

Computational Geometry Software

- 1 Computational Geometry
 - problems and software
 - getting started
- 2 Solving Geometric Problems
 - convex hulls
 - Voronoi diagrams
 - Delaunay triangulations
- 3 the Lake Superior Polygon
 - convex hull and Delaunay triangulation
 - improved Delaunay triangulations

MCS 507 Lecture 30
Mathematical, Statistical and Scientific Software
Jan Verschelde, 30 October 2023

Computational Geometry Software

1 Computational Geometry

- problems and software
- getting started

2 Solving Geometric Problems

- convex hulls
- Voronoi diagrams
- Delaunay triangulations

3 the Lake Superior Polygon

- convex hull and Delaunay triangulation
- improved Delaunay triangulations

computational geometry

Computational geometry studies algorithms for geometric problems.

We distinguish three classes of problems:

- **Static** : compute an object from given inputs.
Examples: convex hulls, triangulations, Voronoi diagrams.
- **Query** : given a search space and a query,
find objects in the search space that satisfy the query.
- **Dynamic** : similar problems as static and query,
but the input changes and objects are inserted and deleted.

Applications : finite element methods, robot motion planning, etc.

computational geometry software

We follow *Mastering SciPy* by Francisco J. Blanco-Silva, Packt Publishing 2015.

We introduce three software packages:

- The `spatial` package of `scipy` for spatial algorithms and data structures: queries, Delaunay triangulations, convex hulls, and Voronoi diagrams.
- Mayavi: 3D scientific data visualization and plotting, provides `mlab`, for 3D plotting in Python.
- Triangle: a mesh generator by Jonathan Richard Shewchuck, used through the wrapper written by Dzhelil Rufat, available at <http://dzhelil.info/triangle/index.html>.

The Computational Geometry Algorithms Library

<http://www.cgal.org>

CGAL is a software project that provides easy access and reliable geometric algorithms in the form of a C++ library. CGAL provides an extensive collection of examples.

Its development is funded by academia and companies, dual licensing:

- open source: Lesser GPL (LGPL) and GPL;
- a commercial license from the GeometryFactory.

Installation:

- CGAL is available via package managers.
- There is a Python interface, install with pip or conda.

CGAL Made More Accessible, by Nir Goren, Efi Fogel, and Dan Halperin, [arxiv:2202.13889v2](https://arxiv.org/abs/2202.13889v2) [cs.CG] 7 Jun 2023.

Computational Geometry Software

1 Computational Geometry

- problems and software
- **getting started**

2 Solving Geometric Problems

- convex hulls
- Voronoi diagrams
- Delaunay triangulations

3 the Lake Superior Polygon

- convex hull and Delaunay triangulation
- improved Delaunay triangulations

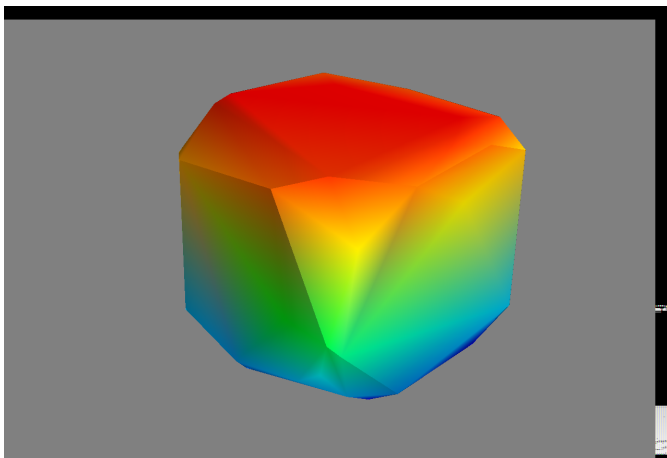
plotting the convex hull of random points

The script below shows the output of the ConvexHull of `scipy.spatial` on a random set of 320 points in 3-space using Mayavi.

```
import numpy as np
from scipy.spatial import ConvexHull
from mayavi import mlab

points = np.random.rand(320, 3)
hull = ConvexHull(points)
X = hull.points[:, 0]
Y = hull.points[:, 1]
Z = hull.points[:, 2]
mlab.triangular_mesh(X, Y, Z, hull.simplices)
mlab.show()
```

the plot of a convex hull in 3-space



Clicking on the picture allows for a change of view point.

Computational Geometry Software

1 Computational Geometry

- problems and software
- getting started

2 Solving Geometric Problems

- **convex hulls**
- Voronoi diagrams
- Delaunay triangulations

3 the Lake Superior Polygon

- convex hull and Delaunay triangulation
- improved Delaunay triangulations

the convex hull problem

Problem statement:

Input: a set S of points.

Output: the smallest convex set that contains S .

We distinguish between the planar and the spatial case.

- The convex hull in the plane consists of vertices and edges.
- In three dimensions, we have vertices, edges, and simplices.

The `ConvexHull` of `scipy.spatial` computes the convex hull with the Qhull library, see <http://www.qhull.org>.

C. Bradford Barber, David P. Dobkin, and Hannu Huhdanpaa.

The Quickhull algorithm for convex hulls.

ACM Trans. Math. Softw. 22(4):469-483, 1996.

the convex hull of 20 random points in the plane

```
import numpy as np
from scipy.spatial import ConvexHull
from scipy.spatial import convex_hull_plot_2d
from matplotlib import pyplot as plt

points = np.random.rand(20, 2)
hull = ConvexHull(points)
print('The vertices :')
print(hull.vertices)
print('The edges :')
for simplex in hull.simplices:
    print(simplex)
convex_hull_plot_2d(hull)
plt.show()
```

output of the script

The vertices :

```
[ 3  5  6 12  1 18 16]
```

The edges :

```
[3 5]
```

```
[18 1]
```

```
[12 1]
```

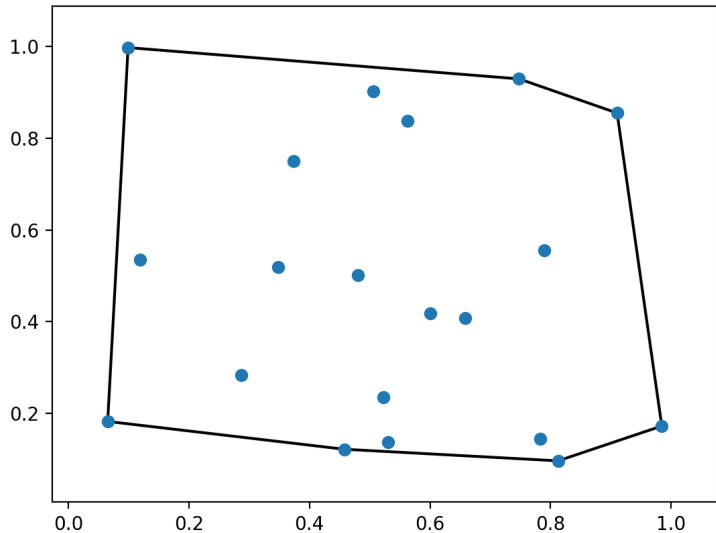
```
[16 3]
```

```
[16 18]
```

```
[6 5]
```

```
[ 6 12]
```

the plot of a convex hull in the plane



Euler's formula

Euler's formula states: $\#vertices - \#edges + \#faces = 2$.

Let us verify this formula for some random points.

```
import numpy as np
from scipy.spatial import ConvexHull

points = np.random.rand(320, 3)
hull = ConvexHull(points)
nv = hull.vertices.size
ns = hull.simplices.shape[0]
print('number of vertices :', nv)
print('number of simplices :', ns)
```

computation of all edges

```
edges = []
for simplex in hull.simplices:
    x, y, z = simplex
    edge = ((x, y) if x < y else (y, x))
    if edge not in edges: edges.append(edge)
    edge = ((y, z) if y < z else (z, y))
    if edge not in edges: edges.append(edge)
    edge = ((z, x) if z < x else (x, z))
    if edge not in edges: edges.append(edge)
ne = len(edges)
print('number of edges :', ne);
print('nv - ne + ns :', nv - ne + ns)
```

Computational Geometry Software

1 Computational Geometry

- problems and software
- getting started

2 Solving Geometric Problems

- convex hulls
- **Voronoi diagrams**
- Delaunay triangulations

3 the Lake Superior Polygon

- convex hull and Delaunay triangulation
- improved Delaunay triangulations

Voronoi diagrams

Problem statement:

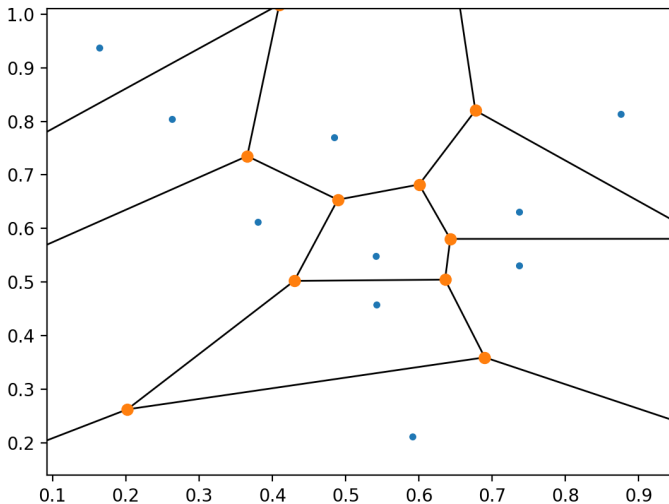
Input: a set S of points (sites) in the plane.

Output: a partition of the plane in cells, so that all points in the same cell are closest to the same site.

```
import numpy as np
from scipy.spatial import Voronoi, voronoi_plot_2d
from matplotlib import pyplot as plt

points = np.random.rand(10, 2)
vor = Voronoi(points)
voronoi_plot_2d(vor)
plt.show()
```

the Voronoi diagram of 10 random sites



the attributes of a Voronoi diagram

```
vor = Voronoi(points)
print('the input points :')
print(vor.points)
print('the vertices :')
print(vor.vertices)
print('the ridge vertices :')
print(vor.ridge_vertices)
print('the ridge points :')
print(vor.ridge_points)
print('the cells :')
for region in vor.regions:
    print(region)
print('the point_region :')
print(vor.point_region)
```

Computational Geometry Software

1 Computational Geometry

- problems and software
- getting started

2 Solving Geometric Problems

- convex hulls
- Voronoi diagrams
- **Delaunay triangulations**

3 the Lake Superior Polygon

- convex hull and Delaunay triangulation
- improved Delaunay triangulations

Delaunay triangulations

Problem statement:

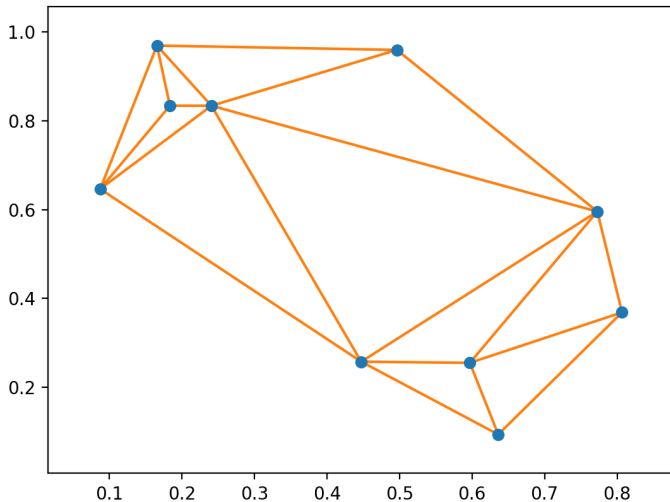
Input: a set S of points in the plane.

Output: the triangulation that maximizes the minimal angle over all triangulations of S .

```
import numpy as np
from scipy.spatial import Delaunay, delaunay_plot_2d
from matplotlib import pyplot as plt

points = np.random.rand(10, 2)
tri = Delaunay(points)
delaunay_plot_2d(tri)
plt.show()
```

the Delaunay triangulation of 10 random points



some attributes of a Delaunay triangulation

```
points = np.random.rand(10, 2)
tri = Delaunay(points)
print('the input points :')
print(tri.points)
print('indices of the simplices :')
for simplex in tri.simplices:
    print(simplex)
print('indices to neighbor simplices :')
for neighbor in tri.neighbors:
    print(neighbor)
print('the equations :')
for equation in tri.equations:
    print(equation)
```

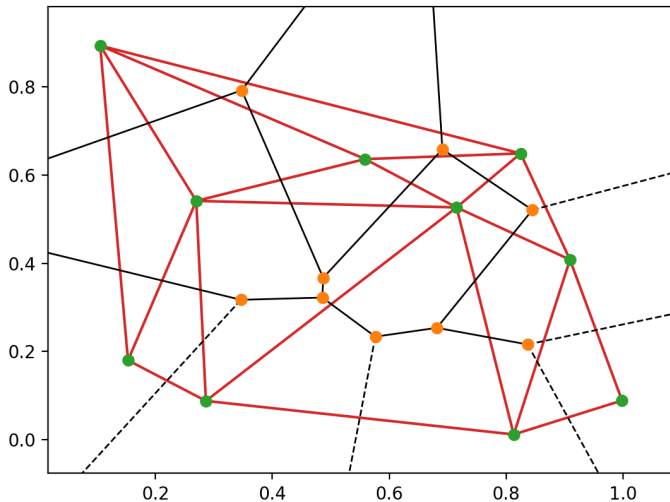
duality

The dual of a Voronoi diagram is a Delaunay triangulation and the dual of a Delaunay triangulation is a Voronoi diagram.

```
import numpy as np
from scipy.spatial import Voronoi, voronoi_plot_2d
from scipy.spatial import Delaunay, delaunay_plot_2d
from matplotlib import pyplot as plt

points = np.random.rand(10, 2)
vor = Voronoi(points)
tri = Delaunay(points)
fig, axs = plt.subplots()
voronoi_plot_2d(vor, ax=axs)
delaunay_plot_2d(tri, ax=axs)
plt.show()
```

Voronoi diagram and Delaunay triangulation



Computational Geometry Software

1 Computational Geometry

- problems and software
- getting started

2 Solving Geometric Problems

- convex hulls
- Voronoi diagrams
- Delaunay triangulations

3 the Lake Superior Polygon

- convex hull and Delaunay triangulation
- improved Delaunay triangulations

the Lake Superior Polygon

Following the *Mastering Scipy* book of Francisco J. Blanco-Silva, Packt Publishing 2015, we examine the Lake Superior polygon, available for download at

```
https://github.com/blancosilva/Mastering-Scipy/blob/master/chapter6/superior.poly.
```

This polygon has 7 holes (for the islands), 518 vertices, and 518 edges.

The .poly format comes from the Triangle software, available at <http://www.cs.cmu.edu/~quake/triangle.html>

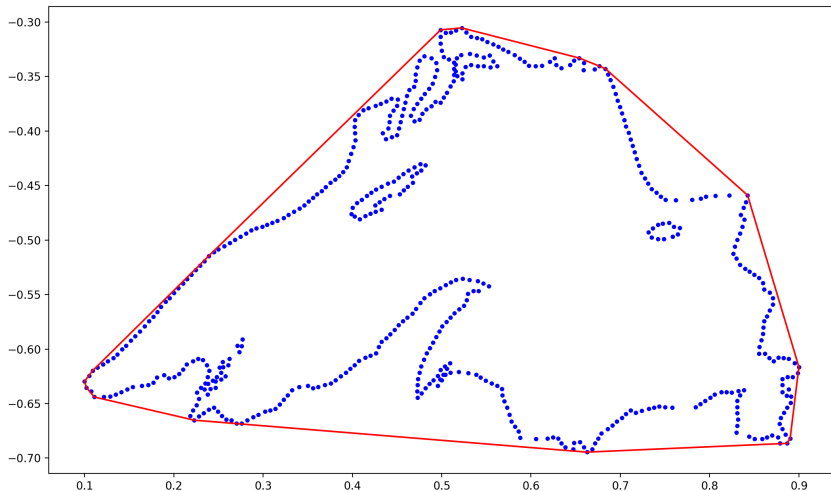
We use the `read_poly()` function of *Mastering SciPy*.

plotting the convex hull

```
from read_poly import read_poly
from matplotlib import pyplot as plt
from scipy.spatial import ConvexHull, convex_hull_plot_2d

lake_superior = read_poly('superior.poly')
points = lake_superior['vertices']
print('the number of vertices : ', end='')
print(points.size)
hull = ConvexHull(points)
plt.plot(points[:,0], points[:,1], 'b.')
for simplex in hull.simplices:
    plt.plot(points[simplex, 0], points[simplex, 1], 'r-')
plt.show()
```

the convex hull

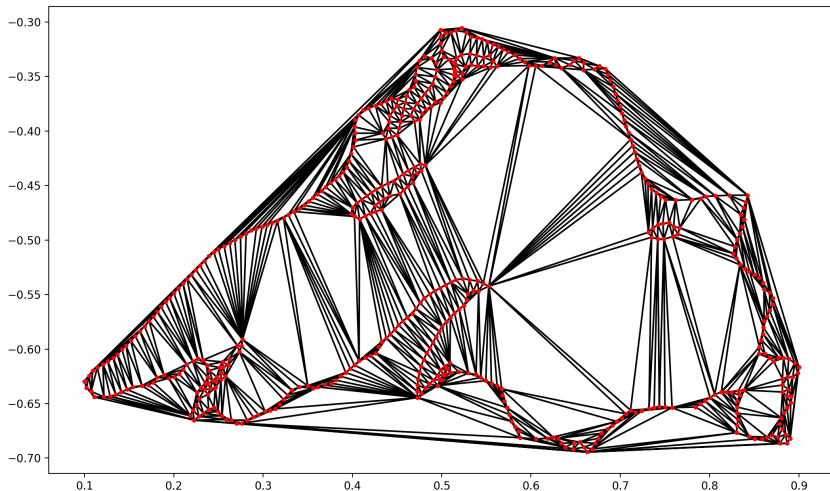


computing the Delaunay triangulation

```
from read_poly import read_poly
from matplotlib import pyplot as plt
from scipy.spatial import Delaunay

lake_superior = read_poly('superior.poly')
points = lake_superior['vertices']
print('the number of vertices : ', end='')
print(points.size)
tri = Delaunay(points)
plt.triplot(points[:,0], points[:,1],
            tri.simplices, 'k-')
plt.plot(points[:,0], points[:,1], 'r.')
plt.show()
```

the Delaunay triangulation



Computational Geometry Software

1 Computational Geometry

- problems and software
- getting started

2 Solving Geometric Problems

- convex hulls
- Voronoi diagrams
- Delaunay triangulations

3 the Lake Superior Polygon

- convex hull and Delaunay triangulation
- improved Delaunay triangulations

prize winning software

The scripts in the following slides contain code of the *Mastering SciPy* book of Francisco J. Blanco-Silva, using the software Triangle.

Jonathan Richard Shewchuk, ***Triangle: Engineering a 2D Quality Mesh Generator and Delaunay Triangulator***, in *Applied Computational Geometry: Towards Geometric Engineering*, edited by Ming C. Lin and Dinesh Manocha, volume 1148 of Lecture Notes in Computer Science, pages 203-222, Springer-Verlag, Berlin, May 1996.

Received the 2003 James H. Wilkinson Prize for Numerical Software.

Usable in Python through wrappers written by Dzhelil Rufat, available via `pip install triangle`.

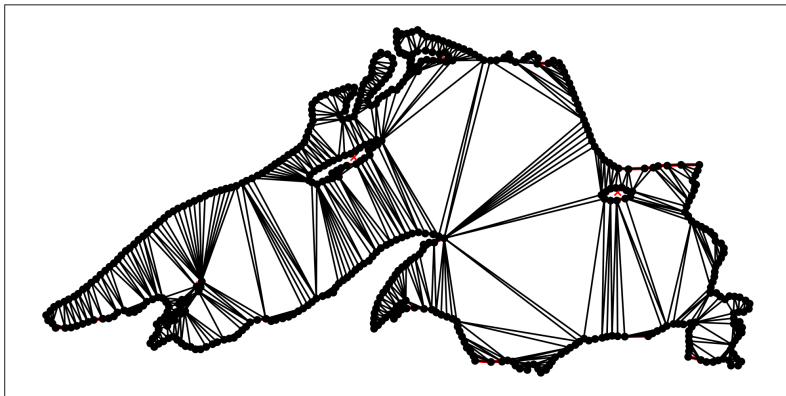
computing a constrained Delaunay triangulation

A constrained Delaunay triangulation (cndt) is computed with the flag `p` to indicate that the source is a planar straight line graph.

```
from read_poly import read_poly
from matplotlib import pyplot as plt
from triangle import triangulate
from triangle import plot as tplot

lake_superior = read_poly('superior.poly')
cndt = triangulate(lake_superior, 'p')
ax = plt.subplot(111)
tplot(ax, **cndt)
plt.show()
```

the constrained Delaunay triangulation



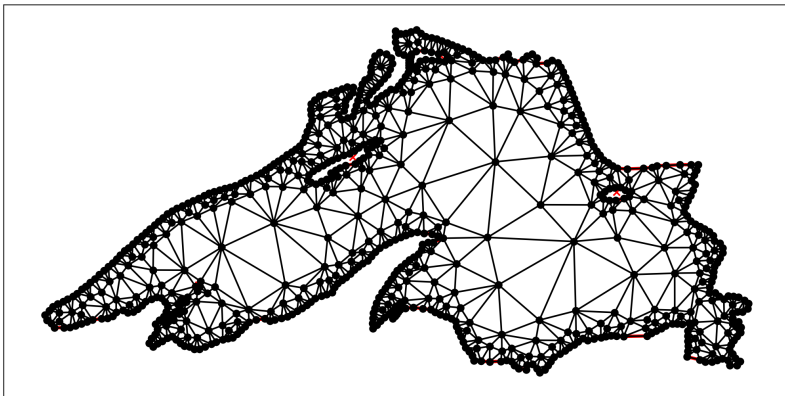
computing a constrained Delaunay triangulation

A constrained conforming Delaunay triangulation (cfdt) is computed with the flag `p` to indicate that the source is a planar straight line graph and the flag `D` to enforce Steiner points. Furthermore, we require all triangles to have a minimum angle of at least 20 degrees.

```
from read_poly import read_poly
from matplotlib import pyplot as plt
from triangle import triangulate
from triangle import plot as tplot

lake_superior = read_poly('superior.poly')
cncfq20dt = triangulate(lake_superior, 'pq20D')
ax = plt.subplot(111)
tplot(ax, **cncfq20dt)
plt.show()
```

the conforming Delaunay triangulation



imposing a maximum area on triangles

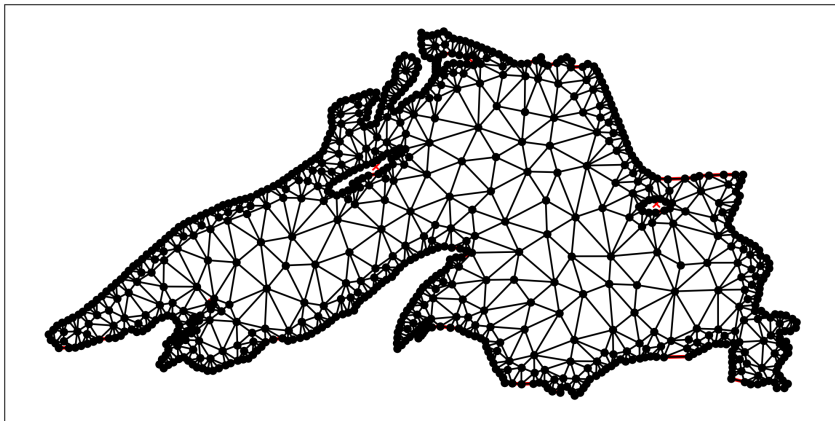
A constrained conforming Delaunay triangulation (cfdt) is computed with the flag `p` to indicate that the source is a planar straight line graph and the flag `D` to enforce Steiner points.

We furthermore require that all triangles have a minimum angle of at least 20 degrees and impose a maximum area on triangles.

```
from read_poly import read_poly
from matplotlib import pyplot as plt
from triangle import triangulate
from triangle import plot as tplot

lake_superior = read_poly('superior.poly')
cncfq20adt = triangulate(lake_superior, 'pq20a.001D')
ax = plt.subplot(111)
tplot(ax, **cncfq20adt)
plt.show()
```

constraints on degrees and areas



Exercises

- 1 Consider the cube spanned by the points in the list L .

$$L = \begin{bmatrix} [0, 0, 0], & [1, 0, 0], & [0, 1, 0], & [1, 1, 0], \\ [0, 0, 1], & [1, 0, 1], & [0, 1, 1], & [1, 1, 1] \end{bmatrix}$$

Apply `ConvexHull` to this list and compute the edges.

Verify Euler's formula.

Explain and give an interpretation for the results.

- 2 Consider five collinear points in the plane, compute and plot their Voronoi diagram. Describe and explain the plot.

- 3 A Delaunay triangulation can be defined as the projection of the lower convex hull of the points $(x, y, x^2 + y^2)$. Choose a good example to make a drawing to show this definition.