

Dynamic Planar Convex Hull

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31st May 2002

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Outline of the talk

- Planar convex hull
- Duality: Lower Envelope
- Application: k -level
- Overall structure of the data structure
- Some key ingredients
- Lower bounds

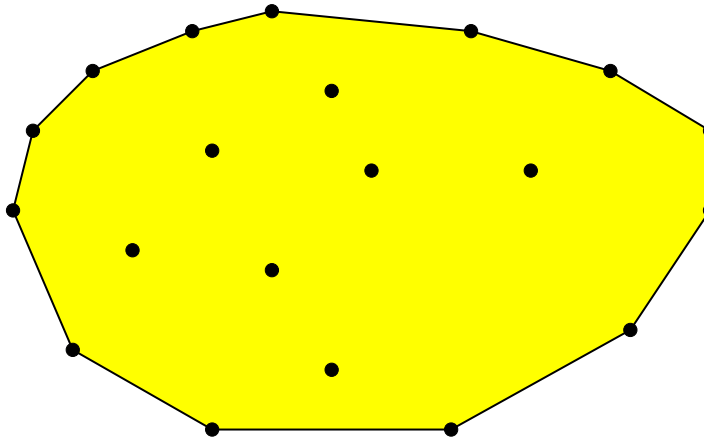
Planar Convex Hull

Input

A set of points $S \subseteq \mathbb{R}^2$

Output

The points on the convex hull $\text{CH}(S)$ in clockwise order ↻



$$n = |S|$$

$$h = |\text{CH}(S)|$$

Known results

Optimal $O(n \log n)$

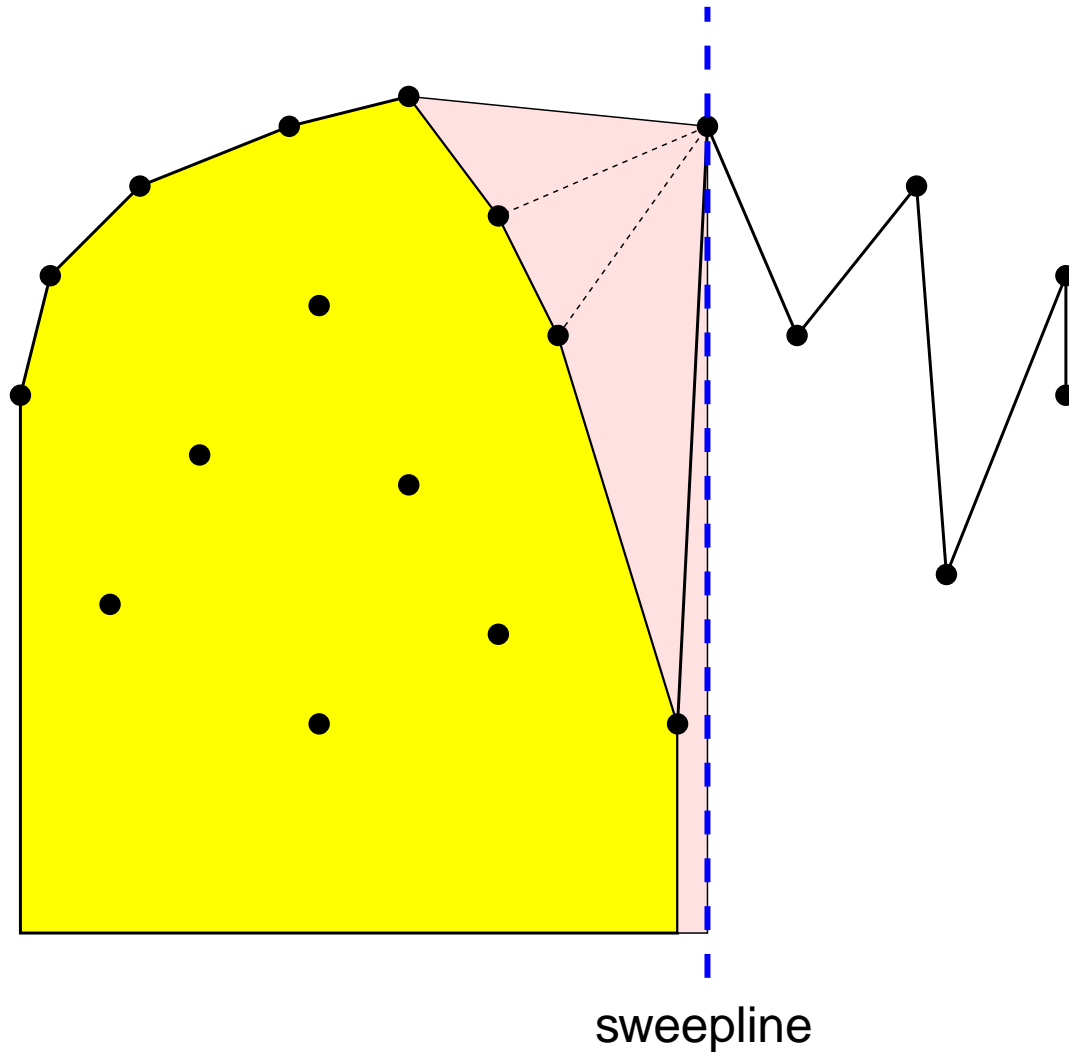
Graham 1972; ...

Output-sensitive $O(n \log h)$

Kirkpatrick, Seidel 1986; Chan 1996

Graham's Scan

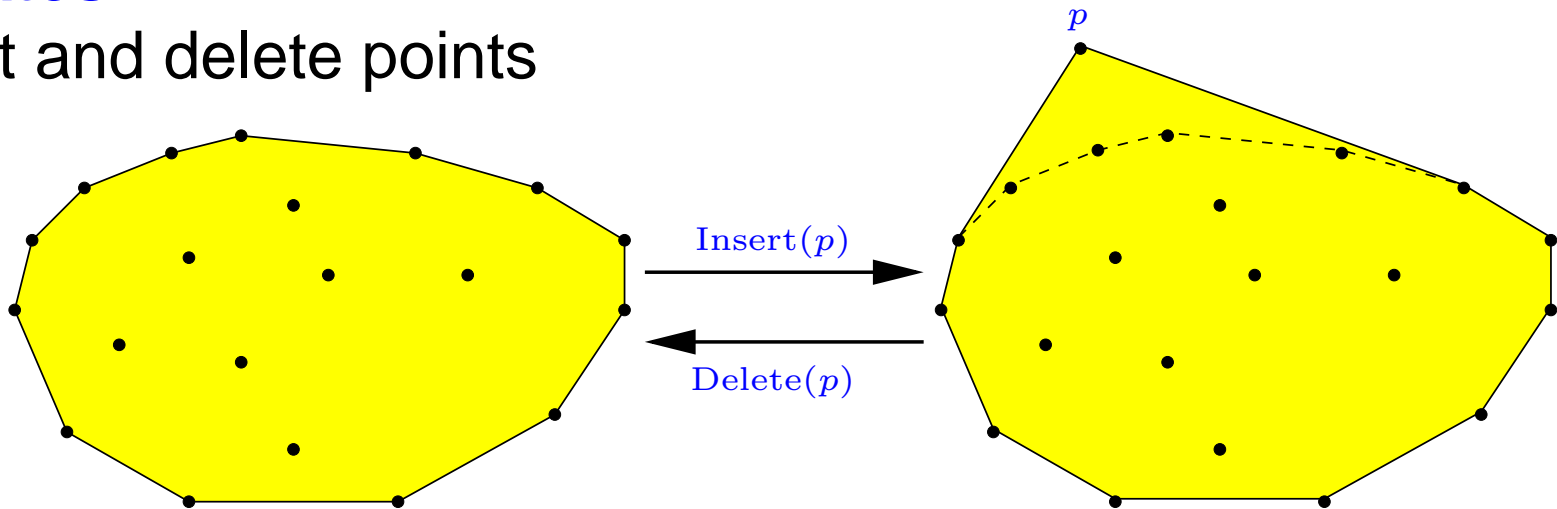
Andrew's variant for upper hull



Dynamic Planar Convex Hull

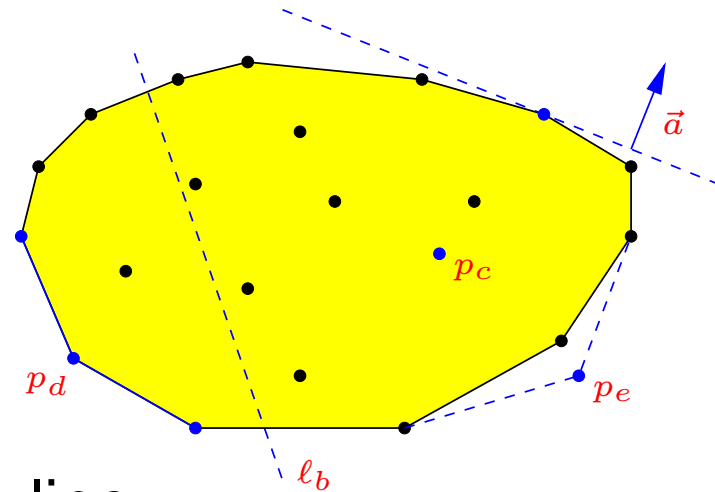
Updates

Insert and delete points



Queries

- (a) The extreme point in a direction
- (b) Does a line intersect $\text{CH}(S)$?
- (c) Is a point inside $\text{CH}(S)$?
- (d) Neighbor points on $\text{CH}(S)$
- (e) Tangent points on $\text{CH}(S)$
- (f) The edges of $\text{CH}(S)$ intersected by a line



Dynamic Planar Convex Hull Results

Insertions only

Update

Query

Preparata 1979

$O(\log n)$

$O(\log n)$

Deletions only

Hershberger, Suri 1992

$O_A(\log n)$

$O(\log n)$

Offline

Hershberger, Suri 1996

$O_A(\log n)$

$O(\log n)$

Fully dynamic

Overmars, van Leeuwen 1981

$O(\log^2 n)$

$O(\log n)$

Chan 1999

$O_A(\log^{1+\epsilon} n)$

$O(\log n)$

Brodal, Jacob 2000

Kaplan, Tarjan, Tsioutsoulis '01

} $O_A(\log n \cdot \log \log n)$ $O(\log n)$

O_A =Amortized

Query=Queries (a)–(e)

Dynamic Planar Convex Hull Results

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$O_A(\log n \cdot \log \log n)$

$O(\log n)$

this Thesis

$O_A(\log n)$

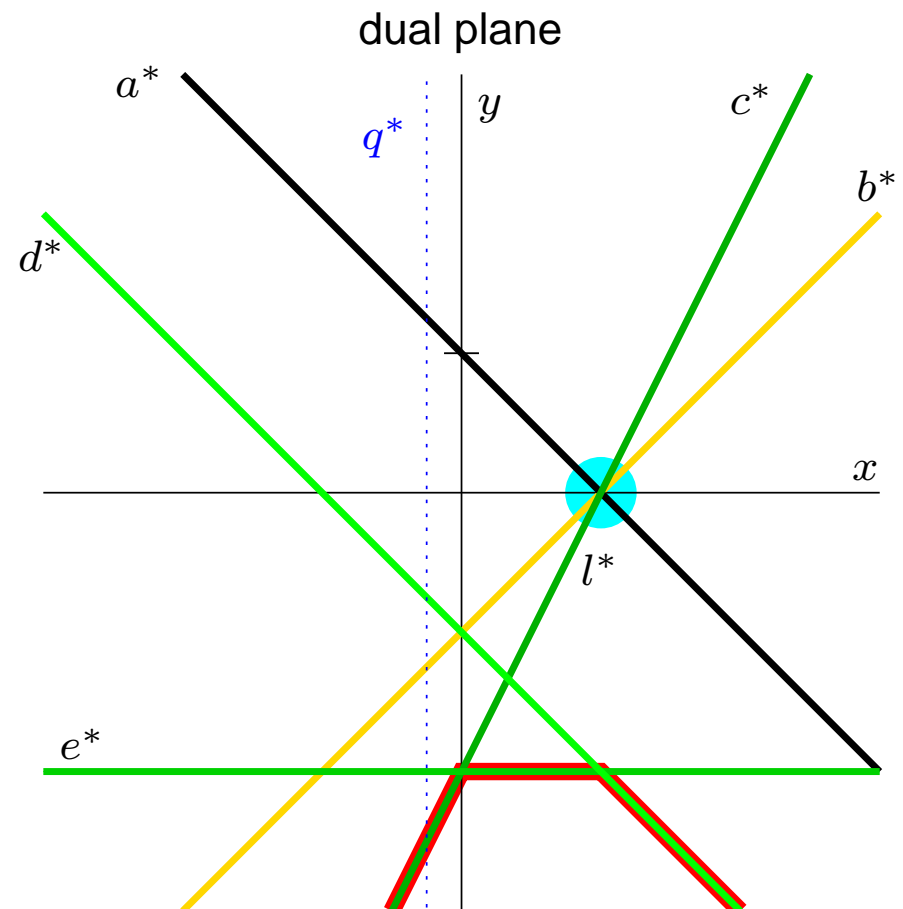
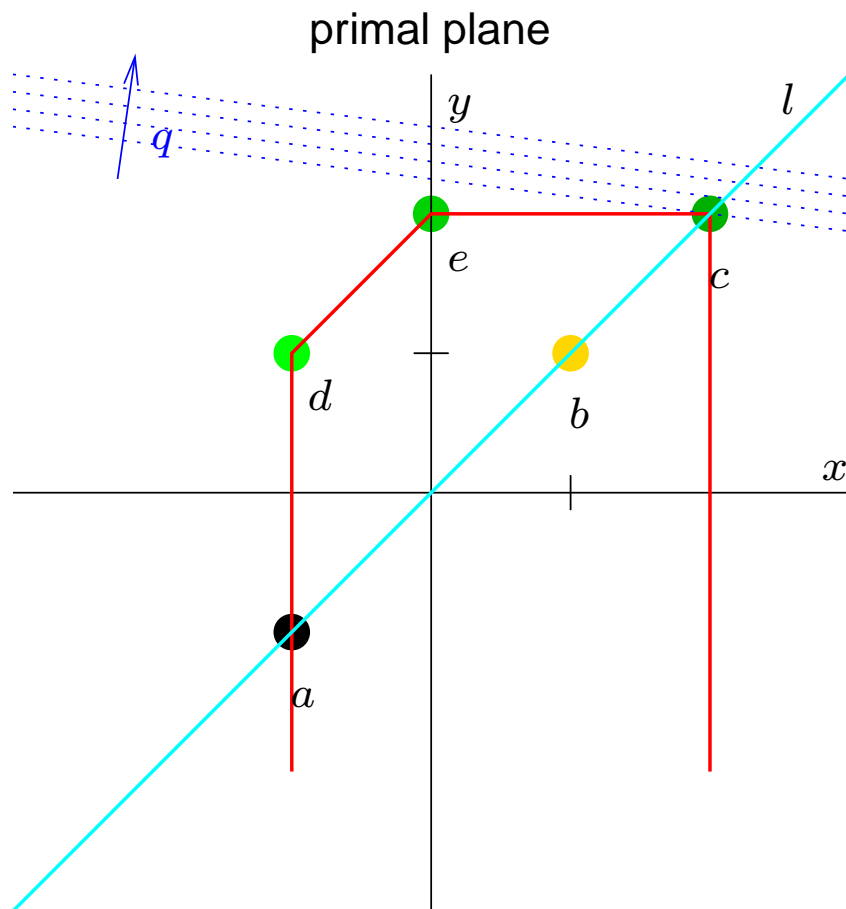
$O(\log n)$

O_A =Amortized

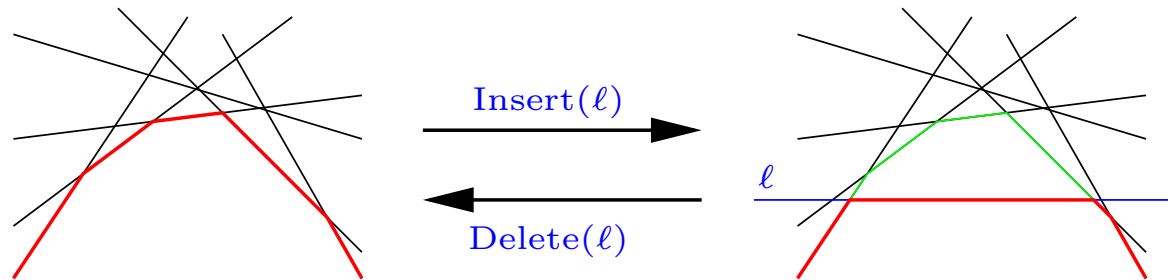
Query=Queries (a)–(e)

Duality Transformation

$p = (a, b) \in \mathbb{R}^2$ maps to $p^* := (a \cdot x - b = y)$



Dynamic Lower Envelope

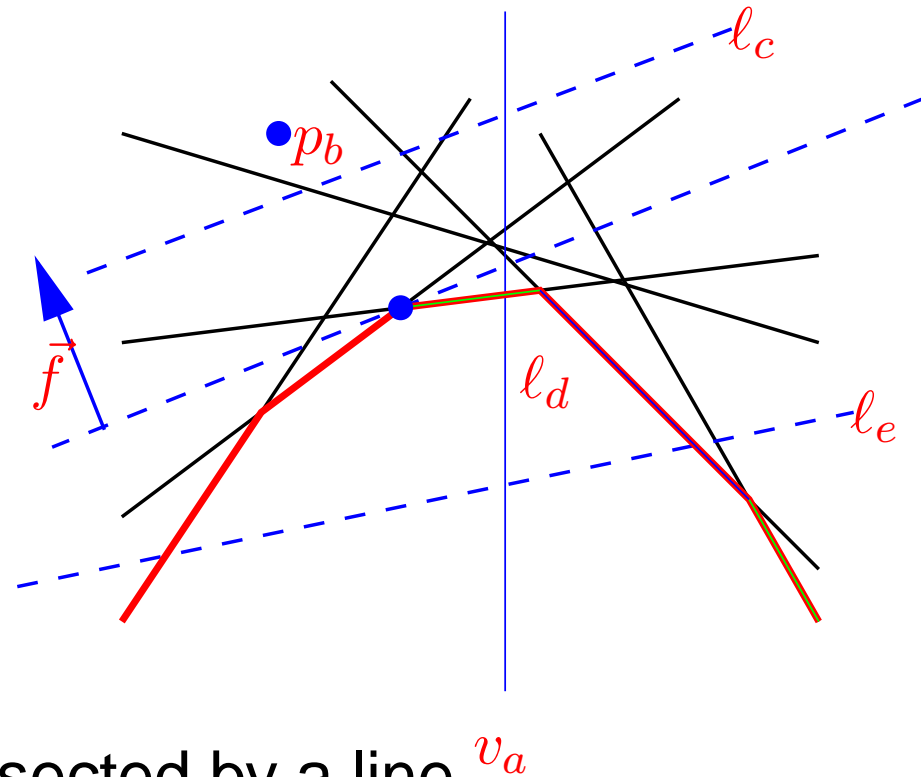


Updates

Insert and delete lines

Queries

- (a) Vertical line intersection
- (b) Is a point above $LE(S)$
- (c) Is a line above $LE(S)$
- (d) Next segments on $LE(S)$
- (e) The segments of $LE(S)$ intersected by a line
- (f) The extreme point of $LE(S)$ in some direction

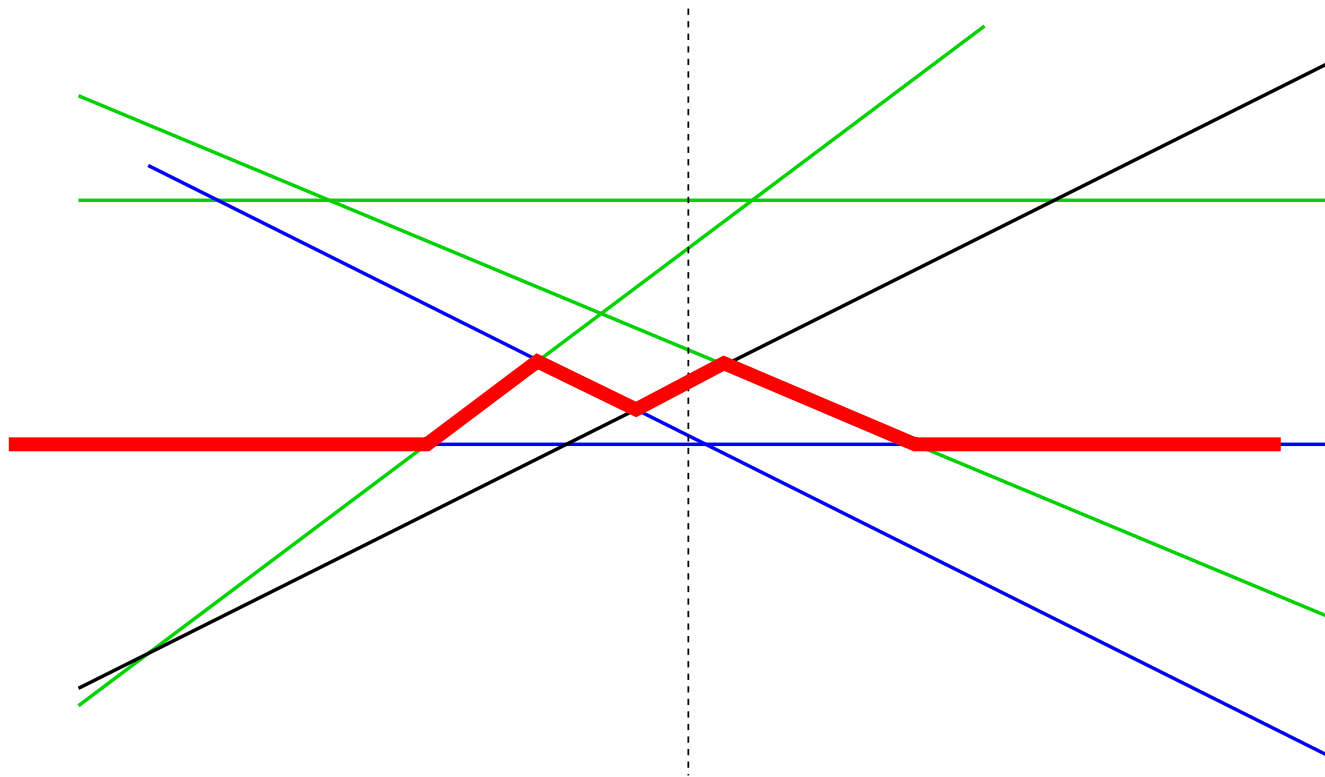


Application: k -level in the plane

Sweep-line algorithm Edelsbrunner and Welzl 1986

Previously $O((n + m)\alpha(n) \log n)$ expected time Har-Peled 1998

Now $O((n + m) \log n)$ for m segments on the k -level

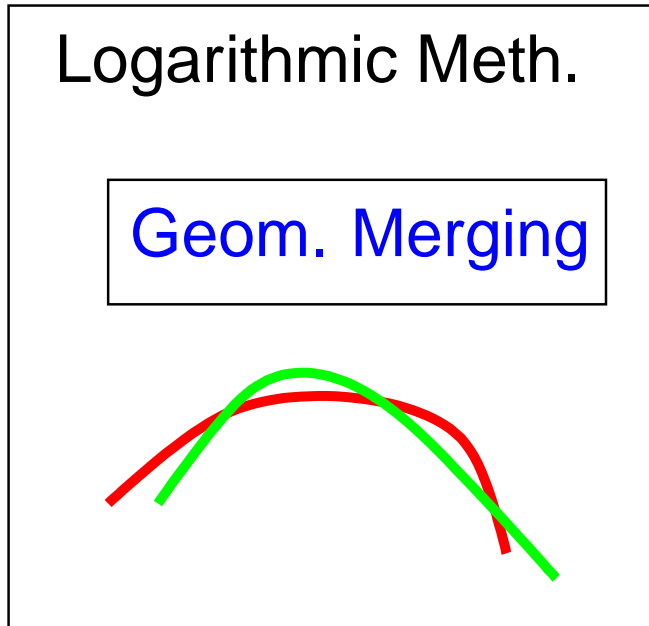


The 3-level of the 6 lines is depicted in thick red.

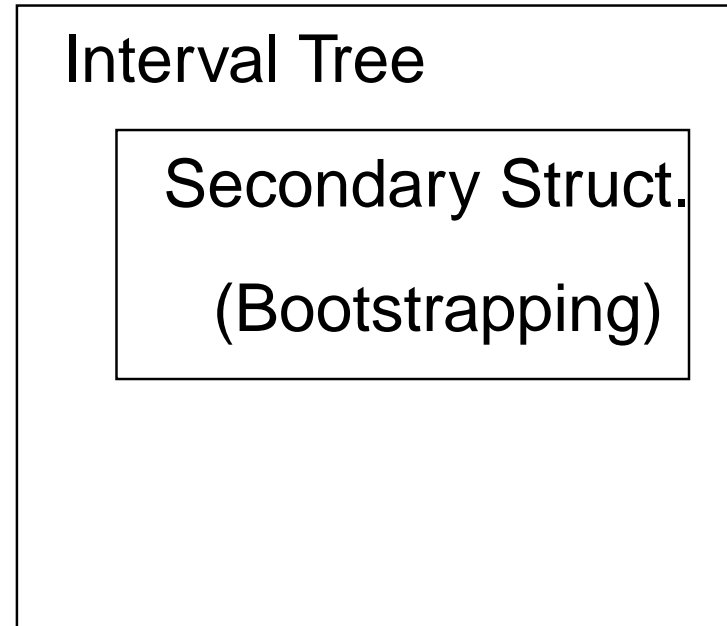
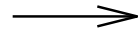
Overall Structure

Updates

Queries

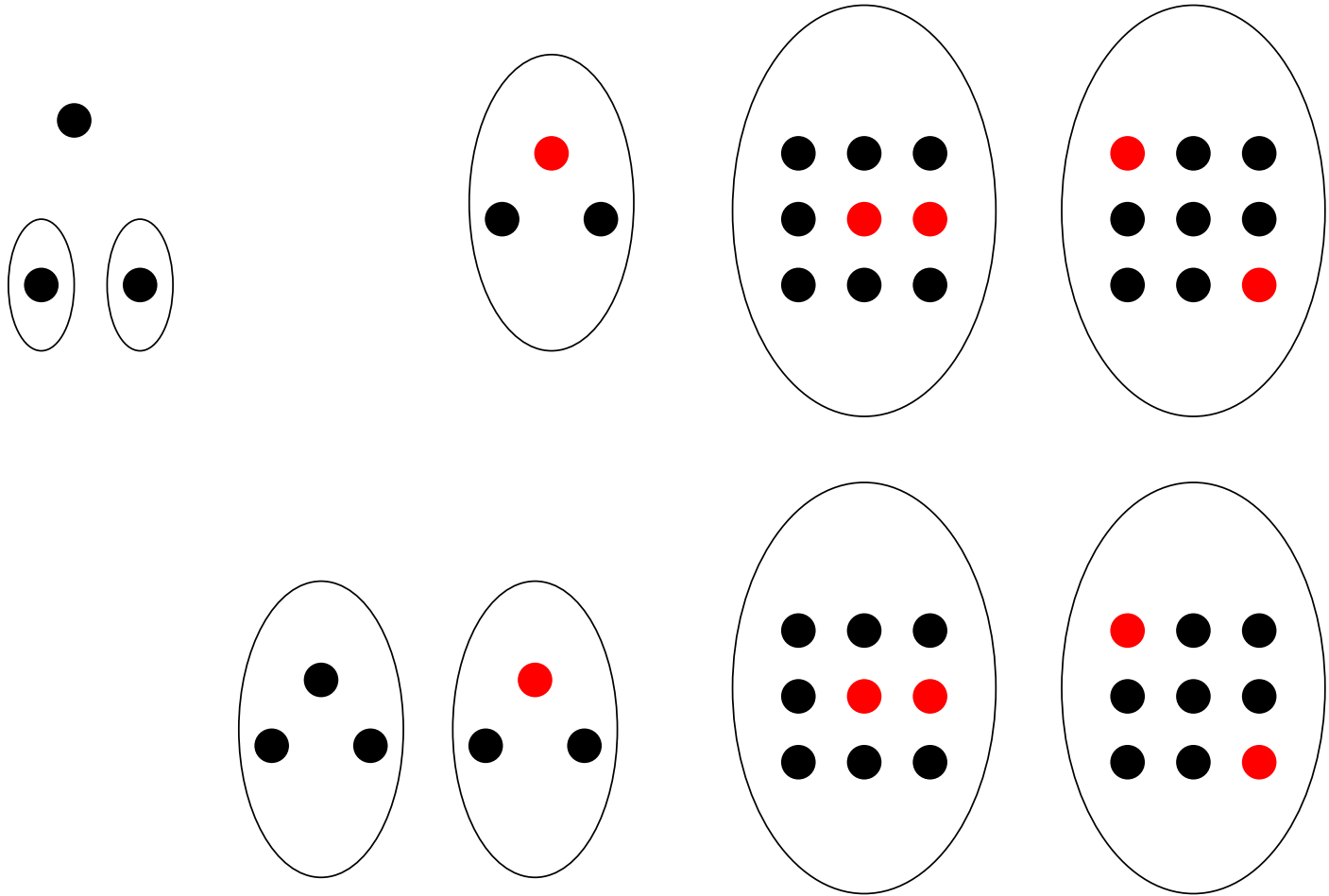


$\log^2 n$ explicit hulls,
deletion-only



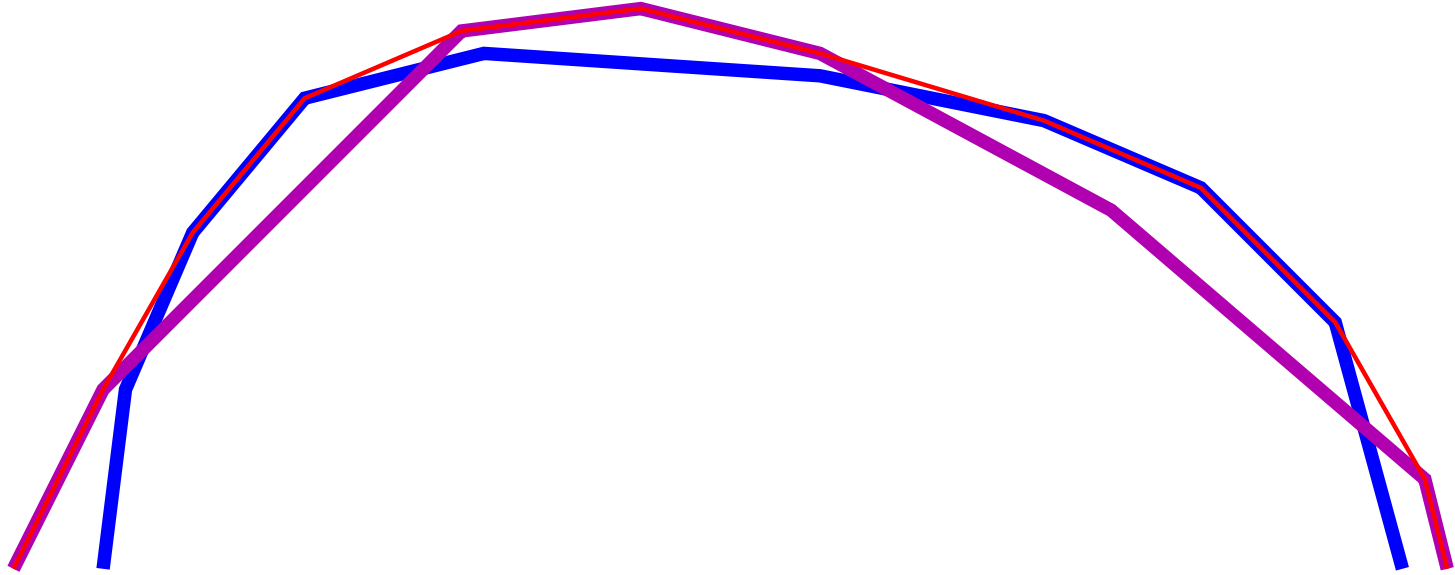
2× Bootstrapping:
Insert and Query $\log n$
Delete: $n \rightarrow \log^5 n \rightarrow \log n$

Logarithmic Method



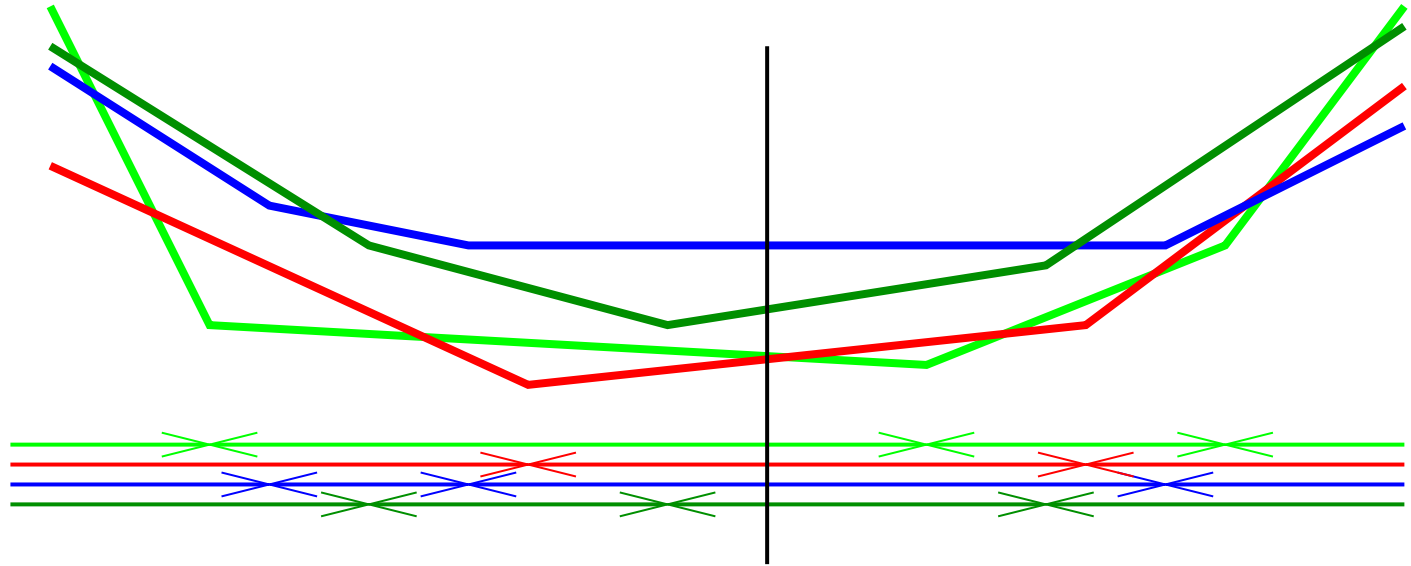
Bentley and Saxe 1980

Static Geometric Merging



Difficult: maintain under deletions of points

Combining Queries: Interval Tree



Task: combine the search on several lower envelopes into one search.

Follows ideas from [Chan 1999](#); different choice of parameters, save some work by relaxed placement and lazy movements:

exploit knowledge about the (dynamic of the) intervals

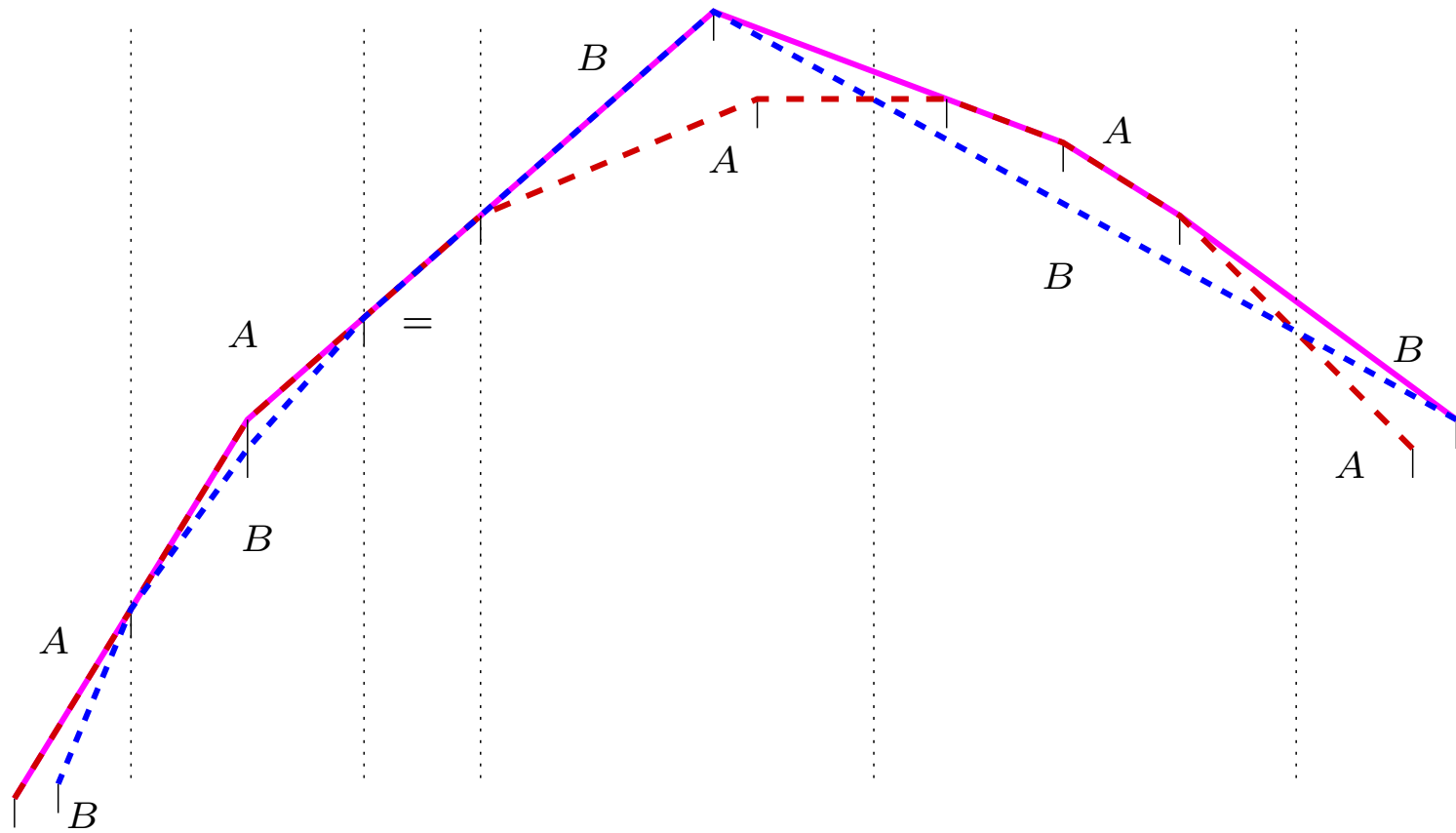
Semidynamic Merging

- Create Set(p) Create singleton set
- Merge(A, B) Combine data structures for A and B into one new for $A \cup B$
- Delete(r) Delete r from all merging structures

- Maintains list of points on the upper hull
- Works on binary merging forest
- Performance: $O_A(1)$ per element in the set

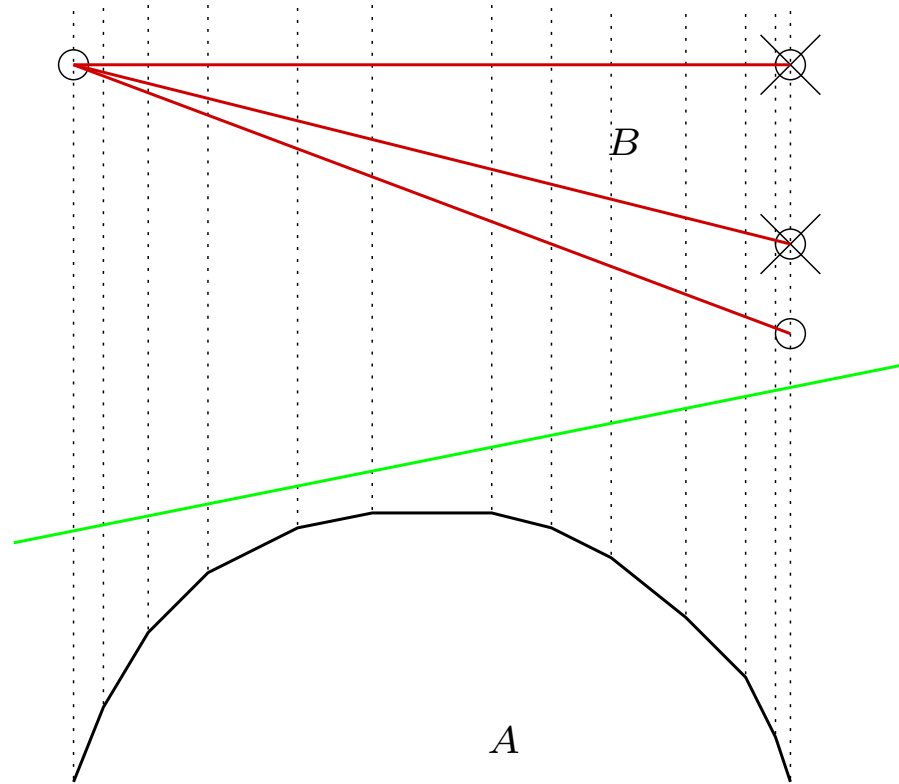
Core Problem

Maintain the equality points of the merging



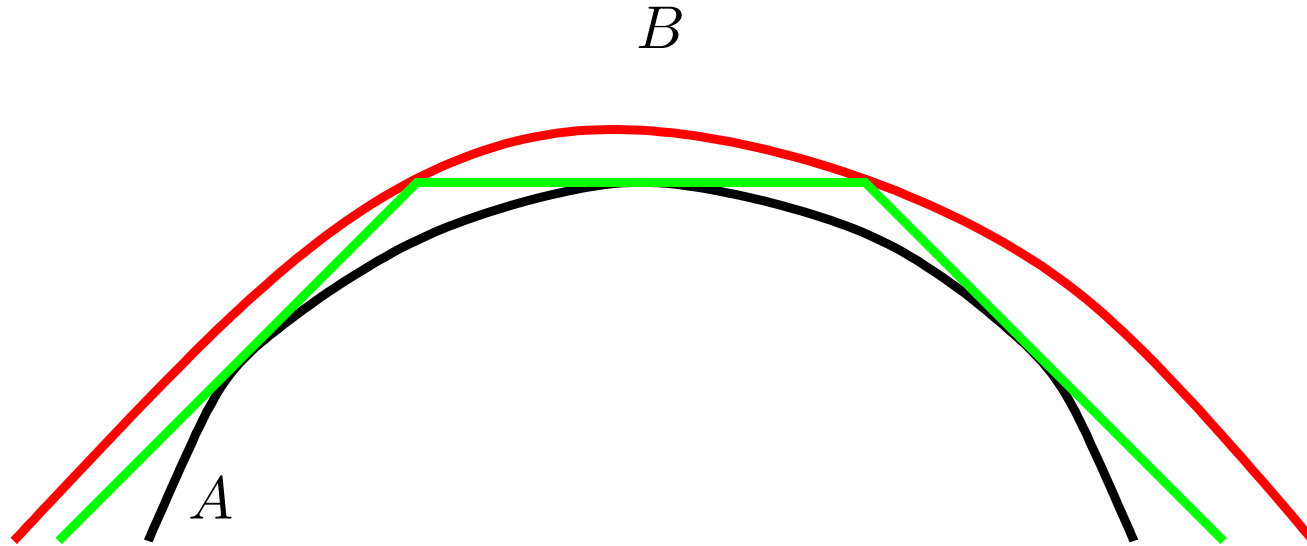
An equality oracle allows $O_A(1)$ per element hull maintenance

Separation Certificate



vertical certificates (sweep line): $O(n)$ per deletion
parallel tangent search: $O(\log n)$ per deletion
suspended search: in some variant $O(1)$ per deletion

Greedy Separation



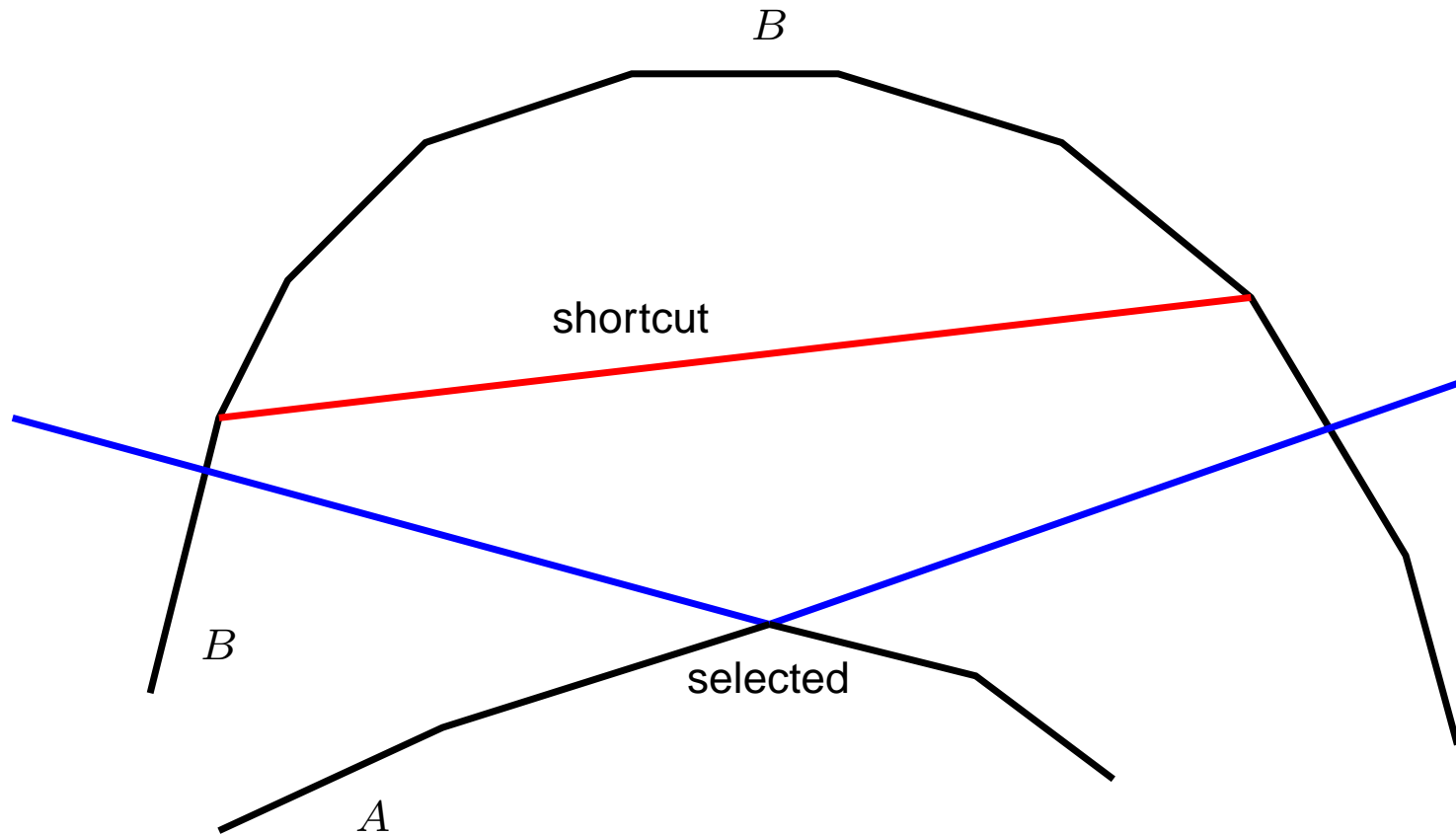
Greedily choosing tangent lines

Seems too rigid and sensitive for changes

Truss Bridge

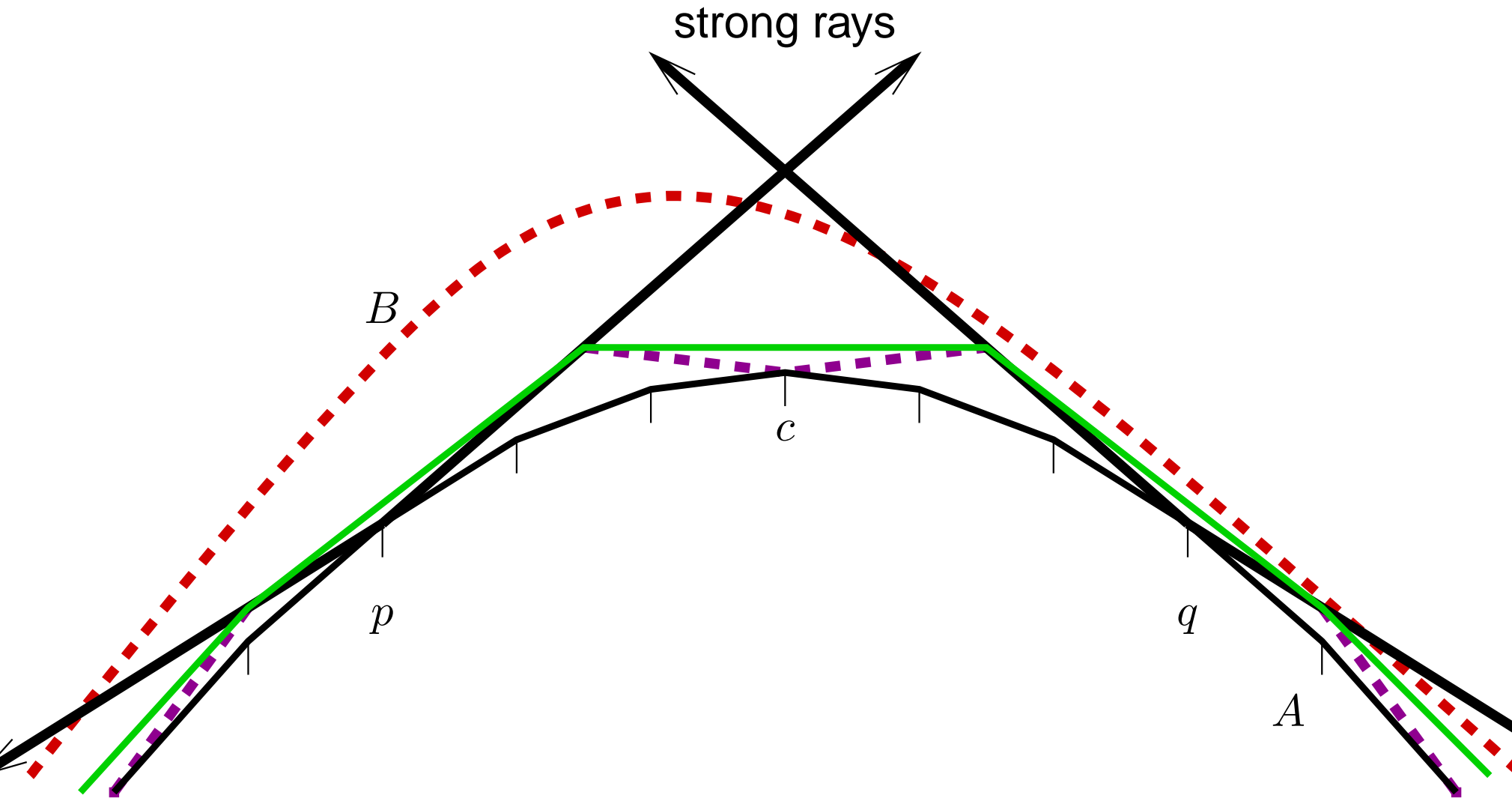
We call the construction **Truss**

Shortcuts



Shortcuts: Reducing the complexity of the outer hull

Dangling search



One deletion affects only constantly many strong rays

Splitters

Data structure(s) keeping (family of) sorted sequences

Elements e_i from a totally ordered universe

Build(e_1, \dots, e_n) $O_A(n)$

Split(t) $O_A(1)$

Extend(e_{n+1}) $O_A(1)$

Hoffmann, Mehlhorn, Rosenstiehl, Tarjan 1986

Split includes searching;

Dangling searches are suspended searches;

promise to split when finishing the search.

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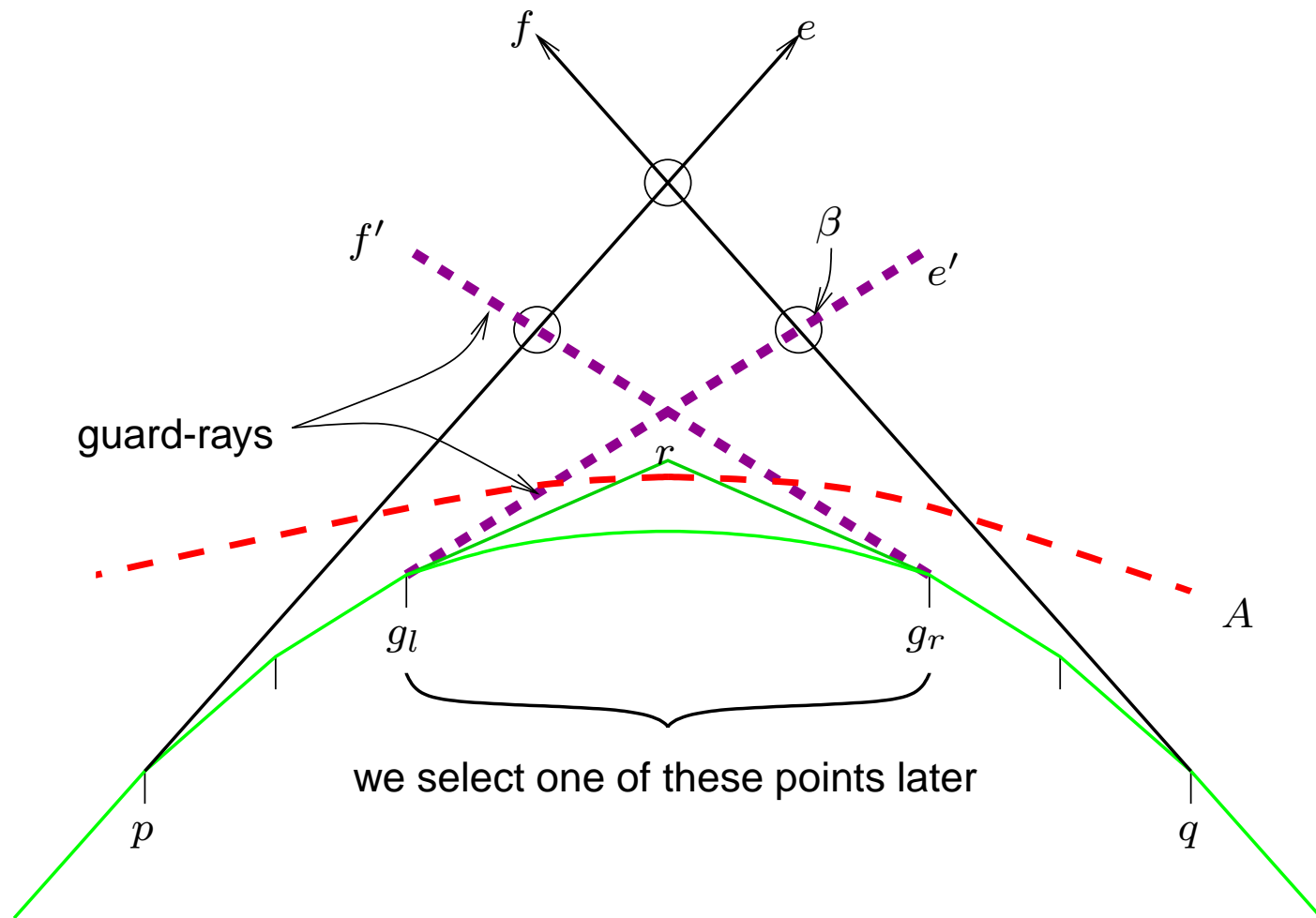
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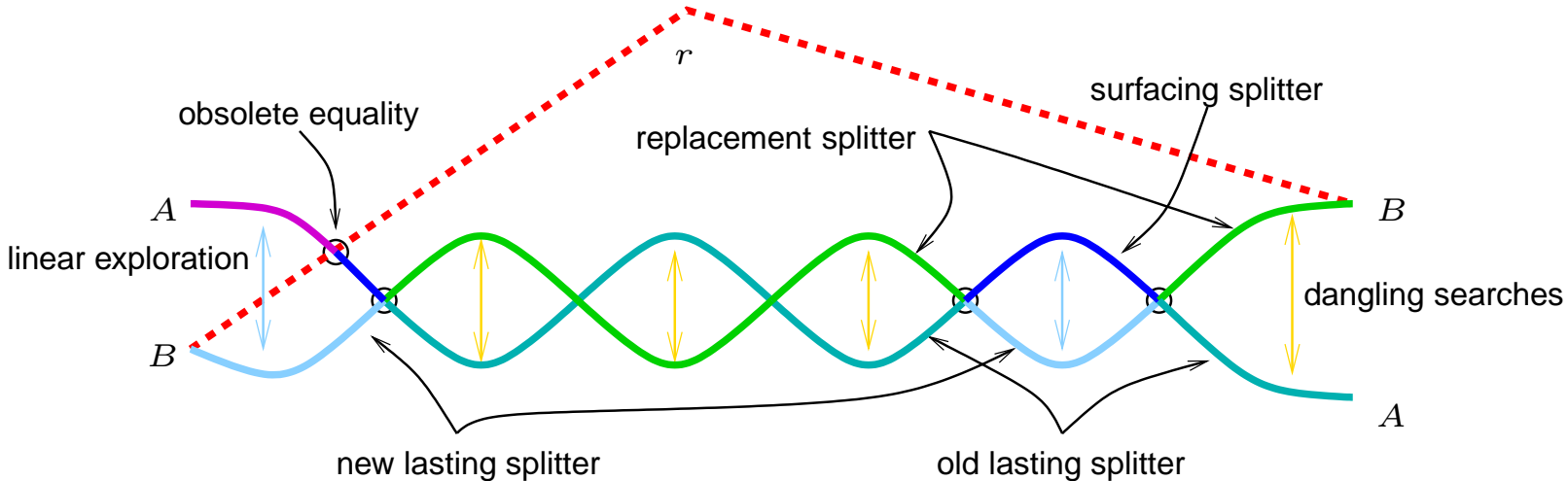
Dangling searches are suspended searches;
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Joining Strips



The geometric situation of loosing equality points.
We join the splitter over a dangling search:
Feasible because we promise to split.

Account: Life-cycle of a point $p \in A$



Schematic: the two hulls after one deletion

1. p becomes part of $UH(A)$.
 p in replacement splitter.
2. we realize $p \in UC_0(B)$.
 p in lasting splitter.
3. we decide to select p .
4. Delete on B : $p \notin UC(B)$.
 p in surfacing splitter.
5. p is hidden by a shortcut.
6. p is hidden by a bridge.
7. p gets on $UH(C)$.
8. The point p gets deleted.

New Techniques

Splitter: suspended (dangling) searches,
restricted join over search

Dynamization: Reuse of existing data structures

Geometric Merging: Focus on equality points,
selected points, dangling search, over-
approximation, shortcuts, truss

Interval Tree: Relaxed placement of intervals,
lazy movement, location justifier

Linear Space: Separators

Tight Lower Bounds

$q(n)$ be (amortized) query time

$I(n)$ amortized insertion time

$$q(n) = \Omega(\log n) \quad \text{and} \quad I(n) = \Omega\left(\log \frac{n}{q(n)}\right).$$

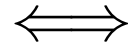
on algebraic real-RAM, off-line usage of data structure, reduction based.

Applies to Membership and Predecessor as well.

Decision Problem

For $k < n$

$$(x_1, \dots, x_n, y_1, \dots, y_k) \in \text{DISJOINTSET}_{n,k} \subset \mathbb{R}^{n+k}$$



for all i, j we have $x_i \neq y_j$

$\text{DISJOINTSET}_{n,k}$ has $\Omega(k^n)$ connected components

An algebraic computation tree has height $\Omega(n \log k)$.

(Ben-Or 83)

Summary and Open Problems

Presented

Dynamic Planar Convex Hull Data structure

Query $O(\log n)$

Insert $O_A(\log n)$

Delete $O_A(\log n)$

and a matching lower bound.

Open Problems

- Make it simple
- worst-case instead of amortized bounds
- more general queries
- explicit maintenance of the hull