

CONVEX HULLS

Chan's optimal output sensitive
algorithms for convex hulls

Alberto Parravicini

Université libre de Bruxelles

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What's on the menu?

- **Basic notions**
- **Algorithms for convex hulls**
- **Optimal 2D algorithm**
- **Optimal 3D algorithm**

Crash course on convex hulls

A *convex set* S is a set in which, $\forall x, y \in S$, the segment $xy \subseteq S$.

Given a set of points P in d dimensions, the **Convex Hull** $\text{CH}(P)$ of P is:

- the *minimal convex set* containing P .
- the union of *all convex combinations* of points in P , i.e. the points $\text{CH}(P)$ are s.t.

$$\sum_{i=1}^{|P|} w_i \cdot x_i, \quad \forall x_i \in P, \quad \forall w_i : w_i \geq 0 \text{ and } \sum_{i=1}^{|P|} w_i = 1$$

Output sensitive algorithm

The complexity of an *output-sensitive* algorithm is a function of both the *input size* and the *output size*.

Jarvis's march

- **Core idea:** given an edge pq of the convex hull, the next edge qr to be added will maximize the angle $\angle pqr$
- At each step we scan all the n points.
- How many steps? h , size of the hull.
- **Complexity:** $O(nh)$

One step, visually

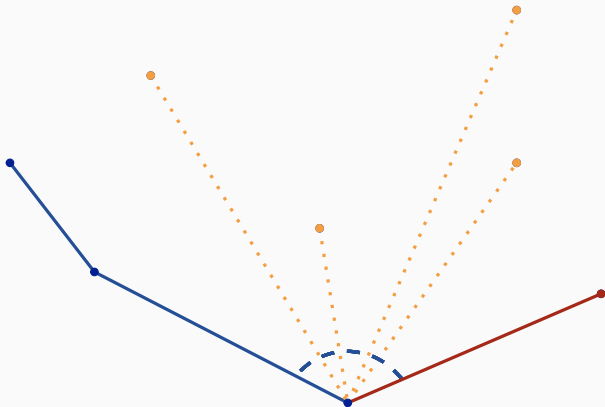


Figure: A step of the Jarvis March. The *red* line is the next segment that will be added to the hull.

Jarvis's march

Algorithm 1: Jarvis March

Input: a list S of bidimensional points.

Output: the convex hull of the set, sorted counterclockwise.

$hull = []$

$x_0 =$ the leftmost point.

$hull.push(x_0)$

Loop $hull.last() \neq hull.first()$

$candidate = S.first()$

foreach p in S **do**

if $p \neq hull.last()$ and p is on the right of the segment
 " $hull.last(), candidate$ " **then**

$candidate = p$

if $candidate \neq hull.first$ **then** $hull.push(candidate)$ **else break**

return $hull$

Graham's scan

- Sort the points in clockwise order around the leftmost point - **Cost:** $O(n \log n)$
- Keep the current hull in a stack
 $H = \{h_0, \dots, h_{curr}\}.$
- Inspect each point p in order - **Cost:** $O(n)$
 - While $h_{curr-1}h_{curr}p$ is a right turn, pop h_{curr} .
 - Push p to H .
- **Overall complexity:** $O(n \log n)$

Graham's scan

Algorithm 2: Graham Scan

Input: a list S of bidimensional points.

Output: the convex hull of the set, sorted counterclockwise.

$hull = []$

x_0 = the leftmost point.

Put x_0 as the first element of S .

Sort the remaining points in counter-clockwise order, with respect to x_0 .

Add the first 3 points in S to the hull.

forall *the remaining points in S* **do**

while $hull.second_to_last(), hull.last(), p$ form a right turn **do**

$hull.pop()$

$hull.push(p)$

return $hull$

Can we do better?

So far:

→ Jarvis's march: $O(nh)$

→ Graham's scan: $O(n \log n)$

→ How do we compare their complexity?

→ We would like to do even better!

Can we get to $O(n \log h)$?

Chan's 2D algorithm

- **Idea:** Some points will never be in the hull! Discard them to speed up *Jarvis's march*.
- Combine *Jarvis's march* and *Graham's scan*.
- Algorithm (**Chan 2D**):
 - Split the n points in groups of size m ($\lceil n/m \rceil$ groups).
 - Compute the hull H_i of each group**Cost:** $O((n/m) \cdot (m \log m)) = O(n \log m)$
 - Until the hull is complete, repeat:
 - Given the current hull $H_{tot} = \{h_0, \dots, h_{curr}\}$, find for each H_i the point right tangent to h_{curr} (binary search, $O(\log m) \lceil n/m \rceil$).
 - Pick the tangent point p that maximizes $\angle h_{curr-1} h_{curr} p$.

One step, visually

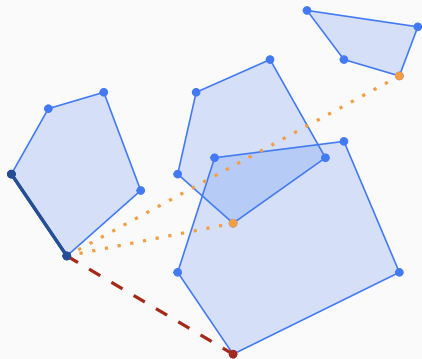


Figure: A step of Chan's algorithm. In *blue*, the existing hull, in *orange*, the tangents, in *red*, the new edge that will be added.

Chan's 2D algorithm, part 2

Complexity: $O(n \log m + (h(n/m) \log m))$

What about m ? Let's pretend the size h of the final hull is known a-priori.

If we put $m = h$, we get complexity

$$O(n \log h + (h(n/h) \log h)) = O(n \log h)$$

But h is not known!

- Reiterate the algorithm multiple times, with $m = 2^{2^i}$
- Iteration cost: $O(n \log H) = O(n 2^i)$
- $O\left(\sum_{i=1}^{\lceil \log \log h \rceil} n 2^i\right) = O(n 2^{\lceil \log \log h \rceil + 1}) = O(n \log h)$

Chan's 2D pseudo-code

Algorithm 3: ChanHullStep, a step of Chan's algorithm

Input: a list S of bidimensional points, the parameters m, H

Output: the convex hull of the set, sorted counterclockwise, or an empty list, if H is $< h$

Partition S into subsets $S_1, \dots, S_{\lceil n/m \rceil}$.

for $i = 1, \dots, \lceil n/m \rceil$ **do**

 Compute the convex hull of S_i by using Graham Scan, store the output in a counter-clockwise sorted list.

$p_0 = (0, -\infty)$

p_1 = the leftmost point of S .

for $k = 1, \dots, H$ **do**

for $i = 1, \dots, \lceil n/m \rceil$ **do**

 Compute the points $q_i \in S$ that maximizes $\angle p_{k-1}p_kq_i$, with $q_i \neq p_k$, by performing binary search on the vertices of the partial hull S_i .

p_{k+1} = the point $q \in \{q_1, \dots, q_{\lceil n/m \rceil}\}$.

if $p_{k+1} = p_t$ **then** return $\{p_1, \dots, p_k\}$

return *incomplete*

Chan's 2D pseudo-code, reprise

Algorithm 4: Chan's algorithm

Input: a list S of bidimensional points

Output: the convex hull of the set

for $i = 1, 2, \dots$ **do**

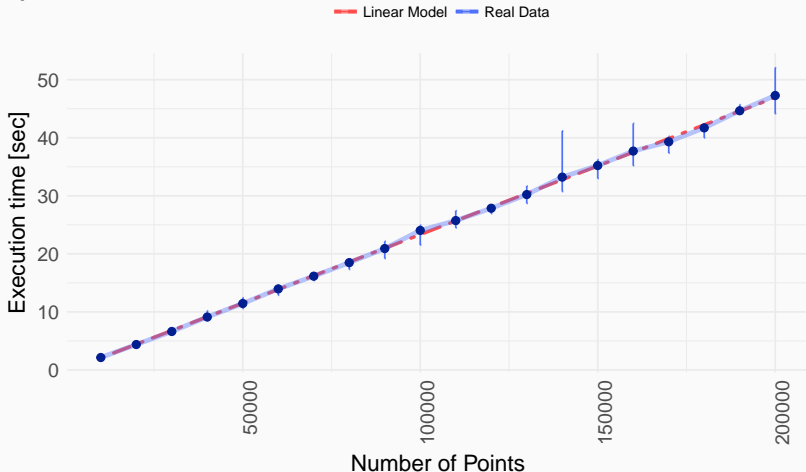
$L = \text{ChanHullStep}(S, m, H)$, where $m = H = \min\{|S|, 2^{2^i}\}$

if $L \neq \text{incomplete}$ **then** return L

Chan's 2D - Empirical analysis

Is a real implementation $O(n \log h)$?

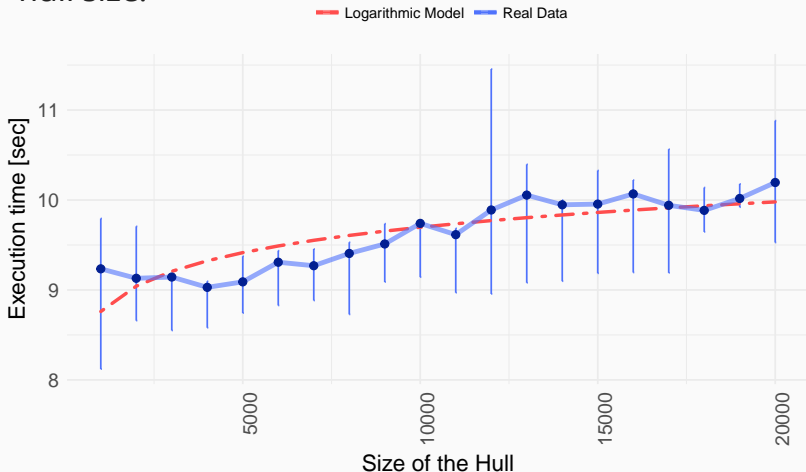
Test 1: fixed hull size (1000), increasing number of points.



Chan's 2D - Empirical analysis

Is a real implementation $O(n \log h)$?

Test 2: fixed number of points (40000), increasing hull size.



THANK YOU!