# Selection of representative subsets of link key candidates

Nacira Abbas, Alexandre Bazin, Jérôme David, Amedeo Napoli



Montbonnot, France Firstname.Lastname@inria.fr https://moex.inria.fr

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# Outline

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#### Data interlinking

Link keys

Link key candidates extraction (with FCA)

Link key candidate reduction and selection

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# The problem: RDF data interlinking

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- Numerical specifications (Link Specifications)
  - Express or learn a similarity from RDF data
  - Generate links using frameworks such as SILK or LIMES
- NLP/IR based approaches
  - Change representation: from RDF space to VSM, or embedding spaces
  - Compute or learn a similarity on this new space
- Logical link specifications
  - Key-based: combine keys and alignments for deducing links
  - Link keys: cross dataset, generalization of keys without requiring alignment between properties or concepts

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#### They may be several expressions having the form of link keys

D (Employés)					D' (Staff)					
id	prenom	datenaiss	poste	bât.	firstname	birthdate	position	building	id	
i <sub>2</sub>	Paul	1967	Dir.	B2	Paul		Dir.	B2	<i>z</i> <sub>2</sub>	
i <sub>3</sub>	Mary	1963	Dir.	B1	Mary		Dir.	B1	$Z_3$	
<b>i</b> 4	John	1963	Pr.	B1	John		Pr.	B1	<b>Z</b> 4	
<b>i</b> 6	Bill	1980	Pr.	B1	William	1980	Pr.		Z6	
i7	Ana	1947	Dir.	B2	Ana	1947	Dir.		<b>Z</b> 7	
i <sub>8</sub>	John	1967	Pr.	B2	John	1967	Pr.		$Z_8$	

Example of link key expressions:

- $k = \langle \{\}, \{\langle datenaiss, birthdate \rangle\}, \langle Employe, Staff \rangle \rangle$
- $h = \langle \{ \langle datenaiss, birthdate \rangle \}, \{ \langle poste, position \rangle \} \langle Employe, Staff \rangle \rangle$
- $\blacktriangleright l = \langle \{ \langle datenaiss, birthdate \rangle, \langle poste, position \rangle \}, \{ \langle poste, position \rangle \}, \langle Employe, Staff \rangle \rangle$

And generated links (if used as link keys):

$$\begin{array}{l} \blacktriangleright \ \ L_k^{D,D'} = \{\langle i_7, z_7 \rangle, \langle i_8, z_8 \rangle, \langle i_6, z_6 \rangle, \langle i_2, z_8 \rangle \} \\ \blacktriangleright \ \ \ L_l^{D,D'} = L_h^{D,D'} = \{\langle i_7, z_7 \rangle, \langle i_8, z_8 \rangle, \langle i_6, z_6 \rangle \} \end{array}$$

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<i>i</i> 4	John	1963	Pr.	B1	N	John		Pr.	B1	$Z_4$
<b>i</b> 6	Bill	1980	Pr.	B1	$\leftrightarrow$	$\forall \forall$ William	1980	Pr.		<i>Z</i> 6
i7	Ana	1947	Dir.	B2	$\leftarrow \parallel$	🕂 Ana	1947	Dir.		<b>Z</b> 7
i <sub>8</sub>	John	1967	Pr.	B2	$\leftarrow \parallel$	→ John	1967	Pr.		<i>Z</i> 8

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#### Data interlinking

Link keys

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# Problem: How to induce link keys from data?

The number of set of pairs of properties is exponential

- Our approach:
  - restrict on link key expressions that would generate at least one link between the datasets
  - consider only closed expressions : those which are maximal for a set of links
- we call such expressions "link key candidates" (LKC) and we extract them using Formal Concept Analysis

The formal context for link key candidates  $\langle G, M, I \rangle$  is:

- ► *G* : the set of pairs of objects from each dataset
- *M* : two sets of pairs of properties from each dataset
  - ►  $\exists$  : if the objects share at least one value ( $\langle o, o' \rangle I \exists \langle p_i, p'_i \rangle$ )
  - ►  $\forall$ : if the object have the same values ( $\langle o, o' \rangle I \forall \langle p_i, p'_i \rangle$ )

# Lattice of extracted link key candidates

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FCA link key extraction is implemented in Linkex
 (https://gitlab.inria.fr/moex/linkex)

Characteristics/functionalities:

- Fully unsupervised: only two RDF datasets as input.
- Normalization of textual content: lowercase, remove diacritics, tokenization, sort
- Can compute inverse and composition of properties
- Compute the class expression of concepts (i.e. covering instances)
- Different output formats: Alignment Format, GraphViz (dot), tabular file, etc.
- Implementation of diverse quality measures (discriminability, coverage, etc)

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# The problem: link key selection

#### Many link key candidates can be extracted!

ex: OAEI Spimbench task, > 2k candidates



How to reduce the lattice and select the interesting/representative ones?

- 1. Reduce
  - Identify redundant LKC according to owl:sameAs semantics
  - Representative link key candidates based on clustering
- 2. Evaluate
  - Quality measures: discriminability and coverage
- 3. Combine: Disjunctions based on antichains
  - Explore the antichains of the LKC lattice

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# Pattern structure for link key candidate discovery



#### Pattern structure: $(G, (E, \Box), \delta)$

PS objects	Descriptions ( $\delta$ )
G	Ε
$(a_1, b_1)$	$\{\exists (p_1, q_1), \exists (p_2, q_2)\}$
$(a_1, b_2)$	$\{\exists (p_2, q_2)\}$
$(a_2, b_1)$	$\{\exists (p_1, q_1)\}$
$(a_2, b_2)$	$\{\exists (p_1, q_1), \exists (p_2, q_2)\}$
$(a_3, b_3)$	$\{ orall (p_1, q_1), \exists (p_1, q_1), \exists (p_2, q_2) \}$
$(a_4, b_4)$	$\{ \forall (p_3, q_3), \exists (p_3, q_3) \}$
$(a_4, b_5)$	$\{ \forall (p_4, q_4), \exists (p_4, q_4) \}$
$(a_5, b_4)$	$\{ orall (p_4, q_4), \exists (p_4, q_4) \}$
$(a_5, b_5)$	$\{ \forall (p_3, q_3), \exists (p_3, q_3) \}$

Nacira Abbas, Jérôme David, Amedeo Napoli: Discovery of Link Keys in RDF Data Based on Pattern Structures: Preliminary Steps. CLA 2020: 235-246

## Pattern Structures lattice

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# Partition pattern Structures lattice

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#### Idea: use Partition pattern structures to detect redundant LKC



#### Link key candidates having the same partition are in the same concept

Nacira Abbas, Alexandre Bazin, Jérôme David, Amedeo Napoli: A Study of the Discovery and Redundancy of Link Keys Between Two RDF Datasets Based on Partition Pattern Structures. CLA 2022: to appear

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This works but ... this is not so useful in practice :-(

•										
Interlinking task datasets		#subj	#prop	#LKC	#NRLKC					
db:Actor	94 606	5 807	16	2 109	2 177 (↓ 1%)					
yago:Actor	1 029 580	108 415	16	2 190						
db:Album	594 144	85 002	5	44	44					
yago:Album	762 238	136 848	5	44						
db:Book	247 372	29 846	7	0.2	82					
yago:Book	185 032	41 849	7	82						
db:Film	1 369 600	82 099	9	10 7 10	17 643 (↓ 5%)					
yago:Film	1 067 084	123 822	9	10 / 10						
db:Mountain	135 442	16 397	5	20	39					
yago:Mountain	233 562	32 874	5	39						
db:Museum	15 940	1 826	7	49	49					
yago:Museum	163 342	21 050	7	40	40					
db:Organization	4 487 205	183 665	17	1 425	1 425					
yago:Organization	4 410 854	430 071	17	1423						
db:Scientist	128 360	18 409	8 409 10		962					
yago:Scientist	671 266	92 828	18	002	002					
db:University	241 838	10 352	9 212		212					
yago:University	263 624	23 334	9	213	215					
	datasets db:Actor yago:Actor db:Album yago:Album db:Book yago:Book db:Film yago:Film db:Mountain yago:Mountain db:Museum yago:Museum yago:Museum yago:Museum db:Organization yago:Crganization db:Scientist yago:Scientist db:University yago:University	datasets         #triple           db:Actor         94 606           yago:Actor         1029 580           db:Album         594 144           yago:Album         594 144           yago:Album         762 238           db:Book         247 372           yago:Book         185 032           db:Film         1 369 600           yago:Film         1 067 084           db:Mountain         135 442           yago:Museum         15 940           yago:Museum         15 940           yago:Museum         163 342           db:Auseum         163 342           db:Scientist         128 360           yago:Scientist         671 266           db:University         241 838           yago:Scientist         671 266	datasets         #triple         #subj           db:Actor         94 606         5 807           yago:Actor         1029 580         108 415           db:Album         594 144         85 002           yago:Album         594 144         85 002           yago:Album         762 238         136 848           db:Book         247 372         29 846           yago:Book         185 032         41 849           db:Film         1 369 600         82 099           yago:Film         1 067 084         123 822           db:Mountain         135 442         16 397           yago:Mountain         133 562         32 874           db:Museum         15 940         1826           yago:Museum         163 342         21 050           db:Corganization         4 487 205         183 665           yago:Scientist         128 360         18 409           yago:Scientist         671 266         92 828           db:University         241 838         10 352           yago:Scientist         673 263 23 334         10 352	datasets         #triple         #subj         #prop           db:Actor         94 606         5 807         16           yago:Actor         1029 580         108 415         16           db:Album         594 144         85 002         5           yago:Album         762 238         136 848         5           db:Book         247 372         29 846         7           yago:Book         185 032         41 849         7           db:Film         136 9600         82 099         9           yago:Film         1067 084         123 822         9           db:Mountain         135 442         16 397         5           yago:Mountain         233 562         32 874         5           db:Museum         15 940         1 826         7           yago:Museum         163 342         21 050         7           db:Cinganization         4 470 854         430 071         17           db:Scientist         128 360         18 409         10           yago:Scientist         671 266         92 828         18           db:University         241 838         10 352         9           yago:University         263 624 23 333 49	datasets         #triple         #subj         #prop         #LKC           db:Actor         94 606         5 807         16         2 198           db:Actor         1029 580         108 415         16         2 198           db:Alburn         594 144         85 002         5         44           db:Book         247 372         29 846         7         82           db:Film         1369 600         82 099         9         18 718           db:Mountain         1369 600         82 099         9         18 718           db:Mountain         135 442         16 397         5         39           db:Mountain         133 562         32 874         5         39           db:Museum         163 342         21 050         7         48           db:Corganization         4 487 205         183 665         17         1425           yago:Scientist         128 360         18 409         10         862           db:Scientist         128 360         18 409         10         862           yago:Scientist         671 266         92 828         18         862           db:University         241 838         10 352         9					

#LKC: number of link key candidates. #NRLKC: number of "non redundant" link key candidates.

#### Datasets from

Danai Symeonidou, Luis Galárraga, Nathalie Pernelle, Fatiha Saïs, Fabian M. Suchanek: VICKEY: Mining Conditional Keys on Knowledge Bases. ISWC (1) 2017: 661-677

# Similarity between link keys

However, a lot of link key candidates generate almost the same links...



We propose to select a subset of representative candidates.

The procedure is as follows:

- 1. similarity is computed thanks to Jaccard index (between sets of links)
- 2. LKC are clustered with hierarchical agglomerative clustering
- 3. clusters are extracted from the resulting hierarchy
- 4. a representative LK is selected for each cluster
  - the LKC that minimizes the distance to the other

# LKC clustering

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#### Cut at 0.5:

- 4 clusters:  $\{k_1, k_2, k_3\}, \{k_4\}$  and  $\{k_6\}$
- representatives (core):  $\{k_3, k_4, k_6, k_7\}$
- ▶ This gives a compression ratio of 0.43 while preserving 77% of links.

## Results on benchmarks

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- Almost 50% of LKC can be removed without loss
- Core LKC are good representative (and preserve F-Measure).

# Conclusion

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The context:

- Link keys are symbolic and meaningful tool for interlinking data
- Fully unsupervised extraction from data with minimal input
- But... a lot of candidates can be discovered

Contributions:

- A pattern structure for non redundant link key w.r.t. owl:sameAs
- A clustering based strategy for selecting representative subset of LKC

Perspectives:

- Combine this approach with selecting disjunctions of LKC
- Generalize this clustering approach to FCA lattice reduction
  - Goal: reduce the # of concepts while preserving the order
  - adapt the similarity measure to FCA lattices  $\frac{|L(a \land b)|}{|L(a \lor b)|}$ instead of  $\frac{|L(a) \cap L(b)|}{|L(a) \cup L(b)|}$
  - design an optimised algorithm

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Nacira . Abbas Jerome . David Amedeo . Napoli

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