

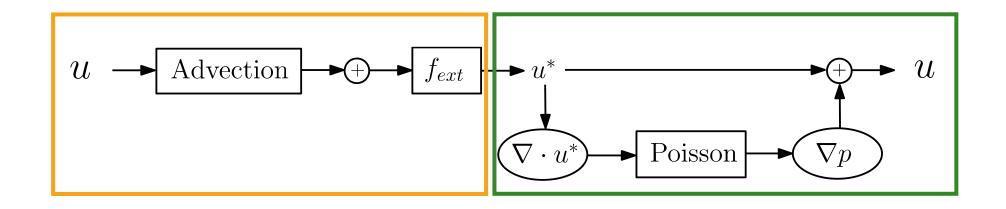
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# Fluid Solver

- Incompressible Navier Stokes
- Based on **PhiFlow** (python version of Mantaflow (Thuerey,2016))
  - Computer-vision orientated
- Two main steps:
  - Advection
  - Pressure projection

$$\begin{vmatrix} \frac{\partial u}{\partial t} = -u \cdot \nabla u - \frac{1}{\rho} \nabla p + f \\ \nabla \cdot u = 0 \end{vmatrix}$$

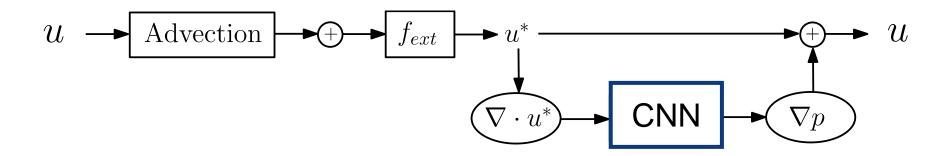


- Resolution methods for Eulerian approach of the Poisson Step (iterative methods):
  - Jacobi methods
    - $\uparrow$  iterations =  $\uparrow$  accuracy =  $\uparrow$  computational cost
  - Preconditioned Conjugate Gradient (PCG)
- It takes up to 80% of the calculation time!



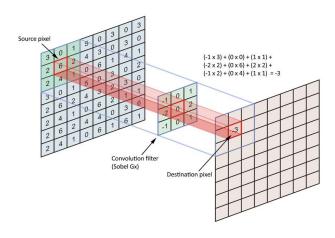
# Deep Learning

• Instead of using those methods, solve the **Poisson step** using **Convolutional Neural Networks** (following Tompson's work)

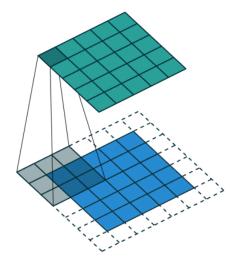


#### Convolutional layers:

- For each input we apply a series of filters of "small" size (typically kernels of 3x3)
- These filters perform a linear operation, sliding through the whole domain

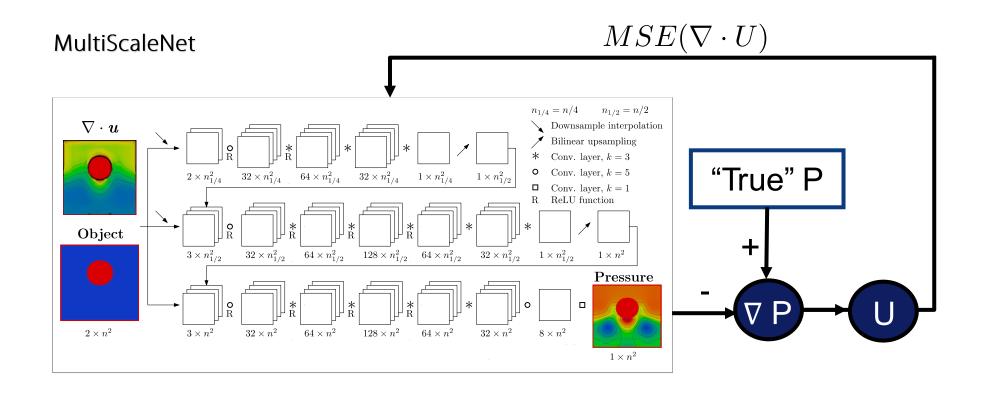




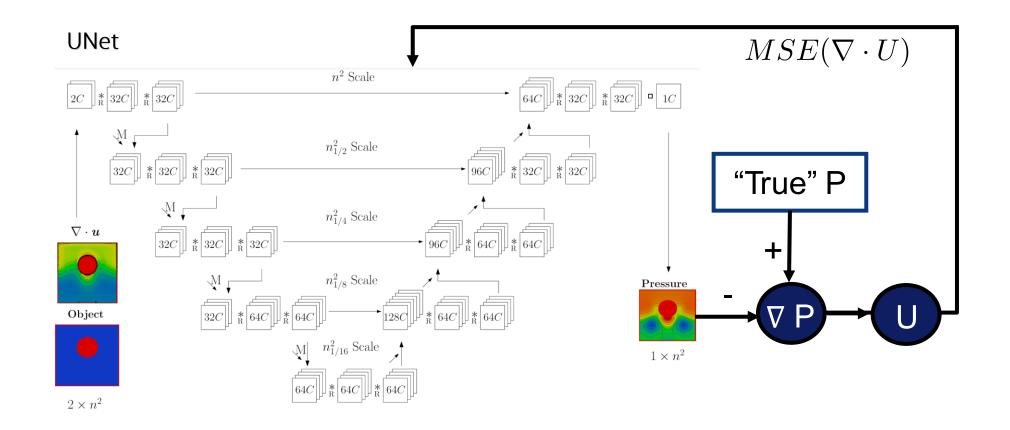


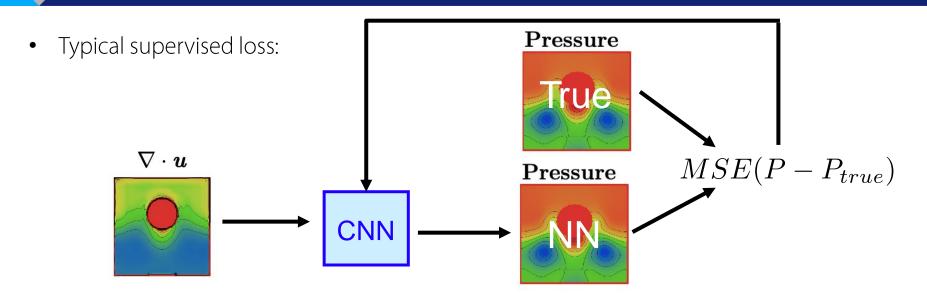
https://github.com/vdumoulin/conv\_arithmetic

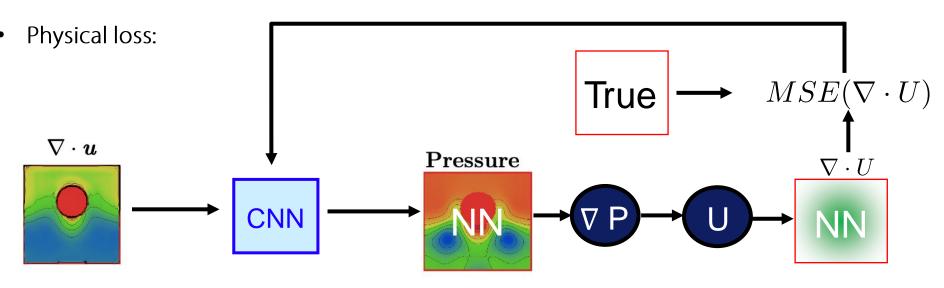
- The specific architecture chosen is:
  - MultiScaleNet (Mathieu, 2015) or UNet (Ronnenberger, 2015)
  - Account for the physical loss



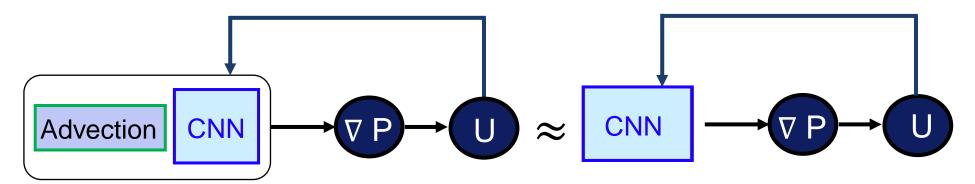
- The specific architecture chosen is:
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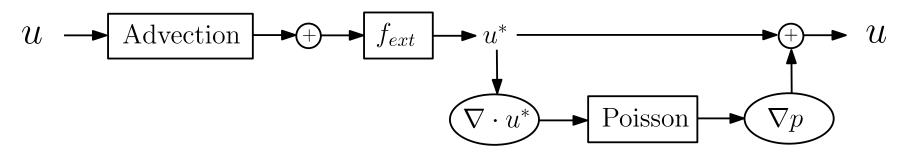




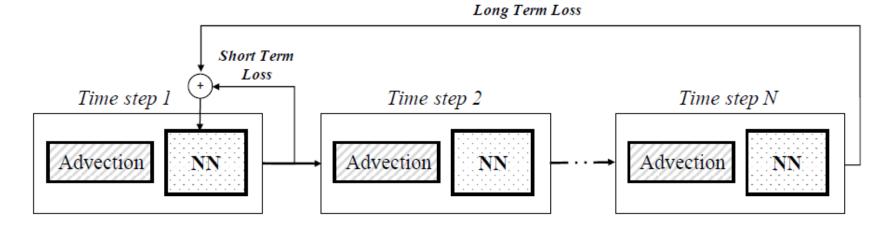
Initially, network can be treated independently to solver during training



During testing, or inference, use the regular configuration

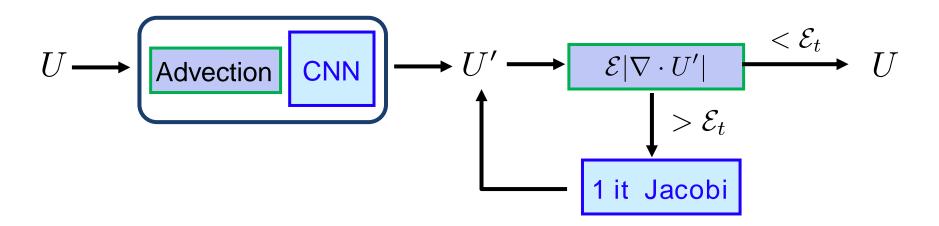


• To reduce the error of the CNN, we use the Short and Long term loss:



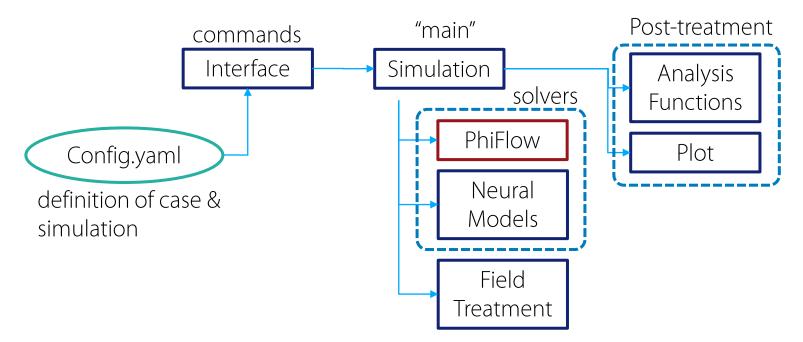
• Each being the physical loss:  $MSE(\nabla \cdot U)$ 

- How can an Al-based solver be **robust** and **reliable**?
  - Improve the learning (loss, architecture, etc)
  - Bigger database
  - New hybrid CFD-Al strategy -> Ensures reliability



# **Code Structure**

- In house-made software base on PhiFlow
- Characteristics:
  - Python based
  - GPU accelerated using torch tensors + differentiable
  - Works around PhiFlow types. Mainly StaggeredGrid and CenteredGrid
- Structure of the code:



# **Folder Structure**

#### Main Folders Structure

cases/
Ca3C3/

- > trainings/
- > neurasim/
  - > analysis/
  - > doc/
  - > engines/phi/
  - > interface/
  - neural\_models/
  - > simulation/
  - > util/
    - > operations/
    - > plot/
    - unit\_test/
    - > temp/

- -> config\_simulation.yaml + launcher.slurm
- -> saved models
- -> análisis functions
- -> documentation (under construction)
- -> phiflow
- -> commands.py, parsers, I/O files, etc
- -> training, architecture classes, etc
- -> VonKarman.py, etc
- -> field operations
- -> plot field, distribution, GIF, etc
- -> if some unit test of future utility put here on its folder
- -> temporal files

# **Configuration Files**

config\_simulation.yaml
 Type of Simulation Class

```
#########################
                                                    SIMULATION
                                                         GEOMETRY
                                                    simClass: VonKarman rotative 
                                                    #Domain
GPU: True
                                                    Lx: 224 #[m] or in mm if consistently changed
save field: False
                                                    Ly: 224 #[m]
sim_method: PHI 			 solver
                                                    #BC
#in dir: './'
                                                    BC domain x: OPEN
out dir: './results pre 1e-3/'
                                                    BC domain y: STICKY
##########################
                                                    #Cilinder
# DISCRETIZATION #
                                                    D: 15
#########################
                                                    xD: 75
Nx: 448 #[] number of control volumes in x direction
Ny: 448 #[] number of control volumes in y direction
                                                    Nt: 15000 #[] number of time steps to simulate
                                                    # PHYSICAL FORCES #
                                                    #########################
# CFL
                                                    Reynolds: 100.0
CFL: 0.2
                                                    Alpha: 1.5
```

Packaging of a meta simulation/training caller used for the iterate command.

Packaging of a meta simulation/training caller used for the iterate command.

#### **#ITERABLE CASE PARAMETERS**

Notice it is not a trivial problem since it consist of an undetermined number of nested for loops inside the previous ones. And managing the corresponding indexes, etc.

The solution adopted consist of using a recursive algorithm to create each combination possible.

# **Command Interface**

• Interface commands

#### Simulate

<u>\*</u>Usage: simulate [-key <arg>] [-long\_key <arg>]

<b>Argument Description</b>	Usage (Short)	Usage (long)	Default Value
Path of the configuration yaml files directory	-cd <'path'>	-conf_dir <'path'>	·./'
Path of the output results files directory	-co <'path'>	-out_dir <'path'>	'./results/'
Path of the input results files directory	-id <'path'>	-in_dir <'path'>	'./results/'

• Interface commands

# Analyze

<u>\*</u>Usage: analyze [-key <arg>] [-long\_key <arg>]

Argument Description	Usage (Short)	Usage (long)	Default Value
Path of the configuration yaml files directory	-cd <'path'>	-conf_dir <'path'>	·./'
Path of the output results files directory	-co <'path'>	-out_dir <'path'>	'./results/'
Path of the input results files directory	-id <'path'>	-in_dir <'path'>	'./results/'
Type of analysis to perform. Append as many as needed	-a <type></type>	- analysis_type <type></type>	_

• Interface commands

#### Train

<u>\*</u>Usage: train [-key <arg>] [-long\_key <arg>]

Argument Description	Usage (Short)	Usage (long)	Default Value
Path of the configuration yaml files directory	-cd <'path'>	-conf_dir <'path'>	·./'
Path of the output results files directory	-co <'path'>	-out_dir <'path'>	'./results/'
Path of the input results files directory	-id <'path'>	-in_dir <'path'>	'./results/'

Interface commands

#### Iterate

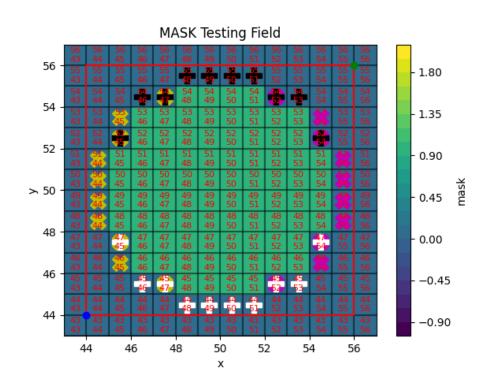
<u>\*</u>Usage: iterate [-key <arg>] [-long\_key <arg>]

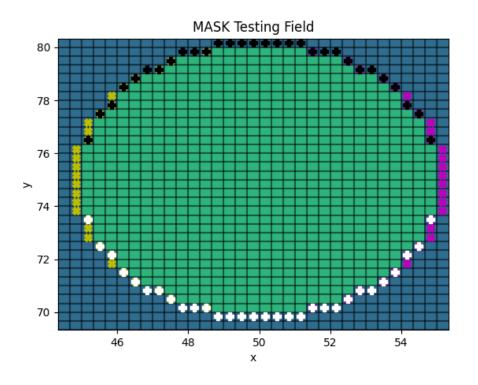
Argument Description	Usage (Short)	Usage (long)	Default Value
Path of the configuration yaml files directory	-cd <'path'>	-conf_dir <'path'>	·./'
Path of the output results files directory	-co <'path'>	-out_dir <'path'>	'./results/'
Path of the input results files directory	-id <'path'>	-in_dir <'path'>	'./results/'
Callable(function, command) or script to iterate over	-е <callable></callable>	-execute <callable></callable>	simulate

<sup>&</sup>lt;u>\*</u>Example: iterate -e "analyze -a velocity"

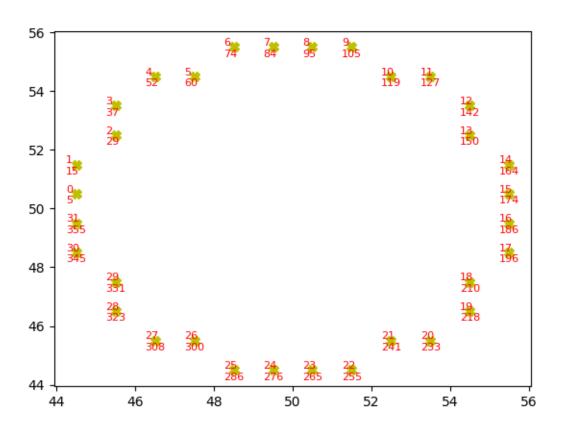
# **Field Operate Functions**

• get\_exterior\_edges(object\_mask)



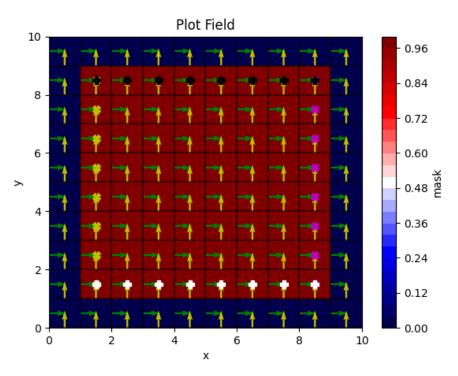


get\_line\_distribution(object\_mask=None, edges=None)



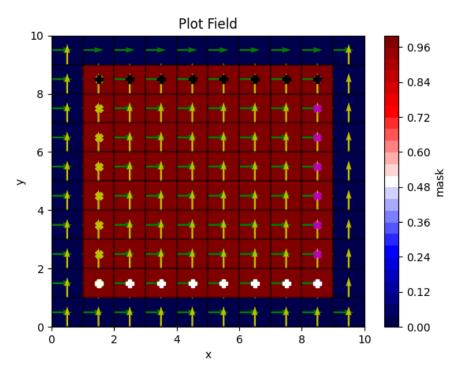
# node degrees

set\_normal\_bc()



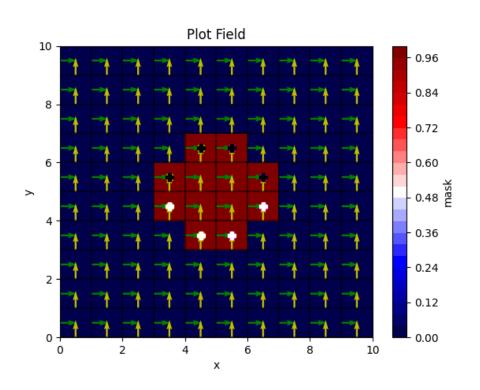
**Input** velocity field + object mask

NOTICE: there are two types of set\_XX\_bc(). Indicated by a 2 at the end. Being the later more efficient but with more inputs precomputed.

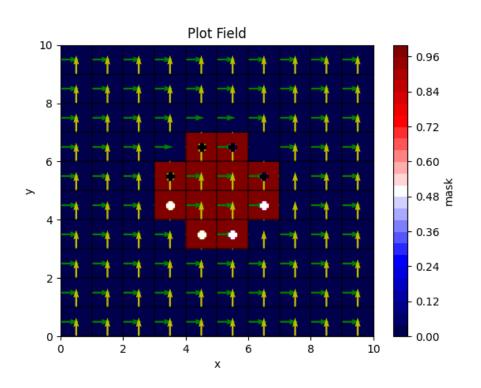


Output velocity field + object mask

set\_normal\_bc()

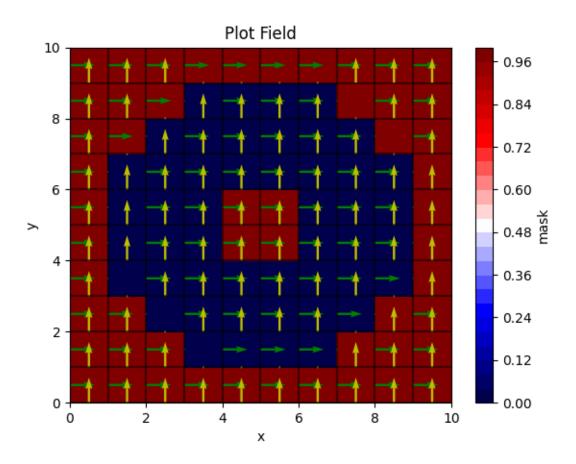


**Input** velocity field + object mask

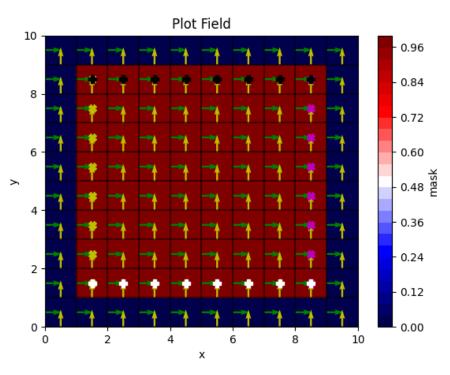


Output velocity field + object mask

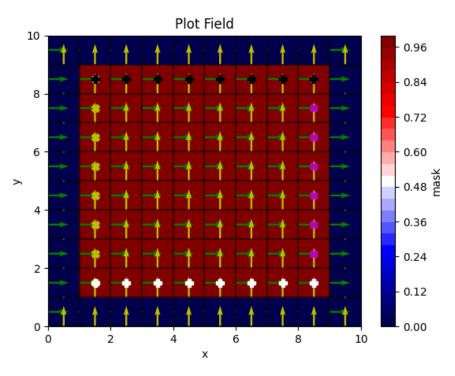
set\_normal\_bc() for **Inverse geometry** 



set\_tangential\_bc()

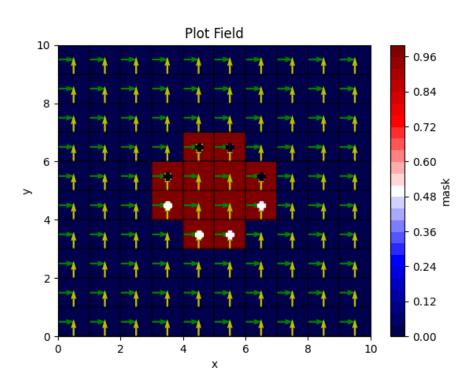


**Input** velocity field + object mask

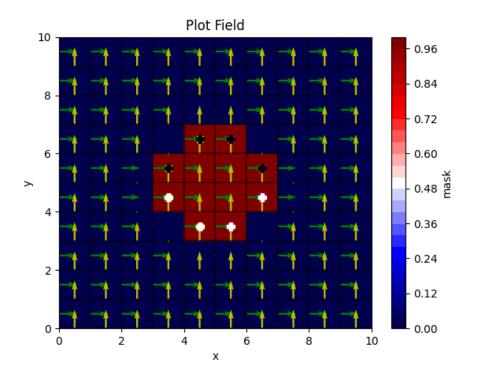


Output velocity field + object mask

set\_tangential\_bc()

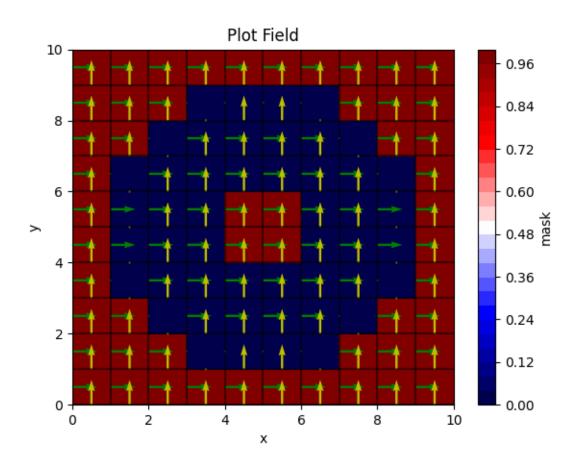


**Input** velocity field + object mask

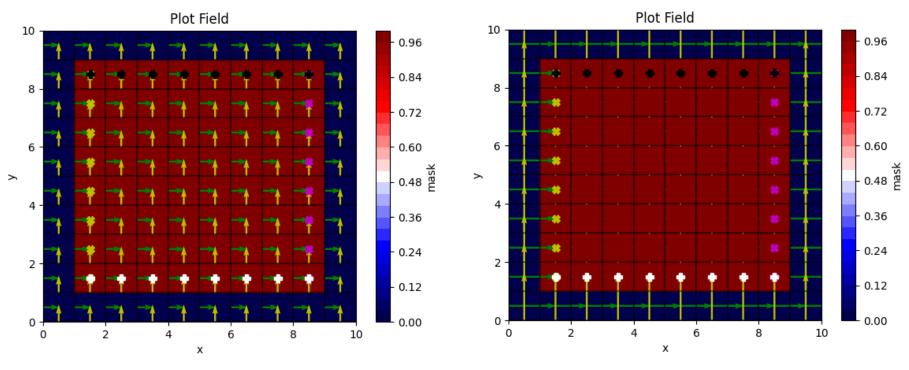


Output velocity field + object mask

set\_tangential\_bc() for Inverse geometry



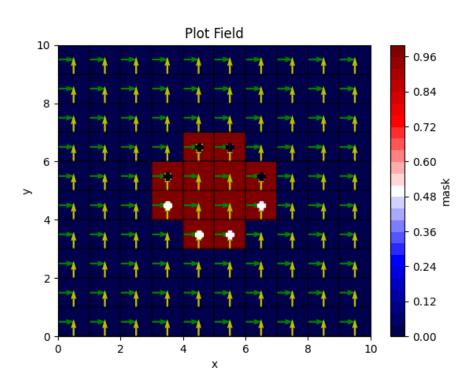
set\_internal\_bc()



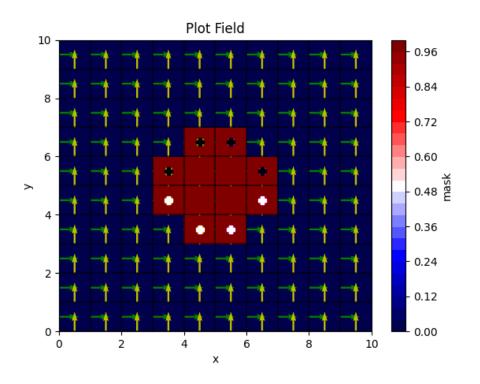
**Input** velocity field + object mask

Output velocity field + object mask

set\_internal\_bc()

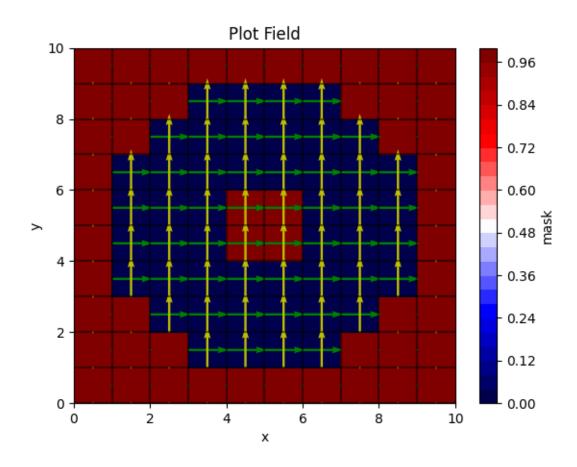


**Input** velocity field + object mask



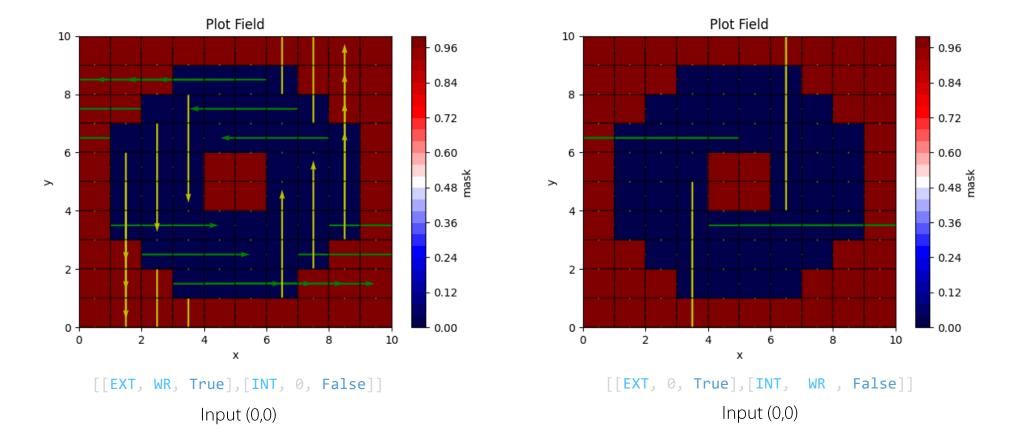
Output velocity field + object mask

• set\_internal\_bc() for Inverse geometry

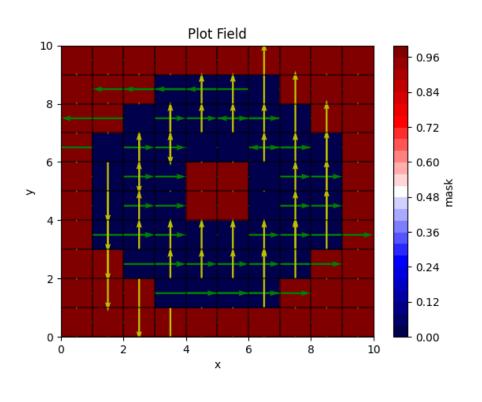


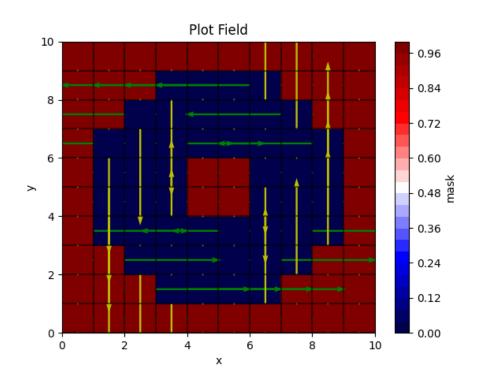
• get\_obstacles\_bc(obstacles)

bool obstacles: [ [obstacle\_mask, 'wr', 'inv\_geom'], [obstacle\_mask, 'wr', 'inv\_geom'], ...]



• get\_obstacles\_bc(obstacles)



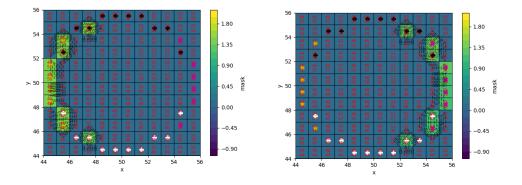


[[EXT, WR, True],[INT, 0, False]] [[EXT, WR, True],[INT,-0.5\*WR, False]]

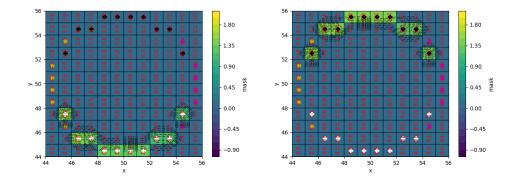
Input (1,1)

## **Analyze Functions**

- calculate\_forces(pressure, object\_mask, dx, dy)
  - > Force\_h = np.sum( pressure[left] pressure[right] )\*dy



> Force\_v = np.sum( pressure[bottom] - pressure[top] )\*dx



• calculate\_forces\_with\_momentum(pressure, velocity, object\_mask, factor=1, rho=1, dx=1, dy=1)

# **Plot Functions**

plot\_field()

plot\_field(field, plot\_type=['surface'], options=[], Lx=None, Ly=None, dx=None, dy=None, lx='x', ly='y', lbar='field', ltitle='Plot Field', save=False, filename='./field.png', fig=None, ax=None)

```
general_plot_options:
    -edges -> [ [edge_hl_x, edge_hl_y], [edge_hr_x, edge_hr_y], [edge_vb_x, edge_vb_y], [edge_vt_x, edge_vt_y] ]
    -square -> [x1,x2,y1,y2]
    -aux_contourn -> True/False

    -limits -> [min, max]
    -vector_axis -> 0/1

    -indeces -> True/False
    -grid -> True/False
    -velocity -> StaggeredGrid

    -zoom_position -> [x1,x2,y1,y2]
    -full_zoom -> True/False
```

• GIF

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