PARALLEL UPDATE OF MINIMUM SPANNING TREES IN LOGARITHMIC TIME

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UNCLASSIFIED CAR-TR-97 AFOSR-TR-85-0069
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COMPUTER SCIENCE
TECHNICAL REPORT SERIES

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20742

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ABSTRACT

Parallel algorithms are presented for updating a minimum spanning tree when the cost of an edge changes or when a new node is inserted in the underlying graph. The machine model used is a parallel random access machine which allows simultaneous reads but prohibits simultaneous writes into the same memory location. The algorithms described in this paper for updating a minimum spanning tree require $O(\log n)$ time and $O(n^2)$ processors. These algorithms are efficient when compared to previously known algorithms for initial construction of a minimum spanning tree that require $O(\log^2 n)$ time and use $O(n^2)$ processors.
Parallel algorithms are presented for updating a minimum spanning tree when the cost of an edge changes or when a new node is inserted in the underlying graph. The machine model used is a parallel random access machine which allows simultaneous reads but prohibits simultaneous writes into the same memory location. The algorithms described in this paper for updating a minimum spanning tree require $O(\log n)$ time and $O(n)$ processors. These algorithms are efficient when compared to previously known algorithms for initial construction of a minimum spanning tree that require $O(\log^2 n)$ time and use $O(n^2)$ processors.
1. Introduction

Incremental graph algorithms deal with recomputing properties of graph after an incremental change is made to the graph, such as addition and deletion of vertices and edges, as well as changes in the costs or capacities (if any) associated with the edges of the graph. Such recomputations are also referred to as "updating" graph properties.


The problem of updating an MST involves reconstructing the new MST from the current MST when the cost of an edge has changed or a vertex along with all its incident edges is inserted or deleted from the underlying graph. We refer to these two subproblems as the edge update and the vertex update problem respectively. Frederickson [6] describes an $O(\sqrt{m})$ algorithm for the edge update problem, where $m$ is the number of edges in the graph. Spira and Pan [12] and Chin and Houck [2] present an $O(n)$ algorithm for updating the MST of an $n$ vertex graph when a new vertex is inserted into the graph.

Parallel algorithms for updating an MST have not been studied so far. In this paper we present parallel algorithms for updating an MST. Our model of computation is the single instruction multiple data stream (SIMD) model. We assume that all processors have access to a common memory and that simultaneous reads from the same loca-
tion are allowed but simultaneous writes to the same location are prohibited. Fortune and Wylie [5] call such model a parallel random access machine (PRAM). Savage and Ja'Ja' [11] and Chin et al. [3] have described an $O(\log^2 n)$ algorithm for constructing an MST on PRAMs. Our algorithm for the edge update problem requires $O(\log n)$ time. By using a novel approach to reconstruct an MST we also solve the vertex insertion problem in $O(\log n)$ time.

The rest of the paper is organized into four sections. In Section 2 we describe some graph-theoretic preliminaries adopting the framework in [13]. In Section 3 we describe the edge update algorithm, and the vertex insertion algorithm is described in Section 4.

2. Preliminaries

Let $G=(V,E)$ denote a graph where $V$ is a finite set of vertices and $E$ is a set of pairs of vertices called edges. If the edges are unordered pairs then $G$ is undirected else it is directed. Throughout this paper we assume that $V=\{1,2,\ldots,n\}$, $|V|=n$ and $|E|=m$. We denote the undirected edge from $a$ to $b$ by $(a,b)$ and the directed edge between them by $<a,b>$. We say that an undirected graph $G$ is connected if for every pair of vertices $u$ and $v$ in $V$, there is a path in $G$ joining $u$ and $v$. Each connected maximal subgraph of $G$ is called a component of $G$. An adjacency matrix $A$ of $G$ is an $n \times n$ Boolean matrix such that $A[u,v]=1$ if and only if $(u,v) \in E$. A tree is a connected undirected graph with no cycles in it. Let $T=(V',E')$ be a directed graph. $T$ is said to have a root $r$, if $r \in V'$ and every vertex $v \in V'$ is reachable from $r$ via a directed path. If the underlying undirected graph of $T$ is a tree then $T$ is called a directed tree. If the edges of $T$ are all reversed then the resulting graph is called an inverted tree. We denote the "undirected"
path from vertex a to vertex b by [a-b] and directed path by [a→b]. Let T be a directed
tree with u,v ∈ V'. Then the lowest common ancestor (LCA(u,v)) of u and v in T is the
vertex w ∈ V' such that w is a common ancestor of u and v, and any other common
ancestor of u and v in T is also an ancestor of w in T. Let C:E→R denote a function
that associates a cost with the edges of G. A minimum spanning tree of G is a spanning
tree of G such that sum of the costs of the edges in the tree is minimum over all span-
ning trees for G.

As we will see later on, our algorithm for updating an MST (vertex insertion in par-
ticular) requires the paths from all vertices to the root in an inverted tree. Tsin and
Chin [13] have described a technique due to Savage [10] to compute all such paths. For
completeness we now describe their technique.

Let T=(V',E') be an inverted tree with V'={1,2,...,n} and |V'|=n. Let r be the
root of this tree. For a directed edge <a,b> we say that vertex b is the father of vertex a.

**Definition:** F:V'→V' is a function such that F(i)=the father of vertex i in T for i≠r
and F(r)=r.

The function F can be represented by a directed graph F which can be constructed
from T by adding a self-loop to the root r.

From the function F, we define F^k, k≥0 as follows.

**Definition:** F^k:V'→V' (k≥0) such that F^0(i)=i, for all i ∈ V' and F^k(i)=F(F^{k-1}(i)), for
all i ∈ V' and k>0.

If i is a vertex in T, F^k(i) is the k'\textsuperscript{th} ancestor of i in the inverted tree.
Definition: For each \( i \in V \), \( \text{depth}(i) = \min\{k | F^k(i) = r \text{ and } 0 \leq k < n\} \).

Lemma 2.1: Given the function \( F \) of an inverted tree, \( F^k \) can be computed in \( O(\log n) \) time using \( O(n^2) \) processors.

Proof: To compute \( F^k \) (\( 0 \leq k < n \)) we proceed as follows. We assume that the processors are indexed as \( P(1.1), P(1.2), \ldots, P(n.n) \). The instructions within "pardo...dopar" are executed in parallel and comments are enclosed within ///://.

1. for all \( i \ (1 \leq i \leq n) \) pardo \( F^0(i) = i, F^1(i) = F(i) \) dopar; ///Processor \( P(1,i) \) executes the instruction within pardo...dopar.///

2. for \( t := 0 \) to \( \log(n-1)-1 \) do
   
   for all \( s \ (1 \leq s \leq 2^t) \) and for all \( i \ (1 \leq i \leq n) \) pardo \( F^{s+t} = F^{s}(F^t(i)) \) dopar; ///Processor \( P(s,i) \) executes the instruction within pardo...dopar.///

Now step (1) can be done in constant time using \( n \) processors. To do the \( i^{th} \) iteration of step (2) in constant time we require \( 2^i n \) processors. As there are \( \log(n-1)-1 \) iterations of step (2), we therefore require \( O(n^2) \) processors.

The actual computations of \( F^k(i) \) (\( 1 \leq i \leq n, 1 \leq k < n \)) are performed in an array \( F^+ \) in which \( F^+[i,k] \) contains \( F^k(i) \). Once the \( F^+ \) array is computed, \( \text{depth}(i) \ (1 \leq i \leq n) \) can be found by performing a binary search on the \( i^{th} \) row. We search for the left-most occurrence of \( r \). This takes \( \log n \) time by assigning a processor per row. However, it can be done in constant time by assigning a processor to each element in \( F^+ \). This is done as follows. Every processor compares its element with the elements in its left and right neighbors. There is exactly one processor which does not have all the three elements identical or distinct and this processor locates the left-most occurrence of \( r \). The depth information is stored in a one-dimensional array \( D^+ \).
After the computations for $D^+$ are finished, each row of $F^+$ is right shifted so that all the $r$'s except the left-most one are eliminated. As a consequence, the right-most column of the array contains only the root $r$. Fig. 2.1 below illustrates an inverted tree and its array $F^+$ after the rows have been shifted right.

Undefined entries are left blank.

Fig 2.1
Lemma 2.2: We can compute the lowest common ancestors of \( \binom{n}{2} \) vertex pairs in the inverted tree in \( O(\log n) \) time using \( O(n^2) \) processors.

Proof: We make use of the array \( F^+ \) to design a parallel algorithm for finding the lowest common ancestors. Let \( a \) and \( b \) be a vertex pair. If \( c \) is their lowest common ancestor, then row \( a \) and row \( b \) of \( F^+ \) will have identical contents for column \( n-1 \), column \( n-2 \), ..., down to the column containing \( c \). After this column the contents of rows \( a \) and \( b \) differ. As a result, to determine \( c \), we can perform a binary search on row \( a \) and row \( b \) simultaneously in the following way. If the two entries being examined in row \( a \) and row \( b \) (in the same column) are different, the search is continued on the right half, otherwise it is continued on the left half. It takes \( (\log n) + 1 \) time steps to find \( c \) with one processor.

Having obtained the lowest common ancestor we can now identify the unique path between any two vertices (passing through their lowest common ancestor). We now describe how to compute the maximum cost edge on the unique path between any two vertices.

Let \( E_m(e_1, e_2) \) denote the maximum cost edge between edges \( e_1 \) and \( e_2 \). Let \( F_m(i) \) \((1 \leq i \leq n)\) be the maximum cost edge on the path from \( i \) to its \( k \)th ancestor in \( T \). Then

1. \( F_m(i) \) is the edge \((i, F^i(i))\)

2. \( F_m^k(i) \) is the edge \( E_m(F_m^k(i), (F_{k-1}^k(i), F^k(i))) \), \( k > 1 \).

We assume that the cost of the edge \((r, r)\) in \( T \) is \(-\infty\).

Lemma 2.3: We can compute \( F_m^k(i) \) for all \( i \) \((1 \leq i \leq n)\) in \( O(\log n) \) time using \( O(n^2) \) processors.
Proof: We describe an algorithm to compute $F^+_m(i)$.

1. For all $i$ ($1 \leq i \leq n$) pardo $F^+_m(i) = (i, F^+(i))$ dopar;

2. For $t := 0$ to $\log(n-1)-1$ do

   For all $s$ ($1 \leq s \leq 2^t$) and for all $i$ ($1 \leq i \leq n$) pardo $F^+_{m^2}(i) = E_m(F^+_{m^2}(i), F^+_{m^2}(F^+_{2i}(i)))$ dopar;

   The analysis is similar to the proof of Lemma 2.1. □

The computations of $F^+_m(i)$ are done in a two-dimensional array $F^+_m$. $F^+_m(i,k)$ is the maximum cost edge on the path from $i$ to its $k^{th}$ ancestor.

Lemma 2.4: Given $F^+, E^+_m, and D^+$, we can find the maximum cost edge on the path $[u-v]$ (for all $u,v \in V'$) in $O(\log n)$ time using $O(n^2)$ processors.

Proof: First we find LCA($u,v$). By Lemma 2.2, this can be done in $O(\log n)$ time using $O(n^2)$ processors. Let $p$ be the LCA($u,v$). Let $x = D^+|u| - D^+|p|$ and $y = D^+|v| - D^+|p|$ (that is, $p$ is the $x^{th}$ ancestor of $u$ and the $y^{th}$ ancestor of $v$). Finding $x$ and $y$ for all pairs $u,v$ takes constant time using $O(n^2)$ processors. Finally, the maximum cost edge on the path $[u-v]$ is $E_m(F^+_m(u), F^+_m(v))$. This again can be computed for all pairs $u,v$ in constant time using $O(n^2)$ processors. □

The maximum cost edge on the path $[u-v]$ for all pairs $u,v$ is stored in a two-dimensional array $M^+$. Note that if $F$ represents a forest of inverted trees then $F^+, D^+$ and $M^+$ contain information about ancestors, depth and maximum cost edges for all the trees in the forest.
3. Edge Update Algorithms

The edge update problem is concerned with reconstructing the new MST when the cost of an edge in the underlying graph changes. There are several cases to be handled in edge-cost updating. The cost of an edge may either increase or decrease and this edge may currently be either in the tree or not in the tree. If the cost of a tree edge decreases, or the cost of a non-tree edge increases, then the old MST will not undergo any change. On the other hand, it may undergo changes when the cost of a tree edge increases or the cost of a non-tree edge decreases. However, in both cases, at most one edge will enter the tree and another edge will leave it.

If the cost of a tree-edge \((x,y)\) increases then the new MST is recomputed as follows.

1.Delete the tree-edge \((x,y)\). This creates a forest of two subtrees.
2.Identify the vertices in each of these subtrees.
3.Find the minimum cost edge connecting them.

If the cost of a non-tree edge \((u,v)\) decreases, then we proceed to recompute the new MST as follows.

1. Add \((u,v)\) to the old MST. The edge \((u,v)\) induces a cycle in the old MST.
2. Remove the maximum cost edge on this cycle.

See Chin and Houck [2] for a proof of correctness of both these algorithms.

We assume that the update algorithms operate on an MST in the form of an inverted tree (see Section 2) with an arbitrary vertex as the root. After an edge update, the algorithms ensure that the reconstructed MST is also preserved as an inverted tree.

Using the technique of Tsin and Chin [13], the parallel algorithm for constructing the
MST in [3.11] can be easily modified to yield an MST in the form of an inverted tree. Such an algorithm requires $O(\log^2 n)$ time and uses $O(n^2)$ processors.

We now describe a parallel algorithm to update the MST when the cost of a tree edge $(x,y)$ increases. Let $r$ be the root of the inverted MST. The steps are as follows.

1. We assume, without loss of generality, that the direction of edge $(x,y)$ is from $x$ to $y$. Now set $F^1(x) = x$. This deletes the directed edge $<x,y>$ from the inverted MST and creates a forest of two subtrees, one of which is rooted at $r$ and the other at $x$. This step can be done in constant time with a single processor.

2. Compute the array $F^r$. By Lemma 2.1, this can be done in $O(\log n)$ time using $O(n^2)$ processors. At the end of this step, all vertices in the subtree rooted at $r$ will have $r$ in their last column in $F^r$ and all vertices in the subtree rooted at $x$ will have $x$ in their last column. We therefore can identify the vertices in the two subtrees.

3. Determine the minimum cost edge connecting these two subtrees. This involves the following steps.

   3a. For each vertex $i$ find the minimum cost edge $(i,j)$ such that $i$ and $j$ are not in the same subtree. Since there are at most $n$ edges incident on $i$, step (3a) can be done in $O(\log n)$ time using $O(n^2)$ processors by assigning $n$ processors to each vertex.

   3b. The minimum cost edge connecting these two subtrees can now be found by selecting the minimum cost edge among the edges selected in step (3a). As there are at most $n$ edges, such a selection can again be done in $O(\log n)$ time using $O(n)$ processors.
Let \((u,v)\) be the edge selected in step (3). If edge \((u,v)\) is the same as edge \((x,y)\) then let \(F'(x) = y\) (that is, the old MST does not change). On the other hand, if edge \((u,v)\) is not the same as edge \((x,y)\) then the two subtrees and the edge \((u,v)\) form the new MST.

5. Finally, we must maintain the new MST is an inverted tree. To do so, we proceed as follows.

Assume, without loss of generality, that \(u\) is in the subtree rooted at \(x\) and \(v\) is in the subtree rooted at \(r\). Now orient the edge \((u,v)\) from \(u\) to \(v\). To do so set \(F'(u) = v\). In step (2) we found the path from vertex \(v\) to \(x\). Now reverse the directions of the edges on the directed path \([v \rightarrow x]\) in the old inverted MST. For instance, if the directed edge \(-ab-\) was on the directed path \([v \rightarrow x]\) then set \(F'(b) = a\). This path can have at most \(n\) edges and hence the reversal can be done in constant time using \(O(n)\) processors.

This completes the description of the parallel algorithm to update the MST when the cost of a tree edge \((x,y)\) increases. We now describe a parallel algorithm to update the MST when the cost of a non-tree edge \((u,v)\) decreases. Again, let \(r\) be the root of the inverted MST. The steps then are as follows.

1. Compute arrays \(F^1, F_n^1\) and \(D^1\). By Lemma 2.4, we can find the maximum cost edge \((x,y)\) on the path \([u-v]\) in the MST is \(O(\log n)\) time using \(O(n^2)\) processors.

2. If the cost of edge \((x,y)\) is less or equal to the cost of edge \((u,v)\) then the old MST does not change. Otherwise, the edge \((x,y)\) must be deleted from the MST and edge \((u,v)\) must be added.

3. Assume, without loss of generality, that the direction of edge \((x,y)\) in the inverted MST is from \(x\) to \(y\). Now set \(F'(x) = x\) and compute \(F^1\). If \(u\) is in the subtree
rooted at x then direct edge \((u,v)\) from \(u\) to \(v\), else direct it from \(v\) to \(u\). Computing \(F^+\) takes \(O(\log n)\) time using \(O(n^2)\) processors.

4. Finally, we have to maintain the new MST as an inverted tree. This can be done in constant time using \(O(n)\) processors (see step (5) of the previous algorithm).

Note that edge insertion and edge deletion can be easily handled by our algorithms. Assign large positive costs \((+\infty)\) to edges not in the underlying graph. If such an edge is inserted into the graph then we can consider this as equivalent to decreasing the cost of a non-tree edge. Similarly edge deletion from the MST can be handled by again assigning a large positive cost to that edge and this in turn is equivalent to increasing the cost of a tree edge.

4. **Vertex Update Algorithm**

The vertex update problem involves reconstructing the new MST when a vertex is either inserted or deleted from the underlying graph. We now describe our method of handling the vertex update problem when a new node is inserted into the underlying graph. The other case of reconstructing the MST when a vertex is deleted from the graphs appears difficult to handle. For instance, if the MST is in the form of a "star" (that is, there exists a vertex on which all the edges in the MST are incident), the deletion of such a vertex deletes all the edges in the tree. Updating the MST then requires reconstructing it all over again (that is, by examining all the remaining edges in the graph).

Spira and Pan [12] update the MST in \(O(n)\) time when a vertex is inserted in the graph. Their algorithm constructs the MST all over again by examining the \(n-1\) edges in the old MST and the new edges (there can be at most \(n\) of them) brought in by the
inserted vertex. The best sequential algorithm for constructing an MST requires $O(n^2)$
time for dense graphs [9] and $O(m \log \log n)$ time for sparse graphs [14] (recall that $m$ is
the number of edges in the graph). The $O(n)$ time complexity obtained by Spira and
Pan to update the MST is primarily due to the smaller number of edges that need to be
examined. However, parallel algorithms to construct an MST [3,11] by just examining
the edges in the old MST and the new edges brought in by the inserted vertex still
requires $O(\log^2 n)$ time. Chin and Houck [2] also describe a sequential algorithm of time
complexity $O(n)$ for the vertex update problem when a new vertex is inserted in the
graph. Their algorithm, however, is inherently sequential.

Our solution to this problem requires a novel way of examining the old tree edges
and the new edges brought in by the inserted vertex. Every pair of edges incident on
the new vertex induces a cycle in the old MST. At most $n$ such edges are incident on
the inserted vertex thereby creating $nC_2$, that is $O(n^2)$ cycles. We break all these cycles
simultaneously by removing the maximum-cost edge on each cycle. We show later on
that the resulting graph is a minimum spanning tree. The details of our algorithm are
as follows.

Let $z$ be the new vertex inserted in the graph.

1. Compute array $M^+$ for the old MST. By Lemma 2.4, this can be done in $O(\log n)$
time using $O(n^2)$ processors.

2. Find out the maximum cost edge on each cycles induced by $z$ in the old MST. For
instance, let $u$ and $v$ be any two vertices in the MST and let $(z,u)$ and $(z,v)$ be the
two new edges incident on $u$ and $v$. Now $M^+(u,v)$ is the maximum cost edge on the
path $[u-v]$. The maximum cost edge on the cycle formed by the edge $(z,u)$, the path
$[u-v]$ and the edge $(v,z)$ is obtained by selecting the maximum cost edge among the
edges \((z,u), (v,z)\) and \(M^*[u,v]\) (recall that \(M^*[u,v]\) is the maximum cost edge on the path \([u-v]\) in the MST). This selection can be done in constant time using \(O(n^2)\) processors.

3. Delete the maximum cost edges selected in step (2). For instance, let \((x,y)\) be such an edge. Assume, without loss of generality, that its direction in the inverted tree is from \(x\) to \(y\). Then, to delete this edge from the inverted tree set \(F^+(x) = x\). This is done by setting \(F^+(x, 1) = x\) in the \(F^+\) array. Since an edge may be selected for deletion by more than one processor, a write conflict may arise. However, such a conflict can be avoided by using the buddy-system technique due to Hirschberg [7]. This technique deletes all the selected edges without write conflicts in \(O(\log n)\) time using \(O(n^2)\) processors.

4. Finally, we must maintain the new MST as an inverted tree. To do so, we proceed as follows. The subtrees created by the deletion of the edges in the old MST are now connected to each other through edges incident on \(z\). Let \(x_1, x_2, \ldots, x_k\) be the roots of the \(k\) such subtrees formed in step (3). Let \(w_1, w_2, \ldots, w_k\) be the vertices in the subtrees rooted at \(x_1, x_2, \ldots, x_k\) respectively that have edges incident on \(z\). Compute array \(F^+\). This array contains the paths from \(w_1\) to \(x_1\), \(w_2\) to \(x_2\), \ldots, \(w_k\) to \(x_k\). Now reverse the direction of all the edges on these paths. Next, orient all the edges \((w_1, z), (w_2, z), \ldots, (w_k, z)\) towards \(z\). Thus, \(z\) becomes the root of the inverted tree representing the new MST. By Lemma 2.1, computation of the \(F^+\) array requires \(O(\log n)\) time using \(O(n^2)\) processors. Also, reversal of the edges can be done in constant time using \(O(n^2)\) processors.

This completes the description of the algorithm. We will now show that our algorithm indeed produces an MST.
Theorem 4.1: Our algorithm computes the new MST after a vertex insertion.

Proof: Let $T'$ be the graph obtained after steps (1), (2), (3) and (4) of our algorithm are executed.

First, $T'$ is acyclic as all the cycles are broken in step (3) of our algorithm. We next show that $T'$ is connected. Consider a vertex $u$ in $T'$ and let $e_1, e_2, \ldots, e_k$ be the edges incident on it (see Fig. 4.1).

![Fig 4.1](image)

Let edge $e_j$ ($j \leq k$) be the minimum cost edge among them. Now consider all the cycles containing $e_j$ that pass through $u$. They must also contain some other $e_i$ ($e_i \neq e_j$). When these cycles are broken in step (3) of our algorithm $e_j$ is retained as it is the minimum cost edge incident on $u$. (If at all any edge incident on $u$ is selected for deletion for cycles passing through $u$, it is not $e_j$.) Therefore the minimum cost edge incident on each vertex is retained. This in turn creates at most $\frac{n}{2}$ components.

Now, assume that there are $k$ ($k > 1$) such components. Using a similar argument based on cycles passing through each component it is easy to see that the minimum cost edge incident on each component is retained. This in turn creates at most $\frac{k}{2}$ components. As the number of components monotonically decrease, we are eventually left with one component. Therefore $T'$ is connected. $T'$ is a tree as it is acyclic and con-
Finally, it is well known that every non-tree edge is the maximum cost edge on the cycle it induces in the MST. Our algorithm deletes the maximum cost edge on each cycle induced by \( z \). Hence \( T' \) is an MST.

**Theorem 4.2:** Our algorithm takes \( O(\log n) \) time and requires \( O(n^2) \) processors.

**Proof:** Steps (1), (2), (3) and (4) of our algorithm.

5. Conclusions

Incremental graph algorithms deal with recomputing properties of a graph after an incremental change has been made to the graph. In this paper we have examined the problem of updating a minimum spanning tree. We have described a parallel algorithm to update an MST when the cost of an edge changes or a new node is inserted in the underlying graph. Our algorithm requires \( O(\log n) \) time using \( O(n^2) \) processors. It is therefore efficient when compared to parallel algorithms for initial construction of a minimum spanning tree which take \( O(\log^2 n) \) time and use \( O(n^2) \) processors.

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