

# **GPU Computing with CUDA**

## **Lecture 9 - Applications - CFD**

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# Outline of lecture

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- ▶ Overview of CFD
  - Navier Stokes equations
  - Types of problems
  - Discretization methods
- ▶ “Conventional” CFD
- ▶ Port CFD codes to CUDA
- ▶ Efforts
- ▶ Example problem: implicit heat transfer

# CFD - Introduction

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- ▶ Numerical modeling of fluid systems
- ▶ Navier-Stokes equation: momentum conservation

$$\frac{\partial \mathbf{u}}{\partial t} + (\mathbf{u} \cdot \nabla) \mathbf{u} = -\frac{\nabla p}{\rho} + \nu \nabla^2 \mathbf{u}$$

- ▶ Type of problems:
  - Incompressible
  - Compressible (non-viscous approximation)
  - Shallow water
  - Biphasic flows.....

# CFD - Introduction

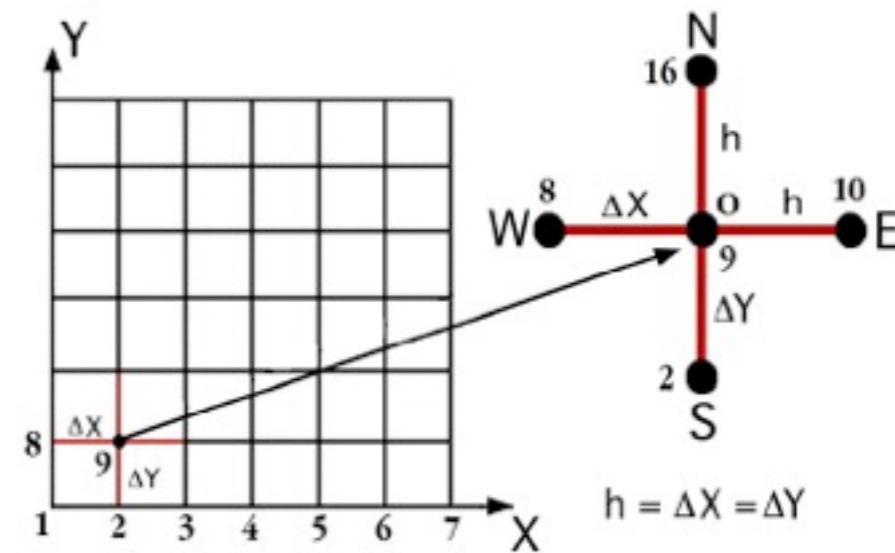
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- ▶ Earliest: Richardson (1910)
  - Human computers
  - Quickest averaged 2000 operations a week
- ▶ CFD development tied with computers!
  - 50s-60s: use of digital computers, finite difference methods
  - 70s: finite element methods, spectral methods
  - 80s: finite volume methods
  - 90s: application to diverse industries

# CFD - Main discretization methods

## ► Finite difference

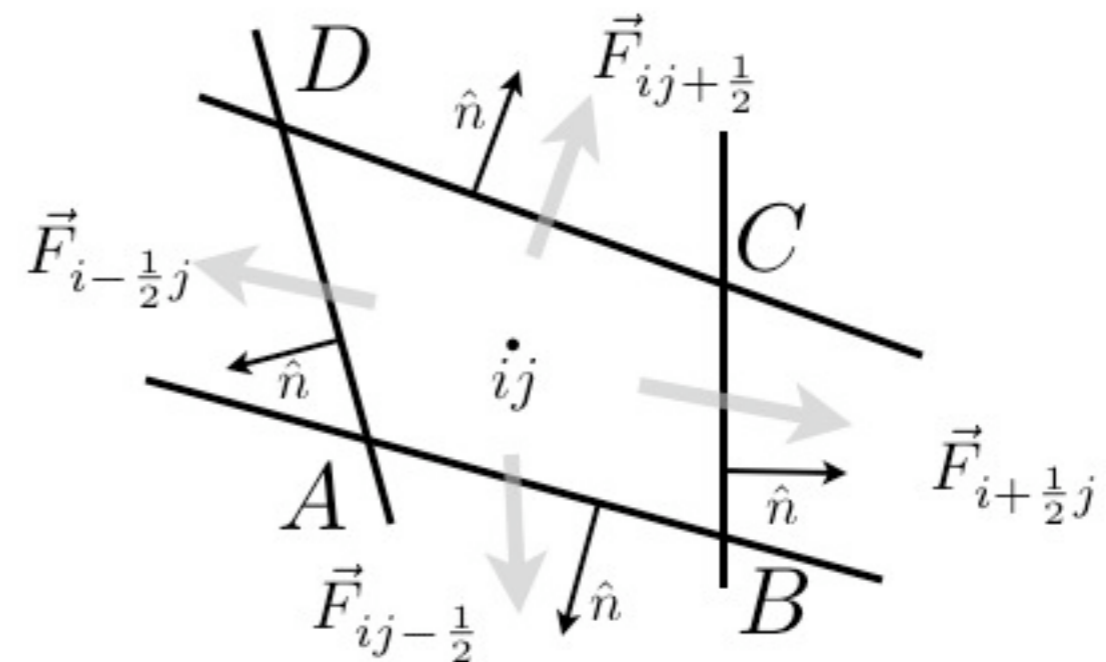
$$\frac{\partial u}{\partial x} = \frac{u_{i+1,j} - u_{i,j}}{\Delta x}$$



## ► Finite volume

$$\frac{\partial \vec{U}}{\partial t} + \frac{\partial \vec{F}}{\partial x} = 0$$

$$\frac{\partial}{\partial t} \int_{\Omega} \vec{U} d\Omega + \oint_{\partial\Omega} \vec{F} \hat{n} ds = 0$$



# CFD - Main discretization methods

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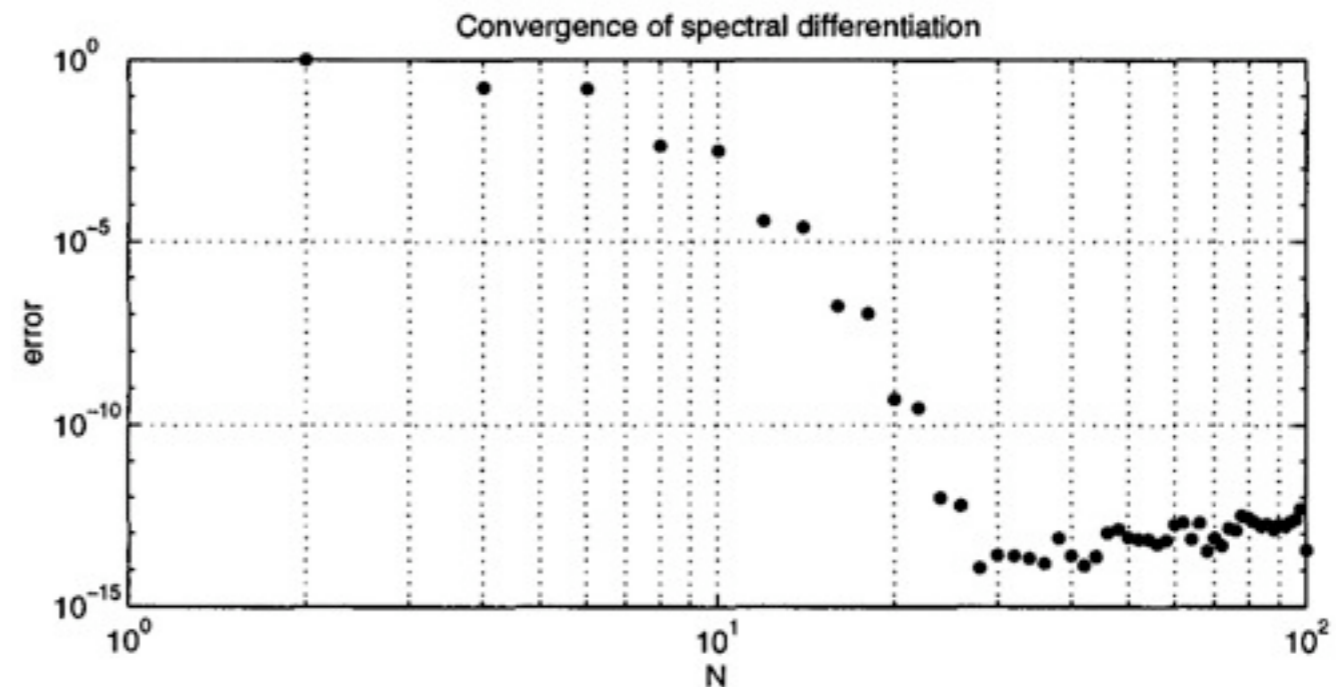
- ▶ Finite element method

$$\int_{\Omega} \nabla u \cdot \nabla v \, dx = \int_{\Omega} f v \, dx.$$



- ▶ Spectral methods

$$\widehat{\frac{\partial u}{\partial x}} = ik\hat{u}$$



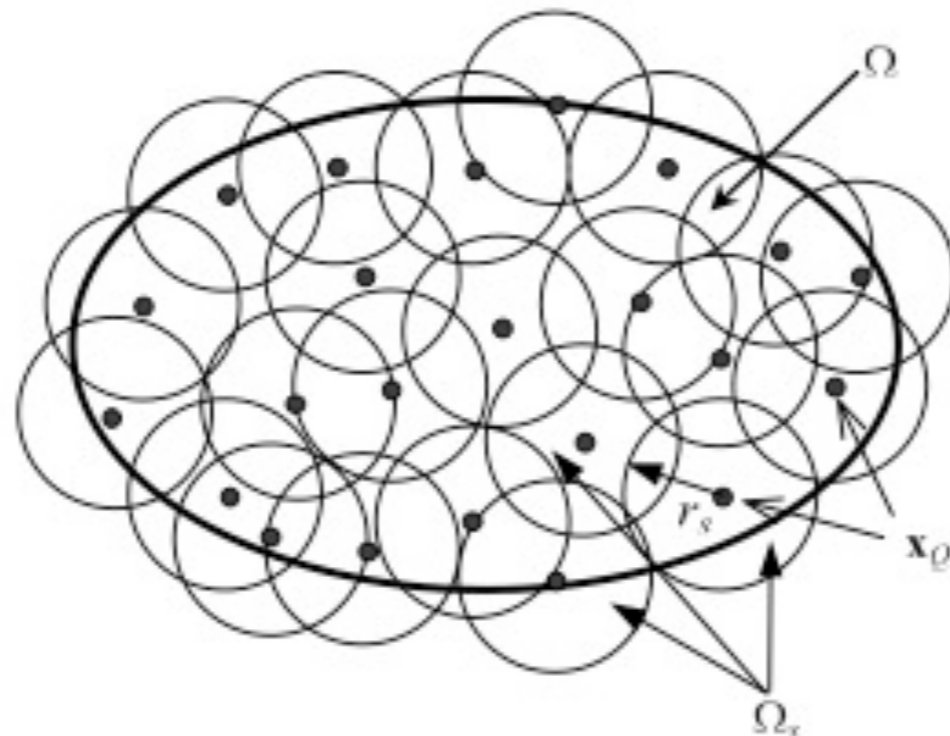
# CFD - Main discretization methods

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## ► Mesh free methods

- Smoothed Particle Hydrodynamics
- Vortex methods
- Radial Basis Functions
- ...

$$\frac{\partial u}{\partial x} = \sum_{i=0}^N \alpha_i \frac{\partial \phi_i}{\partial x}$$

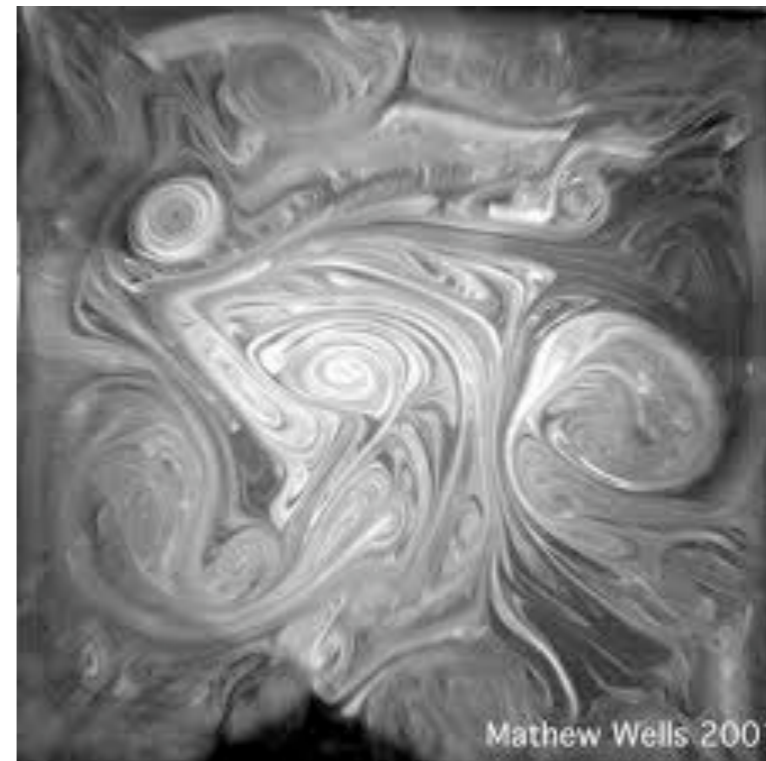


# CFD - Fluid Modeling

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- ▶ Fluid flow is a multi-scale phenomena
  - We need  $Re^3$  mesh points to reproduce all scales!
  - Turbulence modeling
  - Approximate turbulence effects

$$Re = \frac{VD}{\nu}$$





# Conventional CFD

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- ▶ Unstructured grids
  - Unstructured sparse matrices
- ▶ Incompressible  $\nabla \cdot \mathbf{u} = 0$ 
  - Projection methods
- ▶ Implicit
  - Linear solvers
- ▶ Modeled turbulence
  - Reduced number of points

# Conventional CFD

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- ▶ CFD is a tough problem for the GPU:
  - Memory bound problems
- ▶ Also, needs to convince people
  - Old legacy codes
  - How to port old codes to the GPU?
- ▶ On the other hand, CFD codes are
  - SIMD
  - Single precision
  - Large data sets

# Porting a code to GPU

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- ▶ Option 1: accelerate the existing code
- ▶ Option 2: Rewrite code from scratch
- ▶ Option 3: Rethink algorithms

Potential  
acceleration



# Option 1: Accelerate existing code

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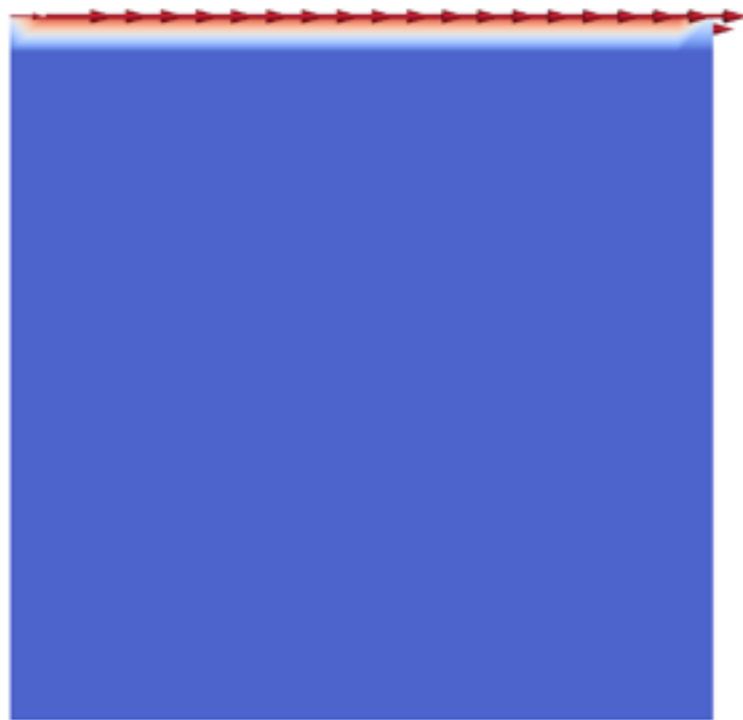
- ▶ Easiest way
- ▶ Probably not huge speedup
- ▶ Libraries like Cusp or CUFFT may be useful

# Option 1: Accelerate existing code - SpeedIT

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## ► SpeedIT (OpenFOAM)

- Ported linear solvers to GPU
- Supports multi-GPU



Mesh	Speedup
20x20	-100x
96x96x96	2.4x
128x128x32	2.0x

Xeon X5650 CPU  
M2050 GPU

# Option 1: Accelerate existing code - FEAST

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- ▶ FEAST (Finite Element Analysis and Solution Tools)
  - High level abstraction approach
  - Isolate “acceleratable” parts of code
  - Ports solver to GPU: Multigrid

Acceleration fraction: 75%

Local speedup: 11.5x

Global speedup: 3.8x

Opteron 2214 4 nodes CPU

GTX 8800 GPU

Strzodka, Goddeke, Behr (2009)

## Option 2: Rewrite whole code

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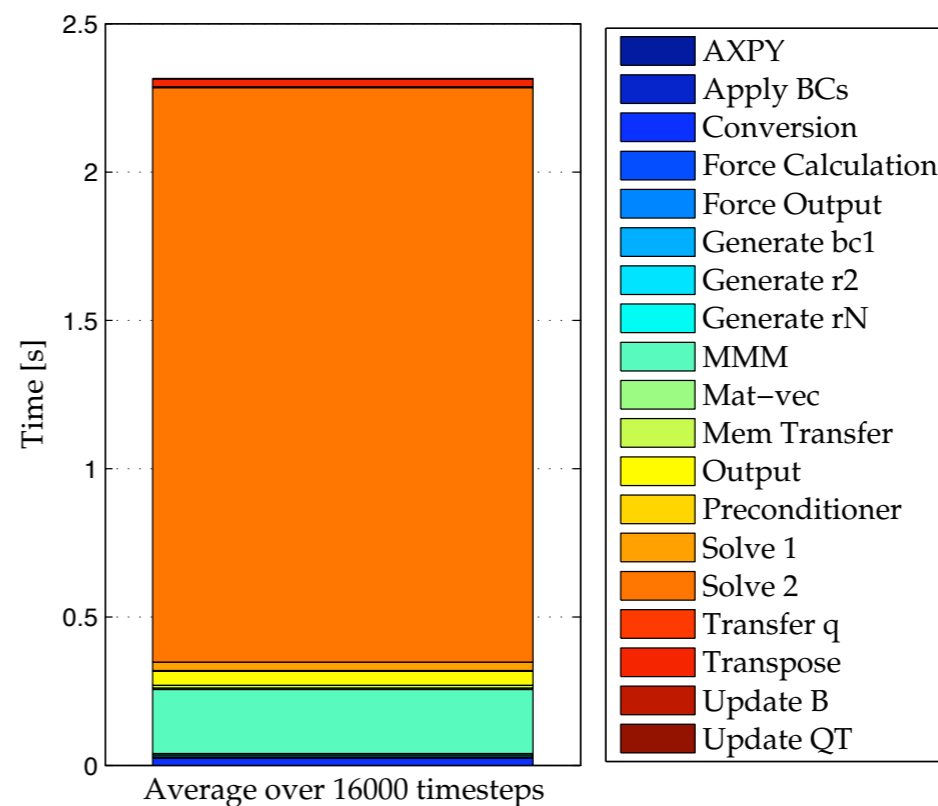
- ▶ First need to think about
  - What is the total application speedup that you can get
  - How does rewrite compare to accelerator approach
  - Good design
  - What global optimizations are possible

# Option 2: Rewrite whole code - cuIBM

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## ► Immersed Boundary Method on GPU (cuIBM)

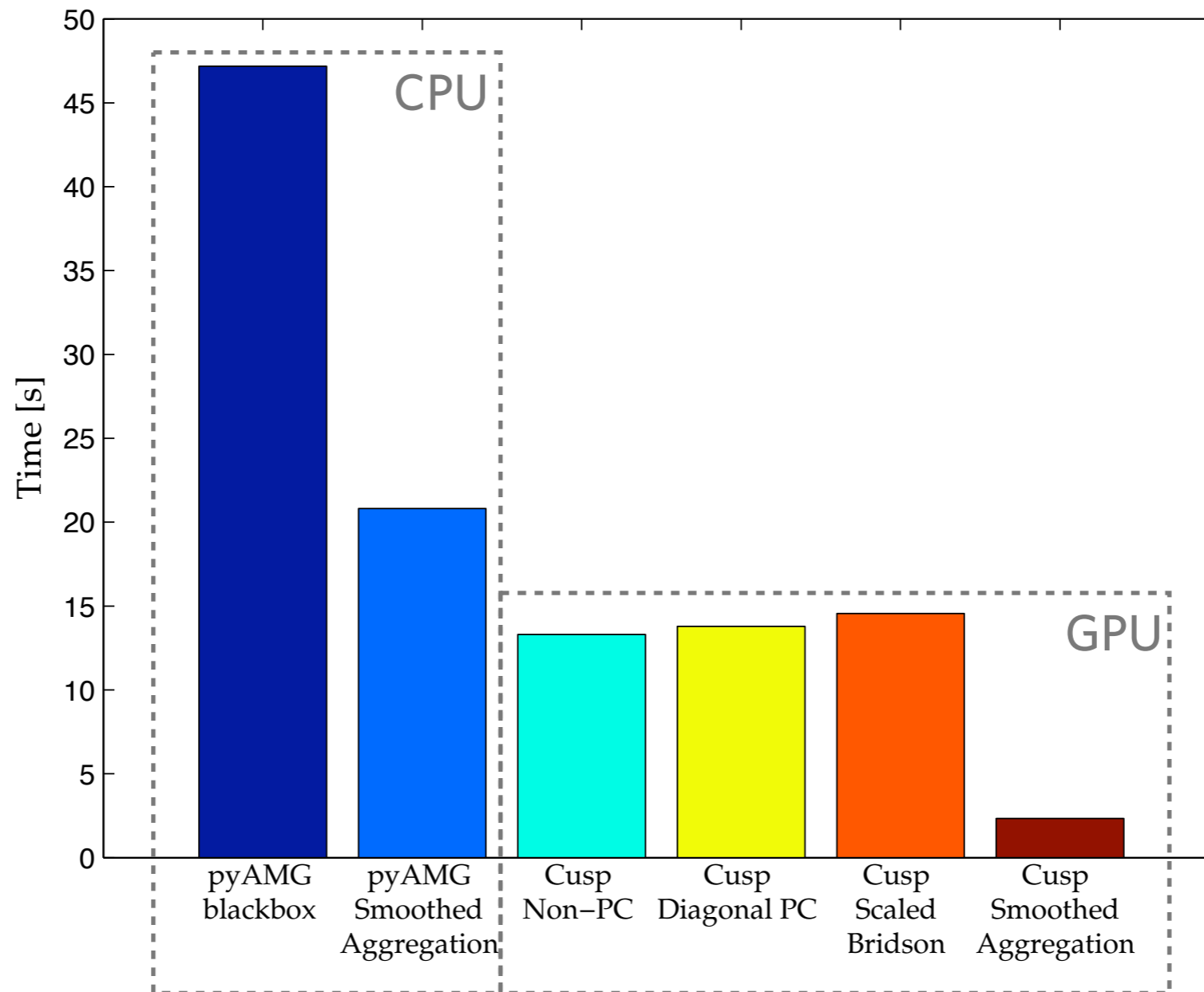
- Finite difference code with immersed boundary no slip condition
- 2 linear systems: implicit diffusion and projection
- Reported speedup: 7x





## Option 2: Rewrite whole code - cuIBM

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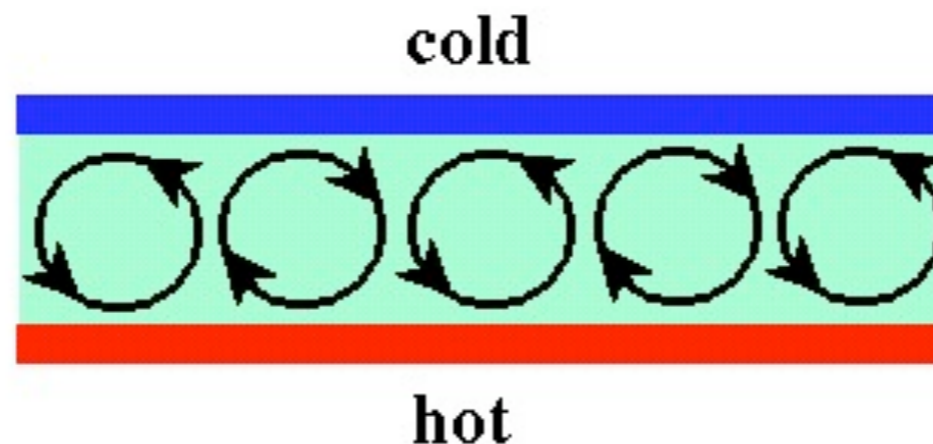


With good pre-conditioner, GPU is 9x faster, not much difference in other cases (best is 1.6x faster)

## Option 2: Rewrite whole code - Open Current

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- ▶ Developed by Jonathan Cohen in NVIDIA
- ▶ Compared a highly optimized CPU code and GPU code
  - CPU: Fortran, 8-core 2.5 GHz Xeon (8 threads with MPI and OpenMP)
  - GPU: CUDA, Tesla C1060
- ▶ Solved the Rayleigh-Bernard with a finite difference code



## Option 2: Rewrite whole code - Open Current

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Resolution	CUDA time/step ms	Fortran time/step ms	Speedup
64x64x32	24	47	2.0x
128x128x64	79	327	4.1x
256x256x128	498	4070	8.2x
384x384x192	1616	13670	8.5x

## Option 3: Rethink numerical algorithms

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- ▶ Most time consuming alternative!
- ▶ Maybe new architectures require new numerics
- ▶ Find methods that map well to the hardware
  - Maybe we overlooked something in the past because it was impractical

# Option 3: Rethink numerical algorithms - DG

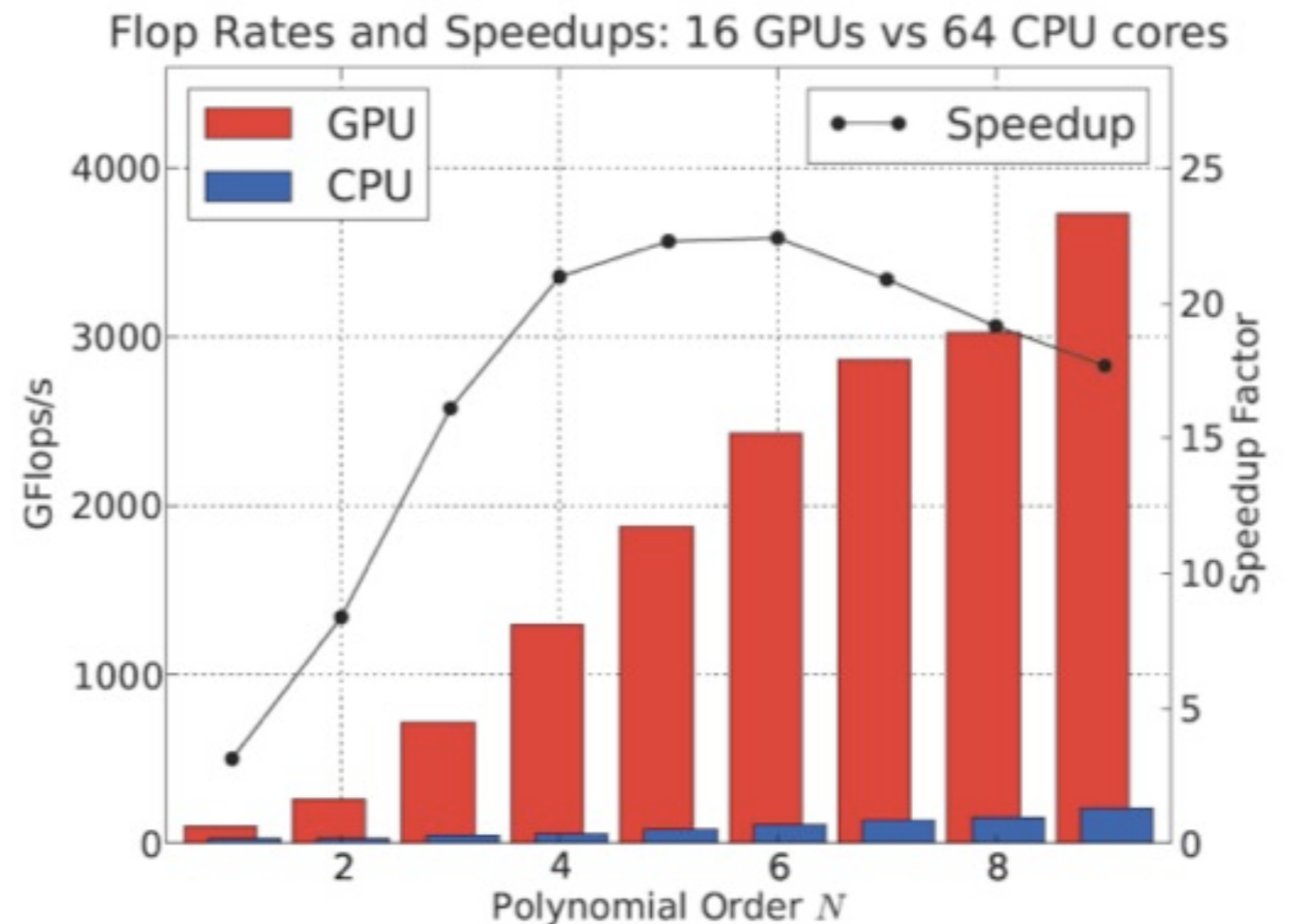
## ► Discontinuous Galerkin Methods

- Arithmetically intensive
- Mainly local

## ► Klockner et al. used DG to solve conservation laws

GPU T1060

CPU: Xeon E5472



# Implicit heat equation solver

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- ▶ Conventional CFD usually is dominated by Poisson type solvers
  - Projection methods
  - Implicit solvers to avoid stability constraints
- ▶ Heat equation with Crank-Nicolson  $\frac{\partial u}{\partial t} = \alpha \nabla^2 u$ 
  - No stability constraint!

$$\frac{\alpha k}{h^2} < 0.5$$

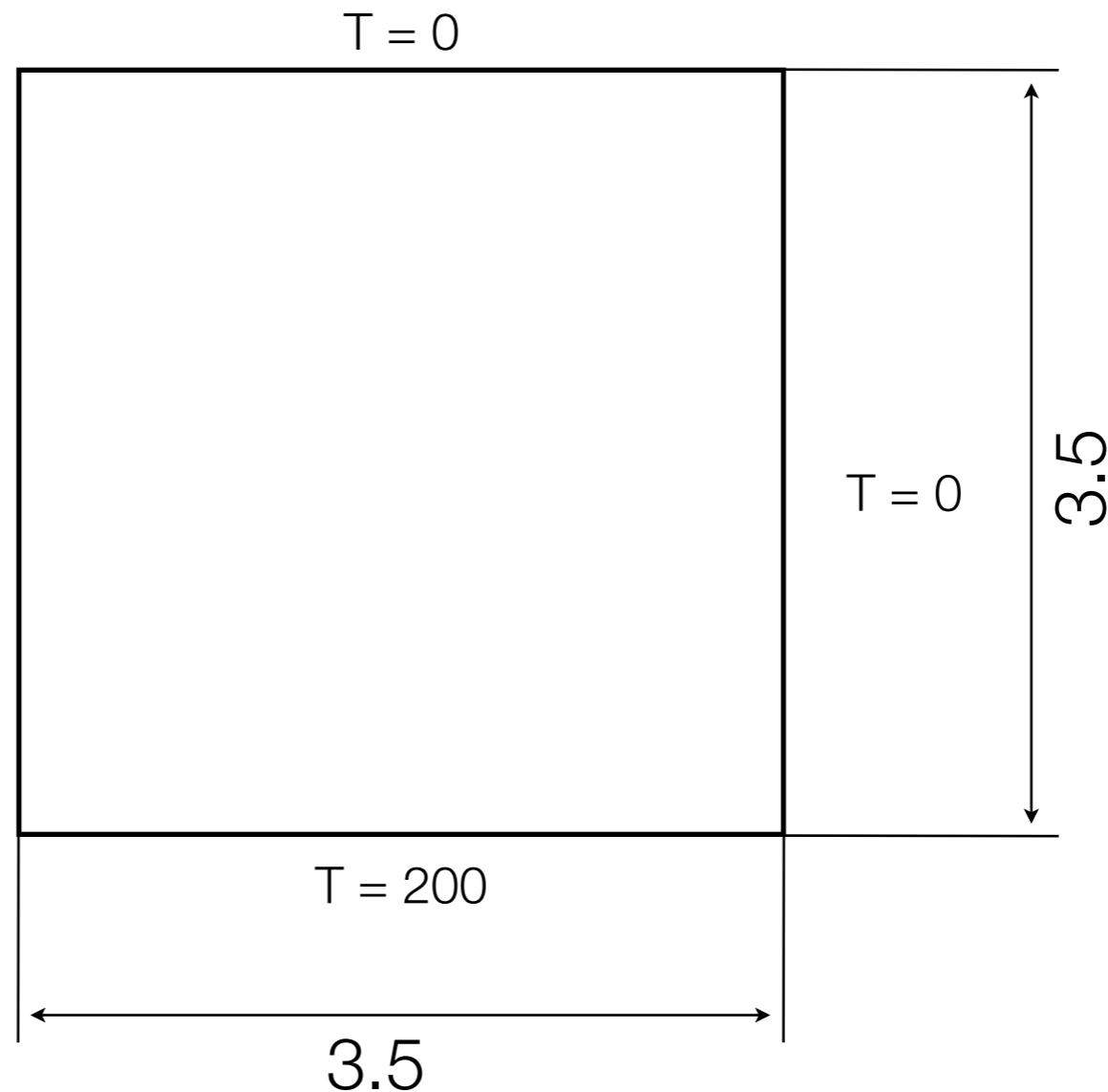
# Implicit heat equation solver

$$\frac{\partial u}{\partial t} = \alpha \nabla^2 u$$

$$\alpha = 0.645 \quad T = 200$$

$$k = 1e-5$$

$$N = 128$$



$$\frac{u_{i,j}^{n+1} - u_{i,j}^n}{k} = \frac{\alpha}{2} \left( \frac{u_{i,j+1}^n + u_{i,j-1}^n + u_{i+1,j}^n + u_{i-1,j}^n - 4u_{i,j}^n}{h^2} + \frac{u_{i,j+1}^{n+1} + u_{i,j-1}^{n+1} + u_{i+1,j}^{n+1} + u_{i-1,j}^{n+1} - 4u_{i,j}^{n+1}}{h^2} \right)$$

# Implicit heat equation solver

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$$au_{i,j-1}^{n+1} + au_{i,j+1}^{n+1} + au_{i-1,j}^{n+1} + au_{i+1,j}^{n+1} + bu_{i,j}^{n+1} = RHS_{i,j}$$

$$a = -\frac{\alpha k}{2h^2} \qquad b = 1 - 4a = 1 + \frac{2\alpha k}{h^2}$$

$$RHS_{i,j} = u_{i,j}^n + \frac{\alpha k}{2h^2} (u_{i,j-1}^n + u_{i,j+1}^n + u_{i-1,j}^n + u_{i+1,j}^n - 4u_{i,j}^n) - BC^{n+1}$$



# Implicit heat equation solver

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$$[A]\mathbf{u}^{n+1} = \mathbf{RHS}$$

$$[A] = I + \frac{\alpha k}{2h^2} \cdot [\text{Poisson}]$$

$$[A] \text{ size } (N-2)^2 \times (N-2)^2$$

$$\mathbf{RHS} \text{ size } (N-2)^2$$

$$\mathbf{u}^{n+1} \text{ size } (N-2)^2$$