1 Multi-dimensional Range Trees

In d-dimensional computational geometry, when one is dealing with sets of points in d-space, one could ask a d-space range query. Consider a spatial interval $(\langle x_1^1, x_2^1 \rangle \dots \langle x_1^d, x_2^d \rangle)$: this represents an axis-aligned rectilinear solid in d-space. A query could ask for the points that occur within this solid. Such problems are solved efficiently in computational geometry using so-called Range Trees (See [1], Chapter 5). We draw an analogy between this problem and a multi-attribute database query.

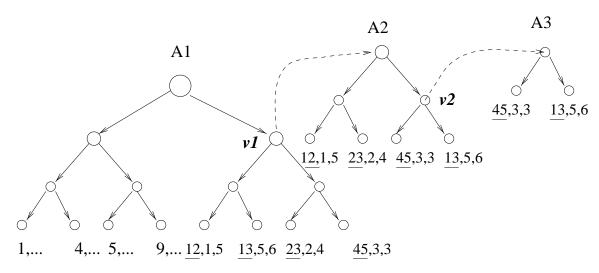


Figure 1: Excerpt of a 3-dimensional range tree, sorted by attributes A_1 , A_2 and A_3

Consider the example mdrt for points in 3D shown in Figure 1. Consider the first, 3-dimensional mdrt (labeled A1). This is simply a tree which sorts the values according to x (first) coordinate (x-coordinate values are underlined). Each interior node in tree A1 is the ancestor of a set of values. Consider such an interior node v1, which is the ancestor of values with primary key values 12, 13, 23, and 45. An mdrt now contains a link from v1 to the root of an associated tree A2, denoted as $T_{assoc}(v1, A_2)$. This 2-dimensional mdrt A2 contains the same set of values with x-values 45, 11, 23, and 13; however, in this tree, they are sorted according to attribute y-coordinate. Likewise each interior node v_i in A1 is the ancestor to a set of values, and contains a pointer to an associated 2-dimensional mdrt $T_{assoc}(v_i, A_2)$ which sorts the values in the subtree below v_i by y-coordinate. In general, each node v_i^j of a $\{d-j+1\}$ -dimensional mdrt contains a pointer to a $\{d-j\}$ -dimensional mdrt. The nodes of the final 1-dimensional tree, do not have such pointers.

Given a 2-dimensional range query which asks for all points (x, y) such that $x_1 \le x \le x_2$ and $y_1 \le y \le y_2$ the structure is used as follows. First, the points q with x-values in the range (x_1, x_2) is identified using tree A1. For simplicity, let us assume (we relax this assumption later) that the values in q form the leaves of a balanced tree with root LCA(q). In an n node tree this range can be found in time $O(\log_2 n)$, the time it takes to find the two end-points of the interval in the first tree. We now follow the link to the associated tree for q this tree sorts just the value set q according to q-coordinate So we can find the subset of q satisfying

 $y_1 \leq y \leq y_2$ also in $O(\log_2 n)$ time. This gives the intuition behind the efficient processing of conjunctive range queries using mdrts. We now relax the assumption that the result of the first selection q includes just the leaves of a balanced tree.

Let us call the leaves of the subtree rooted at node v the canonical subset of v, denoted as P(v). If v is a leaf, P(v) = v. In [1] (pp 103-107) it is shown that any subset of leaves which lies in a range can be expressed as a disjoint union of $O(\log_2 n)$ canonical subsets for the given range query. The roots of these can be found in $O(\log_2 n)$ time in the process of finding the bounding paths for the interval. Given an x-range $\langle x, x' \rangle$, we search for them in the tree until we find node v_{split} where the paths split. Now we search for x (x') in the left (right) subtree. At every point in the search for x where the path goes left, the right subtree belongs to the range; the search for x' goes just the opposite way. The result is a quick identification of roots of the canonical subtrees that precisely cover the leaves whose values are in the interval (see Figure 2).

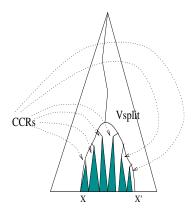


Figure 2: Finding the canonical covering roots (CCRs)

We call these the covering canonical roots (CCRs). Consider, for example, a 2-dimensional range query. First, the CCRs in tree A1 for the given x-range are found; there are $O(\log n)$ of these. Then for each of the CCRs, we go to the associated y-tree, and find the CCRs in that tree for the given y-range. This results in $O(\log^2 n)$ CCRs in each y-tree examined. The union of all the leaves under these CCRs in y-trees constitute the answer. In general, it is shown [1] that d-dimensional range queries which have a total of p points satisfying them can be computed in time $O(\log^d(n) + p)$. Range trees require $O(n \log^{d-1} n)$ storage space, and can be constructed in time $O(n \log^{d-1} n)$.

An enhancement of mdrts called fractional cascading [1] reduces the time to answer a query to $O(\log^{d-1}(n) + p)$ (but does not reduce the space).

References

[1] M. D. Berg, M. V. Kreveld, M. Overmars and O. Schwarzkopf. *Computational Geometry*. Springer-Verlag, New York.