# Semantic Web Technologies

# Topic: Data Cleaning

**Olaf Hartig** 

olaf.hartig@liu.se





## Terminology and Methodologies

 Data cleaning (data cleansing, data scrubbing) "deals with detecting and removing errors and inconsistencies from data in order to improve the quality of data."

[Rahm and Do 2000]

- There are a number of methodologies, for instance:
  - 1. Audit the data to identify quality issues
  - 2. Choose methods to automatically detect and remove the issues
  - 3. Apply the methods
  - 4. Post-processing / control step [Müller and Freytag 2003]

Rahm and Do: Data Cleaning: Problems and Current Approaches. IEEE Data Eng. Bull. 23(4): 3-13, 2000

Müller and Freytag: *Problems, Methods, and Challenges in Comprehensive Data Cleansing*. Technical Report, Humboldt-Universität zu Berlin, HUB-IB-164, 2003.



# **Data Quality**



#### More Terminology

- Data quality: commonly understood as "fitness for use" for a particular application or use case
  - Hence, even a dataset with quality issues may be fully useful for use cases not affected by the issue
- Data quality assessment: process of measuring the quality of some data and, ultimately, identifying whether the data is fit for use
- Data quality dimensions: accuracy, timeliness, completeness, relevancy, objectivity, believability, understandability, consistency, conciseness, etc.
  - Different authors consider different dimensions under different names, and group them into different groups



# Data Quality Dimensions (with a Focus on Semantic Web Data)

Accessibility Representation Rep.-Security\* Interpretability Availability Aspects of Conciseness access, authenticity, Performance\* Interlinking\* Versatility\* Interoperability and retrieval Licensing\* of data Intrinsic Contextual Syntactic Conciseness Validity Relevancy Depend on Trustworthiness Consistency the context Understandability of the task Semantic Completeness **Timeliness** at hand Accuracy Two dimensions Dim2 Dim1 are related



#### Intrinsic Dimensions

- Aspects that are independent of the user's context
- Syntactic validity: degree to which a file conforms to the specification of the serialization format
- Semantic accuracy: degree to which data values correctly represent the real world facts
- Consistency: degree to which there are no logical contradictions w.r.t. the knowledge representation
- Conciseness: degree to which there is no redundancy of entities at the schema level and the data level
- Completeness: degree to which all required information is present in the data



#### Intrinsic Dimensions

- Aspects that are independent of the user's context
- Syntactic validity: degree to which a file conforms to the specification of the serialization format
- Possible metrics for syntactic validity:
  - No syntax errors in the file
- Co conformance to a given schema)
- No malformed datatype literals
   of Emmes at the semental level and the semental server.
- Completeness: degree to which all required information is present in the data



#### Representational Dimensions

- Capture aspects related to the design of the data
- Representational-conciseness: degree to which the representation of the data is compact and well formatted
- Interoperability: degree to which the format and structure conforms to previously returned data and to data from other sources
- Interpretability: degree to which data is represented using appropriate notation and whether the machine is able to process the data
- Versatility: availability of the data in different representations and in an internationalized way



## **Tools**



### Goal of Data Cleaning

- Fix data quality issues in given sets of (semantic) data
- Such quality issues may ...
  - ... be in source datasets (e.g., inaccurate or wrong data items, outdated data items)
  - ... result from imperfections of a data integration process (e.g., data items that have been incorrectly linked with each other)
  - ... reveal themselves only after the data integration (e.g., duplicates, inconsistencies)
- Hence, data cleaning may be relevant both for
  - original datasets before combining/integrating, and
  - datasets resulting from an integration



#### **Options**

- Tools that allow users to identify quality issues (e.g., by highlighting outliers or similarities)
- Tools that identify quality issues (semi-)automatically
- Tools that fix theses issues in an automated process





#### **RDFUnit**

http://rdfunit.aksw.org/

- **S**Unit
- Test driven data-debugging framework
- Test cases are executed as SPARQL queries using a pattern-based transformation approach



## RDFUnit (cont'd)

http://rdfunit.aksw.org/

- **S**Unit
- Test driven data-debugging framework
- Test cases are executed as SPARQL queries using a pattern-based transformation approach
- Test cases that can be created manually, or generated automatically (based on a schema)
  - Supported schemas: OWL, SHACL, IBM Resource
     Shapes, Dublin Core Set Profiles
- Tested data loaded from a specified file or accessed via a SPARQL endpoint
- Report of a test suite can be obtained as an HTML page, but also as RDF data



## RDFUnit (cont'd)

#### Testing

Run tests

Completed! (S: 0 / F:13 / T: 28 / E: 24603)

Cancel

est	Results			
S	Test	Errors	Prevalence	
F	http://databugger.aksw.org/tests#foaf-INVFUNC-0a77ce81bec99608d28790eb695d11fa	<u>25</u>	-1	
F	http://databugger.aksw.org/tests#foaf-INVFUNC-105e1374ad211491979c95caa27ba2f5	53	786	
	http://databugger.aksw.org/tests#foaf-INVFUNC-11eb481f2e37c9e1fd18066d637bc013	-	-	
F	http://databugger.aksw.org/tests#foaf-INVFUNC-18fb0cf9dc8ff9ad9d42982e0434db2c	476	1214	
F	http://databugger.aksw.org/tests#foaf-INVFUNC-2e2b3b0e569d5316d760bdf30f9ecf48	<u>34</u>	87	
F	http://databugger.aksw.org/tests#foaf-INVFUNC-4fe77a880206d4b9a00b9972176043b1	84	244	
F	http://databugger.aksw.org/tests#foaf-INVFUNC-58e73e30a1082f24e75ecb7c394415d9	21219	366471	
F	http://databugger.aksw.org/tests#foaf-INVFUNC-9e12004a97dd6757449f9a1acf86b2a0	165	482	
	http://databugger.aksw.org/tests#foaf-INVFUNC-a81976fee7973a3c722c1cedc2ede84f	-	-	
F	http://databugger.aksw.org/tests#foaf-INVFUNC- b009723769eb05dcb5d67594816a6dba	<u>69</u>	168	
F	http://databugger.aksw.org/tests#foaf-INVFUNC- b6b5b018064e92966bd79a6648b369a7	2474	21301	
	http://databugger.aksw.org/tests#foaf-INVFUNC-ece13a3f9c3919a10d56b18599412cc0	-	-	
F	http://databugger.aksw.org/tests#foaf-OWLCARD-0cab7cf9453873d6fdd60fac66544246	1	7566	
F	http://databugger.aksw.org/tests#foaf-OWLCARD- 28319b6c1b670d59d90438819fe7e3b4	1	484	



#### Sieve

- Uses metadata to assess data quality of RDF datasets and to filter the data http://sieve.wbsg.de/
- Input:
  - a dataset, given as a set of Named Graphs
  - provenance data associated with these graphs
- Main functionality:
  - computes various, configurable quality scores for the graphs (based on the provenance data)
  - these scores are represented as RDF data
- Data fusion component
  - merges parts of the data of the Named Graphs
  - filters out some data based on the quality scores



#### Sieve Configuration Example

```
<QualityAssessment name="Recent and Reputable is Best">
    <AssessmentMetric id="sieve:reputation">
        <ScoringFunction class="ScoredList">
           <Param name="list"
                  value="http://en.wikipedia.org
                         http://es.wikipedia.org
                         http://fr.wikipedia.org"/>
        </scoringFunction>
    </AssessmentMetric>
    <AssessmentMetric id="sieve:recency">
        <ScoringFunction class="TimeCloseness">
            <Param name="timeSpan" value="50000"/>
            <Input path="?GRAPH/ldif:lastUpdate"/>
        </ScoringFunction>
    </AssessmentMetric>
</QualityAssessment>
```



## Generic "Data Wrangling" Tools

"Data wrangling is the process of taking data in its native format and making it usable for analysis." —https://www.trifacta.com/

- OpenRefine (formerly Google Refine, open source)
  - http://openrefine.org/



TRIFACTA Wrangler

- Trifacta Data Wrangler (commercial)
  - https://www.trifacta.com/products/wrangler/
- Tamr (commercial)
  - http://www.tamr.com/product/





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