



# INTRODUCTION TO OHDSI

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Laura Verbeij, Fleur Vereijken & Julia Kurps



# WELCOME!

EUROPE

## EUROPEAN OHDSI SYMPOSIUM Rotterdam

### 18 - 20 April 2026



How to start your OHDSI journey & Patient Reported Outcomes

**Laura**



OHDSI tools, prediction modelling & individual prediction instability

OHDSI community, All things OMOP CDM

**Julia**



**Fleur**



# THIS MORNING...

Topic	Presenter
History & philosophy behind OHDSI	Laura
How does the community work?	Julia
Where and how can you learn more about OHDSI?	Laura
<b>Q&amp;A</b>	
All things OMOP CDM (model, ETL, DQ, conventions)	Julia
What can be done currently with the OHDSI tools?	Fleur
What does it take to be able to use the tools?	Fleur
<b>Q&amp;A</b>	

More info?

# WHAT IS OHDSI?





# THIS IS OHDSI!



*OHDSI is a fun way to collaborate with amazing people across the globe to collectively advance science and improve the lives of patients around the world.*

- Since 2000 rapidly increasing number of drug side effects
- FDA Amendments Act → Risk Identification and Analysis System

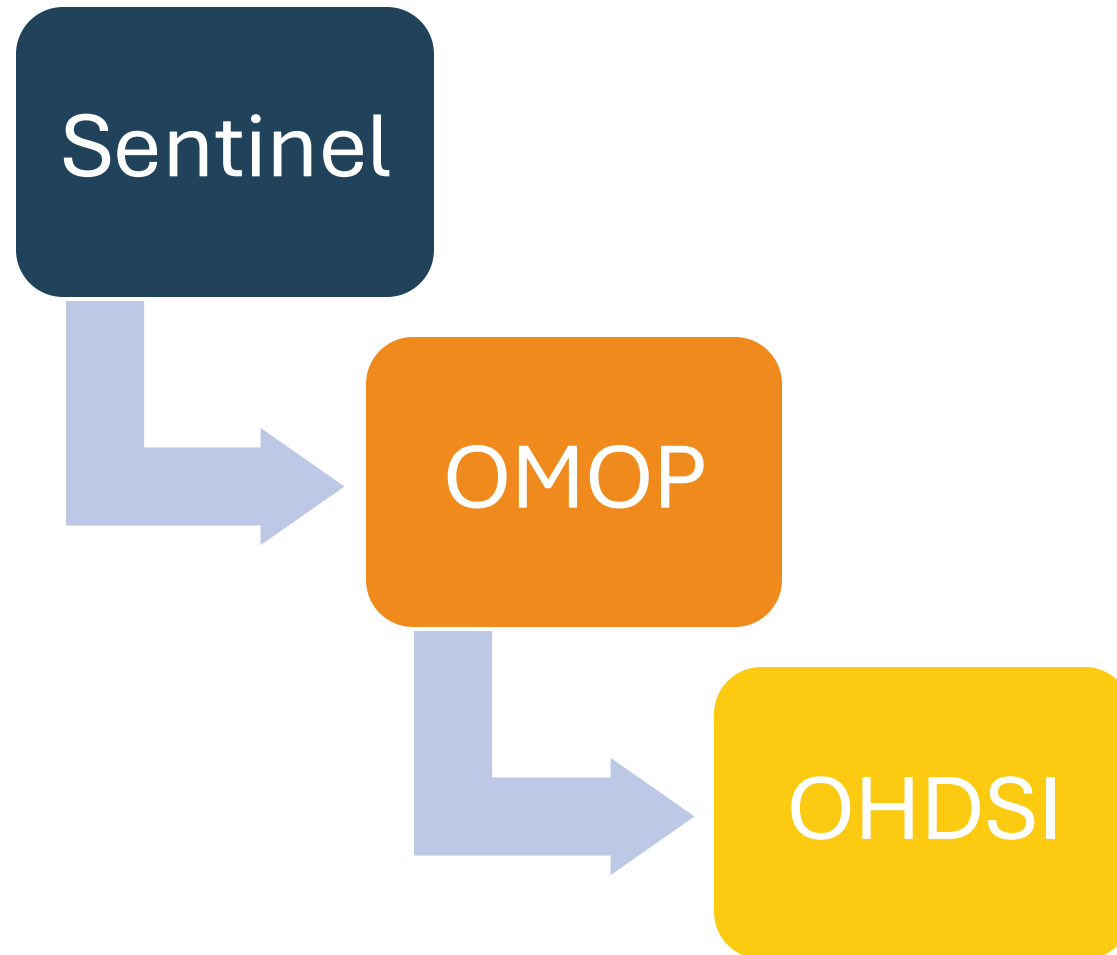


# OMOP EXPERIMENT (2009 →)

- Empirically reproduce a set of known (medical) facts using positive and negative test cases. The assumption was that this could be achieved with the best database and the best methods.
- 10 databases; 14 methods

- Different database, same method, potentially different outcomes
- Analysis choices are crucial







## Vision

A world in which observational research produces a comprehensive understanding of health and disease.

## Mission

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

- Interdisciplinary teams
- International collaboration
- All tools should remain public and freely available

### More info?

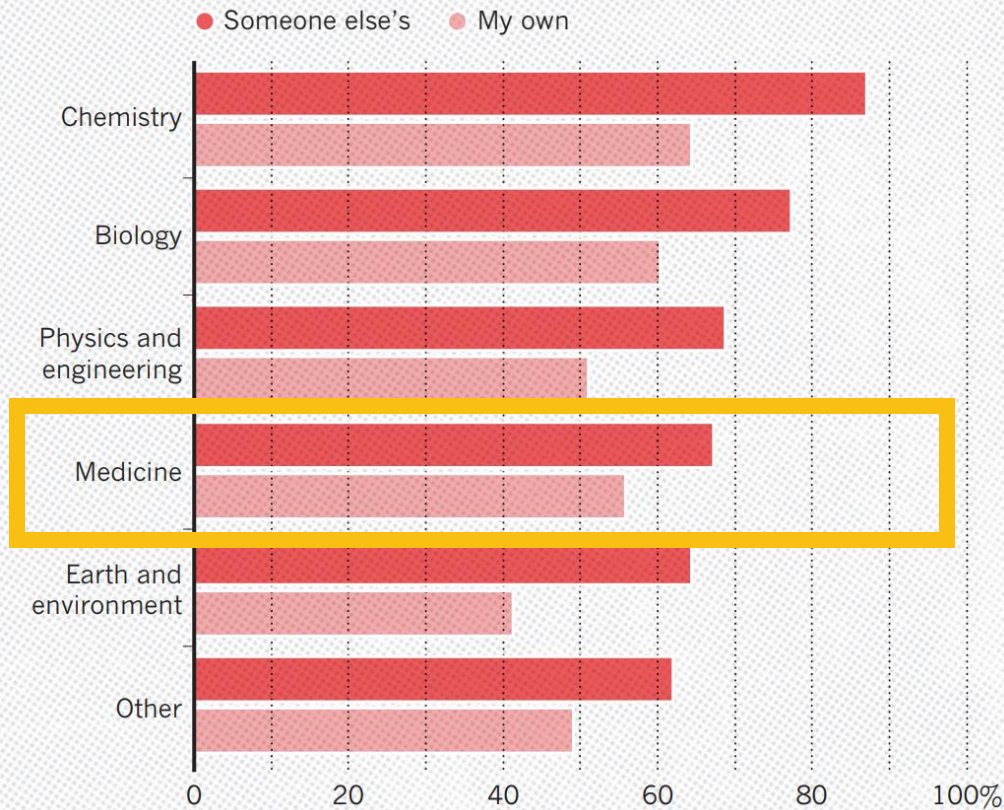
- **The Book of OHDSI**  
chapter 1.2
- **EHDEN academy course**  
“OMOP CDM and  
Standardised  
Vocabularies”

- **Innovation:** Observational research is a field which will benefit greatly from disruptive thinking. We actively seek and encourage fresh methodological approaches in our work.
- **Reproducibility:** Accurate, reproducible, and well-calibrated evidence is necessary for health improvement.
- **Community:** Everyone is welcome to actively participate in OHDSI, whether you are a patient, a health professional, a researcher, or someone who simply believes in our cause.
- **Collaboration:** We work collectively to prioritize and address the real world needs of our community's participants.
- **Openness:** We strive to make all our community's proceeds open and publicly accessible, including the methods, tools and the evidence that we generate.
- **Beneficence:** We seek to protect the rights of individuals and organizations within our community at all times.

# REPRODUCIBILITY CRISIS (NATURE 2016)

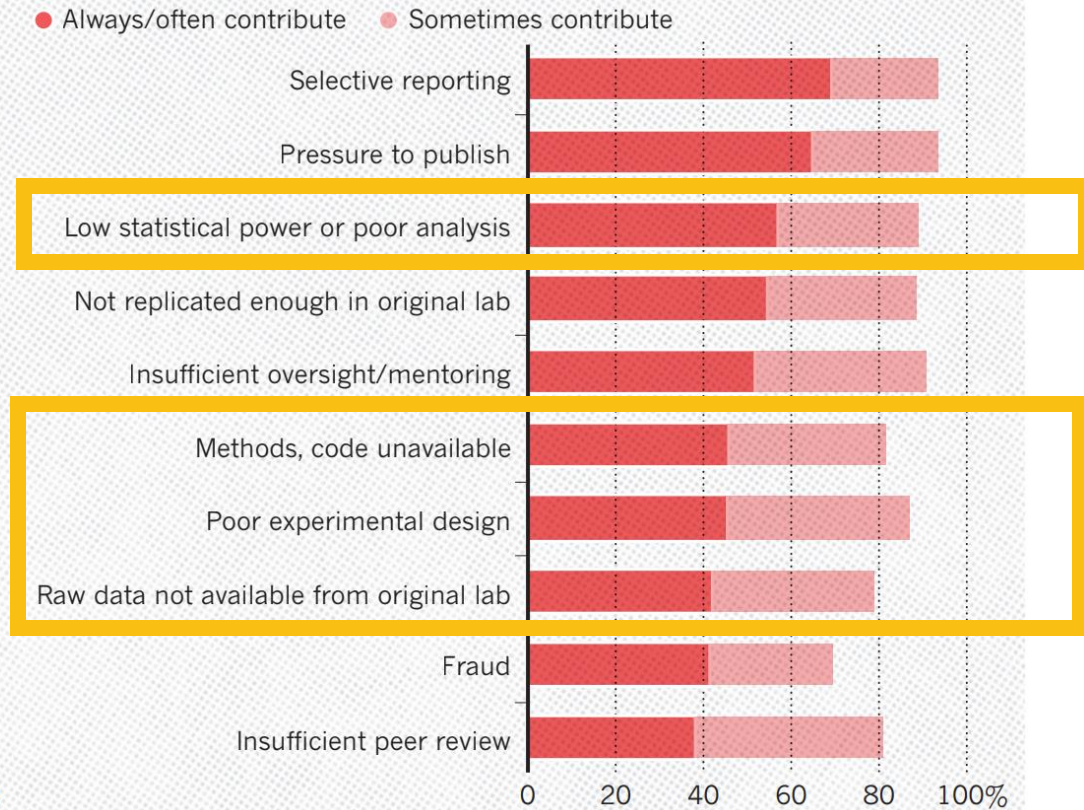
## HAVE YOU FAILED TO REPRODUCE AN EXPERIMENT?

Most scientists have experienced failure to reproduce results.

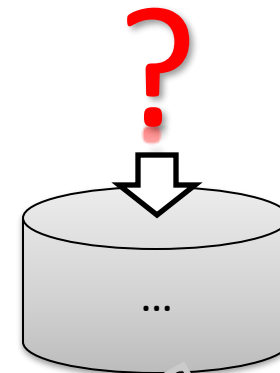
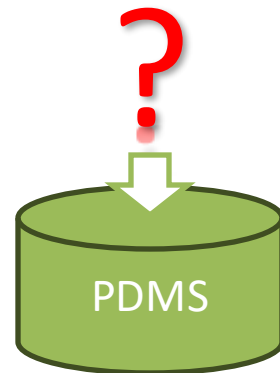
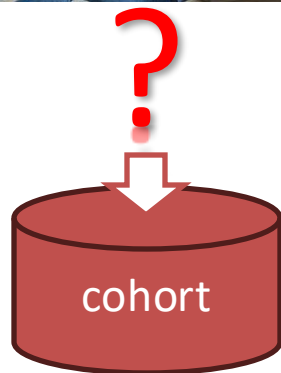


## WHAT FACTORS CONTRIBUTE TO IRREPRODUCIBLE RESEARCH?

Many top-rated factors relate to intense competition and time pressure.



Baker, M. [1,500 scientists lift the lid on reproducibility](#). *Nature* 533, 452–454 (2016)



**Why Most Published Research Findings Are False**

Open access. Free available online.

Since the 1950s, researchers and their reviewers have been misled by the publication of false research findings. This is because of the way that research is conducted and reported. The most common reason for this is that researchers only report the results that they want to see. This is known as 'cherry-picking'.

It can be proven that most published research findings are false.

**Abstract**

**Introduction**

**Discussion**

**Conclusion**

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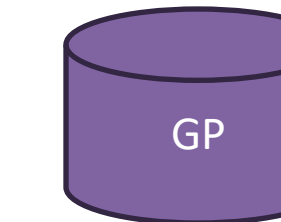
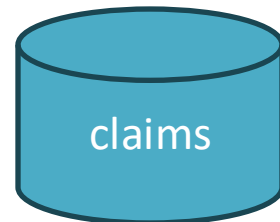
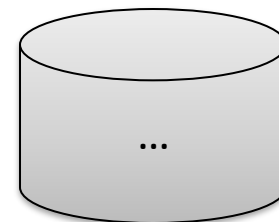
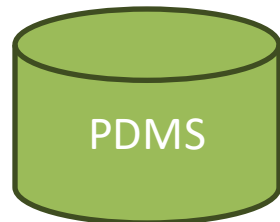
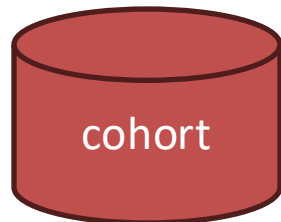
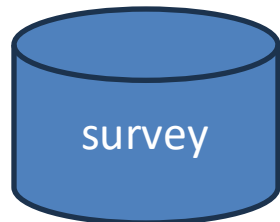
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**Discussion**

**Conclusion**



# OLD-SCHOOL RESEARCH GONE WRONG



**Why Most Published Research Findings Are False**

**Published research findings are more likely to be false than true.**

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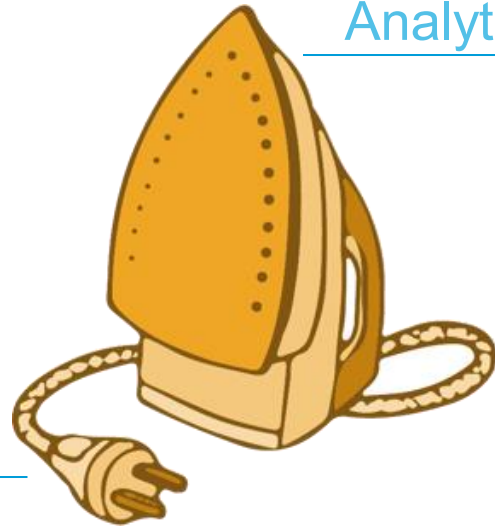
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# THE CHALLENGES OF REAL-WORLD DATA

Analytical method



Link to data

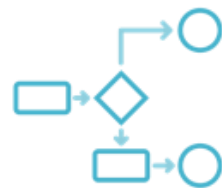
The data...



## What will it require?



Data interoperability



Standardized analytics



Data network



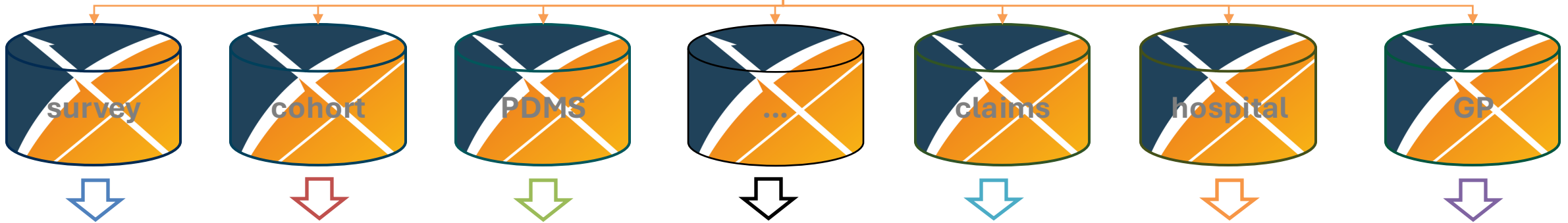
Strong community

# OBSERVATIONAL RESEARCH OHDSI STYLE



Strong community

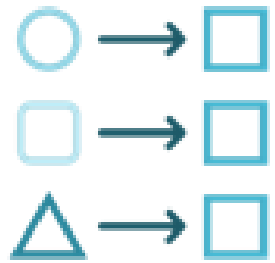
Standardized analytics



Data aggregation & analysis



# DATA INTEROPERABILITY



Data interoperability



OMOP CDM

Syntactic

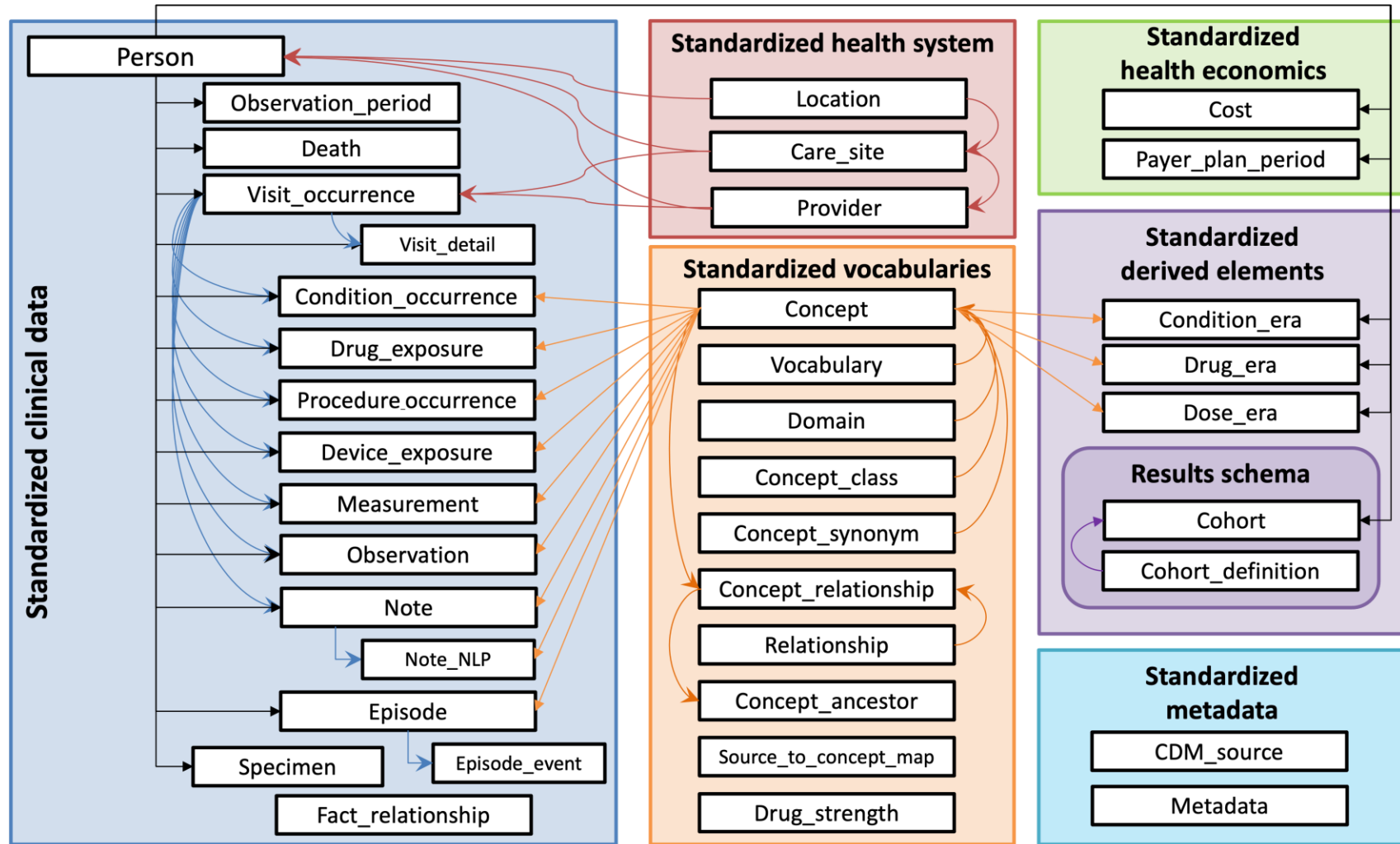
Interoperability

Semantic



# OMOP COMMON DATA MODEL (CDM) v5.4

Syntactic interoperability



More info?  
 - The Book of OHDSI chapter 4



# (STANDARDIZED) VOCABULARIES



Semantic  
interoperability



SEARCH

DOWNLOAD

LOGIN



## Search

Search

1. Usage of quotation marks forces an exact-match search
2. In case of a typo, or if there is a similar spelling of the word, the most similar result will be presented

## Explore domains



**Drugs**

5,511,696



**Conditions**

705,086



**Procedures**

740,521



**Devices**

504,264



**Observations**

906,578



**Measurements**

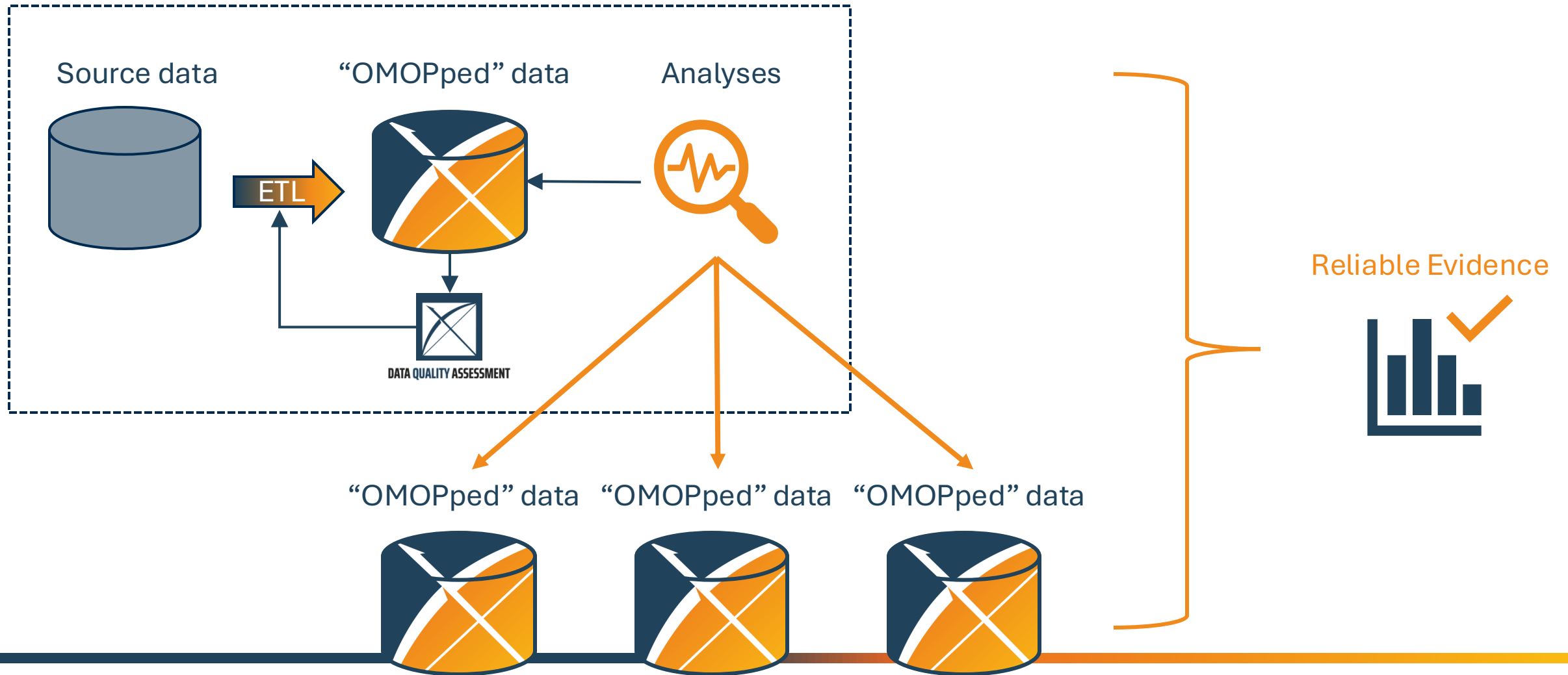
370,099

More info?

- The Book of OHDSI chapter 5
- [athena.ohdsi.org](http://athena.ohdsi.org)

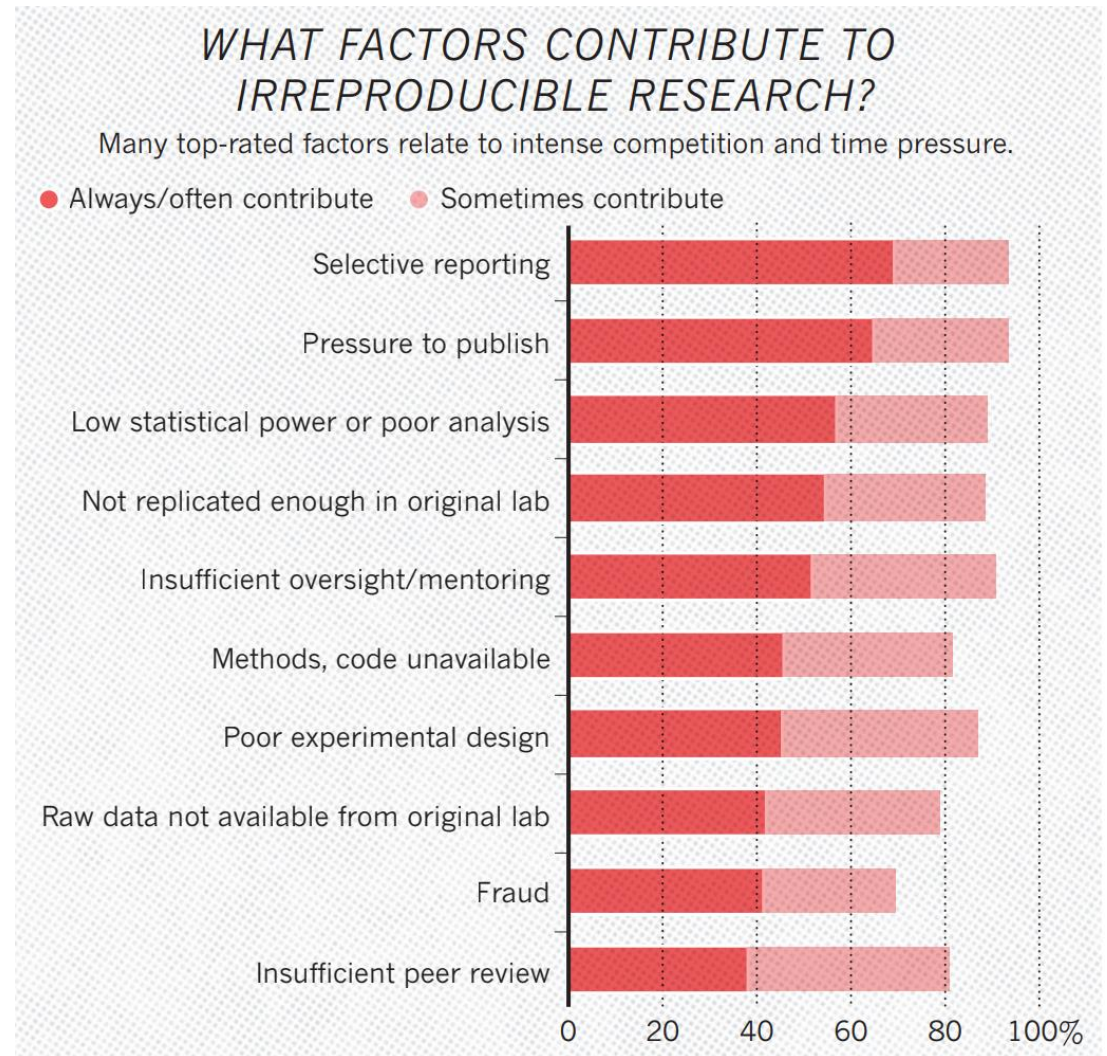


# STANDARDIZED ANALYTICS CAN BE REUSED OVER DATA SOURCES



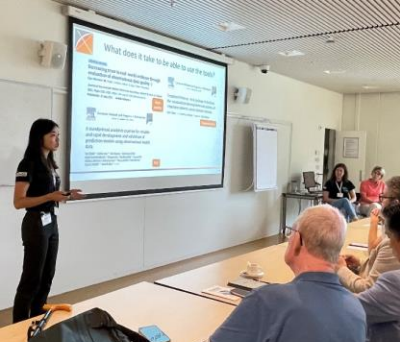
# IN SUMMARY, WHAT IS OHDSI?

- An awesome community
- A common data model
- Standardized open-source tools
- A gateway to a large international data network



- Take a few minutes to get to know your neighbour, perhaps you could talk about:
  - **What do you do?**
  - **What do you hope that OMOP/OHDSI will bring you?**
  - **What are you hoping to find at the symposium?**

# HOW DOES THE COMMUNITY WORK?





## OHDSI COLLABORATORS

### Map of Collaborators

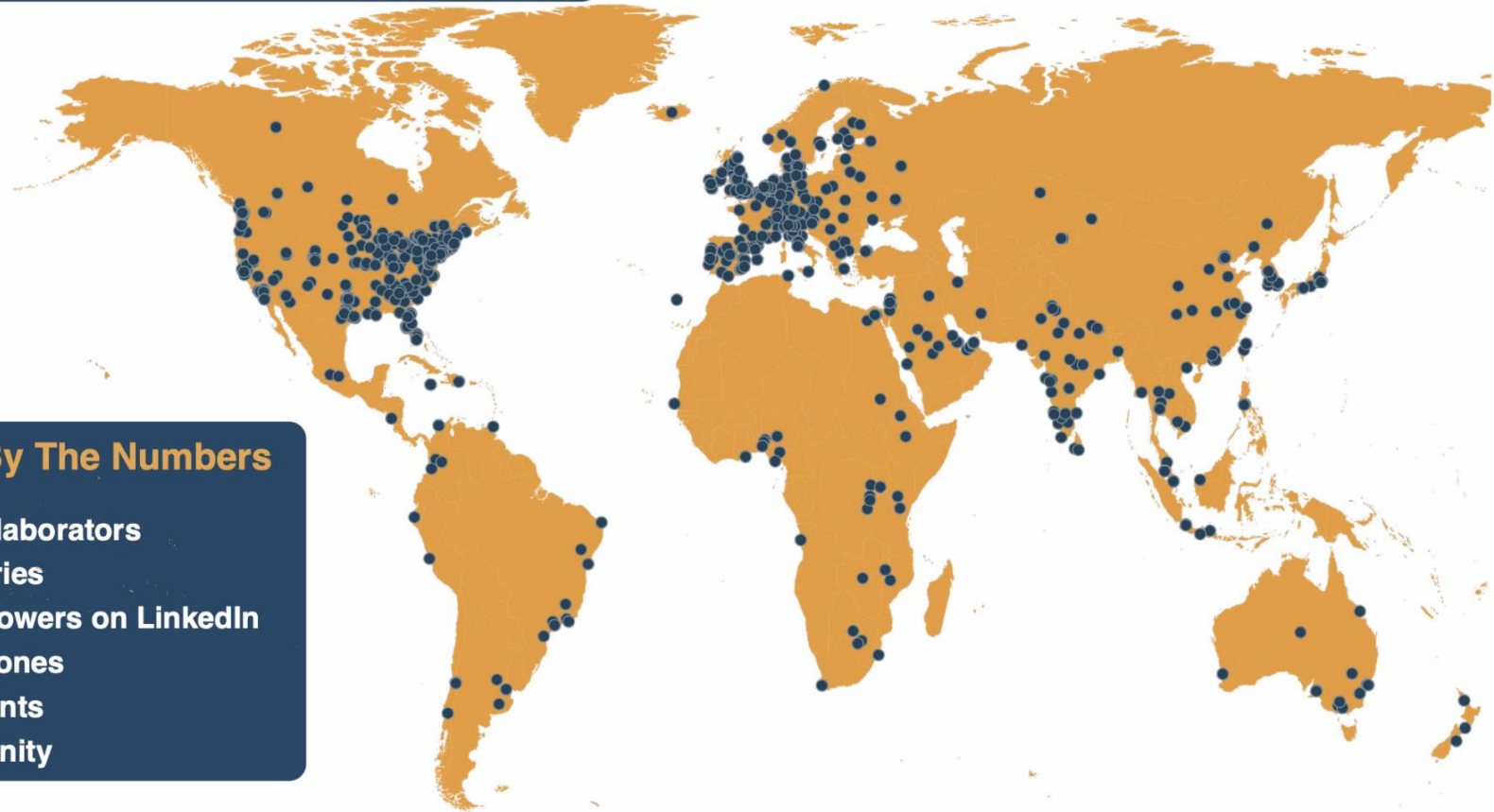
The OHDSI community brings together volunteers from around the world to establish open community data standards, develop open-source software, conduct methodological research, and apply scientific best practices to answer public health questions by generating reliable clinical evidence.

## OHDSI COLLABORATORS

Our community is ALWAYS seeking new collaborators. Do you want to focus on data standards or methodological research? Are you passionate about open-source development or clinical applications? Do you have data that you want to be part of global network studies? Do you want to join a global community that truly values the benefits of open science? Add a dot to the map below and JOIN THE JOURNEY!

### OHDSI By The Numbers

- 4,751 collaborators
- 88 countries
- 9,004 followers on LinkedIn
- 21 time zones
- 6 continents
- 1 community





# LARGE NETWORK OF DATA SOURCES



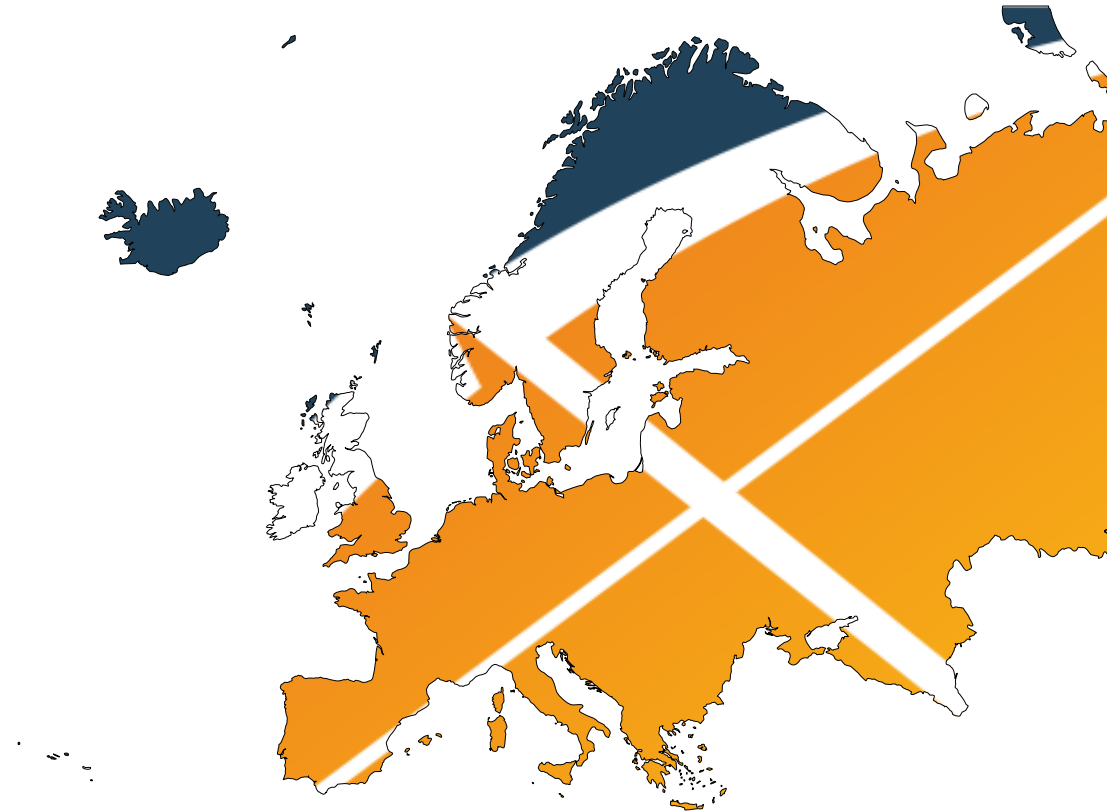
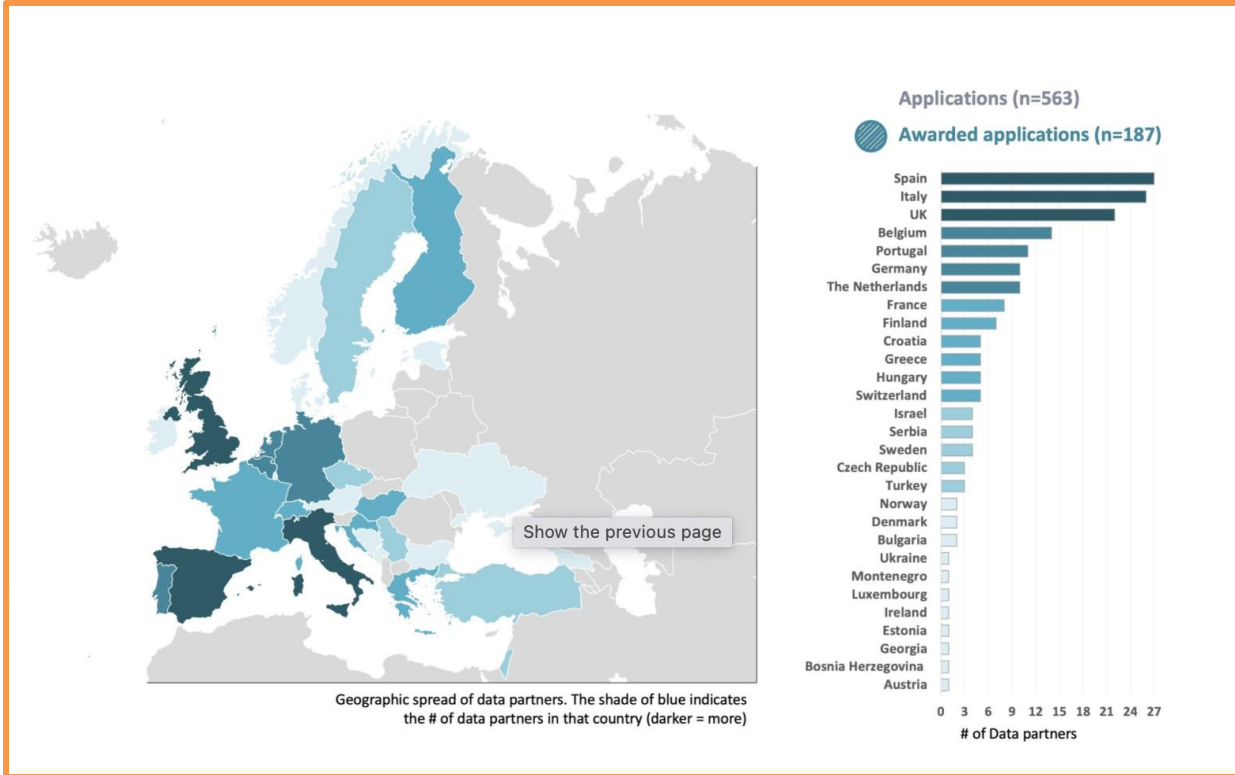
**544 database from 54 countries**

**More than 974 million unique patient records**

**~ 12% of the global population**



## From EHDEN to a network of national nodes



207 databases from 30 countries  
More than 455 million patients



# COMMUNITY CALLS



## Upcoming Community Calls

Date	Topic
Apr. 14	Phenotype April, Week 2: Phenotype Validation Using KEEPER
Apr. 21	NO MEETING / EUROPE SYMPOSIUM
Apr. 28	Phenotype April, Week 4: Final Evaluation and Learnings
May 5	Europe Symposium Review/Phenotype April Finale
May 12	Collaborator Showcase Brainstorm (Submission Deadline is June 5)
May 19	MEDS (Medical Event Data Standard) & Potential Collaborations with OHDSI

#JoinTheJourney

[www.ohdsi.org](http://www.ohdsi.org)



## Weekly community calls



## Monthly Europe community calls



## Upcoming Europe Community Calls

Date	Topic
May 7	Recap OHDSI EU Symposium / Welcome Newcomers
June 11	Oncology Studies
July 9	Deep Dive Topics OHI-Factor

#JoinTheJourney

[www.ohdsi.org](http://www.ohdsi.org)





# JUST ASK @ OHDSI FORUM!!

Huge corpus of background information

No barriers, everyone can sign-up & ask a question

The screenshot shows the OHDSI Forum website. At the top left is the OHDSI logo with the tagline 'OBSERVATIONAL HEALTH DATA SCIENCES AND INFORMATICS'. To the right are 'Sign Up' and 'Log In' buttons. A search icon is also present. On the left side, there is a navigation menu with 'Topics' and 'More' options, followed by a 'Categories' section with sub-items: General, Implementers, Developers, Researchers, and CDM Builders. Below this is an 'All categories' option and a 'Tags' section with items like 'cdm', 'atlas', 'vocabularies', 'patientprediction', and 'webapi'. The main content area features a light blue banner with a welcome message for new members. Below the banner are filter buttons for 'categories', 'tags', 'Categories', and 'Latest'. A table lists forum categories: General (2.2k topics), Implementers (1.3k topics), Developers (1.1k topics), and Researchers (784 topics). To the right of the table, a list of recent posts is shown, including 'OHDSI Forum Best Practices' (2 posts, Jun 2025), 'Welcome to OHDSI' (83 posts, Oct 2025), 'Welcome to OHDSI! - Please introduce yourself' (1.0k posts, 1h), and 'Themis WG meetings' (0 posts, 5h).

See an area where you want to contribute? Please Join The Journey!

Join A Workgroup

Meeting Schedule

Workgroup Tips

## Get to Know the OHDSI Workgroups

Local groups

[Chapter](#)

[WebAPI](#)

[Trials](#)

[Common Data Model](#)

[CDM Summary Subgroup](#)

[Subgroup](#)

Medical domains

[Researchers](#)

[Electronic Animal Health Records](#)

[Eye Care & Vision Research](#)

[FHIR and OMOP](#)

Data Standards/Interoperability

[Latin America](#)

[Medical Devices](#)

Methods

[AI & Analytics in Healthcare \(GAIA\)](#)

[Geographic Information System](#)

[Systems](#)

Analytics

Technology

[Oncology](#)

[Open-Source Community](#)

[Level Prediction](#)

[and Reproductive Health](#)

[Development & Evaluation](#)

[Rehabilitation](#)

[Registry](#)

[Steering Group](#)

[Surgery and Perioperative Medicine](#)

[Themis](#)

[Vaccine Vocabulary](#)

## Community-wide events:

- Phenotype Phebruary (April)

## Focus events:

- Studyathons
- Datathons





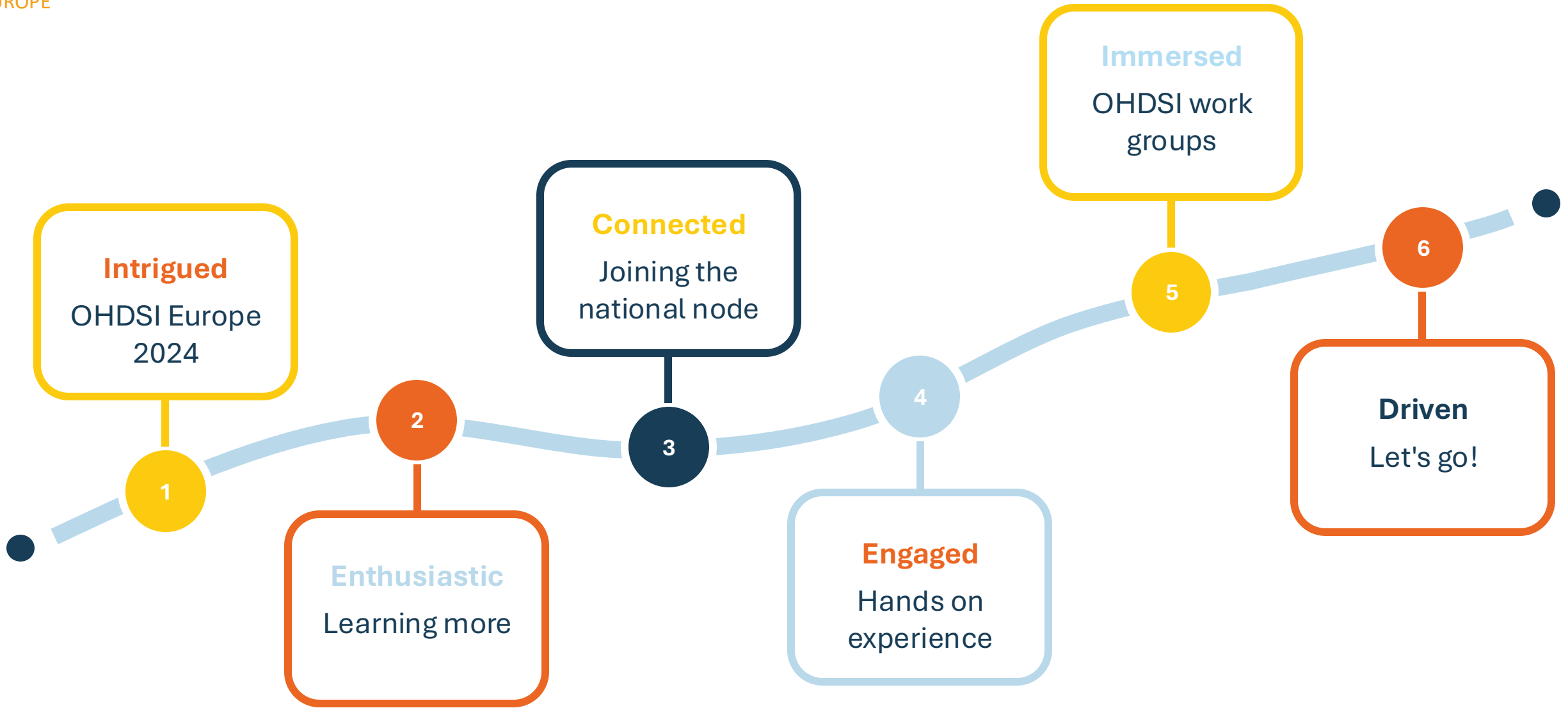
Registration & Call for Participation  
are **OPEN**

**2026 OHDSI Global Symposium**  
**Oct. 20-22 • New Brunswick, N.J. • Hyatt Regency  
Hotel**

# HOW CAN YOU LEARN MORE ABOUT OHDSI



# MY FIRST YEAR OHDSI VOYAGE





New job & starting PhD



Invitation to join OHDSI Europe symposium



First impression



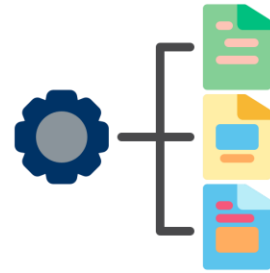
Introduction to OHDSI



## A comprehensive guide to



The OHDSI  
Community



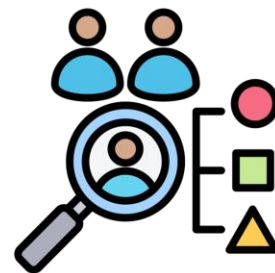
Uniform Data  
Representation



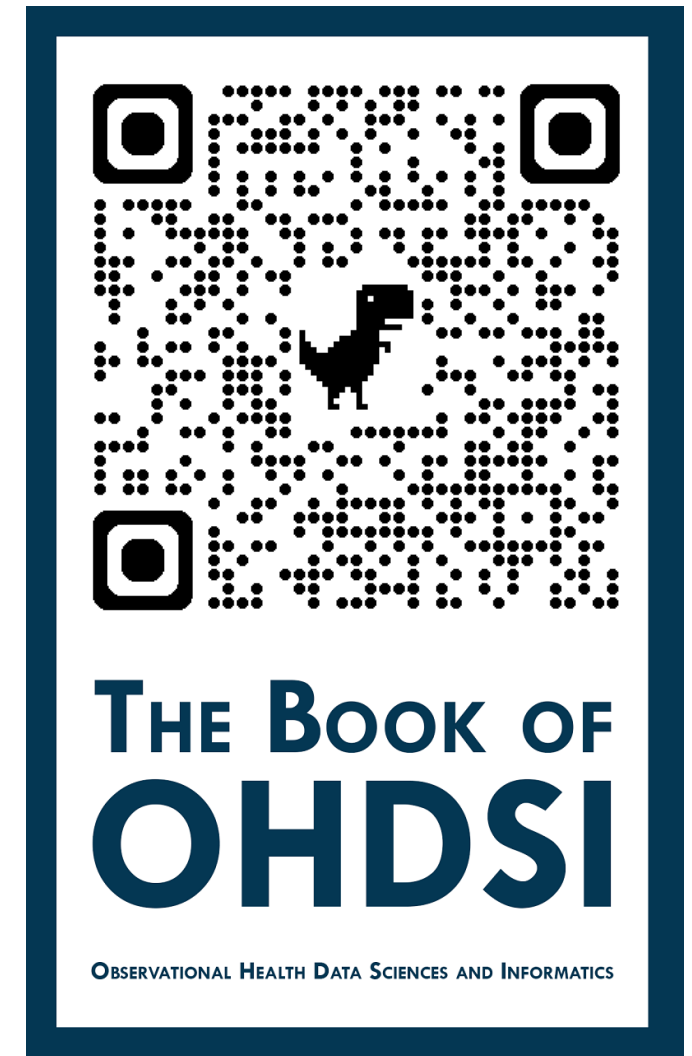
Data Analytics



Evidence Quality



OHDSI studies





- Free online educational resource
- More than 5,350 active learners across 108 countries

Getting Started	EHDEN Foundation	Patient Organisations: Real World Data and Real World Research	OMOP CDM and Standardised Vocabularies	ATLAS	Infrastructure
Extract, Transform and Load	Introduction to Usagi & Code Mappings for an ETL	OHDSI-in-a-Box	ETL Learning Pathway: Data Partner & SME Real World Use Cases	10 Minute Tutorial: PheValuator	10 Minute Tutorial: ATHENA
Open Science & FAIR Principles	Introduction to Data Quality	Phenotype Definition, Characterisation and Evaluation	Population-level Effect Estimation	Patient-Level Prediction	R for Patient-level Prediction
Applied Cost-Effectiveness Modeling with R	Assessing healthcare using outcomes that matter to patients	OHDSI2022 Tutorial - Creating Cohort Definitions	OHDSI2022 Tutorial - OMOP Common Data Model/Vocabulary	One hour of your time: The Phenotyping Problem	Health Technology Assessment



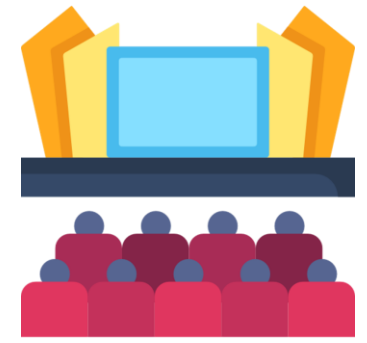
Community call meetings



Tool tutorials



Workshops



Previous events



- Meeting great people with lots of humor, knowledge and good ideas
- Joining national and international collaborations
- Getting involved in studies
  - Study-a-thon
- Showcasing work to a group spanning much of the healthcare system





# CONNECT TO YOUR NATIONAL NODE!

EUROPE

- |                |                    |
|----------------|--------------------|
| 1. Belgium     | 10. Norway         |
| 2. Denmark     | 11. Portugal       |
| 3. Estonia     | 12. Spain          |
| 4. Germany     | 13. United Kingdom |
| 5. Greece      | 14. Finland        |
| 6. Israel      | 15. Hungary        |
| 7. Italy       | 16. Ireland        |
| 8. Luxemburg   | 17. Sweden         |
| 9. Netherlands | 18. Austria        |



<https://www.ohdsi-europe.org/index.php/national-nodes>

## Pan-European Health Outcomes Observatory

- Alignment (common approach) on data and outcome standards (what should be measured)
- Regular publication of aggregated results
- Technological governance for interoperability and reproducibility

## Pan-European Observatory

## National H2O observatories, in four countries for the three disease areas (diabetes, IBD, cancer)

- Separate legal entities
- Mapped to the common data model (OMOP)
- Providing regular publication of aggregated results

## National Observatories



**Other data users** (academic and industry researchers, health authorities)

Data access as appropriate according to legal/ethical governance, reports, decision support

Anonymized patient-level data

Identifiable patient-level data

**Patients and healthcare professionals**

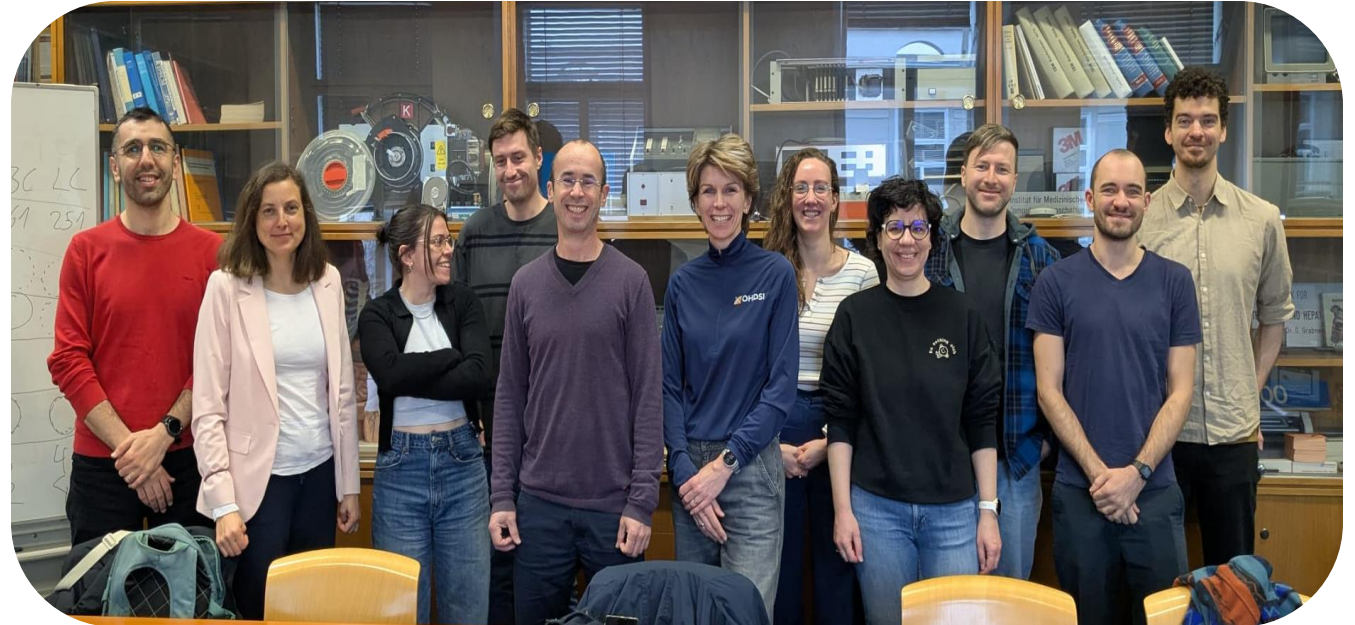
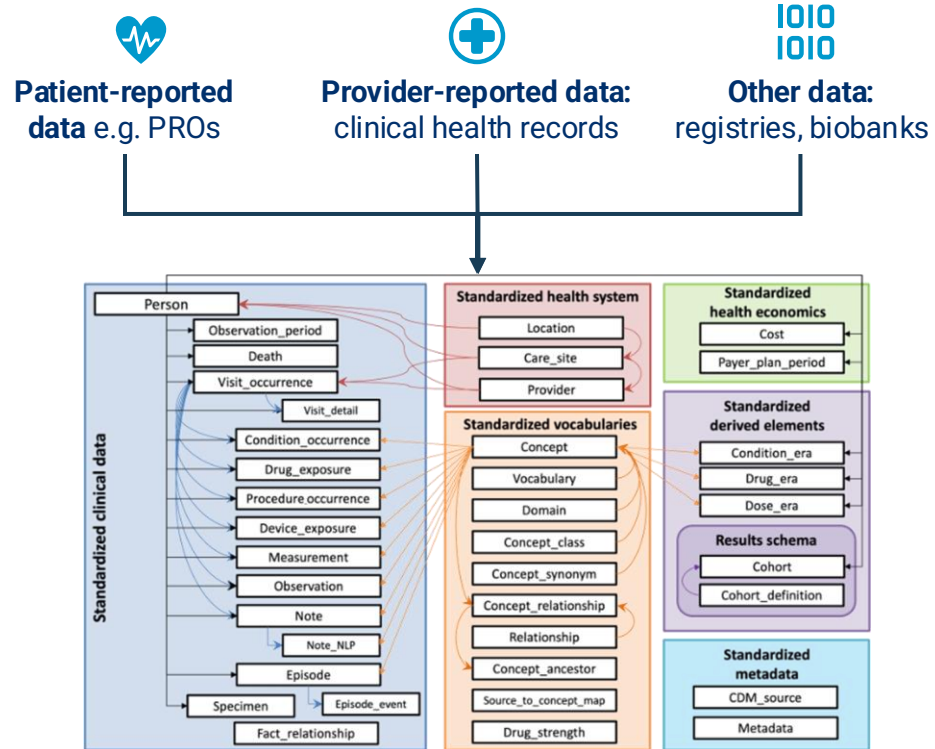
## Data collection layer

Protected under local/national legal and ethical frameworks

  
**Patient-reported data** e.g. PROs

  
**Provider-reported data:** clinical health records

  
**Other data:** registries, biobanks

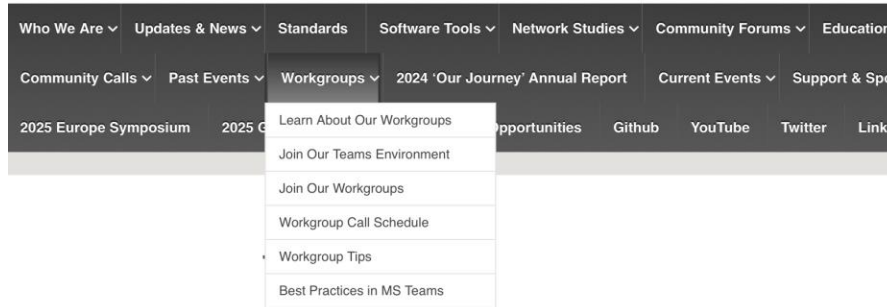




# OHDSI

OBSERVATIONAL HEALTH DATA SCIENCES AND INF

See an area where you want to contribute? Please Join The Journey!



Join A Workgroup

Meeting Schedule

Workgroup Tips

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- [Africa Chapter](#)
- [APAC](#)
- [ATLAS/WebAPI](#)
- [Clinical Trials](#)
- [Common Data Model](#)
- [CDM Survey Subgroup](#)
- [CDM Vocabulary Subgroup](#)
- [Dentistry](#)
- [Early-Stage Researchers](#)
- [Electronic Animal Health Records](#)
- [Eye Care & Vision Research](#)
- [FHIR and OMOP](#)

- [Generative AI & Analytics in Healthcare \(GAIA\)](#)
- [GIS – Geographic Information System](#)
- [HADES](#)
- [Health Equity](#)
- [Healthcare Systems](#)
- [Industry](#)
- [Latin America](#)
- [Medical Devices](#)
- [Medical Imaging](#)
- [Methods Research](#)
- [Natural Language Processing](#)
- [Network Data Quality](#)

- [Oncology](#)
- [Open-Source Community](#)
- [Patient-Level Prediction](#)
- [Perinatal and Reproductive Health](#)
- [Phenotype Development & Evaluation](#)
- [Psychiatry](#)
- [Rehabilitation](#)
- [Registry](#)
- [Steering Group](#)
- [Surgery and Perioperative Medicine](#)
- [Themis](#)
- [Vaccine Vocabulary](#)

## In-person



### 2025 OHDSI Global Symposium

Oct. 7-9 • New Brunswick, N.J. • Hyatt Regency Hotel



### 2024 OHDSI APAC Symposium

December 4-8 • Marina Bay Sands & National University of Singapore (NUS)



## Online

Every Tuesday @ 17:00

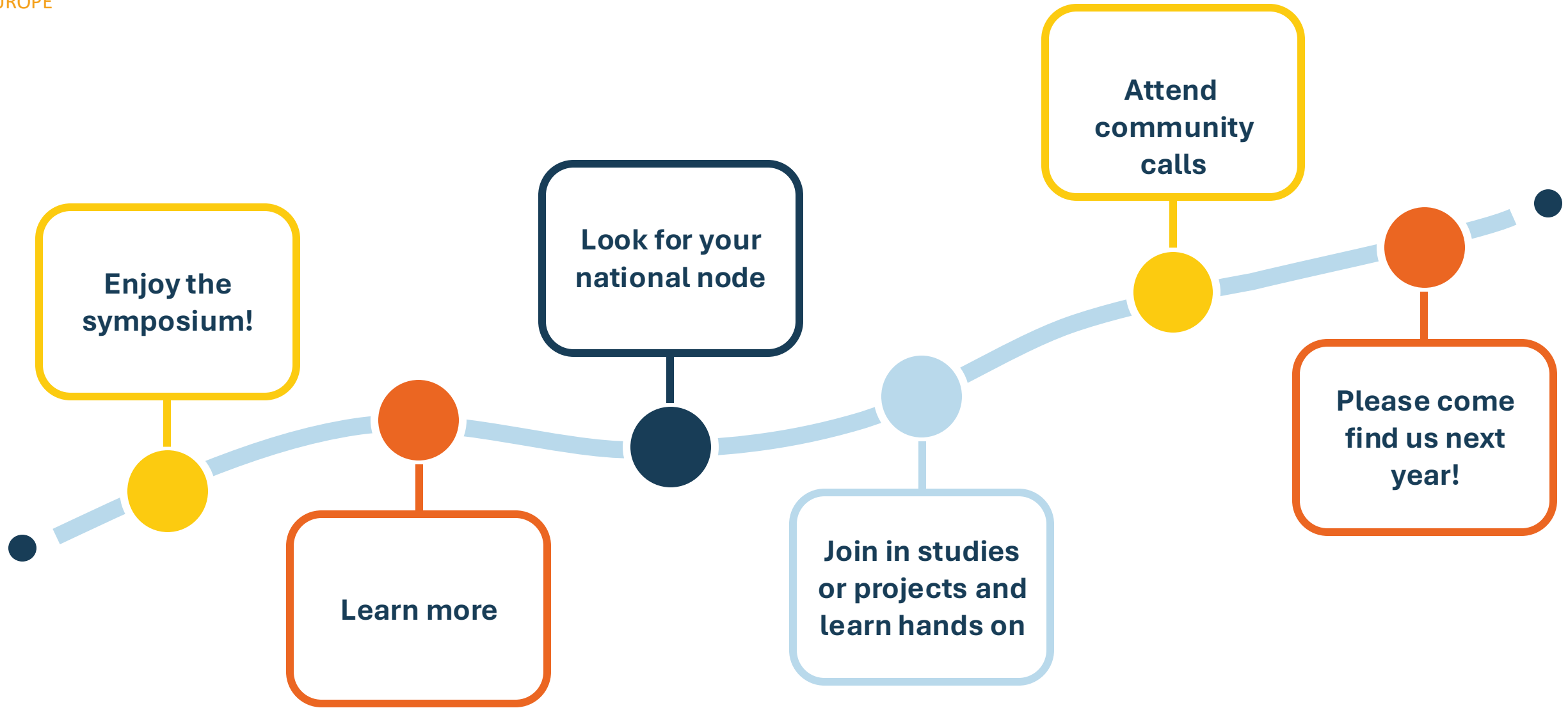


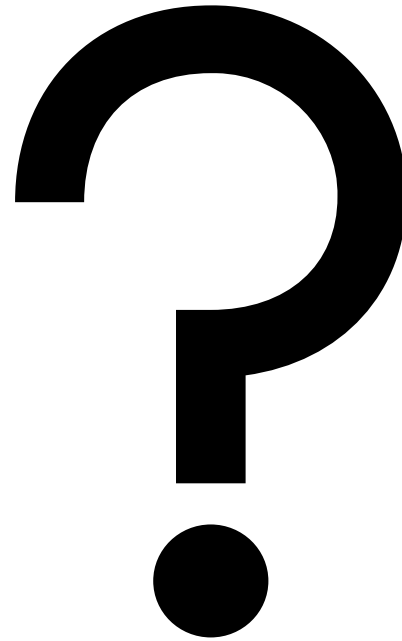
### Upcoming Community Calls

Date	Topic
June 24	ATLAS Deepdive: Characterization, Cohort Pathways, Incidence
July 1	ATLAS Deepdive: Technical and Administrative Capabilities
July 8	No Meeting – Europe Symposium
July 15	Europe Symposium Review
July 22	OMOP/OHDSI Research Spotlight
July 29	Asia-Pacific Regional Updates
Aug. 5	No Meeting
Aug. 12	Newcomer Introductions

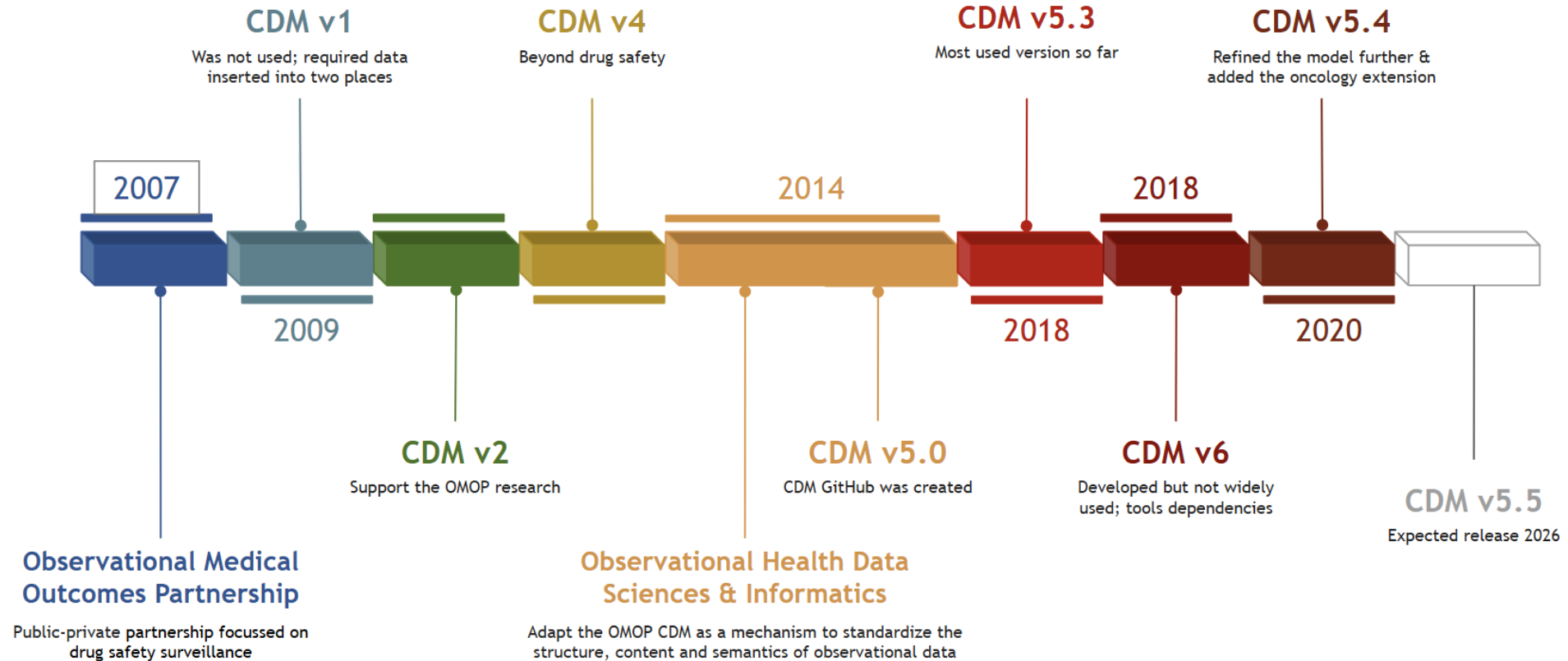


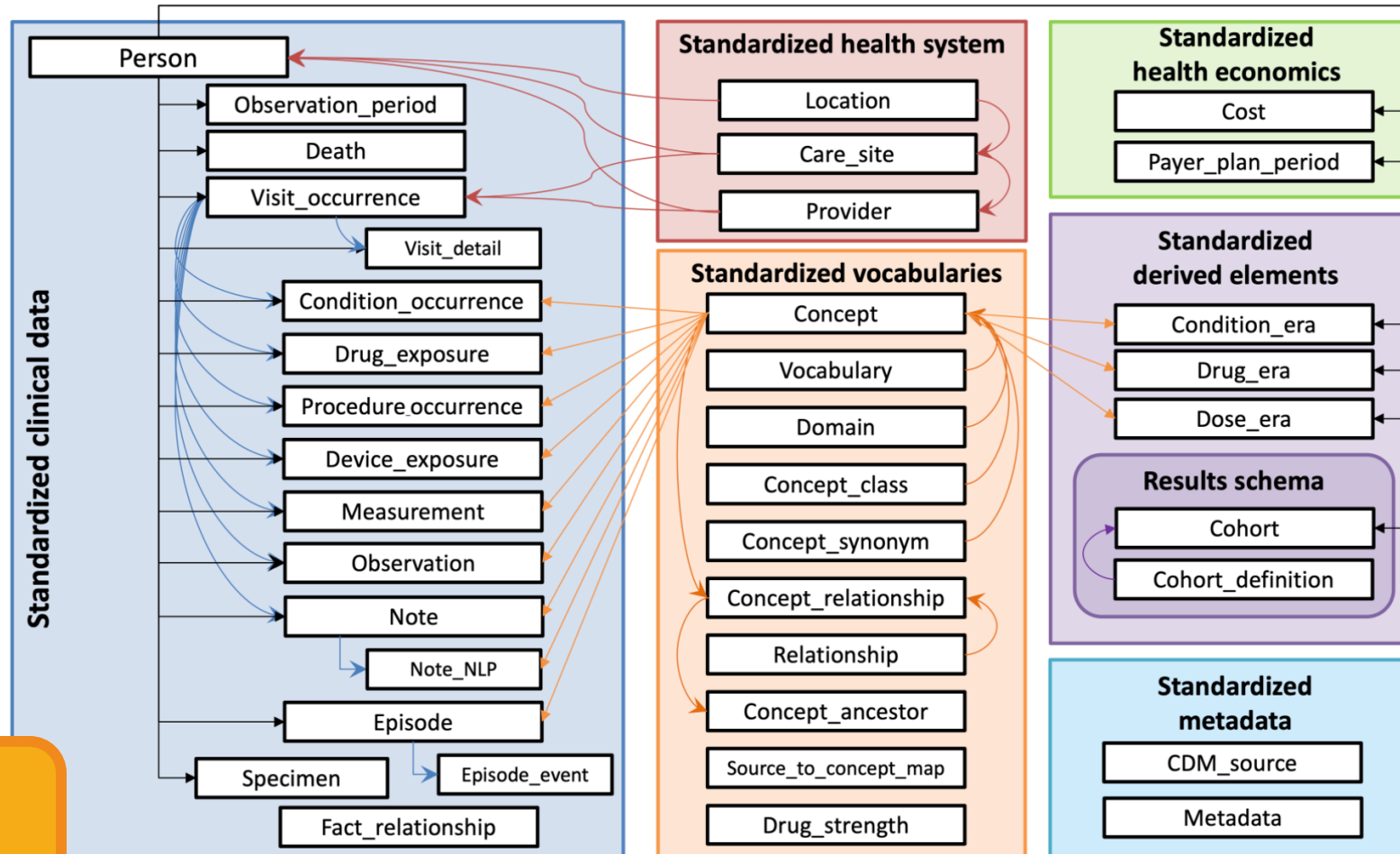
# YOUR FIRST YEAR OHDSI VOYAGE





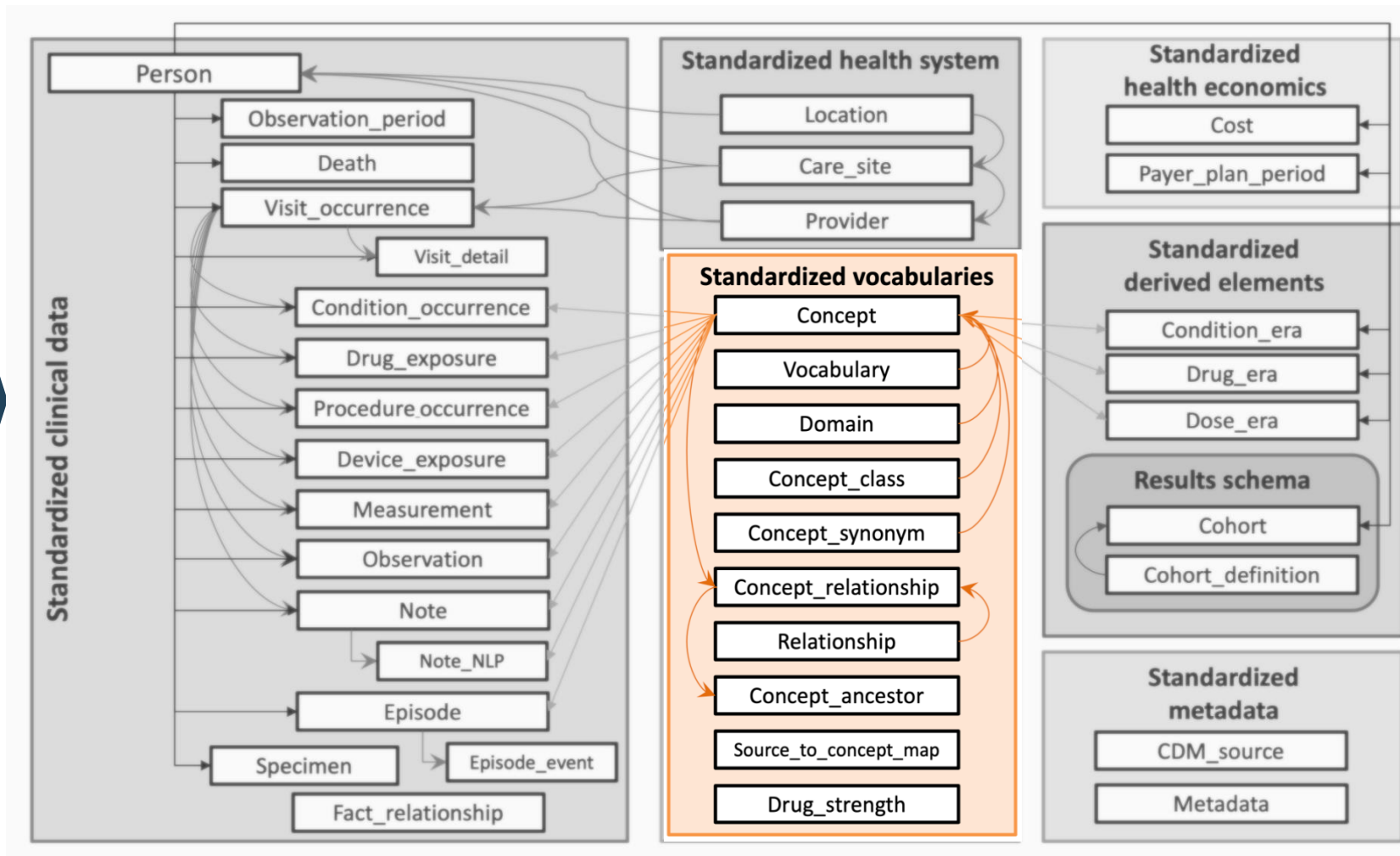
# HISTORY OF THE OMOP CDM





More info?  
- The Book of OHDSI chapter 4

# OHDSI STANDARDIZED VOCABULARIES





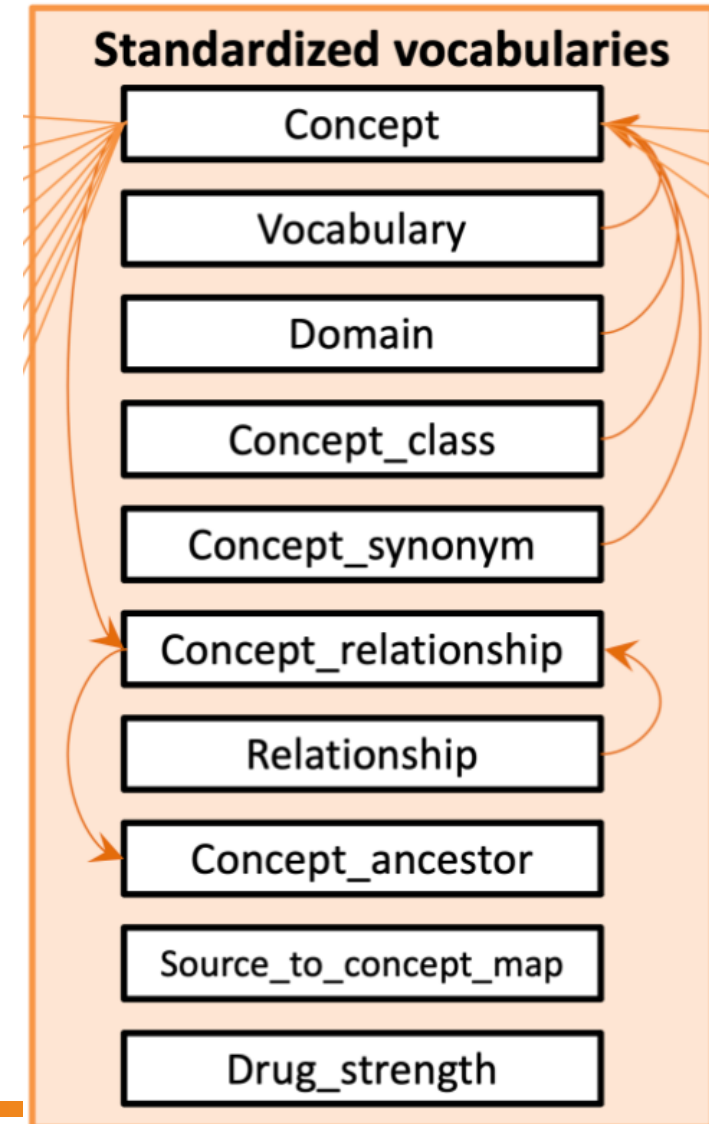
# STANDARDIZED VOCABULARIES

Foundational part of OMOP OHDSI for standardizing data

>100 vocabularies harmonized to a compiled standard

Continuous maintenance and improvements

Updates in **February and August** – Roadmap Vocabulary WG





# STANDARDIZED VOCABULARIES

- Existing vocabularies (e.g. SNOMED, ICD-O-3, etc.)
- Internally produced *administrative* vocabularies (eg. Episode Type)
  
- **Standard concept:** One concept representing the meaning of each clinical event
- **Non-standard concepts:** rest of the concepts of the standardized vocabularies which represent the same clinical event
  
- **CONCEPT\_RELATIONSHIP table:** Mapping between non-standard and standard concepts

## Regional enteritis of unspecified site

Concept\_id: 44822028  
Vocabulary\_id: **ICD9CM**  
Concept\_code: 555.9  
Concept: Non-standard

## Crohn's disease NOS

Concept\_id: 45446786  
Vocabulary\_id: **Read**  
Concept\_code: J40z.11  
Concept: Non-standard

## Crohn's disease

Concept\_id: 201606  
Vocabulary\_id: **SNOMED**  
Concept\_code: 34000006  
Concept: **Standard**

## Crohn disease, unspecified

Concept\_id: 45562490  
Vocabulary\_id: **ICD10**  
Concept\_code: K50.9  
Concept: Non-standard

## Crohn's disease of the small bowel NOS

Concept\_id: 4055022  
Vocabulary\_id: **SNOMED**  
Concept\_code: 196980005  
Concept: Non-standard



# IF YOU WANT TO DISCUSS VOCABS ...

## OHDSI office hours: Q&A with experts

*Open Q&A with community experts to discuss methods, best practices, and real-world experiences across topics such as ETL, data mapping, study design, and OHDSI research workflows.*

Sunday

15:30  
–  
17:00

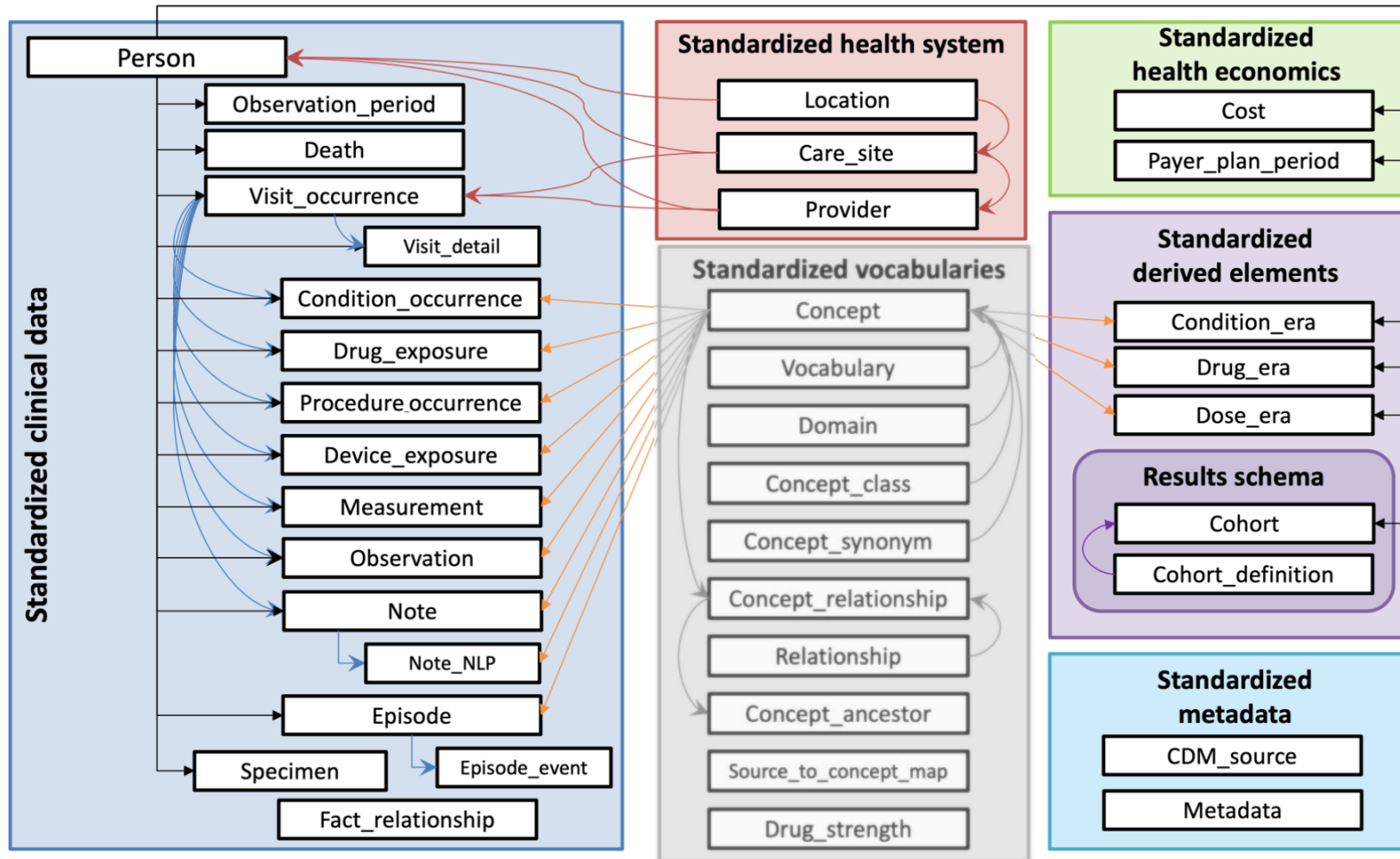
**ETL / Mapping**  
Anne van Winzum &  
Stefan Payralbe  
(The Hyve)

**OHDSI  
Standardized  
Vocabularies**  
Anna Ostropolets

**OHDSI studies**  
Ross Williams

**European Node  
leads meeting\***  
Renske Los

*\*(only for NN leads)*



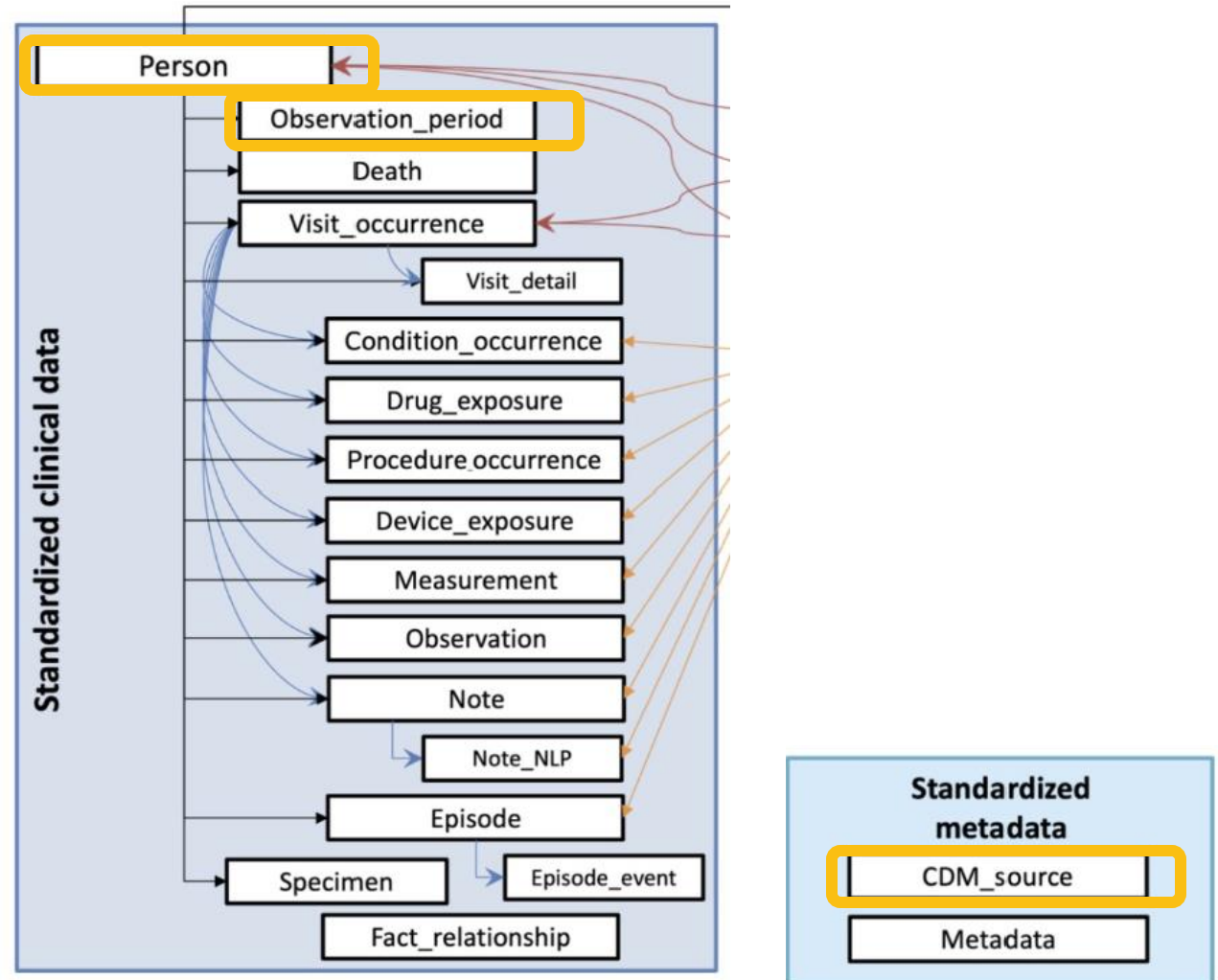
**Person** (mandatory)

**Observation period** (mandatory)

**CDM source** (mandatory)

## Medical Events

- Death
- Visit occurrence
- Clinical event domains (8 tables)









## White Rabbit (WR)

### Source data profiling

- Run on relational database or delimited text files

### Output summary statistics

- Overview of all tables and variable names
- Variable sparsity, most frequently occurring values

### Synthetic data generation



**White Rabbit (WR)** is an application that is used to profile and summarize the source data

Accepts **csv, SQL databases, SnowFlake** etc. as input

Output is a **Scan Report** containing descriptive statistics on table- and field level:

- Data type (e.g. string/numeric)
- Max. length per field
- Emptiness per field
- No. of unique values

	A	B	C	D	E	F	G	H	I	J
1	Table	Field	Description	Type	Max length	N rows	N rows ch	Fraction e	N unique values	
2	Tabel A	Veld 1		string	32	2.24E+09	100000	0.0%	12263	
3	Tabel A	Veld 2		string	32	2.24E+09	100000	0.0%	86623	
4	Tabel A	Veld 3		string	3	2.24E+09	100000	60.5%	44	
5										
6	Tabel B	Veld 11		string	32	2.24E+09	100000	0.0%	99675	
7	Tabel B	Veld 12		string	32	2.24E+09	100000	0.0%	99675	
8	Tabel B	Veld 13		string	1	2.24E+09	100000	0.0%	2	
9										

	A	B	C	D	E	F	G	H
1	patient_id	Frequency	geboortedatum	Frequency	seks	Frequency	plaats	Frequency
2	List truncated...		1998-08-01	5	0	900	"Heerlen"	1565
3			1953-12-15	5	1	1100	"Maastricht"	400
4			List truncated...				"Venlo"	10
5							"Sittard"	5
6							List truncated...	
7								



## Rabbit-in-a-Hat (RiaH)

### Define source data to OMOP mapping

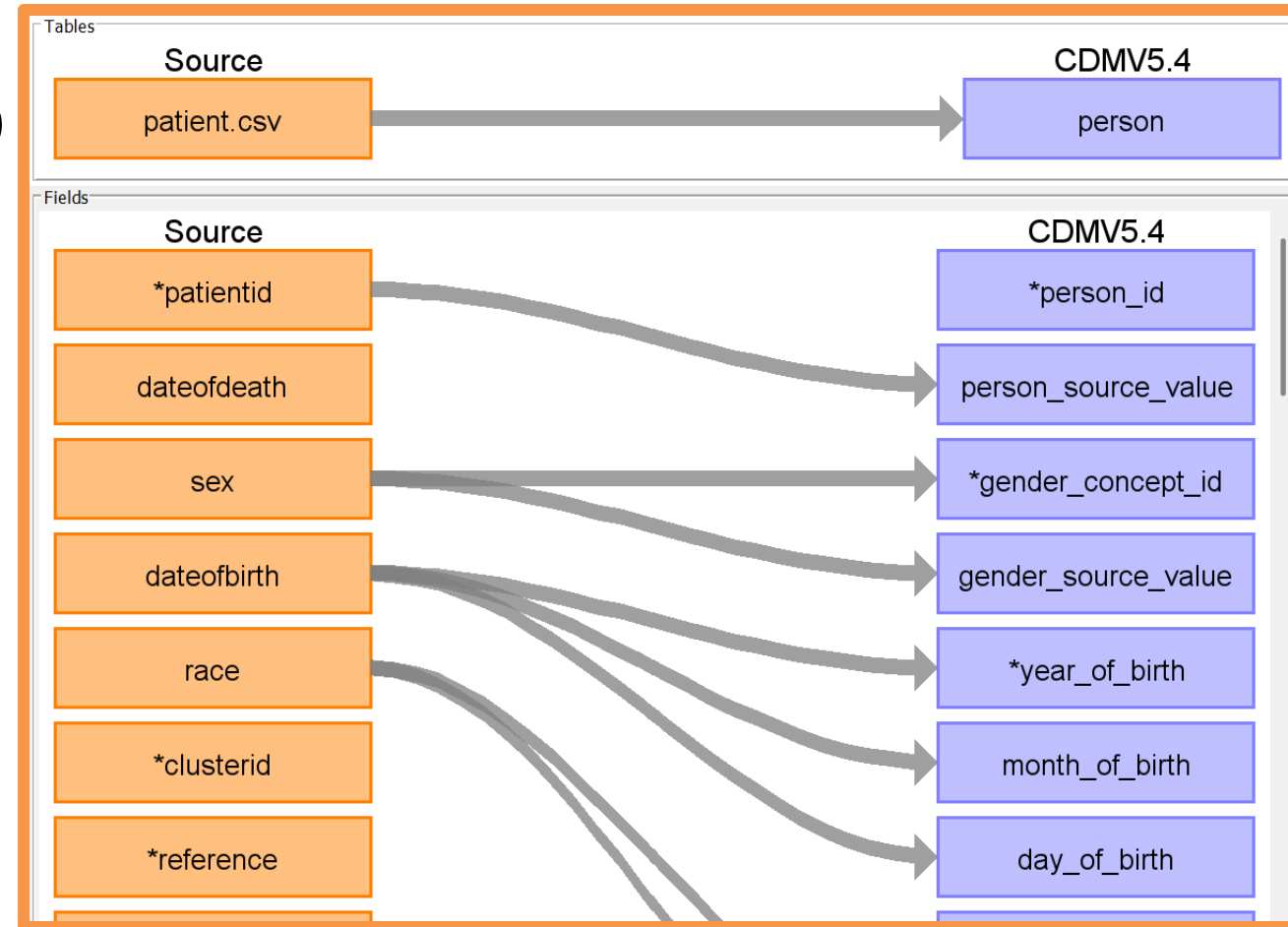
- Created with White Rabbit scan as input
- Table to table and column to column

### Outputs the foundations for the ETL

- Document describing the needed transformations
- Test framework for ETL verification

**Rabbit-in-a-Hat** is a tool used for syntactically mapping source data to target (OMOP) tables and fields

- Input to RiaH is a WR Scan Report
- Output is syntactic mapping documentation
- Enables documenting ETL logic in words (not in screenshot)
- Allows for generation of unit testing framework





## Usagi

### Map source codes/values to OMOP standards

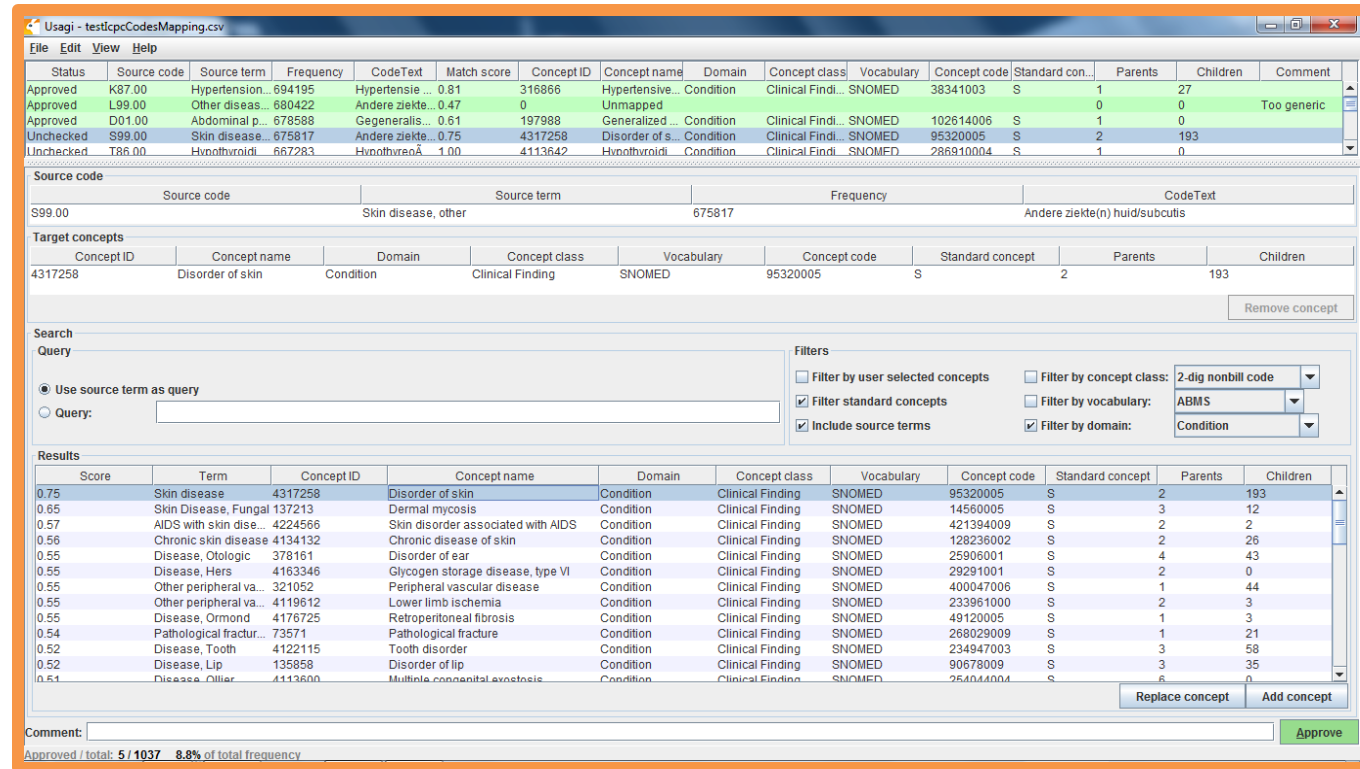
- Usagi will run term similarity approach
- May require domain expertise

### Export into mapping table used in ETL

- Does not have to cover 100% of source values

Usagi is a helper tool that facilitates **semantically mapping** source values to OMOP concepts, saving a lot of manual work

- Requires loading of vocabularies, to be selected by the user
- Will select most similar concept from loaded vocabularies to map from source term
- Initial mapping can be changed using the `query` function
- Facilitated querying using filters
- Collaboration: allows for approval of certain mappings and flagging of uncertain ones



The screenshot shows the Usagi application window titled "Usagi - testfcpCodesMapping.csv". It displays a table of source codes and their mappings to target concepts. Below the table, there is a search interface with a query field and several filters. The results table shows a list of concepts with their scores and details.

Status	Source code	Source term	Frequency	CodeText	Match score	Concept ID	Concept name	Domain	Concept class	Vocabulary	Concept code	Standard con...	Parents	Children	Comment
Approved	K87.00	Hypertension	694195	Hypertensie ...	0.81	316866	Hypertensie	Condition	Clinical Findi...	SNOMED	38341003	S	1	27	
Approved	L99.00	Other diseases	680422	Andere ziekte...	0.47	0	Unmapped					S	0	0	Too generic
Approved	D01.00	Abdominal p...	678588	Gegeneralis...	0.61	197988	Generalized...	Condition	Clinical Findi...	SNOMED	102614006	S	1	0	
Unchecked	S99.00	Skin disease	675817	Andere ziekte...	0.75	4317258	Disorder of s...	Condition	Clinical Findi...	SNOMED	95320005	S	2	193	
Unchecked	T86.00	Hvnothuroidi	667283	Hvnothurenā	1.00	4113642	Hvnothuroidi	Condition	Clinical Findi...	SNOMED	286810004	S	1	0	

Source code	Source code	Source term	Frequency	CodeText
S99.00	S99.00	Skin disease, other	675817	Andere ziekte(n) huid/subcutis

Target concepts	Concept ID	Concept name	Domain	Concept class	Vocabulary	Concept code	Standard concept	Parents	Children
4317258	4317258	Disorder of skin	Condition	Clinical Finding	SNOMED	95320005	S	2	193

Search Query:

Filters:

- Filter by user selected concepts
- Filter by concept class: 2-dig nonbill code
- Filter standard concepts
- Filter by vocabulary: ABMS
- Include source terms
- Filter by domain: Condition

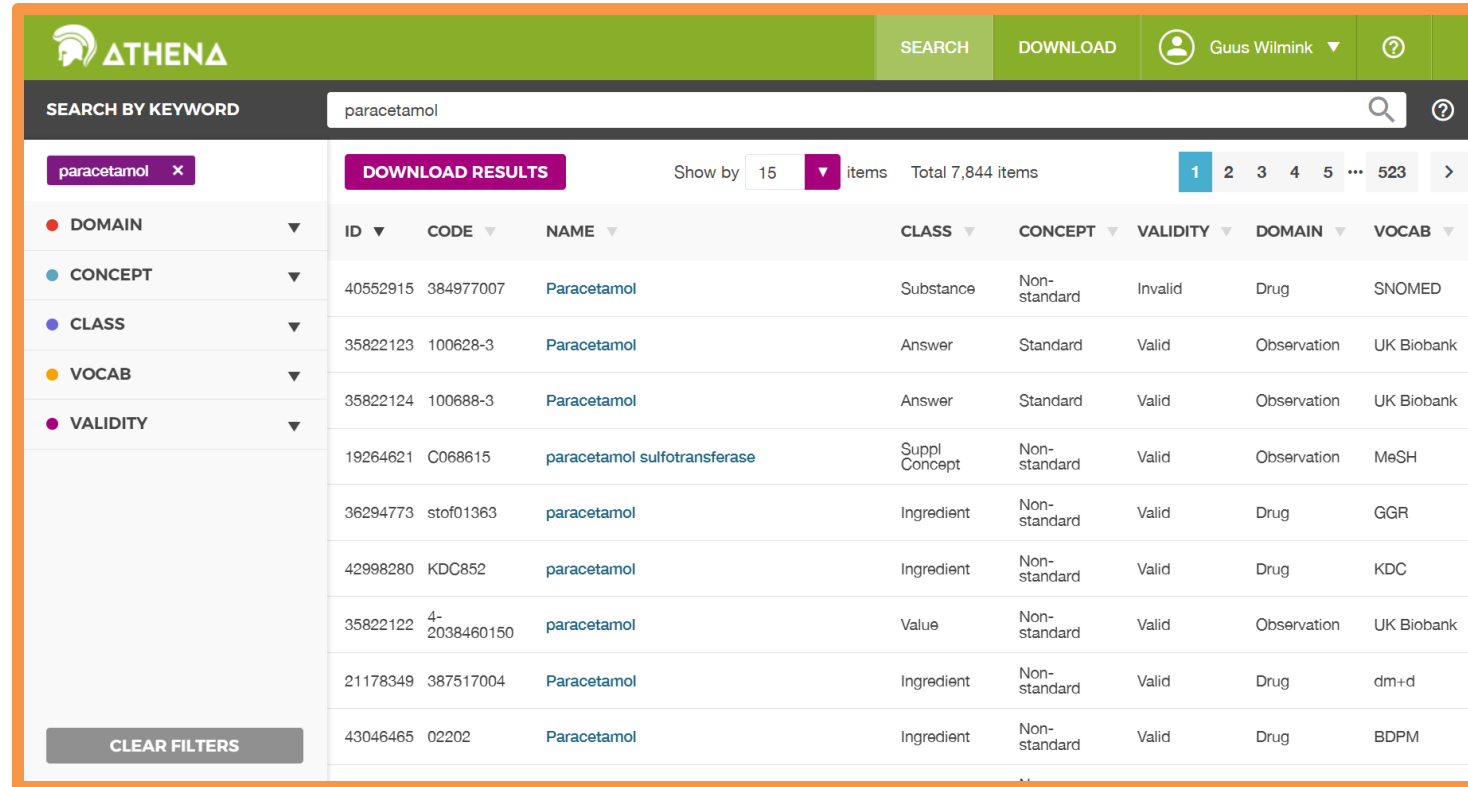
Score	Term	Concept ID	Concept name	Domain	Concept class	Vocabulary	Concept code	Standard concept	Parents	Children
0.75	Skin disease	4317258	Disorder of skin	Condition	Clinical Finding	SNOMED	95320005	S	2	193
0.65	Skin Disease, Fungal	137213	Dermal mycosis	Condition	Clinical Finding	SNOMED	14560005	S	3	12
0.57	AIDS with skin dise...	4224566	Skin disorder associated with AIDS	Condition	Clinical Finding	SNOMED	421394009	S	2	2
0.56	Chronic skin disease	4134132	Chronic disease of skin	Condition	Clinical Finding	SNOMED	128236002	S	2	26
0.55	Disease, Otolgic	378161	Disorder of ear	Condition	Clinical Finding	SNOMED	25906001	S	4	43
0.55	Disease, Hers	4163346	Glycogen storage disease, type VI	Condition	Clinical Finding	SNOMED	29291001	S	2	0
0.55	Other peripheral va...	321052	Peripheral vascular disease	Condition	Clinical Finding	SNOMED	400047006	S	1	44
0.55	Other peripheral va...	4119612	Lower limb ischemia	Condition	Clinical Finding	SNOMED	233961000	S	2	3
0.55	Disease, Ormond	4176725	Retroperitoneal fibrosis	Condition	Clinical Finding	SNOMED	49120005	S	1	3
0.54	Pathological fractur...	73571	Pathological fracture	Condition	Clinical Finding	SNOMED	268029009	S	1	21
0.52	Disease, Tooth	4122115	Tooth disorder	Condition	Clinical Finding	SNOMED	234947003	S	3	58
0.52	Disease, Lip	135858	Disorder of lip	Condition	Clinical Finding	SNOMED	90679009	S	3	35
0.51	Disease, Ollar	4113600	Multiple congenital exostosis	Condition	Clinical Finding	SNOMED	264044004	S	6	0

Comment:

Approved / total: 5 / 1037 8.8% of total frequency

Athena is a **web application** used for **querying the OMOP vocabularies**

- No option to load source terms; only manual querying possible
- Allows for many filters on OMOP domains, vocabularies, classes, standardness, etc.
- OMOP vocabularies can be accessed and downloaded from Athena
- Required for Usagi usage



The screenshot shows the Athena web application interface. At the top, there is a navigation bar with the Athena logo, a search bar, and buttons for 'SEARCH', 'DOWNLOAD', and a user profile dropdown for 'Guus Wilmink'. Below the navigation bar, the search results are displayed. The search term 'paracetamol' is entered in the search bar. The results are shown in a table with columns: ID, CODE, NAME, CLASS, CONCEPT, VALIDITY, DOMAIN, and VOCAB. The table contains 8 rows of results for 'paracetamol'. A 'DOWNLOAD RESULTS' button is visible above the table. The table also includes a 'CLEAR FILTERS' button at the bottom left.

ID	CODE	NAME	CLASS	CONCEPT	VALIDITY	DOMAIN	VOCAB
40552915	384977007	Paracetamol	Substance	Non-standard	Invalid	Drug	SNOMED
35822123	100628-3	Paracetamol	Answer	Standard	Valid	Observation	UK Biobank
35822124	100688-3	Paracetamol	Answer	Standard	Valid	Observation	UK Biobank
19264621	C068615	paracetamol sulfotransferase	Suppl Concept	Non-standard	Valid	Observation	MeSH
36294773	stof01363	paracetamol	Ingredient	Non-standard	Valid	Drug	GGR
42998280	KDC852	paracetamol	Ingredient	Non-standard	Valid	Drug	KDC
35822122	4-2038460150	paracetamol	Value	Non-standard	Valid	Observation	UK Biobank
21178349	387517004	Paracetamol	Ingredient	Non-standard	Valid	Drug	dm+d
43046465	02202	Paracetamol	Ingredient	Non-standard	Valid	Drug	BDPM

More info?

- The Book of OHDSI chapter 5
- [athena.ohdsi.org](http://athena.ohdsi.org)

Session  
**ETL development and updating**



After lunch!

<p>13:00          –          15:00</p>		<p><b>ETL development and updating</b>          Maxim Moinat (EMC),          Anne van Winzum, Stefan Payralbe (The Hyve)</p> <p><u>Description</u></p> <ul style="list-style-type: none"> <li>• Short introduction to OMOP ETL conventions and ETL implementation examples.</li> <li>• Latest developments in available tooling to assist with ETL/mappingTools</li> <li>• Importance and best practices for maintaining and updating ETL/mapping after initial conversion</li> </ul> <p><u>Target audience</u></p> <p>New and current data holders, ETL developers, data engineers responsible for OMOP CDM conversions and ongoing maintenance.</p>
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## SYNTHEA SYNTHETIC HEALTH DATABASE

OVERVIEW

METADATA

RESULTS

ABOUT

## DATA QUALITY ASSESSMENT

### SYNTHEA SYNTHETIC HEALTH DATABASE

Results generated at 2019-08-22 14:15:06 in 29 mins

	Verification				Validation				Total			
	Pass	Fail	Total	% Pass	Pass	Fail	Total	% Pass	Pass	Fail	Total	% Pass
Plausibility	159	<b>21</b>	180	88%	283	0	283	100%	442	<b>21</b>	463	95%
Conformance	637	<b>34</b>	671	95%	104	0	104	100%	741	<b>34</b>	775	96%
Completeness	369	<b>17</b>	386	96%	5	<b>10</b>	15	33%	374	<b>27</b>	401	93%
Total	1165	<b>72</b>	1237	94%	392	<b>10</b>	402	98%	1557	<b>82</b>	1639	<b>95%</b>

# DQ ASSESSMENT – CUSTOMIZABLE CHECKS & THRESHOLDS

% 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





- R package
- Data table counts
- Vocabulary counts
- Technical Infrastructure Chec
- Creates report

### 3. Clinical data

#### 3.1. Record counts per OMOP CDM

**Table 2.** The number of records in all clinical data tables

Table	Count
measurement	
drug_exposure	
procedure_occurrence	
visit_detail	
visit_occurrence	
observation	
condition_occurrence	
condition_era	
drug_era	
person	
observation_period	
provider	
cdm_source	
episode_event	
fact_relationship	
episode	
cost	
death	
device_exposure	
dose_era	
note	
payer_plan_period	
specimen	
care_site	
location	
note_nlp	
metadata	

Query executed in 3.69 seconds

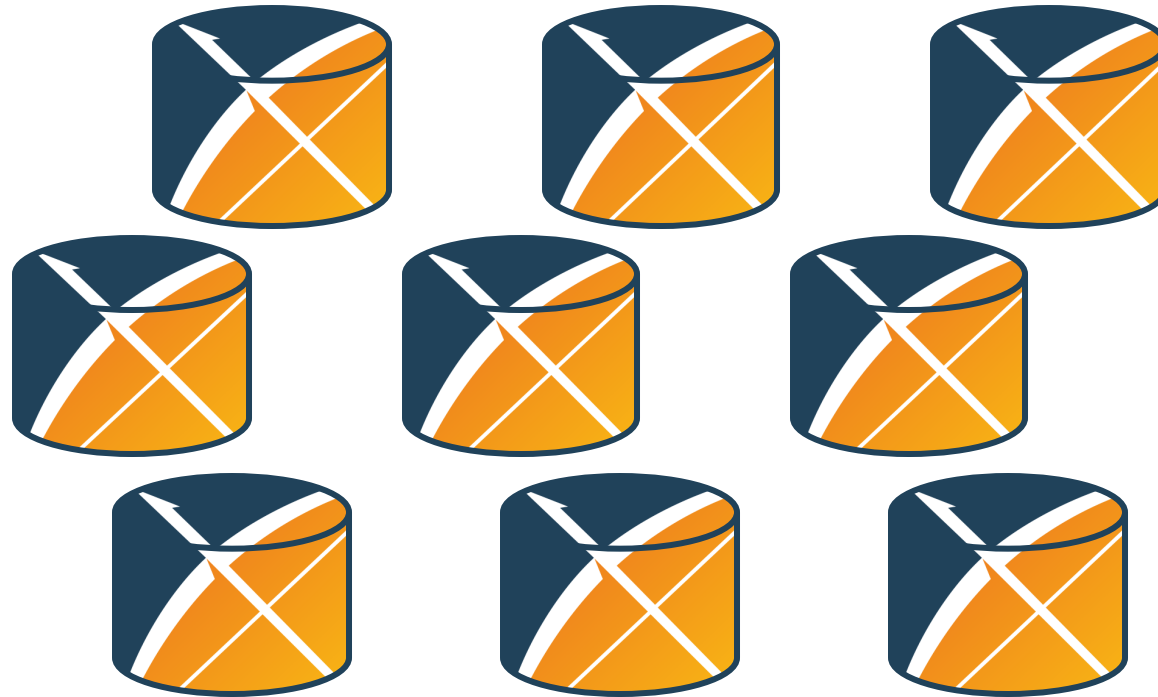
#### 4.4. Mapped Codes

**Table 25.** Top 25 mapped drugs. Counts are rounded up to the nearest hundred. Values with a record count <=5 are omitted. ©

#	Concept id	Concept Name	#Records	%Records
1	19080128	lisinopril 10 MG Oral Tablet	228,200	15.4%
2	19078106	hydrochlorothiazide 25 MG Oral Tablet	188,600	12.7%
3	19073094	amlodipine 2.5 MG Oral Tablet	179,000	12.1%
4	19009384	insulin isophane, human 70 UNT/ML / insulin, regular, human 30 UNT/ML Injectable Suspension [Humulin]	120,600	8.1%
5	19030765	1 ML epoetin alfa 4000 UNT/ML Injection [Epogen]	88,100	5.9%
6	1511248	NDA020503 200 ACTUAT albuterol 0.09 MG/ACTUAT Metered Dose Inhaler	51,500	3.5%
7	19126352	nitroglycerin 0.4 MG/ACTUAT Mucosal Spray	41,900	2.8%
8	1539463	simvastatin 10 MG Oral Tablet	39,800	2.7%
9	1539411	simvastatin 20 MG Oral Tablet	36,600	2.5%
10	40163924	24 HR metformin hydrochloride 500 MG <u>Extended Release</u> Oral Tablet	35,600	2.4%
11	1332419	amlodipine 5 MG Oral Tablet	34,700	2.3%
12	19075601	clopidogrel 75 MG Oral Tablet	32,900	2.2%
13	40163554	warfarin sodium 5 MG Oral Tablet	25,900	1.7%
14	19018935	digoxin 0.125 MG Oral Tablet	25,900	1.7%
15	40171902	verapamil hydrochloride 40 MG	25,100	1.7%
16	40173590	alendronic acid 10 MG Oral Tablet	20,300	1.4%
17	40169217	120 ACTUAT fluticasone propionate 0.044 MG/ACTUAT Metered Dose Inhaler [Flovent]	16,500	1.1%
18	1154380	albuterol 5 MG/ML Inhalation Solution	15,900	1.1%
19	40169281	60 ACTUAT fluticasone propionate 0.25 MG/ACTUAT / salmeterol 0.05 MG/ACTUAT Dry Powder Inhaler	15,700	1.1%



Individual OMOP database



Network of OMOP databases



*Coffee break*

15:30

–

17:30

## Data Quality Assessment Framework & Tools

Clair Blacketer, Anthony Sena (J&J)

### Description

- Data Quality Dashboard (DQD) and other recent developments for tools to assess data quality
- Hands-on exercise for running DQD to identify and address ETL conversion issues
- Data quality considerations for network studies

### Target audience

Anyone responsible for assessing and improving the quality of OMOP CDM ETL or data quality for study

# WHAT CAN (CURRENTLY) BE DONE WITH THE OHDSI TOOLS?

# REAL-WORLD EVIDENCE GENERATION USING OMOP CDM

Different structure  
& terminology

Standardized data



ETL



ETL



ETL



Conversion to  
OMOP CDM



Share aggregated  
results

Run standard  
analytic tools locally

**Standardized data enables standardized analytics!**

Common format  
& terminology

Standard input & output  
Parametrized choices

# BENEFITS OF STANDARDIZED DATA & ANALYTICS

## 1. Large-scale evidence

Federated analysis allows multi-database studies without compromising privacy

## 2. Rapid response

Large time savings through data readiness and pre-developed tooling

## 3. Scalable & reproducible

Tooling can be reused across databases and disease domains

=> **Reliable evidence**

# ATTRIBUTES OF RELIABLE EVIDENCE

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

Source: <https://ohdsi.github.io/TheBookOfOhdsi/EvidenceQuality.html>

# QUESTIONS ASKED ACROSS THE PATIENT JOURNEY

## Clinical characterization

Which patients take which treatments?

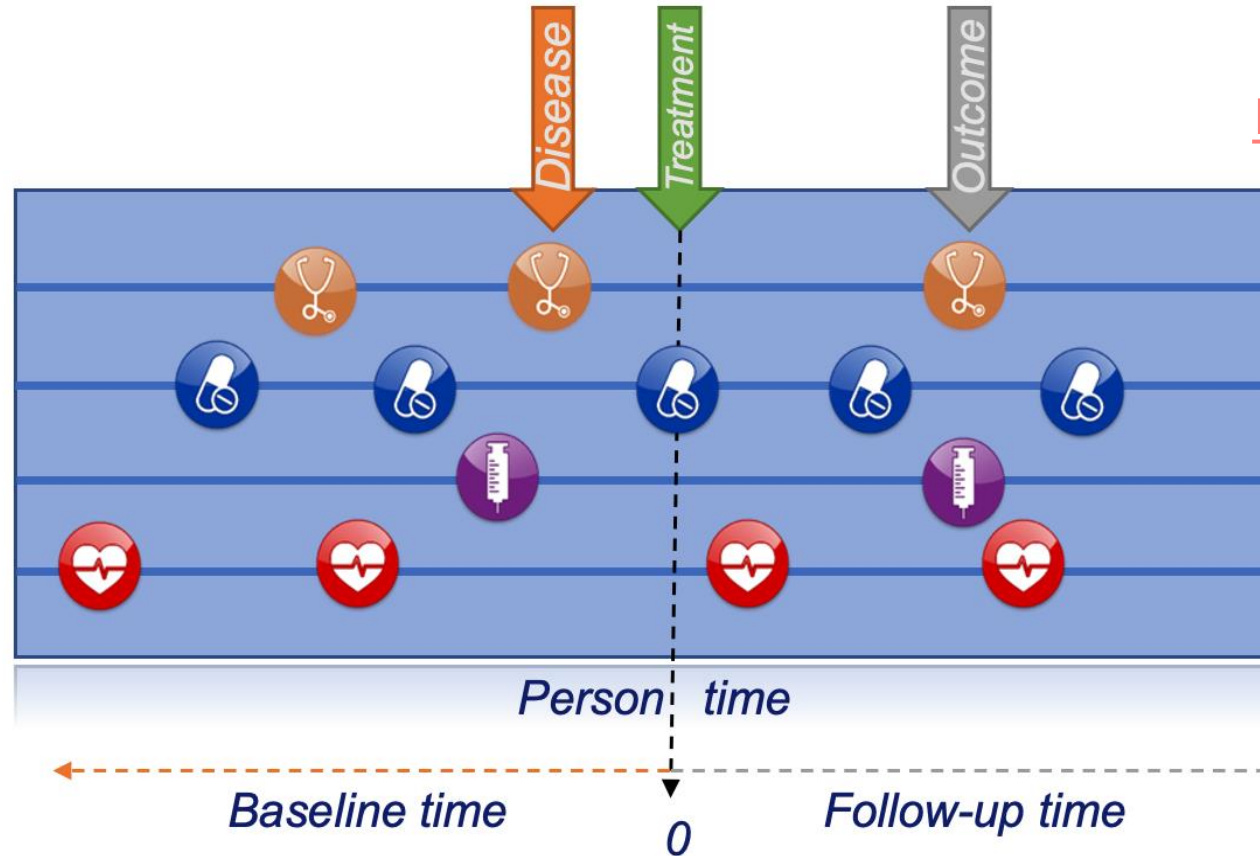
How many patients experienced complications?

Measurements

Conditions

Drugs

Procedures



## Patient-level prediction

What is the probability that I will develop a given disease?

What is the probability that I will experience an (adverse) outcome?

## Population-level effect estimation

Does treatment cause an outcome?

Does one treatment lead to a better outcome than the alternative?

More info?

- The Book of OHDSI chapter 7

## Population-Level Effect Estimation

- What are the causal effects?

## Patient-Level Prediction

- What will happen to me?

## Clinical Characterisation

- What happened to the patients?



Hernán MA, Hsu J, Healy B. [A Second Chance to Get Causal Inference Right: A Classification of Data Science Tasks](#). CHANCE 2019;32(1): 42-49, DOI: [10.1080/09332480.2019.1579578](#).

- Growing number of studies (also used in DARWIN EU<sup>®</sup> now)

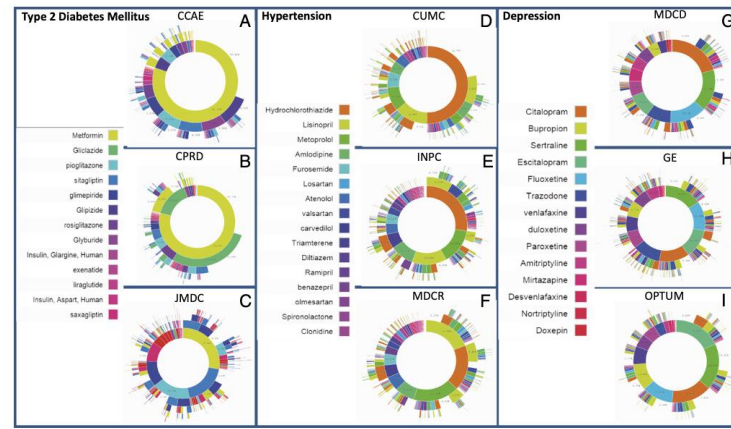


Fig. 3. For each disease, diabetes (A-C), hypertension (D-F), and depression (G-I), the inner circle shows the first relevant medication that the patient took, the second circle shows the second medication, and so forth. Three data sources are shown for each disease; the data source abbreviations are defined in Table 2.

Hripcsak et al. (2016)

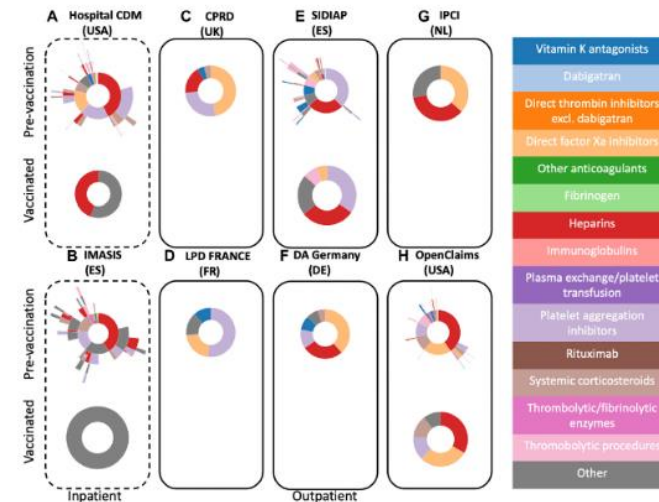
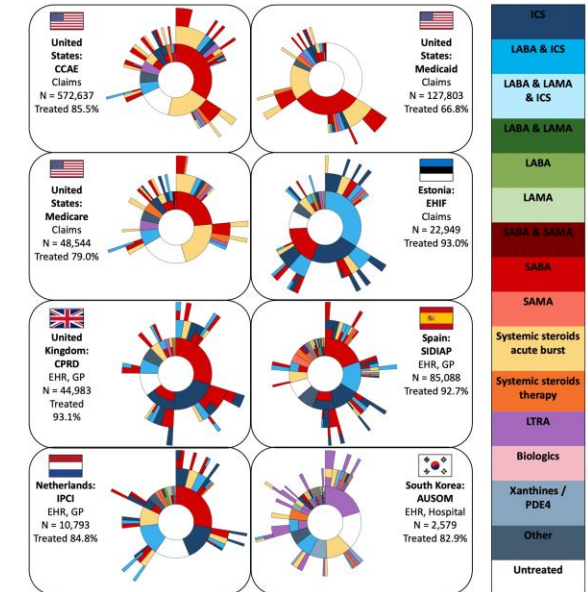


FIGURE 1 Sunburst plots visualizing treatment pathways for TTS patients in pre-vaccination cohorts (top) versus vaccinated cohorts (bottom). Inpatient databases are depicted with a dashed line frame, whilst outpatient ones have a solid frame.

Markus et al. (2023)



Markus et al. (2024)

- **A lot of potential:** use the OHDSI data network to get high-level insight in current treatment practices across different *patient populations, care settings, and countries* in Europe (and globally)



Computer Methods and Programs in  
Biomedicine  
Volume 225, October 2022, 107081



### TreatmentPatterns: An R package to facilitate the standardized development and analysis of treatment patterns across disease domains

Aniek F. Markus<sup>a</sup>, Katia M.C. Verhamme<sup>a, b</sup>, Jan A. Kors<sup>a</sup>, Peter R. Rijnbeek<sup>a</sup>

Show more

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<https://doi.org/10.1016/j.cmpb.2022.107081>

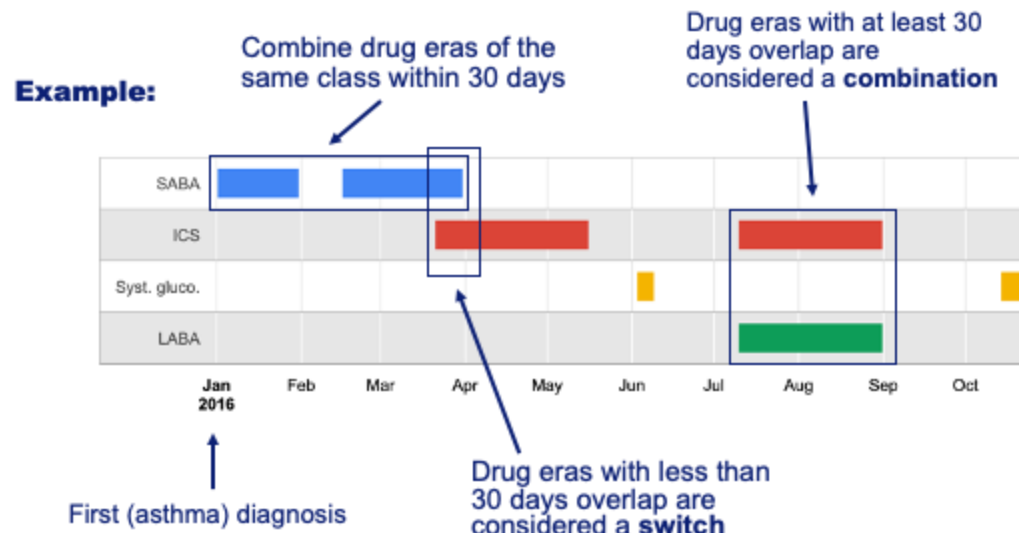
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#### Highlights

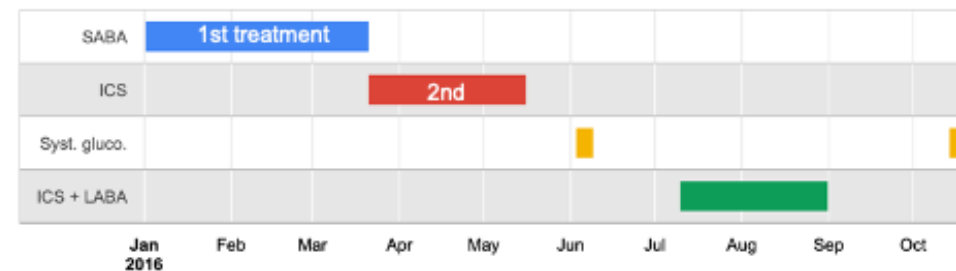
- TreatmentPatterns R package for development and analysis of treatment patterns.
- Formally defines and implements the process of constructing treatment pathways.
- Accessible, standardized, and interpretation friendly tool for a broad audience.
- Supports the accumulation of knowledge on treatment patterns across disease domains.
- Example study in Dutch IPCI database to demonstrate functionalities of the package.



**More info?**

- The Book of OHDSI chapter 11.3 + 11.9
- EHDEN academy course in progress

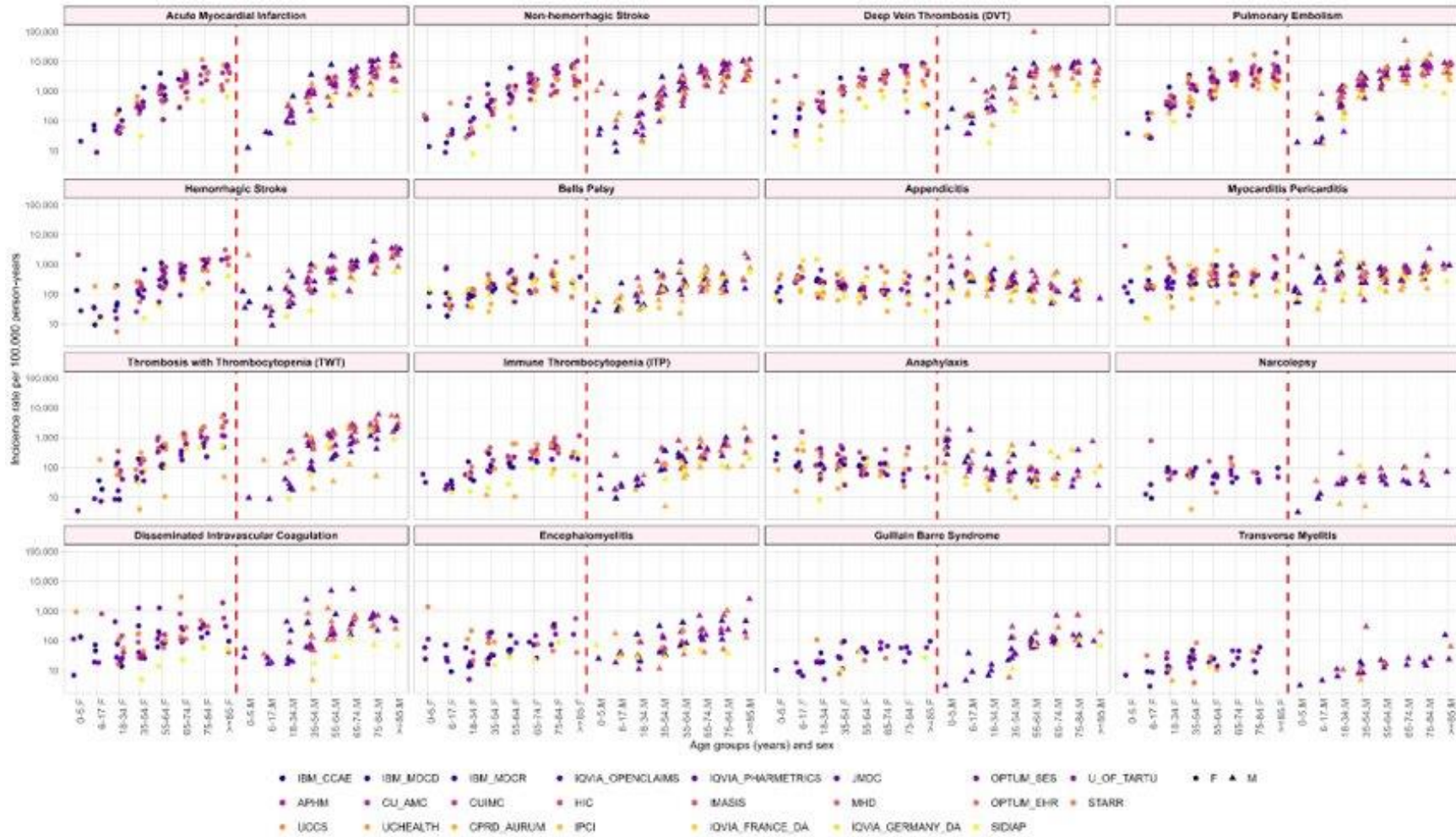
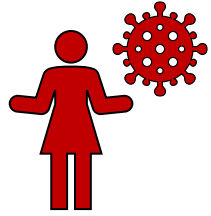
### Example: SABA – ICS – Systemic glucocorticoids – ICS + LABA



<https://darwin-eu.github.io/TreatmentPatterns/>



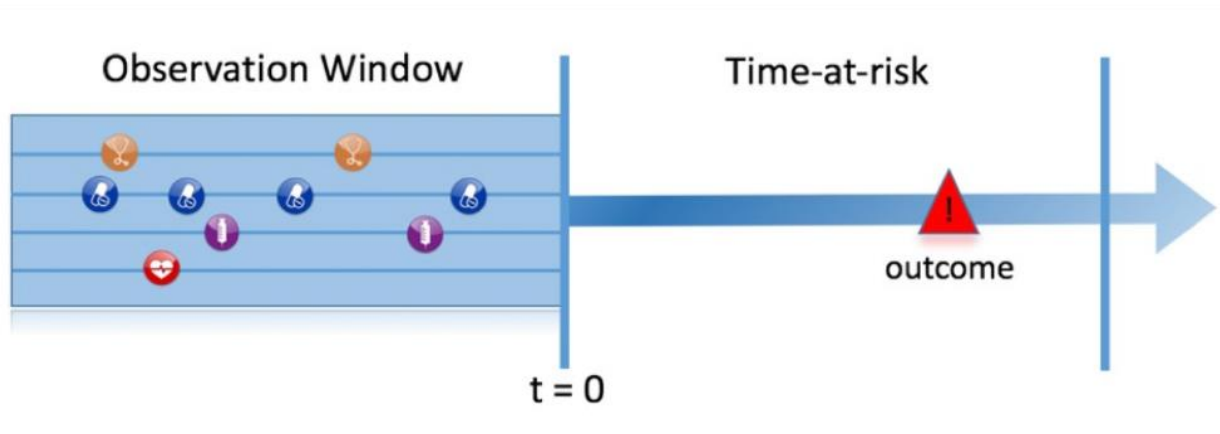
# CHARACTERIZING VACCINE ADVERSE EVENTS IN COVID-19 PATIENTS ACROSS THE OHDSI NETWORK \* = LARGE-SCALE EVIDENCE



Voss et al. (2023)

26 databases  
11 countries

- What will happen to me?



## Patient-level prediction

### PatientLevelPrediction

Build and evaluate predictive models for user-specified outcomes, using a wide array of machine learning algorithms.

[Learn more...](#)

### DeepPatientLevelPrediction

Performing patient level prediction using deep learning

[Learn more...](#)

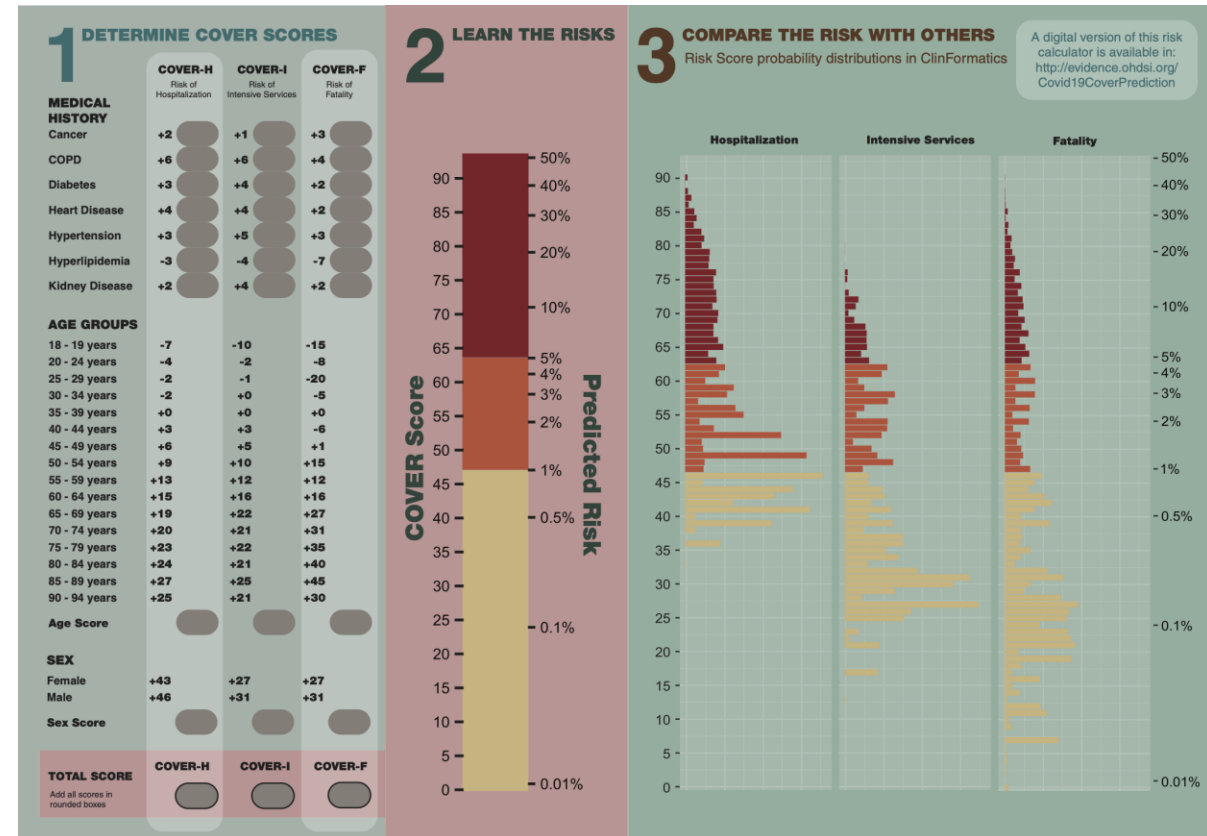
### EnsemblePatientLevelPrediction

Building and validating ensemble patient-level predictive models.

[Learn more...](#)

Objective: develop and externally validate **COVID-19 Estimated Risk** scores that quantify a patient's risk of hospital admission, hospitalization requiring intensive services or fatality.

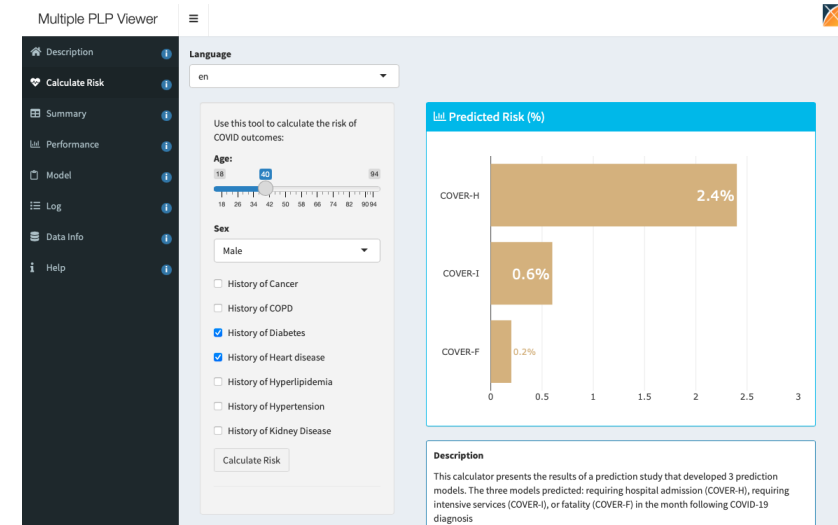
- 14 data sources from 6 countries
- Externally validated in 44,507 COVID cases from 5 data sources in South Korea, Spain, USA



<https://data.ohdsi.org/Covid19CoverPrediction/>

Williams, R.D., Markus, A.F., Yang, C. et al. Seek COVER: using a disease proxy to rapidly develop and validate a personalized risk calculator for COVID-19 outcomes in an international network. BMC Med Res Methodol 22, 35 (2022). <https://doi.org/10.1186/s12874-022-01505-z>

- COVER interactive [website](#) to provide risk scores.
- **Impact:** Health minister of Catalonia Spain explicitly mentioned the COVER index as one of the indicators they used to measure the impact of a given outbreak.



### 3. Indicadors

► El Pla es basa en la mesura de **10 indicadors principals** que permeten una fotografia acurada de la realitat epidèmica a Catalunya.



► En la interpretació dels indicadors s'aplicaran **factors de correcció** com: índex socioeconòmic complex, envelliment de la població o la densitat poblacional.



International Journal of Medical Informatics  
Volume 163, July 2022, 104762



## Logistic regression models for patient-level prediction based on massive observational data: Do we need all data?

Luis H. John<sup>a</sup>, Jan A. Kors<sup>a</sup>, Jenna M. Reps<sup>b</sup>, Patrick B. Ryan<sup>b</sup>, Peter R. Rijnbeek<sup>a</sup>

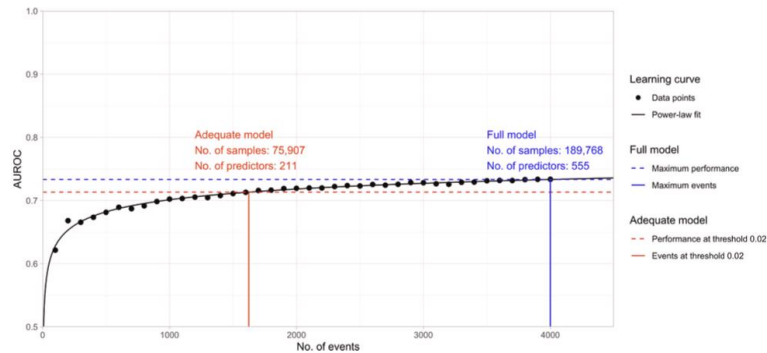


Fig. 1. Learning curve for the prediction of venous thromboembolic events in patients with hypertension using data from MDCR. The horizontal lines indicate the maximum performance of the fitted curve (blue) and the performance at a threshold of 0.02 (red). The vertical lines denote the maximum number of events (blue) and the adequate number of events (red). Number of samples and predictors shown for the adequate model pertain to the model at the closest data point. (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)

"Our results suggest that in most cases **only a fraction of the available data was sufficient** to produce a model close to the performance of one developed on the full data set, but with a substantially reduced model complexity."

Reps et al. *BMC Med Inform Decis Mak* (2021) 21:43  
<https://doi.org/10.1186/s12911-021-01408-x>

BMC Medical Informatics and Decision Making

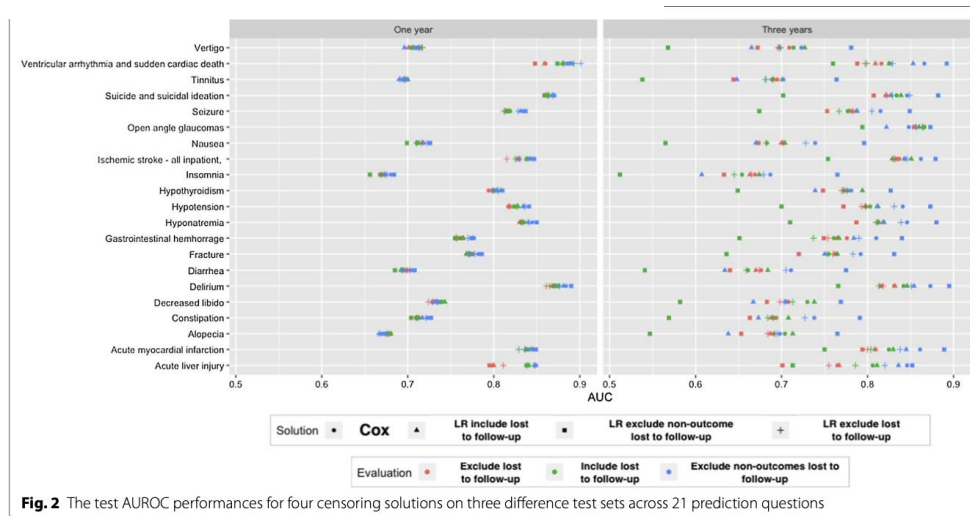
RESEARCH ARTICLE

Open Access



## An empirical analysis of dealing with patients who are lost to follow-up when developing prognostic models using a cohort design

Jenna M. Reps<sup>1</sup>, Peter Rijnbeek<sup>2</sup>, Alana Cuthbert<sup>3</sup>, Patrick B. Ryan<sup>1</sup>, Nicole Pratt<sup>4</sup> and Martijn Schuemie<sup>1</sup>



"Based on this research it appears that it is best to develop models using data that **includes patients that are lost to follow-up**. However, recalibration is likely to be required as this strategy appears to result in models that under-estimate risk."

## Design and implementation of a standardized framework to generate and evaluate patient-level prediction models using observational healthcare data

Jenna M Reps, Martijn J Schuemie, Marc A Suchard, Patrick B Ryan, Peter R Rijnbeek

Journal of the American Medical Informatics Association, Volume 25, Issue 8, August 2018, Pages 969–975, <https://doi.org/10.1093/jamia/ocy032>

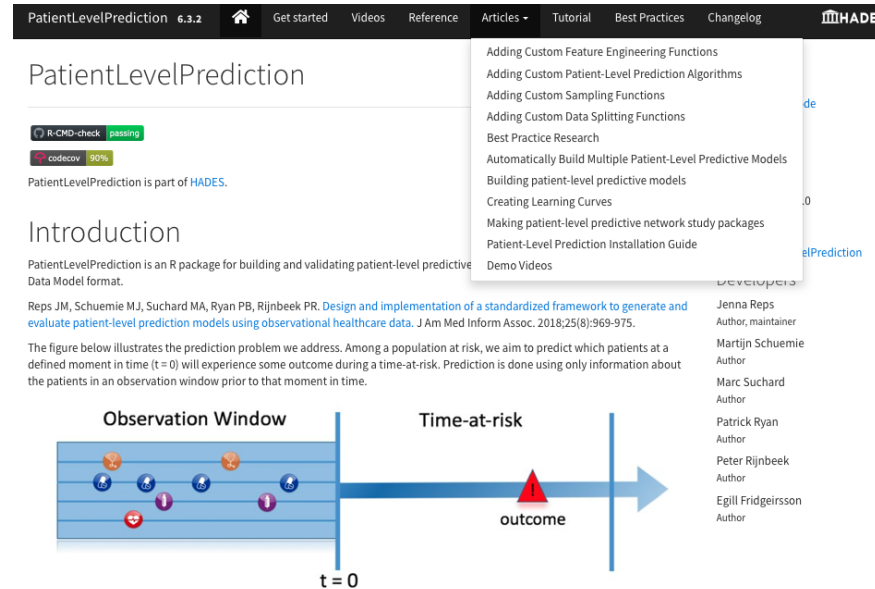
Published: 27 April 2018 [Article history](#)

PDF Split View Cite Permissions Share

### Abstract

#### Objective

To develop a conceptual prediction model framework containing standardized steps and describe the corresponding open-source software developed to consistently implement the framework across computational environments and observational healthcare databases to enable model sharing and reproducibility.



## Best practice publications using the OHDSI PatientLevelPrediction framework

Topic	Research Summary	Link
Problem Specification	When is prediction suitable in observational data?	Guidelines needed
Data Creation	Comparison of cohort vs case-control design	<a href="#">Journal of Big Data</a>
Data Creation	Addressing loss to follow-up (right censoring)	<a href="#">BMC medical informatics and decision making</a>
Data Creation	Investigating how to address left censoring in features construction	<a href="#">BMC Medical Research Methodology</a>
Data Creation	Impact of over/under-sampling	Paper under review
Data Creation	Impact of phenotypes	Study Done - Paper submitted
Model development	How much data do we need for prediction - Learning curves at scale	<a href="#">International Journal of Medical Informatics</a>
Model development	What impact does test/train/validation design have on model performance	<a href="#">BMJ Open</a>
Model development	What is the impact of the classifier	<a href="#">JAMIA</a>
Model development	Can we find hyper-parameter combinations per classifier that consistently lead to good performing models when using claims/EHR data?	Study needs to be done
Model development	Can we use ensembles to combine different algorithm models within a database to improve models transportability?	Study Complete
Model development	Can we use ensembles to combine models developed using different databases to improve models transportability?	<a href="#">BMC Medical Informatics and Decision Making</a>
Evaluation	How should we present model performance? (e.g., new visualizations)	<a href="#">JAMIA Open</a>

### More info?

- The Book of OHDSI chapter 13
- EH DEN academy course Patient Level Prediction + R for Patient Level Prediction

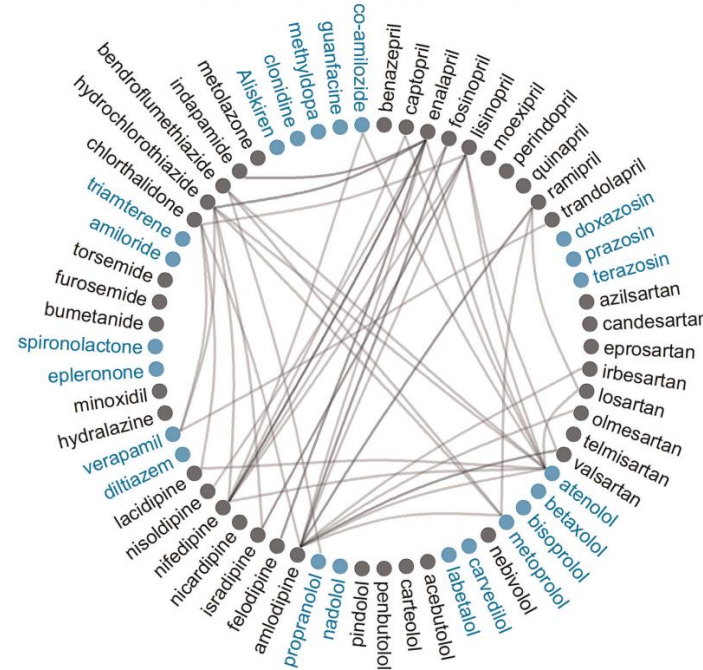
R package: <https://ohdsi.github.io/PatientLevelPrediction/>

## Research and Applications

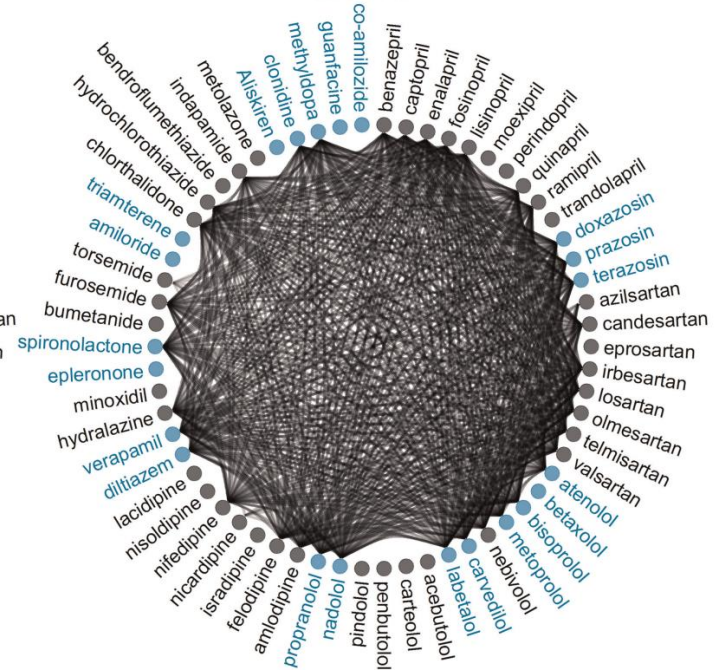
### Large-scale evidence generation and evaluation across a network of databases (LEGEND): assessing validity using hypertension as a case study

Martijn J Schuemie<sup>1,2</sup>, Patrick B Ryan<sup>1,3</sup>, Nicole Pratt<sup>4</sup>, RuiJun Chen<sup>3,5</sup>, Seng Chan You<sup>6</sup>, Harlan M Krumholz<sup>7</sup>, David Madigan<sup>8</sup>, George Hripcsak<sup>3,9</sup> and Marc A Suchard<sup>2,10</sup>

Randomized controlled trials



LEGEND



**Figure 3.** Comparisons of single-drug hypertension treatments in randomized controlled trials (left) and in LEGEND (right). Each circle represents an ingredient. Color groupings indicate drug classes. A line between circles indicates the 2 drugs are compared in at least 1 study.

**Results:** From 21.6 million unique antihypertensive new users, we generate 6 076 775 effect size estimates for 699 872 research questions on 12 946 treatment comparisons. Through propensity score matching, we achieve balance on all baseline patient characteristics for 75% of estimates, observe 95.7% coverage in our effect-estimate 95% confidence intervals, find high between-database consistency, and achieve transitivity in 84.8% of triplet hypotheses. Compared with meta-analyses of RCTs, our results are consistent with 28 of 30 comparisons while providing narrower confidence intervals.

Perspective

## Principles of Large-scale Evidence Generation and Evaluation across a Network of Databases (LEGEND)

Martijn J. Schuemie<sup>1,2</sup>, Patrick B. Ryan<sup>1,3</sup>, Nicole Pratt<sup>4</sup>, RuiJun Chen<sup>3,5</sup>, Seng Chan You<sup>6</sup>, Harlan M. Krumholz<sup>7</sup>, David Madigan<sup>8</sup>, George Hripcsak<sup>3,9</sup>, and Marc A. Suchard<sup>2,10</sup>



## Towards Reliable Evidence ...

Some current practices across the broader research community

Examining one target-comparator pair at a time

Not using appropriate methods to control for bias

Modify the design until significant results are found



Large-scale evidence generation across a network of databases (LEGEND)

- \* Pre-specified fixed design and dissemination of the results regardless of the estimates (avoid publication bias)
- \* Systematic process across all research questions

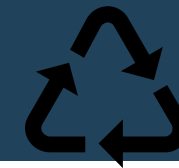
- \* Large-scale: looking at thousands of target-comparator pairs at a time
- \* Use of best practices: LSPS, extensive diagnostics, negative and positive controls

### More info?

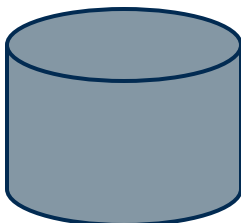
- OHDSI website  
<https://www.ohdsi.org/legend-oct2021-update/>
- Various publications  
e.g. [10.1093/jamia/ocaa103](https://doi.org/10.1093/jamia/ocaa103)



# OHDSI TOOLS (OPEN-SOURCE)



Source data



Summarize the source data  
 WhiteRabbit

Create ETL design  
 RABBIT IN A HAT

Semi-automated vocabulary mapping  
 USAGI

OMOP-CDM data



Analyses



Reliable Evidence



Generate descriptive statistics on OMOP CDM  
**ACHILLES**

Perform data quality checks  
 DATA QUALITY ASSESSMENT

**ARES**

Phenotype development

PHOEBE

Codelist Generator

Exploring vocabulary  
 ATHENA

Cohort Diagnostics R package

PheValuator

Phenotype Library

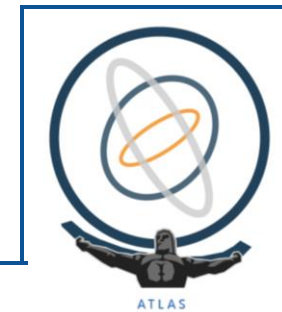
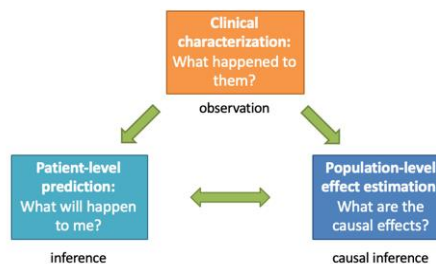
Study design

Characterization R package

PatientLevel Prediction R package

CohortMethodR package

+

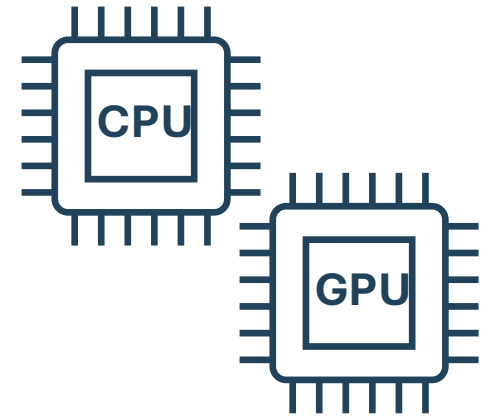
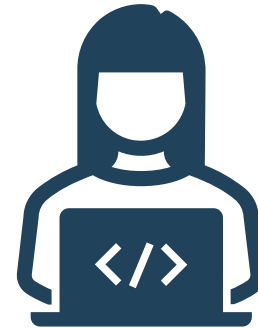


**ATLAS:** An open-source software tool for data exploration and scientific analyses on OMOP CDM data

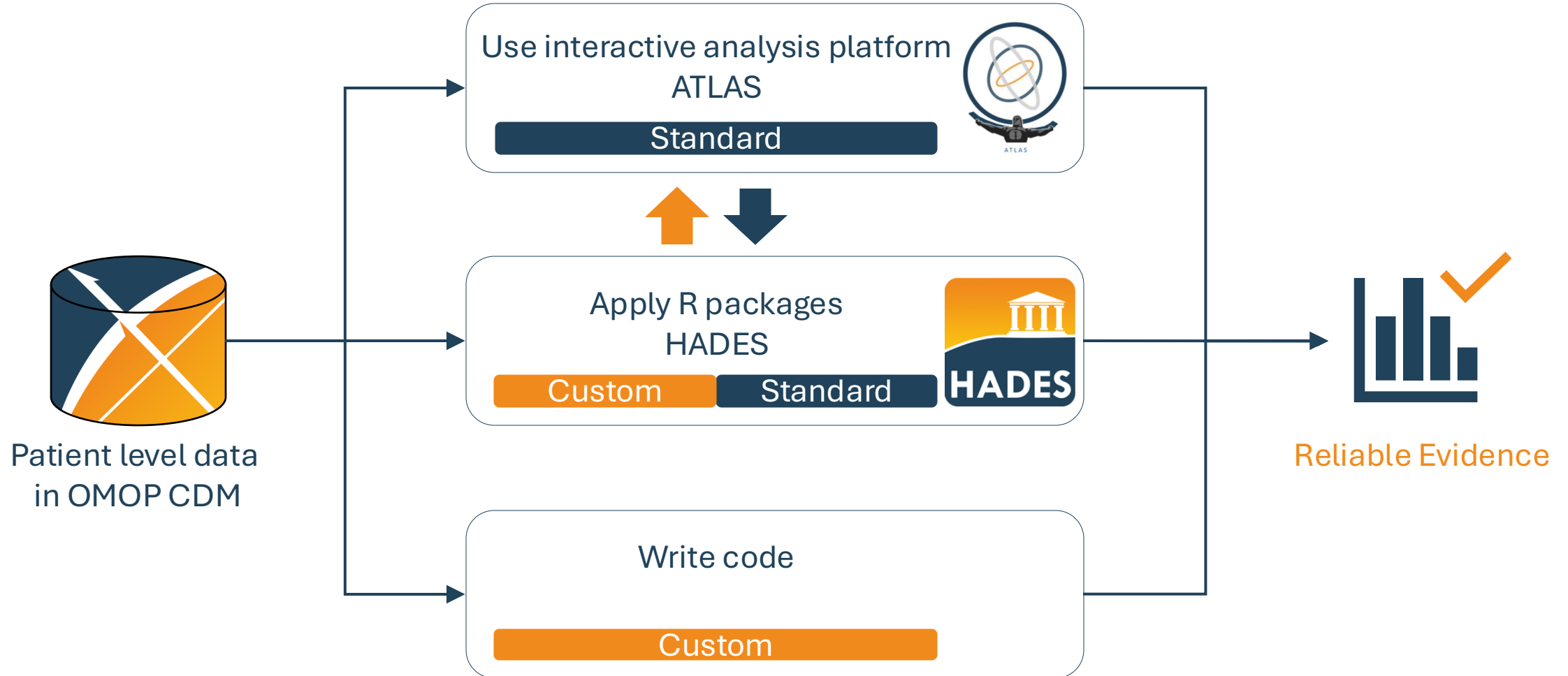
- Given what you know now, discuss with your neighbour:
  - **Which opportunities do you see for using the OHDSI tools?**
  - **Which tools can you see yourself using?**
  - **Are you ready to start using the tools? Why (not)?**

# WHAT DOES IT TAKE TO BE ABLE TO USE THE TOOLS?

# WHAT DOES IT TAKE TO BE ABLE TO USE THE TOOLS?



# OHDSI ROADS TO RELIABLE EVIDENCE





ATLAS is a free, publicly available web-based, open-source software application developed by the OHDSI community to support the design and execution of observational analyses to generate real world evidence from patient level observational data. Atlas is an open science analytics platform that can be installed locally within your institution to perform analyses across one or more observational databases which have been standardized to the [OMOP Common Data Model V5](#) and can facilitate exchange of analysis designs with any other organizations across the OHDSI community who have adopted the same open science community standards and tools.

The screenshot shows the ATLAS web application interface. On the left is a dark sidebar menu with the ATLAS logo at the top and various navigation items: Home, Data Sources, Search, Concept Sets, Cohort Definitions, Characterizations, Cohort Pathways, Incidence Rates, Profiles, Estimation, Prediction, Reusables, Jobs, Configuration, and Feedback. The main content area has a dark header with 'English' and a notification bell. Below the header, it says 'Welcome to ATLAS.' and provides a brief description. Under 'Documentation', it links to the user guide. The 'Getting Started' section features two green buttons: 'Define a New Cohort' (with the text 'Begin performing research by defining the group of people you intend to study') and 'Search the Vocabulary' (with the text 'Search the different ontologies used to describe patient level data around the world'). Below this is a 'Release Notes' section with links to 'ATLAS Version 2.12.1 Release Notes' and 'WebAPI Version 2.12.1 Release Notes', followed by a list of updates. A language dropdown menu is open on the right, showing 'English' (checked), 'Русский', '한국어', and '中文'.

**More info?**  
<https://atlas-demo.ohdsi.org/>  
[https://youtu.be/dr9FhE\\_kf04o?si=V5Cl4jC3bVLbnFi3](https://youtu.be/dr9FhE_kf04o?si=V5Cl4jC3bVLbnFi3)

# EXAMPLE: DEFINING A COHORT (E.G. STUDY POPULATION)

More info?  
- The Book of  
OHDSI chapter 10

**Cohort Entry Events** ⓘ

Events having any of the following criteria: + Add Initial Event... -

a condition occurrence of **ESRD** + Add attribute... - Delete Criteria

✖ occurrence start is: **between** 2010-01-01 and 2020-01-01

with continuous observation of at least **0** days before and **0** days after event index date

Limit initial events to: **earliest event** per person.

Restrict Initial events

---

**Inclusion Criteria** ⓘ

New inclusion criteria

Limit qualifying events to: **earliest event** per person.

---

**Cohort Exit** ⓘ

**Event Persistence:**  
Event will persist until: **end of continuous observation**

**Censoring Events:**  
Exit Cohort based on the following criteria: + Add Censoring Event... -

No censoring events selected.

---

**Cohort Eras**

- Specify era collapse gap size: **0** days
- [add trimming options...](#)



Population-level estimation

Patient-level prediction

Characterization

Cohort construction and evaluation

Evidence Quality

Supporting packages

## Packages

Below are the packages included in HADES. For each package a link is provided with more information, including instructions on how to install and use the package.

### Population-level estimation

#### CohortMethod

New-user cohort studies using large-scale regression for propensity and outcome models.

[Learn more...](#)

#### SelfControlledCaseSeries

Self-Controlled Case Series analysis using few or many predictors, includes splines for age and seasonality.

[Learn more...](#)

#### SelfControlledCohort

A self-controlled cohort design, where time preceding exposure is used as control.

[Learn more...](#)

#### EvidenceSynthesis

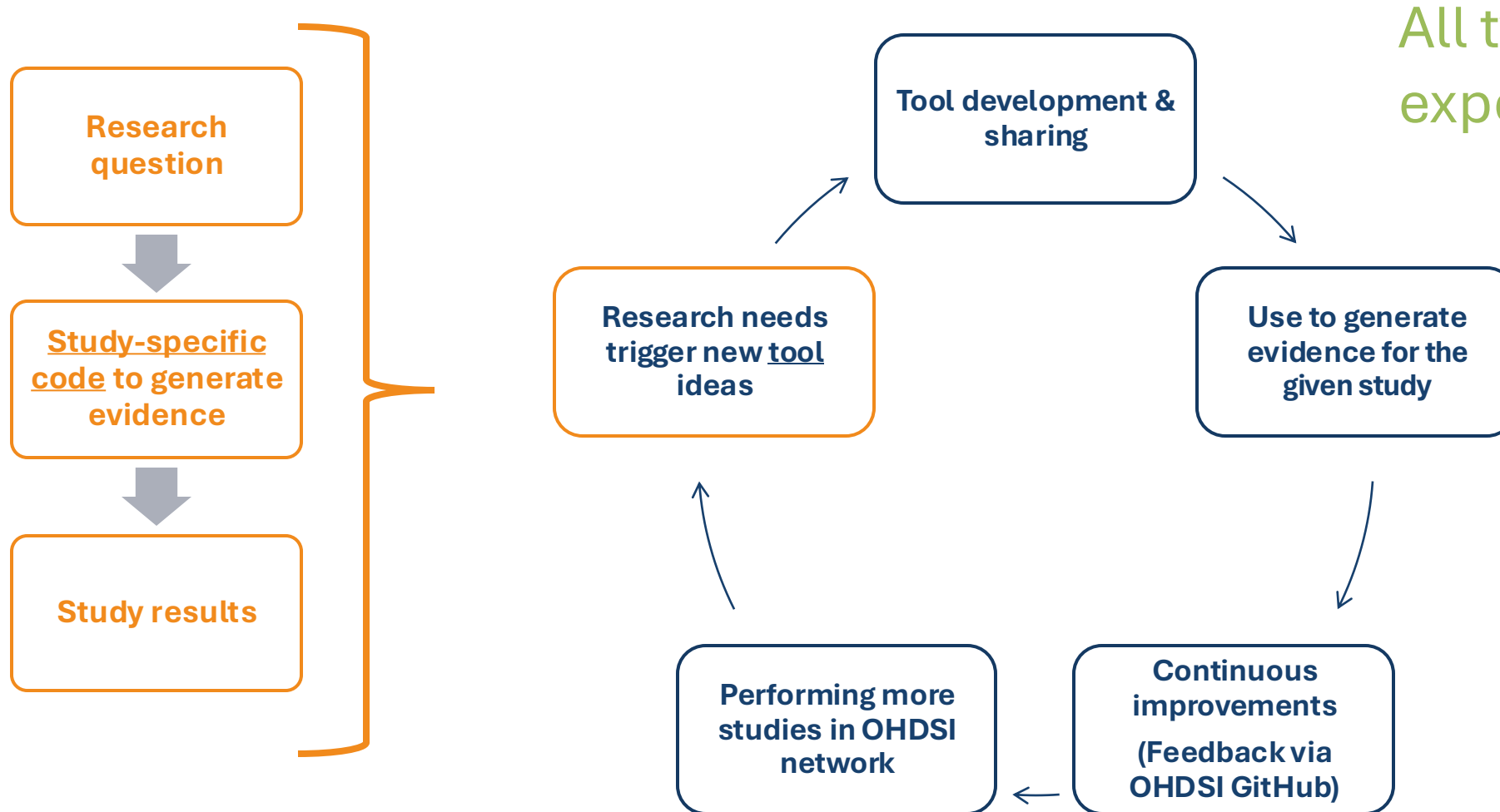
Routines for combining causal effect estimates and study diagnostics across multiple data sites in a distributed study.

[Learn more...](#)

More info?

<https://ohdsi.github.io/Hades/>

# OHDSI = TEAMWORK



All types of expertise needed!

- Any questions for us?

Tackle us, or our colleagues  
in the next few days



# TIPS FOR OHDSI 2026 SYMPOSIUM

- Visit your national node (poster area on Monday)
- Mingle! (the unscheduled events are often most valuable ;-))





# OHDSI EUROPE 2026 – WHAT'S AFTER LUNCH?

<p>13:00 – 15:00</p>		<p><b>ETL development and updating</b> Maxim Moinat (EMC), Anne van Winzum, Stefan Payralbe (The Hyve)</p> <p><u>Description</u></p> <ul style="list-style-type: none"> <li>• Short introduction to OMOP ETL conventions and ETL implementation examples.</li> <li>• Latest developments in available tooling to assist with ETL/mappingTools</li> <li>• Importance and best practices for maintaining and updating ETL/mapping after initial conversion</li> </ul> <p><u>Target audience</u></p> <p>New and current data holders, ETL developers, data engineers responsible for OMOP CDM conversions and ongoing maintenance.</p>	<p><b>Phenotyping in OHDSI: pipelines, steps &amp; tools</b> Anna Ostropolets (J&amp;J), Maria Khitrin, Azza Shoaibi, Dmitro Dymshyts, Anna Saura Lazaro</p> <p><u>Description</u></p> <ul style="list-style-type: none"> <li>• Explore existing pipelines for phenotyping within OHDSI</li> <li>• Explore open-source tools that exists to help researchers with individual phenotyping tasks, including creating clinical description for the clinical idea, literature review and concept sets, phenotype evaluation and storage &amp; phenotype maintenance across different OHDSI Standardized Vocabularies versions</li> </ul> <p><u>Target audience</u></p> <p>Researchers who design studies and build phenotypes and anyone who want to learn more about phenotyping</p>
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- What is the current version of the OMOP-CDM?
- To participate in a network study, you need to share your source data. True or false?
- If you don't have data, you can't join OHDSI. True or false?
- Treatment patterns are awesome. True or false?
- OHDSI will add value to my organization. True or false?

## Learn more:

- Book of OHDSI: <https://ohdsi.github.io/TheBookOfOhdsi/>
  - Follow tutorials on the EHDEN academy: <https://academy.ehden.eu>
  - Read publications about OHDSI tools and pipelines
- ➔ Start getting hands-on experience!

## Stay updated: <https://www.ohdsi.org/ohdsi-news-updates/>

- Weekly community calls – Tuesday 17:00-18:00
- Plenty of workgroups (<https://www.ohdsi.org/workgroups/>)