

Figure: CMOP's "Virtual Columbia River"

The Problem of Semantics in the Metadata Mess

V.M. Megler
David Maier
Portland State University

With thanks to the scientists at Center for Coastal Margin Observation and Prediction (CMOP). This work is supported by NSF award OCE-0424602.

Agenda

- Our “Big Data” Search Engine
- The Metadata Mess
- Reducing Semantic Diversity
- “Metadata Wrangling”
- Current State

Our “Big Data” Search Engine

- Problem: finding relevant data in a “big data” archive
 - Many datasets, dataset shapes and sizes, physical locations, formats, tools (Megler and Maier, 2011; 2012; 2013)
 - Example information need:
“observations collected near [lat = 45.5, lon = -124.4] in mid-2010, with temperature between 5-10C”
- Solution: a data search engine that operates over big data archives

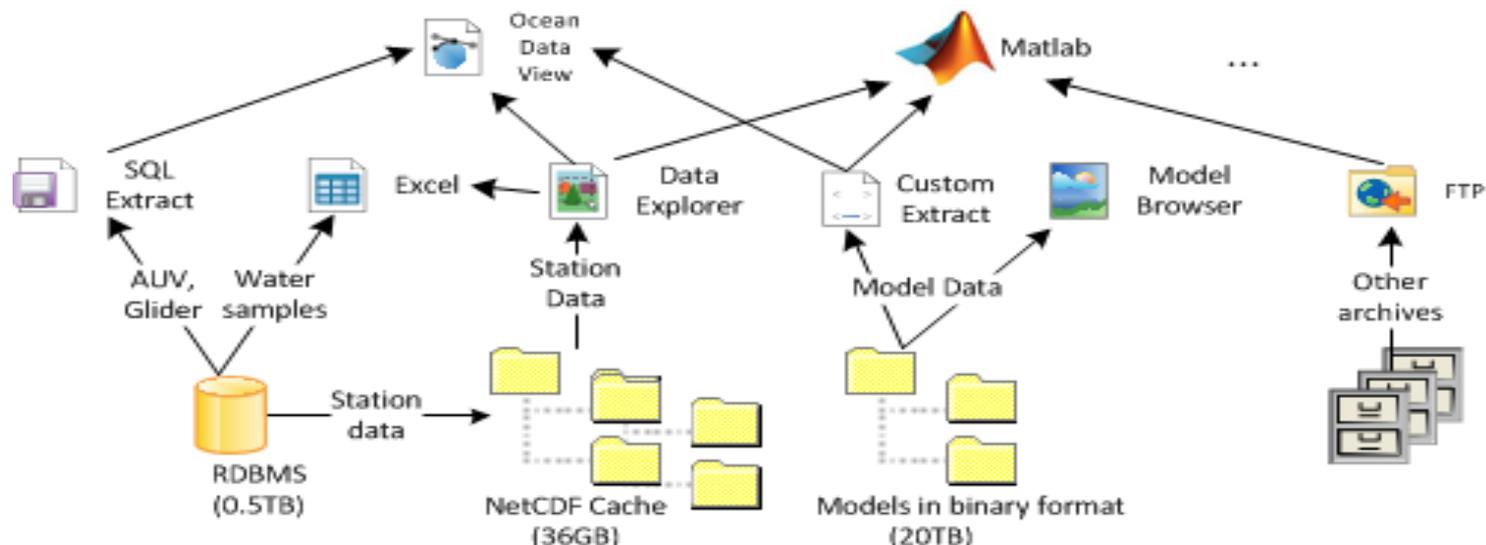


Figure: Heterogeneity of Data Formats and Data Access Tools in One Scientific Archive

IR Architecture Adapted to Scientific Data Search

- Approach:
 1. Scan (heterogeneous) data; extract summary features
 2. Search over features, with real-time response
 - Return ranked results, with links to data and tools

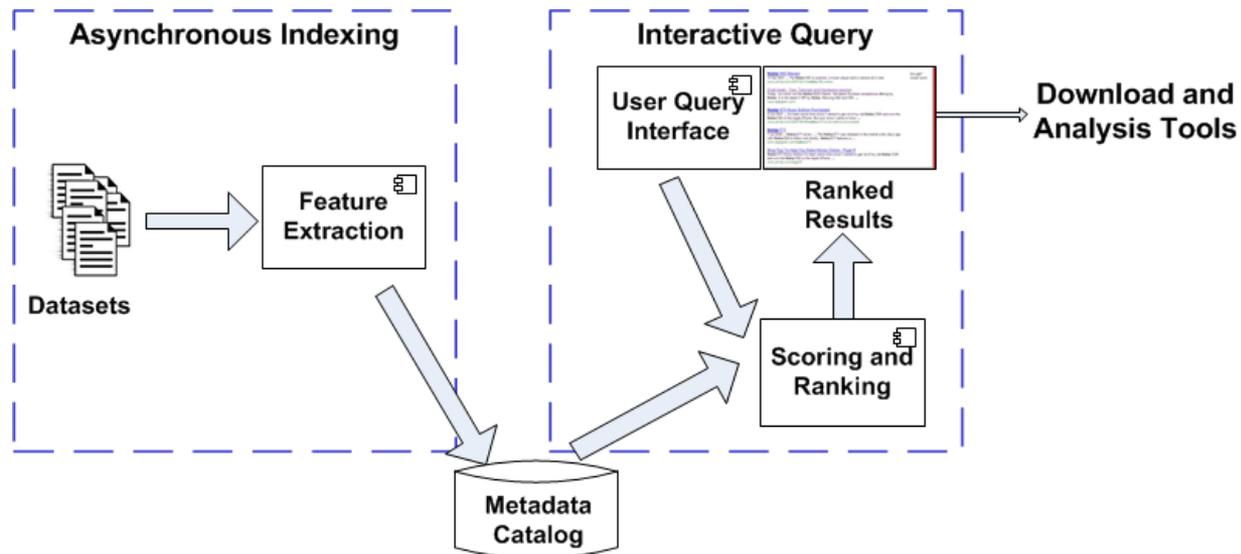


Figure: Information Retrieval Architecture, adapted to data search
(from Megler and Maier, 2012)

Ranked Search Over Data: Location, Time, Variables

Data Near Here V0.6 (Research Edition)

Please enter the following parameters:

Categories: Quality:

SW Corner: NE Corner:
[dec.deg] [dec.deg]

Depth: from [m] Depth to: [m]

Start date: End date:

with variable: Range: - Units:

Min. Obs. Count:

There were 24 results returned; all are listed, and 24 initially shown on map. Temp was found in 24 entries.

Display	Type	Collection	Quality	Start Time	End Time	From Depth	To Depth	temp	Observations	Data Location	Score	DNH
<input checked="" type="checkbox"/>	Cruise	Cruise, May-June 2010, Wecoma, 2010-07-16, Segment 3	preliminary	2010-07-16 05:16 PDT	2010-07-16 05:29 PDT	-5	-5	9.89:12.14 c	14	Download	98	<input type="button" value="DNH"/>
<input checked="" type="checkbox"/>	Cruise	Cruise, April 2010, Wecoma, 2010-04-17, Segment 4	preliminary	2010-04-17 04:06 PDT	2010-04-17 04:26 PDT	-5	-5	10.60:10.85 c	21	Download	97	<input type="button" value="DNH"/>
<input checked="" type="checkbox"/>	Cruise	Cruise, April 2010, Wecoma, 2010-04-17, Segment 11	preliminary	2010-04-17 18:52 PDT	2010-04-17 23:59 PDT	-5	-5	10.88:11.21 c	244	Download	96	<input type="button" value="DNH"/>
<input checked="" type="checkbox"/>	Cruise	Cruise, April 2010, Wecoma, 2010-04-18, Segment 1	preliminary	2010-04-18 00:00 PDT	2010-04-18 01:15 PDT	-5	-5	10.88:11.07 c	77	Download	96	<input type="button" value="DNH"/>

Figure: "Data Near Here" Search Interface (from Megler & Maier, 2011)

Detailed Search Result: Variable Information

Data Near Here V0.5: Dataset Details

Dataset Summary

Agency	Center for Coastal Margin Observation and Prediction		
Description	Forerunner Daily, Forerunner, 2009-05-28		
Type	Cruise		
Data Format	CSV		
Quality	raw_data		
Time: Start		2009-05-28 08:05 PDT	
Time: End		2009-05-28 16:05 PDT	
Depth: Min		0m (free surface)	
Depth: Max		0m (free surface)	
# of Values		2,775	
Data Location	Download		
Last Updated	2011-12-01 08:12 PST		



- Search result leads to “dataset summary”
- Displays dataset variable information from [metadata catalog](#)
 - Features produced via one-time scan per dataset

[Click here for this dataset's parent.](#)

Variables

Variable	Description	Units	Datatype	Minimum	Maximum	Number
conductivity		unknown	double precision			2,774
salinity		unknown	double precision			2,774
temperature		c	double precision			2,774

Variables

Variable	Description	Units	Datatype	Minimum	Maximum	Number
conductivity		unknown	double precision	0	0.32	2,774
salinity		unknown	double precision	0.06	26.54	2,774
temperature		c	double precision	12.23	18.02	2,774

Additional Information

This entry has a next level of detail available in the [parent dataset](#).

- [Forerunner Daily, Forerunner, 2009-05-28, Segment 1](#): with count 92
- [Forerunner Daily, Forerunner, 2009-05-28, Segment 10](#): with count 201
- [Forerunner Daily, Forerunner, 2009-05-28, Segment 11](#): with count 234
- [Forerunner Daily, Forerunner, 2009-05-28, Segment 12](#): with count 212
- [Forerunner Daily, Forerunner, 2009-05-28, Segment 13](#): with count 3
- [Forerunner Daily, Forerunner, 2009-05-28, Segment 14](#): with count 138
- [Forerunner Daily, Forerunner, 2009-05-28, Segment 15](#): with count 127
- [Forerunner Daily, Forerunner, 2009-05-28, Segment 16](#): with count 172
- [Forerunner Daily, Forerunner, 2009-05-28, Segment 2](#): with count 94
- [Forerunner Daily, Forerunner, 2009-05-28, Segment 3](#): with count 117
- [Forerunner Daily, Forerunner, 2009-05-28, Segment 4](#): with count 147
- [Forerunner Daily, Forerunner, 2009-05-28, Segment 5](#): with count 169
- [Forerunner Daily, Forerunner, 2009-05-28, Segment 6](#): with count 161
- [Forerunner Daily, Forerunner, 2009-05-28, Segment 7](#): with count 161
- [Forerunner Daily, Forerunner, 2009-05-28, Segment 8](#): with count 161
- [Forerunner Daily, Forerunner, 2009-05-28, Segment 9](#): with count 161

Motivation for This Work

Emerging problem: Many names for same environmental variable*

- E.g.: temperature, temp, water_temperature
- “Semantic diversity”
- Similar problems in other areas, e.g. variable units

* Hereafter called “variables”

The Metadata Mess

- Working assumption: each named column in a (publicly available) file / dataset represents a valid variable
- Result: Ever increasing number of variables (over 300 at CMOP)
- Problem:
 - Hard for searchers to navigate, locate desired variable
 - Not what the archive wants to expose – “metadata mess”

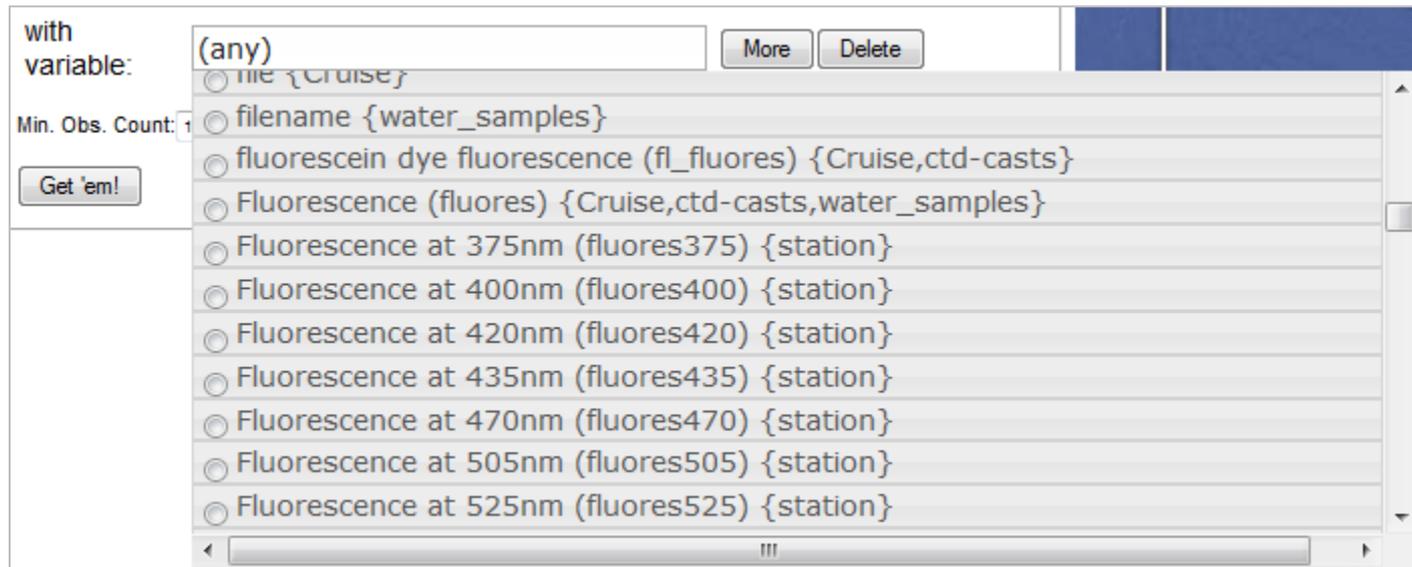


Figure: Variable List as Exposed in Search Tool

Characterizing the Metadata Mess

- Archive curator's goal: to present the metadata he wishes he had
- Sources of the mess:
 - Poor, unenforced or multiple naming standards
 - Data from multiple or external sources or systems
 - Changes in systems, standards and personnel over time
 - Many researchers, from different fields
 - Changing research foci
- Can't we repair the archive?
 - Datasets must be modified or regenerated – not practical
 - May require changing code, systems – expensive, limited payoff
 - Names may be set by vendors or external data providers
 - Time-consuming, error-prone – and problems recur
 - Change is constant

The Metadata Mess (2)

- Alternative approach: **compensate** for the mess
- How?
 - **Reduce** semantic diversity
Perfection not needed
 - Provide **transformation layer** from “what is” to “what should be”

Categories of Semantic Diversity

Category	Example
Minor variations and misspellings	<i>air_temperature, air_temperatrue, airtemp</i>
Synonyms	<i>C, degC, Centigrade</i>
Abbreviations	<i>MWHLA</i>
Excess variables	Quality assurance variables: <i>qa_level</i>
Ambiguous usages	<i>temp: temporary or temperature?</i>
Source-context naming variations	<i>temperature</i> may mean <i>air_temperature</i> or <i>water_temperature</i> , depending on source context
Concepts at multiple levels of detail	<i>Fluorescence, vs. fluores375, fluores400</i>

Semantic Diversity: Overall Approach

➤ Principles:

- No one approach sufficient
- All approaches must be simple; robust; tolerant of continued growth and ambiguity
- “Refunds and exchanges available”
 - Provide defaults
 - Improve results via overrides, modifications, adjustments
 - Be non-destructive: re-doable metadata processing

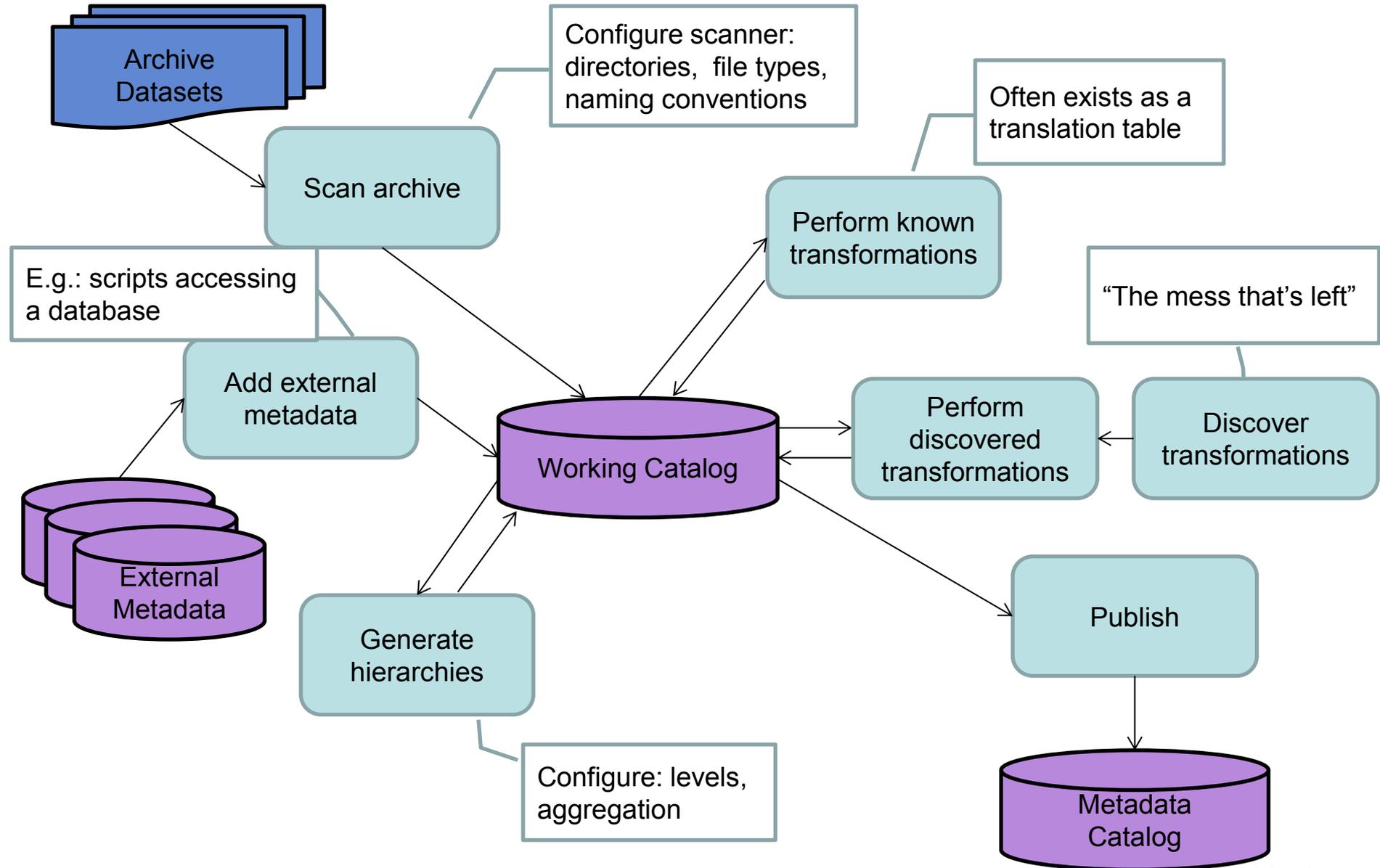
➤ “Semi-curated” model

- Curator performs some work for each new type of data indexed
- Curator can review, adjust and override currently-used defaults and prior decisions

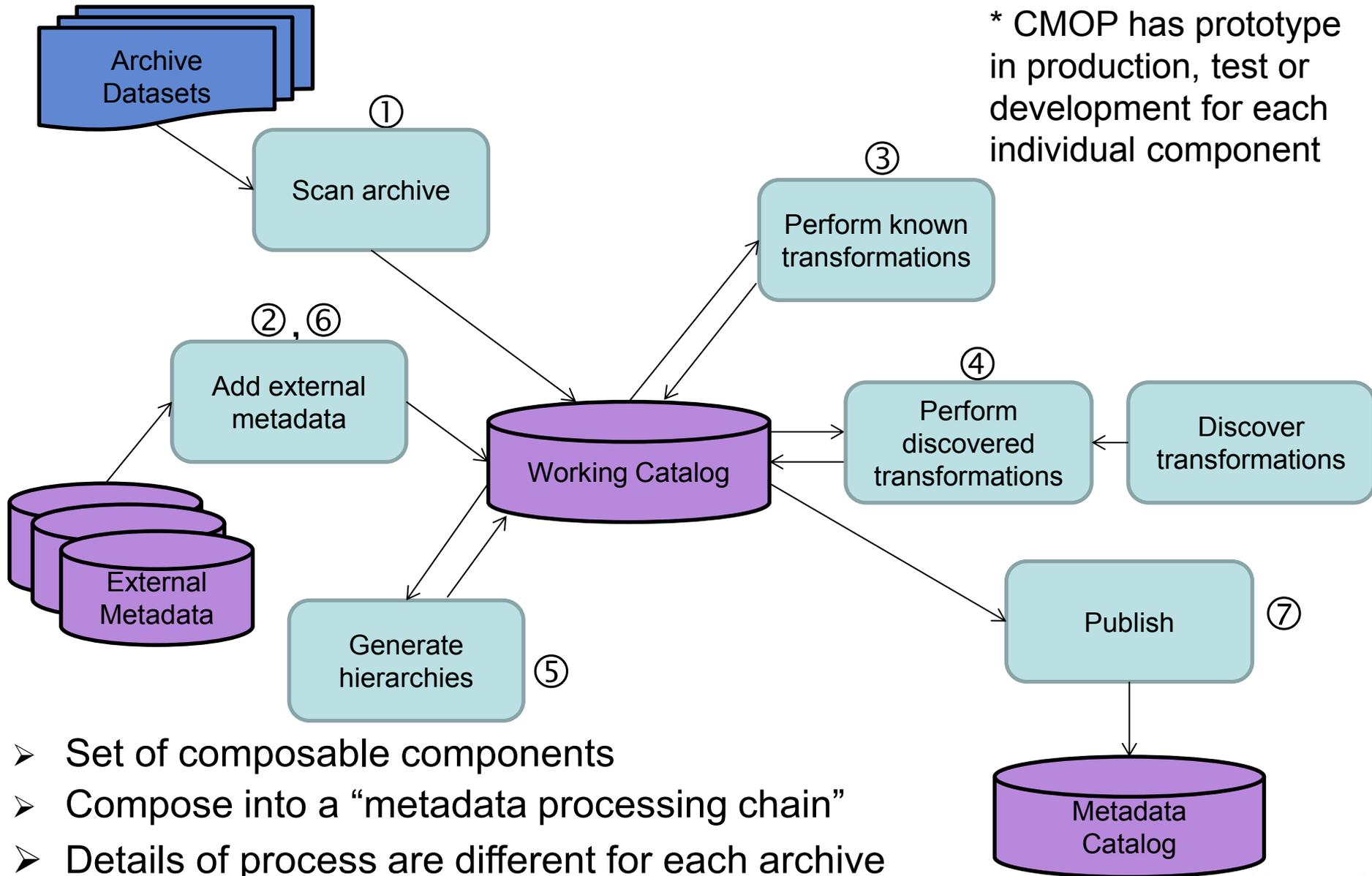
Reducing Variable-Name Diversity: Possible Approaches

Category	Example	Desired Result	Possible Technical Approach
Minor variations and misspellings	<i>air_temperature, air_temperatrue, airtemp</i>	Make them the same	Translate current to desired name
Synonyms	<i>C, degC, Centigrade</i>	Make them the same	Translate current to desired name
Abbreviations	<i>MWHLA</i>	Use full/canonical variable name	Translate current to desired name
Excess variables	Quality assurance variables: <i>qa_level</i>	Exclude from search Show in detailed dataset views	Mark variables Exclude from search
Ambiguous usages	<i>temp: temporary or temperature?</i>	Identify and expose variables. Allow curator to: <ul style="list-style-type: none"> clarify where possible hide variable leave as is 	Provide interface to specify options
Source-context naming variations	<i>Temperature: air_temperature or water_temperature</i> depending on source context	Specify context of variable Make context accessible to user	Link to multiple taxonomies
Concepts at multiple levels of detail	<i>Fluorescence, vs. fluores375, fluores400</i>	Collapse or expose as needed	Allow variables to be grouped Support hierarchical menus

Components of “Metadata Wrangling”



Metadata Wrangling Process



Current State

- Diversity of variable names is an issue – even within a single archive
 - Even larger issue when searching over federated archives
- Metadata wrangling is an ongoing activity
- We have:
 - Analyzed the problem for our archive (CMOP) and data included from other archives
 - Suggested possible approaches to address
 - Experimented with components of the process (scanner; hierarchy generator; scripts to add metadata; discovering & applying transformations)
- Giving a data curator tools to manage what she exposes – to manage her metadata mess – we enable easier use of her data archive.
- By combining this work with our search engine, we allow more effective discovery, access and use of the archive's contents.

References

- [1] J. P. Ahrens, B. Hendrickson, G. Long, S. Miller, R. Ross, and D. Williams, “Data-Intensive Science in the US DOE: Case Studies and Future Challenges,” *Computing in Science Engineering*, vol. 13, no. 6, pp. 14–24, Dec. 2011.
- [2] B. Howe, H. Green-Fishback, and D. Maier, “Scientific Mashups: Runtime-Configurable Data Product Ensembles,” in *Scientific and Statistical Database Management*, 2009, pp. 19–36.
- [3] E. Perlman, R. Burns, Y. Li, and C. Meneveau, “Data exploration of turbulence simulations using a database cluster,” in *Proc. of the ACM/IEEE conf. on Supercomputing*, 2007, pp. 1–11.
- [4] E. Stolte and G. Alonso, “Efficient exploration of large scientific databases,” in *Proc. of VLDB*, 2002, p. 633.
- [5] S. L. Pallickara, S. Pallickara, M. Zupanski, and S. Sullivan, “Efficient metadata generation to enable interactive data discovery over large-scale scientific data collections,” in *2nd IEEE International Conference on Cloud Computing Technology and Science*, 2010, pp. 573–580.
- [6] A. Rajasekar and R. Moore, “Data and metadata collections for scientific applications,” in *High-Performance Computing and Networking*, 2010, pp. 72–80.
- [7] V. M. Megler and D. Maier, “Finding Haystacks with Needles: Ranked Search for Data Using Geospatial and Temporal Characteristics,” in *Scientific and Statistical Database Management*, 2011, vol. 6809.
- [8] V. M. Megler and D. Maier, “When Big Data Leads to Lost Data,” in *PIKM 2012: 5th Workshop for Ph.D. Students at CIKM*, 2012.
- [9] D. Maier, V. M. Megler, A. Baptista, A. Jaramillo, C. Seaton, and P. Turner, “Navigating Oceans of Data,” in *Scientific and Statistical Database Management*, 2012, vol. 7338, pp. 1–19.
- [10] V.M. Megler, “Managing the Metadata Mess”, in *ICDE 2013: Workshop for Ph.D. Students at ICDE*, 2013.
- [11] P. Lord and A. Macdonald, “e-Science Curation Report,” 2003.
- [12] J. K. Batcheller, “Automating geospatial metadata generation – An integrated data management and documentation approach,” *Computers & Geosciences*, vol. 34, no. 4, pp. 387–398, 2008.
- [13] T. Hey and A. E. Trefethen, “The Data Deluge: An e-Science Perspective,” in *Grid Computing: Making the Global Infrastructure a Reality* (eds F. Berman, G. Fox and T. Hey), John Wiley & Sons, Ltd, Chichester, UK, 2003, pp. 809–824.
- [14] P. Cornillon, J. Gallagher, and T. Sgouros, “OPeNDAP: Accessing Data in a Distributed, Heterogeneous Environment,” *Data Science Journal*, vol. 2, no. 0, pp. 164–174, 2003.
- [15] J. Parsons and Y. Wand, “Attribute-based semantic reconciliation of multiple data sources,” *Journal on Data Semantics I*, pp. 21–47, 2003.
- [16] E. Rahm and P. A. Bernstein, “A survey of approaches to automatic schema matching,” *the VLDB Journal*, vol. 10, no. 4, 2001.

Major Curatorial Activities

1. Creating process
2. Running (or rerunning) process
3. Improving process
E.g., modifying a hierarchy, adding entries to a synonym table, specifying an additional directory to scan
4. Validating process results
E.g., verifying that all files in a certain directory were indeed of the same type; checking that all harvested variables names occur in the current synonym table as preferred or alternate terms; determining that expected datasets do indeed show up.

Managing “the Mess that’s Left”

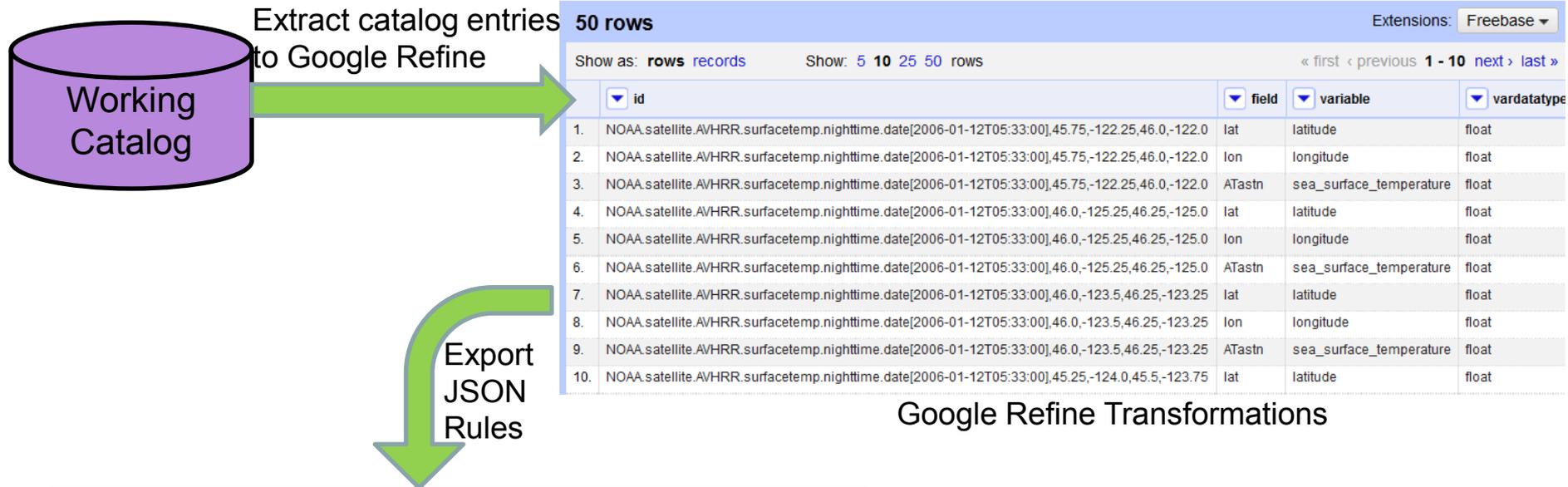
- “Discovered transformations” – discovered by reviewing results so far
 - Experimenting with Google Refine*

- Archive curator:
 1. Accesses list of variables (along with sample datasets they appear in)
 2. Reviews list
 3. Generates set of variable-name transformations and rules
 4. Applies rules and checks results for validity
 5. Exports rules and “applies”

- Transformation Engine:
 - Reruns at intervals: as new datasets are scanned
 - Applies rules to existing metadata

- Search engine:
 - Searches over “cleaned” metadata

Discovering Transformations with Google Refine



```
{
  "op": "core/mass-edit",
  "description": "Mass edit cells in column field",
  "engineConfig": { "facets": [],
    "mode": "row-based" },
  "columnName": "field",
  "expression": "value",
  "edits": [ {
    "fromBlank": false,
    "fromError": false,
    "from": [ "ATastn" ],
    "to": "sea surface temperature" } ] },
```

Run rules against metadata

id	field	variable	vardatatype	va
NO	lat	latitude	float	de
NO	lon	longitude	float	de
NO	ATastn	sea surface temperature	float	de
NO	lat	latitude	float	de
NO	lon	longitude	float	de
NO	ATastn	sea surface temperature	float	de