compGeometeR: an R package for computational geometry

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Abstract

Computational geometry algorithms and data structures are widely applied across numerous scientific domains, and there a variety of R packages that apply computational geometry. However, these packages often work in specific numbers of dimensions, have incompatible data structures, and include additional non-computational geometry functionality that can be domain specific. Our objective in developing the compGeometeR package is to produce a systematic implementation of the most commonly used computational geometry algorithms so that they can be easily integrated into domain specific scientific workflows. We briefly explain the algorithms available in compGeometeR, and identify priorities for future development.

Keywords: alpha complex, alpha shape, convex hull, convex layers, Delaunay triangulation, digital, discrete.

1 Introduction

Geometry is an ancient field of study in which geometer examines the distance, shape, size, and position of objects in space (Gowers, 2003). Computational geometry is a more recent field of study that emerged alongside technological developments in computing. The two main area of computation geometry are algorithmic and Numerical geometry. The algorithmic geometry seeks to develop efficient algorithms and data structures for geometric objects, and these algorithms and data structures have been found to be widely applicable across numerous scientific domains (De Berg et al., 2008), while the numerical geometry is aim to represent real-world objects which are suitable for computer computations (Kimmel, 2012). The algorithmic geometry is made up of the discrete and digital geometry. The discrete algorithmic geometry study the shape and methods of discrete geometric objects, while the digital algorithmic geometry deals with discrete sets by studying the geometric properties of a grid (or lattice) of points (Boissonnat and Yvinec, 1998).

Given the importance of computational algorithmic geometry it is no surprise that within the R (R Core Team, 2019) computing ecosystem there are a variety of options for applying various computational geometry algorithms. However, these packages often work in specific numbers of dimensions, have incompatible data structures, and include additional non-computational geometry functionality that can be domain specific. Also, these packages contains only one type of computational geometry algorithmic, which is the discreet geometry. The digital part of the computational geometry algorithmic are usually not included and it is up to the user to further derived the needed data from the output of the discreet geometry (see Table 1).

Our objective in developing the compGeometeR (a computational geometer using R) package is to produce a systematic implementation of some of the most commonly used computational geometry algorithms (Discrete and Digital) across various of branches of geometry so that they can be easily integrated into domain specific scientific workflows. Our hope is that by producing compGeometeR we can encourage the use of geometry in R for scientific study.

2 Algorithms

There are a variety of computational geometry algorithms available in compGeometeR, some of which are also available in other R packages. By documenting these differences (Table 1) we hope to help potential users to assess if compGeometeR is the best option for their needs as there may be other options that are more suitable.

The functions of compGeometeR interconnect and build upon one another, but it is important to recognise that virtually all of the compGeometeR functions are dependent on computational geometry foundations provided by the Qhull C++ interface software (Barber et al., 1996, http://www.qhull.org/). Next, we introduce the discrete and digital algorithms included in compGeometeR

2.1 Discrete geometry

Discrete geometry studies the shapes and interactions between sets of discrete objects such as points, lines, circles, and polygons in 2-dimensional Euclidean space – and the hyperdimensional equivalents of these objects in n-dimensional Euclidean space. All of the discrete computational geometry algorithms in compGeometeR take as an input a set of points P in n-dimensional Euclidean space \mathbb{R}^n .

A convex hull (Barber et al., 1996) defines the smallest subset of \mathbb{R}^n that contains P and for which the subset is convex so any two points within the convex hull can be connected by a straight line also contained by the convex hull (Figure 2a). As an example of the simplicity of compGeometeR a convex hull can be generated and visualised with minimal code (Listing 1).

```
library(compGeometeR)

# Generate point data

set.seed(2) # to reproduce figure exactly

x = rgamma(n = 20, shape = 3, scale = 2)

y = rnorm(n = 20, mean = 10, sd = 2)

p = cbind(x, y)

# Create convex hull

ch = convex_hull(p)
```

Table 1: Computational geometry software options in R. For each computational geometry function, the R software (and version) that can apply this algorithm is noted by reference to the number of dimensions in which the algorithm will work.

	$\texttt{compGeometeR} \; (1.0.0)$	R (3.5.3)	alphahull (2.2)	alphashape3d $(1.3.1)$	$\mathtt{deldir}\;(0.1\text{-}25)$	$\texttt{geometry}\ (0.4.5)$	$\mathtt{spatstat}\ (1.63\text{-}3)$	${ m tripack} \; (1.3 9.1)$
Discrete geometry								
Alpha complex	n							
Alpha shape			2	3				
Convex hull	n	2				n	2	2
Convex layers	n							
Delaunay triangulation	n		2		2	n	2	2
Voronoi diagram			2		2		2	2
Digital geometry								
Alpha complex	n							
Alpha shape	n							
Convex hull	n							

```
# Plot point data and convex hull
plot(x, y, yaxt="n", xaxt="n", xlab="", ylab = "", pch=16, cex=0.75)
polygon(ch$hull_vertices, col="orange", border="firebrick")
points(p, pch=16, cex=0.75)
```

Listing 1: Example R code to create a discrete convex hull with compGeometeR

Convex layers were first presented by Huber (1972) and Barnett (1976) who both gave unreferenced credit for this idea to Tukey. Convex layers are a nested sequence of convex hulls produced by repeating the process of constructing a convex hull for P and then removing the points forming the vertices of the convex hull from P before producing the next convex layer. The first convex layer is equivalent to the convex hull, with each successive convex layer representing ever smaller region of space (Figure 2b).

The Delaunay triangulation (Delaunay, 1934) produces a set of simplices (triangles in 2-dimensions or tetrahedrons in 3-dimensions) for which no point in P is inside the the circumhypersphere (a circle in 2-dimensions or a sphere in 3-dimensions) of each simplex (Figure 2c).

The alpha complex (Edelsbrunner and Mücke, 1994) is a subset of the Delaunay triangulation that contains only those simplices for whose circumhypersphere radius is smaller than a specified α parameter value ranging $0 < \alpha < \infty$ (Figure 2d). The related alpha shape would be a polytope that combines all of the simplices in an alpha complex (Edelsbrunner and Mücke, 1994).

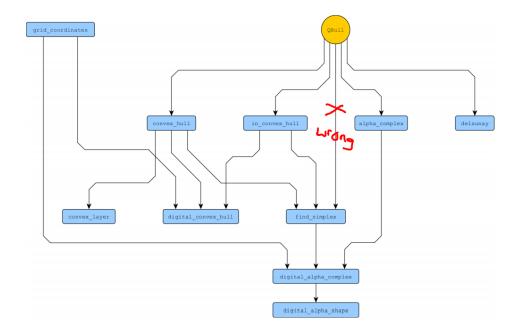


Figure 1: Connections and dependencies of the compGeometeR functions.

2.2 Digital geometry

Digital geometry is a branch of geometry that studies the geometric properties of a grid (or lattice) of points in Euclidean space, and usually involves the digitisation of discrete lines and regions (Rosenfeld and Melter, 1989). Digitisation occurs by representing Euclidean space \mathbb{R}^n as a rectangular orthogonal grid \mathbb{G}^n . The elements of \mathbb{G}^2 are called pixels, and the elements of \mathbb{G}^3 are called voxels, and each element in \mathbb{G}^n has a grid coordinate for its centre (Klette and Rosenfeld, 2004) and a value that denotes if the element belongs to a digitised geometric object, or when required, which part of the digitised geometric object. compGeometeR has implemented digital versions of the convex hull (Figure 3a), alpha complex (Figure 3b), and alpha shape (Figure 3c) that are unique functions amongst R software (Table 1).

The code required for digital versions of the discrete algorithms is no more complicated, and simply requires additional parameters to define the extent and resolution of \mathbb{G}^n . For example, the code required for a digital convex hull (Listing 2) does not differ much from that required for a discrete convex hull (Listing 1).

Listing 2: Example R code to create a digital convex hull with compGeometeR

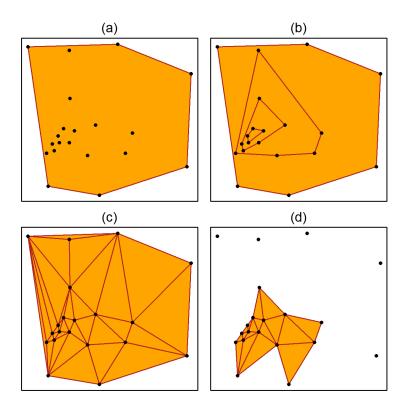


Figure 2: Two-dimensional examples of discrete geometry algorithms currently available in compGeometeR. (a) Convex hull, (b) convex layers, (c) Delaunay triangulation, and (d) alpha complex.

3 Future work

compGeometeR is very much a work in progress, and while there is sufficient functionality for it to be of wide use in the computational sciences, there is some functionality that has been identified as being particularly useful for future development.

The Voronoi diagram (Voronoi, 1908; Okabe et al., 2000) partitions n-dimensional space into a set of polytopes for which each polytope delineates the region of n-dimensional space that is closest to each point in P. The Voronoi diagram would be an obvious addition for the discrete geometry algorithms given its lack of implementation in other R geometry packages (see Table 1).

As well as expanding the number of discrete geometry algorithms, and implementing more discrete algorithms in digital form, there are other kinds of computational geometry algorithms that we think would be of particular value.

There are a variety of useful graph structures that can be produced based on geometric principles. For example, the Delaunay triangulation (Delaunay, 1934) can also be represented as a graph structure, and for which there are several useful subgraphs such as the Gabriel graph (Gabriel and Sokal, 1969), Urquhart graph (Urquhart, 1980), and relative neighbourhood graph (Toussaint, 1980).

In some contexts it will be important to recognise that there is uncertainty in the position of points in space and the parameters defining a given algorithm. In such situations it may be helpful to adopt fuzzy geometry (Rosenfeld, 1998) view in which, foe example, memberships of points in space are not crisp being either 0 or 1, but rather are fuzzy on a scale from 0

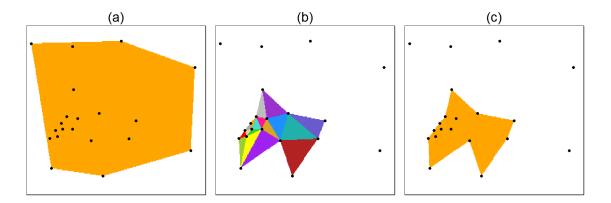


Figure 3: Two-dimensional examples of digital geometry algorithms currently available in compGeometeR. (a) Convex hull, (b) alpha complex, and (c) alpha shape.

to 1. Creating fuzzy versions of the compGeometeR algorithms could result in very useful functionality where quantifying uncertainty of any computational geometry is important. The existing digital geometric algorithms could be easily extended to fuzzy forms as all that is required is to assign a membership value to each pixel (Klette and Rosenfeld, 2004).

All the algorithms in compGeometeR work in Euclidean space, but computational geometry algorithms can also be usefully applied in spherical or hyperbolic space (Gowers, 2003). For example, the Fortran STRIPACK software (Renka, 1997) could be wrapped by R to provide functionality to compute Delaunay triangulations and Voronoi diagrams on the surface of a sphere.

4 Software availability

compGeometeR is open source software made available under a General Public License license. Installation instructions can be found at the GitHub repository https://github.com/manaakiwhenua/compGeometeR.

We are also developing a cookbook of examples of compGeometeR use via the GitHub repository wiki https://github.com/manaakiwhenua/compGeometeR/wiki.

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