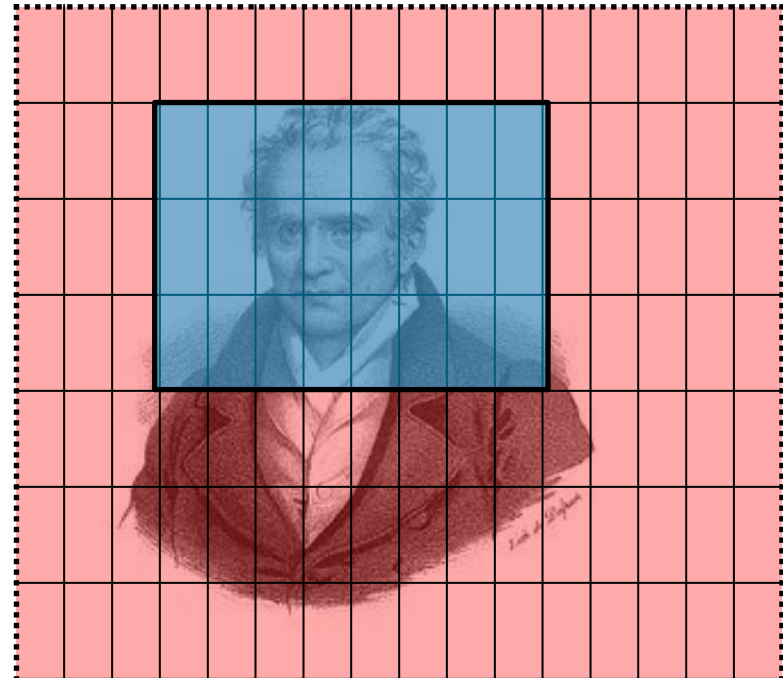


Data structures for totally monotone matrices

Submatrix maximum queries in (inverse) Monge matrices

Input: $n \times n$ (inverse) Monge matrix
 M represented implicitly
(constant time oracle access)

Output: data structure answering
maximum queries in any rectangular
submatrix of M



Submatrix maximum queries

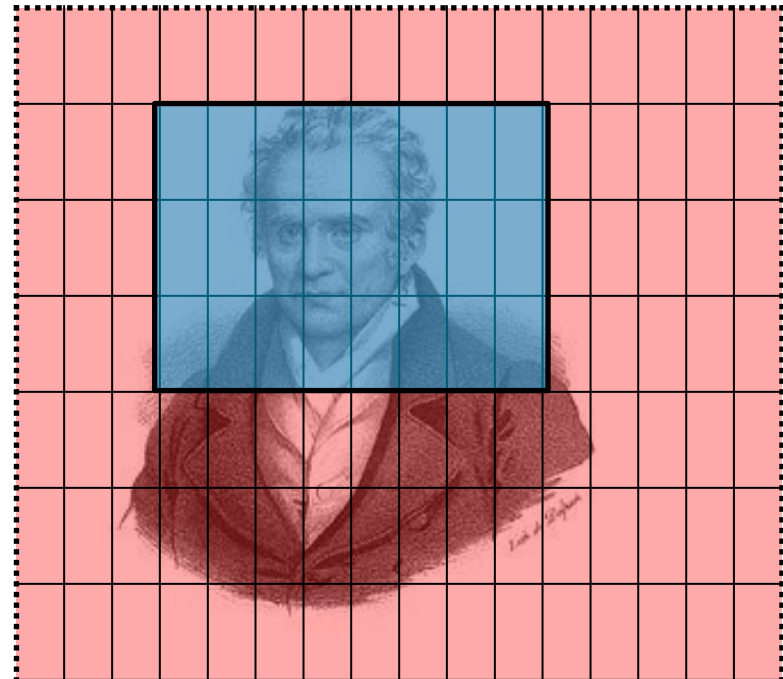
Main Theorem:

A data structure with the following performance:

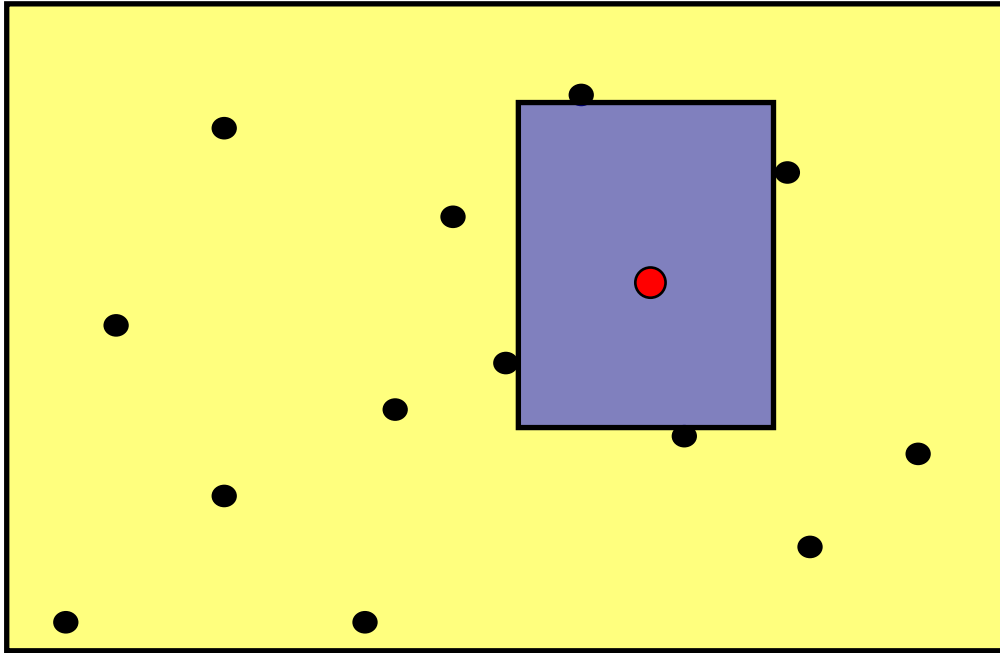
preprocessing time	$O(n \log^2 n)$
space	$O(n \log n)$
query time	$O(\log^2 n)$

Extends to **rectangular** matrices
and certain **partial** matrices

Application first...



Largest Empty Rectangle

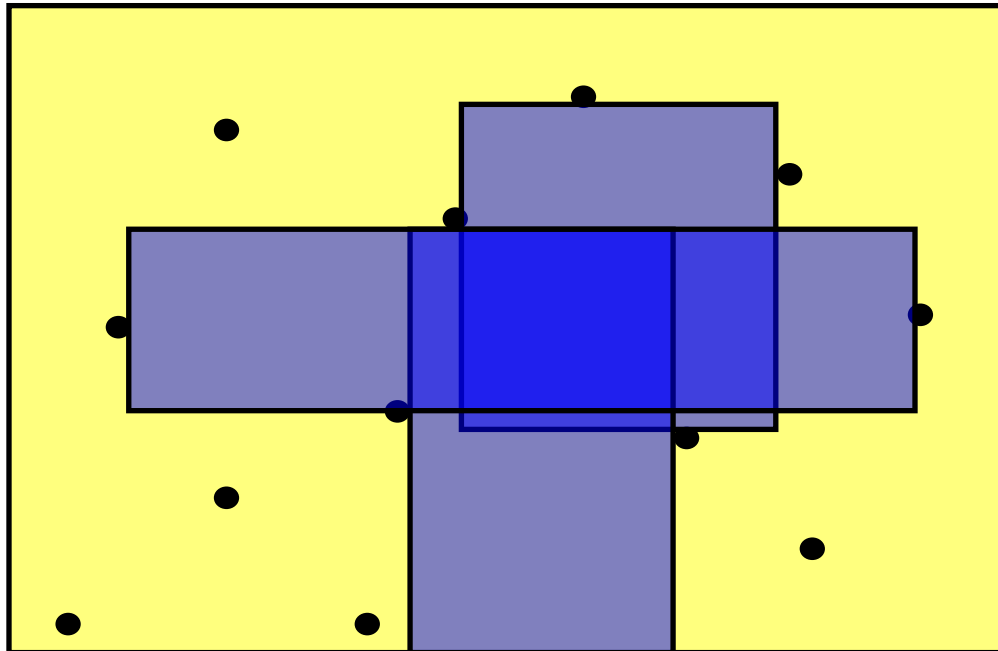


For a set P of n points in the plane build a data structure such that:

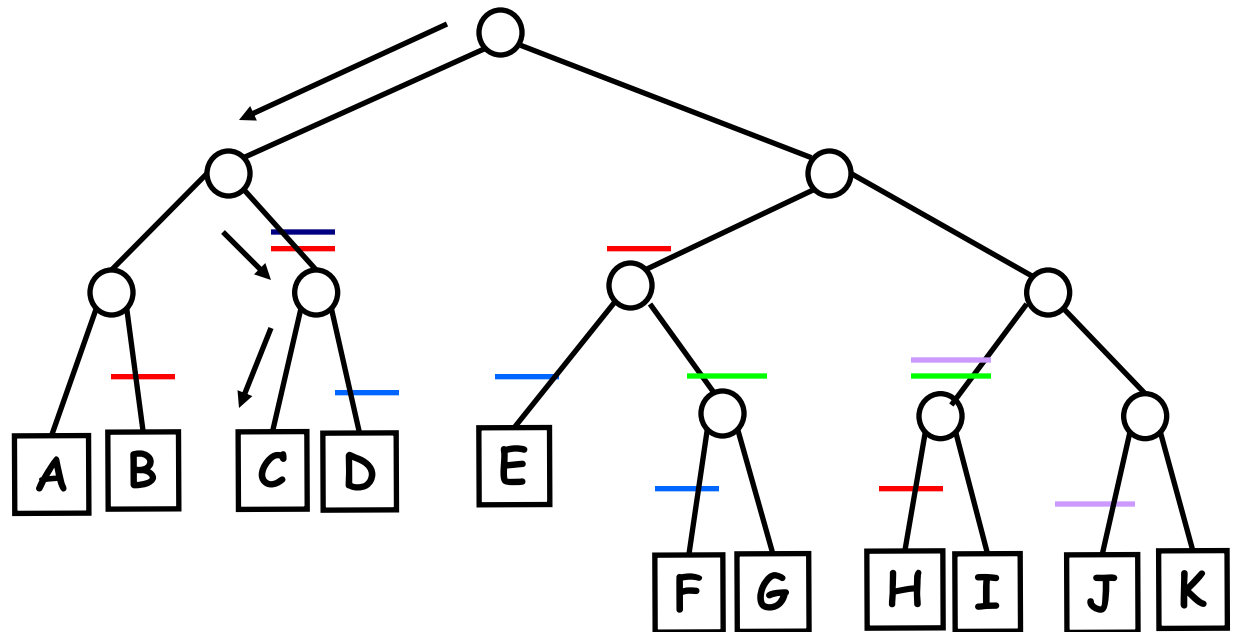
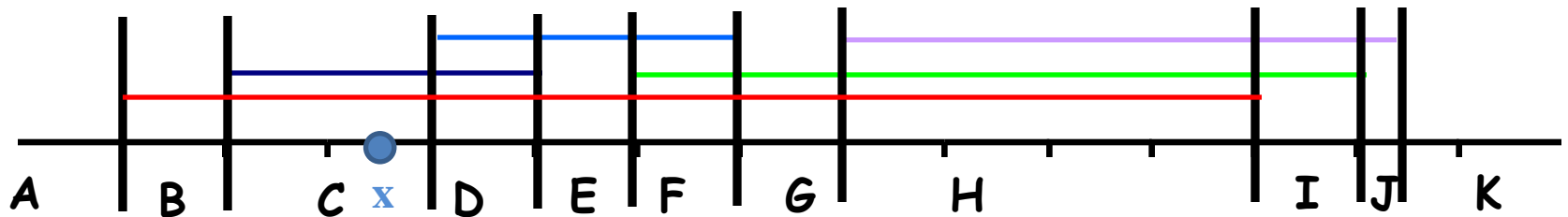
Given a query q , efficiently finds the largest empty (of pts of P) rectangle containing q

Naïve Approach

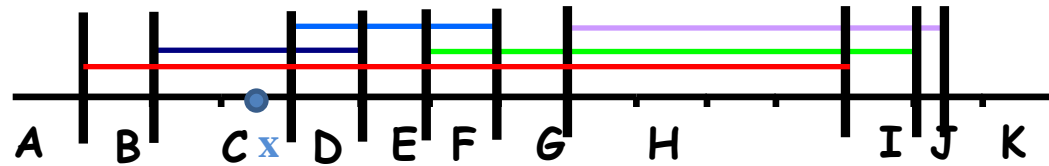
Store all N maximal empty rectangles in a two-dimensional **segment tree**.



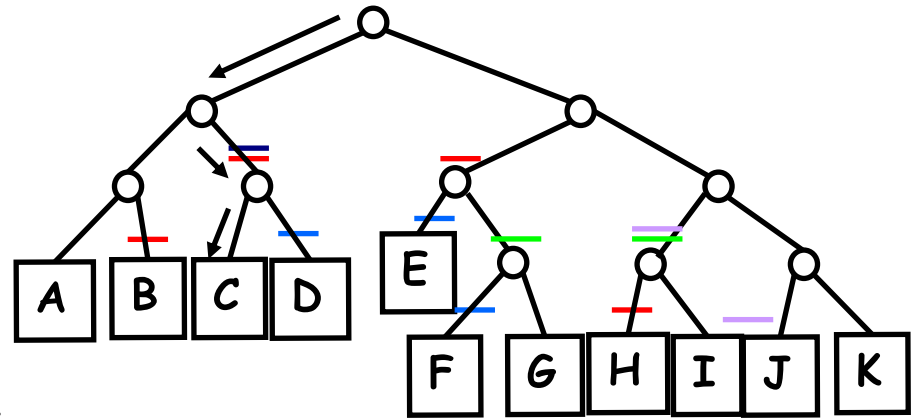
(detour) A 1D Segment tree



A 2D Segment tree

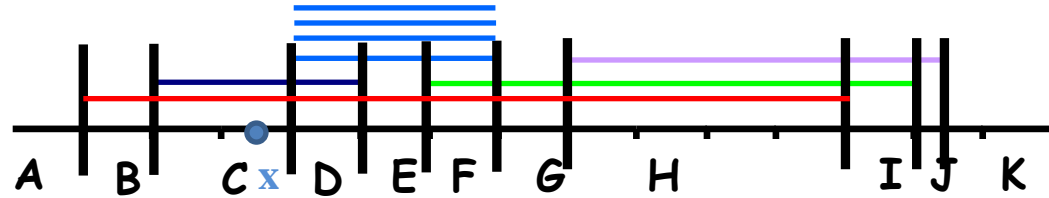


The size of the primary tree is prop. to the number of different x-coordinates

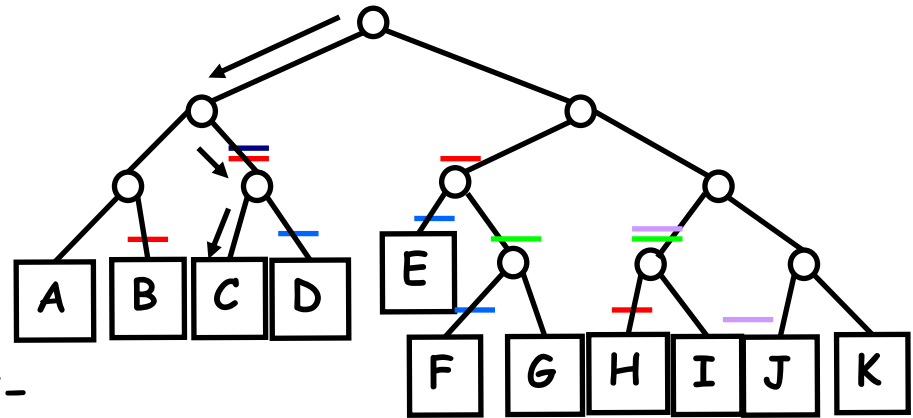


Since rectangles share x-coordinates there could be many rectangles mapped even to deep nodes.

A 2D Segment tree



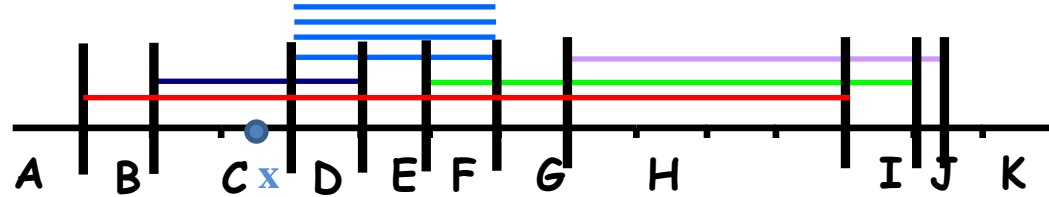
The size of the primary tree is prop. to the number of different x-coordinates



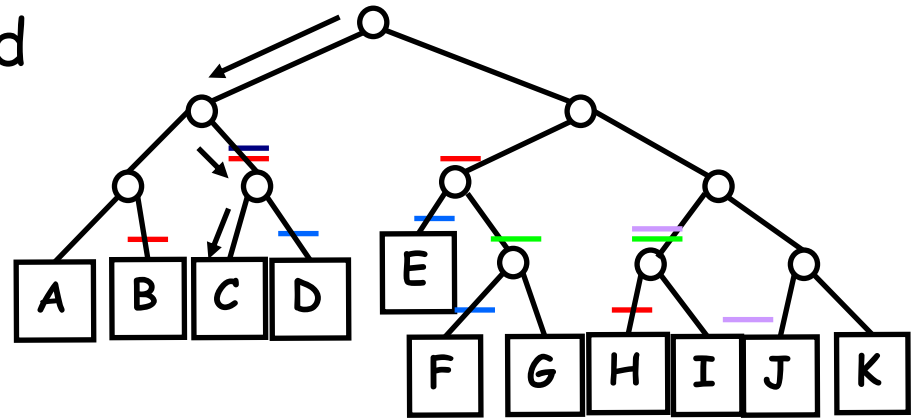
Since rectangles share x-coordinates there could be many rectangles mapped even to deep nodes.

In particular many rectangles can share their x-projections

A 2D Segment tree



Each rectangle is mapped to $O(\log(n))$ primary nodes



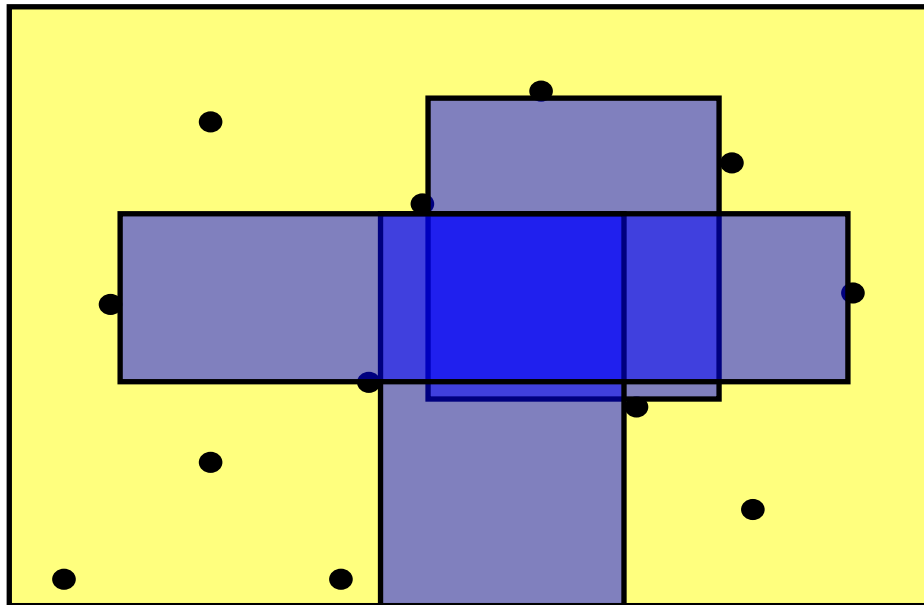
So if we have N rectangles the total size of the secondary trees is $O(N \log(n))$

Naïve approach

Store all N maximal empty rectangles in a
two-dimensional segment tree.

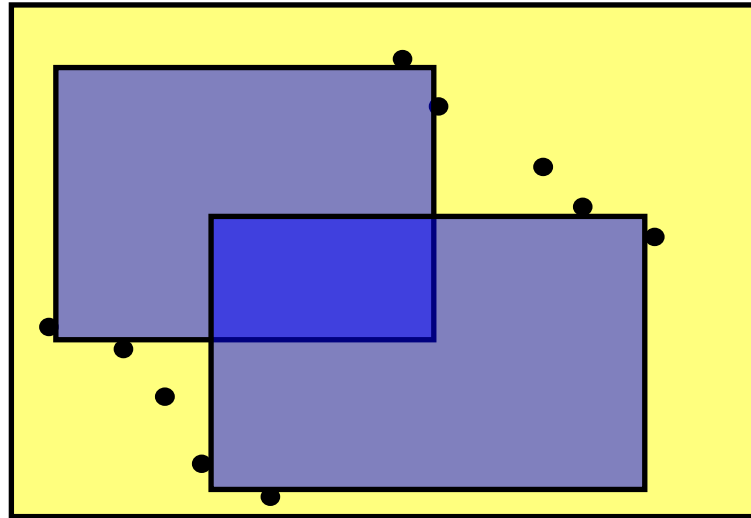
Space required is $O(N \log N)$, preprocessing time $O(N \log^2 N)$
and query time is $O(\log^2 N)$.

How many such rectangles could there be ?



Naïve approach

Caveat: There may be $N = \Theta(n^2)$ maximal empty rectangles

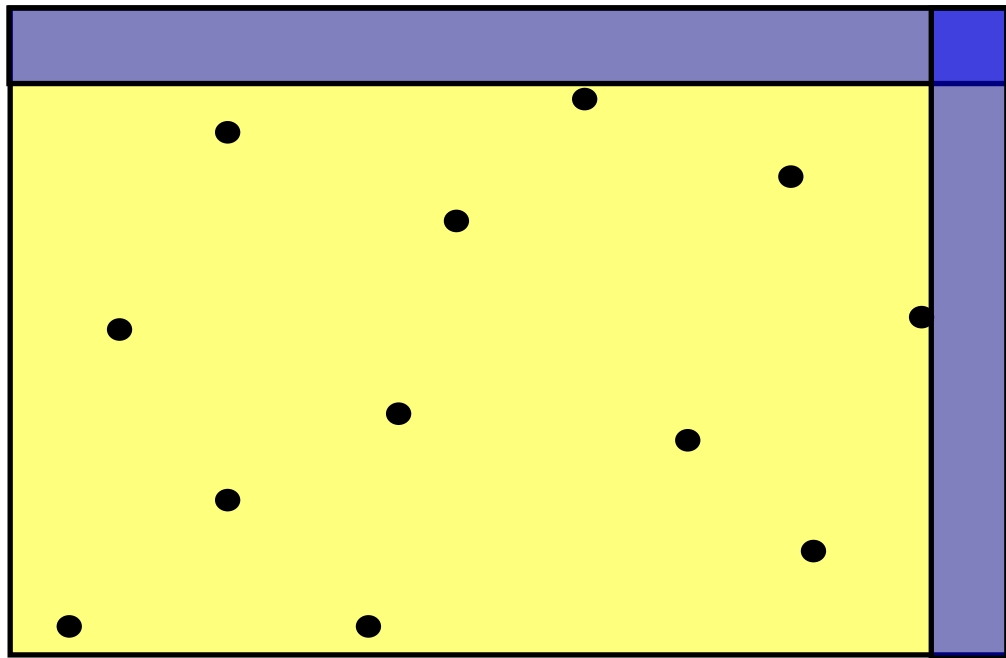


There could be a Quadratic number of rectangles supported by two points in each of the 1st and 3rd quadrants (symmetrically, 2nd and 4th quadrants)

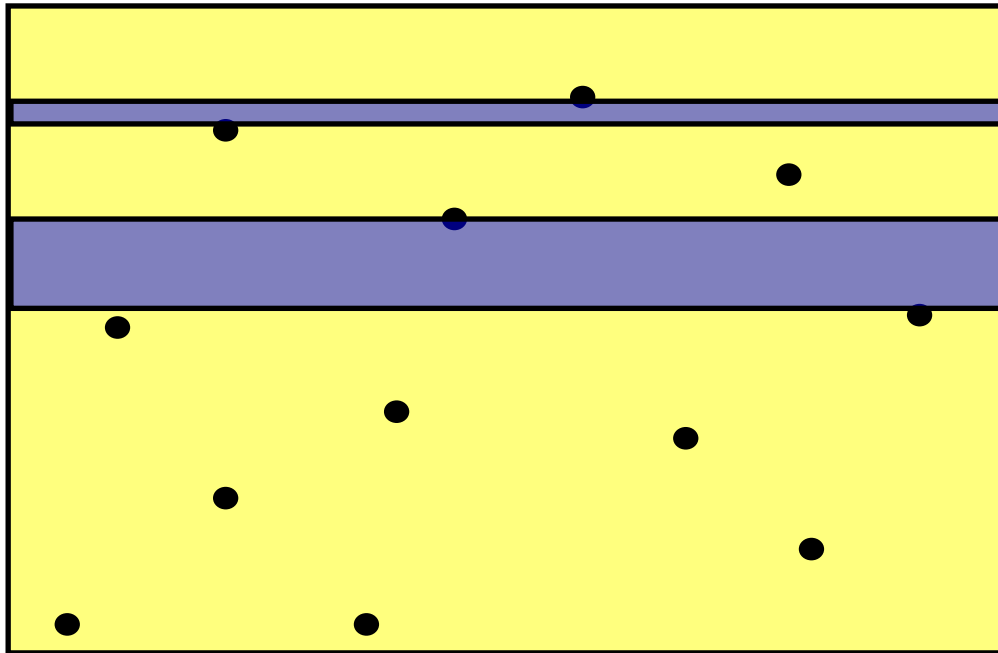
Different kinds of rectangles

- We will classify them, starting with rectangles that have an edge on the boundary

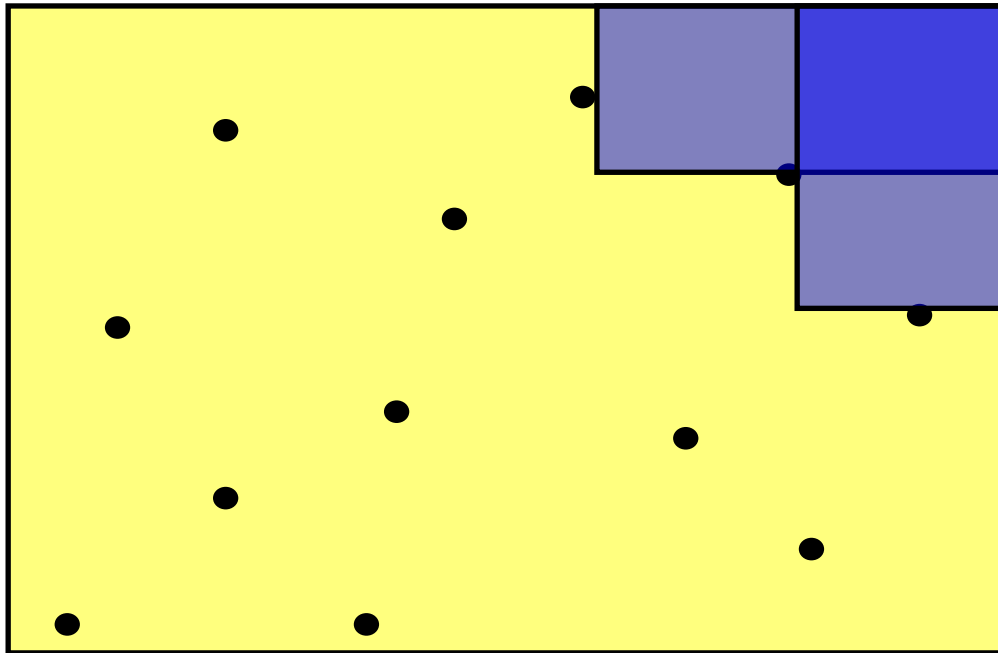
3 Edges on the boundary



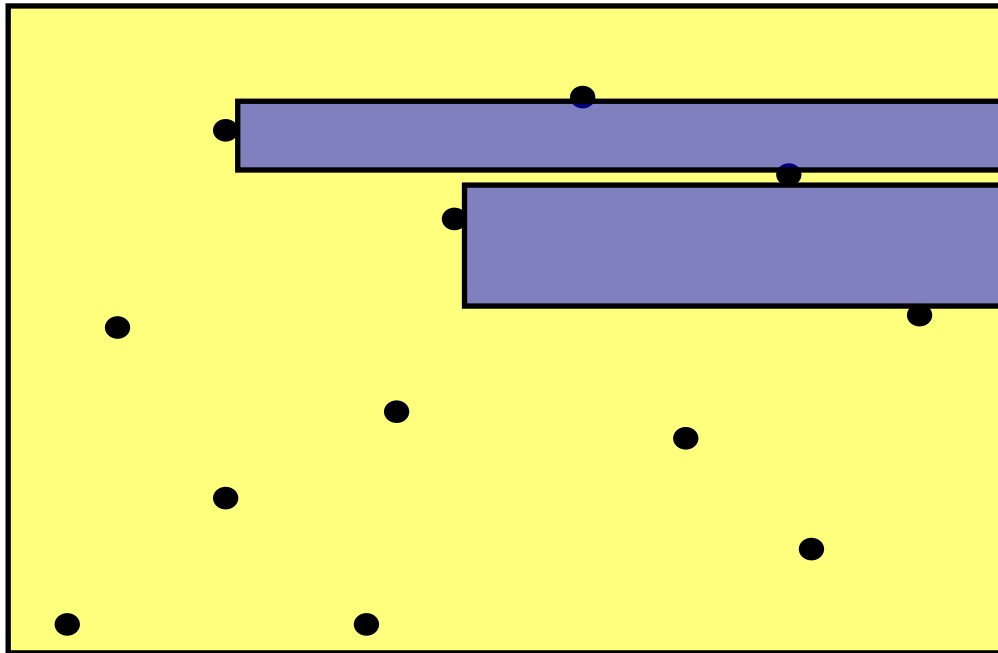
2 edges on the boundary (type 1)



2 Edges on the boundary (type 2)



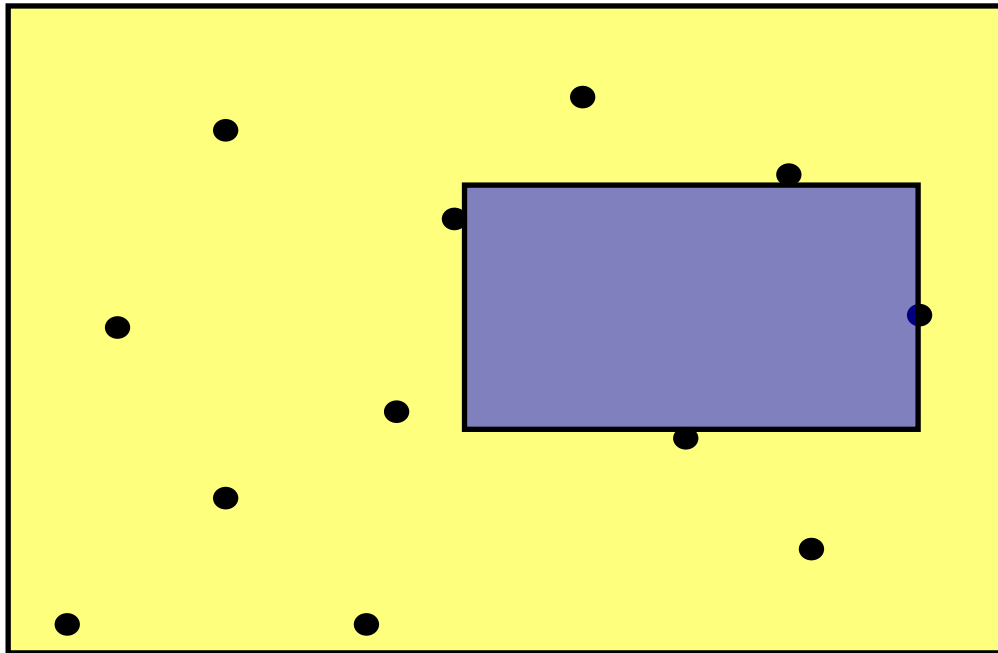
1 Edge on the boundary



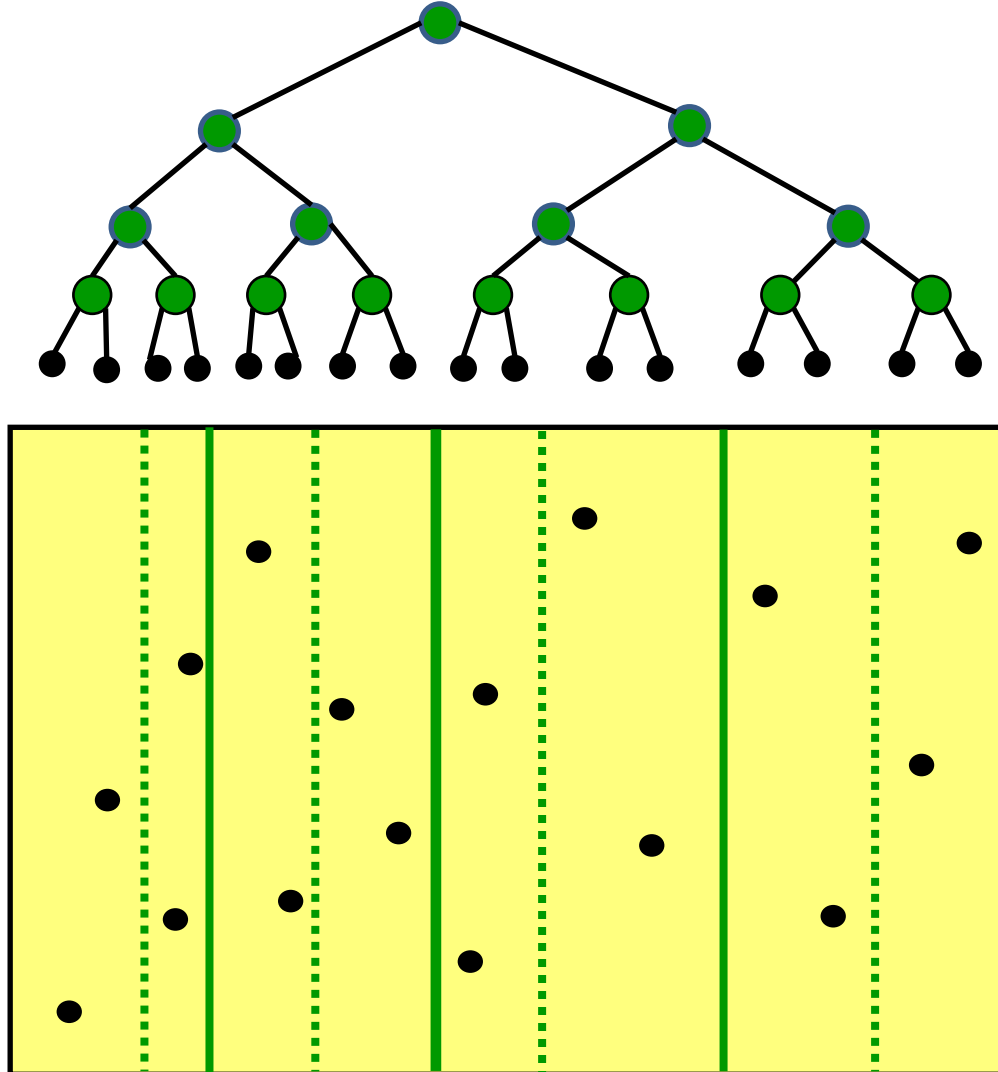
Rectangle with edges on the boundary (summary)

- $O(n)$ rectangles, compute them in $O(n \log(n))$ time and put them in the segment tree in $O(n \log^2(n))$

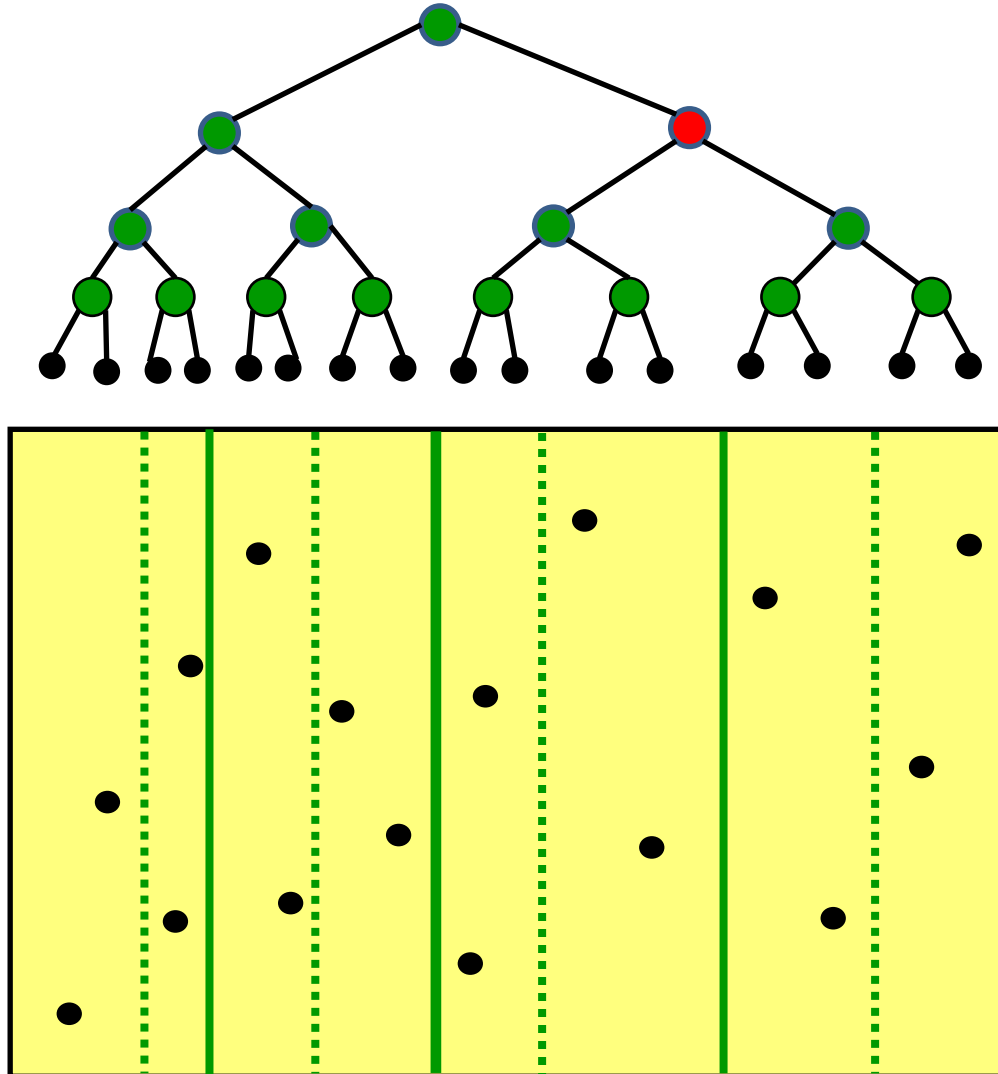
Rectangle without edges on the
boundary



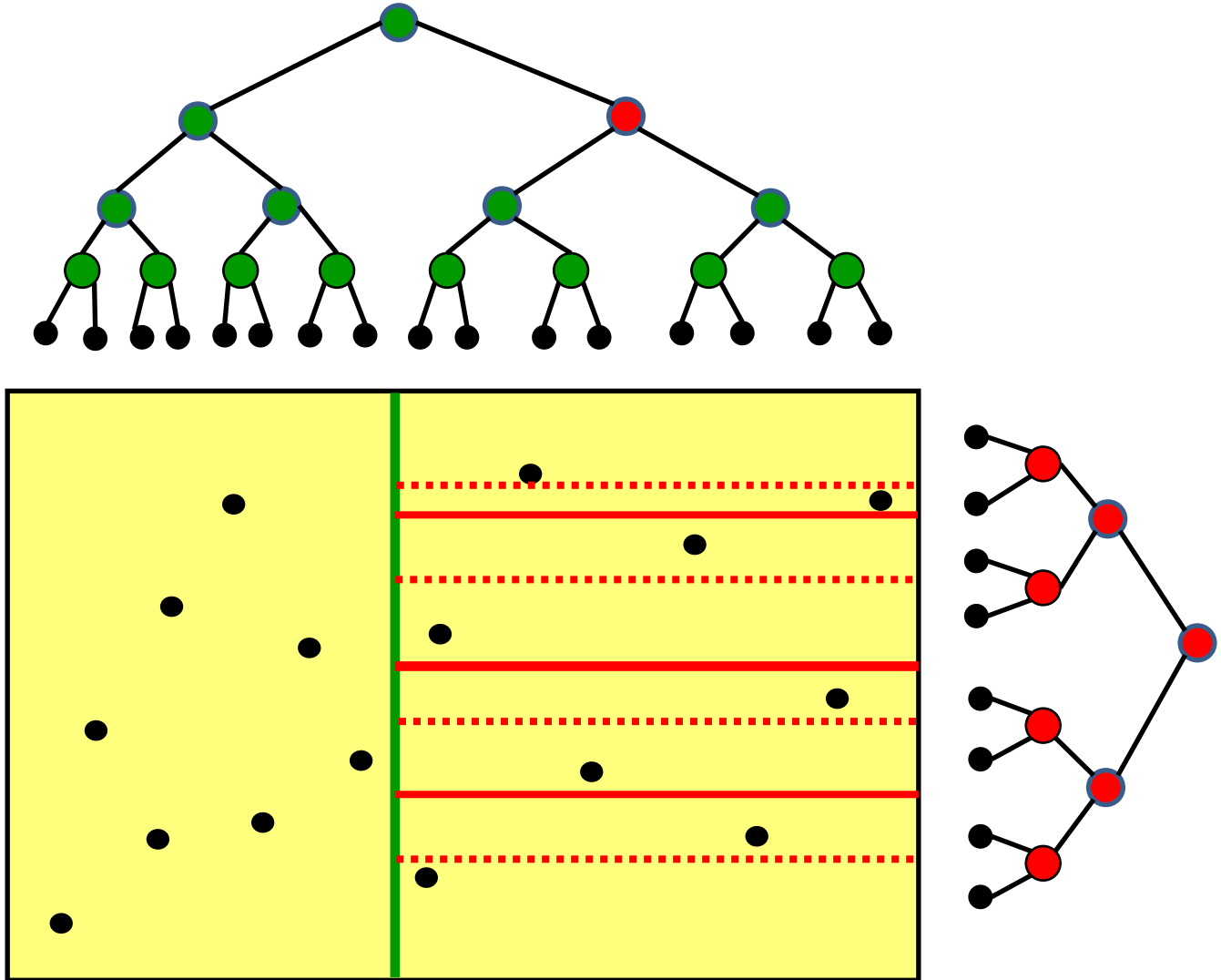
2D Range tree



2D Range tree

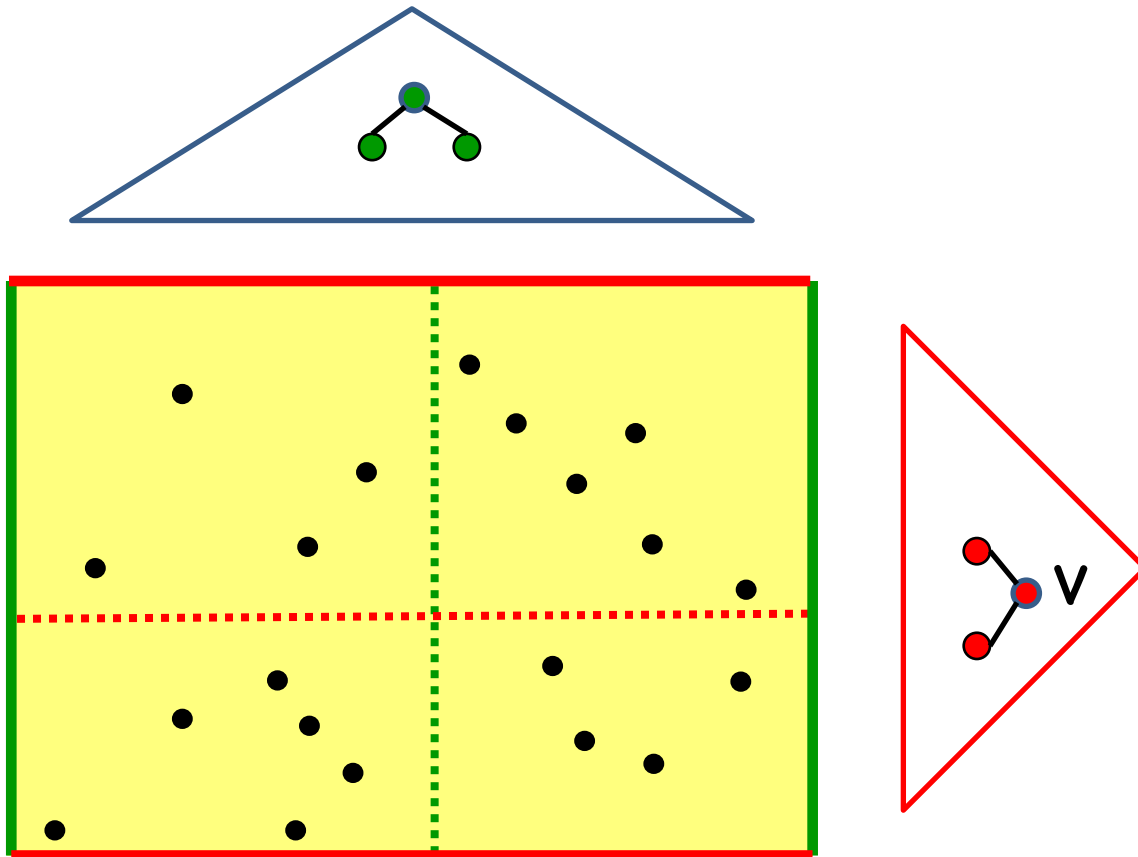


2D Range tree



An internal secondary node in the range tree

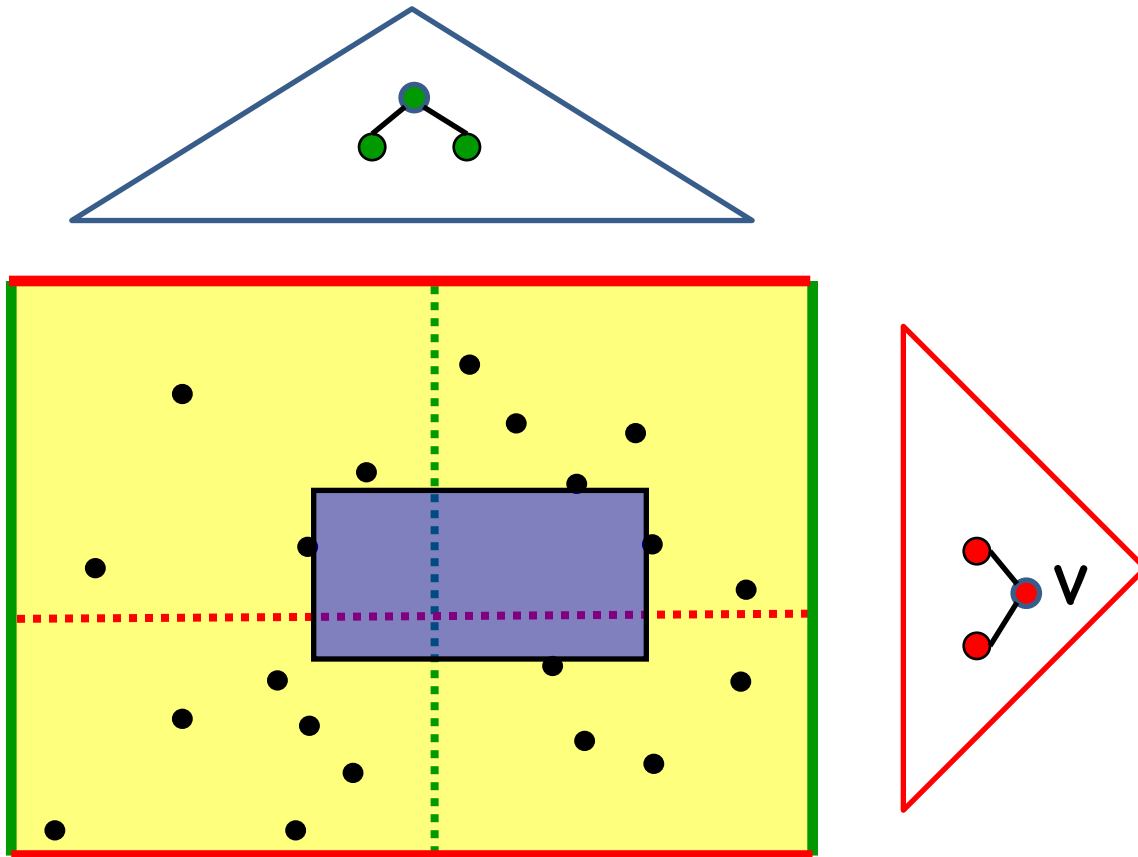
Corresponds to some rectangle with two dividers



A maximal empty rectangle

Contains the "origin" in some internal node v

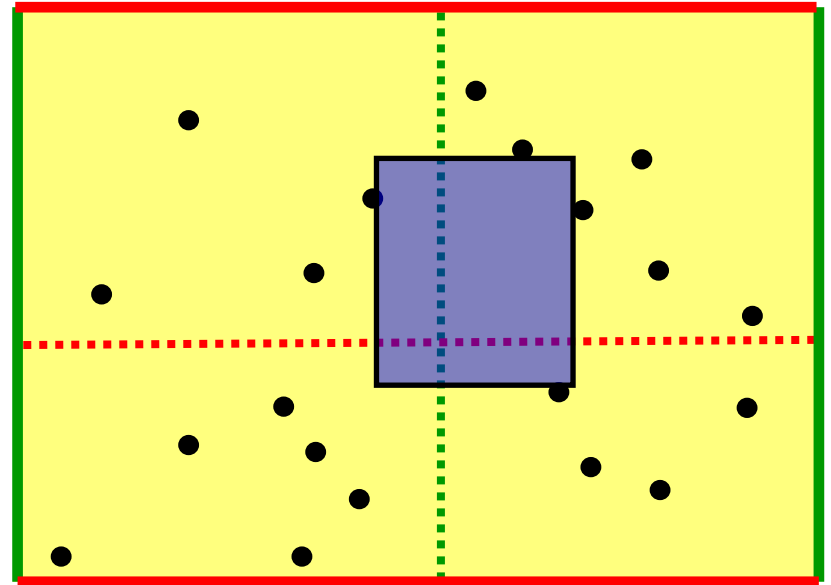
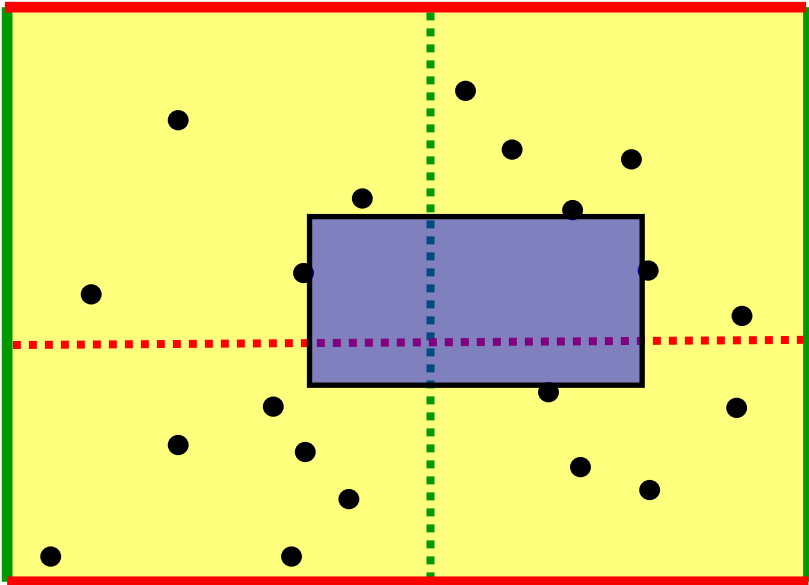
Each node is responsible for the maximal empty rectangles containing its "origin"



Classify maximal empty rectangles containing the origin

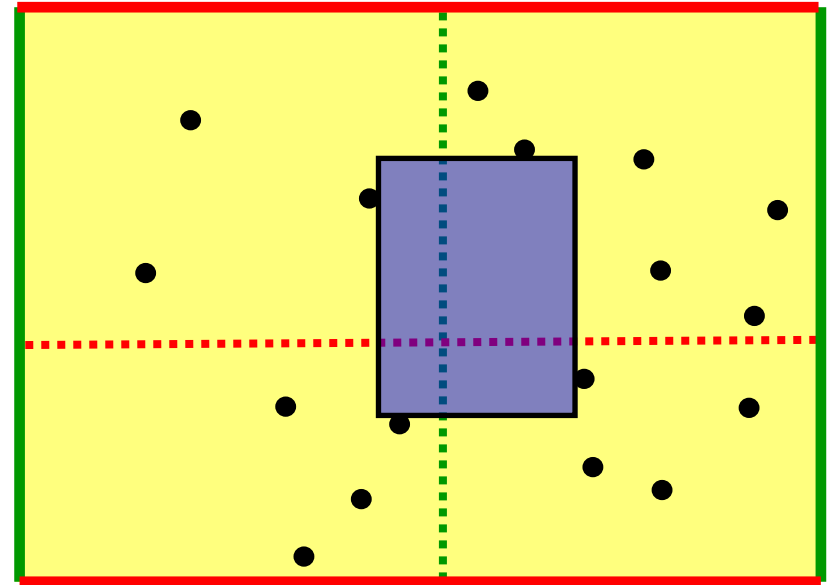
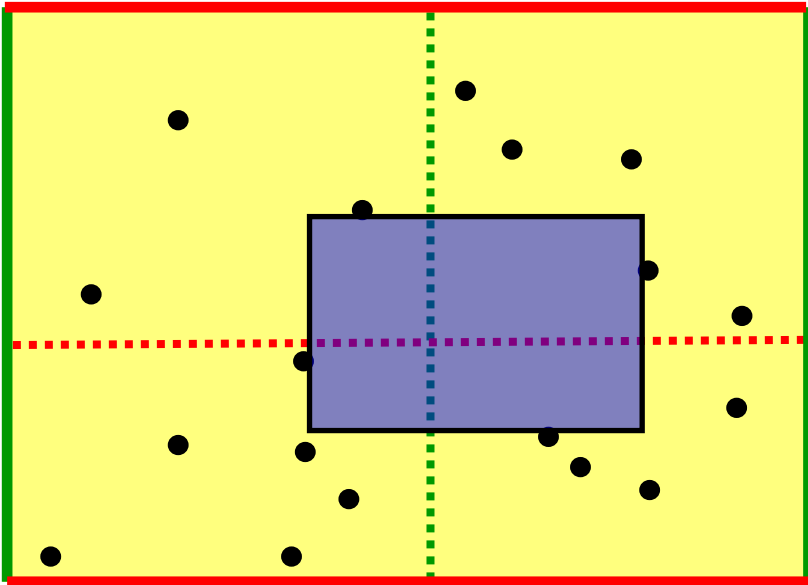
We are now focusing on one secondary node v of the range tree, n_v is the size of the subtree of v

3 points on one side of a divider



$O(n_v)$ of these, can compute then in $O(n_v \log n_v)$ time

1 point in each quadrant

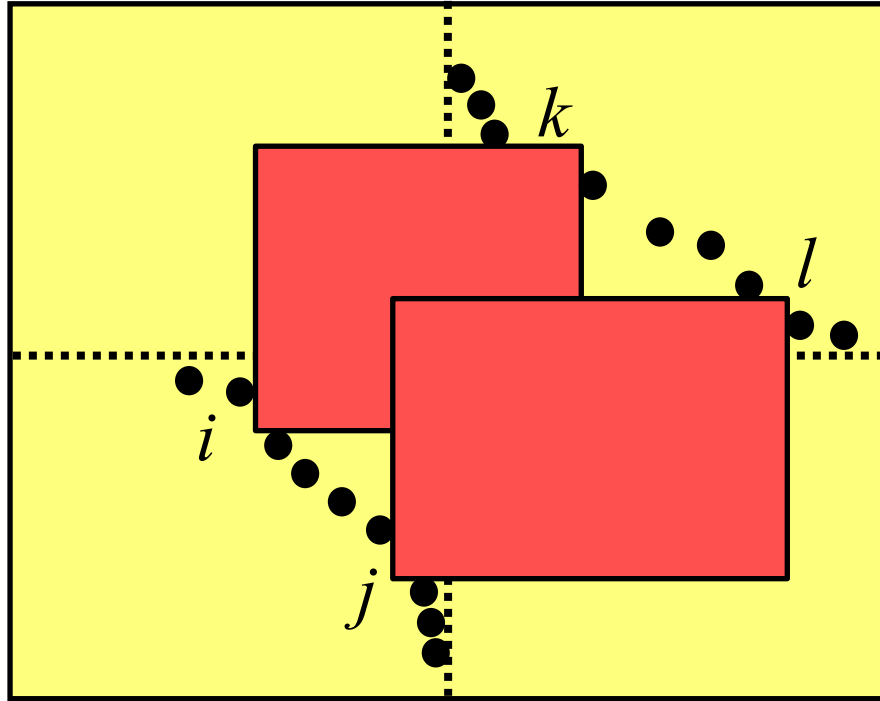


Linearly many $O(n_v)$, can compute then in $O(n_v \log n_v)$ time

Summary so far

- Summing up over the nodes of the range tree we have $O(n \log^2(n))$ rectangles that we compute in $O(n \log^3(n))$ time
- We insert them into the segment tree in $O(n \log^4(n))$ time
- The segment tree takes $O(n \log^3 n)$ space and can answer a query in $O(\log^2 n)$ time

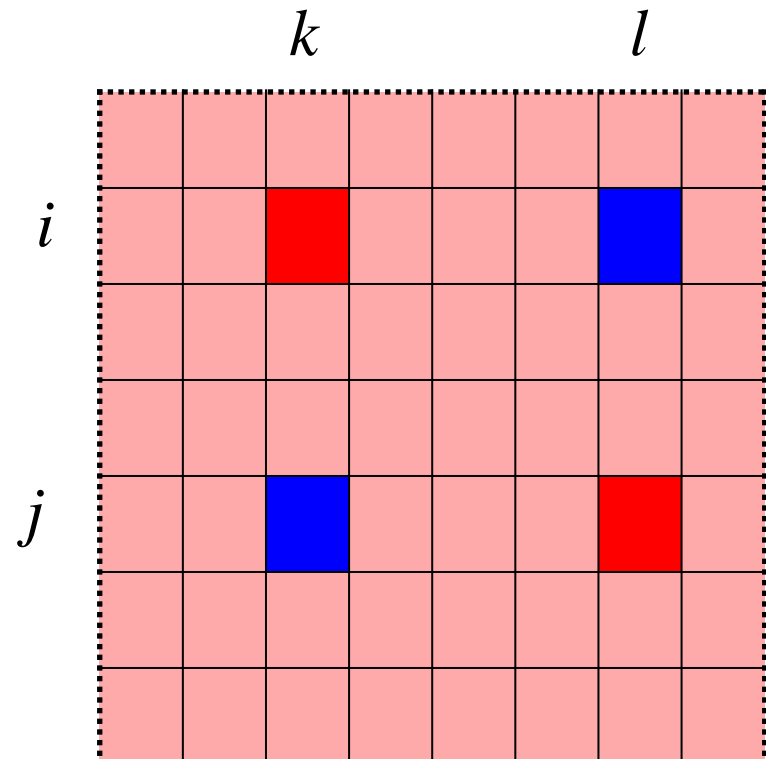
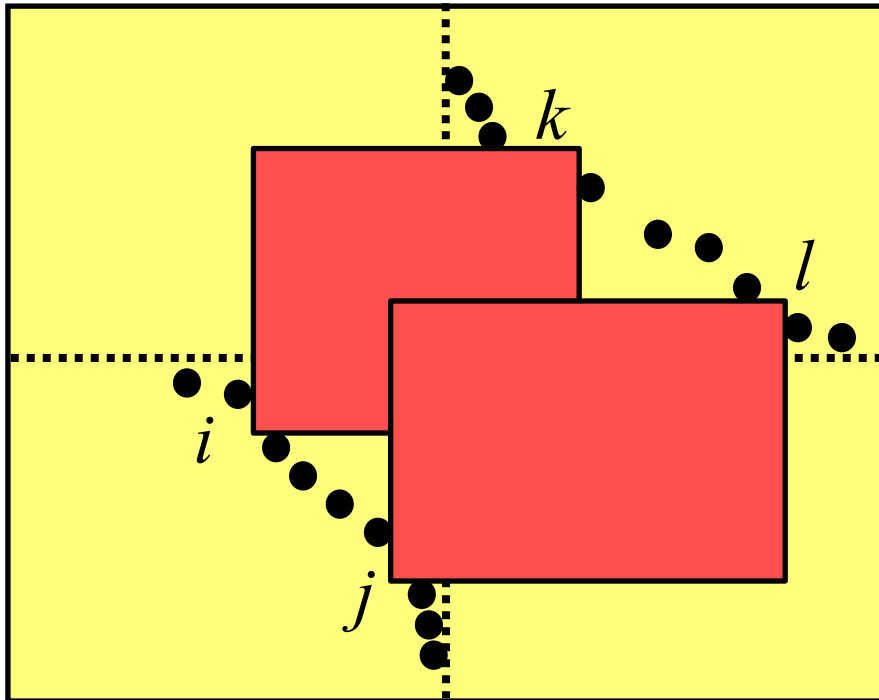
2 Points in each quadrant



We can compute the chains in $O(n_v \log(n_v))$ time, but there are too many rectangles to put in the segment tree

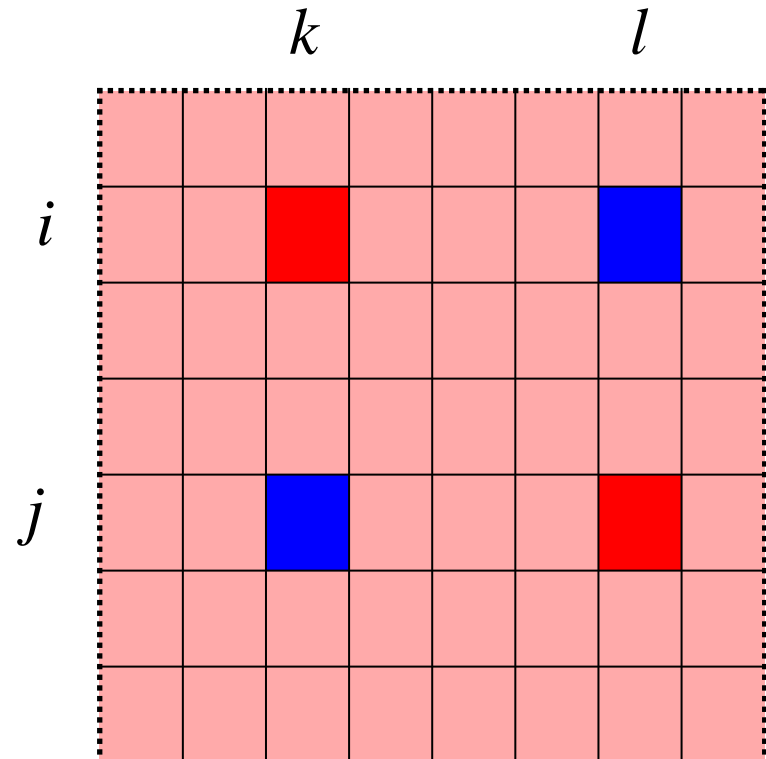
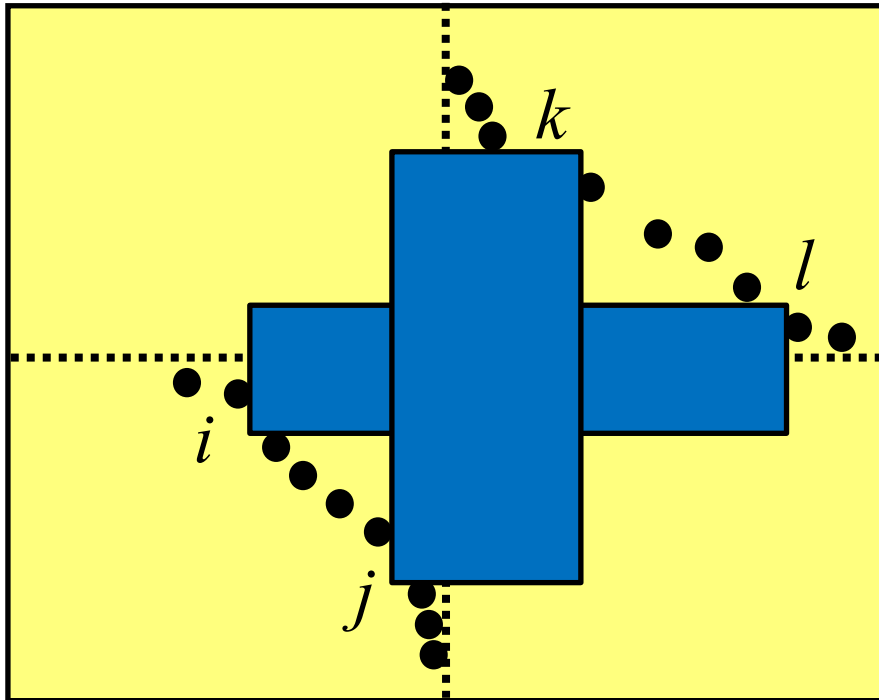
The (inverse) Monge property

Define a matrix M such that M_{ik} is the area of maximal rectangle supported by i^{th} pair in 3rd quadrant and k^{th} pair in 1st quadrant



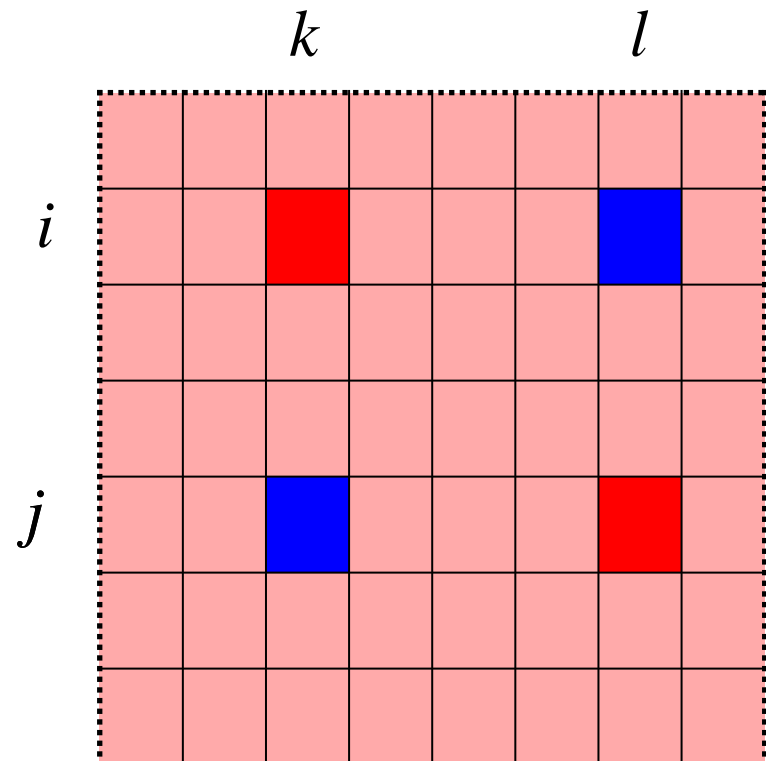
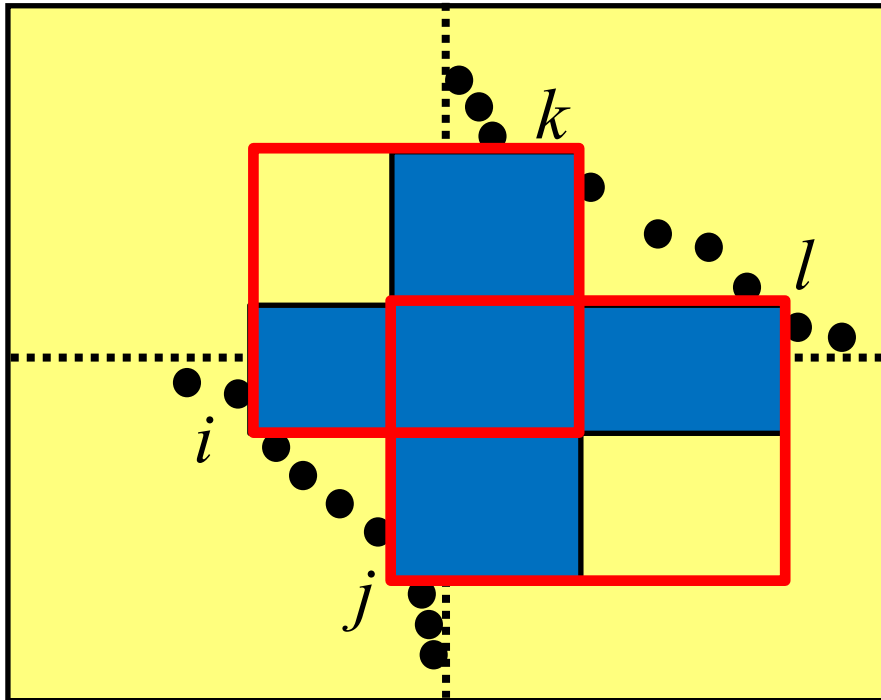
The (inverse) Monge property

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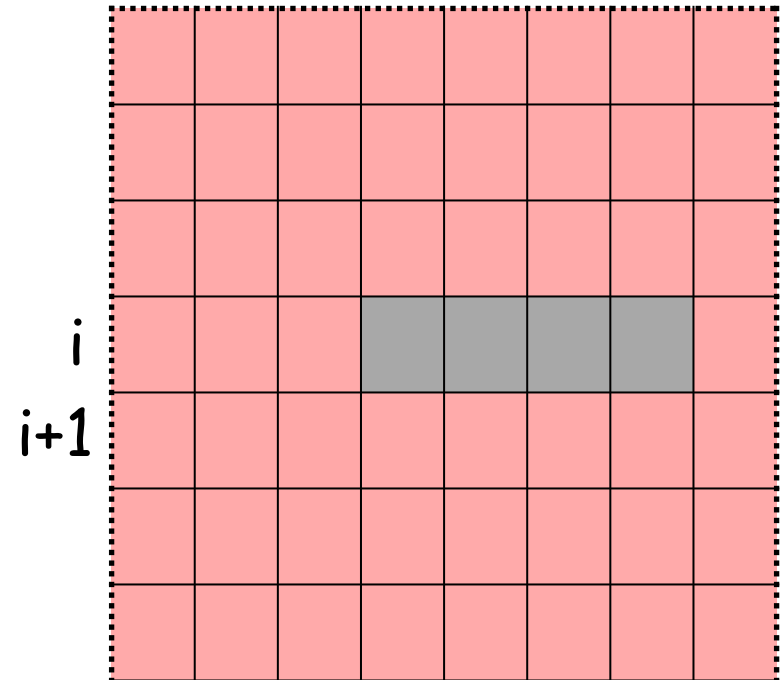
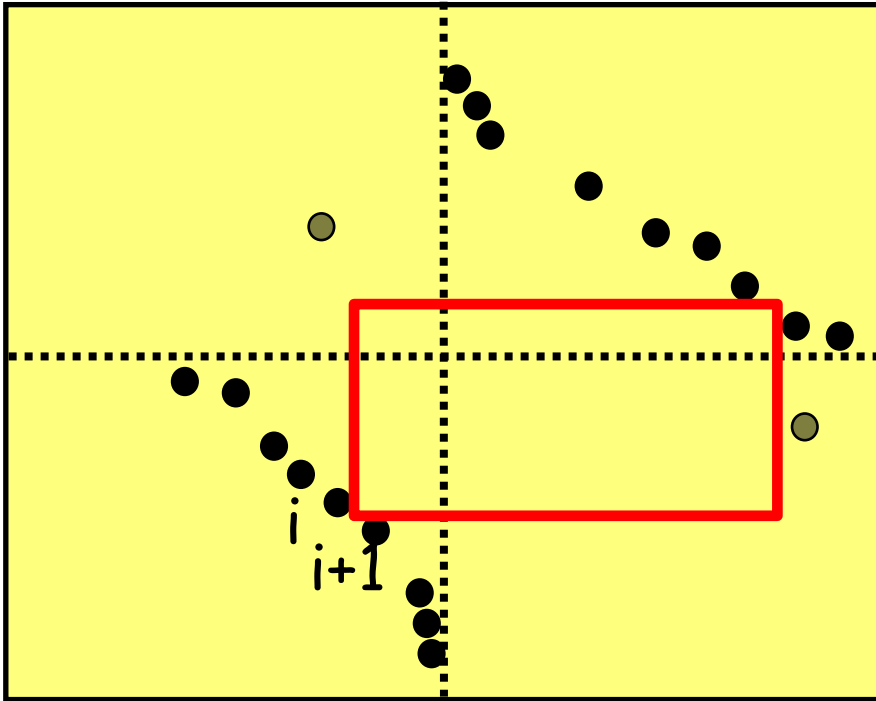


The (inverse) Monge property

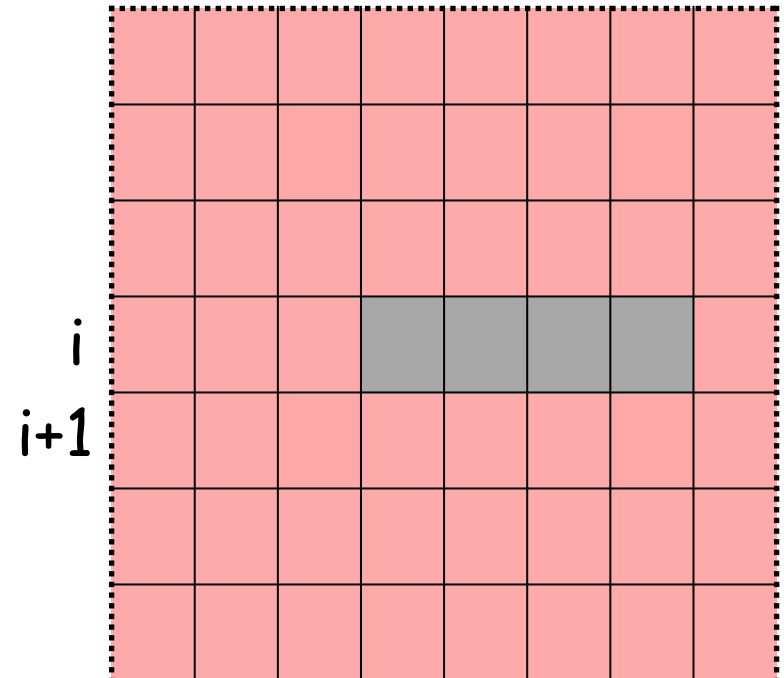
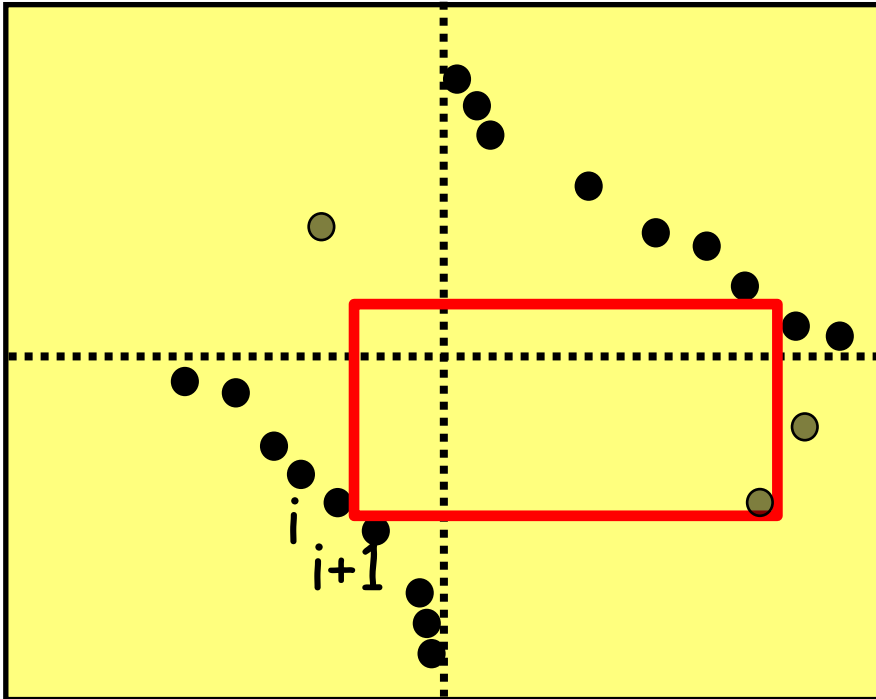
For every $i < j, k < l$, $M_{ik} + M_{jl} \geq M_{il} + M_{jk}$



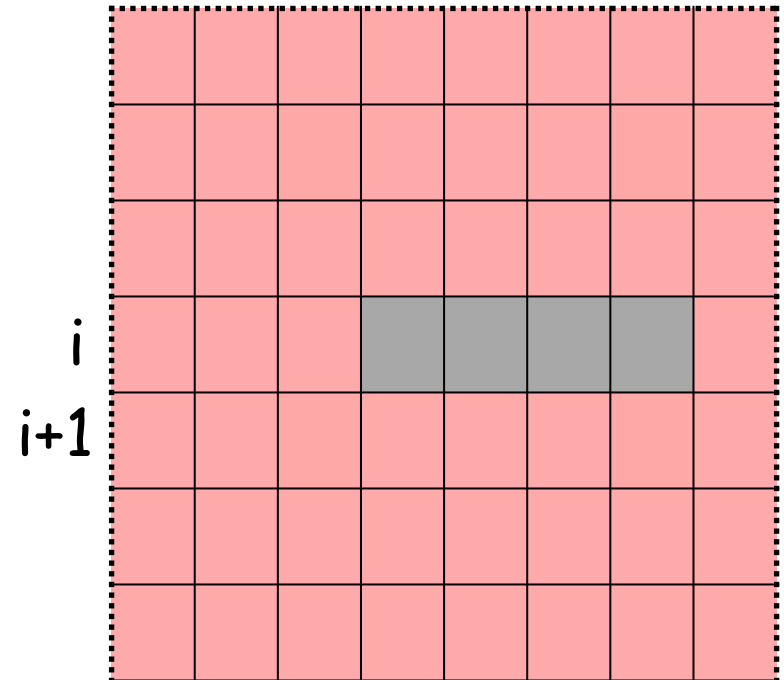
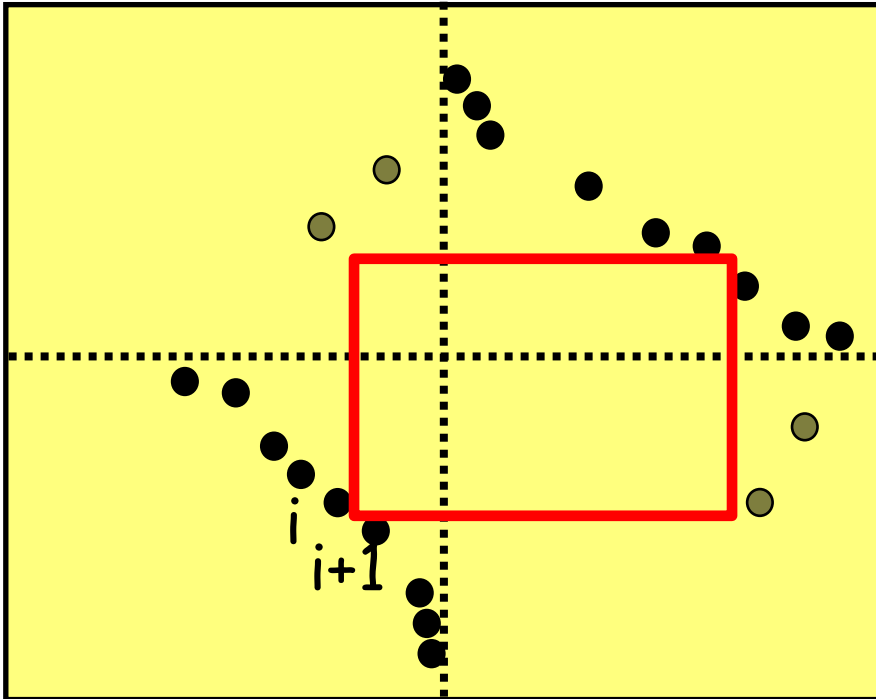
The relation between the defined parts of consecutive rows



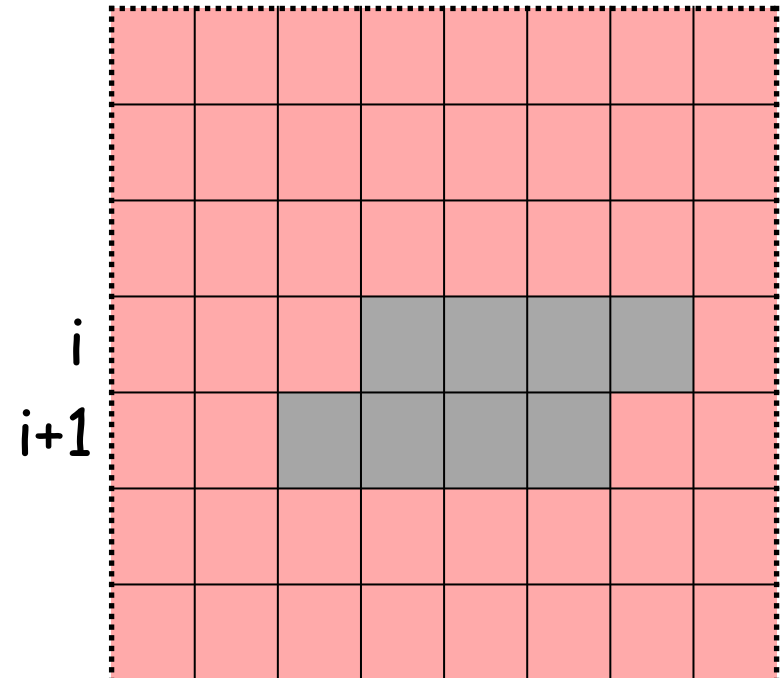
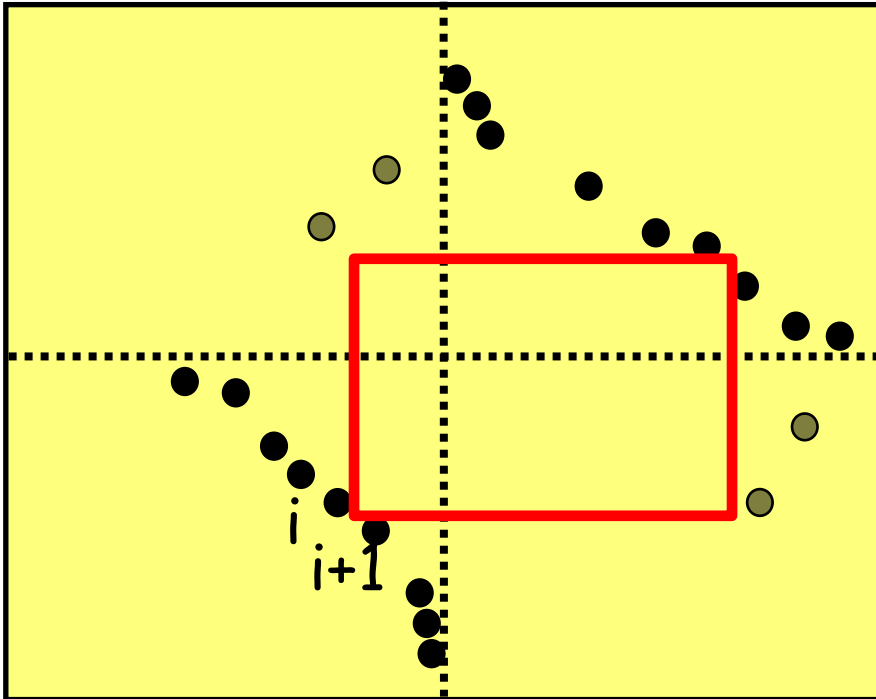
The relation between the defined parts of consecutive rows



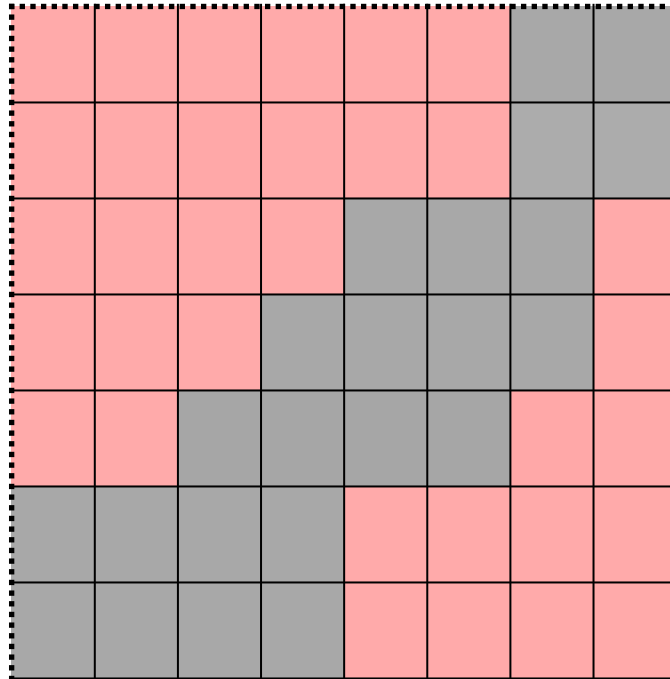
The relation between the defined parts of consecutive rows



The matrix is "staircase"



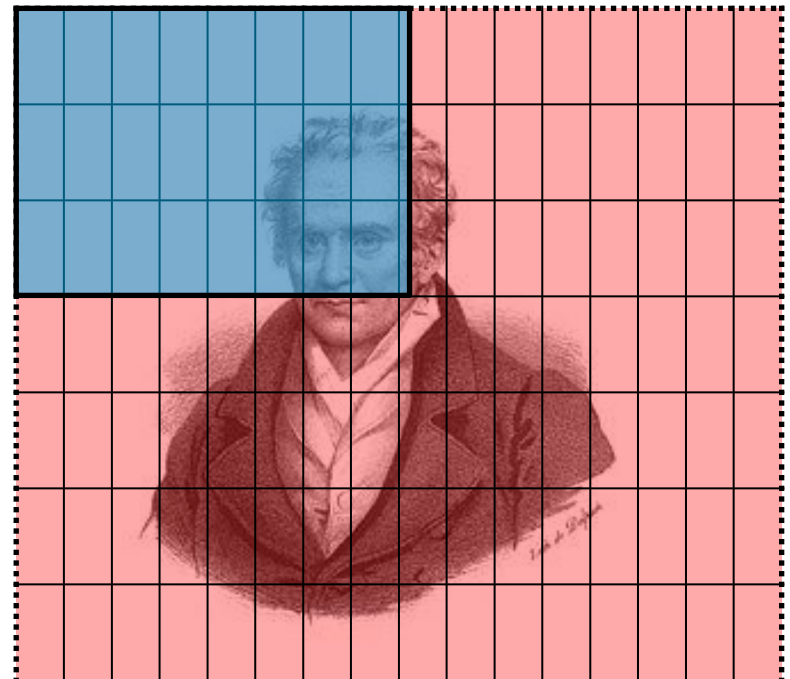
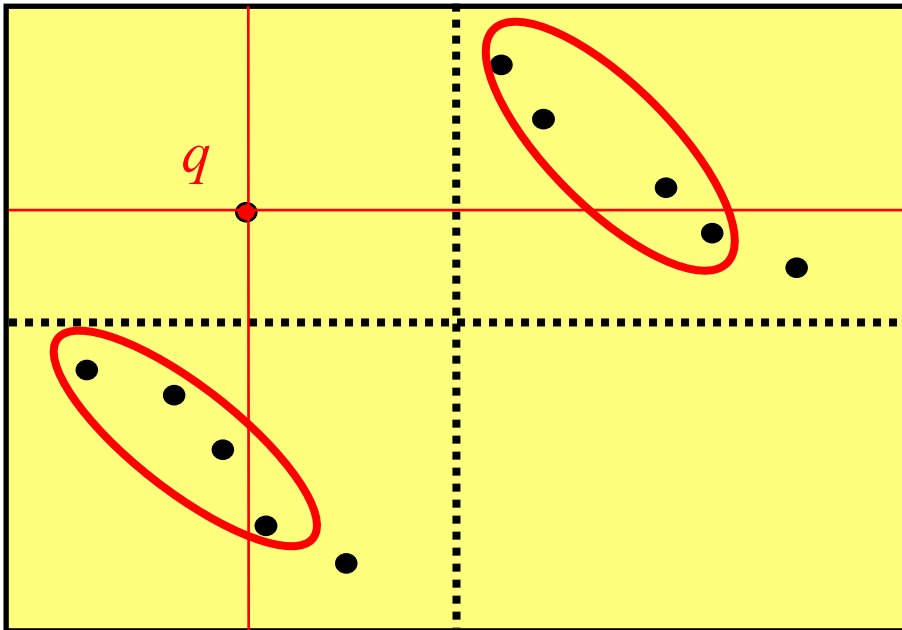
The matrix is "staircase"



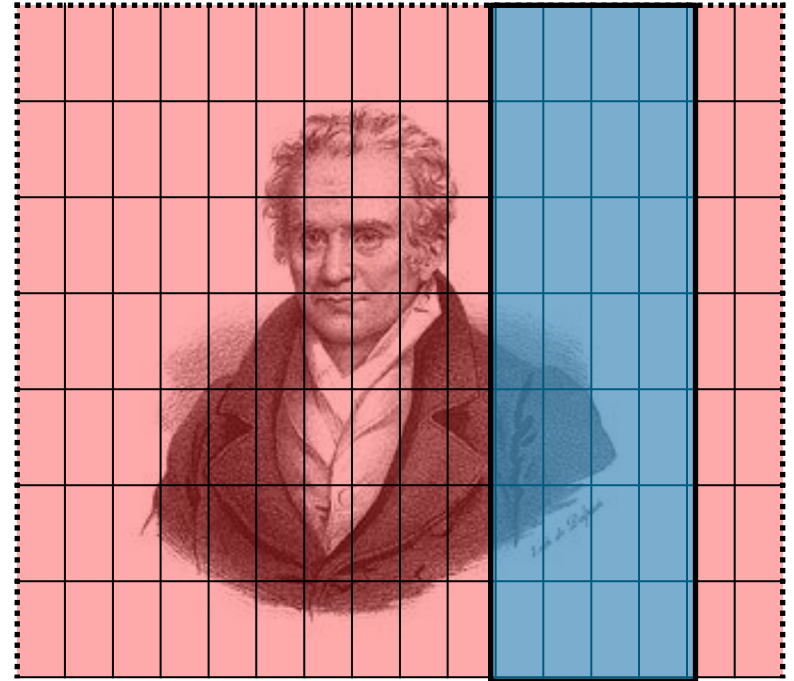
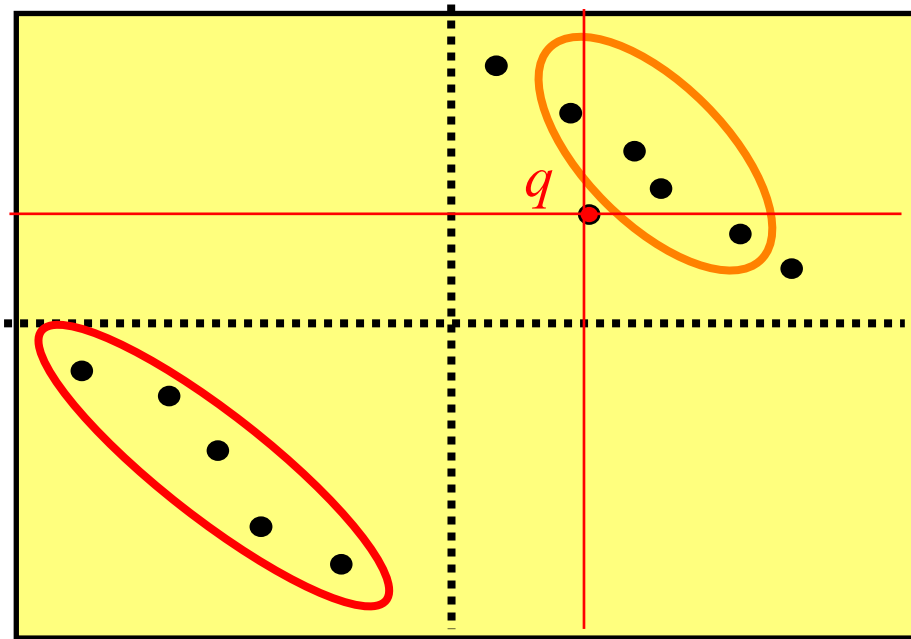
Query

A query point q defines ranges of relevant pairs in the 1st and 3rd quadrants

Use the **Monge submtarix maximum** data structure to find the maximum-area rectangle containing q in the relevant range



Query



Analysis

It takes $O(n_v a(n_v) \log(n_v))$ space $O(n_v a(n_v) \log^2 n_v)$ preprocessing time per secondary node v

Summing over all secondary nodes of the range tree we get that the total space is $O(na(n)\log^3(n))$ and the preprocessing time is $O(na(n)\log^4(n))$

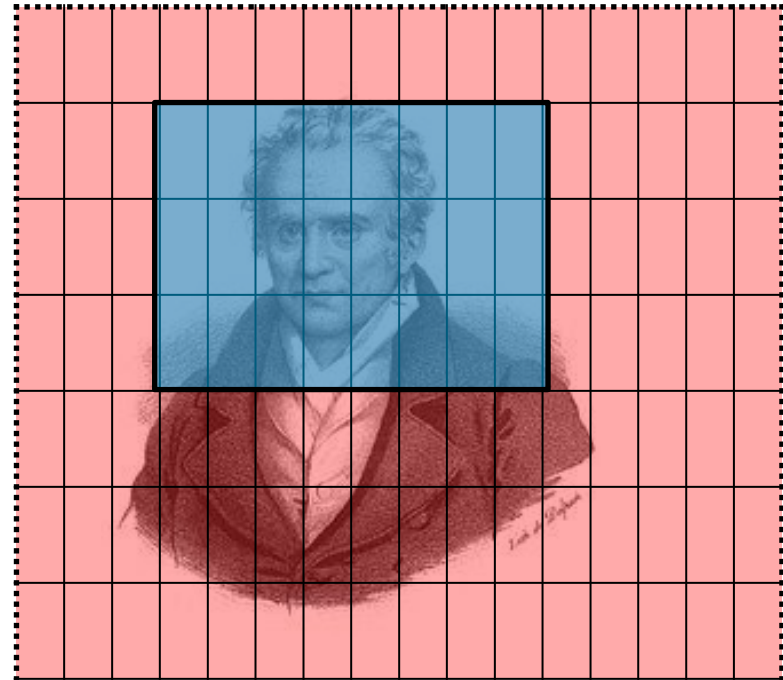
Query time is $O(\log^2 n_v)$ at a secondary node v and $O(\log^4 n)$ overall

Thats it as far as the empty
rectangle application is concerned

Back to our data structure

Input: $n \times n$ Monge matrix M
represented implicitly
(constant time oracle access)

Output: Data structure answering
maximum queries in any rectangular
submatrix of M

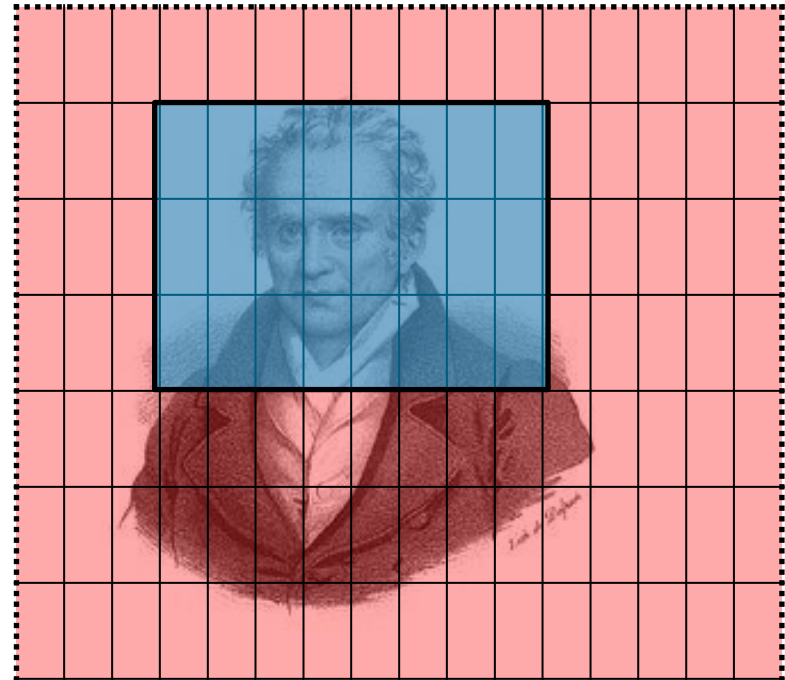


Submatrix maximum queries in Monge matrices

Main Theorem:

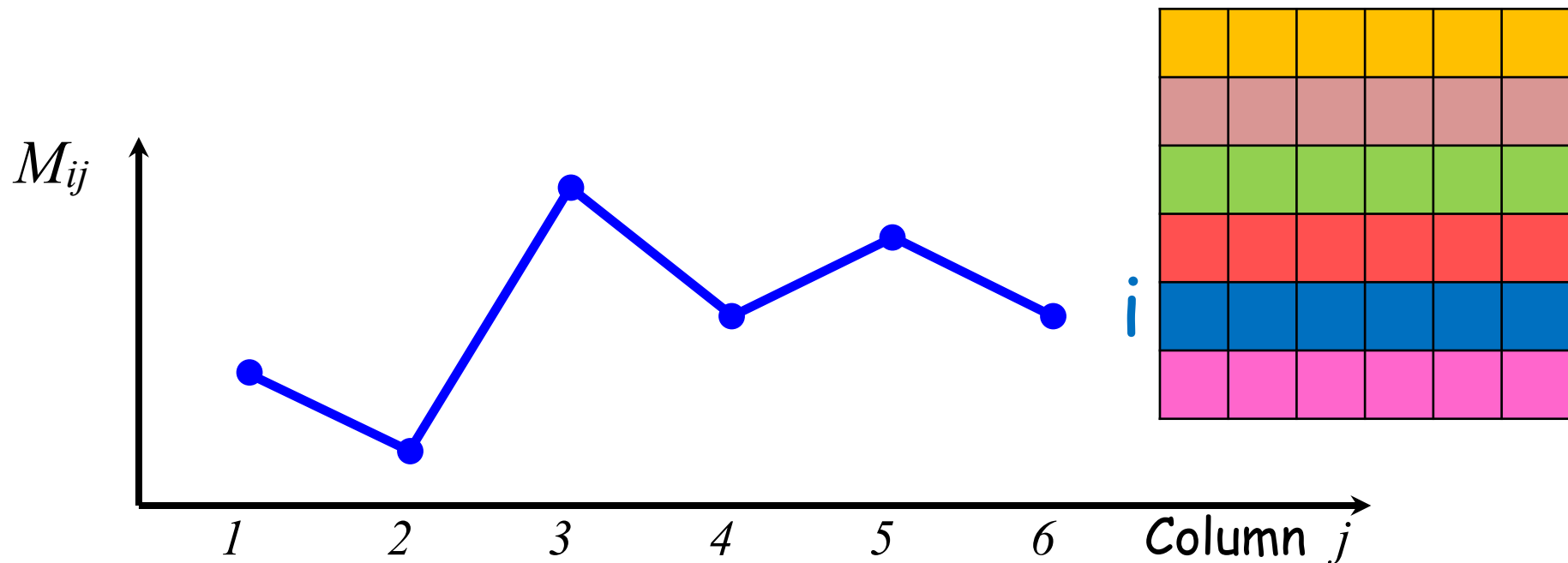
A data structure with the following performance:

preprocessing time	$O(n \log^2 n)$
space	$O(n \log n)$
query time	$O(\log^2 n)$



Totally monotone matrices and pseudo-lines

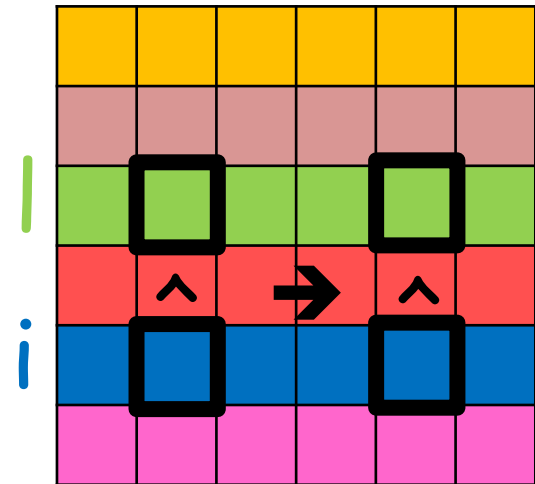
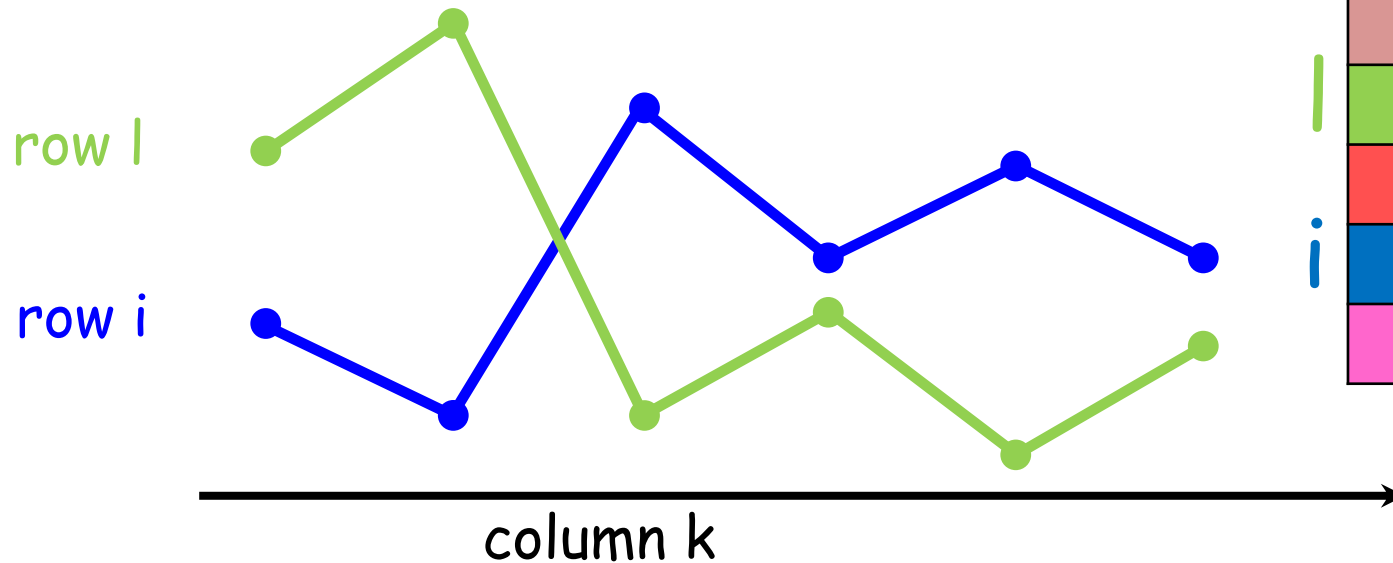
Think of row i of M as a discrete function mapping column j to M_{ij} . Extend the domain to the interval $[1, n]$ by linear interpolation



Totally monotone matrices and pseudo-lines

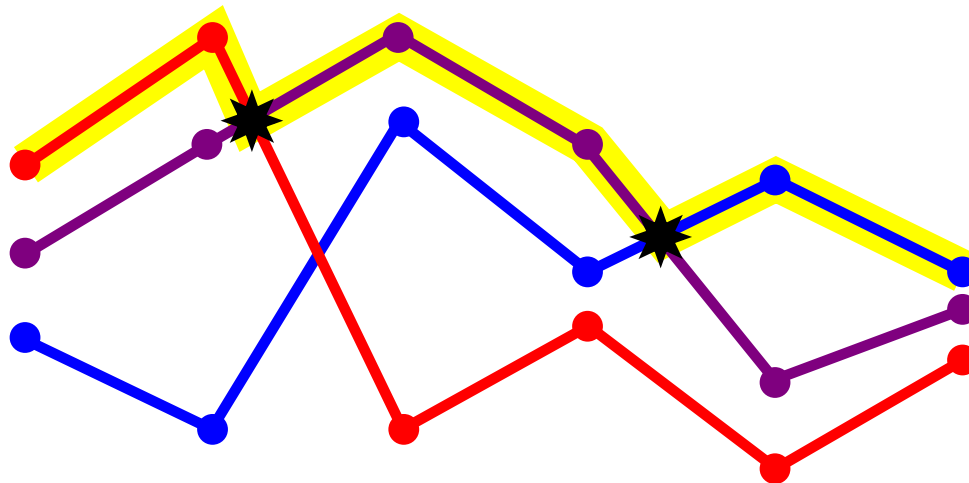
These are pseudo-lines:

Every pair of curves intersect in at most one point

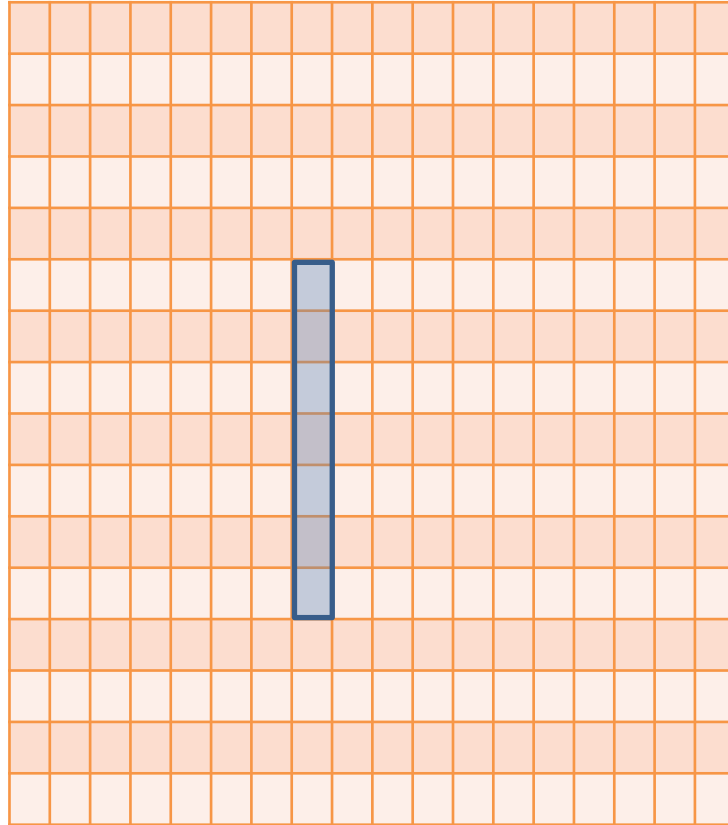


Upper envelopes of pseudo-lines

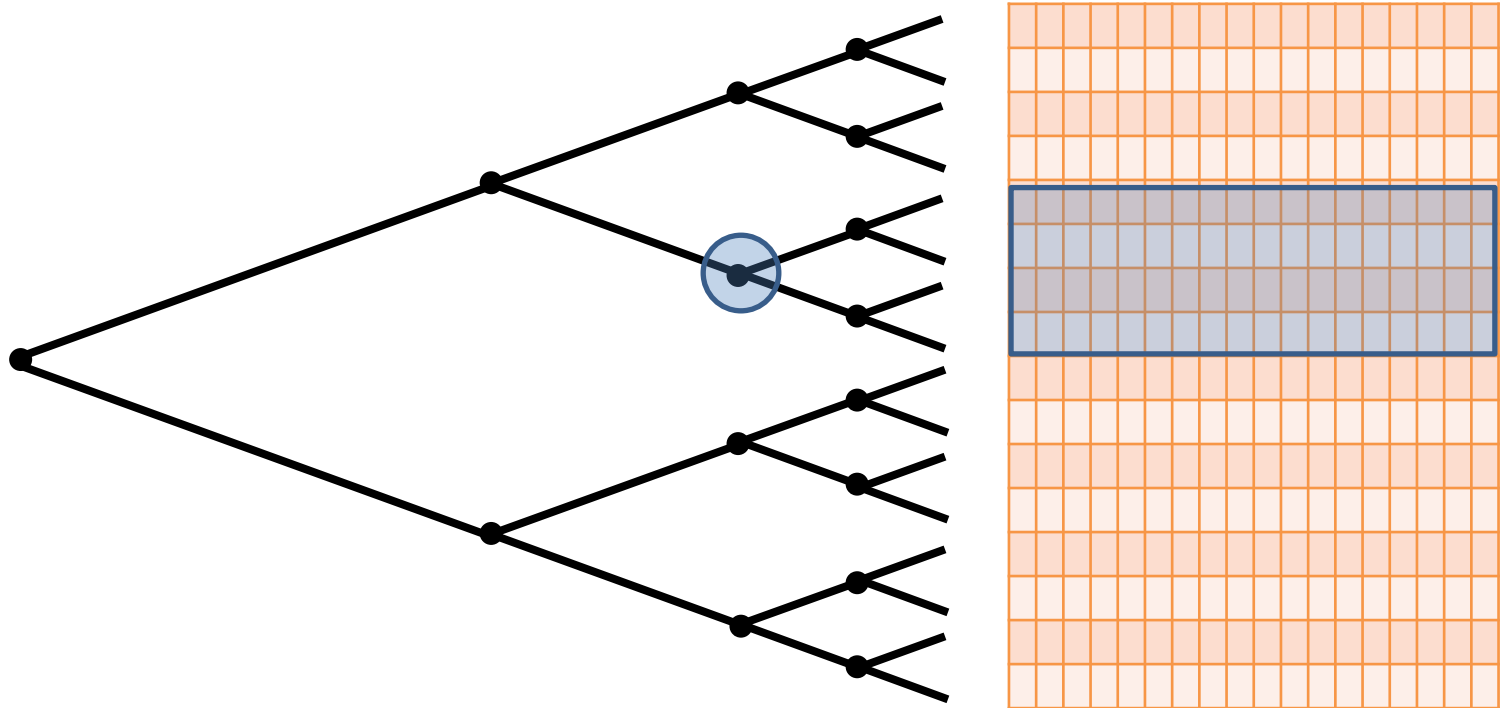
- Column maxima are on the **upper envelope** of the pseudo-lines
- For m rows there are **$O(m)$ breakpoints**
- Represent an upper envelope by a search tree over the breakpoints



Column maxima in a range of rows

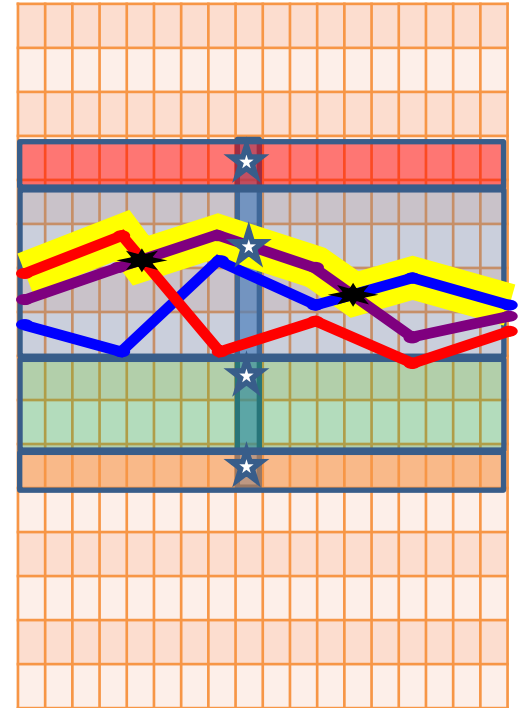
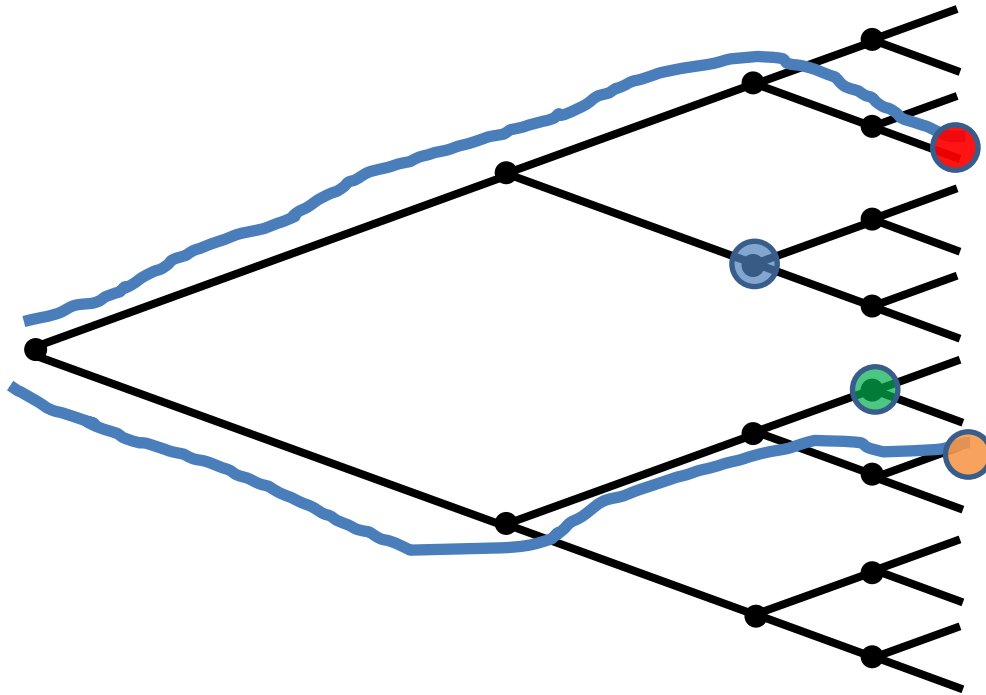


Column maxima in a range of rows



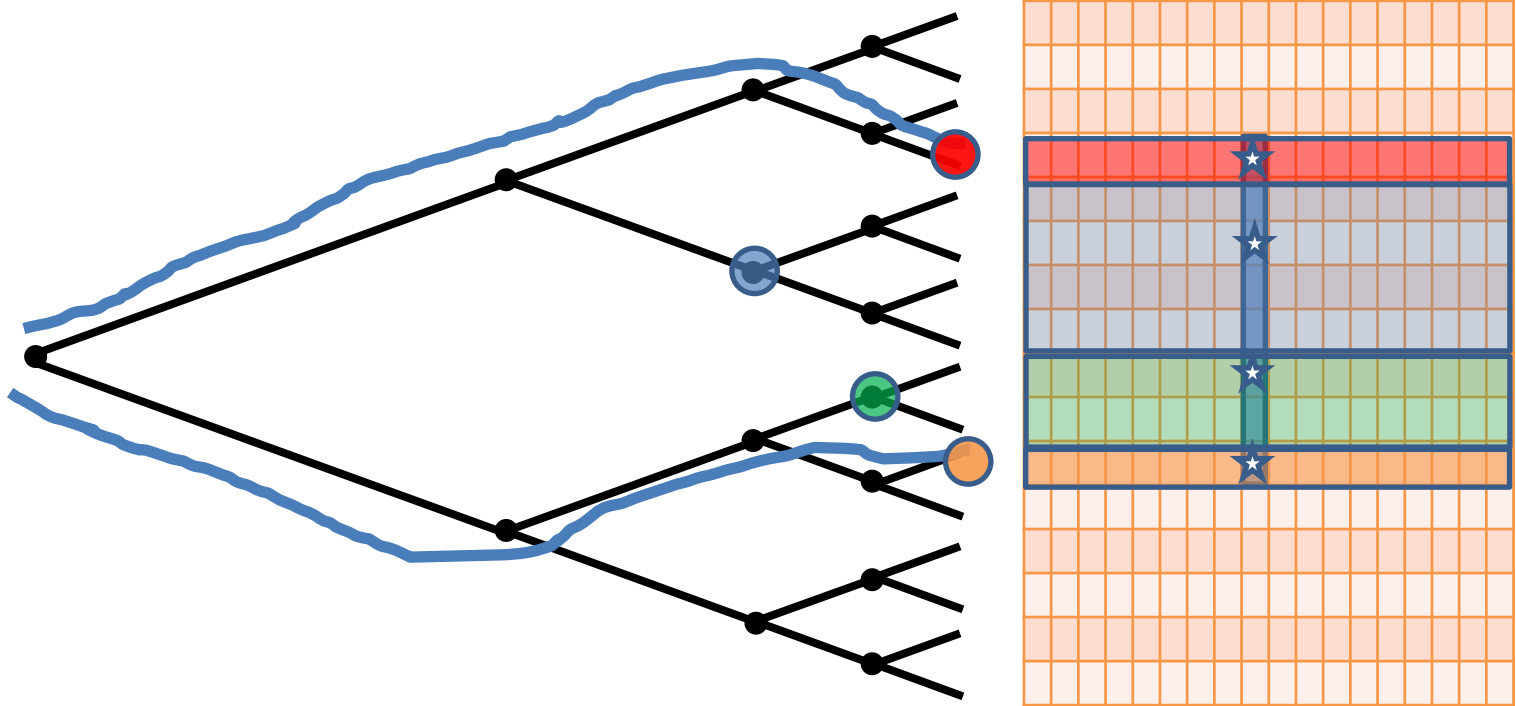
- Balanced search tree over the rows
- Each node stores the upper envelope of its rows
- Takes $O(n \log n)$ time in a bottom-up computation

Column maxima in a range of rows: a query



- A query is "covered" by $O(\log n)$ canonical nodes
- Search the envelope of each canonical node for the interval containing the column
- Return max of entries of the appropriate rows

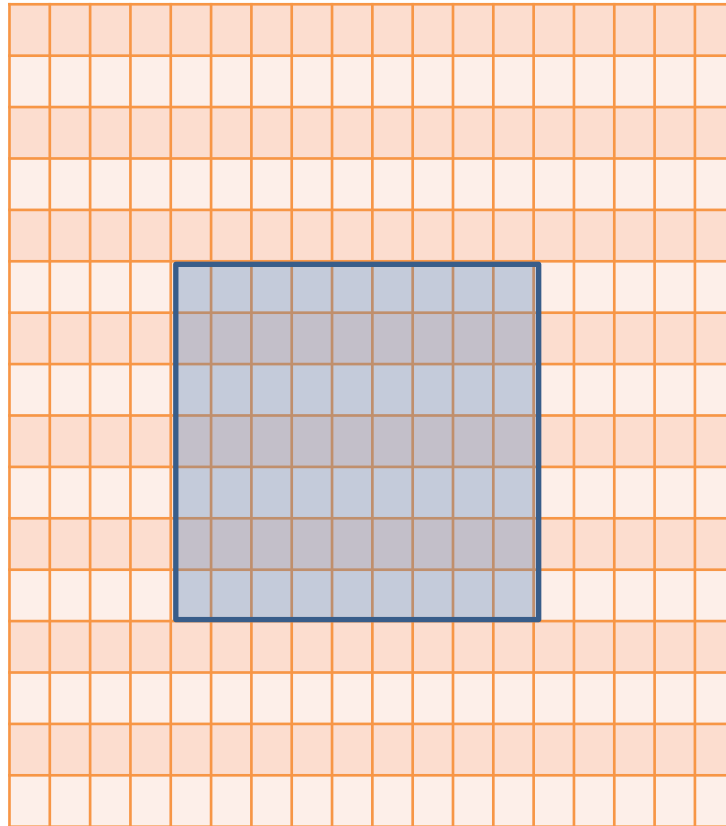
Column maxima in a range of rows: a query



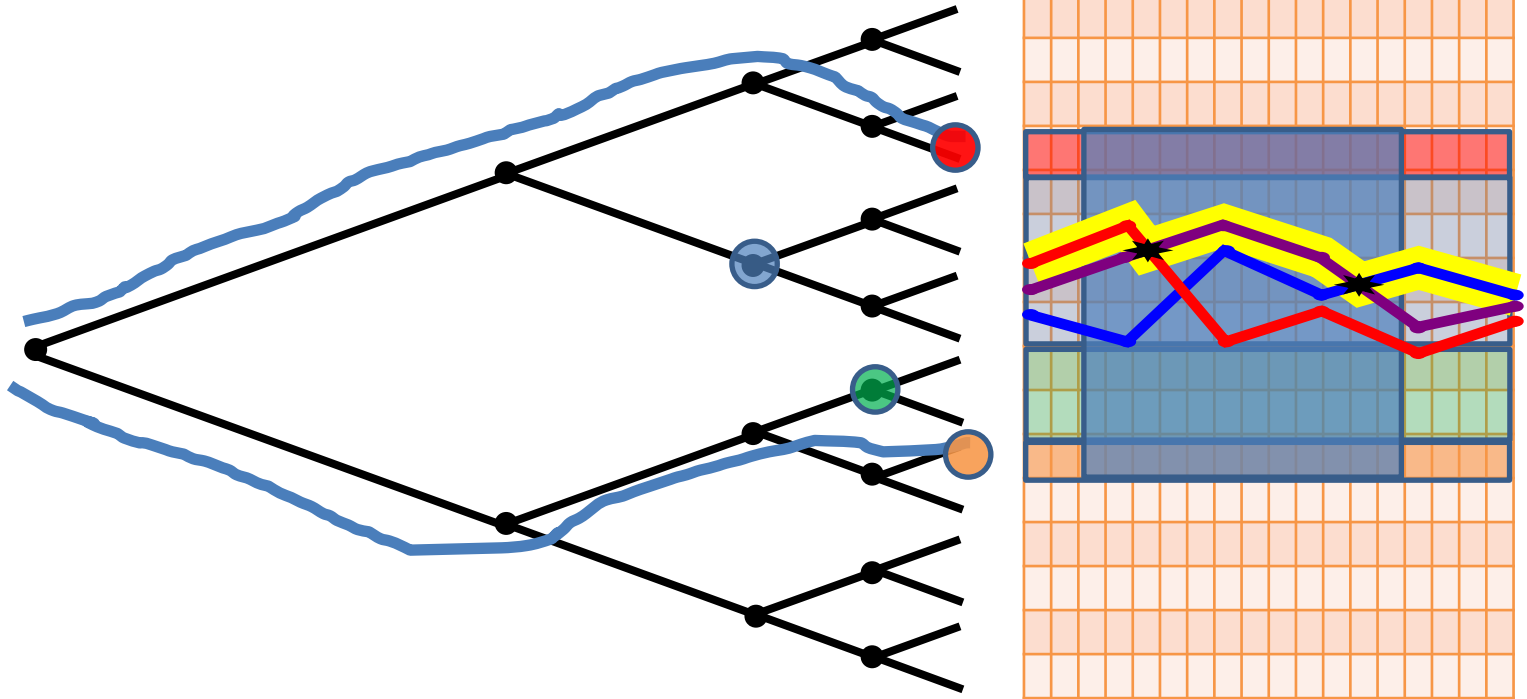
A query can be answered in $O(\log^2 n)$ time by querying each of the canonical nodes in $O(\log n)$ time.

Using fractional cascading reduces query time to $O(\log n)$

Maximum in a general submatrix



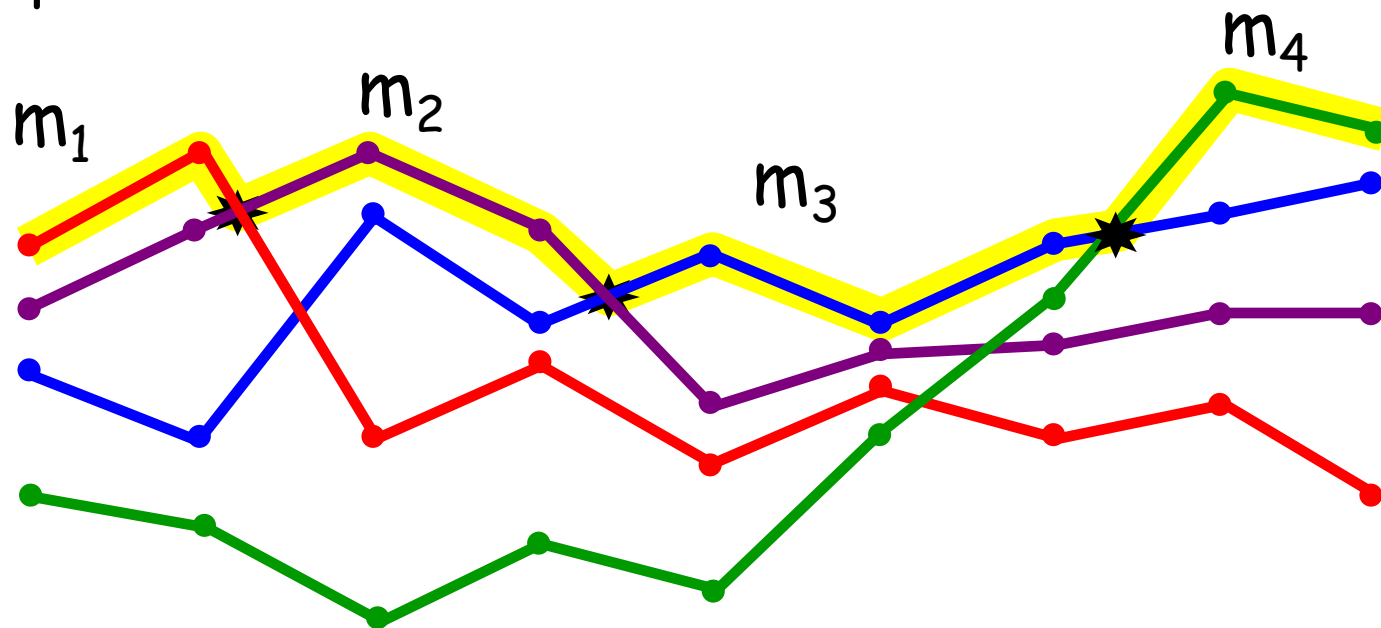
Maximum in a general submatrix



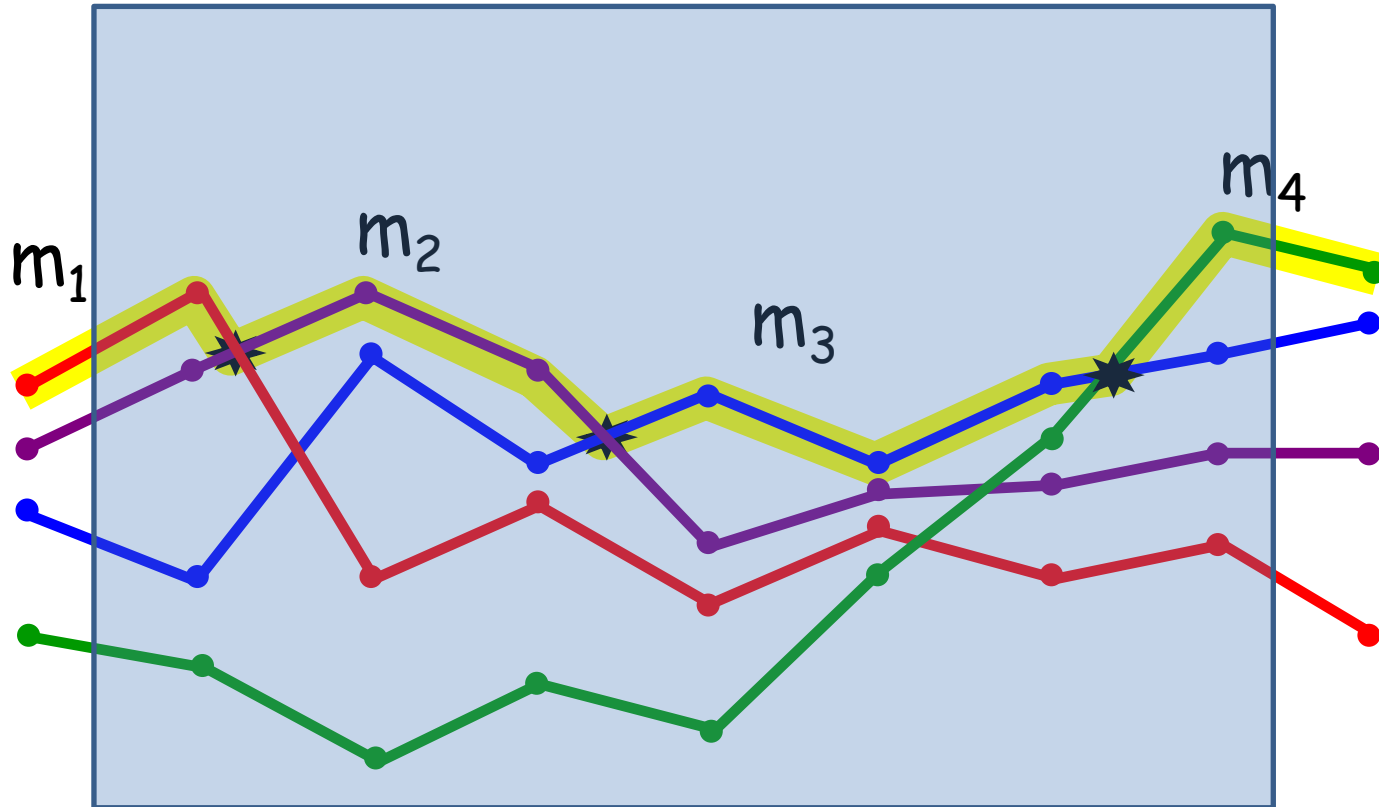
From each envelope we need the maximum in a **sequence of complete intervals** and a **prefix+suffix** of two intervals

Store additional information on the upper envelope

- Compute the maxima in each interval of the envelope
- This is a maxima in a subinterval of a row
- Can compute using a structure as before over the columns
- Store subtree maxima in the search tree of the breakpoints

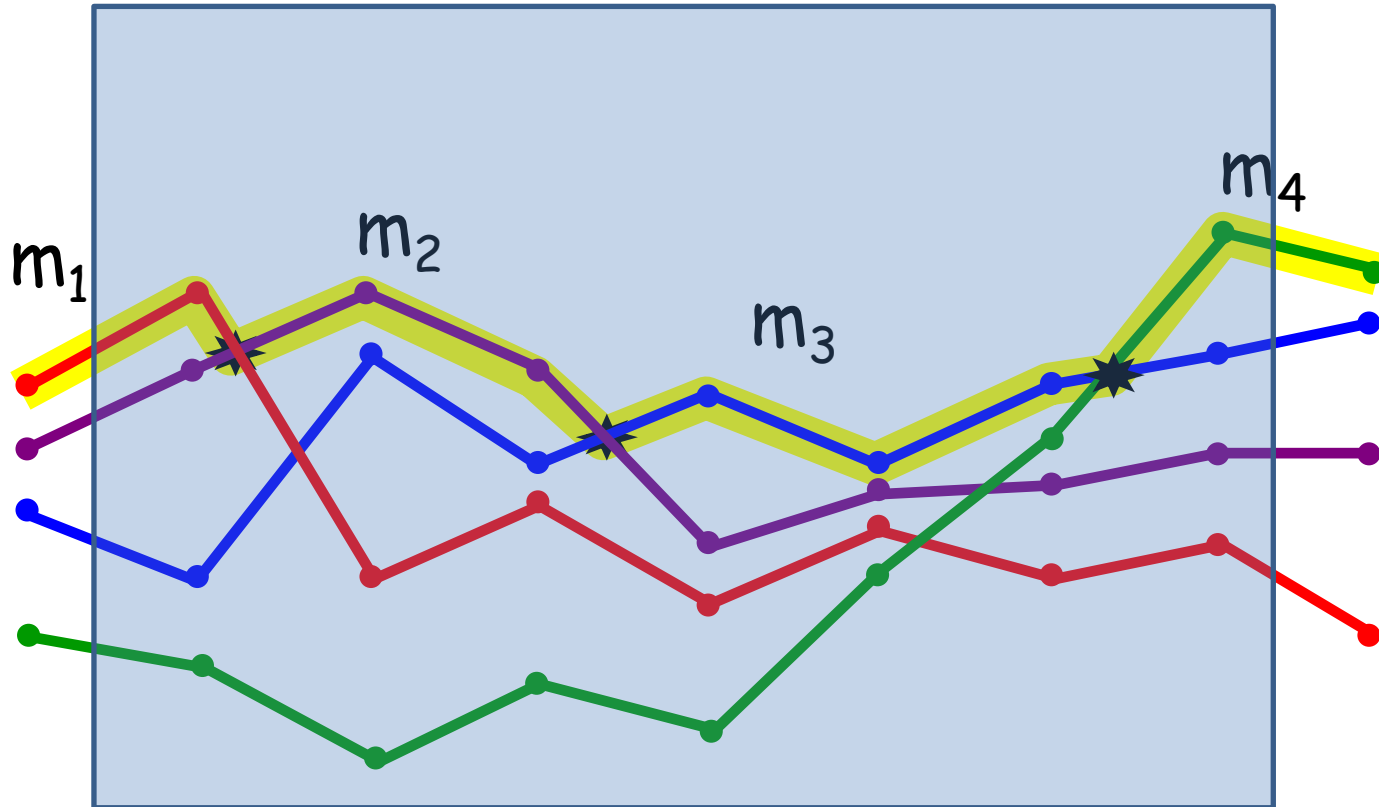


Submatrix query



- Find the maxima in the complete intervals of the envelope contained in the query

Submatrix query



- Find the maxima in the suffix and the prefix by queries to the structure for maxima in subintervals of rows

Submatrix queries: summary

preprocessing time $O(n \log n)$ (the total number of different intervals on all envelopes is $O(n)$)

space $O(n \log n)$

query time $O(\log^2 n)$

Submatrix queries: Extensions

- extension to certain **Monge partial matrices** using **pseudo-segments** instead of pseudo-lines
- preprocessing time $O(na(n)\log^2 n)$
space $O(na(n)\log n)$
query time $O(\log^2 n)$

