Computational topology: Lecture 14

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2 Alpha shapes

Some implementation

Definitions

Given a discrete set S of points:

- intersection of all convex sets containing S;
- minimum convex set containing S
- 3 set spanned by all convex combinations of S points

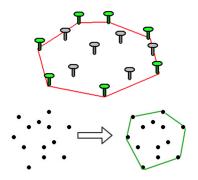
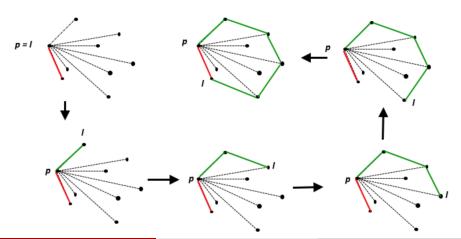


Figure 1: Some examples Computational topology: Lecture 14

Jarvis algorithm (1973)

Gift Wrap Algorithm (Jarvis March Algorithm) to find Convex Hull

O(nh) complexity, where n = #S, and h is the number of points on the convex hull.



Chan's algorithm (1996)

Timothy M. Chan. "Optimal output-sensitive convex hull algorithms in two and three dimensions". *Discrete and Computational Geometry*, Vol. 16, pp.361–368. 1996.

Chan's algorithm in the planar case: the algorithm combines an $O(n \log n)$ algorithm (Graham scan-line, for example) with Jarvis march O(nh), in order to obtain an optimal $O(n \log h)$ time.

Execution demo on Wikipedia

Alpha shapes

Goal: study the shape of a set of points

H. Edelsbrunner, *A short course in computational geometry and topology*, Springer, 2014

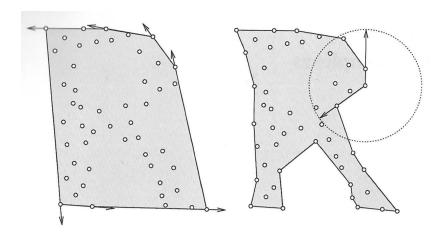


Figure 3: Jarvis construction

α -Hull and α -Shape

H. Edelsbrunner, D. Kirkpatrick and R. Seidel, "On the shape of a set of points in the plane," in *IEEE Transactions on Information Theory*, vol.29, no.4, 1983

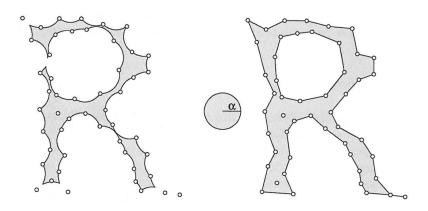


Figure 4: A set of points sampling the letter 'R'

α -Hull and α -Shape

Let $\alpha \geq 0$ be a fixed radius:

- $D_x(\alpha)$ for a closed disk with center x;
- if $D_x(\alpha) \cap S = \emptyset$: then the disk is empty;
- α -hull of S is the complement of the union of empty disks of radius α ;
- $\alpha = 0 \rightarrow S$;
- $\alpha = \infty \to \text{conv}(S)$

Union of disks and Voronoi decomposition

Union of disks of same radius α centered on S points:

$$\mathbb{U}_{\mathcal{S}}(\alpha) = \bigcup_{s \in \mathcal{S}} D_s(\alpha) = \bigcup_{s \in \mathcal{S}} R_s(\alpha), \quad \text{where} \quad R_s(\alpha) = V_{\mathcal{S}} \cap D_s(\alpha)$$

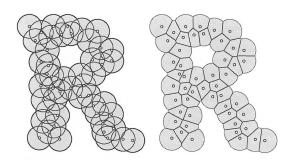


Figure 5: The Voronoi decomposition of the the union

α -Complex (in 2D)

According to Delaunay triangulation, there is an edge between two points if their regions intersect in a common edge, and a triangle between three points if their regions intersect in a common point

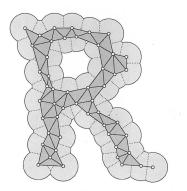


Figure 6: Union of disks decomposed by Voronoi and Delaunay complexes

α -Complex

$$A(\alpha) = \{ \sigma \in K | \alpha_{\sigma} \le \alpha \},$$

where K is the Delaunay triangulation of S

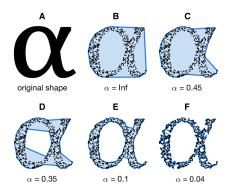


Figure 7: α -Complexes, for varying α

Filtration (to be continued)

aaaa

lpha-Shapes are closely related to lpha-complexes, subcomplexes of the Delaunay triangulation of the point set.

Each edge or triangle of the Delaunay triangulation may be associated with a characteristic radius, the radius of the smallest empty circle containing the edge or triangle.

For each real number α , the α -complex of the given set of points is the simplicial complex formed by the set of edges and triangles whose radii are at most $1/\alpha$.

Some implementation

Introduction

Qhull computes the convex hull, Delaunay triangulation, Voronoi diagram, halfspace intersection about a point, furthest-site Delaunay triangulation, and furthest-site Voronoi diagram

- The source code runs in 2-d, 3-d, 4-d, and higher dimensions
- Qhull implements the Quickhull algorithm for computing the convex hull
- It handles roundoff errors from floating point arithmetic
- It computes volumes, surface areas, and approximations to the convex hull.

http://www.qhull.org

2D example

```
julia> using QHull
julia > p2 = rand(10,2)
10×2 Array(Float64.2):
                                                         0.103684 0.661002
0.59823 0.964113
                                                         0.152175 0.405152
0.500987 0.656277
                                                         0.380007 0.180538
                                                         0.515186 0.288242
0.664168 0.566141
                                                         0.757414 0.20092
0.151405 0.639172
0.735322 0.198219
                                                        iulia> ch2.vertices
0.103684 0.661002
0.152175 0.405152
                                                        # indices to line segments forming the convex hull
0.380007 0.180538
                                                        6-element Array(Int64.1):
0.515186 0.288242
                                                          6
0.757414 0.20092
julia> ch2 = chull(p2)
Convex Hull of 10 points in 2 dimensions
Hull segment vertex indices:
[1, 6, 7, 8, 5, 10]
Points on convex hull in original order:
                                                        julia> ch2.simplices
[0.59823 0.964113; 0.735322 0.198219; ...; ...]
                                                        # the simplexes forming the convex hull
                                                        6-element Array{Array{Int64,1},1}:
                                                        [1, 6]
julia> ch2.points
                          # original points
10×2 Array{Float64,2}:
                                                         [1, 10]
0.59823 0.964113
                                                         [7, 6]
                                                         [7, 8]
0.500987 0.656277
0.664168 0.566141
                                                         [5, 10]
0.151405 0.639172
                                                         [5, 8]
0.735322 0.198219
```

3D example

```
julia> using QHull
julia > p3 = rand(20,3)
iulia > ch3 = chull(p3)
Convex Hull of 20 points in 3 dimensions
Hull segment vertex indices:
[1, 2, 5, 6, 7, 9, 10, 11, 12, 13, 14, 15, 19]
Points on convex hull in original order:
[0.604637 0.0159875 0.0282537; 0.203942 0.883561 0.665614; ...;
0.733554 0.946853 0.736748; 0.689781 0.969748 0.688725]
julia> ch3.vertices'
1×13 LinearAlgebra.Adjoint{Int64.Array{Int64.1}}:
1 2 5 6 7 9 10 11 12 13 14 15 19
iulia> @show p3
p3 = [0.604637 \ 0.0159875 \ 0.0282537; \ 0.203942 \ 0.883561 \ 0.665614;
0.0998071 0.583693 0.387648; 0.322411 0.271389 0.385281; 0.0854707
0.867559 0.407252: 0.544968 0.0609553 0.812328: 0.464687 0.905105
0.74164: 0.588225 0.549453 0.371593: 0.0359141 0.300891 0.357064:
0.138973 0.28496 0.127025; 0.732241 0.90114 0.0442693; 0.684981
0.0743128 0.427316; 0.598535 0.635914 0.881108; 0.623852 0.944785
0.413215; 0.733554 0.946853 0.736748; 0.598131 0.633203 0.293046;
0.37739 0.237786 0.397549: 0.474956 0.483842 0.66596: 0.689781
0.969748 0.688725; 0.487894 0.385103 0.604854]
```

```
julia> @show ch3.simplices;
ch3.simplices = Array{Int64,1}[[3, 8, 7], [3,
[18, 8, 7], [16, 8, 14], [5, 18, 7], [4, 3, 1
18, 11], [9, 18, 8], [9, 16, 11], [9, 16, 8],
13], [20, 13, 11], [20, 16, 11], [20, 12, 13
12, 14], [1, 13, 7], [1, 5, 7], [1, 5, 13],
13], [19, 18, 11], [19, 5, 18]]
```

Convex hull (2D)

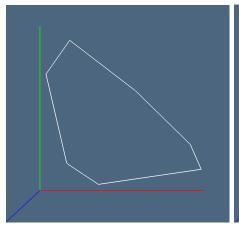
```
using LinearAlgebraicRepresentation, Plasm
Lar = LinearAlgebraicRepresentation

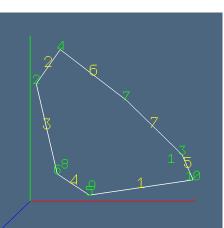
V2 = convert(Lar.Points, p2')
FV2 = ch2.simplices
Plasm.view(V2, FV2)

Plasm.view(Plasm.numbering(0.5)((V2, [[[k]
for k=1:size(V2,2)], FV2])))
```

Convex hull (2D)

Visualization





Convex hull (3D)

```
V3 = convert(Lar.Points, p3')
FV3 = ch3.simplices
Plasm.view(V3, FV3)

EV3 = Lar.simplexFacets(FV3)

Plasm.view( Plasm.numbering(0.25)((V3, [[[k] for k=1:size(V3,2)], EV3, FV3])) )
```

Convex hull (3D)

Visualization

