



# 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., AAAI 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., AAAI 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., AAAI 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>
<b>Max Rate</b>	<b>6.1M t/s</b>		<b>4.2M t/s</b>		<b>4.0M t/s</b>	
<b>Triples</b>	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>
<b>Max Rate</b>	<b>60M t/s</b>		<b>4.2M t/s</b>		<b>4.0M t/s</b>	
<b>Triples</b>	6.7G —> 9.3G		19M —> 539M		113M —> 1.5G	



# Data Scalability on LUBM: RDFox with 1024 threads

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

Motik, B., Nenov, Y., Piro, R., Horrocks, I., Olteanu, D.: Parallel materialisation of datalog programs in centralised, main-memory RDF systems. AAAI 2014

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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
<b>After Import</b>	6.7G	<b>124</b>	19M	<b>80</b>	113M	<b>58</b>
<b>After Materialisation</b>	9.3G	<b>101</b>	539M	<b>37</b>	1.5G	<b>39</b>



# RDFox Architecture

