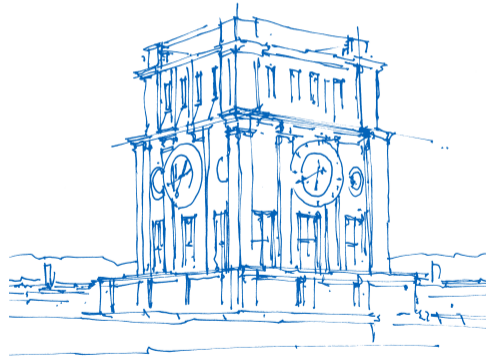


# A Practical Approach to Groupjoin and Nested Aggregates

Philipp Fent, Thomas Neumann

Technische Universität München

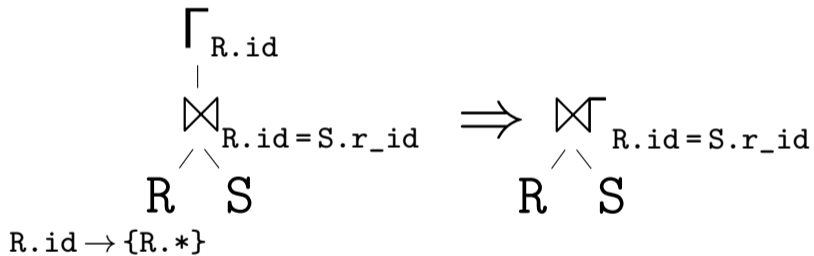
VLDB 2021



*Uhrenturm der TUM*

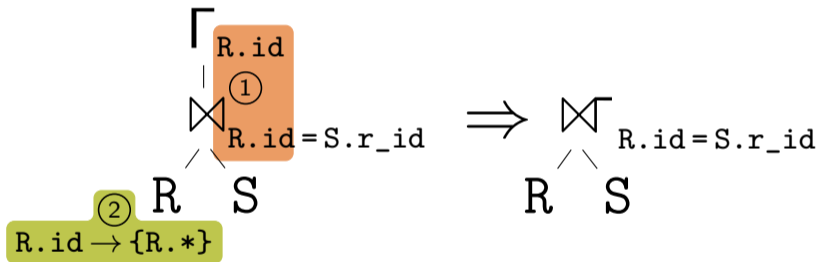
# Groupjoin

Combines Join and Group-By



# Groupjoin

Combines Join and Group-By



## The MD-join : An Operator for Complex OLAP

Damianos Chatziantoniou  
Panakea Software Inc.  
damianos@panakea.com

Michael Akinde  
Dept. of Computer Science\*  
Aalborg University  
strategy@cs.auc.dk

Theodore Johnson  
Database Research Center  
AT&T Labs - Research  
johnsont@research.att.com

Samuel Kim  
Dept. of Computer Science  
Stevens Institute of Tech.  
skim@cs.stevens-tech.edu

## Main Memory Implementations for Binary Grouping

Norman May and Guido Moerkotte

University of Mannheim  
B6, 29  
68131 Mannheim, Germany  
{norman,moez}@informatik.uni-mannheim.de

## Accelerating Queries with Group-By and Join by Groupjoin

Guido Moerkotte  
Universität Mannheim  
Mannheim, Germany  
moerkotte@informatik.uni-mannheim.de

Thomas Neumann  
Technische Universität München  
Munich, Germany  
neumann@in.tum.de

## Getting Swole: Generating Access-Aware Code with Predicate Pullups

Andrew Crotty  
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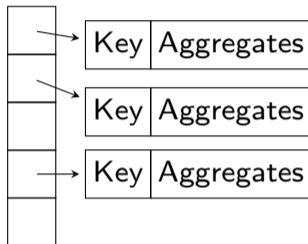
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## Combined Hashtable



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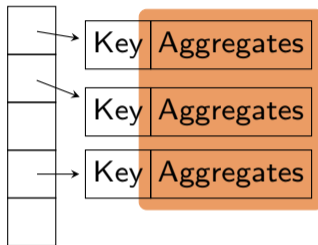
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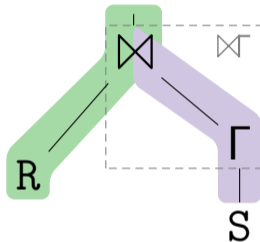
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## Combined Hashtable



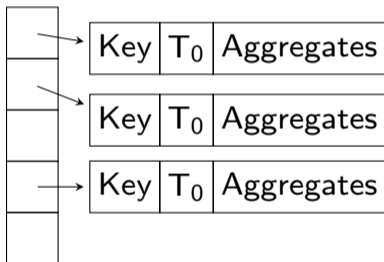
# Groupjoin

Avoiding Contention with Eager Right Execution



- ▶ Re-uses thread-local  $\Gamma$  logic
- ▶ Works well for  $\sigma_S \approx 1$

Global Hashtable



Thread-local Hashtables

T<sub>1</sub>

$$\Gamma_1 = \{\}$$

T<sub>2</sub>

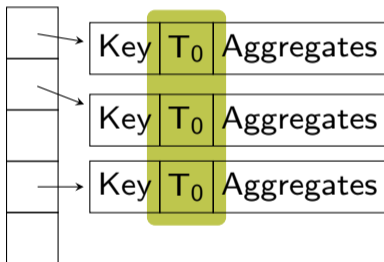
$$\Gamma_2 = \{\}$$

T<sub>3</sub>

$$\Gamma_3 = \{\}$$



Global Hashtable



Thread-local Hashtables

T<sub>1</sub>

$$\Gamma_1 = \{\}$$

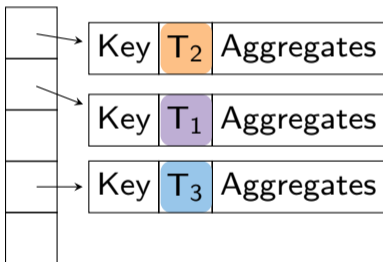
T<sub>2</sub>

$$\Gamma_2 = \{\}$$

T<sub>3</sub>

$$\Gamma_3 = \{\}$$

Global Hashtable



Thread-local Hashtables

T<sub>1</sub>

$$\Gamma_1 = \{\}$$

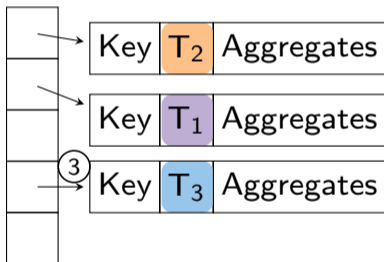
T<sub>2</sub>

$$\Gamma_2 = \{\}$$

T<sub>3</sub>

$$\Gamma_3 = \{\}$$

### Global Hashtable



### Thread-local Hashtables

**T<sub>1</sub>**

$$\Gamma_1 = \{\textcircled{3} : \gamma_1\}$$

**T<sub>2</sub>**

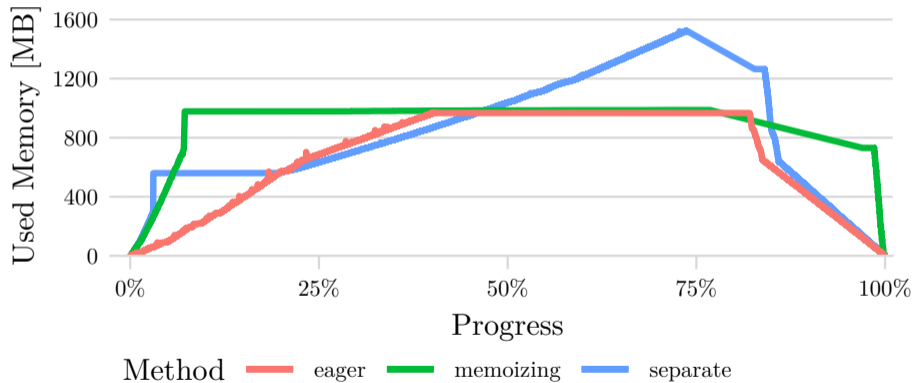
$$\Gamma_2 = \{\textcircled{3} : \gamma_2\}$$

**T<sub>3</sub>**

$$\Gamma_3 = \{\}$$

# Groupjoin

## Comparison

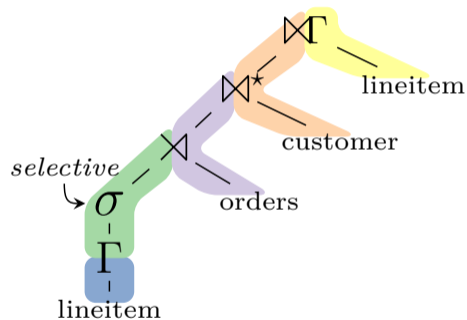


# Nested Aggregates

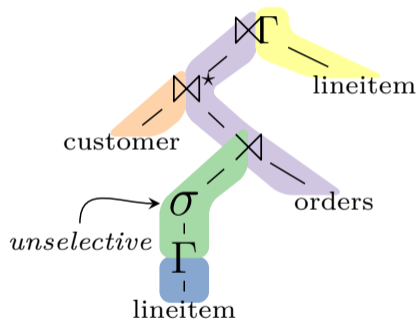
- ▶ Decorrelation supported by Groupjoin
- ▶ HAVING predicates are hard to estimate

# Nested Aggregates

- ▶ Decorrelation supported by Groupjoin
- ▶ HAVING predicates are hard to estimate



a) Selective  $\sigma$ -Predicate



b) Unselective  $\sigma$ -Predicate

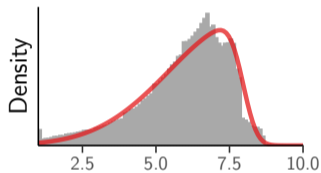
# Estimating Aggregates

- ▶ Numerical columns  $\sim \mathcal{N}(\mu, \sigma^2)$
- ▶ Cheap and generalizes nicely ... but inherently symmetric

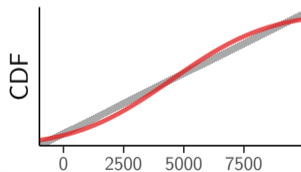
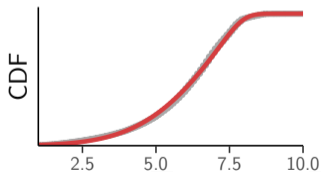
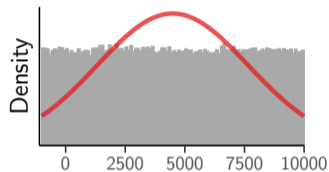
# Estimating Aggregates

- ▶ Numerical columns  $\sim \mathcal{N}(\mu, \sigma^2)$
- ▶ Cheap and generalizes nicely ... but inherently symmetric

IMDb movie rating



TPC-H customer balance



■ Observed Data ■ Calculated skew-normal fit



# Estimating Aggregates

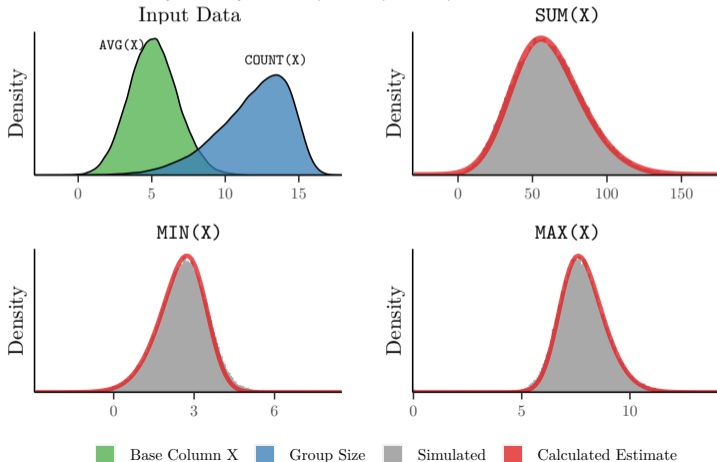
Using a skew-normal distribution

- ▶  $\Pr[x \leq c] = \Phi_{sn}(c)$
- ▶ Also approximates  $(x \circ y)$ , AVG/SUM/MIN/MAX

# Estimating Aggregates

Using a skew-normal distribution

- ▶  $\Pr[x \leq c] = \Phi_{sn}(c)$
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# Practical Groupjoins and Nested Aggregates

Overall impact on TPC-H and TPC-DS

- ▶ Big improvements with cost model advised Groupjoin
- ▶ Slightly better query plans with HAVING estimates

Live queries: <https://umbra-db.com/interface>

