

Transaction Reordering and Grouping for Continuous Data Loading

Gang Luo¹

Jeffrey F. Naughton²

Curt J. Ellmann²

Michael W. Watzke³

IBM T.J. Watson Research Center¹

University of Wisconsin-Madison²

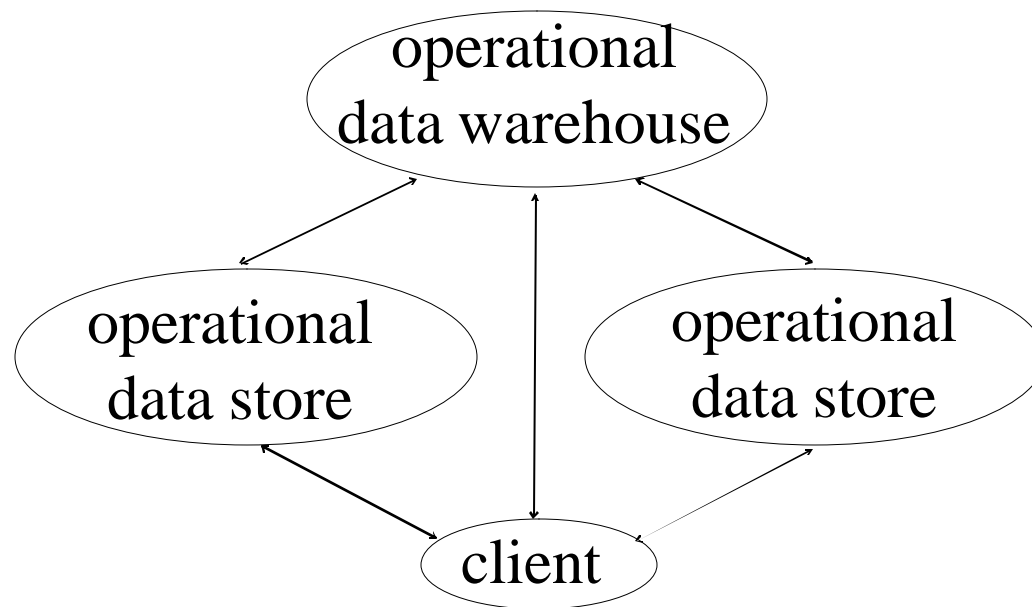
NCR³

luog@us.ibm.com
ellmann@wisc.edu

naughton@cs.wisc.edu
michael.watzke@ncr.com

Continuous Data Loading

- For operational data warehousing, it is critical that the data in the warehouse be as up-to-date as possible
- Commercial RDBMS vendors (Oracle, Teradata, etc.) provide tools for continuous data loading



Immediate Materialized View Maintenance

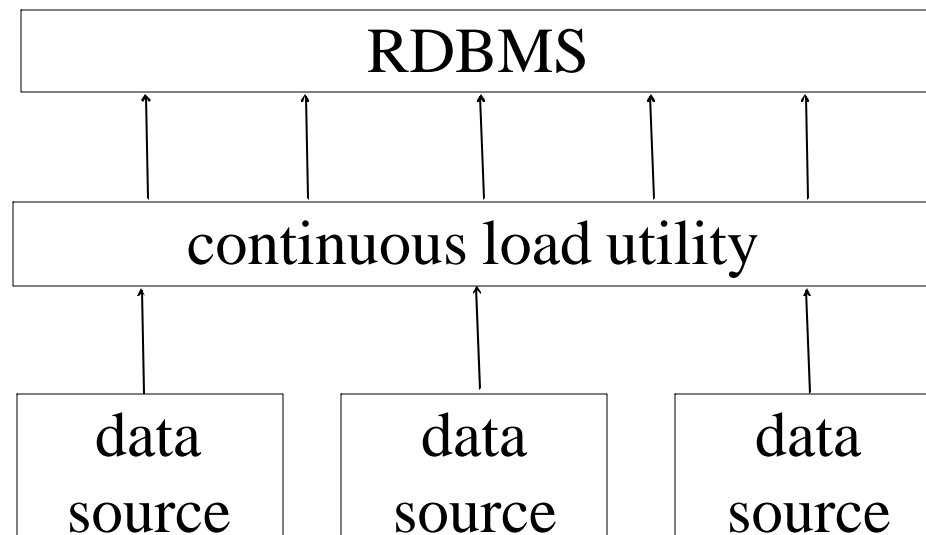
- Necessity:
 - Inconsistency between materialized views and base relations may not be acceptable to all applications (e.g., real-time decision making)
 - Mandated in the TPC-R benchmark
- Problem: Concurrent updates to multiple base relations of the same materialized join view cause deadlocks
 - Reason: Materialized join view maintenance changes update-only transactions to update-read transactions
- Solution: Reorder the updates

Outline

- Existing Continuous Load Utilities
- Problem with Immediate Materialized Join View Maintenance
- Solution with Reordering
- Performance

Continuous Data Loading Architecture

- Data comes in the form of modification operations (insert, delete, update)
- Load data in the form of transactions through sessions



Assumptions

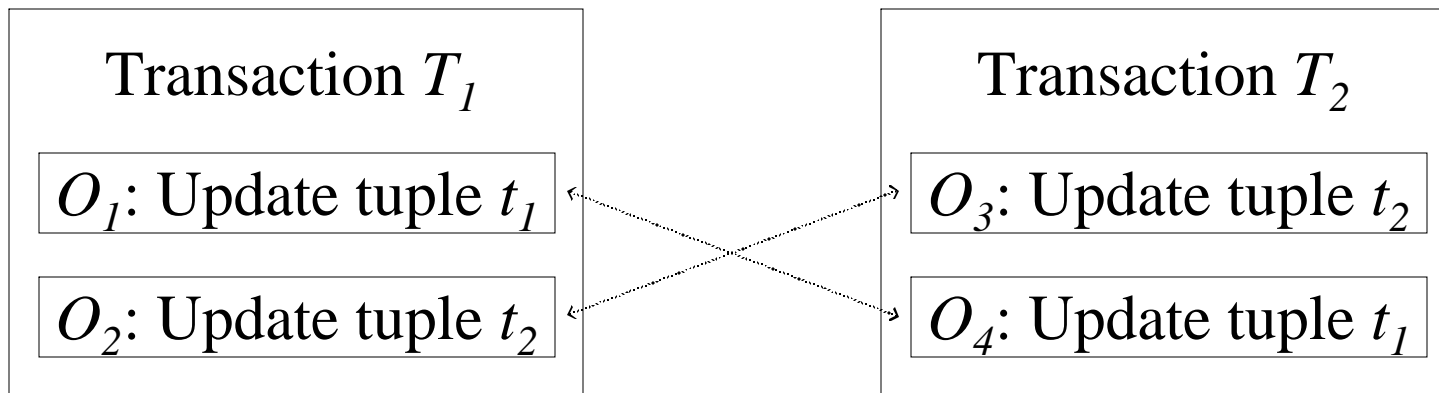
- The system has no control over the order in which modification operations arrive at the RDBMS
- The continuous load utility looks to the RDBMS like a series of transactions, each containing a single modification operation
- No order imposed or assumed for the load transactions
 - The load process can arbitrarily reorder the single-modification-operation transactions
- No requirement on whether multiple modification operations can or cannot commit/abort together
 - The load process can arbitrarily group the single-modification-operation transactions

Assumptions - Continued

- Use strict two-phase locking protocol
- Use multiple-granularity locking protocol
 - Two levels in the locking hierarchy: table-level locks and tuple-level locks
- Each modification operation can be done with tuple-level locks

Common Wisdom

- Grouping to improve efficiency
 - Combine multiple modification operations into a single transaction
- Partitioning to avoid deadlock
 - Partition tuples among different sessions
 - Modification operations on the same tuple are always through the same session



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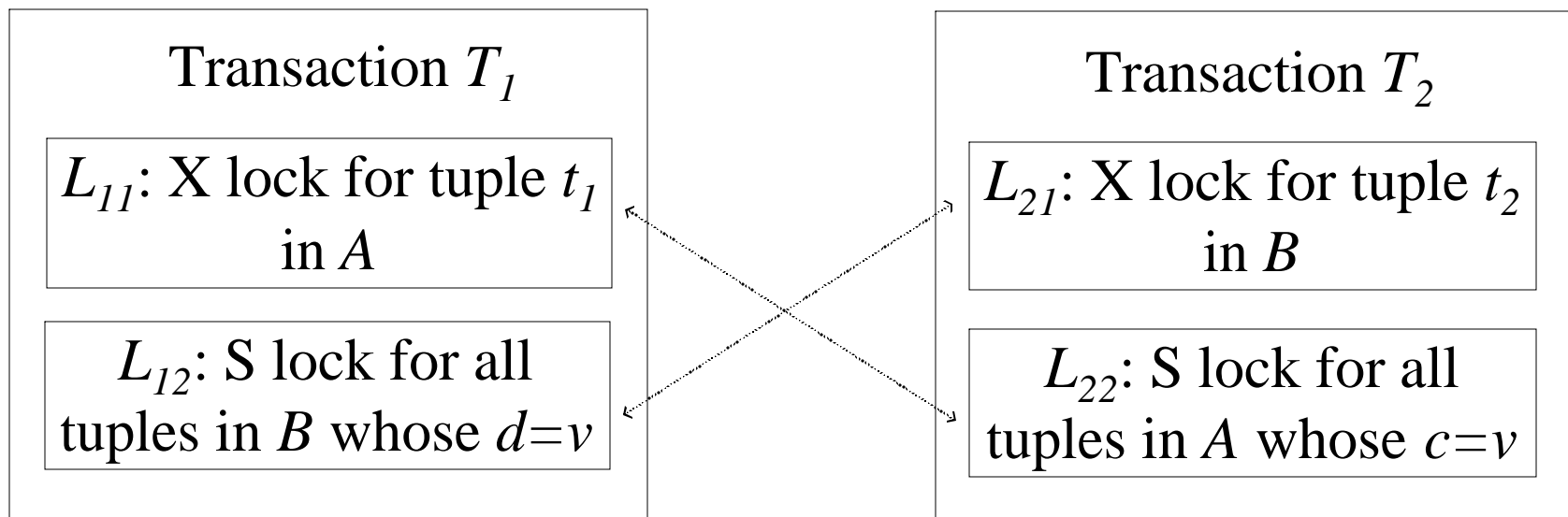
Join View

- A join view JV on two relations A and B pre-computes a selection and projection (and aggregation) of $A \bowtie B$

```
create join view  $JV$  as  
select *  
from  $A, B$   
where  $A.c=B.d$ ;
```

Impact of Immediate Join View Maintenance

- Two transactions
 - T_1 : Modify tuple t_1 in A whose $c=v$
 - T_2 : Modify tuple t_2 in B whose $d=v$



Deadlock Probability

- $k > 1$ concurrent transactions
- Each transaction T contains n modification operations
 - T modifies A with probability p
 - T modifies B with probability $1-p$
 - Each modification operation modifies a random tuple in A (B)
- Totally s distinct values for $A.c$ ($B.d$)
- Deadlock probability: $\min(1, p(1-p)(k-1)n^2/(2s))$

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Intuition for Transaction Reordering

- Only run “compatible” transactions concurrently
 - Reorder transactions so that transactions updating A are executed, then transactions updating B are executed

Rules for Transaction Reordering

- **Rule 1:** At any time, for any join view JV , only allow data to be loaded into one base relation of JV
- **Rule 2:** For those modification operations on the same base relation, use the partitioning method
- **Rule 3:** Use a high concurrency locking protocol (e.g., the V locking protocol) on join views

Justification

- Rules 1 and 2 avoid deadlocks resulting from lock conflicts on the base relations
- Rule 3 avoids deadlocks resulting from lock conflicts on the join views

Data Structures for Transaction Reordering

- For each base relation R_i , maintain a number J_i
 - J_i records the number of running transactions that modify R_i
- For each session S_m , maintain a queue Q_m
 - Q_m records the transactions waiting to be run through S_m

Transaction Reordering Algorithm Outline

- Analyze potential lock conflicts using the data structures
 - Transaction starts execution: Increment J_i
 - Transaction finishes execution: Decrement J_i
- Desirable transaction updating R_i : Does not conflict with any running transaction
 - Check whether or not $J_j = 0$ for each base relation R_j of the same join view as R_i

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Performance

- Testing environment
 - Commercial parallel RDBMS
 - Windows 2000 OS
 - Intel workstation, each data server node has one 400MHz CPU, 250M memory, one 8GB disk
- System configuration in the parallel RDBMS
 - Allocate a processor and a disk for each data server
 - Test system configuration has four data server nodes

Performance - Continued

- Definition of base relations and join view:

demand (partkey, date, quantity, custkey, comment)

inventory (partkey, date, quantity, extended_cost, extended_price)

create join view onhand_demand as

select d.partkey, d.date, d.quantity, d.custkey, i.quantity

from demand d, inventory i

where d.partkey=i.partkey and d.date=i.date

partitioned on d.custkey;

- Modification operations:

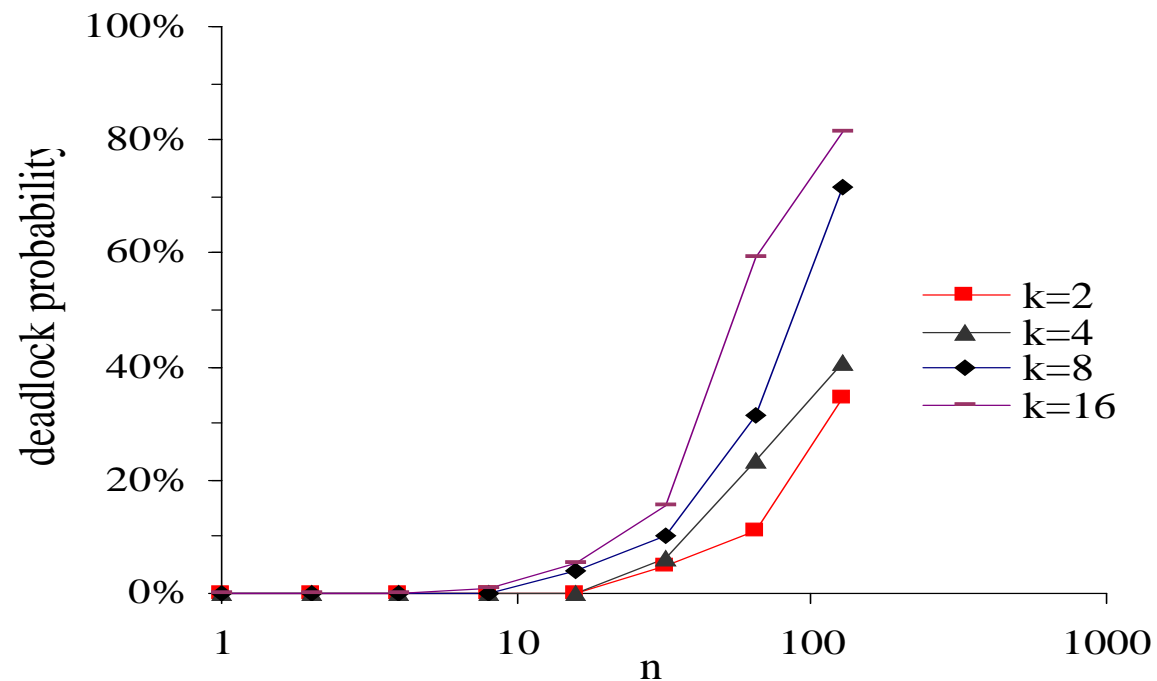
- O_1 : Update one tuple in the *inventory* relation with a specific *partkey* value and today's *date*
- O_2 : Insert one tuple into the *demand* relation with a specific *partkey* value and today's *date*

Performance - Continued

- Compare the naive method vs. the transaction reordering method
 - Run a stream of T_1 's and T_2 's (50% are T_1 's, 50% are T_2 's)
 - Only combine modification operations on the same base relation into a single transaction
 - Each transaction contains the same number n of modification operations
 - 10,000 active parts
 - k sessions
 - Measure throughput of modification operations

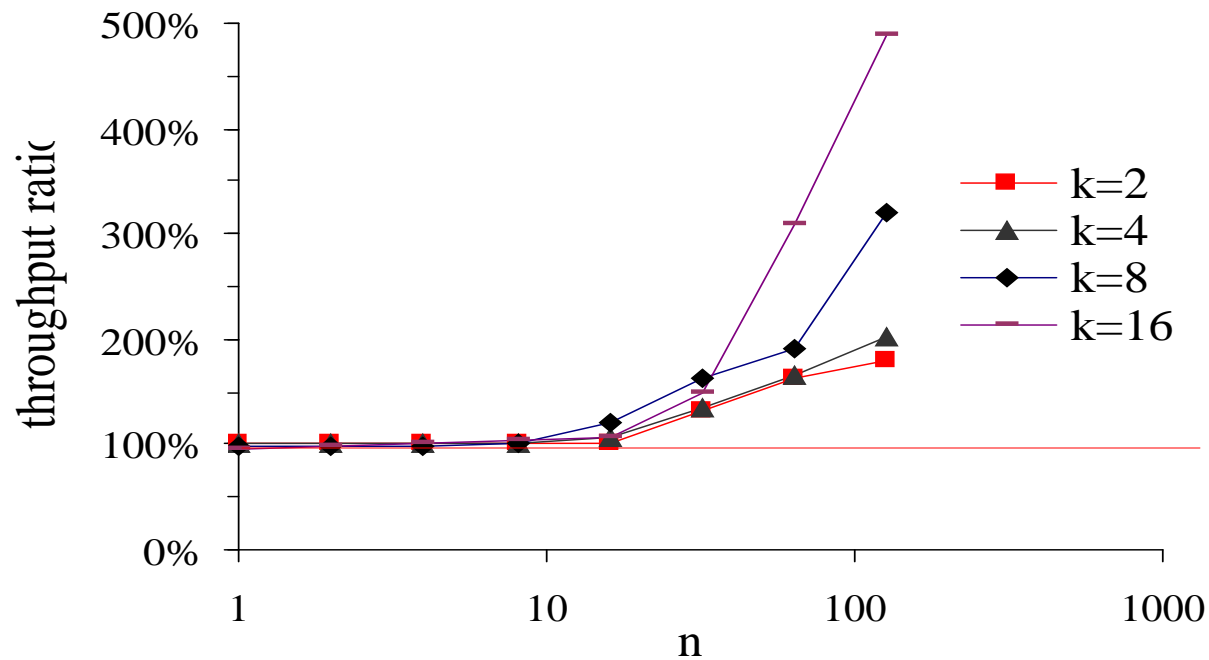
Performance - Deadlock Probability

- Deadlock probability of the naive method increases with both n and k



Performance - Throughput Improvement

- Improvement of throughput gained by transaction reordering increases with both n and k



Questions?