

Describing Diversity of Real-World Data Sources: A New Framework Adapted From the DIVERSE Initiative Scoping Review

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DISCLOSURES

RPA is an employee of RTI Health Solutions, a unit of RTI International, a nonprofit organization that conducts work for government, public, and private organizations, including pharmaceutical companies. RG is employed by ARS Toscana, a publicly owned agency that participates in studies funded by pharmaceutical companies and is compliant with the ENCePP Code of Conduct. The budget of her unit is partially supported by such studies. DB is an employee of Takeda. LL is an employee of Sanofi and may hold shares and/or stock options in the company. CG is an employee of MSD Innovation and Development, Zurich, Switzerland. MSD Innovation and Development did not have any involvement in the study. MBu is an employee of and owns stocks in Merck & Co., Inc., Rahway, NJ, US. SK is an employee of Teva Pharmaceutical Industries Ltd. KM was an employee of IQVIA at the time of conducting the study. GR is a consultant for ARS Toscana. XZ is an employee of Pfizer Inc. and has received stock from Pfizer Inc. RPI has received fees, outside of the scope of this work, for consulting from Biogen, Merck, and Pfizer.

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Registration: The study protocol is publicly available in the HMA-EMA Catalogue of real-world studies (registration number: EUPAS39757).

BACKGROUND

Variation in results across studies may stem from differences in foundational characteristics between real-world data sources (RWDS). This study addresses the reproducibility challenges associated with data diversity across RWDS in pharmacoepidemiologic studies.

OBJECTIVES

To identify and characterize practices, tools, and recommendations for describing and reporting diversity between RWDS used in pharmacoepidemiologic studies.

METHODS

The full methods and results of this study are available online at Pharmacoepidemiology and Drug Safety.¹

A scoping review was conducted. First, a 'core' set of documents was identified by co-authors based on their expert subject knowledge, including:

- Documents offering recommendations, guidelines, or tools for collecting/reporting diversity of data sources.
- Documents describing data sources produced by organizations/networks that conduct multi-database studies.
- Multi-database studies with substantial description of data sources or strategies to leverage data source diversity.

A systematic literature search was conducted in 3 steps, and documents were screened and selected using a standardized selection tool. Information was recorded by reviewers using an extraction tool designed to capture information on topics related to collecting and reporting RWDS diversity. Content analysis was conducted to identify common themes within the topics.

DISCUSSION

- The 9 dimensions provide a framework to describe data sources when reporting future studies.
- 36 common themes highlighted common terminologies and understandings, as well as inconsistencies and misconceptions when describing diverse RWDS.
- Dimensions, such as Content and Data dictionary, had the most consistent and simple themes.
- Others, including Prompt and Inclusion in the population, were more complex and were approached less consistently across articles.
- Describing these dimensions will clarify assumptions that are made when interpreting studies using single or multiple routinely collected health data.

Reference

1. Gini R, Pajouheshnia R, Gardarsdottir H, Bennett D, Li L, Gulea C, et al. Describing diversity of real world data sources in pharmacoepidemiologic studies: the DIVERSE scoping review. Pharmacoepidemiol Drug Saf. 2024 May;33(5):e5787. doi:10.1002/pds.5787.



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RESULTS

Ninety-one articles were included after screening. Mostly review articles and original research articles were included, primarily from North America and Europe (see Figures 1, 2A, 2B).

Nine dimensions to describe RWDS were identified along with 36 associated themes (see Figure 3). No tools to summarize diversity were found for most of the 9 dimensions to describe data source diversity; there were few examples of concepts or ontologies.

Examples from the content analysis:

Example 1: Prompt

Several articles reported that information may be missing from data sources when specific aspects of healthcare are not captured completely. Gaps may arise across data sources due to the specific mechanism that prompts data generation in each data source resulting in information that is missing-not-at-random.

Example 2: Inclusion in the population

Four groups of conditions for in/exclusion from a data source were found, which could form the basis of an ontology: inhabitance of a country/region, enrollment in a health plan, hospital assistance, and primary care assistance.

The concept of date of entry and exit from a data source was commonly missing. However, several articles did distinguish a study population as being nested within an underlying population with the potential to have a record in the data source.

Figure 1. Flowchart of the Selection of the Documents Included in the Scoping Review

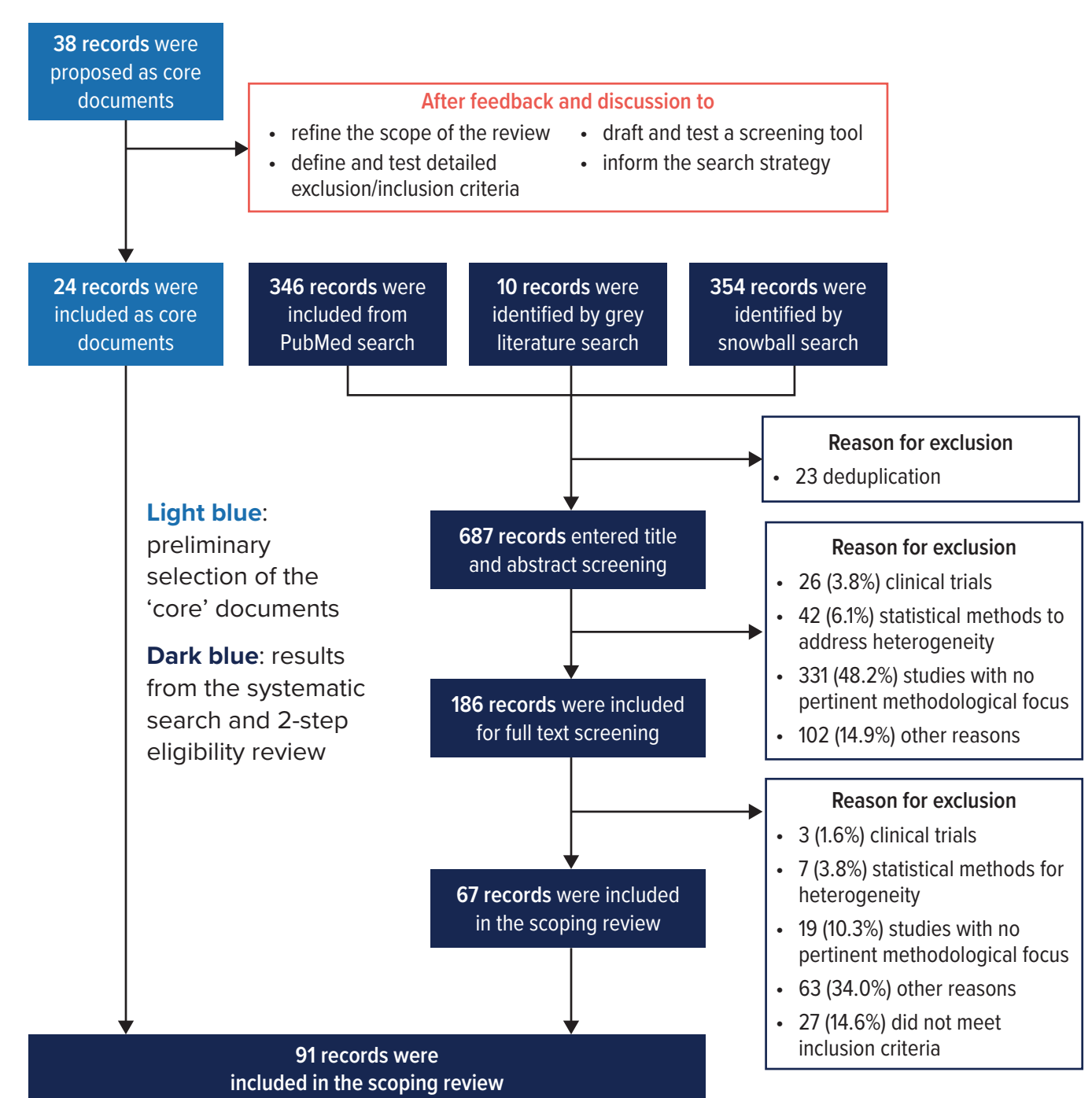


Figure 2. Overview of the Articles Included in the Scoping Review

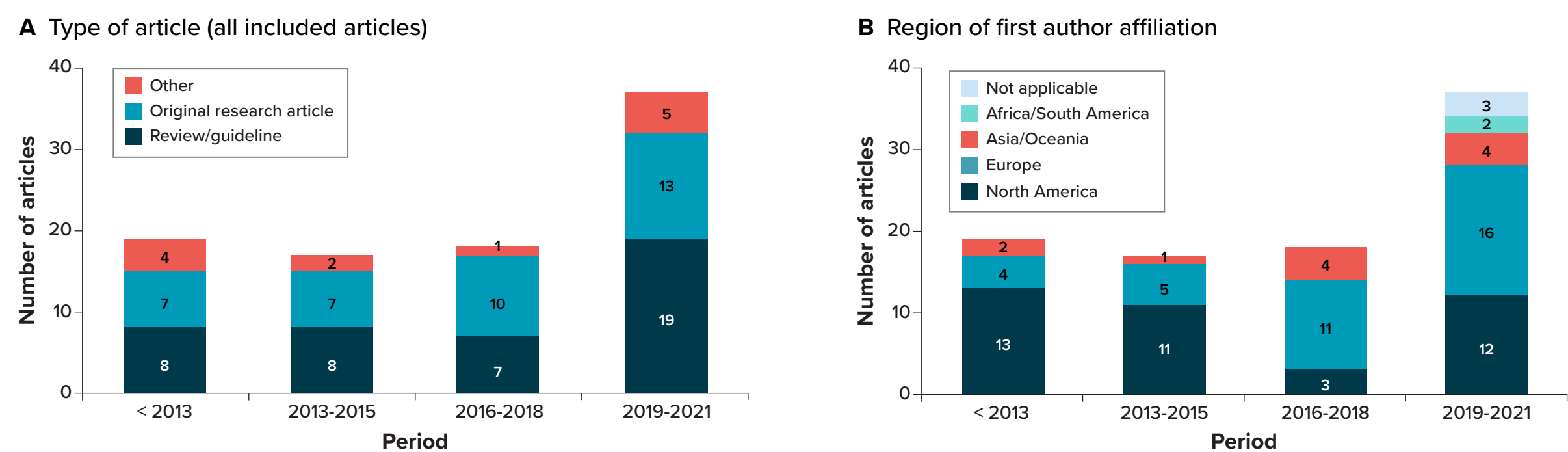
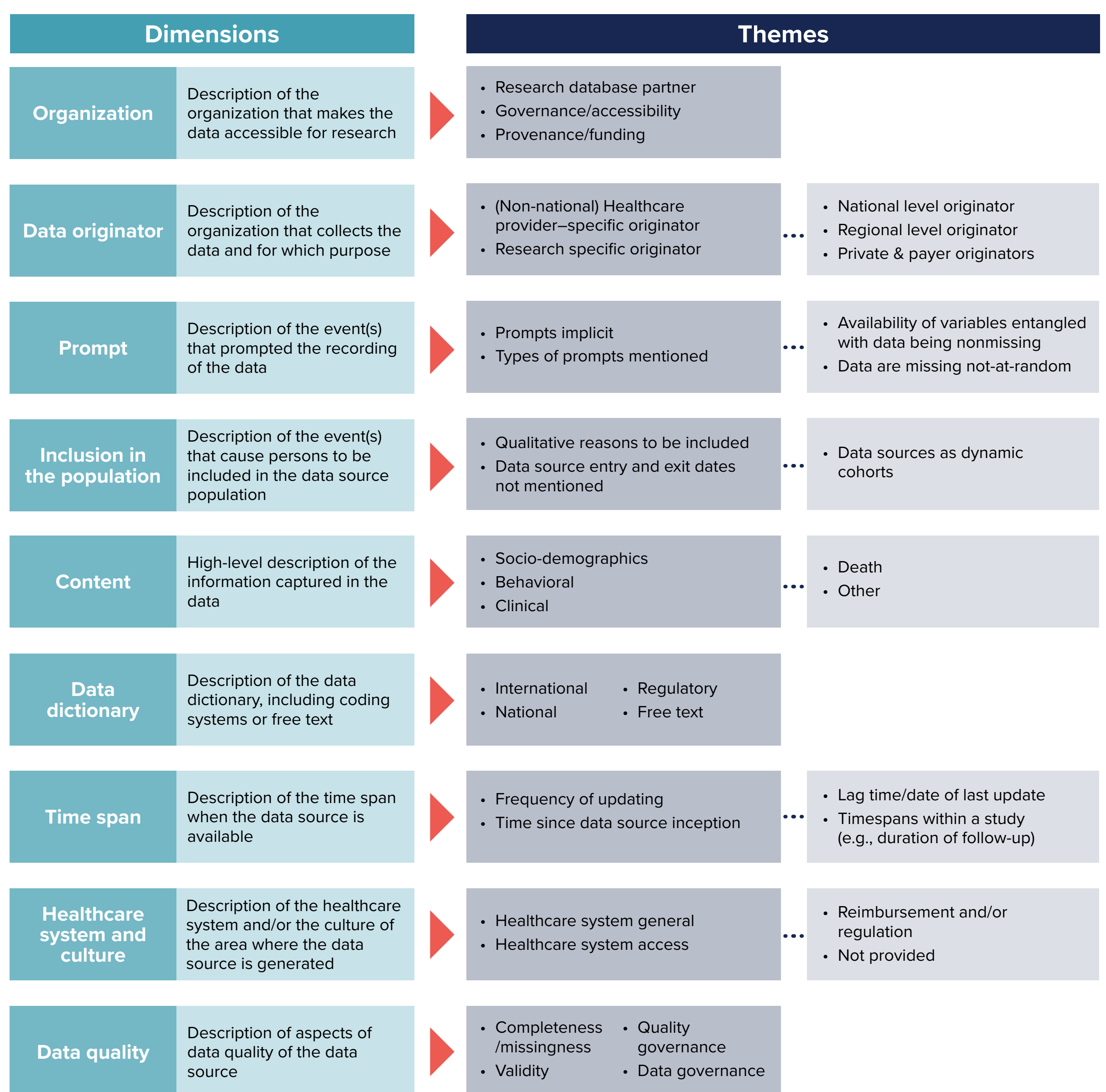


Figure 3. Nine Dimensions to Describe Diverse Data Sources and Related Themes



CONCLUSIONS

- The 9 dimensions provide a framework for describing data sources, thereby further facilitating reproducibility and the interpretation of studies.
- More formal guidance is being developed by the DIVERSE initiative, complementing concurrent initiatives.
- We encourage journals to include the DIVERSE framework in their guidelines for authors for submissions of RWE studies.