

Numerical modelling and High Performance Computing for sediment flows: Part one

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Physics of particle-laden fresh water flows above salted water:



River delta of Irrawady, Myanmar (ESA)

How to build a high performance fluid solver ?

① Introduction

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③ Conclusion and perspectives

Navier-Stokes equations

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Variables of interest

- We consider an incompressible fluid with the following properties:
 - ① A constant **viscosity** μ (resistance to deformation by shear stress).
 - ② A constant **density** ρ (mass per unit volume).

- In **2D**, we are interested in solving the following physical fields:

- ① The fluid **velocity** $\mathbf{u}(\mathbf{x}, t) = \begin{bmatrix} u_x(\mathbf{x}, t) \\ u_y(\mathbf{x}, t) \\ 0 \end{bmatrix}$

- ② The fluid **vorticity** $\boldsymbol{\omega}(\mathbf{x}, t) = \begin{bmatrix} 0 \\ 0 \\ \omega_z(\mathbf{x}, t) \end{bmatrix}$

- ③ The internal **pressure** of the fluid $P(\mathbf{x}, t)$.

We define those variables on a spatial domain Ω with boundaries $\partial\Omega$.

i.e. $\forall \mathbf{x} = (x, y) \in \Omega$ which is a **rectangular domain** $[0, L_x] \times [0, L_y]$.

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We define those variables on a spatial domain Ω with boundaries $\partial\Omega$.

i.e. $\forall \mathbf{x} = (x, y, z) \in \Omega$ which is a **cuboid domain** $[0, L_x] \times [0, L_y] \times [0, L_z]$.

Variables of interest - velocity and vorticity

- The vorticity is a pseudovector field that describes the local spinning motion of the fluid near some point in the whole domain.
- You can directly compute the vorticity $\boldsymbol{\omega}$ by taking the **curl** of the velocity \mathbf{u} . Given $\nabla = \left(\frac{\partial}{\partial x}, \frac{\partial}{\partial y}, \frac{\partial}{\partial z} \right)$ we have $\boldsymbol{\omega} = \nabla \times \mathbf{u}$.
- In 3D, this is **3-components vector field** (like the velocity):

$$\begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} = \begin{bmatrix} \frac{\partial}{\partial x} \\ \frac{\partial}{\partial y} \\ \frac{\partial}{\partial z} \end{bmatrix} \times \begin{bmatrix} u_x \\ u_y \\ u_z \end{bmatrix} = \begin{bmatrix} \frac{\partial u_z}{\partial y} - \frac{\partial u_y}{\partial z} \\ \frac{\partial u_x}{\partial z} - \frac{\partial u_z}{\partial x} \\ \frac{\partial u_y}{\partial x} - \frac{\partial u_x}{\partial y} \end{bmatrix}$$

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- In 2D, this is a **scalar field** (only one component):

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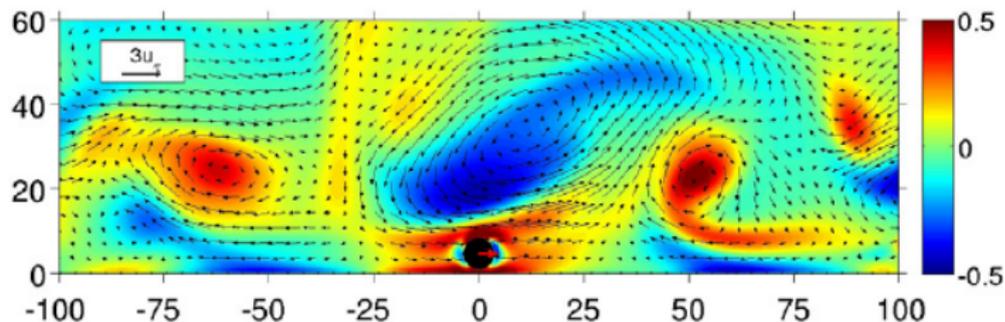
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Example of 2D velocity and vorticity fields

Navier-Stokes - conservation of mass

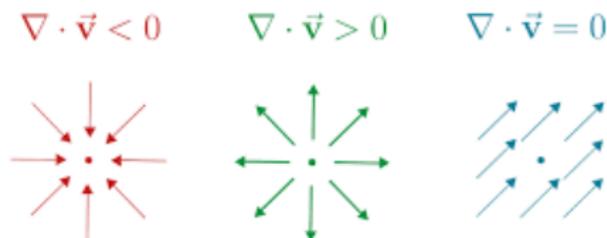
- The first equation relates to the conservation of the mass:

$$\frac{\partial \rho}{\partial t} + \nabla \cdot (\rho \mathbf{u}) = 0$$

- With a constant density ρ this equations reduce to:

$$\nabla \cdot \mathbf{u} = \frac{\partial u_x}{\partial x} + \frac{\partial u_y}{\partial y} + \frac{\partial u_z}{\partial z} = 0$$

- You can see the divergence as a scalar field representing the quantity of a vector field's source at each point:



Navier-Stokes - conservation of momentum

MASS
Density of the fluid

ACCELERATION
How velocity experienced by a particle changes with time

FORCE
All the forces that are acting on the fluid

$$\rho \left(\frac{\partial \mathbf{V}}{\partial t} + \mathbf{V} \cdot \nabla \mathbf{V} \right) = \nabla P + \rho \mathbf{g} + \mu \nabla^2 \mathbf{V}$$

Change in velocity over time

The speed and direction which the fluid is moving

Internal pressure gradient of the fluid (the change in pressure)

External forces acting on the fluid (such as gravity)

Internal stress forces acting on the fluid (taking into consideration viscous effects)

Navier-Stokes Equations

Describe the flow of incompressible fluids.

Navier-Stokes - velocity-vorticity formulation

- What if we do not want to solve the pressure field ?

$$\begin{aligned}\nabla \cdot \mathbf{u} &= 0 \\ \rho \left[\frac{\partial \mathbf{u}}{\partial t} + (\mathbf{u} \cdot \nabla) \mathbf{u} \right] &= \mu \Delta \mathbf{u} - \nabla P + \rho \mathbf{g}\end{aligned}$$

- Just apply the **curl** operator to the second equation:

$$\begin{aligned}\nabla \cdot \mathbf{u} &= 0 \\ \nabla \times \left(\rho \left[\frac{\partial \mathbf{u}}{\partial t} + (\mathbf{u} \cdot \nabla) \mathbf{u} \right] \right) &= \nabla \times (\mu \Delta \mathbf{u} - \nabla P + \rho \mathbf{g})\end{aligned}$$

- In 3D $(\mathbf{u}, \boldsymbol{\omega})$ formulation has 6 unknowns vs. 4 for (\mathbf{u}, P) formulation.

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Target hardware

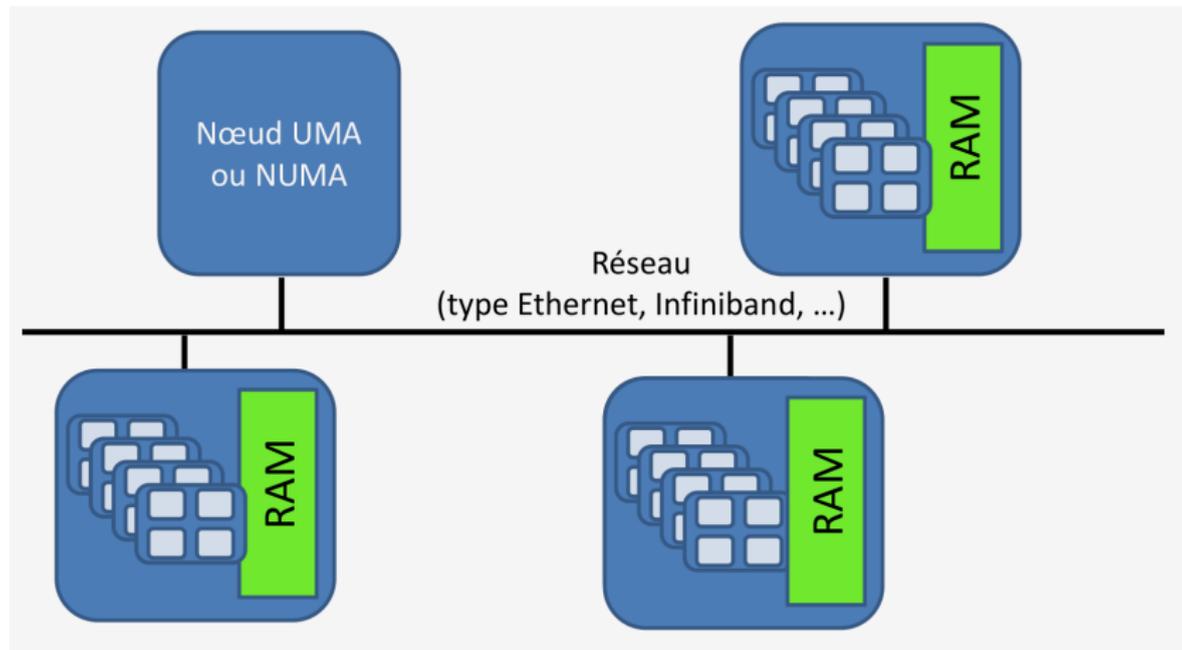
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② How to build a high performance fluid solver ?

- Navier-Stokes equations
- **Target hardware**
- The HySoP library
- Operator splitting

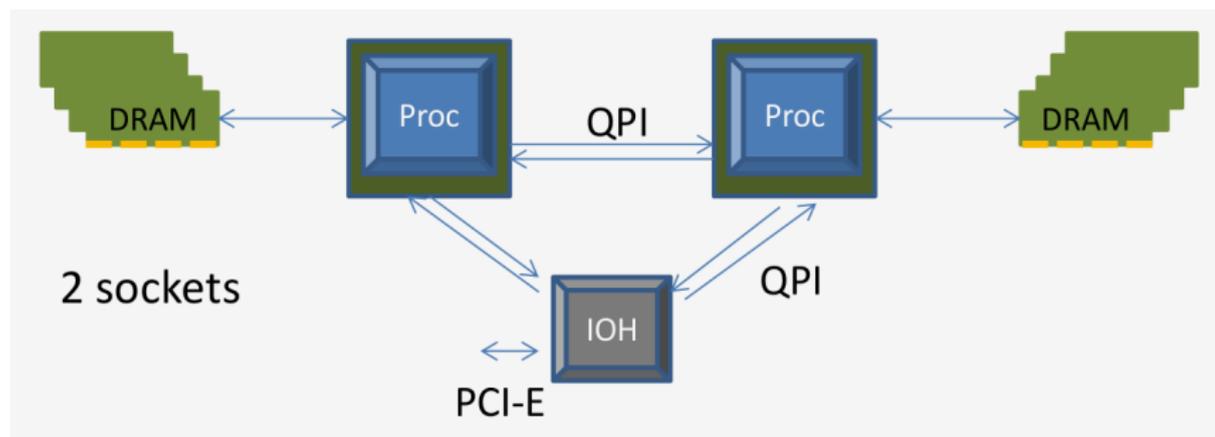
③ Conclusion and perspectives

Global view of a compute cluster



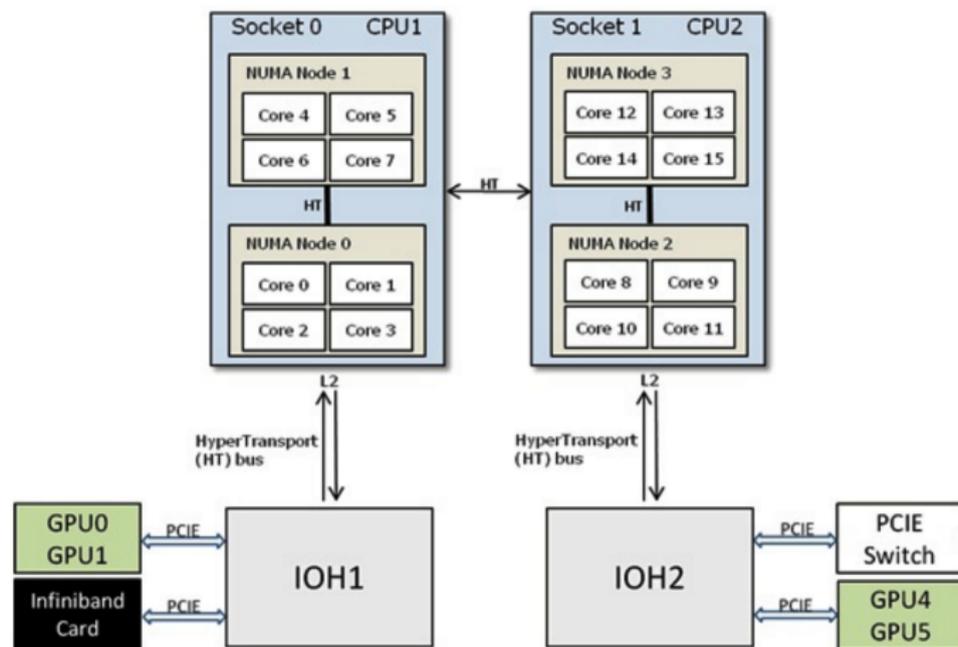
A single compute node

- Each compute node has its own processor(s) and memory banks:

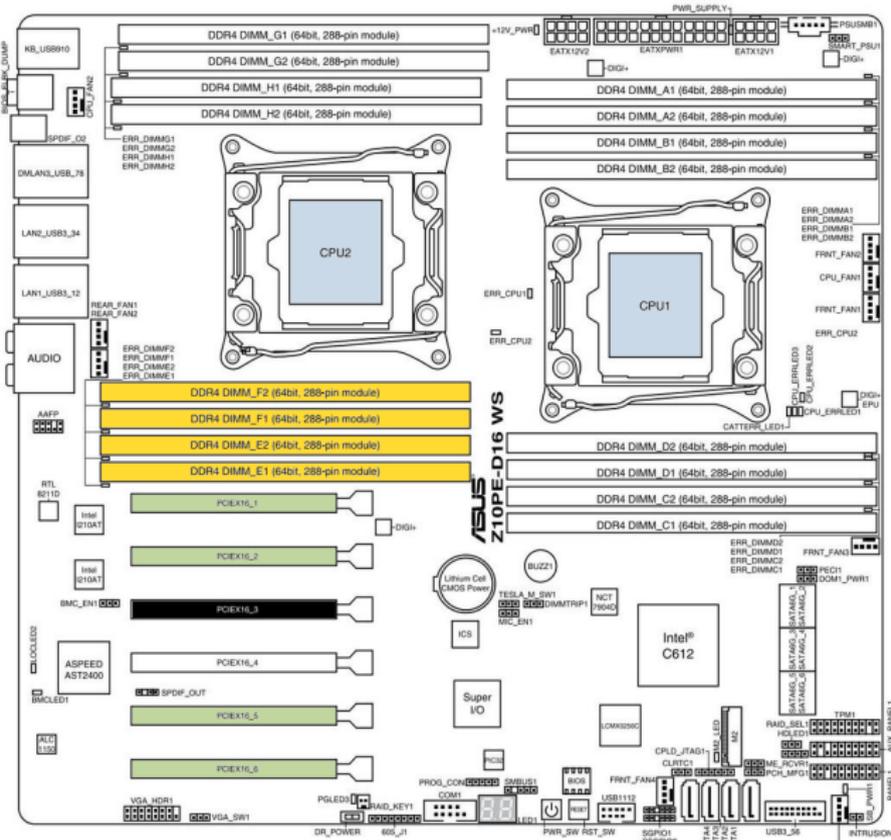


- On conventional architectures you can have 1 to 8 sockets (CPUs).

Zoom on a dual socket compute node



Dual socket motherboard



Accelerators and coprocessors - server grade GPUs



Nvidia Tesla P100



AMD Firepro W9100

This range of products is dedicated to the **double precision compute tasks** (no graphical tasks). Here $FP64 = 1/2 FP32$.

Accelerators and coprocessors - gaming GPUs



Nvidia Titan Xp



AMD RX Vega 64

This range of products is dedicated to the **single precision compute tasks** (mostly graphical tasks). Here $FP64 = \alpha FP32$ where $\alpha \in [1/32, 1/4]$.

Accelerators and coprocessors - Intel's response to GPUs

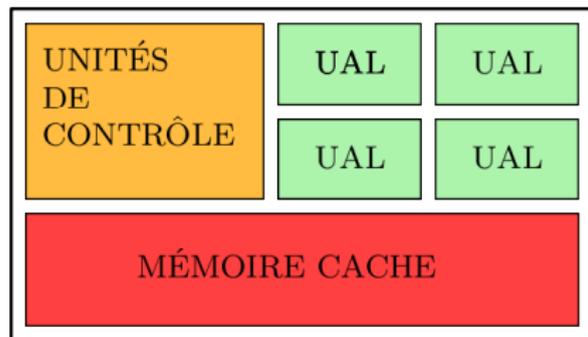


MIC (Many Integrated Core) coprocessor - Xeon-Phi (72 cores)

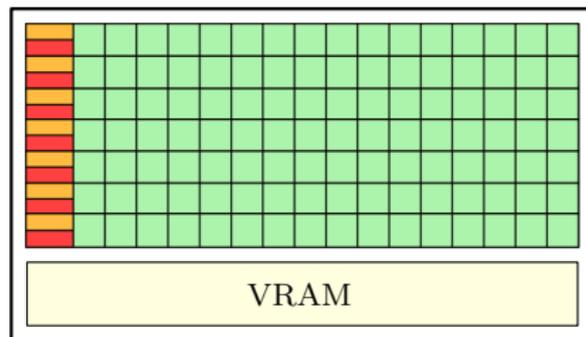
Those cards have been discontinued by Intel since 2017.
Here FP64 = 1/2 FP32.

CPU vs GPU

CPU

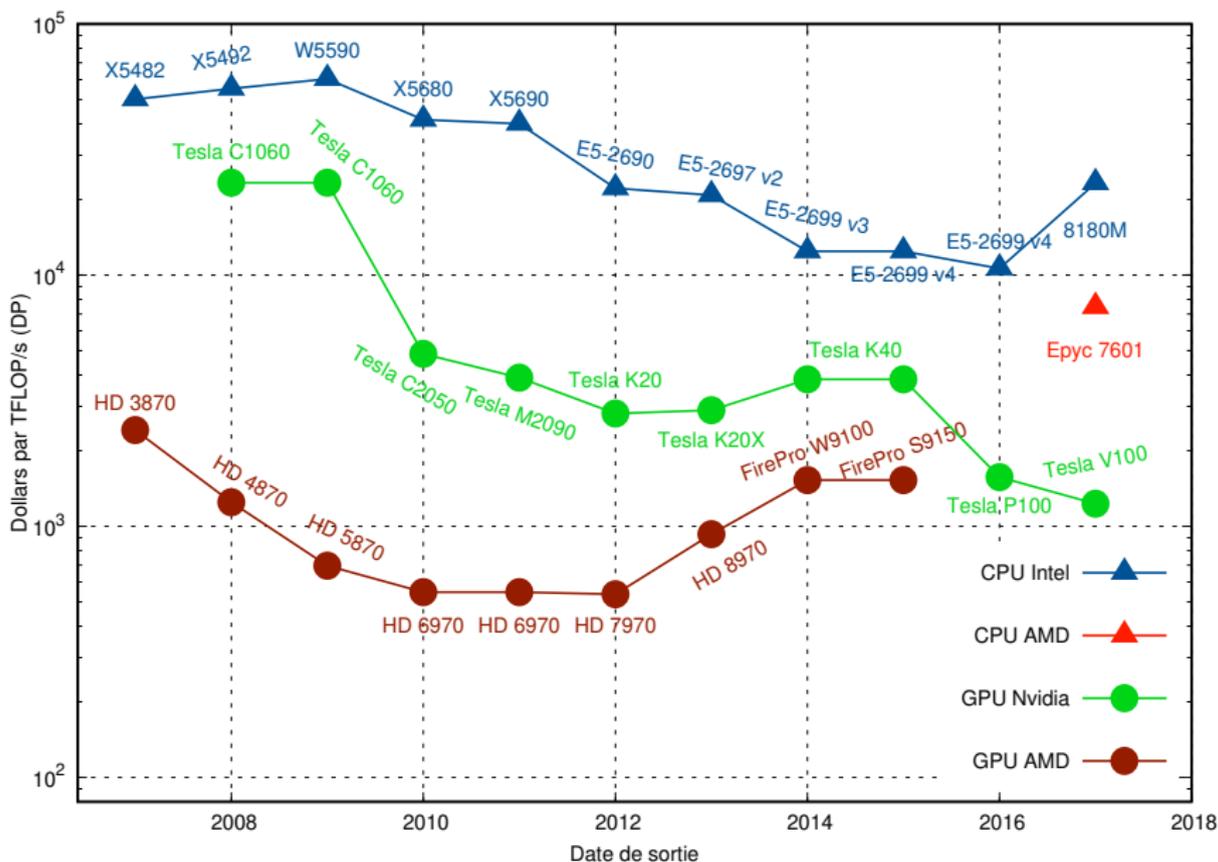


GPU



- More compute cores (but operating in groups) Less cache per core
- Embedded memory

Advantage of coprocessors



Programming models for coprocessors

- **OpenCL (Open Computing Language)**: Framework for writing programs that execute across heterogeneous hardware (CPU, GPU, MIC, FPGA, DSP). This is not synonym of **performance portability**
 - ① OpenCL 1.x: Based on C99 (2008).
 - ② OpenCL 2.x: Based on a subset of C++14 (2013).
- **CUDA (Compute Unified Device Architecture)**: Nvidia only (C++14)
- **OpenMP (Open Multi-Processing)**: Since v.4.0 (2013) you can offload compute task to accelerators (pragma based approach).
- **OpenACC (Open Accelerators)**: Like OpenMP but for accelerators.

	Intel		AMD		Nvidia
	CPU	MIC	CPU	GPU	GPU
OpenCL	2.1	1.2	2.0	1.2	1.2
CUDA	-	-	-	-	10.0
OpenMP	gcc/icc	gcc/icc	gcc	gcc	gcc
OpenACC	-	-	-	-	gcc/pgcc

The HySoP library

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③ Conclusion and perspectives

Discretization of the variables

- Variable are discretized on regular cartesian grids (with ghosts):

$$\Omega = [0, L_x] \times [0, L_y] \times [0, L_z]$$

$$N = N_x \times N_y \times N_z$$

$$\mathbf{dx} = [dx, dy, dz] = \left[\frac{L_x}{(N_x - 1)}, \frac{L_y}{(N_y - 1)}, \frac{L_z}{(N_z - 1)} \right]$$

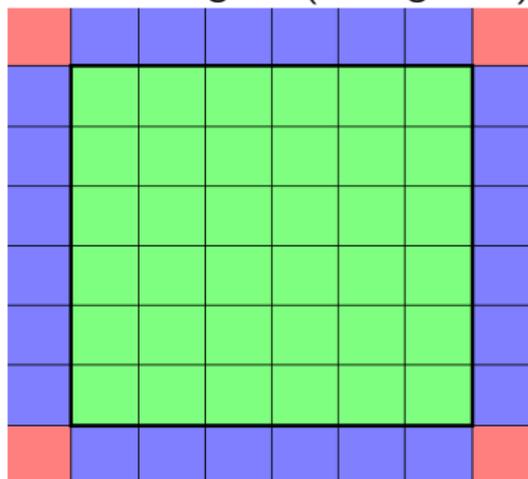
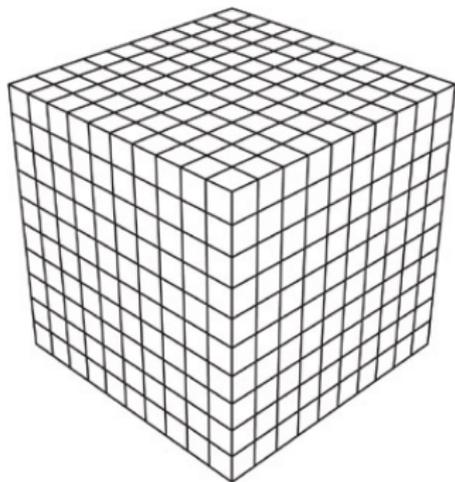
$$F_{ijk}(t) = F(idx, jdy, kdz, t)$$

$$\forall (i, j, k) \in \llbracket 0, N_x - 1 \rrbracket \times \llbracket 0, N_y - 1 \rrbracket \times \llbracket 0, N_z - 1 \rrbracket$$

- This allows us to use efficient methods like:
 - 1 Finite differences
 - 2 Spectral methods

Discretization of the variables

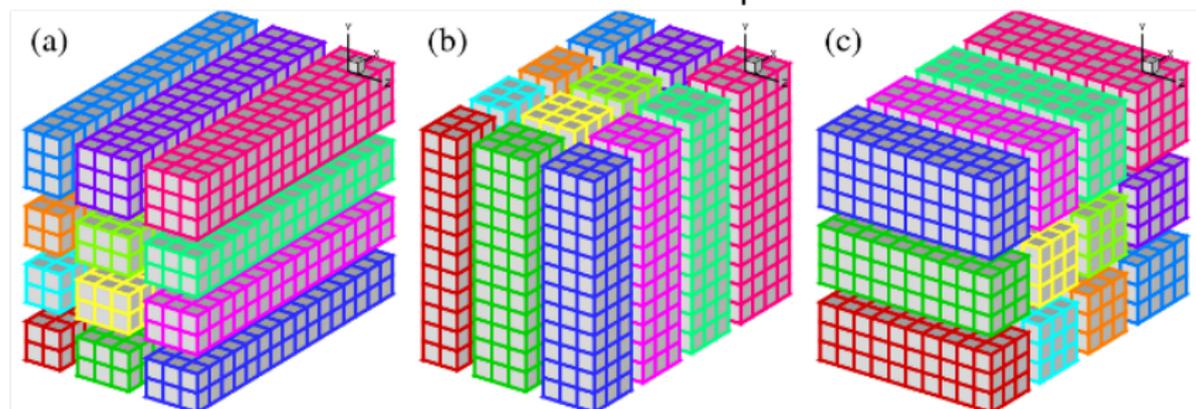
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Discretization of the variables

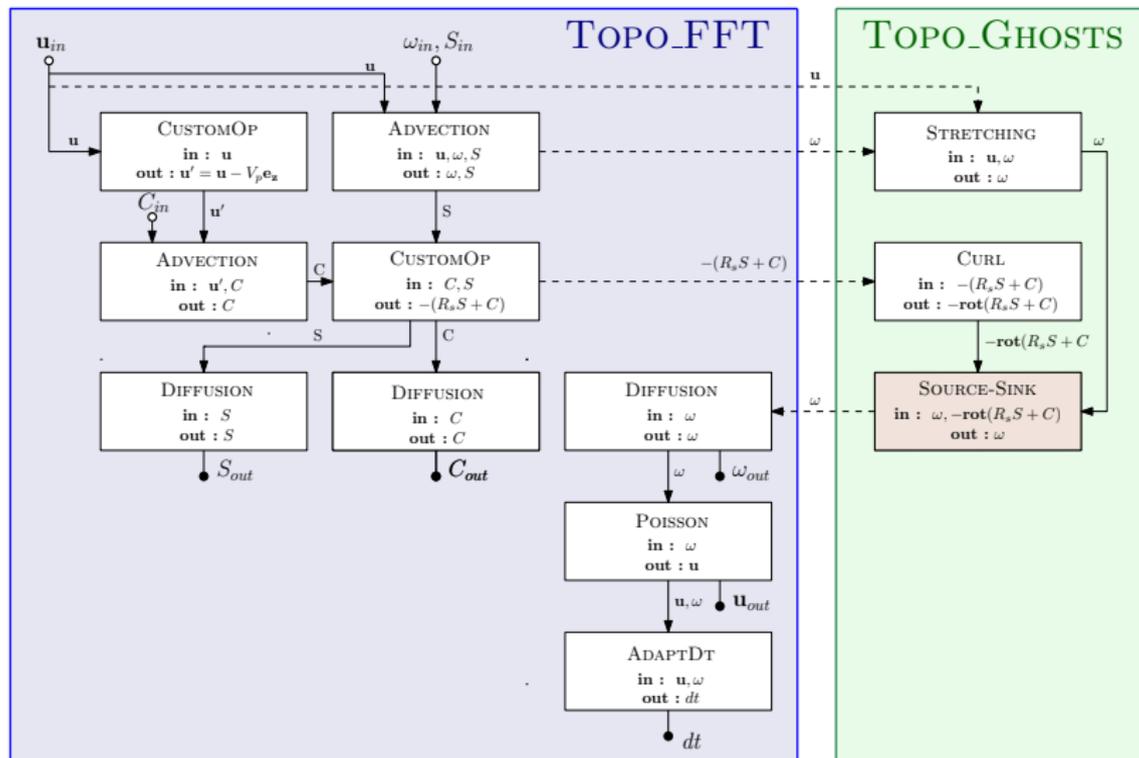
- All the variables are distributed on the compute nodes:



- One variable may have many different topologies depending on operator constraints.
- We use MPI (Message Passing Interface) for inter node communication.

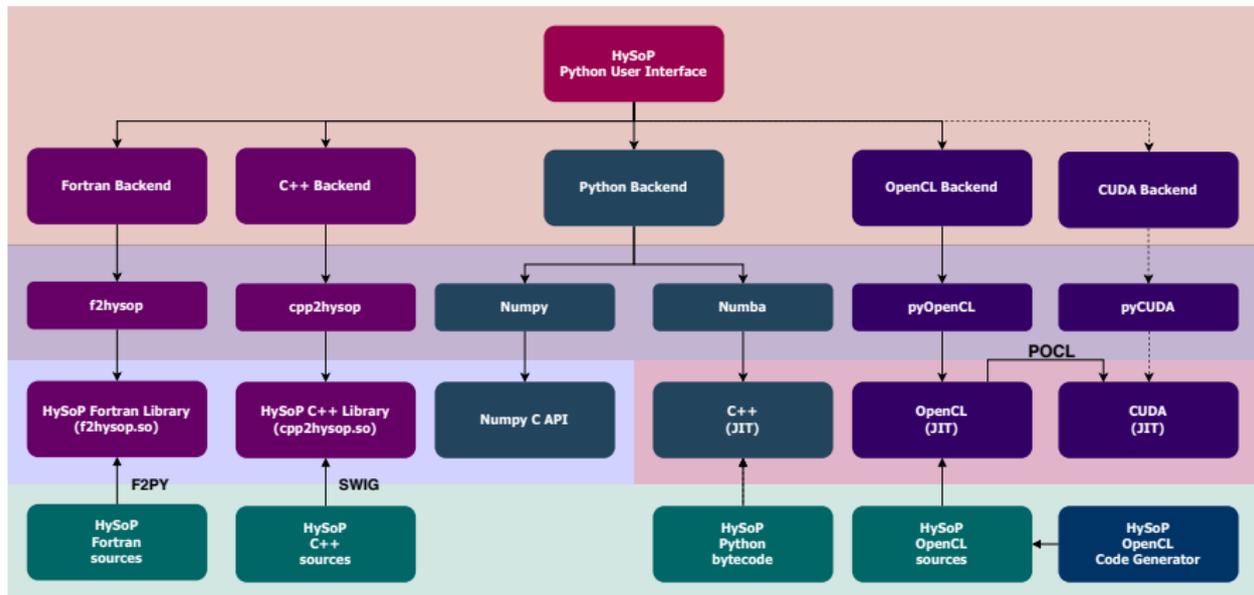
Building a problem

After specifying variables, the user builds a DAG of operators:



HySoP backends

- The user has to choose operators and the backend it will run on:



- Currently we have Python, C++, Fortran and OpenCL support.

Operator splitting

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② How to build a high performance fluid solver ?

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③ Conclusion and perspectives

Operator splitting

- The idea is to **split the Navier-Stokes equations** in $(\mathbf{u}, \boldsymbol{\omega})$ formulation.
- Independent operators are easier to implement, test and debug !

$$\frac{\partial \boldsymbol{\omega}}{\partial t} + (\mathbf{u} \cdot \nabla) \boldsymbol{\omega} = (\boldsymbol{\omega} \cdot \nabla) \mathbf{u} + \nu \Delta \boldsymbol{\omega} + \nabla \times \mathbf{g}$$

- We split the momentum equations as the following:

- 1 Transport: $\frac{\partial \boldsymbol{\omega}}{\partial t} + (\mathbf{u} \cdot \nabla) \boldsymbol{\omega} = 0$
- 2 Stretching: $\frac{\partial \boldsymbol{\omega}}{\partial t} = (\boldsymbol{\omega} \cdot \nabla) \mathbf{u}$ [this term only appears in 3D]
- 3 Diffusion: $\frac{\partial \boldsymbol{\omega}}{\partial t} = \nu \Delta \boldsymbol{\omega}$
- 4 Ext. forces: $\frac{\partial \boldsymbol{\omega}}{\partial t} = \nabla \times \mathbf{g}$

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- Each discretized operator **forces the timestep** to be small enough:

① Transport: $\Delta t_{adv} < \frac{LCFL}{|\boldsymbol{\omega}|_{\infty}}$ with $LCFL < 1$.

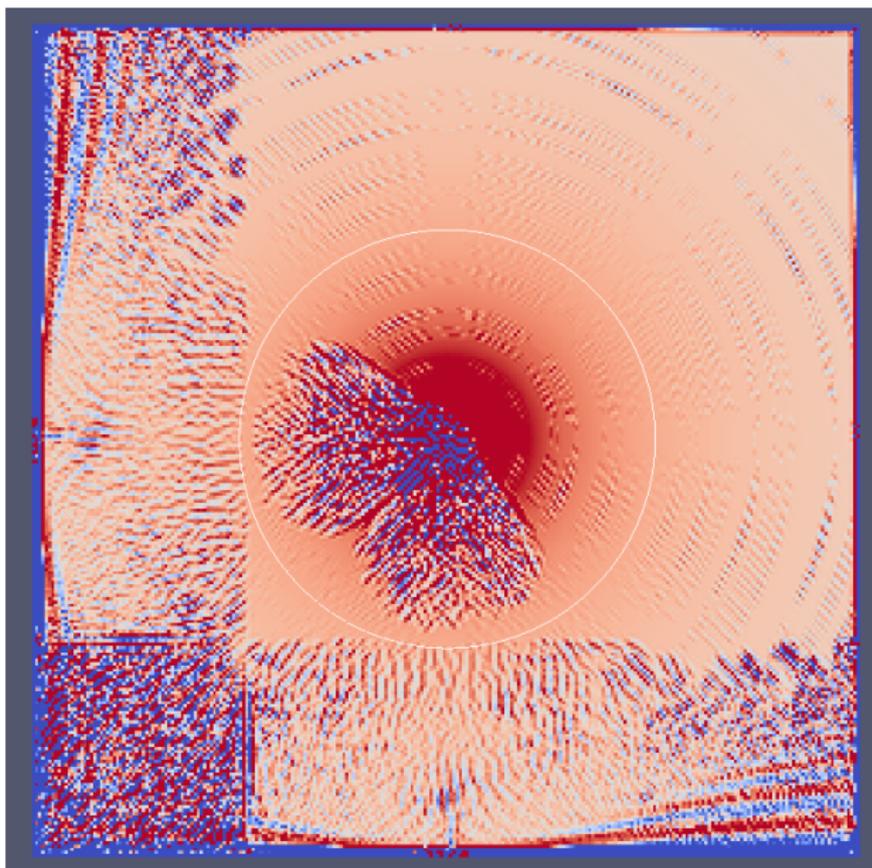
② Stretching: $\Delta t_{str} < \frac{RK}{\max_i \sum_j \left| \frac{\partial \mathbf{u}_i}{\partial \mathbf{x}_j} \right|}$

③ Diffusion: $\Delta t_{diff} < \frac{RK}{\nu} dx^2$

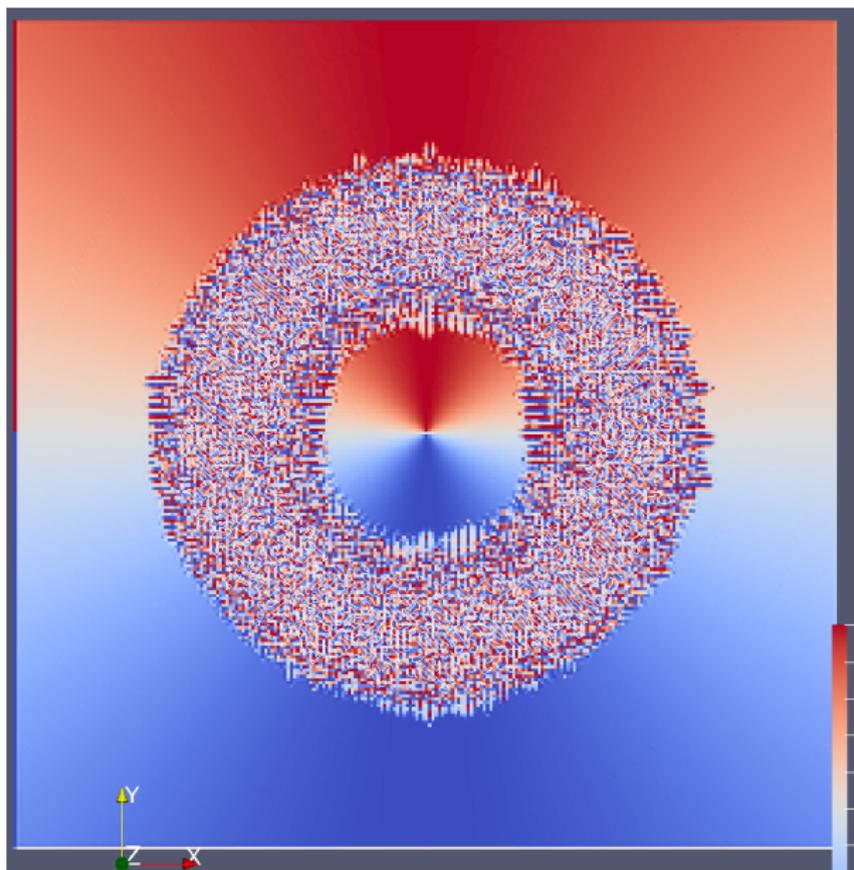
④ Ext. forces: Δt_{fext} depends on the force field

- $\Delta t = \min(\Delta t_{adv}, \Delta t_{str}, \Delta t_{diff}, \Delta t_{fext})$.

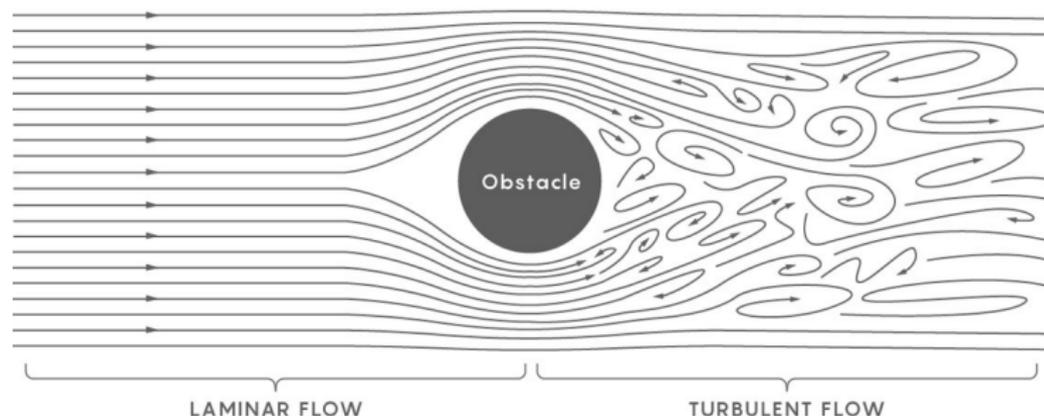
What happens if those conditions are not met ?



What happens if those conditions are not met ?



What about immersed boundaries ?



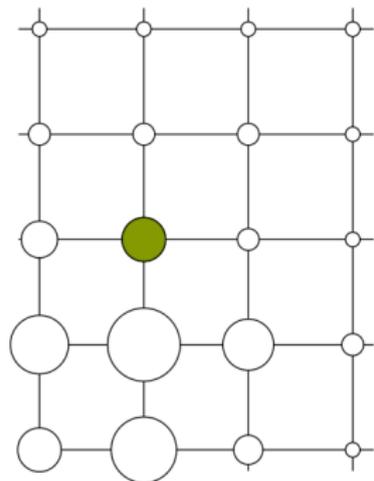
- We just introduce a penalization term to correct the velocity:

$$\frac{\partial u(\mathbf{x}, t)}{\partial t} = \chi_s(\mathbf{x}) \lambda (\mathbf{u}_d - \mathbf{u}(\mathbf{x}, t))$$

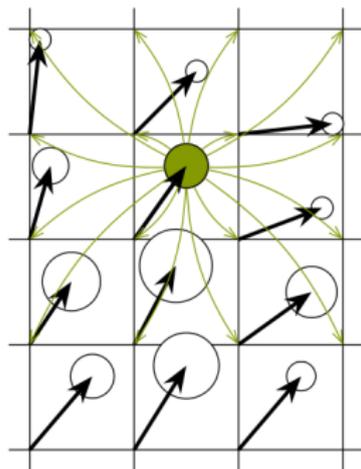
$$\chi_s(\mathbf{x}) = \begin{cases} 1 & \text{if } \mathbf{x} \in S \\ 0 & \text{if } \mathbf{x} \notin S \end{cases}$$

Remeshed particles methods

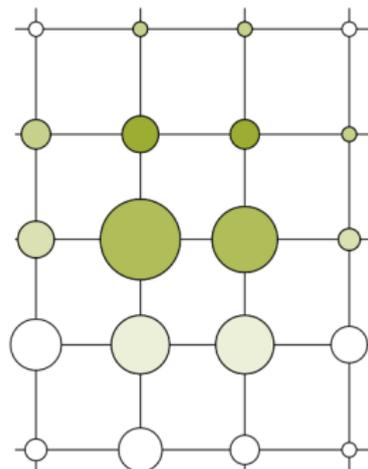
Directional operator splitting example: advection



(a) Instant t^n



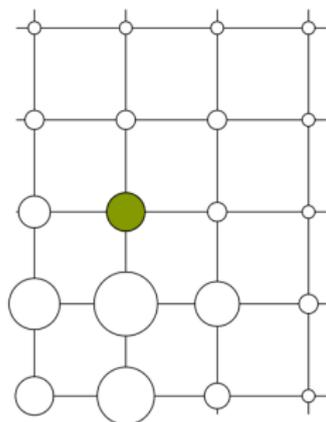
(b) Advection et remillage



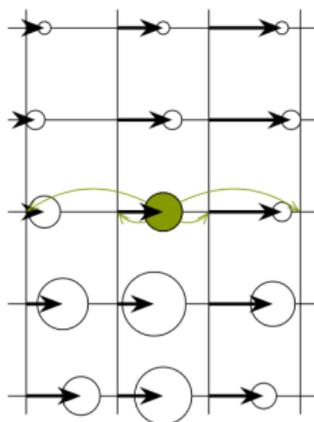
(c) Instant t^{n+1}

Remeshed particles methods - directional splitting

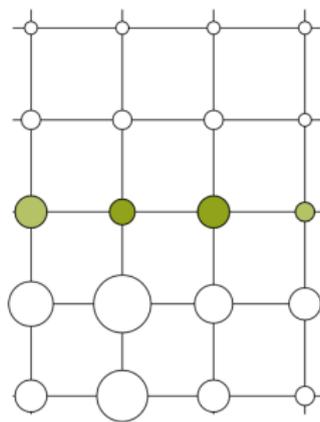
2D advection-remesh using directional splitting (1st axis)



(a) Instant t^n



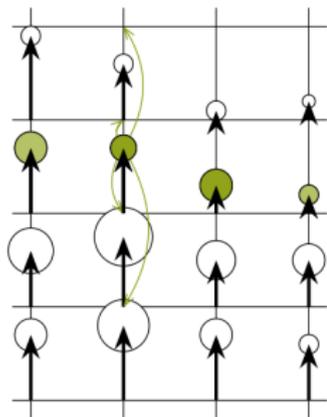
(b) Advection et remaillage, direction X



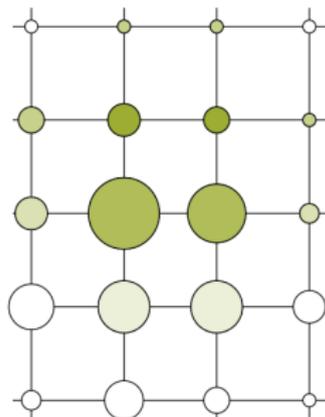
(c) État intermédiaire

Remeshed particles methods - directional splitting

2D advection-remesh using directional splitting (2nd axis)

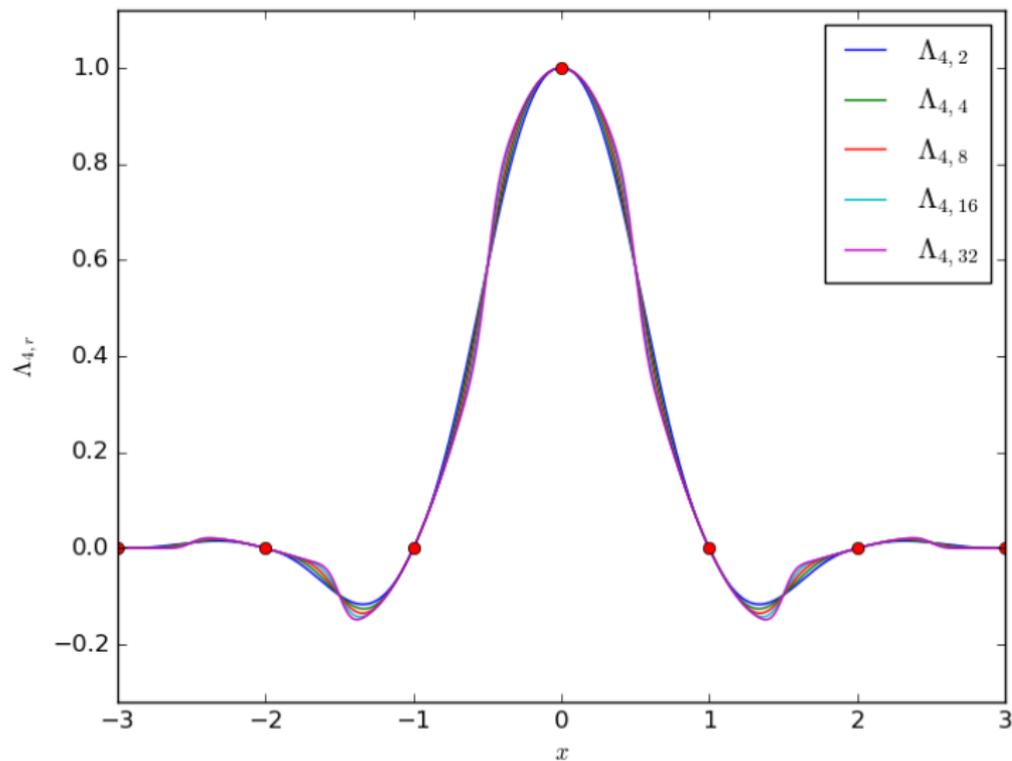


(d) Advection et remaillage, direction Y



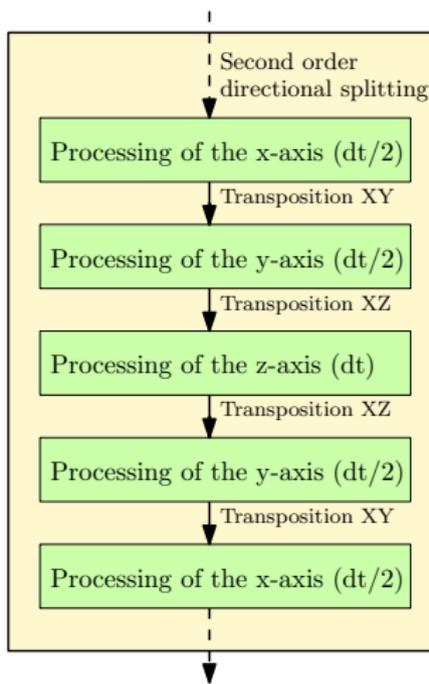
(e) Instant t^{n+1}

Remeshing kernels as obtained with [3]



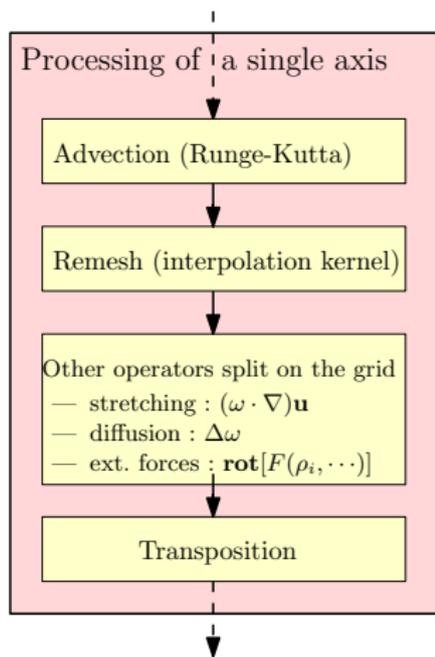
Generalized directional splitting (Strang 2nd order)

- Generalized directional splitting to all splittable operators.
- Between each directional advection-remesh step the data is **transposed** to ensure contiguous memory accesses on the accelerator:



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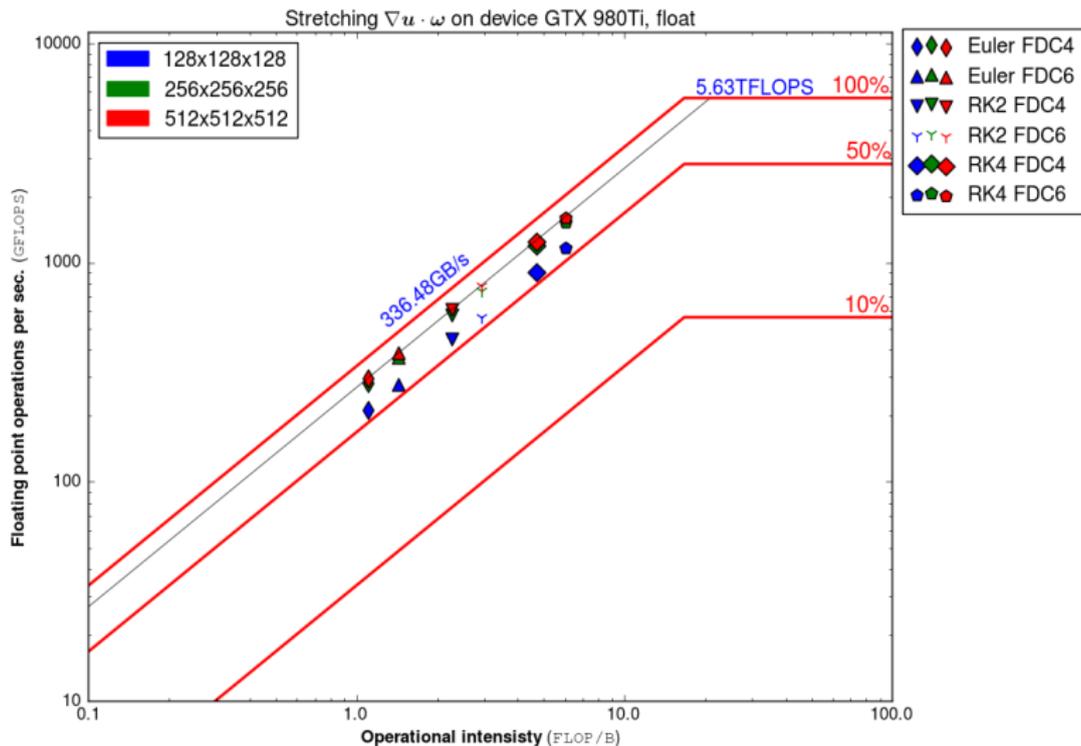
Example of directional splitting for the stretching

- Conservative form: $\frac{\partial \omega}{\partial t} = \mathbf{div} [\mathbf{u} \otimes \omega] = \mathbf{u} \underbrace{(\nabla \cdot \omega)}_0 + (\omega \cdot \nabla) \mathbf{u}.$

$$\frac{\partial \omega}{\partial t} = \mathbf{div} [\mathbf{u} \otimes \omega] = \mathbf{div} \begin{bmatrix} \mathbf{u}_x \omega_x & \mathbf{u}_x \omega_y & \mathbf{u}_x \omega_z \\ \mathbf{u}_y \omega_x & \mathbf{u}_y \omega_y & \mathbf{u}_y \omega_z \\ \mathbf{u}_z \omega_x & \mathbf{u}_z \omega_y & \mathbf{u}_z \omega_z \end{bmatrix}$$

- Directional splitting (3 operators \rightarrow 9 operators)
 - Splitting axe x: $\frac{\partial \omega}{\partial t} = \frac{\partial}{\partial x} [\omega_x \mathbf{u}]$
 - Splitting axe y: $\frac{\partial \omega}{\partial t} = \frac{\partial}{\partial y} [\omega_y \mathbf{u}]$
 - Splitting axe z: $\frac{\partial \omega}{\partial t} = \frac{\partial}{\partial z} [\omega_z \mathbf{u}]$

Performances obtained for the stretching operator



Conclusion and perspectives

Conclusion

- OpenCL is an interesting tool for HPC on heterogeneous compute platforms.
- However you have to ensure the portability of the performances.
- Splitting the Navier-Stokes in many subproblems is the key for simplicity and a natural framework for task parallelisation.
- You can already compute nice simulations with a single compute node (even without GPU).

Future developments

- Implement the multi-scale approach and MPI FFT-based solvers (global transposition of memory accross all processes).
- Full in-core simulation on multiple GPUs should enable high spatial resolution simulations.
- Full release of HySoP v2.0 to the public in 2019.

References

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Thanks for your attention !
Any questions ?