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## I N R I A

## Constructing elimination trees for unsymmetric matrices

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Distributed and High Performance Computing


# Constructing elimination trees for unsymmetric matrices 

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#### Abstract

The elimination tree model for unsymmetric matrices and an algorithm for constructing it have been recently proposed [Eisenstat and Liu, SIAM J. Matrix Anal. Appl., 26 (2005) and 29 (2008)]. The construction algorithm is of time complexity $\mathcal{O}(m n)$ for an $n \times n$ unsymmetric matrix having $m$ nonzeros. We propose another algorithm with a time complexity of $\mathcal{O}(m \log n)$.


Key-words: Elimination tree, sparse matrix factorization

[^0]
## Construire les arbes d'élimination pour des matrices creuses non symétriques

Résumé : Eisenstat et Liu ont récemment décrit l'arbre d'élimination pour des matrices creuses non symétriques et ont proposé un algorithme pour le construire en temps $\mathcal{O}(m n)$ pour une matrice de taille $n$ ayant $m$ éléments non nuls [Eisenstat and Liu, SIAM J. Matrix Anal. Appl., 26 (2005) and 29 (2008)] Nous décrirons un algorithme dont la complexité en temps est $\mathcal{O}(m \log n)$.
Mots-clés : Arbre d'élimination, factorisation des matrices creuses

## 1 Introduction

Arguably, the elimination tree is the single most important data structure in sparse matrix factorization methods. It is used to estimate and optimize the storage and computational requirements during symbolic and numerical factorizations [4]. Although the elimination tree for symmetric (positive definite) systems goes back to 80s [5] and even before (see [1, Section 3.3] and [4]), the elimination tree for unsymmetric matrices is a recent one. Eisenstat and Liu [2] define the the elimination tree for unsymmetric matrices and discuss its properties. In a follow up paper [3], they describe algorithms to construct those trees, to reorder matrices into a special form, and to use the trees to perform symbolic factorization. In this short note, we present an $\mathcal{O}(m \log n)$ time algorithm to construct the elimination tree of an $n \times n$ unsymmetric matrix with $m$ nonzeros. The algorithm by Eisenstat and Liu [3] has a time complexity of $\mathcal{O}(m n)$.

Let $\mathbf{A}$ be a sparse $n \times n$ matrix $\mathbf{A}$ with $m$ nonzeros, and $a_{i j}$ be the entry at its $i$ th row and $j$ th column. The directed graph $G(\mathbf{A})=(V, E)$ of $\mathbf{A}$ contains $n$ vertices, $V=\{1,2, \ldots, n\}$, and $m$ edges, $E=\left\{i j: a_{i j} \neq 0,1 \leq i, j \leq n\right\}$. For our purposes in this paper, we assume that $\mathbf{A}$ has a zero-free diagonal.

Let $G=(V, E)$ be a directed graph (digraph) where $V$ and $E$ are the sets of $n$ vertices and $m$ edges, respectively. An edge from vertex $u$ to vertex $v$ will be denoted as $u v$. If $u=v$, we say that $u v$ is a loop. A path is an alternating sequence of vertices and edges which starts and ends with a vertex. If a path's first and last vertex are the same we say that path is closed. For two vertices $u, v \in V$, we say that $u$ is connected to $v$ in $G$, i.e., $u \xrightarrow{G} v$, if there is path from $u$ to $v$.

A directed graph $G^{\prime}=\left(V^{\prime}, E^{\prime}\right)$ is a subgraph of a larger digraph $G=(V, E)$ if $V^{\prime} \subset V$ and $E^{\prime} \subseteq E \cap\left(V^{\prime} \times V^{\prime}\right)$. A subgraph is said to be a maximal subgraph if $E^{\prime}=E \cap\left(V^{\prime} \times V^{\prime}\right)$. Let $V=\{1,2, \ldots, n\}$ and we have a well defined order on the vertices. Then, $G_{k}$ denotes a special maximal subgraph containing the first $k$ vertices in $V$. That is,

$$
G_{k}=(\{1, \ldots, k\}, E \cap\{1, \ldots, k\} \times\{1, \ldots, k\}) .
$$

If $u \xrightarrow{G} v$ for all $u, v \in V$, the digraph $G$ is called strongly connected. If $G^{\prime}$ is maximally strongly connected, i.e., if there is no strongly connected subgraph $G^{\prime \prime}$ of $G$ such that $G^{\prime}$ is a subgraph of $G^{\prime \prime}$, it is called a strong component (or a strongly connected component) of $G$. Due to the zero-free diagonal assumption, the vertices have self loops. Hence, if $G$ contains a single vertex, it is strongly connected. A directed graph without cycles is called a Dag. The strongly connected components of a Dag are its vertices, i.e., each vertex forms a trivial strong component.

Let $\mathbf{A}=\mathbf{L} \mathbf{U}$ be a nonsingular, sparse, unsymmetric $n \times n$ matrix with a nonzero diagonal where $\mathbf{L}$ and $\mathbf{U}$ be its lower and upper triangular factors, respectively. The elimination tree $T(\mathbf{A})$ of an unsymmetric $\mathbf{A}$ is defined by Eisenstat and Liu $[2,3]$ as follows.

$$
\begin{equation*}
\operatorname{Parent}(i)=\{j: j>i \text { and } j \xrightarrow{G(\mathbf{L})} i \xrightarrow{G(\mathbf{U})} j\} \tag{1}
\end{equation*}
$$

where $\operatorname{Parent}(i)$ shows the parent of vertex $i$ in the elimination tree if $i$ is not a root. Otherwise, Parent $(i)=\infty$. Eisenstat and Liu prove the following the-
orem in order to develop an algorithm for constructing $T(\mathbf{A})$ without knowing $G(\mathbf{L})$ and $G(\mathbf{U})$.

Theorem 1.1 (Theorem 3.3 of [2]) Vertex $k$ is the parent of vertex $i$ in the elimination tree $T(\mathbf{A})$ if and only if $k$ is the first vertex after $i$ such that $k$ and $i$ belong to the same strongly connected component of the subgraph $G_{k}(\mathbf{A})$ of $G(\mathbf{A})$.

Given this theorem, Eisenstat and Liu [3] propose the following algorithm to compute the parent pointers in (1) and construct the elimination tree.

```
Algorithm 1 ETREE
    for vertex \(k=1\) to \(n\) do
        Find the component \(C\) of \(G_{k}(\mathbf{A})\) that contains \(k\)
        for each vertex \(i \in C \backslash\{k\}\) do
            if \(\operatorname{Parent}(i)=\infty\) then
                \(\operatorname{Parent}(i)=k\)
        \(\operatorname{Parent}(k)=\infty\)
```

The strong components of a digraph with $m$ edges and $n$ vertices can be found in $\mathcal{O}(n+m)$ time [6]. Hence, the complexity of the algorithm eTree is $\mathcal{O}(m n)$ for a graph containing $n$ vertices and $m$ edges. Eisentat and Liu reduce the practical running time of the algorithm by using quotient graphs [3] and specialized algorithms to find the strong component containing a given vertex, but report that the time complexity of the improved algorithm is still $\mathcal{O}(m n)$.

## 2 An $\mathcal{O}(m \log n)$ time algorithm

Let $\mathbf{A}$ be an irreducible, unsymmetric, $n \times n$ matrix with a nonzero diagonal. Algorithm 2 shows the proposed algorithm ETREESC which finds the elimination tree $T(\mathbf{A})$ in a bottom-up fashion. There are two inputs for the algorithm: a strongly connected digraph $G$ with $\eta$ vertices and an integer $s<\eta$ such that $G_{s}$ is acyclic. That is, all the strong components of $G_{s}$ are trivial.The algorithm initial call is $\operatorname{ETreeSC}(G(\mathbf{A})=(\{1,2, \ldots, n\}, E), 0)$. We assume that the parent pointers are initialized to $\infty$ before the execution of the algorithm. Our algorithm and its analysis are based on an algorithm by Tarjan [7].

For each call of eTreesC, each vertex $i_{j} \in V$ such that $j \leq s$ represents a subtree in $T(\mathbf{A})$ with root $i_{j}$. The algorithm tries to find the first vertex $i_{k}$ that connects some or all of these subtrees. Since $G$ is strongly connected and $G_{s}$ is known to be acyclic, if $s=\eta-1, i_{\eta}$ must be the vertex that makes $G$ strongly connected. If this is the case, we set the parent pointers appropriately. If this is not the case, the algorithm continues to the search with the vertex $i_{k}$ where $k=\lceil(s+\eta) / 2\rceil$. If $G_{k}$ is strongly connected, the search continues recursively. On the other hand, if $G_{k}$ is not strongly connected, we know that each strong component corresponds to a subtree in the elimination tree. In this case, each recursive call at line 15 is responsible for setting the parent pointers of the vertices in these subtrees, and the one at 20 is responsible for setting the pointers of the roots of these subtrees. For the last call, we create a new graph $G^{\prime}=\left(V^{\prime}, E^{\prime}\right)($ from $G)$, which contains a super-vertex for each strong component of $G_{k}$. That is, the vertex set $V_{\ell}$ of each strong component $C_{\ell}$ is

```
Algorithm \(2 r=\operatorname{ETREESC}\left(G=\left(\left\{i_{1}, i_{2}, \ldots, i_{\eta}\right\}, E\right), s\right)\)
Output \(r\) : the root of the elimination tree of \(G\)
    if \(s=\eta-1\) then
        for \(j=1\) to \(\eta-1\) do
            \(\operatorname{Parent}\left(i_{j}\right)=i_{\eta}\)
        return \(i_{\eta}>i_{\eta}\) is the root
    else
        \(k=\lceil(s+\eta) / 2\rceil\)
        Let \(p\) be the number of strong components of \(G_{k}\)
        if \(p=1\) then
            return \(\operatorname{ETreeSC}\left(G_{k}, s\right)\)
        else
            for \(\ell=1\) to \(p\) do
                Let \(C_{\ell}=\left(V_{\ell}, E_{\ell}\right)\) be the \(\ell\) th strong component of \(G_{k}\)
                if \(\left|V_{\ell}\right|>1\) then
                    \(s_{\ell}=\left|\left\{i_{j}: i_{j} \in V_{\ell}, j \leq s\right\}\right| \nabla\) the first \(s\) vertices form a Dag
                    \(r_{\ell} \leftarrow \operatorname{ETREESC}\left(C_{\ell}, s_{\ell}\right) \quad\) find the root of each str. comp.
                    else
                    \(r_{\ell} \leftarrow\) the vertex in \(V_{\ell}\)
                    Shrunk \(V_{\ell}\) into the vertex \(r_{\ell}\)
            Let \(V^{\prime}=\left\{r_{1}, \ldots, r_{p}, i_{k+1}, i_{k+2}, \ldots, i_{\eta}\right\}\)
            return \(\operatorname{ETrEESC}\left(G^{\prime}=\left(V^{\prime}, E^{\prime}\right), p\right) \quad G_{p}^{\prime}\) is acyclic
```

shrunk into a single super-vertex $r_{\ell} \in V^{\prime}$ for $1 \leq \ell \leq p$. In the elimination tree, $r_{\ell}$ is the root of the subtree containing the vertices in $V_{\ell}$. The edge set $E^{\prime}$ contains an edge for each edge in $E$ except the ones whose endpoints are in the same strong component of $G_{k}$, and a self loop for each super-vertex. If an endpoint of an edge is in a strong component of $G_{k}$, we use the corresponding super-vertex in $V^{\prime}$ for that edge.

For a recursive call with a strong component $C_{\ell}, s_{\ell}$ is equal to the number of vertices in $V_{\ell}$ at an index smaller than or equal to $s$. Hence, the subgraph of $C_{\ell}$ containing the first $s_{\ell}$ vertices is acyclic. On the other hand, for the graph $G^{\prime}$, we know that the first $p$ vertices form a directed acyclic graph since they correspond to the strong components of $G_{k}$.

We adapt the analysis in [7] to find the time complexity of Algorithm 2. Forgetting the recursive calls, the complexity of the body of eTreesC is equal to the complexity of assigning the parent pointers and finding the strong components which is $\mathcal{O}(m)$. In total, line 3 will be executed as many times as the number of vertices in the original graph. For the recursive calls, each edge of $G$ is used in at most one recursive call. Let $\eta-s$ be the rank of the problem, i.e., the number of vertices to be searched. The ranks of the recursive calls at lines 9,15 and 20 are at most $\max (k-s, \eta-k)$. As $k, s$ and $\eta$ are integers, and due to the definition of $k$ at line 6 , for a call with rank $\rho$, the ranks of its recursive calls are at most $2 \rho / 3$. Let $R(m, \rho)$ be the time complexity of a problem (excluding the part for setting the parent pointers) with $m$ edges and rank $\rho$, and $c$ be the number of recursive calls. Then

$$
R(m, \rho)=\mathcal{O}(m)+\sum_{i=1}^{c} R\left(m_{i}, \rho_{i}\right)
$$

By induction, we will show that $R(m, \rho)=\mathcal{O}(m \log \rho)$. We assume that the equation is correct for all $m_{i}$ and $\rho_{i}$. For the base case, when $\rho=1$, the algorithm only sets the parent pointers hence the assumption holds. Since,
$\sum_{i=1}^{c} m_{i} \leq m$, by the assumption,

$$
R(m, \rho)=\mathcal{O}(m)+R\left(m, \frac{2 \rho}{3}\right)=\mathcal{O}(m \log \rho)
$$

where the last equality is due to the well-known master theorem. Hence, the time complexity of the algorithm is $\mathcal{O}(m \log n+n)$. As $m \geq n$, we can conclude that the complexity is $\mathcal{O}(m \log n)$.

Consider the sample matrix from [3] shown in Fig. 1(a) whose graph is shown in Fig. 1(b). The first call with the whole graph finds $k=5, p=3$ with $V_{1}=\{1\}$,


Figure 1: Tracing the algorithm on the sample matrix.
$V_{2}=\{4\}, V_{3}=\{2,3,5\}$, and then recursively calls the algorithm in $V_{3}$; then the algorithm reaches to line 20, yielding the subtree shown in Fig. 1(c) with $V^{\prime}=\left\{r_{1}=1 ; r_{2}=4 ; r_{3}=5 ; 6 ; 7 ; 8 ; 9 ; 10\right\}$. The recursive call is invoked with $s=3$ and then finds $k=6$ (pointing to vertex 8). Three strong components $V_{1}=\{1,4,6,7\}, V_{2}=\{2,3,5\}$, and $V_{3}=\{8\}$ are found, and 7 becomes the root of $V_{1}$ after the recursive call (the subtree is at Fig. 1(d)). Then the algorithm reaches again to line 20 with $V^{\prime}=\left\{r_{1}=7 ; r_{2}=5 ; r_{3}=8 ; 9 ; 10\right\}$ and $s=3$. This time $k=4$ (points at vertex 9 ). Again three strong components are found and 9 becomes the root of the component containing 5 , yielding the tree shown in Fig. 20. The last call results in the elimination tree shown in Fig. 1(f).

## 3 Conclusion

We have proposed an algorithm to compute the elimination tree of an $n \times n$ unsymmetric matrix with $m$ nonzeros in $\mathcal{O}(m \log n)$ time. The previously known algorithm [3] is of time complexity $\mathcal{O}(m n)$.

The proposed algorithm is based on an algorithm by Tarjan [7]. It is interesting to note (as noted by Liu [4, p.142]) that the elimination tree for symmetric matrices can also be constructed by an algorithm by Tarjan and Yannakakis (the fill-in computation algorithm [8, p. 570]). If one removes the line "add $\{x, v\}$ to" and changes the test "index $(x)<i$ " into" $f(x) \neq x$ " then the $f(\cdot)$ pointers become exactly the parent pointers. The modifications we needed for the unsymmetric case are a little more involved than those for the symmetric case, but still this is a curious coincidence.

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