

# Computational Geometry

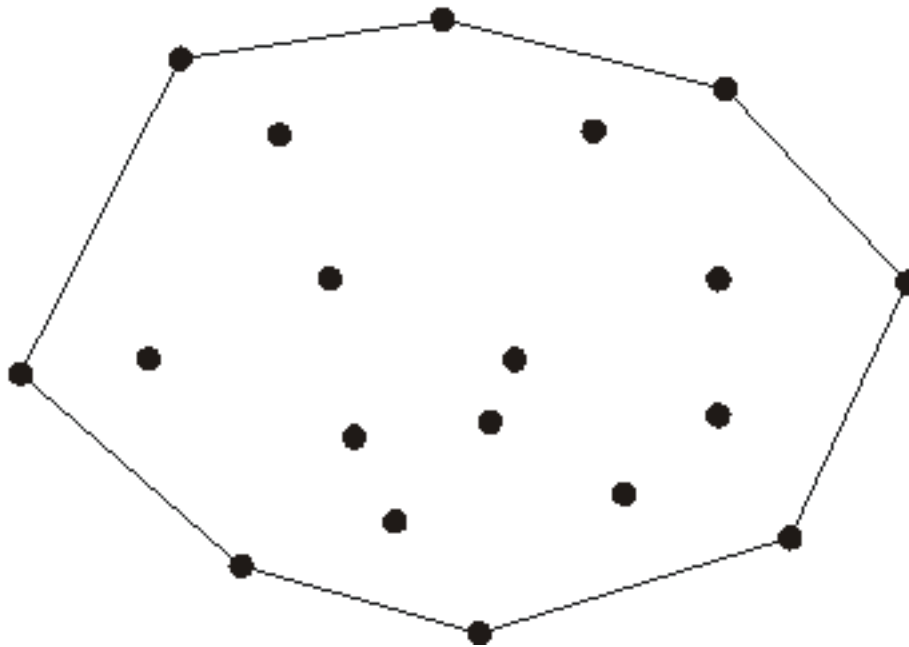
- Introduction
- Algorithms Classification
- Computational complexity
- Degeneracies and Robustness

# What is computational geometry ?

**Computational geometry** is a branch of the theory of computations that studies geometric problems of great size focusing on robust and asymptotically fast algorithms.

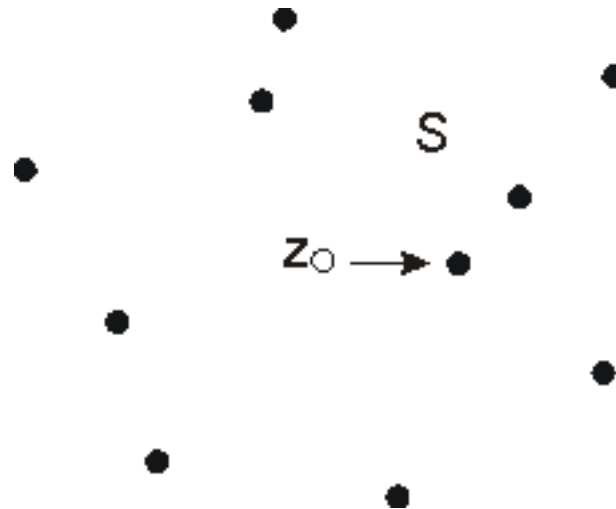
# Example

Given a set  $S$  of  $n$  points on the plane. Find the convex hull of  $S$ .



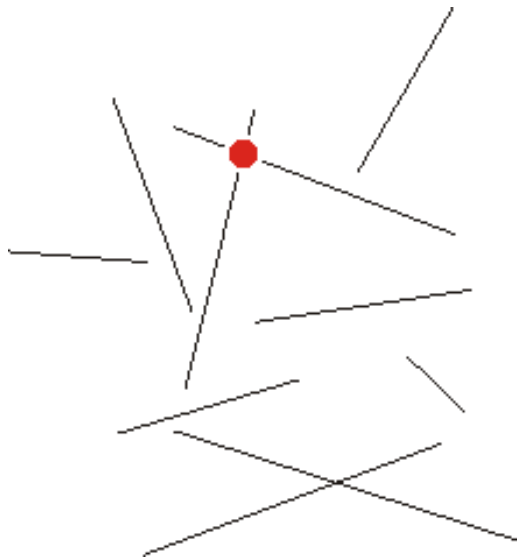
# Example

Given a set  $S$  of  $n$  points on the plane and a point  $z \notin S$ . Find a point in  $S$  closest to  $z$ .



# Example

Given a set of  $n$  line segments on the plane. Do any two segments intersect?



# Computational complexity

- Computational complexity of an **algorithm**: amount of time spending by the algorithm as a function of size of the problem. Usual notation is  $T(n)$ , where  $n$  denotes the size of the problem.
- Computational complexity of a **problem**: complexity of the best algorithm that solves this problem.

# Computational complexity

DETECT-SEGMENT-INTERSECTION ( $S$ )

```
1      for  $i \rightarrow 0$  to  $n - 1$  do  
2          for  $j \rightarrow i + 1$  to  $n$  do  
3              if (INTERSECTS( $S[i]$ ,  $S[j]$ ))  
4                  return true;  
5      return false;
```

$$T(n) = c_1 \frac{n(n-1)}{2} + c_2 n + c_3$$

# Asymptotic complexity

**Asymptotic complexity** is behavior of computational complexity  $T(n)$  if  $n$  approaches the infinity.

**Example:**

If for some constant  $C > 0$  and  $N \geq 0$  such that for all  $n \geq N$

$$T(n) = c_1 \frac{n(n-1)}{2} + c_2 n + c_3 \leq C \mathbf{n^2},$$

then an *asymptotic notation* is used

$$T(n) = O(\mathbf{n^2}).$$

The constants  $C$  and  $N$  can be estimated as

$$C = a + |b| + c, N = \left\lceil \frac{2|b|}{a} \right\rceil,$$

where  $a = c_1/2$ ,  $b = c_2 - a$ ,  $c = c_3$ .



# $O$ -notation

- A notation

$$T(n) = O(f(n))$$

means that there exist constants  $C > 0$  and  $N > 0$  such that for all  $n \geq N$

$$T(n) \leq Cf(n),$$

- By definition, it means that

$$\lim_{n \rightarrow \infty} \frac{T(n)}{f(n)} = 1.$$

- Is used to denote the **worst-case** of algorithm running time.
- We say that, in the worst-case, the algorithm will spend **at most**  $O(f(n))$  time to solve the problem.

# $\Omega$ -notation

- A notation

$$T(n) = \Omega(f(n))$$

means that there exist constants  $C > 0$  and  $N > 0$  such that for all  $n \geq N$

$$T(n) \geq Cf(n),$$

- Is used to denote the **lower-bound** of the complexity of a problem.
- We say that the best algorithm will need **at least**  $\Omega(f(n))$  time to solve this problem.

# $\theta$ -notation

- A notation

$$T(n) = \theta(f(n))$$

means that

$$T(n) = O(f(n)) = \Omega(f(n)).$$

- We say that  $f(n)$  is an **asymptotically tight** bound of  $T(n)$ .
- Is used to denote computational complexity of **optimal** algorithms for the specified problem.

# Growth of functions

$T(n)$	$T_1 = T(1000)$	$T_2 = T(1000000)$	$T_2/T_1$
$\log n$	10	20	2
$n$	$10^3$	$10^6$	$10^3$
$n \log n$	$10^4$	$2 \times 10^7$	$2 \times 10^3$
$n^2$	$10^6$	$10^{12}$	$10^6$
$2^n$	$10^{300}$	$10^{300000}$	$10^{299700}$

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**Linear** growth of complexity  
regarding growth of the input


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**1000 times** growth of complexity  
regarding growth of the input!

# Growth of functions

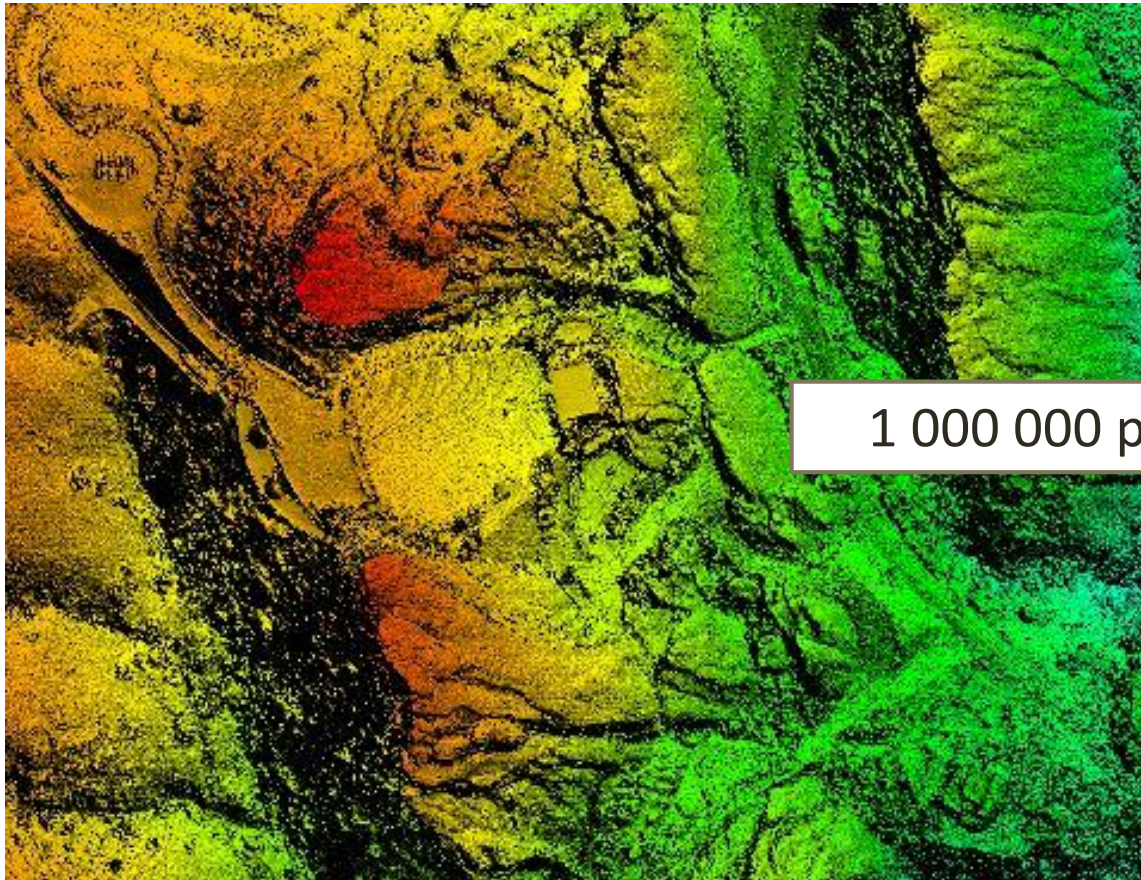
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**1000 times** growth of complexity regarding growth of the input!

Natural limitation for solving problems of great size

# Digital Terrain Model



1 000 000 polygons




# Stanford's Digital Michelangelo



2 000 000 000 polygons

# When and what kind of optimization is appropriate?

- $O(\log n)$ : **real time** request (milliseconds)
  - Search and display of the current information under mouse pointer, onMouseMove( ) event handler.
  - Analysis and display of permanently changing object status.
- $O(n)$ : mouse click request (0.5 – 1 sec).
- $O(n \log n)$ : short time computation (1 sec – few minutes).
- $O(n^k)$ : long computation (few hours – few days).
- $O(2^n)$ :  (years, centuries)

# Algorithm development: three steps

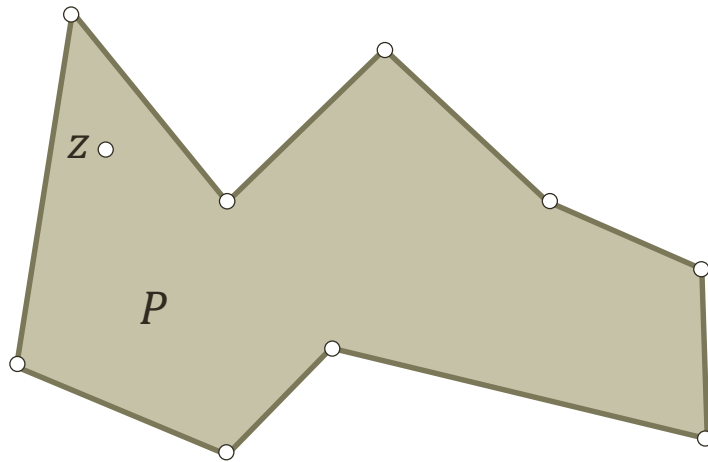
**Step 1.** Design an algorithm with the **best asymptotical complexity**.

**Step 2.** Handle **degeneracies**.

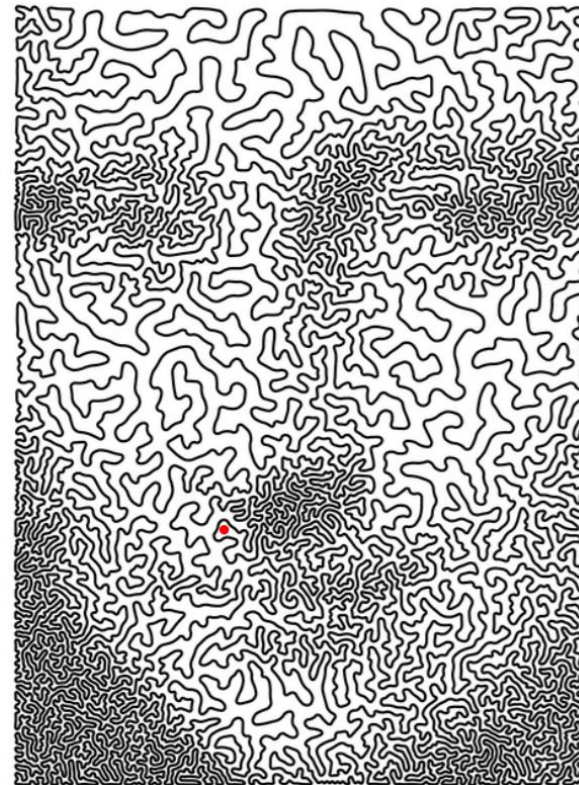
**Step 3.** Provide **computational robustness**.

# Algorithm development: example

Given a simple polygon  $P$  and a point  $z$ , identify whether  $z$  belongs to  $P$ .

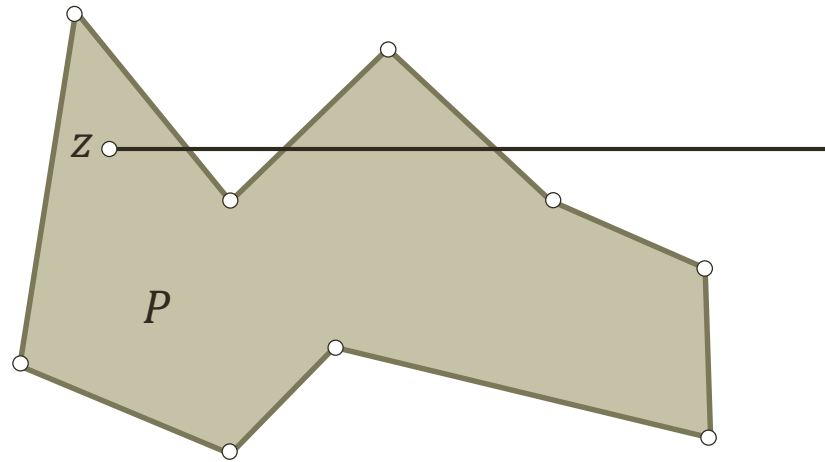


A simple polygon, as well...



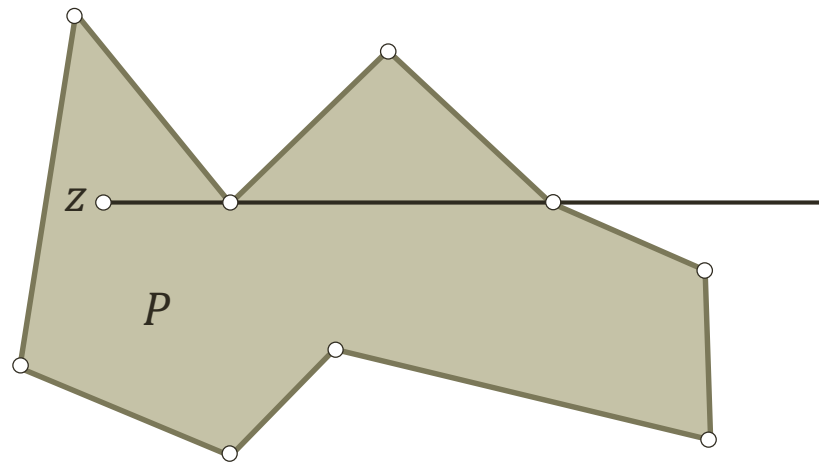
# Algorithm development: example

**Step 1:** design an algorithm with the best asymptotical complexity.



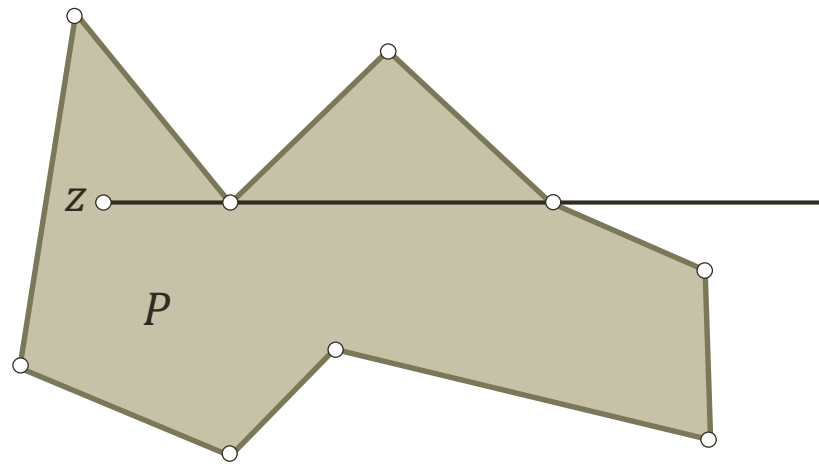
# Algorithm development: example

**Step 1:** design an algorithm with the best asymptotical complexity.



# Algorithm development: example

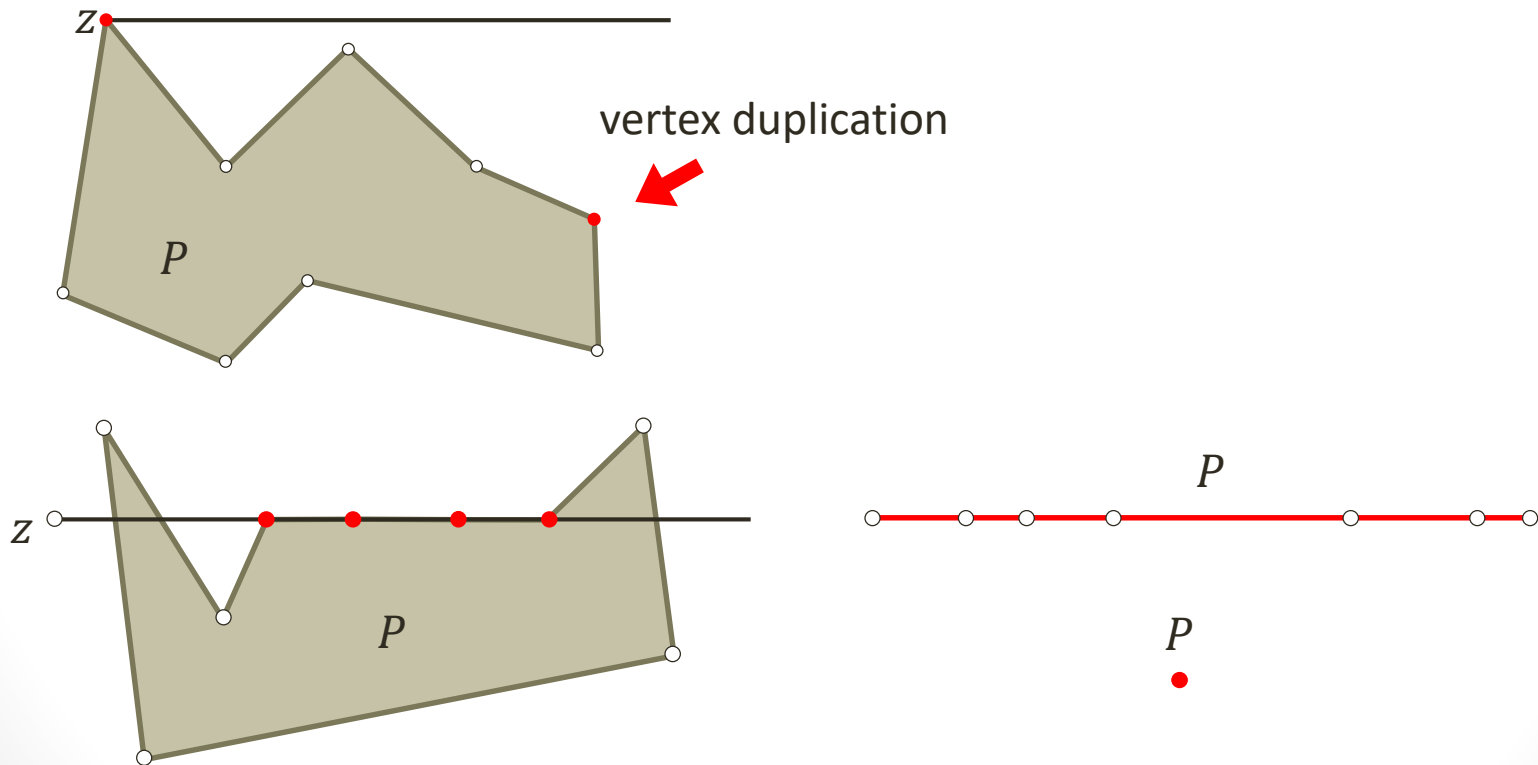
**Step 1:** design an algorithm with the best asymptotical complexity.



Asymptotical complexity:  $T(n) = \theta(n)$

# Algorithm development: example

**Step 2:** handle degeneracies.





# Algorithm development: example

**Step 3:** provide computational robustness.

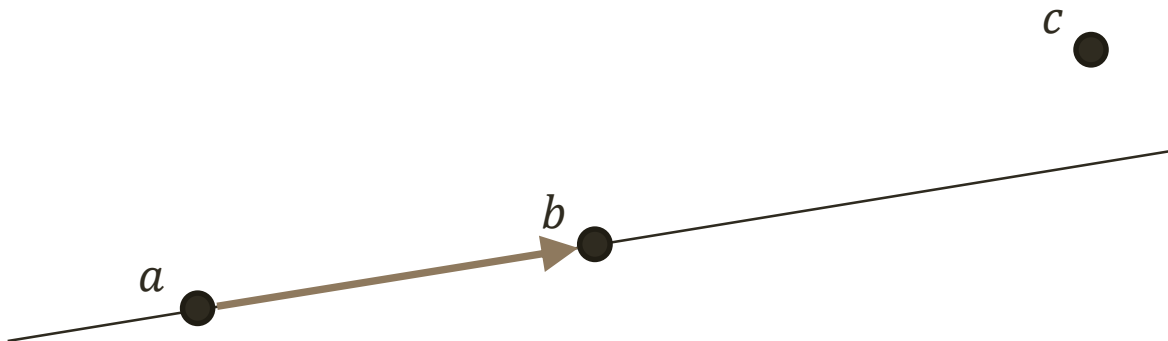
Rounding errors may **crucially impact on algorithm execution.**

# Geometric predicates

Identify location of a point  $c$  on the plane relative to an oriented line, that passes through points  $a$  and  $b$ .

In other words, are the points  $a$ ,  $b$  and  $c$  arranged in clockwise (CW) order, counterclockwise (CCW) order or they are collinear?

- The points are defined by their coordinates of type **double**.
- The solution must be correct for **any** input.



# Geometric predicates

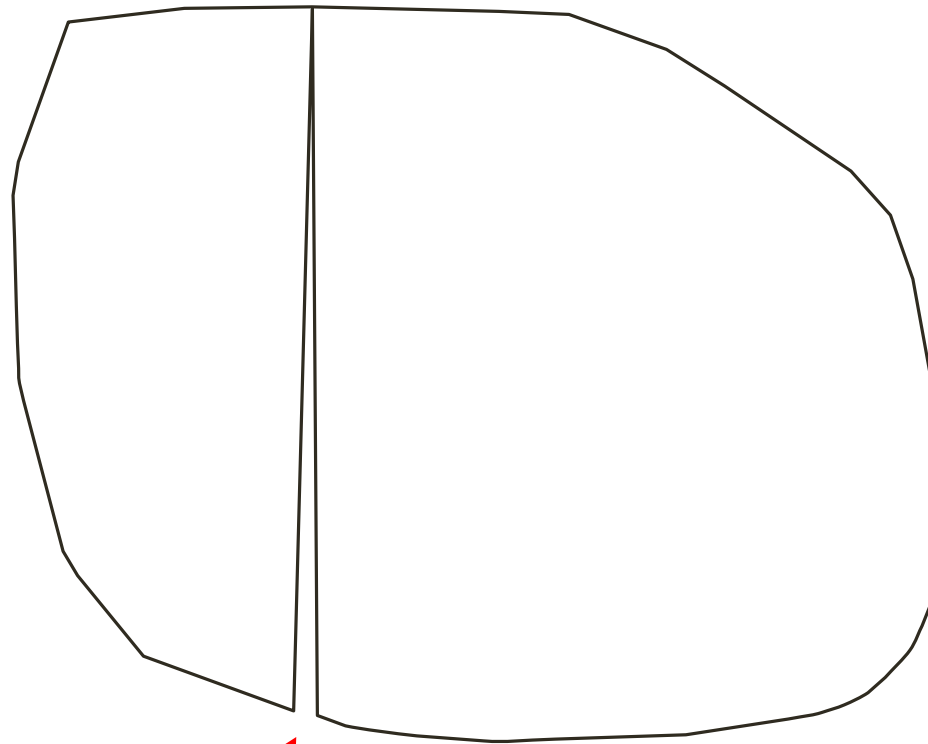
$$D = \begin{vmatrix} a_x & a_y & 1 \\ b_x & b_y & 1 \\ c_x & c_y & 1 \end{vmatrix} = \begin{vmatrix} a_x - c_x & a_y - c_y \\ b_x - c_x & b_y - c_y \end{vmatrix}$$

- $D > 0 \Rightarrow c$  is on the **left** of the line  $(a, b)$
- $D < 0 \Rightarrow c$  is on the **right** of the line  $(a, b)$
- $D = 0 \Rightarrow c$  is **on** the line  $(a, b)$

```
int orient2d(double ax, double ay, double bx, double by, double cx, double cy)
{
    return (ax - cx) * (by - cy) - (bx - cx) * (ay - cy);
}
```

# Geometric predicates

**Convex hull**

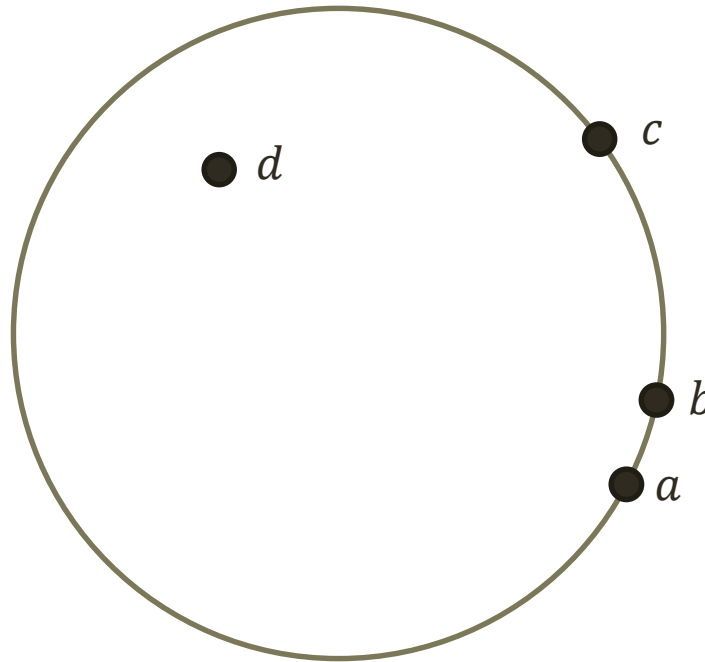


Incorrect identification of points  
arrangement caused by rounding error

# Geometric predicates

Identify location of a point  $d$  on the plane relative to a circle that passes through points  $a$ ,  $b$  and  $c$ .

In other words, is  $d$  **inside**, **outside** or **exactly on** the circle?



# Geometric predicates

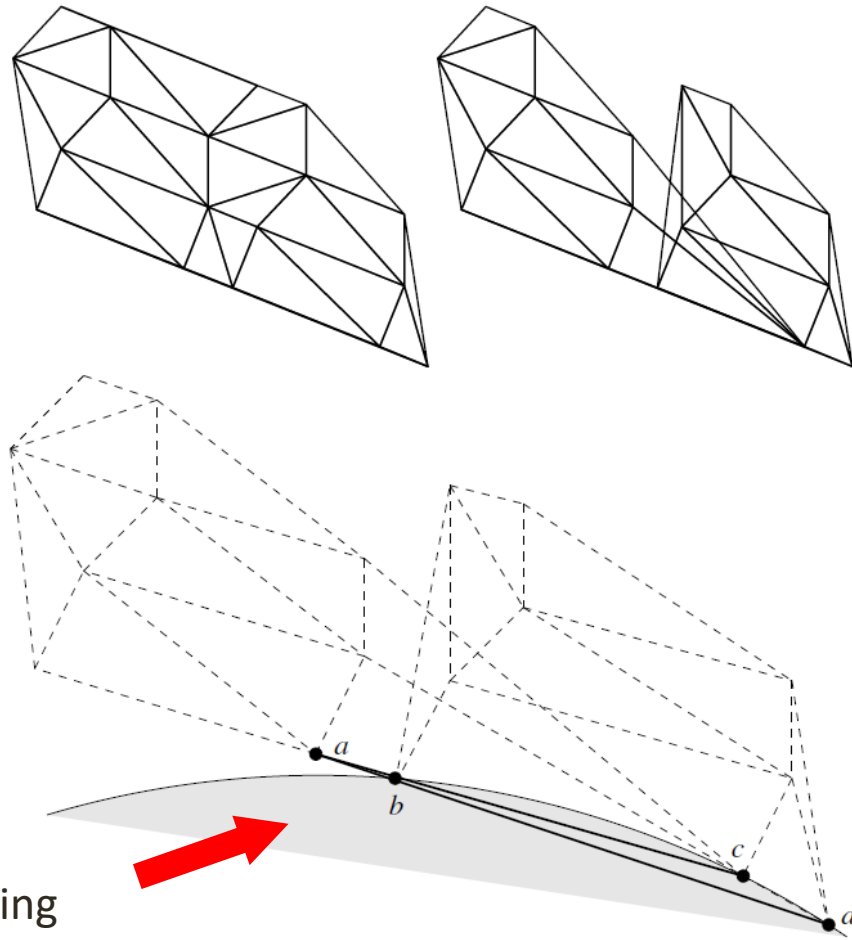
$$D = \begin{vmatrix} a_x & a_y & a_x^2 + a_y^2 & 1 \\ b_x & b_y & b_x^2 + b_y^2 & 1 \\ c_x & c_y & c_x^2 + c_y^2 & 1 \\ d_x & d_y & d_x^2 + d_y^2 & 1 \end{vmatrix}$$

- $D > 0 \Rightarrow d$  is inside the circle,
- $D < 0 \Rightarrow d$  is outside the circle,
- $D = 0 \Rightarrow d$  is on the circle

# Geometric predicates

## Triangulation

[Shewchuk 1999]



Incorrect identification of point location relative to a circle (the Delaunay test) caused by rounding errors

# Computational robustness

- Big numbers
- Exact arithmetic
- Adaptive arithmetic

A particular case: integer arithmetic

- point positions are specified by `__int32` (4 bytes),
- intermediate and final results are stored in `__int64` (8 bytes) and `__int128` (16 bytes).



# Computational robustness

- Big numbers
- Exact arithmetic
- Adaptive arithmetic

## Arbitrary precision floating-point arithmetic:

- a number  $x$  is expressed as an *expansion*

$$x = x_n + \cdots + x_2 + x_1,$$

- $|x_n| > \cdots > |x_1|$ ,
- components  $x_i$  are *nonoverlapping* by digit positions

for example,  $12.3456 = 12. + 0.34 + 0.0056$

**The sign of  $x$  is equal to the sign of the largest component  $x_n$ !**

# Exact arithmetic

Addition rule [Dekker]:

**Fast-Two-Sum**( $a, b$ )

```
1       $x \leftarrow a \oplus b$ 
2       $b_{virt} \leftarrow x \ominus a$ 
3       $y \leftarrow b \ominus b_{virt}$ 
4      return ( $x, y$ )
```

$$(a + b) = (7231. + 56.78)$$

$$\begin{array}{r} a \qquad \qquad \qquad 7231. \\ b \qquad \qquad \qquad \underline{56.78} \end{array}$$

$$x = a \oplus b \qquad 7287.$$

$$\begin{array}{r} a \qquad \qquad \qquad \underline{7231.} \end{array}$$

$$b_{virt} = x \ominus a \qquad 56.$$

$$y = b \ominus b_{virt} \qquad 0.78$$

$$(x + y) = (7287. + 0.78)$$

# Exact arithmetic

Kahan Summation Formula:

**Summation** ( $a_1, \dots, a_n$ )

```
1       $s \leftarrow a_1$ 
2       $y \leftarrow 0$ 
3      for  $i \leftarrow 1$  to  $n$ 
4           $b \leftarrow a_i \ominus y$ 
3           $x \leftarrow s \oplus b$ 
4           $b_{virt} = x - s$ 
5           $y = b_{virt} \ominus b$ 
6           $S = x$ 
7      return  $S$ 
```

Computed sum is equal to

$$\sum x_i(1 + \delta_i) + O(n\epsilon^2)\sum |x_i|,$$

where

$$|\delta_i| \leq 2\epsilon$$

Naïve summation gives

$$\sum x_i(1 + \delta_i),$$

where

$$|\delta_i| < (n - i)\epsilon.$$

# Computational robustness

- Big numbers
- Exact arithmetic
- Adaptive arithmetic

The idea: use exact arithmetic **only when necessary!**

# Adaptive arithmetic



**Jonathan Shewchuk**

Professor in Computer Science University of  
California at Berkeley

```
int orient2d(p, p1, p2);  
int orient3d(p, p1, p2, p3);  
int incircle(p, p1, p2, p3);  
int insphere(p, p1, p2, p3, p4);
```

**More than 4000 lines  
of source code on C**

# Example

Does a point belong to a given triangle?

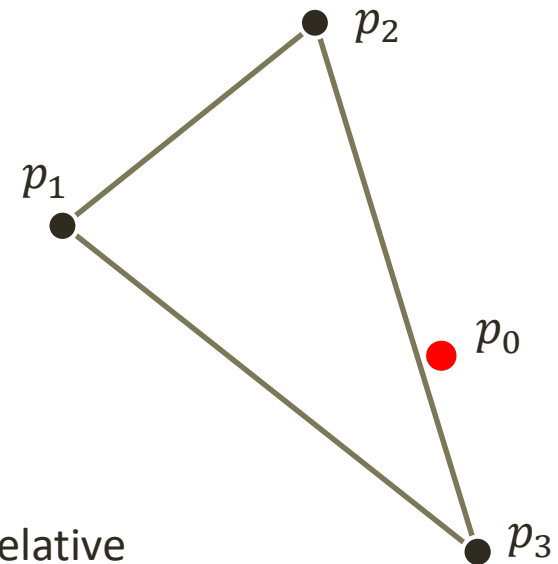
1. Primitive (non-robust) solution: estimate the barycentric coordinates of  $p_0$

$$\begin{aligned}x_0 &= x_1 b_1 + x_2 b_2 + x_3 b_3 \\y_0 &= y_1 b_1 + y_2 b_2 + y_3 b_3 \\1 &= b_1 + b_2 + b_3\end{aligned}$$

```
bool inside = (b1 >= 0 && b1 <= 1) &&  
              (b2 >= 0 && b2 <= 1) &&  
              (b3 >= 0 && b3 <= 1);
```

2. Robust solution: estimate exact position of  $p_0$  relative to the sides of the triangle

```
bool inside = (orient2d(p0, p1, p2) <= 0 &&  
              orient2d(p0, p2, p3) <= 0 &&  
              orient2d(p0, p3, p1) <= 0);
```



# References

1. **Franco P. Preparata, Michael Ian Shamos.** Computational Geometry: An Introduction. Springer-Verlag, 1985
2. **Mark de Berg, et al.** Computational Geometry. *Algorithms and Applications*. Springer, 2008.
3. **Д. М. Васильков.** Геометрическое моделирование и компьютерная графика: вычислительные и алгоритмические основы. Мн., БГУ, 2011.
4. **David Goldberg.** What every computer scientist should know about floating-point arithmetic. *Journal ACM Computing Surveys (CSUR)*, vol. 23 Issue 1, March 1991 (5-48).
5. **J. R. Shewchuk.** Adaptive precision floating-point arithmetic and fast robust geometric predicates. *Discrete & Computational Geometry*, 18:305–363, 1997.
6. <https://www.cs.cmu.edu/~quake/robust.html>