Simulating Additive Manufacturing By Rethinking Simulation Structures

Jay Lofstead, John Mitchell
Enze Chen, Manisha Ganesh, Gavin St. John

gflofst@sandia.gov
Mod/Sim Typical Structure

- Examples:
  - CFD, heat transfer, molecular dynamics

- Compute on every element every timestep

- Mesh types vary, but most are 3D today

- Deploy across machine balancing compute vs. communication
Another class of physical models

- Activity focused in a small subset of simulation domain
  - Welding
  - Additive manufacturing (small parts)
  - Spray
  - Pressure waves in material

- Domain too large to fit into memory
  - Additive manufacturing at scale
    - Satellite fuel tank cap 3D printed (1.16m diameter, 10 cm thick)
Welding example

- 1 site per micron
- Move heat affected zone across powder or piece edges to melt and form new piece
Spray accumulation model

- Spray a material (e.g., paint) and model the accumulation
Additive Manufacturing (single layer)

- Similar to welding, but sweeping and ultimately multi-layer
- Move heat affected zone across powder or piece edges to melt and form new piece

AM demo simulation
What these share

- In all cases, the area that changes after a computation round is highly localized

- Total area can be large leading to 1% or less use of allocated compute resources

- IO costs would dominate and storage space used would be overwhelming with vast unchanging data between timesteps
New Approach

- Idea: Only compute on part of domain that will be affected “soon”
  - Split simulation run into a series of computational volumes and write data progressively

- Reduce computation footprint by 99%+
- Reduce computation time by eliminating IO and reducing communication costs

- Challenges:
  - How do we make the compute approach work?
  - How do we store data to enable analysis later?
Can’t Current Models Work?

- Existing compute models can—with a little help

- Existing IO models cannot since they depend on whole domains being written
  - HDF5, NetCDF, PnetCDF, ADIOS, and others all have this “feature”
Addressing Computation

- Rethink as a series of small problems
- Build “glue” for overlapping areas and to initialize unvisited areas
- Make traversal algorithm and scripting that also handles resilience
Illustrative Example (SPPARKS)

Grain growth across a large domain is simulated using a series of smaller overlapping sub-volumes.

Post-process, visualize and analyze on arbitrary sub-volumes and arbitrary times.
Illustrative Example

- Orange/Purple/Green vs. Blue
Addressing IO

- Rethink how IO should work:
  - Lazy
    - Only track what has been seen so far (i.e., we don’t care about the size of the simulation domain)
  - Minimal
    - Only write what has changed since last output
  - Eventually Consistent
    - Rely on the output to eventually “make sense”
  - Construct arbitrary requested domain on demand

- Reading specifies an arbitrary region and a time; Stitch-IO assembles (‘stitches’) the region state together from various pieces using the newest for every point
Stitch-IO – A New IO Approach

- Stitch-IO changes the rules
  - Does not require global domain setup
  - Offers support for combining data from multiple outputs into a single blob
    - Selecting a region that goes behind the active region will get previously completed data even though it has not been written during the latest output
  - Uses floating point numbers (with absolute and relative tolerances) for identifying a time epoch
  - Uses a standard format for easy, direct access from other tools (SQLite)
  - Supports writing at both the current, new time and older (existing or not) times without ill effect.
  - Simultaneous writing and reading is assumed and fully supported.
Stitch-IO – A New IO Approach

- Written in C with a full capability Python module interface
- Has both a serial (single process) and parallel (MPI-based) interface for C and Python.
  - The only difference is passing a communicator to the `open’ command. Calls are invisibly parallel and efficient if can be optimized.

- Play along at home:
- [https://github.com/gflofst/Stitch-IO](https://github.com/gflofst/Stitch-IO)
- You can build on a basic Linux setup for most clusters trivially with late model Python3, numpy, and a C compiler (and MPI for parallel builds).
Stitch-IO Schema

- SQLite storage format opens easy direct data access
  - Extend functionality without changing library using Python native API
Example Application Use

Digital twin creation for Additive Manufacturing

- Use a .STL file as source
- Use Slic3r to generate g-code to drive simulation
  - Fixup the g-code into an execution script
  - Run the simulation
- Use AM machine slicer to generate machine specific g-code
  - Run the AM machine with the g-code
- Use CT Scanner to get internal images of physical part to compare against simulation
- Use simulation results to test other physics
A series of computation volumes

- Limit simulation domain in memory to just a small part that we will compute over.

- HAZ = Heat Affected Zone
- Laser Path is the AM machine laser path
Slicing

- Slic3r used to generate paths
- Alternating layers. Red outline is perimeter, blue is laser path with a fill angle of 35 degrees.
Path Pruning

- Get rid of places where the laser is turned off just to move to a new starting position
How Stitch-IO Reduces Computation

- Computational volumes limit total domain
- Left: Geometry and Slic3r z-cut path lines
- Middle: Computational volumes bounding boxes
- Right: Simulated layer (red border is no data present)
Staircase example
Forming the Staircase

- 4 different paths
  - (UDLR, RLUD, DURL, LRDU)

- Each colored dot is metallic powder. The larger colored areas are metal grains formed by the laser melt.

- Top graphic shows the laser path. It starts in each corner iteratively and follows a different path.

- Bottom graphic shows errors in the edges of a staircase causing weaknesses and the large color bars indicate large grain growth.
Stitch-CAD for Digital Twins

Simulate rather than build and destructively test

- From CAD file that can be used by the AM machine
  1. Use Slic3r to generate path (G-code)
  2. Split G-code into layers
  3. Prune each layer
  4. For each laser path
     1. Calculate a computational volume
     2. Initialize the computational volume
     3. Run simulation
  5. End for

- See microstructure generation identifying parameter issues that will generate parts with inadequacies.
- Iterate in simulation to find the best parameters
API Basics

- from stitch.libstitch import libstitch
- (rc, file_id) = libstitch.open(“filename”)
- rc = libstitch.close(file)
- (rc, new_time) = libstitch.write_block(file, field_id, timestamp, block, state)
  - Timestamp is a real
  - Block is 6 ints representing the (x,y,z)-(x,y,z) min max pairs
  - State is a linearization of the block (typical memory layout)
  - new_time is a flag to indicate if this time has been used before or not as a sanity check
- (rc, state, new_time) = libstitch.read_block(file, field_id, timestamp, block)
  - Same as write
API Basics

- `(rc, field_id) = libstitch.create_field(file, 'field_name', 1, 1, -1)`
  - Type 1 is 32-bit int (C has an enum)
  - Length 1 is single element (C has an enum)
  - Default value is -1

- `(rc, field_id) = libstitch.query_field(file, 'field_name')`

- `(rc, times) = libstitch.get_times(file)`
  - list of times used

- `(rc, field_ids, labels, t, lengths, no_value_presents) = libstitch.get_fields(file)`
  - The field attributes
API Basics

- $rc = \text{libstitch.set\_parameters} (\text{file, abs\_tol, rel\_tol, nvp})$
  - Optional as these are defaulted
- $(rc, \text{abs\_tol, rel\_tol, nvp, first\_time, last\_time}) = \text{libstitch.get\_parameters} (\text{file})$
  - Mainly to get the min and max time in the file
- $rc = \text{libstitch.set\_field\_no\_value\_present} (\text{file, field\_id, nvp\_val})$ – value to initialize areas with no value
Benefits and Challenges

- Move from 1000s process to 10s (cluster to a laptop)
- Radical data size reduction (about 1% or less losslessly)
- Wall clock time the same or smaller (less output time)
  - Spray model went from > 24 hours to 6-8 hours; latest changes may reduce that to 4 hours

- New simulations are pushing things hard (400+ million blocks)

- Open Source (LGPL) at https://github.com/gflofst/Stitch-IO
- Paper at IPDPS 2020

- Email: gflofst@sandia.gov
Other ideas for using Stitch-IO

- Any image data-based application
- CT scans
  - Explore regions across images
Future Work

- Continue to work on scalability with SPPARKS
- Additional application examples and other domains
- Working with any of you on new problems
- gflofst@sandia.gov