

VIT Chennai, 29th September 2021

Invited Lecture

as part of the series 'Knowledge Representation
and Reasoning'

Auditory Introduction to Query Processing in Semantic Web Databases

Professor Dr. rer. nat. habil. Sven Groppe

<https://www.ifis.uni-luebeck.de/index.php?id=groppe>

Stations of my academic life



Trying to impress (not sure if it's working)...

- > 122 publications
 - 93 co-authors
 - 41 publications (33%) are co-authored by bachelor/master students (being students at time of writing)
- 2 supervised dissertations, 3 current PhD students
- $\approx$ 82 bachelor/master/diploma thesis/student projects
- $\approx$ 60 contributors to LUPOSDATE
- > 35 lectures, in the areas of
 - Semantic Web
 - Cloud- and Web-Technologies
 - Mobile and Distributed Databases
 - Databases
 - Algorithms and Datastructures
 - Compiler

Trying to impress (not sure if it's working)...

- 6 third-party fundings ($\approx$ € 1M)
- Scientific Services
 - Workshop Chairs
 - Semantic Big Data (SBD) @ SIGMOD (2016 - 2020)
 - Big Data in Emergent Distributed Environments (BiDEDE) @ SIGMOD (2021 - 2021)
 - Very Large Internet of Things (VLIoT) @ VLDB (2017 - 2021)
 - Web Data Processing and Reasoning (WDPAR) @ KI 2018
 - General Chair
 - International Semantic Intelligence Conference (ISIC) 2021 - 2022
 - many other scientific services
 - $\approx$ 90 PC memberships
 - reviewer of $\approx$ 30 journals
 - editor of 3 journals

Trying to impress (not sure if it's working)...

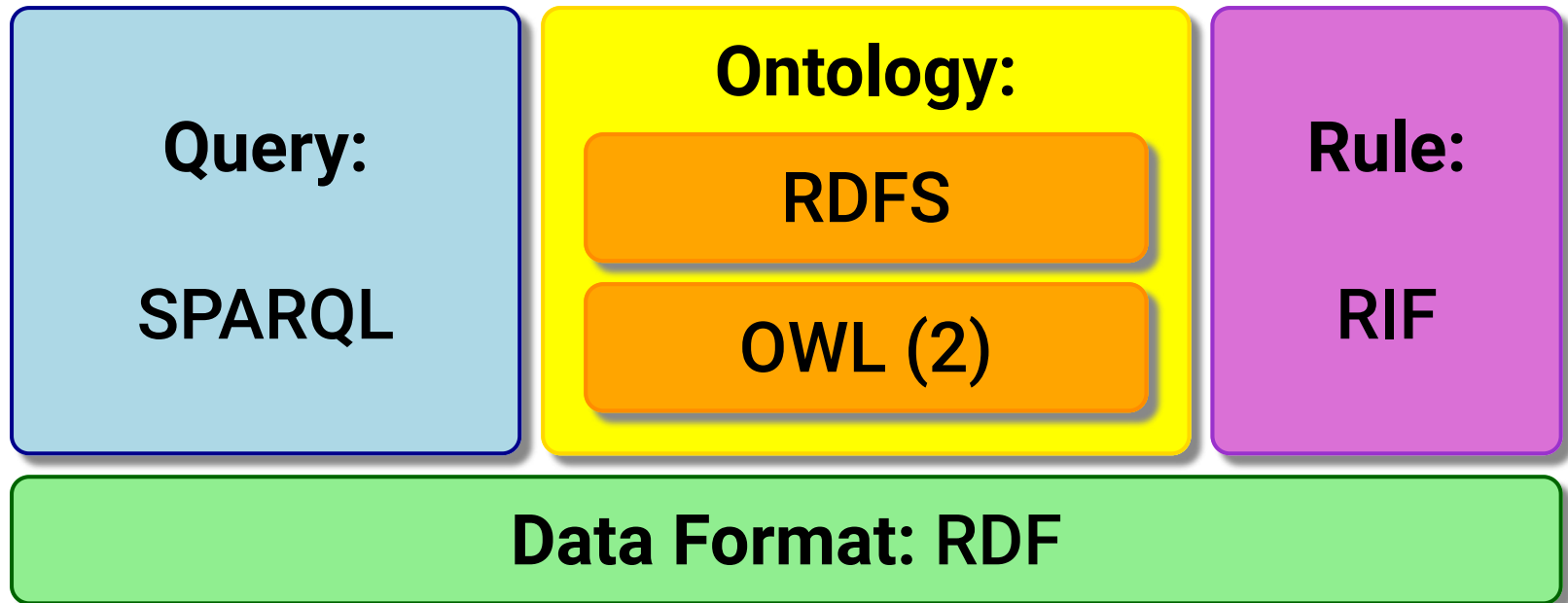
Detailed Information available at

<https://www.ifis.uni-luebeck.de/~groppe>

Agenda: Introduction to Query Processing in Semantic Web Databases

- SPARQL Query Language
 - Transformation to Relational Algebra Expression/Operatorgraph
- Sound of Databases
- RDF3X Indices
- Merge Join
- Sideways Information Passing (SIP) in
 - Merge Joins
 - RDF3X B⁺-Tree
- Hash Join
- Inspiring Sounds

Semantic Web (Core) "Standards"



- Every data model (here Semantic Web) has its own set of languages (data, query, rule, ...)

Recap: SPARQL Query and its Result

SPARQL Query:

SELECT ?person ?name

WHERE {

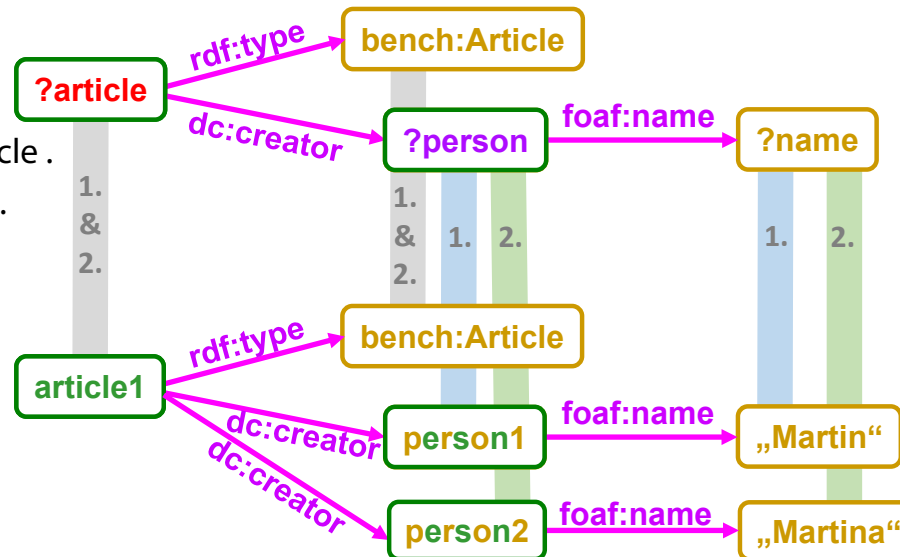
?article rdf:type bench:Article .

?article dc:creator ?person .

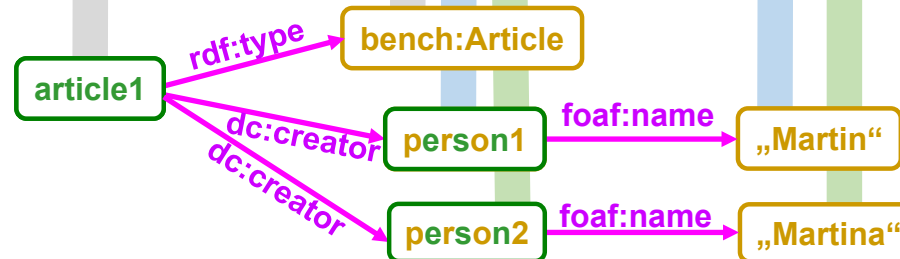
?person foaf:name ?name .

}

Query Graph:



RDF Graph:



Solutions:

1. ?article=article1
?person=person1
?name=„Martin“

2. ?article=article1
?person=person2
?name=„Martina“

Transformation to Relational Algebra Expression

SELECT ?person ?name

WHERE {

?article rdf:type bench:Article .

?article dc:creator ?person .

?person foaf:name ?name .

}

$\pi_{\{?person, ?name\}}$

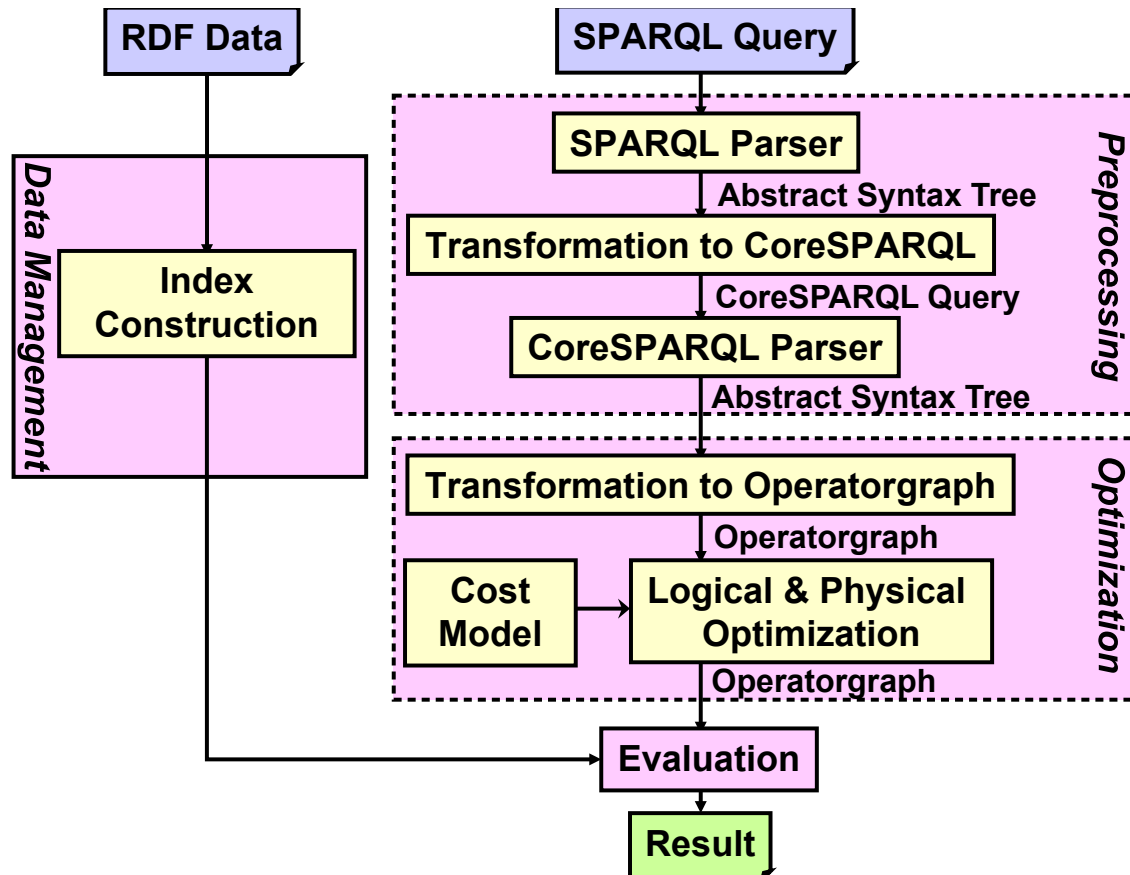


?article rdf:type bench:Article

?article dc:creator ?person

?person foaf:name ?name

Architecture of a Semantic Web Database Management System



Semantic Web DBMS LUPOSDATE

Support of:

- SPARQL Queries
- RIF Rules
- RDF Schema
- OWL (via OWL2RL in RIF)

Indexing:

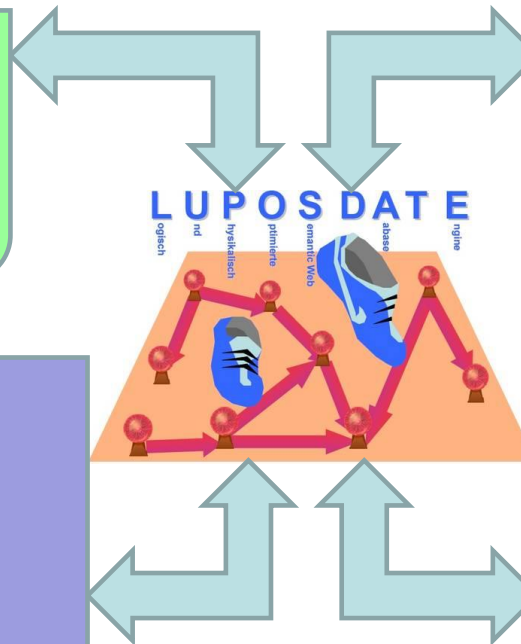
- Stream Processing
- Main memory for small datasets
- Disk-based for large datasets
 - RDF3X
- Cloud: HBase
- P2P

Visualizations:

- Visual Editor
 - Queries (SPARQL)
 - Rules (RIF)
 - Data (RDF) in
 - 2D and
 - 3D
 - Logical Optimization Rules
- Summaries of RDF Data
- Operator graph
- Processing of Queries and Rules
- Optimization Steps

Extra:

- Parallel Processing
- Distributed Processing
- Cloud Computing
- Mobile Computing
- P2P for Internet of Things
- Compression of RDF Data
- Embedding of SW Languages in Programming Languages
- Speeding up by FPGAs

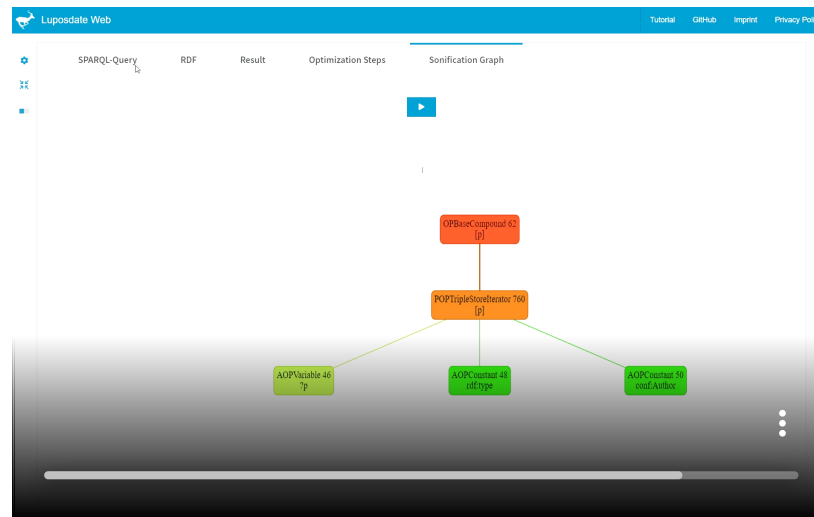


LUPOSDATE3000

- Successor of **LUPOSDATE**
- **Multiplatform** Database
 - programming language **Kotlin**
 - targets: JVM, JS, native binaries (win/linux/macOS/iOS/...)
- Focus on:
 - parallel databases
 - distributed databases for Internet-of-Things
 - Semantic Web: RDF, SPARQL
- Planned:
 - RDFS/OWL Inference
 - multi-model support



Sonification of Triple Pattern Evaluation



[Start Sonification](#)

- What do you **hear**?
 - An **ascending scale** (in pitch)
- **We learn:** Semantic Web data often **stored in indices**
 - Output is **pre-sorted**
 - **Maximize** usage of fast **merge joins**

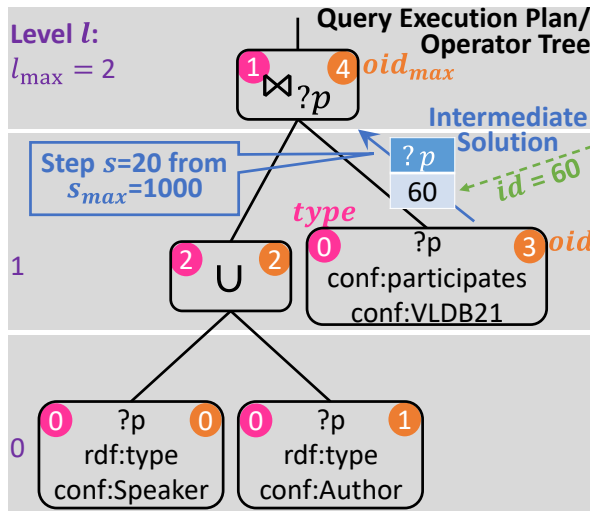


Sonification

- *use of nonspeech audio to convey information [...] for the purposes of facilitating communication or interpretation¹*
- Examples
 - [brain activity](#)
 - [from deep space to cancer research](#)
 - [Sound of Sorting](#)
 - ...
- Motivation
 - reduces barriers for people with poor education
 - presents multidimensional data in an easy-to-understand way
 - additional medium helps with learning new information
 - helps humans with visual impairments



Sound of Databases



Dictionary:

id	value
0	...
...	...
60	conf:Sven
...	...
$id_{\max}=100$	...

Op.Type Mapping:

type	value
Tuple pattern	0
⋈ (Join)	1
U (Union)	2
...	...
...	$type_{\max}=45$

Possible Mappings m_i of one processing step to $[0; 1]$ based on

a) value: $m_0 = \frac{id}{id_{\max}} = \frac{60}{100} = 0.6$

b) depth: $m_1 = \frac{l}{l_{\max}} = \frac{1}{2} = 0.5$

c) operator id: $m_2 = \frac{oid}{oid_{\max}} = \frac{3}{4} = 0.75$

d) operator type: $m_3 = \frac{type}{type_{\max}} = \frac{0}{45} = 0$

e) processing step: $m_4 = \frac{s}{s_{\max}} = \frac{20}{1000} = 0.02$

f) (weighted) combinations of above mappings:

$$m_5 = \sum_{i \in \{0, \dots, 4\}} w_i \cdot m_i, \text{ where } w_i \in [0; 1]$$

chosen, such that $\sum_{i \in \{0, \dots, 4\}} w_i = 1$

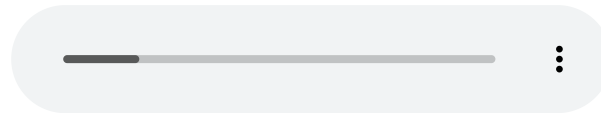
for guaranteeing $m_i \in [0; 1]$

Auditory Dimension
pitch
loudness
instrument
spatialization
duration
brightness
timbre
tempo
spectral power
...

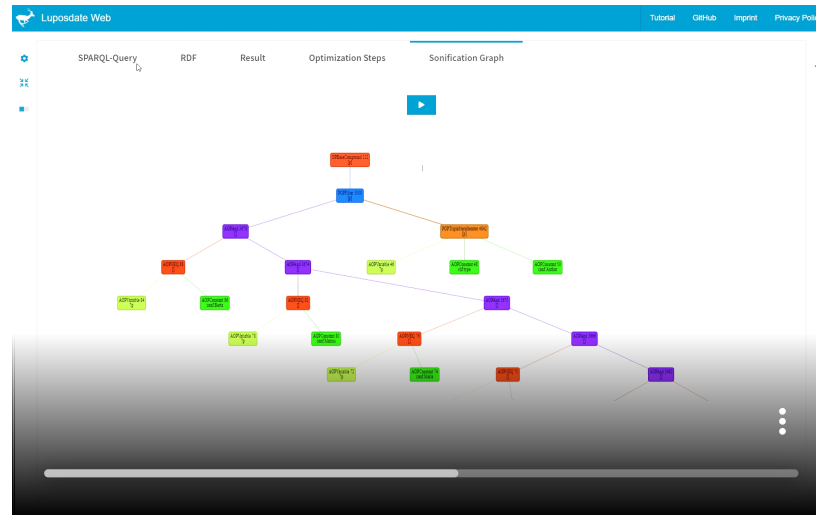
- useful mapping for understanding the operator algorithm (or at least its in- & output):
 - each operator plays another instrument (easy to distinguish from the other)
 - value of intermediate solution maps to pitch
- mapping for melodies: operator depth maps to pitch
- Online: <https://www.ifis.uni-luebeck.de/~groppe/soundofdatabases/>



Please **guess** the kind of operator!



- Solution: **Selection/Filter**



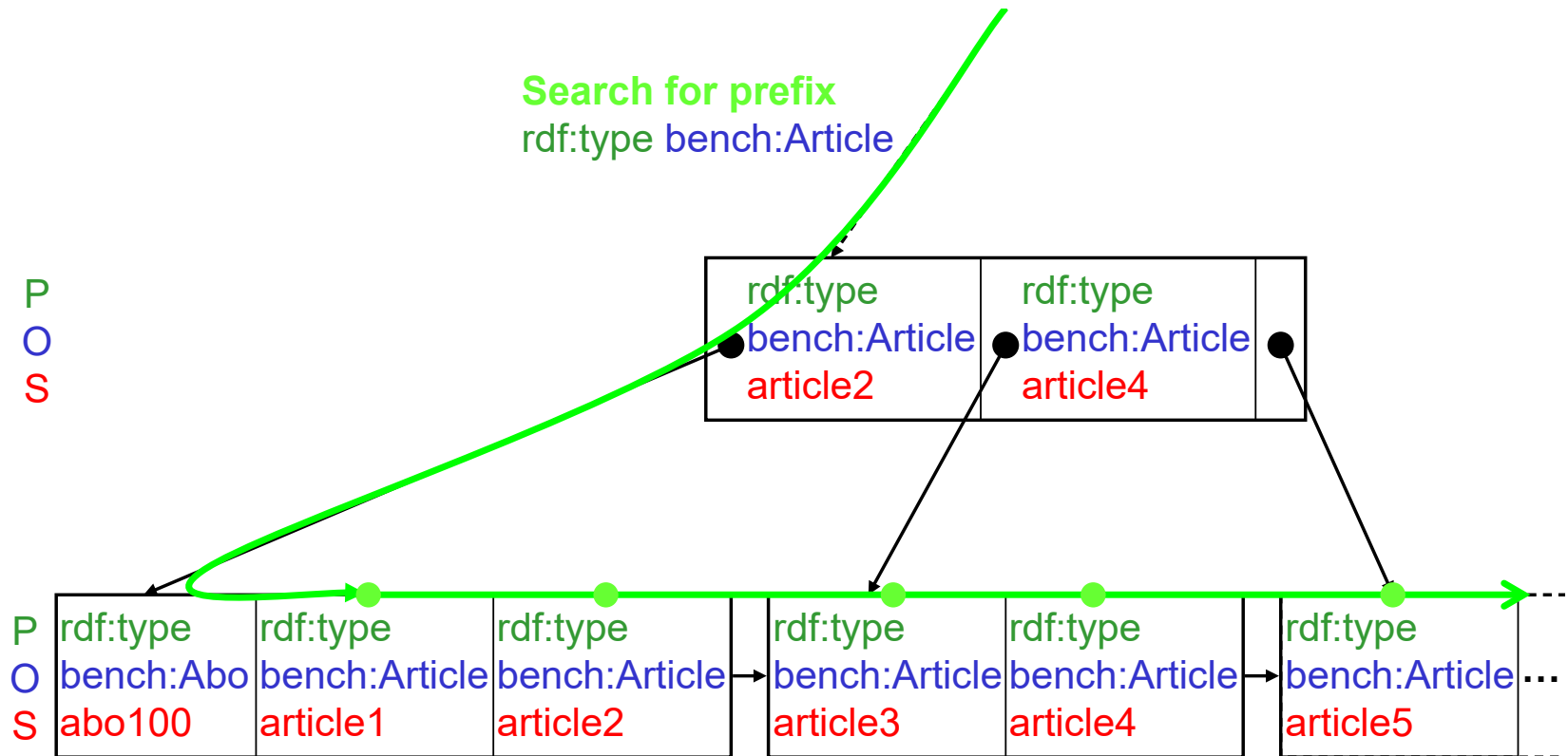
[Start Sonification](#)

- **not all** intermediate results **are passing** the filter operator

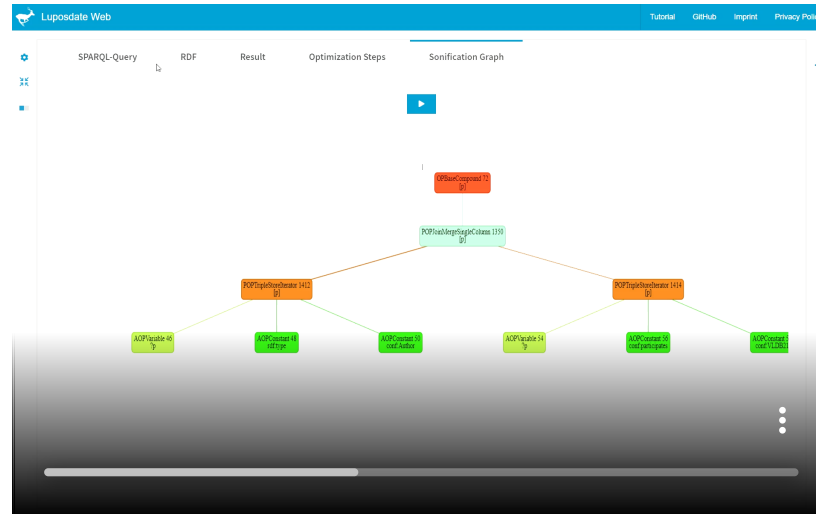
RDF3X Indices for Triple Pattern Evaluation

Triple Pattern	1st Alternative			2nd Alternative		
	Index	Prefix Key	Sort Criterium	Index	Prefix Key	Sort Criterium
?s f:p f:o	POS	f:p f:o	?s	OPS	f:o f:p	?s
f:s ?p f:o	SOP	f:s f:o	?p	OSP	f:o f:s	?p
f:s f:p ?o	SPO	f:s f:p	?o	PSO	f:p f:s	?o
?s ?p f:o	OSP	f:o	?s ?p	OPS	f:o	?p ?s
?s f:p ?o	PSO	f:p	?s ?o	POS	f:p	?o ?s
f:s ?p ?o	SPO	f:s	?p ?o	SOP	f:s	?o ?p
f:s f:p f:o	SPO	f:s f:p f:o	-	All other indices with corresponding keys		

B+-tree



Joining the Result of 2 Triple Patterns



[Start Sonification](#)

- What to **learn**?
 - The **result** is again **sorted** if **merge join** is used

$\bowtie_{=}$ Equi-Join

- Following join algorithms are only suitable for Equi-Joins
- $R \bowtie_C S = \sigma_C(R \times S)$, where C is join condition of the form:
 - $r = s$, where r is attribute of R and s is attribute of S
 - $C_1 \wedge C_2$, where C_1 and C_2 are join conditions

Set Semantics

e.g.: R S $R \bowtie_{R.B=S.B} S$

A	B	B	C	A	R.B	S.B	C
1	2	2	3	1	2	2	3
0	2	2	4	0	2	2	3
3	4	5	6	1	2	2	4
7	8	4	5	0	2	2	4
				3	4	4	5

Multi-Set Semantics

e.g.: R S $R \bowtie_{R.B=S.B} S$

A	B	B	C	A	R.B	S.B	C
1	2	2	3	1	2	2	3
0	2	4	5	0	2	2	3
7	8	4	5	3	4	4	5
3	4	4	5	3	4	4	5
		5	6	3	4	4	5



Merge Join

- **Precondition:** Input sorted according to join columns
 - Otherwise a preceding sort phase necessary → "Sort Merge Join"
- **Algorithm:** step-by-step comparison of both sorted (input) relations
 - If it cannot be joined, then it will be read from the relation with the currently smaller value, because this smaller value does not occur any more in the other relation (utilizing the sorting)

Merge Join as Iterator

```
open() {
    R.open(); r = R.next()
    S.open(); s = S.next()
    B = {} × {}; B.open()
}
```

```
close() {
    R.close()
    S.close()
}
```

Runtime complexity for

- **Input without duplicates** in join columns:
 $O(|R| + |S|)$
- **Input with duplicates** in join columns:
 $O(|R| \times |S|)$

```
next() {
    IF(B.hasNext()) RETURN B.next()
    IF(s == null ∨ r == null) RETURN null
    WHILE( $\pi_{\text{Join Attributes}}(s) \neq \pi_{\text{Join Attributes}}(r)$ ){
        IF( $\pi_{\text{Join Attributes}}(s) < \pi_{\text{Join Attributes}}(r)$ ){
            s = S.next()
            IF(s == null) RETURN null
        } ELSE { // r < s (see WHILE- and IF-condition!)
            r = R.next()
            IF(r == null) RETURN null
        }
    }
    cmp =  $\pi_{\text{Join Attributes}}(s)$  // =  $\pi_{\text{Join Attributes}}(r)$  (see WHILE-condition)
    ts = {}; WHILE( $\pi_{\text{Join Attributes}}(s) == \text{cmp}$ ) { ts = ts ∪ s; s = S.next() }
    tr = {}; WHILE( $\pi_{\text{Join Attributes}}(r) == \text{cmp}$ ) { tr = tr ∪ r; r = R.next() }
    B = ts × tr; B.open()
}
```



Merge Join 1/11

- Result $R \bowtie_{R.lang=S.lang} S$:
- Buffer:
- Input:

 R

first	lang
hallo	de
hello	en
hi	en
hola	sp

 S

lang	second
de	Student
en	student
en	collegian
cn	daxuesheng

Precondition: Inputs must be sorted according to join columns

Merge Join 2/11

- Result $R \bowtie_{R.lang=S.lang} S$:

- Buffer:

R		S	
first	lang	lang	second
hallo	de	de	Student

Load rows with smallest values in join columns into buffer

- Input:

R		S	
first	lang	lang	second
hello	en	en	student
hi	en	en	collegian
hola	sp	cn	daxuesheng

Merge Join 3/11

- Result $R \bowtie_{R.lang=S.lang} S$:

- Buffer:

$$R \bowtie_{R.lang=S.lang} S$$

first	$R.lang$	$S.lang$	second
hallo	de	de	Student

Compute **join result** for current content of buffer

- Input:

R

first	lang
hello	en
hi	en
hola	sp

S

lang	second
en	student
en	collegian
cn	daxuesheng

Merge Join 4/11

- Result $R \bowtie_{R.lang=S.lang} S$:

first	$R.lang$	$S.lang$	second
hallo	de	de	Student

Emit join row

- Buffer:
- Input:

R

first	lang
hello	en
hi	en
hola	sp

S

lang	second
en	student
en	collegian
cn	daxuesheng

Merge Join 5/11

- Result $R \bowtie_{R.lang=S.lang} S$:

first	$R.lang$	$S.lang$	second
hallo	de	de	Student

- Buffer:

R		S	
first	lang	lang	second
hello	en	en	student
hi	en	en	collegian

Load rows with smallest values in join columns into buffer

- Input:

R		S	
first	lang	lang	second
hola	sp	cn	daxuesheng

Merge Join 6/11

- Result $R \bowtie_{R.lang=S.lang} S$:

first	$R.lang$	$S.lang$	second
hallo	de	de	Student

- Buffer:

first	$R.lang$	$S.lang$	second
hello	en	en	student
hello	en	en	collegian
hi	en	en	student
hi	en	en	collegian

Compute **join result** as cartesian product $\times$ of the buffer

- Input:

R		S	
first	lang	lang	second
hola	sp	cn	daxuesheng

Merge Join 7/11

- Result $R \bowtie_{R.lang=S.lang} S$:

first	$R.lang$	$S.lang$	second
hallo	de	de	Student
hello	en	en	student
hello	en	en	collegian
hi	en	en	student
hi	en	en	collegian

Emit join rows

- Buffer:
- Input:

R

first	lang
hola	sp

S

lang	second
cn	daxuesheng

Merge Join 8/11

- Result $R \bowtie_{R.lang=S.lang} S$:

first	$R.lang$	$S.lang$	second
hallo	de	de	Student
hello	en	en	student
hello	en	en	collegian
hi	en	en	student
hi	en	en	collegian

- Buffer:

S

lang	second
cn	daxuesheng

Load rows with smallest values in join columns into buffer

- Input:

R

first	lang
hola	sp

Merge Join 9/11

- Result $R \bowtie_{R.lang=S.lang} S$:

first	$R.lang$	$S.lang$	second
hallo	de	de	Student
hello	en	en	student
hello	en	en	collegian
hi	en	en	student
hi	en	en	collegian

- Buffer: Discard buffer

- Input:

R

first	lang
hola	sp

Merge Join 10/11

- Result $R \bowtie_{R.lang=S.lang} S$:

first	$R.lang$	$S.lang$	second
hallo	de	de	Student
hello	en	en	student
hello	en	en	collegian
hi	en	en	student
hi	en	en	collegian

- Buffer:

R

first	lang
hola	sp

Load rows with smallest values in join columns into buffer

- Input:

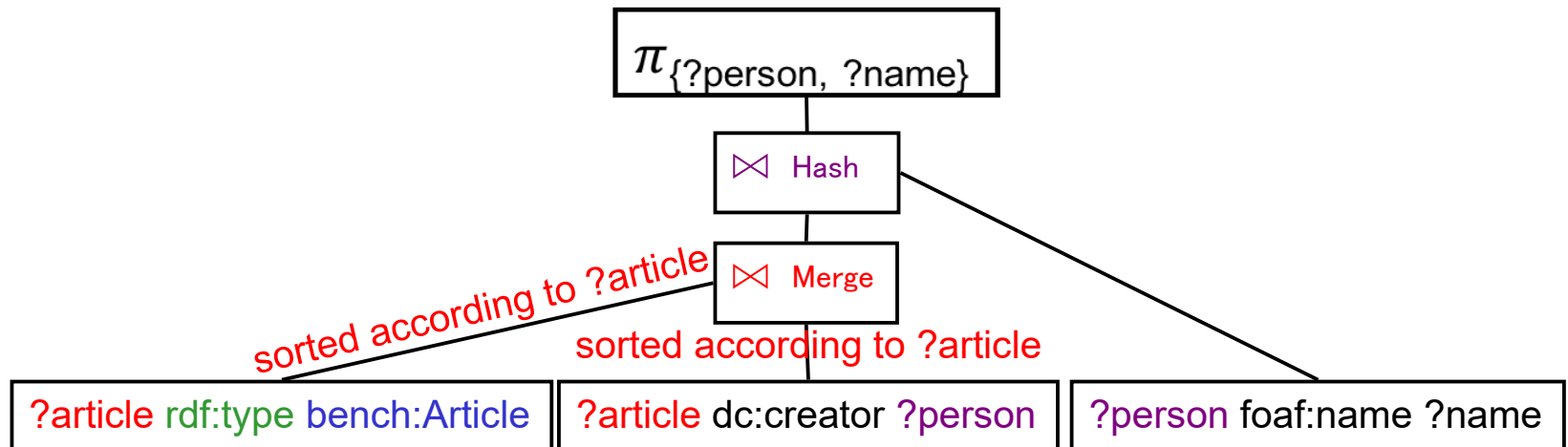
Merge Join 11/11

- Result $R \bowtie_{R.lang=S.lang} S$:

first	$R.lang$	$S.lang$	second
hallo	de	de	Student
hello	en	en	student
hello	en	en	collegian
hi	en	en	student
hi	en	en	collegian

- Buffer: Discard buffer
- Input:

Merge Join

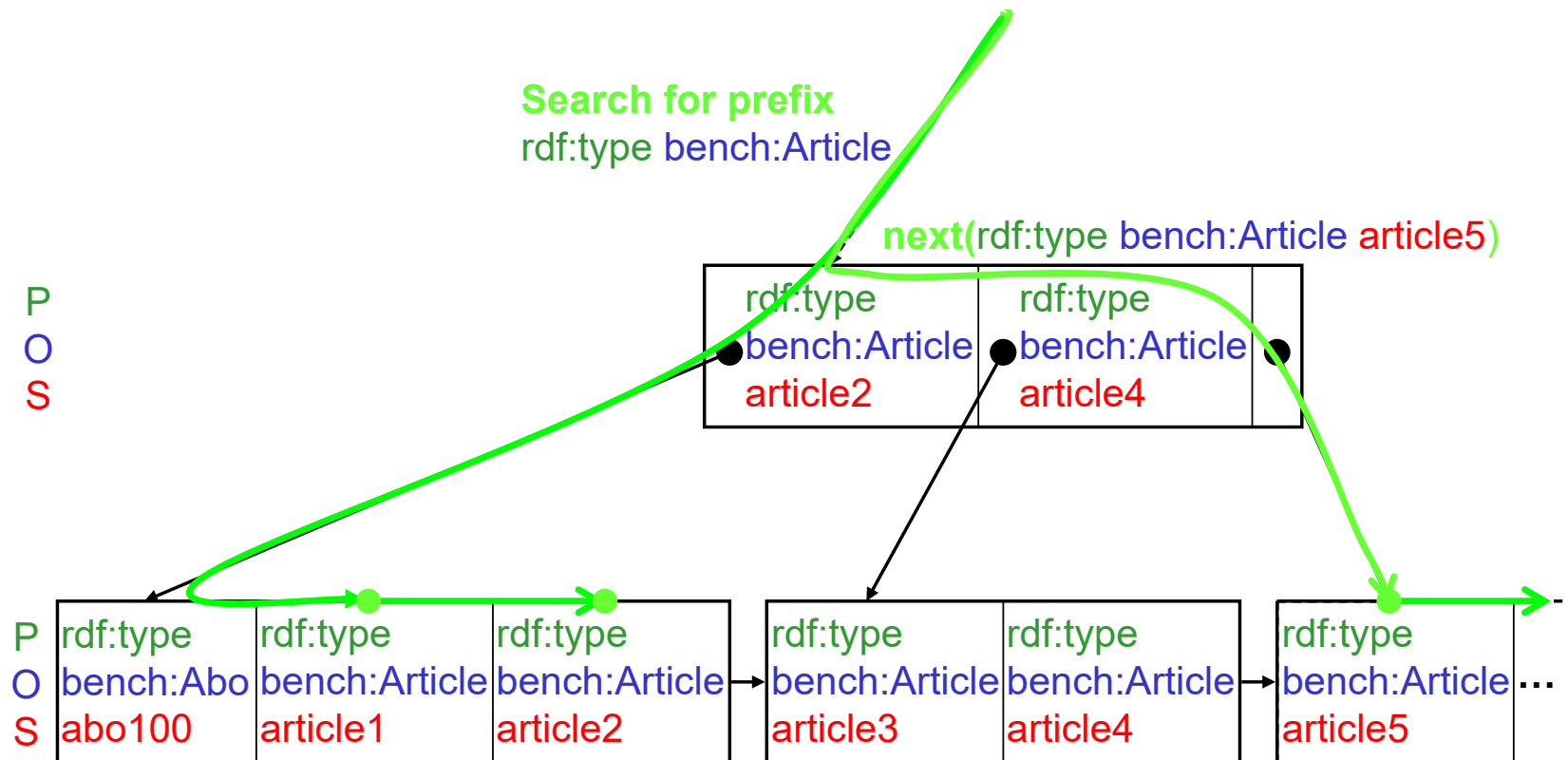


Search for prefix
dc:creator

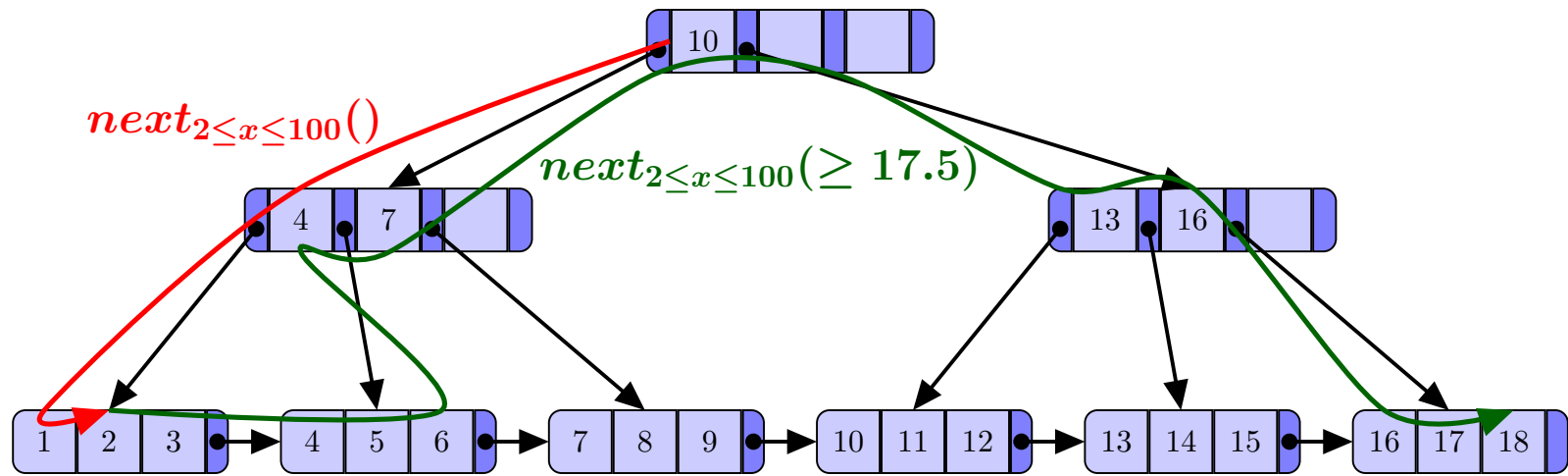
in index sorted according to P S O:

?article = article1 = ?article = article1 ?person=p1
 ?article = article2 ↓ ?article = article2 ?person=p2
 ?article = article3
 ?article = article4
 ?article = article5 ↓ ?article = article5 ?person=p1

Sideways Information Passing (SIP) in RDF3X B⁺-Tree



SIP in B⁺-Tree



- Too much overhead if the inner nodes are searched every time and the value to search for is close to the current one
- **Heuristic:** If the SIP value is not in the current and in the next leaf node, then search via the inner nodes (starting from the leaf node to the root via the previous search path)



Sideways Information Passing (SIP)

```

open() {
    R.open(); r = R.next()
    S.open(); s = S.next()
    B = {} × {}; B.open()
}
close() {
    R.close()
    S.close()
}

```

Runtime Complexity with SIP

In most cases (in practice) $O(|R \bowtie S|)$

```

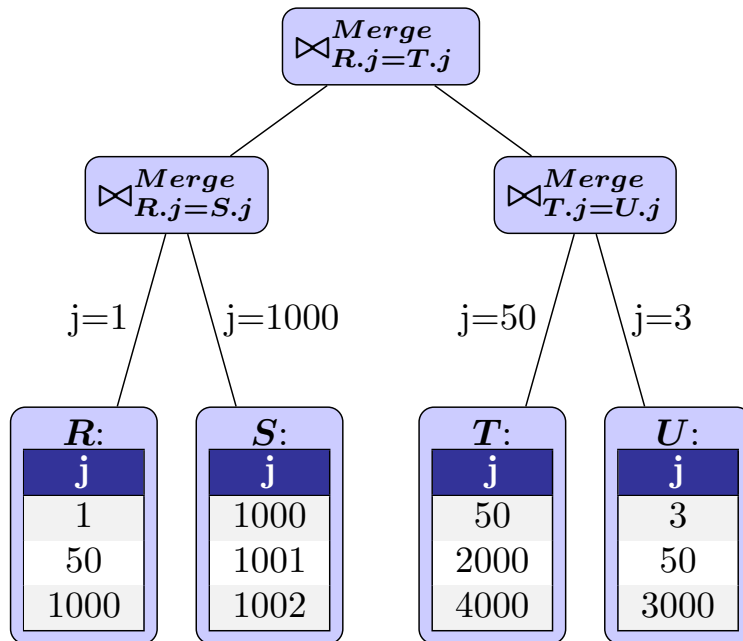
next( sip ) { // modified for SIP! Return result ≥ sip (not guaranteed)!
    IF(B.hasNext( sip )) RETURN B.next( sip ) // SIP!
    IF(s == null ∨ r == null) RETURN null
    WHILE(πJoin Attributes(s) ≠ πJoin Attributes(r) ∨
           πJoin Attributes(r) < sip ∨ πJoin Attributes(s) < sip ){ // SIP!
        IF(πJoin Attributes(s) < πJoin Attributes(r)){
            s = S.next( max(πJoin Attributes(r), sip) ) // SIP!
            IF(s == null) RETURN null
        } ELSE {
            r = R.next( max(πJoin Attributes(s), sip) ) // SIP!
            IF(r == null) RETURN null
        }
    }
    cmp = πJoin Attributes(s) // = πJoin Attributes(r) (see WHILE-condition)
    ts = {}; WHILE(πJoin Attributes(s) == cmp) { ts = ts ∪ s; s = S.next() }
    tr = {}; WHILE(πJoin Attributes(r) == cmp) { tr = tr ∪ r; r = R.next() }
    B = ts × tr; B.open()
    RETURN B.next()
}

```

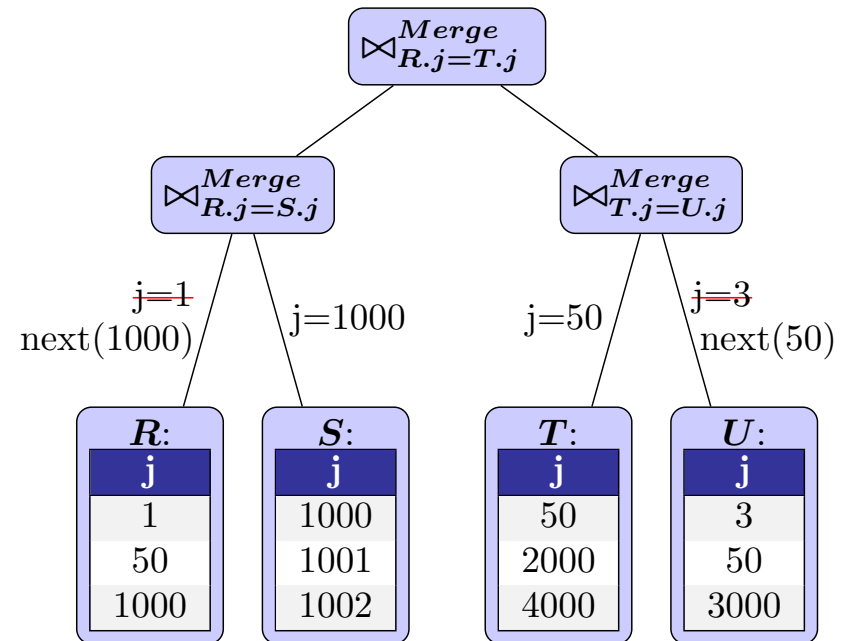
Sideways Information Passing (SIP)

with several Merge Joins 1/2

1.



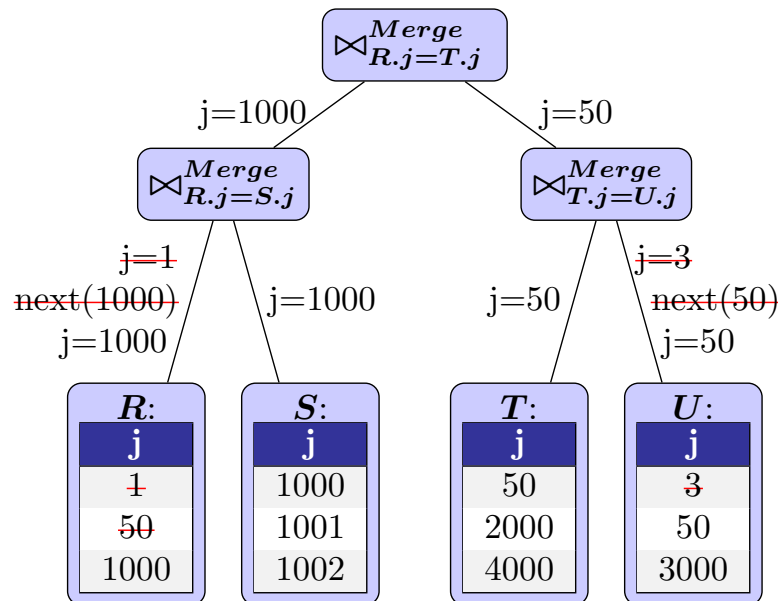
2.



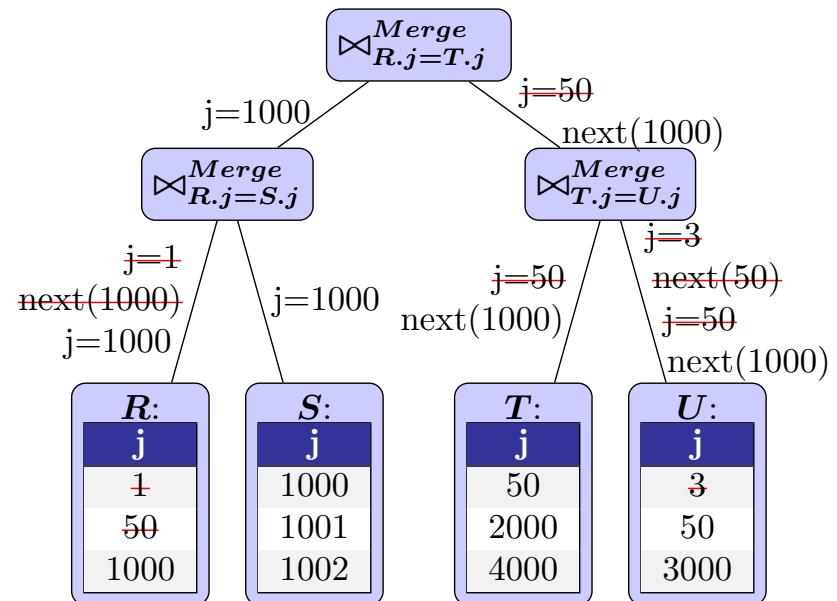
Sideways Information Passing (SIP)

with several Merge Joins 2/2

3.

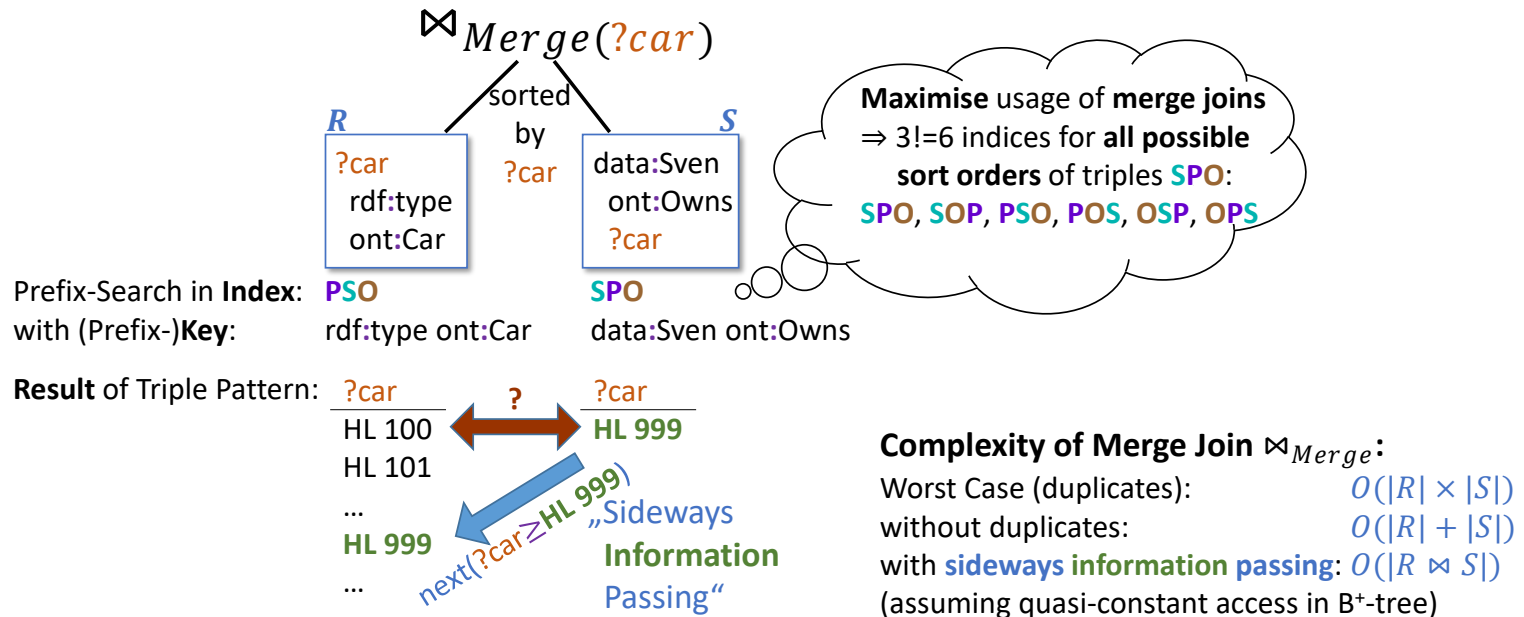


4.

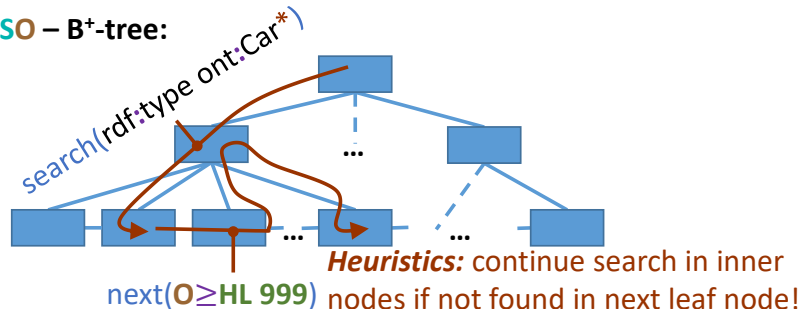


- Information $\text{next}(1000)$ passes sideways from *S* to *R* and also over the root node to *T* and *U*

RDF3X - Indexing Scheme for large-scale RDF triple stores

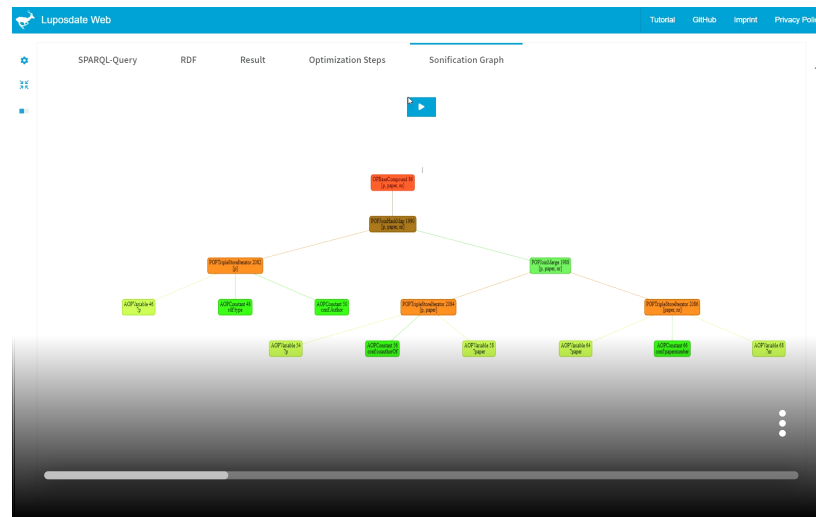


Search in **PSO** – B⁺-tree:



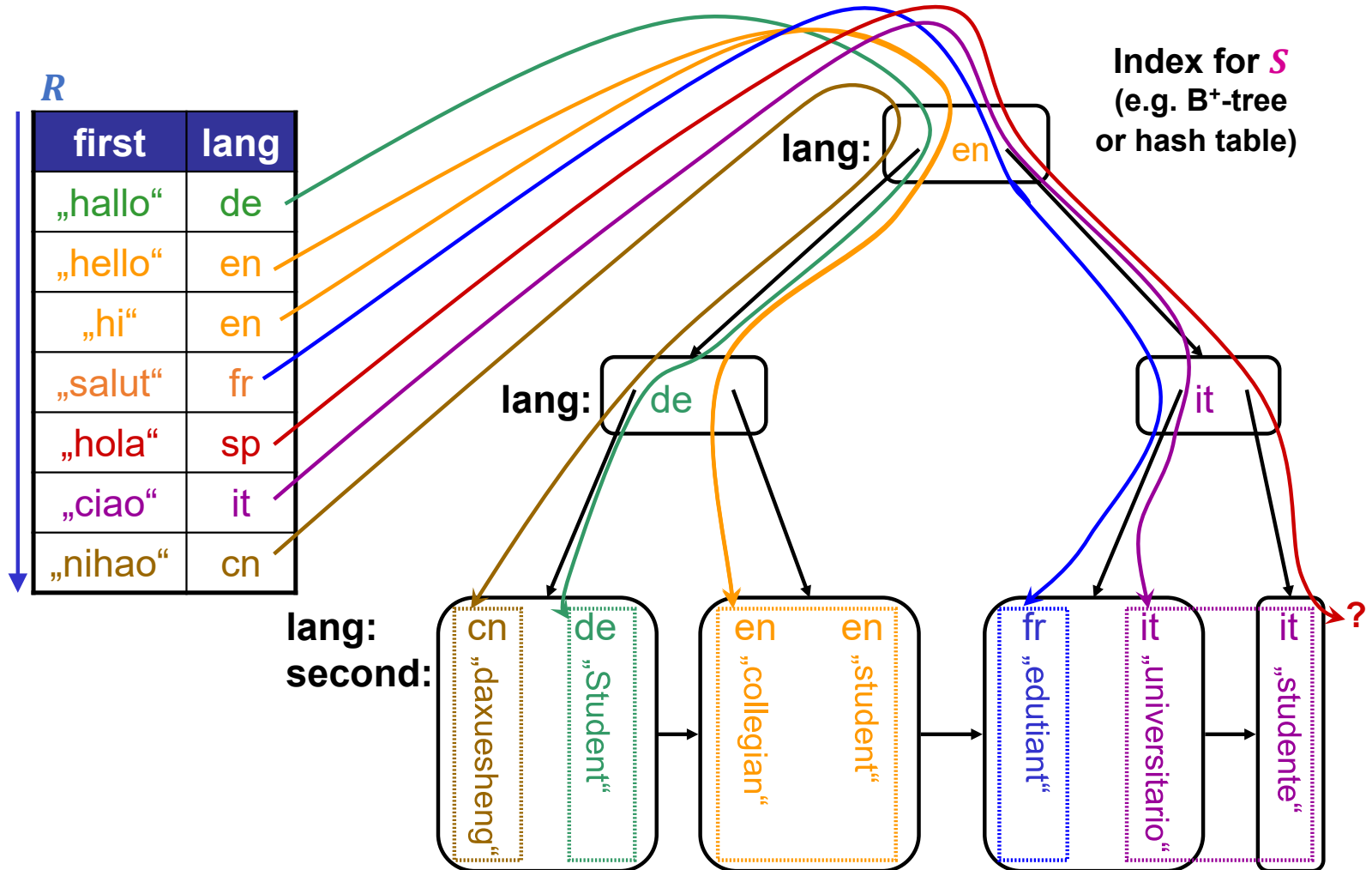


(In-Memory) Hash Join/Index Join with Hash Table



Start Sonification

Index Join 1/2



Index Join 2/2

- Runtime $O(\text{Indexing}(S) + |R| \times \text{Lookup}(\text{Index}(S)))$
 - $\text{Indexing}(S)$ is omitted in case of existing (primary/secondary) indices

	$O(\text{Indexing}(S))$	$O(\text{Lookup}(\text{Index}(S)))$
Hashtable	$O(S \times h(\max(\pi_{key}(s) s \in S)))$	$O(h(\max(\pi_{key}(s) s \in S)))$
B⁺-Tree	$O(S \times \log(S) \times \max(\pi_{key}(s) s \in S))$	$O(\log_k(S))$ respectively $\approx O(1)$ for large k (Practice!)

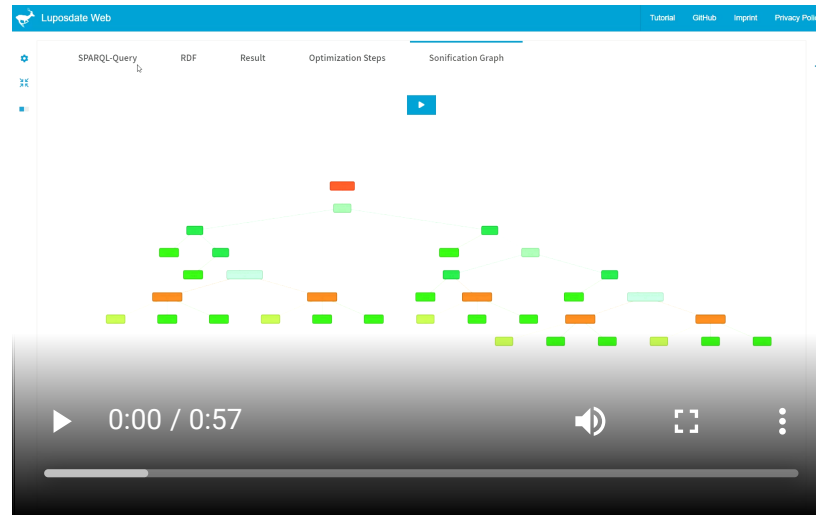
With long keys, the time for hashing, comparing, etc. is significant, otherwise $O(1)$
 h Runtime for hash function (often linear to key length), k Out degree B⁺-tree node
 B⁺-tree can be created after sorting by $1 \times$ traversing the sorted entries

Summary and Conclusions

- SPARQL Query Language
 - Transformation to Relational Algebra Expression/Operatorgraph
- Sound of Databases
- RDF3X Indices
- Merge Join
- Sideways Information Passing (SIP) in
 - Merge Joins
 - RDF3X B⁺-Tree
- Hash Join
- Inspiring Sounds



Whole Pieces of Music from your Database



[Start Sonification](#)