

# GPU Porting of CAS3D and SOLEDGE3X codes using OpenACC/OpenMP offload

# Transition towards GPU codes

There are 3 main approaches:

Pragma directives (OpenMP offload / OpenACC)

Cuda

Library encapsulation (e.g. Kokkos)

- **ASCOT5** 
- **CAS3D** 
- **FELTOR** 
- **GBS** 
- **GRILLIX** 
- **GyselaX** 
- **ORB5** 
- **Soledge3X** 
- SCITAS

# Transition towards GPU codes

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Pragma directives (OpenMP offload / OpenACC)

Cuda / Hip / ...

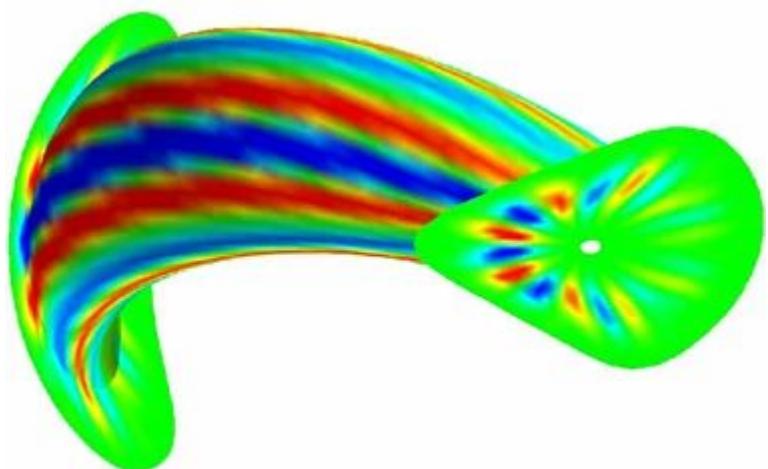
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- GRILLIX 
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- Soledge3X 

G. Fourestey, M. Peybernes (SCITAS),  
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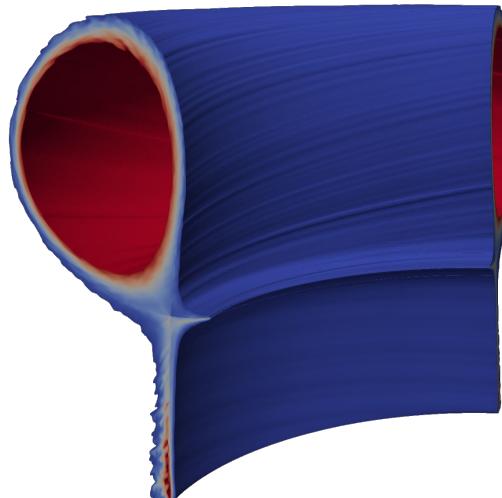
- CAS3D is used to study the ideal magnetohydrodynamic (MHD) properties of fusion toroidal plasmas
- The numerical scheme uses:
  - **Finite elements** along the radial direction
  - **Fourier decomposition** along the poloidal and toroidal directions
  - Inverse iteration algorithm to solve **eigenvalues problem** for a sparse symmetric matrix.



# Soledge3X

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- Soledge3x is used to study tokamak edge plasma
- The numerical scheme uses:
  - WENO methods for the advection
  - 1D, 2D, and 3D Sparse matrix problems
  - linear solvers
- 5 implicit solvers in SOLEDGE3X (typical share of computing time):
  - Parallel viscosity terms (~12% / 7% with FN)
  - Parallel heat conduction terms (~17% / 9% with FN)
  - Vorticity equation (~22% / 12% with FN)
  - (optional) fluid neutrals (~48% with FN!!)
  - (optional) potential filter (negligible)



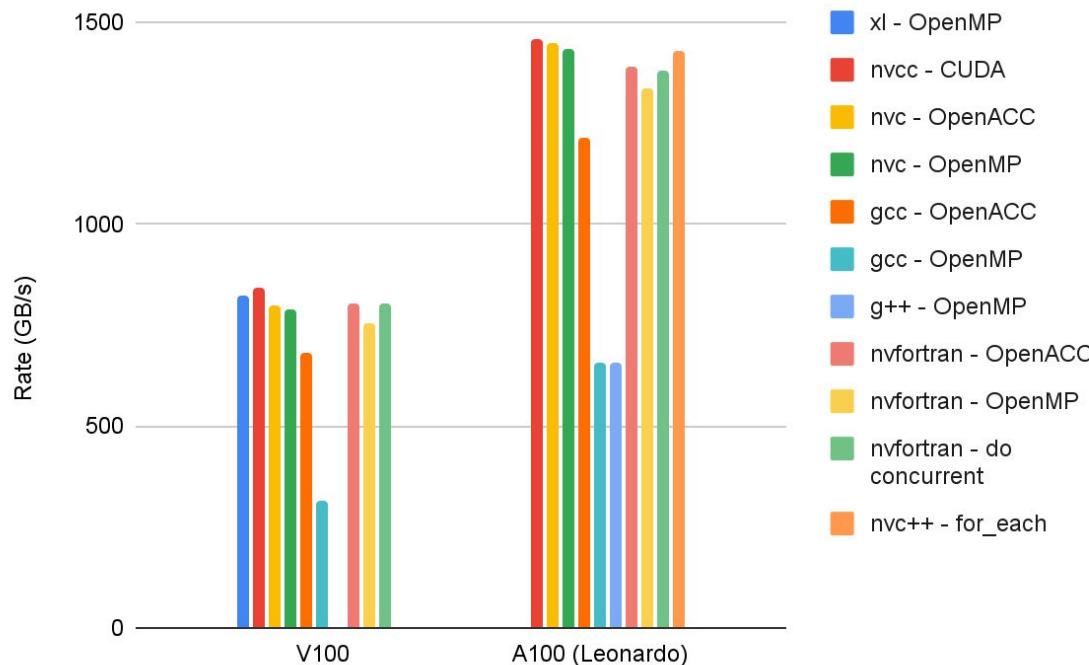
# Porting codes to GPU with pragmas

General strategy to port CAS3D and SOLEDGE3X codes to GPU:

- Porting by directives with OpenMP offload and/or OpenACC
- This choice allows us
  - to keep only one version of the code with readability and durability
  - to perform tests on NVIDIA and AMD GPUs (and INTEL) for multiple compilers:  
xl (M100-Nvidia) ; gnu (Leonardo-Nvidia) ; cce (Piz-Daint-Nvidia, then ALPS & Adastra-AMD & LUMI-AMD) ; nvhpc (Leonardo-Nvidia)
- This strategy can show good performance (not always!), in particular on Marconi100 with the IBM XL compiler with OpenMP offload
- However, other compilers, like GCC, present weak performances with OpenMP offload while OpenACC appears to be more efficient
- This lack of performance portability led to the introduction of generic pragmas to use OpenMP or OpenACC

# Porting codes to GPU with pragmas

STREAM benchmark:



# Porting codes to GPU with pragmas

```
!$omp target teams distribute collapse(2)
do ichunk = 1, split%Nchunks
  do ispec = 0, Nspecies
    ... some work ...
    !$omp parallel do simd collapse(3)
    do ipsi = ipsimin, ipsimax
      do itheta= ithetamin, ithetamax
        do iphi = iphimin, iphimax
```

OpenMP

```
!$acc loop gang collapse(2)
do ichunk = 1, split%Nchunks
  do ispec = 0, Nspecies
    ... some work ...
    !$acc loop worker vector collapse(3)
    do ipsi = ipsimin, ipsimax
      do itheta= ithetamin, ithetamax
        do iphi = iphimin, iphimax
```

OpenACC

```
GPU_LOOP_L1 collapse(2)
do ichunk = 1, split%Nchunks
  do ispec = 0, Nspecies
    ... some work ...
    GPU_LOOP_L2_L3 collapse(3)
    do ipsi = ipsimin, ipsimax
      do itheta= ithetamin, ithetamax
        do iphi = iphimin, iphimax
```

Pragmas

```
#ifndef gpu_commands
#define gpu_commands
#endif _OPENMP

#define GPU_MAP_TO_DEVICE !$omp target enter data map(to:
#define GPU_MAP_FROM_DEVICE !$omp target exit data map(from:

#define GPU_LOOP_L1 !$omp target teams distribute
#define GPU_END_LOOP_L1 !$omp end target teams distribute

#define GPU_LOOP_L2_L3 !$omp parallel do simd

... etc ...
#elif _OPENACC

#define GPU_MAP_TO_DEVICE !$acc enter data copyin(
#define GPU_MAP_FROM_DEVICE !$acc exit data copyout()

#define GPU_LOOP_L1 !$acc loop gang
#define GPU_END_LOOP_LEVEL_1 !$acc end loop

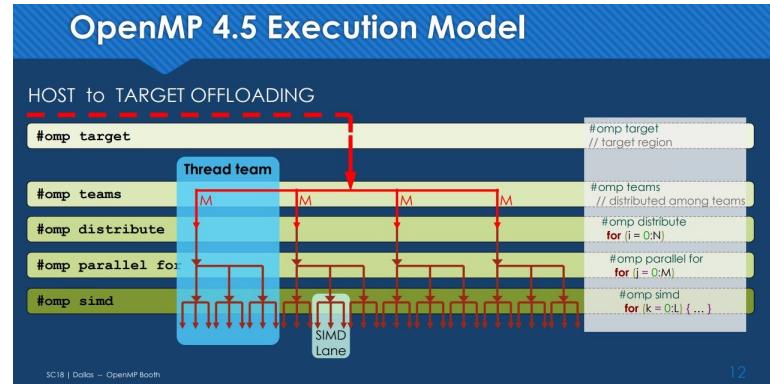
#define GPU_LOOP_L2_L3 !$acc loop worker vector

... etc ...
#endif
```

gpu\_transfers.inc

# Levels of Parallelism

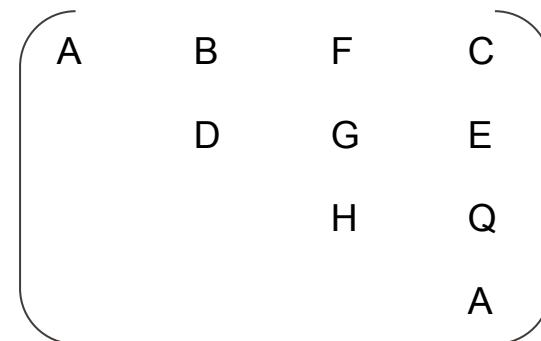
- OpenMP and OpenACC have similar coding paradigms under the hood
- In theory (according to the standards) the implementation of the levels adapts to the hardware, even if in reality some compilers struggle with certain parallelisation levels



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OpenMP	OpenACC	HIP	CUDA	Mapping AMD GPU	Mapping NVIDIA GPU
Parallel	Kernel/parallel	Kernel	Kernel	GPU/GCD	GPU
Team	gang	Work group	Thread bloc	Comput Unit	SM (Symmetric Multiprocessor)
Thread	worker	Work item	thread	Scalar Unit Part of Vector (SIMD) Unit	Comput Unit
SIMD	Vector	Wavefront (64 work items)	Warp (32 threads)	64-wide work item	32-wide thread

- The expensive part is the computation of matrices (memory latency bound)
- Each Matrix block has the following dimensions
  - A-block:  $modx \cdot modx$
  - B-block:  $modx \cdot mody$
  - F-block:  $modx \cdot modz$
  - C-block:  $modx \cdot modx$
  - D-block:  $mody \cdot mody$
  - G-block:  $mody \cdot modz$
  - E-block:  $mody \cdot modx$
  - H-block:  $modz \cdot modz$
  - Q-block:  $modz \cdot modx$



where  $modx$ ,  $mody$ ,  $modz$  depend on the number of Fourier coefficients used during the discretisation

—> typically,  $modx, mody, modz$  in the order of  $10^2$

- Each GPU kernel computes a matrix block ( $N \sim 10^4$ ) :
  - each team/gang calls *matrixa* subroutine
  - each team/gang spawns its threads to parallelize the loop in *matrixa* subroutine
  - need to refactor initial code to collapse 2 loops on Fourier coefficients to extract more parallelism over the teams/gangs

Parallelization over  
gangs/teams



```
if (irun>=1 ... then
GPU_LOOP_L1 COLLAPSE(2) PRIVATE(Tab1(N), Tab2(N), ...)
    do jmod = 1, mody
        do imod = 1, modx
            call matrixa (wpot, wkin, imod, jmod)
            Tab1(k) = ...
            a (imod,jmod,igrid) = fachalf*pot
            arhs(imod,jmod,igrid) = fachalf*wki
```

Parallelization over  
workers/threads



```
GPU_LOOP_L2_L3
    do jt = 1, N
        wpot = ...
        wki = ...
```

# CAS3D - Gang-private array workaround

- On some machines/compilers (nvfortran < 23.1) private arrays are not correctly handled
- A workaround is to allocate shared arrays over the number of gangs
  - avoids also dynamical allocation and improves performance
- This requires accessing a gang/team id.
  - In OpenMP this is available
  - In OpenACC with **nvhpc** we can use pragmas exposed via C (not compiler-portable):

```
#ifdef __PGI
#include "openacc.h"

#pragma acc routine worker
int get_gang_id() {
    return __pgi_gangidx() + 1;
}

#pragma acc routine vector
int get_worker_id() {
    return __pgi_workeridx() + 1;
}

#pragma acc routine seq
int get_vector_id() {
    return __pgi_vectoridx() + 1;
}

#endif
```

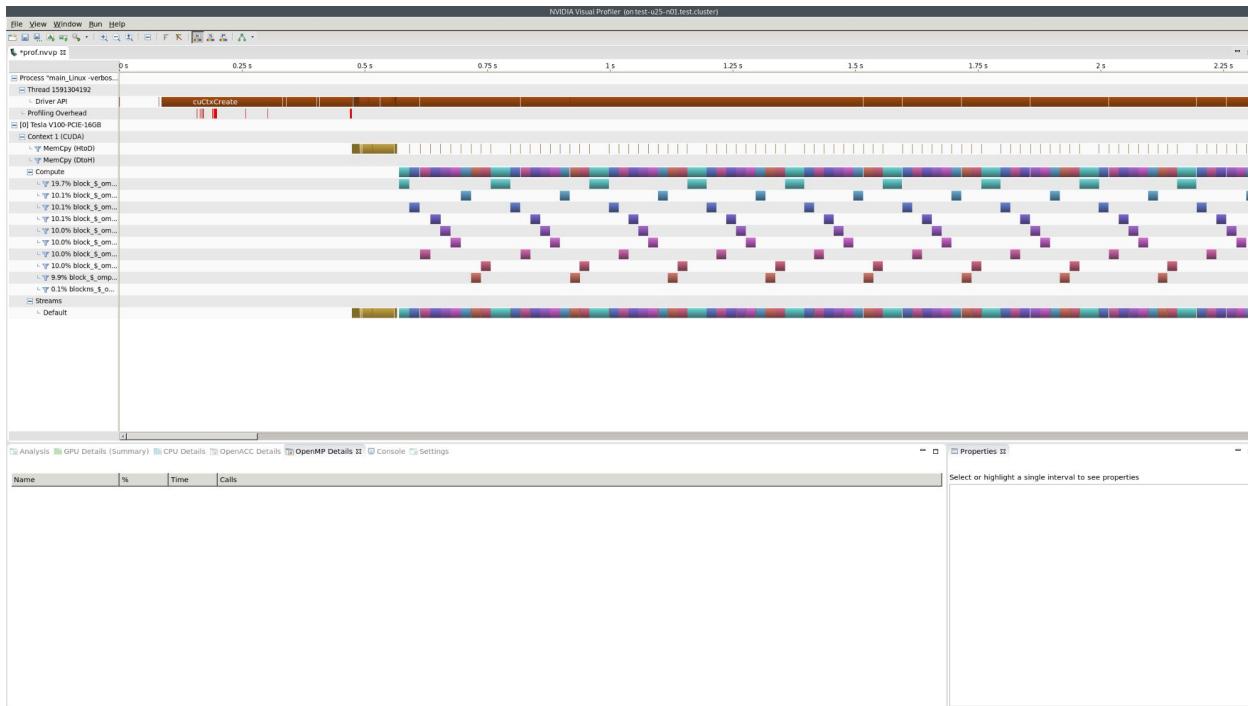
```
allocate(Tab1(N, nbgangs))
allocate(Tab2(N, nbgangs))

if (irun>=1 ... then
    GPU_LOOP_L1 COLLAPSE(2) PRIVATE(Tab1(N), Tab2(N), ...)
        do jmod = 1, mody
            do imod = 1, modx
                call matrixa (wpot, wkin, imod, jmod)
                Tab1(k,gangId) = ...
                a (imod,jmod,igrid) = fachalf*wpot
                arhs(imod,jmod,igrid) = fachalf*wki
```

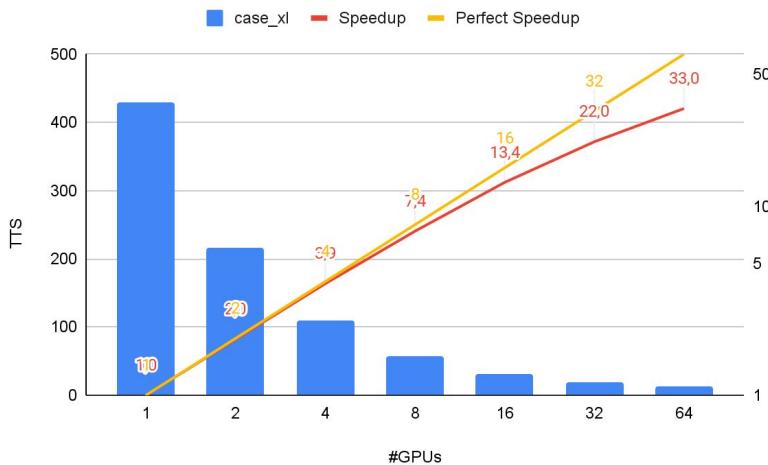
```
GPU_LOOP_L2_L3
    do jt = 1, N
        wpot = ...
        wki = ...
```

- GPU

- Profiling with Nvidia Nsight Systems
- Transfers optimized between kernels

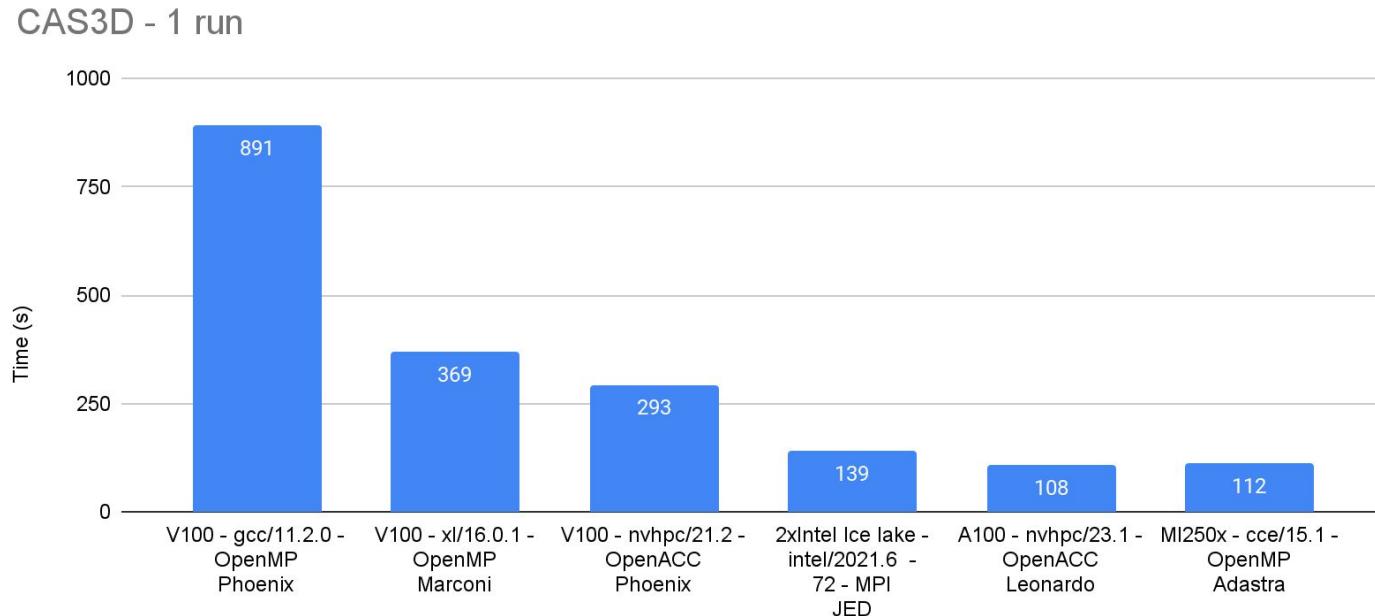


- Results on Marconi 100 (P9 + V100 GPUs)
  - Good scaling
  - Performance increases compared to the CPU version for large test cases



XL-compiler on M100		1 node (4GPUs)	2 nodes	4 nodes	8 nodes	16 nodes	GPU-serial (640 teams - 128 threads)	CPU: 1node (32-cores)
#GPUs	4		8	16	32	64	1	0
case_mm	4 s		3,2 s	3 s	3,1 s	3,3 s	6 s	5 s
case_ll	78 s		42 s	24 s	15 s	14,8 s	299 s	190 s
case_xl	110 s		58 s	32 s	19,5 s	13 s	410 s	418 s

- GPU - OpenMP-Offload vs OpenACC



# Soledge3X - Porting strategy

- 3 level domain decomposition:
  - Structured zones for magnetic topology
  - MPI blocks: prioritized by flux surface across zones
    - If  $NMPI \leq NFS$  each MPI process in charge of a set of FS
    - If  $NMPI > NFS$  largest flux surfaces will be shared by a team MPI processes
  - Thread chunks: no direction priority, aiming at load balance between chunks ; CPU OpenMP loops are on chunks and species
- GPU porting
  - Open ACC/MP for advection and matrix construction
  - PETSC (with CUDA/HIP) for linear solvers

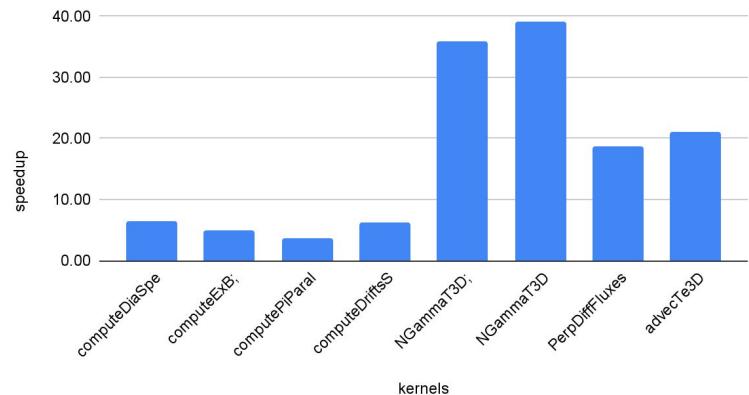
```

!$OMP DO SCHEDULE (RUNTIME) COLLAPSE (2)
GPU_PARALLEL_LOOP_L1 collapse(2)
do ichunk = 1, split%Nchunks
  do ispec = 0, Nspecies
    ... some work ...
    GPU_LOOP_L2_L3 collapse(3)
    do ipsi = ipsimin, ipsimax
      do itheta= ithetamin, ithetamax
        do iphi = iphimin, iphimax

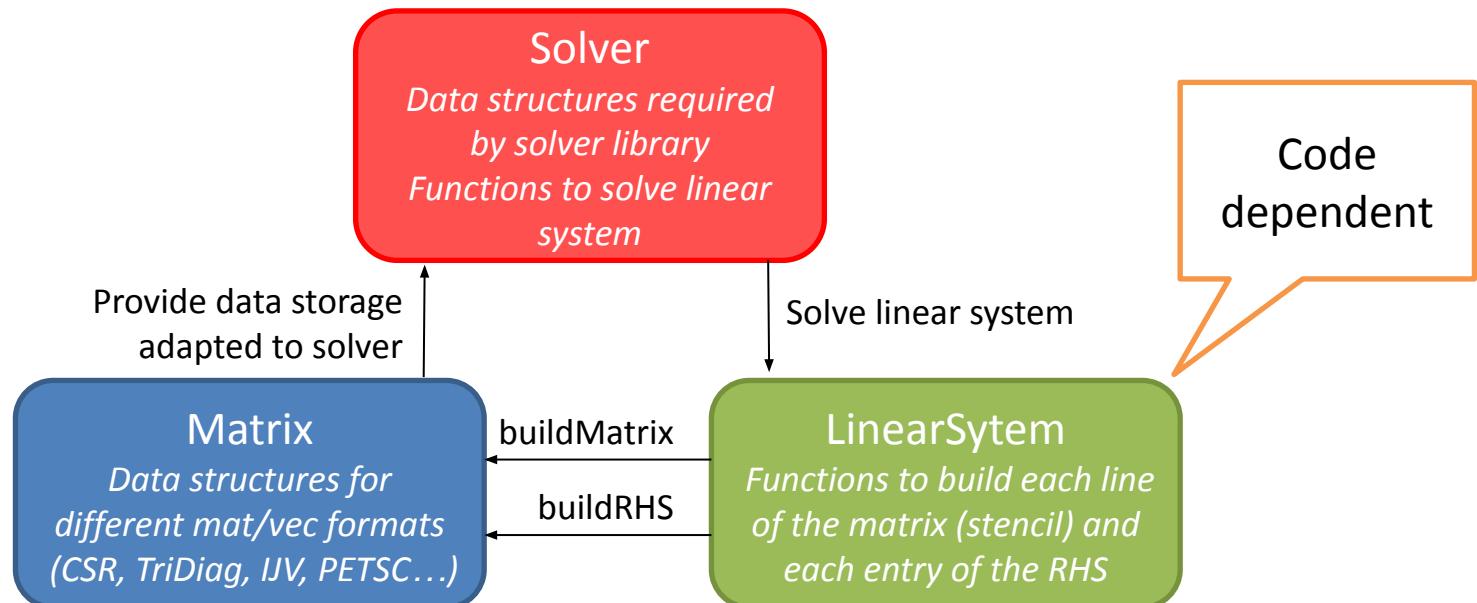
```

*Each kernel profiled with nsys-nvtx on CPU/GPU*

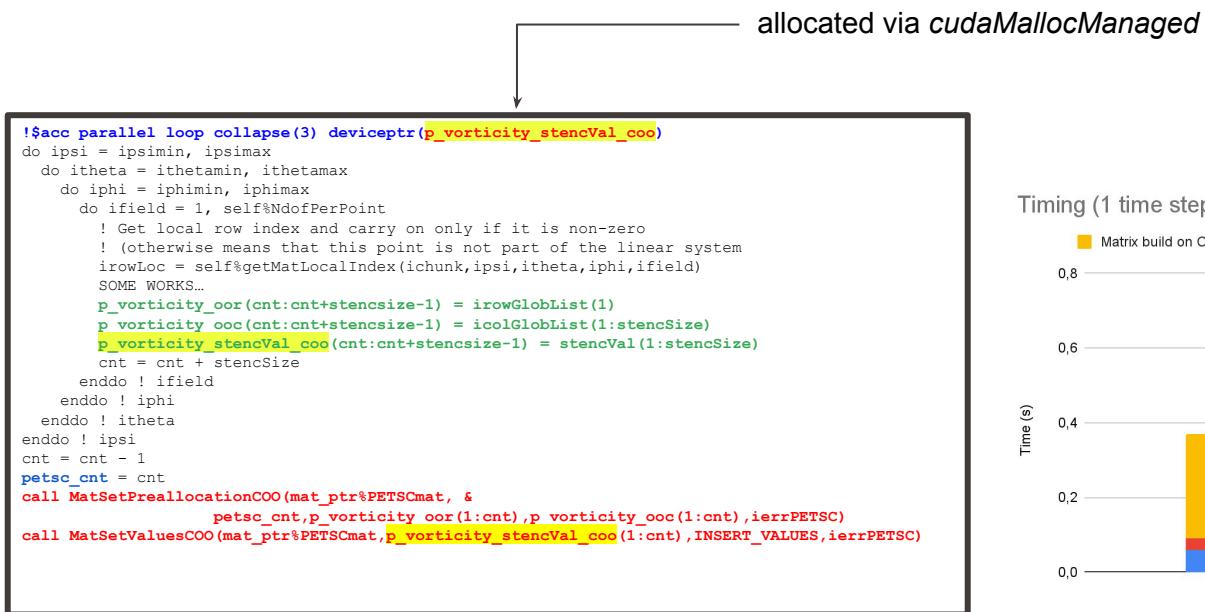
Speedup 1-GPU (OpenACC) vs 1-CPU node (32 cores)



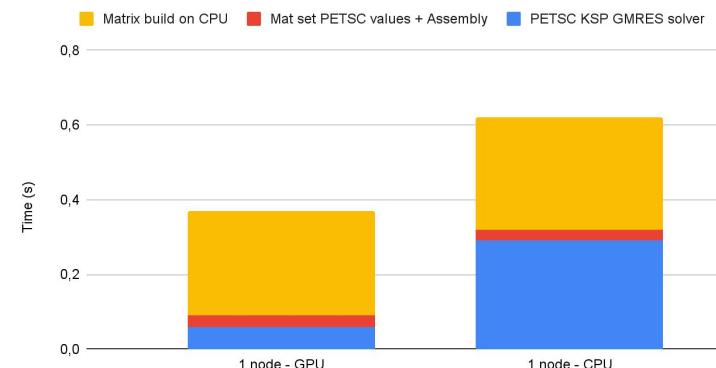
- Initially strong duplication of effort with risk of bugs
  - implementation of each solver library needs to be done separately for each implicit linear system, including matrix construction (bugs!)
- Solution recast of solvers management based on 3 Fortran classes



- OpenACC Interoperability with CUDA



Timing (1 time step) for 3D Vorticity Implicit solver



Preliminary results on Leonardo

# Conclusions

- Porting by directives with OpenMP offload and/or OpenACC
- Pragmas allow for a single code base with different solutions
- (Performance) portability is strongly dependent on hardware, compiler etc.
- For the moment, OpenACC is more mature than OpenMP offload
- Depending on the compiler, we can get quite good performance for CAS3D and Soledge3X codes
- Leonardo (Nvidia A100 - OpenACC) is ~3x faster than Marconi 100 (Nvidia V100 - OpenMP offload) for CAS3D
- Most CPU algorithms have to be modified further to get better performance on GPU
- GPU porting of Soledge3x needs some workarounds due to use of “recent” Fortran features like function pointer or class not handled by the most recent nvhpc compilers (23.9)

—> ongoing work ...