



Coordination Centre

Data quality metrics in the context of DARWIN EU®

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2023-06-26

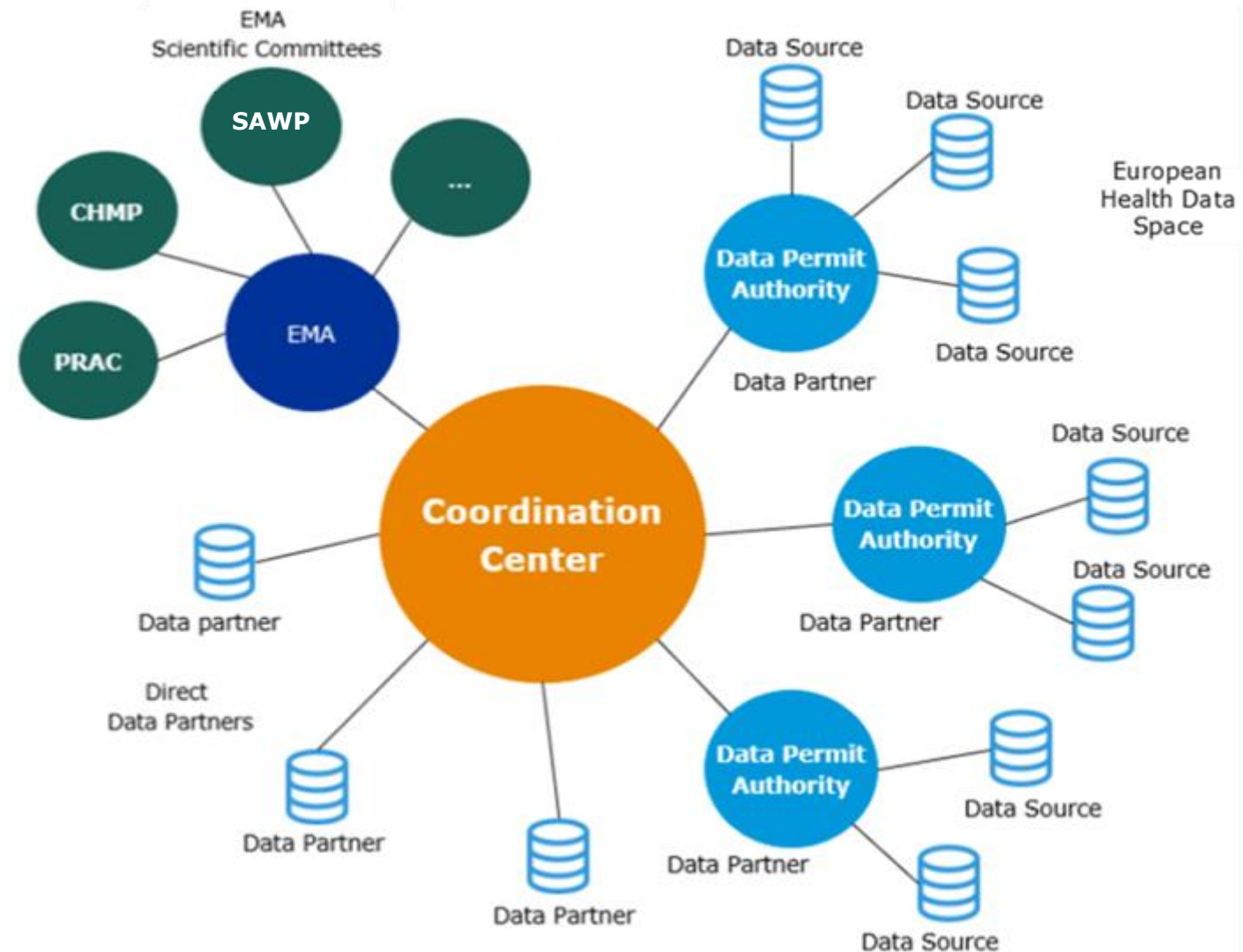
Disclosure

This presentation represents the views of the DARWIN EU® Coordination Centre only and cannot be interpreted as reflecting those of the European Medicines Agency or the European Medicines Regulatory Network.

DARWIN EU® is a federated **network of data, expertise and services** that supports better decision-making throughout the product lifecycle by generating reliable **evidence from real world healthcare data**

FEDERATED NETWORK PRINCIPLES

- Data stays **local**
- **Use of the OMOP CDM** to perform studies in a timely manner and increase consistency of results



Kahn Framework for DQA

Kahn et al.: Harmonized data quality terminology



A Harmonized Data Quality Assessment Terminology and Framework for the Secondary Use of Electronic Health Record Data

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ABSTRACT

Objective: Harmonized data quality (DQ) assessment terms, methods, and reporting practices can establish a common understanding of the strengths and limitations of electronic health record (EHR) data for operational analytics, quality improvement, and research. Existing published DQ terms were harmonized to a comprehensive unified terminology with definitions and examples and organized into a conceptual framework to support a common approach to defining whether EHR data is 'fit' for specific uses.

Materials and Methods: DQ publications, informatics and analytics experts, managers of established DQ programs, and operational manuals from several mature EHR-based research networks were reviewed to identify potential DQ terms and categories. Two face-to-face stakeholder meetings were used to vet an initial set of DQ terms and definitions that were grouped into an overall conceptual framework. Feedback received from data producers and users was used to construct a draft set of harmonized DQ terms and categories. Multiple rounds of iterative refinement resulted in a set of terms and organizing framework consisting of DQ categories, subcategories, terms, definitions, and examples. The harmonized terminology and logical framework's inclusiveness was evaluated against ten published DQ terminologies.

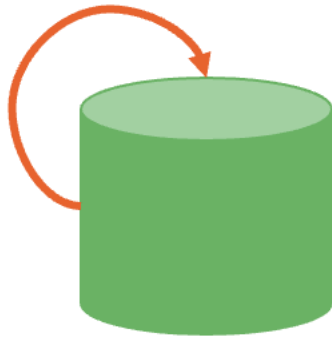
Volume 4 (2016) Issue Number 1

Table 1. Harmonized DQ Terms, Definitions, and Examples: Organized by Verification and Validation Contexts Within Categories and Subcategories

VERIFICATION		VALIDATION	
DEFINITION	EXAMPLE	DEFINITION	EXAMPLE
CONFORMANCE: DO DATA VALUES ADHERE TO SPECIFIED STANDARDS AND FORMATS?			
VALUE CONFORMANCE			
a. Data values conform to internal formatting constraints.	a. Sex is only one ASCII character.	a. Data values conform to representational constraints based on external standards.	a. Values for primary language conform to ISO standards.
b. Data values conform to allowable values or ranges.	b. Sex only has values "M," "F," or "U."		
RELATIONAL CONFORMANCE			
a. Data values conform to relational constraints.	a. Patient medical record number links to other tables as required.	a. Data values conform to relational constraints based on external standards.	a. Data values conform to all not-NULL requirements in a common multi-institutional data exchange format.
b. Unique (key) data values are not duplicated.	b. A medical record number is assigned to a single patient.		
c. Changes to the data model or data model versioning.	c. Version I data does not include medical discharge hour.		
COMPUTATIONAL CONFORMANCE			
a. Computed values conform to computational or programming specifications.	a. Database- and hard-calculated Body Mass Index (BMI) values are identical.	a. Computed results based on published algorithms yield values that match validation values provided by external source.	a. Computed BMI percentiles yield identical values compared to test results and values provided by the CDC.
COMPLETENESS: ARE DATA VALUES PRESENT?			
a. The absence of data values at a single moment in time agrees with local or common expectations.	a. The encounter ID variable has missing values.	a. The absence of data values at a single moment in time agrees with trusted reference standards or external knowledge.	a. The current encounter ID variable is missing twice as many values as the institutionally validated database.
b. The absence of data values measured over time agrees with local or common expectations.	b. Gender should not be null.	b. The absence of data values measured over time agrees with trusted reference standards or external knowledge.	b. A drop in ICD-9CM codes matches implementation of ICD-10CM.
c. Medical discharge time is missing for three consecutive days.			
PLAUSIBILITY: ARE DATA VALUES BELIEVABLE?			
UNIQUENESS PLAUSIBILITY			
a. Data values that identify a single object are not duplicated.	a. Patients from a single institution do not have multiple medical record numbers.	a. Data values that identify a single object in an external source are not duplicated.	a. An institution's CMS facility identifier does not refer to a multiple institutions.

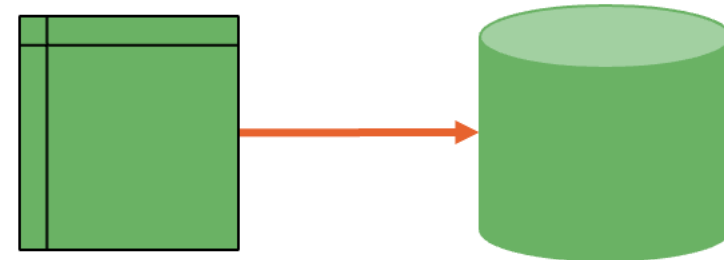
Verification

*"focuses on how data values match expectations with respect to metadata **constraints**, **system assumptions**, and **local knowledge**"*



Validation

*"focuses on the alignment of data values with respect to relevant **external benchmarks**"*



Conformance

*"focuses on DQ features that describe the **compliance** of the representation of data against internal or **external formatting, relational, or computational** definitions", e.g.:*

- ✓ Diagnosis is linked to person table via valid foreign key
- ✓ All persons have a birthdate
- ✓ Given BMI is equal to calculated BMI value

Completeness

*"focuses on features that describe the **frequencies of data attributes present** in a data set without reference to data values", e.g.:*

- ✓ At least one diagnosis code given for a healthcare visit
- ✓ Fewer hospitalisations during the COVID lockdown

Plausibility

*"focuses on features that describe the **believability or truthfulness of data values**", e.g.:*

- ✓ Height and weight values are positive
- ✓ Medication administrated after birthdate

Where to begin with Data Quality?

CATEGORIES

CONTEXTS

	Verification	Validation
Plausibility	?	?
Conformance	?	?
Completeness	?	?

	Verification	Validation
Plausibility	?	?
Conformance	?	?
Completeness	?	?

The number and percent of records with a value in the *year_of_birth* field of the *person* table less than 1850.

	Verification	Validation
Plausibility	?	?
Conformance	?	?
Completeness	?	?

The number and percent of records with a value in the *CDM field* of the *CDM table* less than *a low value*.

	Verification	Validation
Plausibility	?	?
Conformance	?	?
Completeness	?	?

The number and percent of records with a value in the *quantity* field of the *drug_exposure* table less than 0.

	Verification	Validation
Plausibility	?	?
Conformance	?	?
Completeness	?	?

The number and percent of values which are mapped to a standard concept in the **CONDITION_CONCEPT_ID** field of the **CONDITION_OCCURRENCE** table.

...which we can make more generic

	Verification	Validation
Plausibility	?	?
Conformance	?	?
Completeness	?	?

The number and percent of values which are not mapped to a standard concept in the *CDM field* of the *CDM table*.

Resulting in a set of Check Types

Conformance

- Primary Key
- Foreign Key
- Data type
- CDM Field Exists
- Required Field

Database constraints

- Standard Valid Concept
- Domain
- Class

Completeness

- Records per person
- Field value
- Standard Concept
- Source Concept
- Source Value

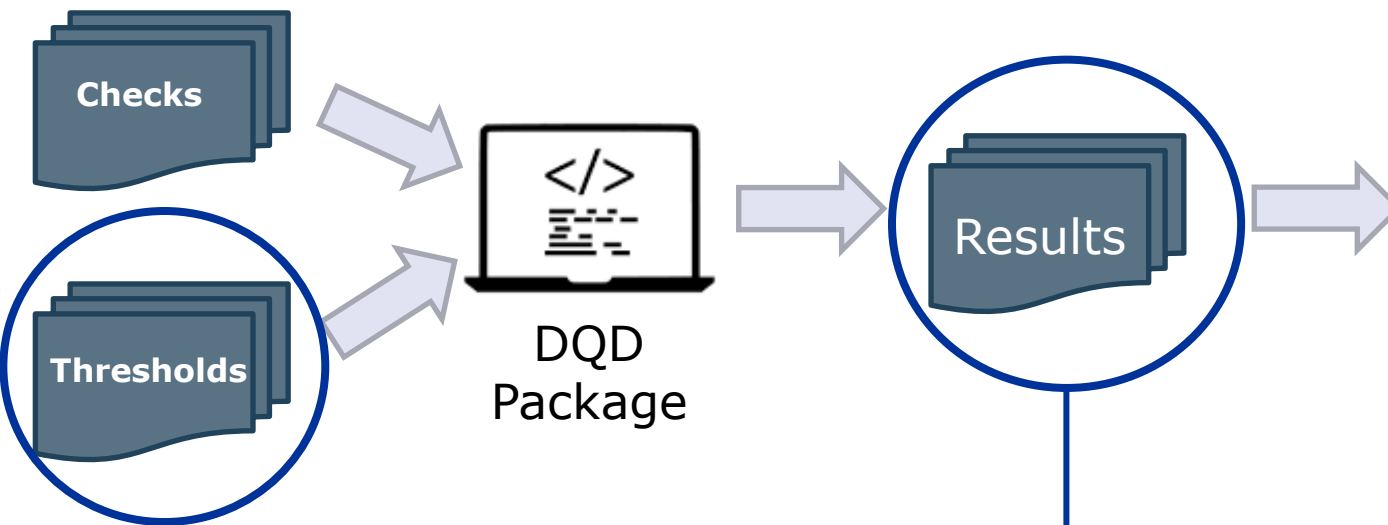
Plausibility

- Gender
- Low Value
- High Value
- Unit concept
- During Life
- Temporal after

Examples of validation context

A total of >3000 checks

	Verification	Validation
Plausibility	1878	287
Conformance	681	104
Completeness	386	15



To be shared as part
of the onboarding
process



DATA QUALITY ASSESSMENT

MEDICARE CLAIMS SYNTHETIC PUBLIC USE FILES (SYNPUFS)

DataQualityDashboard Version: 1.0.0
Results generated at 2021-11-25 05:41:37 in 14 hours

	Verification				Validation				Total			
	Pass	Fail	Total	% Pass	Pass	Fail	Total	% Pass	Pass	Fail	Total	% Pass
Plausibility	1859	1	1860	100%	89	198	287	31%	1948	199	2147	91%
Conformance	588	3	591	99%	97	0	97	100%	685	3	688	100%
Completeness	332	6	338	98%	12	2	14	86%	344	8	352	98%
Total	2779	10	2789	100%	198	200	398	50%	2977	210	3187	93%

<https://data.ohdsi.org/DataQualityDashboardSynpuf/>

How to decide if result are good enough?

% of records that **violate**
check

0% = any violation → FAIL
100% = always PASS

Check Category	Check Description	Check Result	Decision Threshold	Pass / Fail	
Verification - Plausibility	The number and percent of records with a value in the YEAR_OF_BIRTH field of the PERSON table less than 1850.	0%	0%	PASS	
Verification - Plausibility	The number and percent of records with a value in the DAYS_SUPPLY field of the DRUG_EXPOSURE table less than 0.	0%	1%	PASS	
Verification - Plausibility	For Hemoglobin A1c percent, the number and percent of records with a value in the VALUE_AS_NUMBER field of the MEASUREMENT table less than 4.	0.01%	5%	PASS	
Verification - Completeness	The number and percent of records with a value of 0 in the standard concept field CONDITION_CONCEPT_ID in the CONDITION_OCCURRENCE table.	0.02%	5%	PASS	
Verification - Completeness	The number and percent of records with a value of 0 in the standard concept field UNIT_CONCEPT_ID in the MEASUREMENT table.	93.66%	5%	FAIL	

OHDSI Data Quality Dashboard (DQD)

OHDSI / DataQualityDashboard

Unwatch

7

Star

5

Fork

6

Code

Issues 16

Pull requests 0

Actions

Projects 1

Wiki

Security

Insights

Settings

A tool to help improve data quality standards in observational data science. <https://ohdsi.github.io/DataQualityDa...>

Edit

data-quality

Manage topics

150 commits

4 branches

0 releases

1 environment

6 contributors

Apache-2.0

Branch: master

New pull request

Create new file

Upload files

Find File

Clone or download

alondhe

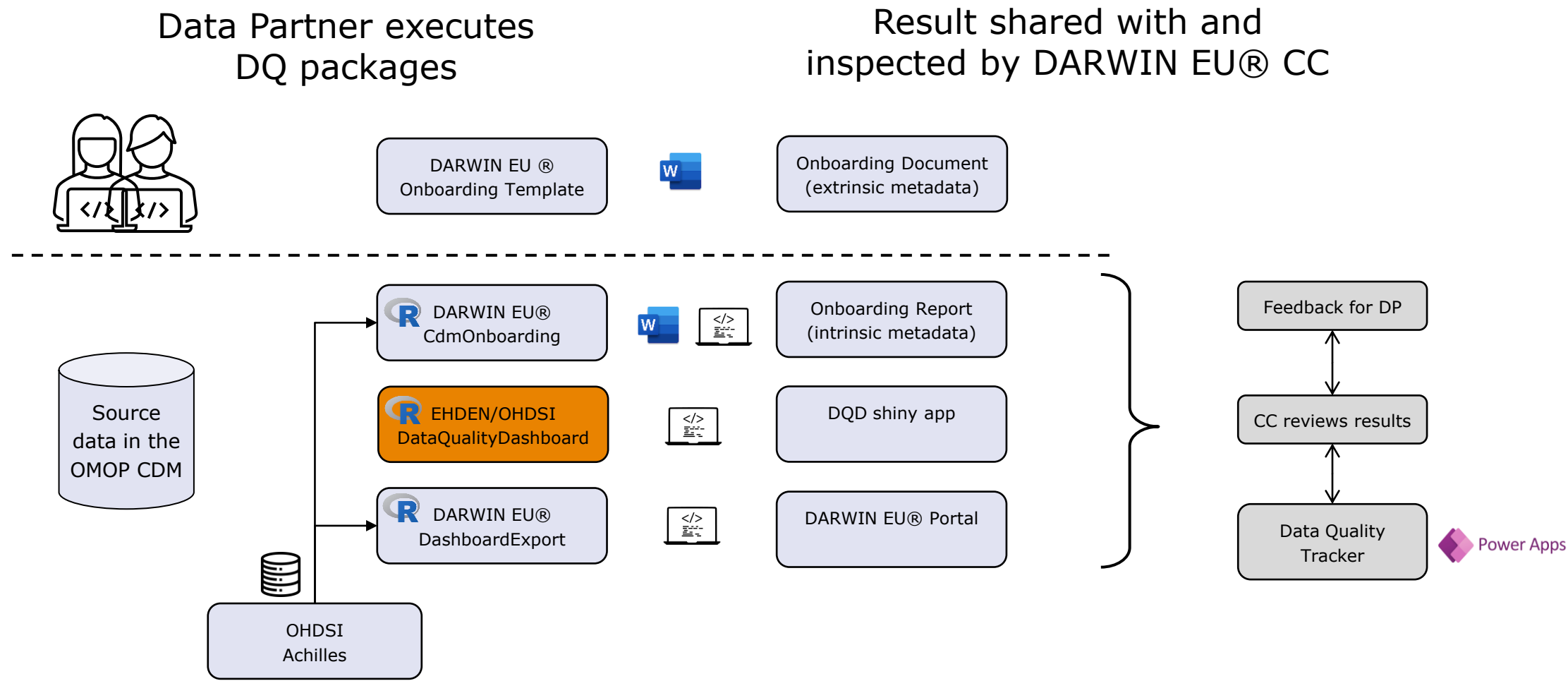
Disable scientific notation in execution

✓ Latest commit 4676653 12 hours ago

R	Disable scientific notation in execution	12 hours ago
docs	Added tablesToExclude parameter to allow skipping of tables if known ...	13 days ago
extras	Added screenshot	18 days ago
inst	Remove duplicates from concept level	4 days ago
man	Added tablesToExclude parameter to allow skipping of tables if known ...	13 days ago
tests	Added missing dbType in test	18 days ago
.Rbuildignore	updated R package wrappers	2 months ago
.gitignore	Fixes #30, making autocommit setting specific to connectionDetails db...	18 days ago
.travis.yml	Added devtools to travis. Added travis and codecov statuses to readme...	18 days ago
DESCRIPTION	Added pkgdown documentation. Added one testthat test.	19 days ago
DataQualityDashboard.Rproj	updated R package wrappers	2 months ago
LICENSE	Initial commit	3 months ago
NAMESPACE	Added pkgdown documentation. Added one testthat test.	19 days ago



<https://github.com/OHDSI/DataQualityDashboard>



More Information



DQ Packages and documentation:

<https://github.com/ohdsi/DataQualityDashboard>

<https://github.com/darwin-eu/CdmOnboarding>

<https://github.com/darwin-eu/DashboardExport>



[Coordination Centre website](https://darwin-eu.org/)

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For questions to the Coordination Centre,
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