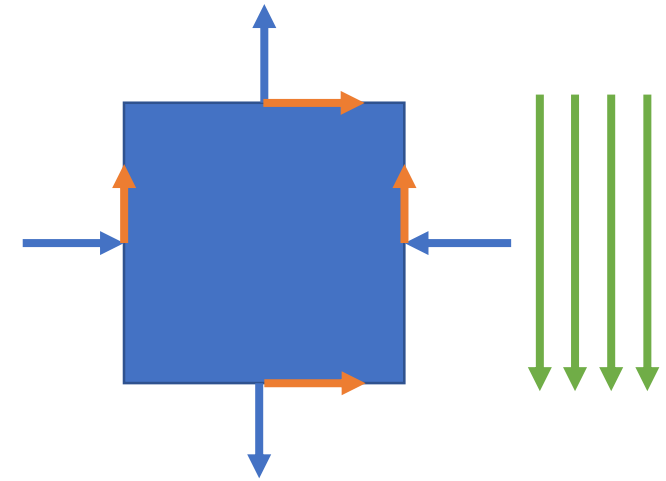
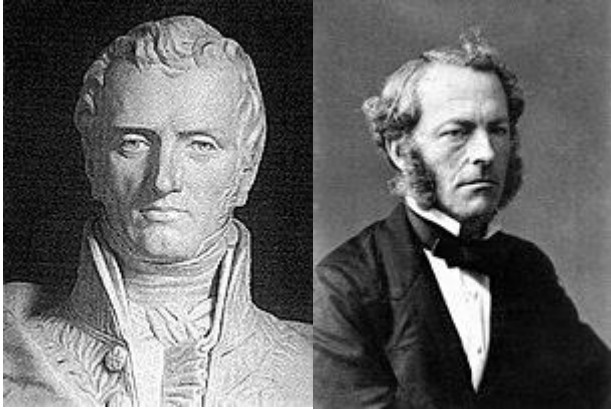


$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$
$$\nabla \cdot v = 0$$



Temporal discretization

Incompressible Navier-Stokes equation



The ρ in $p = k(\rho - \rho_0)$ is unknown!

$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v \quad \leftarrow \text{The momentum equation}$$

$$\nabla \cdot v = 0 \quad \leftarrow \text{The mass conserving condition}$$

$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$

$$\nabla \cdot v = 0$$

- The divergence free condition $\nabla \cdot v = 0$ comes to the rescue:

- $\nabla \cdot v = 0 \iff \frac{D\rho}{Dt} = 0$

- Integrate the incompressible Navier-Stokes equation in steps:

- Step 1: input v_n , output $v_{n+0.5}$

- $\rho \frac{Dv}{Dt} = \rho g + \mu \nabla^2 v$

- Step 2: input $v_{n+0.5}$, output v_{n+1}

- $\rho \frac{Dv}{Dt} = -\nabla p$

- $\nabla \cdot v = 0$

This integration method is sometime referred as “*Operator splitting*” or “*Advection-Projection*” in different contexts

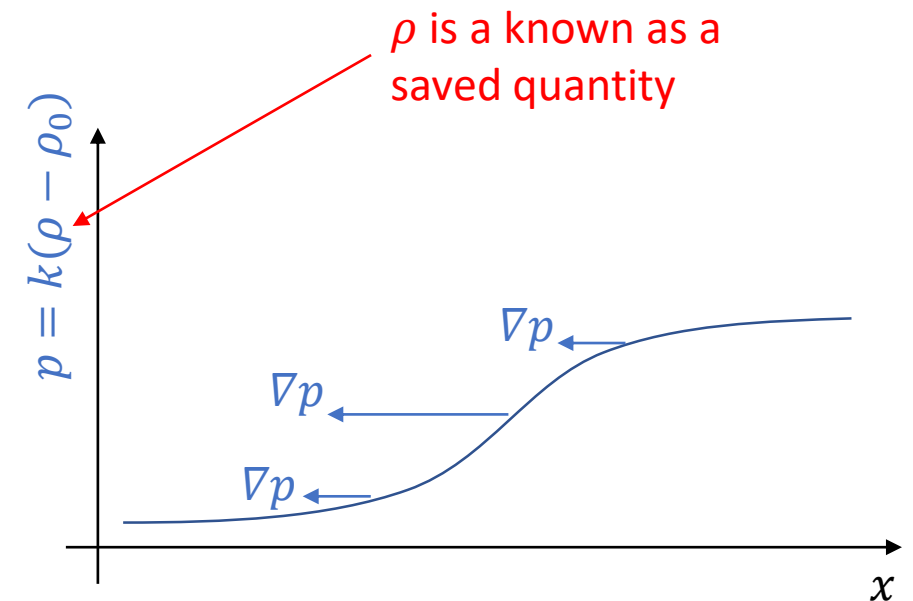
Full time integration

- Given x_n, v_n :
 - Step 1: Advection / external and viscosity force integration
 - Solve: $dv = g + \nu \nabla^2 v_n$
 - Update: $v_{n+0.5} = v_n + \Delta t dv$
 - Step 2: Projection / pressure solver
 - Solve: $dv = -\frac{1}{\rho} \nabla(k(\rho - \rho_0))$ and $\frac{D\rho}{Dt} = \nabla \cdot (v_{n+0.5} + dv) = 0$
 - Update: $v_{n+1} = v_{n+0.5} + \Delta t dv$
 - Step 3: Update position
 - Update: $x_{n+1} = x_n + \Delta t v_{n+1}$
 - Return x_{n+1}, v_{n+1}

The weakly compressible assumption

- Storing the density ρ as an individual variable that advect with the velocity field:

$$\frac{Dv}{Dt} = g - \frac{1}{\rho} \nabla p + \nu \nabla^2 v$$
$$\nabla \cdot v = 0$$



Integrate with the weakly compressible assumption

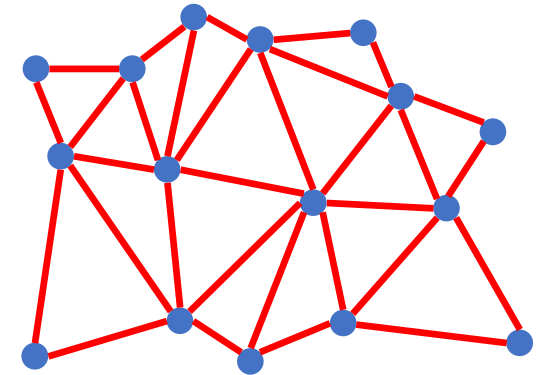
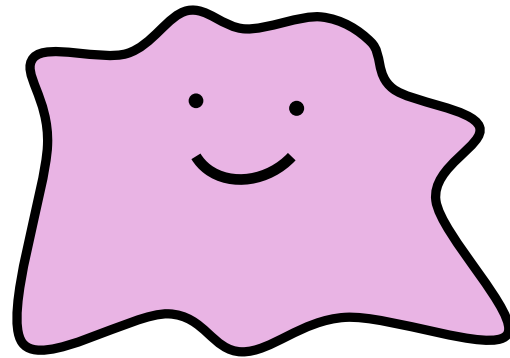
$$\frac{Dv}{Dt} = g - \frac{1}{\rho} \nabla p + \nu \nabla^2 v$$

- Given x_n, v_n :
 - Step 1: Advection / external and viscosity force integration
 - Solve: $dv = g + \nu \nabla^2 v_n$
 - Update: $v_{n+0.5} = v_n + \Delta t \, dv$
 - Step 2: Projection / pressure solver
 - Solve: $dv = -\frac{1}{\rho} \nabla (k(\rho - \rho_0))$
 - Update: $v_{n+1} = v_{n+0.5} + \Delta t \, dv$
 - Step 3: Update position
 - Update: $x_{n+1} = x_n + \Delta t \, v_{n+1}$
 - Return x_{n+1}, v_{n+1}

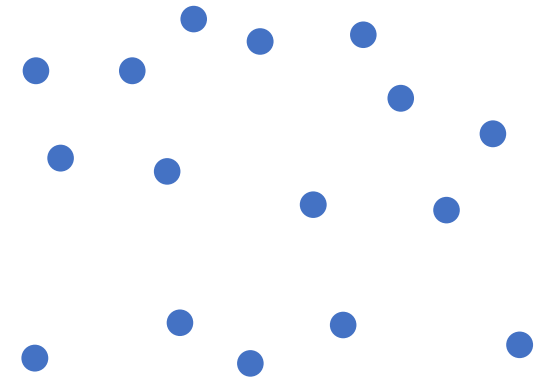
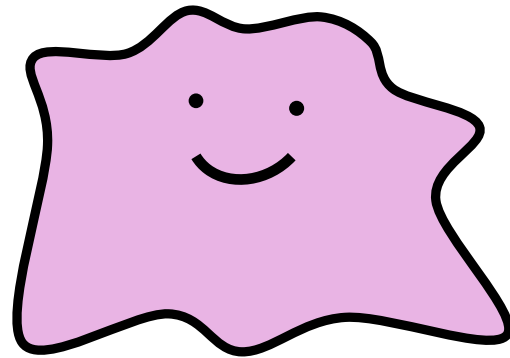
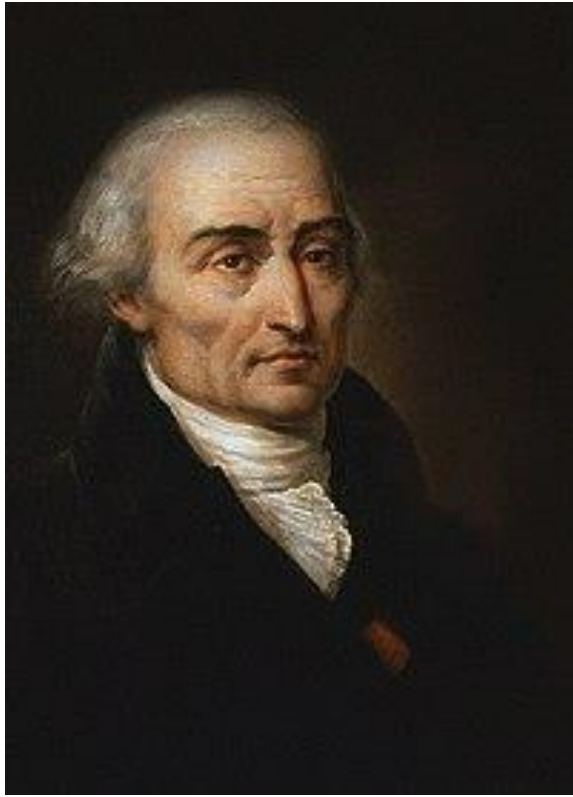
This is nothing but *Symplectic Euler* integration.

Spatial discretization (Lagrangian view)

Previously in this course (mesh-based simulation)



Today in this course (mesh-free simulation)

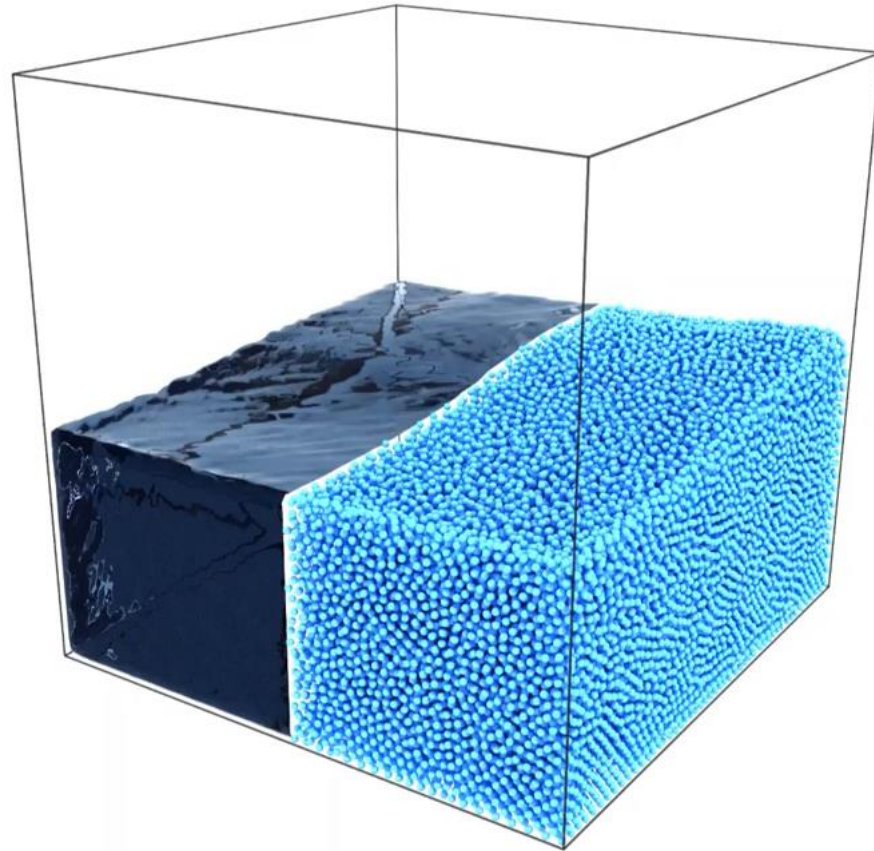


Fluid can be discretized using particles as well



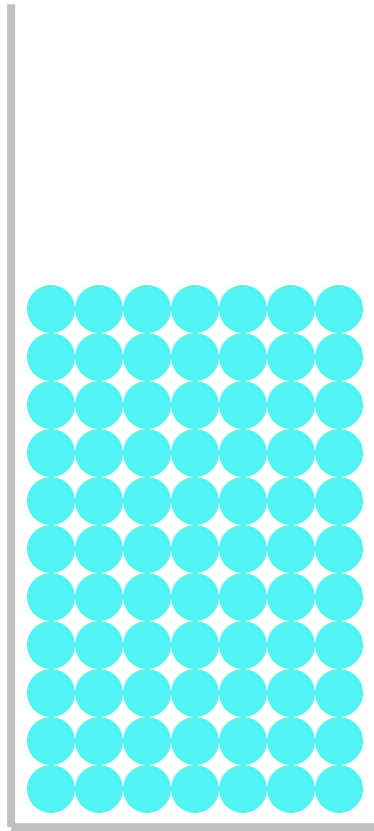
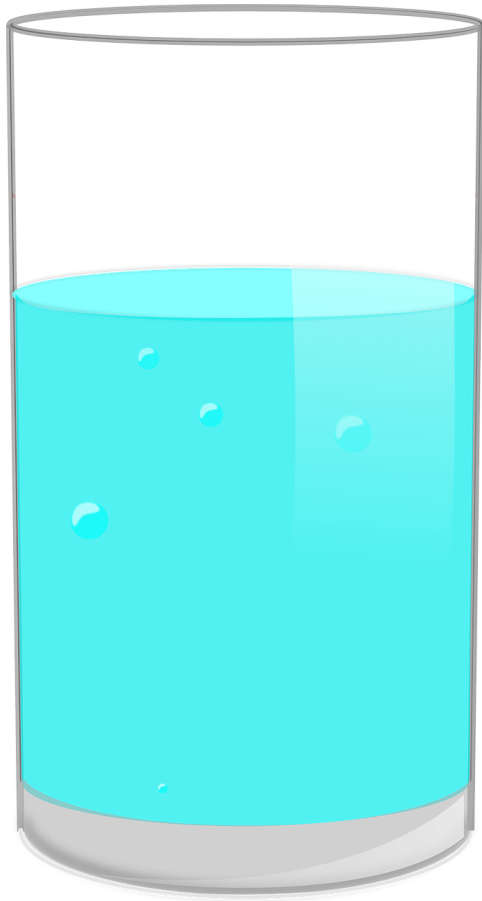
[The Big Bang Theory]

Fluid can be discretized using particles as well

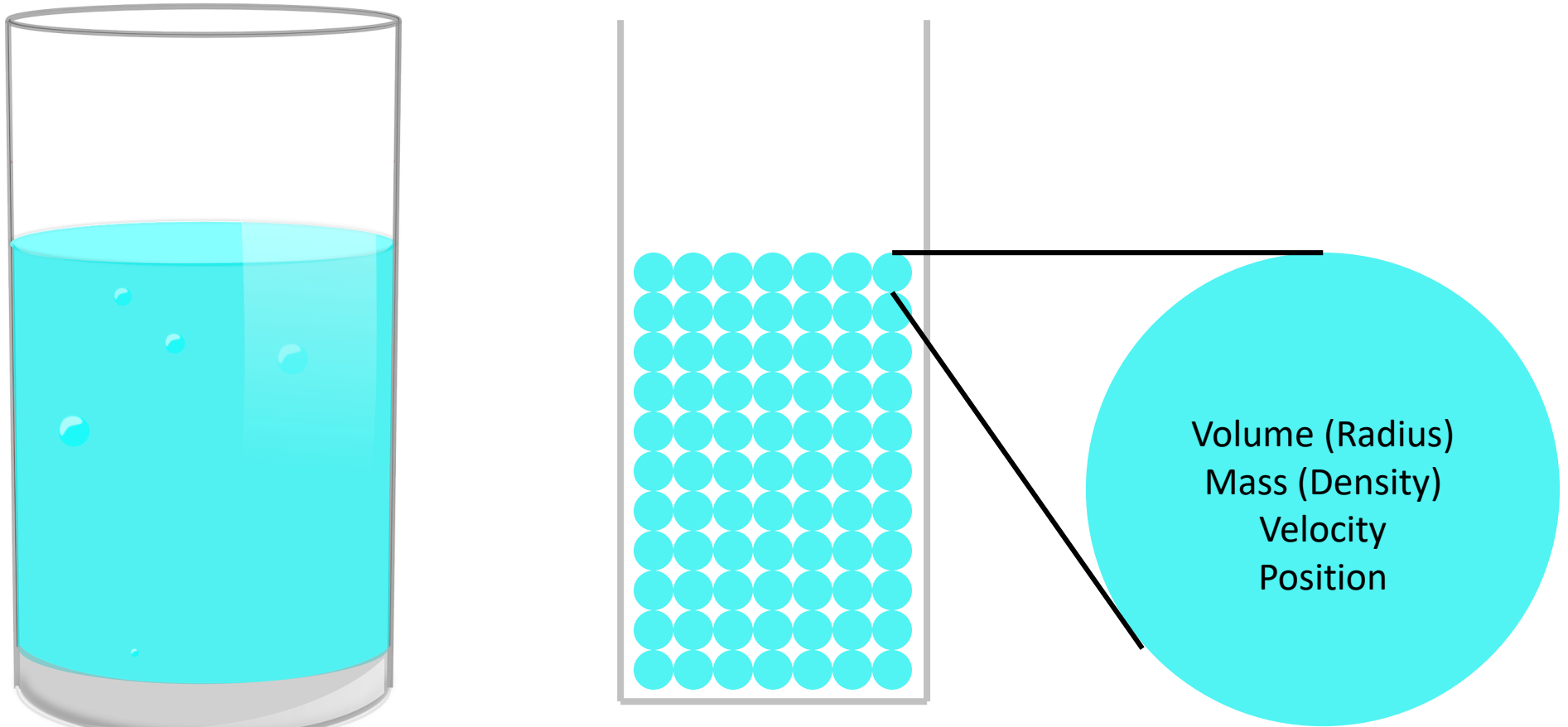


[www.dive-solutions.de]

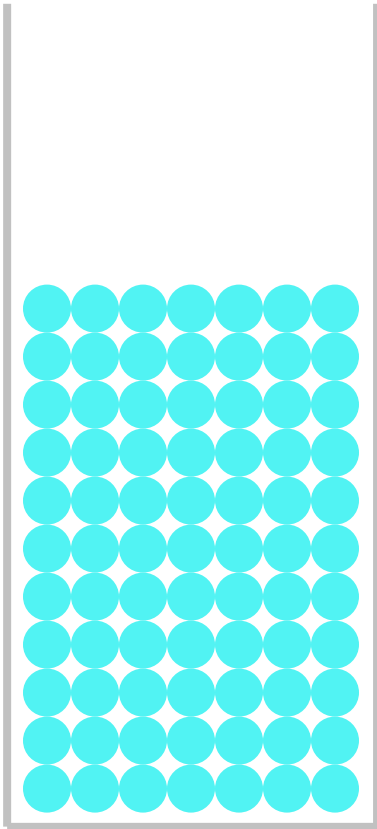
Fluid -> Particles



What do we store in a particle?



Fluid dynamics with particles (weakly compressible)



Continuous view:

$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$

$$\nabla \cdot v = 0$$

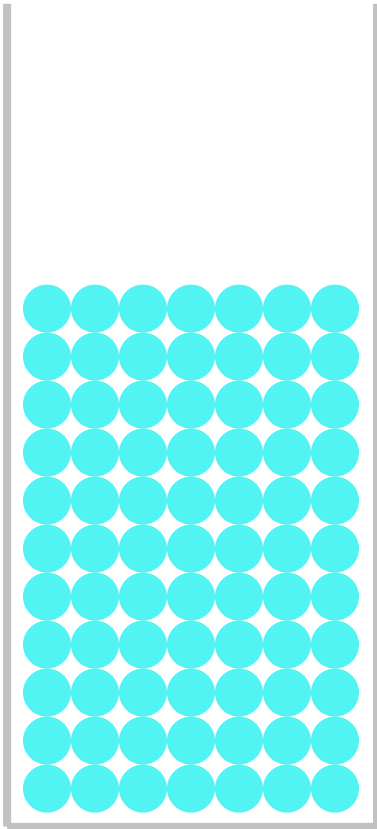
Discrete view (using particles):

$$\frac{dv_i}{dt} = \underbrace{g - \frac{1}{\rho} \nabla p(x_i) + v \nabla^2 v(x_i)}_{a_i}, \text{ where } v = \frac{\mu}{\rho_0}$$

Time integration (Symplectic Euler):

- for i in particles:
 - $v_i = v_i + \Delta t a_i$
- for i in particles:
 - $x_i = x_i + \Delta t v_i$

Fluid dynamics with particles (weakly compressible)



Continuous view:

$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$

$$\nabla \cdot v = 0$$

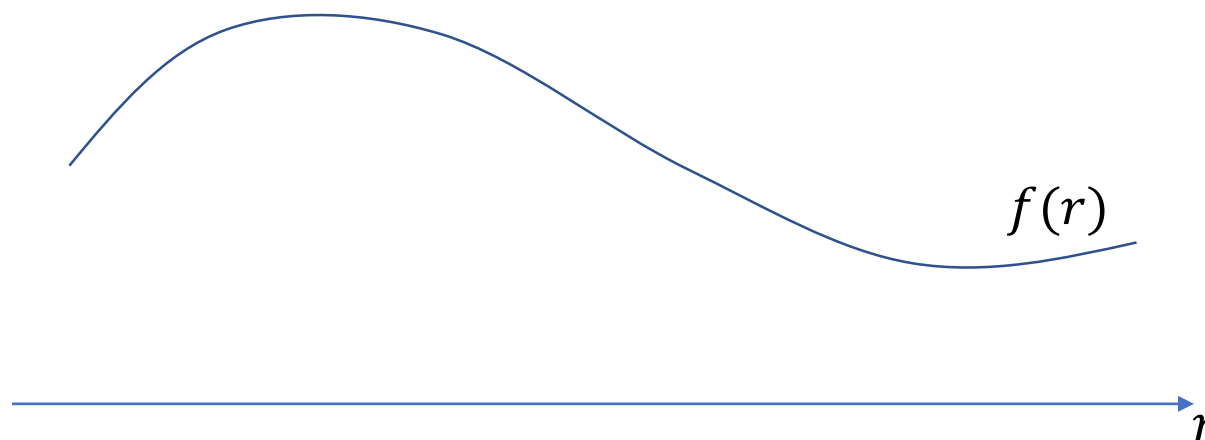
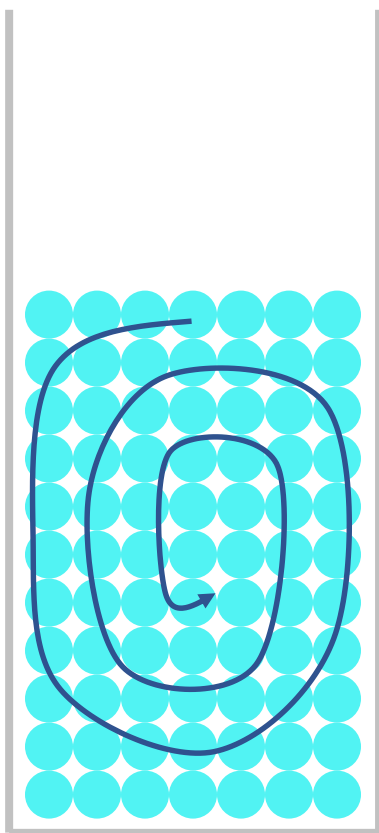
Discrete view (using particles):

$$\frac{dv_i}{dt} = g - \frac{1}{\rho} \nabla p(x_i) + \underbrace{\nu \nabla^2 v(x_i)}_{a_i}, \text{ where } \nu = \frac{\mu}{\rho_0}$$

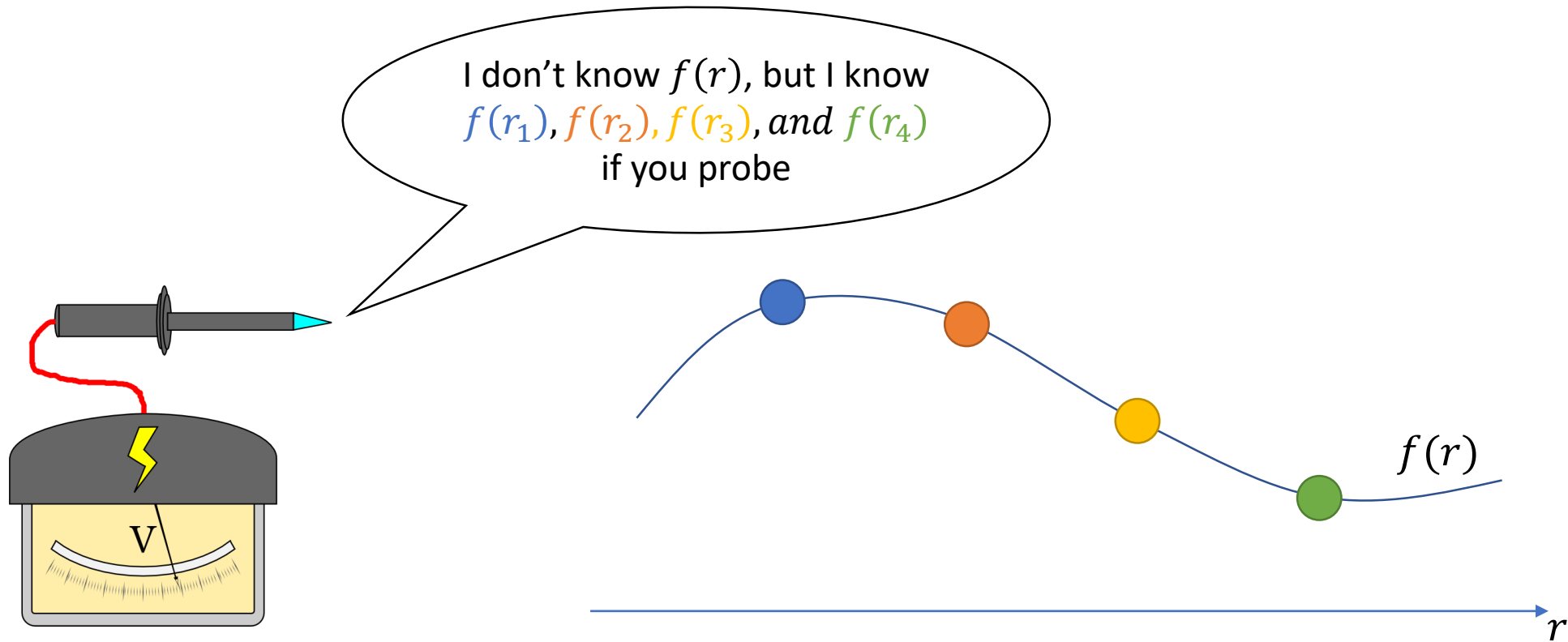
Time integration (Symplectic Euler):

- for i in particles:
 - $v_i = v_i + \Delta t a_i$
- for i in particles:
 - $x_i = x_i + \Delta t v_i$

How to evaluate a function? $\rho(x_i)$ $\nabla p(x_i)$ $\nabla^2 v(x_i)$



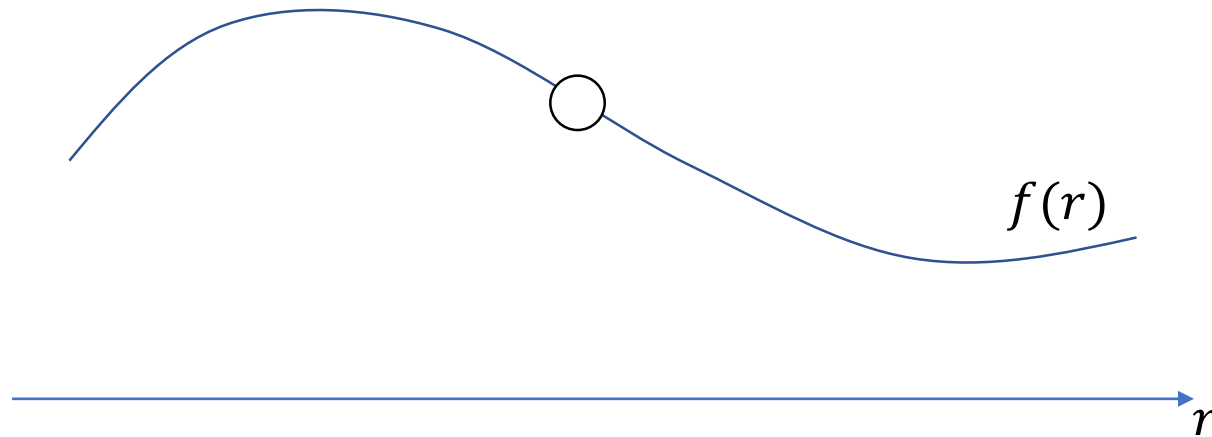
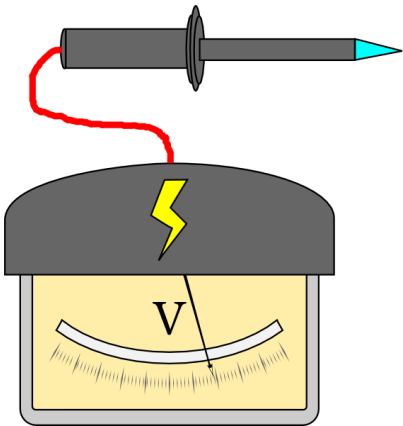
Imaging a magic probe...



A trivial identity with Dirac delta

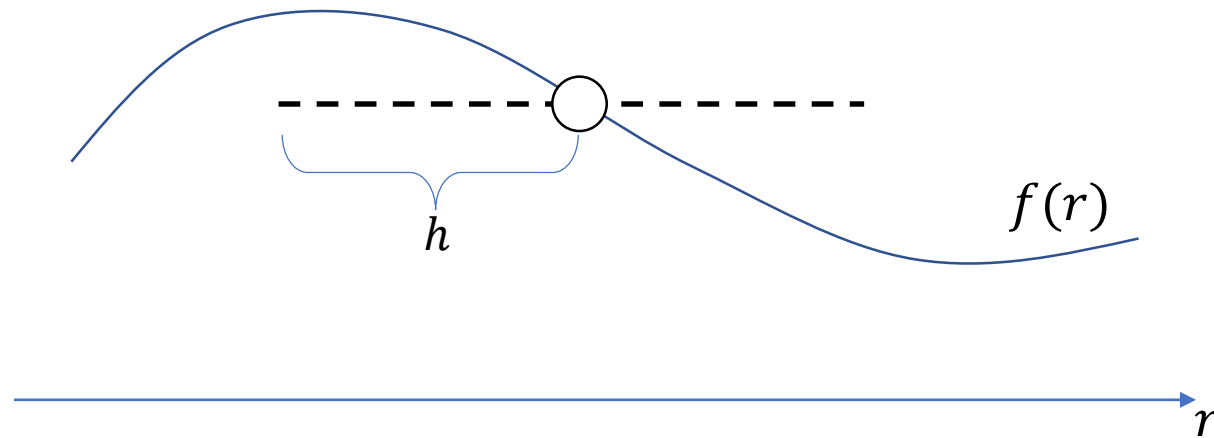
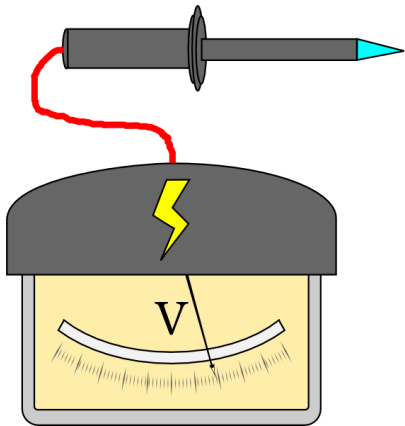
$$f(r) = \int_{-\infty}^{\infty} f(r') \delta(r - r') dr'$$

$$\delta(r) = \begin{cases} +\infty, & \text{if } r = 0 \\ 0, & \text{otherwise} \end{cases} \text{ and } \int_{-\infty}^{\infty} \delta(r) dr = 1$$



Let us widen the Dirac delta

$$f(r) \approx \int f(r')W(r-r',h)dr' , \text{ where } \lim_{h \rightarrow 0} W(r,h) = \delta(r)$$

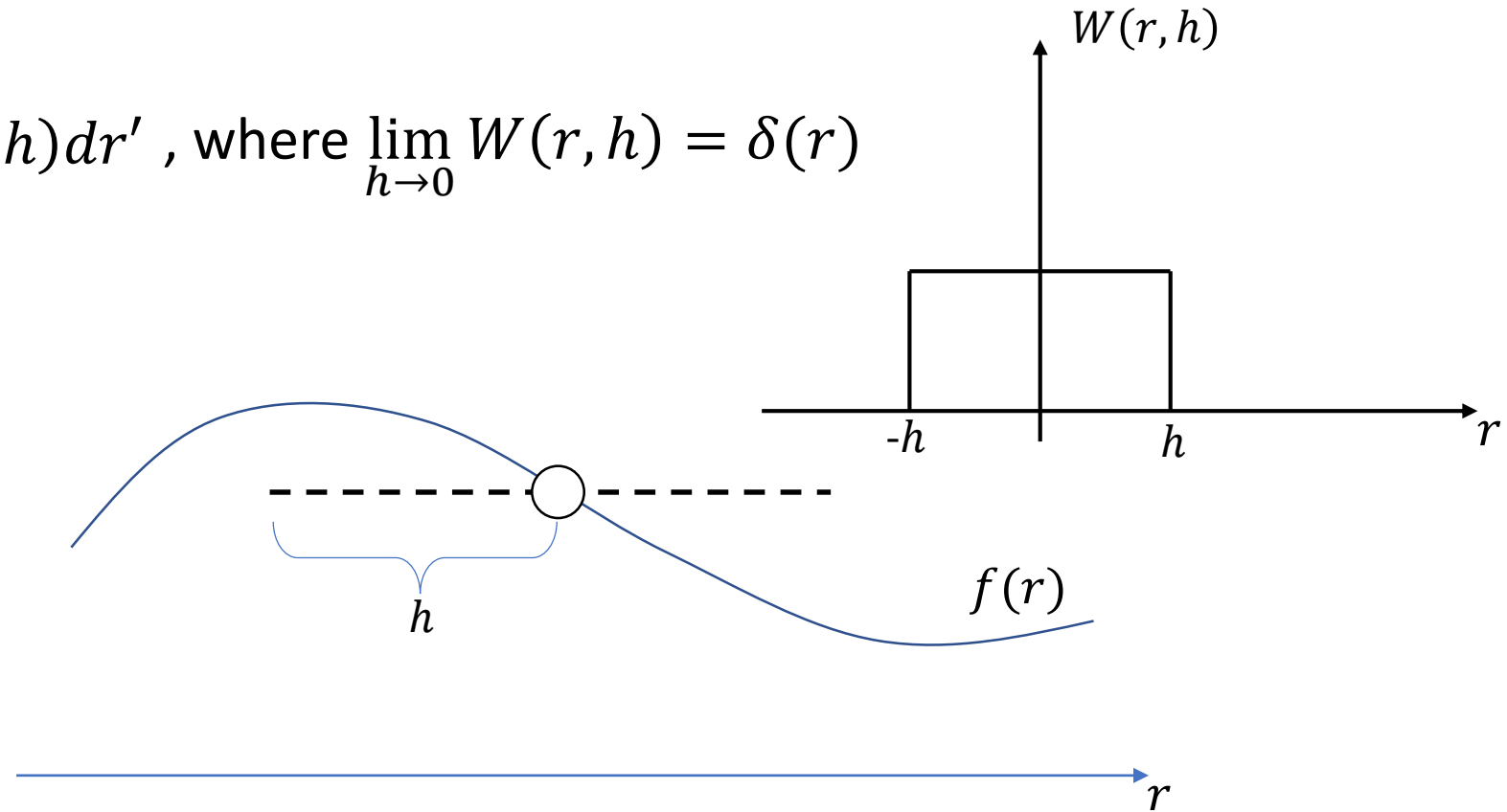
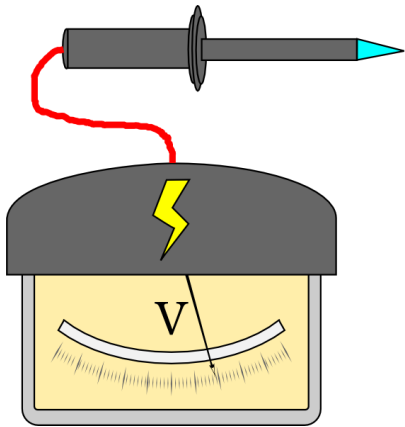


$W(r,h)$: kernel function

- Sum to unity: $\int W(r,h)dr = 1$
- Symmetric: $W(r,h) = W(-r,h)$
- Compact support: $W(r,h) = 0$ if $|r| > h$

A trivial kernel function

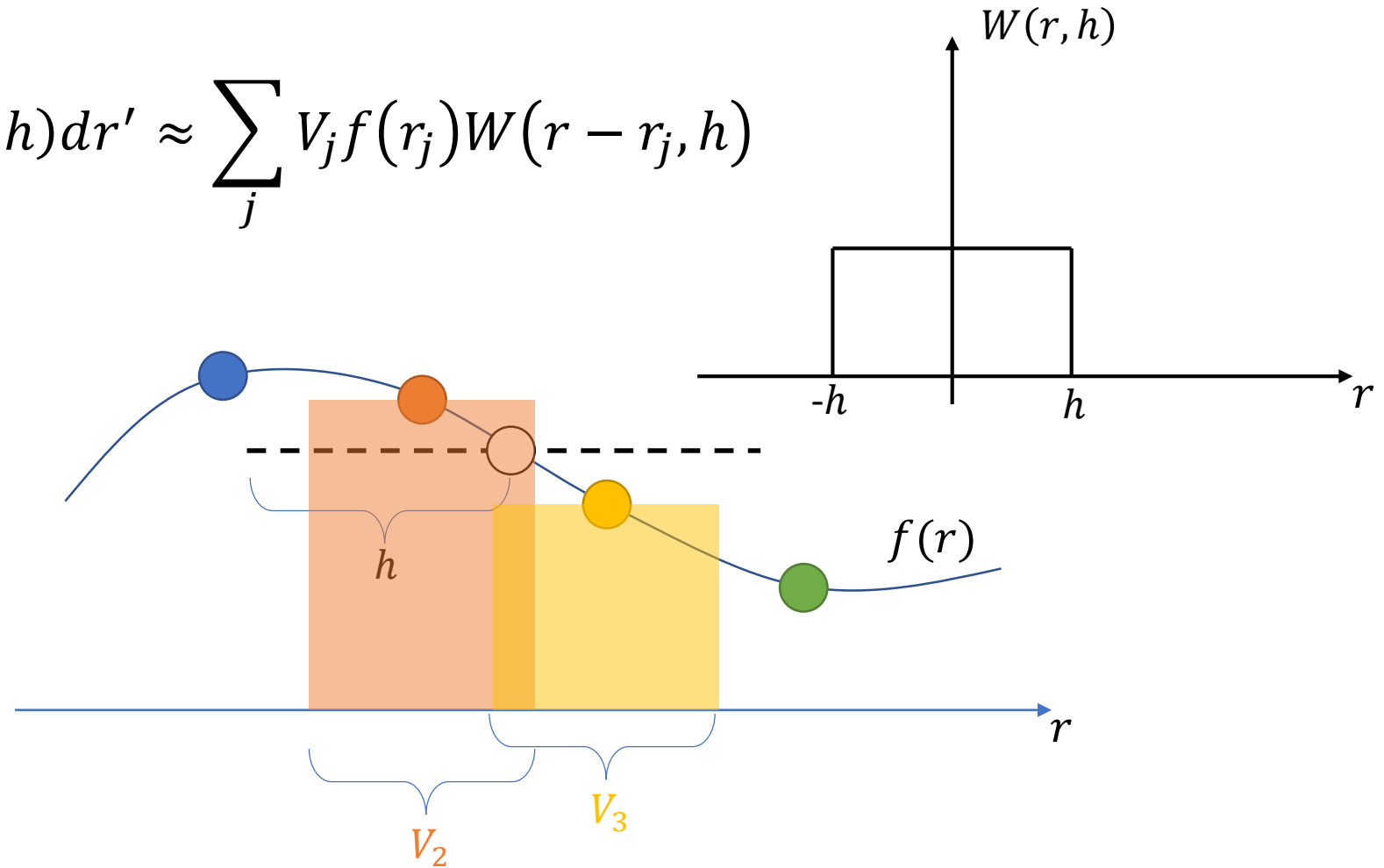
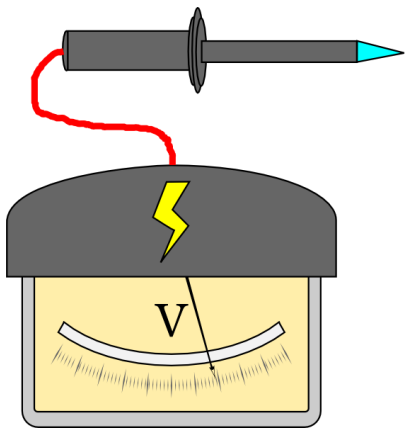
$$f(r) \approx \int f(r') W(r - r', h) dr' , \text{ where } \lim_{h \rightarrow 0} W(r, h) = \delta(r)$$



$$W(r, h) = \begin{cases} \frac{1}{2h}, & \text{if } |r| < h \\ 0, & \text{otherwise} \end{cases}$$

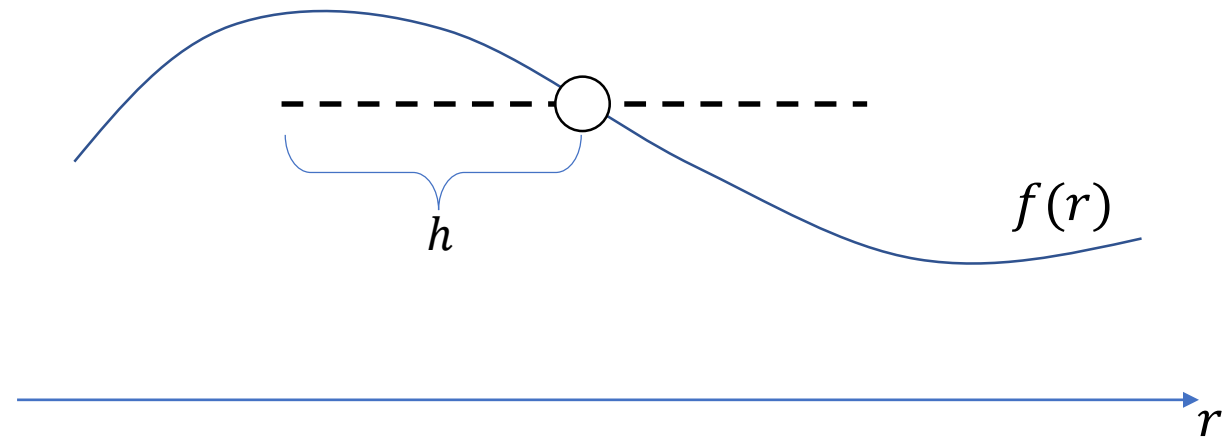
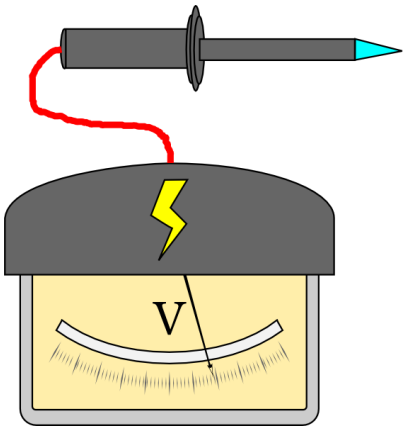
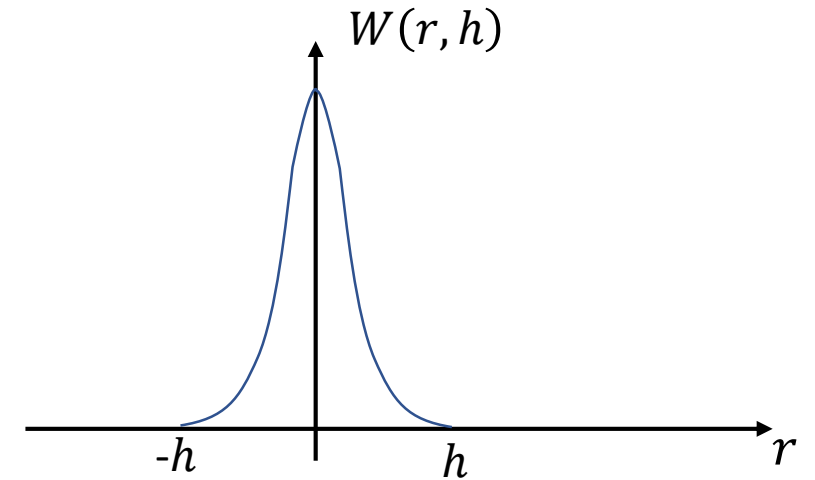
Finite probes: from integration to summation

$$f(r) \approx \int f(r')W(r-r',h)dr' \approx \sum_j V_j f(r_j)W(r-r_j,h)$$



A smoother kernel function

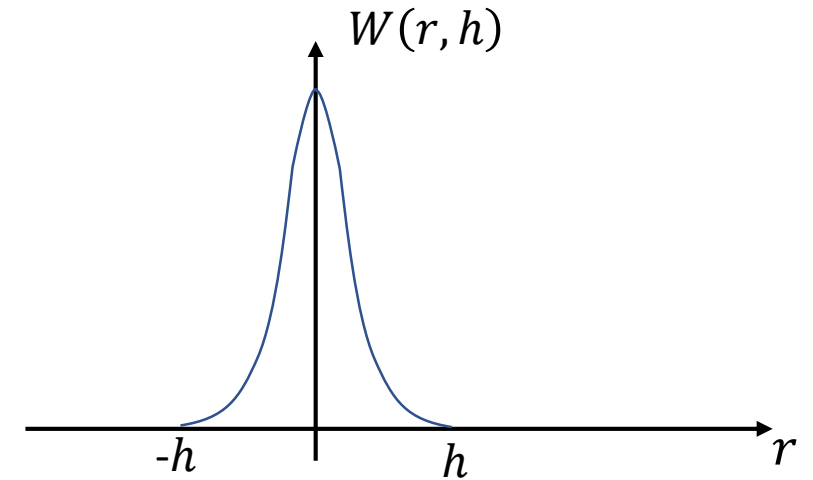
$$f(r) \approx \sum_j V_j f(r_j) W(r - r_j, h)$$



Smooth $W(r, h)$: “we trust the closer probes better”

A smoother kernel function

$$f(r) \approx \sum_j V_j f(r_j) W(r - r_j, h)$$



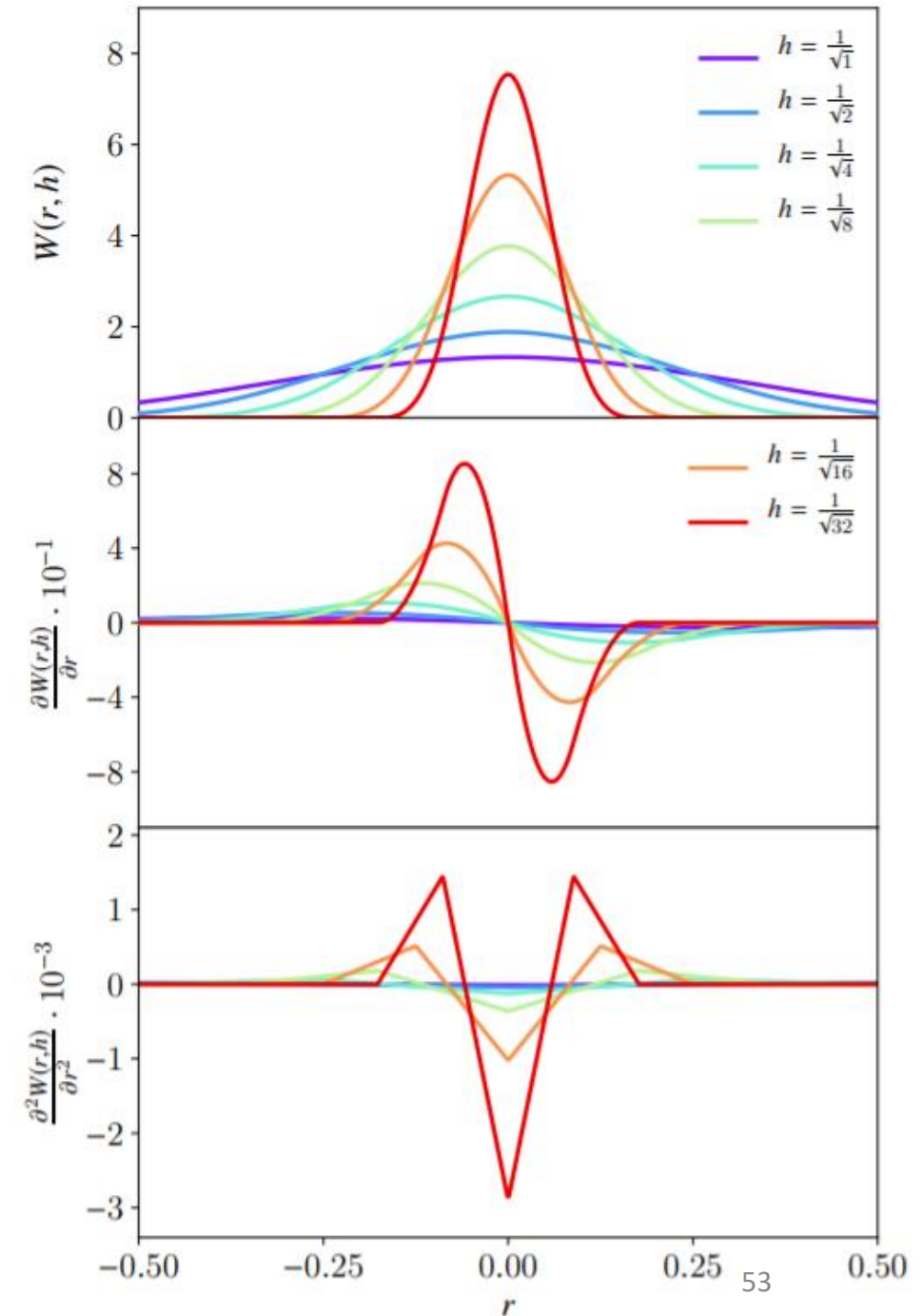
$$W(r, h) = \sigma_d \begin{cases} 6(q^3 - q^2) + 1 & \text{for } 0 \leq q \leq \frac{1}{2} \\ 2(1 - q)^3 & \text{for } \frac{1}{2} \leq q \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

$$\text{with } q = \frac{1}{h} \|r\|, \sigma_1 = \frac{4}{3h}, \sigma_2 = \frac{40}{7\pi h^2}, \sigma_3 = \frac{8}{\pi h^3}$$

A smoother kernel function

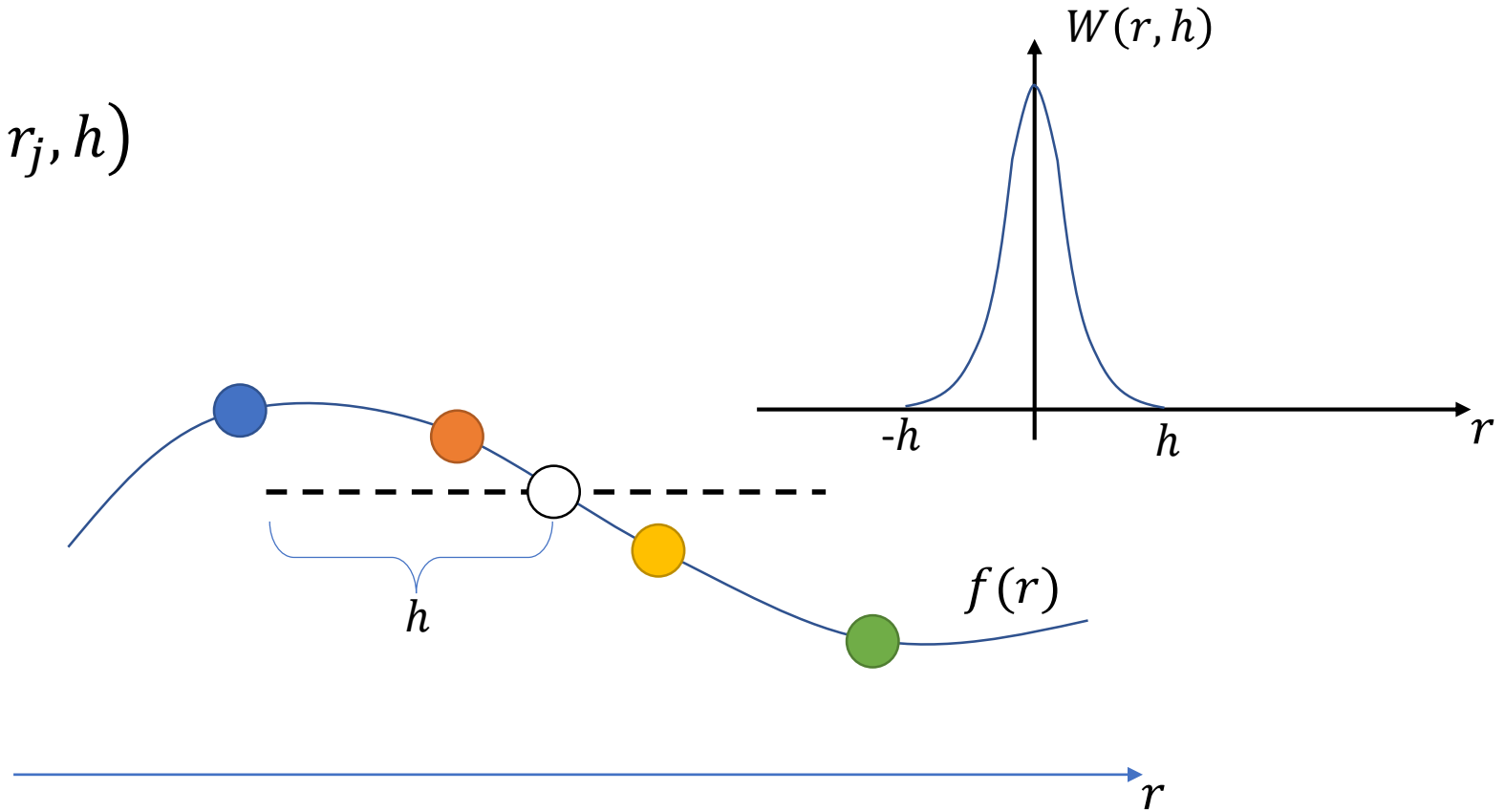
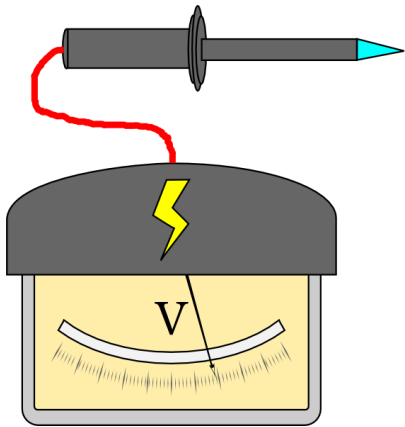
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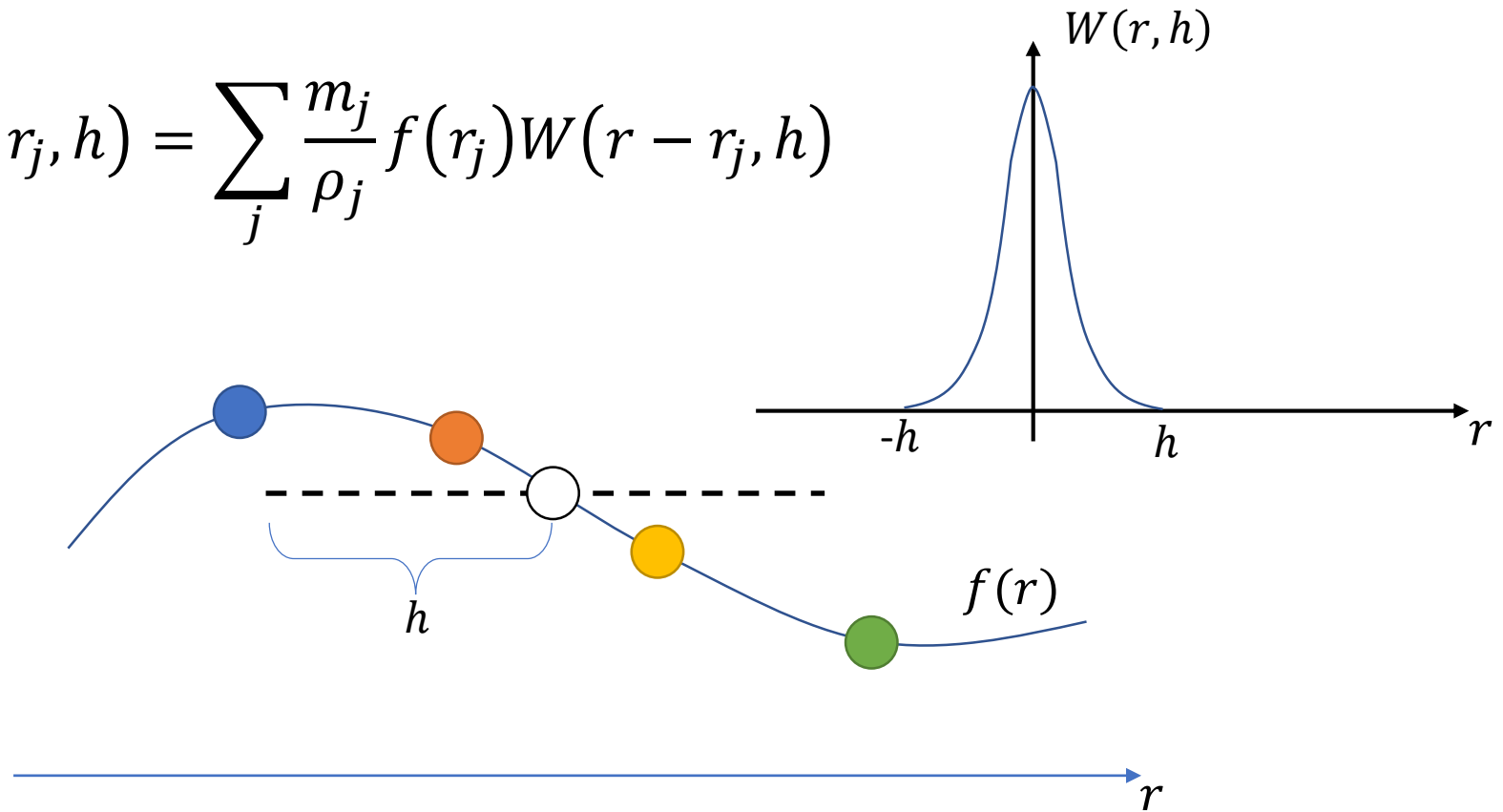
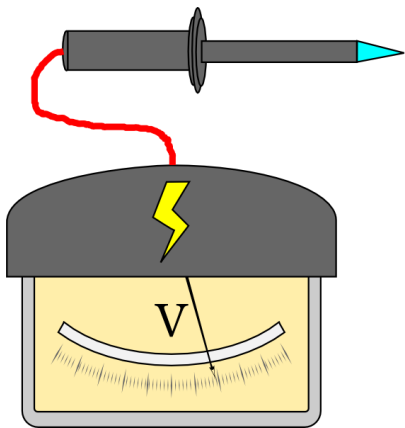
A Smoothed particle

$$f(r) \approx \sum_j V_j f(r_j) W(r - r_j, h)$$



Smoothed particle hydrodynamics (SPH)

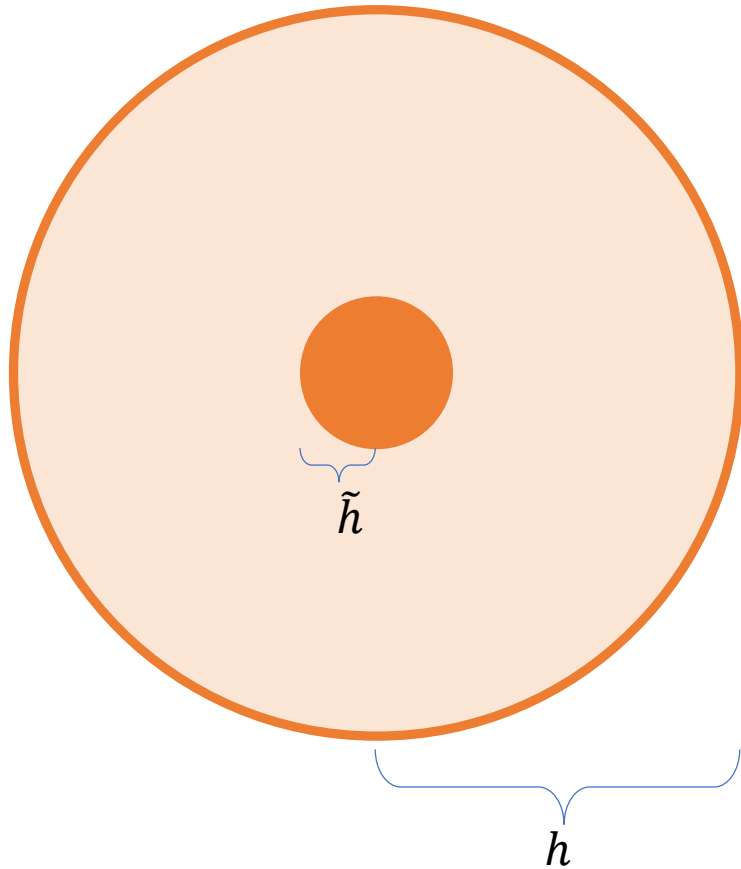
$$f(r) \approx \sum_j V_j f(r_j) W(r - r_j, h) = \sum_j \frac{m_j}{\rho_j} f(r_j) W(r - r_j, h)$$



Smoothed particle hydrodynamics: theory and application to non-spherical stars

[Gingold and Monaghan 1977][[Link](#)]

A smoothed particle in 2D



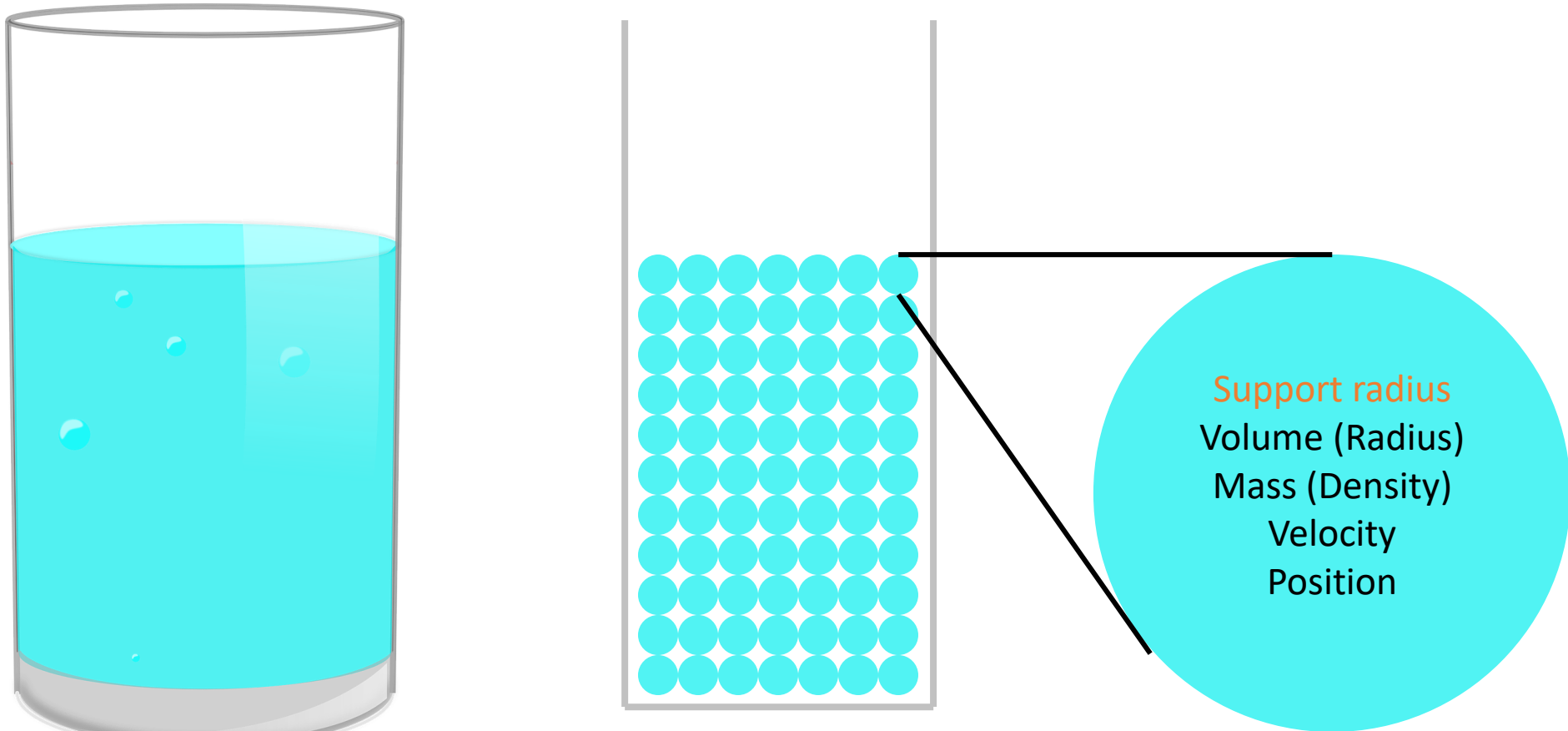
Intrinsic quantities:

- h : support radius
- \tilde{h} : particle radius $\rightarrow V$: particle volume

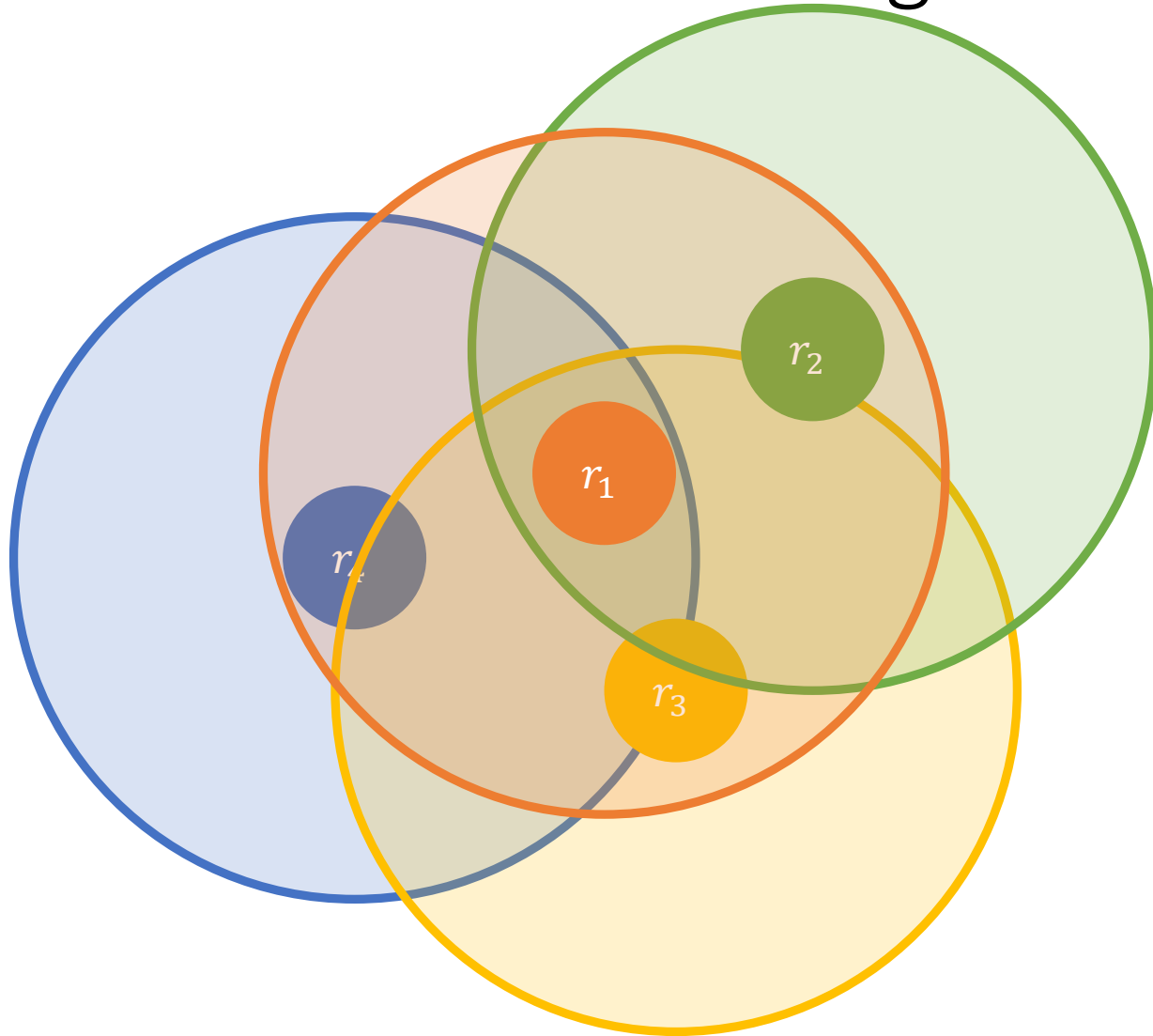
Time varying quantities:

- ρ : density
- v : velocity
- x : position

What do we store in a **smoothed** particle?



Evaluate 2D fields using the smoothed particles

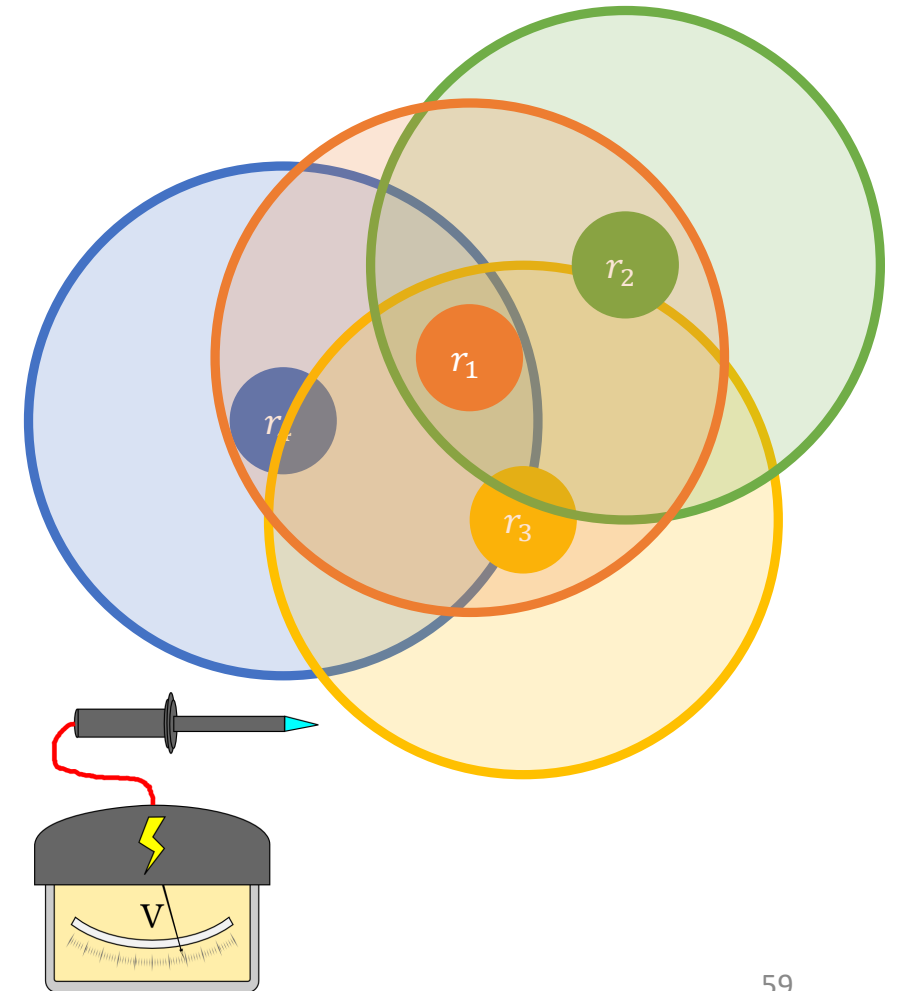


$$f(r) \approx \sum_j \frac{m_j}{\rho_j} f(r_j) W(r - r_j, h)$$

$$\begin{aligned} f(r_1) &\approx \frac{m_2}{\rho_2} f(r_2) W(r_1 - r_2, h) \\ &+ \frac{m_3}{\rho_3} f(r_3) W(r_1 - r_3, h) \\ &+ \frac{m_4}{\rho_4} f(r_4) W(r_1 - r_4, h) \end{aligned}$$

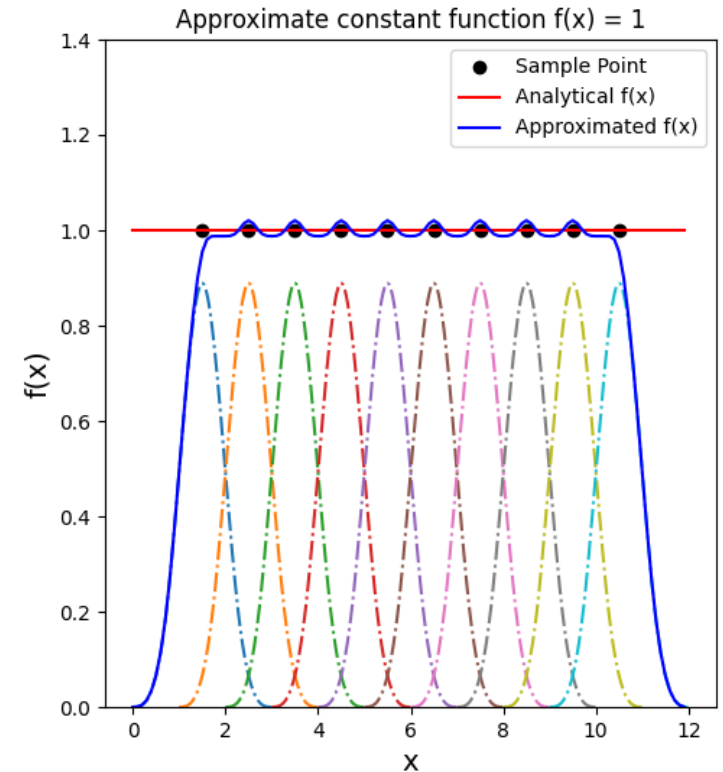
Smoothed particle hydrodynamics (SPH)

- SPH discretization:
 - $f(r) \approx \sum_j \frac{m_j}{\rho_j} f(r_j) W(r - r_j, h)$
- SPH spatial derivatives:
 - $\nabla f(r) \approx \sum_j \frac{m_j}{\rho_j} f(r_j) \nabla W(r - r_j, h)$
 - $\nabla \cdot \mathbf{F}(r) \approx \sum_j \frac{m_j}{\rho_j} \mathbf{F}(r_j) \cdot \nabla W(r - r_j, h)$
 - $\nabla \times \mathbf{F}(r) \approx -\sum_j \frac{m_j}{\rho_j} f(r_j) \times \nabla W(r - r_j, h)$
 - $\nabla^2 f(r) \approx \sum_j \frac{m_j}{\rho_j} f(r_j) \nabla^2 W(r - r_j, h)$



Improving approximations for spatial derivatives

- $f(r) \approx \sum_j \frac{m_j}{\rho_j} f(r_j) W(r - r_j, h)$
- $\nabla f(r) \approx \sum_j \frac{m_j}{\rho_j} f(r_j) \nabla W(r - r_j, h)$
- Let $f(r) \equiv 1$, we have:
 - $1 \approx \sum_j \frac{m_j}{\rho_j} W(r - r_j, h)$
 - $0 \approx \sum_j \frac{m_j}{\rho_j} \nabla W(r - r_j, h)$




Improving approximations for spatial derivatives

- Since $f(r) \equiv f(r) * 1$, we have:
 - $\nabla f(r) = \nabla f(r) * 1 + f(r) * \nabla 1$
 - Or equivalently: $\nabla f(r) = \nabla f(r) - f(r) * \nabla 1$

Improving approximations for spatial derivatives

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$$\nabla f(r) \approx \sum_j \frac{m_j}{\rho_j} f(r_j) \nabla W(r - r_j, h)$$

Improving approximations for spatial derivatives

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 - $\nabla f(r) = \nabla f(r) * 1 + f(r) * \nabla 1$
 - Or equivalently: $\nabla f(r) = \nabla f(r) - f(r) * \nabla 1$
- $\nabla f(r) \approx \sum_j \frac{m_j}{\rho_j} f(r_j) \nabla W(r - r_j, h) - f(r) \sum_j \frac{m_j}{\rho_j} \nabla W(r - r_j, h)$

Improving approximations for spatial derivatives

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 - $\nabla f(r) = \nabla f(r) * 1 + f(r) * \nabla 1$
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- $\nabla f(r) \approx \sum_j m_j \frac{f(r_j) - f(r)}{\rho_j} \nabla W(r - r_j, h)$

The anti-symmetric form

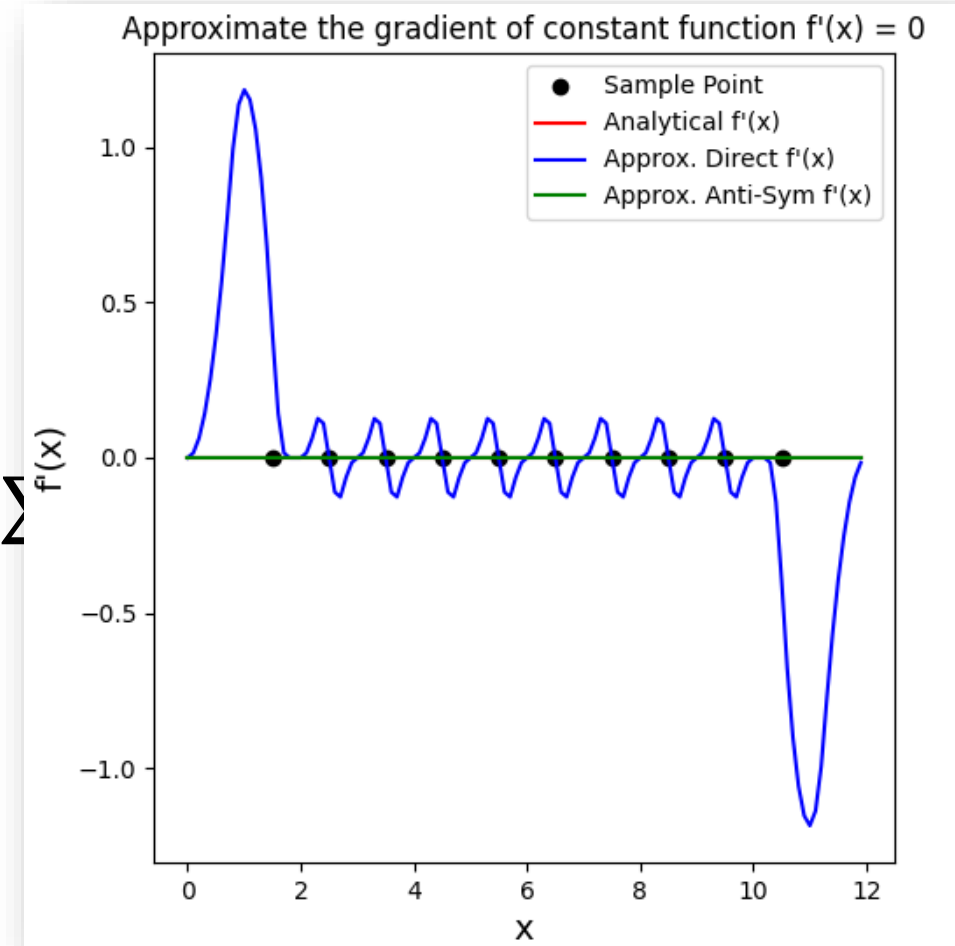
Improving approximations for spatial derivatives

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- $\nabla f(r) \approx \sum_j m_j \frac{f(r_j) - f(r)}{\rho_j} \nabla W(r - r_j, h)$

The anti-symmetric form



Improving approximations for spatial derivatives

- A more general case:

- $\nabla(f(r)\rho^n) = \nabla f(r) * \rho^n + f(r) * n\rho^{n-1}\nabla\rho$


- $\Rightarrow \nabla f(r) = \frac{1}{\rho^n} (\nabla(f(r) * \rho^n) - f(r) * n * \rho^{n-1}\nabla\rho)$

Improving approximations for spatial derivatives

- A more general case:

- $\nabla(f(r)\rho^n) = \nabla f(r) * \rho^n + f(r) * n\rho^{n-1}\nabla\rho$

- $\Rightarrow \nabla f(r) = \frac{1}{\rho^n} (\nabla(f(r) * \rho^n) - f(r) * n * \rho^{n-1}\nabla\rho)$


$$\nabla f(r) \approx \sum_j \frac{m_j}{\rho_j} f(r_j) \nabla W(r - r_j, h)$$

Improving approximations for spatial derivatives

- A more general case:

- $\nabla(f(r)\rho^n) = \nabla f(r) * \rho^n + f(r) * n\rho^{n-1}\nabla\rho$

- $\Rightarrow \nabla f(r) = \frac{1}{\rho^n} (\nabla(f(r) * \rho^n) - f(r) * n * \rho^{n-1}\nabla\rho)$

- $\nabla f(r) \approx \frac{1}{\rho^n} \left(\sum_j \frac{m_j}{\rho_j} f(r_j) \rho_j^n \nabla W(r - r_j, h) - n\rho^{n-1} f(r) \sum_j \frac{m_j}{\rho_j} \rho_j \nabla W(r - r_j, h) \right)$

Improving approximations for spatial derivatives

- A more general case:

- $\nabla(f(r)\rho^n) = \nabla f(r) * \rho^n + f(r) * n\rho^{n-1}\nabla\rho$

- $\Rightarrow \nabla f(r) = \frac{1}{\rho^n} (\nabla(f(r) * \rho^n) - f(r) * n * \rho^{n-1}\nabla\rho)$

- $\nabla f(r) \approx \frac{1}{\rho^n} \left(\sum_j \frac{m_j}{\rho_j} f(r_j) \rho_j^n \nabla W(r - r_j, h) - n\rho^{n-1} f(r) \sum_j \frac{m_j}{\rho_j} \rho_j \nabla W(r - r_j, h) \right)$

- $\nabla f(r) \approx \sum_j m_j \left(\frac{f(r_j) \rho_j^{n-1}}{\rho^n} - \frac{n f(r)}{\rho} \right) \nabla W(r - r_j, h)$

Improving approximations for spatial derivatives

- $\nabla f(r) \approx \sum_j m_j \left(\frac{f(r_j) \rho_j^{n-1}}{\rho^n} - \frac{nf(r)}{\rho} \right) \nabla W(r - r_j, h)$
- When $n = -1$
 - Type equation here.
 - $\nabla f(r) \approx \rho \sum_j m_j \left(\frac{f(r_j)}{\rho_j^2} + \frac{f(r)}{\rho^2} \right) \nabla W(r - r_j, h)$

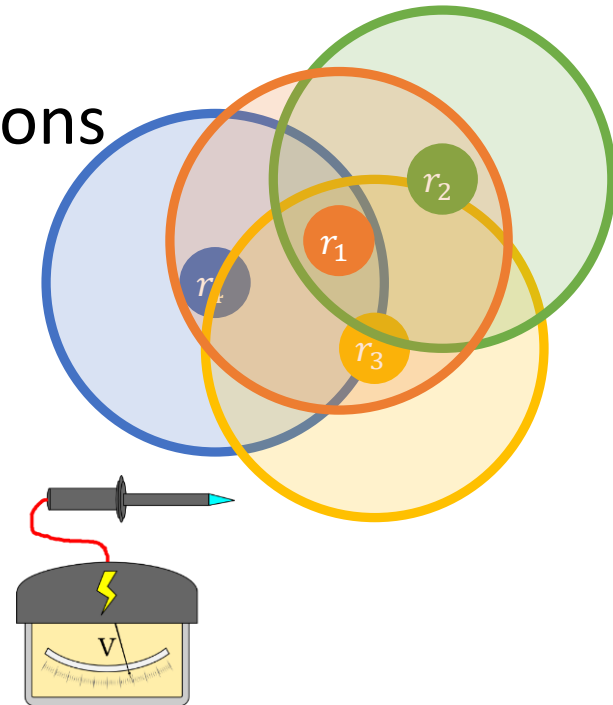
The symmetric form

Smoothed particle hydrodynamics (SPH)

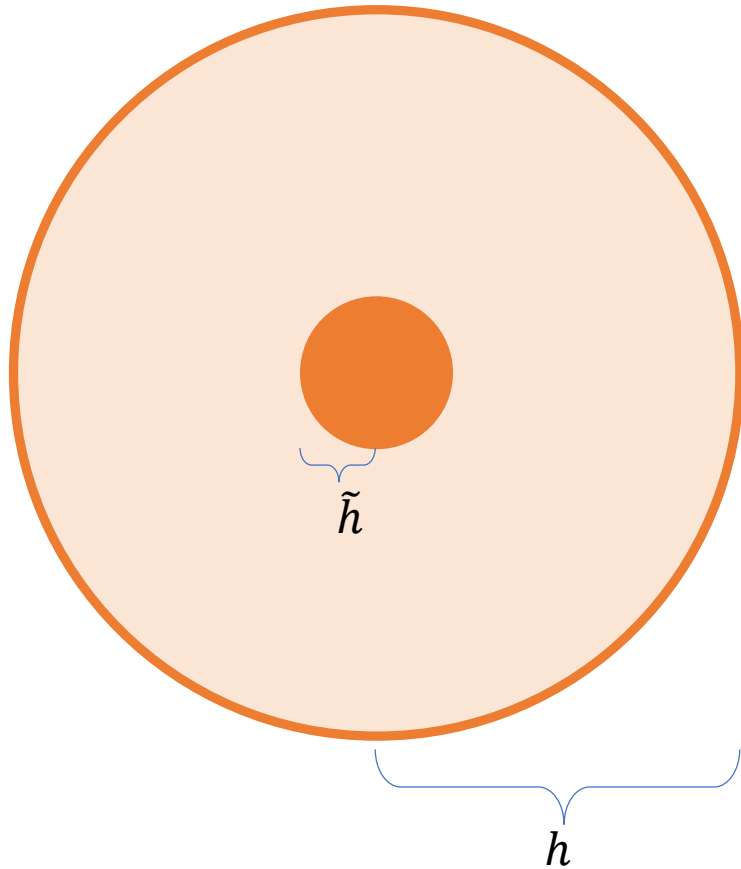
$$f(r) \approx \sum_j \frac{m_j}{\rho_j} f(r_j) W(r - r_j, h)$$

- Approximate a function $f(r)$ using finite probes $f(r_j)$
- The degree of freedom (r) goes inside the kernel functions

- Anti-sym: $\nabla f(r) \approx \sum_j m_j \frac{f(r_j) - f(r)}{\rho_j} \nabla W(r - r_j, h)$
- Sym: $\nabla f(r) \approx \rho \sum_j m_j \left(\frac{f(r_j)}{\rho_j^2} + \frac{f(r)}{\rho^2} \right) \nabla W(r - r_j, h)$



Quiz: which one is the **smoothed particle** in SPH?



The bigger circle?
or
The smaller circle?

Intrinsic quantities:

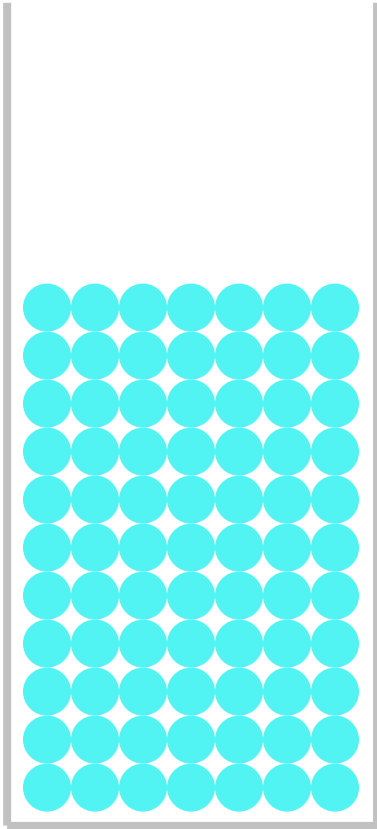
- h : support radius
- \tilde{h} : particle radius $\rightarrow V$: particle volume

Time varying quantities:

- ρ : density
- v : velocity
- x : position

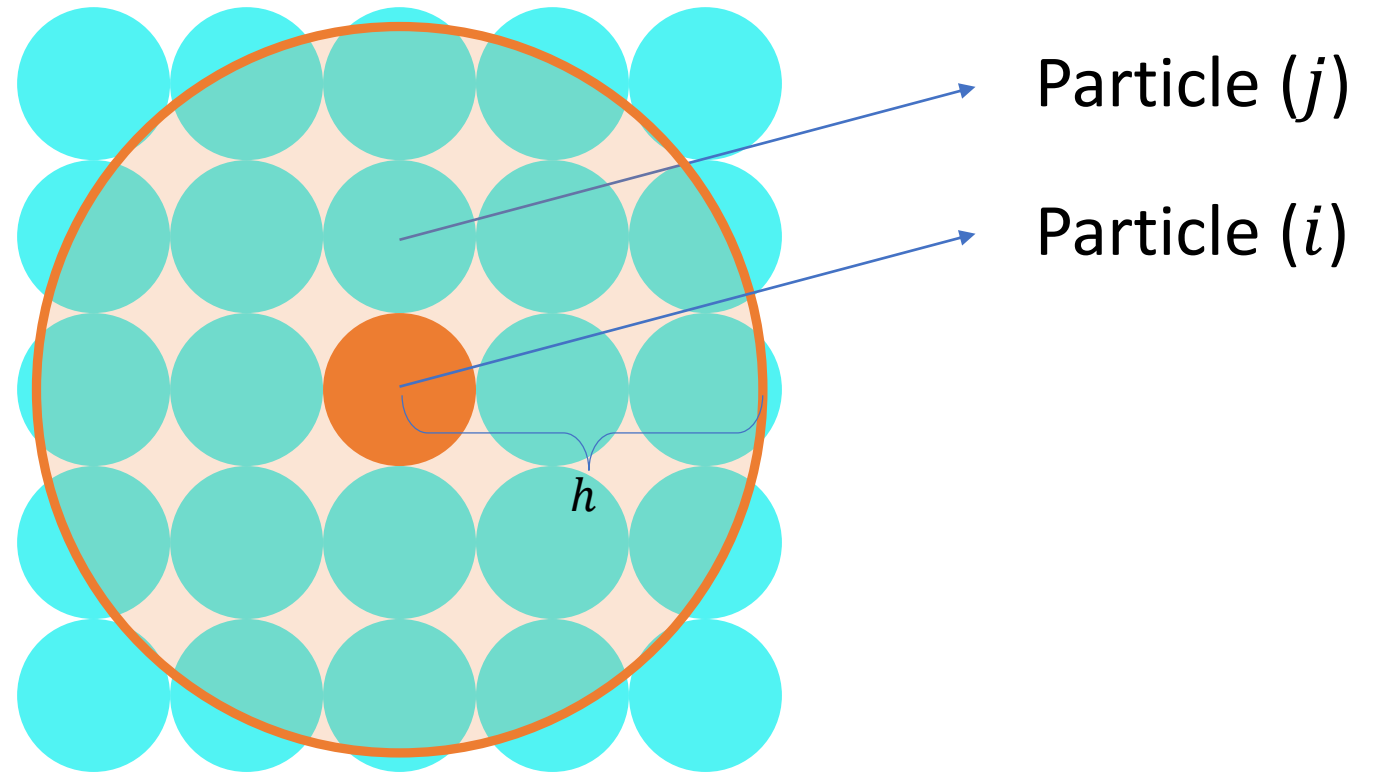
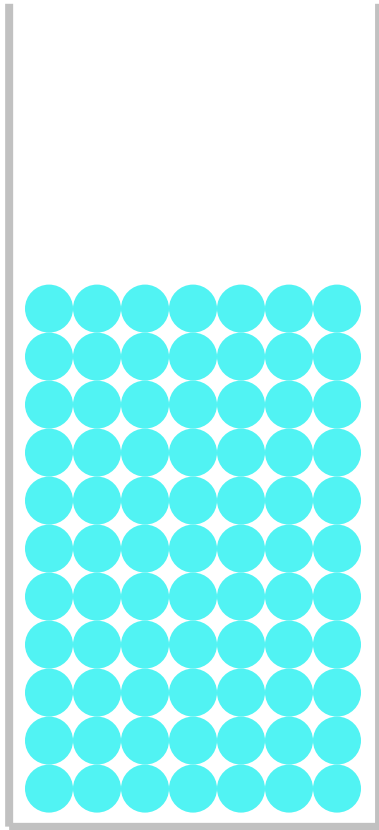
Implementation details (WCSPH)

Simulation pipeline



- for i in particles:
 - Search for neighbors j
- for i in particles:
 - Sample the velocity/density field using SPH
 - Compute force/acceleration using Navier-Stokes equation
- for i in particles:
 - Update velocity using acceleration
 - Update position using velocity

Find a particle of interest (i) and its neighbors (j) within its support radius h



Compute the acceleration for particle (i)

- for i in particles:

- Step 1: Evaluate density

- $\rho_i = \sum_j \frac{m_j}{\rho_j} \rho_j W(r_i - r_j, h) = \sum_j m_j W_{ij}$

- Step 2: Evaluate viscosity

- $\nu \nabla^2 v_i = \nu \sum_j m_j \frac{v_j - v_i}{\rho_j} \nabla^2 W_{ij}$

- Step 3: Evaluate pressure gradient

- $-\frac{1}{\rho_i} \nabla p_i = -\frac{\rho_i}{\rho_i} \sum_j m_j \left(\frac{p_j}{\rho_j^2} + \frac{p_i}{\rho_i^2} \right) \nabla W_{ij}, \text{ where } p = k(\rho_j - \rho_0)$

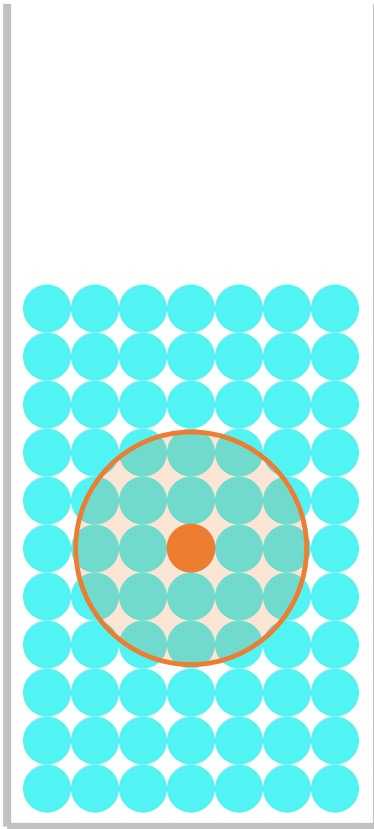
- $\frac{dv_i}{dt} = g - \frac{1}{\rho_i} \nabla p_i + \nu \nabla^2 v_i$

Time integration (Symplectic Euler)

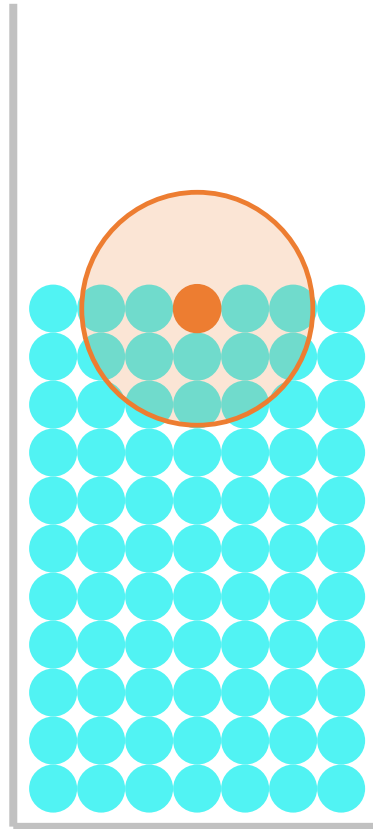
- for i in particles:
 - $v_i = v_i + \Delta t * \frac{dv_i}{dt}$
 - $x_i = x_i + \Delta t * v_i$

Boundary conditions

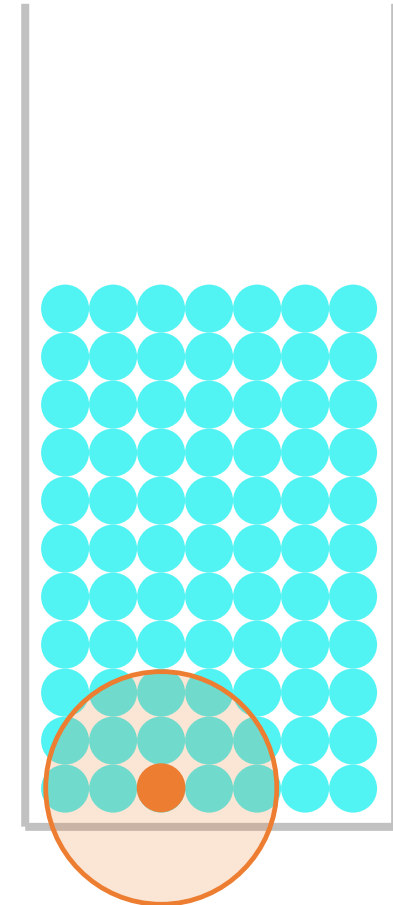
fluid



free surface

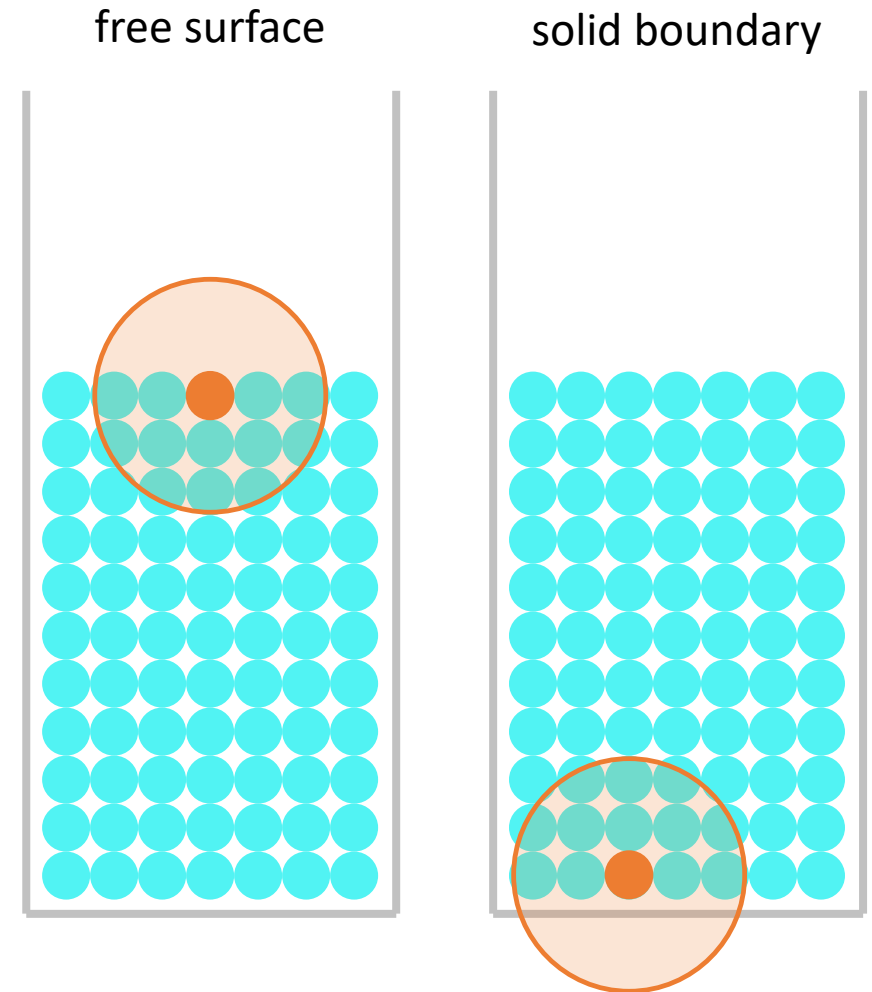
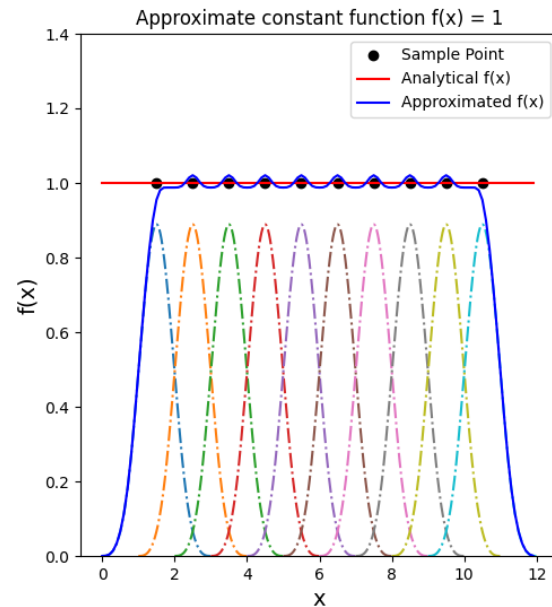


solid boundary



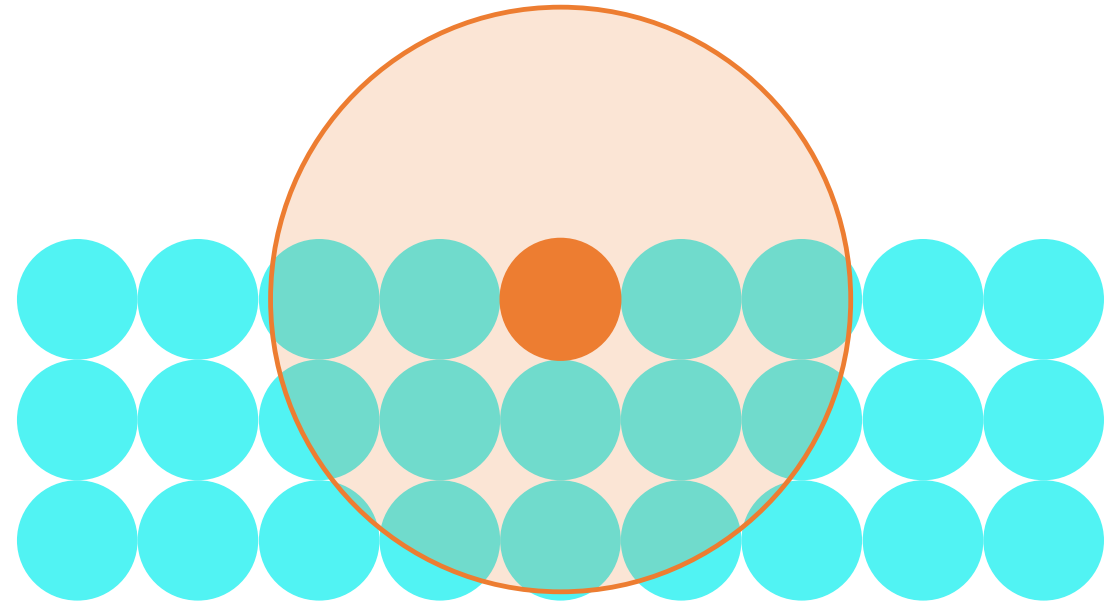
Problems of boundaries

- Not enough samples within the supporting radius
 - Density: ↓
 - $\rho_i = \sum_j m_j W_{ij}$



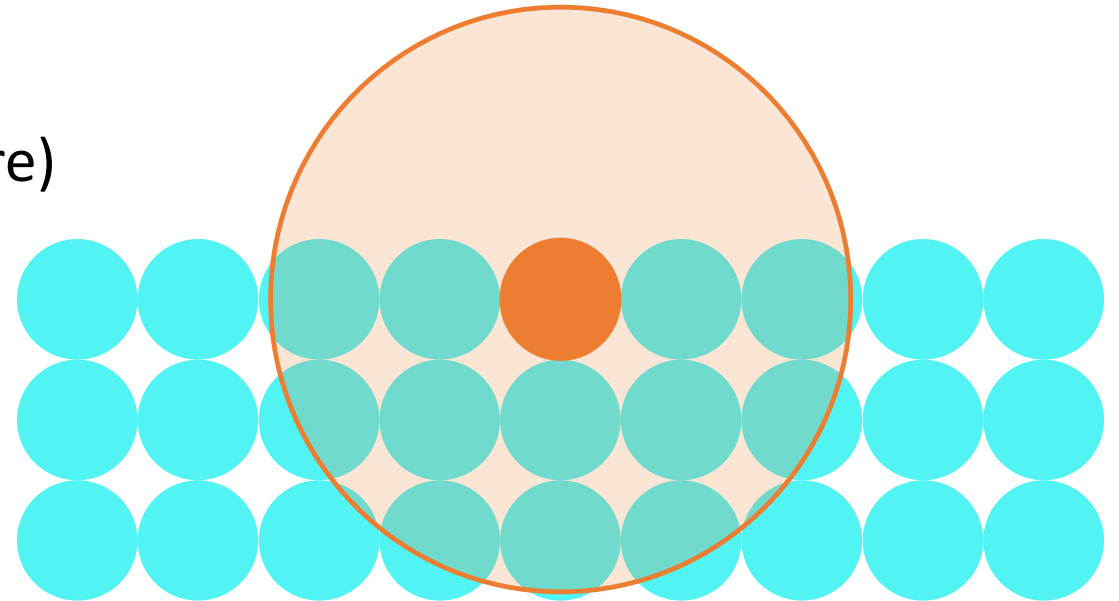
Free surface

- Problem:
 - Density \downarrow Pressure \downarrow
 - Generate outward pressure



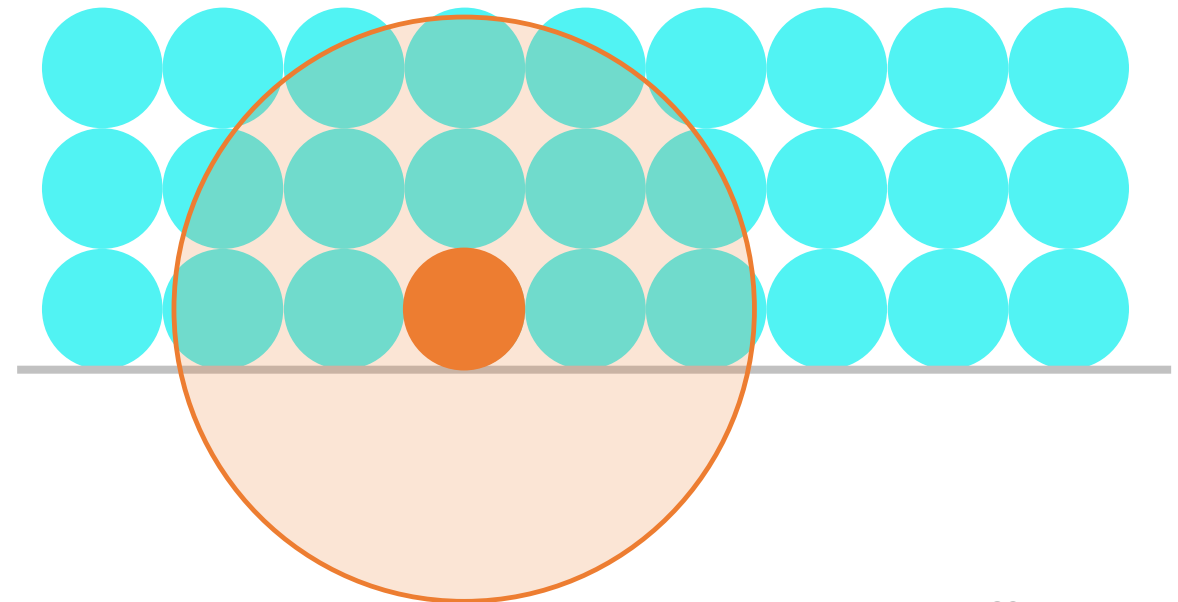
Free surface

- Problem:
 - Density \downarrow Pressure \downarrow
 - Generate outward pressure
- Solution:
 - Clamp the negative pressure (everywhere)
 - $p = \max(0, k(\rho - \rho_0))$



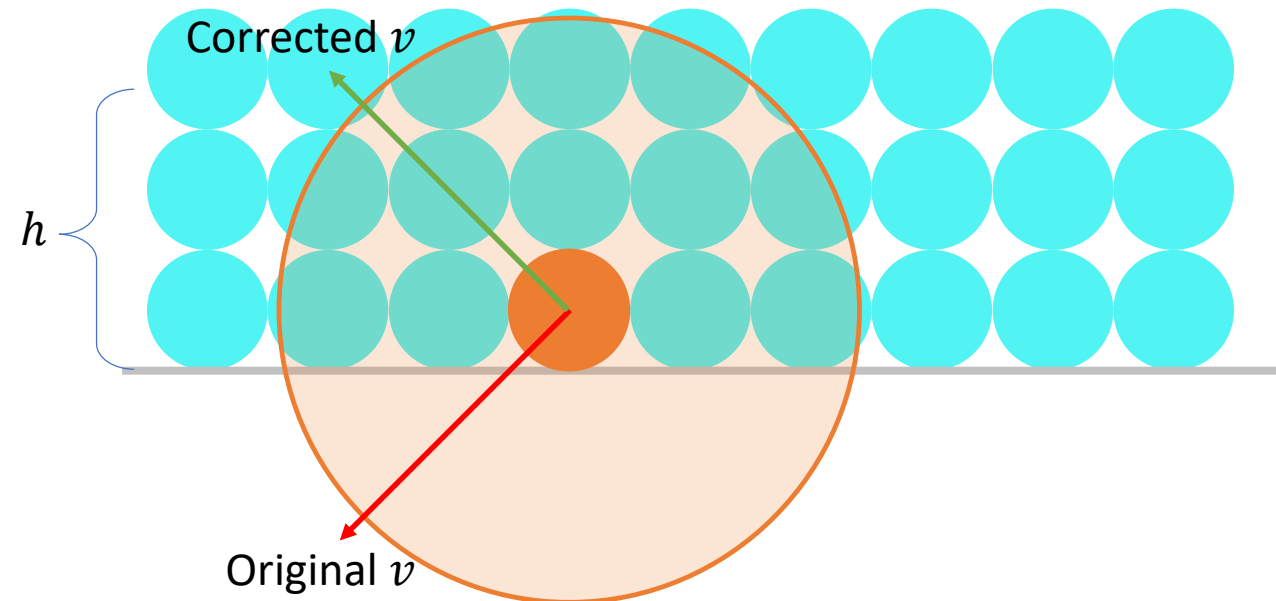
Solid boundary

- Problem:
 - Density \downarrow Pressure \downarrow
 - Fluid leakage (due to outbound velocity)



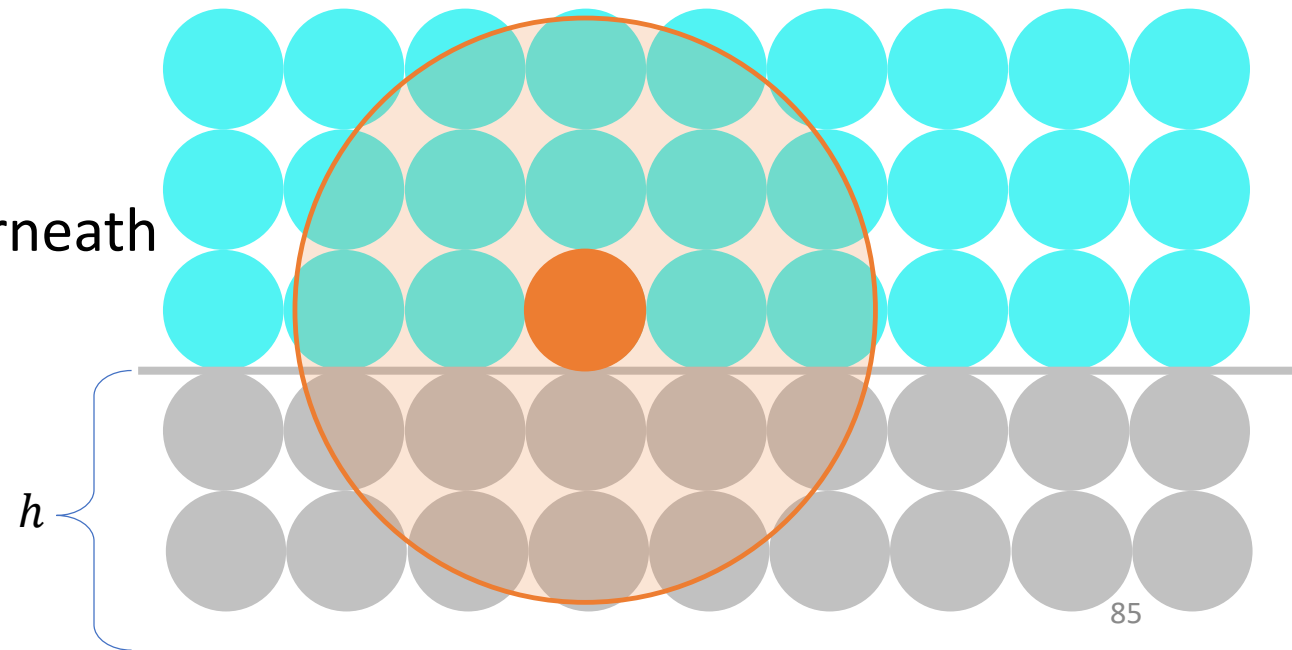
Solid boundary

- Problem:
 - Density \downarrow Pressure \downarrow
 - Fluid leakage (due to outbound velocity)
- $p = \max(0, k(\rho - \rho_0))$
- Solution 1 for leakage:
 - Reflect the outbound velocity when close to boundary



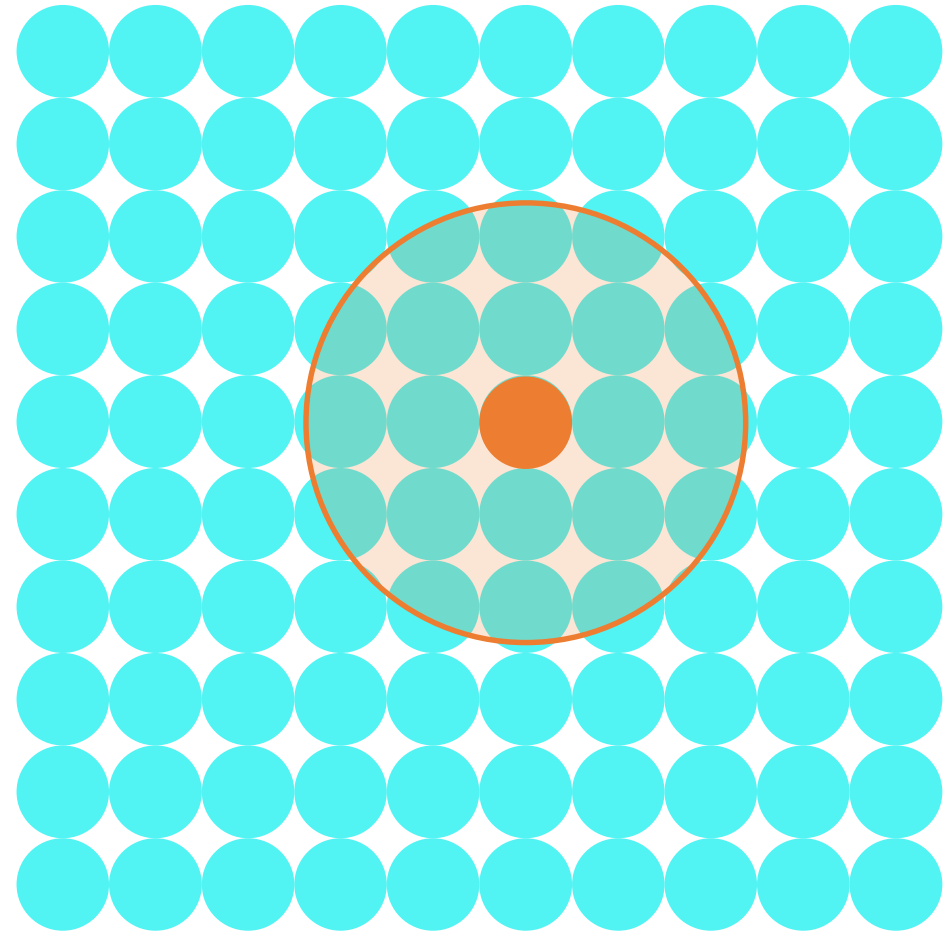
Solid boundary

- Problem:
 - Density \downarrow Pressure \downarrow
 - Fluid leakage (due to outbound velocity)
- $p = \max(0, k(\rho - \rho_0))$
- Solution 2 for leakage:
 - Pad a layer of solid particles underneath the boundaries
 - $\rho_{solid} = \rho_0$
 - $v_{solid} = 0$



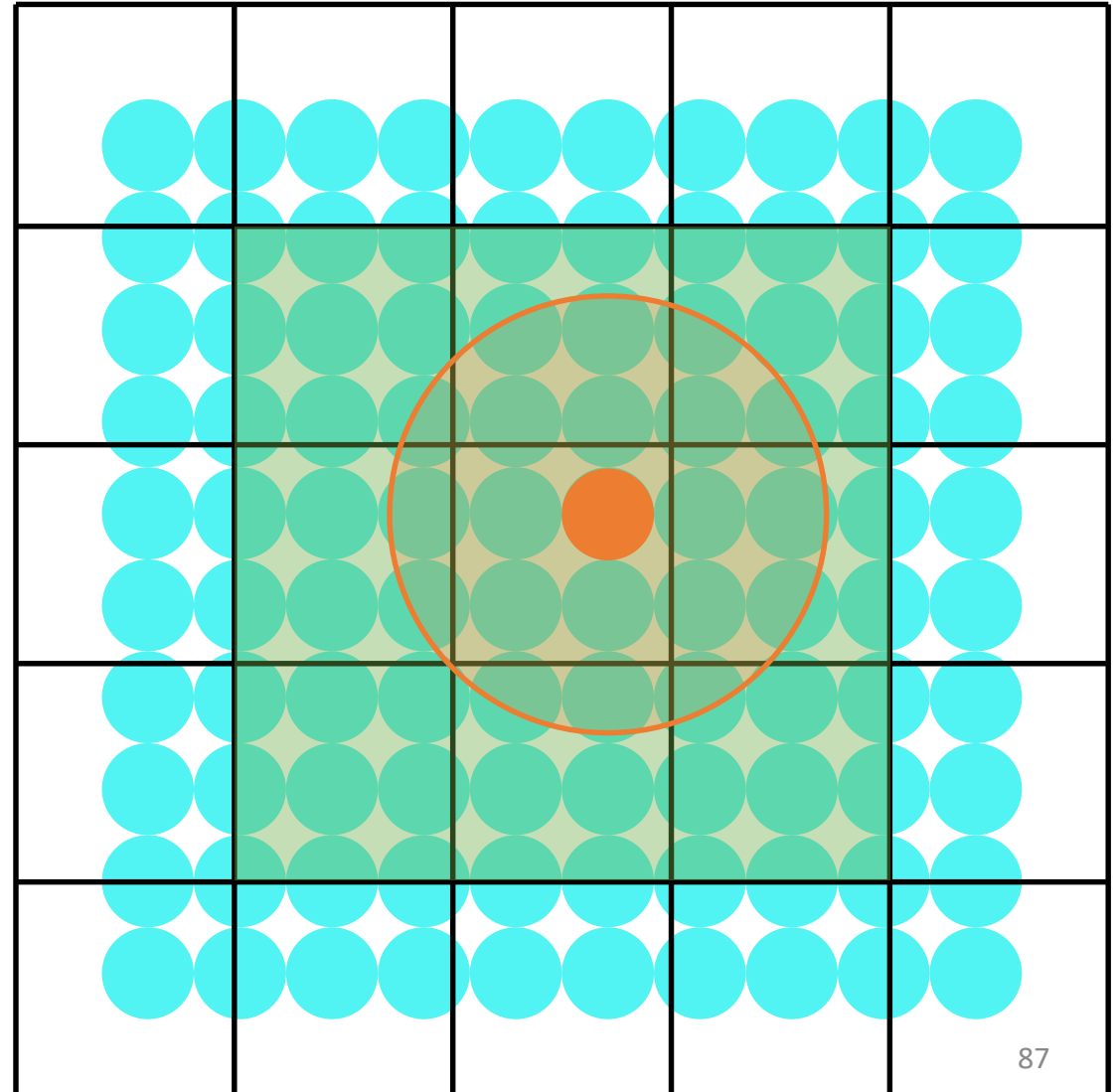
Neighbor search

- Naïve search methods takes $\mathcal{O}(n^2)$ time



Neighbor search

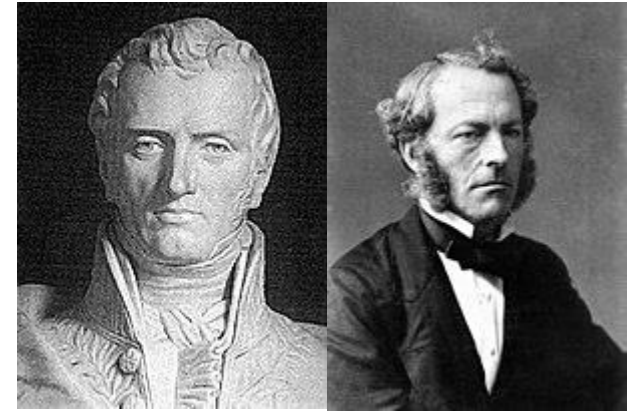
- Naïve search methods takes $\mathcal{O}(n^2)$ time
- A background grid can help
 - Common grid size = h (the support radius in SPH)
 - Each neighbor search takes 9 grids in 2D and 27 grids in 3D



Remark

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- Incompressible fluid dynamics
 - Incompressible Navier–Stokes equations
- Time discretization
 - Operator splitting
 - Integration with the weakly compressible assumption
- Spatial discretization
 - Smoothed particle hydrodynamics (SPH)
- Implementation details (WCSPH)
 - Simulation Pipeline
 - Boundary conditions
 - Neighbor search



$$\rho \frac{Dv}{Dt} = \rho g - \nabla p + \mu \nabla^2 v$$

$$\nabla \cdot v = 0$$

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$$\rho \frac{Dv}{Dt} = \rho g + \mu \nabla^2 v$$

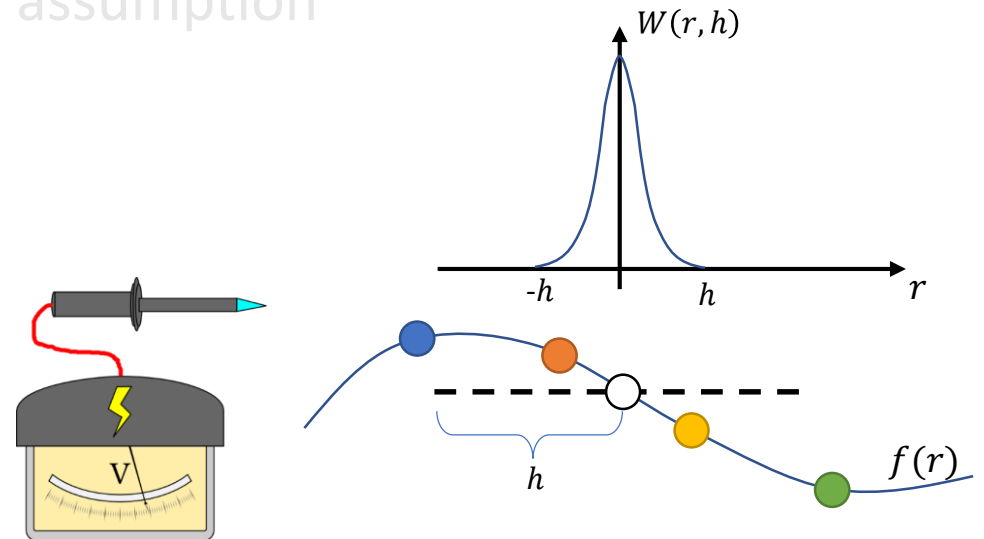
$$\rho \frac{Dv}{Dt} = -\nabla p$$

$$\nabla \cdot v = 0$$

Remark

- Incompressible fluid dynamics
 - Incompressible Navier–Stokes equations
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 - Neighbor search

$$f(r) \approx \sum_j \frac{m_j}{\rho_j} f(r_j) W(r - r_j, h)$$

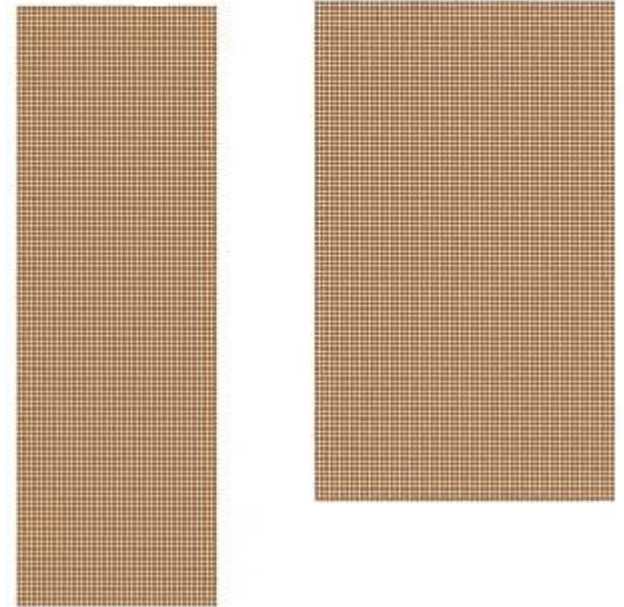


Remark

- Incompressible fluid dynamics
 - Incompressible Navier–Stokes equations
- Time discretization
 - Operator splitting
 - Integration with the weakly compressible assumption
- Spatial discretization
 - Smoothed particle hydrodynamics (SPH)
- **Implementation details (WCSPH)**
 - Simulation Pipeline
 - Boundary conditions
 - Neighbor search

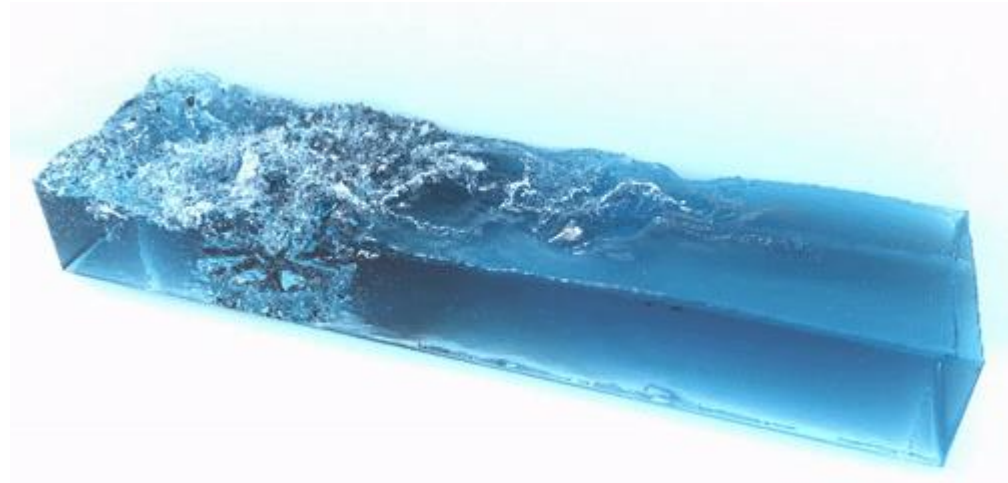
Remark

- Incompressible fluid dynamics
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- Implementation details (WCSPH)
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Further readings

- *Smoothed Particle Hydrodynamics* [Monaghan 2005][[Link](#)]
- *Smoothed particle hydrodynamics and magnetohydrodynamics* [Price 2012][[Link](#)][[Preprint](#)]
- *Smoothed Particle Hydrodynamics Techniques for the Physics Based Simulation of Fluids and Solids* [Eurographics Tutorial, 2019][[Link](#)][[Code](#)]



Next week



Homework

Homework Today

- Download the repo (taichi_sph):
 - https://github.com/taichiCourse01/taichi_sph
- Try:
 - Designing your own scene
 - Implementing a particle based boundary handling
 - Changing the dense grid in the codebase to a sparse grid [[03讲](#)]

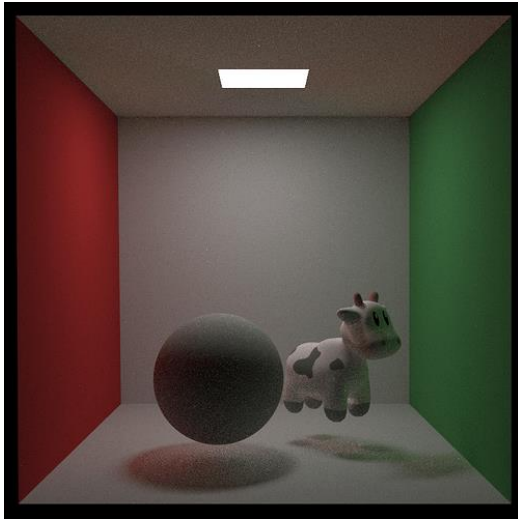
Candidate projects for your final

- Candidate topics:
 - Try advanced pressure (Poisson) solvers: IISPH/PCISPH/DFSPH [[Chapter 4](#)]
 - Put a statue/kinematically-controlled fan into your pond (One way coupling)
 - Throw a rubber duck into your pond (Two way coupling)
 - Render your fluid using your own path tracer (You may want a marching cube/square to construct the water surface)
- Both 2D and 3D projects are great!
 - As long as your pictures look great 😊

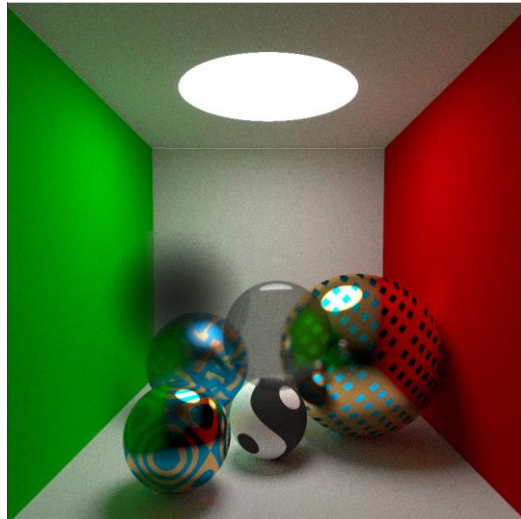
Final project

- 死线：2022年1月3日
- 要求：
 - 使用[作业模板](#)
 - 需要有设计文档，如果有参照代码也需要标明
- 题材：
 - 任何使用Taichi完成的内容（图形学更佳）
 - 可以参考每节图形课后给出的大作业选题灵感 [参考第[07](#),[09](#),10讲]
 - 鼓励实现任意图形学论文/图形学课程内容
- 形式：
 - 使用 GitHub/Gitee提交并邀请tgc01@taichi.graphics加入你的代码仓
 - 允许三人以下合作，记得管理多人合作的git commits
- 奖励：
 - 太极图形课第一季结业证书一份+神秘Taichi礼物一份

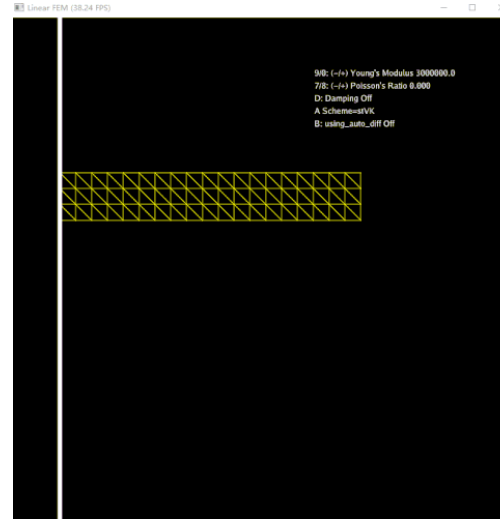
Excellent homework assignments



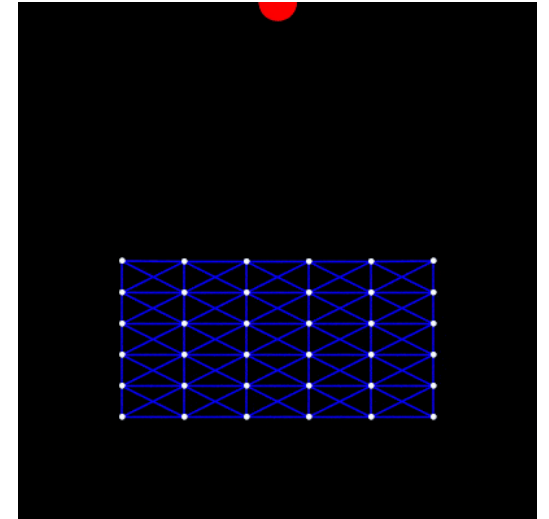
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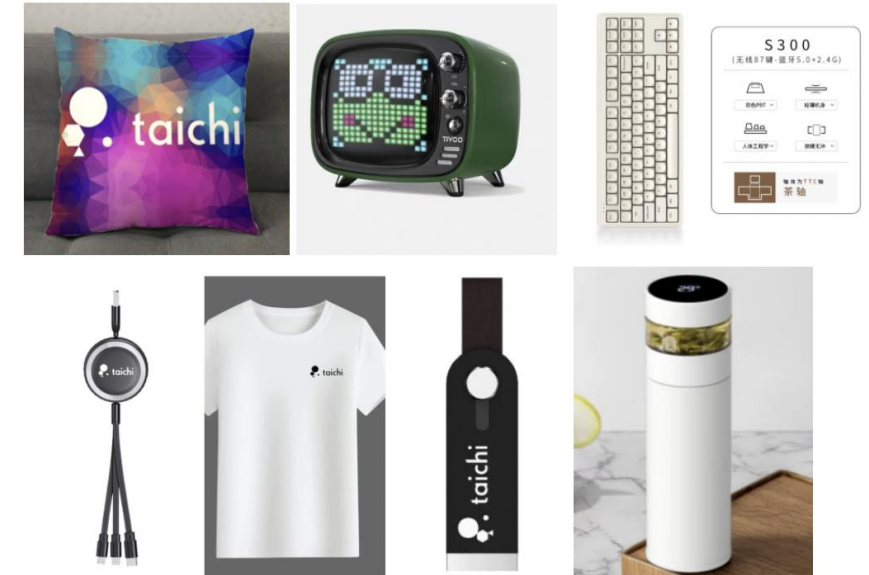
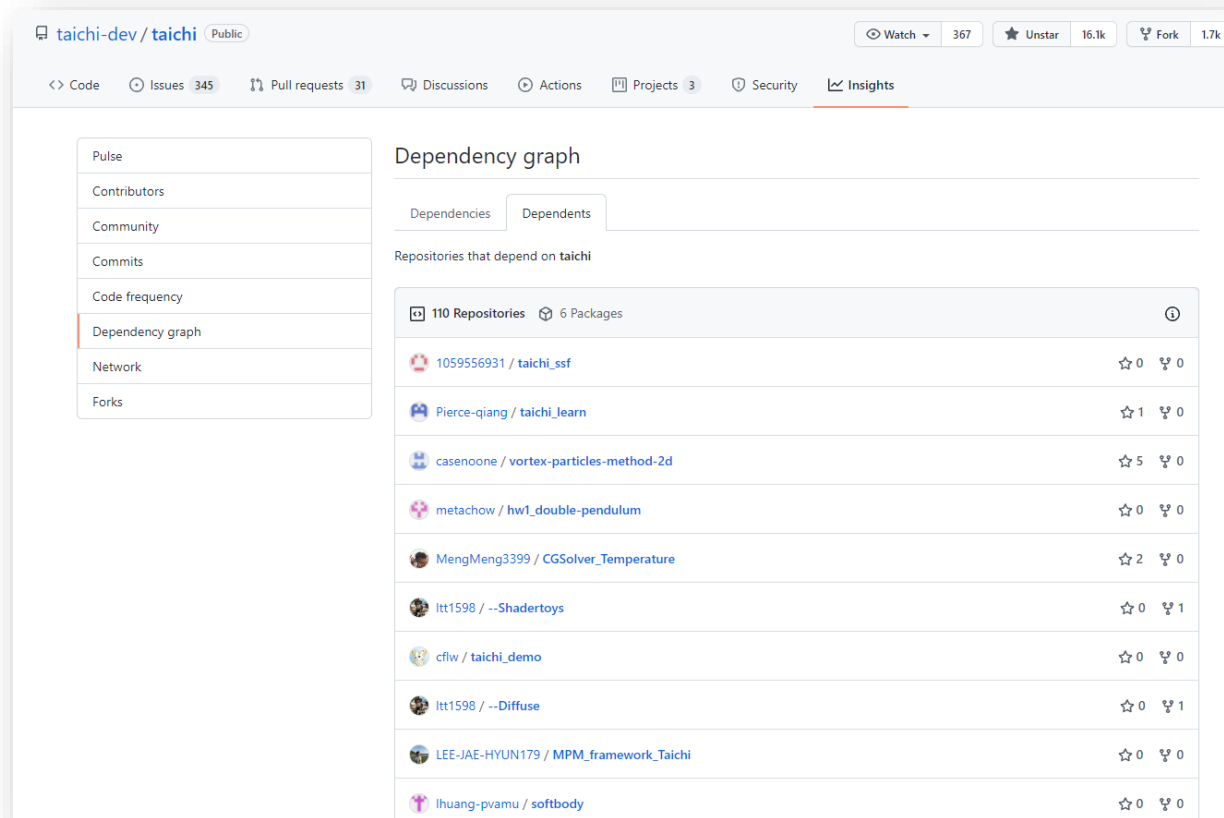
[@MengMeng3399]



[@runck]

Gifts for the gifted

- Use [Template](#) for your homework
- Next check Dec. 14, 2021



Questions?

本次答疑：12/02 ◀ 作业分享也在这里

下次直播：12/07

直播回放：Bilibili 搜索「太极图形」

主页&课件：<https://github.com/taichiCourse01>

主页&课件(backup)：<https://docs.taichi.graphics/tgc01>