GPU Accelerated Pressure Solver for TIGER-F code



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(NTUST CFD Lab)



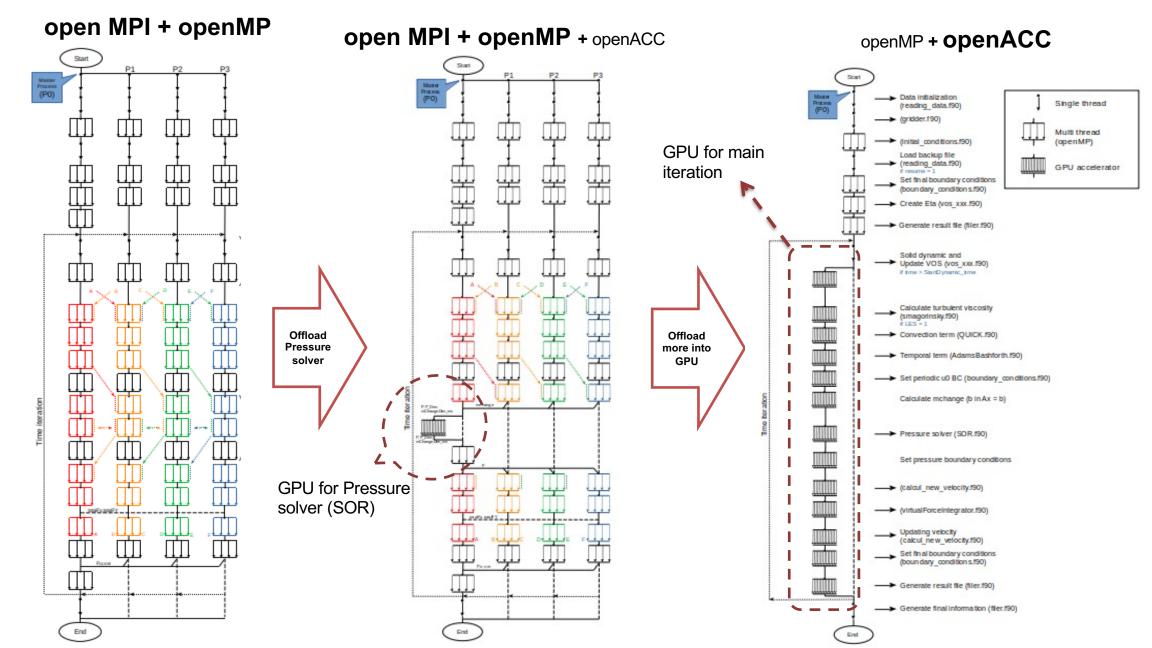
TIGER-F

- Problem trying to be solved
 3D-CFD using the Direct Forcing Immersed Boundary (DFIB) method, LES turbulence model.
- Scientific driver for the chosen algorithm
 The flexibility to solve large scale FSI problems.
- What's the algorithmic motif?
 - The use of simple and fixed Cartesian grids eliminates the need for remeshing, even as the structure is in motion.
- What parts are you focusing on?
 Initially: Computation intensive parts (pressure solver).
 But in the end, we need to offload all processes in the main iteration.

Evolution and Strategy

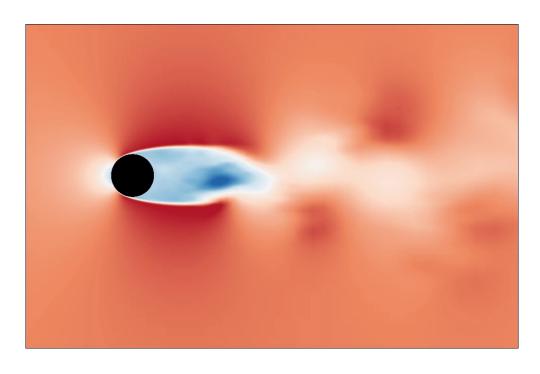
- What was your goal coming here?
 to speedup our in-house solver using openACC by offloading and optimizing the computation intensive processes (e.g. pressure solver) into GPU
- What was your initial strategy?
 - To optimize the GPU-accelerated pressure solver and to use the Nsight systems for profiling.
 - to offload more subroutines into GPU.
- How did this strategy change?

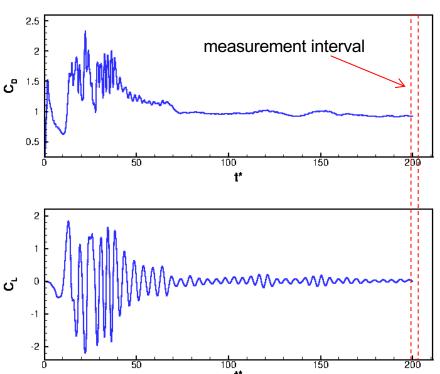
The majority of computational time is consumed by data transfer overhead, which could be diminished by relocating **all** processes into GPU.



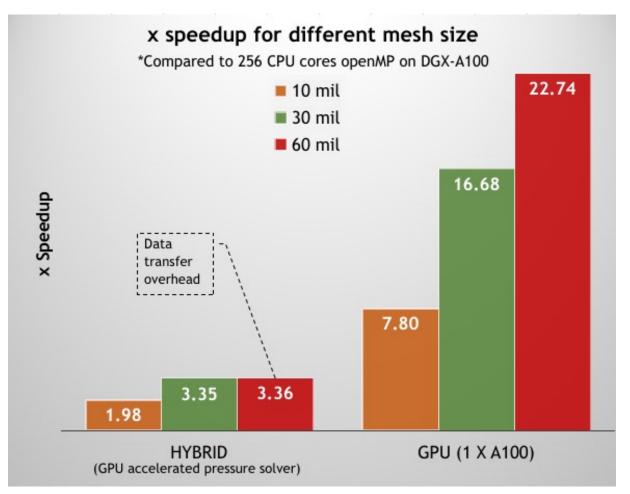
Test Case

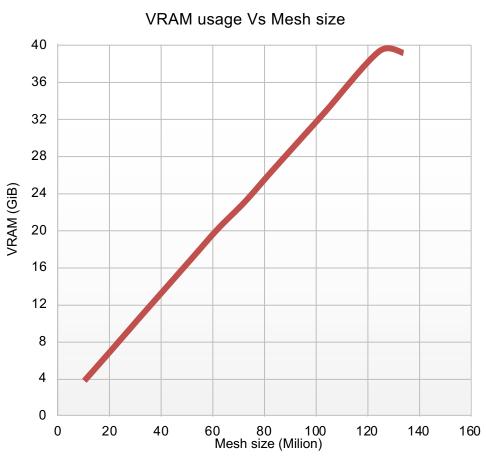
The test case was the flow past a stationary cylinder at Re = 5000, running for 200 sec of dimensionless time to give enough time to reach a **steady number** of pressure iterations. Speedup was measured for the next 1 second.





Results and Final Profile





Results and Final Profile

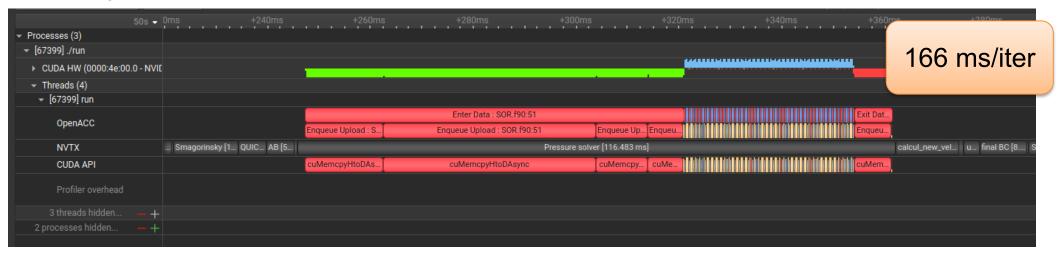
00_DEFAULT 01 SOR 1 02 AB 03 QUICK 04 SMAGO 05 CALC NEW V 06_VIRT_FORCE 07 UPDATE VEL 08 MANAGE DATA 09 MANAGE DATA1 10 FINAL BC 11 PINNED 12_MANAGE_DATA2 Tall 13 VEC LENGTH 14 VEC SOR 64 15 VEC SOR 256 16_GANGVECTOR_SOR 17_VEC_SOR_512 18 UPDATE CLAUSE 19_OUTPUT_FILE_TIME

20 ASYNC01

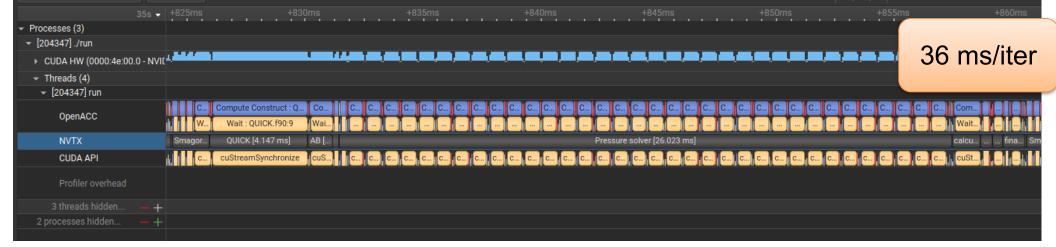
- What did you learn?
 - Data transfer management.
 - OpenACC's three levels of parallelism (gang, worker, vector).
 - Profiling tools: Nsight systems, NVTX.
- Achieved new scientific goals?
 - Apply the red-black SOR algorithm for 3D domain, and run in GPU.
 - 100 Million of mesh is possible to be done (very) much faster --> one step closer to simulate real world cases.

Results and Final Profile

HYBRID 10mil



GPU 10mil



Energy Efficiency

INPUTS		
# CPU Cores	256	
# GPUs (A100)	1	
Application Speedup	16.7x	

Node Replacement 266.9x

GPU NODE POWER SAVINGS			
	AMD Dual Rome 7742	8x A100 80GB SXM4	Power Savings
Compute Power (W)	293,568	6,500	287,068
Networking Power (W)	12,393	93	12,300
Total Power (W)	305,961	6,593	299,368

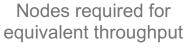
Node Power efficiency 46.4x

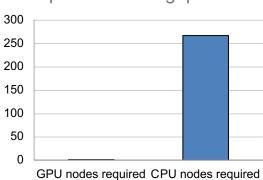
ANNUAL ENERGY SAVINGS PER GPU NODE			
	AMD Dual Rome 7742	8x A100 80GB SXM4	Power Savings
Compute Power (kWh/year)	2,571,656	56,940	2,514,716
Networking Power (kWh/year)	108,563	814	107,749
Total Power (kWh/year)	2,680,219	57,754	2,622,465

\$/kWh	\$ 0.34
Annual Cost Savings	\$ 891,638.13
3-year Cost Savings	\$ 2,674,914.39

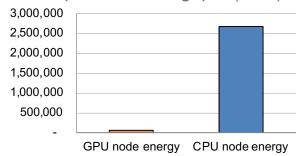
Metric Tons of CO ₂	1,859
Gasoline Cars Driven for 1 year	401
Seedlings Trees grown for 10 years	30,735

(source: Link)



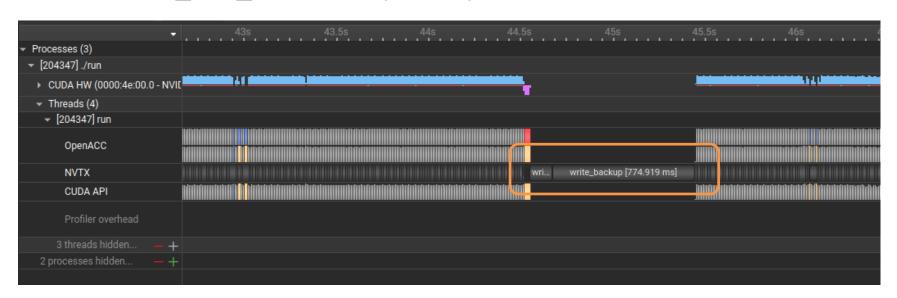


Annual energy required for equivalent throughput (kWh)



What problems have you encountered?

- Unsuccessful Async between CPU file writing and GPU processes
- Unable to use "cache" clause
- Nsight compute could not be used because CAP_SYS_ADMIN capability was disabled.



Wishlist

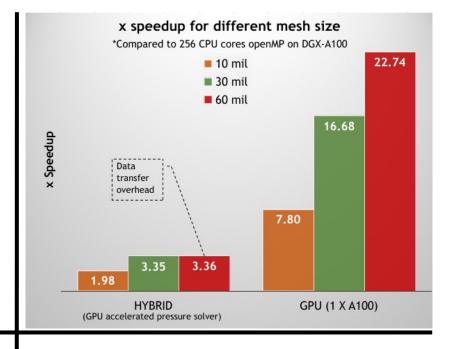
- What do you wish existed to make your life easier?
 - Multi GPU resource availability.
 - Updated training especialy on multi GPU utilization.
 - An integrated cluster with a balanced
 CPU+GPU performance.

Was it worth it?

- Was this worth it? Definitely YES!
- Will you continue development? YES
 - problems encountered on this event (async file writing, cache)
 - Use Nsight compute to optimize kernels
 - Multi-GPU process
 - Employing a readily available linear system solver with multi-GPU support. (AmgX)
- What sustained resources/support will be critical for your work after the event?
 - Multi-GPU resource and support.

Application Background

- A 3D-CFD in-house code using the DFIB method, LES turbulence model.
- Pressure solver is the most computational intensive subroutine.



Hackathon Objectives and Approach

- Offloading into GPU
- Reducing data transfer overhead
- Profiling / hot spots
- Performance tuning

Technical Accomplishments and Impact

- Learn more openACC directives & clauses
- Learn how to use Nsight systems
- Offload more processes into GPU
- Reduce data transfer overhead
- Speedup

Please use 100 words to summarize your team's achievements during this Hackathon

We began considering GPU offloading around 7 months ago and successfully accelerated our pressure solver onto the GPU. This hackathon not only boosted our learning curve but also enhanced the performance of our code. From this event, we are able to:

- Utilize openACC directives and their clauses
- Manage host-device data transfer better
- Optimize parallelism using openACC's three levels of parallelism (gang, worker, vector)
- Utilize Nsight systems with NVTX to profile the code
- Offload all processes in the main iteration to the GPU.
- Achieve a speedup of 16.7x (and even higher with increased computational load)

This marks another milestone for our in-house code as large-scale computations are required to simulate real-world cases. The significant speedup achieved is a substantial leap forward for us in reaching that goal.



Our mentors:

Bharat, Shijie, Kuan-Ting

Host	Jay	Floating Mentor	Bharat, Aswin
Host	CK	Marketing	Jinny
Infra Setup	Kuan-Ting	Event logistics	Apoorva
Account Manager	Vincent	NCHC Contact Window	Zhoujin Wu

All mentors and all participants in this hackathon

NAR Labs 國家實驗研究院

國家高速網路與計算中心

National Center for High-performance Computing

