

CFD Code Development Frameworks

Python, C, OpenFOAM, Fenics

Dr. Rodolfo Monico, Taher Chegini, M. Sarraf Josaghani

CTFM 2018
University of Houston

rostilla@uh.edu
tchegini@uh.edu
m.sarraf.j@uh.edu

June 2, 2018



Table of Content

1 Artificial Compressibility Method

- Navier-Stokes Equations
- Deriving ACM

2 Development: from scratch

- Solution Algorithm
- Grid
- Discretization

3 Development: OpenFOAM

- OpenFOAM
- Formulation

4 Problem Statement

- Workshop
- Projects Structure



Governing Equations

Incompressible Navier-Stokes Equations (INSE) in Eulerian form:

$$\partial_t \rho + \rho \nabla \cdot \mathbf{u} = 0 \quad (1)$$

$$\partial_t \mathbf{u} + \mathbf{u} \cdot \nabla \mathbf{u} = -\frac{1}{\rho} \nabla p + \nu \nabla^2 \mathbf{u} + \mathbf{g} \quad (2)$$

Eqn. 1: Density can be omitted, but let's keep it for now.

Eqn. 2:

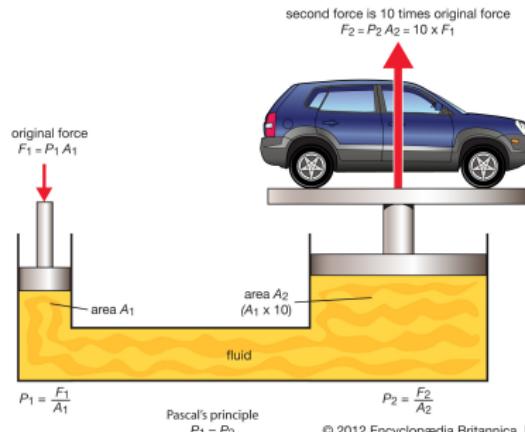
- LHS is acceleration which is intrinsically Lagrangian and when it is translated to Eulerian coordinates it becomes non-linear
- RHS describes influence of pressure gradient, viscosity and body forces.



Complexities

Difficulties for numerical solution:

- Nonlocality of pressure gradient (Solution: None)
- Nonlinearity of acceleration term (Solution: e.g. Picard)
- Pressure and velocity are coupled and there is no separate equation for pressure (Solution: e.g. SIMPLE)





Deriving ACM

From INSE to ACM

ACM first was developed by A.J. Chorin in 1967 and assumes a small compressibility for the fluid and isothermal condition for the flow:

$$\rho = \rho(p) \therefore \partial_t \rho = \frac{\partial \rho}{\partial p} \frac{\partial p}{\partial t} = \frac{1}{c^2} \frac{\partial p}{\partial t}$$

where c is artificial sound speed. So INSE becomes:

$$\partial_t P + c^2 \nabla \cdot \mathbf{u} = 0$$

$$\partial_t \mathbf{u} + \mathbf{u} \cdot \nabla \mathbf{u} = -\nabla P + \nu \nabla^2 \mathbf{u} + \mathbf{g}$$

where $P = \frac{p}{\rho}$.

Remarks on ACM

- Limited scope of applicability: incompressible steady state problems.
- Small time step: t is pseudo-time step and depends on c . Therefore, if an explicit discretization method is used time step should have a small value.
- Efficient parallelization: no elliptic PDE

Although applicability of the method is very limited, ACM provides a simple yet informative framework for pedagogical purposes



Table of Content

1 Artificial Compressibility Method

- Navier-Stokes Equations
- Deriving ACM

2 Development: from scratch

- Solution Algorithm
- Grid
- Discretization

3 Development: OpenFOAM

- OpenFOAM
- Formulation

4 Problem Statement

- Workshop
- Projects Structure

Solution Algorithm

Pseudo Code

- 1 Initialization of the domain and the variables (u, v, p)
- 2 While convergence is reached do:
 - 1 Solve momentum equation in x -direction to get u
 - 2 Solve momentum equation in y -direction to get v
 - 3 Update the boundary conditions for u and v
 - 4 Solve continuity equation to get P
 - 5 Check convergence criteria
- 3 Output the fields data

Note: a flowchart is provided in the repository
(UHWorkshop/workshop3/flowchart.pdf).

Solution Algorithm

Boundary Conditions

The two common boundary conditions can be formulated as follows:

- Dirichlet: $\phi_g = 2\phi_b - \phi_i$
- Neumann: $\phi_g = \phi_i - \left(\Delta n \frac{\partial \phi}{\partial n} \right)_b$

Indices: g , b and i denote a ghost, boundary and inner node, respectively

Variables: ϕ is a quantity of interest (u , v or P) and n represents a direction normal to the boundary cell face

○
○○
○○○

○
○○●
○○○○○
○○○○○○○

○
○○

○○○
○○○○
○○○○○

Solution Algorithm

Convergence Criteria

$$1 \quad E_u = \sqrt{(\Delta t \Delta x \Delta y) \sum_{i,j} (u_{i,j}^{n+1} - u_{i,j}^n)^2}$$

$$2 \quad E_v = \sqrt{(\Delta t \Delta x \Delta y) \sum_{i,j} (v_{i,j}^{n+1} - v_{i,j}^n)^2}$$

$$3 \quad E_p = \sqrt{\frac{\Delta t \Delta x \Delta y}{c^2} \sum_{i,j} (P_{i,j}^{n+1} - P_{i,j}^n)^2}$$

$$4 \quad E_\nabla = (\Delta t \Delta x \Delta y) \nabla \cdot \mathbf{u}$$

$$5 \quad E_{total} = \max\{E_u, E_v, E_p, E_\nabla\} < \varepsilon$$

where ε is the desired tolerance.



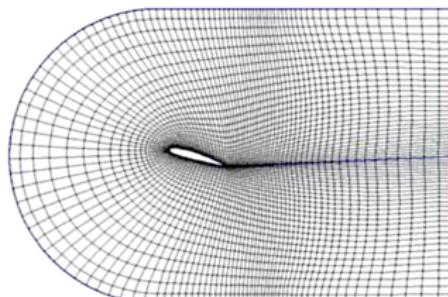
Grid



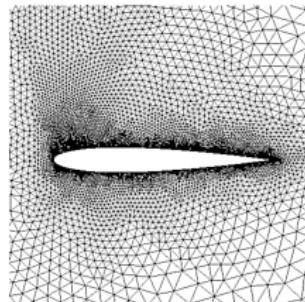
Structured Mesh (SM) vs. Unstructured Mesh (UM)

- Node connection: SM has regular connectivities while UM has irregular ones
- Quality: SM gives better results if applicable
- Memory usage: SM requires less memory and converge faster while UM requires storage of cell-to-cell pointers so take up more memory and is slower

Structured



Unstructured



○
○○
○○○

○
○○○
●●●○
○○○○○○○

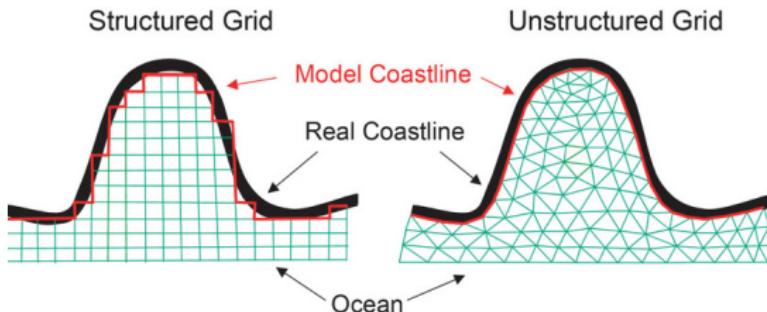
○
○○

○○○
○○○
○○○○

Grid

Structured Mesh (SM) vs. Unstructured Mesh (UM)

- Generation: UMs are usually highly automated and take less time to generate while generating SMs could be cumbersome
- Complex geometry: UM can fit complex geometries much better



Chen, C., R.C. Beardsley, and G. Cowles. 2006. An unstructured grid, finite-volume coastal ocean model (FVCOM) system. *Oceanography* 19(1):78–89, <https://doi.org/10.5670/oceanog.2006.92>.

○
○○
○○○

○
○○○
○○●○○
○○○○○○○

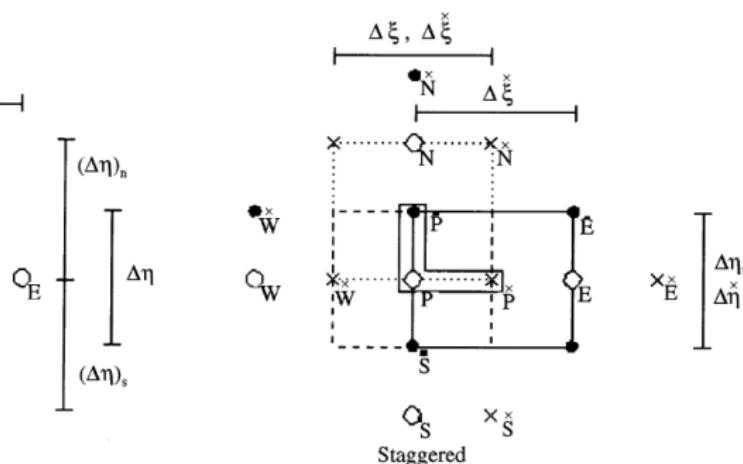
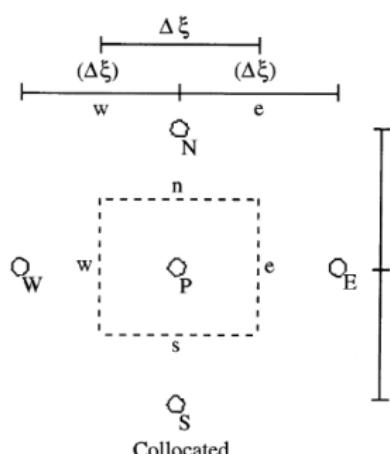
○
○○
○○○

○○○
○○○
○○○○

Grid

Staggered (SG) vs. Collocated (CG)

- Variables: for CG, all at node centers while for SG scalars at node centers and vectors at face centers



Meier, H. F., Alves, J. J. N., & Mori, M. (1999). Comparison between staggered and collocated grids in the finite-volume method performance for single and multi-phase flows. Computers & Chemical Engineering

Staggered (SG) vs. Collocated (CG)

- Accuracy: SG is more accurate while CG has issues such as checkerboard (Rhee–Chow interpolation is one of the remedies)
- Implementation: CG is easier to implement and extend

We use SG for this project:

- Simple geometry
- No additional remedy is required for implementing the numerical schemes
- Get more accurate results

○
○○
○○○

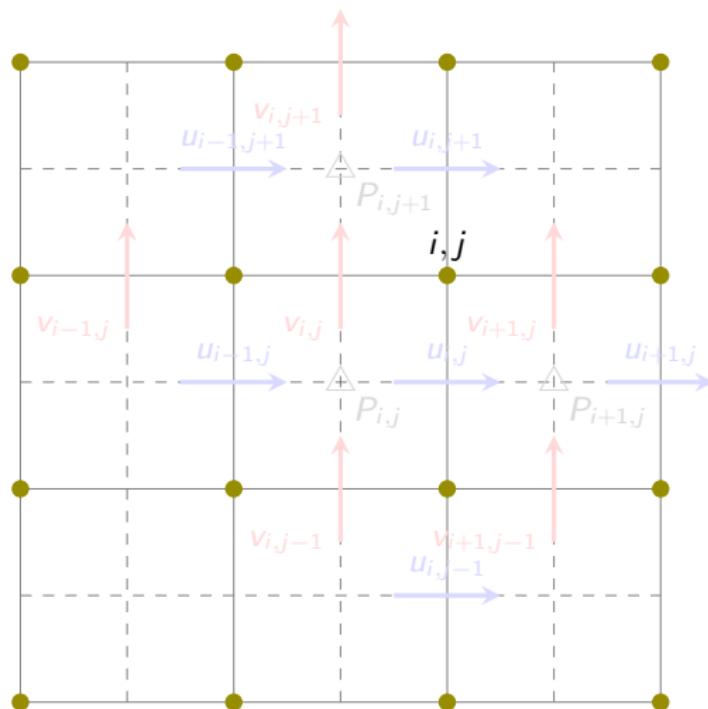
○
○○○
○○○●
○○○○○○

○
○○
○○○

○○○
○○○
○○○○

Grid

More On Staggered Grid



○
○○
○○○

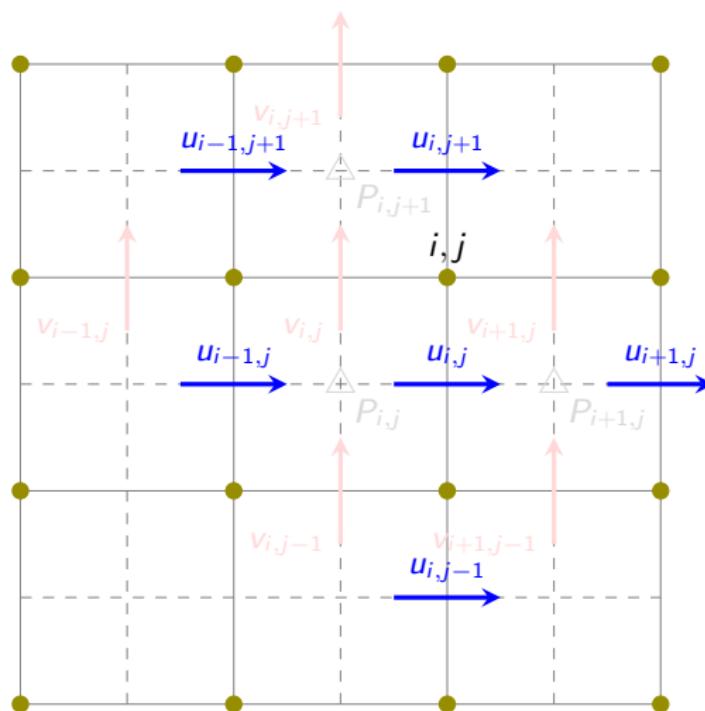
○
○○○
○○○●
○○○○○○

○
○○
○○○

○○○
○○○
○○○○

Grid

More On Staggered Grid



○
○○
○○○

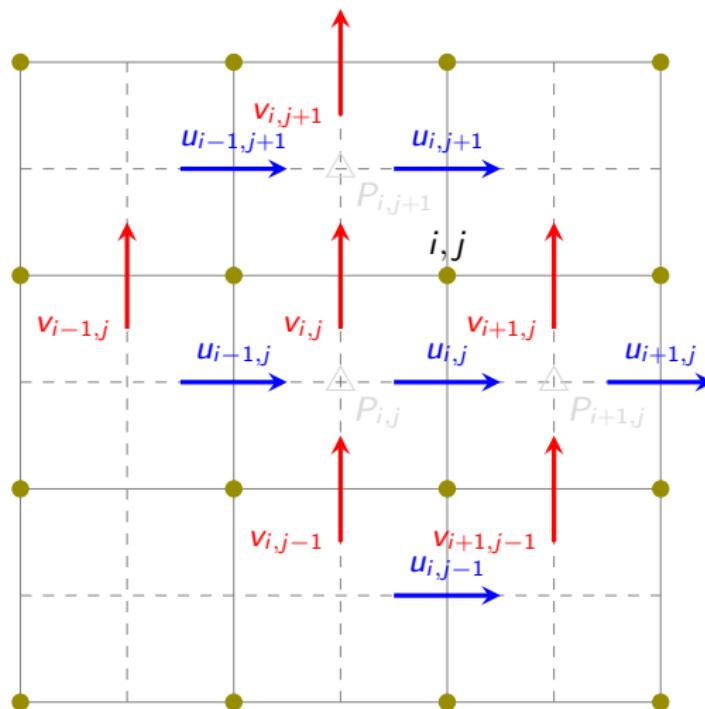
○
○○○
○○○●
○○○○○○

○
○○

○○○
○○○
○○○○

Grid

More On Staggered Grid



○
○○
○○○

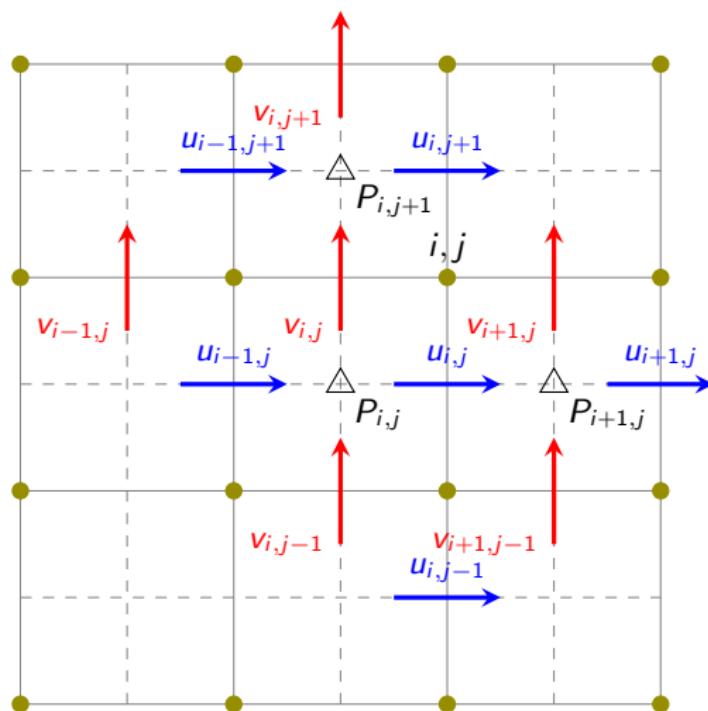
○
○○○
○○○●
○○○○○○

○
○○
○○○

○○○
○○○
○○○○

Grid

More On Staggered Grid



○
○○
○○○

○
○○○
○○○○
●○○○○○

○
○○

○○○
○○○
○○○○

Discretization

Formulation of discretized form

Spatial: central difference, second order

$$\partial_x \phi = \frac{\phi_{i+1} - \phi_{i-1}}{2\Delta x}$$

Temporal: backward difference, first order

$$\partial_t \phi = \frac{\phi^{n+1} - \phi^n}{\Delta t}$$

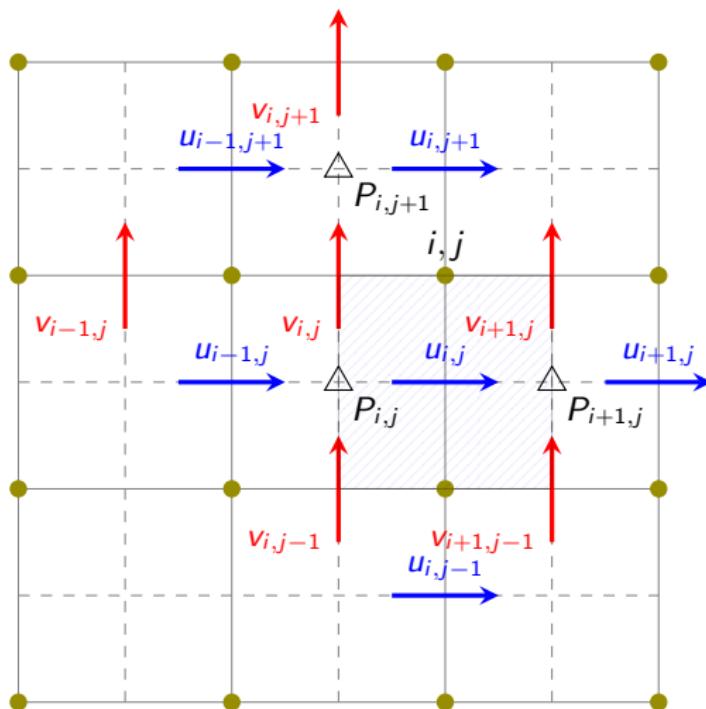
Note: for brevity we use the superscript $n + 1$ to denote the variable in the current time step and no superscript for the previous time step.

Let's discretize the equations based on a staggered grid node numbering.



Discretization

Control Volume for Momentum Equation in x -direction



Discretization

Momentum Equation in x -direction

$$\partial_t u + \partial_x(uu) + \partial_y(uv) = -\partial_x P + \nu (\partial_{xx} u + \partial_{yy} u)$$

$$\frac{u_{i,j}^{n+1} - u_{i,j}}{\Delta t} + \frac{\left(\frac{u_{i+1,j} + u_{i,j}}{2}\right)^2 - \left(\frac{u_{i,j} + u_{i-1,j}}{2}\right)^2}{\Delta x} + \frac{\left(\frac{u_{i,j+1} + u_{i,j}}{2}\right) \left(\frac{v_{i+1,j} + v_{i,j}}{2}\right) - \left(\frac{u_{i,j} + u_{i,j-1}}{2}\right) \left(\frac{v_{i+1,j-1} + v_{i,j-1}}{2}\right)}{\Delta y}$$

=

$$-\frac{P_{i+1,j} - P_{i,j}}{\Delta x} + \nu \left(\frac{u_{i+1,j} - 2u_{i,j} + u_{i-1,j}}{\Delta x^2} + \frac{u_{i,j+1} - 2u_{i,j} + u_{i,j-1}}{\Delta y^2} \right)$$

○
○○
○○○

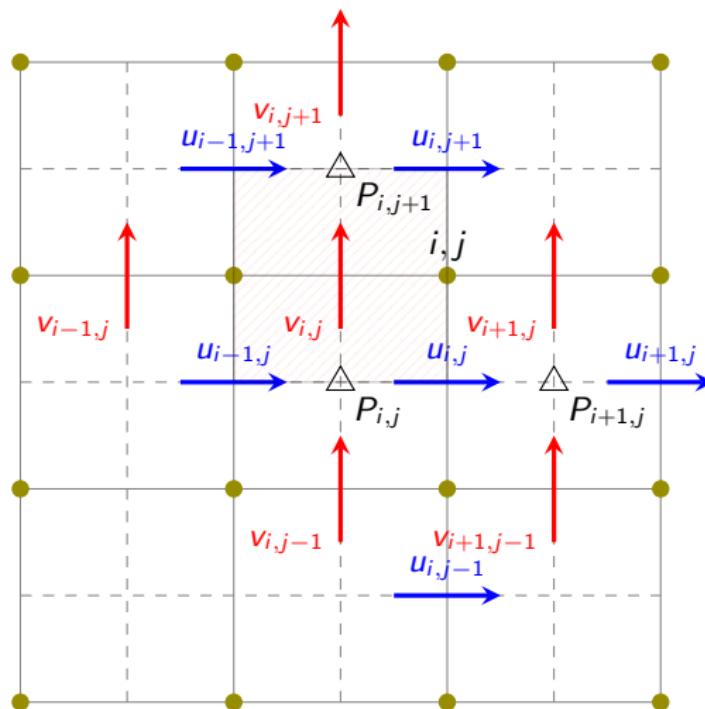
○
○○○
○○○○○
○○○●○○○

○
○○

○○○
○○○○
○○○○○

Discretization

Control Volume for Momentum Equation in y -direction



Artificial Compressibility Method

```

○
○○
○○○

```

Development: from scratch

```

○
○○○
○○○○○
○○○○●○○

```

Development: OpenFOAM

```

○
○○
○○○

```

Problem Statement

```

○○○
○○○
○○○○

```

Discretization

Momentum Equation in y -direction

$$\partial_t v + \partial_x(uv) + \partial_y(vv) = -\partial_y P + \nu (\partial_{xx} v + \partial_{yy} v)$$

$$\frac{v_{i,j}^{n+1} - v_{i,j}}{\Delta t} + \frac{\left(\frac{u_{i,j+1} + u_{i,j}}{2}\right) \left(\frac{v_{i+1,j} + v_{i,j}}{2}\right) - \left(\frac{u_{i-1,j+1} + u_{i-1,j}}{2}\right) \left(\frac{v_{i,j} + v_{i-1,j}}{2}\right)}{\Delta x}$$

$$+ \frac{\left(\frac{v_{i,j+1} + v_{i,j}}{2}\right)^2 - \left(\frac{v_{i,j} + v_{i,j-1}}{2}\right)^2}{\Delta y}$$

=

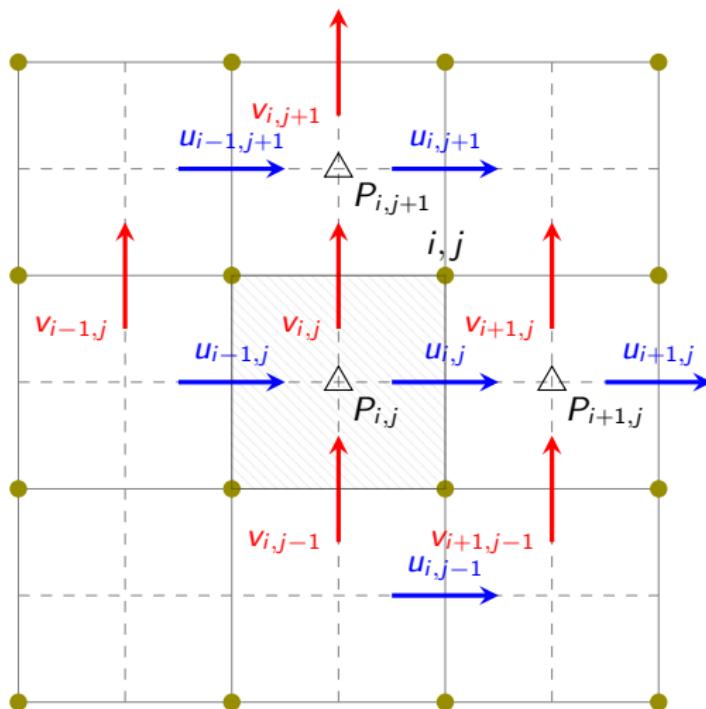
$$- \frac{P_{i,j+1} - P_{i,j}}{\Delta y}$$

$$+ \nu \left(\frac{v_{i+1,j} - 2v_{i,j} + v_{i-1,j}}{\Delta x^2} + \frac{v_{i,j+1} - 2v_{i,j} + v_{i,j-1}}{\Delta y^2} \right)$$



Discretization

Control Volume for Continuity Equation



Artificial Compressibility Method



Discretization

Development: from scratch



Development: OpenFOAM



Problem Statement



Continuity Equation

$$\partial_t P + c^2 (\partial_x u + \partial_y v) = 0$$

$$\frac{P_{i,j}^{n+1} - P_{i,j}}{\Delta t} + c^2 \left(\frac{u_{i,j}^n - u_{i-1,j}^n}{\Delta x} + \frac{v_{i,j}^n - v_{i,j-1}^n}{\Delta y} \right) = 0$$

Table of Content

1 Artificial Compressibility Method

- Navier-Stokes Equations
- Deriving ACM

2 Development: from scratch

- Solution Algorithm
- Grid
- Discretization

3 Development: OpenFOAM

- OpenFOAM
- Formulation

4 Problem Statement

- Workshop
- Projects Structure

OpenFOAM

OpenFOAM offers a framework that can be used in three levels:

- User: use it as a blackbox to perform simulations
- Dev: manipulate solvers in high-level (modifying equations etc.)
- Hacker: manipulate the underlying engine (modifying formation of matrices etc.)



Momentum Equation

$$\partial_t \mathbf{u} + \mathbf{u} \cdot \nabla \mathbf{u} = -\nabla P + \nu \nabla^2 \mathbf{u}$$

```
fvVectorMatrix UEqn
(
    fvm::ddt(U)
    +
    fvm::div(phi, U)
    +
    turbulence->divDevReff(U)
    ==
    -fvc::grad(p)
);
```

Continuity Equation

$$\partial_t P + c^2 \nabla \cdot \mathbf{u} = 0$$

```
fvScalarMatrix pEqn
(
    fvm::ddt(p)
    ==
    -c2*fvc::div(phi)
);
```

Table of Content

1 Artificial Compressibility Method

- Navier-Stokes Equations
- Deriving ACM

2 Development: from scratch

- Solution Algorithm
- Grid
- Discretization

3 Development: OpenFOAM

- OpenFOAM
- Formulation

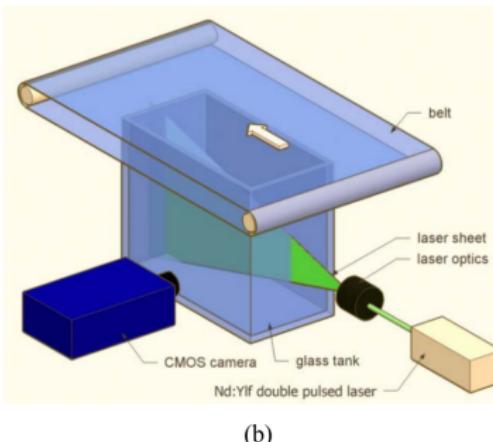
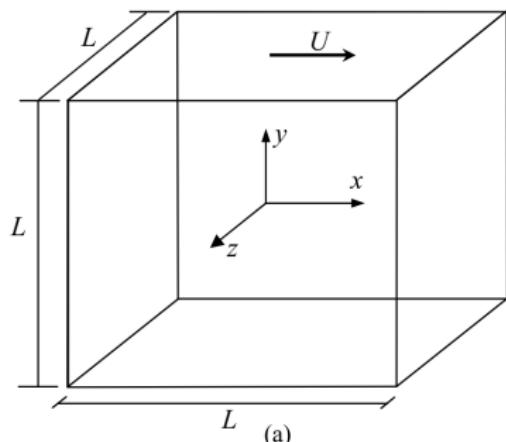
4 Problem Statement

- Workshop
- Projects Structure



Lid-Driven Cavity

- Characteristics: steady-state, incompressible
- Boundary Conditions: walls everywhere (top wall is moving)
 - Velocity: Dirichlet ($\mathbf{U} = 0$ except for the top wall where $\mathbf{U}_x = 1$ and $\mathbf{U}_y = 0$)
 - Pressure: Neumann ($\nabla p = 0$)



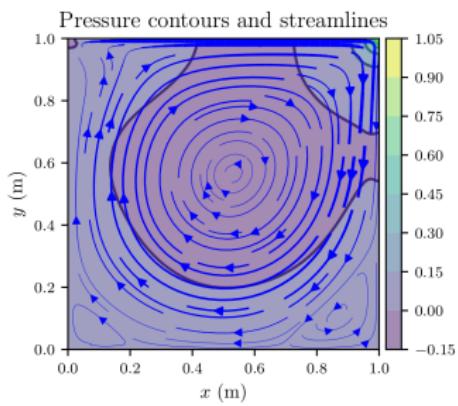
○
○○
○○○

○
○○○
○○○○○○
○○○○○○○○

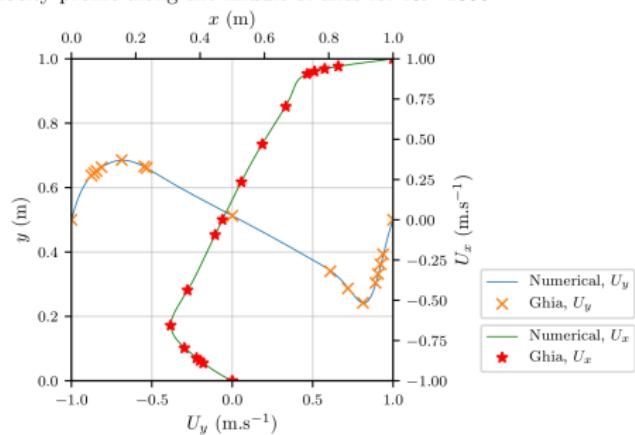
○
○○
○○○

○○●
○○○○
○○○○○

Outputs



Velocity profile along the middle of axes for $Re=1000$





Road map

A numerical solution to this benchmark problem will be presented based on:

- Finite Difference: Coding from scratch: Python, C
- Finite Volume: Modifying one of the solvers in OpenFOAM
- Finite Element: Using libraries provided by FEniCS



Exercise: Taylor–Green Vortex

A numerical solution to Taylor–Green Vortex benchmark problem:

- Initial Condition:

$$u(x, y) = \sin(x) \cos(y)$$

$$v(x, y) = -\cos(x) \sin(y)$$

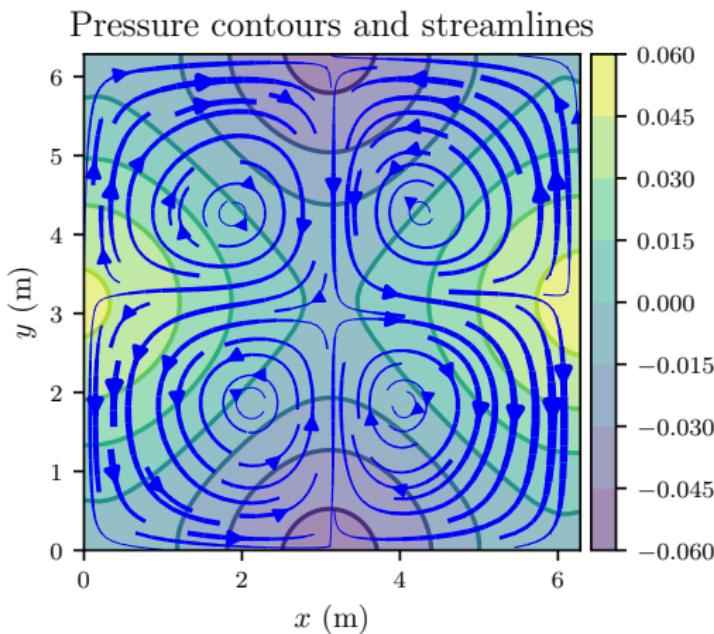
- Boundary condition: walls all around
- Parameters: $L = 2\pi$, $\nu = 0.2$, $\Delta t = 10^{-4}$ and $Re = 1000$.
- Analytical Solution:

$$u(x, y) = \sin(x) \cos(y) e^{-2\nu t}$$

$$v(x, y) = \cos(x) \sin(y) e^{-2\nu t}$$

$$p(x, y) = \frac{\rho}{4} (\cos(2x) + \cos(2y)) e^{-4\nu t}$$

Exercise: Output



Projects Structure

Tools

The following tools are required:

- Python 3.0+: Python 2 is deprecated
- Numpy: array operation and manipulation
- Numba: code speed-up
- Matplotlib: post-processing
- Flake8: PEP8 style guide enforcement
- CMake: managing the build process



Projects Structure

Folders Content

```
c
  └── bin
      └── CMakeFiles
          └── lidCavity
  └── build
      ├── CMakeCache.txt
      ├── CMakeFiles
      ├── cmake_install.cmake
      └── Makefile
  └── CMakeLists.txt
  └── data
      ├── Central_U
      ├── Central_V
      ├── residual
      └── xyuvp
  └── output
      ├── pressure_stream.pdf
      └── velocity.pdf
  └── run
  └── src
      ├── functions.h
      └── lidCavity.c
  └── log.plt
```

```
python
  └── data
      ├── Central_U
      ├── Central_V
      ├── residual
      └── xyuvp
  └── output
      ├── pressure_stream.pdf
      └── velocity.pdf
  └── run
  └── src
      ├── functions.py
      ├── functions.pyc
      └── lidCavity.py
  └── __pycache__
      └── setup.pyc
```



Final Remarks

Some other interesting open source CFD projects to look into:

- **Basilisk**: solution of partial differential equations on adaptive Cartesian meshes. (C)

<http://basilisk.fr>

- **FluidDyn**: some Python packages specialized for different tasks, in particular for 2D and 3D Fast Fourier Transforms, numerical simulations, laboratory experiments and processing of images of fluid.

<https://fluiddyn.readthedocs.io/en/latest>

- **Firedrake**: an automated system for the solution of partial differential equations using the finite element method. (Python)

<https://firedrakeproject.org>

Let's Get Started!

```
$ mkdir UHWS  
  
$ cd UHWS  
  
$ git clone https://github.com/taataam/UHWorkshop.git  
  
$ cd UHWorkshop/workshop3
```

