EEE/ACM CCGRID 2023

The 23rd IEEE/ACM International Symposium on Cluster, Cloud and Internet Computing

Highly Scalable Large-Scale Asynchronous Graph Processing using Actors

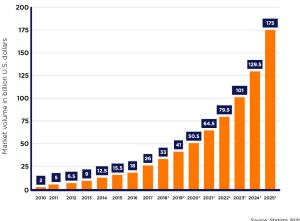
Youssef Elmougy*, Akihiro Hayashi, and Vivek Sarkar Georgia Institute of Technology, Atlanta GA USA

* Corresponding Author and Presenting Author

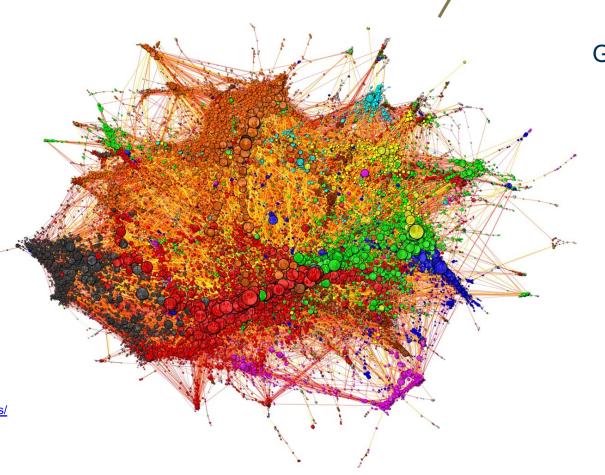


The Growth of Big Data

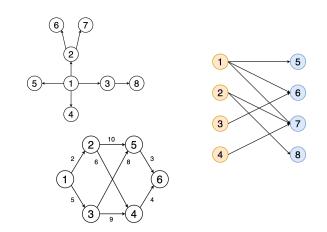
INFORMATION CREATED GLOBALLY 2010-2025

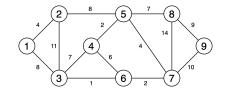


Picture borrowed from: https://www.iteratorshq.com/blog/big-data-business-impacts/



Graph algorithms have become increasingly important for solving problems in many computational domains





The scale of these graphs present difficulties to their processing and analysis!





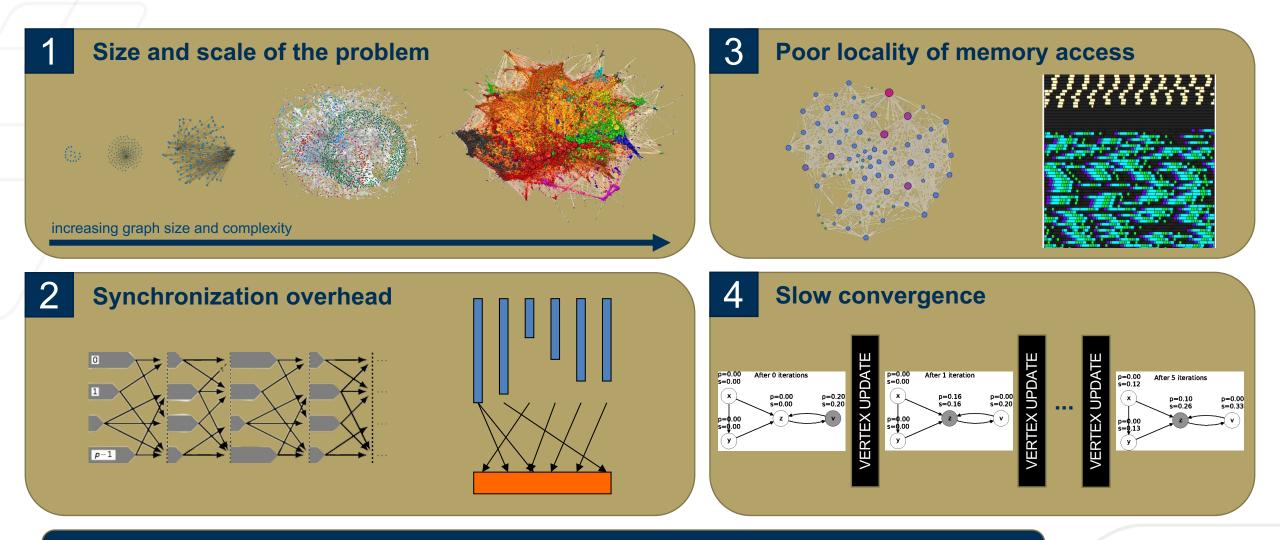
□ There is a need for parallel computing resources to meet the computational and memory requirements

Existing algorithms and software that perform well for mainstream parallel scientific applications are not necessarily efficient for large-scale graph applications

It is critical to develop lightweight and scalable systems to efficiently process large-scale graphs!



The 4 Overarching Challenges in Parallel Graph Processing



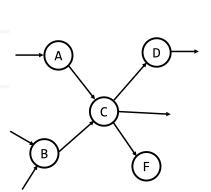
This paper studies the scalability of our novel actor-based programming system to overcome the inherent challenges of large-scale graph processing!

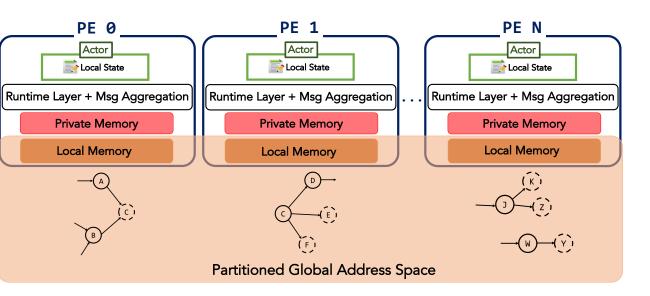


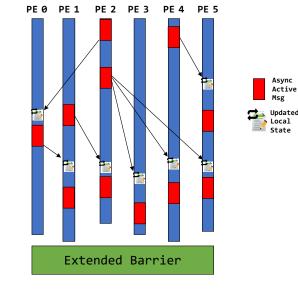
Sample Graph



Execution Model







- Presents a lightweight, asynchronous computation model
- Utilizes fine-grained asynchronous actor messages to express point-to-point remote operations
- Treats actors as primitives of computation, where actors are inherently isolated and share no mutable state
- Actors process messages sequentially within its mailbox, thereby avoiding data races and synchronization

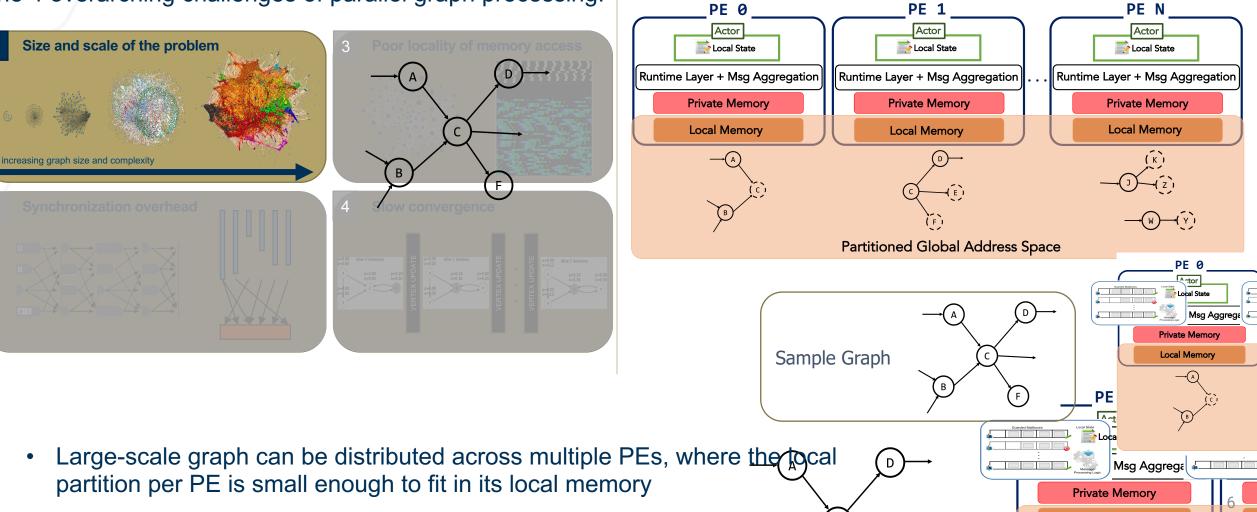
NOTE: "Actor" and "Selector" will be used interchangeably



SOLUTION

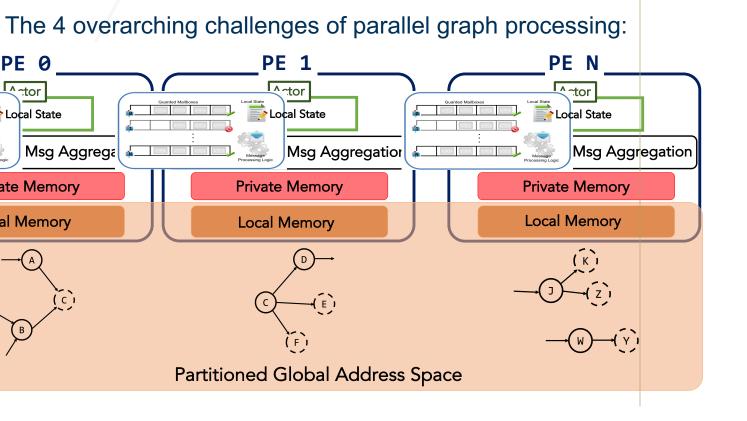
Distributed Actor Runtime

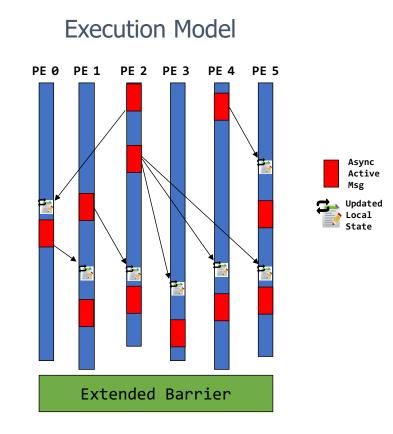
Local Memory



The 4 overarching challenges of parallel graph processing:

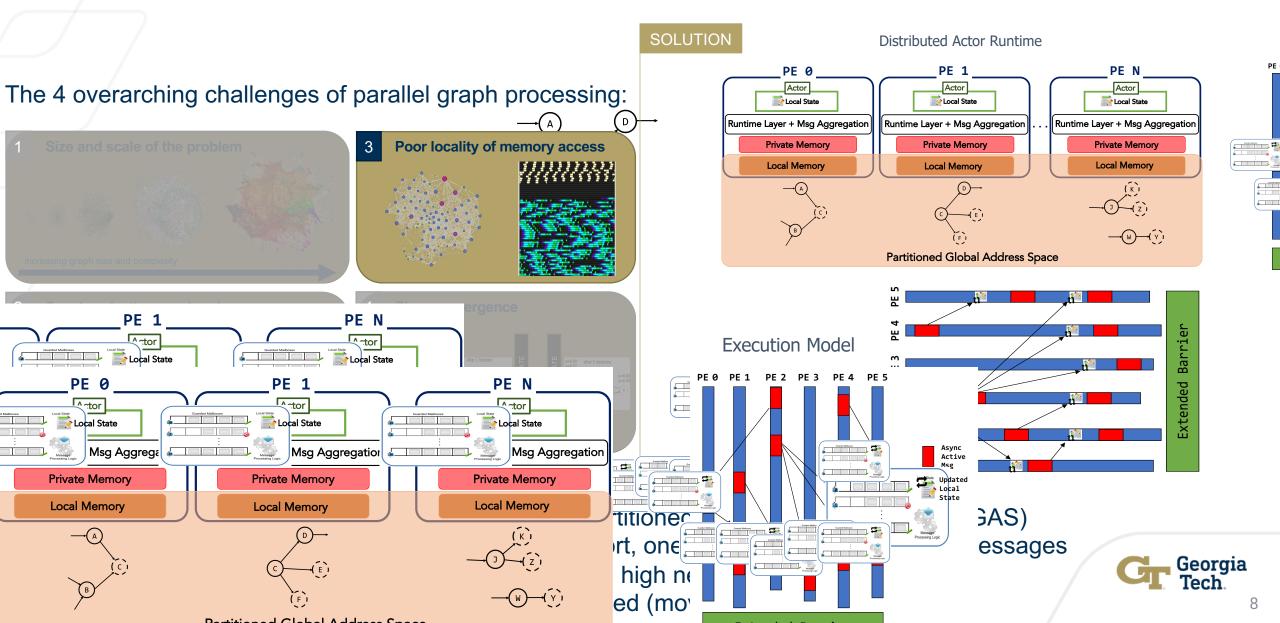
SOLUTION



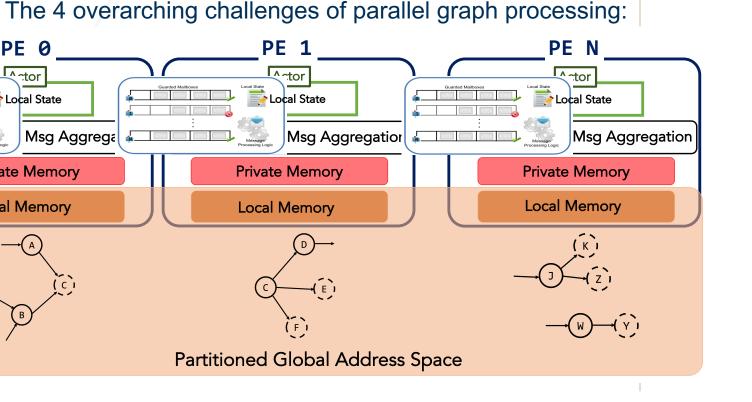


- Fine-grained Asynchronous Bulk-Synchronous (FABS) Parallelism model
- Reduces barriers and time spent idling at barriers, further reducing stall cycles

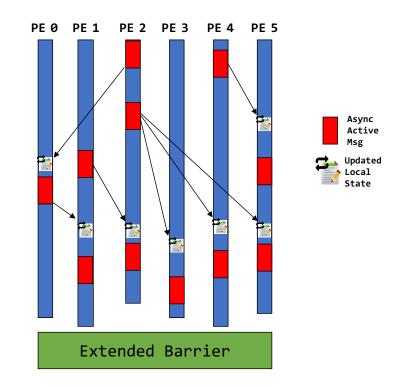




SOLUTION



Execution Model



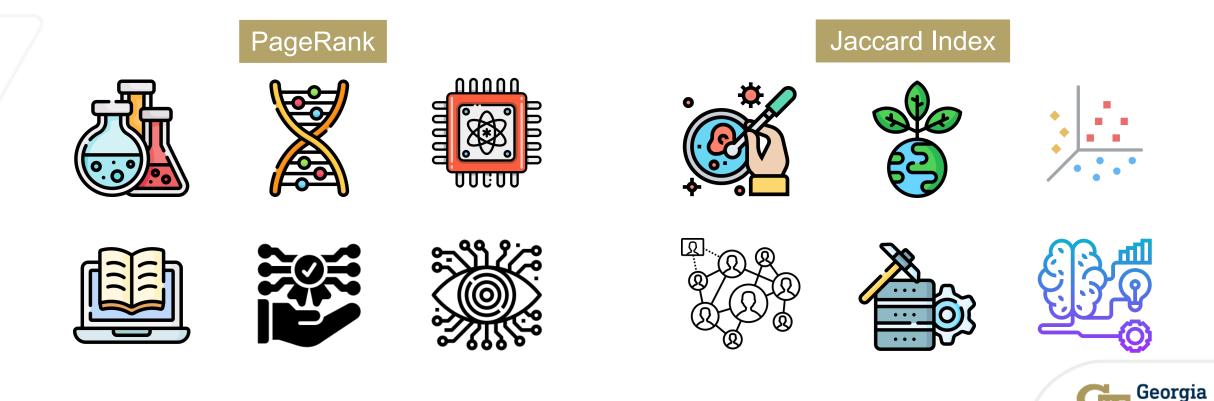
- As actor messages are executed, the local state of the actor is updated, allowing the next received actor message to utilize the updated state within the same super-step
- Updates the neighboring vertices with most recent values within the same iteration



Showing the System's Extreme Scalability with PageRank & Jaccard Index

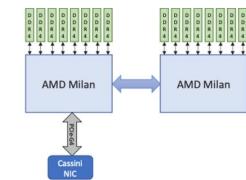
We focus on PageRank and Jaccard Index due to two reasons:

- 1. They show iterative vs non-iterative application types
- 2. They have been applied to many real-world problems with social impact



Experimental Setup and Architecture

- Metrics: execution time (in seconds) and traversed edges per second (TEPS)
- Experiments conducted on the CPU nodes of the **Perlmutter supercomputer** at the National Energy Research Scientific Computing Center (NERSC)
 - 2x AMD EPYC 7763 (Milan) CPUs
 - 64 physical cores per CPU
 - 512 GB memory
 - 1x HPE Cray Slingshot Interconnect
- Results for different dimensions of scalability are presented



Package 1#0			Package L#1			
Group0	Groupi		Group0		GroupD	
MIMANode L#0 P#0 (62(38)	MUMANode L#1 P#1 (63CB)		MUMANode L#4 P#4 (63G	3)	MUMANode L#5 P#5 (63GB)	
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Picture borrowed from: https://docs.nersc.gov/systems/perlmutter/architecture/



Dimensions of Scalability

Scaling performance is shown using three experiment types:

(1) **SCALE1**



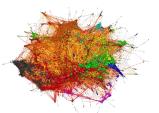
increasing graph size (WEAK SCALING, 5000 vertices per core)



(2) **SCALE2**

constant graph size (STRONG SCALING, 10.2M vertices & 696.4M edges)

(3) **SCALE3**



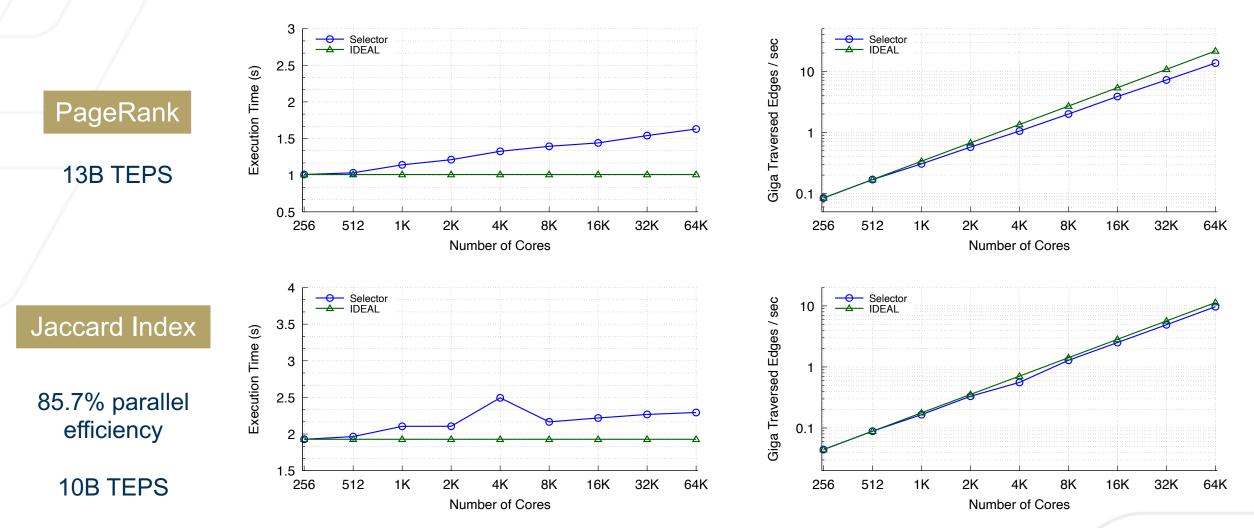
constant graph size (STRONG SCALING, 10.2M vertices & 696.4M edges)

	increasing #cores per node (<u>SCALE-UP</u> , from 1 core/node to
increasing #nodes (<u>SCALE-OUT</u> , from 256 to 65,53 from 1 to 1024	Gr Georgia Tech

• 11		

increasing #nodes (SCALE-OUT, from 256 to 65,536 cores from 1 to 1024 nodes)

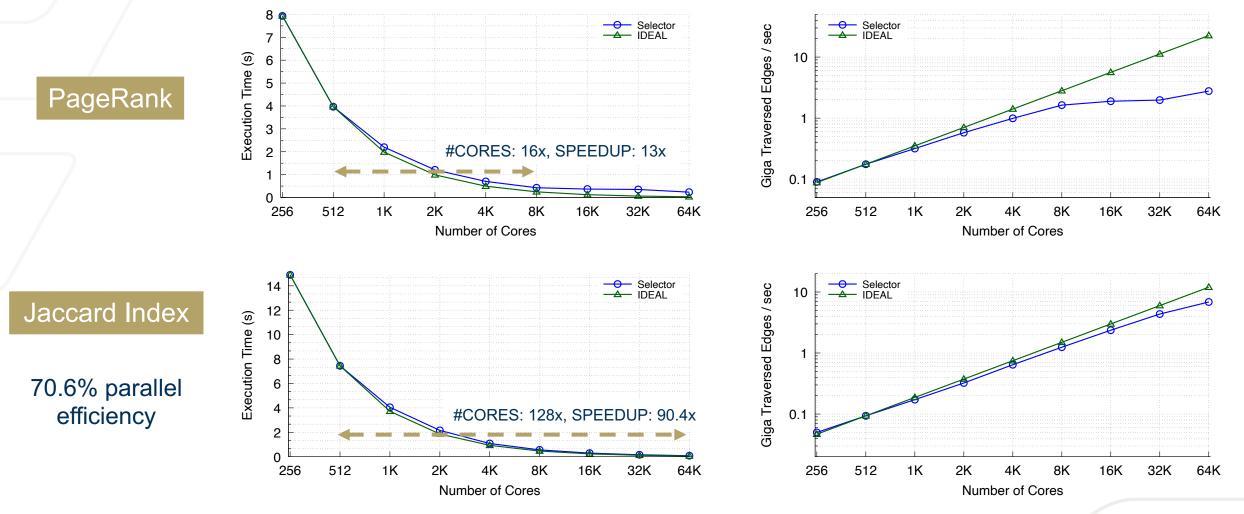
Performance Results: <u>SCALE1</u>



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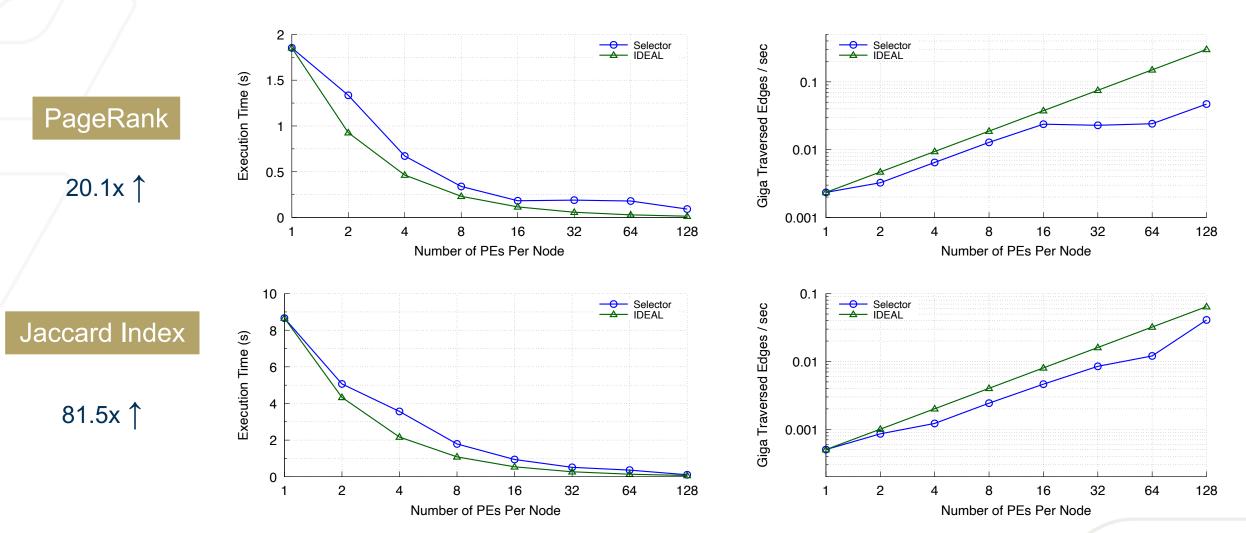
Performance Results: <u>SCALE2</u>







Performance Results: <u>SCALE3</u>



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Contrasting to Related Approaches

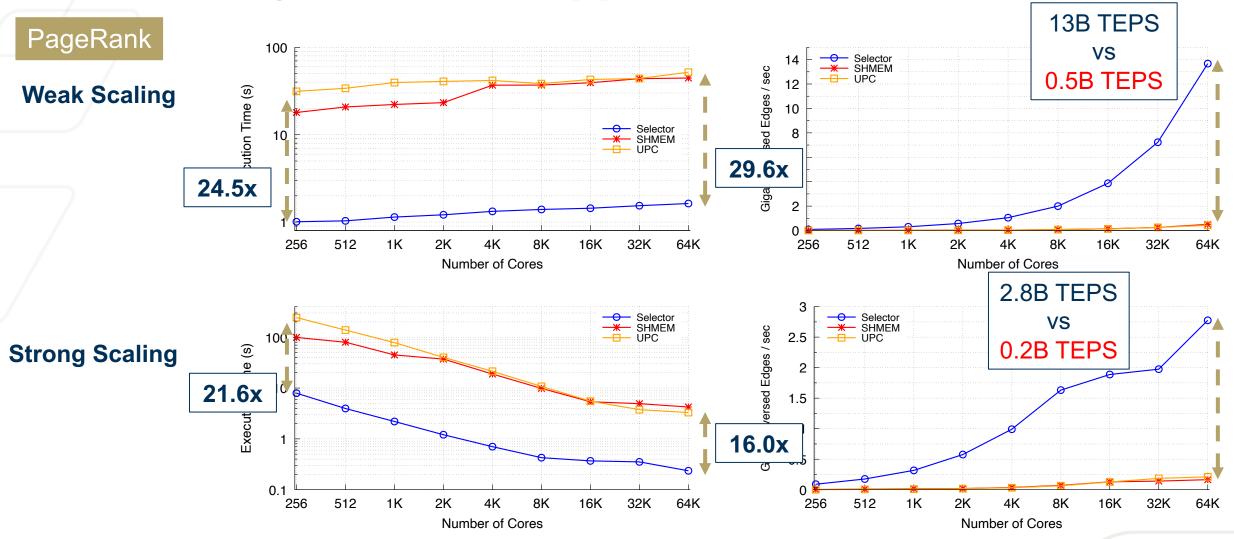
- We contrast with respect to remote atomics performance and graph application performance
- Related approaches: OpenSHMEM, UPC, MPI3-RMA, YGM

Communication System	Non-Blocking	Read / Get	Update / Set	
OpenSHMEM (cray-shmem 7.7.19)	N	40.06	N.A.	
OpenSHMEM NBI (cray-shmem 7.7.19)	Y	4 .79	4.99 🦷	
UPC (Berkley-UPC 2022.5.0)	N	30.37	30.03	
UPC NBI (Berkley-UPC 2022.5.0)	Y	20.58	N.A.	
MPI3-RMA (OpenMPI 4.1.2)	1.8x	25.44	142.04	7.6
MPI3-RMA (cray-mpich 7.7.19)	Y	9.67	59.47	
YGM	Y	N.A.	> 600	
Actors (cray-shmem 7.7.19)	Y	2.70	0.66	

Remote Atomics

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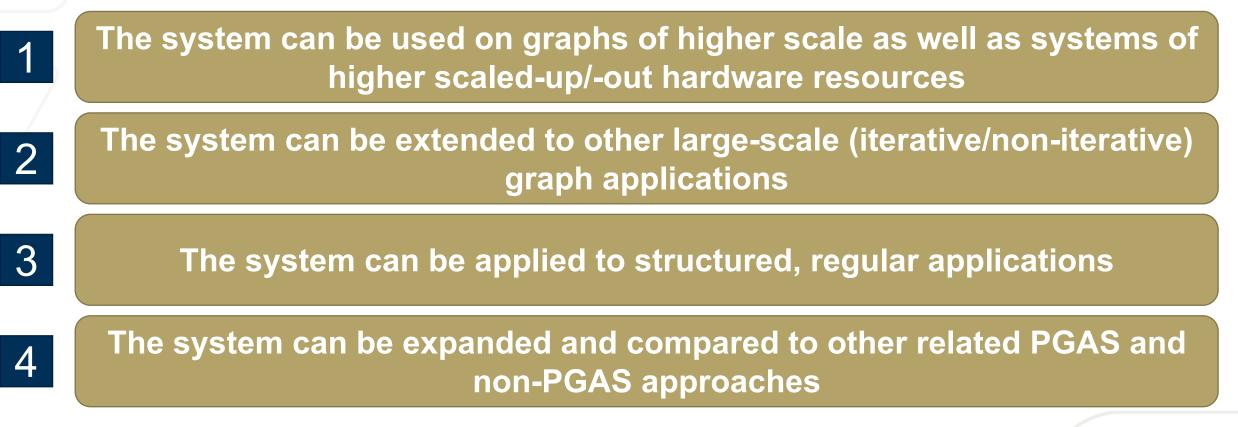
Contrasting to Related Approaches



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Impact of the Solution

- The actor-based system has shown **scalability**, **productivity**, and **performance**
- The extensibility of this system can front four impacts:





You can try this at home... Just visit hclib-actor.com !

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HClib-Actor Documentation Home	Bulk Synchronous Parallel	60	Table of contents What is the bulk synchronous parallel model?
Background Theory Bulk Synchronous Parallel Partitioned Global Addres:	 What is the bulk synchronous par The Bulk Synchronous Parallel (BSP) model is one 		Single Program Multiple Data (SPMD) Programming Further Readings
Space Actor Model Practice	models. The model consists of:		
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Thank you for your attention!

