



# The Journey of OHDSI: Where have we been and where can we go together?

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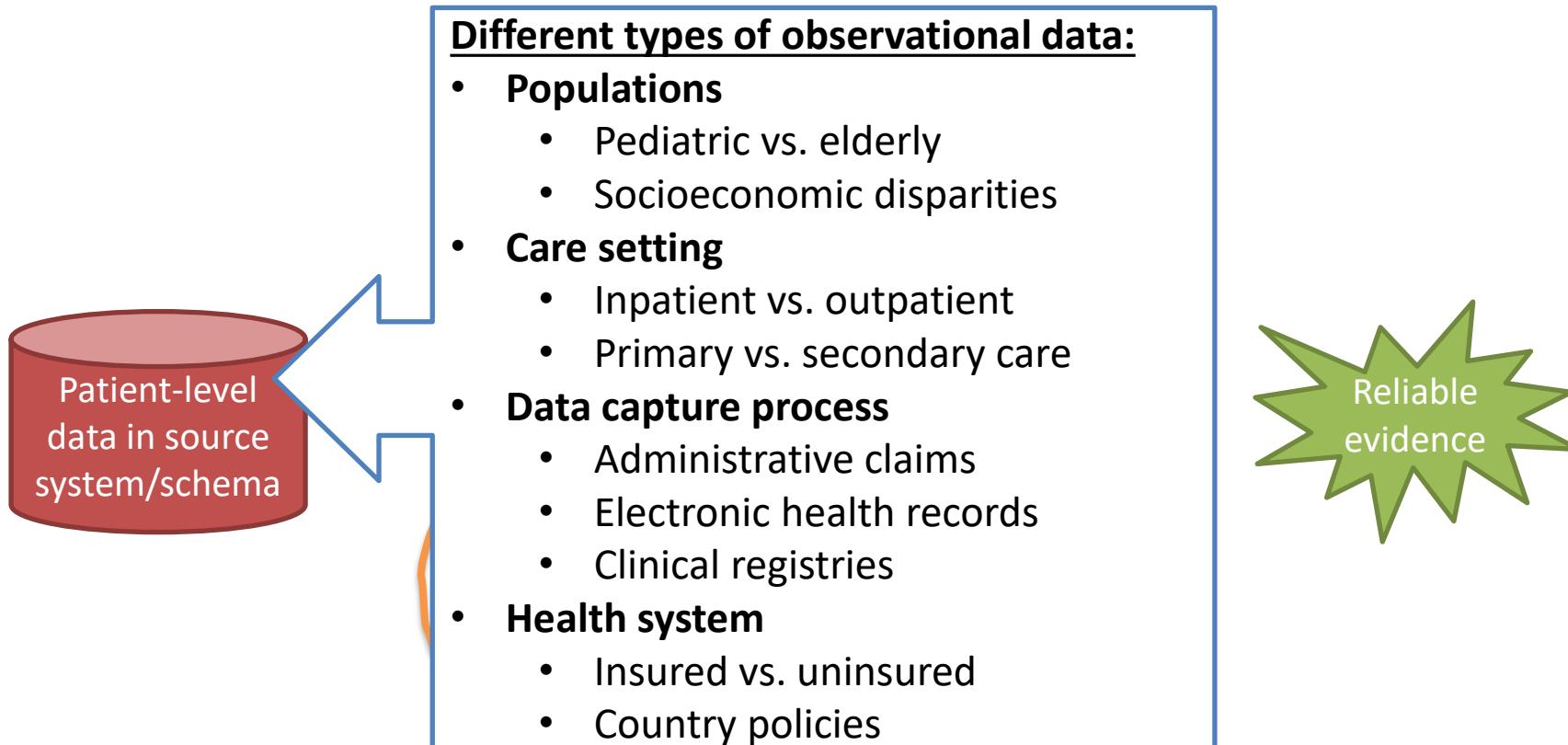


# The journey to real-world evidence



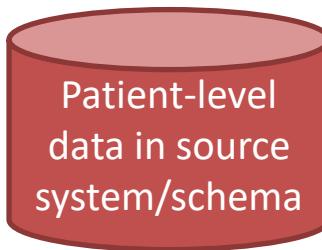


# The journey to real-world evidence





# The journey to real-world evidence



## **Types of evidence desired:**

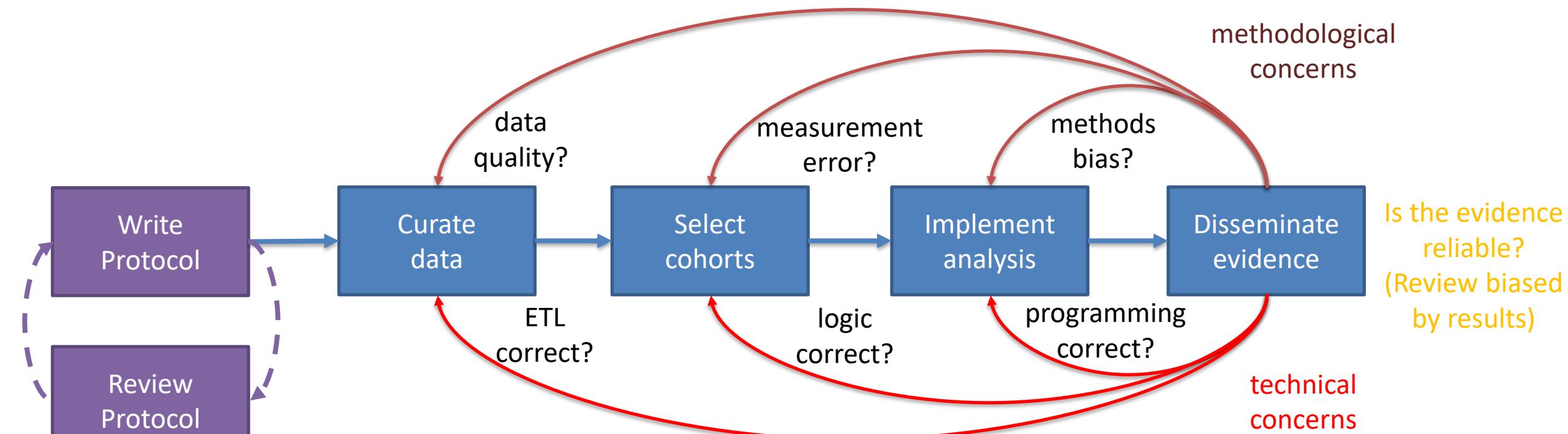
- **Clinical characterization**
  - Clinical trial feasibility
  - Treatment utilization
  - Disease natural history
  - Quality improvement
- **Population-level effect estimation**
  - Safety surveillance
  - Comparative effectiveness
- **Patient-level prediction**
  - Precision medicine
  - Disease interception

Reliable  
evidence



# Current status quo in observational research makes it challenging to build trust in evidence

Does the study provide an unbiased effect estimate?  
Are the findings generalizable to the population of interest?



Can the study be fully reproduced?  
Does the analysis actually do what the protocol said it would do?



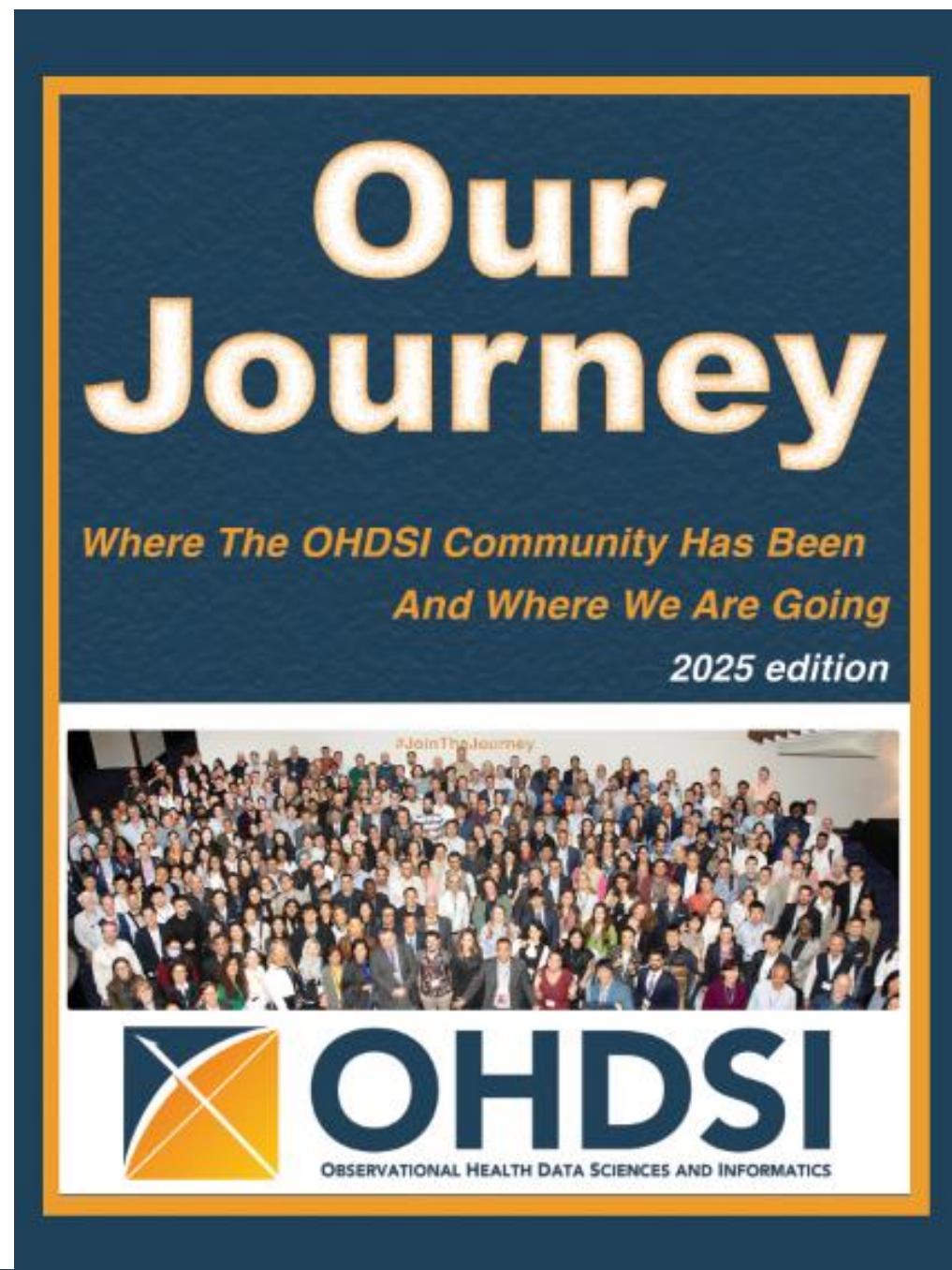
# Desired attributes for reliable evidence

Desired attribute	Question	Researcher	Data	Analysis	=	Result
Repeatable	Identical	Identical	Identical	Identical	=	Identical
Reproducible	Identical	Different	Identical	Identical	=	Identical
Replicable	Identical	Same or different	Similar	Identical	=	Similar
Generalizable	Identical	Same or different	Different	Identical	=	Similar
Robust	Identical	Same or different	Same or different	Different	=	Similar
Calibrated	Similar (controls)	Identical	Identical	Identical	=	Statistically consistent



## OHDSI's mission

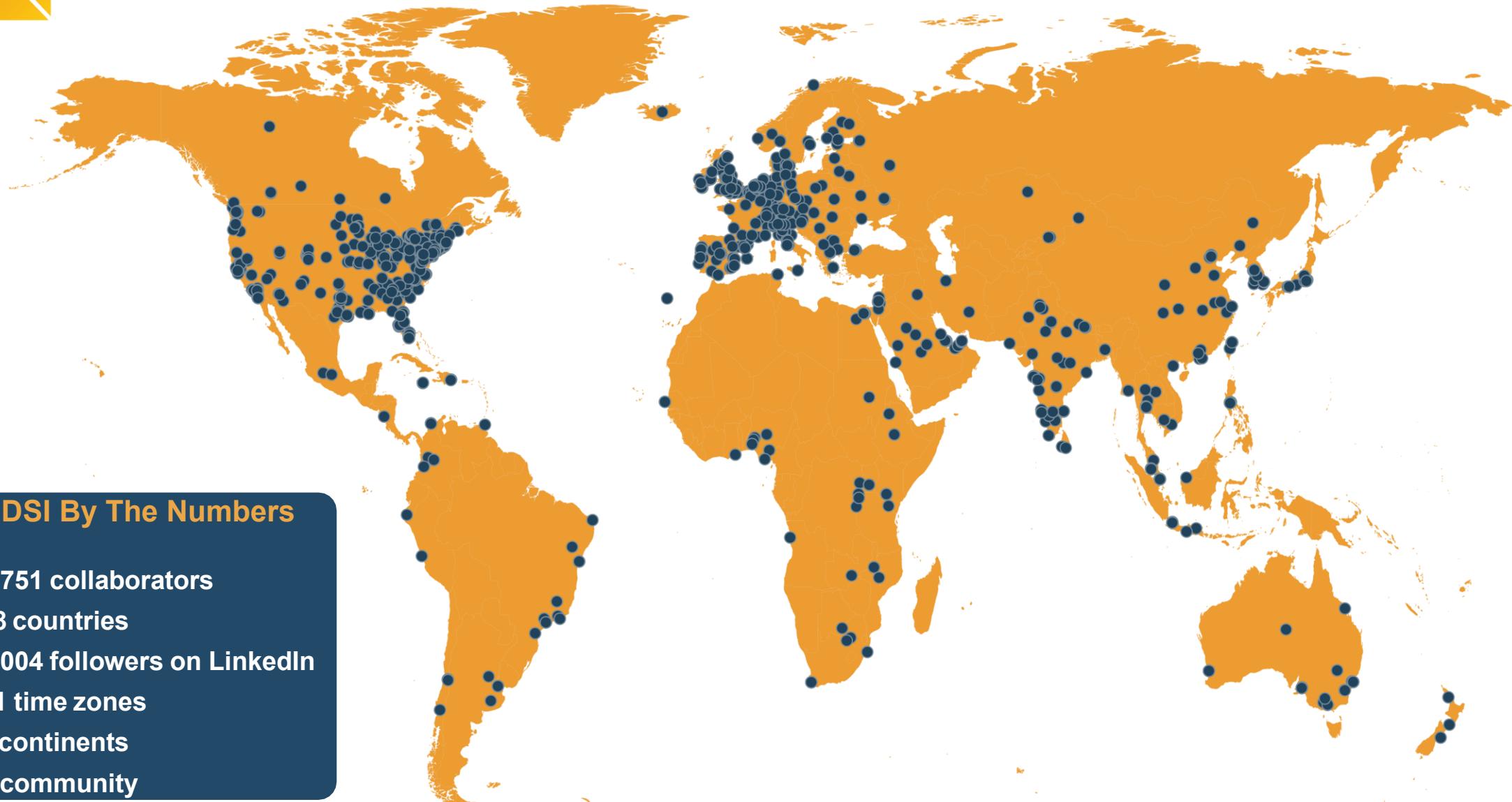
To improve health by empowering a community to collaboratively generate the evidence that promotes better health decisions and better care



<https://www.ohdsi.org/wp-content/uploads/2025/10/OurJourney-2025.pdf>



# OHDSI collaborators



Join the Journey at <https://ohdsi.org/>



# Workgroups led by community

ATLAS/WebAPI	Clinical Trials	Common Data Model	CDM Survey	CDM Vocabulary	Medical Imaging	Methods Research	Natural Language Processing	Network Data Quality			
Christopher Knoll	Alexey Manoylenko	Mike Hamidi	Zhen Lin	Clair Blacketer	Nicole Gerlano	Anna Ostropolski	Song Chan You	Martijn Schuemie			
Databricks Users	Dentistry	Early-Stage Researchers	Electronic Animal Health Records	Oncology	Open-Source Community	Patient-Level Prediction (PLP)	Perinatal and Reproductive Health				
John Gresh	Robert Koski	Shounak Chattopadhyay	Ben Martin	Harry Reyes Nieva	Manlik Kwong	Wayde Shipman	Asieh Golozar				
Evidence Network Partners	Eye Care and Vision Research	FHIR and OMOP	Perinatal and Reproductive Health	Phenotype Development & Evaluation	Psychiatry	Rare Diseases					
Clair Blacketer	Paul Nagy	Sally Baxter	Cindy Cai	Kerry Goetz	Michelle Hribar	Davera Gabriel	Louisa Smith				
FHIR and OMOP	Generative AI & Analytics in Healthcare	GIS - Geographic Information Systems	HADES	Rehabilitation	Steering	Surgery and Perioperative Medicine	Themis				
Ben Hamlin	Guy Tsafnat	Martijn Schuemie	Robert Miller	Kyle Zollo-Veneek	Anthony Sena	Martijn Schuemie	Esther Janssen				
Health Economics and Value Assessment	Health Equity	Healthcare Systems	Industry	Medical Devices	Transplant	Vaccine Vocabulary	Women of OHDSI				
Gaurav Dravida	Gowtham Rao	Atif Amin	Melanie Philofsky	Paul Dougall	Sarah Seager	Asiyah Lin	Michal Mankowski				
Sarah Seager	Oliver Ho	Asiyah Lin	Sarah Seager	Oliver Ho	Asiyah Lin	Sarah Seager	Sarah Seager				
<b>Workgroups Homepage</b>											
In OHDSI, there is a home for you. Please visit our workgroups homepage to learn more about each group, find the meeting schedule and sign up to one or several workgroups!											
<a href="http://www.ohdsi.org/workgroups">www.ohdsi.org/workgroups</a>											



# Regional chapters and national nodes

Africa



Agnes Kiragga



Cynthia Sung

Asia-Pacific (APAC)



Mui Van Zandt

Australia



Nicole Pratt

China



Hua Xu

Europe



Peter Rijnbeek



Swetha Kiranmayi Jakkuv

India



Vikram Patil

Japan



Parthiban Sulur



Tatsuo Hiramatsu

Latin America



Julio Oliveira



Rae Woong Park

Republic of Korea



Seng Chan You

Singapore



Mengling 'Mornin' Feng

Taiwan



Jason Hsu

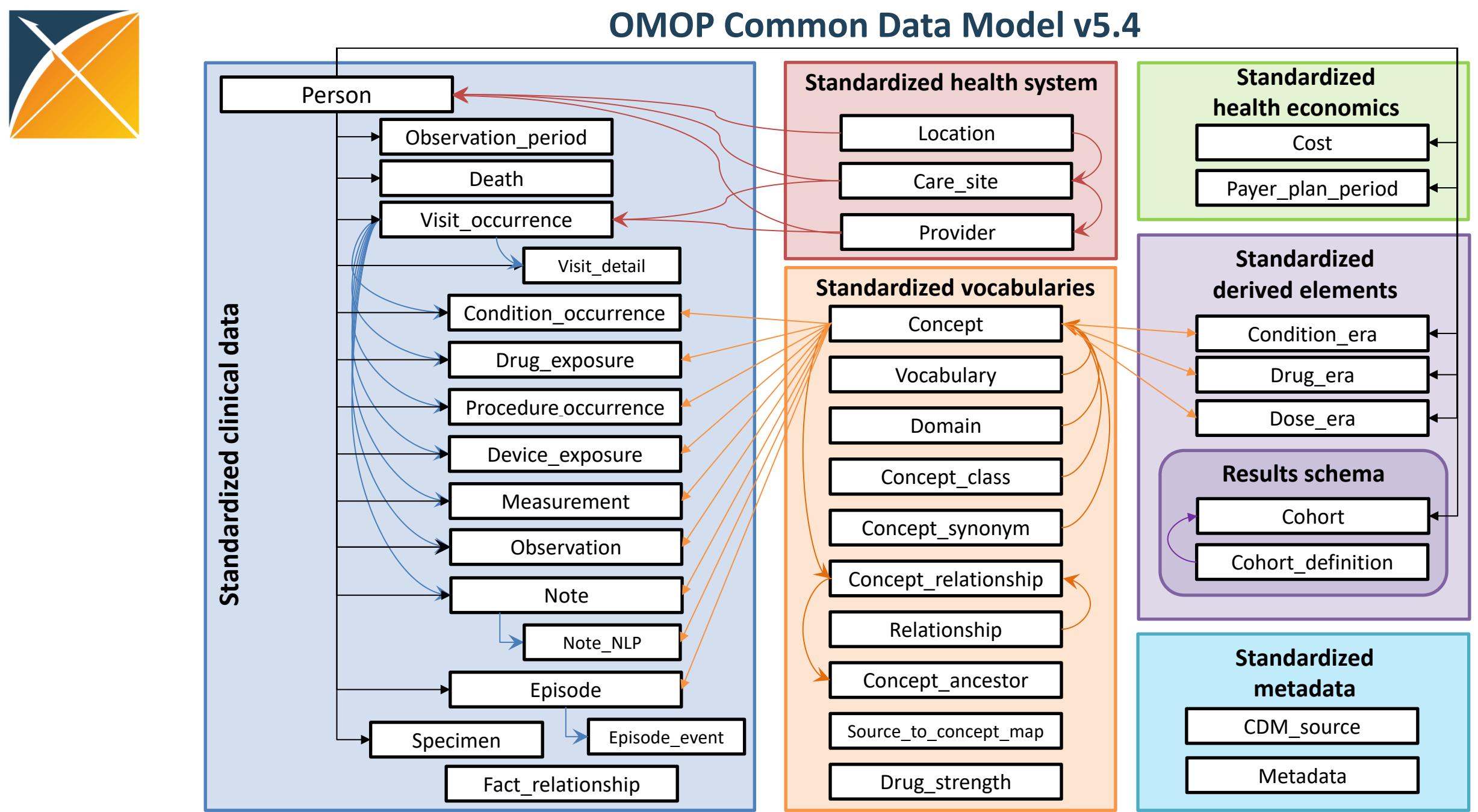
## Node.....

## Lead(s)

Belgium .....	Liesbet Peeters, Annelies Verbiest, Ilse Vermeulen
Denmark .....	Ismail Gögenur, Martin Höyer Rose, Andreas Weinberger Rosen
Estonia .....	Raivo Kolde, Sulev Reisberg
Finland .....	Eric Fey, Gustav Klingstedt
Germany .....	Ines Reinecke, Michele Zoch
Greece .....	Anastasia Farmaki, Pantelis Natsiavas, Grigoris Papapostolou
Hungary .....	Zsolt Bagyura, Ágota Mészáros
Ireland .....	Aedin Culhane, Mark Lawler, Catherine Mahoney
Israel .....	Chen Yanover
Italy .....	Lucia Sacchi, Matteo Gabbetta
Luxembourg .....	Claudine Backes, Andreas Kremer, Maria Quaranta
Netherlands .....	Renske Los, Aniek Markus
Norway .....	Espen Enerly, Siri Larønningen
Portugal .....	Patricia Couceiro, Carmen Nogueira
Spain .....	Miguel Angel Mayer, Talita Duarte Salles
Switzerland .....	Olga Endrich, Karen Triep
United Kingdom .....	Dani Prieto-Alhambra
<i>coming soon .....</i>	<i>Austria Sweden</i>

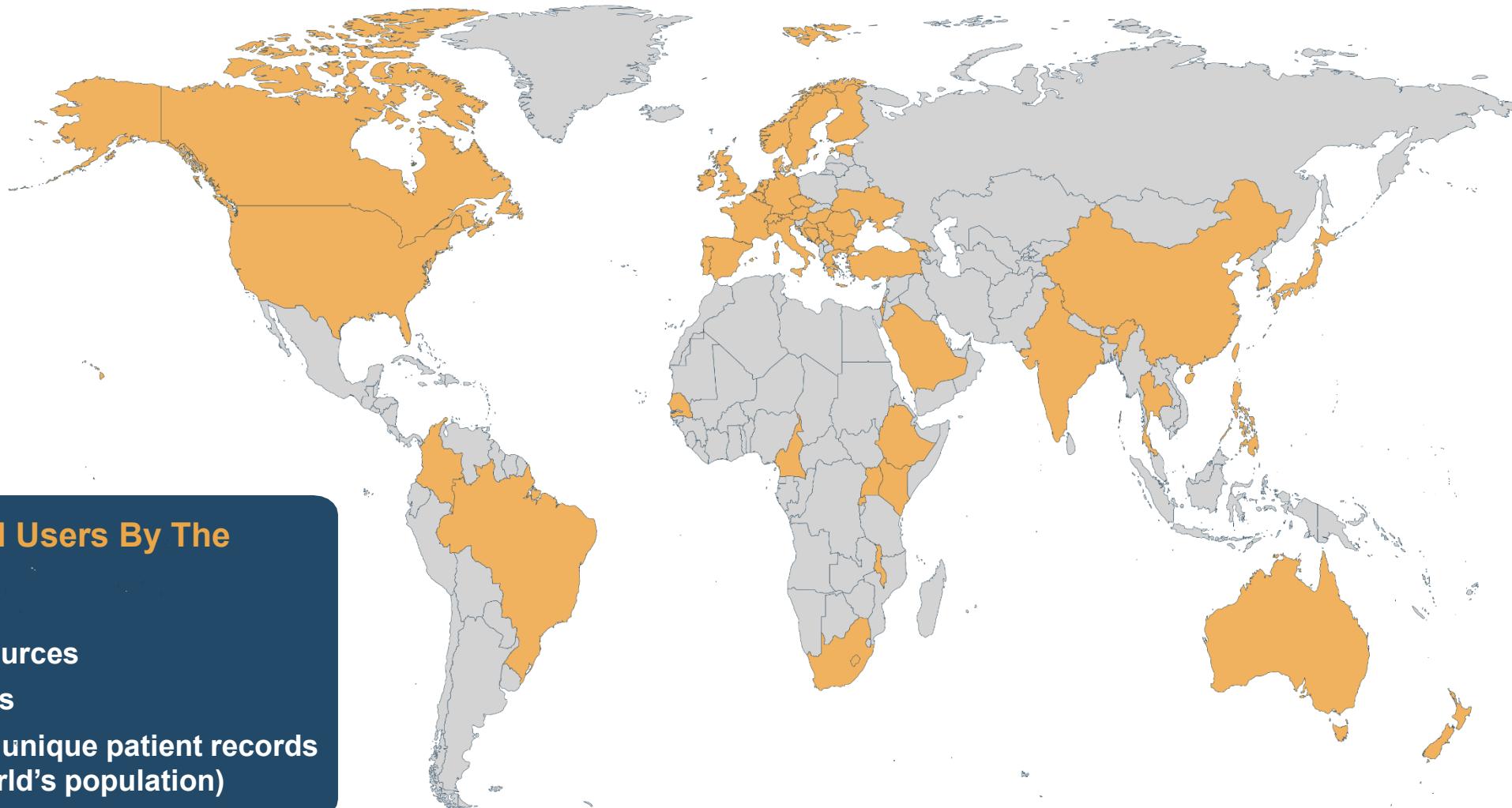


# OMOP Common Data Model v5.4



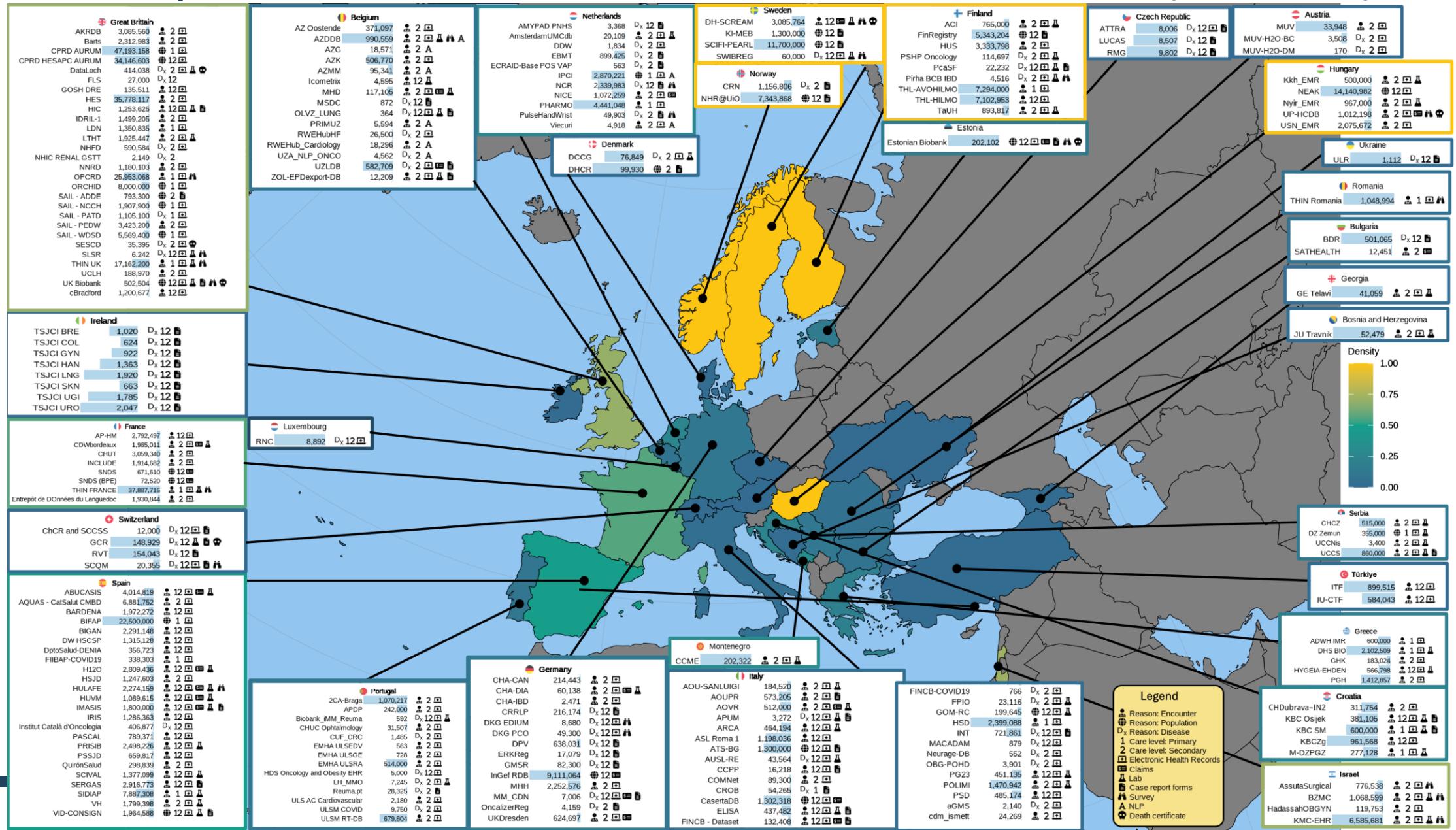


# OMOP Common Data Model adoption



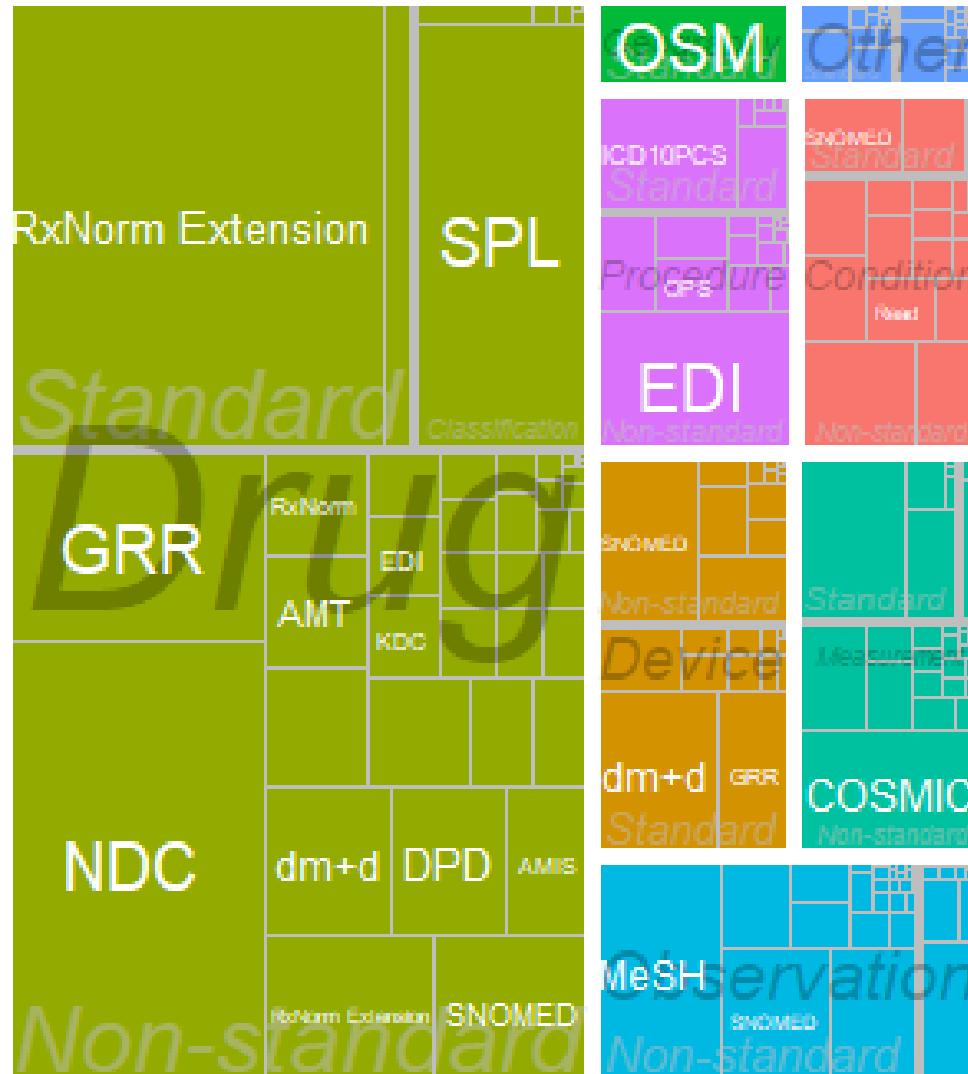


# European Health Data & Evidence Network (EHDEN)





# OHDSI standardized vocabularies

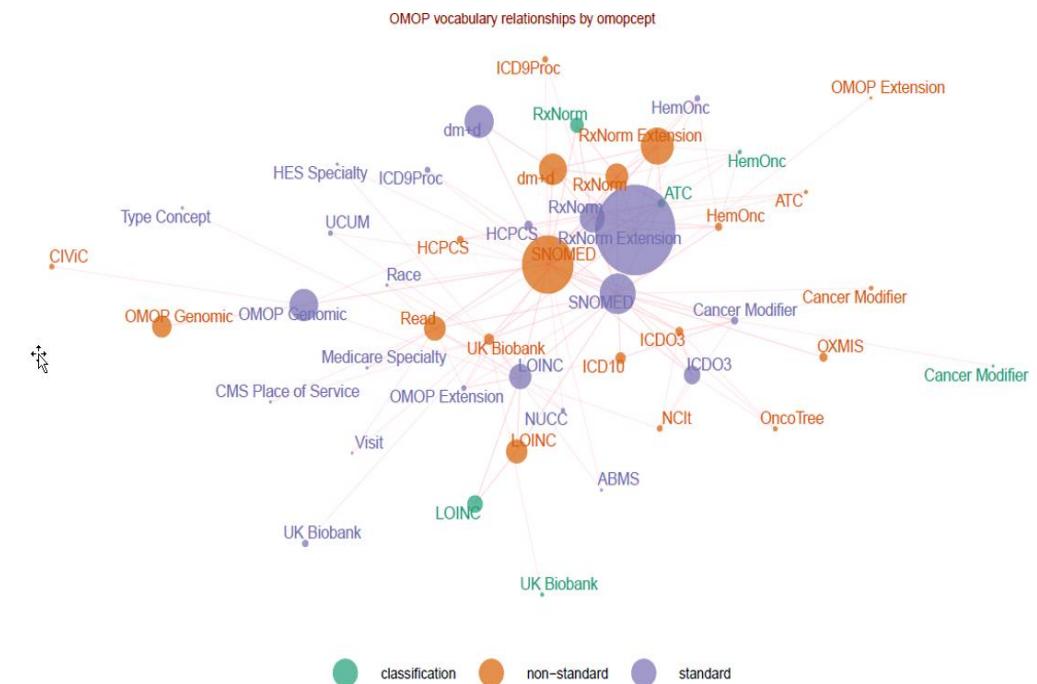


## OHDSI Vocabularies By The Numbers

as of August 2025 release

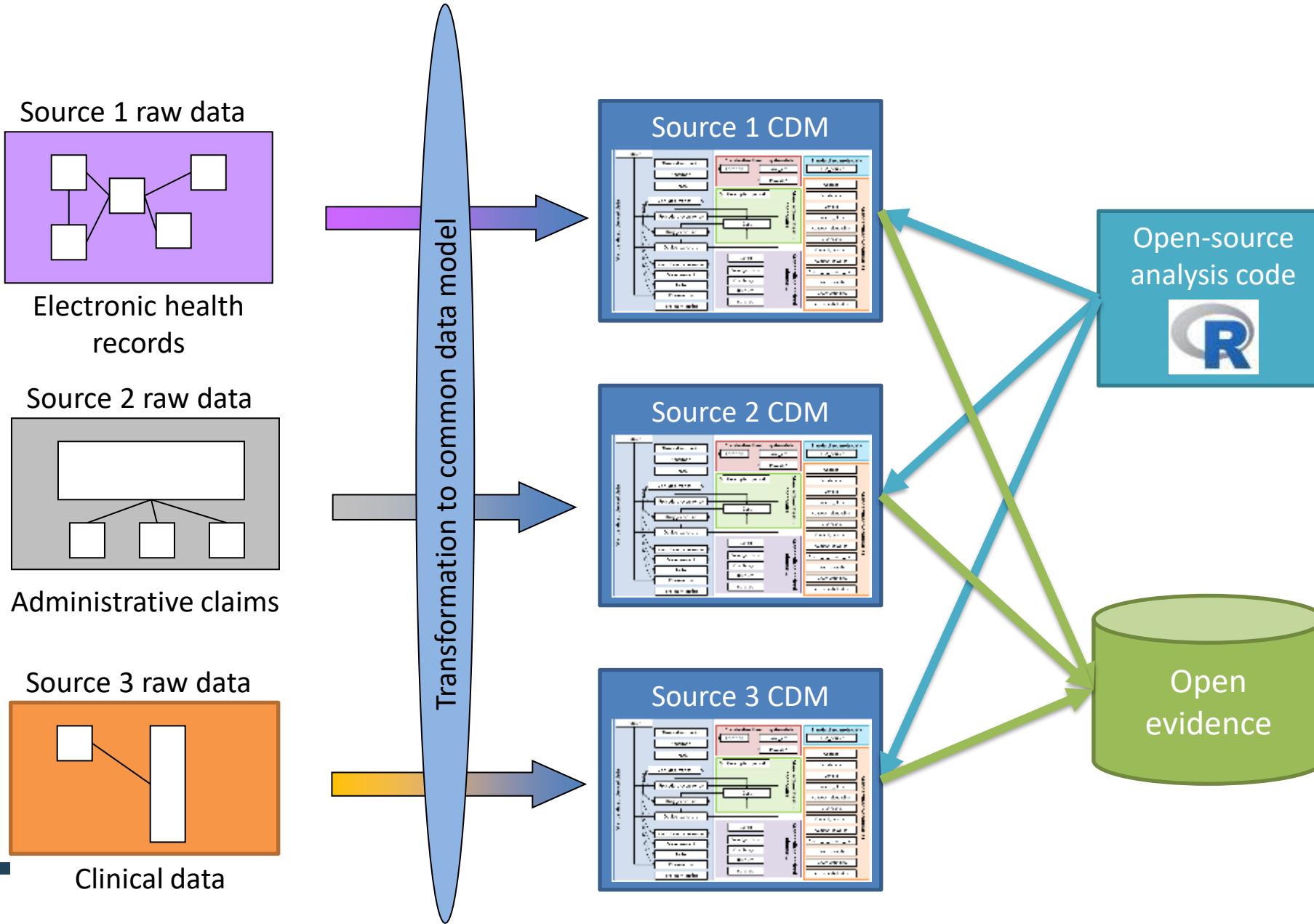
- 11,804,307 concepts
    - 3,784,263 standard concepts
    - 971,914 classification concepts
  - 145 vocabularies
  - 43 domains
- 87,948,636 concept relationships
- 101,696,159 ancestral relationships
- 6,028,711 concept synonyms

1 Shared Resource to Enable Data Standards

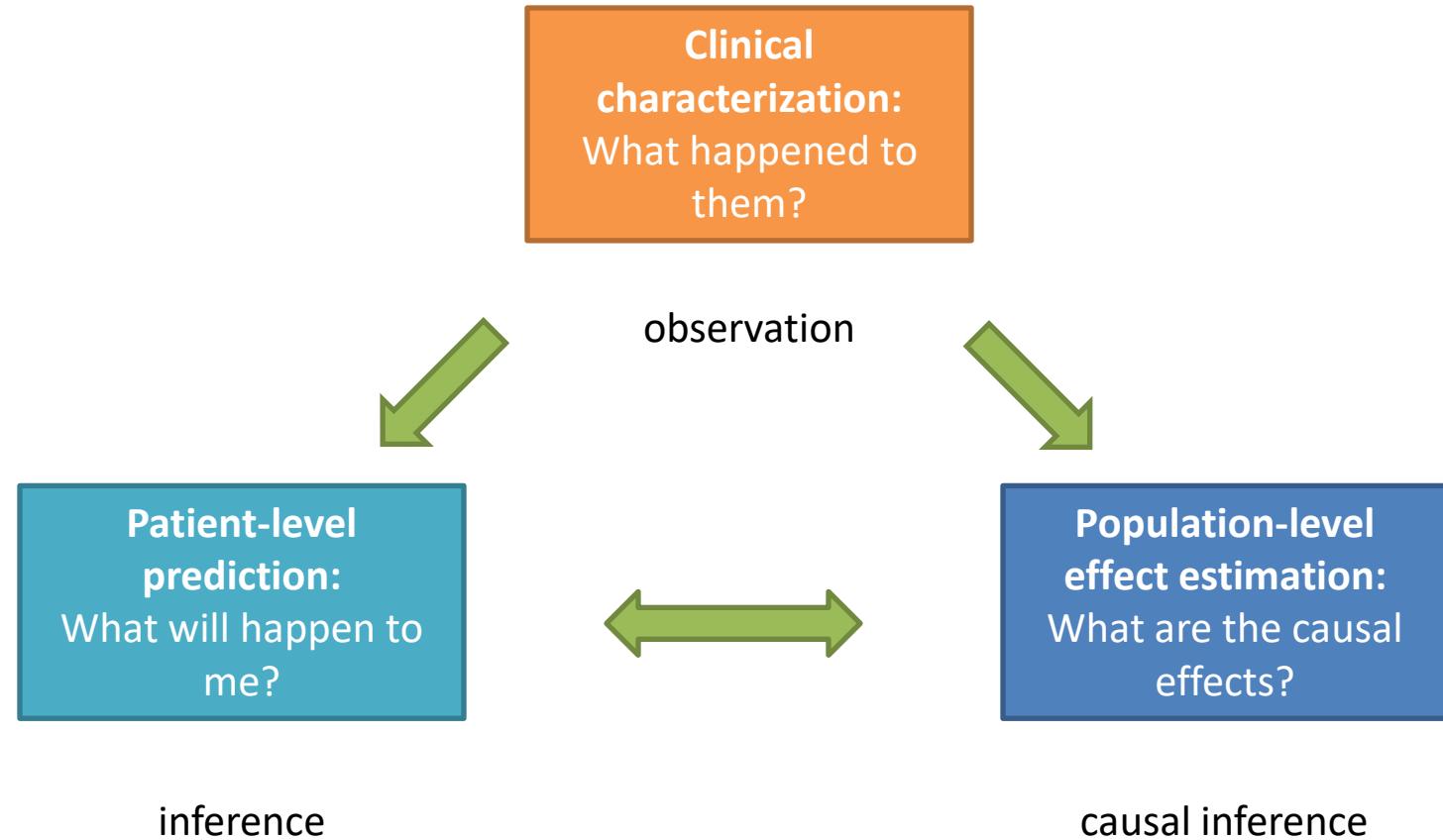




# Common data model can enable standardized analytics across a distributed data network



# Complementary evidence to inform the patient journey

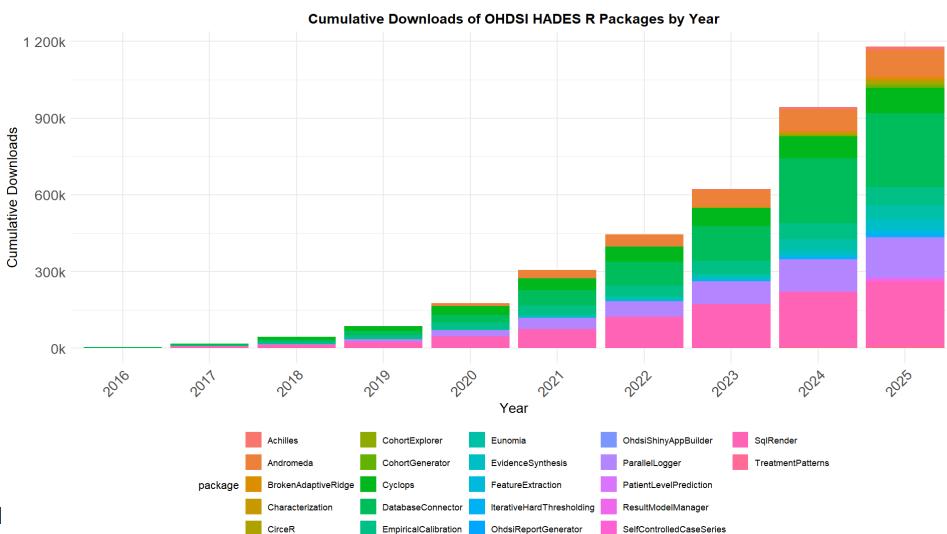


Analytic use case	Type	Structure	Example
Clinical characterization	Disease Natural History	Amongst patients who are diagnosed with <insert your favorite disease>, what are the patient's characteristics from their medical history?	Amongst patients with <b>rheumatoid arthritis</b> , what are their demographics (age, gender), prior conditions, medications, and health service utilization behaviors?
	Treatment utilization	Amongst patients who have <insert your favorite disease>, which treatments were patients exposed to amongst <list of treatments for disease> and in which sequence?	Amongst patients with <b>depression</b> , which treatments were patients exposed to <b>SSRI, SNRI, TCA, bupropion, esketamine</b> and in which sequence?
	Outcome incidence	Amongst patients who are new users of <insert your favorite drug>, how many patients experienced <insert your favorite known adverse event from the drug profile> within <time horizon following exposure start>?	Amongst patients who are new users of <b>methylphenidate</b> , how many patients experienced <b>psychosis</b> within <b>1 year of initiating treatment</b> ?
Population-level effect estimation	Safety surveillance	Does exposure to <insert your favorite drug> increase the risk of experiencing <insert an adverse event> within <time horizon following exposure start>?	Does exposure to <b>ACE inhibitor</b> increase the risk of experiencing <b>Angioedema</b> within <b>1 month after exposure start</b> ?
	Comparative effectiveness	Does exposure to <insert your favorite drug> have a different risk of experiencing <insert any outcome (safety or benefit)> within <time horizon following exposure start>, relative to <insert your comparator treatment>?	Does exposure to <b>ACE inhibitor</b> have a different risk of experiencing <b>acute myocardial infarction</b> while <b>on treatment</b> , relative to <b>thiazide diuretic</b> ?
Patient level prediction	Disease onset and progression	For a given patient who is diagnosed with <insert your favorite disease>, what is the probability that they will go on to have <another disease or related complication> within <time horizon from diagnosis>?	For a given patient who is <b>newly diagnosed with atrial fibrillation</b> , what is the probability that they will go onto to have <b>ischemic stroke in next 3 years</b> ?
	Treatment response	For a given patient who is a new user of <insert your favorite chronically-used drug>, what is the probability that they will <insert desired effect> in <time window>?	For a given patient <b>with T2DM who start on metformin</b> , what is the probability that they will <b>maintain HbA1C&lt;6.5% after 3 years</b> ?
	Treatment safety	For a given patient who is a new user of <insert your favorite drug>, what is the probability that they will experience <insert adverse event> within <time horizon following exposure>?	For a given patients who is a <b>new user of warfarin</b> , what is the probability that they will have <b>GI bleed in 1 year</b> ?



# Open-source software development

- HADES is an ecosystem of 41 R packages to support standardized analytics for the OMOP CDM and across OHDSI network
- OHDSI CRAN packages (n=22) have been downloaded >1m times



Package	Version	Maintainer(s)	Availability
Achilles	v2.1.2	Frank DeFalco	CRAN
Andromeda	v3.1.1	Martijn Schuemie	CRAN
Birkn	v3.0.2	Martijn Schuemie	Github
BrokenAdaptiveRidge	v1.1.1	Marc Suchard	CRAN
Caer	v2.1.0	Martin Lavalee	Github
Characterization	v2.2.0	Jenna Reps	CRAN
CirceR	v1.1.3	Chris Knoll	CRAN
CohortDiagnostics	v3.4.2	Jamie Gilbert	Github
CohortExplorer	v0.1.0	Gowtham Rao	CRAN
CohortGenerator	v0.12.3	Anthony Sena	CRAN
CohortIncidence	v4.1.0	Chris Knoll	Github
CohortMethod	v5.5.0	Martijn Schuemie	Github
Cycles	v3.5.0	Marc Suchard	CRAN
DatabaseConnector	v6.1.0	Martijn Schuemie	CRAN
DataQualityDashboard	v2.7.0	Katy Sadowski	Github
DecoratePatientLevelPrediction	v2.1.0	Egill Fridgeirsson	Github
EmpiricalCalibration	v3.1.4	Martijn Schuemie	CRAN
EnsemblePatientLevelPrediction	v1.3.2	Jenna Reps	Github
Eunomia	v2.1.0	Frank DeFalco	CRAN
EvidenceSynthesis	v1.0.0	Martijn Schuemie	CRAN
FeatureExtraction	v1.1.6	Ger Inberg	CRAN
Hydra	v0.4.0	Anthony Sena	Deprecated
IterativeHardThresholding	v1.0.3	Marc Suchard	CRAN
Keoper	v0.1.1	Anna Ostropolski	Github
MethodEvaluation	v2.4.0	Martijn Schuemie	Github
OhdsiReportGenerator	v1.1.1	Jenna Reps	CRAN
OhdsiShaper	v0.1.2	Lee Evans	Github
OhdsiShinyAppBuilder	v1.0.0	Jenna Reps	CRAN
OhdsiShinyModules	v1.1.0	Jenna Reps	Github
ParallelLogger	v3.5.0	Martijn Schuemie	CRAN
PatientLevelPrediction	v6.5.0	Egill Fridgeirsson & Jenna Reps	CRAN
PhenotypeLibrary	v1.16.0	Gowtham Rao	Github
Phenotuator	v2.2.15	Joel Swerdel	Github
ResultModelManager	v0.5.13	Jamie Gilbert	CRAN
SelfControlledCaseSeries	v1.1.3	Gowtham Rao	Github
SelfControlledOhort	v1.6.0	Jamie Gilbert	Github
ShowcaseBuilder	v2.2.0	Jenna Reps	Deprecated
SciRender	v1.16.3	Martijn Schuemie	CRAN
Stratus	v1.4.1	Anthony Sena	Github
TreatmentPatterns	v3.3.1	Maarten van Kessel	CRAN

The open-source tools that empower OHDSI research are not only available to the community, but they are DEVELOPED by the community. We thank the many developers and maintainers who empower our research initiatives around the world!

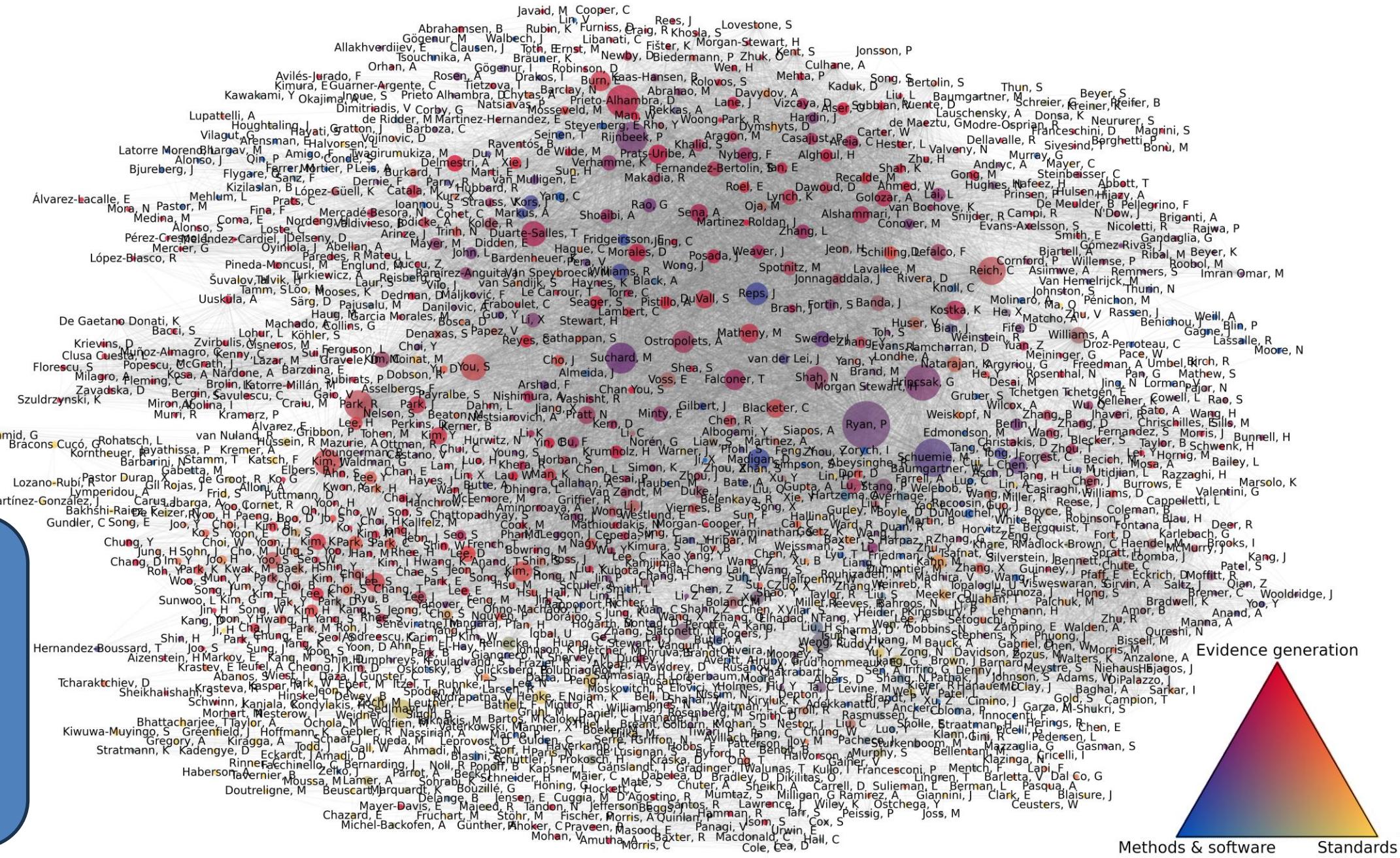




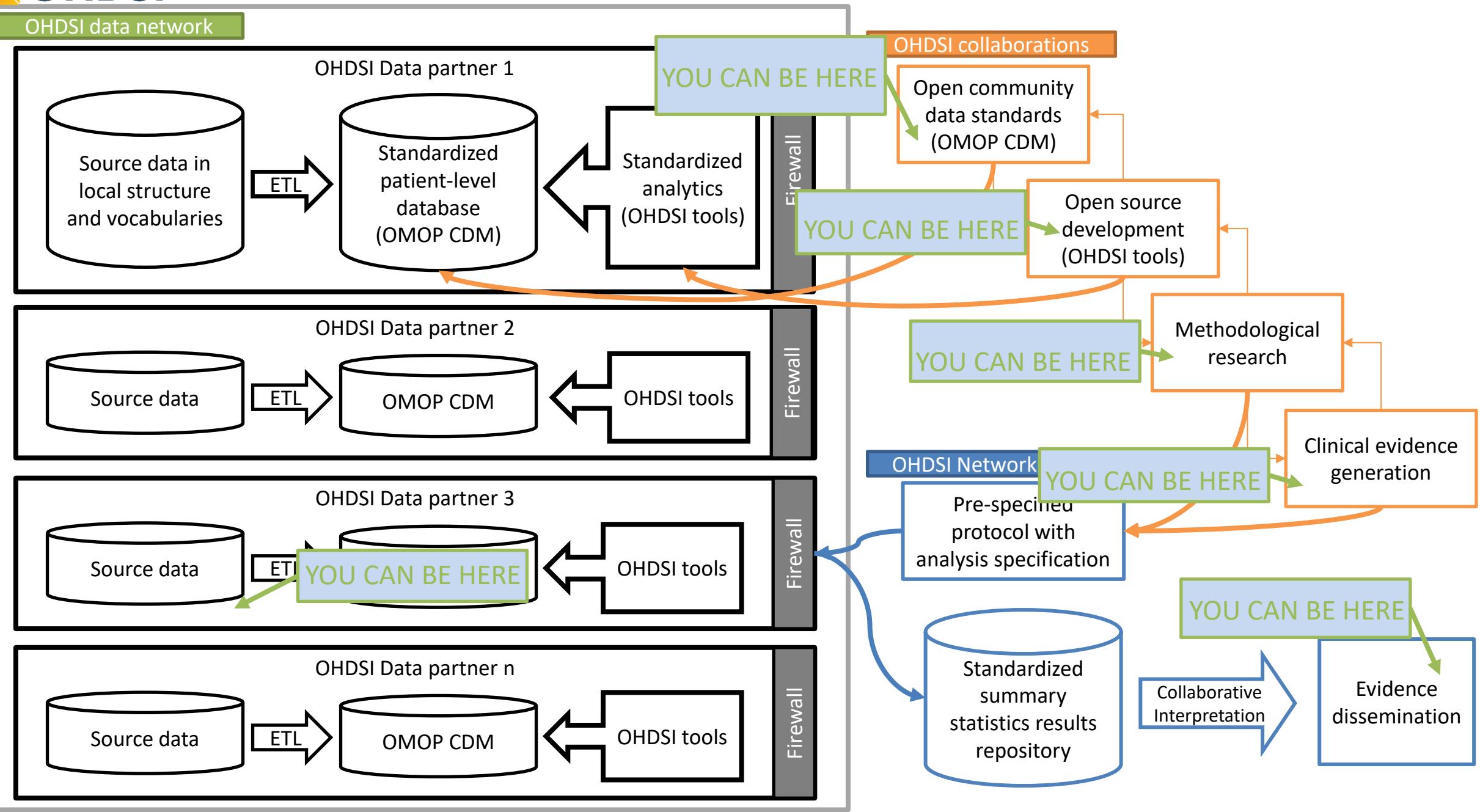
# OHDSI collaborations in scholarship

>900

publications, including  
in top clinical journals  
(JAMA, BMJ, Lancet,  
JAMA Internal  
Medicine, JACC) and  
leading methodological  
journals (JAMIA, JBI,  
Nature Digital  
Medicine)



## OHDSI data network





# Why OHDSI and the world needs Sweden

## The Legatum Prosperity Index™ Ranks 1–56

European Journal of Epidemiology (2025) 40:563–579  
<https://doi.org/10.1007/s10654-025-01226-9>

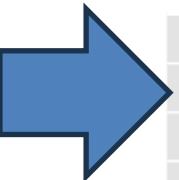
DATA RESOURCE



### The healthcare system in Sweden

Jonas F. Ludvigsson<sup>1,2,3</sup> · David Bergman<sup>1</sup> · Catharina Ihre Lundgren<sup>4,5</sup> · Kristina Sundquist<sup>6</sup> · Jean-Luc af Geijerstam<sup>7</sup> · Anna H. Glenngård<sup>8</sup> · Marie Lindh<sup>9</sup> · Johan Sundström<sup>10,11</sup> · Johan Kaarme<sup>12,13</sup> · Jialu Yao<sup>1</sup>

Received: 10 March 2025 / Accepted: 23 March 2025 / Published online: 19 May 2025  
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2013 rank	2022 rank	2023 rank	Country
4	1	1	Denmark
1	3	2	Sweden
2	2	3	Norway
5	4	4	Finland
3	5	5	Switzerland
6	6	6	Netherlands
8	7	7	Luxembourg
14	9	8	Iceland
9	8	9	Germany
7	10	10	New Zealand

#### Abstract

The Swedish population is characterized by high life expectancy and low avoidable mortality rates. This review outlines the Swedish healthcare system, which offers universal access to all residents and has a long tradition of reforms for social equity. Responsibility for healthcare is shared between the state, the regions, and the municipalities. The Ministry of Health and Social Affairs provides the overall healthcare framework; additionally, several governmental agencies are directly involved in healthcare and public health initiatives. The 21 regions organize, finance, and provide most primary, secondary, and tertiary care, as well as health information channels. Resources for primary care are less plentiful than in many other countries. The 290 municipalities deliver care to elderly people and those with functional impairment. The Swedish healthcare system is primarily tax-funded, with 86% of total healthcare expenditures from public expenses and <1% from voluntary health insurance. The gross domestic product (GDP) share of healthcare expenditures, 10.5% in 2022, is above the EU average. The level of unmet needs in the population is low, due to universal coverage and caps on user charges except for dental care. Sweden's healthcare system performs well on care quality and patient satisfaction, but suffers from workforce shortage and care fragmentation. Limitations in care coordination can be attributed to a siloed digital infrastructure and care governance, a low number of hospital beds per capita, and a compensation system that often does not incentivize coordination. Despite these challenges, life expectancy is high and avoidable mortality rates are low in Sweden.

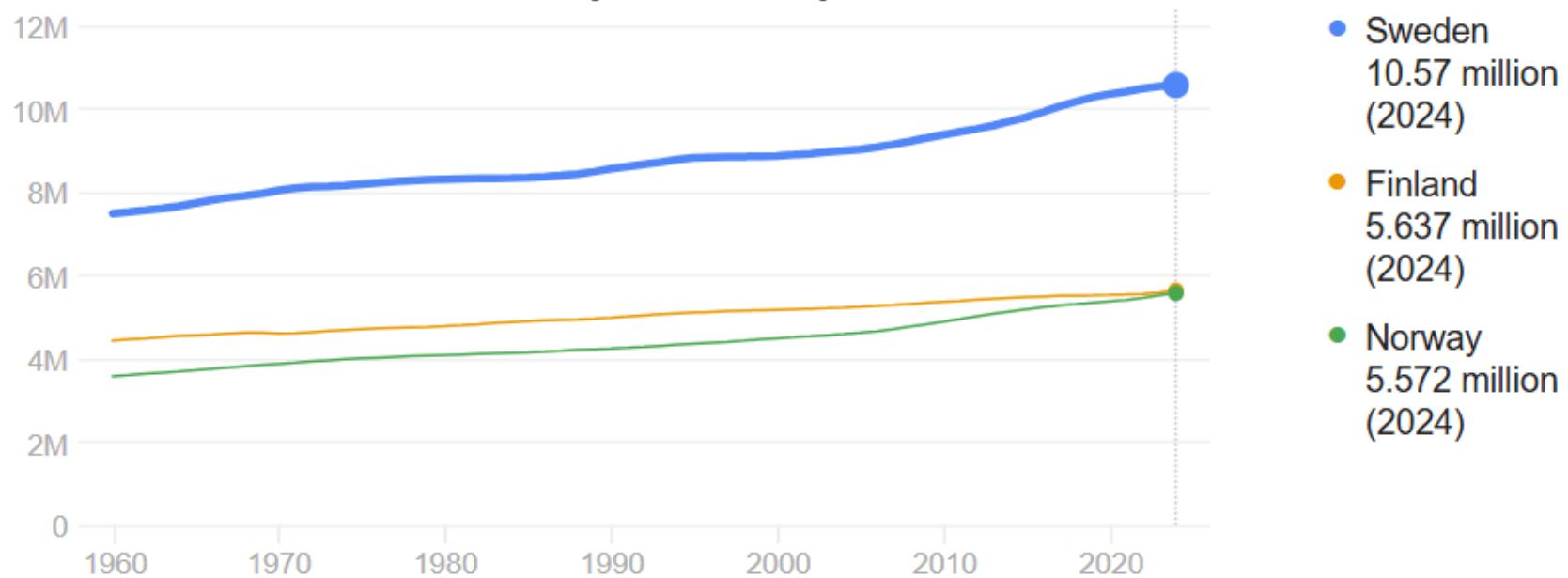
Health	Education	Natural Environment
16	5	5
9	8	1
7	10	9
15	2	2
10	4	6
11	6	39
12	36	10
8	11	13
13	19	12
25	18	11



# Why Sweden needs OHDSI and the rest of the world

Sweden / Population

10.57 million (2024)





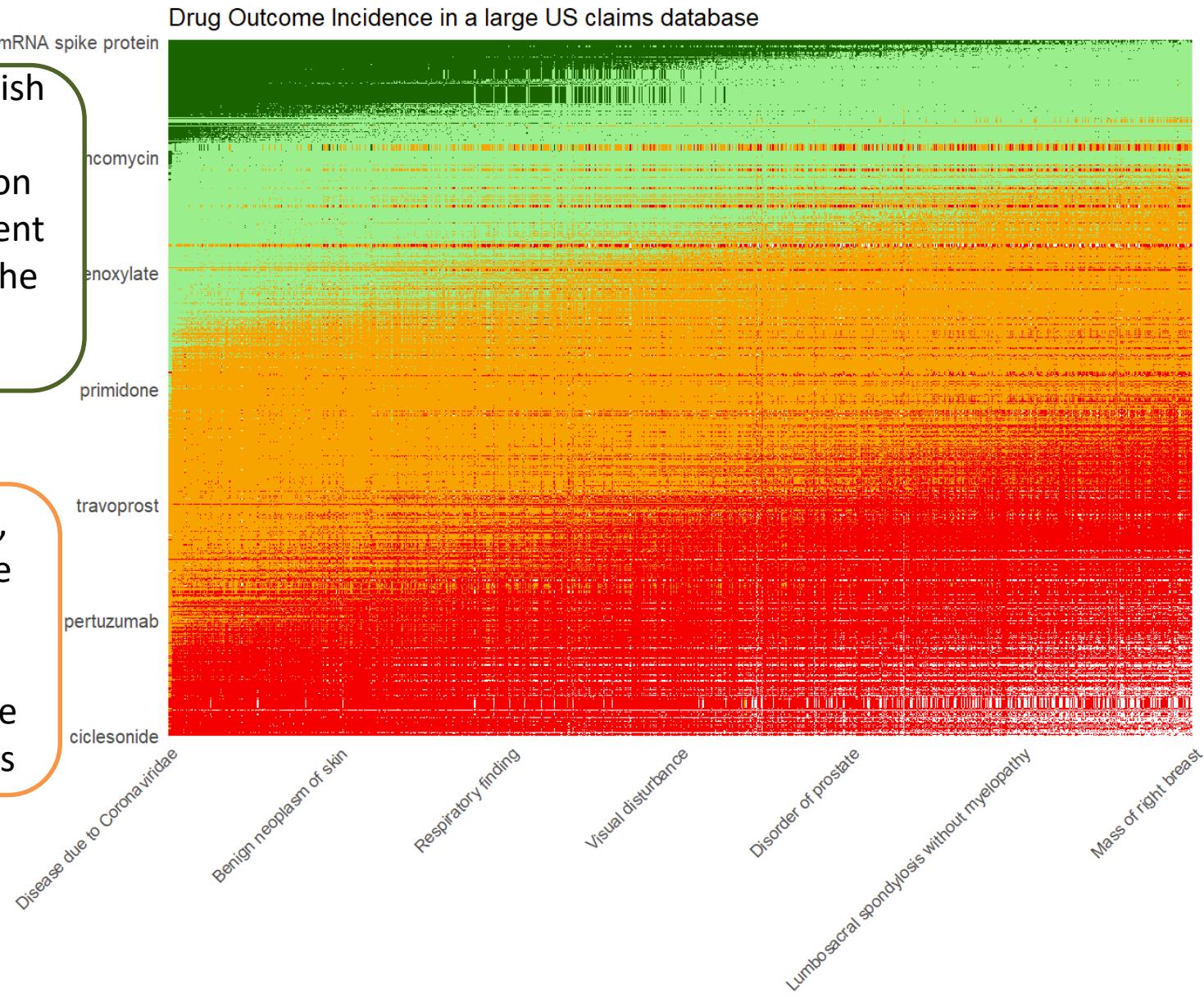
# Why Sweden needs OHDSI and the rest of the world

SARS-CoV-2 (COVID-19) vaccine, mRNA spike protein

If full health data for the entire Swedish population of 10 million were accessible, then ~10% of questions on drug-outcome pairs may have sufficient statistical power to answer alone if the incidence are comparable to US population....

Top 100

...but even with a national database, >70% of drug-outcome questions are likely to have insufficient data for prediction or estimation, so an international network study would be required to generate reliable answers





# Join the Journey!

<https://www.ohdsi.org/join-the-journey/>

Looking forward to collaborating with all of you!

Email me: [ryan@ohdsi.org](mailto:ryan@ohdsi.org)