# A Forward Scan based Plane Sweep Algorithm for Parallel Interval Joins

Panagiotis Bouros<sup>1</sup> and Nikos Mamoulis<sup>2</sup>

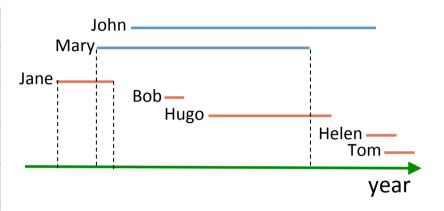
<sup>1</sup> Aarhus University, Denmark <sup>2</sup> University of Ioannina, Greece

employee	start	end
John	1994	2006
Mary	1992	2002

employee	start	end
Jane	1990	1993
Bob	1995	1996
Hugo	1997	2003
Helen	2005	2007
Tom	2006	2008

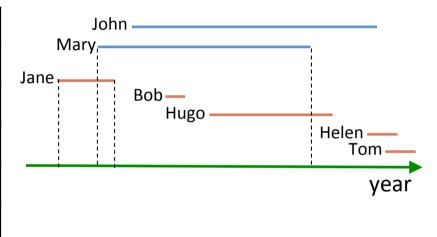
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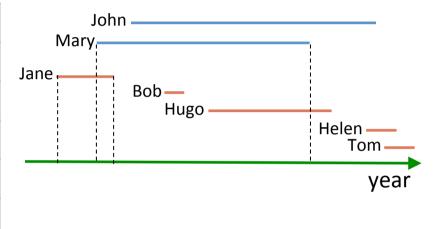
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Find all pair of employees whose period of work on D1 and D2 intersect

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- Applications
  - Temporal databases
  - Multidimensional data management
  - Uncertain data management

## Our focus

- Efficient evaluation of interval joins
  - Single-threaded processing
    - Simple plane sweep based method
    - Competitive to state-of-the-art
  - Parallel processing
    - Partitioning-based join
    - Share nothing

### SINGLE-THREADED PROCESSING

## Related work

## Nested loops & Sort-merge join

[Segev and Gunadhi, VLDB'89] [Gunadhi and Segev, ICDE'91]

### Partitioningbased

[Soo et al., ICDE'94] [Dignös et al., SIGMOD'14] [Cafagna and Böhlen, VLDBJ'17]

#### Index-based

[Zhang et al., ICDE'02] [Enderle et al., SIGMOD'04]

### Plane-sweep based

[Brinkhoff al., SIGMOD'93] [Arge et al., VLDB'98] [Piatov et al., ICDE'16]

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# Plane sweep methods

Endpoint-Based Join (EBI/LEBI)

[Piatov et al., ICDE'16]

- Sweep line stops both on start and end
- Backwards scan, Gapless hash map buffers open intervals

#### **Pros**

- ✓ No domain-point comparisons
- √ Tailored to modern hardware
- ✓ Main memory cache-aware
- ✓ Fast

### Cons

Special structure needed

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- Forward Scan based (FS)
  - Sweep lines stops only on start

### **Pros**

- ✓ Simple
- ✓ No special structure needed

### Cons

|R| + |S| + |R| S| comparisons in total

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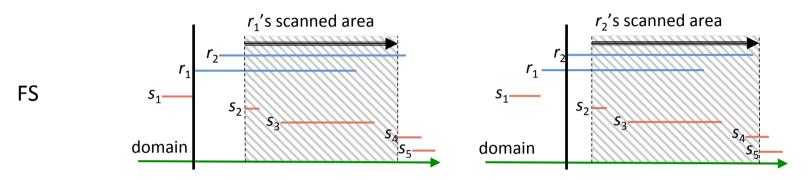
Pros

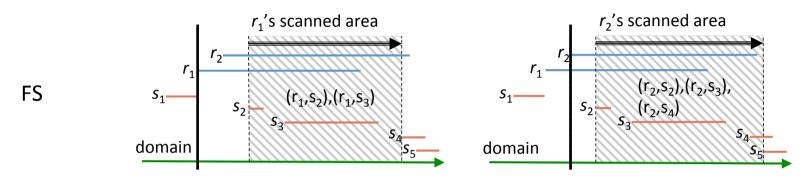
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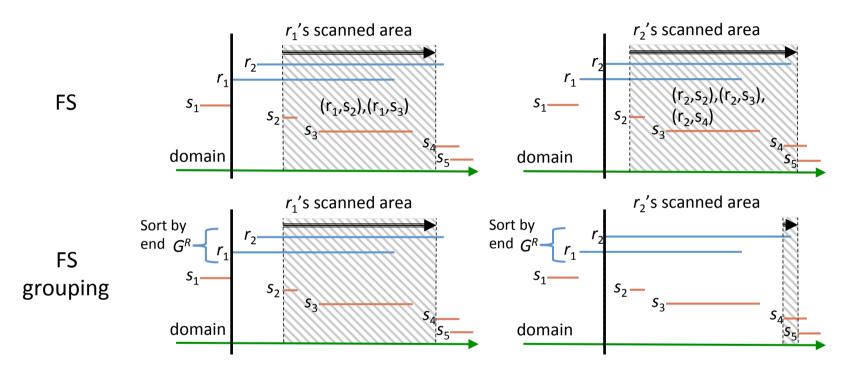
[Brinkhoff et al., SIGMOD'93]

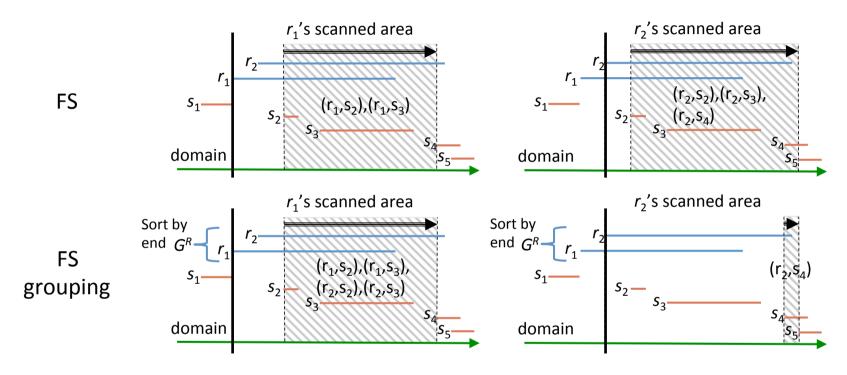
Cons

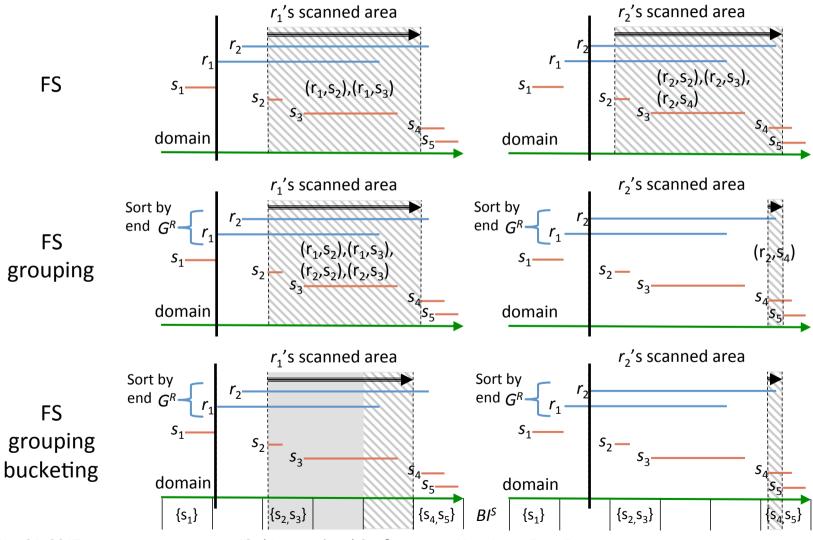
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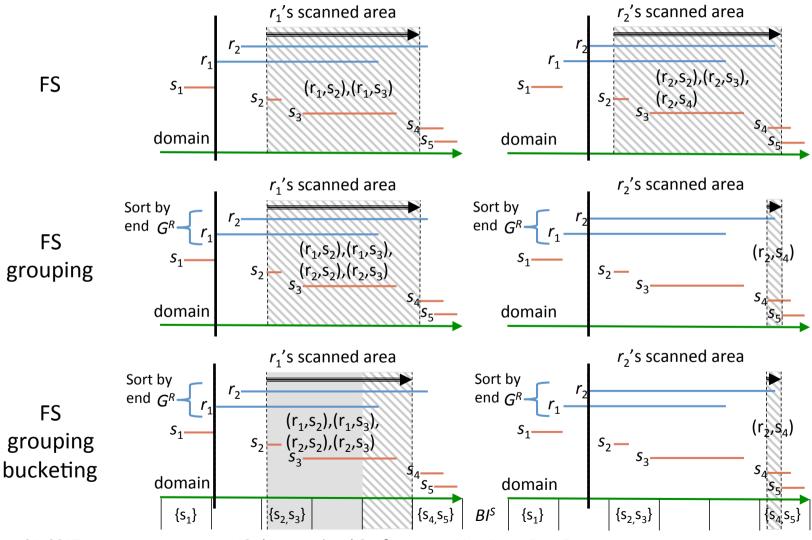








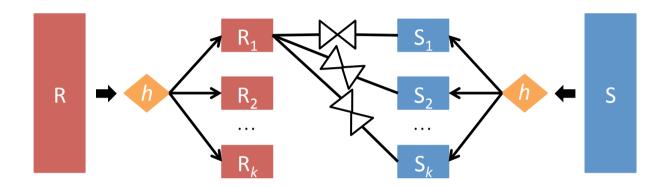




## **PARALLEL PROCESSING**

# Hash-based partitioning

[Piatov et al., ICDE'16]

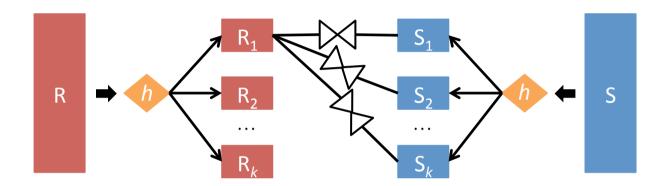


### Idea

- Randomly split each input into k partitions using hash function h
- Evaluate  $k^2$  independent partition joins

# Hash-based partitioning

[Piatov et al., ICDE'16]



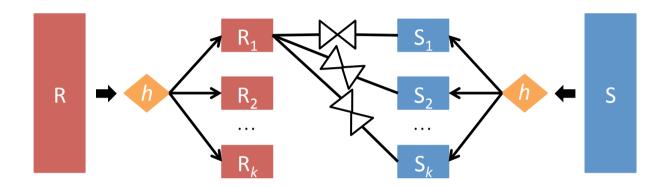
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### **Pros**

- ✓ Simple
- ✓ Load balancing

# Hash-based partitioning

[Piatov et al., ICDE'16]



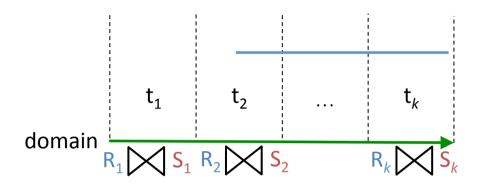
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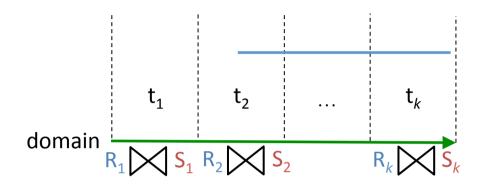
- ✓ Simple
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### Cons

- X Domain-point comparisons rise
  - $-2 \cdot k \cdot (|R|+|S|)$  for EBI/LEBI,  $k \cdot (|R|+|S|)$  for FS
- Degree of parallelism
  - n CPU cores →  $k = \sqrt{n}$  partitions



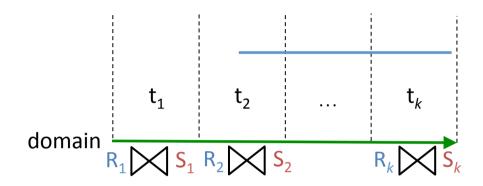
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  - Split domain into k tiles
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### **Pros**

- ✓ Degree of parallelism
  - $n \text{ CPU cores } \rightarrow k = n \text{ partitions}$
- ✓ Automatic duplicate elimination



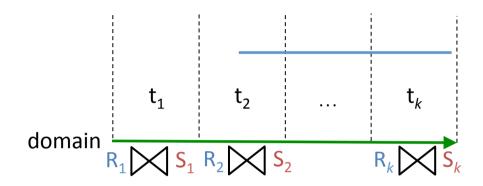
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- X Load balancing



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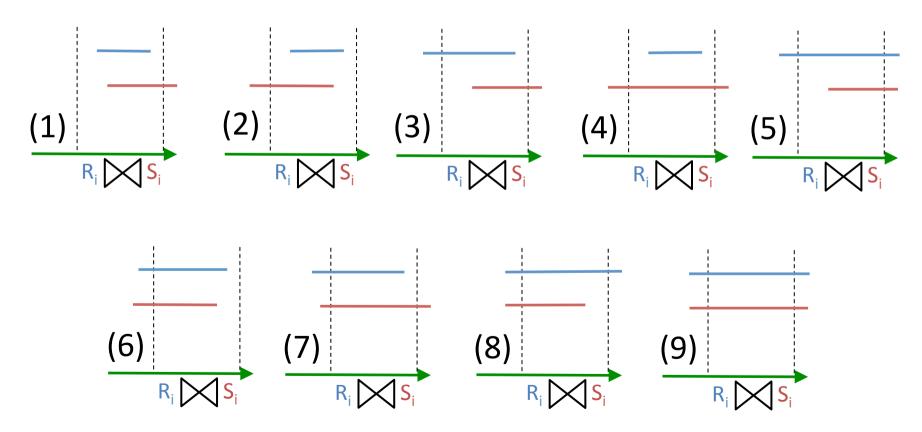
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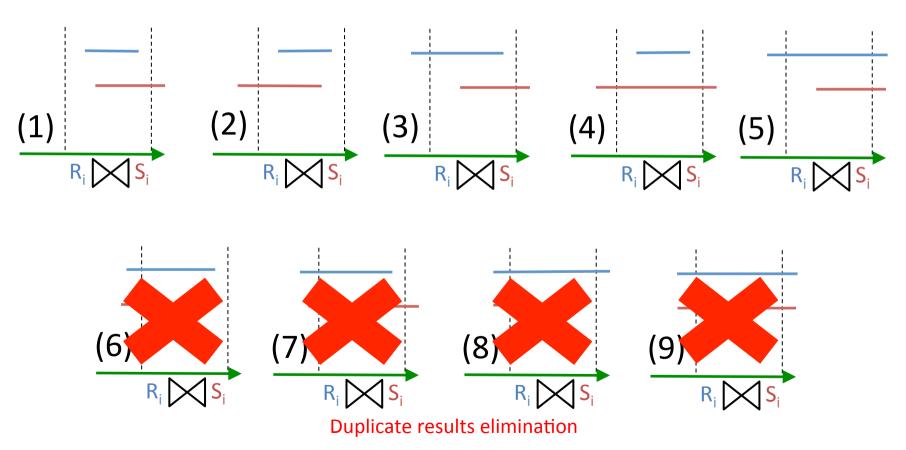
## Mini-joins break down

3 types of intervals → 9 types of mini-tasks



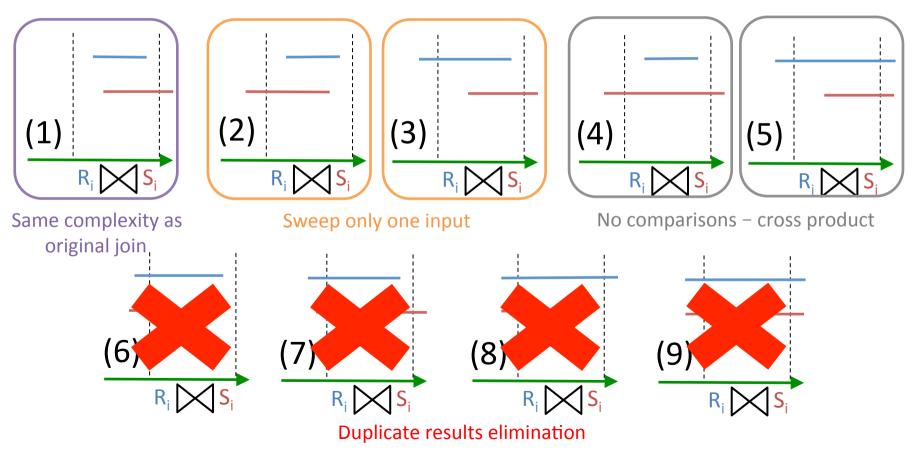
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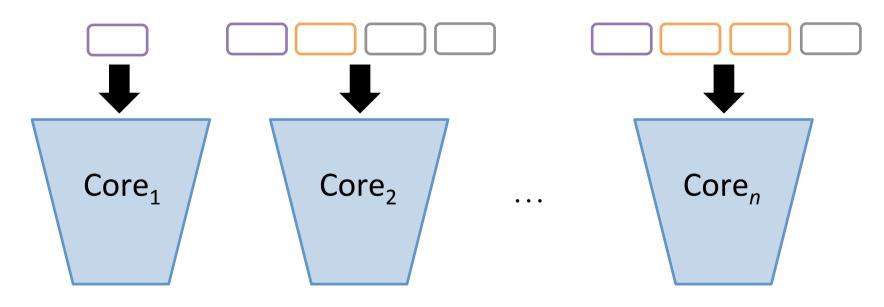


# Greedy scheduling

- Idea
  - Distribute  $1 + 5 \cdot (k-1)$  mini-joins to different cores
  - Evenly distribute load → minimize max load
  - NP-hard problem
    - Greedy approximation algorithm

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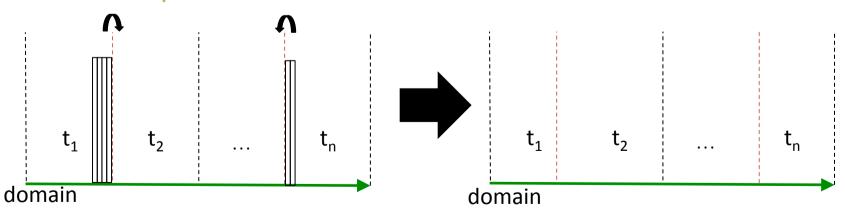
# Adaptive partitioning

- Idea
  - Create an initial uniform partitioning
  - Employ a very fine tiling granules
  - Move load between neighboring tiles
    - Move granules between neighboring tiles
    - Reposition borders of tiles

# Adaptive partitioning

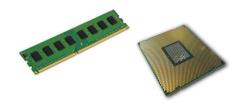
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### **EXPERIMENTAL ANALYSIS**

## Setup



### Hardware

- dual 10-core Intel(R) Xeon(R) CPU E5-2687W v3 @ 3.10 GHz with 128 GBs of RAM
- Hyper-threading enabled, up to 40 threads



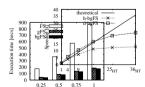
### Software

- Workload [ICDE'16] → XOR of start
- Loop unrolling forced, OpenMP for multi-threading



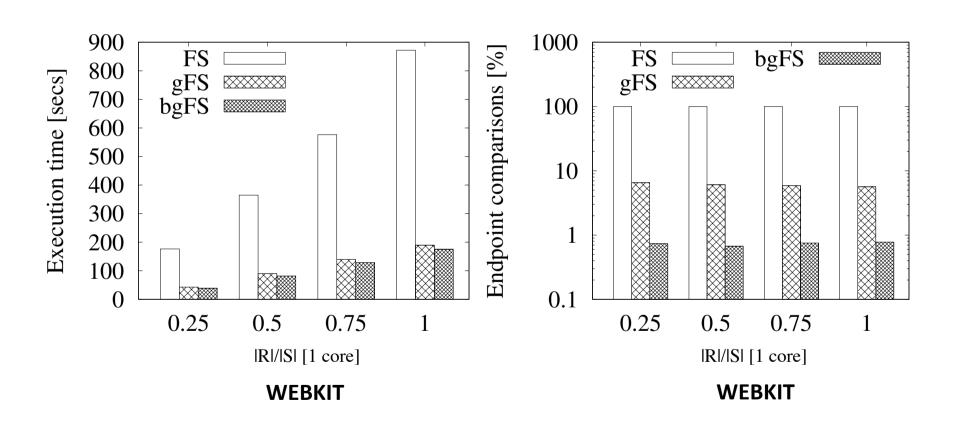
#### Datasets

- WEBKIT git repo, interval = period of time file unchanged
- BOOKS Aarhus libraries, interval = period of time book lent
- Synthetic
  - Interval duration follows exponential distribution, uniformly distributed start plus peaks



### Experiments

- Execution time, # comparisons, memory footprint
- Both self joins and non-self joins
- Vary |R|/|S|, # cores (threads)



### Single-threaded processing 30 bgFS<sup>\*</sup> 3 Patic/fixed LEBI DIP OIP 199000 static/fixed 10000 10000 1000 1000 100 10 [1 core]10000 0.2530.0.5 0.25Figure 6: 0.5SS 0.6 | IRI. SS 0.5 | UNITED BY OUT OF STREET STRE tatic/fixed IRI/ISI [4 core] 1000 GREENDPING WEBKIT Aug 31, 2017

Execution time [secs]

# Optimizme Domethedased Paradigm bqFS ( mj ntomic/uniform LEBI atomic/uniform atomic/adap<u>tiv</u>e 7 Ked 36<sub>HT</sub> Avg idle tim Avg idle 36<sub>HT</sub> 20

0

9

16

# cores

0.25

0.5

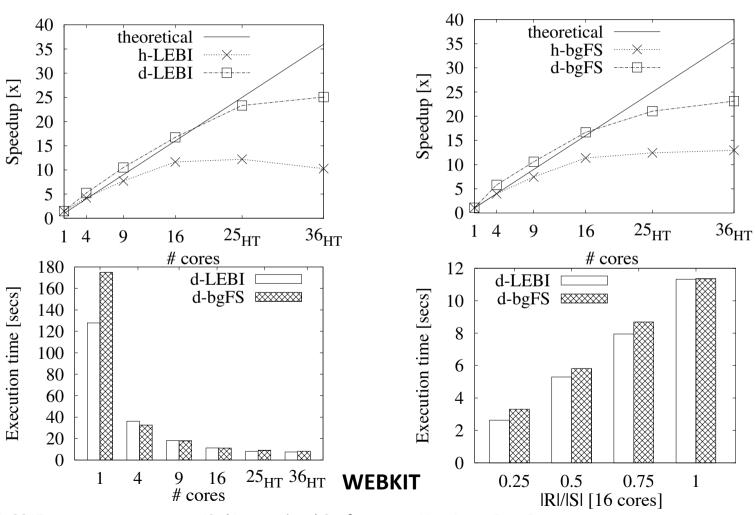
0.75

10000

**WEBKIT** 

25<sub>HT</sub> 36<sub>HT</sub>

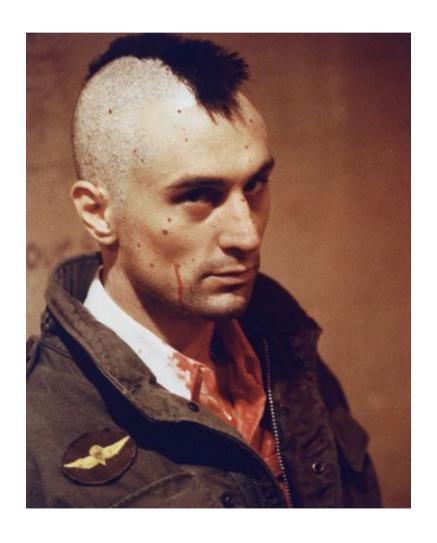
## Parallel processing



## To sum up

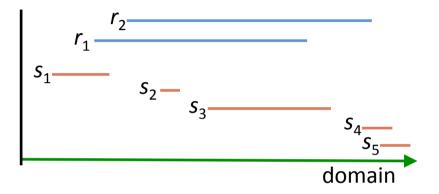
- Contributions
  - Efficient evaluation of interval joins
  - Single-threaded processing
    - Optimized bgFS, competitive to state-of-the-art EBI/LEBI
    - Lower memory footprint
  - Parallel processing
    - Novel domain-based partitioning paradigm
    - Higher speedup
- Future work
  - Other types of temporal joins
  - Other types of temporal operators
  - Parallel processing
    - Data-level parallelism, share data between threads

# Questions?



### **EXTRAS**

[Piatov et al., ICDE'16]



### **Endpoint indices**

$$\begin{split} &\mathsf{EI^R} = \{\mathsf{r}_1.start,\,\mathsf{r}_2.start,\,\mathsf{r}_1.end,\,\mathsf{r}_2.end\} \\ &\mathsf{EI^S} = \{\mathsf{s}_1.start,\,\mathsf{s}_1.end,\,\mathsf{s}_2.start,\,\mathsf{s}_2.end,\,\mathsf{s}_3.start,\,\mathsf{s}_3.end,\,\mathsf{s}_4.start,\,\mathsf{s}_5.start,\,\mathsf{s}_4.end,\,\mathsf{s}_5.end\} \end{split}$$

#### **Active sets**

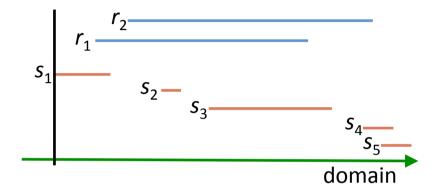
$$A^R = \{\}$$

$$A^S = \{\}$$

#### Result

{}

[Piatov et al., ICDE'16]



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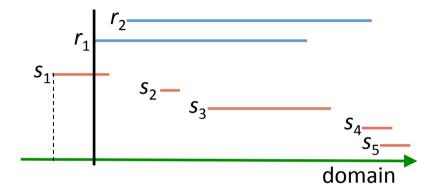
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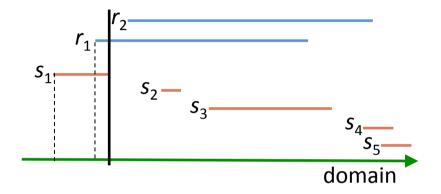
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 $\{(r_1,s_1)\}$ 

[Piatov et al., ICDE'16]



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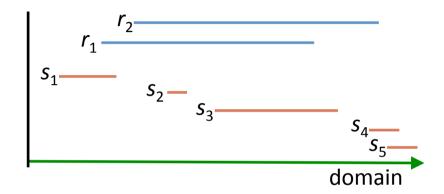
#### **Pros**

- ✓ No domain-point comparisons when producing results
- ✓ Tailored to modern hardware
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#### Cons

Special data structure needed for active sets, support for efficient updates and scans

[Brinkhoff et al., SIGMOD'93]



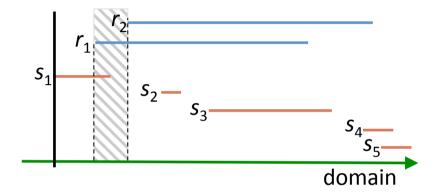
### **Sorted inputs**

R = {
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,  $r_2$ }  
S = { $s_1$ ,  $s_2$ ,  $s_3$ ,  $s_4$ ,  $s_5$ }

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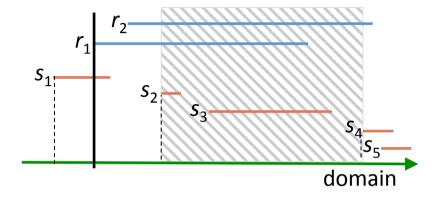
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#### Result

$$\{(r_1,s_1), (r_1,s_2), (r_1,s_3)\}$$

[Brinkhoff et al., SIGMOD'93]

### **Pros**

- ✓ Simple
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### Cons

 $\nearrow$  Each join result requires a domain-point comparison,  $|R| + |S| + |R| \otimes |S|$  comparisons in total

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