

The Quickhull Algorithm for Higher Dimensional Convex Hulls

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Western
UNIVERSITY CANADA



Saskatchewan is flat as a pancake, this is first time seeing a hill. In Québec, July 1st is also colloquially called 'moving day', because many leases for apartments get over in this time. Alaska is being tied and gagged, because it's very

Canadian dream to kidnap Alaska!

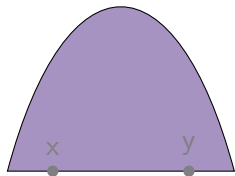
2013 - r/polandball



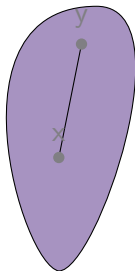
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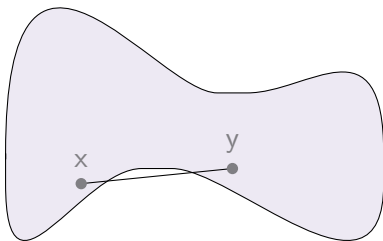




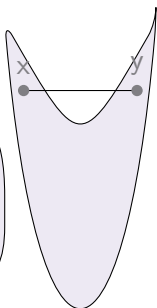
Convex



Convex

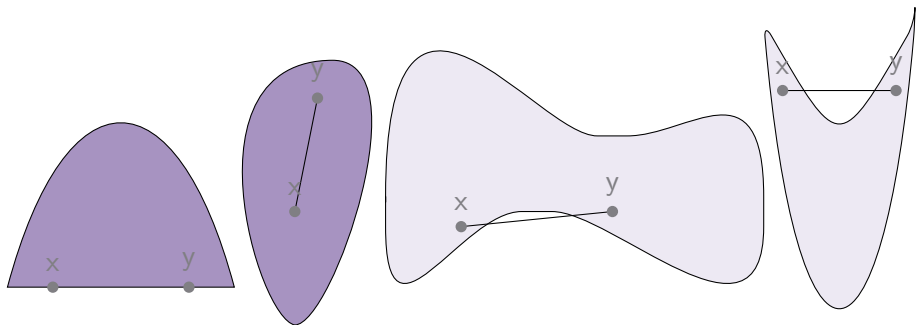


Non-Convex



Non-Convex





Convex

Convex

Non-Convex

Non-Convex

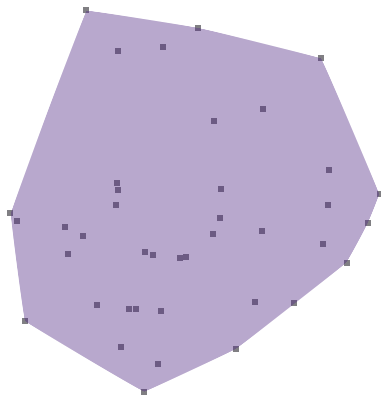
A set S is called **convex** if the line joining any two points in S is in S , i.e.,

$$\forall x, y \in S, \forall \lambda \in [0, 1], \lambda x + (1 - \lambda)y \in S.$$

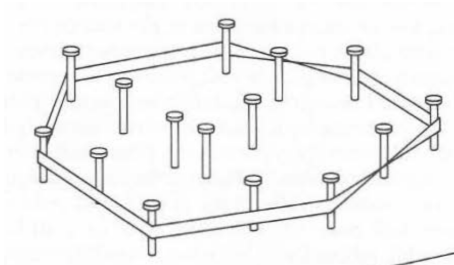


The **convex hull** of a set S is the set of all convex combinations of S , i.e.,

$$\text{conv}(S) := \left\{ \sum_{i=1}^n \lambda_i x_i \mid \sum_{i=1}^n \lambda_i = 1, \lambda_i \in [0, 1] \right\}.$$

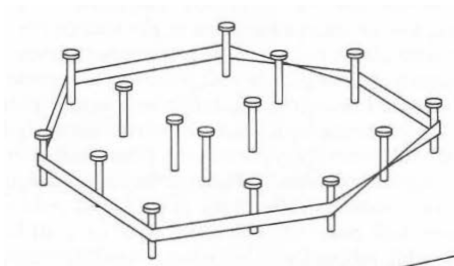


Existence and uniqueness of a convex hull.



1984 – Dewdney's Analog Gadgets

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1984 – Dewdney's Analog Gadgets

The $\text{conv}(S)$ is the smallest convex set containing S , i.e. it is the intersection of all convex sets containing S .

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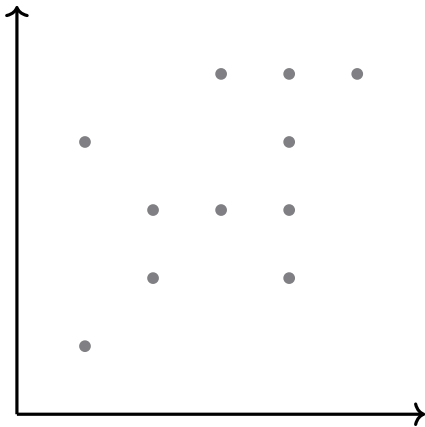
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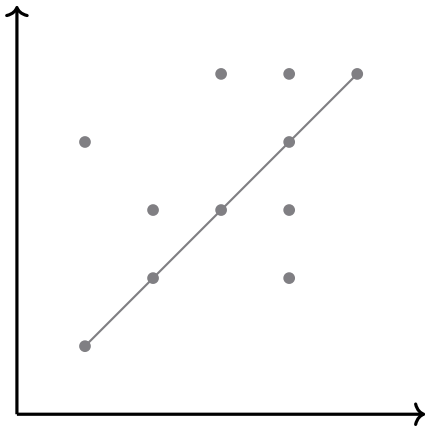
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- Best optimal output-sensitive algorithm created by Chan in 1996.
- Higher dimensional Quickhull was invented in 1996 by C. Bradford Barber, David P. Dobkin, and Hannu Huhdanpaa.



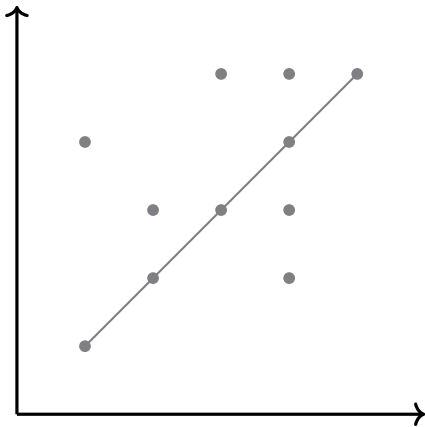
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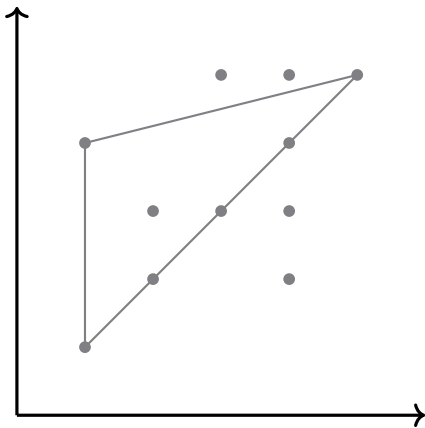
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- 2 Join these two points with a **line L** that divides the shape into two parts.



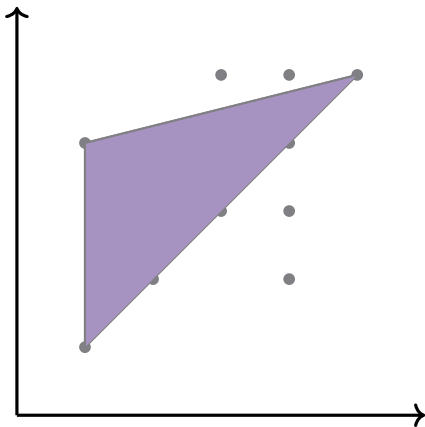
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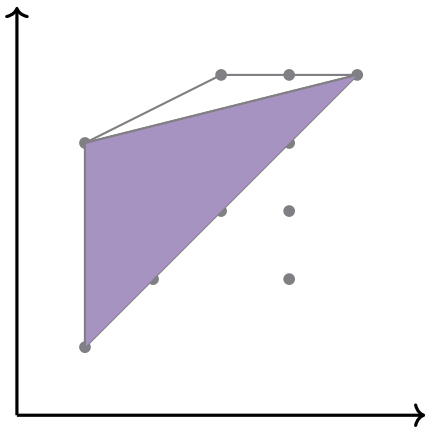
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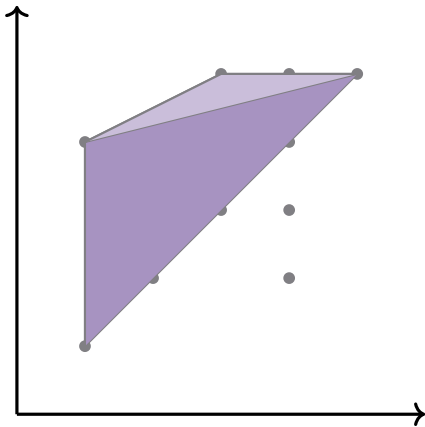
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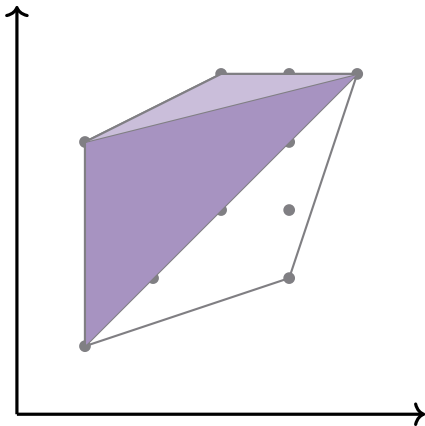
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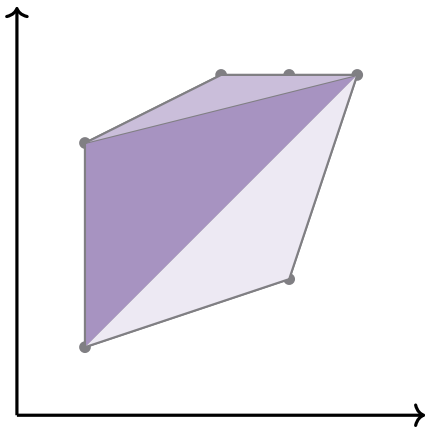
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A nice example of QuickHull algorithm in Python, which randomly generates a set of points and finds the convex hull can be found at [AnantJoshiCZ/QuickHull.git](#).



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An $(n - 1)$ -dimensional face of a polytope is called a **facet** and

$(n - 2)$ -dimensional face is called a **ridge**.



The main geometric operation used in the quickhull algorithm is the **signed distance** from a point $\vec{p} \in \mathbb{R}^n$ to a hyperplane:

$$\frac{\vec{n} \cdot \vec{p} - b}{\|\vec{n}\|}$$



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Normal vector $\vec{n} = (a_1, \dots, a_n)$. Let, \vec{x}_0 be a point in the hyperplane then, $\vec{q} = \vec{p} - \vec{x}_0$. Hence, $proj_{\vec{n}} \vec{q} = \frac{\vec{n} \cdot \vec{q}}{\|\vec{n}\|} = \frac{\vec{n} \cdot (\vec{p} - \vec{x}_0)}{\|\vec{n}\|} = \frac{\vec{n} \cdot \vec{p} - \vec{n} \cdot \vec{x}_0}{\|\vec{n}\|} = \frac{\vec{n} \cdot \vec{p} - b}{\|\vec{n}\|}$, as $\vec{n} \cdot (\vec{x} - \vec{x}_0) = 0 \iff \vec{n} \cdot \vec{x} = \vec{n} \cdot \vec{x}_0 = b$.

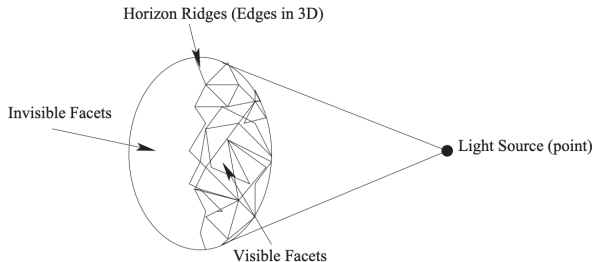


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If the distance is positive, we say the point is **above** the hyperplane, else the point is **below** the hyperplane.

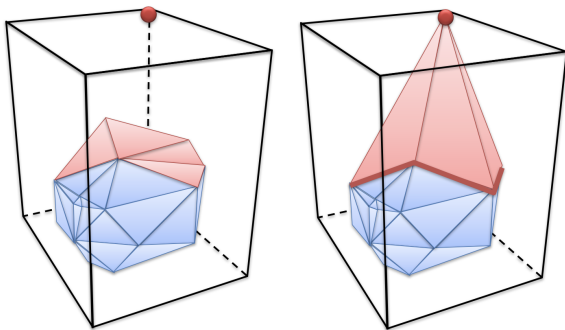


2005 - Agarwal

Theorem (Simplified Beneath-Beyond, 1961 - Grünbaum)

If S is a convex hull in \mathbb{R}^n and P be a point in $\mathbb{R}^n \setminus S$, then F is a facet of $\text{conv}(P \cup S)$ if and only if:

- F is a facet of S , and P is below F , or
- F is not a facet of S , and its vertices are P and the vertices of the ridges of S with one facet below P and other facet above P .



2010 - Farag

- ① Create initial convex hull $Hull$ by choosing $(n + 1)$ -points which do not share a plane or a hyperplane.
- ② For each facet F in $Hull$, find all unassigned points above it and add them to F 's outside set F^c .
- ③ For each F with non-empty F^c ,
 - ① Find the point P with maximum distance from F and add it to Fac .
 - ② Create a visible set V and initialize it to F .
 - ③ The boundary of V froms the set of horizon ridges H .
 - ④ Add the facets created from P and H to Fac .
 - ⑤ Delete all the points of F^c which are created from facets of V .
 - ⑥ For each $Fac \setminus Hull$ go to step 3.2
 - ⑦ Delete the internal facets from $Hull$ and add $Fac \setminus Hull$ to $Hull$.
- ④ Repeat step 2 and step 3 until all points are points are inside the convex hull.



A nice visual implementation of QuickHull algorithm in 3-dimension can be found at [carolhmj/quickhull-3d.git](https://github.com/carolhmj/quickhull-3d).



Lemma

Every extreme point of the input is added to the convex hull irrespective of which outside set it has been assigned to.



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Prove by contradiction: assume an extreme point P which does not belong to any outside set and hence do not belong to $Hull$. Since, P is an extreme point it must belong to at least one outside set for F in the initial $Hull$. By assumption, there is a point Q with P in its visible outside sets but not in its new outside sets. By definition P is above a visible facet and below all new visible facets for Q , which implies that P is in $Hull$ and hence not an extreme point.



Theorem

The quickhull algorithm produces a convex hull of a set of points in \mathbb{R}^n .



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The algorithm begins with a convex hull of $(n + 1)$ -points and partitions points into outside sets and chooses the point with maximum distance. By last lemma and the Simplified Beneath-Beyond theorem quickhull algorithm produces convex hull of processed points.



Graham Scan
sorted points

$$\mathcal{O}(n \log n)$$

$$\mathcal{O}(n)$$

Jarvis March or Gift wrapping
all points on convex hull

$$\mathcal{O}(nh)$$

$$\mathcal{O}(n^2)$$

Quickhull for dimension ≤ 3
for dimension $d > 3$ [1996 - Klee]

$$\mathcal{O}(n \log h)$$

$$\mathcal{O}\left(\frac{n}{h}(\mathcal{O}(h^{\lfloor d/2 \rfloor} / \lfloor d/2 \rfloor!))\right)$$

Kirkpatrick–Seidel algorithm

$$\mathcal{O}(n \log h)$$

Chan's algorithm

$$\mathcal{O}(n \log h)$$



- Image Processing: Quickhull and Chan's algorithm because they can handle large point sets, allowing for the determination of object boundaries or region of interests in images.



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- **Data Visualization:** Graham's scan, Quickhull, and Chan's algorithm are popular choices due to their efficiency and ease of implementation.

