

INFS4205/7205 Advanced Techniques for High Dimensional Data

Spatial Query Processing 1

Semester 1, 2021

University of Queensland

+ Advanced Techniques for High Dimensional Data

- □ Course Introduction
- □ Introduction to Spatial Databases
- Spatial Data Organization
- Spatial Query Processing
- Managing Spatiotemporal Data
- Managing High-dimensional Data
- Other High-dimensional Data Applications
- When Spatial Temporal Data Meets Al
- Route Planning
- □ Trends and Course Review

+ Learning Objectives

- What we will cover
 - The filter-and-refine approach
 - Some basic computational geometry algorithms
 - Intersection join algorithms using various spatial indexes
 - Efficient processing of some advanced spatial queries

Goals

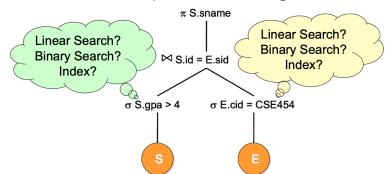
- Understand efficient processing of spatial queries using spatial indexes and the filter-and-refine approach
- Enhance the understanding of spatial databases from in-depth knowledge of advanced spatial processing
- Provide a brief view of the frontier of spatial database research

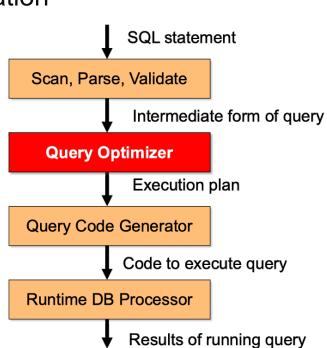
+ Readings

- R. Güting, An Introduction to Spatial Database Systems,
 The VLDB Journal, 3:4, 1994
- Oracle Business and Technical White Papers
- V. Gaede and O Günther, Multidimensional Access Methods, ACM Computing Surveys, 30:2, 1998
- J. Orenstein and F. Manola, PROBE Spatial Data Modeling and Query Processing in an Image Database Application, IEEE Transactions on Software Engineering, 14:5,1988
- T. Brinkhoff, H.-P. Kriegel and B. Seeger, Efficient Processing of Spatial Joins Using R-Trees, SIGMOD'93

+ Query Processing

- Process a query accurately and in the minimum amount of time possible
 - Design and fine-tune algorithms for each of the basic operators
 - Map high-level queries into a composition of these basic operators and optimize with their information
- RDBMS Query Processing
 - Projection / Selection / Join
 - Linear Scan / Binary Search / Index
 - Nested Loop / Sort Merge / Hash Join



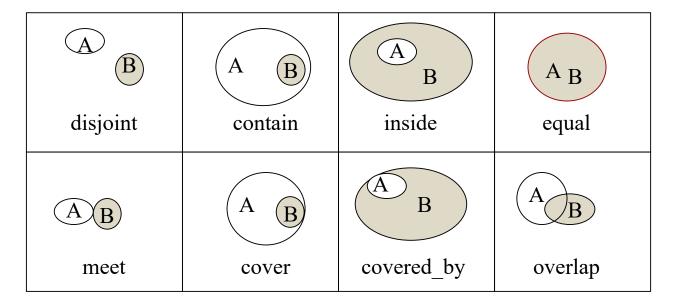


+ Spatial Operations

- Spatial operations are intrinsically more expensive than relational operations
 - Large amount of complex data
 - → higher I/O cost
 - Computational geometry algorithms
 - → higher CPU cost
- Spatial query processing is different from RDBMS query processing
 - Different types of indexes
 - CPU costs typically not considered in RDBMS query optimization

+ Spatial Relations

Topological



- Directional
 - Above, Left, North of,...
- Metric
 - Distance, Length, Area,...

+ Spatial Operations

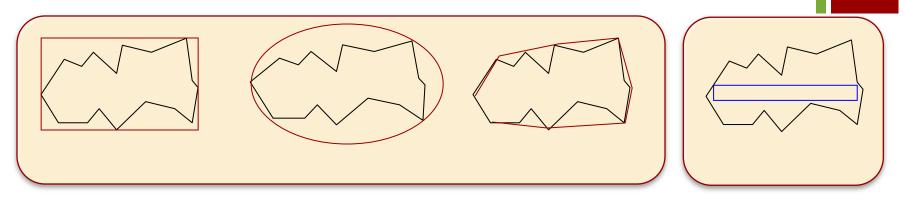
- Update
 - Create, Delete, Modify...
- Selection
 - Point Query
 - Given a point p
 - Find all the spatial objects *O* that *contains* it
 - Find all the spatial objects *0* that within 10km in Euclidean/Network distance
 - Range Query
 - Given a query polygon *Q*, find all spatial objects *O* that *intersects* it
 - Window query: when *Q* is a rectangle
- Spatial Join
 - \blacksquare R and S join on a spatial predicate θ

+ Filter-and-Refine

- A most commonly used processing strategy
- Motivation
 - Avoid expensive spatial processing as much as possible
- Basic Idea
 - A filter step, followed by a refinement step
 - Filter step: applying simple operations on approximations of spatial objects
 - Refinement step: applying the actual spatial operations on the full geometry of spatial objects

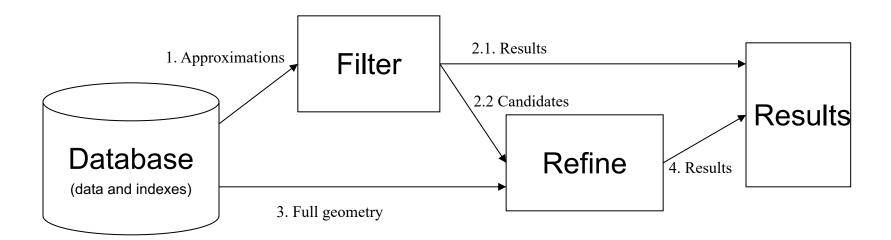
+ Object Approximation

Different types of approximation



- Conservative approximation: containing the object (min)
 - Approximation don't overlap → impossible for objects to overlap
 - Approximation do overlap → objects may overlap (need to refine!)
- Progressive approximation: contained by the object (max)
 - Approximation do overlap → objects overlap too!
 - Approximation don't overlap → objects may overlap (need to refine!)

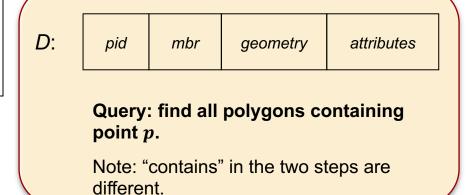
+ The Filter-and-Refine Workflow



+ An Algorithm Using MBR Filter

■ The Filter step

```
C = \phi;
for each d in D
if d.mbr contains p
add d.pid to C;
```



■ The Refine step

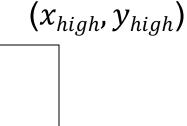
```
for each c in C select geometry from D where pid = c into geo; if geo contains p output c;
```

+ Computational Geometry Algorithms

- Point in Rectangle
- Rectangle intersection
- Point in Polygon
- Polyline intersection
- Polygon intersection

Reading: Preparata and Shamos, Computation Geometry: An Introduction Springer-Verlag 1985.

+ Point in Rectangle

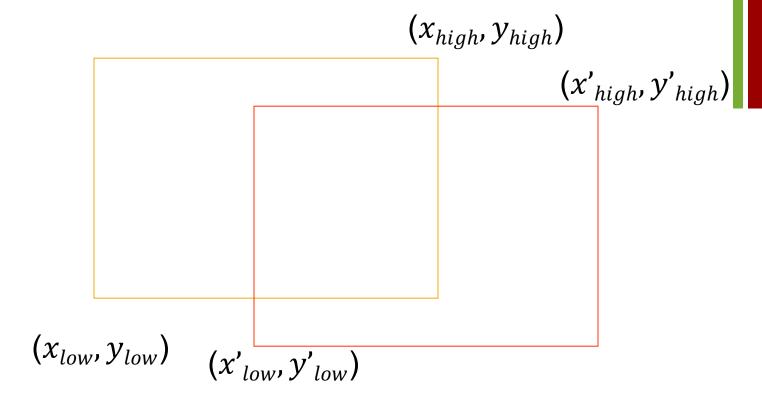


(x,y)

 (x_{low}, y_{low})

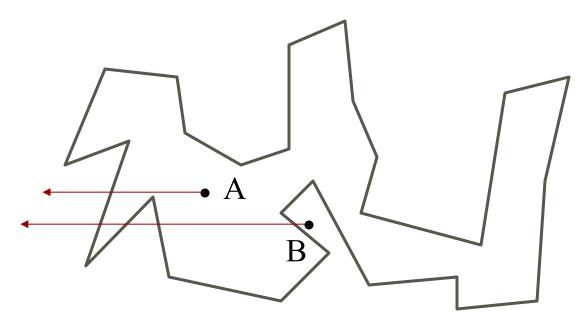
...the condition?

+ Rectangle Intersection



- Four *point-in-rectangle* check?
- When do not overlap?
 - One is on top: $y'_{low} > y_{high}$ or $y_{low} > y'_{high}$
 - One is left/right: $x_{low} > x'_{high}$ or $x'_{low} > x_{high}$

+ Point in Polygon



The ray-cutting algorithm:

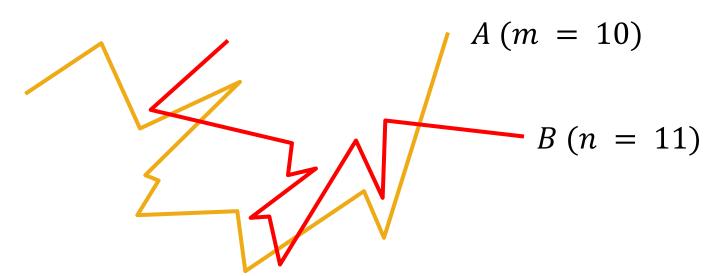
#Intersection: odd = inside, even = outside

...how to represent a line using equations?

...how to check if two lines interest each other?

+ Polyline Intersection

- Naïve: for each line segment in *A*, test against each and every line segment in *B*
 - $O(m \times n)$, where A and B have m and n line segments respectively
 - When each segment intersects with all the others, $\Omega(m \times n)$ is inevitable
 - But in practice, the total number of intersections is much smaller
 - Output / Intersection sensitive?



+ The Plane-Sweep Algorithm

- Extensively used in computational geometry to compute intersections of geometric objects
 - Move a vertical line (*the sweep line*) from left to right
 - Only test those whose x-interval overlaps
 - End points are the event point: Only test at these points
 - Only those *A* and *B* lines which intersect with the sweep line at the same time can possibly intersect



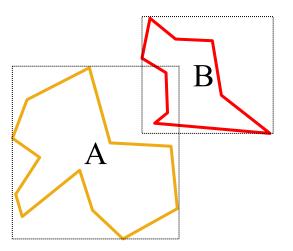
+ The Plane-Sweep Algorithm

- Only those A and B lines which intersect with the sweep line at the same time can possibly intersect
 - Add into test buffer when the sweep line is at the left end
 - Delete when the sweep leaves at the right end
 - But the two segments that intersect the sweep line can still be far apart in the vertical direction
 - Order the segments from top to bottom as they intersect the sweep line
 - $O((m+n)\log(m+n))$...much better than O(m*n)!



+ Polygon Intersection

- Step1: MBR Intersection
 - Filter the impossible
- Step2: Point-in-Polygon testing
 - "A[i] in B" or "B[j] in A"
 - *i*, *j*: arbitrary point
- Step3: Polyline Intersection

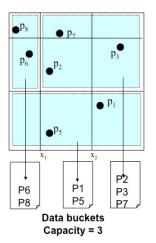


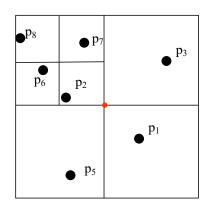
+ Computational Algorithms vs Spatial Database Algorithms

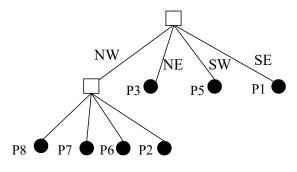
- Computational algorithms
 - Concerning two complex objects
 - Goal: (in memory) computational complexity
- Spatial database algorithms
 - Concerning one or two very large sets of object, which can be simple or complex
 - Goal: *the total cost* (I/O and CPU)
- A spatial database algorithm uses some sort of computational algorithms, but focuses on making data retrieval more efficient (e.g., using spatial indexes)

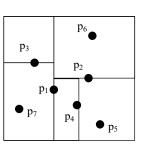
+ Spatial Selection Query

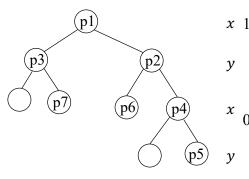
- Point / Range / Within Distance Query
 - Grid Files
 - Quadtree
 - Kd-Tree
 - Z-Value
 - R-Tree

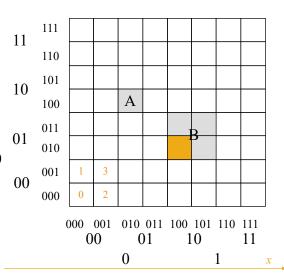


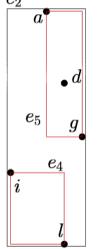


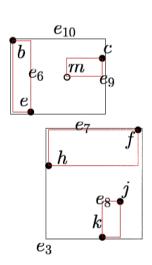










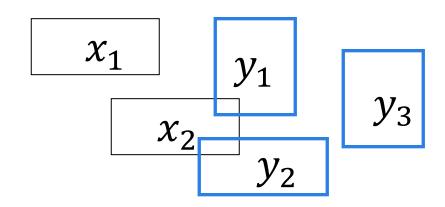


+ Spatial Join Algorithms

- One of the most important spatial operations
- Can be very time-consuming
 - Complex data, complex operations
 - It's not an equijoin!
- Three types of join algorithms:
 - Nested-loops
 - Sort-merge
 - Hash-based

+ Spatial Join Example

Intersection join



- Join results: $(x_2, y_1), (x_2, y_2)$
- Other spatial join operations
 - Topological: intersection, adjacent, contains...
 - Metrical or directional: within_distance...
 - More advanced: nearest....

+ Processing Framework

Filter step

- Find a set of candidates $C = \{(p, s): p \in R \text{ and } s \in S\}$ quickly
 - Using approximations (e.g., MBR) and indexes
 - Other filter steps possible (eg, using progressive approximation)

Housekeeping step

- Process C such that the IO cost for the refinement step can be further minimized
- E.g., Removing duplicates; Performing refinement in optimal order

Refine step

Fetch full geometry for the objects in each candidate, and apply a full test to drop "false hits"

+ Simple Nested-Loops Join

```
for each r in R

for each s in S

if (r,MBR \text{ intersect } s.MBR)

put (r,s) to the candidate set;
```

+ Properties with R/R+-Trees

- All tree nodes contain a set of (MBR, pointer) pairs
 - For each entry in a leaf node:
 - The pointer points to a disk page
 - The MBR covers all the polygons on the disk page
 - For each entry in an internal node:
 - The pointer points to the node of a sub-tree
 - The MBR covers all the polygons (or their MBRs) in that sub-tree

Therefore

- If the MBRs of two entries are disjoint, so must be all their children
- If the MBRs of two entries intersect, then some children pairs might intersect somewhere down the tree/on the page

+ Indexed Nested-Loops Join

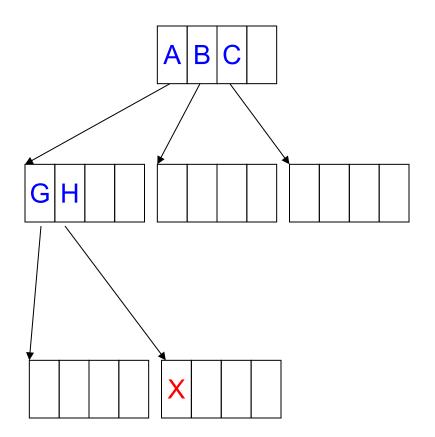
Using a window query against S

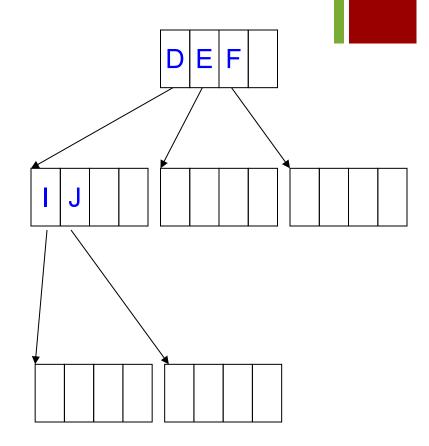
for each r in R

Find all s in S such that s.MBR intersect r.MBR

put (r, s) to the candidate set;

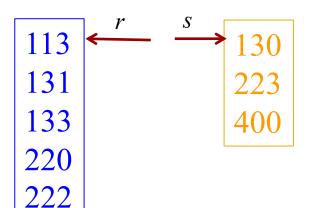
+ Nested-Loops With R/R+-Trees





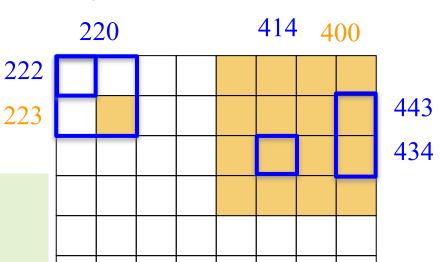
Is it possible to produce the same candidate (p_1, p_2) multiple times?





Differences

- 1. $131 \neq 130$
- Cannot move from 130 to 223 immediately



Algorithm Sketch:

414

434

443

- 1) Two sorted lists and two pointers
- 2) Synchronized traversal
 - overlap(*r*, *s*)?
 - increase min(*r*, *s*)
- 3) Some values in stack

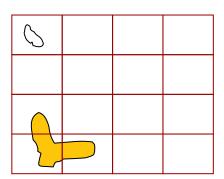
113 131 133 130

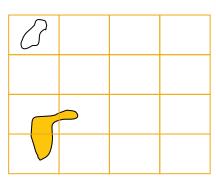
+ Spatial Hash Join

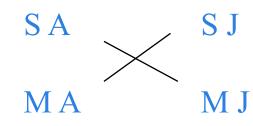
- Bucket Assignment
 - SA: a polygon is mapped to a cell by its centroid
 - MA: a polygon is mapped to all cells it overlaps with

Bucket Join

- SJ: one R bucket joins with one corresponding S bucket
- MJ: one R bucket joins with many S buckets
 - Question: which buckets to join with?

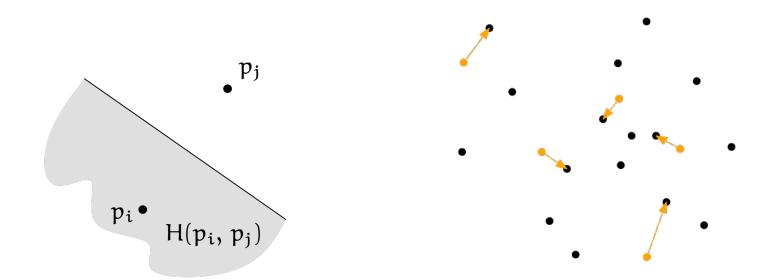






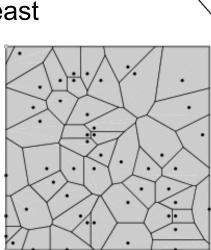
+ The Post Office Problem

- Given n post offices $P = \{p_1, ..., p_n\}$ in a city, find the nearest post office of a given location q
 - The locations of post offices are known and do not change frequently
 - How to answer the query efficiently?



+ Voronoi Diagram

- $\forall p_i, p_j \in S$, p_i dominates p_j is the sub-plane being at least as close to p_i as to p_j
 - $dom(p_i, p_j) = \{q \in R^2 | dist(q, p_i) \le dist(q, p_j)\}$
 - Region $reg(p_i) = \bigcap_{p_i \in S \{p_i\}} dom(p_i, p_j)$
 - Intersection of n-1 half-planes
 - The boundary has at most n-1 edges / vertices
 - Each point on an edge is equidistance from two points is
 - Each vertex on an edge is equidistance from at least three points
 - This polygonal partition is Voronoi diagram Vor(S)
 - Contains exactly n regions



+ Voronoi Diagram Everywhere









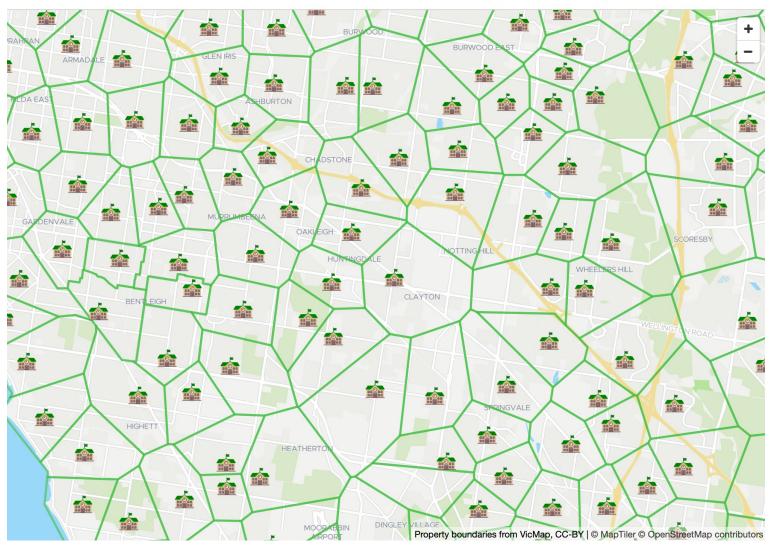


+ Voronoi Diagram Everywhere



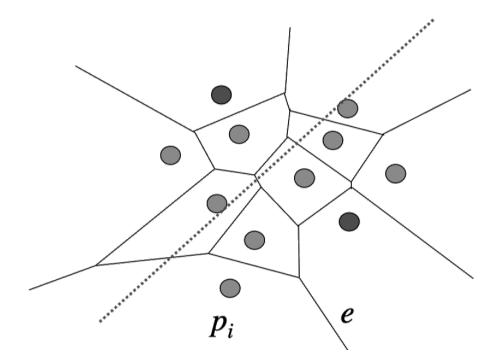


Education and Training



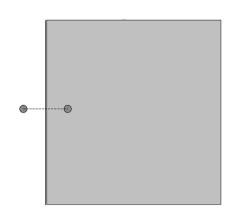
+ Voronoi Diagram

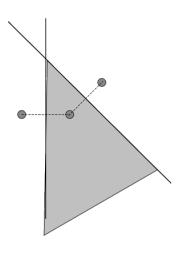
- Linear Size |V|, |E| = O(n)
 - Intuition: Not all bisectors are Voronoi edges
 - Euler's Formula: |V| |E| + f = 2
 - |V| ≤ 2n 5
 - $|E| \le 3n 6$

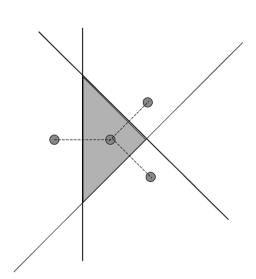


+ Voronoi Diagram Construction

■ Half Plane Intersection



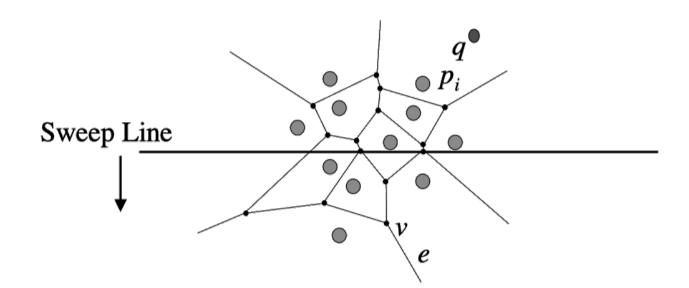




 $O(n^2 log n)$

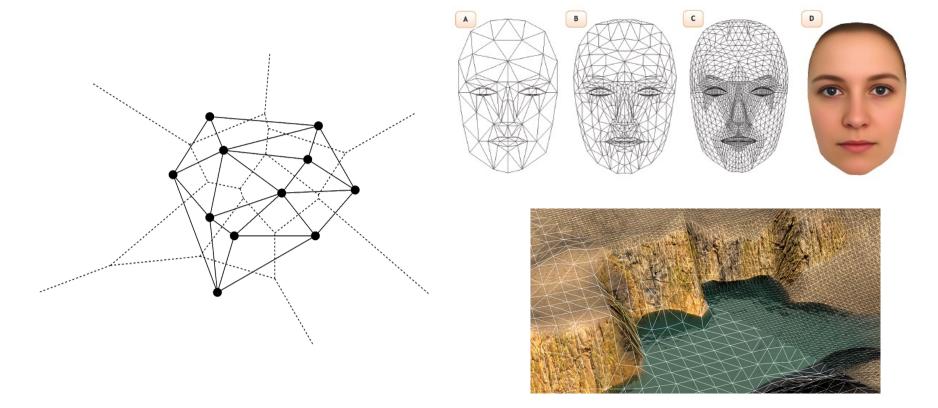
+ Voronoi Diagram Construction

- Fortune's Algorithm
 - Sweep line
 - Horizontal line from top to bottom
 - Incremental construction
 - lacksquare O(nlogn)



+ Voronoi Diagram

- Dual Problem: Delaunay Triangulation
 - Given a set P of discrete points in a plane, no point in P is inside of any triangle in DT(P).



+ Advanced Spatial Queries

- Other types of spaces
 - Euclidean space simple but not always realistic
 - Network space and surface space
 - Computing the distance between two points becomes very time consuming
- Other types of relationships
 - Point query and Range query
 - Selection query vs Join query
 - Nearest neighbor and skyline queries
 - Very different query processing strategies
 - With applications beyond spatial databases