

LINEAR VERIFICATION FOR SPANNING TREES

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Given a rooted tree with values associated with the n vertices and a set A of directed paths (queries), we describe an algorithm which finds the maximum value of every one of the given paths, and which uses only

$$5n + n \log \frac{|A| + n}{n}$$

comparisons.

This leads to a spanning tree verification algorithm using $O(n+e)$ comparisons in a graph with n vertices and e edges.

No implementation is offered.

0. Introduction

1. Finding the minimal spanning tree in an undirected network is a well-researched area of computer science. The classical algorithms of Kruskal and Prim have been modified and improved several times. For a study of several spanning tree algorithms, see [1].

The best known result has been the $O(|E| \log \log |V|)$ algorithm of Yao [6] until recently.

A few weeks ago, Fredman and Tarjan [2] developed a method which applies to both the shortest path and the spanning tree problems, leading to an $O(|E| \beta(|E|, |V|))$ algorithm for the latter one; where $\beta(m, n) = \min \{i | \log^{(i)} n \leq m/n\}$. On the other hand, the only verification result we know of is the $O(|E| \alpha(|E|, |V|))$ algorithm of Tarjan [5]; (here α is the inverse Ackermann function).

Here we describe an algorithm which finds maxima over various paths of a tree, which leads to a minimal spanning tree verification algorithm with a *linear number of comparisons*.

We want to emphasize, however, that the only cost we deal with is the total number of comparisons made, for we could not find an effective implementation with a linear overhead cost. In other words, our result is of an information theoretical nature.

We remark that the problem is a particular instance of the following general question that is discussed in [3].

(Q) Given an n element set $E=(e_1, \dots, e_n)$, and a list of m subsets of $\{1, 2, \dots, n\}$, $L=(S_1, \dots, S_m)$. Find the maxima

$$M_i = \max_{j \in S_i} e_j, \quad i = 1, 2, \dots, m.$$

(If E is the set of edges of the spanning tree and the elements of S_i are the circuit edges created by the i -th outside edge, we get the tree verification problem.)

Of course, sorting the whole list E provides all the necessary information for finding all M_i (overhead is not counted!), but one would hope for an algorithm using only $O(m+n)$ comparisons.

Fredman proved (see [3]) that the number of possible outcomes is not more than $\binom{m+n-1}{n-1} < 2^{m+n}$, thus the above hope is realistic.

A family of paths on trees provides enough structure to make the problem easier to attack.

The general question (Q) is still unanswered.

2. Given an undirected network G (a graph with n vertices, e edges and real values associated with the edges) and a spanning tree T of G , we want to test whether T is minimal among all spanning trees of G .

Any edge x of G not in T creates a unique circuit C_x with edges of T ; and it is well-known that T is minimal if and only if, for any outside edge x , the value of x is not smaller than any value in C_x .

Thus, we only need to know, for all outside x , the maximum value on the path $C_x - x$, so that we can compare this maximum with the value of x . Note that $C_x - x$ consists entirely of edges of T .

Let us root T by a leaf of T , and consider it a directed tree with edges directed away from the root. Any path of T is the union of at most two directed paths, and so it is sufficient to find the maxima on the directed "half-paths" corresponding to the outside edges.

By reassigning the values of the edges to their lower endpoints (and deleting the root), we get a more attractive model, in which the values are associated with the vertices.

Whatever cost we obtain for this directed path problem, we only need $2(e-n+1)$ extra comparisons for the spanning tree verification problem.

Notation

T is a rooted tree with edges directed away from the root.

$V(T)$ = set of vertices of T

$T-U$ for a vertex set $U \subset V(T)$ is the graph T restricted to $V(T) - U$

$x \cong y$ for $x, y \in V(T)$ means x is a predecessor of y on T

$x \succ y$ means $x \cong y$ and $x \neq y$

$\text{deg}(x)$ is the number of outgoing edges from x

path means directed path of T

$p(x, y)$ stands for the directed path $(u|x \cong u \cong y)$

$\log x$ is binary logarithm

Formulation of the problem

Given a rooted tree T with real values associated with the vertices $f: V(T) \rightarrow \mathbf{R}$ and a set A of directed paths of T (queries)

$$A \subset Q = \{p(x, y) | x, y \in V(T), x \cong y\}.$$

Find the maximum

$$\max_{x \cong u \cong y} f(u)$$

for all $p(x, y) \in A$.

We will present a solution using less than

$$C = 5n + n \log \frac{|A| + n}{n}$$

comparisons.

Note that $C \ll |A|$ for $|A| \gg n$, thus the linear term $O(n)$ for the tree verification problem ($|A| = e - n + 1$) comes from the comparisons made between the maxima in A and the outside edges.

The paper is structured as follows. First we will describe two completely different algorithms for two particular cases: when T is a string (Section 1, this is even implementable) and when T is a full branching tree (Section 2), and then we will show how a general tree can be interpreted as a mixture of these two extremes.

1. When T is a string

In other words, we have an array, $[f(i); 1 \leq i \leq n]$, and want to find maxima over intervals $[f(i); s \leq i \leq t]$. Although there are $\binom{n+1}{2}$ such intervals, we give an algorithm that uses less than $2n$ comparisons, and still can find the answers for all cn^2 queries (with a bounded overhead per query).

This easy part of our algorithm may be folklore, but we could not trace it in the literature.

Symmetric order heaps

Given an array $[f(i); 1 \leq i \leq n]$, we construct a binary tree SH on n nodes with the following properties:

1. $f(i)$ is assigned to node v_i of SH ,
2. SH is a heap; i.e. $v_i > v_j$ implies $f(i) \geq f(j)$,
3. v_i are in symmetric order; i.e., if v_j belongs to the left (right) subtree of v_i

then $j < i$ ($j > i$).

Clearly, these properties uniquely determine SH , namely the root of SH must be $f(m) = \max_{1 \leq i \leq n} f(i)$, and the elements $f(i), 1 \leq i < m$ ($f(i), m < i \leq n$) should form the left subtree (right subtree) of v_m ; proceed recursively inside these subtrees.

Once the tree SH has been constructed, finding the maximum over a subarray $[f(i); s \leq i \leq t]$ reduces to determining the lowest common ancestor of v_s and v_t . Harel [4] has an ingenious algorithm which, after an $O(n)$ cost preprocessing, will process any number of lowest common ancestor queries for a constant cost each.

Construction of the tree SH

The binary tree *SH* will be represented by the standard LEFTCHILD (*LC*) and RIGHTCHILD (*RC*) arrays.

Definition. The right shoulder of a binary tree is the (maximal) array $S=[S(0), S(1), \dots, S(k)]$ of nodes of the tree in which $S(0)$ is the root of the tree, and $S(i)$ is the right child of $S(i-1)$.

For an easier formal description, we will use an auxiliary node v_0 with value $f(0) = +\infty$, and *SH* will be the right subtree of v_0 . Starting with this v_0 , we insert the elements of $f(i)$ into the tree one-by-one, by keeping track of not only the tree structure (*LC*, *RC*) but also the right shoulder array *S*. Note that *S* forms a monotone stack ($S(k) \cong S(k-1) \cong \dots \cong S(0)$ and $f(S(k)) \leq \dots \leq f(S(0))$).

We compare the new value $f(i)$ with the stack values $f(S(j))$, $j=k, k-1, \dots$, until we find a $f(S(j)) \cong f(i)$. Then we remove the subtree *T* with roots $S(j+1)$ from $S(j)$, add a new node v_i (with value $f(i)$) to the tree as the new right child of $S(j)$, and attach the above *T* back as the left subtree of v_i . It is clear that both the symmetric order and the heap property are preserved.

The formal description uses the arrays

$$\begin{aligned} [f(i); 0 \leq i \leq n], & \quad f(0) = +\infty \\ [LC(i), RC(i); 0 \leq i \leq n], & \quad LC(0) = A \\ [S(i); 0 \leq i \leq n], & \quad S(0) = 0. \end{aligned}$$

Procedure SYMHEAP (*n*, *f*, *RC*, *LC*)

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S(0) = 0
RC(0) = A
k = 0
for i=1 to n do
  j ← k
  while f(i) > f(S(j)) do
    j ← j-1
  repeat
    k ← j+1
    LC(i) ← RC(S(j))
    RC(i) ← A
    RC(S(j)) ← i
    S(j+1) ← i
  end for
end SYMHEAP

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2. When T is a full branching tree

A rooted tree is a full branching tree, if all leaves are on the same level and non-leaves have at least 2 children. For a full binary tree, processing all path-queries has a cost $\Theta(n \log \log n)$, so here we have to restrict our attention to the set A of actual queries.

Let $A(y)$ be the set of actual queries which go through y :

$$A(y) = \{p(x, z) \in A \mid x \cong y \cong z\}$$

and $A^*(y)$ the set of restrictions of these paths to the interval $[\text{root}(T), y]$.

Starting with the root, we go down level by level and successively find the maxima over all paths in the sets $A^*(y)$.

Assume that we know these maxima down to the i -th level, and let us find the maxima on paths in $A^*(y)$ for a y on level $i+1$. If \bar{y} is the parent of y , then we know the maxima on the restrictions to $[\text{root}, \bar{y}]$ of all paths in $A(y)$. It remains to compare $f(y)$ with these maxima. Since the maxima are known as an *ordered* list (maximum on $p(x, \bar{y})$ is at least as large as maximum on $p(x', \bar{y})$ if $x > x'$), we can simultaneously compare $f(y)$ with them using binary insertion. Cost $\cong \lceil \log(A(y)+1) \rceil$.

Write L_i for the set of nodes on level i , and $l_i = |L_i|$. For a full branching tree, $\sum_{j < i} l_j < l_i$ and thus for the entropy

$$\sum_i (l_i/n) \log(n/l_i) < \sum_i i 2^{-i} = 2.$$

Since L_i is an antichain, the sets $A(y), y \in L_i$ are pairwise disjoint, thus Jensen's inequality leads to

$$\sum_{y \in L_i} \lceil \log(A(y)+1) \rceil < \sum_{y \in L_i} (1 + \log(A(y)+1)) \cong l_i + l_i \log \frac{|A| + l_i}{l_i}.$$

Thus, we get for the total cost:

$$\begin{aligned} \text{Cost} &< \sum_i \left(l_i + l_i \log \frac{|A| + l_i}{l_i} \right) \\ &\cong n + n \log \frac{|A| + n}{n} + \sum l_i \log \frac{n}{l_i} \\ &\cong 3n + n \log \frac{|A| + n}{n}. \end{aligned}$$

3. General T

Scalping a rooted tree

The scalp of a tree T is defined as

$$S(T) = \{v \in V(T) \mid \text{deg}(u) \cong 1 \text{ for all } u \cong v\}.$$

The subgraph of T spanned by $S(T)$ splits into vertex-disjoint paths (of length $\cong 0$) called hairs or fringes.

We set $T_0 = T$ and inductively define the trees $T_{i+1} = T_i - S(T_i)$, $i = 0, 1, \dots$ and write k for the index of the last non-empty T_i (hence $S(T_k) = T_k$). Note that every vertex $v \in T_i$ ($i \geq 1$) has a child in T_{i-1} .

For a given i , $1 \leq i \leq k$, and a vertex $v \in T_i$, we define the *root of v in T_i* as the element

$$R_i(v) = \min \{u \in T_i \mid u > v\}.$$

For $v \in S(T_{i-1})$, the "natural" root $R_i(v)$ will be denoted by $R^*(v)$. We will write R for the set of all roots:

$$R = \{u \mid R_i(v) = u \text{ for some } i, v\}.$$

Clearly, $R = \{u \mid \deg(u) \geq 2\}$. (Note that the root of the whole tree in the traditional sense need not be in R .)

The restriction of A (actual queries) to T_i is defined in the natural way (by restricting the paths in A to T_i), and is denoted by A_i . We also define $A^*(y) = A_i(y)$ for $y \in S(T_i)$, and $A^+(y)$ as the set of all queries in A , whose restriction to T_i ends at $y \in S(T_i)$. (I.e., for $y \in S(T_i)$, the restrictions to T_i of queries in $A^+(y)$ form the set $A^*(y)$.) (The rest of the section is not needed for the description of the algorithm, only for the cost analysis.) In other words, a query $p(x, z) \in A^+(y)$ (where $y \in S(T_i)$) iff $R_i(z) = y$. But then $R(v) = y$ for all $v, y \geq v \geq z$. Hence, if $y_1 \in S(T_i)$, $y_2 \in S(T_j)$, $i > j$, and a query $p(x, z)$ belongs to both $A^+(y_1)$ and $A^+(y_2)$, then $R_i(y_2) = y_1$.

Thus, we obtained the following lemma.

Lemma 1. *If Y is a set of vertices such that no element in Y is a root of another, then the sets $A^+(y)$, $y \in Y$, are pairwise disjoint. ■*

For an element $u \in S(T_i)$, the set $C(u) = \{v \mid R_i(v) = u\}$ is called the *court* of u . Now we are going to partition R by coloring its elements with k colors. The elements of $S(T_1) \cap R$ get color 1. Having colored the elements of

$$\bigcup_{1 \leq j < i} [S(T_j) \cap R]$$

we color the elements of $S(T_i) \cap R$ as follows: u gets the smallest color (smallest positive integer) that does not appear in its court $C(u)$. Clearly, in the obtained coloring an element in $S(T_i) \cap R$ gets a color not exceeding i . We will write R_i for the set of $u \in R$ with color i , R_0 for the set of leaves of T_0 , r_i for $|R_i|$, and r for $|R|$.

The following crucial lemma will be proved at the end of the paper.

Lemma 2. A) *For any fixed $i \geq 1$, the sets $A^+(y)$, $y \in R_i$, are disjoint.*

B) *For $i \geq 0$, we have the exponential decay*

$$\sum_{j>i} r_j < r_i$$

whence $r < n/2$, and

$$\sum_{j>i} r_j < r 2^{-i}, \quad i = 1, 2, \dots$$

consequently, for the entropy

$$\sum_{i \geq 1} (r_i/r) \log(r/r_i) < \sum_{i \geq 1} i 2^{-i} = 2.$$

In other words, the coloring defined above provides a finite-entropy decomposition of R into antichains in the partial order $u \gg v$ if $u = R_i(v)$ for some i .

The Algorithm

We start with the decomposition

$$V(T) = \sum_{i=0}^k S(T_i) \quad (\text{disjoint union}).$$

Since every scalp $S(T_i)$ represents a disjoint union of paths, $V(T)$ is decomposed into disjoint sets, each of which spans a (directed) path of T .

Perform SYMHEAP on each of these sets. Total cost is less than $2n$. Next, starting with A_k , we will inductively find the answers for all queries in A_i , $i=k, k-1, \dots, 0$ ($A_0=A$ is the original set of queries).

For $i=k$, the tree T_k is but a string, and SYMHEAP provided us with the maximum on all possible paths of T_k .

Having answered all queries in A_{i+1} , we proceed to get the answers for queries in $A_i - A_{i+1}$. For a given vertex $y \in S(T_i)$ there are $|A_i(y)|$ queries in A_i that end at y . For a particular query $p(x, y) \in A_i$ we already know an index m_{xy} such that

$$f(m_{xy}) = \max_{x \cong u \cong R^*(y)} f(u)$$

(for $x < R^*(y)$ we may interpret $f(m_{xy})=0$), and from SYMHEAP, we know an index m_y such that

$$f(m_y) = \max_{R^*(y) > u \cong y} f(u).$$

It remains to compare $f(m_y)$ with $f(m_{xy})$. For fixed y , we know the order of the values $f(m_{xy})$, since $x_1 > x_2$ implies $f(m_{x_1y}) \cong f(m_{x_2y})$. Thus, we can make the comparisons of $f(m_y)$ with the various values $f(m_{xy})$ simultaneously, by merging the value $f(m_y)$ into the ordered sequence $f(m_{x_1y}) \cong f(m_{x_2y}) \cong \dots$. Cost is at most $\lceil \log(|A_i(y)|+1) \rceil = \lceil \log(|A^*(y)|+1) \rceil$ for every $y \in S(T_i)$.

Cost Analysis

Theorem. *The obtained total cost*

$$C = 2n + \sum_y \lceil \log(|A^*(y)|+1) \rceil$$

is less than

$$5n + n \log \frac{|A|+n}{n}.$$

Proof. We will separately handle terms with $y \notin R$ and those with $y \in R$.

By Lemma 1, the sets $A^+(y)$, $y \notin R$, are pairwise disjoint, thus applying the Jensen inequality

$$\sum_{i=1}^t \log x_i \cong t \log \frac{\sum x_i}{t}$$

we get

$$\sum_{y \notin R} \lceil \log(|A^*(y)|+1) \rceil < \sum \lceil t + \log(|A^*(y)|+1) \rceil \cong m + m \log \frac{|A|+m}{m}$$

where $m=n-r$ is the number of non-roots.

It remains to estimate the sum

$$C^1 = \sum_{i=1}^k \sum_{y \in R_i} [\log(|A^*(y)| + 1)].$$

Here we may deal with different restrictions of the same path, so multiplicities may occur. By part *A* of Lemma 2, however, we can use the above estimation for the inner sums:

$$C^1 < \sum_{i=1}^k \left[r_i + r_i \log \frac{|A| + r_i}{r_i} \right].$$

The exponential decay of the sizes r_i does the rest (part *B* of Lemma 2):

$$C^1 < r + r \log \frac{|A| + r}{r} + \sum_{i \geq 1} r_i \log \frac{r}{r_i} < 3r + r \log \frac{|A| + r}{r}.$$

Now $m + r = n$, thus (using $r < n/2$ and the log-sum inequality

$$x \log \frac{a}{x} + y \log \frac{b}{y} \cong (x + y) \log \frac{a + b}{x + y}, \quad x, y > 0$$

we get for the total cost

$$C < 2n + m + 3r + m \log \frac{|A| + m}{m} + r \log \frac{|A| + r}{r} < 5n + n \log \frac{|A| + n}{n}$$

as stated. ■

Proof of Lemma 2. The sets R are clearly antichains of T in the partial order \gg , so part *A* follows from Lemma 1. To establish the exponential decay of $r_i = |R_i|$, we need two more lemmas:

Lemma 3. *Leaves of T_i ($i \geq 1$) have color i . Furthermore, for $i \geq 2$ and $u \in R_i$, the court $C(u)$ contains exactly the colors $\{1, \dots, i-1\}$.*

Lemma 4. *Every node $u \in \bigcup_{j \geq i} R_j$ must have at least two children in T_i .*

To get the exponential decay, it remains to apply the following simple fact: In any rooted tree, the number of vertices of degree ≥ 2 is less than the number of leaves. ■

Proof of Lemma 3. For $i = 1$ the statement is trivial. Assuming its validity for all values less than a certain i , let us prove it for i .

We know that a leaf u of T_i (actually any vertex of T_i) has a child in T_{i-1} , thus there is a leaf v of T_{i-1} such that $R_i(v) = u$. Since (by induction) the color of v is $i-1$, and $C(v)$ (a subset of $C(u)$) contains all colors $\{1, 2, \dots, i-2\}$, u must have color $\geq i$. But $u \in S(T_i)$, so its color cannot exceed i , thus it is i .

Furthermore, any vertex $u \in R_i$ has (by definition) a vertex $v \in R_{i-1}$ in its court $C(u)$. By induction, $C(v)$ contains all colors $\{1, \dots, i-2\}$, thus the relation $C(v) \subset C(u)$ implies the second statement of the lemma. ■

Proof of Lemma 4. If $u \in \bigcup_{j \geq i} R_j$, then u has a vertex $v \in R_i$ in its court $C(u)$. $R_i \subset T_i$ implies $v \in T_i$. Restricting ourselves to the tree T_i , we found a vertex v to which u is a root. Thus, the degree of u within T_i must be at least 2. ■

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