

# Enabling Reproducibility for Small and Large Scale Research Data Sets

Stefan Pröll, Andreas Rauber

Secure Business Austria &  
Vienna University of Technology

[sproell@sba-research.org](mailto:sproell@sba-research.org)

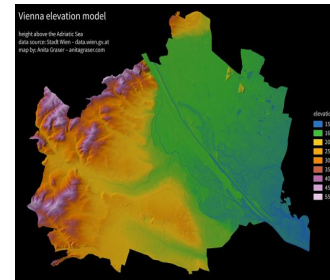
- 
- Research data challenges
  - Reproducible data sets and subsets
  - Small and large scale data settings
  - Summary
-

## ■ Results are based on data

- Data analysis
- Data mining
- Visualisation

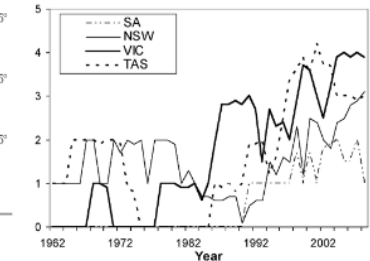
Table 1. Definition of the meteorological data and associated analyses.

Variable	Source	Temporal resolution	Spatial resolution
Precipitation	Analysis of rain gauge data	Daily and monthly total	0.05°×0.05°
Daily maximum temperature	Analysis of thermometer data	Daily and monthly average	0.05°×0.05°
		Daily and monthly average	0.05°×0.05°
		Daily and monthly average at 9 am and 3 pm	0.05°×0.05°



Source: open.wien.gv.at

Fig. 4 The average number of high-elevation stations operating in January of the listed year. High-elevation stations are defined as those above 1500 metres in NSW and Victoria, above 1000 metres in Tasmania and above 700 metres in South Australia.



## ■ Publications often contain just an aggregated result

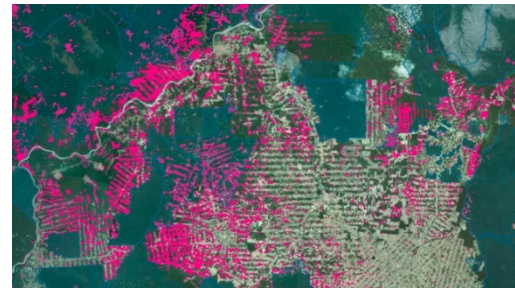
- Images
- Tables
- Graphs
- Visualisations

<sup>1</sup> This dataset is available from: <http://www.cs.utexas.edu/users/ml/nldata.html>  
<sup>2</sup> There is also a dataset consisting of 250 questions available from the University of Texas, but this is merely a subset of the larger dataset.  
<sup>3</sup> <http://www.w3.org/TR/owl-features/>

## ■ Results are based on subsets

The datasets consists of a set of 880 test questions (actually 883 questions) and was collected through a web interface hosted at the University of Austin in Texas<sup>2</sup>. We used the 883 test questions for our analysis. After downloading the

- Increasingly large amounts of data
  - Sensors
  - Streaming data
  - Time series
  - Satellite images
  - Real time analytics
- Data is heterogenous across domains but homogenous within a domain
  - Silos prevent data and methods exchange
- How to improve and maintain accessibility to data?



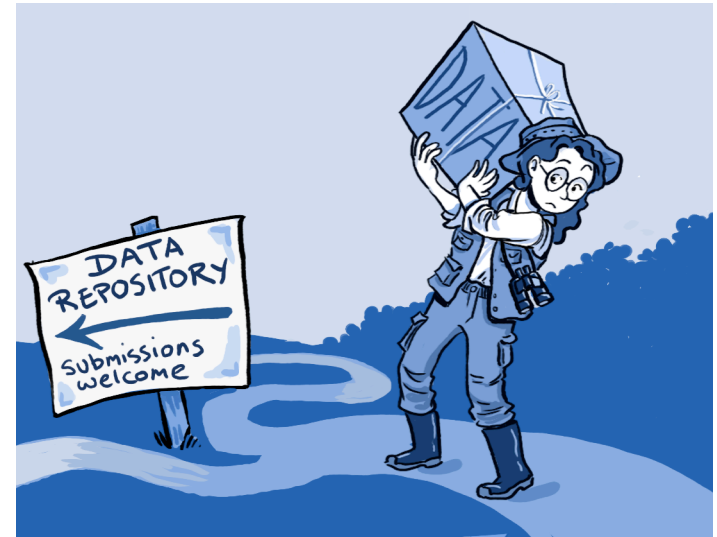
Source: CartoDB

- Data as a “1<sup>st</sup>-class citizen” in science
- We need to be able to
  - preserve data and keep it accessible
  - cite data to give credit and show which data was used
  - **identify precisely the subset of data used** in a study/ process for repeatability, verifiability,...
- Why is this difficult?  
(after all, it's being done...)



# Main Challenges

- Scalability
  - More and more data sets
  - Growing amounts of data
  - Granularity
- Infrastructure
  - Sophisticated data management is not always available
  - Processes not defined well
- Dynamics
  - Frequent updates
  - Evolving data
- Precise identification
  - Ambiguity?



Src: CC BY 4.0, <https://commons.wikimedia.org/w/index.php?curid=30978545>

# Granularity of Subsets

- What about the **granularity** of data to be identified?
  - Databases collect enormous amounts of data over time
  - Researchers use specific subsets of data
  - Need to identify precisely the subset used
- Current approaches
  - Storing a copy of subset as used in study -> scalability
  - Citing entire dataset, providing textual description of subset  
-> imprecise (ambiguity)
  - Storing list of record identifiers in subset -> scalability,  
not for arbitrary subsets (e.g. when not entire record selected)
- Would like to be able to identify precisely the  
**subset of (dynamic) data used** in a process

- Citable datasets have to be static
  - Fixed set of data, no changes:  
no corrections to errors, no new data being added
- But: (research) data is **dynamic**
  - Adding new data, correcting errors, enhancing data quality, ...
  - Changes sometimes highly dynamic, at irregular intervals
- Current approaches
  - Identifying entire data stream, without any versioning
  - Using “accessed at” date
  - “Artificial” versioning by identifying batches of data (e.g. annual), aggregating changes into releases (time-delayed!)
- Would like to identify precisely the **data as it existed at a specific point in time**



# Goals

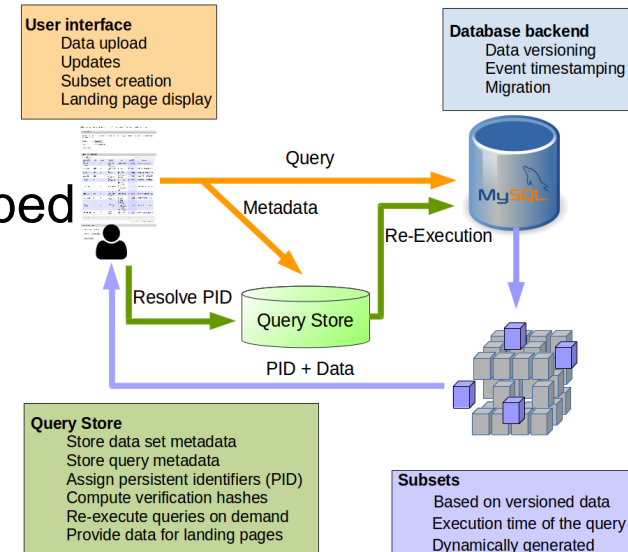
- Would like to cite precisely the **data as it existed at certain point in time**, without delaying release of new data
- We want to be able to **access previously existing version of a subset** and track the changes
- We want to **identify precisely the subset of (dynamic) data used** in a study
- Improve scalability

## Idea: Versioned data + timestamped queries

- Data: timestamped and versioned (aka history)
- Query: Timestamped
- Access: Re-execute query on versioned data with the appropriate timestamp.
- Trick: **Assign the PID to the query**
- Store queries enhanced with:
  - Time-stamping for re-execution against versioned DB
  - Normalize queries for detecting duplicates
  - Apply unique sorting
  - Compute hash of the result-set for verifying identity/correctness

# Query Store

- The Query Store is a central concept
  - Stores queries, parameters and metadata
  - Identifies all queries and data sets with unique PIDs
  - Establishes a link between the timestamped query and the versioned data set
  - Allows to re-execute queries and access the data
  - Provides information for landing pages
  - Allows to verify data sets and subsets
  - Analyse data usage
  - Can enforce policies
  - ...



- Researcher uses workbench or tool to identify subset of data
- Upon executing selection („download“) user gets
  - Data (package, access API, ...)
  - PID (e.g. DOI) (Query is time-stamped and stored)
  - Hash value computed over the data for local storage
  - Recommended citation text (e.g. BibTeX)
- PID resolves to landing page
  - Provides detailed metadata, link to parent data set, subset,...
  - Option to retrieve original data OR current version OR changes
- Upon activating PID associated with a data citation
  - Query is re-executed against time-stamped and versioned DB
  - Results as above are returned
- **Query store** aggregates data usage

- Note: query string provides excellent provenance information on the data set!
- Data (package, access API, ...)
- PID (e.g. DOI) (Query is time-stamped and stored)
- Hash value computed over the data for local storage
- Recommended citation text (e.g. BibTeX)
- PID resolves to landing page
  - Provides detailed metadata, link to parent data set, subset,...
  - Option to retrieve original data OR current version OR changes
- Upon activating PID associated with a data citation
  - Query is re-executed against time-stamped and versioned DB
  - Results as above are returned
- Query store aggregates data usage

- Note: query string provides excellent provenance information on the data set!

- Data (package)
- PID (e.g. DOI)
- Hash value
- Recommended citation text (e.g. BibTeX)

This is an important advantage over traditional approaches relying on, e.g. storing a list of identifiers/DB dump!!!

- PID resolves to landing page
  - Provides detailed metadata, link to parent data set, subset,...
  - Option to retrieve original data OR current version OR changes
- Upon activating PID associated with a data citation
  - Query is re-executed against time-stamped and versioned DB
  - Results as above are returned
- Query store aggregates data usage

- Note: query string provides excellent subset of data
- provenance information on the data set! user gets

- Data (package)
- PID (e.g. DOI)
- Hash value
- Recommended citation text (e.g. PID text)

This is an important advantage over traditional approaches relying on, e.g. storing a list of identifiers/DB dump!

- PID resolves
    - Provides details
    - Option to retrieve
- Identify which parts of the data are used.  
If data changes, identify which queries (studies) are affected

- Upon activating PID associated with a data citation
  - Query is re-executed against time-stamped and versioned DB
  - Results as above are returned

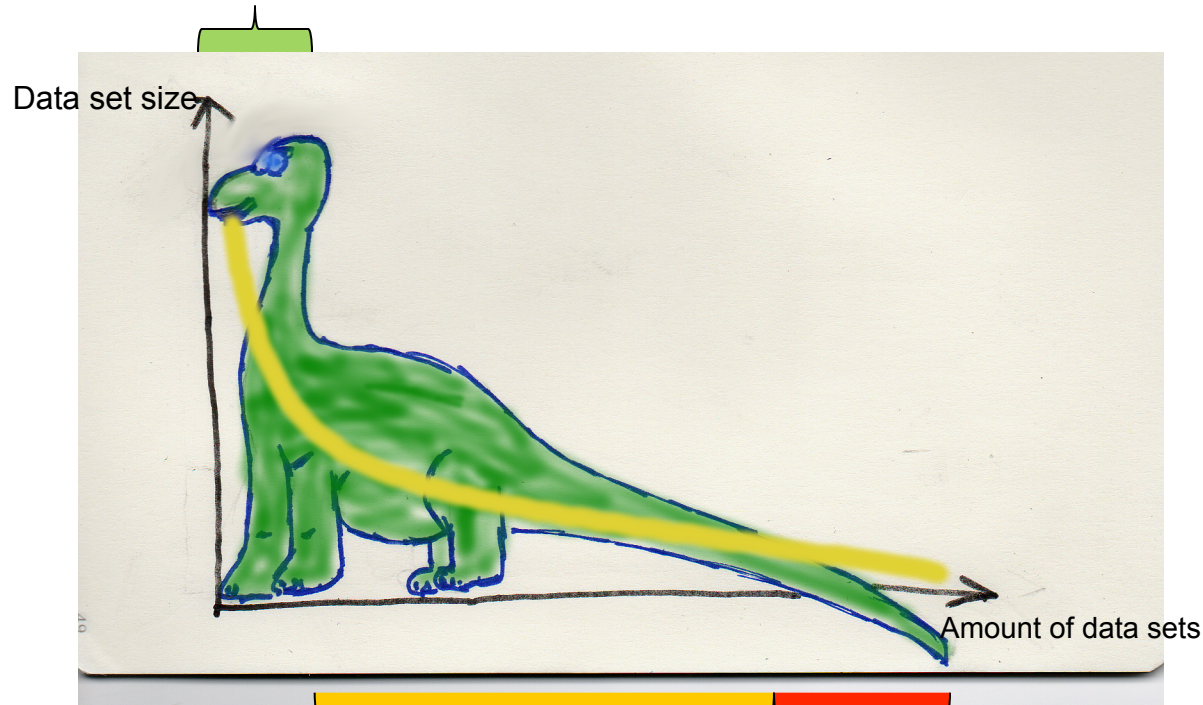
- Query store aggregates data usage





# Long Tail Research Data

Big data,  
well organized,  
often used and cited



Less well organized, “Dark data”  
non-standardised  
no dedicated infrastructure

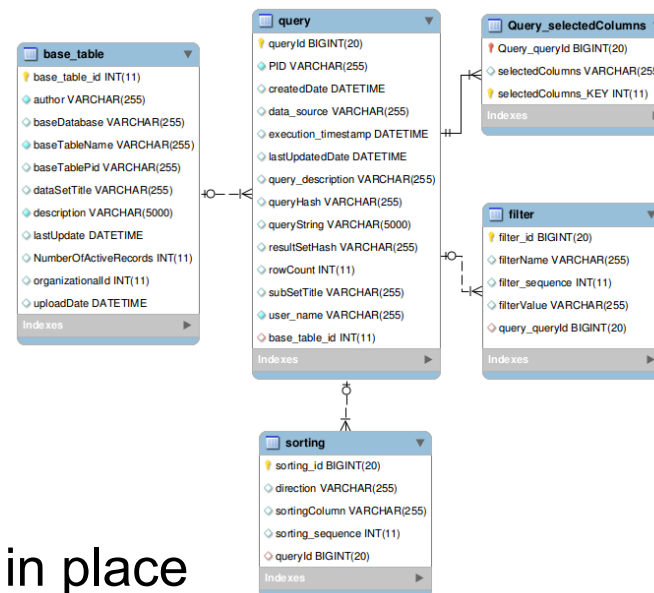


## ■ Advanced data infrastructure

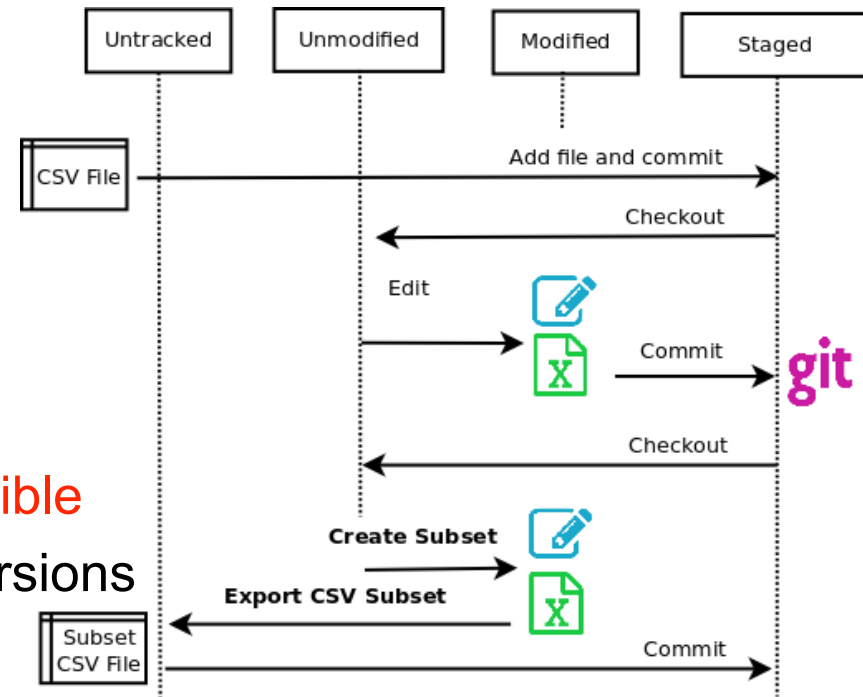
- Big data
- Database driven
- Defined interfaces
- Trained experts available

## ■ Required adaptations

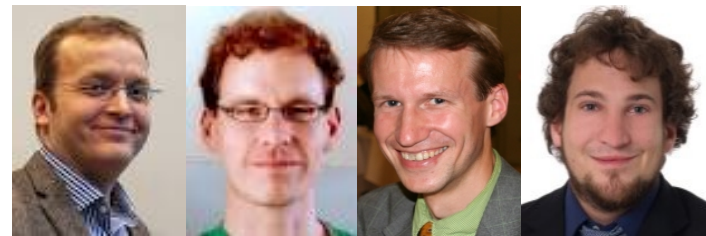
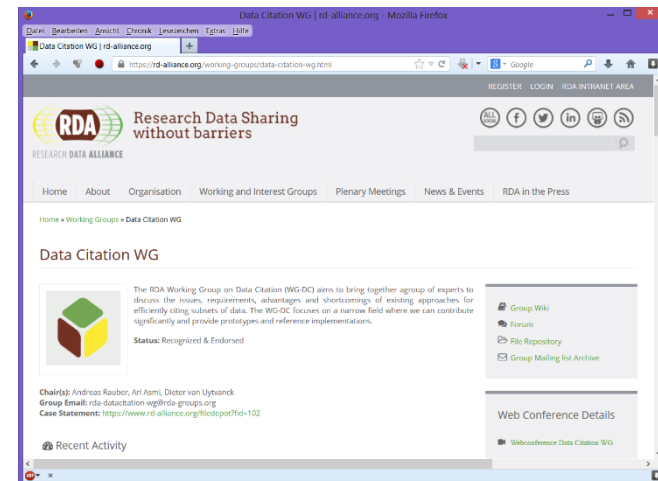
- Introduce versioning, if not already in place
- Capture subset process
- Implement dedicated query store



- Local workstations
  - Smaller data sets
  - Local storage and tools
  - Scripting languages
- Required adaptions
  - Data versioning, e.g. with Git
  - Store scripts versioned as well
  - **Make subset creation reproducible**
  - Document software and OS versions
  - Share repositories



- Research Data Alliance
- WG on **Data Citation: Making Dynamic Data Citeable**
- WG officially endorsed in March 2014
  - Concentrating on the problems of **large, dynamic (changing) datasets**
  - Focus! Identification of data!  
Not: PID systems, metadata, citation string, attribution, ...
  - Liaise with other WGs and initiatives on data citation  
(CODATA, DataCite, Force11, ...)



- <https://rd-alliance.org/working-groups/data-citation-wg.html>

- 14 Recommendations grouped into 4 phases:
  - Preparing data and query store
  - Persistently identifying specific data sets
  - Resolving PIDs
  - Upon modifications to the data infrastructure
- 2-page flyer
- More detailed Technical Report: <https://rd-alliance.org/group/data-citation-wg/wiki/wgdc-recommendations.html>
- Reference implementations (SQL, CSV, XML) and Pilots



Name	Data	Type	Status	Notes
Timbus	RDBMS	research	finished	Sensor data, pilot
XML-Reference	XML	research	finished	eXist-DB
DEXHELPP	CSV/RDBMS	research	running	Social security data
CSV-Reference	CSV/RDBMS	reference	running - $\beta$	Reference implem.
GIT-Reference	<ASCII>	reference	running - $\alpha$	Reference implem.
VAMDC	SQL/NoSQL/ ASCII -> XML	deployment	running	Distributed data center
CBMI@wustl	RDBMS	deployment	starting	integration into i2b2
CCCA	NetCDF	deployment	starting	climate data
ENVRIplus	NetCDF	deployment	starting	ICOS: Carbon Obs.Infr.
ARGO	NetCDF	deployment	starting	ODIP-II, RDA-Europe
BCO-DMO	CSV	deployment	starting	RDA-US
VMC (Vermont)	VMC data cat.	deployment	starting	Forest Research Data
<a few others>	CSV, RDBMS	deployment	planned	Conceptual evaluation, seeking funding

If you are interested in joining the discussion, contributing a pilot, wish to establish a data citation solution, ...

- Register for the RDA WG on Data Citation:

- Website:

<https://rd-alliance.org/working-groups/data-citation-wg.html>

- Mailinglist:

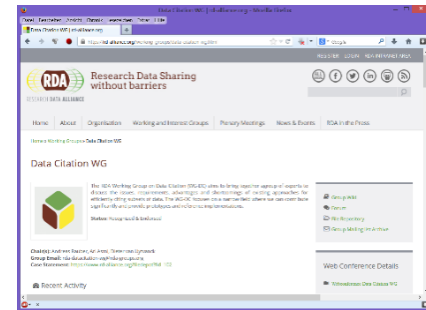
<https://rd-alliance.org/node/141/archive-post-mailinglist>

- Web Conferences:

<https://rd-alliance.org/webconference-data-citation-wg.html>

- List of pilots:

<https://rd-alliance.org/groups/data-citation-wg/wiki/collaboration-environments.html>



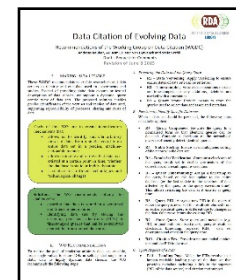
# Thank you!

- Questions?
- Comments?

Thank you very much for your attention!

[sproell@sba-research.org](mailto:sproell@sba-research.org)

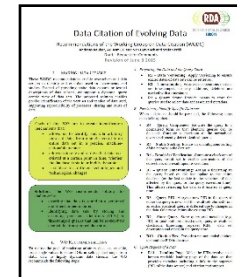
 @stefanproell



## A) Preparing the Data and the Query Store

- **R1 – Data Versioning:** Apply versioning to ensure earlier states of data sets the data can be retrieved
- **R2 – Timestamping:** Ensure that operations on data are timestamped, i.e. any additions, deletions are marked with a timestamp
- **R3 – Query Store:** Provide means to store the queries used to select data and associated metadata

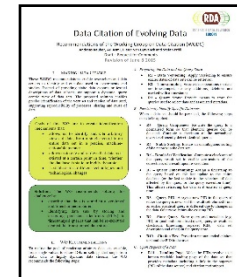




## B) Persistently Identify Specific Data sets (1/2)

*When a data set should be persisted:*

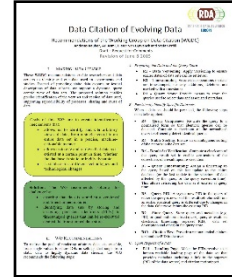
- **R4 – Query Uniqueness:** Re-write the query to a normalised form so that identical queries can be detected. Compute a checksum of the normalized query to efficiently detect identical queries
- **R5 – Stable Sorting:** Ensure an unambiguous sorting of the records in the data set
- **R6 – Result Set Verification:** Compute a checksum of the query result set to enable verification of the correctness of a result upon re-execution
- **R7 – Query Timestamping:** Assign a timestamp to the query based on the last update to the entire database (or the last update to the selection of data affected by the query or the query execution time). This allows retrieving the data as it existed at query time



## B) Persistently Identify Specific Data sets (2/2)

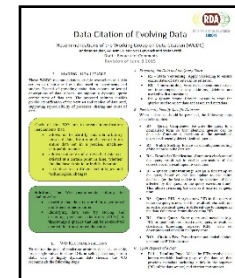
*When a data set should be persisted:*

- **R8 – Query PID:** Assign a new PID to the query if either the query is new or if the result set returned from an earlier identical query is different due to changes in the data. Otherwise, return the existing PID
- **R9 – Store Query:** Store query and metadata (e.g. PID, original and normalised query, query & result set checksum, timestamp, superset PID, data set description and other) in the query store
- **R10 – Citation Text:** Provide a recommended citation text and the PID to the user



## C) Upon Request of a PID

- **R11 – Landing Page:** Make the PIDs resolve to a human readable landing page of the data set that provides metadata including a link to the superset (PID of the data source) and citation text snippet
- **R12 – Machine Actionability:** Make the landing page machine-actionable, allowing to retrieve the data set by re-executing the timestamped query



## D) Upon Modifications to the Data Infrastructure

- **R13 – Technology Migration:** When data is migrated to a new representation (e.g. new database system, a new schema or a completely different technology), migrate also the queries and associated checksums
- **R14 – Migration Verification:** Verify successful query migration should, ensuring that queries can be re-executed correctly