A Work-Efficient Parallel Breadth-First Search Algorithm (or How to Cope with the Nondeterminism of Reducer Hyperobjects)

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Breadth-first search (BFS)



Problem: Given an unweighted graph G = (V, E) and a designated starting vertex v_0 , find the shortest path distance from v_0 to all other $u \in V$.

- Guarantee that the vertices are visited in *breadth-first order*: For all distances *d*, all vertices that are *d* away from v_0 must be visited before any vertex of distance d + 1.
- We want a parallel algorithm to solve this problem.

Serial BFS

SERIAL-BFS($G = (V, E), v_0$) for each vertex $u \in V - \{v_0\}$ 1 2 u. dist = ∞ 3 v_0 . dist = 0 4 $Q = \{v_0\}$ 5 while $Q \neq \emptyset$ 6 $u = \mathsf{DEQUEUE}(Q)$ 7 for each $v \in V$ such that $(u, v) \in E$ 8 if v. dist == ∞ 9 v. dist = u. dist + 110 ENQUEUE(Q, v)

- The queue *Q* is a FIFO queue.
- The distance of vertex
 u = DEQUEUE(Q) in
 line 6 is monotonically
 increasing.
- Consequently, vertices are visited in breadth-first order.

This algorithm does not parallelize well.

- FIFO queue is a serial bottleneck.
- Parallelizing the **for** loops gives O(E/V) parallelism, which is puny for sparse graphs.

Summary of results



- We have designed a parallel breadth-first search algorithm, called PBFS, and we have implemented PBFS using Cilk++.
- PBFS obtains 5× to 6× speedup on eight processing cores on many real-world benchmark graphs.
- When run serially, PBFS is competitive with SERIAL-BFS.
- The theoretical running time of PBFS on *P* processors is $O((V + E)/P + D \lg^3(V/D)).$

Outline

Strategy for Parallelizing Breadth-First Search

2 The Bag Data Structure

- Bag Requirements and Usage
- Bag Design

Empirical Results

Theoretical Results

- The DAG Model of Computation
- Modeling Reducers
- Theoretical Analysis of PBFS

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Strategy: consider the graph in layers.

- The dth layer of G is the set V_d of vertices that are all at distance d from v₀.
- Breadth-first ordering: all vertices in V_d are visited before any vertex in V_{d+1}.
- We shall examine the layers V_d serially, but
- For each layer V_d , we shall process all vertices in V_d in parallel.



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Problem: We need a data structure to handle a single layer. Specifically, we need a data structure that does the following:

- It must store an unordered set of elements.
- It must support efficient parallel traversal of the stored elements.
- It must allow parallel workers to add elements simultaneously to the same structure.

Solution: Use a *bag* — a multi-set data structure, which supports the following operations:

- BAG-CREATE Create a new, empty bag.
 - BAG-INSERT Add an element to a bag.
 - BAG-SPLIT Divide a bag into two equal-sized bags.
 - BAG-UNION Combine the contents of two bags into a single bag.

```
PROCESS-LAYER(in-bag, out-bag, d)
```

```
if BAG-SIZE(in-bag) < GRAINSIZE
11
12
         for each u \in in-bag
13
              parallel for each v \in Adi[u]
14
                  if v dist == \infty
15
                        v.dist = d + 1 // benign race
16
                        BAG-INSERT(out-bag, v)
17
         return
18
    new-bag = BAG-SPLIT(in-bag)
19
    spawn PROCESS-LAYER(new-bag, out-bag, d)
    PROCESS-LAYER(in-bag, out-bag, d)
20
21
    sync
```

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```

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Cilk++ reducers

Cilk++ supports a type of parallel data structure, called a *reducer*.

1	<i>x</i> = 10	1	<i>x</i> = 10	1	<i>x</i> = 10
2	X ++	2	X ++	2	<i>x</i> ++
3	<i>x</i> += 3	3	<i>x</i> += 3	3	<i>x</i> += 3
4	<i>x</i> + = −2	4	<i>x</i> + = −2		<i>x</i> ′ = 0
5	<i>x</i> += 6	5	<i>x</i> += 6	4	x' + = -2
6	<i>X</i> ——		<i>x</i> ′ = 0	5	<i>x</i> ′ += 6
7	<i>x</i> + = 4	6	<i>x</i> ′	6	<i>x</i> ′
8	<i>x</i> += 3	7	x' + = 4		<i>x</i> ′′′ = 0
9	X ++	8	<i>x</i> ′ + = 3	7	x'' + = 4
10	<i>x</i> += -9	9	x'++	8	<i>x</i> ″ += 3
		10	x' + = -9	9	x''++
			x + = x'	10	x'' + = -9
					x + = x'

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x + = x''

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2	<i>x</i> ++	2	<i>x</i> ++	2	<i>x</i> ++
3	<i>x</i> += 3	3	<i>x</i> += 3	3	<i>x</i> += 3
4	<i>x</i> + = −2	4	<i>x</i> += -2		<i>x</i> ′ = 0
5	<i>x</i> += 6	5	<i>x</i> += 6	4	<i>x</i> ′ + = −2
6	<i>x</i> ——		<i>x</i> ′ = 0	5	x' + = 6
7	<i>x</i> + = 4	6	<i>x</i> ′	6	<i>x</i> ′
8	<i>x</i> += 3	7	<i>x</i> ′ + = 4		<i>x''</i> = 0
9	<i>x</i> ++	8	<i>x</i> ′ + = 3	7	x'' + = 4
10	x + = -9	9	x'++	8	x'' + = 3
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4	<i>x</i> + = −2	4	<i>x</i> += -2		<i>x</i> ′ = 0
5	<i>x</i> += 6	5	<i>x</i> += 6	4	<i>x</i> ′ + = −2
6	<i>x</i> ——		<i>x</i> ′ = 0	5	x' + = 6
7	<i>x</i> + = 4	6	<i>x</i> ′	6	<i>x</i> ′
8	<i>x</i> += 3	7	<i>x</i> ′ + = 4		<i>x''</i> = 0
9	<i>x</i> ++	8	<i>x</i> ′ + = 3	7	x'' + = 4
10	<i>x</i> += -9	9	<i>x</i> ′++	8	x'' + = 3
		10	x' + = -9	9	x''++
			x + = x'	10	x'' + = -9
					x + = x'

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x + = x''

We use the bag as a Cilk++ reducer to solve our malignant race.

- After stealing a task, a worker starts executing the task with a local, "identity" copy of a new reducer.
- Each worker freely manipulates its local copy with write-only update operations.
- As tasks return, the workers' local copies are combined together into a single data structure using REDUCE operations.
- If REDUCE is associative, then the program has serial semantics.

We use the bag as a Cilk++ reducer to solve our malignant race.

- After stealing a task, a worker starts executing the task with a local, "identity" copy of a new reducer.
 - For bags, the identity is an empty bag.
- Each worker freely manipulates its local copy with write-only update operations.
 - For bags, the update operation is BAG-INSERT.
- As tasks return, the workers' local copies are combined together into a single data structure using REDUCE operations.
 - For bags, REDUCE = BAG-UNION.
- If REDUCE is associative, then the program has serial semantics.
 - For bags, BAG-UNION is not strictly associative, since the order of elements within a bag is nondeterministic. Bags have a notion of "logical associativity," which is sufficient for PBFS's correctness.

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The bag data structure



- A bag is made up of *pennants* complete binary trees with extra root nodes, which store the elements.
 - Pennants may be split and combined in *O*(1) time by changing pointers.
 - A pennant is only combined with another pennant of the same size.
- A bag is an array of pointers to pennants.
 - For all *i*, the *i*th entry in the array is either null or points to a pennant of size 2^{*i*}.
 - Intuitively, a bag acts much like a binary number.

Inserting an element works similarly to incrementing a binary number.



BAG-INSERT runs in O(1) amortized time and $O(\lg n)$ worst-case time.

The bag data structure — BAG-INSERT





Splitting a bag works similarly to an arithmetic right shift.





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BAG-SPLIT runs in O(lg n) time.

Unioning two bags is works similarly to adding two binary numbers.



BAG-UNION works in $O(\lg n)$ time.

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Name Description	Spy Plot	V E D	Parallelism	PBFS T ₁ SERIAL BFS T ₁	PBFS T_1/T_8
Kkt_power Optimal power flow, nonlinear opt.		2.05 M 12.76 M 31	104.09	0.705	6.102
Freescale1 Circuit simulation		3.43 M 17.1 M 128	153.06	1.120	5.145
Cage14 DNA electrophoresis		1.51 M 27.1 M 43	246.35	1.060	5.442
Wikipedia Links between Wikipedia pages		2.4 M 41.9 M 460	179.02	0.804	6.833
Grid3D200 3D 7-point finite-diff mesh		8 M 55.8 M 598	79.27	0.747	4.902
RMat23 Scale-free graph model		2.3 M 77.9 M 8	93.22	0.835	6.794
Cage15 DNA electrophoresis		5.15 M 99.2 M 50	675.22	1.058	5.486
Nipkkt160 Nonlinear optimization		8.35 M 225.4 M 163	331.57	1.138	6.096

C.E. Leiserson, T.B. Schardl (MIT CSAIL) A Work-Efficient Parallel BFS Algorithm

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The dag model of computation



- We can model a Cilk++ program with a dag (directed acyclic graph) A.
- Each vertex in A corresponds to a strand a sequence of serially executed instructions.
- Edges in *A* describe control dependencies between strands.

The dag model of computation



- Work W(A): The sum of the lengths of all of the strands in A.
- **Span** S(A): The length of the longest path in A.
- The Cilk++ scheduler guarantees that A runs in $T_P(A) \le W(A)/P + O(S(A))$.
- **Parallelism** of A: W(A)/S(A).
- This model does not accurately represent runtime system operations on reducers.

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Modeling reducers





For a computation A:

- Consider the *user dag* User(A) a dag where runtime system operations on reducers are not represented.
 - This dag models the computation as the user understands it.
- Insert REDUCE strands performed by the runtime system into User(A) before the sync strand that requires its completion to get a *performance dag* Perf(A).
- A delay-sequence argument proves that the performance of A is $T_P(A) \leq W(\operatorname{Perf}(A))/P + O(S(\operatorname{Perf}(A))).$
- We want a performance bound in terms of User(A).



Let τ be the worst-case running time of any REDUCE operation. In the paper, we show that:

- $S(\operatorname{Perf}(A)) = O(\tau \cdot S(\operatorname{User}(A)))$ and
- $W(\operatorname{Perf}(A)) = W(\operatorname{User}(A)) + O(\tau^2 P \cdot S(\operatorname{User}(A))).$

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4

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- PBFS's user dag has O(V + E) work and O(Dlg(V/D)) span for bounded-degree input graphs.
- The worst-case cost of any BAG-UNION in PBFS is $O(\lg(V/D))$.
- Relating the user and performance dags for PBFS, we have $S(PBFS) = O(D \lg^2(V/D))$ and $W(PBFS) = O(V + E) + O(PD \lg^3(V/D)).$
- Consequently, we have $T_P(PBFS) = O((V + E)/P + D \lg^3(V/D))$.
- If O((V + E)/P) ≫ O(D lg³(V/D)), then we expect linear speedup from PBFS. We define the *effective parallelism* of PBFS to be O (V+E)/D ≥ O(E)/D ≥ O(E).

- We have seen a parallel breadth-first search algorithm implemented in Cilk++, which uses a novel Cilk reducer for unordered sets.
- Future work includes:
 - Comparing the performance of PBFS versus an implementation that uses thread-local storage instead of reducers.
 - Augmenting PBFS to return a deterministic BFS tree.
 - Parallelizing other graph algorithms, such as weighted SSSP, max flow, or min-cost flow.
 - Parallelizing other non-numeric algorithms using Cilk technologies.

Questions?

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Reducers are implemented in Cilk++ with one additional optimization.

- After stealing a task, a worker starts executing the task with a local *null* copy of a reducer.
 - For example, this null copy may be a NULL pointer to a bag.
- The first time the worker tries to manipulate its local copy of the reducer after a steal, the runtime system initializes the reducer using a CREATE-IDENTITY operation.

• For bags, CREATE-IDENTITY = BAG-CREATE.

• Modeling CREATE-IDENTITY calls in theory is similar (and simpler) than modeling REDUCE calls.

Optimizing the bag data structure

We can improve the real-world efficiency of the bag by storing an array of data at each node.



- Each node in a pennant stores a fixed-size array of data, which is guaranteed to be full.
- The bag stores an extra fixed-size array of data, called the *hopper*, which may not be full.
- Inserts first attempt to insert into the hopper. Once the hopper is full, a new, empty hopper is created while the old hopper is inserted into the bag using the original algorithm.

With this optimization, the common case for BAG-INSERT is exactly like enqueueing a vertex in a FIFO queue.

Locked PBFS

To simplify theoretical analysis, we analyze a locked version of PBFS.

PROCESS-LAYER(*in-bag*, *out-bag*, *d*) 11 if BAG-SIZE(*in-bag*) < GRAINSIZE 12 for each $u \in in$ -bag 13 parallel for each $v \in Adj[u]$ 14 if v dist == ∞ 15 if TRY-LOCK(v) if v dist == ∞ 16 17 v dist = d + 118 BAG-INSERT(*out-bag*, *v*) 19 RELEASE-LOCK(v) 20 return 21 new-bag = BAG-SPLIT(in-bag)22 **spawn** Process-Layer(*new-bag*, *out-bag*, *d*) 23 PROCESS-LAYER(*in-bag*, *out-bag*, *d*) 24 sync

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Lemma

Consider a computation A, and let τ be the worst-case running time of any REDUCE or CREATE-IDENTITY operation in A. We have $S(\text{Perf}(A)) = O(\tau \cdot S(\text{User}(A)))$ in expectation.

Proof.

- Every successful steal may force a CREATE-IDENTITY operation.
- Every successful steal may force a REDUCE operation.

(4) (5) (4) (5)

Relating the user and performance dags





Proof cont'd.

- Consider a critical path *p* in Perf(*A*).
- This path p corresponds to some path q in User(A), which has length at most S(User(A)).
- Since at most every node in *q* corresponds to a steal, the length of *p* is *O*(*τ* · *S*(User(*A*))) in expectation.

Lemma

Consider a computation A, and let τ be the worst-case running time of any REDUCE or CREATE-IDENTITY operation in A. We have $W(\text{Perf}(A)) = W(\text{User}(A)) + O(\tau^2 P \cdot S(\text{User}(A)))$ in expectation.

Proof.

- The computation A contains all of the strands in User(A).
- At most $O(P \cdot S(Perf(A)))$ steals occur during A's execution.
- Consequently, REDUCE and CREATE-IDENTITY strands contribute O(*τ*P ⋅ S(Perf(A))) additional work to User(A).
- From the previous lemma, we have $S(\text{Perf}(A)) = O(\tau \cdot S(\text{User}(A))).$
- Therefore, we have $W(\text{Perf}(A)) = W(\text{User}(A)) + O(\tau^2 P \cdot S(\text{User}(A)))$ in expectation.