



# Evaluating Claude Code's Coding and Test Automation for GPU Acceleration of a Legacy Fortran Application: A GeoFEM Case Study

Tetsuya Hoshino, Shunichiro Hayashi, Daichi Mukunoki,  
Takahiro Katagiri (Nagoya Univ.),  
Toshihiro Hanawa (The University of Tokyo)

# Background

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[← Previous](#) [↑ Index](#) [→ Next](#)

## RIKEN launches international initiative with Fujitsu and NVIDIA for "FugakuNEXT" development

Building the next-generation "AI-HPC platform" to solve complex social challenges through computational science

[Japanese Page](#)

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RIKEN is collaborating with Fujitsu Limited(Fujitsu) and NVIDIA to launch an international initiative for the development of Japan's new flagship supercomputer - the next-generation successor to the current "Fugaku" supercomputer - (development codename: "FugakuNEXT"). For the first time in Japan's flagship supercomputing systems, GPUs will be adopted as accelerators, with NVIDIA

- GPU porting of legacy applications is increasingly urgent, but progress is slow due to various challenges.
  - In particular, Fortran programs—common in legacy applications—remain a major challenge.

# Challenges in GPU Porting

- Inconsistent support for parallel programming languages/models across vendors

Support status via each GPU vendor's own compiler

Vendor	OpenACC	OpenMP (GPU)	CUDA	SYCL	HIP	Standard parallelism (stdpar)	OpenCL
NVIDIA	✓	✓	✓	—	—	✓	C only
AMD	—	✓	—	—	C only	C only	C only
Intel	—	✓	—	C only	—	✓	C only

- Increased maintenance cost
  - CPU version + GPU version  $\times$  (# of programming languages/models)
    - In some cases, conversion to C is also required
- ↑ We want to leverage AI to deal with this
  - There are many success stories at the function level (e.g., BLAS)
  - But what about full HPC applications?

# Key Features of Claude Code

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- Claude Code
  - A CLI tool developed by Anthropic
  - Functions as an interactive AI assistant for code development
  - Can develop code with direct access to the file system
  - Integrates with large language models such as Claude Opus
    - Can be instructed in Japanese
- Claude Opus 4.1
  - Anthropic's large language model released in Aug 2025
    - Latest is Opus 4.5 (Nov.28, 2025)
  - With Claude Code, can **autonomously execute an end-to-end workflow from coding to test runs**

↑ We want to evaluate how useful it is for GPU code development

# Using Claude Code on a Supercomputer

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- Install and use on login and compute nodes
  - Inference runs on external servers (requires external network access)
  - Requires a subscription with Anthropic
    - Top personal plan: USD 200/month
- Directly edits and runs source code on the file system
  - By default, permissions are limited to the launch directory and below
  - Requests permission when editing or deleting files
    - Launching with `--dangerously-skip-permissions` skips permission prompts
  - Any command available under your user privileges can be used (e.g., submitting jobs with `qsub`)
    - Write instructions to a file; with `--dangerously-skip-permissions` to skip interaction, the code-development process itself can run as a batch job
- Inference is stochastic; the seed cannot be fixed

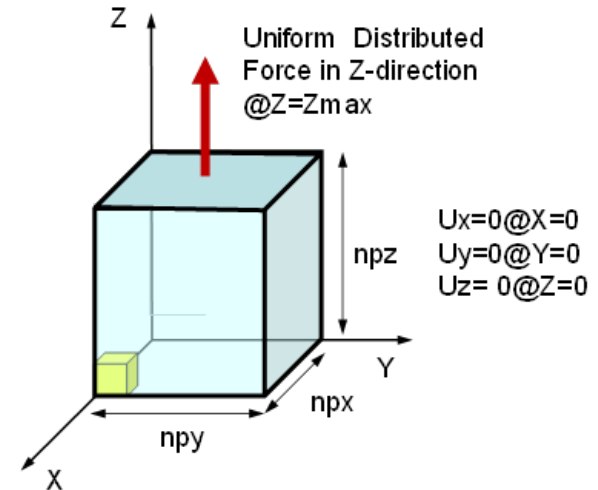
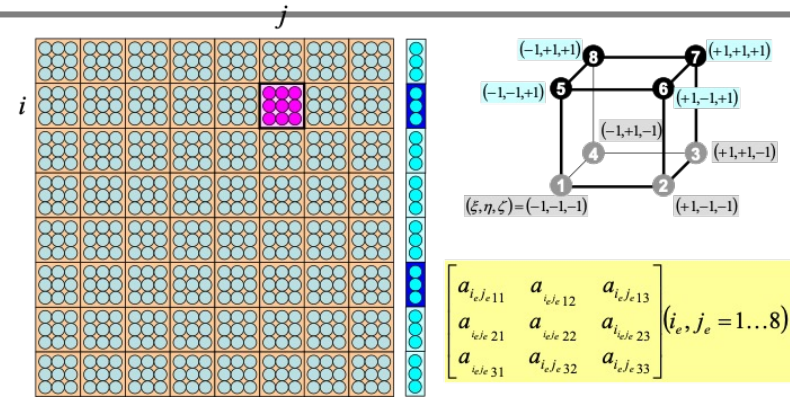
# Evaluation Policy

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- Develop a GPU-enabled version of GeoFEM/Cube using Claude Code
  - GeoFEM/Cube
    - A Fortran-based application parallelized with MPI + OpenMP
    - It has optimization track records across various environments including GPUs (here we experiment from the basic MPI+OpenMP code)
- We evaluate:
  - Performance of the code generated by Claude Code
  - Time spent on Claude Code's code-development process

# GeoFEM/Cube

- Finite-volume solver for the uniform-field Poisson equation
  - Various optimization achievements in HPC environments
  - Grid: unstructured data structure; 7-point stencil
  - Linear system with an SPD sparse coefficient matrix
- Coloring and reordering
  - CM-RCM + Coalesced/Sequential
- Matrix storage schemes
  - CRS, Sliced-ELL, Sell-C- $\sigma$
- Main components: coefficient-matrix generation and the solver



# GeoFEM/Cube File Layout

- The table on the right was generated by asking Claude Code: "Convert the directory structure into a LaTeX table."
- Fixed-form Fortran 90 (.f)
- Uses implicit real
- Parallelized with OpenMP
  - Coefficient-matrix generation files
    - mat\_ass\_\*.f
  - Solver-related files
    - solver\_\*.f
- The doc folder includes application documentation in Word and PDF

表 2: Directory Structure of GeoFEM-Cube-Hybrid\_CG.3

Directory	File	Description
./doc/	GeoFEM-Cube-3.docx	Documentation (Word)
	GeoFEM-Cube-3.pdf	Documentation (PDF)
./run/	go.sh	Execution script
	mesh.inp	Mesh input file
	test.lst	Test list file
./src/	Makefile	Build configuration
	hpcmw_all.f	HPC middleware main module
	hpcmw_fem_cntl.f	FEM control module
	hpcmw_fem_mesh.f	FEM mesh module
	hpcmw_fem_util.f	FEM utility module
	hpcmw_finalize.f	Finalization module
	hpcmw_init.f	Initialization module
	hpcmw_solver_cntl.f	Solver control module
	hpcmw_solver_matrix.f	Solver matrix module
	hpcmw_util.f	Utility module
	input_cntl.f	Input control module
	input_grid.f	Grid input module
	mat_ass_bc.f	Matrix assembly BC module
	mat_ass_init.f	Matrix assembly init module
	mat_ass_main.f	Matrix assembly main module
	mat_con0.f	Matrix construction module 0
	mat_con1.f	Matrix construction module 1
	mat_trans.f	Matrix transformation module
	solver33.f	3x3 solver module
	solver_CG.3_SMP_novec.f	CG solver (SMP, no vectorization)
	solver_SR.3.f	SR solver module
	test1.f	Test program
	util.f	Utility functions



# Coefficient-Matrix Generation

- The loop structure is very complex
- Just getting it to run in parallel is not hard
- But parallelizing efficiently on GPUs is not straightforward

Parallelizable loops →

```
1 do icol= 1, ELMCOLORTot
2   !$omp parallel do private (...)
3   do icel0= ECidx(icol-1)+1, ECidx(icol)
4     ! 中略
5     do ie= 1, 8
6       ip = nodLOCAL(ie)
7       if (ip.le.N) then
8         do je= 1, 8
9           jp = nodLOCAL(je)
10          ! 中略
11          kk= 0
12          iiS= indexU(ip-1) + 1
13          iiE= indexU(ip )
14          do k= iiS, iiE
15            if ( itemU(k).eq.jp ) then
16              kk = k
17              IDlu= 1
18              exit
19            endif
20          enddo
```

→

For efficient GPU parallelization, it's typical to move the je loop right after the ie loop, fuse (ie, je), and parallelize the fused loop

```
21 if (kk.eq.0) then
22   iiS= indexL(ip-1) + 1
23   iiE= indexL(ip )
24   do k= iiS, iiE
25     if ( itemL(k).eq.jp) then
26       kk= k
27       IDlu= -1
28     endif
29   enddo
30 endif
31 ! 中略
32 do kpn= 1, 2
33   do jpn= 1, 2
34     do ipn= 1, 2
35       ! a11, a12, ... a33 を計算
36     enddo
37   enddo
38 enddo
39 if (IDlu.eq.1) then
40   AU(...)= a11, a12,..., a33 を代入
41 endif
42 if (IDlu.eq.-1) then
43   AL(...)= a11, a12,..., a33 を代入
44 endif
45 if (IDlu.eq.0) then
46   D(...)= a11, a12,..., a33 を代入
47 endif
48 enddo
49 endif
50 enddo
51 enddo
52 enddo
```

# CG Solver

- Computation is a combination of standard linear algebra operations
  - SpMV, dot products, AXPY, etc.
  - Parallelization is easy: a mix of trivially parallel loops and reduction loops

```
Compute  $r^{(0)} = b - [A]x^{(0)}$ 
for i= 1, 2, ...
    solve  $[M]z^{(i-1)} = r^{(i-1)}$ 
     $\rho_{i-1} = r^{(i-1)} z^{(i-1)}$ 
    if i=1
         $p^{(1)} = z^{(0)}$ 
    else
         $\beta_{i-1} = \rho_{i-1} / \rho_{i-2}$ 
         $p^{(i)} = z^{(i-1)} + \beta_{i-1} p^{(i-1)}$ 
    endif
     $q^{(i)} = [A]p^{(i)}$ 
     $\alpha_i = \rho_{i-1} / p^{(i)} q^{(i)}$ 
     $x^{(i)} = x^{(i-1)} + \alpha_i p^{(i)}$ 
     $r^{(i)} = r^{(i-1)} - \alpha_i q^{(i)}$ 
    check convergence  $|r|$ 
end
```

# GeoFEM/Cube Output Example

- The following output example is taken from the Word file in the doc folder
  - In actual runs, only the black text is output
  - Unless explicitly told via prompt, Claude Code must interpret the Word file or infer the output from the code

```
128      128      128      npz, npy, npz
 2        2        1      ndx, ndy, ndz
12                                     PEsmptTOT
                                     (The number of OpenMP threads)

### NORMAL
color number:      0
### MATRIX assembly 3. 046744E-01 Elapsed time of Matrix assembly

1295      9. 866630E-09      Number of Iterations, Residual

### min/max/ave 3. 005061E+01 3. 005061E+01 3. 005061E+01
                                     Elapsed time of CG (min, max, ave)

524288 -3. 810000E+01 -3. 810000E+01 1. 270000E+02
                                     Reference point displacement (Ux,Uy,Uz)

* normal termination
```

# Experiment Details

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- Experimental Setup
  - With the same input source code and instruction, generate each case 10 times
  - Measure each generated program 5 times and take the median
- Targets
  - Input source code variants with “GPU-enable this code.”
    - .f: fixed-form Fortran 90, implicit real (original)
    - .f90: free-form Fortran 90, implicit none
    - .f90\_woOMP: .f90 with OpenMP directives removed
  - .f90 as the input, target programming model is specified with prompt
    - omp: GPU porting using OpenMP target based on .f90
    - acc: GPU porting using OpenACC based on .f90
    - cuda: GPU porting using CUDA Fortran based on .f90
  - Further optimization
    - fast1: based on acc, instruct “make it faster”
    - fast2: based on fast1, instruct “make it even faster”
    - fast3: based on fast2, instruct “make it even faster”

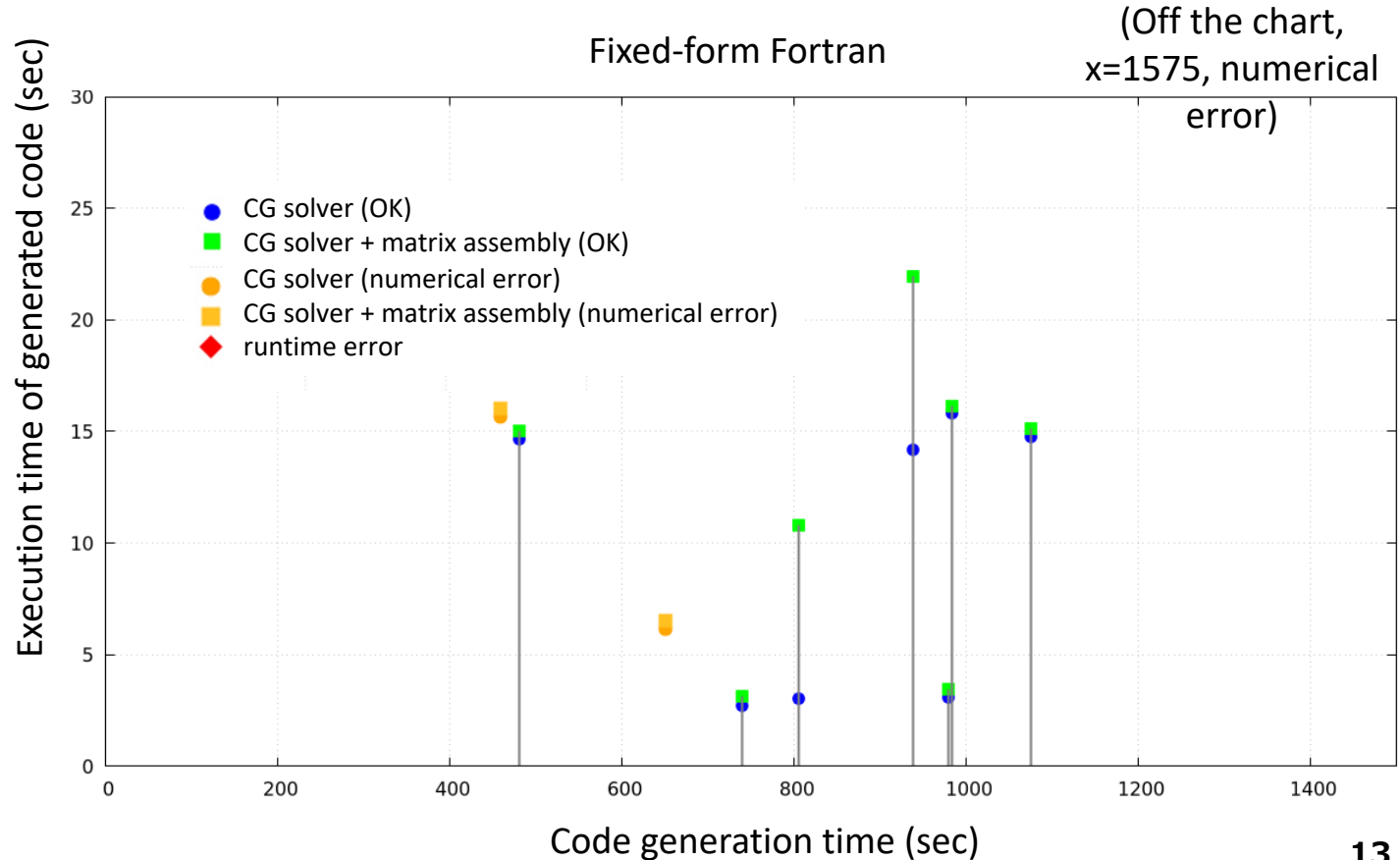
# Evaluation Environment

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- Miyabi (JCAHPC: Univ. of Tokyo & Univ. of Tsukuba)
  - GH200
  - CPU–GPU connected via NVLink C2C (450 GB/s per direction), cache-coherent
- Software environment
  - GPU compiler: nvfortran 24.9 (default)
    - Claude Code could have used other compilers, but in practice it used this for all cases
  - Claude Code ver.1.0.83 ~ 1.0.90
    - Updated frequently, so we could not pin a fixed version
    - All models were Claude Opus 4.1

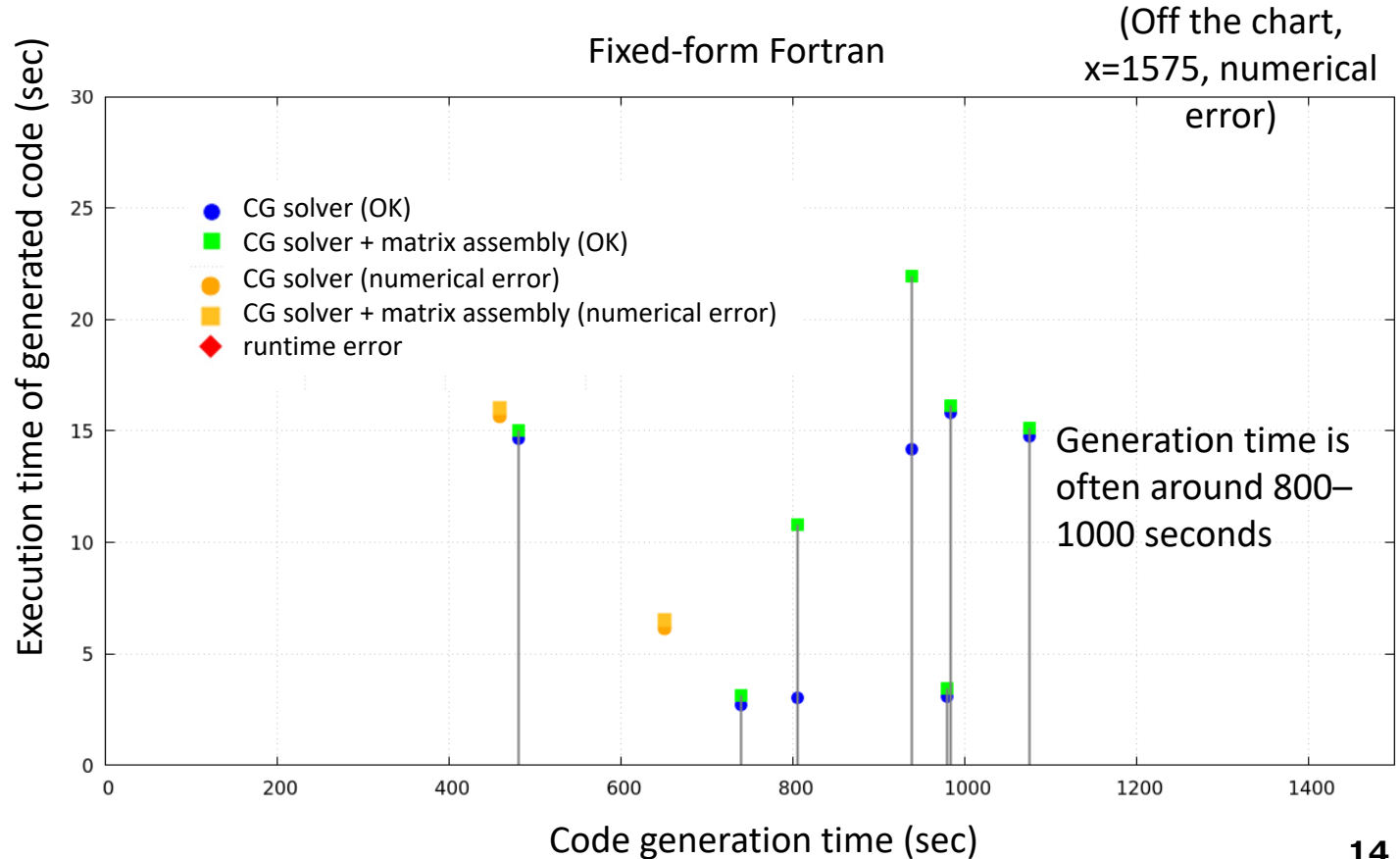
# Input source code variant (.f)

- Blue dots: solver time only
- Green dots: solver + matrix generation
- Yellow/red: did not run correctly
- At least, it judged from the code that the CG solver should be sped up, and it largely succeeded in GPU-ifying it



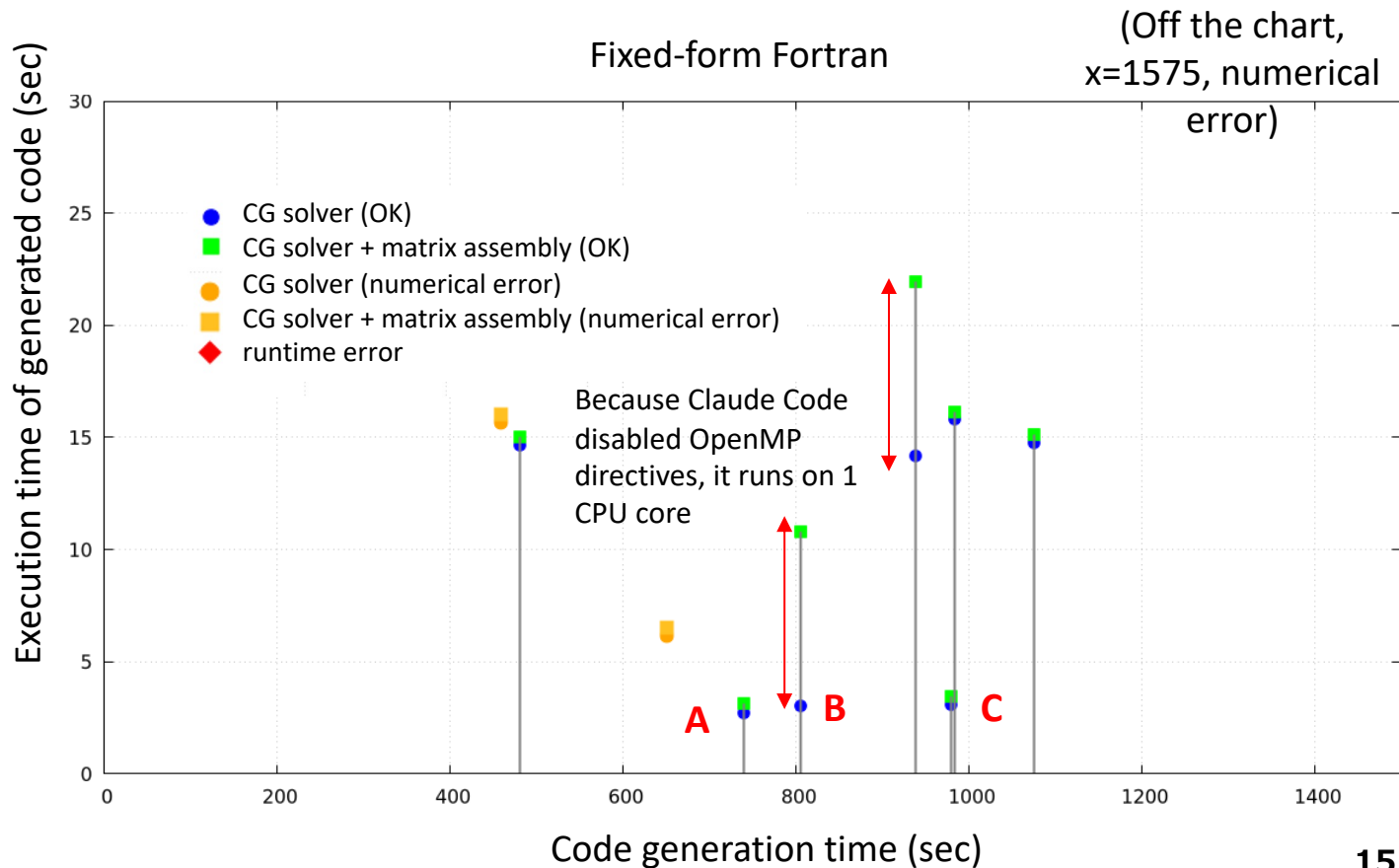
# Input source code variant (.f)

- As with human coding, fixed-form line limits and implicit typing cause bugs, increasing generation time
- Off-chart points indicate it attempted matrix generation and ultimately failed



# Input source code variant (.f)

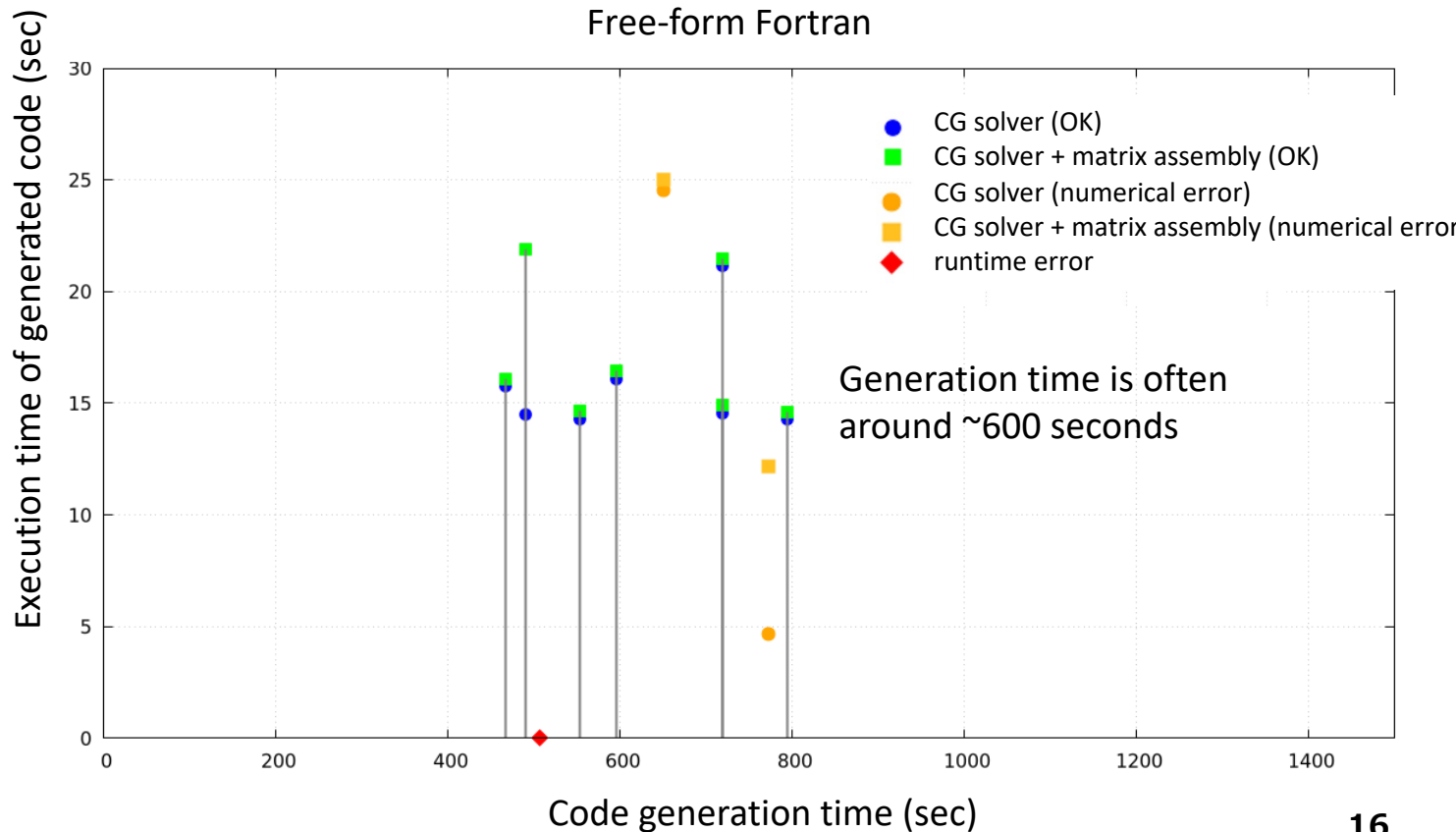
- GPU porting of the matrix-generation part never succeeded
- A and C use OpenACC kernels directives and achieve reasonable performance for the solver part
- B uses OpenACC parallel directives and specifies clauses appropriately
- The others use parallel directives but without appropriate clauses





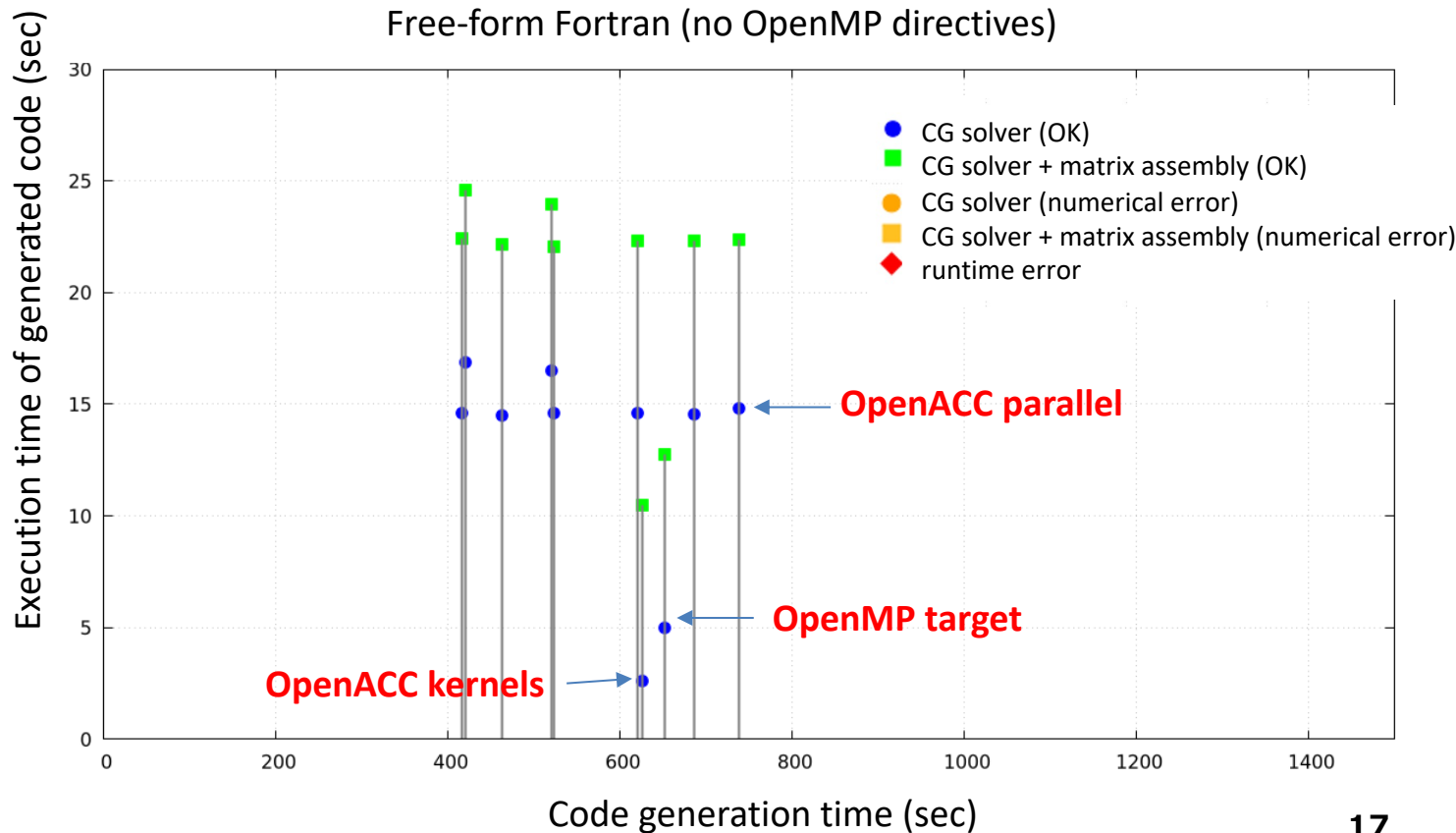
# Input source code variant (.f90)

- Generation time is clearly shorter
- With .f it sometimes outputs performance-appropriate code, but with .f90 it does not (reason unknown)



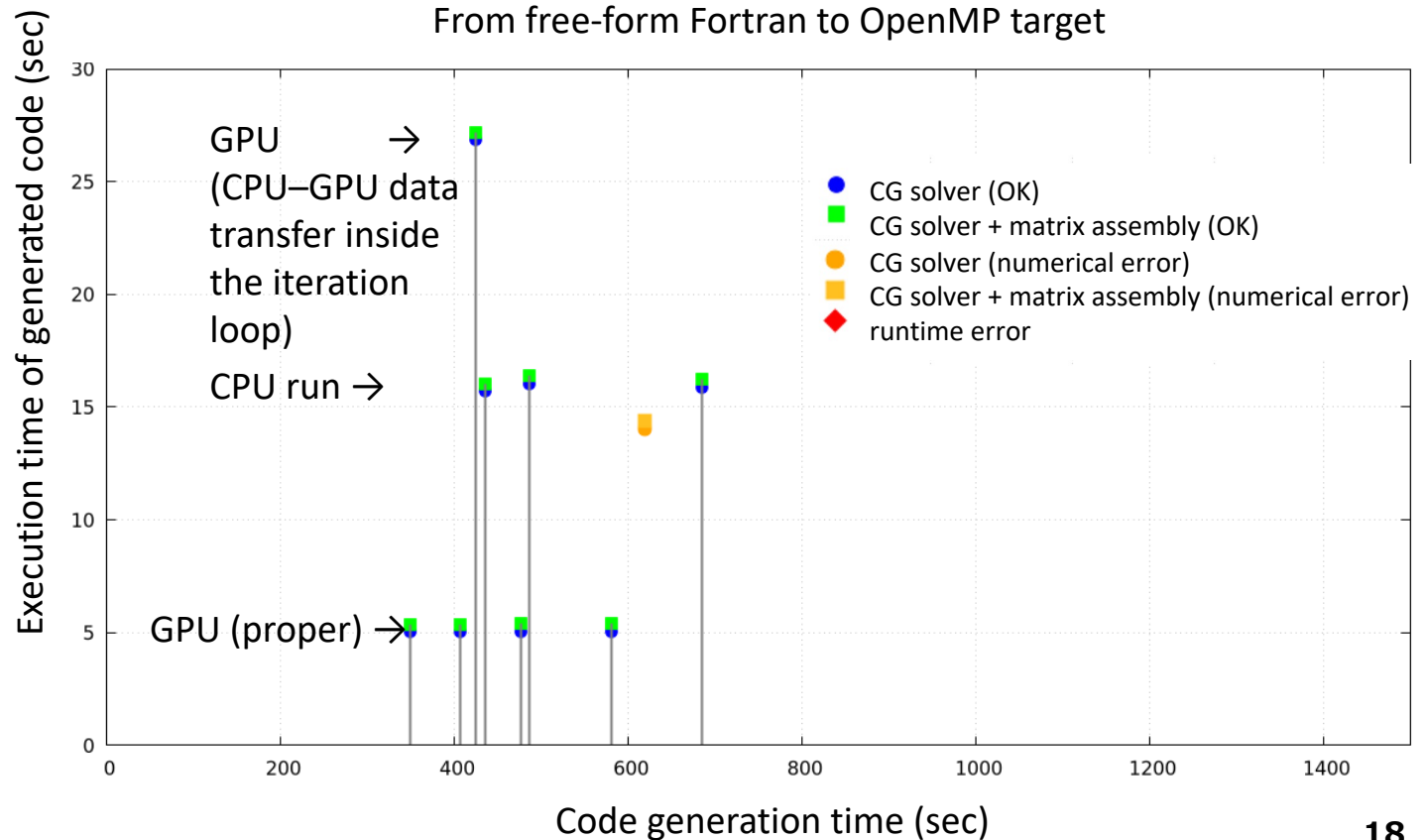
# Input source code variant (.f90 without OpenMP directives)

- With OpenMP directives removed, it stopped trying to GPU-ify matrix generation; as a result, success rate increased and generation time decreased
- The matrix-generation part runs on 1 CPU core



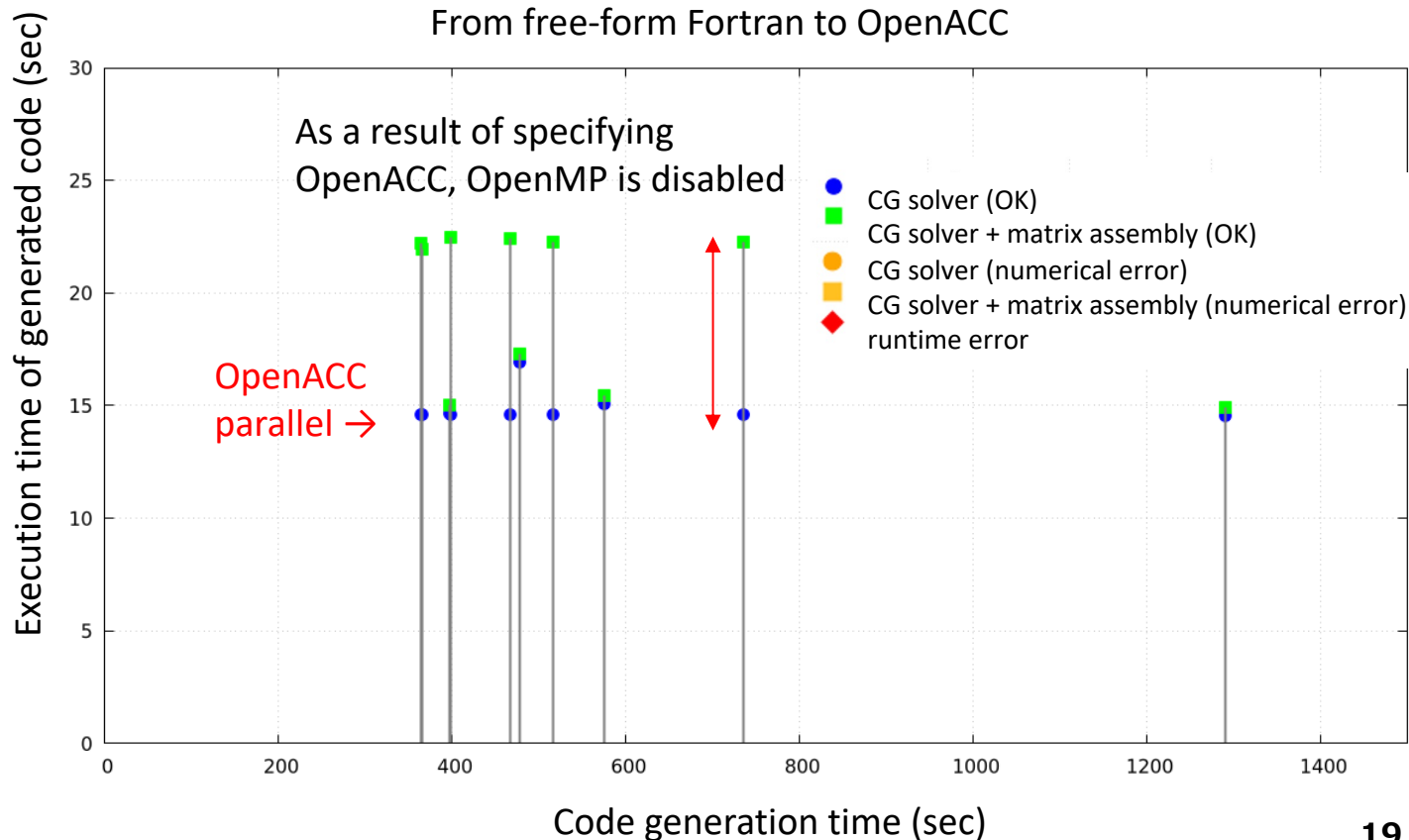
# Target programming model (OpenMP target)

- Only 5 cases actually succeeded in running on the GPU
- Generation is generally fast



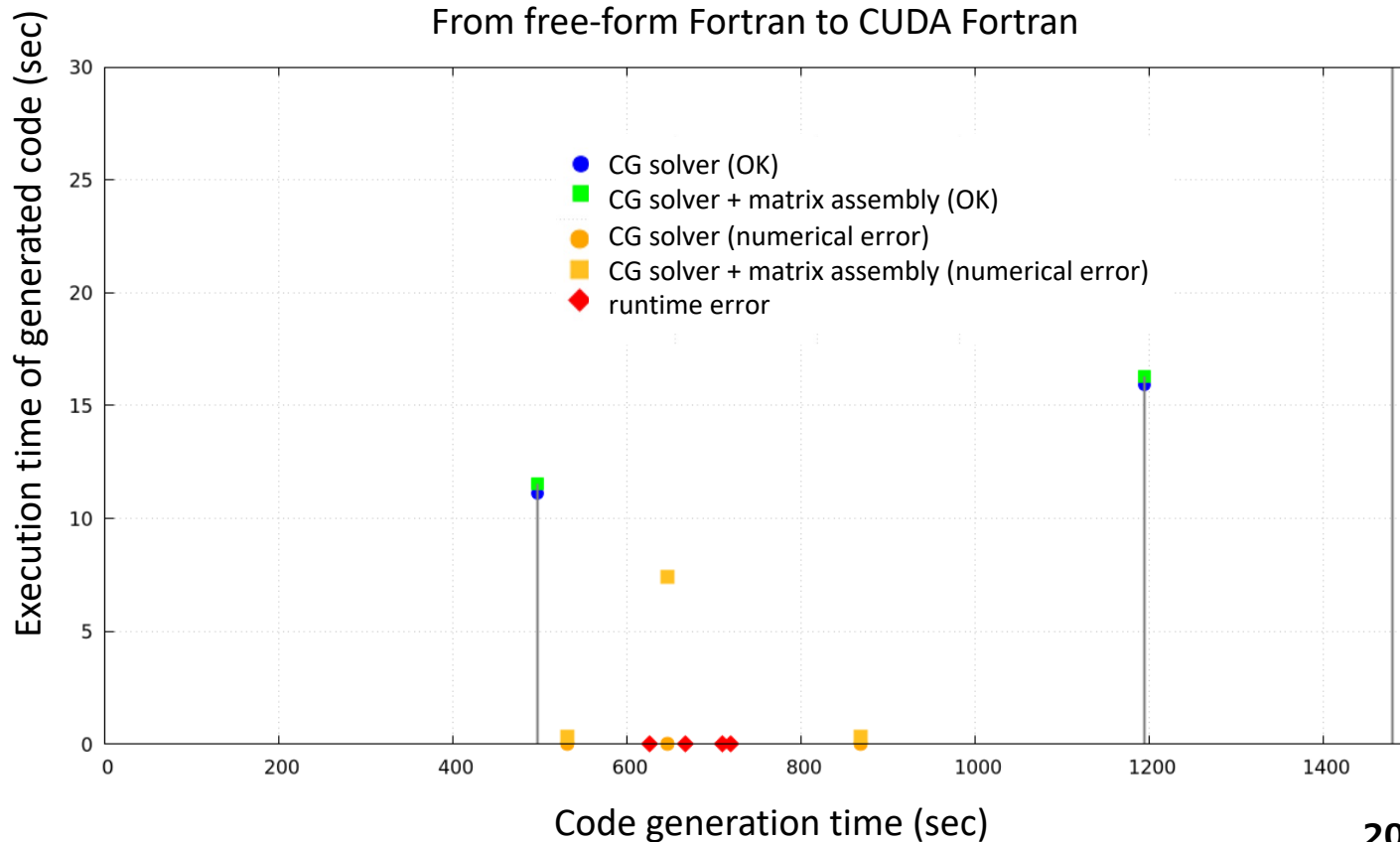
# Target programming model (OpenACC)

- High likelihood of generating runnable GPU code
- All use OpenACC parallel directives; because clauses are not specified appropriately, performance is poor (reason unknown)
- Specifying OpenACC removed OpenMP flags from the Makefile, so much of matrix generation runs on 1 CPU core



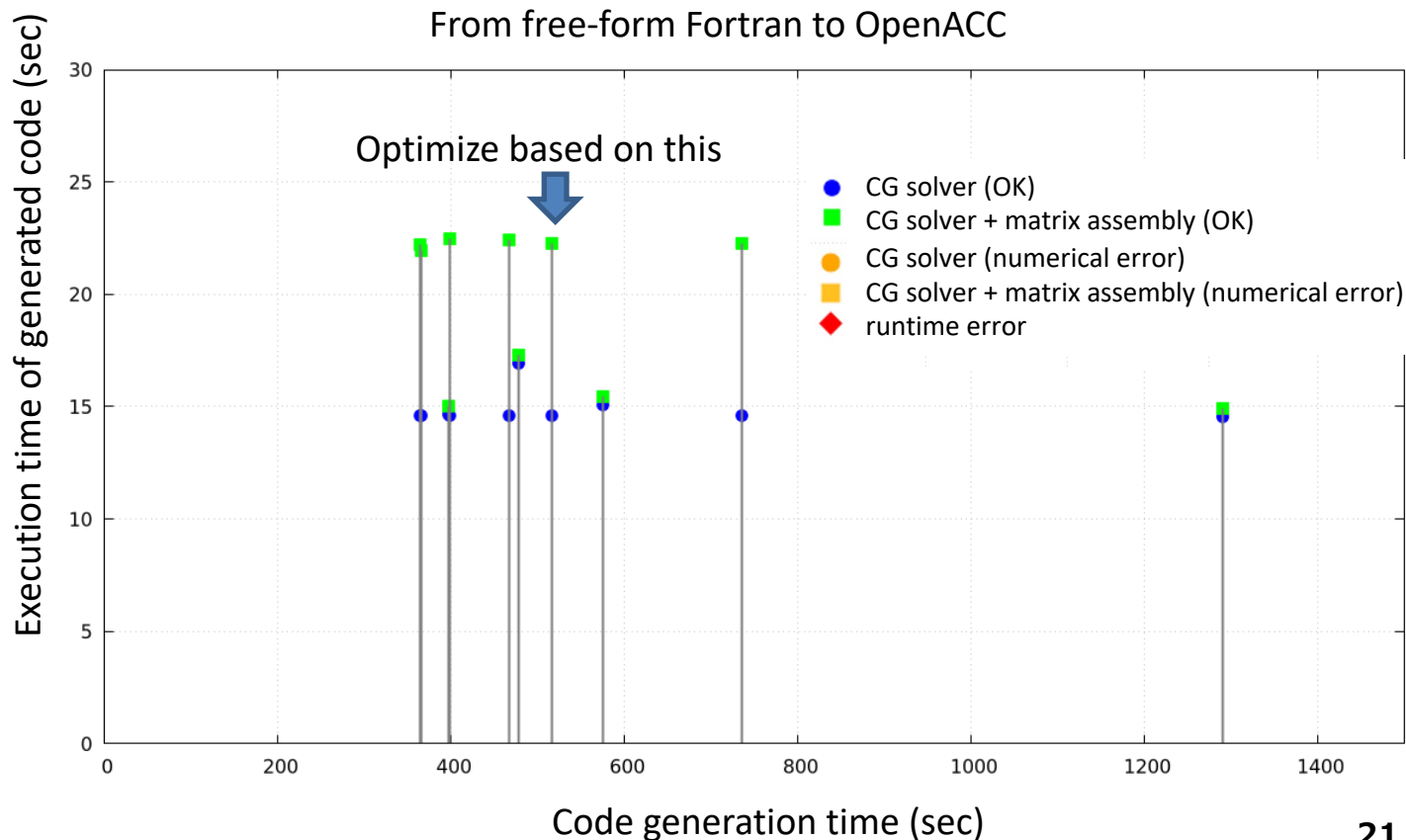
# Target programming model (CUDA Fortran)

- Success rate is very low
- Even successful cases are not fast
- Switching the output to CUDA C might work better, but that is future work



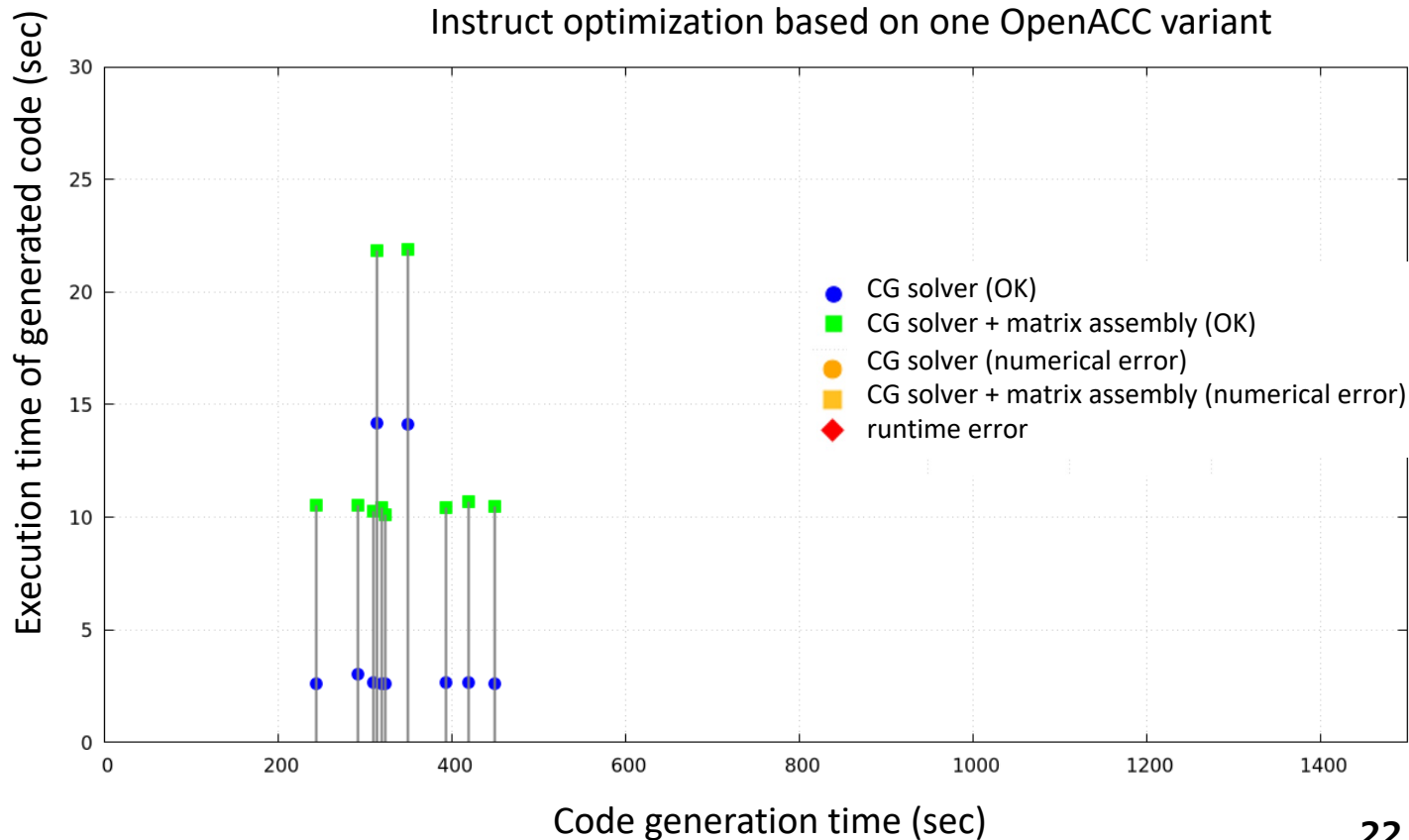
# Further optimization (Level 0)

- Further optimization starting from slow code that uses OpenACC parallel directives with inappropriate clauses



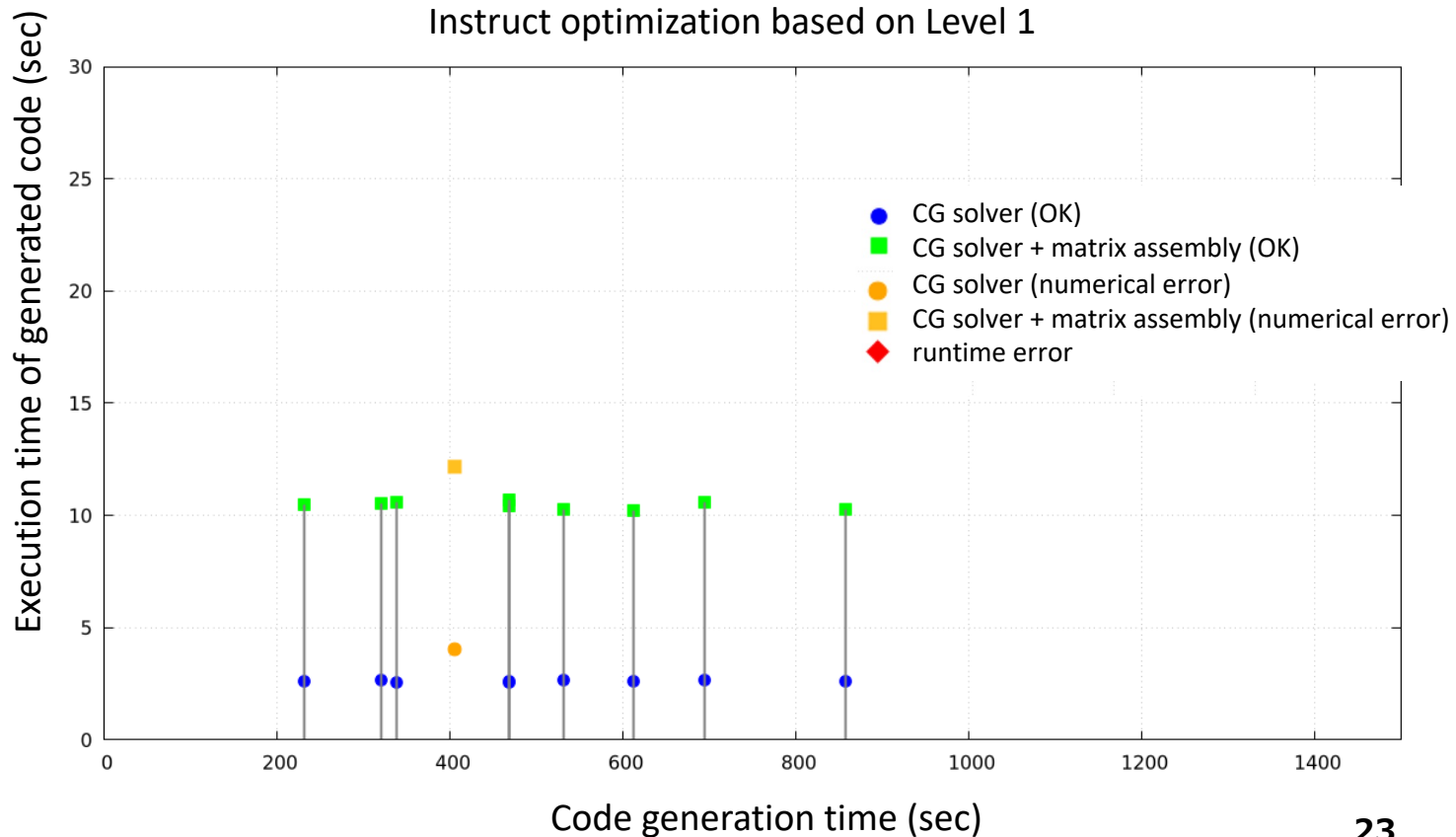
# Further optimization (Level 1)

- In many cases, speedup was achieved by specifying parallel directive clauses appropriately



# Further optimization (Level 2)

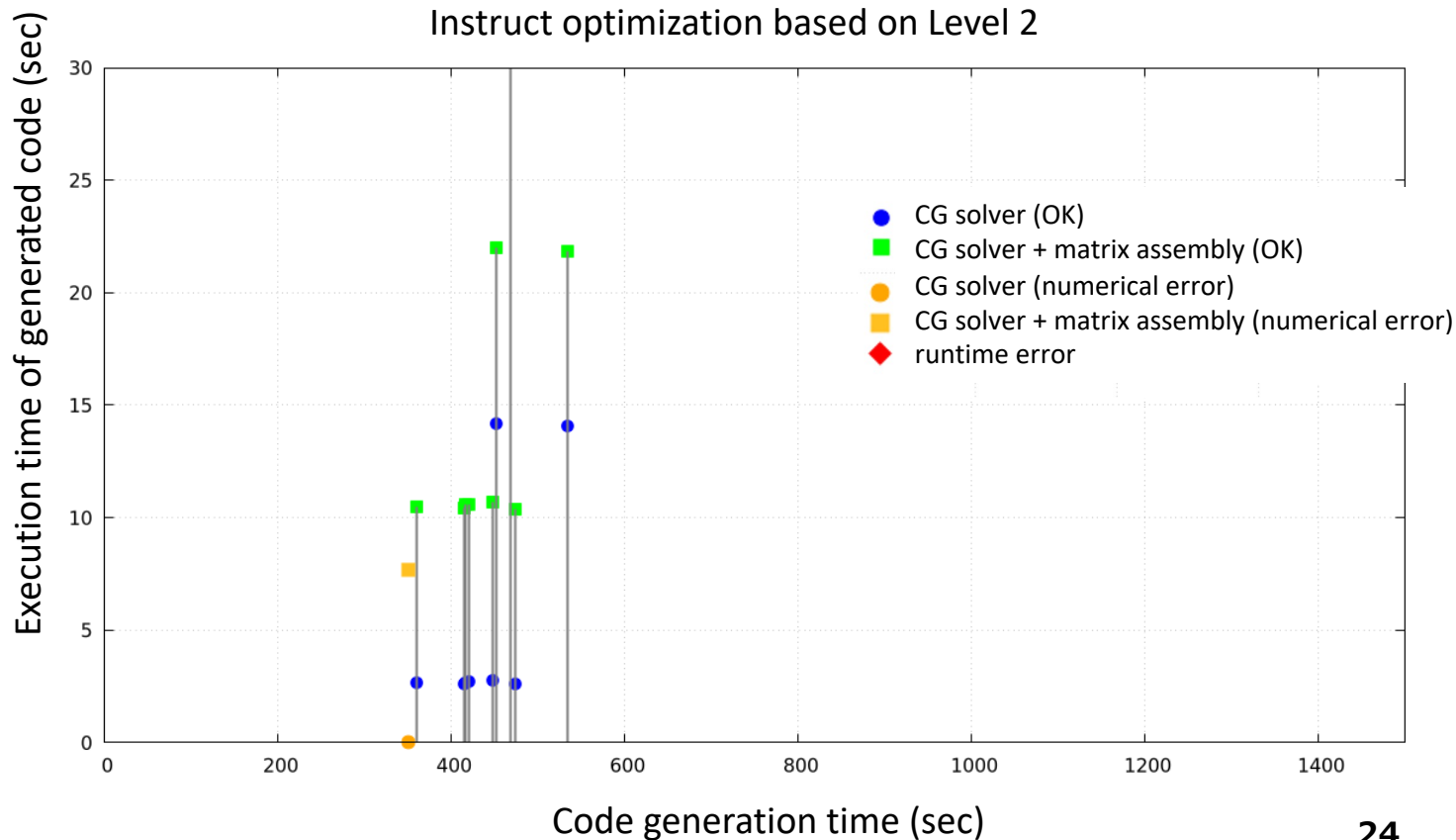
- In many cases, an async clause was added and performance improved slightly (a common technique)
- For the CG part, it successfully inserted OpenACC parallel + gang/vector + async clauses appropriately
- The matrix-generation part is not accelerated



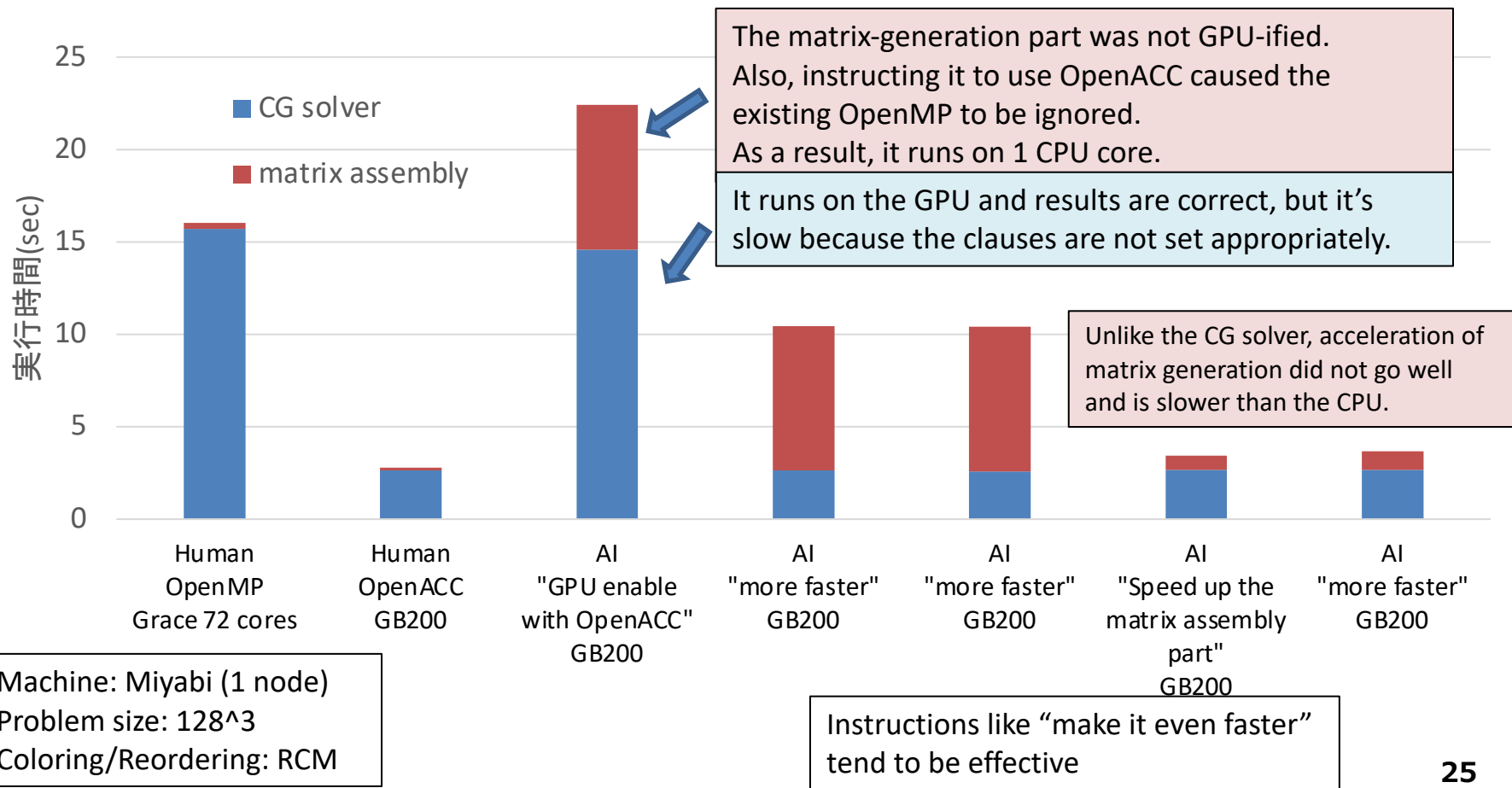


# Further optimization (Level 3)

- Many cases failed and became slower
- Even though matrix generation clearly dominates runtime, no optimization was performed



# GeoFEM GPU Porting by Claude Code Summary



# Discussion

- For the CG solver, it judged it should be accelerated regardless of OpenMP directives and generally succeeded in generating GPU code
  - Success rate is high with OpenACC
    - Likely because there is a lot of existing GPU-ported Fortran code
    - If you instruct “further optimization,” you generally get reasonably appropriate code
  - Because it already “knows” a parallel CG solver?
- For matrix generation, even with OpenMP directives present, parallelization is not done (or fails even if attempted)
  - Because it does not “know” a parallel matrix-generation implementation?

If it were a human,  
“This loop can be parallelized” → apply directives  
A code-generation AI,  
“This is CG” → “the corresponding parallel CG is this”  
...maybe that’s the difference?

# Summary

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- We evaluated an AI code generator's ability to develop a GPU-enabled version based on the MPI+OpenMP parallelized, Fortran-based GeoFEM/Cube
- Even for the same computation, development time varied greatly depending on the input source variant (fixed-form vs free-form Fortran, etc.)
- The CG part was generated well, but the highly bespoke matrix-generation part did not go well
  - If instructed to speed up matrix generation, it can at least generate code that runs on the GPU
- Future work
  - Develop methods to improve GPU code-generation success for highly bespoke parts
    - E.g., add in-code guidance like “parallelize this loop and this loop”

## Additional experiments beyond the paper: GPU porting of matrix generation

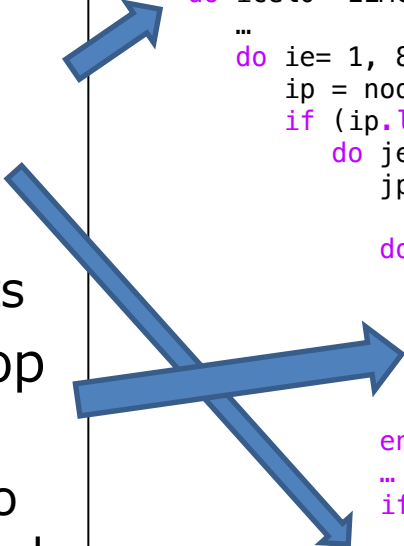
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- Focus on matrix generation and instruct: “speed it up on the GPU”
  - Succeeded in generating code that can run on GPUs
  - But performance is slow
- Instruct further optimization
  - It created v1–v3 on its own, but it didn’t get faster
    - v1: simple parallelization
    - v2: loop unrolling (runnable but slow)
    - v3: memory-access optimization (compile error)

# What's the problem?

- The intent of the OpenMP version's parallelization strategy was not conveyed
  - By coloring,
  - avoid write conflicts
- Doesn't consider loop length (?)
  - $2 * 2 * 2 = 8$  is too short for GPU thread-level parallelism

## OpenMP version

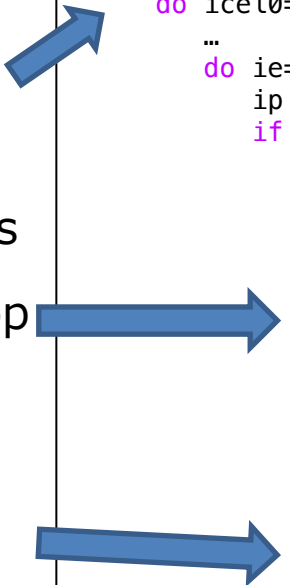


```
do icol= 1, ELMCOLORTot
  !$omp parallel do private (...)
  do icel0= ELMCOLORindex(icol-1)+1, ELMCOLORindex(icol)
    ...
    do ie= 1, 8
      ip = nodLOCAL(ie)
      if (ip.le.N) then
        do je= 1, 8
          jp = nodLOCAL(je)
            ...
            do kpn= 1, 2
              do jpn= 1, 2
                do ipn= 1, 2
                  ...
                enddo
              enddo
            enddo
          ...
          if (IDlu.eq.1) then
            AU(9*kk-8)= AU(9*kk-8) + a11
          endif
          ...
        enddo
      endif
    enddo; enddo; enddo
```

# What's the problem?

- Loop-parallelization strategy is not great
  - Pattern: simply parallelize OpenMP-parallel loops with (gang, vector)
  - Or: use gang for OpenMP-parallel loops and collapse(3) the innermost  $2 \times 2 \times 2$  loop with vector
- Unnecessary atomic operations
  - If the ie/je loops are not parallelized, coloring avoids conflicts

OpenACC version generated by Claude Code



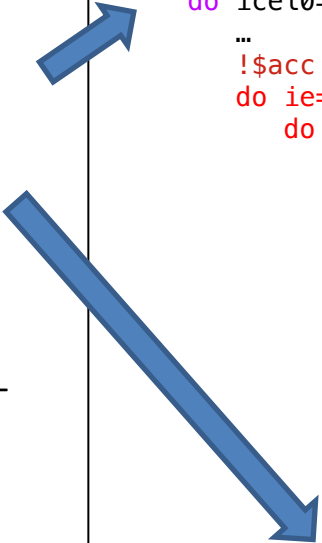
```
do icol= 1, ELMCOLORtot
  !$acc parallel loop gang vector private(...)
  do icel0= ELMCOLORindex(icol-1)+1, ELMCOLORindex(icol)
    ...
    do ie= 1, 8
      ip = nodLOCAL(ie)
      if (ip.le.N) then
        do je= 1, 8
          jp = nodLOCAL(je)
            ...
            do kpn= 1, 2
              do jpn= 1, 2
                do ipn= 1, 2
                  ...
                enddo
              enddo
            enddo
          enddo
        enddo
      enddo
    enddo
    if (IDlu.eq.1) then
      !$acc atomic update
      AU(9*kk-8)= AU(9*kk-8) + a11
    endif
    ...
  enddo
endif
enddo; enddo; enddo
```

# What's the problem?

- Reorder loops so ie/je can be collapsed, then vector-parallelize
- Then use atomic operations

It's important to teach the code-generation AI the preconditions for (non-)parallelizability

Desired OpenACC version



```
do icol= 1, ELMCOLORTot
  !$acc parallel num_gangs(...) vector_length(64) loop gang private(...)
  do icel0= ELMCOLORindex(icol-1)+1, ELMCOLORindex(icol)
    ...
    !$acc loop collapse(2) vector(64)
    do ie= 1, 8
      do je= 1, 8
        ip = nodLOCAL(ie)
        if (ip.le.N) then
          jp = nodLOCAL(je)
          ...
          do kpn= 1, 2
            do jpn= 1, 2
              do ipn= 1, 2
                ...
              enddo
            enddo
          enddo
        ...
        if (IDlu.eq.1) then
          !$acc atomic update
          AU(9*kk-8)= AU(9*kk-8) + a11
        endif
        ...
      enddo
    enddo
  endif
enddo
```