

Computing Maximal Layers Of Points In $E^{f(n)}$

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Work $(p \succ q)$

For two points $p, q \in E^k$ we say p dominates q or p is above q iff $p[i] \geq q[i] \forall i \in [k]$.

Let $P = \{p_1, \dots, p_n\}$ with $p_i \in E^k$, we define the **First Maximal Layer** $M_1(P)$ of P as the set of points, that are not dominated by any other points.

The h^{th} maximal layer is defined recursively:

$$M_h(P) = M_1 \left(P \setminus \bigcup_{i=1}^{h-1} M_i(P) \right)$$

The **height** h of P is the number of non-empty (maximal) layers in P and the **width** w is the size of the largest **anti-chain** in (P, \succ) .

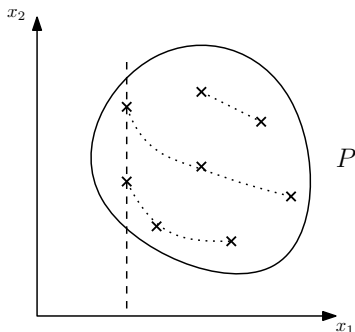


Figure: Dotted chains represent the maximal layers of the point set P .

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P is a random order if P is built by picking points uniformly at random from the unit k -cube $[0, 1]^k$.

Alternatively P is picked uniformly at random from S_n^k , set of all k -tuple of n -permutations

It can be shown that they are equivalent

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- 1 The expected height $\bar{h} = O(n^{1/k})$ and width $\bar{w} = O(n^{1-1/k})$
- 2 It is known that $M_1(P) = O((\log n)^{k-1})$ (for fixed k)
- 3 There is no known asymptotic formula for the size distribution of layers beyond the first layer.
- 4 We show (in ongoing work) the problem is closely related to determining the **longest increasing subsequence** of a random permutation.
- 5 For a random order $\mathbf{Prob}[p[i] = q[i]] = 0$.

Deterministic Setting: (k is fixed)

- 1 In a seminal paper [Kung et al, 1975] proposed an algorithm for finding the first maximal layer in $O(n(\log n)^{k-2})$ time. This uses multi-dimensional divide-and-conquer approach.
- 2 [Jensen, 2003] extended this to compute all the layers in $O(n(\log n)^{k-1})$ time.

($k = n$)

[Matoušek, 1991] gave a $O(n^{1.5+\omega/2})$ algorithm for determining the first maximal layer.

Randomized Algorithm When P Is A Random Order

- 1 [Bentley, 1975] gave expected $O(n)$ algorithm for a fixed k

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- 1 We introduce a new iterative framework for the maximal layers problem.
- 2 We assume that the dimension k may not be constant,
- 3 The proposed algorithm has an expected run-time of $O(kn^{2-\delta(k)})$ ($\delta(k) > 0$).
- 4 The factor k accounts for the fact that checking dominance requires $O(k)$ time
- 5 Later it is extended for an arbitrary P

A Simple Iterative Algorithm For Random Orders

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Main Steps:

- 1 Sort the points in descending order according to the L_∞ norm. (T)
- 2 We pick from T one element at a time in order.
- 3 `Insert()` the element in to the layer it belongs to.
- 4 End when T is empty.

Step 3 is achieved through a new data structure [Hierarchical Search Tree](#) (HST), for each layer.

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If p precedes q in T then q cannot dominate p .

Expected runtime = $n \times t_i + t_n$.

t_i = the expected time to insert an element in to its
HST. (Line 3)

t_n = time to compute T

T can be computed in $O(kn + \log n)$, which we can
ignore

We maintain each layer using an HST

- 1 Layers are totally ordered.
- 2 Hence, we can use a binary search tree (B), to maintain the sorted order
- 3 During `Insert()` we traverse B , querying corresponding HST to see if the current point is dominated by some point in that layers/HST
- 4 If no suitable HST is found, then we create an empty HST in B and insert the point

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An HST Is Defined Recursively

- 1 A singleton node (root) storing a point p .
- 2 A root has a number of non-empty children nodes (up to k) each of which is a root of an HST.
- 3 If node q is in the j^{th} subtree of node p then $p[j] \geq q[j]$

HST is similar in principal to space partition trees.

Hierarchical Search Tree (HST) Queries

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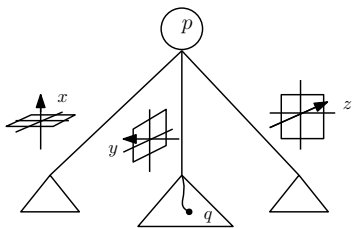


Figure: Example of an HST for the 3-dimensional case.

Above(): Checks whether the query point is dominated by some point inside the layer

Add(): Inserts the query point into the HST

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Above()

- 1 Let q be the query point, let p be the point at the root
- 2 **if** $p \succeq q$ **then** stop (the current layer dominates q)
else
recursively search a subtree for each each half-space
labeled j for which $p[j] \geq q[j]$

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Add()

- 1 We start at the root, and proceed as we did in the case of `Above()`
- 2 If there are more than one choice of subtrees then we choose one uniformly at random
- 3 We stop and insert the point when we reach an empty leaf node.

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Lemma:

$$\eta = \text{Prob}\{ p[j] \geq q[j] \mid p \text{ precedes } q \text{ in } T \} = 1 - \frac{k-1}{2(k+1)}$$

- 1 We see η does not depend on the relative ranks in T of the elements
- 2 $t_i = t_{\text{search}} + t_{\text{add}}$
- 3 t_{search} = the expected time it takes to search B to find the correct layer
- 4 t_{add} = the cost adding the element to an HST
- 5 $t_{\text{search}} \geq t_{\text{add}}$

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- 1 First we compute $a_{m,d}$ the expected number of nodes at **depth d** of a HST with m nodes
- 2 Let $b_{m,k} =$ the number of nodes visited during `Above()`
- 3 We use $a_{m,d}$ to bound the expected value of $b_{m,k}$
- 4 Then $t_{search} = b_{m,k} \times \log h$

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- ① Recall $\eta = 1 - \frac{k-1}{2(k+1)}$ is the probability that a node p , already in some HST, dominates the query point q . Then,

$$b_{m,k} = \sum_{i=0}^{m-1} a_{m,i} \eta^i$$

- ② We compute $a_{m,d}$ using the following recurrence:

$$a_{m,d} = \left(\frac{1}{k^{d-1}} \right) a_{m-1,d-1} + \left(1 - \frac{1}{k^d} \right) a_{m-1,d}$$

- ③ It is unlikely that the above recurrence has a closed form as the recurrence for the binomial coefficient is a special form of it.

$$a_{m,d} = \left(\frac{1}{k^{d-1}} \right) a_{m-1,d-1} + \left(1 - \frac{1}{k^d} \right) a_{m-1,d}$$

- 1 Fortunately $a_{m,d}$ is unimodal and is bell-shaped with respect to d when w and k are fixed
- 2 Using bounds on the maxima of $a_{m,d}$ we show that

$$b_{m,k} = O\left(m^{1 - \frac{1}{\log k} + \log_k \left(1 + \frac{1}{k}\right)}\right)$$

- 3 $m, h \leq n$
- 4 $t_{search} = b_{m,k} \times \log h = O(kn^{1-\delta(k)} \log n)$
- 5 Total runtime = $O(kn^{2-\delta'(k)})$

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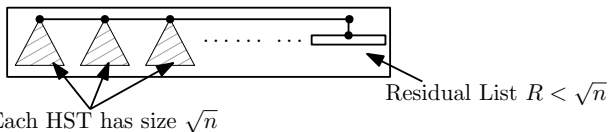
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List-HST with $\leq \sqrt{n}$ nodes



- 1 The [Above\(\)](#) is modified to iterate over this list of HSTs instead of a single one and the list R
- 2 When adding to the List-HST, we first check if R is full or not. If R is full we create an HST and add it to the list. Otherwise we just add the point to the R .

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Creating an HST

We randomly permute the points in R before creating a new HST

So we can carry the same probabilistic assumptions from the previous case

- 1 It can be shown that SEARCH on the list of HSTs takes $O(kn^{1/2+(\log_k(k-1))/2})$
- 2 Which gives the total runtime of $O(k^2n^{3/2+(\log_k(k-1))/2} \log n) = O(k^2n^{2-\xi(k)})$
- 3 $0 < \xi(k) < 1/2$

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- 1 In the deterministic setting, can we reduce the upper bound to $o(n^2)$ independent of k .
- 2 Alternatively improve the trivial lower bound of $(n \log n)$
- 3 Given a random order, determine the expected value of $M_h(P)$.

Questions?



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