MIRGE-Com

Y3 Scaling Status

Fr@CEESD - October 13, 2023



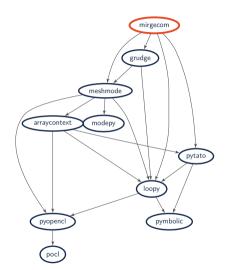
Mike Campbell & CEESD





Outline

- ► MIRGE-Com Overview
- ► Performance & Scalability
- ► Recent Challenges
- Conclusion & Future Directions



https://github.com/illinois-ceesd/mirgecom/



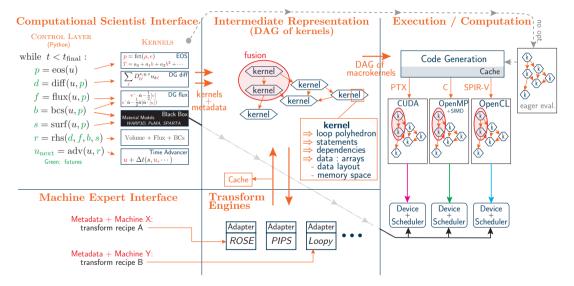


MIRGE-Com in Y3





Architecture Overview







Simulation Infrastructure

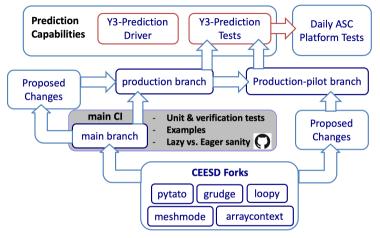
- Infrastructure provides:
 - DG operators (e.g., mass, stiffness, grad, div)
 - Multiple parallel discrete geometries

 [M. Smith]
 - Compute device access
 - Symbolic infrastructure great for verification

- ► *MIRGE-Com* library provides:
 - Conservation-law-specific data structures
 - Simulation / driver API
 - Prediction-relevant
 - * RHS operators
 - * Model-specific constructs (e.g., EOS, transport, reactions)
 - * Boundary and numerical fluxes

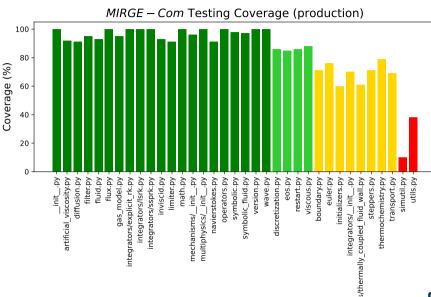


Prediction-supporting Development



- Prediction-supporting development process
- Prediction-targeted testing mechanism
- Continuous integration and daily ASC platform testing





Performance & Scalability





Performance & Scalability

- Computational performance: Speed or FLOPS
- Parallel performance: Scalability or efficiency
 - Strong scaling: Fixed problem size, scaling resources
 - Weak scaling: Fixed work/resource, scaling problem size with resource
 - Mesh scaling: Scaling work, fixed resource
- ▶ I/O and memory performance: Bandwidth, bytes/second
- Cost performance
 - Energy efficiency
 - Total time to solution (TTS)
 - FLOPS/Fidelity
 - Usability/Productivity
- Why should we care about scalability? (enables prediction!)
 - Weak scaling out to prediction-scale
 - Versatility, portability, and resource options



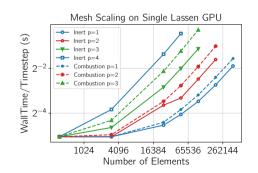


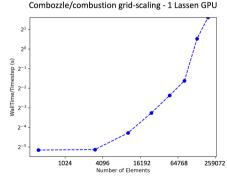
Performance & Scalability

New: prediction-enabling performance

- Scaling as expected (mostly)
- Small problems are expensive

- ► OOM: SVM/Unified memory
- ► Mem growth: Garbage collection





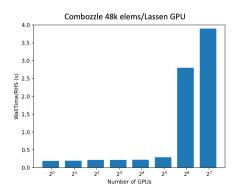




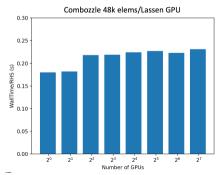
DAG Splat Effect

New: prediction-enabling performance

- ► DAG Splat: DAG for each boundary
- Limits weak scaling DAG for each neighbor



- lacktriangle Mitigation: Metis ightarrow 1D decomp
- Real fix: Function calls in the DAG (aka function outlining)





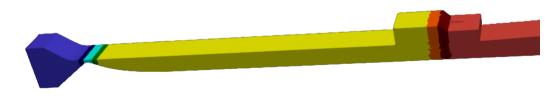


DAG Splat Mitigation

New: prediction-enabling performance

- DAG Splat: DAG for each boundary
- Limits weak scaling DAG for each neighbor

- ightharpoonup Mitigation: Metis ightharpoonup 1D decomp
- Real fix: Function calls in the DAG (a.k.a. outlining)



Y2 Domain Decomposition



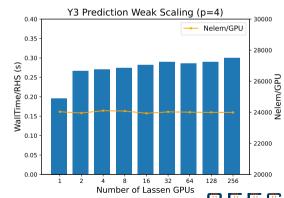


Scaling with 1D Partitioning

New: prediction-enabling performance

- DAG Splat: DAG for each boundary
- Limits weak scaling DAG for each neighbor
- Y2 KS3D Weak Scaling 0.200 100000 Nelem/GPU 0.175 95000 0.150 WallTime/RHS (s) 0.125 0.100 0.075 90000 85000 80000 0.050 75000 0.025 0.000 70000 32 Number of Lassen GPUs

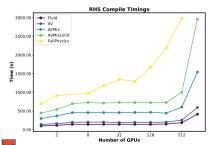
- Mitigation: Metis vs. 1D decomp
- ▶ Real fix: Function calls in the DAG (a.k.a. outlining) [M. Smith]

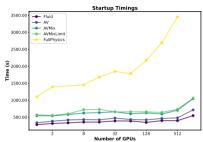


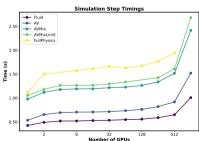


Prediction Performance Snapshots: Lassen DAT

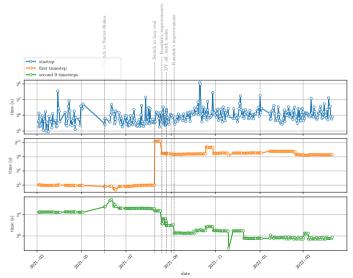
- ▶ Weak scaling: 24k p4 per GPU
- ► Fluid: NS-only, 2 species
- ► AV: Fluid + Artificial Viscosity
- ► AVMix: AV + 7 species mixture EOS
- AVMixLimit: AVMix + species limiter
- FullPhysics: AVMixLimit + PLTransport+ Wall









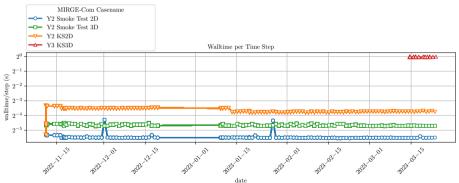






- Data on key capabilities collected nightly
- ► Intended to track performance vs. code/features
- $ightharpoonup \Delta$'s indicate change in performance

- ► New this cycle:
 - Multi-case/comparitive plotting
 - Tracking parallel cases







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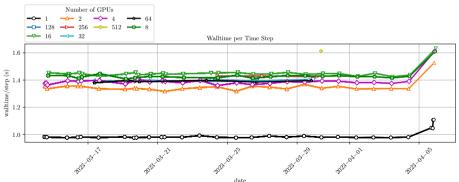






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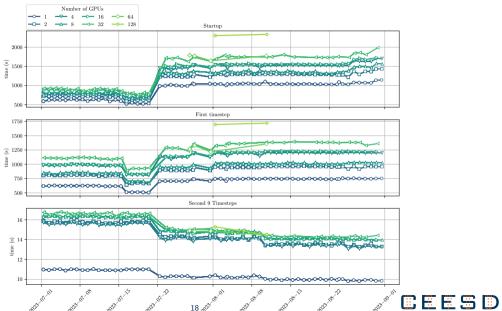
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 - Multi-case/comparitive plotting
 - Tracking parallel cases







Recent Y3 Monitoring on Lassen





Recent Challenges





Recent Challenges and Looming Threats

- ► Loop nest error and indeterminism [M. Diener, M. Smith]
- DAG splat 1D part seen as looming threat
- Super long compile times for essential models
- ► Mesh I/O and processing
 - Mesh distribution fell down at scale (mpi4py@pkl5) [M. Diener]
 - Gmsh ingest uses far too much memory
 - * Prevented prediction scaling past 512 ranks, precluded machine-scale runs
 - * Pre-partitioning work-around gets us to 1024, will 2048 test today!
- M-to-N restart: needed for simulation portability (now working for prediction)





Wrapping Up & Looking Ahead





Summary and Next Steps

- MIRGE-Com has Y3 prediction-supporting performance (poised to deliver more)
- Understanding MIRGE-Com performance is the next major focus

Next steps

- Understanding and improving performance:
 - Instrumentation (Mem & Tags) [M. Diener]
 - Code-to-kernel correspondence improvements: [M. Diener]
 - Auto-tuning [Nick Christensen]

- DAG Splat [M. Smith]
- Performance model
- Upcoming enhancements:
 - Workflow: Parsl [D. Friedel]
 - Hexahedral elements [Addison Alvey-Blanco]
 - M-to-N restart (done!)







Questions?

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