JLESC Project: Toward taming large and complex data flows in data-centric supercomputing

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November 29, 2016





Project's status

Toward taming large and complex data flows in data-centric supercomputing

- Summary
 - Collaboration between Argonne National Lab. and Inria Bordeaux
 - Started in the beginning of 2015
 - Understand, characterize, and transform application data flows over a range of architectures
- Visits and meetings
 - E. Jeannot visited ANL on March 2015 and June 2016
 - I stayed 10 days at ANL on March 2015
- Impact and publications
 - I moved from Inria to ANL in February 2016 for a postdoc position
 - Our first results have been published and presented at the COM-HPC Workshop few weeks ago (SC'16)

Topology and Affinity Aware Hierarchical and Distributed Load-balancing in Charm++

- ▶ JLPC Inria-PPL project completed two years ago
- Recent publication at the COM-HPC Workshop (SC'16) thanks to the JLESC for Blue Waters allocations

Topology-Aware Data Aggregation for Intensive I/O on Large-Scale Supercomputers

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Data Movement at Scale

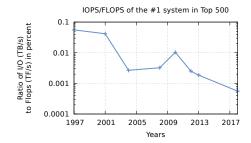
Context

- Computational science simulation such as climate, heart and brain modelling or cosmology have large I/O needs
 - Typically around 10% to 20% of the wall time is spent in I/O

Table: Example of I/O from large simulations

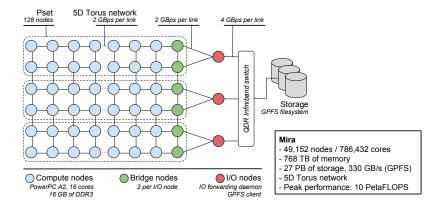
Scientific domain	Simulation	Data size	
Cosmology	Q Continuum	2 PB / simulation	
High-Energy Physics	Higgs Boson	10 PB / year	
Climate / Weather	Hurricane	240 TB / simulation	

 Increasing disparity between computing power and I/O performance in the largest supercomputers



Complex Architectures

- ► Complex network topologies: multidimensional tori, dragonfly, ...
- ► Partitioning of the architecture to reduce I/O interference
 - IBM BG/Q with I/O nodes (Figure), Cray with LNET nodes
- ► New tiers of storage/memory for data staging
 - MCDRAM in KNL, NVRAM, Burst buffer nodes

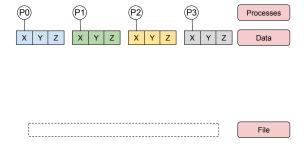


Two-phase I/O

- ► Available in MPI I/O implementations such as ROMIO
- Improves I/O performance by writing larger data chunks
- Selects a subset of processes to aggregate data before writing it to the storage system

Limitations:

- Poor for small messages (from experiments)
- Inefficient aggregator placement policy
- Fails to take advantage of data model, data layout and memory hierarchy

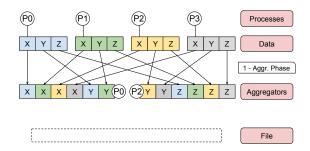


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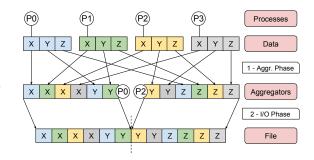


Figure: Two-phase I/O mechanism

Outline

- Context
- 2 Approach
- 3 Evaluation
- 4 Conclusion and Perspectives

Approach

Improved aggregator placement while taking into account:

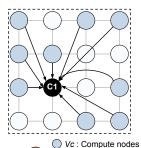
- ► The topology of the architecture
- ► The data access pattern

Efficient implementation of the two-phase I/O scheme

- Captures the data model and the data layout to optimize the I/O scheduling
- ▶ Pipelining of aggregation phase and I/O phase to optimize data movement
- ▶ Leverage one-sided communication
- Uses non-blocking operation to reduce synchronization

- ω(u, v): Amount of data exchanged between nodes u and v
- ightharpoonup d(u, v): Number of hops from nodes u to v
- I: The interconnect latency
- ▶ $B_{i\to j}$: The bandwidth from node i to node j.

$$\mathbf{C_2} = I \times d(A, IO) + \frac{\omega(A, IO)}{|V_C| \times B_{A \to IO}}$$



IO : I/O node

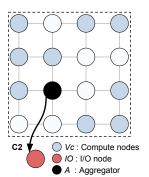
A : Aggregator

Objective function:

- ▶ TopoAware(A) = $min(C_1 + C_2)$
- ▶ Computed by each process independently in O(n), $n = |V_C|$

- $\omega(u, v)$: Amount of data exchanged between nodes u and v
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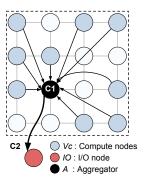


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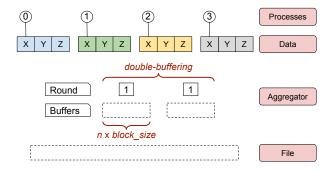
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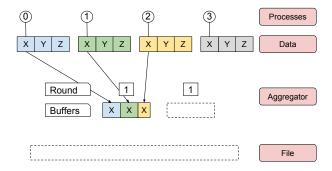
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- Initialization: allocate buffers, create MPI windows, compute tuples {round, aggregator, buffer} for each process P
 - Let's say P1 is the aggregator
- ▶ P0, P1 and P2 put data in buffer 1 (round 1) of P1. P3 waits (fence)
- ▶ P1 writes buffer 1 in file and aggregates data from all the ranks in buffer 2
- ▶ 2nd round. P1 writes buffer 2 and aggregates data from P1, P2 and P3
- and so on...
- ▶ Limitations: MPI_Comm_split, one aggr./node at most

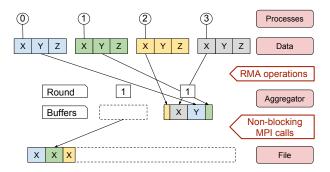


Conclusion

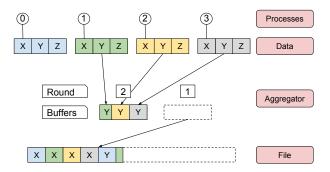
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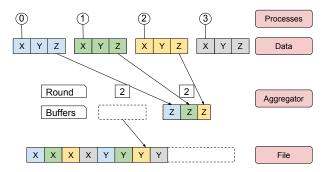
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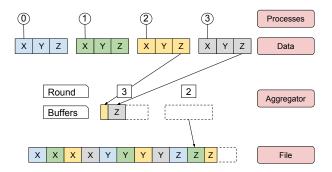
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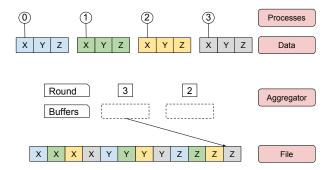
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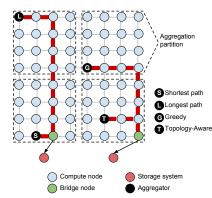
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- ▶ Evaluation on Mira (BG/Q), 512 nodes, 16 ranks/node
- ► Each rank sends a data buffer chosen randomly between 0 and 2 MB
- ► Writes to /dev/null of the I/O node (aggregation and I/O phases only)
- ▶ Aggregation settings: 16 aggregators, 16 MB buffer size
- Four tested strategies
 - Shortest path: smallest distance to the I/O node
 - Longest path: longest distance to the I/O node
 - **Greedy**: lowest rank in partition (similar to the default MPICH strategy)
 - Topology-aware



Micro-benchmark - Placement strategies

- ► Evaluation on Mira (BG/Q), 512 nodes, 16 ranks/node
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Table: Impact of aggregators placement strategy

Strategy	I/O Bandwidth (MBps)	Aggr. Time/round (ms)
Greedy	1927.45	421.33
Longest path	2202.91	370.40
Shortest path	2484.39	327.08
Topology-Aware	2638.40	310.46

▶ I/O bandwidth increased by 37% in comparison to the Greedy strategy and 6% over the Shortest Path approach

HACC-IO

- ▶ I/O part of a large-scale cosmological application simulating the mass evolution of the universe with particle-mesh techniques
- ► Each process manage particles defined by 9 variables (38 bytes)
 - XX, YY, ZZ, VX, VY, VZ, phi, pid and mask
- ▶ One file per *Pset* (128 nodes) vs. one single shared file
- ▶ Aggregation settings: 16 aggregators per *Pset*, 16 MB buffer size (MPICH)
- Average and standard deviation on 10 runs

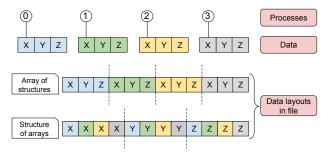
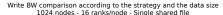
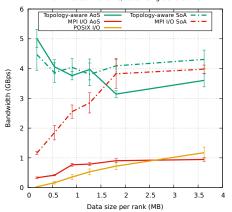


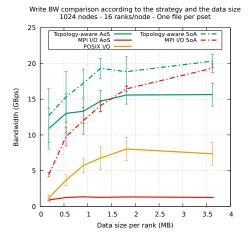
Figure: Data layouts in HACC-IO





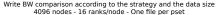
- ▶ Peak is estimated to 22.4 GBps (theoretical: 28.8 GBps)
- ▶ Our approach achieves higher performance than the default strategies
 - 5K particles (190 KB) and AoS data layout: 15× faster than MPI I/O

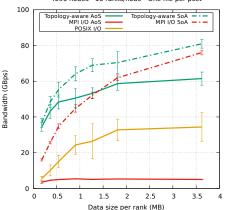
HACC-IO - 1024 nodes - 16K ranks - Sub-filing (One file per *Pset*)



- ▶ Sub-filing is an efficient approach for improved I/O performance
- Our topology-aware strategy achieves 90% of the peak I/O bandwidth (22.4 GBps)
 - Significant improvement particularly for small messages

Approach





- ▶ Peak is estimated to 89.6 GBps (theoretical: 115.2 GBps)
- ▶ 90% of the peak I/O bandwidth achieved as on 1024 nodes
- ▶ Improved I/O performance for both AoS and SoA layouts and significant improvement on smaller messages for the SoA case (up to 43%)

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Conclusion and Perspectives

Conclusion

- Optimized two-phase I/O library incorporating
 - Topology-aware aggregator placement
 - Optimized data movement and buffering (double-buffering, one-sided communication, block size awareness)
- Very good performance at scale, outperforming standard approaches
- \blacktriangleright On the I/O part of a cosmological application, up to $12\times$ improvement on 65K ranks
- ► Architecture characteristics are critical for performance at scale

Next steps

- ► Take the routing policy into account
- ► Incorporate additional data models and layouts (2D, 3D-arrays)
- ▶ Hierarchical approach to tackle different tiers of storage

Conclusion

Acknowledgments

- ► Argonne Leadership Computing Facility at Argonne National Laboratory
- ▶ DOE Office of Science, ASCR
- NCSA-Inria-ANL-BSC-JSC-Riken Joint-Laboratory on Extreme Scale Computing
- European Union Seventh Framework Program

Conclusion

Thank you for your attention! ftessier@anl.gov

Micro-benchmark - #Aggr and buffer size

- ► Evaluation on Mira (BG/Q), 1024 nodes, 16 ranks/node
- ► Each rank writes 1 MB
- Write to /dev/null of the I/O node (performance of just aggregation and I/O phases)

Table: I/O Bandwidth (in MBps) achieved on a simple benchmark with a topology-aware aggregator placement while varying the number of aggregators and the buffer size.

#Aggr/Pset	Buffer size			
#Aggi/Fset	8 MB	16 MB	32 MB	
8	7652.49	8848.28	9050.71	
16	7318.15	8774.58	9331.84	
32	6329.95	7797.12	8134.41	