# CONVEX HULLS Chan's optimal output sensitive

Chan's optimal output sensitive algorithms for convex hulls

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## What's on the menu?

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

#### **BASIC NOTIONS**

### 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 CH(P)** of *P* is:

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

$$\sum_{i=1}^{|P|} \mathbf{w}_i \cdot \mathbf{x}_i, \ \forall \mathbf{x}_i \in P, \ \forall \mathbf{w}_i : \ \mathbf{w}_i \ge 0 \ \text{and} \ \sum_{i=1}^{|P|} \mathbf{w}_i = 1$$

## Output sensitive algorithm

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

#### ALGORITHMS FOR CONVEX HULLS

#### Jarvis's march

→ Core idea: given an edge pq of the convex hull, the next edge qr to be added will maximize the angle ∠pqr

- $\rightarrow$  At each step we scan all the *n* points.
- $\rightarrow$  How many steps? h, size of the hull.
- $\rightarrow$  Complexity: O(nh)

## One step, visually

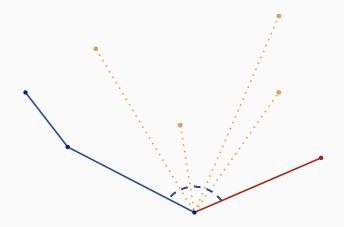


Figure: A step of the Jarvis March. The red line is the next segment that will be added to the hull.

#### 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() != hull.first()
   candidate = S.first()
   foreach p in S do
       if p != hull.last() and p is on the right of the segment
        "hull.last(), candidate" then
          candidate = p
   if candidate != 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)
- $\rightarrow$  Keep the current hull in a stack  $H = \{h_0, \ldots, h_{curr}\}$ .
- $\rightarrow$  Inspect each point p in order **Cost:** O(n)
  - $\rightarrow$  While  $h_{curr-1}h_{curr}p$  is a right turn, pop  $h_{curr}$ .
  - $\rightarrow$  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:

- $\rightarrow$  Jarvis's march: O(nh)
- $\rightarrow$  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

## 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  $H_{tot} = \{h_0, \ldots, h_{curr}\}$ , find for each  $H_i$  the point right tangent to  $h_{curr}$  (binary search,  $O(\log m) \lceil n/m \rceil$ ).
    - $\rightarrow$  Pick the tangent point p that maximizes  $\angle h_{curr-1}h_{curr}p$ .

## One step, visually

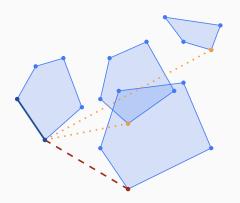


Figure: A step of Chan's algorithm. In blue, the existing hull, in orange, the tangents, in red, the new edge that will be added.

## 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 know.

With 
$$m = h$$
, we get complexity  $O(n \log h + (h(n/h)\log h)) = O(n \log h)$ 

#### But h is not known!

- $\rightarrow$  Reiterate the algorithm, with  $m = 2^{2^i}$
- → Iteration cost:  $O(n \log H) = O(n2^i)$
- $\rightarrow O\left(\sum_{i=1}^{\lceil \log \log h \rceil} n2^{i}\right) = O(n2^{\lceil \log \log h \rceil + 1}) = O(n \log h)$

## 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, \ldots, S_{\lceil n/m \rceil}.
for i = 1, \ldots, \lceil n/m \rceil do
    Compute the convex hull of S<sub>i</sub> 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, \ldots, H do
    for i = 1, \ldots, \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, \ldots, q_{\lceil n/m \rceil}\}.
    if p_{k+1} = p_t then return \{p_1, \ldots, p_k\}
return incomplete
```

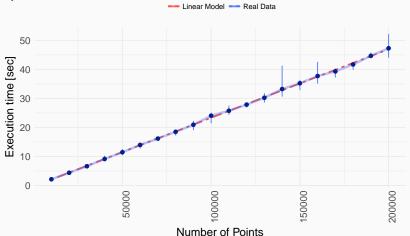
## Chan's 2D pseudo-code, reprise

#### Algorithm 4: Chan's algorithm

## Chan's 2D - Empirical analysis

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

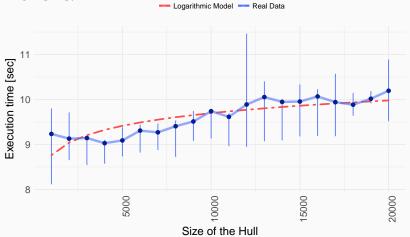
**Test 1:** fixed hull size (1000), increasing number of points.



## Chan's 2D - Empirical analysis

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

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



## CHAN'S 3D ALGORITHM

## Chan's 3D algorithm

→ Idea: The structure of the algorithm doesn't change, just the ingredients.

Given a set *S* of *n* points with an hull of size *h*:

- $\rightarrow$  Jarvis's march  $\rightarrow$  Chand and Kapur's gift wrapping O(nh).
- → Graham's scan  $\rightarrow$  **Preparata and Hong 3D** Algorithm  $O(n \log n)$ .
- → Finding tangents in 2D → Supporting planes in 3D with Dobkin-Kirkpatrick hierarchy -O(log n).

The overall complexity is still  $O(n \log h)$ .

### Chan's 3D algorithm

#### → 3D Gift wrapping

- $\rightarrow$  Pick an edge  $e_i$  of a face f of the partial hull.
- $\rightarrow$  Find p that maximizes the angle between f and the new face given by  $e_j$  and p. Repeat for the 3 edges of f.
- → Use *breadth first visit* to build the entire hull.
- Preparata and Hong 3D algorithm, a divide-and-conquer algorithm for convex hulls
  - $\rightarrow$  Split the set of points S in  $S_1$  and  $S_2$ .
  - → Recursively split until the base case, then build the convex hull.
  - → Merge the partial hulls.

## Chan's 3D algorithm

- → To find supporting planes, store each partial hull as a **Dobkin-Kirkpatrick hierarchy** 
  - → Sequence  $P_0, \ldots, P_k$  of increasingly smaller approximations of a polyhedron  $P = P_0$ .
  - $\rightarrow$  Build  $P_{i+1}$  from  $P_i$  by picking a maximal set of independent vertices of  $P_i$ .
  - $\rightarrow$  Overall, building the hierarchies takes O(n).

#### → Finding supporting planes in 3D

- → **Goal:** given an edge  $e_j$  and the DK hierarchy of a partial hull  $H_i$ , find the plane passing through  $e_j$  and tangent to  $H_i$  in  $p_t$ .
- $\rightarrow$  Find  $p_t$  in constant time in  $P_k$
- $\rightarrow$  Step up in the hierarchy: if  $p_t$  changes, it will move to a neighbour (constant time check).
- $\rightarrow$  The DK hierarchy has O(log m) height.

## **THANK YOU!**

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**Beamer theme:** Presento, by Ratul Saha. The research system in Germany, by Hazem Alsaied