Locality-Sensitive Operators for Parallel Main-Memory Database Clusters

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Scale Out

- HyPer: High-performance in-memory transaction and query processing system
- Scale out to process very large inputs
- Aim at clusters with large main memory capacity
- A server with 20 cores and 256 GB RAM costs ~\$7,500



Running Example (1)

- Focus on analytical query processing in this talk
- TPC-H query 12 used as running example
- Runtime dominated by join orders ⋈ lineitem



Running Example (2)

- Relations are equally distributed across nodes
- We make no assumptions on the data distribution
- Thus, tuples may join with tuples on remote nodes
- Communication over the network required



CPU vs. Network



CPU speed has grown much faster than network bandwidth

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Scale Out: Network is the Bottleneck

- Single node: Performance is bound algorithmically
- Cluster: Network is bottleneck for query processing
- We propose a novel join algorithm called Neo-Join
- Goal:

Increase local processing to close the performance gap



Neo-Join: Network-optimized Join

1. **Open Shop Scheduling** Efficient network communication

2. Optimal Partition Assignment Increase local processing

3. Selective Broadcast Handle value skew

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Open Shop Scheduling

Efficient network communication

Standard Network Model

Star topology

Nodes are connected to a central switch

- Fully switched All links can be used simultaneously
- Fully duplex Nodes can both send and receive at full speed



Bandwidth Sharing



- Simultaneous use of a single link creates a bottleneck
- Reduces bandwidth by at least a factor of 2

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Naïve Schedule



- Node 2 and 3 send to node 1 at the same time
- Bandwidth sharing increases network duration significantly

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Open Shop Scheduling (1)

Avoiding bandwidth sharing translates to **open shop scheduling**:

- A job consists of one task per processor
- A processor can perform at most one task at a time
- At most one task of a job can be processed at a time



Open Shop Scheduling (2)

Avoiding bandwidth sharing translates to **open shop scheduling**:

- A sender has one transfer per receiver
- A receiver should receive at most one transfer at a time
- A sender should send at most one transfer at a time



Open Shop Scheduling (3)

Compute optimal schedule:

- Edge weights represent total transfer duration
- Scheduler repeatedly finds perfect matchings
- Each matching specifies one communication phase
- Transfers in a phase will never share bandwidth



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Optimal Schedule



- Open shop schedule achieves minimal network duration
- Schedule duration determined by maximum straggler

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Optimal Partition Assignment

Minimize network duration for distributed joins

Distributed Join

- Tuples may join with tuples on remote nodes
- Repartition and redistribute both relations for local join
- Tuples will join only with the corresponding partition
- Using hash, range, radix, or other partitioning scheme
- In any case: Decide how to assign partitions to nodes



fragmented

redistributed

Running Example: Hash Partitioning



Assign Partitions to Nodes (1)

Option 1: Minimize network traffic

- Assign partition to node that owns its largest part
- Only the small fragments of a partition sent over the network
- Schedule with minimal network traffic may have high duration

hash partitioning (x mod 3)



Assign Partitions to Nodes (2)

Option 2: Minimize response time:

- Query response time is time from request to result
- Query response time dominated by network duration
- To minimize network duration, minimize maximum straggler

hash partitioning (x mod 3)







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Minimize Maximum Straggler

- Formalized as mixed-integer linear program
- Shown to be NP-hard (see paper for proof sketch)
- In practice fast enough using CPLEX or Gurobi (< 0.5 % overhead for 32 nodes, 200 M tuples each)
- Partition assignment can optimize any partitioning

minimize w, subject to $w \ge \sum_{j=0}^{p-1} h_{ij} (1 - x_{ij}) \qquad 0 \le i < n$ $w \ge \sum_{j=0}^{p-1} \left(x_{ij} \sum_{k=0, i \ne k}^{n-1} h_{kj} \right) \qquad 0 \le i < n$ $1 = \sum_{i=0}^{n-1} x_{ij} \qquad 0 \le j < p$

Running Example: Locality



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Locality

- Running example exhibits time-of-creation clustering
- Radix repartitioning on most significant bits retains locality
- Partition assignment can exploit locality
- Significantly reduces query response time

radix partitioning (MSB) • n1 15 1 0 • n2 1 11 1 • n3 0 2 11 P1 P2 P3



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Selective Broadcast

Handle value skew

Running Example: Skew



Skew

- Skewed partition P₂ has to be assigned, e.g., to node 3
- Node 3 will receive much more than its fair share
- May balance skewed partitions by creating more partitions
- However: More expensive and high skew is still a problem

hash partitioning (mod 3)



Broadcast

- Alternative to data repartitioning
- Replicate the smaller relation between all nodes
- Larger relation remains fragmented across nodes



Selective Broadcast

- Decide per partition whether to assign or broadcast
- Broadcast orders for P₂, let line items remain fragmented
- Assign the other partitions taking locality into account
- Improves performance for high skew and many duplicates

hash partitioning (mod 3)





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Role Reversal

- Selective broadcast allows for role reversal
- Broadcast different partitions by different relations

Example:

- Large suppliers produce a large variety of parts
- Important parts available from many suppliers



Evaluation

Experimental Setup

- Cluster of 4 nodes
- Core i7, 4 cores, 3.4 GHz, 32 GB RAM
- Gigabit Ethernet
- Tuples consist of 64 bit key, 64 bit payload

Locality

- Vary locality from 0 % (uniform distribution) to 100 % (range partitioning)
- Neo-Join improves join performance from 29 M to 156 M tuples/s (> 500 %)
- 3 nodes, 600 M tuples



Skew

Zipfian distribution models realistic data skew

- Using more partitions alleviates the problem
- Selective broadcast actually improves performance for skewed inputs
- 4 nodes, 400 M tuples

	Zipf factor s				
partitions	0.00	0.25	0.50	0.75	1.00
16	27 s	24 s	23 s	29s	44s
512	23 s	23 s	23 s	23 s	33s
16 (SB)	24 s	24 s	23 s	20s	10s

TPC-H Results (scale factor 100)

- Results for three selected TPC-H queries
- Broadcast outperforms hash for large relation size differences
- Neo-Join always performs better due to selective broadcast and locality
- 4 nodes, scale factor 100



Summary

Motivation:

- Scale out to handle very large inputs
- Network is the bottleneck
- Thus, reduce network duration

Contributions:

- Maximize bandwidth usage with Open Shop Scheduling
- Exploit locality with Optimal Partition Assignment
- Handle skewed inputs with Selective Broadcast