





Deriving an Emergent Relational Schema from RDF Data

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Main Problems in RDF Data Management [1]

- Bad query plans
- Low storage locality
- Lack of user schema insight

RDF de-phresisses method of structure in the data



Recovering the Emergent Schema of RDF data

Emergent schema = "rough" schema to which the majority of triples conforms Recognize:

- Classes (Characteristic Sets CS's) recognize "classes" of often co-occurring properties
- Relationships (CS) recognize often-occurring references between such classes
- + give logical names to these

```
<http://www.w3.org/1999/02/22-rdf-syntax-ns#type>
<http://rdfs.org/sioc/ns#num_replies>
<http://purl.org/dc/terms/title>
<http://rdfs.org/sioc/ns#has_creator>
<http://purl.org/dc/terms/date>
<http://purl.org/dc/terms/created>
<http://purl.org/dc/terms/created>
<http://purl.org/rss/1.0/modules/content/encoded>
```

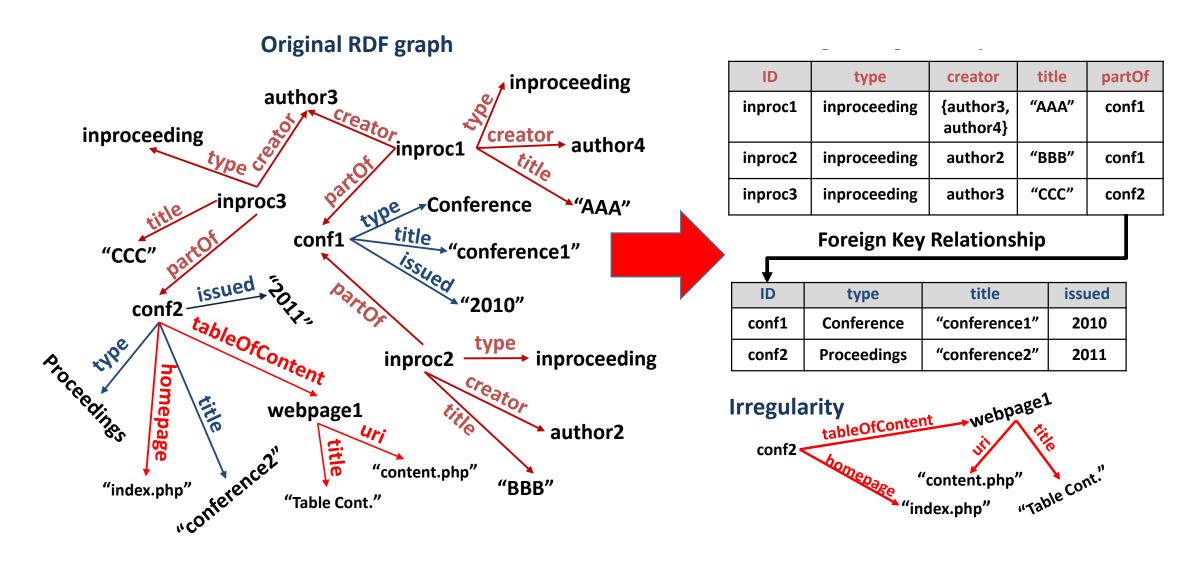
"Book"

<has_creator>

```
<http://www.w3.org/1999/02/22-rdf-syntax-ns#type>
```

<http://xmlns.com/foaf/0.1/name> <http://xmlns.com/foaf/0.1/page>

"Author"



Example of structure recognized from RDF graph



What does "schema" mean?

Relational Schema

Describes the structure of the occurring data Concept mixing (for convenience) Designed for one database (=dataset)

Semantic Web Schema

Purpose: knowledge representation

Describing a concept universe (regardless data)

Designed for interoperability in many contexts

Statement: it is useful to have both an (Emergent) Relational and Semantic Schema for RDF data

- useful for systems (higher efficiency)
- useful for humans (easier query formulation)



When is a Emergent Schema of RDF data useful?

- Compact Schema
 - as few tables as possible
 - homogeneous literal types (few NULLs in the tables)
- Human-friendly "Labels"
 - URIs + human-understandable table/column/relationship names
- High "Coverage"
 - the schema should match almost all triples in the dataset
- Efficient to compute
 - as fast as data import



Basic CS discovery

```
(s1, offers, offer1)
(s1, region, region1)
(s2, offers, offer2)
(s2, offers, offer3)
(s2, region, region1)
(offer1, availableDeliveryMethods, DHL)
(offer1, description, "Offer data")
(offer1, hasBusinessFunction, "Sell")
(offer1, hasEligibleQuantity, 1)
(offer1, hasInventoryLevel, 1)
(offer1, hasStockKeepingUnit, 112)
(offer2, availableDeliveryMethods, DHL)
(offer2, hasPriceSpec, price1)
(offer2, hasStockKeepingUnit, 112)
(offer2, type, Offering)
(price1, hasCurrency, "EUR")
(price1, hasCurrencyValue, "35.99")
(price1, hasUnitOfMeasurement, "C62")
(price1, valueAddedTaxIncluded, "false")
(price1, eligibleTransactionVolume, 0)
(price1, ...
```

' <Example RDF triples>



Characteristic Sets in some well-known RDF datasets

Datasets	#triples*	#CS's	#CS's to cover 90%	Avg. #prop.	#multi-type properties	
LUBM	100M	17	7	5.71	0	
BSBM	100M	49	14	12.61	0	
SP2Bench	100M	554	7	9.8	0	
synthetic	data	created b	y benchmar	rk data g	generator	
MusicBrainz	179M	27	10	4.7	0	
EuroStat	70K	44	8	7.77	0	
DBLP	56M	249	8	13.70	0	
PubMed	1.82B	3340	35	19.27	0	
relational	RDF (data fron	n a relation	al datab	ase dump	
WebData.	90M	13354	930	7.94	551	
DBpedia	404M	439629	85922	24.36	1507	
native		real data originating as RDF				



Partial and Mixed Use of Ontologies

	mixed	
	number of	9
	ontology	
	classes	1
dataset	used per CS	us
LUBM	1.94	
BSBM	3.96	
SP2Bench	4.94	
MusicBrainz	3.93	
EuroStat	3.14	
DBLP	6.58	
PubMed	4.94	
WebData.	2.27	
DBpedia	8.35	L

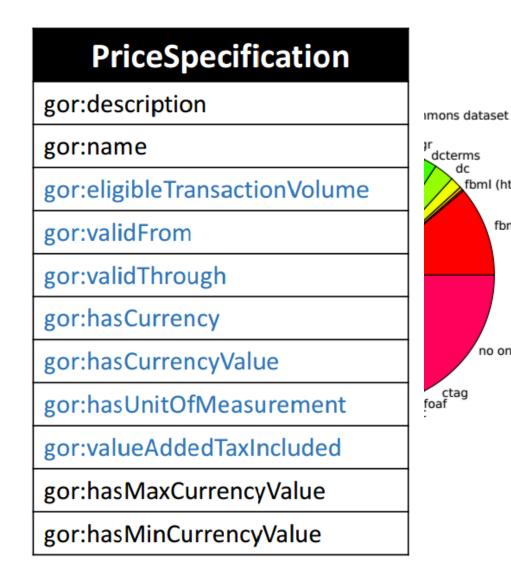
	cs ₄
_	dc:description
9	gor:validFrom
I us	gor:validThrough
	gor:hasCurrency
	gor:hasCurrencyValue
	gor:hasUnitOfMeasurement
	gor:valueAddedTaxIncluded
	gor:eligibleTransactionVolume

(prefix gor:

http://purl.org/goodrelations/v1#

prefix dc:

http://purl.org/dc/elements/1.1/)



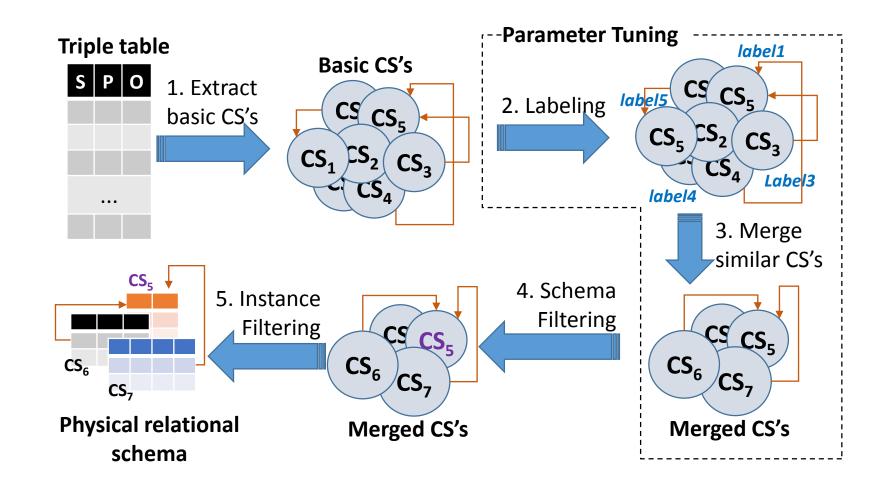
fbml (https)

fbml

no ontology



Emerging a Relational Schema





Results: compact schemas with high coverage

	Number of tables			Coverage – Metric C (%)		
Datasets	before	after	remove	remove	prune	final
	merging	merging	small	small	infreq.	schema
			tables	tables	prop.	
LUBM	17	13	12	100	100	100.00
BSBM	49	8	8	100	100	100.00
SP2B	554	13	10	99.99	99.65	99.65
MusicBrainz	27	12	12	100	99.9	99.60
EuroStat	44	10	5	99.73	99.53	99.53
DBLP	249	9	6	100	99.68	99.60
PubMed	3340	14	12	100	99.75	99.73
WebData.	13354	3000	253	98.17	94.37	92.79
DBpedia	439629	542	234	99.12	96.68	95.82



Results: understandable labels & performance

labels	WebData.	DBpedia
top 3	3.6	3.8
final	4.1	4.6

Table 3: Human survey results on Likert scale

Likert Score: 1=bad 5=excellent

	Query 3			Query 5		
RDF Store	Cold	Hot	Opt. Time	Cold	Hot	Opt. Time
Virt-Quad	4210	53	40.2	3842	1350	18.6
Virt-CS	2965	9	5.4	2130	712	4.2

Table 5: Query time (msecs) w/wo the recognized schema (Cold: First query runtime after re-starting the server

Hot: Run the query 3 times and get the last runtime

Opt. Time: Query optimization time)

