



# RDFox: A Highly-Scalable RDF Store

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# Motivation

- ▶ Semantic Web applications commonly represent data using the Resource Description Framework (RDF)
- ▶ Such applications rely on *RDF stores* to:
  - efficiently handle large amounts of RDF data
  - handle data from heterogeneous sources
  - manage and reason with background knowledge
  - answer queries w.r.t. *both* data and knowledge



# Motivation

Hardware trends open new possibilities for highly-scalable main-memory RDF stores

- ▶ Main memory

- personal computers: 4GB-16GB
- small servers: hundreds of gigabytes
- higher-end servers: terabytes

- ▶ Processing units

- personal computers: 2-4 cores
- small servers: tens of cores
- higher-end servers: thousands of cores



# RDFox Overview

- ▶ A cross-platform, centralised, main-memory RDF store
- ▶ Supports parallel import and efficient storage of large amounts of RDF data
- ▶ Supports highly-scalable materialisation-based reasoning;
  - incremental reasoning;
  - native handling of owl:sameAs
- ▶ Provides efficient query answering (tree decomposition)
- ▶ Provides versatile modes of use (C, Java and Python APIs, SPARQL endpoint)
- ▶ Developed at the University of Oxford and available under an academic licence at <http://rdfox.org>



# Storage Scheme

- ▶ RDFox uses a flexible and highly-efficient storage scheme that supports highly-scalable parallel updates
  - it stores RDF data in a triple table
  - it has a configurable indexing scheme for efficient pattern matching (default [s,p,o], [s,p], [p], [o,p])
  - it supports ‘mostly’ lock-free updates
  - it is crucial for scalable parallel importation and parallel reasoning



# Datalog Reasoning

- ▶ RDFox supports reasoning with datalog ontologies
- ▶ RDFox incorporates a materialisation-based datalog engine that implements state-of-the-art reasoning algorithms
- ▶ Datalog is a rule-based language that can capture OWL 2 RL and SWRL rules
  - each OWL 2 RL ontology can be encoded as datalog rules
  - one can also use the fixed datalog program that corresponds to the rules in the OWL 2 RL specification
  - SWRL rules are a syntactic variant of datalog rules



# Datalog Materialisation

- ▶ Precompute the consequences of the input RDF data and datalog rules
- ▶ Queries are answered over the materialised consequences
  - no need for reasoning during query evaluation
- ▶ RDFox uses a shared-memory parallel algorithm **[Motik et al., AAI 2014]**
  - a 'triple-at-a-time' variant of the semi-naive algorithm
  - no repetition of work (every rule instance is considered at most once)
  - even partitioning of reasoning into small subtasks (one per triple)
  - allows high scalability
  - no need of explicit load balancing



# Incremental Reasoning

- ▶ Semantic Web data changes continuously
- ▶ RDFox supports incremental reasoning for efficient updating of datalog materialisations **[Motik et al., AAI 2015]**
  - uses the **FBF** algorithm, which efficiently identifies the triples that need to be deleted and added
  - requires no extra information collected during the initial materialisation (no counts, no dependencies, no proofs)
  - improves the well-known DRed algorithm known from the database community by performing **exact** deletions



# Native Equality Reasoning

- ▶ owl:SameAs is used to assert equalities between resources
- ▶ Equality reasoning can affect performance: increased memory consumption and reasoning times
- ▶ RDFox uses rewriting **[Motik et al., AAI 2015]**
  - assigns a common representative to equal individuals
  - handles correctly rewritten constants in rules
- ▶ RDFox implements the **first** incremental algorithm for equality reasoning with rewriting **[Motik et al., IJCAI 2015]**



# Evaluation

- ▶ RDFox on mid-range servers (previous work)
  - stores 1.5 G triples in 52 GB of RAM
  - 14x speedup on using 16 physical cores
  - efficient incremental reasoning for small and medium sized updates (with and without native equality reasoning)
- ▶ We evaluate RDFox on an **Oracle SPARC T5-8**
  - 4 TB of RAM
  - 8 SPARC V9 processors at 3.6 GHz
  - 128 physical threads and 1024 virtual threads



# Parallelisation Scalability

|          | LUBM-50k        |            | Claros          |            | DBpedia         |            |
|----------|-----------------|------------|-----------------|------------|-----------------|------------|
|          | Time(s)         | Speedup    | Time(s)         | Speedup    | Time(s)         | Speedup    |
| 1        | 27000           |            | 10,000          |            | 31,000          |            |
| 64       | 727             | <b>37x</b> | 375             | <b>27x</b> | 1,200           | <b>26x</b> |
| 128      | 387             | <b>70x</b> | 226             | <b>44x</b> | 698             | <b>45x</b> |
| 256      | —               | —          | 226             | <b>44x</b> | 684             | <b>46x</b> |
| 512      | —               | —          | 154             | <b>65x</b> | 432             | <b>72x</b> |
| 1024     | —               | —          | 125             | <b>80x</b> | 359             | <b>87x</b> |
| Max Rate | <b>6.1M t/s</b> |            | <b>4.2M t/s</b> |            | <b>4.0M t/s</b> |            |
| Triples  | 6.7G —> 9.3G    |            | 19M —> 539M     |            | 113M —> 1.5G    |            |



# Parallelisation Scalability

|          | LUBM-9k        |             | Claros          |            | DBpedia         |            |
|----------|----------------|-------------|-----------------|------------|-----------------|------------|
|          | Time(s)        | Speedup     | Time(s)         | Speedup    | Time(s)         | Speedup    |
| 1        | 1600           |             | 10,000          |            | 31,000          |            |
| 64       | 50             | 32x         | 375             | 27x        | 1,200           | 26x        |
| 128      | 28             | 58x         | 226             | 44x        | 698             | 45x        |
| 256      | 17             | 97x         | 226             | 44x        | 684             | 46x        |
| 512      | 8              | 190x        | 154             | 65x        | 432             | 72x        |
| 1024     | 8              | <b>213x</b> | 125             | <b>80x</b> | 359             | <b>87x</b> |
| Max Rate | <b>60M t/s</b> |             | <b>4.2M t/s</b> |            | <b>4.0M t/s</b> |            |
| Triples  | 6.7G —> 9.3G   |             | 19M —> 539M     |            | 113M —> 1.5G    |            |



# Data Scalability on LUBM: RDFox with 1024 threads

|           | Input | Mat.         | Time (s)   |
|-----------|-------|--------------|------------|
| LUBM 20k  | 3.1G  | 4.2G         | 42         |
| LUBM 40k  | 5.5G  | 7.5G         | 85         |
| LUBM 60k  | 8.0G  | 10.9G        | 118        |
| LUBM 80k  | 11.0G | 15.0G        | 179        |
| LUBM 100k | 13.5G | 18.4G        | 228        |
| LUBM 120k | 15.9G | <b>21.7G</b> | <b>251</b> |



# Conclusion

RDFox provides a unique combination of versatility, rich functionality, high performance and scalability:

- ▶ a highly efficient and flexible storage scheme
- ▶ state of the art datalog reasoning algorithms
- ▶ versatile modes of access

Suitable for data-intensive applications requiring expressive and highly scalable reasoning

- ▶ storage of up to **21G** triples
- ▶ reasoning speeds of up to **60M t/s**
- ▶ speedups of up to **213** times using 1024 threads



# Outlook

- ▶ Improving our query answering algorithms
- ▶ Extending our support to full SPARQL 1.1
- ▶ Adding support for named graphs
- ▶ Reasoning with aggregation and non-monotonic negation
- ▶ Distributing storage, querying and reasoning in RDFox



# Thank you!

## References

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# RDFox: Use Cases

- ▶ **Kaiser Permanente**—a US health care consortium
  - uses RDFox to analyse patient data records
- ▶ **E´lectricite´ de France (EDF)**—a French electric utility company
  - uses RDFox to manage and analyse information about their electricity distribution network
- ▶ **Statoil ASA**—a Norwegian multinational oil and gas company
  - uses RDFox as part of a large-scale Semantic Web application that facilitates the integration and analysis of oil production and geological survey data



# Evaluation: Memory

|                       | LUBM-50k |            | Claros  |           | DBpedia |           |
|-----------------------|----------|------------|---------|-----------|---------|-----------|
|                       | Triples  | B/t        | Triples | B/t       | Triples | B/t       |
| After Import          | 6.7G     | <b>124</b> | 19M     | <b>80</b> | 113M    | <b>58</b> |
| After Materialisation | 9.3G     | <b>101</b> | 539M    | <b>37</b> | 1.5G    | <b>39</b> |



# RDFox Architecture

