

github.com/diatomic/LowFive



# LowFive: In Situ Data Transport for High-Performance Workflows

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"Somewhere, something incredible is waiting to be known."

-Carl Sagan

HDF5 Users Group Meeting '23 August 17, 2023 Tom Peterka tpeterka@mcs.anl.gov Mathematics and Computer Science Division

# **Design Choices**

A balance between user's view of data (productivity) and the workflow's efficient movement of data (performance)

Design Criteria	LowFive Choices
User's view of data (model or schema)	HDF5 data model
In situ transport mechanism (direct, staging)	Direct, parallel, MPI point to point messages
Software stack intercept location	High-level HDF5 metadata
Software design	Standalone HDF5 VOL plugin

# In Situ Data Transport Mechanism

### Staging

- Dedicated resources for transport
- Decouple producer from consumer (could allow overlap)
- May require launching a separate service
- Shared access (could also involve locking)

#### Direct

- No additional resources or services
- Simple, point-to-point communication
- Tightly coupled producer and consumer (synchronous)
- A staging area could still be a producer/consumer task



# Software Stack Intercept Location



• No metadata

Software Stack Intercept Location



• No metadata

### Software Stack Intercept Location



• No metadata



# LowFive Metadata Tree



HDF5 Data Model

- Hierarchical data model much like a UNIX file system
- Root is the file
- Internal nodes are groups
- Leaves are datasets or other objects (e.g., attributes)

#### LowFive Data Model

- Our in-memory replica of HDF5 metadata
- One object for every HDF5 object
- Shallow or deep data pointer or copy

# Data Redistribution



Example of data redistribution from a producer task with 6 processes decomposed row-wise to a consumer task with 4 processes decomposed column-wise. The problem is that neither the producer nor the consumer task knows anything about the other's decomposition.

# Synthetic Benchmarks



Total # MDI	# Pro-	# Con-	Total	Total #	Total Data	
# MPI	ducer	sumer	# Grid	Parti-	Data	
Procs.	Procs.	Procs.	Points	cles	Size	
					(GiB)	
4	3	1	3.0e6	3.0e6	0.06	
16	12	4	1.2e7	1.2e7	0.22	
64	48	16	4.8e7	4.8e7	0.99	
256	192	64	1.9e8	1.9e8	3.54	
1024	768	256	7.7e8	7.7e8	14.34	
4096	3072	1024	3.0e9	3.0e9	55.88	
16384	12288	4096	1.2e10	1.2e10	223.51	

Number of processes and data sizes for synthetic benchmark experiments

Different experiment scenarios

### Synthetic Benchmarks: In Situ vs. Storage



Weak Scaling LowFive File vs Memory Mode

Time to write/read grid and particles between I producer task and I consumer task, comparing LowFive file and memory modes, in a weak scaling regime.

# Synthetic Benchmarks: Overhead of Using LowFive vs. Pure HDF5 for File I/O



Weak Scaling LowFive File Mode vs. HDF5

**Number Processes** 

Time to write/read grid and particles, comparing LowFive file mode with pure HDF5 file, in a weak scaling regime.

# Synthetic Benchmarks: Overhead of Using LowFive vs. Pure MPI for Message Passing



Number Processes

Time to write/read grid and particles comparing LowFive memory mode, with pure MPI communication, in a weak scaling regime.

### Synthetic Benchmarks: IOX Data Size

- 10<sup>7</sup> regularly structured grid points + 10<sup>7</sup> particles per producer process
- I90 MiB of data per producer process
- 0.55 GiB of data per consumer process (3:1 producer:consumer procs)
- Total data size at the largest scale tested is 0.55 TiB.



Time to write/read large size grid and particles, comparing LowFive memory mode, DataSpaces, and pure MPI, in a weak scaling regime

# Science Workflow: Cosmology



Data Size	LowFive	LowFive	HDF5 Write	HDF5 Read	Plotfiles	LowFive	LowFive
	Write Time	Read Time	Time	Time	Write Time	vs HDF5	vs Plotfiles
$256^{3}$	2.87	0.106	5.46	0.37	4.42	1.9	1.54
$512^{3}$	2.00	0.287	104.20	0.69	18.10	52.01	9.03
$1024^{3}$	2.87	0.628	920.44	3.02	35.00	320.00	12.17
$2048^{3}$	7.69	3.205	х	x	154.52	x	20.09

Time to write/read data between Nyx and Reeber using LowFive memory mode, HDF5 files, and AMReX plotfiles demonstrates that LowFive in situ data transport is 20X faster at scale than the best I/O solution (AMReX plotfile format). 15

# Recap

#### LowFive

- In situ data transport layer for workflows
- HDF5 data model
- Built as an HDF5 VOL plugin
- Allows bypassing storage and sending data over MPI
- Redistributes data between producer and consumer tasks
- Standalone software library that workflow systems can use

# Next Steps

- Finish implementing missing functions in our metadata
- Continue to test on applications and their software stacks
- Producer consumer synchronization and flow control
- Integrate in workflow systems driving further development
  - Henson can use LowFive (Nyx + Reeber use case)
  - We are also developing a new workflow system---Wilkins---on top of Henson and LowFive



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