## Ligra: A Lightweight Graph Processing Framework for Shared Memory

**Julian Shun and Guy Blelloch 2013** 

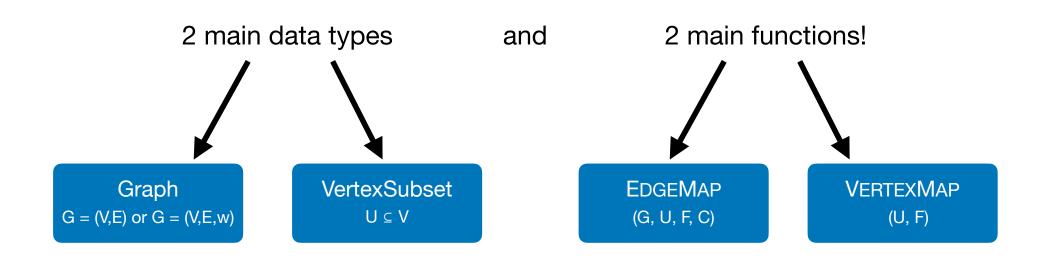
Presented by Sarah Zhao, R244 10.26.2022

#### **Motivation**

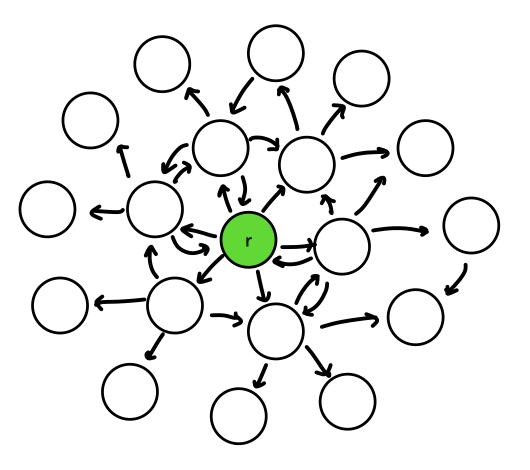
- 1. Shared Memory vs Distributed Memory
  - Many graph-processing frameworks designed for distributed memory systems (e.g. Pregel)
  - Advancements in technology -> enough storage in shared-memory machines (can handle graphs 100 billion edges in main memory)
  - Data locality, cheaper communication costs
- 2. Beamer et al, 2011 and 2012: hybrid approach to BFS exploiting variation in number of vertices and edges (i.e. frontier size) computed in each iteration of a parallel process
  - Can we generalize to other algorithms?

#### What is Ligra?

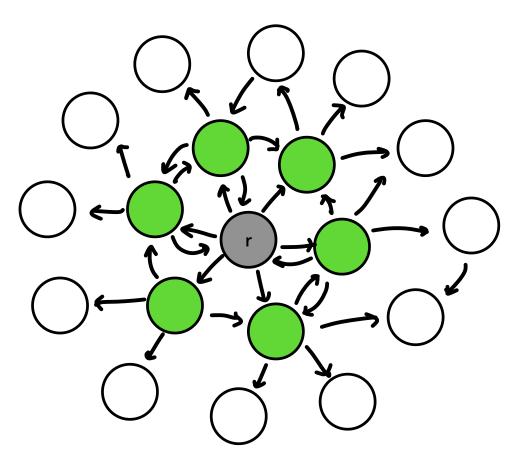
Simple shared-memory parallel graph-processing framework:



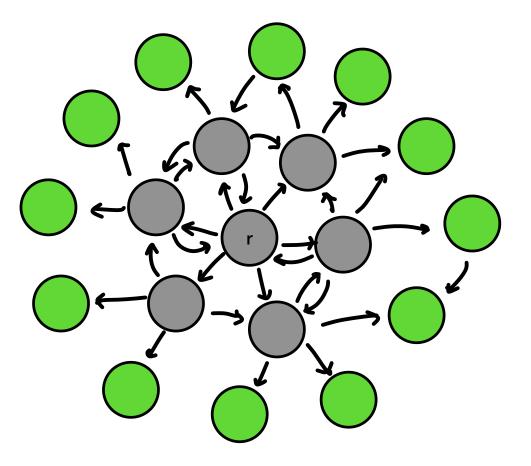
```
1: Parents = \{-1, \ldots, -1\}
                                                      \triangleright initialized to all -1's
2:
3: procedure UPDATE(s, d)
       return (CAS(&Parents[d], -1, s))
4:
5:
6: procedure COND(i)
       return (Parents[i] == -1)
7:
8:
9: procedure BFS(G, r)
                                                              \triangleright r is the root
        Parents[r] = r
10:
        Frontier = \{r\}
11:
                                \triangleright vertexSubset initialized to contain only r
        while (SIZE(Frontier) \neq 0) do
12:
           Frontier = EDGEMAP(G, Frontier, UPDATE, COND)
13:
```



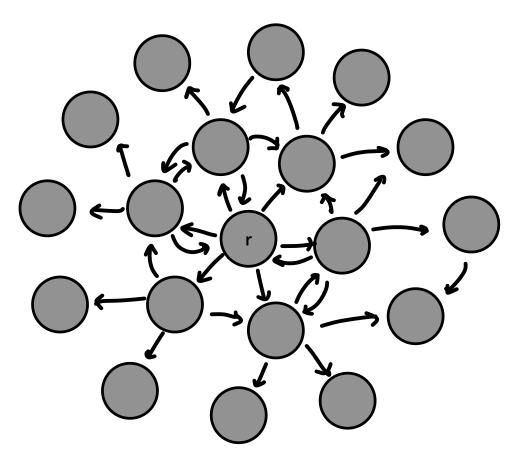
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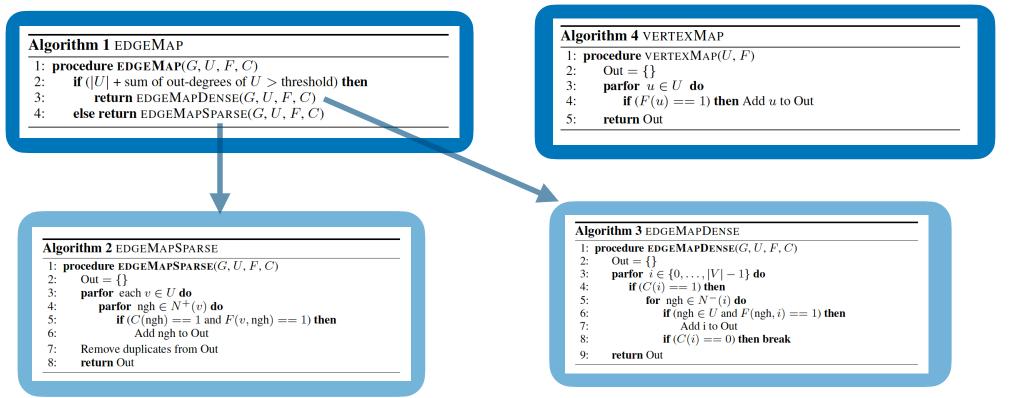
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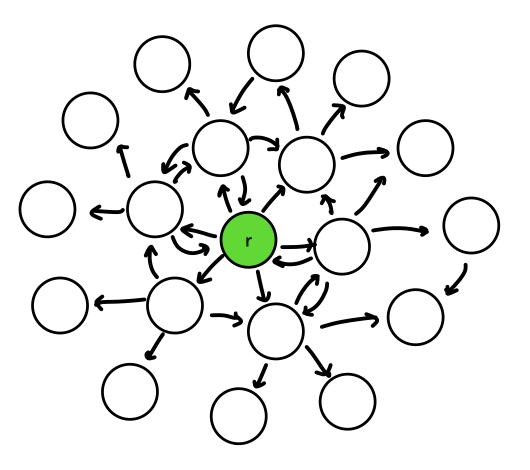
## **Hybrid Model for Varying Frontier-Size**



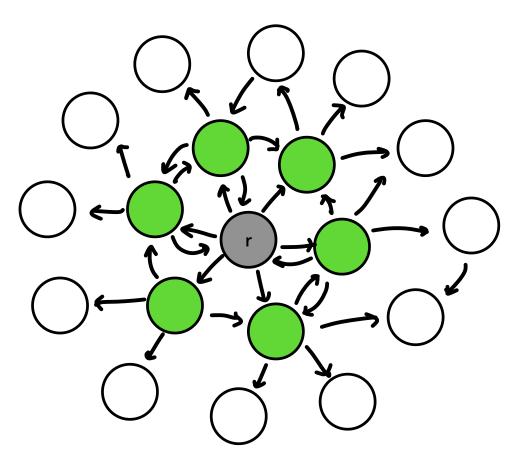
Check out-neighbors for each vertex in the frontier

Check in-neighbors for each target vertex

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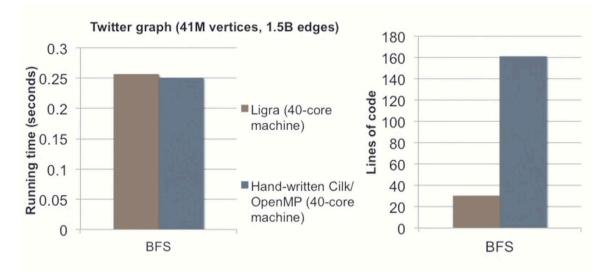


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#### **BFS Evaluation**

 Implemented on a 40-core Intel machine with 256GB of RAM (and with multithreading)



· Comparing against direction-optimizing code by Beamer et al.

https://www.youtube.com/watch?v=W5mDx\_G45RQ&ab\_channel=MMDSFoundation

#### **Other Applications and Results**

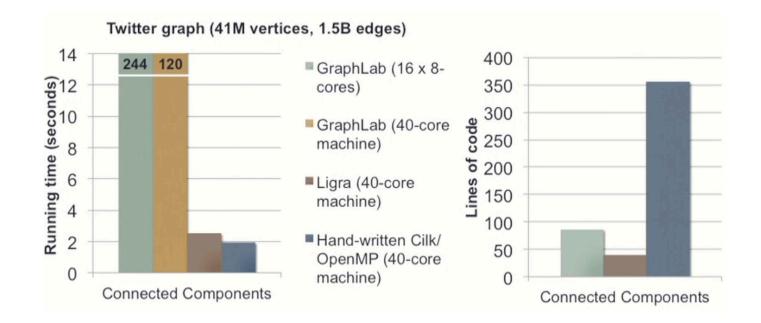
Input	Num. Vertices	Num. Directed Edges
3D-grid	107	$6 \times 10^7$
random-local	107	$9.8 \times 10^7$
rMat24	$1.68 \times 10^7$	$9.9  imes 10^7$
rMat27	$1.34 \times 10^{8}$	$2.12 \times 10^{9}$
Twitter	$4.17 \times 10^7$	$1.47 \times 10^{9}$
Yahoo*	$1.4 \times 10^{9}$	$12.9 \times 10^{9}$

**Table 1.** Graph inputs. \*The original asymmetric graph has  $6.6 \times 10^9$  edges.

Application		3D-grid	l	rai	ndom-lo	cal		rMat24			rMat27			Twitter			Yahoo	
	(1)	(40h)	(SU)	(1)	(40h)	(SU)	(1)	(40h)	(SU)	(1)	(40h)	(SU)	(1)	(40h)	(SU)	(1)	(40h)	(SU)
Breadth-First Search	2.9	0.28	10.4	2.11	0.073	28.9	2.83	0.104	27.2	11.8	0.423	27.9	6.92	0.321	21.6	173	8.58	20.2
Betweenness Centrality	9.15	0.765	12.0	8.53	0.265	32.2	11.3	0.37	30.5	113	4.07	27.8	47.8	2.64	18.1	634	23.1	27.4
Graph Radii	351	10.0	35.1	25.6	0.734	34.9	39.7	1.21	32.8	337	12.0	28.1	171	7.39	23.1	1280	39.6	32.3
Connected Components	51.5	1.71	30.1	14.8	0.399	37.1	14.1	0.527	26.8	204	10.2	20.0	78.7	3.86	20.4	609	29.7	20.5
PageRank (1 iteration)	4.29	0.145	29.6	6.55	0.224	29.2	8.93	0.25	35.7	243	6.13	39.6	72.9	2.91	25.1	465	15.2	30.6
Bellman-Ford	63.4	2.39	26.5	18.8	0.677	27.8	17.8	0.694	25.6	116	4.03	28.8	75.1	2.66	28.2	255	14.2	18.0

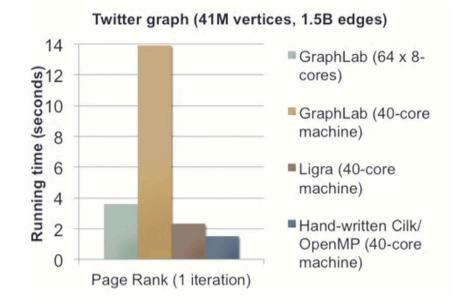
**Table 2.** Running times (in seconds) of algorithms over various inputs on a 40-core machine (with hyper-threading). (SU) indicates the speedup of the application (single-thread time divided by 40-core time).

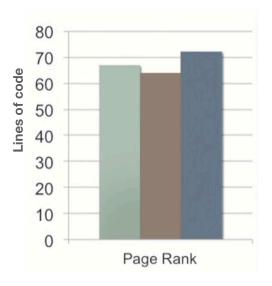
#### **Experimental Results** Connected Components



https://www.youtube.com/watch?v=W5mDx\_G45RQ&ab\_channel=MMDSFoundation

#### **Experimental Results** PageRank





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## Summary

- Lightweight parallel graph processing framework
- Dependence on shared-memory systems (no communication overhead as compared to distributed systems)
- Designed for frontier-based algorithms
- Comparable to or outperformed then graph-processing frameworks

#### Discussion

- Incomplete evaluation
- Performance to be hardware dependent
  - Worse performance with a different set-up (64-core AMD Opteron machine)
- Scalability and speedup limited by tech
  - Expansion to GPUs? [Shun et al. 2013]
- Exploring applications in other graph algorithms (e.g. max flow, biconnected components, belief propagation, Markov clustering) [Shun et al. 2013]
- Dynamic graph data, graph data stream processing?

#### **Extensions**

- Ligra+ (2015) compressed graphs -> require less memory
- Julienne (2017) bucketing-based algorithms (generalization of frontier-based algorithms)
- Hygra (2020) support for hyper graphs

Code at https://github.com/jshun/ligra

# Thank you!



#### References

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