

Ad-Hoc Top-k Query Answering for Data Streams

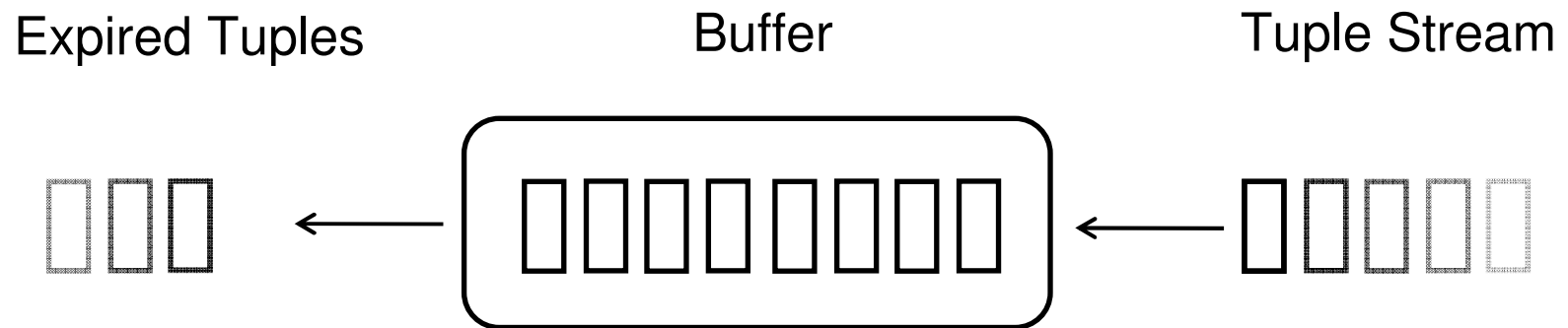
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Nick Koudas, Univ. of Toronto

Nikos Sarkas, Univ. of Toronto

Data Stream



Top-k Queries

- Top-k queries on the contents of the buffer
- Previous work [MBP06]
 - Top-k query maintenance
 - *Static* queries
- Current work
 - Ad-hoc top-k queries
 - *Dynamic* queries

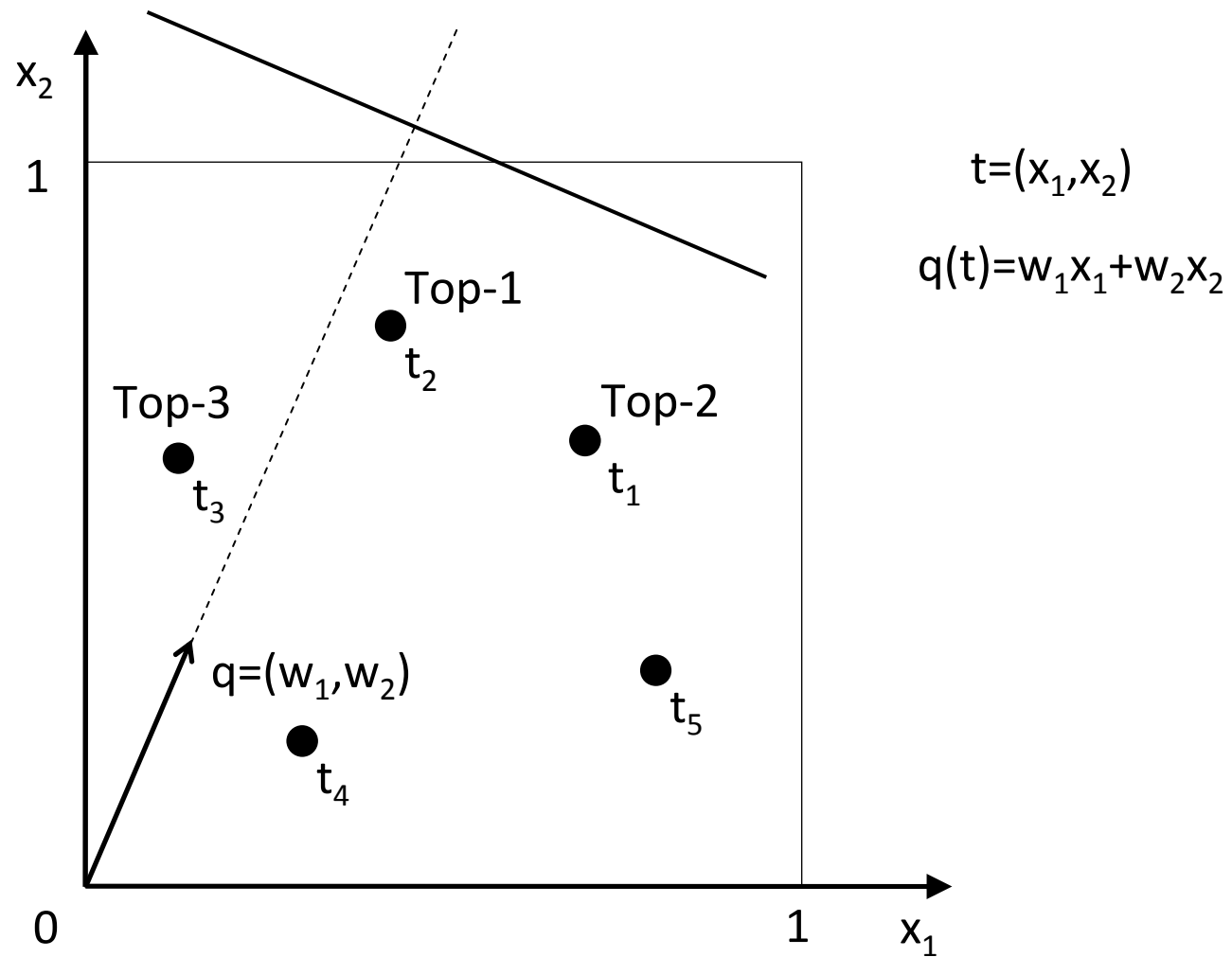
Outline

- Top-k query answering
 - Primal Plane
 - Dual Plane
- Arrangements
 - Representation
 - Operations
- Tuple Pruning
 - Principles
 - Implementation
- Experimental Evaluation

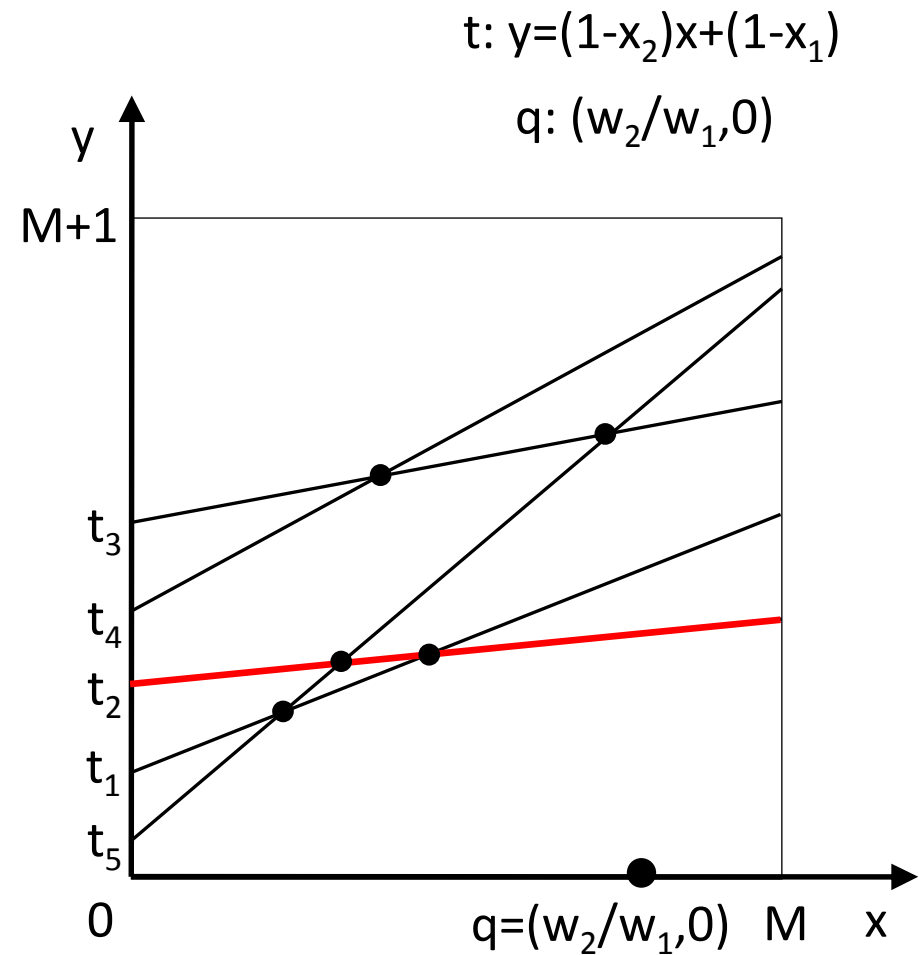
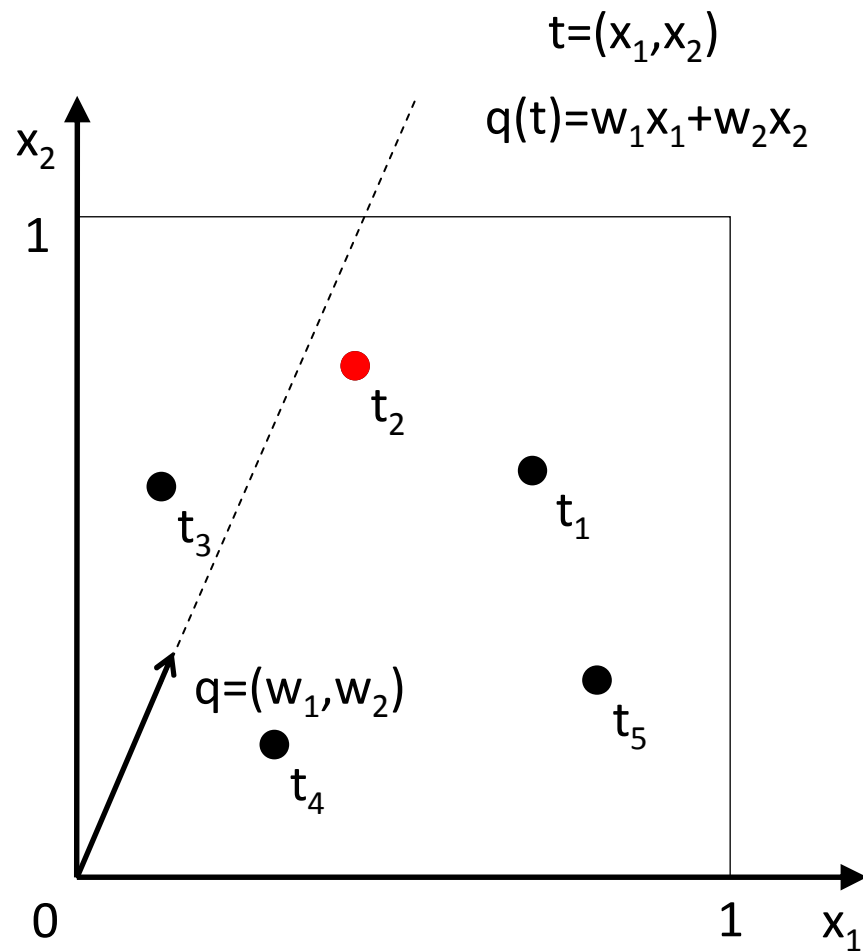
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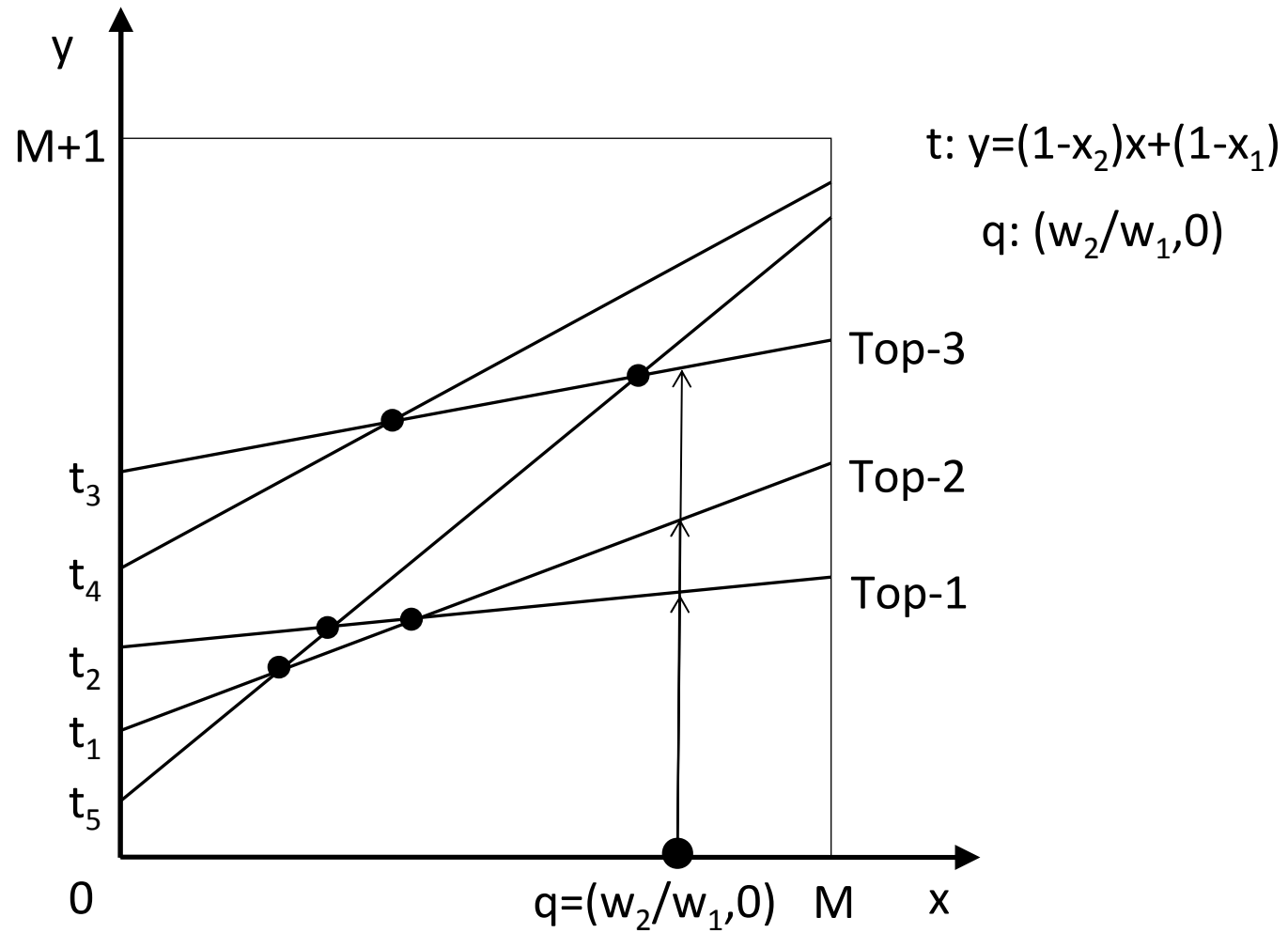
Primal Plane



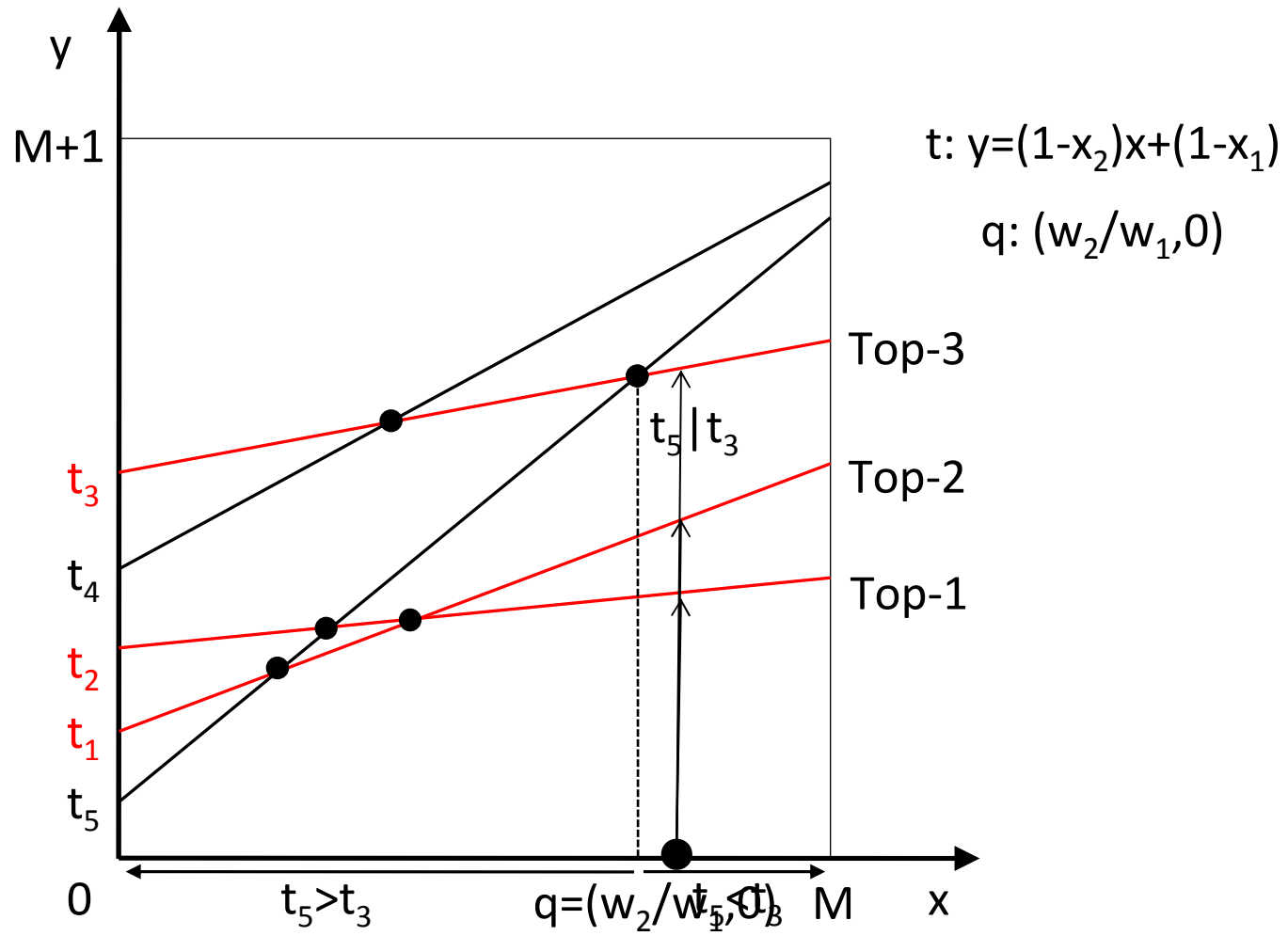
Primal-Dual Transformation



Dual Plane



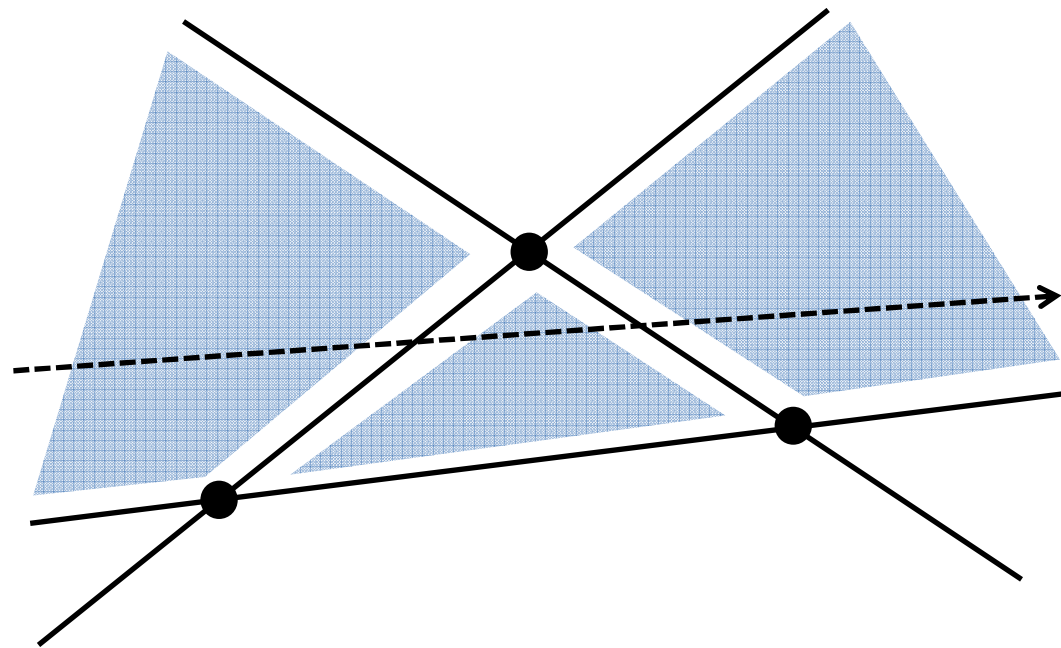
Dual Plane



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Arrangements: Introduction



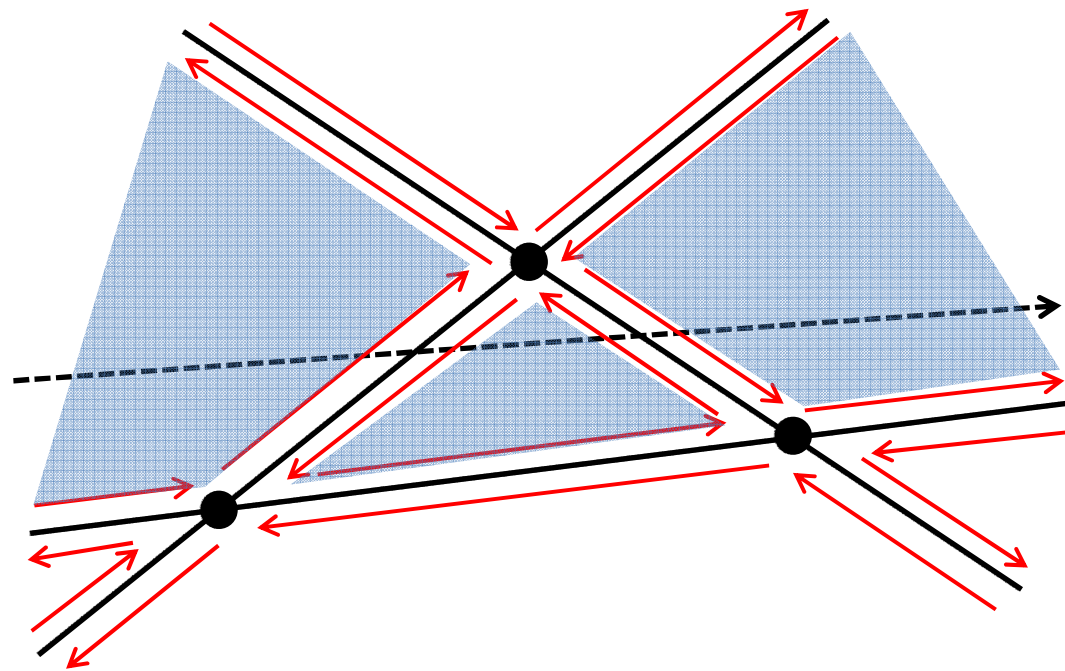
$O(n^2)$ vertices

$O(n^2)$ edges

$O(n^2)$ faces

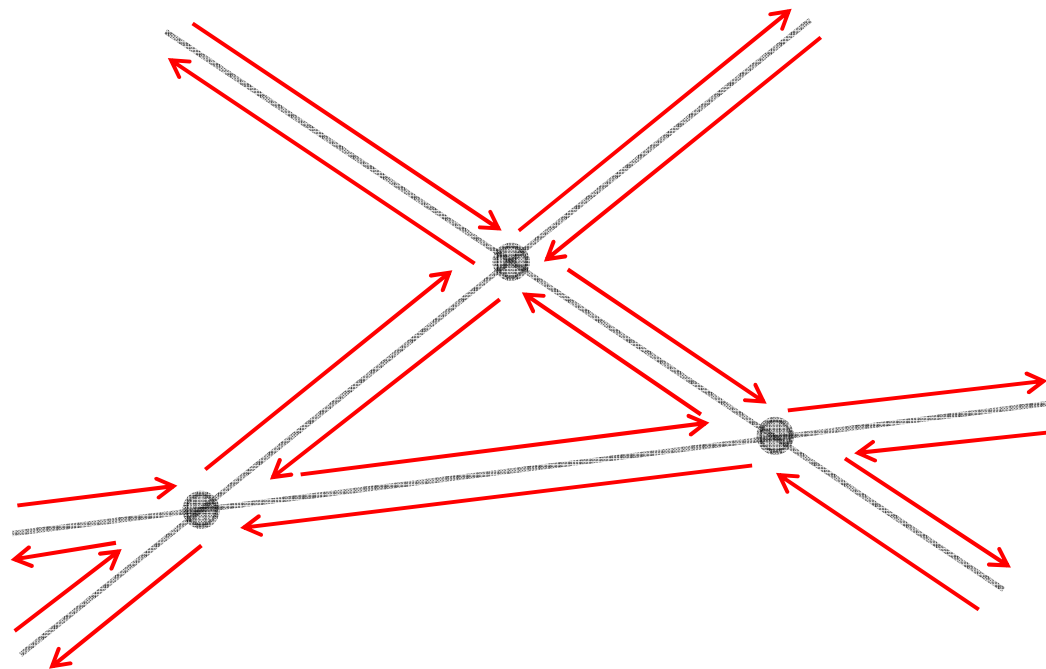
Zone: $O(n)$ edges

Arrangements: Representation



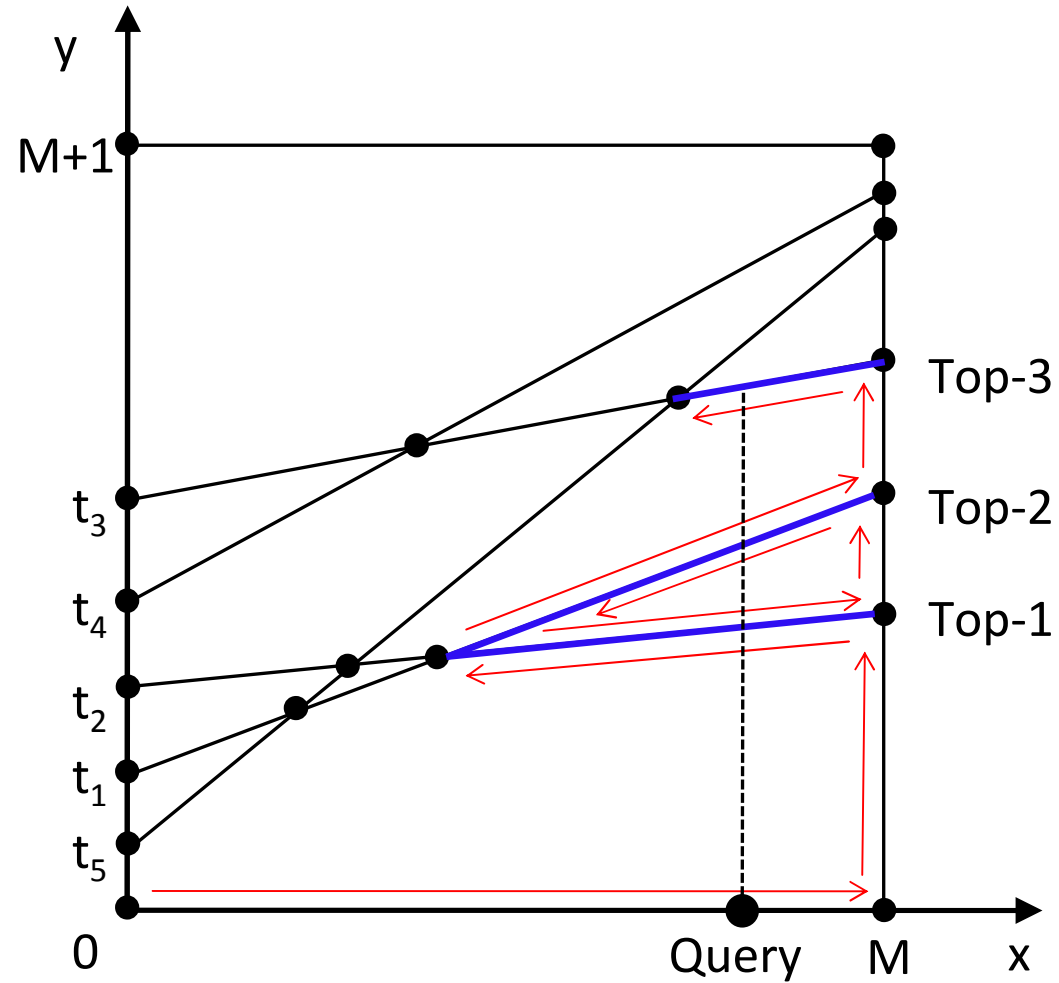
$O(n^2)$ vertices
Vertices
 $O(n^2)$ edges
Two Half Edges
 $O(n^2)$ boundaries
Connected Boundaries
Zone: $O(n)$ edges

Arrangements: Representation

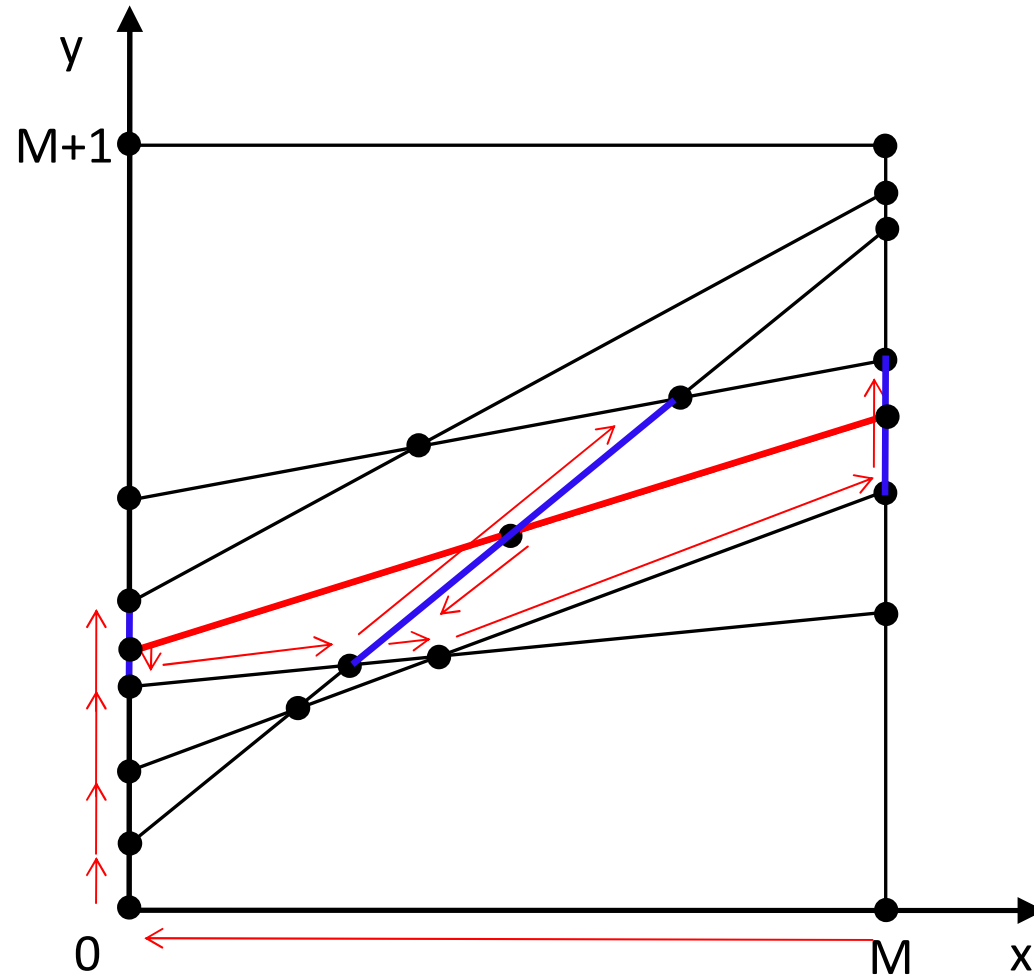


Vertices
Twin Half-edges
Connected Boundaries

Arrangements: Top-k Query Answering



Arrangements: Tuple Insertion



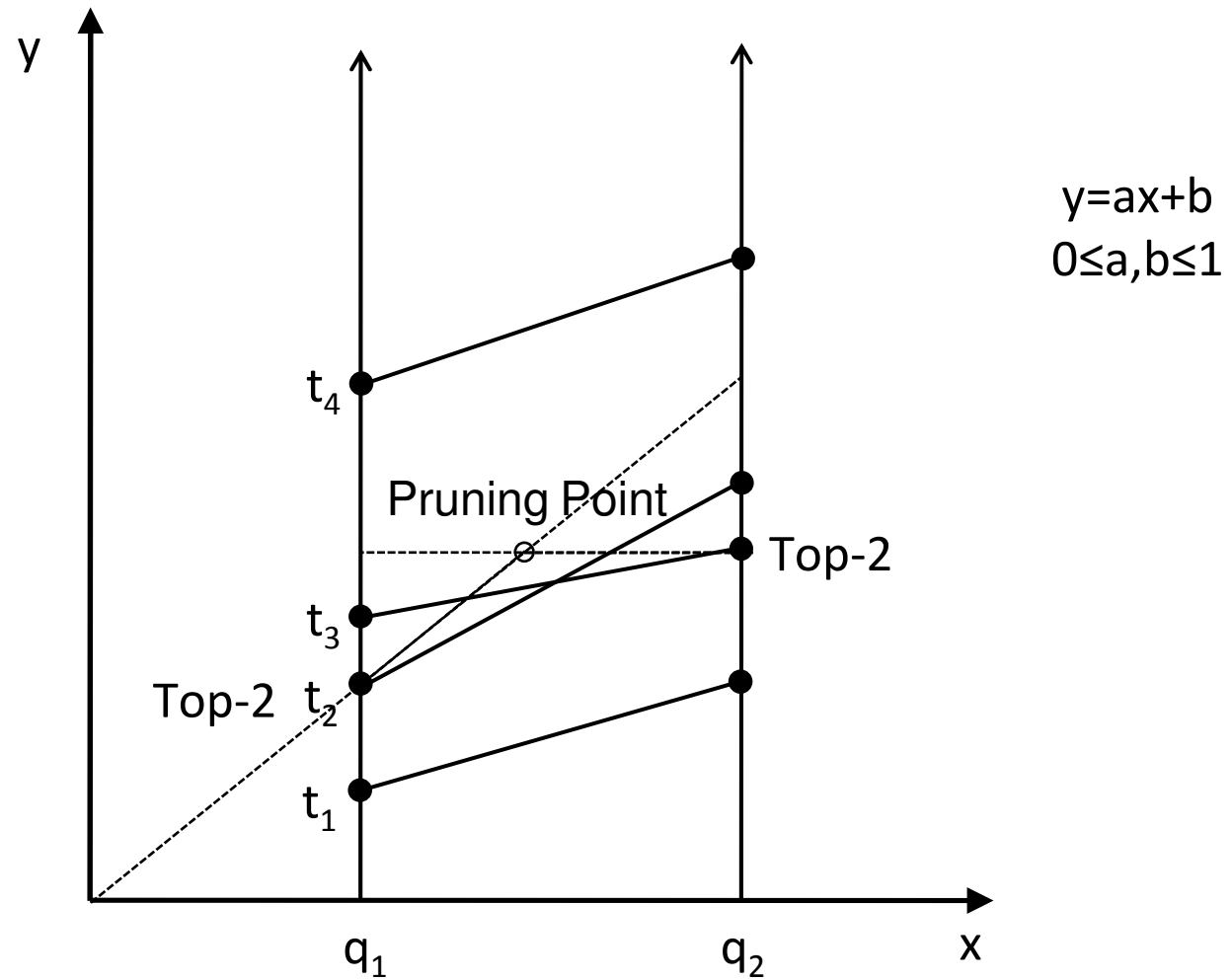
Contributions so far

- Dual representation of top-k problem
- Use of arrangements and development of algorithms
 - $O(n)$ query answering, $O(k)$ in practice
 - $O(n)$ insertion and deletion
 - $O(n^2)$ space overhead
- Benefits
 - *Non-redundant, self-organizing representation of the ranking of all possible top-k queries*
- Still, we can do much better
 - $O(k \log n)$ operations
 - $O(k^2 \log^2 n)$ space overhead

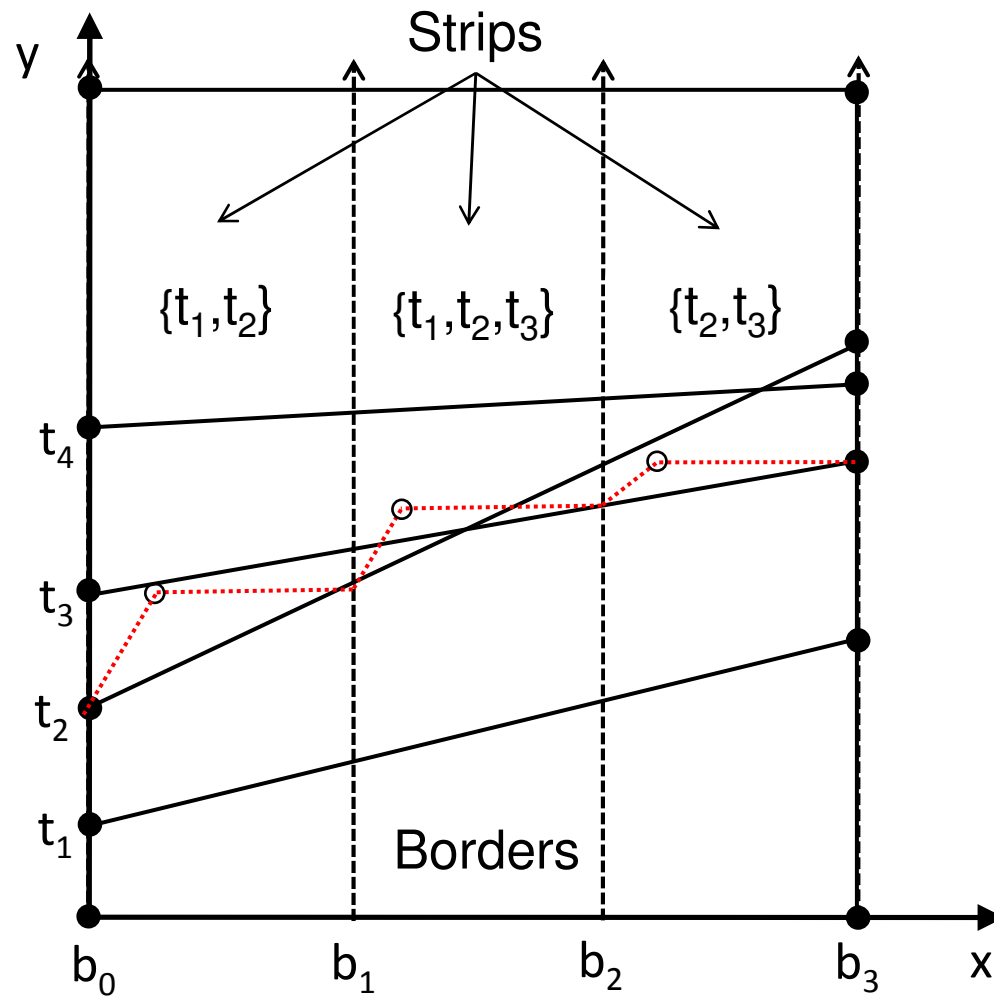
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Tuple Pruning

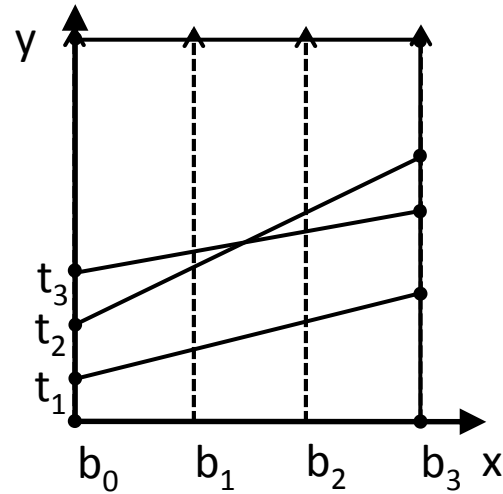


Tuple Pruning

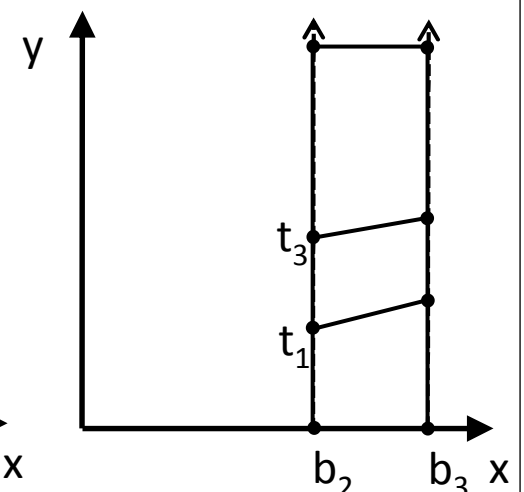
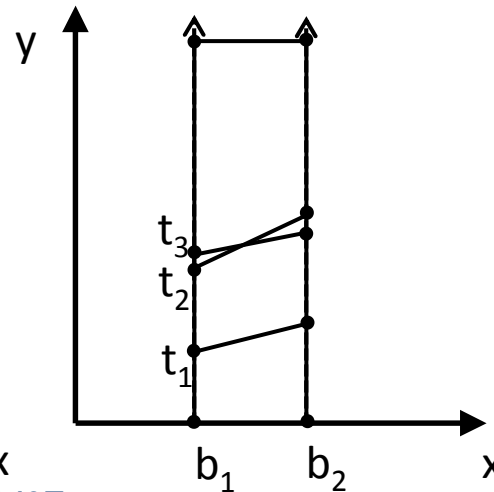
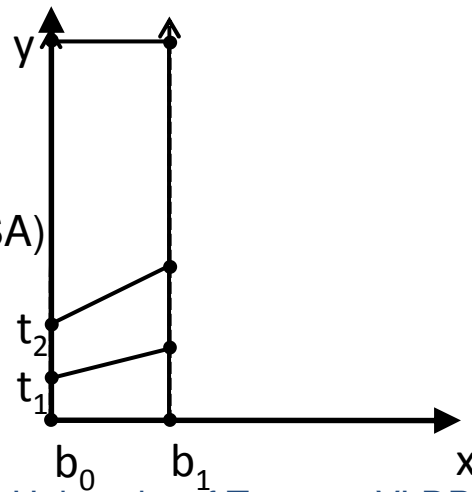


Storing the Pruned Dataset

Full Arrangement (FA)



Strip Arrangements (SA)



Pruning Efficiency

- Size of the filtered dataset is $O(k \log n)$
- Thus, $O(k \log n)$ operations on the arrangement
- Example
 - Top-20 queries
 - 1 million 2d uniformly distributed tuples
 - 16 borders
 - Only 250 tuples need to be stored in the arrangement!

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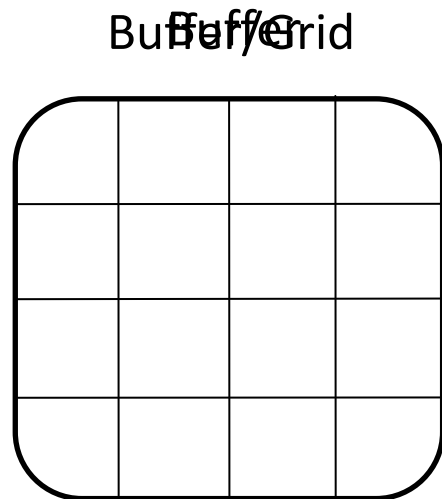
Challenges

- Maintain relevant tuples in the presence of streaming updates
- Procedure
 - Update the top-k results along the borders
 - Update the pruning points
 - For each strip, update the tuples that fall below the corresponding pruning point
 - Update arrangement

Solutions

- Maintain the top-k result along the borders
 - Top-k query maintenance techniques [MBP06]
- For each strip, update the tuples that fall below the corresponding pruning point
 - Half-space range searching in the primal plane!
- Index the buffer using a grid

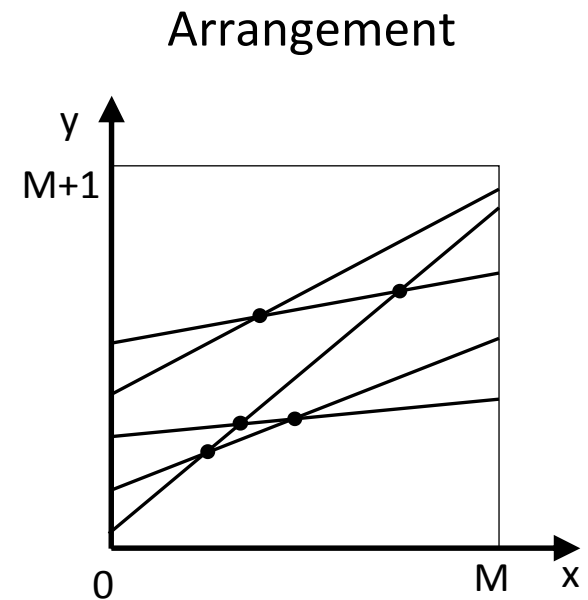
Maintaining the Pruned Dataset



Border Maintenance

→

Half-space
Range Searching



Placing the Borders

- Increasing the number of borders increases the pruning efficiency and overhead
- Objective
 - Equi-depth partitioning
- Heuristic
 - Iteratively split strips until strips have less than a certain number of vertices (*strip complexity*)

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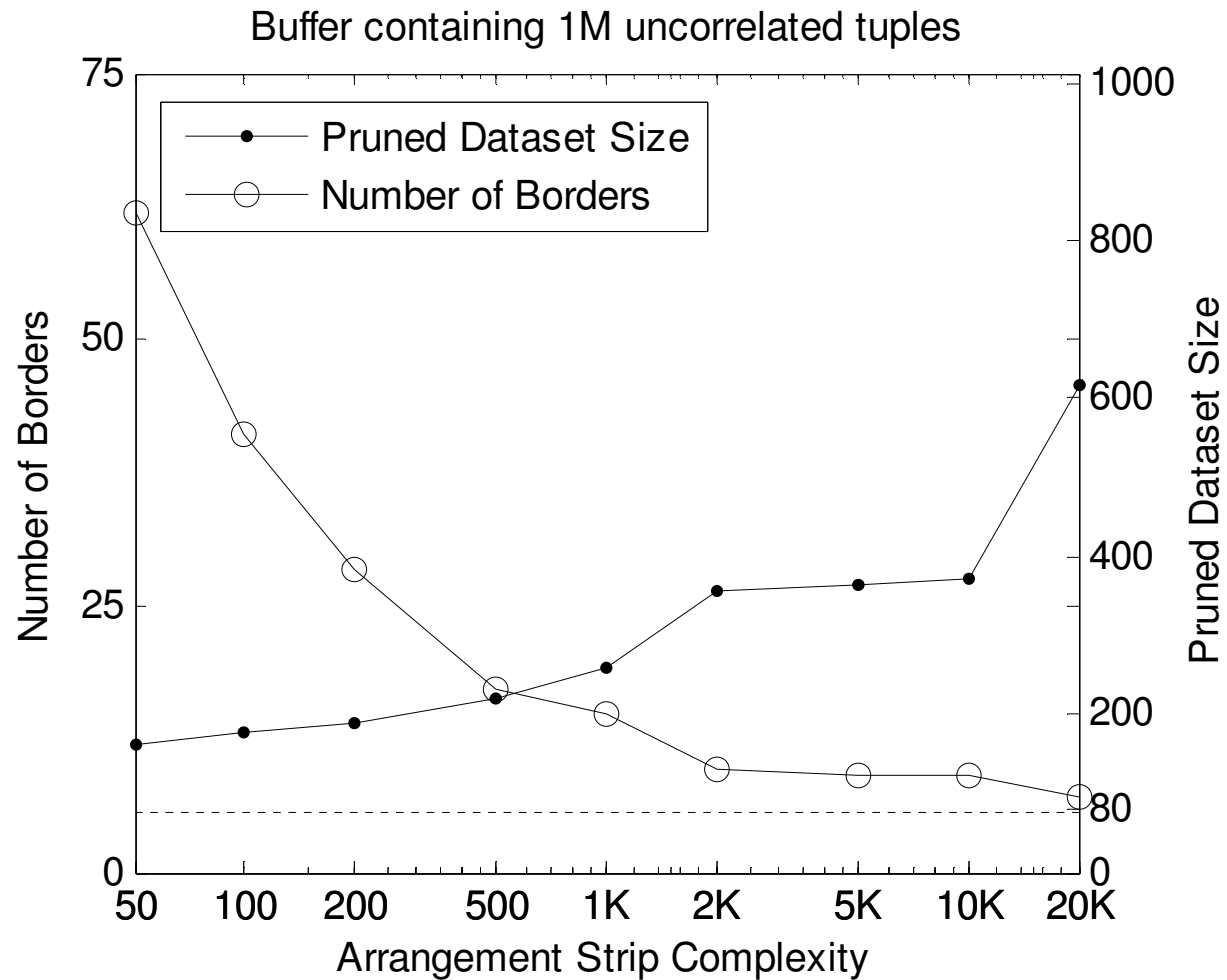
Experimental Setting

- Data sets
 - Synthetic: uniform, correlated, anti-correlated
 - Real: Intel Lab data
- Experiments
 - Pruning Efficiency
 - Memory overhead
 - Variable buffer size, stream rate, query results (k), query frequency, dimensionality

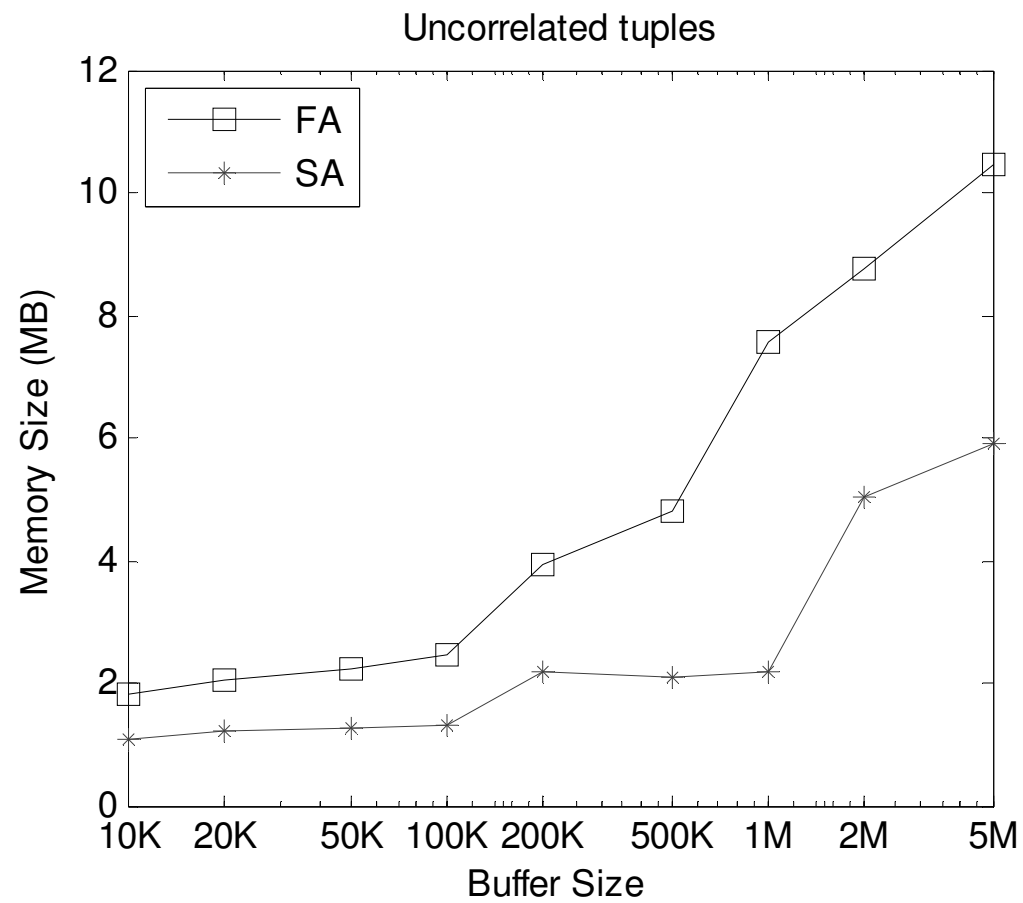
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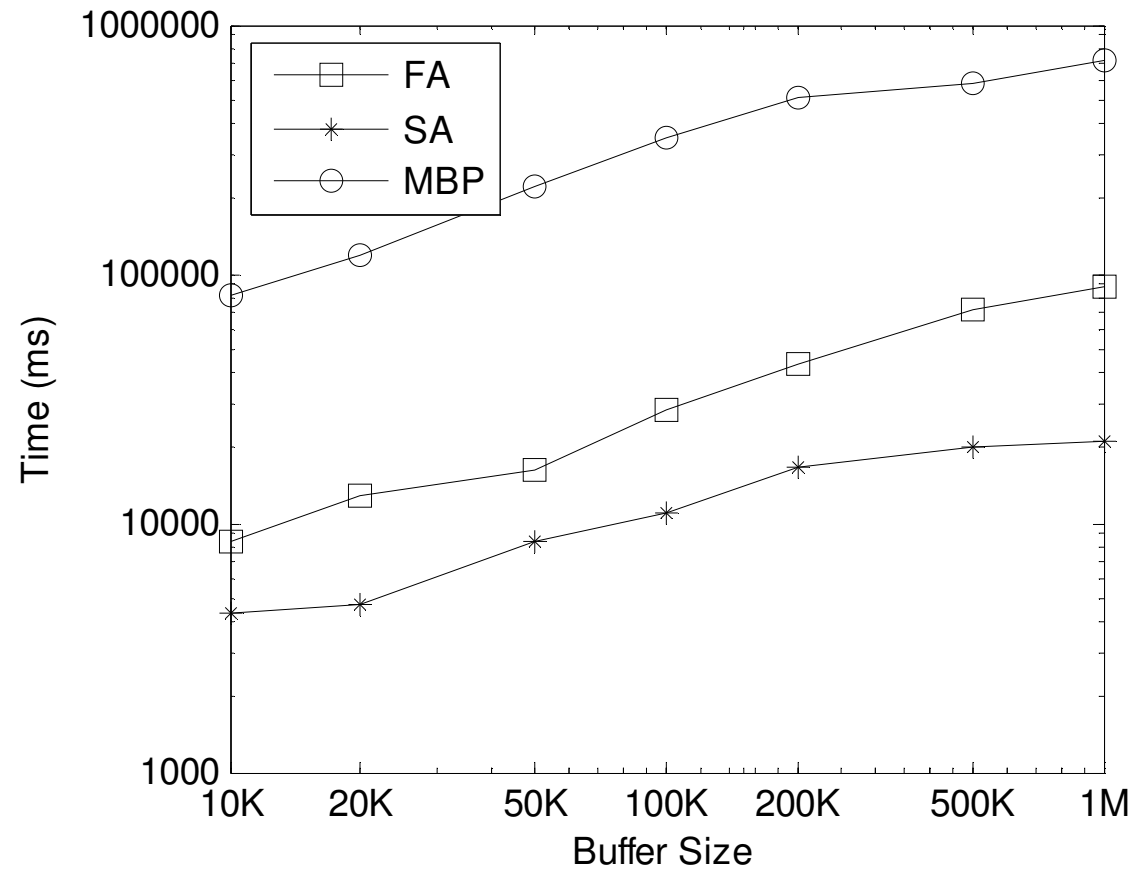
Pruning Efficiency



Memory Overhead



Real Data



Conclusions

- Dual space representation of the top-k problem
- Use of arrangements
- Tuple pruning technique

Thank you!

References

- [MBP06], K. Mouratidis, S. Bakiras, D. Papadias: Continuous Monitoring of Top-k Queries over Sliding Windows, *SIGMOD* 2006.

Synthetic Data

