# LOCATING THE VERTICES OF A STEINER TREE IN AN ARBITRARY METRIC SPACE 

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#### Abstract

Given a tree each of whose terminal vertices is associated with a given point in a compact metric space, the problem is to optimally associate a point in this space to each nonterminal vertex of the tree. The optimality criterion is the minimization of the sum of the lengths, in the metric space, over all edges of the tree. This note shows how a dynamic programming solution to this problem generalizes a number of previously published algorithms in diverse metric spaces, each of which has direct and significant applications to biological systematics or evolutionary theory.


Given $T$ a tree with vertex set $V(T)=\left\{X_{1}, \ldots, X_{m}, Y_{1}, \ldots, Y_{n}\right\}$ and edge set $E(T)$. Let ( $S, d$ ) be a compact metric space where to each vertex $X_{i} \in V(T)$ is associated a given point $x_{i} \in S$. The problem is to associate to each $Y_{j} \in V(T)$ some point $y_{j} \in S$ so as to minimize the edge-length of $T: \Sigma_{W Z \in E(T)} d(w, z)$, where $w, z \in S$ are associated with vertices $W, Z \in V(T)$. This arises in connection with the larger and generally intractable Steiner problem, in which the only information given is the position of $x_{1}, \ldots, x_{m} \in S$ and where both the structure of $T$ and the positions of the $y_{1}, \ldots, y_{n} \in S$ must be determined. This type of problem arises frequently in numerical taxonomy where the $X_{i}$ represent different organisms, the $x_{i}$ their positions in some space ( $S, d$ ) of characters or features, and the $Y_{j}$ represent hypothetical ancestral organisms. The minimality criterion corresponds to the economy or likelihood of the evolutionary explanation represented by the tree $T$.

Here we present a dynamic programming solution for the more restricted problem of finding the $y_{i}$ 's, given $T$. Though a solution to a full Steiner problem can always be considered to involve a tree with vertices of degree one or three only (allowing for edges of length zero), our method also applies in contexts where the given tree has vertices of
higher degree. Our formulation of the problem and its solution subsumes and generalizes a number of previously published versions, in particular metric spaces, and these will furnish some of our examples.

It suffices to consider the case where the $X_{i}$ are all terminal vertices (i.e. of degree one) of $T$, and the $Y_{j}$ are all non-terminal (degree $\geqslant 3$ ). Other cases are easily reducible to this one, e.g. by decomposing $T$ into the subtrees coincident with any non-terminal $X_{i}$ and optimizing each one separately, and by constraining $y_{j}$ for any $Y_{j}$ of degree $<3$ to take on the same value as one of its neighbours.

Choose any $Y_{r}, r=1, \ldots, n$ to be the root of the tree. Then for any vertex $Z$ of $T$, all vertices on the unique path between $Z$ and $Y_{r}$, including $Z$ and $Y_{r}$, are said to dominate $Z$. The vertices dominated by any vertex $Z$ determine the subtree $T_{Z}$ dominated by $Z$. Those vertices $Z_{1}, \ldots, Z_{p(Y)}$ which are both dominated by $Y$ and share an edge with $Y$ are said to be immediately dominated by $Y$.

We then construct a number of functions $f_{Z}$ on $S$, one for each vertex $Z \in V(T)$. For $i=1, \ldots, m$ we set $f_{X_{i}}\left(x_{i}\right)=0$ and $f_{X_{i}}(x) \equiv \infty$ if $x \neq x_{i}$. The $f_{Y}$ are then defined so that $f_{Y}(x)$ is the minimal edge-length of $T_{Y}$ given that $Y$ is associated with $x$. From the principle of optimality it follows that

$$
f_{Y}(x)=\min _{\left(z_{1}, \ldots, z_{p(Y)}\right)} \sum_{i=1}^{p(Y)}\left[f_{Z_{i}}\left(z_{i}\right)+d\left(z_{i}, x\right)\right]
$$

Then $\min _{x \in S} f_{Y_{r}}(x)=f_{Y_{r}}\left(y_{r}^{*}\right)$ is the minimal edge-length of $T$ and the usual backtrack process starting from $Y_{r}$ determines optimum locations $y_{i}^{*}$ of all $Y_{i}$.

It might seem that in some metric spaces, this dynamic programming approach would be of little use, due to the difficulty or impossibility of computing the $f_{Y}$ explicitly. However, in a diverse selection of spaces, such as those described below, it can be shown that the $f_{Y}$ must be calculated only over small subsets of $S$, leading to a feasible and rapid algorithm.

The Manhattan metric space

$$
S=\mathbf{R}^{N}, \quad d(u, v)=\sum_{J=1}^{N}|u(J)-v(J)| .
$$

In this space, the coordinates $y_{1}(J), \ldots, y_{n}(J)$ may be found sepa-
rately for each $J=1, \ldots, N$, so that it suffices to consider $S=\mathbf{R}^{1}$. Suppose $Y_{i}$ dominates immediately $X_{1}, \ldots, X_{p}$ where $x_{1}<\ldots<x_{p}$. If $p$ is even, set $q=\frac{1}{2} p-1$, and $I\left(Y_{i}\right)=\left[x_{\frac{1}{2} p}, x_{\frac{1}{2} p+1}\right]$. If $p$ is odd set $q=\frac{1}{2}(p-1)-1$, and $I\left(Y_{i}\right)=\left\{x_{\frac{1}{2}(p+1)}\right\}$. Then for $y \in I\left(Y_{i}\right)$

$$
f_{Y_{i}}(y)=\sum_{j=0}^{q}\left(x_{p-j}-x_{j+1}\right)
$$

and it will not be necessary to calculate $f_{Y_{i}}$ elsewhere. Now suppose $Y_{i}$ immediately dominates $Z_{1}, \ldots, Z_{p}$. Let $r_{1} \leqslant \ldots \leqslant r_{p}$ be the right-hand end-points of $I\left(Z_{u(1)}\right), \ldots, I\left(Z_{u(p)}\right)$, and $t_{1} \leqslant \ldots \leqslant t_{p}$ the left-hand endpoints of $I\left(Z_{v(1)}\right), \ldots, I\left(Z_{v(p)}\right)$, where $u$ and $v$ are suitable permutations of $(1, \ldots, p)$. If $I\left(Z_{k}\right)$ is a single point, it is listed both as a $t$ and an $r$. Suppose $r_{1}<t_{p}, r_{2}<t_{p-1}, \ldots, r_{1+q}<t_{p-q}$, but the remaining $r_{j}$ are all greater than or equal to the remaining $t_{k}$. Define

$$
I\left(Y_{i}\right)=\left[t_{p-q-1}, r_{q+2}\right]
$$

and for $y \in I\left(Y_{i}\right)$

$$
\begin{aligned}
f_{Y_{i}}(y)= & \sum_{j=0}^{q}\left[\left(t_{p-j}-r_{j+1}\right)+f_{Z_{u(j+1)}}\left(r_{j+1}\right)+f_{Z_{v(p-j)}}\left(t_{p-j}\right)\right] \\
& +\sum_{j=q+1}^{p-q-2} f_{Z_{u(j+1)}}(y) .
\end{aligned}
$$

The $y_{i}^{*}$ can then be chosen, since they will all be in the $I\left(Y_{i}\right)$, $i=1, \ldots, n$. A version of this algorithm, specific to the case where all non-terminal vertices have degree 3, was published by Farris [2], and is routinely applied in the study of evolution using continuous characters. Other results on Steiner trees in Manhattan space of dimension two are given in [5,7].

The space of qualitative characters

$$
S=\{1, \ldots, N\}, \quad d(I, J)=1 \text { if } I \neq J
$$

The case of 3-valent non-terminal nodes was investigated by Fitch [3], and the more general case by Hartigan [6]. Suppose $Y_{i}$ dominates $X_{1}, \ldots, X_{p}$. If $J \in S$ occurs as frequently or more frequently than any other $K \in S$ among $x_{1}, \ldots, x_{p}$, say $\alpha$ times,

$$
f_{Y_{i}}(J)=p-\alpha
$$

If $K \in S$ occurs $\alpha-1$ times,

$$
f_{Y_{i}}(K)=p-\alpha+1
$$

and it will not be necessary to calculate the remaining values of $f_{Y_{i}}$. We say $J$ is a best value and $K$ a next best value.

Consider a $Y_{i}$ which dominates just $Z_{1}, \ldots, Z_{p}$. Suppose $J$ is a best value most frequently among the $Z_{1}, \ldots, Z_{p}$, say $\alpha$ times. Then

$$
f_{Y_{i}}(J)=p-\alpha
$$

and if $K$ is a best value only $\alpha-1$ times, then

$$
f_{Y_{i}}(K)=p-\alpha+1 .
$$

Then for each $Y_{i}$ it will be possible to choose $y_{i}^{*}$ from among the best or next best values.

This algorithm is widely used in studying protein and DNA evolution and has been somewhat generalized in [11].

A space of finite sequences
Let $S$ be the set of $(N-1)$-ary sequences, and for $\boldsymbol{a}=(a(1), \ldots, a(r)$ ), $\boldsymbol{b}=(b(1), \ldots, b(s))$, where $r \leqslant s$

$$
d(\boldsymbol{a}, \boldsymbol{b})=r+s-\max _{\substack{0 \\ 0 \leqslant \lambda \leqslant i_{1}<\ldots<\lambda_{\lambda} \leqslant r \\ 1 \leqslant j_{1}<\ldots<j_{\lambda} \leqslant s}} \sum_{k=1}^{\lambda}\left[1+\delta\left(a\left(i_{k}\right), b\left(j_{k}\right)\right)\right],
$$

where $\delta(u, v)=1$ if $u=v$, and $\delta(u, v)=0$ otherwise. This metric, representing the mutational distance between the two sequences, arises in the study of molecular evolution as discussed by Ulam [16] and Sellers [15]. See [17] for another interpretation of this metric. In [12] we show that the search for $y_{1}^{*}, \ldots, y_{n}^{*}$ can be reduced to a large number (about $(2 r)^{m}$, where $r$ is the length of the longest sequence among the $\left.\boldsymbol{x}_{i}=\left(x_{i}(1), \ldots, x_{i}\left(r_{i}\right)\right), i=1, \ldots, m\right)$ of applications of the algorithm in the simple metric space ( $S=\{1, \ldots, N\}, d(I, J)=1-\delta(I, J)$ ) described in the preceding section. An application of these methods to infer the structure of RNA in ancestral organisms is presented in [13,14].

Euclidean space

$$
S=\mathbf{R}^{N}, \quad d(u, v)=\left(\sum_{j=1}^{N}(u(J)-v(J))^{2}\right)^{1 / 2}
$$

It is, of course, in this metric space, especially the case $N=2$, that the Steiner problem is classically posed. Making use of the fact that in a Steiner tree the $Y_{i}$ must all be of degree 3 and the $y_{i}$ must all be vertices of three angles of $\frac{2}{3} \pi$ radians, Melzak [9,10] and Gilbert and Pollak [4] have given an inductive algorithm which produces at most $2^{n}$ possible configurations for the set of $n$ non-terminal vertices of a given tree topology.

For arbitrary $T$, however, where the $Y_{i}$ may have degree $>3$, no noniterative algorithm is known. For $n=1$, the problem is a version of the Weber problem, or Fermat's problem and Kuhn [8] has shown that a well-known gradient-based iteration converges to the unique solution for almost all starting approximations. This algorithm can be extended to apply to our problem when $n>1$, though proofs of convergence would seem to be more difficult.

The dynamic programming formulation can also be employed in an iterative manner in Euclidean space. Although not as simple as the gradient method, we sketch briefly the procedure for $N=2$, for the sake of comparison with the other metric spaces we have studied.

The principle of the algorithm is to find regions $R_{i}$ which necessarily contain the $y_{i}^{*}, i=1, \ldots, n$, and to shrink these regions as much as possible in successive iterations.
(0) Divide a rectangle $R$ containing all of the $x_{i}, i=1, \ldots, m$, into squares of side $\sqrt{ } 2 s$, and set $R_{1}=\ldots=R_{n}=R$.
(i) For $Y$ dominating only terminal nodes $X_{k_{1}}, \ldots, X_{k_{p}}, \bar{f}_{Y}$ and $f_{Y}$ are calculated just at the center point $c$ of each square as follows

$$
\begin{aligned}
& \bar{f}_{Y}(c)=\sum_{j=1}^{p} d\left(x_{k_{j}}, c\right)+p s \\
& f_{Y}(c)=\sum_{j=1}^{p} d\left(x_{k_{j}}, c\right)-p s
\end{aligned}
$$

For $Y$ immediately dominating $Z_{1}, \ldots, Z_{p}$, set

$$
\begin{aligned}
& \bar{f}_{Y}(c)=\min _{\left(c_{1}, \ldots, c_{p}\right)} \sum_{k=1}^{p}\left[\bar{f}_{Z_{k}}\left(c_{k}\right)+d\left(c_{k}, c\right)\right]+2 p s, \\
& \underline{f}_{Y}(c)=\min _{\left(c_{1}, \ldots, c_{p}\right)} \sum_{k=1}^{p}\left[f_{Z_{k}}\left(c_{k}\right)+d\left(c_{k}, c\right)\right]-2 p s,
\end{aligned}
$$

where the minima are taken over all those $c_{i}$ falling in the appropriate $R_{j}$. The functions $\bar{f}_{Y}$ and $f_{Y}$ represent upper and lower bounds for $f_{Y}$ valid not only at $c$ but throughout the square of which $c$ is the center.
(ii) All squares for which $f_{Y_{r}}(c)>\min _{c} \bar{f}_{Y_{r}}(c)$ could not possibly contain $y_{r}^{*}$, i.e., a $y$ which minimizes $f_{Y_{r}}$, so they are ignored in all further calculations of $\bar{f}_{Y_{r}}$ and $f_{Y_{r}}$. Let $R_{r}^{\prime}$ be the remaining region.
(iii) If $Y_{j}$ immediately dominates $Y_{k}$ and a new region $R_{j}^{\prime}$ has already been delimited, then define $R_{k}^{\prime}$ as follows. Any square $\in R_{k}$ with center $c$ for which

$$
f_{Y_{k}}(c)+\min _{e \in R_{j}^{\prime}} d(e, c)-2 s>\min _{c \in R_{k}}\left[\bar{f}_{Y_{k}}(c)+\max _{e \in R_{j}^{\prime}} d(e, c)+2 s\right]
$$

could not contain $y_{k}^{*}$. The remainder constitute $R_{k}^{\prime}$. This step is repeated until all $Y_{i}$ are exhausted, $i=1, \ldots, n$.
(iv) Redefine $R_{i}$ to be $R_{i}^{\prime}$ for $i=1, \ldots, n$. If all the $R_{i}$ are now smaller than some critical area, stop. Otherwise divide the $R_{i}$ into squares of side $\sqrt{ } 2 s$, where $s$ is suitably smaller than in the present cycle, and return to step (i).

See [1] for a biological interpretation of Steiner trees in Euclidean space.

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