Gluon: A Communication-Optimizing Substrate for Distributed Heterogeneous Graph Analytics

PLDI 2018

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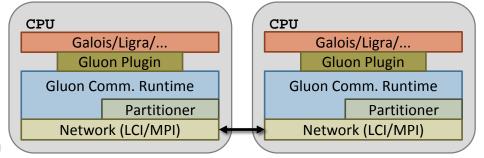


What is Gluon?

- Substrate: enable single address space applications to run on distributed,
 heterogeneous clusters
- Provides:
 - Partitioner
 - High-level synchronization API
 - Communication-optimizing runtime

How to use Gluon?

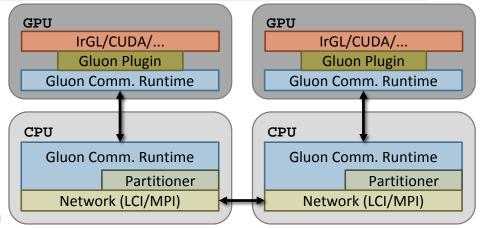
- Programmers:
 - Write shared-memory applications
 - Interface with Gluon using API
- Gluon transparently handles:
 - Graph partitioning
 - Communication and synchronization



Galois [SoSP'13] Ligra [PPoPP'13] IrGL [OOPSLA'16] LCI [IPDPS'18]

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Outline

- Gluon Synchronization Approach
- Optimizing Communication
 - Exploiting Structural Invariants of Partitions
 - Exploiting Temporal Invariance of Partitions
- Experimental Results

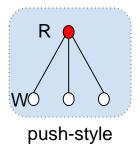
Gluon Synchronization Approach

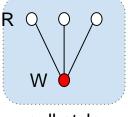
Vertex Programming Model

- Every node has a label
 - e.g., distance in single source shortest path (SSSP)
- Apply an operator on an active node in the graph
 - e.g., relaxation operator in SSSP

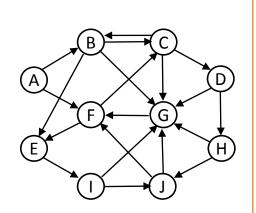


- Push-style: reads its label and writes to neighbors' labels
- Pull-style: reads neighbors' labels and writes to its label
- Applications: breadth first search, connected component, pagerank, single source shortest path, betweenness centrality, k-core, etc.





pull-style

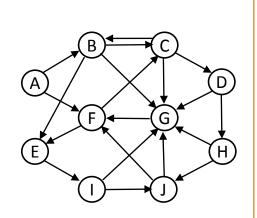


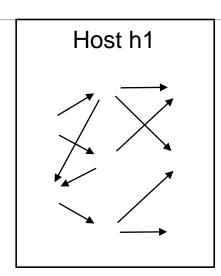
Host h1

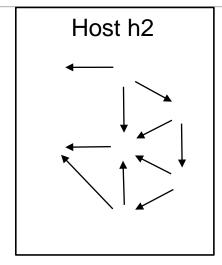
Host h2

Original graph

Partitions of the graph



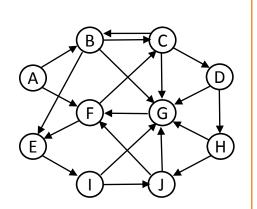


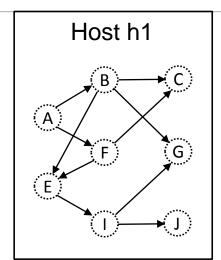


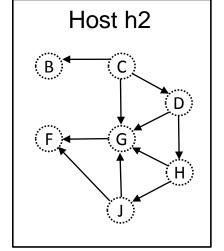
 Each edge is assigned to a unique host

Original graph

Partitions of the graph



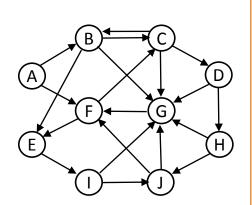




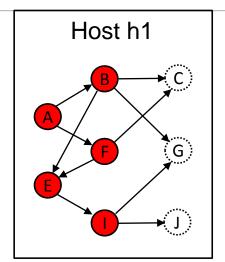
- Each edge is assigned to a unique host
- All edges
 connect proxy
 nodes on the
 same host

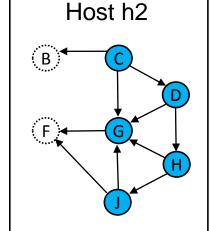
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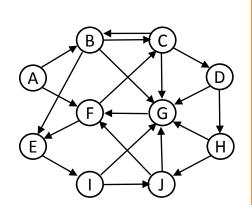


: Master proxy

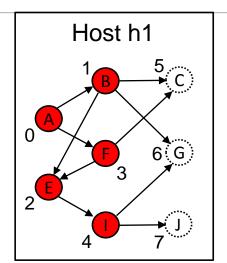
: Mirror proxy

Partitions of the graph

- Each edge is assigned to a unique host
- All edges
 connect proxy
 nodes on the
 same host
- A node can have multiple proxies: one is master proxy; rest are mirror proxies



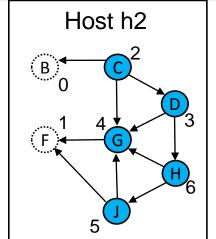
Original graph



A-J: Global IDs

0-7: Local IDs

Partitions of the graph



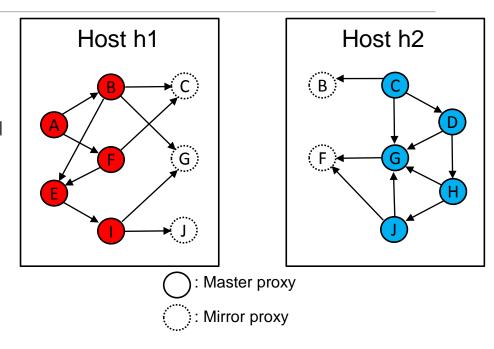
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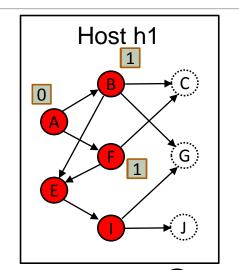
How to synchronize the proxies?

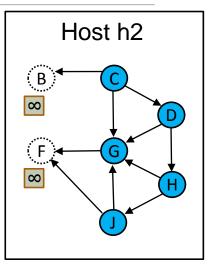
- Distributed Shared Memory (DSM) protocols
 - Proxies act like cached copies
 - Difficult to scale out to distributed and heterogeneous clusters



How does Gluon synchronize the proxies?

- Exploit domain knowledge
 - Cached copies can be stale as long as they are eventually synchronized





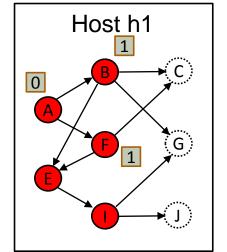
: Master proxy

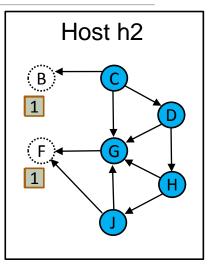
: Mirror proxy

: distance (label) from source A

How does Gluon synchronize the proxies?

- Exploit domain knowledge
 - Cached copies can be stale as long as they are eventually synchronized
- Use all-reduce:
 - Reduce from mirror proxies to master proxy
 - Broadcast from master proxy to mirror proxies



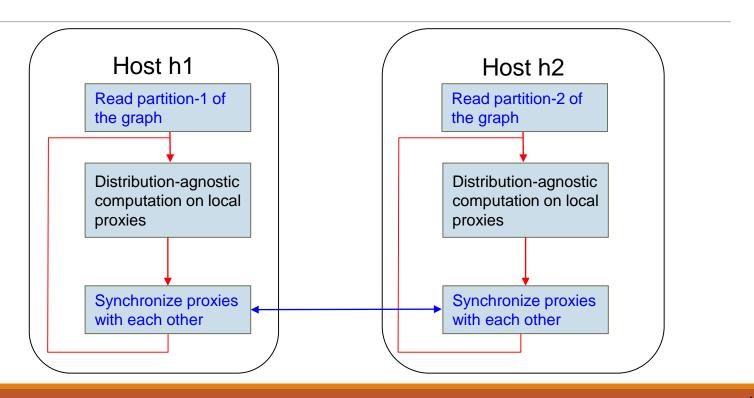


): Master proxy

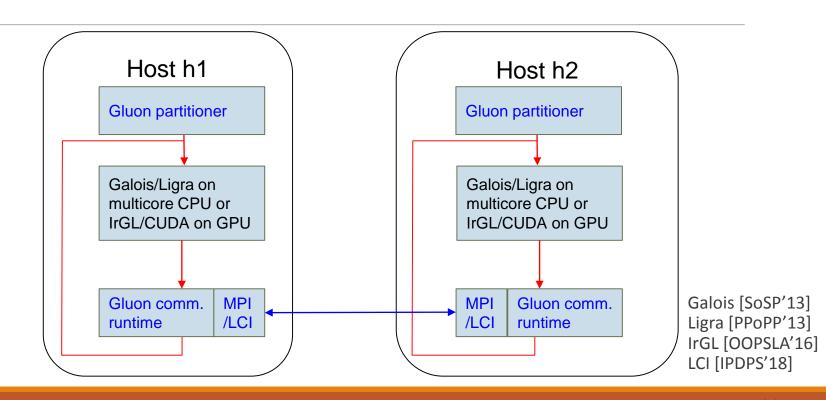
: Mirror proxy

: distance (label) from source A

When to synchronize proxies?



Gluon Distributed Execution Model



Gluon Synchronization API

- Application-specific:
 - What: Field to synchronize
 - O When: Point of synchronization
 - O How: Reduction operator to use
- Platform-specific:
 - Access functions for fields (specific to data layout)

Exploiting Structural Invariants to Optimize Communication

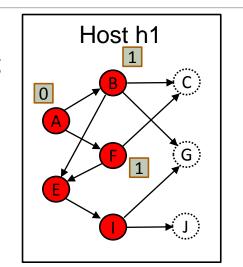
Structural invariants in the partitioning

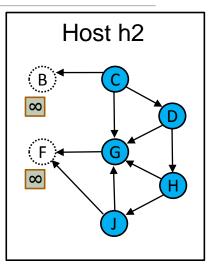
Structural invariants in this partitioning:

 Mirror proxies do not have outgoing edges

As a consequence, for sssp:

- Mirror proxies do not read their distance label
- Broadcast from master proxy to mirror proxies is not required





: Master proxy

: Mirror proxy

: distance (label) from source A

Partitioning: strategies, constraints, invariants

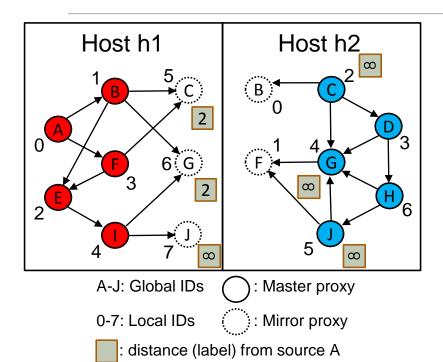
Algorithm invariant in SSSP:

R ● → W

Strategy	Constraints and Invariants	SSSP: Invariants	SSSP: Sync
Outgoing Edge-Cut (OEC)	Mirrors: no outgoing edges	Mirrors: label not read	Reduce
Incoming Edge-Cut (IEC)	Mirrors: no incoming edges	Mirrors: label not written	Broadcast
Cartesian Vertex-Cut (CVC)	Mirrors: either no outgoing edges or no incoming edges	Mirrors: either label not read or label not written	Reduce-partial & Broadcast-partial
Unconstrained Vertex-Cut (UVC)	None	None	Reduce & Broadcast

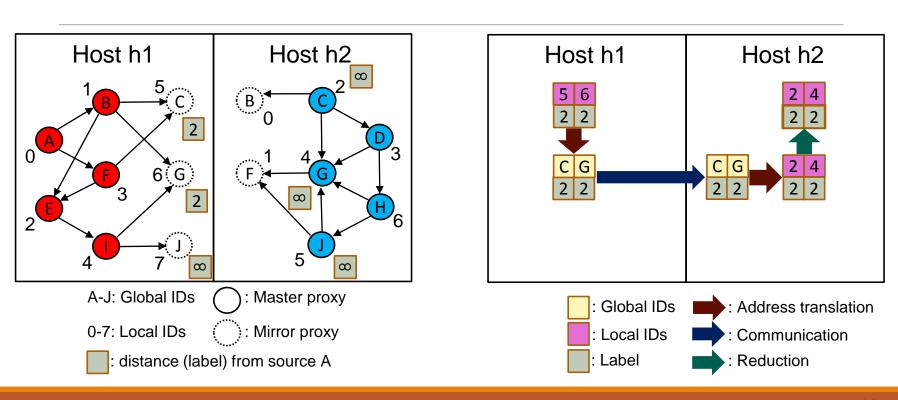
Exploiting Temporal Invariance to Optimize Communication

Bulk-communication

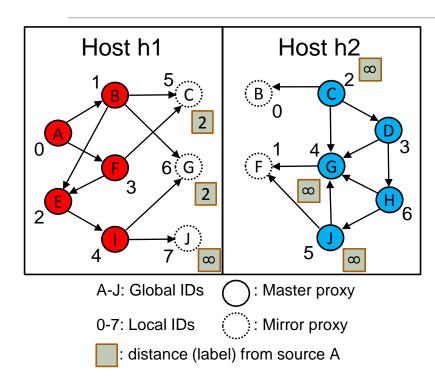


- Proxies of millions of nodes need to be synchronized in a round
 - Not every node is updated in every round
- Address spaces (local-IDs) of different hosts are different
- Existing systems: use address translation and communicate global-IDs along with updated values

Bulk-communication in existing systems



Bulk-communication in Gluon



- Elides address translation during communication in each round
- Exploits temporal invariance in partitioning
 - Mirrors and masters are static
 - e.g., only labels of C, G, and J can be reduced from h1 to h2
- Memoize address translation after partitioning

Experimental Results

Experimental setup

Systems:

- D-Ligra (Gluon + Ligra)
- D-Galois (Gluon + Galois)
- D-IrGL (Gluon + IrGL)
- Gemini (state-of-the-art)

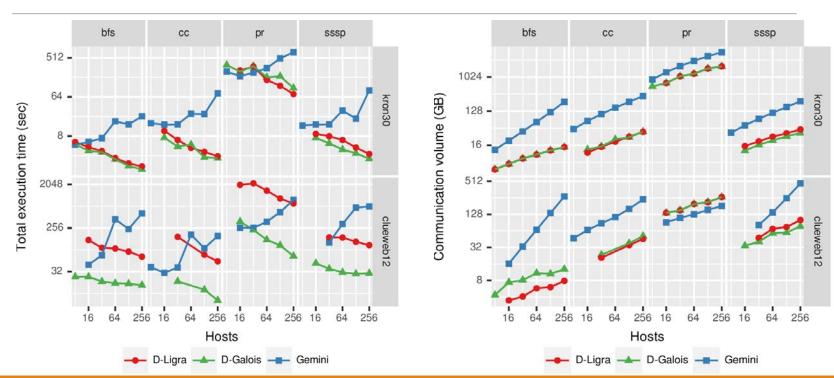
Inputs	rmat28	kron30	clueweb12	wdc12
V	268M	1073M	978M	3,563M
E	4B	11B	42B	129B
E / V	16	16	44	36
Size (CSR)	35GB	136GB	325GB	986GB

Benchmarks:

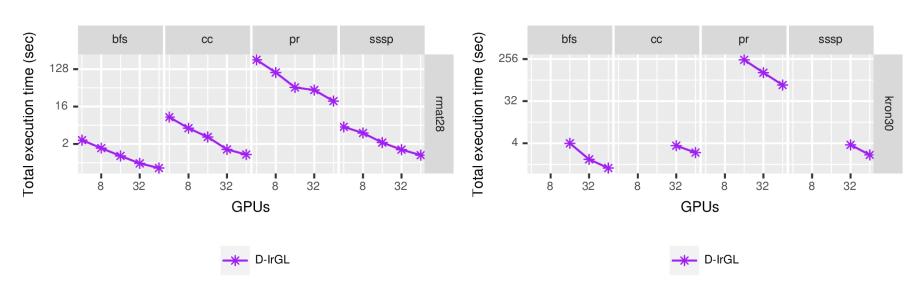
- Breadth first search (bfs)
- Connected components (cc)
- Pagerank (pr)
- Single source shortest path (sssp)

Clusters	Stampede (CPU)	Bridges (GPU)
Max. hosts	256	64
Machine	Intel Xeon Phi KNL	4 NVIDIA Tesla K80s
Each host	272 threads of KNL	1 Tesla K80
Memory	96GB DDR3	12GB DDR5

Strong scaling on Stampede (68 cores on each host)



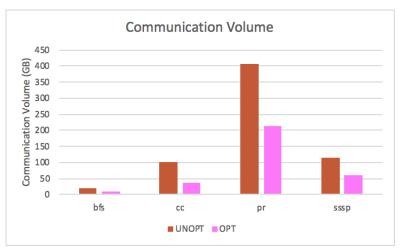
Strong scaling on Bridges (4 GPUs share a physical node)

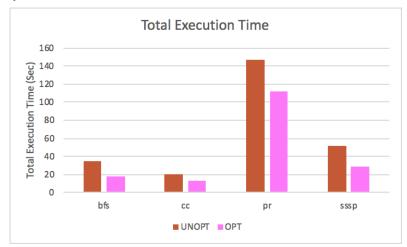


D-IrGL scales well

Impact of Gluon's communication optimizations

D-Galois on 128 hosts of Stampede: clueweb12 with CVC



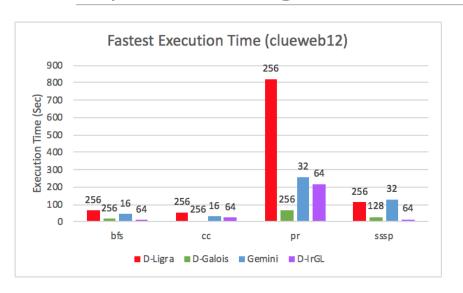


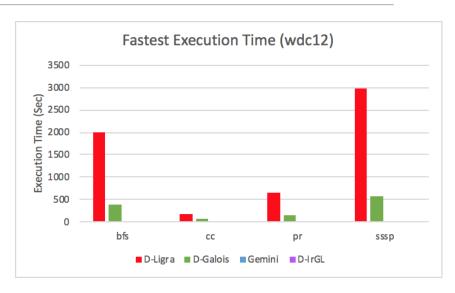
Improvement (geometric mean):

Communication volume: 2x

Execution time: **2.6x**

Fastest execution time (sec) of all systems using best-performing number of hosts/GPUs





D-Galois and D-IrGL are faster than Gemini by factors of 3.9x and 4.9x

Conclusions

- Novel approach to build distributed, heterogeneous graph analytics systems: scales out to 256 multicore-CPUs and 64 GPUs
- Novel communication optimizations: improve execution time by 2.6x
- [EuroPar'18] Abelian compiler: shared-memory Galois apps ---> distributed, heterogeneous (D-Galois + D-IrGL) apps
- Gluon, D-Galois, and D-IrGL: publicly available in Galois v4.0

http://iss.ices.utexas.edu/?p=projects/galois



Use Gluon to scale out your shared-memory graph analytical framework