Computational Geometry

1. Introduction

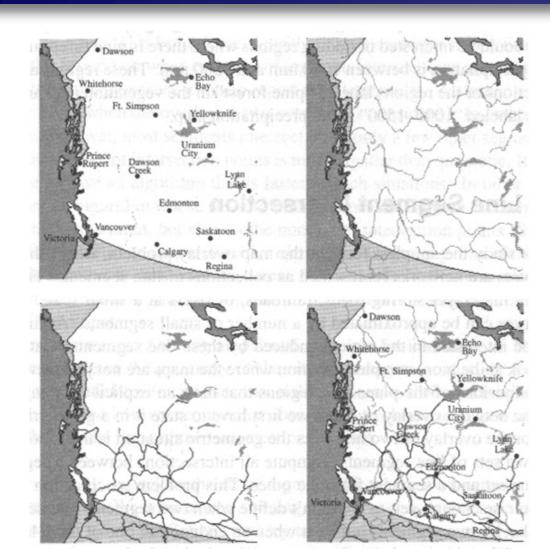
1. Introduction

- Computational Geometry is concerned with the design and the analysis of algorithms and data structures for geometric problems from
 - computer graphics,
 - visualization,
 - geographic information systems (GIS),
 - computer vision and pattern recognition,
 - robotics, etc.
- Computational Geometry was established as discipline of its own right between 1975 and 1980.

- a) In a geographic information systems (GIS) the information is layered in separated maps:
 - Maps for streets, rivers, train tracks, villages, land use, etc.
 - Every map corresponds to one information layer.
 - Question a): Find all bridges, i.e. compute all intersections of streets with rivers.
 - If both are represented as polygon chains this requires computing all intersections of all polygon chains of both maps.
 - Naïve approach: Intersect all lines representing streets with all lines representing rivers...
 - ... infeasible for real world data.
 - Line segment intersections: Chapter 2

1. Introduction

1.1 Typical problems



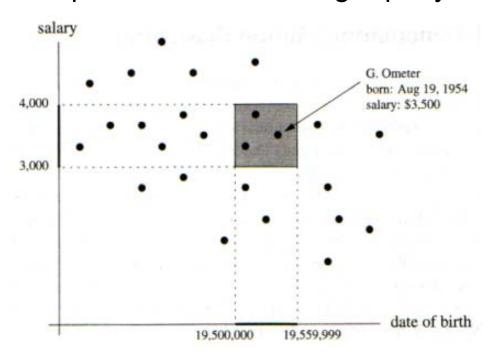
Cities, rivers, railroads, and their overlay in western Canada

- b) In a data base the data of the employees of a company are stored: name, date of birth, monthly salary.
 - Question b): Find all employees born between 1950 and 1955 with a salary between 3.000 \$ and 4.000 \$.
 - To sort the date of birth compute the numbers

$$10.000 \, \text{x year} + 100 \, \text{x month} + \text{day}.$$

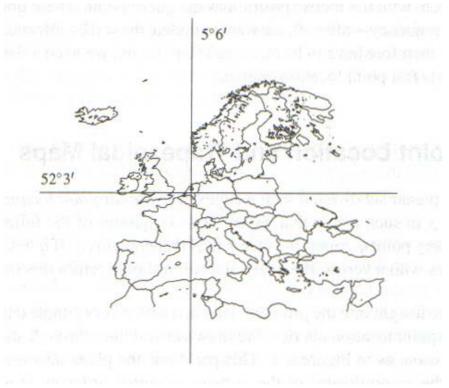
- In combination with the salary this yields a 2d range query:
 - The first dimension is the date of birth.
 - The second dimension is the salary.

Geometric interpretation of a 2d range query



- Remark: Range queries run usually for very high dimensional data.
- Range query: Chapter 4

c) Question c): Compute from the geographic position the country in a map.

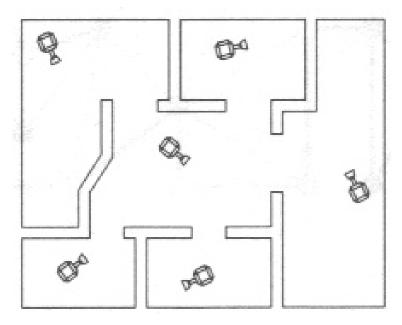


For humans this task is easy on 2d maps, but...

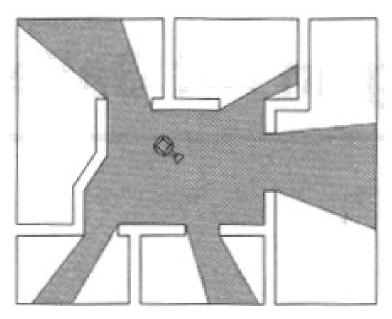
- ... for the computer this is difficult.
- The performance and usability of visualization systems depends on a fast answer to this **point query** for large data sets.
- Point location queries must run in real time, e.g.:
 - navigation systems
 - mouse clicks in windows or graphical object on the screen
 - ...
- Point location: Chapter 5

d) For security reasons an art gallery shall be equipped with pivoting surveillance cameras.

Question d): How many cameras do you need at least and where do you place them?



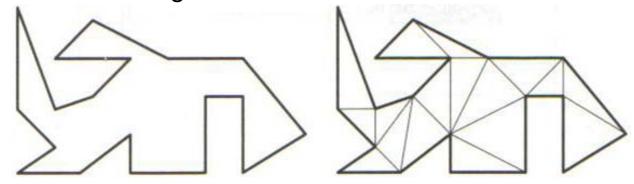
Art gallery with cameras



Area controlled by one camera

This leads to triangulations of arbitrary (simple) polygons.

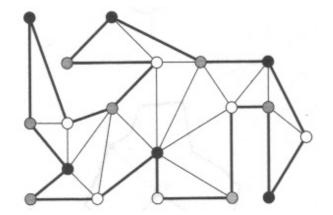
- A polygonal area represents the art gallery
- Camera positions correspond to points inside the polygon.
- Extreme case: Convex polygons can be controlled by one camera from an arbitrary point inside the polygon.
- → This implies that you need at least as many cameras as there are triangles on the triangulation.



A simple polygon and a possible triangulation of it.

Instead of placing in each triangle one camera, one can look for better solutions with less cameras:

- place cameras on particular diagonals, or
- place cameras in particular corners polygon.



- The connection to computer graphics is obvious because since GPUs process triangles.
 - Every CAD system must be able to triangulate arbitrary polygons automatically.
- Polygon triangulation: Chapter 6

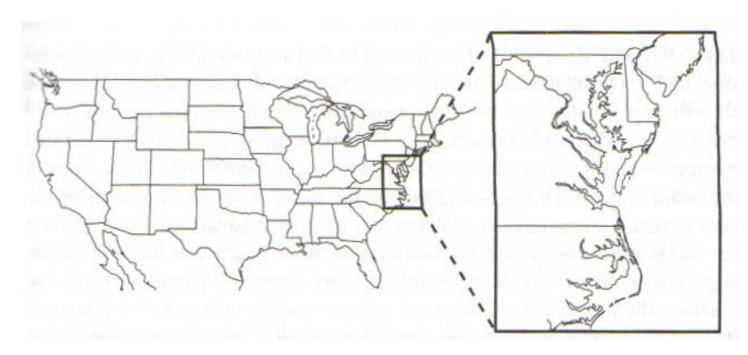
e) You need to make an urgent telephone call on campus but you do not have a cell phone.

However, you know some telephone boxes near by.

- Question e): Which is the closest telephone box to your position?
- A solution would be a map subdivided into regions, where each region contains the closest telephone box.
- Further questions:
 - What is the shape of these regions?
 - How can these regions be computed?
- Voronoi diagrams & Delaunay triangulations: Chapter 7

f) In a navigation system all streets, villages, rivers, etc. of the USA are stored.

However, on the display only a small section around the current position is visible.



Question f): Compute all entities that are visible within a given viewing window.

- To solve this a window query is computed on the data base of points, line segments and polygons.
- Remark: This window query must run a cheap hardware with very little resources.
- Further applications:
 - Flight simulators, displaying only a small section (viewing volume) of a 3d landscape.
 - Control of chip layouts.
- Window query: Chapter 9

1. Introduction

1.2 Basics

- 1.2.1 Geometric objects
- **1.2.2** Primitive geometric operations
- 1.2.3 Computation model
- 1.2.4 Runtime of algorithms
- **1.2.5** O-calculus
- 1.2.6 Master-Theorem for recurrences
- 1.2.7 Example: 2d-convex hull

1.2.1 Geometric objects

1.2 Basics

- *Points* in *d*-dimensional Euclidian space E^d
- Fixed Cartesian coordinate system with origin O.
 - Points are represented by coordinate tuples:

$$p = (p_1, ..., p_d), q = (q_1, ..., q_d), ... \in \mathbb{R}^d.$$

A coordinate tuple p represents also a vector from the origin O to the point p.

1. Introduction

1.2.1 Geometric objects

1.2 Basics

■ For points $p,q \in \mathbb{R}^d$, $\lambda \in \mathbb{R}$, we have the operations:

• Addition:
$$p+q=(p_1,...,p_d)+(q_1,...,q_d):=(p_1+q_1,...,p_d+q_d).$$

- Multiplication: $\lambda p = \lambda(p_1,...,p_d) := (\lambda p_1,..., \lambda p_d).$
- Norm: $|p| := \sqrt{p_1^2 + ... + p_d^2}$.
- Scalar product: $< p,q> := (p_1 q_1 + ... + p_d q_d) = |p| \cdot |q| \cos(^pq)$.
- Cross product: $|p \times q| = |p| \cdot |q| \sin(^{\bullet}pq)$.
- Orthogonality: Two vectors are orthogonal, if

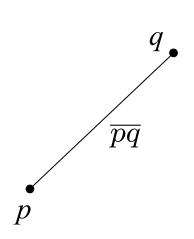
$$< p, q > = 0.$$

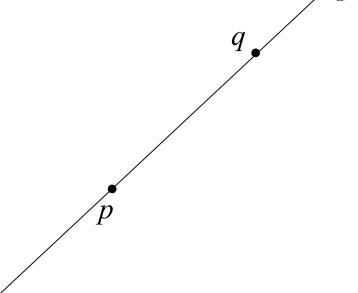
1.2.1 Geometric objects

1. Introduction

1.2 Basics

- Line segments (Strecke)
 - Two points p and q with $\overline{pq} := \{(1-\lambda)p + \lambda q \mid \lambda \in [0,1]\}.$
- Lines (Gerade)
 - Two points p and q with $g := \{(1-\lambda)p + \lambda q \mid \lambda \in \mathbb{R}\}.$

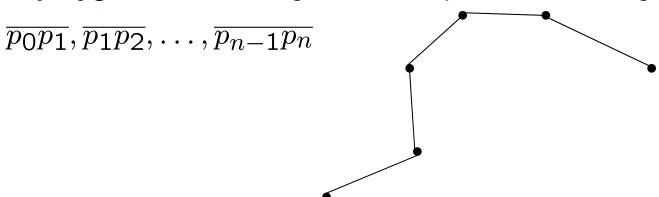




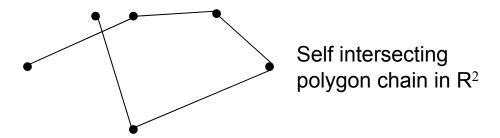
1.2 Basics

Polygon chains

A polygon chain of length n is a sequence of line segments with



A polygon chain is called simple, if it has no self intersections.

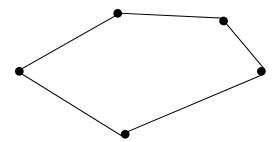


1.2.1 Geometric objects

1. Introduction

1.2 Basics

- Polygons (closed polygon chain)
 - A polygon chain with $p_0 = p_n$ is called **polygon.**



1.2.2 Primitive operations

1.2 Basics

- Primitive geometric operations involve only objects of constant size, e.g.:
 - Intersection of two lines.
 - Point lies in half-space of a hyper-plane.
 - Not primitive: Point lies in polygon, intersection of two polygons.
- Primitive operations are treated in the computation model (see §1.2.3) as operations with constant run time and memory consumption.

1.2.3 Computation model

1.2 Basics

Real RAM

- Countable infinitely many memory cells, that are addressable by natural numbers.
- Memory cells can take an arbitrary number of real numbers and integers.
- The following operations are primitive and take only constant time:
 - arithmetic operations (+, , × , / , mod , div),
 - comparisons of real numbers,
 - indirect addressing,
 - roots, trigonometric functions, logarithms.

1.2.4 Runtime of algorithms

1.2 Basics

- Runtime of an algorithm = Number of used primitive operations!
 - One primitive operation corresponds to one time step.
- Runtime depends on size n of input!
- Thus, the runtime is a function of the input size!
 - E.g. $T(n) = a n^2 + b n + c$ for constants a, b and c.
 - T(n) is also called *runtime function in n*.

1.2.4 Runtime of algorithms

1.2 Basics

- Reminder:
 - Best-Case-Analysis
 - Worst-Case-Analysis
 - Average-Case-Analysis
- The complete runtime functions is often very complicated.
- For simplicity only the order of runtime is considered!!
 - The order of $T(n) = a n^2 + b n + c$ is e.g. n^2 .
 - The order of T(n) = a n + b is n.

1.2 Basics

The O-Calculus (Landau-Symbols)

- Runtime functions can be very complicated.
- To determine the complexity of a problem runtime functions must be compared.
- The asymptotic growth of a function is most relevant.
- → Idea: Compare the growth of two functions g(n) and f(n) only for large n.
 - "Which function wins the race?"
- Advantages:
 - Constant factors are irrelevant.
 - For polynomials only the larges exponent counts.



Edmund Landau, 1877-1938. Quelle: Wikipedia

1.2 Basics

Definition: (pronounced: "Big-Oh")

 $O(g(n)) = \{f(n) : \text{ There are positive constants } c \text{ and } n_0, \text{ such that } 0 \le |f(n)| \le c |g(n)| \text{ for all } n \ge n_0\}.$

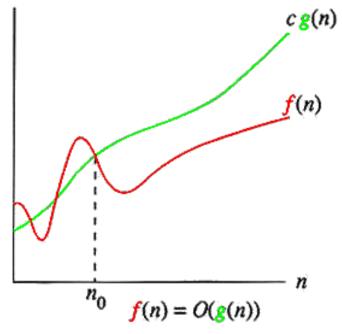
- Notation: f(n) = O(g(n)) instead of $f(n) \in O(g(n))$.
- Examples:

•
$$3n = O(n^2)$$

•
$$n^2/2 - 3n = O(n^2)$$

•
$$6n^3 \neq O(n^2)$$

• f(n) = O(g(n)) resembles " $a \le b$ ".

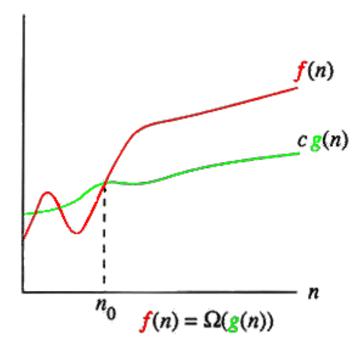


1.2 Basics

Definition: (pronounced: "Big-Omega")

$$\Omega(g(n)) = \{f(n) : \text{ There are positive constants } c \text{ and } n_0, \text{ such that } 0 \le c |g(n)| \le |f(n)| \text{ for all } n \ge n_0\}.$$

- Examples:
 - $3n \neq \Omega(n^2)$
 - $n^2/2 3n = \Omega(n^2)$
 - $6n^3 = \Omega(n^2)$
- $f(n) = \Omega(g(n))$ resembles " $a \ge b$ ".

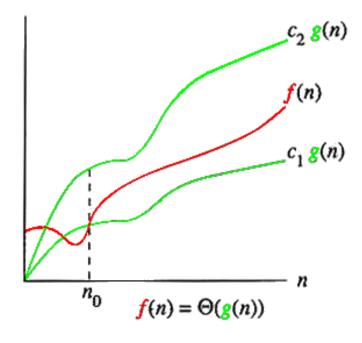


1.2 Basics

Definition: (pronounced: "Big-Theta")

 $\Theta(g(n)) = \{f(n) : \text{ There are positive constants } c_1, c_2 \text{ and } n_0, \text{ such that } 0 \le c_1 |g(n)| \le |f(n)| \le c_2 |g(n)| \text{ for all } n \ge n_0\}.$

- Examples:
 - $3n \neq \Theta(n^2)$
 - $n^2/2 3n = \Theta(n^2)$
 - $6n^3 \neq \Theta(n^2)$
- $f(n) = \Theta(g(n))$ resembles "a = b".



1.2 Basics

Definition: (pronounced: "Small-Oh")

 $o(g(n)) = \{f(n) : \text{ For all constants } c > 0 \text{ there is a } n_0 > 0, \text{ such that } 0 \le |f(n)| \le c |g(n)| \text{ for all } n \ge n_0\}.$

Examples:

- $3n = o(n^2)$
- $n^2/2 3n \neq o(n^2)$
- $6n^3 \neq o(n^2)$
- f(n) = o(g(n)) resembles "a < b".
- An equivalent condition for f(n) = o(g(n)): $\lim_{n \to \infty} \frac{f(n)}{g(n)} = 0$.

1.2 Basics

Definition: (pronounced: "Small-Omega")

$$\omega(g(n)) = \{f(n) : \text{ For all constants } c > 0 \text{ there is a } n_0 > 0, \text{ such that } 0 \le c |g(n)| < |f(n)| \text{ for all } n \ge n_0 \}.$$

Examples:

- $3n \neq \omega(n^2)$
- $n^2/2 3n \neq \omega(n^2)$
- $6n^3 = \omega(n^2)$
- $f(n) = \omega(g(n))$ resembles "a > b".
- An equivalent condition for $f(n) = \omega(g(n))$: $\lim_{n \to \infty} \frac{f(n)}{g(n)} = \infty$.

1.2.6 Master-Theorem

1.2 Basics

- For a divide-and-conquer-algorithm ("teile und herrsche") with runtime function T(n) the problem of input size n
 - 1. is divided into a sub-problems of size n/b with runtime f(n),
 - 2. every sub-problem is solved with runtime T(n/b) and
 - 3. all sub-solutions are merged ("conquer") with runtime f(n).
- What is the total runtime of such an algorithm.
- In many cases an answer is given by the

"Master-Theorem".

1.2.6 Master-Theorem

1.2 Basics

Theorem (Master Theorem for Recursions)

Let T(n) be a runtime function of the form $T(n) = a \cdot T(n/b) + f(n)$ with

- constants $a \ge 1$ and b > 1,
- a function f(n), and
- n/b is interpret as $\lceil n/b \rceil$ or $\lfloor n/b \rfloor$.

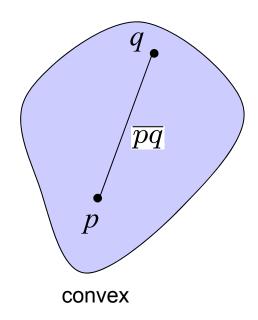
Then T(n) satisfies one of the following estimates:

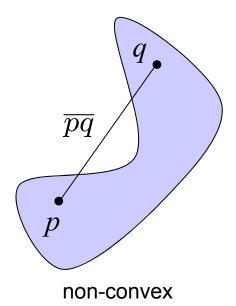
1. If
$$f(n) = O(n^{\log_b a - \varepsilon})$$
 for $\varepsilon > 0$, then $T(n) = \Theta(n^{\log_b a})$.
2. If $f(n) = \Theta(n^{\log_b a})$, then $T(n) = \Theta(n^{\log_b a})$ log n .
3. If $f(n) = \Omega(n)$ for $\varepsilon > 0$ and $a \cdot f(n/b) \le c \cdot f(n)$ for $c < 1$ and sufficiently large n , then $T(n) = \Theta(f(n))$.

1.2 Basics

Definition (Convex Set)

A set $S \subseteq \mathbb{R}^d$ is called **convex**, iff for all points $p,q \in S$ the connecting line segment $\overline{pq} = \{x: x = \lambda \ p + (1-\lambda)q, \lambda \in [0,1] \}$ is contained in S, i.e. $\overline{pq} \subseteq S$.





1.2 Basics

Definition (Convex Hull)

The **convex hull** CH(S) of a set $S \subseteq \mathbb{R}^d$ is the smallest convex set containing S, i.e. it is the intersection of all convex sets containing S.

Rubber band metaphor

1.2 Basics

For d = 2 there is an alternative (informal) definition:

The convex hull CH(P) of a finite set P of n points is the unique, convex polygon with knots from P containing all points of P.

This definition is used to develop an algorithm computing the convex hull of a set of points.

Remark: Convex hull algorithms are to Computational Geometry what sorting is to discrete algorithms.

1. Introduction

1.2 Basics

Remark:

The complexity of the convex hull problem in 2d is $O(n \log n)$.

Remark:

Four approaches:

1. Naïve approach: $O(n^3)$.

2. Divide-and-conquer approach: $O(n \log n)$.

3. Scan line approach, Graham's scan: $O(n \log n)$.

4. Output-size sensitive approach, Jarvis' march: O(h n).