

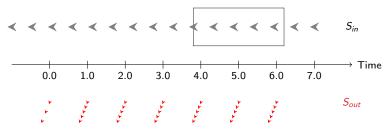
Stream Processing with Local Temporal Reasoning

Özgür L. Özçep

Workshop Stream Reasoning, Vienna November 9, 2015

Local Reasoning on Streams

Taming the Potential Infinity of Streams

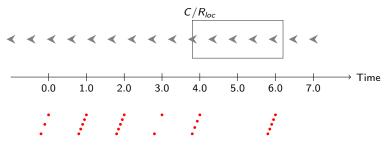


Window operator as a means to cope with potential infinity

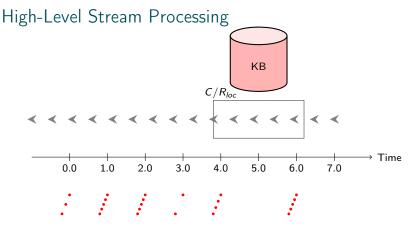
Grab finite portion of stream and do something on it

Local Reasoning Service

. . .



- Local calculation/reasoning C/R_{loc}
 - arithmetics, timeseries-analysis operations
 - Entailment, satisfiability, query answering, abduction, revision,



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- arithmetics, timeseries-analysis operations
- Entailment, satisfiability, query answering, abduction, revision,
- Background knowledge KB: static data, historical data, learned data

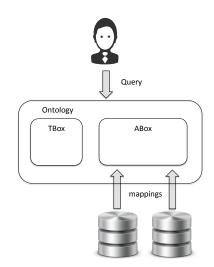
OBDA within **OPTIQUE**

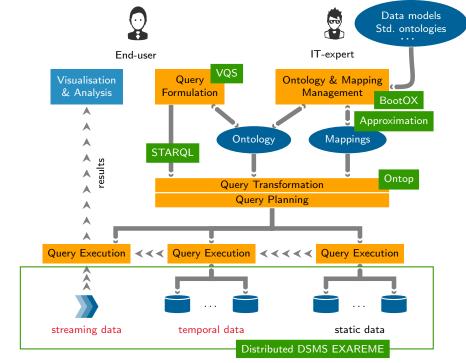
OPTIQUE

- EU 7th framework program (http://www.optique-project.eu/)
- Two big data use cases from industrial partners
 - STATOIL SAS: Querying data on wellbore related DBs
 - SIEMENS: Querying sensor and event data from (gas) turbines
- Cycle of constructing query, issueing it, and getting answers is bottleneck in both use cases
- Optique platform: OBDA with user support + optimizations on different levels + Streaming
- ► Lübeck (R. Möller, C. Neuenstadt, Ö.Ö.) responsible for stream-temporal OBDA module ⇒ STARQL

Ontology-Based Data Access

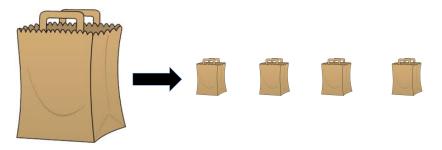
- Use ontologies as interface ...
- ▶ to access (here: query)
- data stored in some format ...
- using mappings
- Classical OBDA
 - Relational data
 - ABox is virtual
 - Query answering by rewriting/unfolding ("Reasoning by rewriting")
 - Weak ontology language (no qualified existentials on left-hand of inclusions)





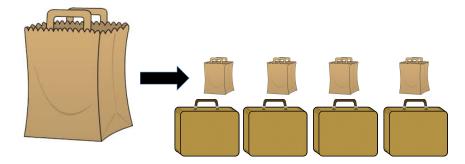
Reasoning within STARQL

Window Semantics in STARQL



- Group elements according to specified criterion (including timestamps) into mini-bags
- ► Technically: Result is a sequence of ABoxes/RDF graphs

Incorporating the Background Knowledge



Sequencing in STARQL

Information Need

Output every 1 minute those temperature sensors having value above 90 over the last minute

```
CREATE STREAM S_out AS

CONSTRUCT { ?sens rdf:type :tooHigh }<NOW>

FROM S_in [ NOW , NOW - 1 minute ]-> 1 minute,

ABOX, TBOX

WHERE { ?sens rdf:type TempSens }

SEQUENCE BY StdSeq AS seq

HAVING FORALL i IN seq FORALL ?x

IF { ?sens :hasVal ?x }<i> THEN ?x > 90
```

Why at all Bother with State Sequences?

- Building microcosm for LTL like temporal reasoning on states
- But note
 - Temporal logic frameworks presuppose state sequences
 - In contrast, sequence construction is part of STARQL query
- Use case may require different types of states
 - cluster states using machine learning techniques
 - states corresponding to consistent ABoxes

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Determining certain answers

- ► Has to incorporate TBox (e.g. *BTTempSens* \sqsubseteq *TempSens*)
- Handled by rewriting

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Local temporal reasoning on states

```
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Reasoning involved in **constructing the state sequence** (in particular for checking consistency of mini ABoxes)

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Theoretical Results

State elimination

- State abstraction means additional layer in OBDA stack
- Nonetheless, it can be eliminated (Ö., Möller, Neuenstadt 2014, 2015)
- Relation to LTL approaches
 - Backend systems mostly have domain independent languages
 - LTL like query languages not domain independent
 - ► TCQs: CQs combined with LTL (Borgwardt et al. 13)
 - A fragment of STARQL embeds a safe fragment of TCQs (Ö., Möller, Neuenstadt 2015)

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Practical Results

- Implemented STARQL sub-module with optimizations
- Transformation realized to backend EXAREME
- Optimizations for distributed stream processing in EXAREME
- Multiple Query/multiple stream handling
 - Monitor different components (turbines, sensors)
 - Monitor different hand-crafted well-proven patterns
- Specific statistical and time-series operators
 - Pearson-correlation (e.g. for detecting out faulty sensors)
 - Calls for specific optimizations (local-sensitive hashing)

Future Work

Stream Reasoning for NLP

- Intention: Use stream semantics and techniques for natural language processing (NLP)
- One of the (very few) application scenarios where stream-processing historical data makes sense
 - You could read a text in "parallel" but here, "order really matters":
 - Meaning of sentence depends on meanings of preceding sentences
- Discoure representation theory (DRT): Capture super-sentence meaning by discourse structures
- Calls for state-based stream processing with a scopus storing discourse structures

Challenges

- Need for state-based stream processing with a scopus storing discourse structures
 - Discourse structure dynamically updated
 - In general may grow arbitrarily
- Need for abduction style reasoning (Sherlock Holmes style reasoning)
 - From observations to possible explanations
 - Have to constrain search space, anytime abduction
- Different orders to incorporate
 - sentence (arrival) ordering (so)
 - causal ordering (co)
 - temporal ordering(s) (to)

Example

- Bob cried. Alice consoled him.
- Bob cried. Alice insulted him.

(so corresponds to to) (so corresponds to co)

Thank you for your attention!