Issues in combined static and dynamic data management

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Overview

- Introduction
- Some existing approaches
 - for dynamic data management
 - for static data management
- Federated Architecture
- Issues
- Outlook



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Example Application Domain: Safe Offshore Operations

 Goal: Improve the safety of offshore operation for people, machines, and processes (4 year project, more @ http://soop.offis.de)





Static and Dynamic Data Management ...



Data Base Management Systems

- "store, then process"
- analytical queries
- one-time queries
- well-known technology
- user/application is active, data management passive



Data Stream Management Systems

- "on-the-fly" data processing
- real-time reaction on changes
- store only what needs to be kept
- register long-running queries
- user/application is passive, data management pushes data
- Complex event processing for temporal patterns

Typical queries in database management systems



Typical queries in data stream processing systems universität oldenburg





see Friday 27, Session 4, 11:30 - 13:00 Trust, Privacy and Security. NexusDSS: A System for Security Compliant Processing of Data Streams



DSM – Challenges

- Data stream
 - Continuously arriving data
 - Potential infinite
- Challenges
 - Blocking behaviour of query operators not adequate for stream processing
 - E.g., how to compute the average?
 - \rightarrow unblocking by windows
 - Temporal relationship of streaming data

Window definitions

- Monotone window operator to split stream into segment
- Window size can be based on:
 - number of elements (e.g., last 100 elements)
 - time (e.g., last 5 seconds)
 - predicates of elements (e.g., value between two threashholds)
- Window stride: how it moves
 - jumping or tumbling (no overlaps, data processed once)
 - sliding (continously, overlaps, data processed more than once)
 - sampling (no overlaps but gaps, some data is not processed
- Window realization (next slide)
 - interval appraoch
 - positive/negative approach



Features of Data Stream Management Systems

- Programming Abstraction
 - declarative: query
 - functional: flow graph
 - \rightarrow enables optimizations
 - \rightarrow better maintanance of systems
 - → using a DSMS on data streams is like using a DBMS instead of files
- Data flow vs. event bus (as in many CEP engines)
 - execution of data flow processes only data items that are needed by a query
- Parallel execution of operators in graph
 → no shared memory
- Data streams can be unbounded:
 - issues with sorting, joins, aggregation
 - \rightarrow approximate answers
 - \rightarrow window semantics



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However ...



Data Base Management Systems

- ACID properties, persistant storage
- well-established semantics (rel. algebra)
- well-established optimizers
- often deeply integrated into enterprise IT infrastructure
- can also cope with dynamic management (triggers, repeated optimized queries)



Data Stream Management Systems

- main memory processing
- often non-deterministic results (depend on timing of streams)
- no standard query semantics (often relational algebra with extras, or many user-defined operators)
- new system in IT infrastructure
- has to be combined with static data management



Can't we do that with a DBMS?

- Active Data Management in DBMS (Triggers, CQN)
- Magic Updates: The Bonn Approach

Dynamic Data in DBMS



- Database triggers
 - Are designed to intercept database change requests and transaction requests
 - Are part of a transaction synchronous
 - Allow to check validity of changes
 - Widely used in applications
 - But: since triggers can trigger triggers, complex logic gets hard to maintain
- Database Log Miner (e.g., Oracle)
 - The log miner is designed to select specific journal entries in batch or real time mode
 - Works asynchronously
 - Shows 'dirty' data
 - Works closely with dissemination services such as Oracle Streams and Oracle GoldenGate
 - Used by many infrastructure services such as replication, logical standby, real time warehousing, security auditing, real time analytics through TimesTen (In Memory Database), CEP

Continuous Query Notification - CQN



- CQN is designed to notify users/applications if the result set of a query has changed
 - Requests are defined as registered queries normal query plus a notification destination
 - Works asynchronously
 - Shows only 'visible' data
- The following events generate notifications:
 - Committed DML Transactions
 - Committed DDL Statements
 - Deregistration
 - Global Events

Continuous queries 'made' in Bonn:

- We're experts for update propagation techniques.
- A <u>considerable</u> amount of stream applications can be realized using conventional DBS.
- Continuous queries are implemented using materialized views.
- Incremental techniques are used for their maintenance.
- In order to process
 <u>static</u> domain knowledge with
 <u>dynamic</u> stream data,
 we employ Magic Updates!





Data Stream research today:

- It is widely believed that conventional DBS are <u>not</u> suited for processing continuous queries.
- Instead, main memory-based DSMS are employed:
 - no transactions (no recovery, no rollback,...)
 - Limited multi-user access
 - no history recorded
 - etc.
- \Rightarrow Basically, only SQL survived
- DSMS can process fast stream data but static domain knowledge only in a limited way.



Case study: View-based analysis of aircraft movements using the Airspace Monitoring System (AIMS)



- Which aircrafts approach each other critically?
- Which aircrafts are currently landing?
- Which planes are late?
- Which flight are entering areas with bad weather conditions?
- Which planes are changing their course unexpectedly?
- What is the average number of flights for an arbitrarily chosen area?

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Architecture of AIMS



dynamic track data

Flights

- · Call sign of a flight
- Aircraft type
- Registration
- Transponder code
- · Position by latitude, longitude, altitude
- Vertical speed
- · Horizontal speed
- · Heading of the aircraft in degrees
- · Time stamp
- ...

static domain data

Airports

- · Name of airport, city, country
- IATA / ICAO-Code
- Position by latitude, longitude, altitude
- Region code like "E" for Europe
- ...

Flight plan

- · Waypoints: beacon-code, position, speed, time
- Plan ID, source, country code
- Callsign, aircraft type, destination airport
- ...





- its vertical velocity is negative
- its flight level is below zero
- it is currently approaching an airport less than 20 miles away.





Aircraft positions frequently change:

CallSign	Туре	Longitude	Latitude	 Date	1
ACA876	B763	-4,27333	52,50833	 23-03-2007 14:17:03	
ACA876	B763	-4,82512	51,33	 23-03-2007 14:17:07	
ACA876	B763	-5,92113	51,1211	 23-03-2007 14:17:11	
-	• •		• • •		

Time stamped data stream

CQ: Which aircrafts are currently landing?





landed

to land

CREATE	AC AC MATERIAL	 2A873 14 2A876 14 2A876 14 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2	 17:03 17:05 EW la .Time	ndeAnflu	+ g As	INSI SE: FI WI	S Tim V069 14:17 ERT INTO LECT nf.C ROM new HERE nf.I 	new landeAnfl allSign, flight_da ncline <	deriv ug nf.T: ta A: 0 ANI	ation ime S nf D
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SELE FROM WHER	flight 2 5 f.Incl:	AS f ine < 0 i	AND							
SELE FROM WHER	flight 2 f.Incl: Longitude	AS f ine < 0 i		Time		CS	Longitude	Latitude		Tim
SELEG FROM WHER CS ACA873	Longitude -4.556667	Latitude 52,578333	AND	Time 14:17:03		CS BAW06	Longitude 9 1,521667	Latitude		Tim 14:17
SELEG FROM WHER CS ACA873 ACA876	Longitude -4,273333	Latitude 52,578333 52,508333	AND 	Time 14:17:03 14:17:05		CS BAW06	Longitude 9 1,521667	Latitude 51,35254		Tim 14:17



Performance achieved so far with AIMS:

- AIMS typically processes 2000 tupels (20 attributes) every 3-4 seconds
- Determination of landing/starting planes <1 second
- Determination of critical approximations <1 second

Fertig

• Counting of flights within a user-defined region <2 seconds





Update propagation is suited for analyzing a wide spectrum of stream applications.

Air Traffic Control



In cooperation with

Stochastic Analysis of Radar Data



In cooperation with



Analyzing Stock Market Data



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DBMS approaches assessment



- Consider updates and the change of queries
- Are good up to a couple thousand of tuples/second
- However, there are application that need much higher throughput
 - real-time multi-sensor fusion
 - dynamic context models

Sensor Fusion Example (project SaLsA)

- Context-Model Generation for Safe Autonomous Transport Vehicles
 - Autonomic vehicles perform mobility operations with walking speed to ensure safety
 - Faster mobility requires specific knowledge of the environment
 - Realization via a dynamic context model based on sensor data with uncertainties
 - Using Data Stream
 Management
 Technology for Sensor
 Fusion

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afe Project





http://www.offis.de



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DBMS approaches assessment



- Consider updates and the change of queries
- Are good up to a couple thousand of tuples/second
- However, there are application that need much higher throughput
 - real-time multi-sensor fusion (54000 points/second/ sensor)
 - dynamic context models
 - distributed data processing
 - specialized operators
 - not all data needs to be stored

\rightarrow a federated approach might be a good idea!

TO BOLDEY GO FOR A FEDERATION



Query Semantics

- How do we formulate federated queries?
- DBMS:
 - SQL (but no standardized window semantics)
 - ECA rules, stored procedures, CQN
- DSMS:
 - CQL (Arasu, Babu, and Widom 2006)
 - StreamSQL (StreamBase Systems 2012)
 - Sase (Gyllstrom et al. 2007)
 - Aurora's boxes and arrows (Abadi et al. 2003)
 - IBM Infosphere Stream's Stream Programming Language SPL (Biem et al. 2010)

→ for federated queries, we need a common query language and semantics!



Optimized Data Flow

- How do we decide about the data flow?
- Data from active resources is often produced decentrilized
- In classical federation approaches, data is sent to the federation software
 - might not be the best idea here
- Decide:
 - where to process which operator?
 - send data from DBMS to application or from DSMS?

... data flow must be part of the processing optimization.



Optimized Cont. Query Distribution

- How do we decide about the optimal distribution of the continuous queries?
- Depends on
 - data location
 - update rate of data
 - required report rate of query result
 - required persistance of data
 - if system crashes and recovers, is the application interested in old (recovered) results?

Options range from CQN/triggers to pure DSMS queries to mixed executions





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Adaptation and Migration

- If the parameters used for the query distribution change, how do we adapt the continous query to the new situation?
 - monitoring of
 - data location
 - update rate of data
 - required report rate of query result
 - required persistance of data
- If we decide to adapt how do we migrate from old state to new state?
 - have to take care of stateful operators

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	ls	sues
	sem	antics?
	data	a flow?
	optim	nization?
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Adaptation and Migration

- Some works has been done for migrating pure DSMS queries:
 - F: Fresh start: start a new continuous query
 - I: Immediate stop: the query stops processing immediately
 - D: Drain stop: the query will be graduately stopped, finishing partially results
- Resulting stateflow diagram for continuous queries:



Kyumars Sheykh Esmaili, Tahmineh Sanamrad, Peter M. Fischer, Nesime Tatbul: Changing flights in mid-air: a model for safely modifying continuous queries. SIGMOD 2011:613-624



Data and Query Provenance

- How can we know why a notification was produced (or not?)
- Many applications would code complex decision trees or situation recognition in federated DSMS/DBMS query plans

Data provenance:

- for a given federated result, which data tuples contributed to it?
- Query provenance:
 - when was / is which query plan active?
 - can be used to re-construct data provenance



Data and Query Provenance

- Use Cases
 - Ad-hoc human inspection
 - Continuous query debugging
 - Indicator-based assurance
 - Event warehousing
- Challenges
 - Infinity, Performance, and Aggregation
 - Non-determinism (of some DSMS)
 - Order (of DSMS processing)

Some work for DSMS:

Boris Glavic, Kyumars Sheykh Esmaili, Peter M. Fischer, Nesime Tatbul: The Case for Fine-Grained Stream Provenance. BTW Workshops 2011:58-61

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	Issues
	semantics?
	data flow?
	optimization?
	adaptation?
	provenance?
	persistance?
	interaction?
	distribution?

Persistance and Intelligent Archiving

- How can we realize persistance in a world of high volume data streams?
 - not everything has to be persistant
 - but how to decide?
- 1. manually
 - by registering archiving queries
- 2. randomly
 - by sampling parts of the stream
- 3. automatically intelligent = model-based & adaptive
 - by learning patterns / models in the stream, e.g., by stream data mining
 - archive only new patterns in hig h resolution
 - count the other (= store aggregated information)





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DSMS and DBMS interaction

- How do the DSMS and the DBMS interact?
 - who controls whom?
 - how tightly are they coupled?
- 1. The federation controls both system
 - low coupling between the systems
 - only running queries are affected if a system changes (interface, schema, ...)
- 2. The DSMS controls the DBMS
 - it issues queries to the DBMS to access static data
 - this function is integrated in most DSMS
- 3. The DBMS controls the DSMS
 - would require concepts to include DSMS queries in, e.g., trigger definitions



Distributed Continuous Query Processing

- How to optimize the distribution of the query processing?
- Many active data sources are physically distributed or even mobile (sensor networks, mobile sensors in smart phones or trucks)
- Many sensing devices have (some) processing power
- Should be considered in query optimization
- e.g.: project "Resource constrained Distributed Stream Processing "@Universität Erlangen, Germany



Michael Daum, Frank Lauterwald, Philipp Baumgärtel, Niko Pollner, Klaus Meyer-Wegener: Efficient and cost-aware operator placement in heterogeneous streamprocessing environments. DEBS 2011:393-394



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Conclusion

- To combine static and dynamic data management, many issues have to be addressed
 - DBMS has to be extended for temporal semantics
 - DSMS have to mature in processing semantics, determinism and persistance concepts
 - Together, a data management biotop has to have a
- In the long run, DSMS functionalities will be part of a DBMS system
- Outlook: consideration of NoSQL-concepts





Thank's for the fish!





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