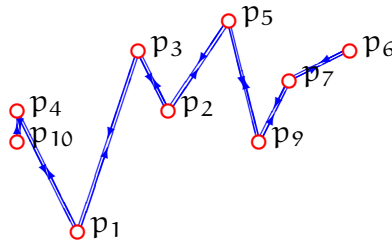


2. Chan's Algorithm

Lecture on Thursday 24th September, 2009 by Michael Hoffmann <hoffmann@inf.ethz.ch>

2.1 Graham Scan (Successive Local Repair)

Sort points lexicographically and remove duplicates: (p_1, \dots, p_n) .



$p_{10} p_4 p_1 p_3 p_2 p_5 p_9 p_7 p_6 p_7 p_9 p_5 p_2 p_3 p_1 p_4 p_{10}$

As long as there is a (consecutive) triple (p, q, r) s.t. q is left of or on the directed line \overrightarrow{pr} , remove q from the sequence.

Theorem 2.1 *The convex hull of a set $P \subset \mathbb{R}^2$ of n points can be computed using $O(n \log n)$ geometric operations.*

Proof.

1. Sorting and removal of duplicate points: $O(n \log n)$.
2. At begin: $2n - 2$ points; at the end: h points. $\Rightarrow 2n - h - 2$ shortcuts/positive orientation tests. In addition at most $2n - 2$ negative tests. Altogether at most $4n - h - 4$ orientation tests.

In total $O(n \log n)$ operations. Note that the number of orientation tests is linear only, but $O(n \log n)$ lexicographic comparisons are needed. \square

There are many variations of this algorithm, the basic idea is due to Graham [4].

2.2 Lower Bound

Theorem 2.2 $\Omega(n \log n)$ geometric operations are needed to construct the convex hull of n points in \mathbb{R}^2 (in the algebraic computation tree model).

Proof. Reduction from sorting (for which it is known that $\Omega(n \log n)$ comparisons are needed in the algebraic computation tree model). Given n real numbers x_1, \dots, x_n , construct a set $P = \{p_i \mid 1 \leq i \leq n\}$ of n points in \mathbb{R}^2 by setting $p_i = (x_i, x_i^2)$. This construction can be regarded as embedding the numbers into \mathbb{R}^2 along the x -axis and

then projecting the resulting points vertically onto the unit parabola. The order in which the points appear along the lower convex hull of P corresponds to the sorted order of the x_i . Therefore, if we could construct the convex hull in $o(n \log n)$ time, we could also sort in $o(n \log n)$ time. \square

Clearly this simple reduction does not work for the Extremal Points problem. But using a more involved construction one can show that $\Omega(n \log n)$ is also a lower bound for the number of operations needed to compute the set of extremal points only. This was first shown by Avis [1] for linear computation trees, then by Yao [5] for quadratic computation trees, and finally by Ben-Or [2] for general algebraic computation trees.

In fact, the argument is based on a lower bound of $\Omega(n \log n)$ operations for *Element Uniqueness*: Given n real numbers, are any two of them equal? At first glance, this problem appears a lot easier than sorting, but apparently it is not, at least in this model of computation.

2.3 Jarvis' Wrap and Graham Scan in C++

Jarvis' Wrap.

$p[0..N)$ contains a sequence of points.
 p_{start} point with smallest x -coordinate.
 q_{next} some *other* point in $p[0..N)$.

```
int h = 0;
Point_2 q_now = p_start;
do {
    q[h] = q_now;
    h = h + 1;

    for (int i = 0; i < N; i = i + 1)
        if (rightturn_2(q_now, q_next, p[i]))
            q_next = p[i];

    q_now = q_next;
    q_next = p_start;
} while (q_now != p_start);
```

$q[0, h)$ describes a convex polygon bounding the convex hull of $p[0..N)$.

Graham Scan.

$p[0..N)$ lexicographically sorted sequence of pairwise distinct points, $N \geq 2$.

```
q[0] = p[0];
int h = 0;
```

```

// Lower convex hull (left to right):
for (int i = 1; i < N; i = i + 1) {
    while (h>0 && rightturn_2(q[h-1], q[h], p[i]))
        h = h - 1;
    h = h + 1;
    q[h] = p[i];
}

// Upper convex hull (right to left):
for (int i = N-2; i >= 0; i = i - 1) {
    while (rightturn_2(q[h-1], q[h], p[i]))
        h = h - 1;
    h = h + 1;
    q[h] = p[i];
}

```

$q[0,h]$ describes a convex polygon bounding the convex hull of $p[0..N]$.

2.4 Chan's Algorithm

Given matching upper and lower bounds we may be tempted to consider the algorithmic complexity of the planar convex hull problem settled. However, this is not really the case: Recall that the lower bound is a worst case bound. For instance, the Jarvis' Wrap runs in $O(nh)$ time and thus beats the $\Omega(n \log n)$ bound in case that $h = o(\log n)$. The question remains whether one can achieve both output dependence and optimal worst case performance at the same time. Indeed, Chan [3] presented an algorithm to achieve this runtime by cleverly combining the "best of" Jarvis' Wrap and Graham Scan. Let us look at this algorithm in detail.

Divide. *Input:* a set $P \subset \mathbb{R}^2$ of n points and a number $H \in \{1, \dots, n\}$.

1. Divide P into $k = \lceil n/H \rceil$ sets P_1, \dots, P_k with $|P_i| \leq H$.
2. Construct $\text{conv}(P_i)$ for all i , $1 \leq i \leq k$.
3. Construct H vertices of $\text{conv}(P)$. (*conquer*)

Analysis. Step 1 takes $O(n)$ time. Step 2 can be handled using Graham Scan in $O(H \log H)$ time for any single P_i , that is, $O(n \log H)$ time in total.

Conquer.

1. Find the lexicographically smallest point in $\text{conv}(P_i)$ for all i , $1 \leq i \leq k$.

- Starting from the lexicographically smallest point of P find the first H points of $\text{conv}(P)$ oriented counterclockwise (simultaneous Jarvis' Wrap on the sequences $\text{conv}(P_i)$).

Determine in every step the points of tangency from the current point of $\text{conv}(P)$ to $\text{conv}(P_i)$, $1 \leq i \leq k$, using binary search.

Analysis. Step 1 takes $O(n)$ time. Step 2 consists of at most H wrap steps. Each wrap needs to find the minimum among k candidates where each candidate is computed by a binary searches on at most H elements. This amounts to $O(Hk \log H) = O(n \log H)$ time for Step 2.

Remark. Using a more clever search strategy instead of many binary searches one can handle the conquer phase in $O(n)$ time. However, this is irrelevant as far as the asymptotic runtime is concerned, given that already the divide step takes $O(n \log H)$ time.

Searching for h . While the runtime bound for $H = h$ is exactly what we were heading for, it looks like in order to actually run the algorithm we would have to know h , which—in general—we do not. Fortunately we can circumvent this problem rather easily, by applying what is called a *doubly exponential search*. It works as follows.

Call the algorithm from above iteratively with parameter $H = \min\{2^{2^t}, n\}$, for $t = 0, \dots$, until the conquer step finds all extremal points of P (i.e., the wrap returns to its starting point).

Analysis: Let 2^{2^s} be the last parameter for which the algorithm is called. Since the previous call with $H = 2^{2^{s-1}}$ did not find all extremal points, we know that $2^{2^{s-1}} < h$, that is, $2^{s-1} < \log h$, where h is the number of extremal points of P . The total runtime is therefore at most

$$\sum_{i=0}^s cn \log 2^{2^i} = \sum_{i=0}^s cn 2^i = cn(2^{s+1} - 1) < 4cn \log h = O(n \log h).$$

In summary, we obtain the following theorem.

Theorem 2.3 *The convex hull of a set $P \subset \mathbb{R}^2$ of n points can be computed using $O(n \log h)$ geometric operations, where h is the number of convex hull vertices.*

Questions

- How is convexity defined? What is the convex hull of a set in \mathbb{R}^d ? Give at least three possible definitions.
- What does it mean to compute the convex hull of a set of points in \mathbb{R}^2 ? Discuss input and expected output and possible degeneracies.

3. *How can the convex hull of a set of n points in \mathbb{R}^2 be computed efficiently?* Describe and analyze (incl. proofs) Jarvis' Wrap, Successive Local Repair, and Chan's Algorithm.
4. *Is there a linear time algorithm to compute the convex hull of n points in \mathbb{R}^2 ?* Prove the lower bound and define/explain the model in which it holds.
5. *Which geometric primitive operations are needed to compute the convex hull of n points in \mathbb{R}^2 ?* Explain the two predicates and how to compute them.

References

- [1] D. Avis, Comments on a lower bound for convex hull determination, *Inform. Process. Lett.* **11** (1980), 126.
- [2] M. Ben-Or, Lower bounds for algebraic computation trees, in: *Proc. 15th Annu. ACM Sympos. Theory Comput.*, 1983, 80–86.
- [3] Timothy M. Chan, Optimal Output-Sensitive Convex Hull Algorithms in Two and Three Dimensions, *Discrete Comput. Geom.* **16** (1996), 361–368.
- [4] R. L. Graham, An efficient algorithm for determining the convex hull of a finite planar set, *Inform. Process. Lett.* **1** (1972), 132–133.
- [5] A. C. Yao, A lower bound to finding convex hulls, *J. ACM* **28** (1981), 780–787.