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# Drawing rooted trees in linear time



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#### **SUMMARY**

The aim of automatic graph drawing is the development of algorithms for creating nice and easily readable layouts of abstractly given graphs. For the special case of rooted trees of unbounded degree, John Q. Walker II presented a drawing algorithm in this journal in 1990. This algorithm is an extension of the Reingold–Tilford algorithm. It yields very good results and is therefore widely used. Furthermore, it is widely assumed to run in linear time, as the author claims in his article. However, the algorithm in its presented form clearly needs quadratic runtime. We explain the reasons for that and state a revised algorithm that creates the same layouts in linear time. Copyright © 2006 John Wiley & Sons, Ltd.

KEY WORDS: graph drawing; trees; Reingold–Tilford algorithm; Walker's algorithm

#### INTRODUCTION

Automatic graph drawing is concerned with the problem of creating two- or three-dimensional layouts of abstractly given graphs. In these layouts, nodes are usually represented by simple geometric objects, e.g., circles or rectangles, while edges are represented by lines or arbitrary curves connecting the two objects corresponding to the nodes adjacent to that edge. Drawing an abstract graph thus means to choose the positions of the nodes and to route the edges between them.

One might try to create such drawings manually. However, if the graphs to be drawn are large or if there are a lot of them, this approach is tedious and does not in general yield satisfactory results. For this reason, one tries to develop algorithms for creating nice and easily readable layouts automatically. In this approach, the first step is to determine and to formalize the criteria for measuring the quality of the drawing. Typical aims are a small number of crossings or bends of edges, if any, or the display of possible symmetric structure, or a small area consumed by the drawing, and many others.

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The second step is to develop algorithms for computing drawings that realize the specified criteria as far as possible. However, many of the arising optimization problems are NP-hard.

Because of this, one often considers special classes of graphs where these problems become tractable. For the case of rooted binary trees, Wetherell and Shannon [1] presented a linear time algorithm satisfying the following aesthetic requirements: the *y*-coordinate of a node corresponds to its level, so that the hierarchical structure of the tree is displayed; the left child of a node is placed to the left of the right child, i.e. the order of the children is displayed; finally, each parent node is centered over its children.

Nevertheless, this algorithm showed some considerable aesthetic deficiencies. Reingold and Tilford [2] designed a better algorithm by adding the following feature: each pair of isomorphic subtrees is drawn identically up to translation, i.e. the drawing does not depend on the position of a subtree within the complete tree. They also made the algorithm symmetrical: if all orders of children in a tree are reversed, the computed drawing is the reflected original one. The width of the drawing is not always minimized subject to these conditions, but it is close to the minimum in general. The Reingold–Tilford algorithm runs in linear time, too.

Extending this algorithm to rooted ordered trees of unbounded degree in a straightforward way produces layouts where some subtrees of the tree may get clustered on a small space, even if they could be distributed more evenly. This problem was solved in 1990 by the algorithm of Walker [3], which spaces out subtrees whenever possible. Unfortunately, the runtime of the algorithm presented in [3] is quadratic, in contrast to the author's assertion. In the present article, we give an adjustment of Walker's algorithm that does not affect the computed layouts but yields linear runtime. For this, we partly use ideas already applied by Reingold and Tilford and partly develop new techniques.

In the next section, we establish the basic notation about trees and state the aesthetic criteria guiding the algorithms to be dealt with. Then we explain the Reingold–Tilford algorithm, describe the idea of Walker's algorithm and point out the nonlinear parts of the latter. Next we show how to improve these parts in order to get a linear time algorithm. In the appendix, we state the complete revised algorithm and give experimental runtime results.

## **PRELIMINARIES**

We define a (rooted) tree as a directed acyclic graph with a single source, called the root of the tree, such that there is a unique directed path from the root to any other node. The level of a node is the length of this path. For each edge (v, w), we call v the parent of w and w a child of v. If  $w_1$  and  $w_2$  are two different children of v, we say that  $w_1$  and  $w_2$  are siblings. Each node w on the path from the root to a node v is called an ancestor of v, while v is called a descendant of v. A leaf of the tree is a sink of the graph, i.e. a node without children. If  $v_-$  and  $v_+$  are two nodes such that  $v_-$  is not an ancestor of  $v_+$  and vice versa, the greatest distinct ancestors of  $v_-$  and  $v_+$  are defined as the unique ancestors  $v_-$  and  $v_+$  of  $v_-$  and  $v_+$ , respectively, such that  $v_-$  and  $v_+$  are siblings. Each node v of a rooted tree v induces a unique subtree of v with root v.

In a *binary* tree, each node has at most two children. In an *ordered* tree, a certain order of the children of each node is fixed. The first (last) child according to this order is called the *leftmost* (*rightmost*) child. The *left* (*right*) *sibling* of a node v is its predecessor (successor) in the list of children of the parent of v. The *leftmost* (*rightmost*) *descendant* of v on level l is the leftmost (rightmost) node on level l belonging



to the subtree induced by v. Finally, if  $v_1$  is the left sibling of  $v_2$ ,  $w_1$  is the rightmost descendant of  $v_1$  on some level l, and  $w_2$  is the leftmost descendant of  $v_2$  on the same level l, we call  $w_1$  the *left neighbor* of  $w_2$  and  $w_2$  the *right neighbor* of  $w_1$ .

To draw a tree into the plane means to assign x- and y-coordinates to its nodes and to represent each edge (v, w) by a straight line connecting the points corresponding to v and w. When drawing a rooted tree, one usually requires the following aesthetic properties.

- (A1) The layout displays the hierarchical structure of the tree, i.e. the *y*-coordinate of a node is given by its level.
- (A2) The edges do not cross each other and nodes on the same level have a minimal horizontal distance.
- (A3) The drawing of a subtree does not depend on its position in the tree, i.e. isomorphic subtrees are drawn identically up to translation.

If the trees to be drawn are ordered, we additionally require the following.

- (A4) The order of the children of a node is displayed in the drawing.
- (A5) The algorithm works symmetrically, i.e. the drawing of the reflection of a tree is the reflected drawing of the original tree.

Here, the *reflection* of an ordered tree is the tree with reversed order of children for each parent node. Aesthetics (A1) and (A2) were introduced by Wetherell and Shannon, while (A3)–(A5) are due to Reingold and Tilford. Usually, one tries to find a layout satisfying (A1)–(A5) with a small width, i.e. with a small range of *x*-coordinates.

#### REINGOLD AND TILFORD'S ALGORITHM

For ordered binary trees, the first linear time algorithm satisfying (A1)–(A5) was presented by Reingold and Tilford [2]. This algorithm is easy to describe informally; it draws the tree recursively in a bottom-up sweep. Leaves are placed to an arbitrary x-coordinate and to the y-coordinate given by their level. After drawing the subtrees induced by the children of a parent node independently, the right one is shifted so that it is placed as close to the right of the left subtree as possible  $^{\ddagger}$ . Next, the parent is placed centrally above the children, that is, at the x-coordinate given by the average x-coordinate of the children and at the y-coordinate given by its level. Finally, the edges are inserted.

The Reingold–Tilford algorithm obviously satisfies (A1)–(A5). The difficult task is how to perform the steps described above in linear time. The crucial part of the algorithm is the shifting of the second subtree; solving the following problems takes a quadratic runtime in total, if a straightforward algorithm is used: first, computation of the new position of this subtree; second, the shifting of the subtree itself.

<sup>&</sup>lt;sup>‡</sup>For simplicity, we assume throughout this paper that all nodes have the same dimensions and that the minimal distance required between neighbors is the same for each pair of neighbors. Both restrictions can be relaxed easily, since we will always compare a single pair of neighbors.



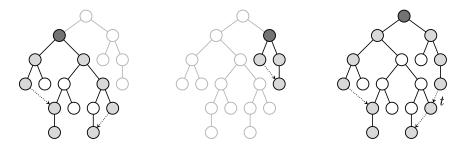


Figure 1. Combining two subtrees and adding a new thread t.

For the first problem, define the *left (right) contour* of a tree as the sequence of leftmost (rightmost) nodes in each level, traversed from the root to the highest level. For an illustration, see Figure 1, where nodes belonging to the contours are shaded. To place the right subtree as close to the left one as possible, we have to compare the positions of the right contour of the left subtree with the positions of the left contour of the right subtree, for all levels occurring in both subtrees. Since each node is traversed as part of the left contour in at most one subtree combination phase, the total number of such comparisons is linear for the complete tree. The runtime problem is how to traverse the contours without traversing (too many) nodes not belonging to the contours. To solve this problem, Reingold and Tilford introduce *threads*. For each leaf of the tree that has a successor in the same contour, the thread is a pointer to this successor. See Figure 1 again, where the threads are represented by dotted arrows. For every node of the contour, we now have a pointer to its successor in the contour: either it is the leftmost (rightmost) child, or it is given by the thread. Finally, to keep the threads up to date, one has to add a new thread whenever two subtrees of different height are combined.

For the second problem, the straightforward algorithm would shift all nodes of the right subtree by the same value. Since this needs quadratic time in total, Reingold and Tilford attach a new value mod(v) to each node v, which is called its modifier (this technique was presented by Wetherell and Shannon [1]). The position of each node is preliminary in the bottom-up traversal of the tree. When moving a subtree rooted at v, only mod(v) and a preliminary x-coordinate prelim(v) are adjusted by the amount of shifting. The modifier of a node v is interpreted as a value to be added to all preliminary x-coordinates in the subtree rooted at v, except for v itself. Thus, the real position of a node is its preliminary position plus the aggregated modifier modsum(v) given by the sum of all modifiers on the path from the parent of v to the root. To compute all real positions in linear time, the tree is traversed in a top-down fashion at the end.

When comparing contour nodes to compute the new position of the right subtree, we need the real positions of these nodes, too. For runtime reasons, we may not sum up modifiers on the paths to the root. Therefore, modifiers are used in leaves as well. A modifier of a leaf v with a thread to a node w stores the difference between modsum(w) and modsum(v). Since new threads are added after combining two subtrees, these modifier sums can be computed while traversing the contours of the two subtrees. We have to traverse not only the inside contours but also the outside contours for computing



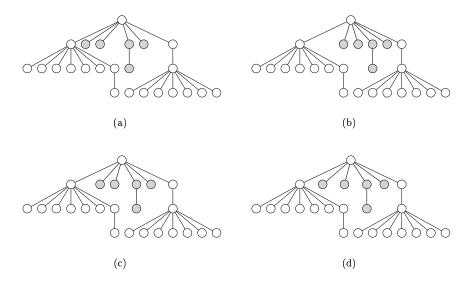


Figure 2. Extending the Reingold–Tilford algorithm to trees of unbounded degree.

the modifier sums, since v is a node of the outside contour. Now the aggregated modifiers can be computed as the sums of modifiers along the contours instead of the paths to the root.

#### WALKER'S ALGORITHM

For drawing trees of unbounded degree, the Reingold–Tilford algorithm could be adjusted easily by traversing the children from left to right, placing and shifting the corresponding subtrees one after another. However, this violates property (A5): the subtrees are placed as close to each other as possible and small subtrees between larger ones are piled to the left; see Figure 2(a). Reingold and Tilford proposed to avoid this effect by adding an analogous second traversal from right to left, see Figure 2(b), and to take average positions after that. This algorithm satisfies (A1)–(A5), but smaller subtrees are usually then clustered; see Figure 2(c). Because of this shortcoming, a new aesthetic requirement is added for non-binary trees:

(A6) The children of a node should be equally spaced.

To obtain a layout where smaller subtrees are spaced out evenly, as for example in Figure 2(d), Walker [3] proposed the following procedure (see Figure 3): the subtrees of the current root are processed one after another from left to right. First, each child of the current root is placed as close to the right of its left sibling as possible. As in Reingold and Tilford's algorithm, the left contour of the current subtree is then traversed top-down in order to compare the positions of its nodes to those of their left neighbors. Whenever two conflicting neighbors  $v_-$  and  $v_+$  are detected, forcing  $v_+$  to be



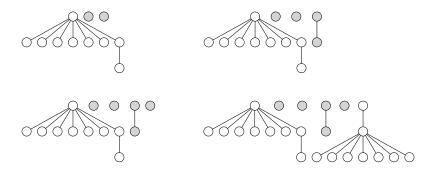


Figure 3. Spacing out the smaller subtrees.

shifted to the right by an amount of *shift*, we apply an appropriate shift to all subtrees between those containing  $v_-$  and  $v_+$ . More precisely, let  $w_-$  and  $w_+$  be the greatest distinct ancestors of  $v_-$  and  $v_+$ . Notice that both  $w_-$  and  $w_+$  are children of the current root. Let *subtrees* be the number of children of the current root between  $w_-$  and  $w_+$  plus 1. Spacing out the subtrees is shifting the subtree rooted at the *i*th child to the right of  $w_-$  by an amount of *i*-shift/subtrees, for  $i=1,\ldots,subtrees$ . Observe that subtrees may be shifted several times by this algorithm, even while adding a single subtree. It is easy to see that this algorithm satisfies (A5).

Unfortunately, many parts of the algorithm presented in [3] do not run in linear time. Some of them are easy to improve, for example by using Reingold and Tilford's ideas, and some require new ideas. All problems concern Walker's procedure APPORTION [3, pp. 695–697]. In the following, we list the critical parts. In the next section, we will explain how to change these in order to obtain linear runtime.

## **Traversing the contours**

A recursive function GETLEFTMOST is used to find the leftmost descendant of a given node v on a given level l. If the level of v is l, the algorithm returns v. Otherwise, GETLEFTMOST is applied recursively to all children of v, from left to right. The aggregated runtime of GETLEFTMOST is not linear in general. To prove this, we present a series of trees  $T_k$  such that the number of nodes in  $T_k$  is  $n \in \Theta(k^2)$ , but the total number of GETLEFTMOST calls is  $\Theta(k^3)$ . Since  $k \in \Theta(n^{1/2})$ , this shows that the total runtime of GETLEFTMOST is  $\Omega(k^3) = \Omega(n^{3/2})$ . The tree  $T_k$  is defined as follows (see Figure 4(a) for k=3). Beginning at the root, there is a chain of 2k nodes, each of the last 2k-1 being the right or only child of its predecessor. For  $i=1,\ldots,k$ , the ith node in this chain has another child to the left; this child is the first node of a chain of 2(k-i)+1 nodes. The number of nodes in  $T_k$  is

$$2k + \sum_{i=1}^{k} (2(k-i) + 1) = 2k + k(k-1) + k \in \Theta(k^2)$$

Now, for each i = 0, ..., k - 1, we have to combine two subtrees when visiting the node on the right contour of  $T_k$  on level i. In this combination, the highest common level of the subtrees is 2k - i - 1,



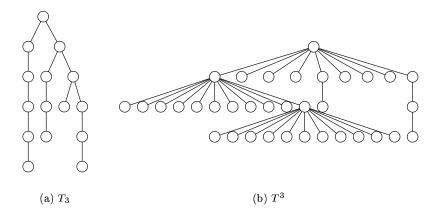


Figure 4. Examples proving the nonlinear runtime of Walker's algorithm.

and by construction of  $T_k$  we always have to apply GETLEFTMOST to every node of the right subtree up to this level. The number of these nodes is

$$k - i + \sum_{j=0}^{k-i-1} (2j) = (k-i) + (k-i)(k-i-1) = (k-i)^2$$

hence the total number of GETLEFTMOST calls for all combinations is

$$\sum_{i=0}^{k-1} (k-i)^2 = \sum_{i=1}^{k} i^2 = k(k+1)(2k+1)/6 \in \Theta(k^3)$$

#### Finding the ancestors and summing up modifiers

This part of the algorithm is obviously quadratic. When adjusting the current subtree to the left subforest, the greatest distinct ancestors of the possibly conflicting neighbors are computed for each level by traversing the graph up to the current root, at the same time computing the modifier sums. Since the distance of the levels grows linearly, the total number of steps is in  $\Omega(n^2)$ .

# Counting and shifting the smaller subtrees

When shifting the current subtree to the right because of a conflict with a subtree to the left, the procedure APPORTION also shifts all smaller subtrees in between immediately. Furthermore, the number of these subtrees is computed by counting them one by one. Both actions have an aggregated runtime of  $\Omega(n^{3/2})$ , as the following example shows. Let the tree  $T^k$  be constructed as follows (see Figure 4(b) for k = 3). Add k children to the root. The ith child, counted  $i = 1, \ldots, k$  from



left to right, is the root of a chain of i nodes. Between each pair of these children, add k children as leaves. The leftmost child of the root has 2k + 5 children, and up to level k - 1, every rightmost child of the 2k + 5 children has again 2k + 5 children. The number of nodes of  $T^k$  is

$$1 + \sum_{i=1}^{k} i + (k-1)k + (k-1)(2k+5) \in \Theta(k^2)$$

Furthermore, by construction of the left subtree, adding the *i*th subtree chain for i = 2, ..., k results in a conflict with the left subtree on level *i*. Hence all (i - 1)(k + 1) - 1 smaller subtrees between the two conflicting ones are counted and shifted. Thus, the total number of counting and shifting steps is

$$\sum_{i=2}^{k} ((i-1)(k+1) - 1) = (k+1)k(k-1)/2 - k + 1 \in \Theta(k^3)$$

As in the last example, we derive that counting and shifting needs  $\Omega(n^{3/2})$  time in total.

#### IMPROVING WALKER'S ALGORITHM TO RUN IN LINEAR TIME

In this section, we explain how to improve the algorithm of Walker to run in linear time without affecting the computed layouts. For a closer look, see Appendix A.

## Traversing the contours and summing up modifiers

This can be done exactly as in the case of binary trees by using threads; see the section on the Reingold–Tilford algorithm. The fact that the left subforest is no tree in general does not create any additional difficulty.

## Finding the ancestors

The problem of finding the greatest distinct ancestors  $w_-$  and  $w_+$  of two nodes  $v_-$  and  $v_+$  can be solved by the algorithm of Schieber and Vishkin [4]. For each pair of nodes, this algorithm can determine the greatest distinct ancestors in constant time, after an O(n) preprocessing step. However, in our application, a much simpler algorithm can be applied. First observe that we know the right ancestor  $w_+$  anyway, it is just the root of the current subtree. Furthermore, as  $v_+$  is always the right neighbor of  $v_-$  in our algorithm, the left one of the greatest distinct ancestors only depends on  $v_-$ . Thus we may shortly call it the ancestor of  $v_-$  in the following. We use a node pointer ancestor(x) for each node x to save its ancestor and initialize it to x itself. Observe that this value is not correct for rightmost children, but we do not need the correct value  $w_-$  of ancestor( $v_-$ ) until the right neighbor  $v_+$  of  $v_-$  is added, i.e. until the current root is the parent node of  $w_-$ . Hence assume that we are placing the subtrees rooted at the children of v from left to right. Since tracing all ancestor(x) consumes too much time, we use another node pointer defaultAncestor. Our aim is to have the following property (\*) for all nodes  $v_-$  on the right contour of the left subforest after each subtree addition: if ancestor( $v_-$ ) is up to date, i.e. is a child of v, then it points to the correct ancestor  $w_-$  of  $v_-$ ; otherwise, the correct ancestor is defaultAncestor. We start with placing the first subtree, rooted at w, which does not require any



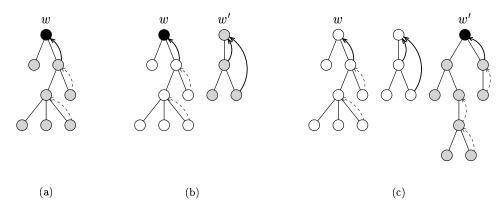


Figure 5. Adjusting ancestor pointers when adding new subtrees: the pointer ancestor(x) is represented by a solid arrow if it is up to date and by a dashed arrow if it is expired. In the latter case, the defaultAncestor is used and drawn black. When adding a small subtree, all ancestor pointers ancestor(x) of its right contour are updated. When adding a large one, only defaultAncestor is updated.

ancestor calculations. After that, we set *defaultAncestor* to w. Since all pointers ancestor(x) of the left subtree either point to w or to a node of a higher level, the desired property (\*) holds, see Figure 5(a). After placing the subtree rooted at another child w' of v, we distinguish two cases: if the subtree rooted at w' is smaller than the left subforest, we can actualize ancestor(x) for all nodes x on its right contour by setting it to w'. By this, we obviously keep (\*) (see Figure 5(b)). Otherwise, if the new subtree is larger than the left subforest, we may not do the same because of runtime. But now it suffices to set defaultAncestor to w', since again all pointers ancestor(x) of the subtree induced by w' either point to w' or to a node of a higher level, and all other subtrees in the left subforest are hidden. Hence we have (\*) again (see Figure 5(c)).

# **Counting the smaller subtrees**

For this, we just have to number the children of each node consecutively; then the number of smaller subtrees between the two greatest distinct ancestors  $w_-$  and  $w_+$  is the number of  $w_+$  minus the number of  $w_-$  minus 1. Hence it can be computed in constant time (after a linear time preprocessing step to compute all child numbers).

# Shifting the smaller subtrees

In order to get a linear runtime, we will shift each subtree once at most when it is not the currently added subtree. The currently added subtree, however, may be shifted whenever it conflicts with a subtree to the left, using the fact that shifting a single subtree is done in constant time (recall that we only have to adjust  $prelim(w_+)$  and  $mod(w_+)$ ). Furthermore, shifting the current subtree immediately is necessary to keep the right contour of the left subforest up to date. All shiftings of non-current



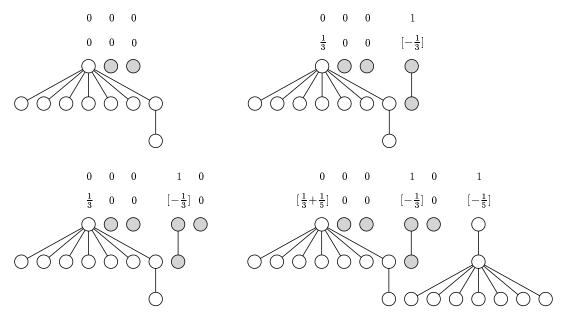


Figure 6. Aggregating the shifts: the top number at node x indicates the value of shift(x), and the bottom number indicates the value of change(x).

subtrees are performed in a single traversal after all subtrees of the current root have been placed. To store the shiftings at the moment they arise, we use real numbers shift(x) and change(x) for each node x and set both to zero at the beginning. Assume that the subtree rooted at  $w_+$  is the subtree currently placed, and that a conflict with the subtree rooted at  $w_{-}$  forces the current subtree to move to the right by an amount of *shift*. Let *subtrees* be the number of subtrees between  $w_-$  and  $w_+$ , plus 1. According to Walker's idea, the ith of these subtrees has to be moved by  $i \cdot shift/subtrees$ . We save this by increasing  $shift(w_+)$  by shift, decreasing  $change(w_+)$  by shift/subtrees, and increasing  $change(w_-)$ by shift/subtrees. The interpretation of this is the following: to the left of node  $w_+$ , the nodes are shifted by an amount initialized to shift, but this amount starts decreasing by shift/subtrees per subtree at node  $w_+$  and ends decreasing at  $w_-$ , where it is zero. The trick is to aggregate the shifts: since the decrease in the amount of shifting is linear, we can add all these decreases in one array; see Figure 6 for an example. Finally, we execute all shifts in a single traversal of the children of the current root as follows (see Figure 7): we use two real values shift and change to store the shifts and the decreases of shift per subtree, respectively, and set both to zero at the beginning. Then we traverse the children from right to left. When visiting child v, we move v to the right by shift (i.e. we increase prelim(v) and mod(v) by shift), increase change by change(v), and increase shift by shift(v) and by change. Then we go on to the left sibling of v. It is easy to see that this algorithm shifts each subtree by the correct amount.



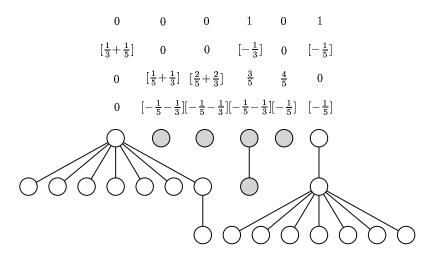


Figure 7. Executing the shifts: the new numbers at node x indicate the values of *shift* and *change* before shifting x, respectively.

#### APPENDIX A. THE COMPLETE REVISED ALGORITHM

In this appendix, we list the complete revised algorithm described in the preceding sections. The algorithm TREELAYOUT first initializes modifiers, threads, and ancestors; then it starts the bottom-up and top-down traversals of the tree (called FIRSTWALK and SECONDWALK following Walker).

```
TREELAYOUT(T)

for all nodes v of T

let mod(v) = thread(v) = 0

let ancestor(v) = v

let r be the root of T

FIRSTWALK(r)

SECONDWALK(r, -prelim(r))
```

Calling FIRSTWALK(v) computes a preliminary x-coordinate for v. Before that, FIRSTWALK is applied recursively to the children of v, as well as the function APPORTION. After spacing out the children by calling EXECUTESHIFTS, the node v is placed to the midpoint of its outermost children.

```
FIRSTWALK(v)

if v is a leaf

let prelim(v) = 0

if v has a left sibling w

let prelim(v) = prelim(w) + distance
```



```
else
let defaultAncestor be the leftmost child of v
for all children w of v from left to right
FIRSTWALK(w)
APPORTION(w, defaultAncestor)
EXECUTESHIFTS(v)
let midpoint = \frac{1}{2}(prelim(\text{leftmost child of }v) + prelim(\text{rightmost child of }v))
if v has a left sibling w
let prelim(v) = prelim(w) + distance
let mod(v) = prelim(v) - midpoint
else
let prelim(v) = midpoint
```

The procedure APPORTION (again following Walker's notation) is the core of the algorithm. Here a new subtree is combined with the previous subtrees. As in the Reingold–Tilford algorithm, threads are used to traverse the inside and outside contours of the left and right subtree up to the highest common level. The vertices used for the traversals are  $v_+^i$ ,  $v_-^i$ ,  $v_-^o$ , and  $v_+^o$ , where the superscript o means outside and i means inside, the subscript — means left subtree and + means right subtree. For summing up the modifiers along the contour, we use respective variables  $s_+^i$ ,  $s_-^i$ ,  $s_-^o$ , and  $s_+^o$ . Whenever two nodes of the inside contours conflict, we compute the left one of the greatest distinct ancestors using the function ANCESTOR and call MOVESUBTREE to shift the subtree and prepare the shifts of smaller subtrees. Finally, we add a new thread (if necessary). Observe that we have to adjust  $ancestor(v_+^o)$  or defaultAncestor to keep property (\*). For function NEXTLEFT and NEXTRIGHT see below.

```
APPORTION(v, defaultAncestor)
```

```
if v has a left sibling w let v_+^i = v_+^o = v let v_-^i = w let v_-^o be the leftmost sibling of v_+^i let s_+^i = mod(v_+^i) let s_+^o = mod(v_+^o) let s_-^o = mod(v_-^o) let s_-^o = mod(v_-^o) while NEXTRIGHT(v_-^i) \neq 0 and NEXTLEFT(v_+^i) \neq 0 let v_-^i = \text{NEXTRIGHT}(v_-^i) let v_-^i = \text{NEXTLEFT}(v_+^i) let v_-^o = \text{NEXTLEFT}(v_-^o) let v_+^o = \text{NEXTLEFT}(v_-^o) let v_+^o = \text{NEXTLEFT}(v_-^o) let v_+^o = \text{NEXTRIGHT}(v_+^o) let v_-^o = \text{NEXTRIGHT}(v_-^o) let v_-^o = \text{NEXRIGHT}(v_-^o) let v_-^o =
```



```
\begin{aligned} &\text{MOVESUBTREE}(\text{ANCESTOR}(v_-^i, v, defaultAncestor}), v, shift) \\ &\text{let } s_+^i = s_+^i + shift \\ &\text{let } s_-^o = s_-^o + shift \\ &\text{let } s_-^i = s_-^i + mod(v_-^i) \\ &\text{let } s_+^i = s_+^i + mod(v_-^i) \\ &\text{let } s_+^o = s_-^o + mod(v_-^o) \\ &\text{let } s_+^o = s_+^o + mod(v_+^o) \end{aligned} if Nextright(v_-^i) \neq 0 and Nextright(v_+^o) = 0 let thread(v_+^o) = \text{Nextright}(v_-^i) let mod(v_+^o) = mod(v_+^o) + s_-^i - s_+^o if Nextleft(v_+^i) \neq 0 and Nextleft(v_-^o) = 0 let thread(v_-^o) = \text{Nextleft}(v_+^i) let mod(v_-^o) = mod(v_-^o) + s_+^i - s_-^o let defaultAncestor = v
```

The function NEXTLEFT(v) is used to traverse the left contour of a subtree (or subforest). It returns the successor of v on this contour. This successor is either given by the leftmost child of v or by the thread of v. The function returns 0 if and only if v is on the highest level of its subtree.

```
NEXTLEFT(v)

if v has a child

return the leftmost child of v

else

return thread(v)
```

The function NEXTRIGHT(v) works analogously.

```
NEXTRIGHT(v)

if v has a child

return the rightmost child of v

else

return thread(v)
```

Shifting a subtree can be done in linear time if performed as explained above. Calling MOVESUBTREE( $w_-, w_+, shift$ ) first shifts the current subtree, rooted at  $w_+$ . This is done by increasing  $prelim(w_+)$  and  $mod(w_+)$  by shift. All other shifts, applied to the smaller subtrees between  $w_-$  and  $w_+$ , are performed later by EXECUTESHIFTS. To prepare this, we have to adjust  $change(w_+)$ ,  $shift(w_+)$ , and  $change(w_-)$ .

```
MOVESUBTREE(w_-, w_+, shift)

let subtrees = number(w_+) - number(w_-)

let change(w_+) = change(w_+) - shift/subtrees

let shift(w_+) = shift(w_+) + shift
```



```
let change(w_{-}) = change(w_{-}) + shift/subtrees
let prelim(w_{+}) = prelim(w_{+}) + shift
let mod(w_{+}) = mod(w_{+}) + shift
```

Now EXECUTESHIFTS(v) only needs one traversal of the children of v to execute all shifts computed and memorized in MOVESUBTREE.

```
EXECUTESHIFTS(v)

let shift=0
let change=0
for all children w of v from right to left
let prelim(w) = prelim(w) + shift
let mod(w) = mod(w) + shift
let change = change + change(w)
let shift = shift + shift(w) + change
```

The function ANCESTOR returns the left one of the greatest distinct ancestors of  $v_{-}^{i}$  and its right neighbor.

```
ANCESTOR(v_-^i,v,defaultAncestor)

if ancestor(v_-^i) is a sibling of v
return ancestor(v_-^i)
else
return defaultAncestor
```

Finally, the SECONDWALK is used to compute all real x-coordinates by summing up the modifiers recursively.

```
SECONDWALK(v, m)

let x(v) = prelim(v) + m

let y(v) be the level of v

for all children w of v

SECONDWALK(w, m + mod(v))
```

### APPENDIX B. EXPERIMENTAL RUNTIME

We compared the runtime of Walker's original algorithm as presented in [3] with the runtime of an implementation of the revised algorithm presented in this paper. The experiments were carried out on a Sun UltraSPARC III 900 MHz, using randomly created trees. The results are displayed in Figure B1. These results underline the nonlinearity of Walker's algorithm also from a practical point of view.



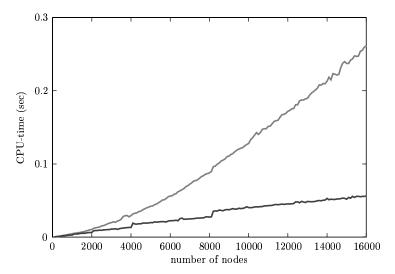


Figure B1. CPU-time needed by the original version of Walker's algorithm (light grey) and by the revised algorithm (dark grey).

## REFERENCES

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