



Lessons Learned in Building Linked Data for the American Art Collaborative

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Project Goals

- Launch the American Art Collaborative
 - Consortium of 14 American art museums
 - Explore the use of Linked Data to make their data available for research, education, and outreach
- Build 5* Linked Data for the museums
 - Map the data about artwork and artists to a common ontology
 - Link the data to other resources
 - Create/extend tools to support the construction of Linked Data
 - Create applications using the data

Outline

- Mapping the data
- Linking the entities
- Using the Linked Data
- Related Work & Discussion

Mapping the Data

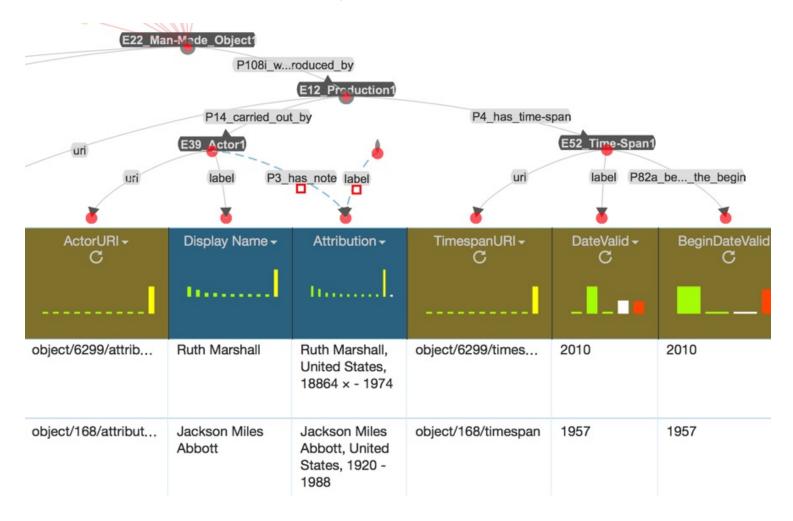
Challenges

- Museums have the data in wildly different formats and use different schemas
- The CIDOC-CRM ontology is a large and complicated ontology

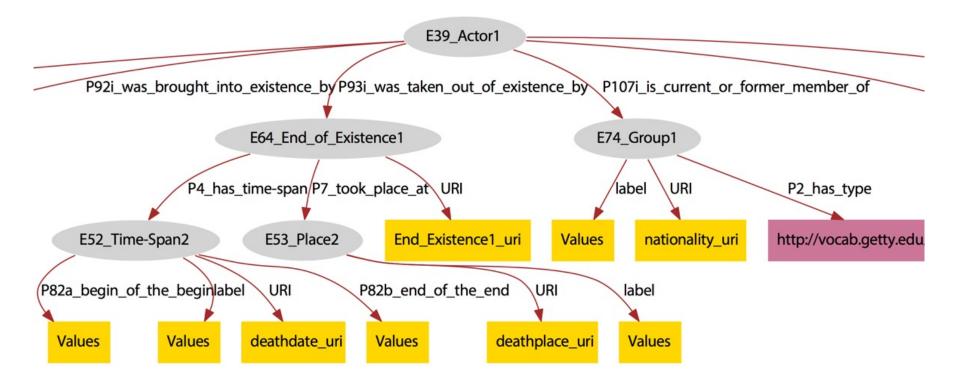
Approach

- Use Github to organize all of the data, mappings, and resulting RDF
- Use Karma to create the mappings of each dataset
- Trained a team of USC students to apply the tools to the datasets

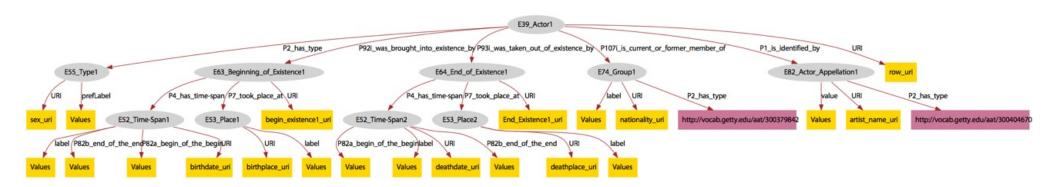
Use Karma to Map the Data to the Ontology



Example Model of Actor for Amon Carter



Complete Model of Actor for Amon Carter



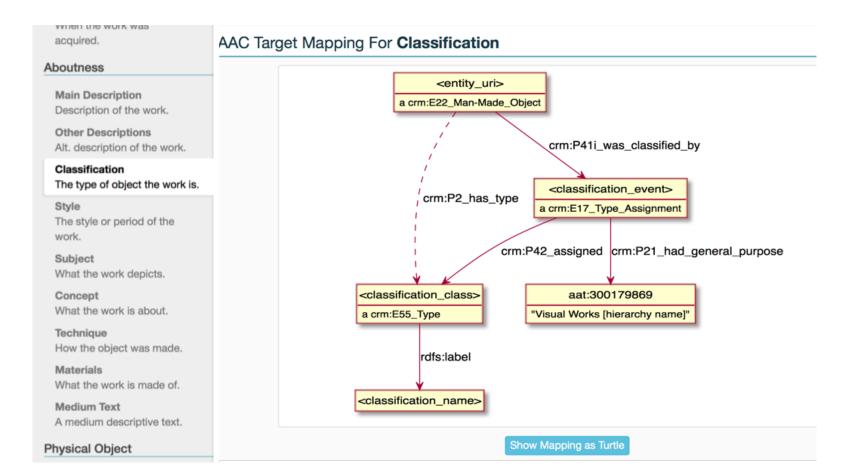
AAC Data Statistics

	Museum	Format	Files	Mappings	People	Commits	Issues
Archives of Amer	ican Art	xls	5	5	5	67	17
Amon Carter	Museum	xml	2	3	7	195	17
Autry	Museum	xlsx	6	6	9	309	68
Crystal Bridges	Museum	csv	8	14	7	572	76
Colby College Museu	m of Art	json	1	2	7	345	31
Dallas Museu	m of Art	csv	2	2	3	250	11
Gilcrease	Museum	xlsx	9	12	5	447	24
Indianapolis Museu	m of Art	json	3	3	6	214	16
National Museum of Wil	dlife Art	csv	2	3	6	196	9
National Portrai	t Gallery	xlsx	11	12	7	334	75
Princeton University Art	Museum	json	10	11	7	421	53
Smithsonian American Art	Museum	csv	11	14	4	408	49
Walters Art	Museum	xml	6	12	6	878	28
	Total	4	76	99		4,636	474

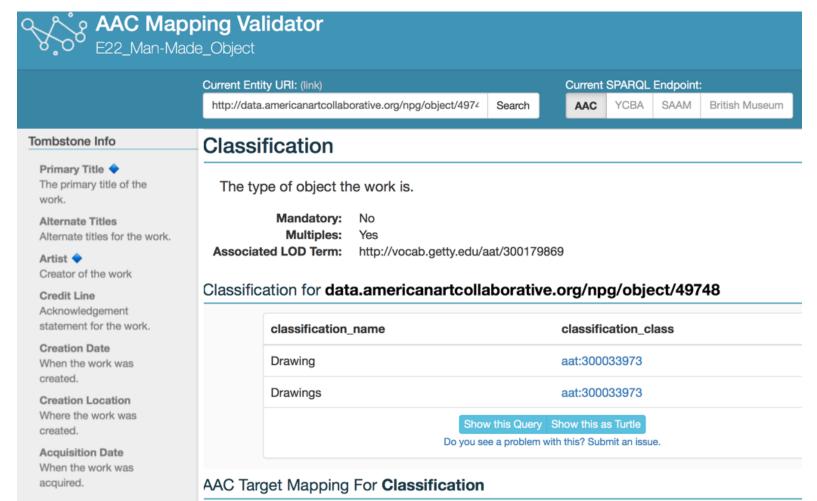
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AAC Target Mappings



AAC Mapping Validator



Statistics on the Mappings

	Data	Structure		Semantio	:
Museum	Trans.	Trans.	Classes	Types	Links
Archives of American Art	46	0	30	65	43
Amon Carter Museum	13	3	13	26	14
Autry Museum	76	0	46	87	49
Crystal Bridges Museum	112	6	74	132	89
Colby College Museum of Art	52	0	36	69	52
Dallas Museum of Art	46	0	27	55	39
Gilcrease Museum	105	5	75	132	109
Indianapolis Museum of Art	87	2	55	101	75
National Museum of Wildlife Art	37	0	24	47	34
National Portrait Gallery	112	2	64	118	69
Princeton University Art Museum	116	5	95	153	115
Smithsonian American Art Museum	88	4	67	114	95
Walters Art Museum	78	8	56	99	71
Total USC Informa	968	35	662	1,198	854

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Statistics on What Was Mapped

Museum	Constituents	Objects	Events	Places	Triples
Archives of American Art	6,944	15,025	7,301	1,592	210,360
Amon Carter Museum	806	6,421	13,164	532	225,528
Autry Museum	148	193	558	0	14,639
Crystal Bridges Museum	514	1,691	3,384	0	96,533
Colby College Museum of Art	2,210	8,217	18,905	0	456,711
Dallas Museum of Art	1,299	2,229	5,639	0	114,184
Gilcrease Museum	1,578	20,904	83,603	4,159	1,851,246
Indianapolis Museum of Art	2,131	22,314	34,560	432	846,952
National Museum of Wildlife Art	376	2,208	2,226	0	83,486
National Portrait Gallery	12,553	16,829	54,097	5,713	1,902,699
Princeton University Art Museum	2,899	13,314	43,828	881	1,253,239
Smithsonian American Art Museum	20,490	43,038	106,534	3,042	2,597,938
Walters Art Museum	182	801	1722	159	60,136
Total	52,130	153,184	375,521	16,510	9,713,651

Mapping Lessons

- Lesson 1: Reproducible Workflows
 - Allow museums to export raw data from their collection management systems
- Lesson 2: Shared Repository
 - Github was invaluable for managing all the data and mapping files
- Lesson 3: Data Cleaning
 - Significant data cleaning was required
 - Integrated as part of the data processing workflow in Karma
- Lesson 4: Mapping Inconsistencies
 - Validation tool was critical to completing a consistent set of mappings
- Lesson 5: Expert Review
 - The outside review was crucial in identifying and resolving mapping inconsistencies

Linking the Data

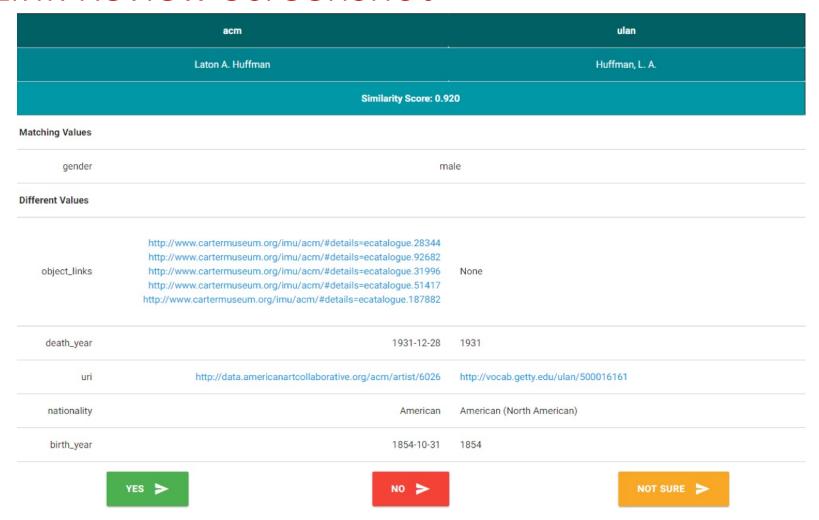
Goals

- Link the entities in the museum data to other resources.
- Capability for museums to curate the automatically generated links
- Demonstration: linking artists to Getty ULAN

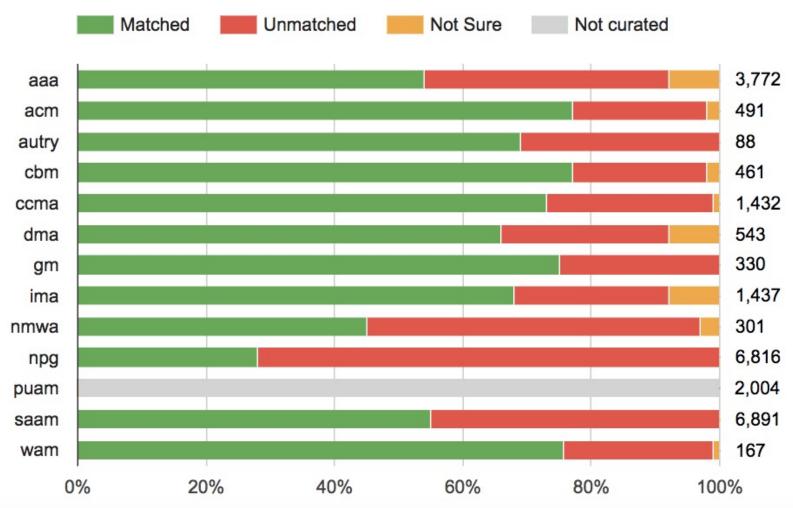
Approach

- Attempted to use existing linking tools, but they either didn't scale or students found them difficult to configure
- Wrote a specialized script to generate high recall & precision candidates
- Built a link review tool that the museums used to curate their links

Link Review Screenshot



Screenshot with Linking Review Results



Summary Statistics on the Linking Process

- Number of constituents in data: 42,685
- Previously existing links to ULAN: 3,349
- Linking based on previously existing links:
 - Precision: .96
 - Recall: .88
 - F1-measure: .92
- Candidate matches: 24,733
- Incorrect links in museums datasets: 19
- New links to ULAN: 9,357
- Incorrect links after review: 2
- Previously existing links not discovered: 136

Lessons Learned on Linking

Lesson 6: Linking Tools

- Difficult to configure and use the existing linking tools and get them to scale to large datasets (e.g., ULAN, DBPedia, & VIAF)
- We need easy to work with and scalable libraries for linking tasks

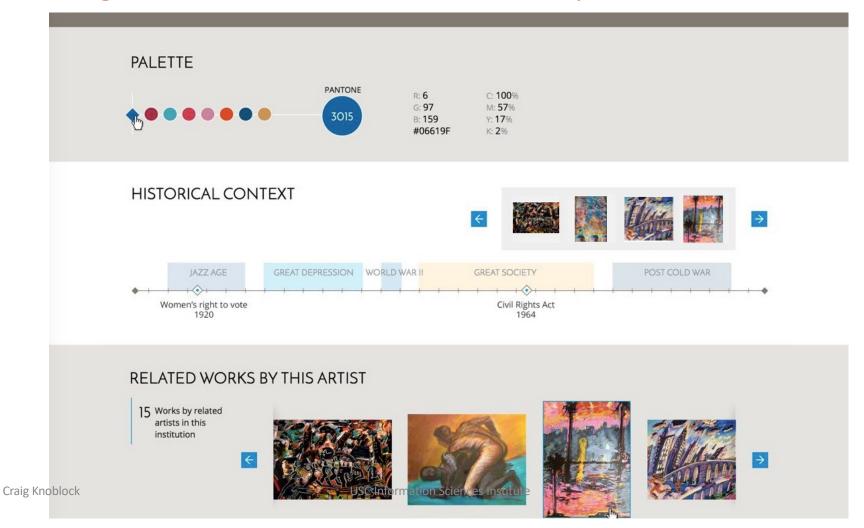
Lesson 7: Manual Review

- Users are willing to invest significant time and energy to ensure the final data is accurate
- The museums reviewed almost 25K links!
- A few weeks of effort almost tripled the number of links to ULAN

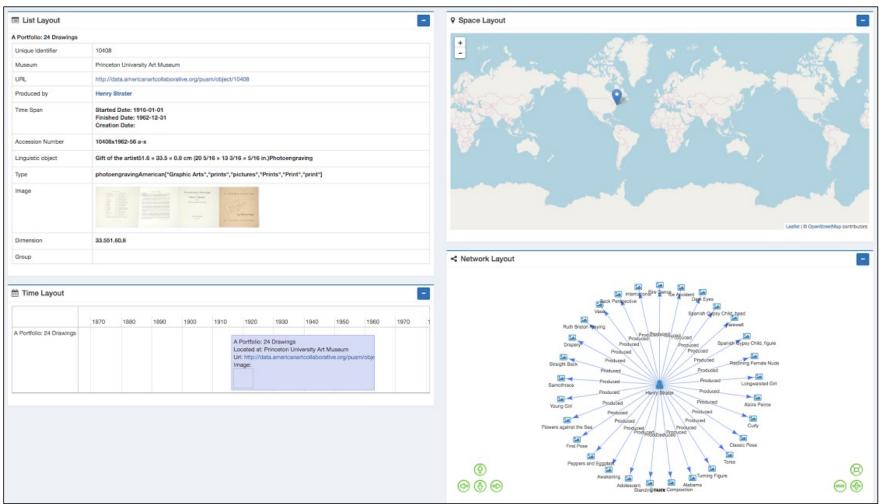
Using the Data: The Browse Application



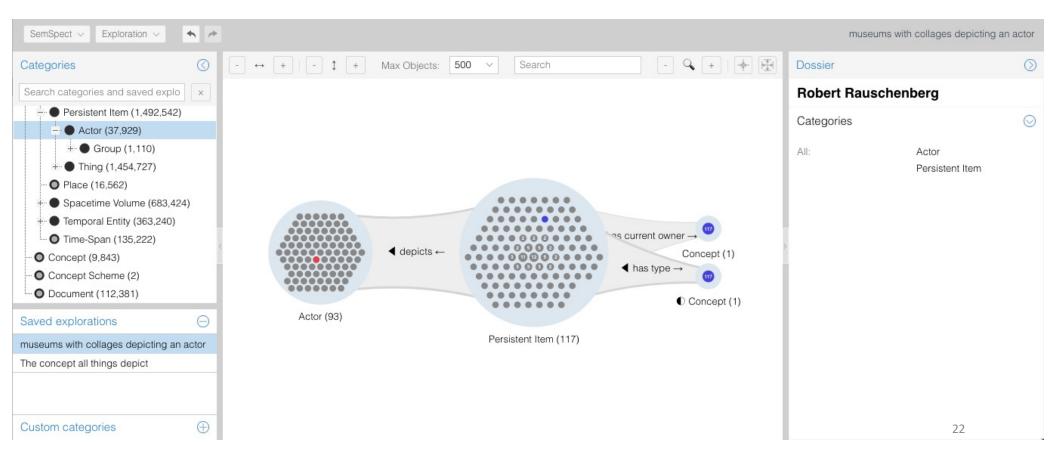
Using the Data: Under Development



SemUI Visualization [Giunchiglia, Ojha, & Das, ICSC 2017]



SemSpect (http://aac.semspect.de/) Thorsten Liebig (derivo)



Lessons Learned in Using the Data

- Lesson 8: Visualization
 - Easy to understand visualization is needed for non-technical users
- Lesson 9: Simple Schema
 - CRM ontology may be useful for research, but challenging for applications
 - Created a set of SPARQL queries to create JSON objects
 - Loaded the objects into Elasticsearch for complex analysis

Related Work

Consortiums of museums

- Europeana 1500 cultural heritage institutions, 17 million items
- CHIN 8 Canadian museums, 85,000 items
- LODAC 114 museums in Japan
- Published a fixed schema and mapped all museums/institutions to the schema

Organizations using the CRM ontology

- Research space (British museum & Yale Center for British Art)
- Pharos 14 historical photo archives
- Each organization is responsible for publishing their own data to CRM

Mapping data to CRM

X3ML – maps XML data to CRM using manually written rules

Linking

- Silk, Dedup, etc. focus is on automatic linking, but no curation of the links
- Mix'n'match, OpenRefine support link review, but targets highly technical users

Discussion

- Collaborated with 14 American art museums to publish 5* Linked Data
- Created a set of tools to create the linked data
 - Karma clean and map the data, publish directly to Gitub
 - Mapping Validation tool review the mappings to ensure consistency
 - Karma execution tool applies Karma mappings and published both RDF and JSON-LD directly to Github
 - Link Review tool allows non-technical users to quickly and easily review links to other sources
 - Browse application allows museum staff, art historians, and the general public to verify and explore the data

Success?

- 14 additional museums have now released their data as linked data
- Three museums have already learned how to use the tools to create their own mappings
 - Indianapolis Museum of Art
 - Smithsonian Archives of American Art
 - Colby College Museum of Art
- Researchers outside the project have applied their visualization tools to the data
 - Sajan Raj Ojha (Univ. of Trento): SemUI
 - Thorsten Liebig (derivo): SemSpect http://aac.semspect.de/

Future Work

- Automate the addition of new museums to the AAC
 - Gather, map, and link the data directly form their online web pages
- Extend the types of information supported
 - E.g., exhibition data & bibliographies
 - Improve the ability in Karma to automate complex mappings
- Link the existing data to other sources
 - E.g., VIAF, Geonames, & DBpedia
 - Build a library of linking functions to support easy and scalable linking

More Info

Karma: karma.isi.edu

AAC: americanartcollaborative.org

Github: github.com/american-art

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Thanks!