

The Semantic Data Dictionary Approach to Data Annotation & Integration

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Motivation

- Standard Data Dictionary
 - A controlled vocabulary [Linnarsson and Wigertz, 1989]
 - Human readable
 - Difficult for a machine to understand
 - Integration tasks are not easily automated
- Semantic Data Dictionary (SDD) Specification
 - Allows for integration of data from multiple domains
 - Uses a common metadata standard
 - Can leverage domain-relevant terminology
 - Helps with the specification, curation and search of data
 - "Compositional" descriptions, rather than 1:1 mappings



Introduction

Some challenges a scientist may face include

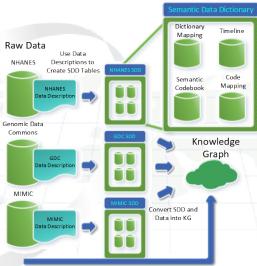
- Finding data across data resources with the same semantic meaning
- Annotation of data without a vast knowledge of ontologies
- A single row in a dataset may contain reference on multiple entities

The Semantic Data Dictionary Specification

- Facilitates finding data that are relevant for comparison
- Creates a more accessible workflow



SDD Annotation Flow Diagram



Use Data to assign values to Knowledge Assertions

Related Work

Data Integration - The ability to unite data from multiple sources in such a way that results in a unified view of the combined data [Lenzerini, 2002]

- Use of ontologies to annotate data is increasing
- Adhering to a common foundational model such as SIO or the OBO Foundry ontologies facilitates data integration

Schema Merging - An approach to integrate information from multiple datasets

- Two general methods commonly used [Buneman et al., 1992]
- Considerations when performing schema merging [McBrien and Poulovassilis, 1998]
- Methods, like CrowdMap [Sarasua et al., 2012], may take advantage of crowd sourcing

Semantic Annotation - The practice of assigning metadata descriptions about entities in a database or text

The Semantic Data Dictionary Specification

The Semantic Data Dictionary Specification - A way to represent implicit entities and their relationships using a general ontology

- Semanticscience Integrated Ontology (SIO)
 - High-level conceptualization of data
 - Three entity world view: attributes, objects, and processes
 [Dumontier et al., 2014]
 - General properties to describe the relationships
- Can use domain-specific ontologies to allow more fine-grained and dataset-specific annotations
- Contains information about the entity types represented in a tabular dataset
- Conveys relationships in both machine-readable and unambiguous manners



Dictionary Mapping Example (Actual)

Table 1: Example Dictionary Mapping (Actual Entry)

Attribute	attributeOf	Unit	Time	inRelationTo
sio:Identifier	??child			
sio:Race	??mother			
sio:Age	??mother	sio:Year	??visit1	
chear:EducationLevel	??mother		??visit1	
chear:BMI	??mother	kgm2	??visit1	
chear:Weight	??mother	kg	??visit1	
sio:Height	??mother	cm	??visit1	
chear:SmokingStatus	??mother		??pregn	
sio:Concentration	??pb_1	mgL	??visit1	??sample1
sio:Concentration	??pb_2	mgL	??visit2	??sample2
chear:GestationalAge	??child	sio:Week	??birth	1
chear:Weight	??child	kg	??birth	
	sio:Race sio:Age chear:EducationLevel chear:BMI chear:Weight sio:Height chear:SmokingStatus sio:Concentration sio:Concentration chear:GestationalAge	sio:Identifier ??child sio:Race ??mother sio:Age ??mother chear:EducationLevel ??mother chear:BMI ??mother chear:Weight ??mother sio:Height ??mother chear:SmokingStatus ??mother sio:Concentration ??pb_1 sio:Concentration ??pb_2 chear:GestationalAge ??child	sio:Identifier ??child sio:Race ??mother sio:Age ??mother sio:Year chear:EducationLevel ??mother chear:BMI ??mother kgm2 chear:Weight ??mother kg sio:Height ??mother cm chear:SmokingStatus ??mother sio:Concentration ??pb_1 mgL sio:Concentration ??pb_2 mgL chear:GestationalAge ??child sio:Week	sio:Identifier ??child sio:Race ??mother sio:Age ??mother sio:Year ??visit1 chear:EducationLevel ??mother kgm2 ??visit1 chear:Weight ??mother kg ??visit1 sio:Height ??mother cm ??visit1 chear:SmokingStatus ??mother cm ??visit1 sio:Concentration ??pb_1 mgL ??visit1 sio:Concentration ??pb_2 mgL ??visit2 chear:GestationalAge ??child sio:Week ??birth

Dictionary Mapping Example (virtual)

Table 2: Example Dictionary Mapping (Virtual Entry)

Column	Entity	Role	Relation	inRelationTo	wasDerivedFron
??mother	sio:Human	chear:Mother		??child	1 / 1
??child ??birth	sio:Human chear:Birth	chear:Child		??mother ??child	
??sample1	S				??mother
??sample2	S				??mother
??pb_1	Pb		sio:isPartOf	??sample1	
??pb_2	Pb		sio:isPartOf	??sample2	

The SDD Specification Table

Table 3: Semantic Data Dictionary Mappings

Column	Related Property	Description
Column		Column header
Label	rdfs:label	Label for the column
Comment	rdfs:comment	Comment for the column
Definition	skos:definition	Text column definition
Attribute	rdf:type	URI of the attribute type
attributeOf	sio:isAttributeOf	Entity having the attribute
Unit	sio:hasUnit	Unit of Measure for attribute
Time	sio:measuredAt	Time point attribute was measured
Entity	rdf:type	Type of the entity
Role	sio:hasRole	Type of the role the entity plays
inRelationTo	sio:inRelationTo	Entity that the role is linked to
wasDerivedFrom wasGeneratedBy	prov:wasDerivedFrom prov:wasGeneratedBy	Entity from which the attribute was derived Activity from which the attribute was produce

Codebook Code Mapping Tables

Table 4: Codebook and Code Mapping Tables

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hnicalSchool

code	uri	label
Pb	chebi:25016	Lead
S	uberon:0001977	Serum
cm	obo:UO_0000015	centimeter
kg	obo:UO_0000009	kilogram
kgm2	obo:UO_0000086	kilogram per square meter
mgL	obo:UO_0000273	milligrams per liter

RDF Knowledge Entry (CHEAR)

```
:birthweight rdf:type chear:Weight;
   sio:isAttributeOf :joe;
   sio:hasValue 3;
   sio:hasUnit uo:kilogram;
   sio:existsAt [ a sio:TimeInterval, chear:BirthTime;
       sio:hasValue 0:
       sio:hasUnit sio:Day;
       sio:inRelationTo :birth ];
   sio:existsAt [ a sio:TimeInterval, chear:GregorianTime;
        sio:hasValue "2016-03-12"^^xsd:dateTime:
        sio:hasUnit sio:Day;
       sio:inRelationTo:birth].
```

Smoking Status Query

```
SELECT DISTINCT * WHERE {
?s rdf:type ?attribute .
OPTIONAL {?s sio:existsAt ?timepoint . }
?s sio:isAttributeOf ?attrOf .
?s sio:hasValue ?attributeVal .
OPTIONAL { ?cohort rdf:type hasco:Cohort .
      ?cohort sio:isAttributeOf ?attrOf .
      ?cohort sio:hasValue ?cohortVal . }
OPTIONAL {?s sio:hasUnit ?unit . }
FILTER (?attributeVal != 'NA') .
FILTER (?attribute = chear:SmokingStatus )
LIMIT 10000
```

Evaluation

- Approach was applied to the National Health and Nutrition Survey (NHANES) data from 2013-2014
 - Able to generate SDD and Codebook starting points
 - 150 documents
 - 6 categories
 - 4818 Dictionary Mapping rows
 - Over 17000 codebook entries
 - Of the 4818 SDD rows, 1148 or 23.83% were mapped to existing concepts in SIO or CHEAR
 - The remaining rows were not mapped to any concepts due to limitations in the extraction algorithm
 - Values for columns were populated through a web scraping script
 - A look-up approach was used to compare NHANES labels with terms in SIO or CHEAR

Discussion

- Human input is still necessary to complete the annotation
- Ongoing effort to manually annotate the remaining NHANES concepts
- The SDD specification is being applied to additional publicly available datasets
- Knowledge Graphs have been created for the subset of NHANES that had been annotated
- Actively used in a Data Analytics course at RPI to demonstrate to students how semantics can be leveraged to perform analytics

Conclusion

- Provides a formal means to annotate dataset columns
- Allows production of OWL-based metadata
- Helps address Semantic Web goal of Interoperability
 - Single conceptual structure
 - Comparison to any other dataset that has also been mapped
- Domain scientists can produce high quality integrated data

Projects -

- Children's Health Exposure Analysis Resource (CHEAR)
- The Center for Architecture Science and Ecology (CASE)
- The Healthy Birth, Growth, and Development (HBGD)
- The Center for Health Empowerment by Analytics, Learning, and Semantics (HEALS)

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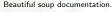


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Questions?

