

# **High Performance Computing**

# Coursework 2



#### **Problem Domain**



- Coursework taken from the field of Computational Fluid Dynamics (CFD)
- ☐ Fluid dynamics based on three fundamental principles: (i) mass is conserved; (ii) Newton's second law; (iii) energy is conserved
- Expressed as partial differential equations, showing how velocity and pressure are related, etc. (called governing equations).

# **Governing Equations**





#### Navier-Stokes Equations 3 - dimensional - unsteady

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Coordinates: (x,y,z)

Velocity Components: (u,v,w)

Time: t Pressure: p

Heat Flux: q

Density: p

Stress: τ

Reynolds Number: Re

Total Energy: Et

Prandtl Number: Pr

$$\frac{\partial \rho}{\partial t} + \frac{\partial (\rho u)}{\partial x} + \frac{\partial (\rho v)}{\partial y} + \frac{\partial (\rho w)}{\partial z} = 0$$

X – Momentum: 
$$\frac{\partial(\rho u)}{\partial t} + \frac{\partial(\rho u^2)}{\partial x} + \frac{\partial(\rho uv)}{\partial y} + \frac{\partial(\rho uv)}{\partial z} = -\frac{\partial p}{\partial x} + \frac{1}{Re_r} \left[ \frac{\partial \tau_{xx}}{\partial x} + \frac{\partial \tau_{xy}}{\partial y} + \frac{\partial \tau_{xz}}{\partial z} \right]$$

Y - Momentum: 
$$\frac{\partial(\rho v)}{\partial t} + \frac{\partial(\rho u v)}{\partial x} + \frac{\partial(\rho v^2)}{\partial y} + \frac{\partial(\rho v w)}{\partial z} = -\frac{\partial p}{\partial y} + \frac{1}{Re_r} \left[ \frac{\partial \tau_{xy}}{\partial x} + \frac{\partial \tau_{yy}}{\partial y} + \frac{\partial \tau_{yz}}{\partial z} \right]$$

Z - Momentum 
$$\frac{\partial(\rho w)}{\partial t} + \frac{\partial(\rho uw)}{\partial x} + \frac{\partial(\rho vw)}{\partial y} + \frac{\partial(\rho w^2)}{\partial z} = -\frac{\partial p}{\partial z} + \frac{1}{Re_x} \left[ \frac{\partial \tau_{xz}}{\partial x} + \frac{\partial \tau_{yz}}{\partial y} + \frac{\partial \tau_{zz}}{\partial z} \right]$$

$$\begin{split} \frac{\partial (E_T)}{\partial t} + \frac{\partial (uE_T)}{\partial x} + \frac{\partial (vE_T)}{\partial y} + \frac{\partial (wE_T)}{\partial z} &= -\frac{\partial (up)}{\partial x} - \frac{\partial (vp)}{\partial y} - \frac{\partial (wp)}{\partial z} - \frac{1}{Re_r Pr_r} \left[ \frac{\partial q_x}{\partial x} + \frac{\partial q_y}{\partial y} + \frac{\partial q_z}{\partial z} \right] \\ &+ \frac{1}{Re_r} \left[ \frac{\partial}{\partial x} (u \, \tau_{xx} + v \, \tau_{xy} + w \, \tau_{xz}) + \frac{\partial}{\partial y} (u \, \tau_{xy} + v \, \tau_{yy} + w \, \tau_{yz}) + \frac{\partial}{\partial z} (u \, \tau_{xz} + v \, \tau_{yz} + w \, \tau_{zz}) \right] \end{split}$$

#### **Problem Domain**



- Coursework taken from the field of Computational Fluid Dynamics (CFD)
  - ☐ Fluid dynamics based on three fundamental principles: (i) mass is conserved; (ii) Newton's second law ;(iii) energy is conserved
  - Expressed as partial differential equations, showing how velocity and pressure are related, etc. (called governing equations).
  - ☐ The coordinates and time are independent variables while velocity and pressure are dependent variables
- Computational Fluid Dynamics is the science of finding the numerical solution to the governing equations of fluid flow, over the discretized space or time

#### CFD code





The code in the coursework, called Karman, calculates the velocity and pressure of a 2D flow

The code writes the solution values into a binary file

**Currently the code is sequential** 

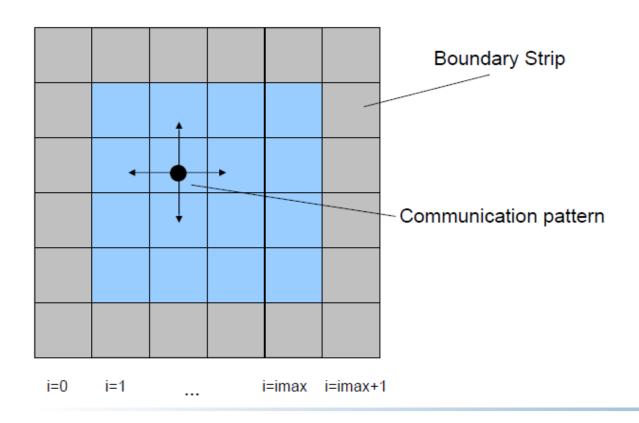
The purpose of the coursework is to parallelize the code

#### **Data and stencil**



The area represented as a 2D Grid (discretize)

#### Calculate one point in each cell



#### **CFD** code





C_B	С_В	С_В	С_В	С_В	С_В	C_B	С_В	С_В	С_В	С_В
C_B	C_F	C_F	C_F	C_F	C_F	C_F	C_F	C_F	C_F	C_B
C_B	C_F <del>*</del>	EQN	→C_F	C_F	С_В	C_F	C_F	C_F	C_F	С_В
C_B	C_F	C_F	C_F	C_F	C_B	C_F	C_F	C_F	C_F	С_В
C_B	C_F	C_F	C_F	C_F	C_F	C_F	C_F	C_F	C_F	С_В
C_B	С_В	С_В	С_В	С_В	С_В	C_B	С_В	С_В	C_B	С_В

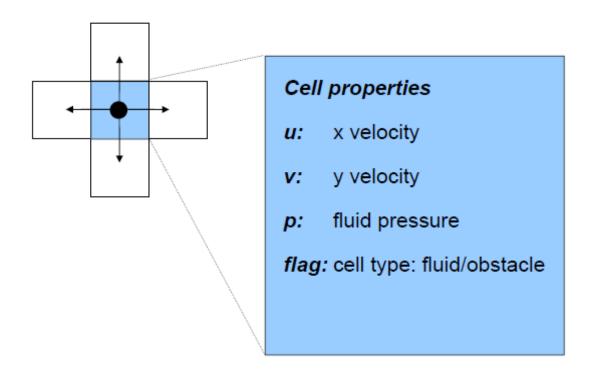
jmax

imax

There are two types of cell: fluid (C\_F) and obstacle (C\_B)

#### **Data and stencil**

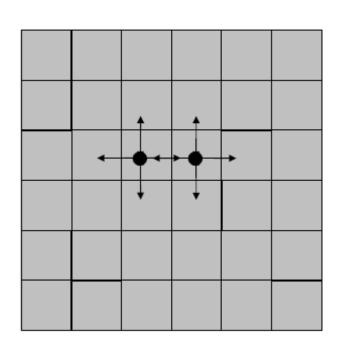




- The properties of a cell in the grid
- Communication pattern: using a five-point stencil to calculate a point

# Numerical method for solving the governing equations





Initialize each cell value

Check if the solution satisfies the governing equations

If not, generate the new solution based on a stencil of current solutions from neighbouring cells

Iterations are advanced until the termination condition is met

# General Steps for using the numerical method to solve the linear equations



- → Aim: solve AΦ=B
- Step 1: Guess an initial solution Φ⁰
- Step 2: If the convergence is reached by checking the residual B-AΦ<sup>i</sup><tolerance, then return the correct solution;
- □ Step 3: For  $\Phi^i$  (i>=0),  $\Phi^{(i+1)} = f(\Phi^{(i)})$ , go to Step 2

e.g., 
$$f(\Phi^{(i)}) = \Phi^{(i)} + (B - A\Phi^{(i)})$$

Key: when we repeat iterative steps, each step generates a better solution

## **Successive Over-Relaxation(SOR)**

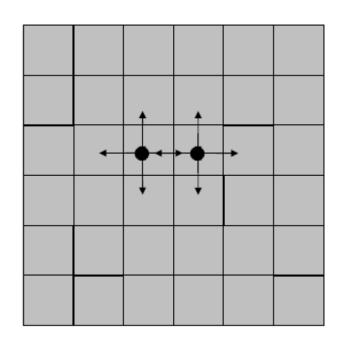


- SOR is a method to generate new solutions: it can speed up convergence
- For a set of linear equations: AΦ=b
- □ let A=D+L+U, where D, L and U denote the diagonal, strictly lower triangular, and strictly upper triangular parts of A, respectively
- → The successive over-relaxation (SOR) iteration is defined by the following

$$(D + \omega L)\phi^{(k+1)} = (-\omega U + (1 - \omega)D)\phi^{(k)} + \omega b.$$
 (\*)

■ Where values of ω > 1 are used to speedup convergence of a slow-converging process, while values of ω < 1 are often help establish convergence of diverging iterative process

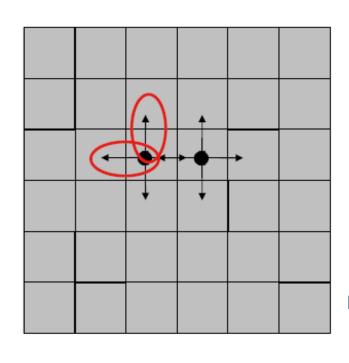




Successive Over-Relaxation (SOR) method is used to speed up convergence

$$(D + \omega L)\phi^{(k+1)} = (-\omega U + (1 - \omega)D)\phi^{(k)} + \omega b.$$
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 SOR is originally written so that p\_new(i, j) depends on

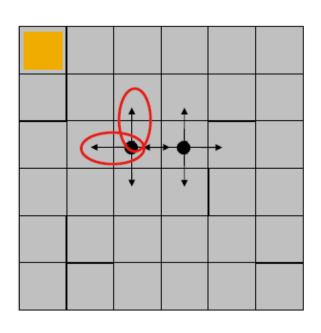
- 1) a stencil of current solutions from neighbouring cells,
- 2) p\_new(i-1, j),
- 3) p\_new(i, j-1).

for all 
$$i \le \text{imax}$$
,  $j \le \text{jmax}$ , do  $p_{ij}^{(k+1)} = f(P^{(k)}, p_{i-1,j}^{(k+1)}, p_{i,j-1}^{(k+1)})$ 

 $P^{(k)}$  is the solution at the iteration k (P is a matrix  $[p_{ij}]$ )

- This introduces a chain of dependencies for each cell
- Fine on sequential execution:
   just calculate values from left to
   right, and top to bottom





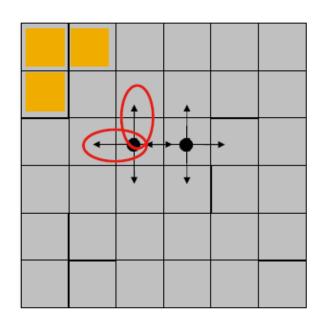
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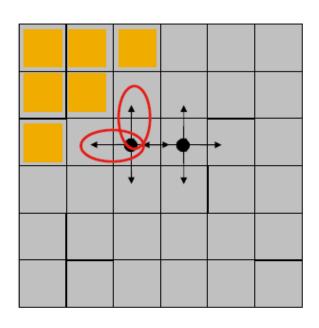
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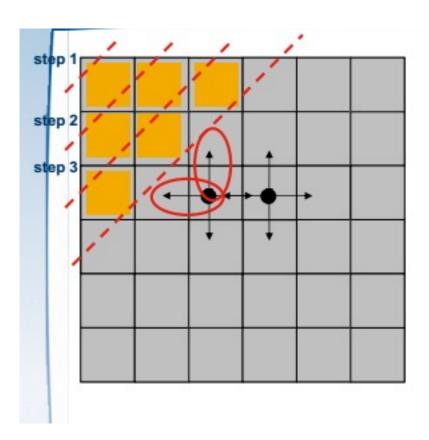
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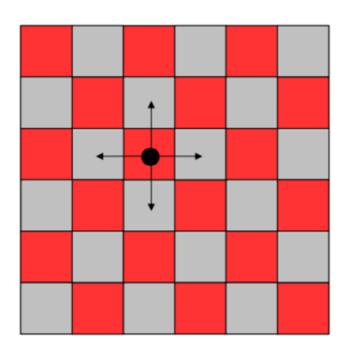




- This approach forms a "wave front"
- It does not make for a good parallel solution
- The fundamental reasons are
  - A chain of dependencies
  - The number of cells to be computed varies in each step (unbalanced workload)

#### Red/Black SOR





- **❖Less dependency**
- Workload can be balanced among the processors

- Another solution that achieves the same result is red/black ordering
- Rewrite the SOR equations so the cells are divided into two populations.
- New red values are calculated first (using current black values)

# **Red cells:** $p_{ii}^{(k+1)} = f(p_{i-1,i}^{(k)}, p_{i,i-1}^{(k)}, p_{i+1,i}^{(k)}, p_{i,i+1}^{(k)})$

 Then black is calculated using the just updated red values.

#### Black cells:

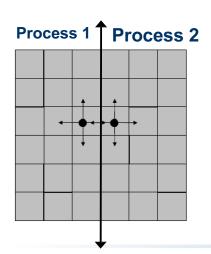
$$p_{ij}^{(k+1)} = f(p_{i-1,j}^{(k+1)}, p_{i,j-1}^{(k+1)}, p_{i+1,j}^{(k+1)}, p_{i,j+1}^{(k+1)})$$
  
Note:  $p_{i-1,j}^{(k+1)}, p_{i,j-1}^{(k+1)}, \dots$  are red cells

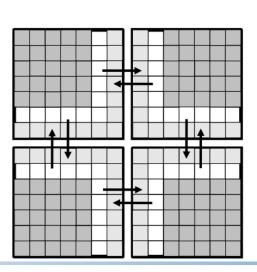
- All red cells can be calculated in parallel,
- Then all black cells can be calculated in parallel

### Decomposition of a grid of cells



- Make each process responsible for updating a block of cells
- A process must send the cells at the edge of its domain to its neighbours, and receive a copy of the edge cells from its neighbours
- 1D decomposition vs. 2D decomposition
  - 1D is OK for this coursework





## The main loop



for 
$$(t = 0.0; t < t_end; t = t + delt)$$
 {

- 1.Calculate an approximate time-step size by seeing how much movement occurred in the last time-step. The discrete approximation is only stable when the maximum motion < 1 cell per time-step.
- 2.For each cell, compute a tentative new velocity (f,g) based on the previous (u,v) values. It takes as input the u, v and flag matrices, and updates the f and g matrices.
- 3.For each cell calculate the RHS of the pressure equation (Poisson equation). This uses two f and g values. It takes as input the f, g, and flag matrices and updates the RHS matrix.
- 4.For the entire pressure matrix, use Red/Black SOR to solve the Poisson equation. This takes a large number iterations of the Red/Black process as shown in the slides. It takes as input the current pressure matrix, flag matrix, and the RHS matrix and outputs a new pressure matrix

### The main loop



- 5. For each cell, update the real (u,v) values based on the pressure matrix and the tentative velocity values (f,g). It takes as input the pressure, f, g and flag matrices and updates the u and v matrices.
- 6. For each cell that is adjacent to an edge cell, the *(u,v)* values of the boundary cells are updated (by taking values from their neighbours). It takes as input the *u, v* and flag matrices and updates the *u* and *v* matrices

Note that it is not going to be worth parallelizing the whole program, just parts of it.

#### Coursework 2 - What to Do in General



- You will be given a serial program, called Karman
- You need to parallelize the Karman code and write a report
- 80% of the full mark comes from the parallelization using MPI
- 20% of the mark comes from the parallelization using both MPI and OpenMP (a hybrid approach), i.e., parallelizing the computations in one machine using OpenMP, while parallelizing the computations across machines using MPI

#### **Coursework 2 – Detailed Requirements for Coding**



- Parallelize the Karman code, using a pure MPI approach (80% of marks) and a hybrid MPI-OpenMP approach (20% marks)
- Profiling which Karman functions are more time consuming
- Design your decomposition strategy, data exchange strategies between neighboring partitions, the MPI functions (e.g., collective operations)
- For a hybrid approach, add OpenMP directives

#### **Coursework 2 – Detailed Requirements for Coding**

- WARWICI
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- Benchmarking the execution time (e.g., using MPI\_Wtime function) of your parallel code as you
  - increase the number of processes
  - and/or change the problem size if it helps you observe the trend
- If you develop a hybrid MPI-OpenMP implementation,
  - benchmark the runtime of the hybrid parallel version
  - Compare the performance between the hybrid approach and the pure MPI approach
- Ensure the simulation results are the same after the parallelization
- Contain adequate comments as good programming practice.

#### **Coursework 2 – Requirements for Report**



- Profiling the functions to determine which functions are more time consuming; discuss the profiling results
- Discuss your MPI implementation, for example,
  - your decomposition strategy and load balancing strategy;
  - your strategy of exchanging the boundary data between neighboring partitions;
  - MPI functions such as collective operations you used;
- Benchmark the change in execution times of your parallel program as you increase the number of processes and/or problem size; present the results in graph or table
- Analyze and discuss the performance of your parallel program

#### **Coursework 2 – Requirements for Report**



- If you develop the hybrid MPI-OpenMP implementation,
  - Discuss OpenMP implementation
  - Benchmark the runtime of the hybrid parallel code and present the results
  - Discuss the performance comparison between the hybrid approach and the pure MPI approach
- Writing skills: pay attention to spelling, punctuation and grammar
- Up to 6 A4 pages

#### Coursework 2 – Submission



- Submit a compressed folder (.zip) to Tabula; the folder should contain
  - your parallel code (the directory structure used in the project should not change)
  - your report (pdf format and place your report in the top level directory in the folder)
- Deadline: 12 noon, March 16<sup>th</sup>, 2023