### Clean Breadth First Search

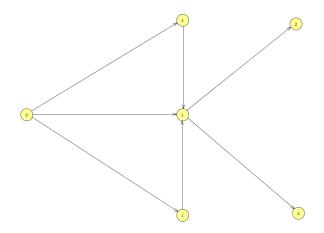
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## BFS Problem Description

- Given
  - Graph  $G = (\mathbb{V}, \mathbb{E})$
  - ▶ Distinguished source vertex  $s \in \mathbb{V}$
- Calculate
  - ▶  $Dist_{u \in \mathbb{V}}$  = length of the shortest path from s to u in G
  - ▶ \* $Parent_{u \in \mathbb{V}} = v \in \mathbb{V}$  s.t.  $(v, u) \in \mathbb{E}$ ,  $Dist_u = Dist_v + 1$

# BFS Example



#### Terms Used in this Presentation

- n = |V| the number of nodes in a graph
- $m = |\mathbb{E}|$  the number of edges in a graph
- ▶  $D = \mathcal{O}(n)$  the diameter of a graph
- ightharpoonup  $\Gamma_u$  the set of vertexes adjacent to u
- $ightharpoonup T_s$  the running time for a serial algorithm
- ► T<sub>P</sub> the running time for a parallel algorithm running on P cores
- ► T<sub>∞</sub> the running time for a parallel algorithm running on infinite cores
- ▶  $T_1 = \Omega(T_s)$  the running time for a parallel algorithm running on one core
- $W_P = \Omega(T_s)$  the total work done by a parallel algorithm running on P cores (excluding idle time)
  - ▶ Reducing  $W_P$  can reduce energy use [1]

### Serial-BFS

- 1. for each vertex  $u \in \mathbb{V}$
- 2.  $Dist_u \leftarrow \infty$
- 3.  $Dist_s \leftarrow 0$
- 4.  $Q \leftarrow \emptyset$
- 5. Enqueue(Q, s)
- 6. while  $Q \neq \emptyset$  do
- 7.  $u \leftarrow \text{Dequeue}(Q)$
- 8. for each vertex v in  $\Gamma(u)$  do
- 9. if  $Dist_v = \infty$  then
- 10.  $Dist_v \leftarrow Dist_u + 1$
- 11. ENQUEUE( Q, v )

## Model of Computation

- Large shared memory
- Consistent caches between cores
- ▶ Synchronizing x tasks takes  $T_{\infty} = \Theta(\log x)$  time
- Cilk has this model with randomized work stealing [3]

# Motivation for (P)BFS

#### BFS is used for

- ► Path Finding
  - Video Games
  - ▶ Google Maps
- Analyzing social networks
- Designing and analyzing VLSI
- Task scheduling
- As a primitive in other algorithms

## Existing Approaches for PBFS

- Assumes somewhat unrealistic PRAM model
- Specialized for specific hardware [7]
  - ▶ GPU
  - CRAY (hardware mutex every 64 bits, atomic add) [2]
- Uses atomic instructions[5]
- Specialized for sparse (dense) graphs only
- Specialized for bounded out-degree (Not scale-free) [4]
- $ightharpoonup T_1$  or  $T_p$  is not asymptotically optimal
- ▶ Room for energy efficiency improvements (non-optimal  $W_P$ )
- Offloads some work to scheduler
  - Work-stealing (randomized) gives at best high probability bounds [4]
- Non level-synchronous[6]

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- distinguished-bfs [6] contracts the graph to make it dense
- cray-bfs [2] uses fast hardware mutexes and atomic increments

## Level-Synchronous BFS

- All nodes at distance d from s are processed before any nodes at distance d' > d.
- $T_P = \Omega (n/p + m/p + D \log P)$ 
  - $T_p = \Omega (T_s/P) = \Omega (n/p + m/p)$
  - ▶ Let  $n_{\ell}$ ,  $m_{\ell}$  be the number of nodes and edges visited at level  $\ell$ .
  - ► Consider a graph where  $\forall_{0<\ell \leq D} n_{\ell} + m_{\ell} = \Theta(P)$
  - ▶ Every level has  $\Theta(P)$  work and uses  $P_{\ell} \leq P$  cores
  - ▶ For each level,  $T_p = \Omega(P/P_\ell + \log P_\ell) = \Omega(\log P)$  time

# Bottlenecks for Parallelizing BFS

- ► FIFO
- Dist array

# Clean BFS - Properties

$$T_s = \mathcal{O}(n+m)$$

$$ightharpoonup T_1 = \mathcal{O}(n+m)$$

$$V_P = \mathcal{O}(n+m)$$

$$T_P = \mathcal{O}\left(n/p + m/p + D\log P\right)$$

- Scale-free
- Deterministic worst case bounds

## Clean BFS - Approach

- Evenly split edges among cores using prefix sum and binary search
- Combine output queues using prefix sum

## CBFS - Prepare to Split Work

- ▶ One input queue  $Q_{in} \subseteq \mathbb{V}$ .
  - ► Each vertex in *Q<sub>in</sub>* is unique\*
  - $\quad \blacktriangleright \ \forall_{u \in Q_{in}} Dist_u = \ell$
- ▶ Generate  $OutDegrees[0 \le i < |Q_{in}|] = |\Gamma(Q_{in}[i])|$  in parallel
- Perform a parallel prefix sum\* on OutDegrees

## CBFS - Split Work and Process Edges

- ▶ Each core *i* processes  $m_{\ell}/P$  edges
  - searches *OutDegrees* for  $1 + \left| \frac{i \cdot m_{\ell}}{P} \right|$  to find starting edge
  - does  $\mathcal{O}\left(\log \frac{n_{\ell}}{P} + \log P\right)$  work\*
  - processes  $\left|\frac{m_{\ell}}{P}\right|$  consecutive edges
    - $ightharpoonup Q_i \leftarrow \emptyset$
    - for each edge (u, v)
    - if  $Dist_v = \infty$  then
    - $Dist_v \leftarrow Dist_u + 1$
    - $Dist_v \leftarrow Dist_u + \cdots$
    - Owner<sub>v</sub>  $\leftarrow$  i
    - Enqueue( $Q_i, v$ )
  - Benign race conditions

# CBFS - Dedup Vertexes and Combine Queues

- ►  $Size_{-1} = 0$
- ► Each core *i* uses *Owner* to ensure each vertex lives in at most one output queue
  - ▶  $Q_i \leftarrow \{u \in Q_i : Owner_u = i\}$
  - ▶  $Size_i \leftarrow |Q_i|$
- ▶ Perform a parallel prefix sum\* on Size
- ► Each core i copies its queue back into Q<sub>in</sub> at offset Size<sub>i-1</sub>

# CBFS - Reducing $W_P$ for binary searches

- $\triangleright$  N  $\leftarrow$  |OutDegrees|
- Each core i
  - ▶ FirstDegree  $\leftarrow$  OutDegrees  $\left[\left|\frac{iN}{P}\right| 1\right]$
  - $\qquad \qquad \textbf{\it FirstDegreeNext} \leftarrow \textit{\it OutDegrees}[\left| \frac{(i+1)N}{P} \right| 1]$
  - ► FirstCore  $\leftarrow \left\lceil \frac{P \cdot FirstDegree}{m_{\ell}} \right\rceil$ ► LastCore  $\leftarrow \left\lceil \frac{P \cdot FirstDegreeNext}{m_{\ell}} \right\rceil$

  - ▶ parallel for  $i \leftarrow FirstCore$  to LastCore
  - $SubList_i \leftarrow i$
- ▶ Using SubList<sub>i</sub>, core i can search only  $n_{\ell}/p$  indexes
- $\triangleright$   $W_P$  goes from  $\mathcal{O}(n+m+DP\log P)$  to  $\mathcal{O}(n+m+DP)$

## CBFS - Reducing $W_P$ for parallel prefix sums

- Immediately after calculating  $m_\ell$
- ▶  $P_{\ell} \leftarrow \min(m_{\ell}, P)$
- ▶ Use at most  $P_{\ell}$  cores until next time  $m_{\ell}$  is calculated.
- ▶ This ensures the  $\mathcal{O}(P_{\ell})$  work every level is  $\mathcal{O}(m_{\ell})$  and can be absorbed into the constant.
- ▶  $W_P$  goes from  $\mathcal{O}(n+m+DP)$  to  $\mathcal{O}(n+m+D) = \mathcal{O}(n+m)$

#### **Future Work**

- Optimize CBFS for the PRAM model
  - CBFS runs in same time for PRAM but is not asymptotically optimal
- Modify CBFS to remove false sharing

#### References I



Race to idle: New algorithms for speed scaling with a sleep state.

🔋 David A. Bader and Kamesh Madduri.

Designing multithreaded algorithms for breadth-first search and st-connectivity on the cray mta-2.

In *Proceedings of the 2006 International Conference on Parallel Processing*, ICPP '06, pages 523–530, Washington, DC, USA, 2006. IEEE Computer Society.

Matteo Frigo, Pablo Halpern, Charles E. Leiserson, and Stephen Lewin-Berlin.

Reducers and other cilk++ hyperobjects.

In Proceedings of the twenty-first annual symposium on Parallelism in algorithms and architectures, SPAA '09, pages 79–90, New York, NY, USA, 2009. ACM.

#### References II



Charles E. Leiserson and Tao B. Schardl.

A work-efficient parallel breadth-first search algorithm (or how to cope with the nondeterminism of reducers).

In Proceedings of the 22nd ACM symposium on Parallelism in algorithms and architectures, SPAA '10, pages 303–314, New York, NY, USA, 2010. ACM.



Erik Saule and Umit V. Catalyurek.

An early evaluation of the scalability of graph algorithms on the intel mic architecture.

In Proceedings of the 2012 IEEE 26th International Parallel and Distributed Processing Symposium Workshops & PhD Forum, IPDPSW '12, pages 1629–1639, Washington, DC, USA, 2012. IEEE Computer Society.

#### References III



J. Ullman and M. Yannakakis.

High-probability parallel transitive closure algorithms.

In *Proceedings of the second annual ACM symposium on Parallel algorithms and architectures*, SPAA '90, pages 200–209, New York, NY, USA, 1990. ACM.



Andy Yoo, Edmond Chow, Keith Henderson, William McLendon, Bruce Hendrickson, and Umit Catalyurek. A scalable distributed parallel breadth-first search algorithm on bluegene/I.

In Proceedings of the 2005 ACM/IEEE conference on Supercomputing, SC '05, pages 25–, Washington, DC, USA, 2005. IEEE Computer Society.