TPC-BiH: A Benchmark for Bitemporal Databases

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Motivation: Inventory Management

OrderedItems

Inventory

OrderID	ProdID	Ship	Time		ProdID	Price	Time
1	mbpro	false	10 – ∞		w530	3000	1 – 15
1	w530	false	16 – 19	DH	mbpro	2000	1 – 15
2	mbair	false	17 – ∞	-1-1-1	mbair	2500	1 – ∞
1	w530	true	<u>19 – ∞</u>	IF IF	w530	3050	15 – ∞
		重型型	AND EAR		mbpro	2100	15 – 17
					mbpro	1900	<u>17 – ∞</u>
					x220	2300	20 – ∞

Analysing temporal data

- 1) What was the total value of the inventory at each point in time?
 - Temporal Aggregation
- 2) What was the price of MacBook Pro on January 15th?
 - Time-Travel
- 3) What was the most expensive order last year?
 - Temporal Join





- Increasing demand for temporal features
 - Dozens of apps require it
 - Adoption of bitemporal features in SQL:2011
 - Support by commercial database vendors
- Users need to assess performance
 - Keeping and querying history not reflected by existing benchmarks
 - Temporal processing is performance-critical
 - Understanding of the performance characteristics of alternative implementations of temporal operators





Outline

- Motivation
- State-of-the-art
- The TPC-BiH Benchmark
- Preliminary Results
- Conclusion







Temporal Data: State of the Art

- Long-Running Research
 - Access Methods for Multiversion Data. David Lomet et al. 1989
 - TSQL Language Specification. Richard Snodgrass et al. 1994
- Currently limited support on language side
 - SQL:2011 is a big step forwards
 - No complete coverage of all use cases
- Commercial systems start adding temporal features
 - Bitemporal data: Oracle, Teradata, DB2 (SQL:2011)
 - System time: SAP HANA





Existing Temporal Benchmarks

- Previous work on benchmarking the temporal dimension:
 - Requirements specifications (Dunham et al., 1995)
 - Functional tests (test suite from the TSQL2 editors)
- Recent activity and closest match: "Adding a Temporal Dimension to the TPC-H Benchmark" (TPCTC 2012)
 - Limited extension of TPC-H schema
 - Sketches of data and queries
 - Focus on TSQL2





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Methodology of the Benchmark

Overall goals

- Comprehensive benchmark for bitemporal query processing
- Benchmark settings reflect
 - real-life customer workloads
 - synthetic tests of temporal properties
- Benchmark is targeted towards SQL:2011

Schema

- Extend TPC-H tables with different types of history classes:
 - fully bitemporal
 - degenerated
 - multiple user times





Methodology of the Benchmark (II)

Data

- Initialization
 - Initial version is regular TPC-H data
 - Derive application time from TPC-H data semantics
- History
 - Evolve through carefully designed update scenarios
 - Maintain overall distributions at each point in tinme

Queries

- Coverage of common temporal DB requirements
- Stressing the system for individual time dimensions
- Correlations among the dimensions whenever relevant



Schema

PART

<u>PARTKEY</u> NAME

MFGR

WIFUK

BRAND

TYPE

SIZE

CONTAINER

RETAILPRICE

COMMENT

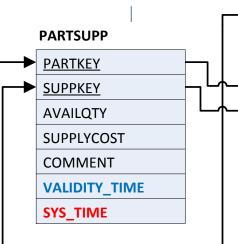
AVAILABILITY_TIME

SYS_TIME

SUPPLIER

SUPPKEY NAME ADDRESS NATIONKEY PHONE ACCTBAL COMMENT

SYS_TIME



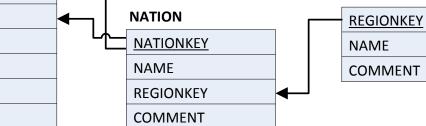
CUSTOMER

CUSTKEY NAME ADDRESS NATIONKEY PHONE ACCTBAL

MKTSEGMENT COMMENT

VISIBLE_TIME

SYS_TIME



LINEITEM ORDERKEY PARTKEY SUPPKEY SUPPKEY LINENUMBER QUANTITY EXTENDEDPRICE DISCOUNT TAX RETURNFLAG LINESTATUS SHIPDATE COMMITDATE RECEIPTDATE

SHIPINSTRUCT

SHIPMODE

COMMENT

SYS TIME

REGION

ACTIVE_TIME

ORDERS

 ORDERKEY

 CUSTKEY

 ORDERSTATUS

 TOTALPRICE

 ORDERDATE

 ORDERPRIORITY

 CLERK

 SHIPPRIORITY

 COMMENT

 ACTIVE_TIME

 RECEIVABLE_TIME

 SYS_TIME

August 26th, 2013





Data Generator

- Use TPC-H dbgen for version 0
- Generate history by adding 9 realistic update scenarios:

Scenario	Updated Tables	Probability
New Order	Customer, Orders, Lineitem	0.3
Cancel Order	Orders, Lineitem	0.05
Deliver Order	Customer, Orders, Lineitem, Partsupp	0.25
Receive Payment	Orders, Customer	0.20
Update Stock	Partsupp	0.05
Delay Availability	Part	0.05
Change Supplier Price	Partsupp	0.05
Update Supplier	Supplier	0.049
Manipulate Order	Orders, Lineitems	0.01





Queries

- The queries cover operations such as time travel, key in time, temporal joins and temporal aggregation
- Investigate many patterns of storage access and time- vs. key-oriented access with varying ranges and selectivity
- Classes of TPC-BiH benchmark queries:
 - 1) Pure-Timeslice Queries (Time Travel)
 - 2) Pure-Key Queries (Audit)
 - 3) Range-Timeslice Queries
 - 4) Bitemporal Queries





1) Pure-Timeslice Queries (Time Travel) Motivation

- Testing "slices" of time, i.e., state of table at a previous time
- Each time dimension can be treated as a point or slice
 - Point: looking at a single time
 - Complete slice: full evolution of a tuple through a time dimension
- Evaluate on application time, system time or both

Example: "Value of all unshipped orders at a given time"

```
SELECT SUM(o_totalprice) AS revenue
FROM ORDERS
FOR SYSTEM_TIME AS OF TIMESTAMP '[TIME1]'
FOR BUSINESS_TIME AS OF '[TIME2]'
WHERE o orderstatus = '0'
```





2) Pure-Key Queries (Audit)

Motivation

- History of a single or small set of tuples
- Investigate how tuples evolve over time
- Evolution along the system time, application time(s) or both

Example: "What is the history of a given order?"

```
SELECT orderstatus, totalprice, orderpriority,
    shippriority, sys_time_start
FROM orders
FOR SYSTEM_TIME FROM '0001-01-01' TO '9999-12-30'
WHERE o_orderkey = [ORDER_KEY]
ORDER BY sys_time_start
```





3) Range-Timeslice Queries

Motivation

- Permit any combination of constraints
- General access pattern: both value and temporal aspects

Example: "When did we have the largest number of unshipped items?"

```
CREATE VIEW unshipped AS
  SELECT COUNT(*) AS total, li.VERSION() as version
  FROM LINEITEM li
  WHERE li.L_LINESTATUS = '0'
  GROUP BY li.VERSION();

SELECT version, total FROM unshipped
WHERE total = (SELECT MAX(total) FROM unshipped)
```





4) Bitemporal Queries

Motivation

- Stress several time dimensions at the same time
- Vary the usage of each time dimension
 - a) point in time: current vs. past
 - b) sequenced/time range
 - c) non-sequenced/agnostic of time
- Complementary query variants to cover all combinations





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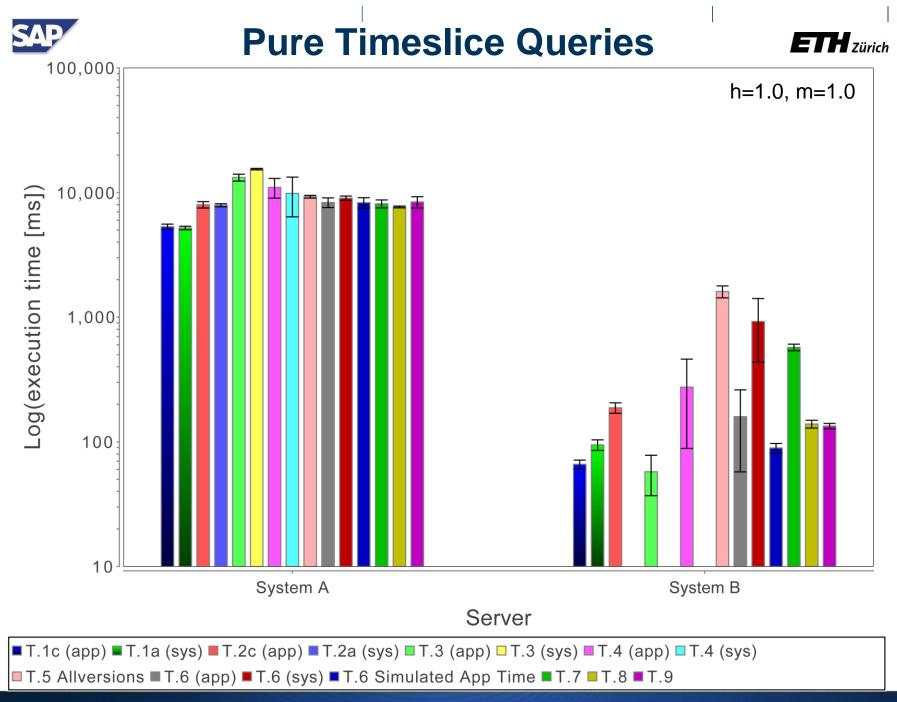




Evaluation Setup

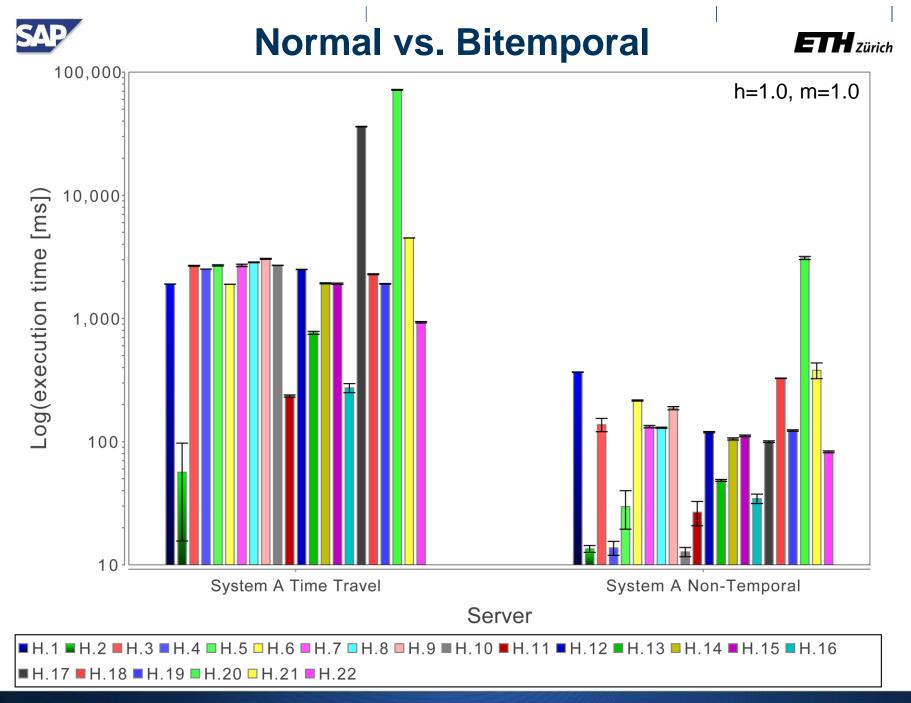
- Out-of-the-box settings (tuning is ongoing work)
- Two DBMS tested
 - System A: RDBMS supporting bitemporal workloads
 - System B: Main memory DBMS supporting system time
- Run on single system
 - 2 Intel Xeon X5675 processors with 6 cores at 3.06 GHz
 - 192GB of DDR3-1066MHz RAM
- Data
 - Initial data scaling factor (as in TPC-H): h=1.0
 - History size m=1.0 (1 million update scenarios)
 - Update scenario: Transaction (no bulk loading, but generate data!)
- At this stage purely performance results, no deep analysis

Note: logarithmic y-axis for all plots!



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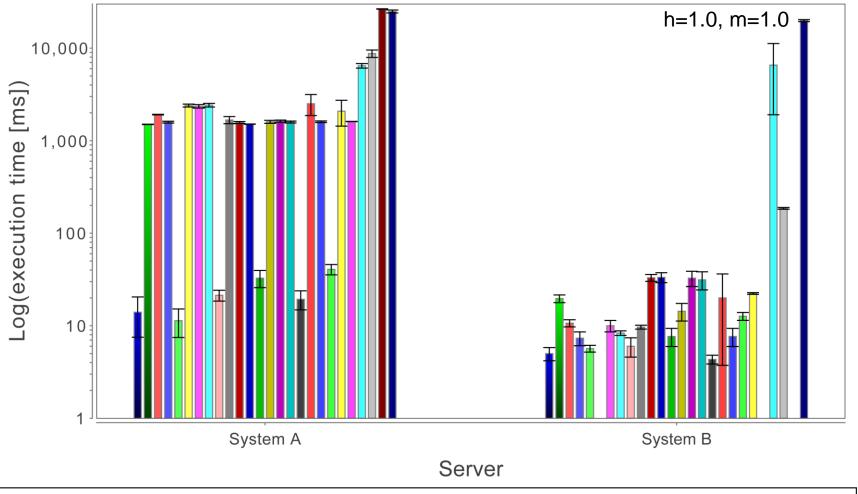
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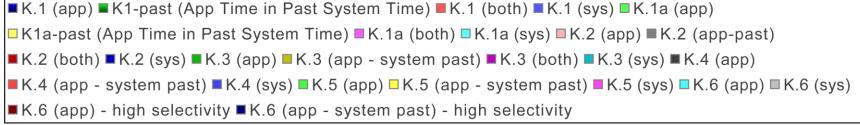


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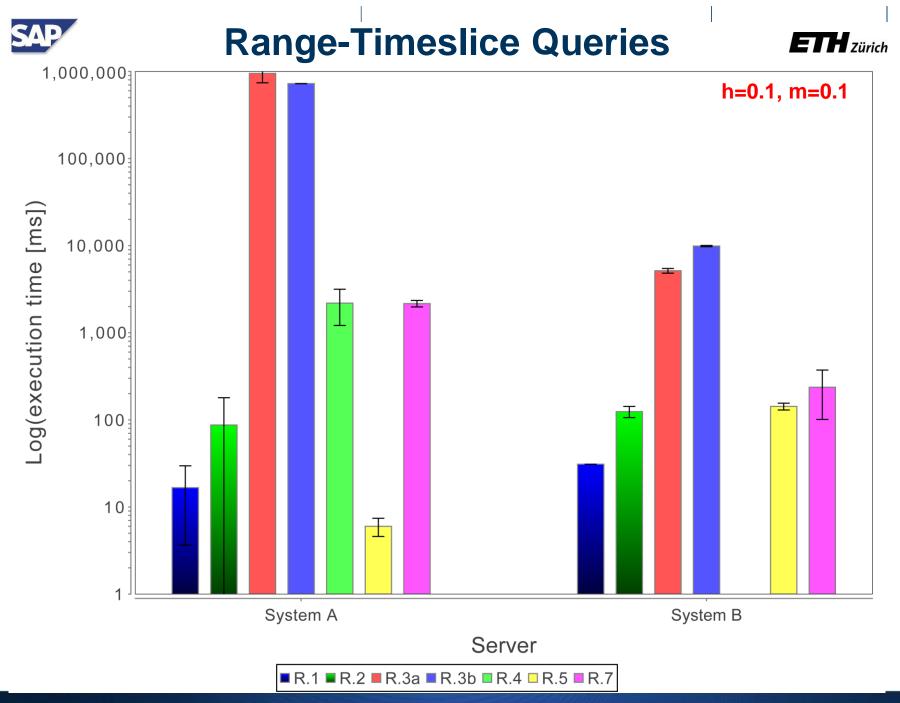
Simple Aggregation over Time





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Ongoing and Future Work

- Better coverage of DBMSs
- (Basic) tuning for temporal workloads, deeper analysis of query plans
- Variance in workloads
- Impact of scaling and value distribution
- Evaluation of update performance





TPC-BiH: Conclusion

Properties

- Builds on existing benchmarks
- Comprehensive coverage of use cases for temporal data
- Data generator based on real-world scenarios

Prelimilary Results for Commercial Systems

- Significant potential for optimization
- Not all common application use cases supported sufficiently

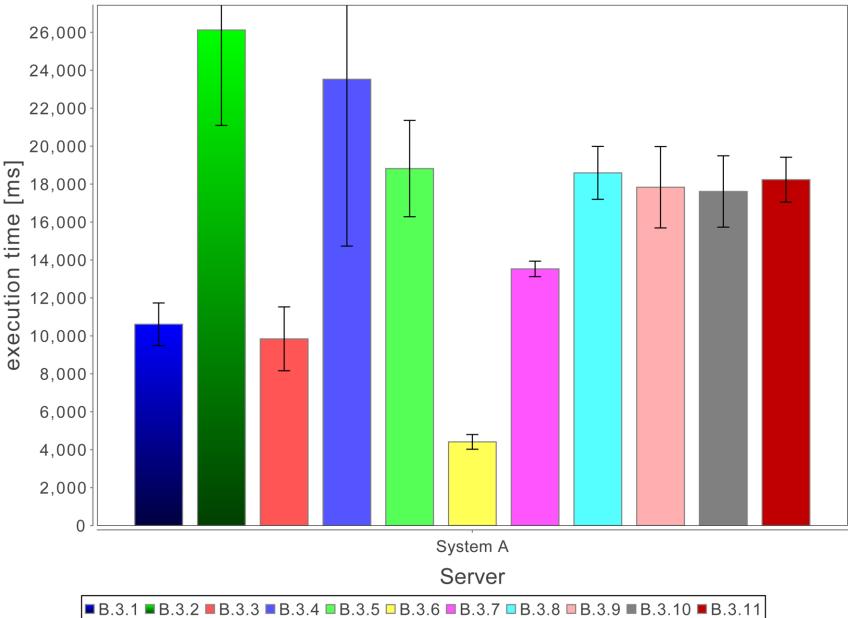




Backup Slides



Covering Bitemporal Dimensions



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