

Aligning Unions of Concepts in Ontologies of Geospatial Linked Data

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INTRODUCTION

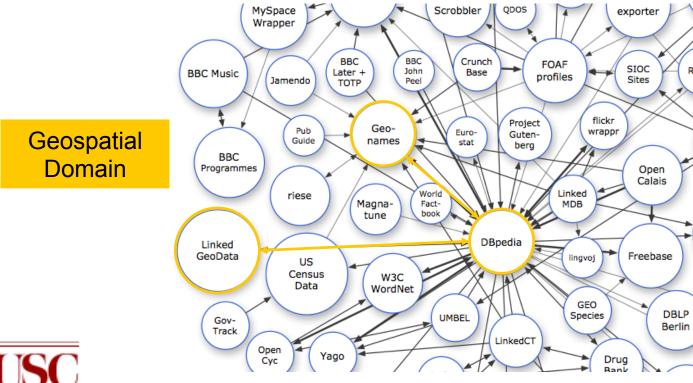






Web of Geospatial Linked Data

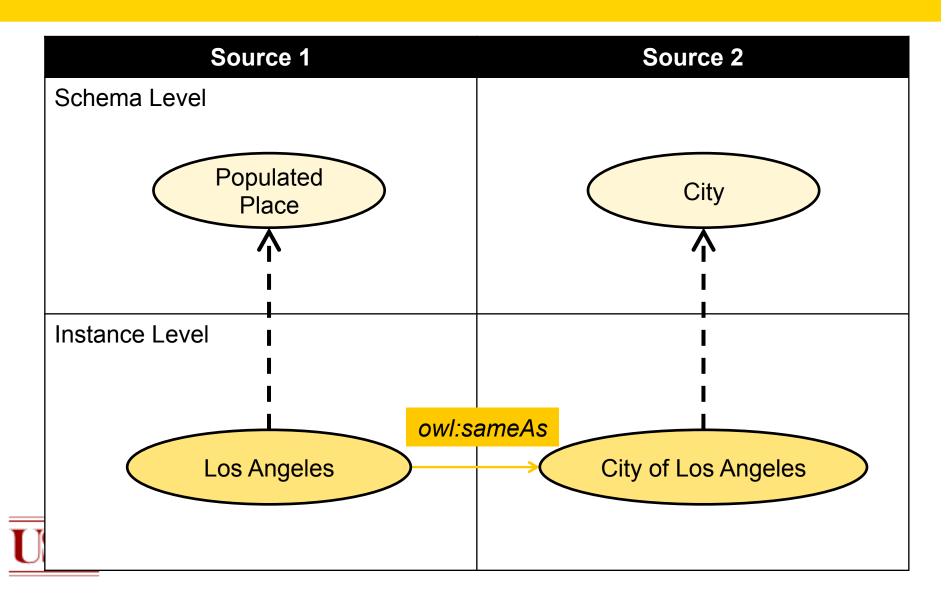
- Different sources with different schemas
- Equivalent instances in the geospatial domain connected with owl:sameAs





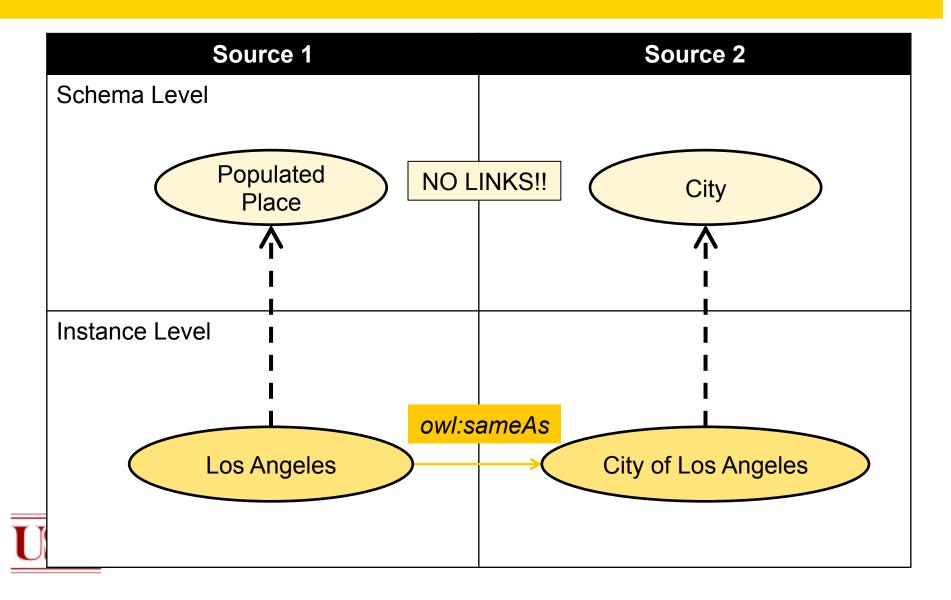


Interlinked instances...



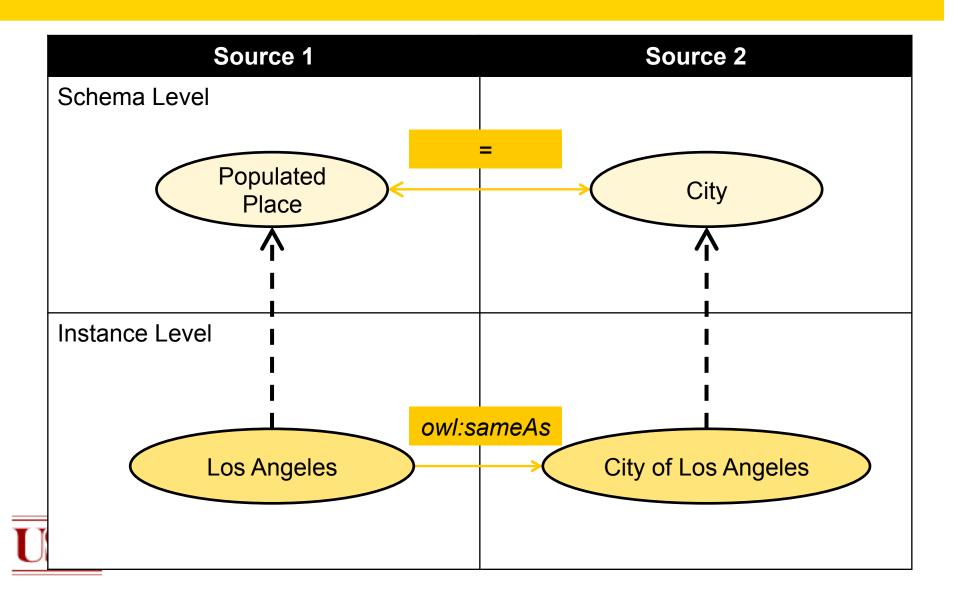


...with disjoint schemas





Can we find schema alignments?





Previous Work @ ISWC 2010

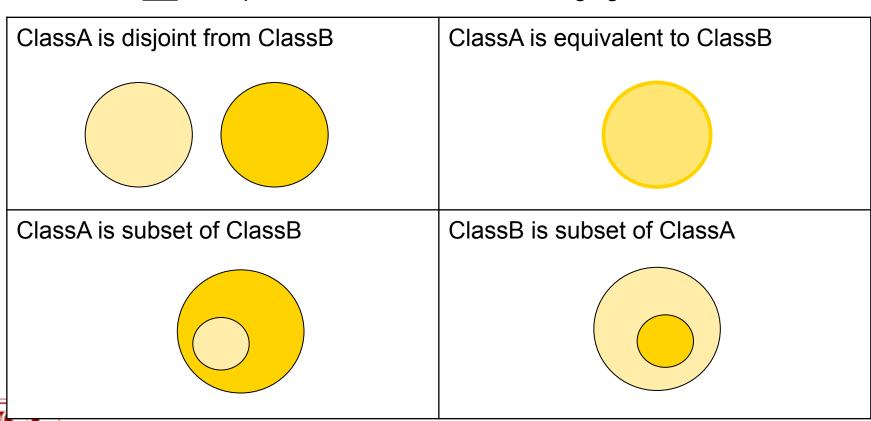
Linking and Building Ontologies of Linked Data





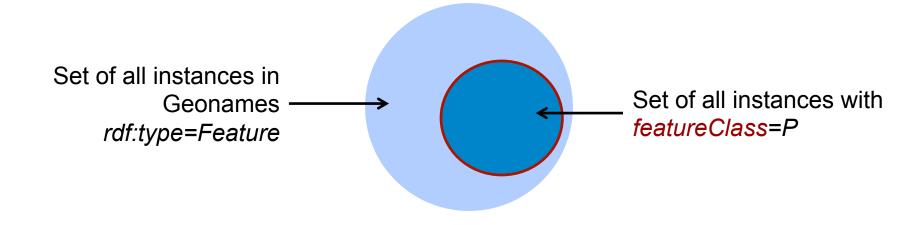
Extensional Approach to Ontology Alignment

	Represents set of instances belonging to ClassA
	Represents set of instances belonging to ClassE





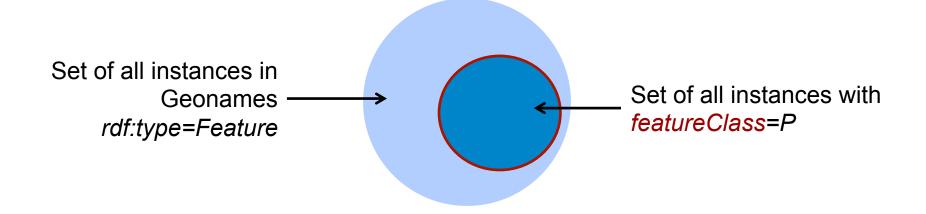
Classes are created extensionally by adding value restrictions on properties

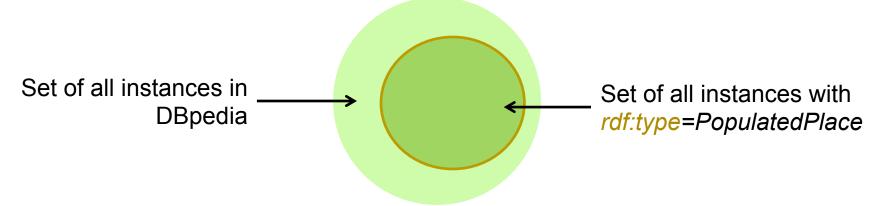






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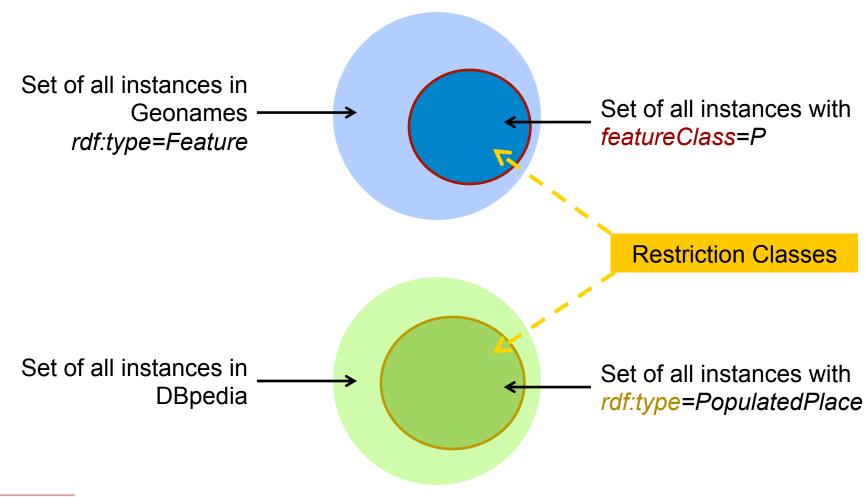








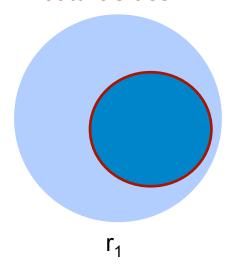
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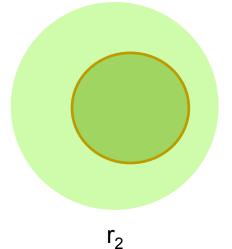




featureClass=P



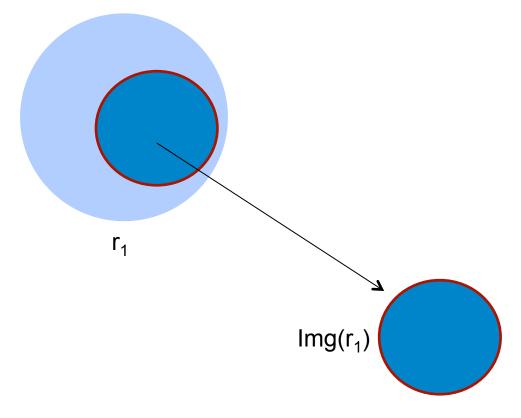
rdf:type=PopulatedPlace





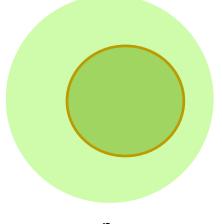


featureClass=P



Set of instances from DBpedia that r_1 is linked to

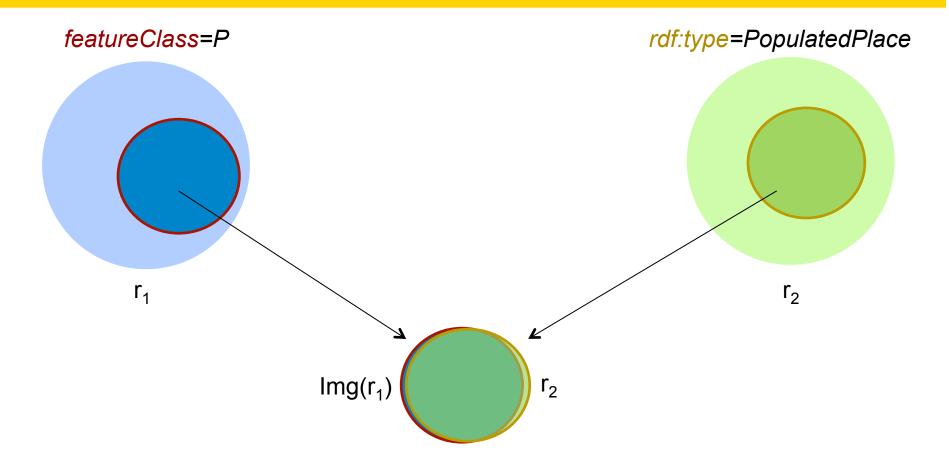
rdf:type=PopulatedPlace



 r_2











Extensionally, when are two classes equal?

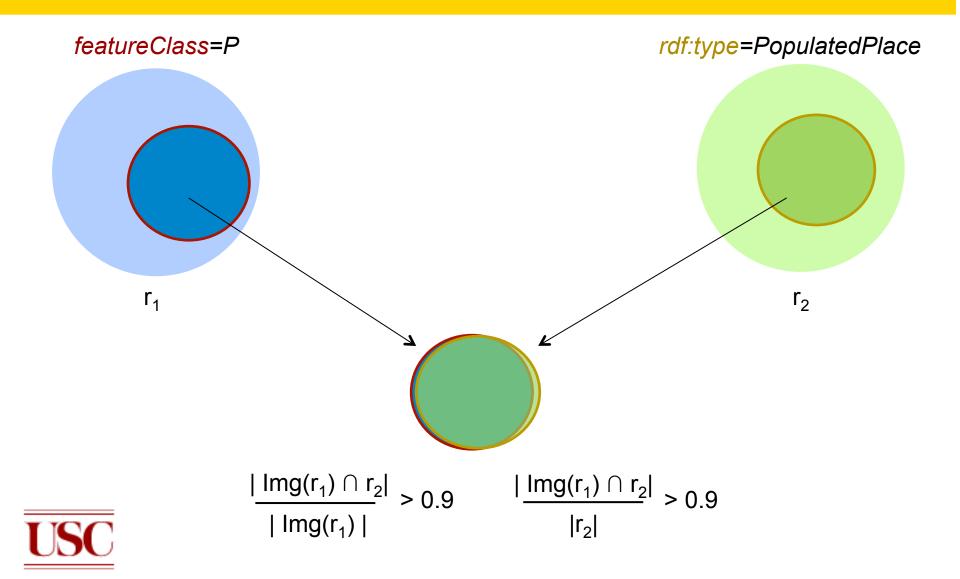
- Represents set of instances belonging to ClassA
- Represents set of instances belonging to ClassB



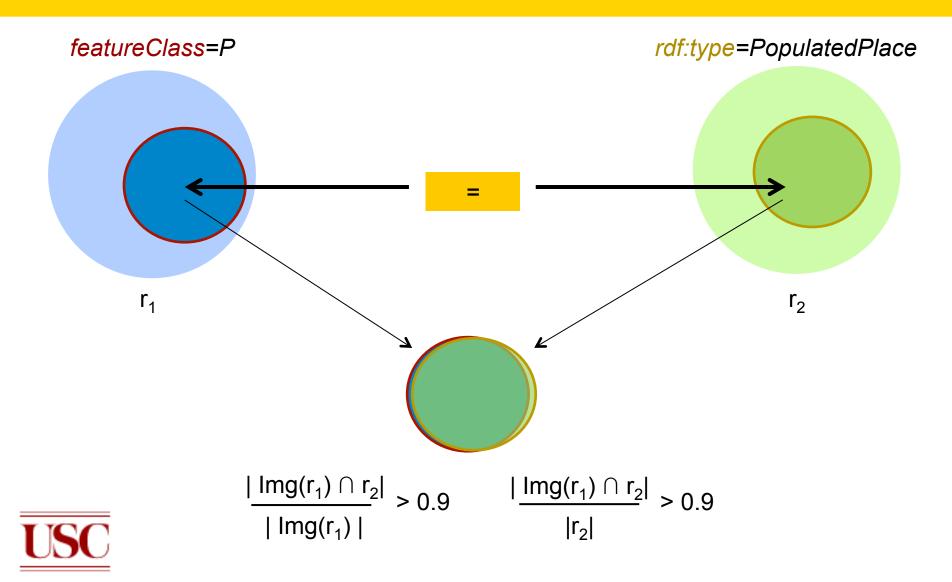
$$\frac{|ClassA \cap ClassB|}{|ClassA|} = \frac{|ClassA \cap ClassB|}{|ClassB|} = 1$$













Alignments Found in the ISWC'10 Paper

- Algorithm was able to
 - Specialize ontologies where original were rudimentary
 - Find complimentary hierarchy across an ontology
- Alignments based on the actual data
 - reflects the semantics of the sources in practice
- Equivalences, Subset alignments before and after removing implied alignments

	Source 1	Source 2	$\#(r_1 = r_2)$	$\#(r_1 = r_2)$	$\#(r_1 \subset r_2)$	$\#(r_1 \subset r_2)$	$\#(r_2 \subset r_1)$	$\#(r_2 \subset r_1)$
	(O_1)	(O_2)	total	best matches	before	after	before	after
	LinkedGeoData	DBpedia	158	152	2528	1837	1804	1627
	Geonames	DBpedia	31	19	809	400	1384	1247
	Geospecies	DBpedia	509	420	9112	2294	6098	4455
Y	MGI	GeneID	10	9	2031	1869	3594	2070
J)	Geospecies	Geospecies	94	88	1550	1201	-	-



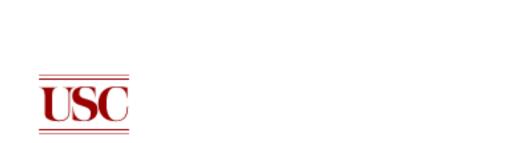
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Can we use the subset relations to find more meaningful alignments?







TerraCognita Workshop - ISWC 2011

Aligning Unions of Concepts in Ontologies of Linked Geospatial Data





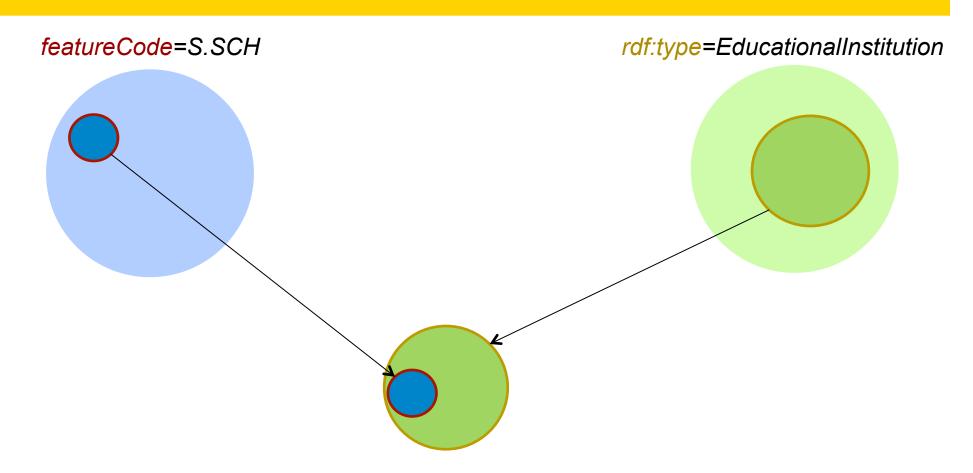
Is there a pattern in the subset relations?

Let's look at 3 of the subset relations we found...





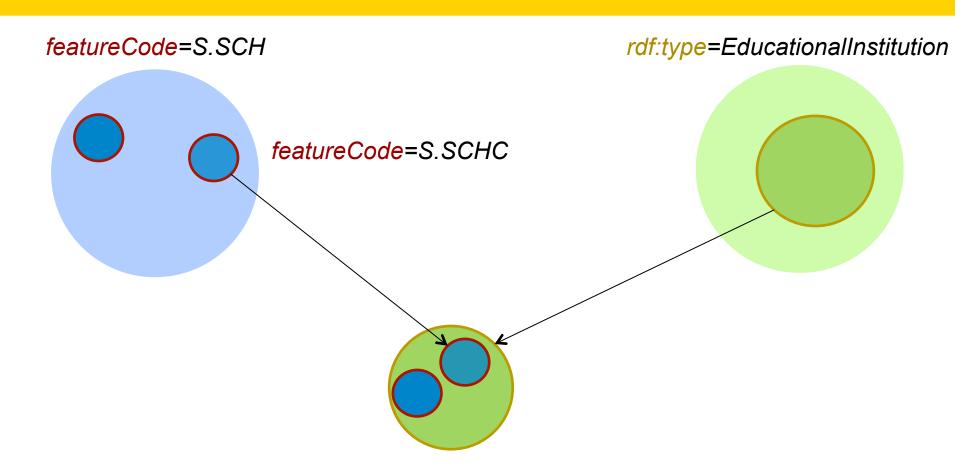
1) Schools in *Geonames* are Educational Institutions in *DBpedia*







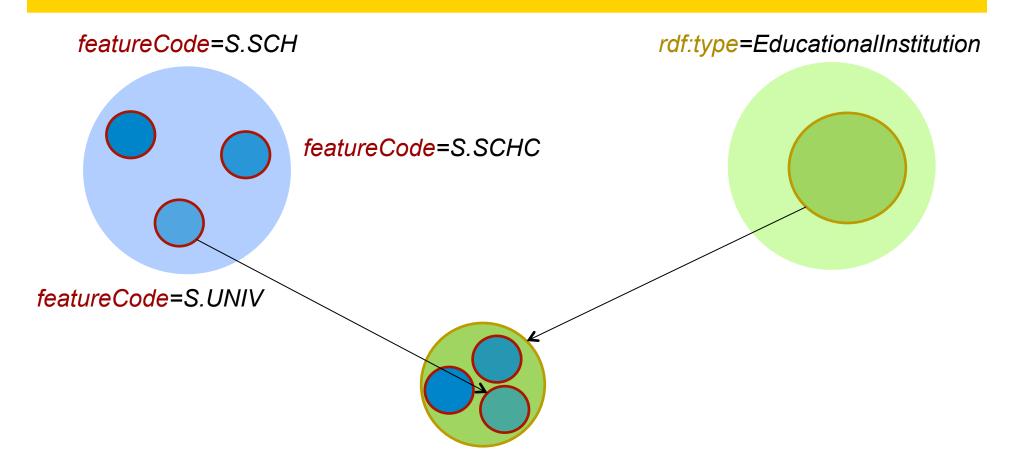
2) Colleges in *Geonames* are Educational Institutions in *DBpedia*







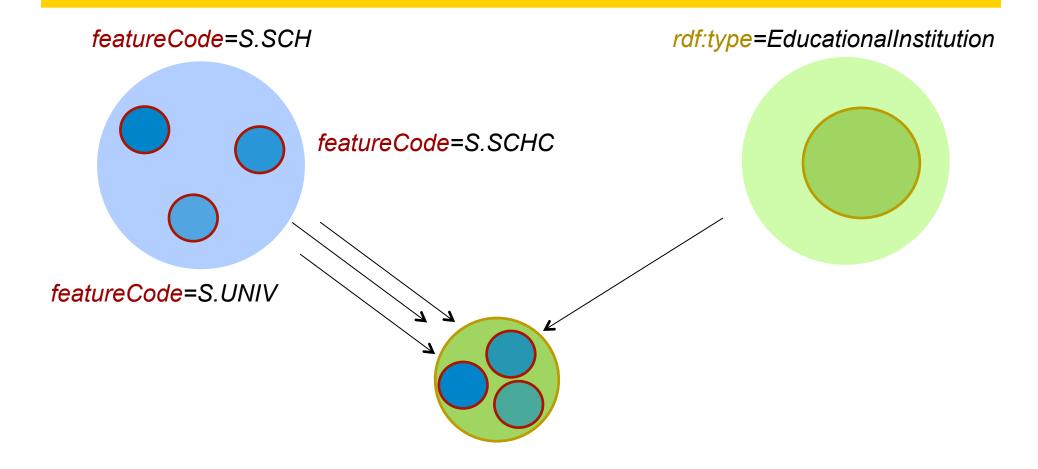
3) Universities in *Geonames* are Educational Institutions in *DBpedia*







Taken by themselves, the subset relations are not useful





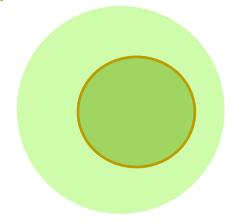


We use the common *featureCode* property as a hint...

featureCode=S.SCHC

featureCode=S.UNIV

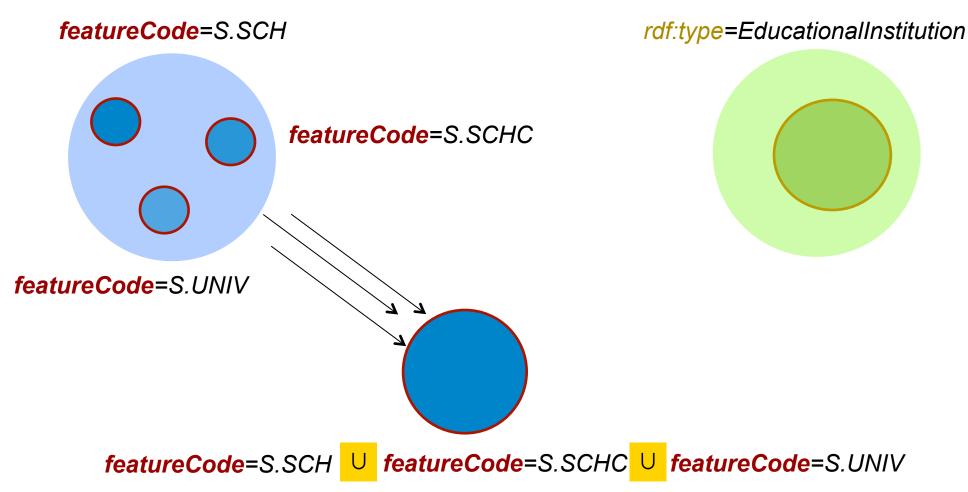
rdf:type=EducationalInstitution







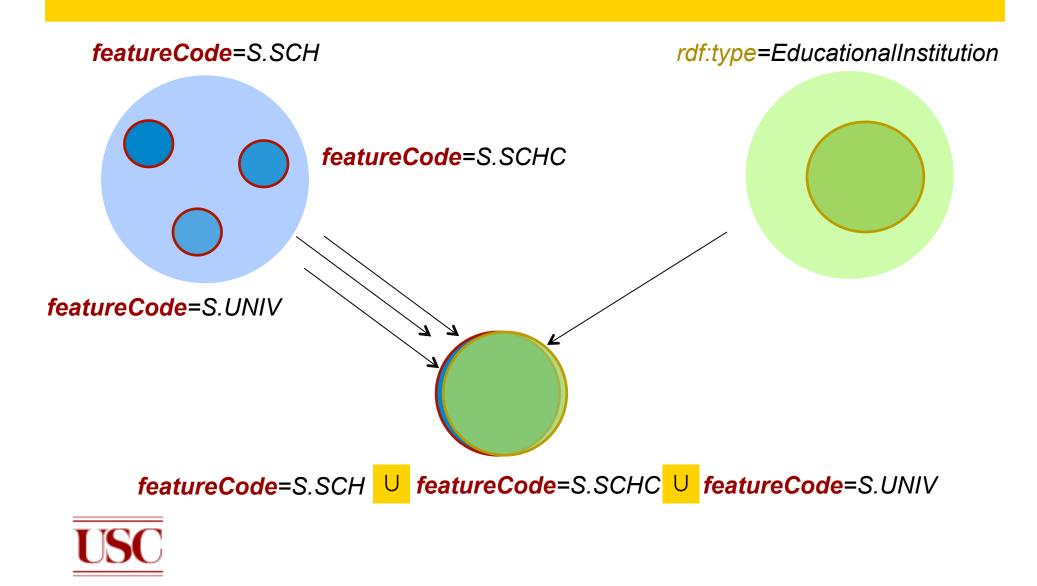
...to form a *Union* of Restriction Classes







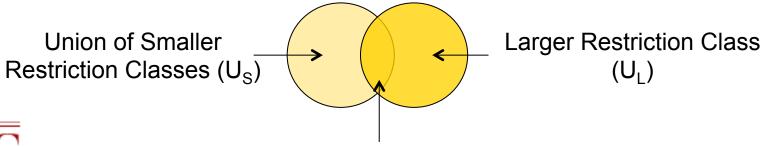
Contribution 1: Find Union Alignments





Finding Union Alignments: Approach

- For all alignments found in the ISWC2010 paper marked as subsets
 - We group all subset alignments according to the common larger restriction class
 - 2. We form a *union concept* such that all restriction classes
 - have the same property
 - have a single property-value pair each
 - 3. We then try to match the *union concept* to the larger class
 - 4. This forms a hypothesis *Union Alignment*

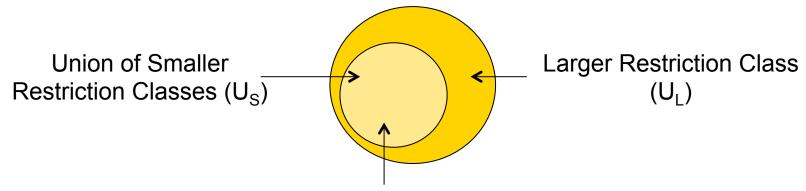




Intersection Set of Linked Instances $(U_A) = U_S \cap U_L$



Finding Union Alignments: Scoring



Intersection Set of Linked Instances $(U_A) = U_S \cap U_L$

$$\frac{\mid U_A \mid}{\mid U_S \mid}$$
 = 1 since by definition, all smaller classes are subsets

So, if
$$\frac{|U_A|}{|U_L|}$$
 = 1, then the larger class U_L is equivalent to U_S



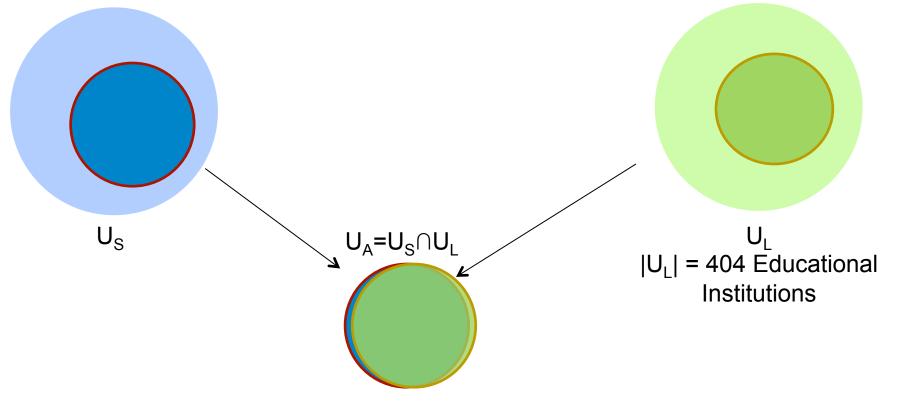
Practically, we use a relaxed subset assumption: $\frac{|U_A|}{|U_S|}$, $\frac{|U_A|}{|U_L|}$ >0.9



Contribution 1: Find Union Alignments



rdf:type=EducationalInstitution





$$\frac{\mid \mathsf{U}_{\mathsf{A}} \mid}{\mid \mathsf{U}_{\mathsf{S}} \mid} > 0.9$$

$$\frac{|U_A|}{|U_L|} = \frac{396}{404} = 0.98 > 0.9$$



What are the other 8 Educational Institutions?





Contribution 2: Find Outliers / Discrepancies

- We are also able to point out where the instances that disagree with the alignment lie
- These instances were not part of the alignment because
 - Their restriction class was not a subset (P'<0.9)
 - Some of these instances are
 - Linked Incorrectly with owl:sameAs
 - Assigned wrong value during RDF generation*
 - Common in both sets (could be debatable)
 - Did not have a minimum support size of 2 instances (set with 1 instance cannot be relied on)



 Outliers help in understanding discrepancies in the Linked Data



What are the other 8 Educational Institutions?

- 1 with featureCode=S.HSP (Hostpitals)
 - There are 31 instances with S.HSP because of which Hospitals are not subsets
- 3 with featureCode=S.BLDG (Buildings)
- 1 with featureCode=S.EST (Establishment)
- 1 with featureCode=S.LIBR (Library)
- 1 with featureCode=S.MUS (Museum)
- 1 doesn't have a featureCode property





RESULTS







Results: Geonames-DBpedia

Larger class from *DBpedia* and union of smaller classes from *Geonames*

#	Sub-group $\{p_1, v_1, p_2\}$	$List(v_2)$	$R'_U = \frac{ U_A }{ U_L }$	$ U_A $	$ U_L $	Outliers	# Explained Instances
1	{rdf:type,	S.SCH, S.SCHC,	0.9801	396	404	S.BLDG (3/122), S.EST (1/13),	403
	dbpedia:EducationalInstitution, geonames:featureCode}	S.UNIV				S.LIBR (1/7), S.HSP (1/31), S.MUS (1/43)	
2	$\{dbpedia: country,$	ES	0.9997	3917	3918	IT (1/7635)	3918
	dbpedia:Spain,						
	geonames:countryCode						
3	$\{dbpedia: region,$	geonames:2989247,	1.0	754	754		754
	dbpedia: Basse-Normandie,	geonames:2996268,					
	geonames:parentADM2	geonames:3029094					
4	$\{rdf:type,$	S.AIRB, S.AIRP	0.9924	1981	1996	S.AIRF (9/22), S.FRMT (1/5),	1996
	dbpedia:Airport,					S.SCH (1/404), S.STNB (2/5)	
	$geonames:featureCode\}$					S.STNM (1/36), T.HLL (1/61)	

Larger class from Geonames and union of smaller classes from DBpedia

#	Sub-group $\{p_1, v_1, p_2\}$	$List(v_2)$	$R'_U = \frac{ U_A }{ U_L }$	$ U_A $	$ U_L $	Outliers	# Explained Instances
5	{geonames:countryCode,	dbpedia:Netherlands,	0.9802	1939	1978	dbpedia:Kingdom_of	1940
	NL,	dbpedia:The_Netherlands,				_the_Netherlands	
	dbpedia:country}	dbpedia:Flag_of_the					
		$_{ m Netherlands.svg}$					
6	{geonames:countryCode,	dbpedia:Jordan	0.95	19	20		20
	JO,	dbpedia:Flag_of_Jordan.svg					
	$dbpedia:country\}$						



Results: LinkedGeoData-DBpedia

Larger class from DBpedia and union of smaller classes from LinkedGeoData

#	Sub-group $\{p_1, v_1, p_2\}$	$List(v_2)$	$R'_U = \frac{ U_A }{ U_L }$	$ U_A $	$ U_L $	Outliers	# Explained Instances
7	$\{dbpedia:bundesland,$	HOM, IGB, MZG,	0.93	46	49		46
	Saarland,	NK, SB, SLS,					
	$lgd: Open GeoDBLicense Plate Number\}$	VK, WND					
8	$\{rdf:type,$	lgd:Amenity, lgd:K2543,	0.9901	2609	2610		2609
	dbpedia: Educational Institution,	lgd:School, lgd:University,					
	rdf:type	lgd:WaterTower					

Larger class from LinkedGeoData and union of smaller classes from DBpedia

#	Sub-group $\{p_1, v_1, p_2\}$	$List(v_2)$	$R'_U = \frac{ U_A }{ U_L }$	$ U_A $	$ U_L $	Outliers	# Explained Instances
9	$\{lgd:gnisST_alpha,$	Atlantic, Burlington,	1.0	214	214		214
	NJ,	Cape May, Hudson,					
	dbpedia:subdivisionName	Hunterdon, Monmoth,					
		New Jersey, Ocean, Passaic					
10	{rdf:type,	dbpedia:Stream,	0.97	33	34	dbpedia:Place(1/94989)	34
	lgd:Waterway,	dbpedia:River					
	rdf:type						





Results: Summary

We find a total of 6595 Union Alignments

Source 1	Source 2	Larger Class from Source 1	Larger Class from Source 2	Total number of Union Alignments found
Geonames	DBpedia	434	318	752
LinkedGeoData	DBpedia	2746	3097	5843

Results also available at http://www.isi.edu/integration/data/UnionAlignments





Related Work

- BLOOMS, BLOOMS+ ([4][5] in paper)
 - Linked Open Data ontologies aligned with 'Proton'
 - Constructs a forest of concepts and computes structural similarity
 - Geonames Proton has "poor performance" because of small number and vague classes in Geonames
- Volker et al. ([8] in paper)
 - Statistical schema induction
 - Mines associativity rules from intermediate 'transaction datasets'
 - Develops OWL2 Axioms



Similarity Metrics on labels of classes





Conclusion and Future Work

Conclusion

- We were able to find *Union Alignments* in the Geospatial Domain
 - Find alignments where no direct equivalence was evident
 - Introduced a disjunction operator to restriction classes
- We were able to find Outliers
 - Help identify inconsistencies in the data

Future work

- Our algorithm is not limited to Geospatial domain. We would like to explore other domains
- Experimental comparison with other approaches
- Preliminary findings suggest patterns in properties like geonames:countryCode and dbpedia:country





Any questions?

THANK YOU



