# Diving into SPH

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# Navier-Stokes Equations (Still N-S!)

The key idea is unchanged. You want to approximate the N-S equations as before, but not discretize your solution on a mesh. This is like any Lagrangian method. To recap the equations for continuum flow:

Continuity Equation:

$$\frac{D\rho}{Dt} = -\rho\nabla\cdot\mathbf{u}$$

Momentum Equation:

$$\frac{D\mathbf{u}}{Dt} = -\frac{1}{\rho}\nabla P + \nu\nabla^2\mathbf{u} + \mathbf{f}$$

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# Smoothed Particle Hydrodynamics (SPH)

What makes it different

- SPH is mesh-free! Treatment of complex and/or moving geometries is easier
- Free surface flows that are challenging in Eulerian methods are more or less natural in SPH
- Exact (and simultaneous) conversion of mass and momentum
- Inherent parallelism



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#### Mathematical Form

In 3D and 4th order (also what I've used)

$$W(q) = rac{21}{64\pi h^3} \cdot egin{cases} (2-q)^4 (1+2q) & ext{if } 0 \le q \le 2 \ 0 & ext{otherwise} \end{cases}$$
 (1)

#### where:

- Kernels just convolve over the function, and aim to be an approximation to the Dirac Delta function
- It has compact support, which means the tail is not infitine, it takes a certain range of particles around it
- A kernel function must be consistent, i.e.:

$$\int_{-\infty}^{\infty} W(q) \, dq = 1 \tag{2}$$

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## **SPH Equations**

Continuity

The approach in SPH is sto sum up contributions from neighbours.

$$\rho_{a} = \sum_{b} m_{b} \frac{W(\mathbf{r}_{ab}, h)}{\rho_{b}} \tag{3}$$

where:

- *W* is the smoothing kernel function.
- $\mathbf{r}_{ab}$  is the vector from particle a to b.
- *h* is the smoothing length.



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### **SPH Equations**

#### Momentum

The kernel functions can be differentiated - and this is used in the gradient for pressure in the momentum equation. Pressure is calculated from the equation of state.

$$\frac{D\mathbf{v}_a}{Dt} = -\sum_b m_b \left(\frac{P_a}{\rho_a^2} + \frac{P_b}{\rho_b^2}\right) \nabla_a W_{ab}(h) + \nu \nabla^2 \mathbf{v}_a + \mathbf{f}_a \tag{4}$$

#### where:

- $P_a$  and  $P_b$ ,  $\rho_a$  and  $\rho_b$  are the pressures and densities of the current particle and its neighbours
- $\nabla_a W_{ab}(h)$  is the gradient of the smoothing kernel between particles a and b with smoothing length h.

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# Numerical quantities

Kernel Weighted Interpolation

- Postprocessing of SPH would involve interpolation of the same nature to obtain a numerical quantity from particle quantities
- We renormalize the interpolation using a Shepherd summation in the denominator
- For instance, the numerical pressure  $P_a$  at particle a can be computed as a weighted sum of neighbour particles

$$P_{a} = \frac{\sum_{b} P_{b} \frac{m_{b}}{\rho_{b}} W_{ab}}{\sum_{b} \frac{m_{b}}{\rho_{b}} W_{ab}}$$

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## A Simple Solver

#### Pressure Force

```
for(const auto& other : particles)
  if(&p != &other)
    double r = distance(p, other);
    if(r > 0)
      double pj = calculatePressureTait(other);
      double rhoj = other.density;
      double w_press = wendlandGradient2D(r);
      pressureForceX +=
        other.mass * (pi / (rhoi * rhoi) + pj / (rhoj * rhoj)) * w_press;
      pressureForceY +=
        other.mass * (pi / (rhoi * rhoi) + pj / (rhoj * rhoj)) * w_press;
```

# A Simple Solver

Density

```
double calculateDensity(const std::vector<Particle>& particles, const
          Particle& p)
{
          double density = 0.0;
          for(const auto& other : particles)
          {
                double r = distance(p, other);
                 density += other.mass * wendlandKernel2D(r);
        }
        return density;
}
```

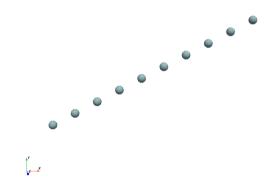
# A Simple Solver

Main loop

```
for(auto& p : particles)
    // From https://www.cs.cmu.edu/~scoros/cs15467-s16/lectures/11-fluids2.pdf
    p.vx += dt * p.fx / p.density;
    p.vv += dt * (p.fv / p.density - g); //add gravity here
    p.x += dt * p.vx;
    p.y += dt * p.vy;
    //recalculate forces in the next step
    p.fx = 0.0:
    p.fy = 0.0;
  // Calculate density and pressure forces
  for(auto& p : particles)
    p.density = calculateDensity(particles, p);
  calculatePressureForce(particles);
```

# Just gravity

(Not much, but I'm glad I could get something basic up)



\*Using PyVista for this

### Dam Break

### PyVista Visualization





Figure: Domain at the end of the run



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### Dam Break

### Generating CV Mesh

```
def write_to_stl():
  x resolution = 10
  v resolution = 10
  x = np.linspace(0.07, 0.17, x_resolution)
  y = np.linspace(0.07, 0.17, y_resolution)
  z = np.zeros(1) # 2D box, so z dimension has only one value
  grid = pv.RectilinearGrid()
  grid.x = x
  grid.y = y
  grid.z = z
  mesh = grid.cast_to_structured_grid().extract_surface()
  mesh.save("output.stl")
  mesh.plot("show_edges=True")
```

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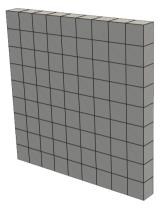


Figure: Creating a control volume of the dimensions and location of the leftmost box



### Dam Break

### Recovering Surface Mesh in Nemo

```
def read_into_nemo(): #note I could only read back an STL, so it's going to be
   triangulated ...
    mesh = n.Mesh.load(n.File("output.stl"))
    segmentcloud = mesh.data
    areas = mesh.get surface area per segment()
    normals = segmentcloud.get_segment_normals()
   print(segmentcloud.point attributes) #-->empty dict, how to proceed?
    print(segmentcloud.point_attributes["face_normal"])# KeyError: 'face_normal
   #Riemann sum of a unit vector
    dot_product = np.sum(normals, axis=1) #horizontally sum across normals -> 1
    riemann_sum = np.sum(dot_product * areas)
    print('Riemann sum is:')
    print(mesh.get_surface_area())
    print(riemann_sum)
```



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Figure: Displaying the triangulated CV volume. Triangulated because I could only import back an STL and not a structured grid cast to a .vtu

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### Interpolation

#### Control Volume

(I think the time was beyond the scope for me to implement this, but this is what I considered. This should apply to all quantities - flux and primitives alike)

### Pseudoalgorithm

- 1 Define a search radius = smoothing length h for each particle (not sure, but could be a start)
- 2 For each grid segment, look for neighbors within the radius *h*. Find atleast one point particle.
- 3 Calculate value of numerical quantity using Shephard normalization and use value for grid segment. This is essentially a projection onto the cell centroid.
- 4 Do Riemann sum over grid for total value on surface

4D > 4B > 4B > 4B > 900