

# **2019 April Business Meeting**



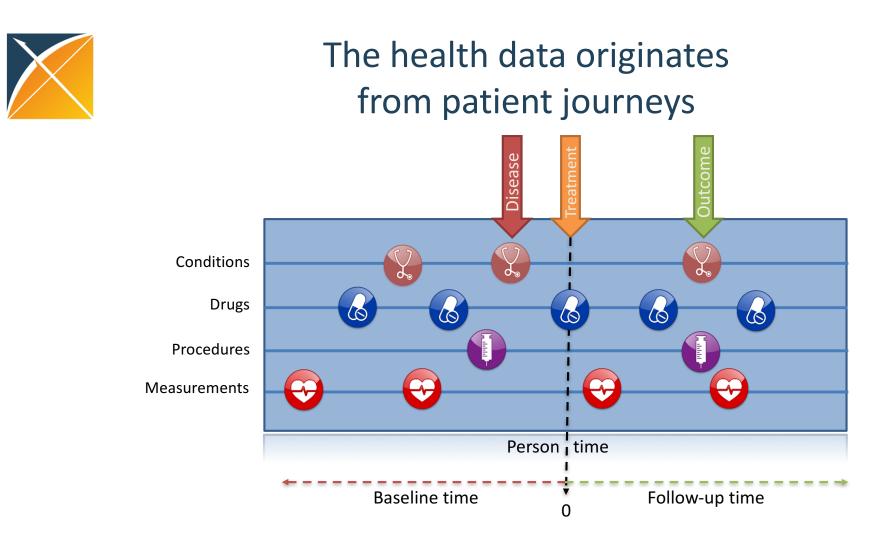
### Enabling Large-Scale Analysis of Electronic Health Records by Standardizing to a Common Data Model

Peter R. Rijnbeek, PhD Associate Professor Health Data Science Department of Medical Informatics Erasmus MC, The Netherlands



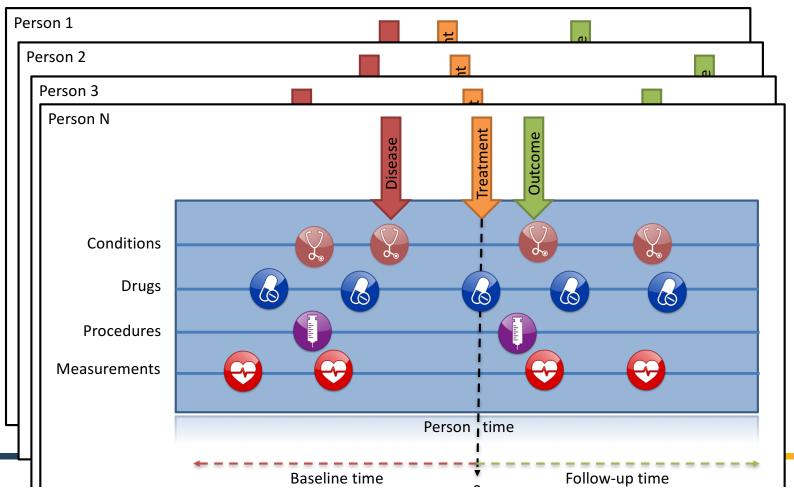
# The Journey to large-scale analytics

- Introduction to the use of a Common Data Model and Standardized Vocabularies
- Example: Large-scale Patient-Level Prediction
- The European OHDSI Symposium
- The European Health Data and Evidence (EHDEN) Project



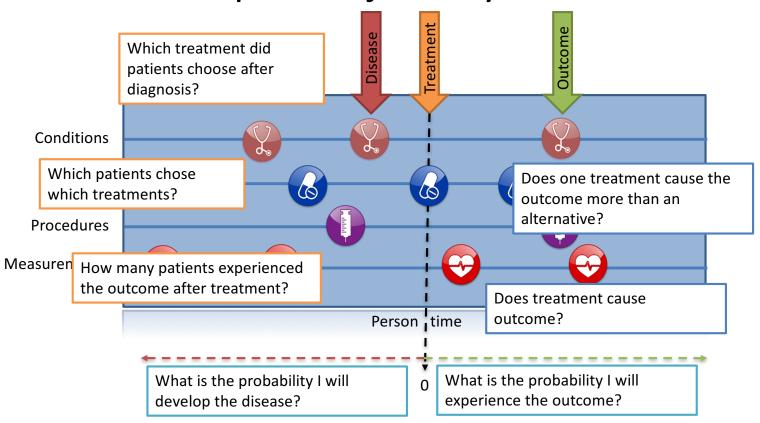


# Each observational database is just an (incomplete) compilation of patient journeys



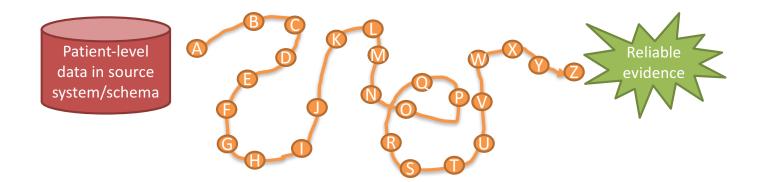


# Questions asked across the patient journey





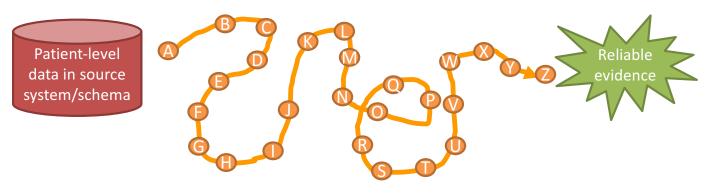
# The journey to real-world evidence: a fully reproducible data flow





# Minimum requirements to achieve reproducibility

Desired attribute	Question	Researcher	Data	Analysis		Result
Reproducible	Identical	Different	Identical	Identical	=	Identical

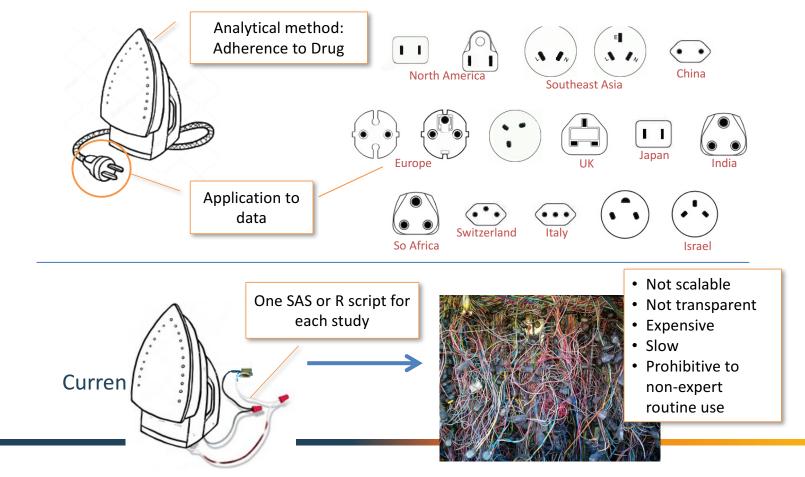


- Complete documented specification that fully describes all data manipulations and statistical procedures
- Full analysis code that executes end-to-end (from source to results) without manual intervention



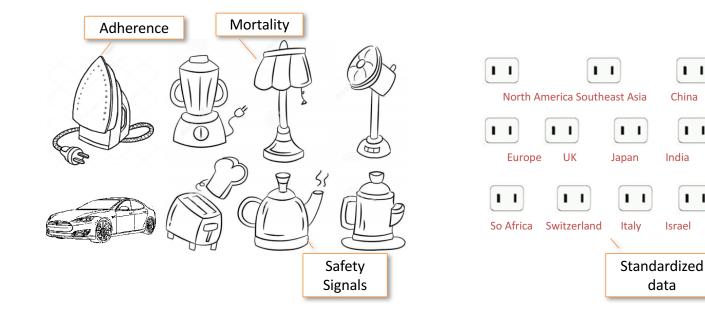
#### Current Approach: "One Study – One Script"

"What's the adherence to my drug of interest?"





### Solution: Data Standardization **Enables Systematic Research**



**Analytical Tools** 

**CDM** 

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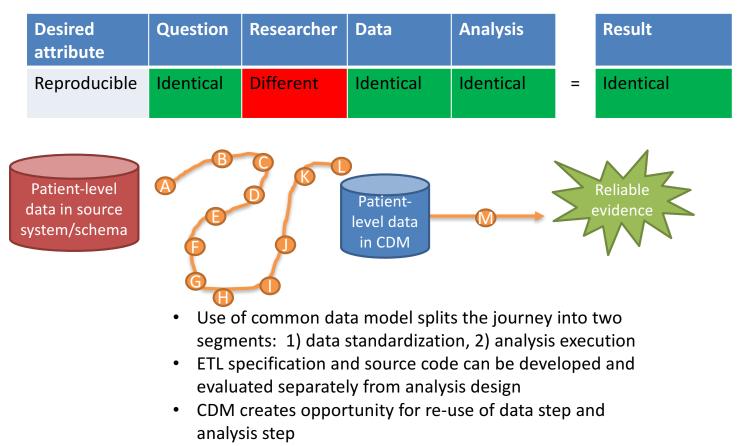
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#### How a common data model + common analytics can support reproducibility





Observational Health Data Sciences and Informatics (OHDSI) mission

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

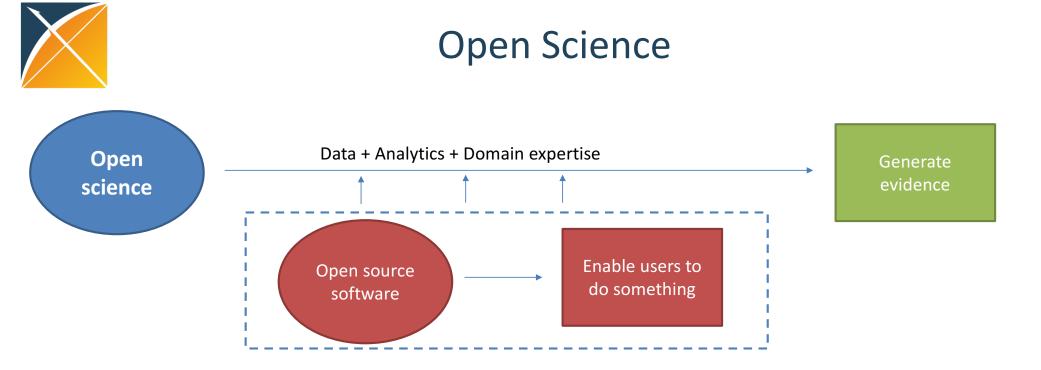


## OHDSI's global research community



- >200 collaborators from 25 different countries
- Experts in informatics, statistics, epidemiology, clinical sciences
- Active participation from academia, government, industry, providers
- Currently records on about 500 million unique patients in >100 databases

http://ohdsi.org/who-we-are/collaborators/

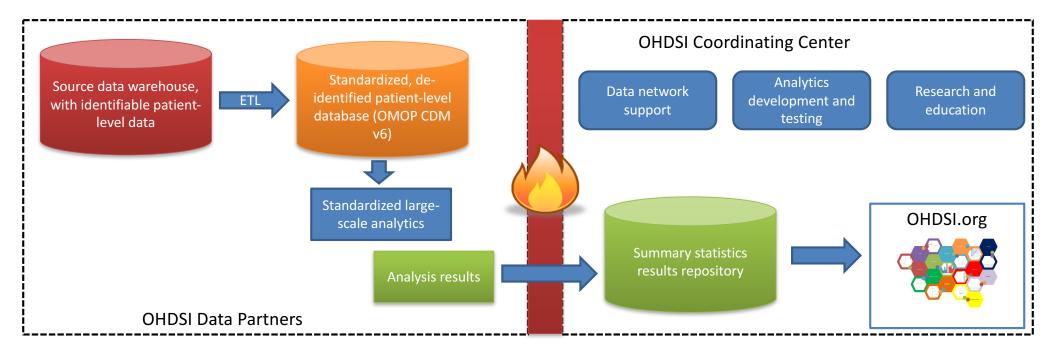


#### Standardized, transparent workflows

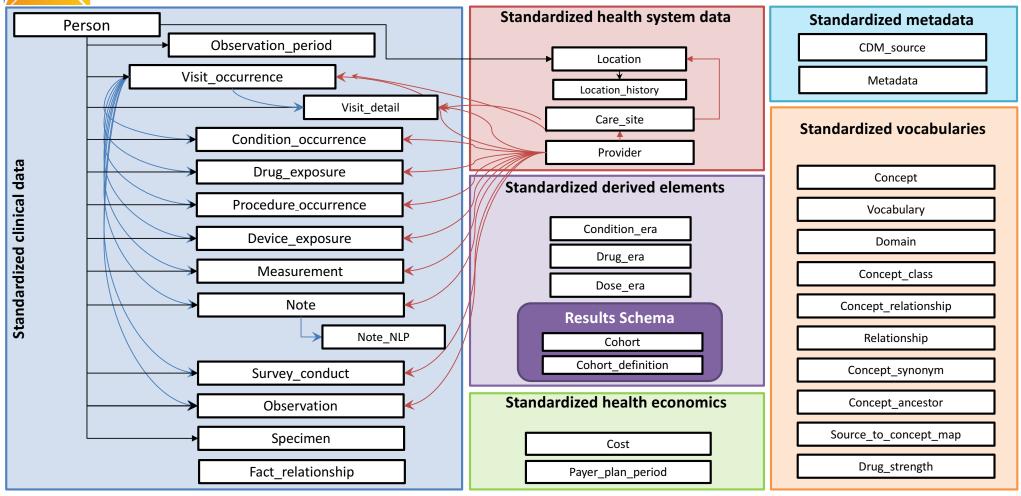
Database summary	Cohort definition	Cohort summary	Compare cohorts	Exposure-outcome summary	Effect estimation & calibration	Compare databases	
	/						

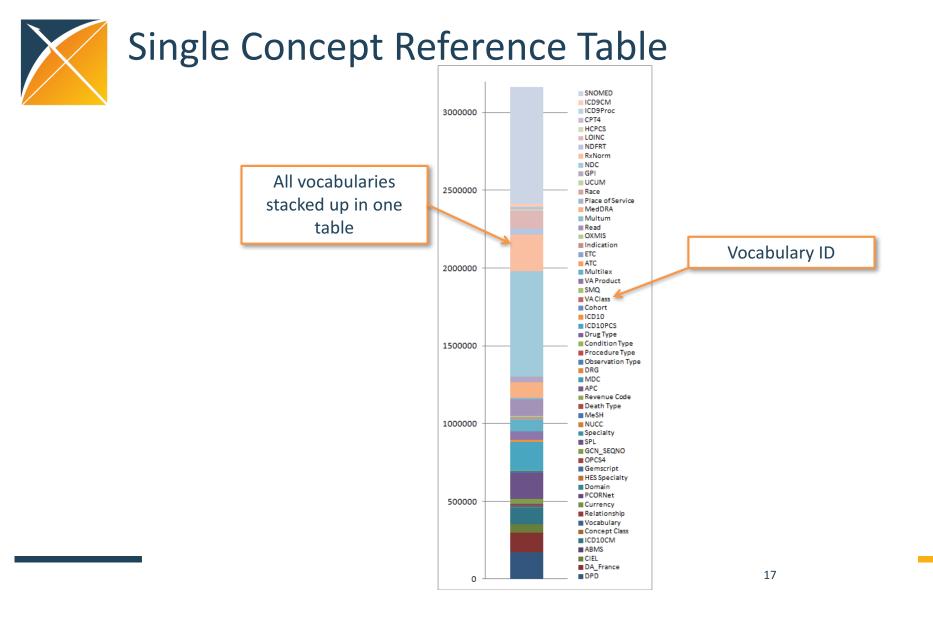


#### How OHDSI Works



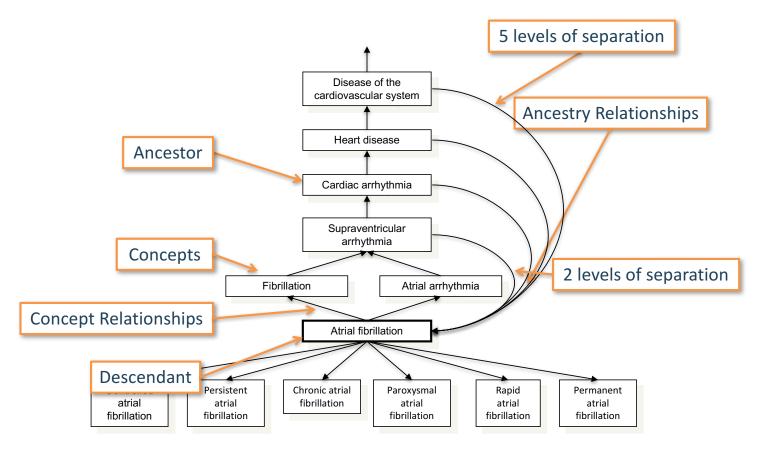
### Deep information model OMOP CDM Version 6







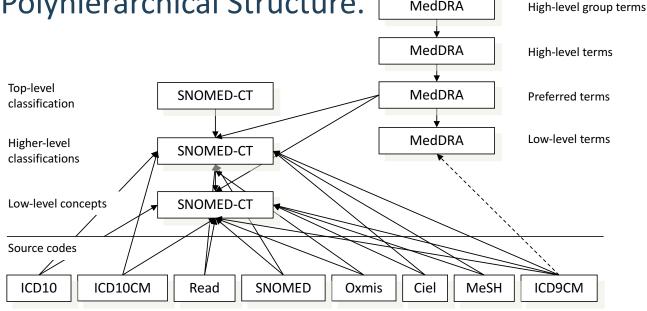
#### Ancestry Relationships: Higher-Level Relationships





# Use of SNOMED in the Standardized Vocabularies

- SNOMED is the standard in several domains, e.g. conditions, procedure. MedDRA System organ class
- Powerful Polyhierarchical Structure.



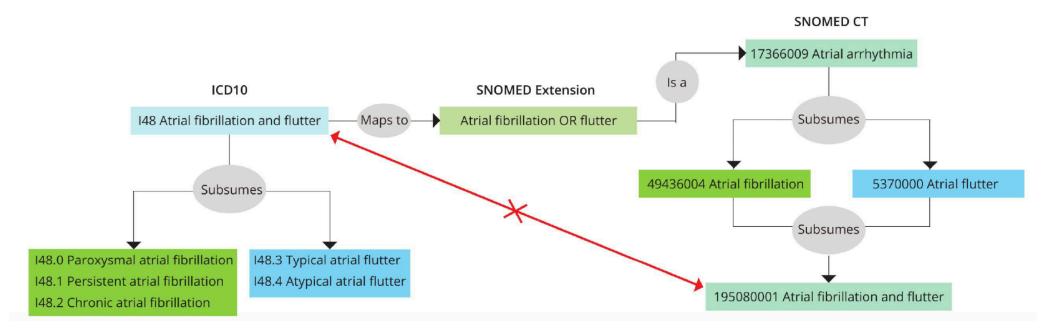


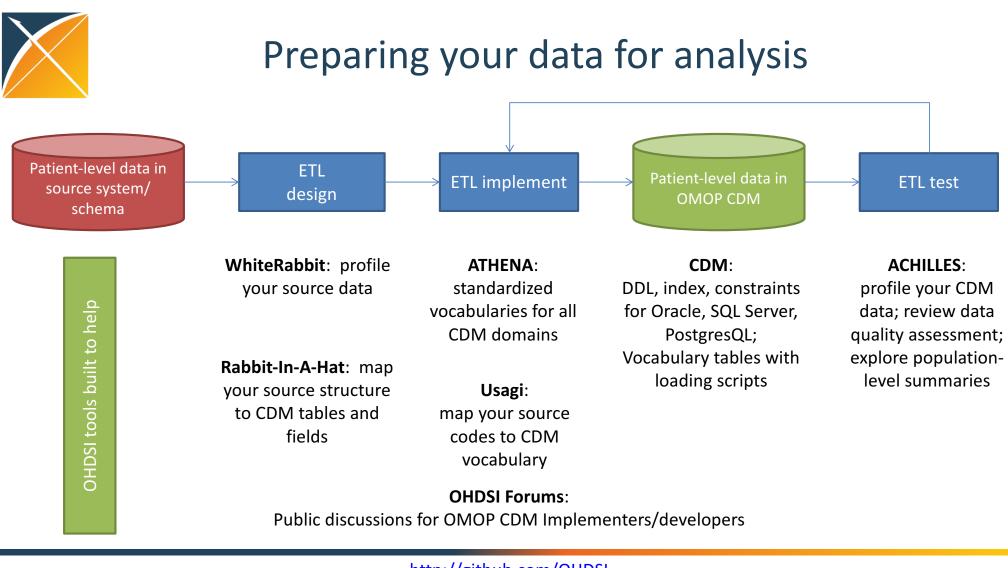
# **SNOMED** Challenges

- We have to making mappings from many source coding systems to SNOMED in Europe.
- We want to use SNOMED across the world: how to deal with countries that do not (yet) have a license?
- We will require SNOMED extensions to accommodate differences in granularity or classification differences.



## **Example: SNOMED Challenges**





http://github.com/OHDSI



## **ACHILLES Heel Data Curation**

Data Quality Messages	
	Search: Show / hide columns
Message Type	▲ Message
ERROR	101-Number of persons by age, with age at first observation period; should not have age < 0, (n=848)
ERROR	103 - Distribution of age at first observation period (count = 1); min value should not be negative
ERROR	114-Number of persons with observation period before year-of-birth; count (n=851) should not be > 0
ERROR	206 - Distribution of age by visit_concept_id (count = 7); min value should not be negative
ERROR	301-Number of providers by specialty concept_id; 224 concepts in data are not in correct vocabulary (Specialty)
ERROR	400-Number of persons with at least one condition occurrence, by condition_concept_id; 115 concepts in data are not in correct vocabulary (SNOMED)
ERROR	406 - Distribution of age by condition_concept_id (count = 753); min value should not be negative



# ATLAS to build, visualize, and analyze cohorts

- People having any of the following: Add Primary Criteria	
a condition occurrence of Delivery  Add Criterion  Coccurrence start is: Between  2005-01-01 and 2013-12-31  With age Between  18 and 55  Kwith a gender of: KFEMALE Add Import	- Delete
with observation at least 180 V days prior and 365 V days after index Limit primary events to: All Events V per person.	
For people matching the Primary Criteria, include:         — People having All ▼ of the following criteria:         Add New Criteria▼	
with At Least 🔻 1 💌 occurrences of:	Add Criterion 🔻
a condition occurrence of Depression	
occurring between 0 ▼ days Before ▼ and 180 ▼ days After ▼ index	Delete Criteria
and with At Most 🔻 0 🔻 occurrences of:	Add Criterion 👻
a condition occurrence of Depression	
occurring between All ▼ days Before ▼ and 0 ▼ days After ▼ index	Delete Criteria



## Characterize the cohorts of interest





## What is OHDSI's strategy to deliver reliable evidence?

#### • Methodological research

- Develop new approaches to observational data analysis
- Evaluate the performance of new and existing methods
- Establish empirically-based scientific best practices

#### Open-source analytics development

- Design tools for data transformation and standardization
- Implement statistical methods for large-scale analytics
- Build interactive visualization for evidence exploration

#### Clinical evidence generation

- Identify clinically-relevant questions that require real-world evidence
- Execute research studies by applying scientific best practices through open-source tools across the OHDSI international data network
- Promote open-science strategies for transparent study design and evidence dissemination



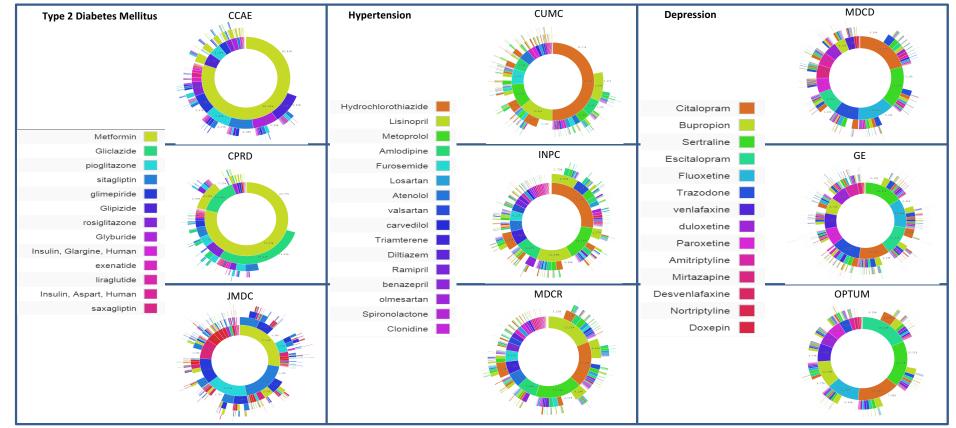
# Evidence OHDSI seeks to generate from observational data

- Clinical characterization
  - Natural history: Who has diabetes, and who takes metformin?
  - Quality improvement: What proportion of patients with diabetes experience complications?
- Population-level effect estimation
  - Safety surveillance: Does metformin cause lactic acidosis?
  - Comparative effectiveness: Does metformin cause lactic acidosis more than glyburide?
- Patient-level prediction
  - Precision medicine: Given everything you know about me, now I started using metformin, what is the chance I will get lactic acidosis?
  - Disease interception: Given everything you know about me, what is the chance I will develop diabetes?





# Clinical Characterization: Population-level heterogeneity across systems, and patient-level heterogeneity within systems



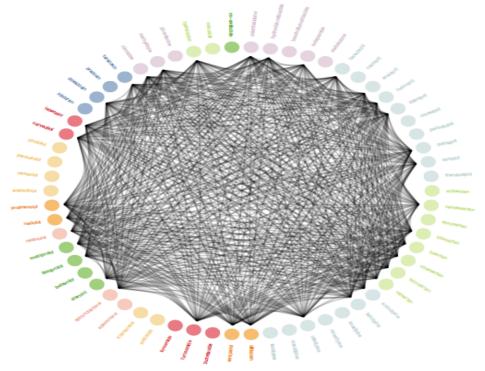


# Population-Level Effect Estimation: LEGEND Study

Head-to-head HTN drug comparisons



- Trials: 40
- *N* = 102 [1148] 33K



- Comparisons: 10,278
- *N* = 3502 [212K] 1.9M

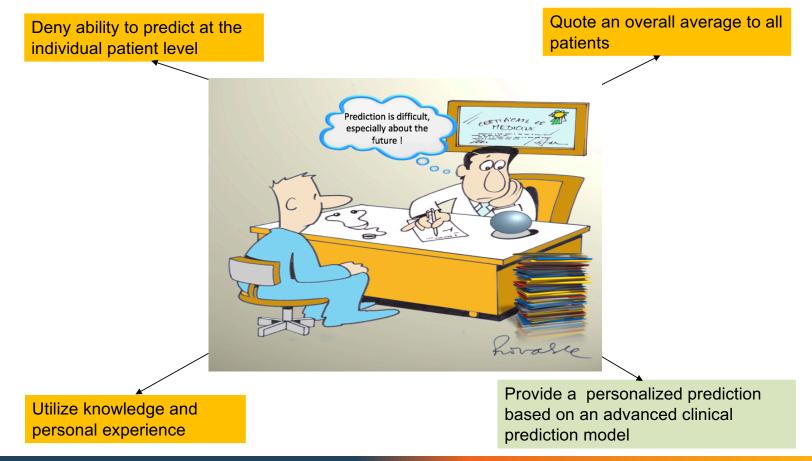


# Journey of Patient-Level Prediction

An example of large-scale analysis enabled by data standardization

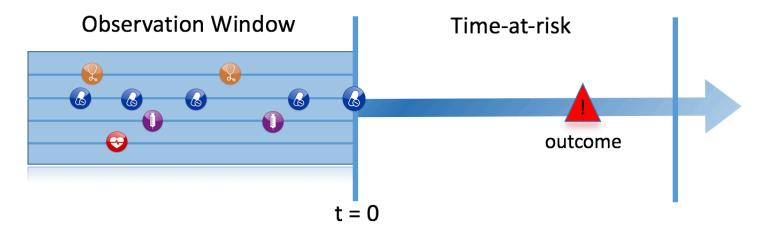


# Clinicians are confronted with prediction questions on a daily basis. What options do they have?





# **Problem definition**



Among a target population (T), we aim to predict which patients at a defined moment in time (t=0) will experience some outcome (O) during a time-at-risk Prediction is done using only information about the patients in an observation window prior to that moment in time.



### What are the key inputs to a patient-level prediction study?

Input parameter	Design choice
Target cohort (T)	
Outcome cohort (O)	
Time-at-risk	
Model specification -which model(s)? -which parameters? -which covariates?	

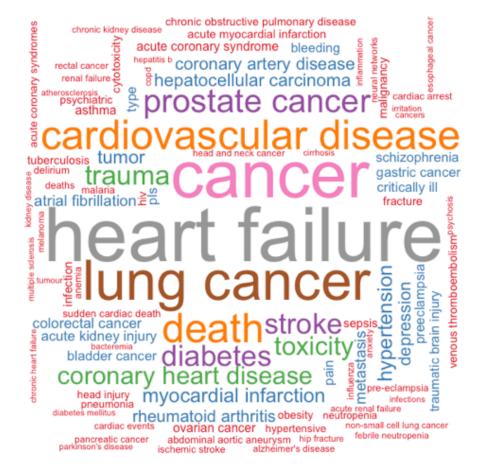


# Types of prediction problems in healthcare

Туре	Structure	Example
Disease onset and progression	Amongst patients who are newly diagnosed with <insert favorite<br="" your="">disease&gt;, which patients will go on to have <another disease="" or="" related<br="">complication&gt; within <time diagnosis="" from="" horizon="">?</time></another></insert>	Among newly diagnosed AFib patients, which will go onto to have ischemic stroke in next 3 years?
Treatment choice	Amongst patients with <indicated disease=""> who are treated with either <treatment 1=""> or <treatment 2="">, which patients were treated with <treatment 1=""> (on day 0)?</treatment></treatment></treatment></indicated>	Among AFib patients who took either warfarin or rivaroxaban, which patients got warfarin? (as defined for propensity score model)
Treatment response	Amongst patients who are new users of <insert chronically-<br="" favorite="" your="">used drug&gt;, which patients will <insert desired="" effect=""> in <time window&gt;?</time </insert></insert>	Which patients with T2DM who start on metformin stay on metformin after 3 years?
Treatment safety	Amongst patients who are new users of <insert drug="" favorite="" your="">, which patients will experience <insert adverse="" event="" favorite="" from<br="" known="" your="">the drug profile&gt; within <time exposure="" following="" horizon="" start="">?</time></insert></insert>	Among new users of warfarin, which patients will have GI bleed in 1 year?
Treatment adherence	Amongst patients who are new users of <insert chronically-<br="" favorite="" your="">used drug&gt;, which patients will achieve <adherence metric="" threshold=""> at <time horizon="">?</time></adherence></insert>	Which patients with T2DM who start on metformin achieve >=80% proportion of days covered at 1 year?



### Growing interest in prediction modelling





#### Current status of predictive modelling

Opportunities and challenges in developing risk prediction models with electronic health records data: a systematic review



Benjamin A Goldstein<sup>1,2</sup>, Ann Marie Navar<sup>2,3</sup>, Michael J Pencina<sup>1,2</sup>, John PA Ioannidis<sup>4,5</sup>

#### ABSTRACT

**Objective** Electronic health records (EHRs) are an increasingly common data source for clinical risk prediction, presenting both unique analytic opportunities and challenges. We sought to evaluate the current state of EHR based risk prediction modeling through a systematic review of clinical prediction studies using EHR data.

**Methods** We searched PubMed for articles that reported on the use of an EHR to develop a risk prediction model from 2009 to 2014. Articles were extracted by two reviewers, and we abstracted information on study design, use of EHR data, model building, and performance from each publication and supplementary documentation.

**Results** We identified 107 articles from 15 different countries. Studies were generally very large (median sample size = 26 100) and utilized a diverse array of predictors. Most used validation techniques (n = 94 of 107) and reported model coefficients for reproducibility (n = 83). However, studies did not fully leverage the breadth of EHR data, as they uncommonly used longitudinal information (n = 37) and employed relatively few predictor variables (median = 27 variables). Less than half of the studies were multicenter (n = 50) and only 26 performed validation across sites. Many studies did not fully address biases of EHR data such as missing data or loss to follow-up. Average c-statistics for different outcomes were: mortality (0.84), clinical prediction (0.83), hospitalization (0.71), and service utilization (0.71).

Conclusions EHR data present both opportunities and challenges for clinical risk prediction. There is room for improvement in designing such studies.

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Goldstein BA, J Am Med Inform Assoc. 2016.



#### Current status of predictive modelling

- Inadequate internal validation
- Small sets of features
- Incomplete dissemination of model and results
- No transportability assessment
- Impact on clinical decision making unknown



Relatively few prediction models are used in clinical practice



# **OHDSI** Mission for Patient-Level Prediction

OHDSI aims to develop a systematic process to learn and evaluate large-scale patient-level prediction models using observational health data in a data network





# **PLP Framework Paper**



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

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

Journal of the American Medical Informatics Association, ocy032, https://doi.org/10.1093/jamia/ocy032 Published: 27 April 2018 Article history ▼

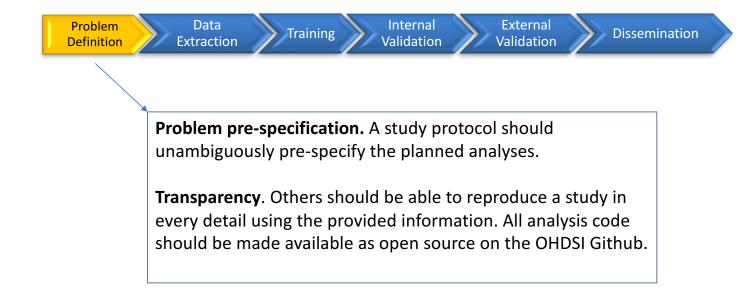
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#### Abstract

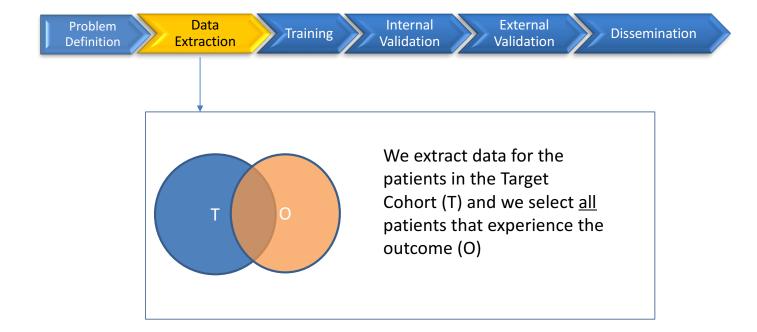
#### Objective

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.

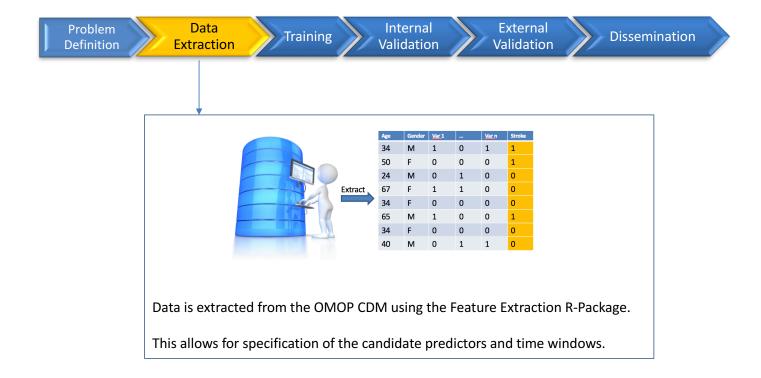




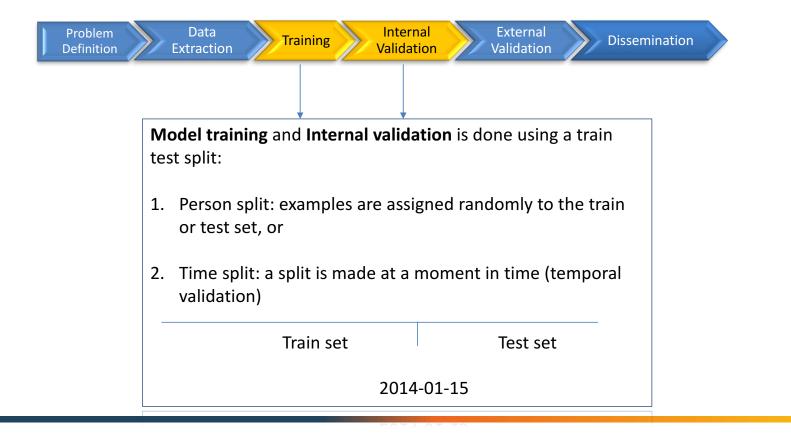






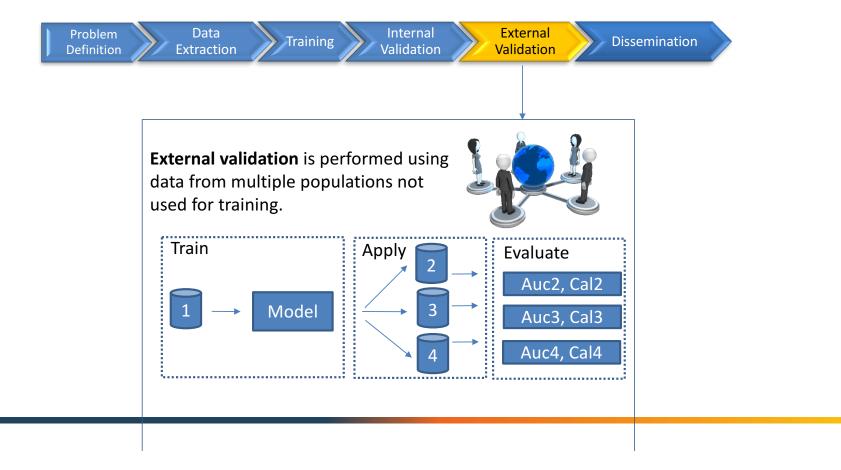






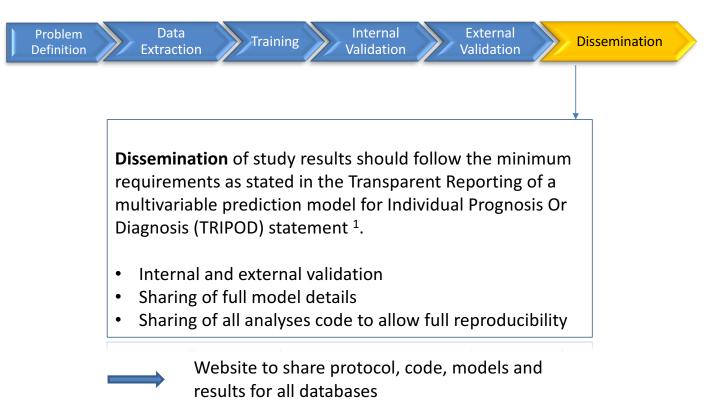


# **External Validation**





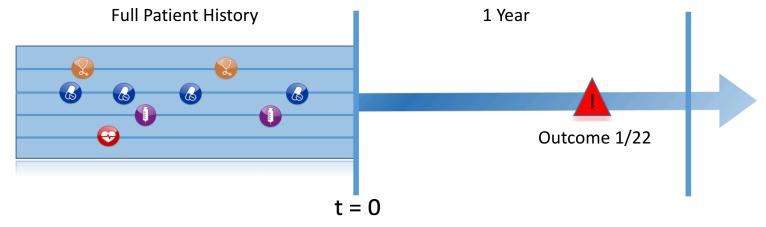
# Dissemination



<sup>1</sup> Moons, KG et al. Ann Intern Med. 2015;162(1):W1-73



### **Example Study**



First Pharmaceutically Treated Depression

Among patients <u>in 4 different databases</u>, we aim to develop prediction models to predict which patients at a defined moment in time (<u>First Pharmaceutically</u> <u>Treated Depression Event</u>) will experience one out of <u>22 different outcomes</u> during a time-at-risk (<u>1 year</u>). Prediction is done using <u>all demographics</u>, <u>conditions</u>, and drug use data prior to that moment in time.



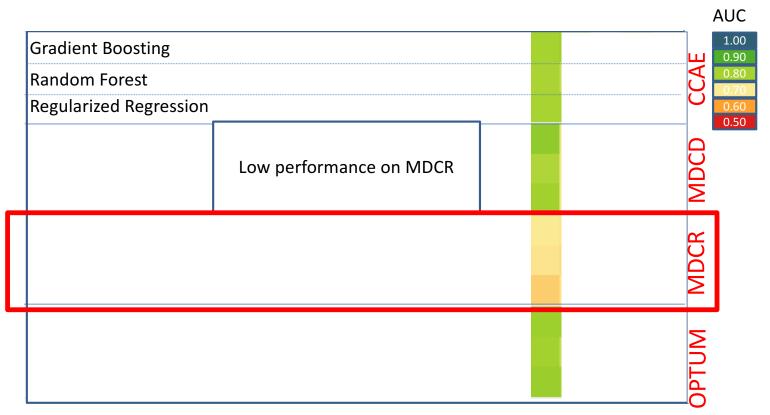
### Model Discrimination Stroke

Acute liver injury	Acute myocardial infanction	Alopecia	Constipation	Decreased libido	Delinium	Diarrhea	Fracture	Gastrointestina hemhorrage	Hyponatremia	H ypotension	Hypothyroidism	Insomnia	Nausea	Open-angle glaucoma	STRC	ЭКЕ	Ventricular anthythmia and sudden cardiac d	Vertigo		AUC
Gra	Gradient Boosting																		ш	1.00 0.90
Random Forest																	CA	0.80		
Regularized Regression																		0.60 0.50		
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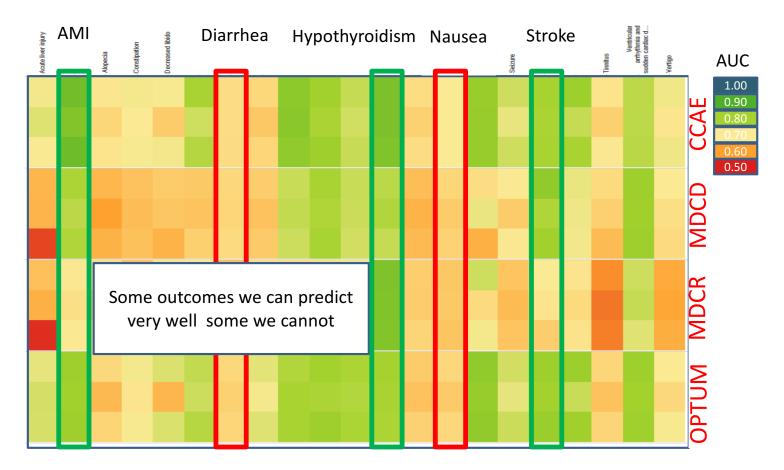
#### **Model Discrimination**

Outcomes





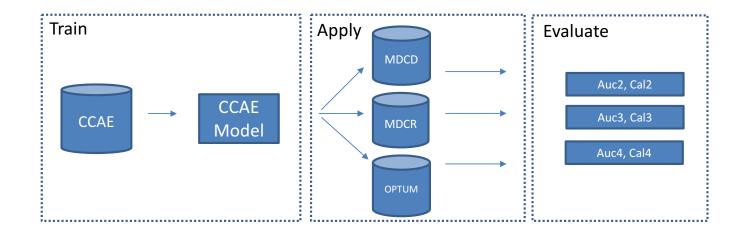
#### **Model Discrimination**





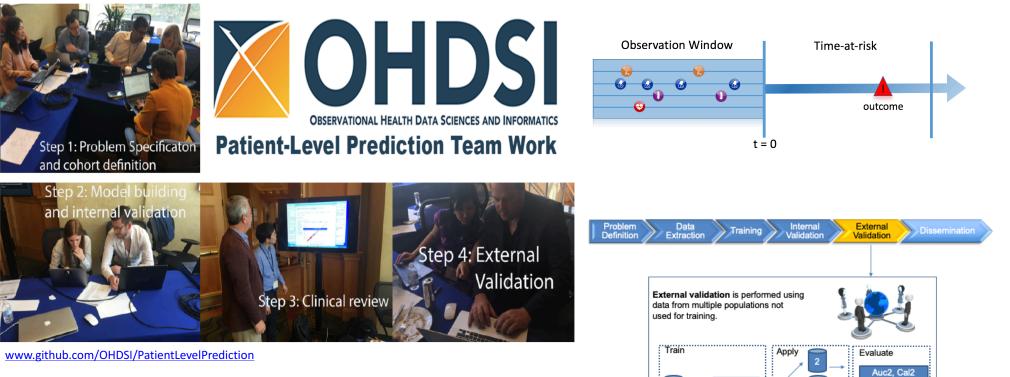
### **Transportability Assessment**

#### How well do the models perform on other databases?





# Patient-Level Prediction Workgroup



Model

Auc3, Cal3

Auc4, Cal4

4

Jenna M Reps, Martijn J Schuemie, Marc A Suchard, Patrick B Ryan, Peter R Rijnbeek; Design and implementation of a standardized framework to generate and evaluate patient-level prediction models using observational healthcare data, Journal of the American Medical Informatics Association, Volume 25, Issue 8, 1 August 2018, Pages 969–975, <a href="https://doi.org/10.1093/jamia/ocy032">https://doi.org/10.1093/jamia/ocy032</a>





### First Annual OHDSI Symposium, March 23th 2018





# Meeting Goals Second OHDSI Symposium



- Provide a platform to stimulate community building
- Demonstrate the OHDSI approach to Reliable and Reproducible Evidence Generation
- Educate and train the community

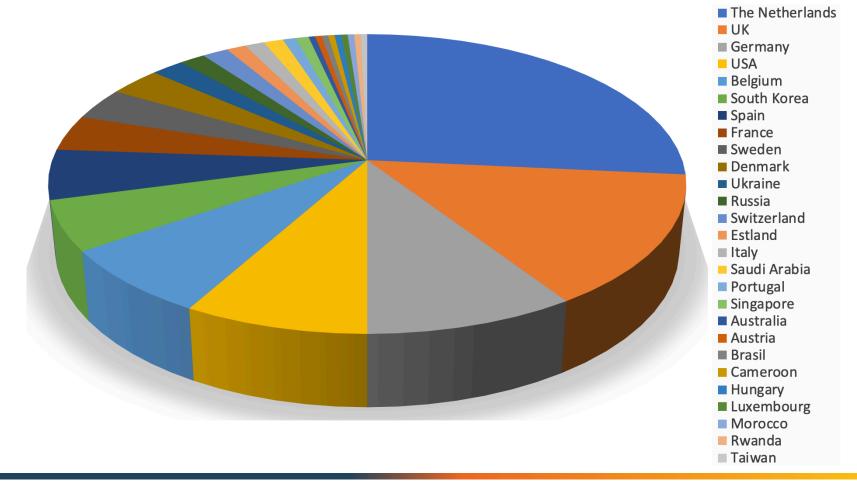


### First Annual OHDSI Symposium, March 23th 2018



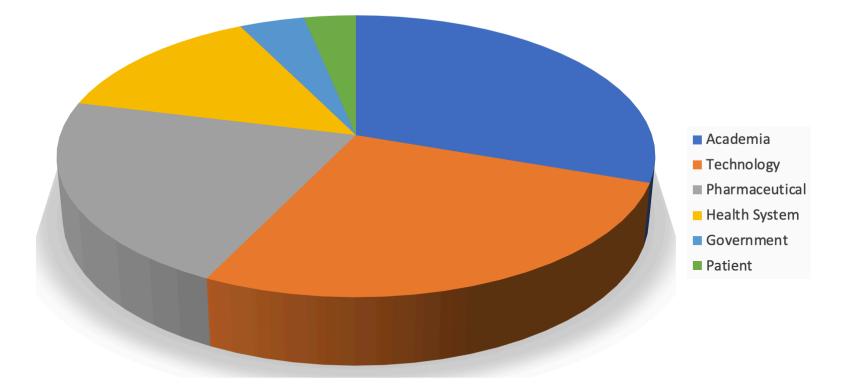


### Breakdown of Participants: Countries



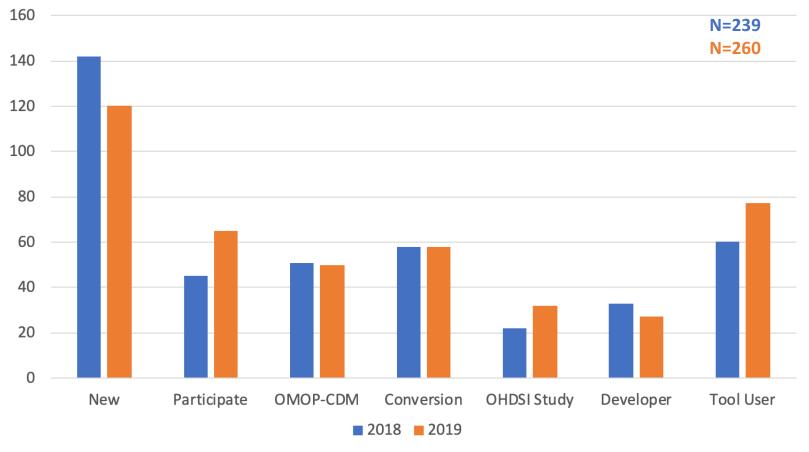


# Breakdown of Participants: Stakeholders





### Relationship with OHDSI





# The Journey From Data to Evidence



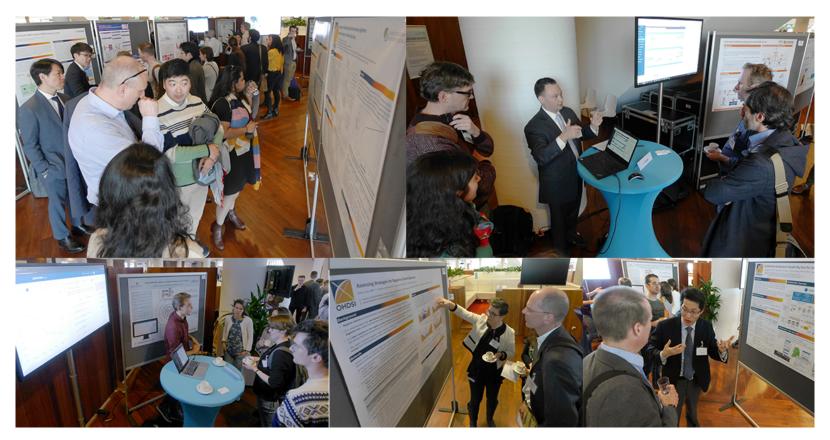


# Lightning Talks





### 35 Posters and 8 Software Demos





### 8 Tutorials with 150 participants

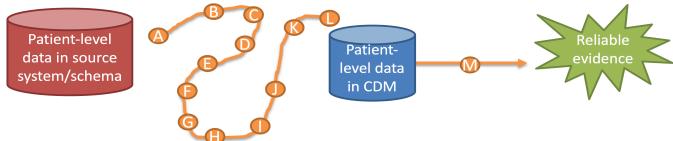


- OMOP-CDM and Vocabularies
- Extract Transform Load
- Cohort Building
- Patient-Level Prediction
- Population-Level Effect Estimation





# We collaboratively need to make the next steps!



- Increase the number of data sources in the open network
- Extend the vocabularies for the European market
- Optimize standardized procedures for quality control
- Further improve and extend the analytical toolset
- Build a sustainable eco-system for mapping support, study execution etc.
- Train all our stakeholders





#### The European Health Data and Evidence Network (EHDEN) Project

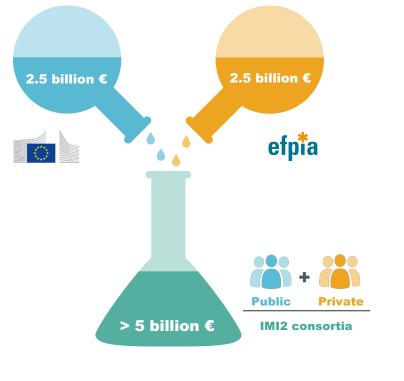
#### Peter Rijnbeek, PhD

Associate Professor Health Data Science Department of Medical Informatics Erasmus MC, The Netherlands



#### INNOVATIVE MEDICINES INITIATIVE

#### www.imi.europa.eu



IMI 1 (2008-2014) 2 bn € budget 59 projects

#### IMI 2 (2014-2024)

3.3 bn € budget More ambitious, more open & greater scope



#### **EHDEN Project Facts**

Start date: 1st Nov 2018 End date: 30th Apr 2024 Total duration: 66 months



**Total budget:** 28,917,357€

IMI2 Funding: 14,105,750€ (7M Harmonization Fund) EFPIA contribution: 14,811,607€ (10M Harmonization Fund)

### **EHDEN** HAD ITS KICK-OFF MEETING IN JANUARY







#### Vision

The European Health Data & Evidence Network (EHDEN) aspires to be the trusted observational research ecosystem to enable better health decisions, outcomes and care

#### Mission

Our mission is to provide a new paradigm for the discovery and analysis of health data in Europe, by building a large-scale, federated network of data sources standardized to a common data model



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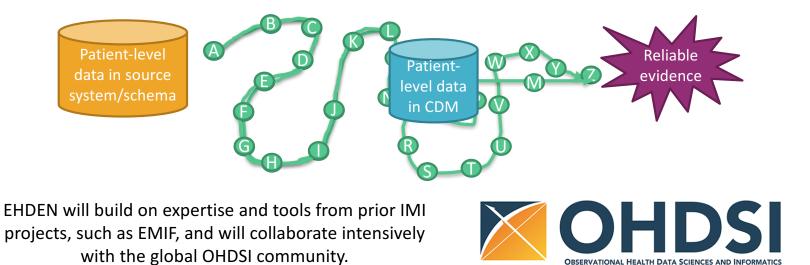






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#### UILDING THE ECO-SYSTEM: HARMONISATION

Ecosystem

The aim is to map 100 million health records across the EU via a common data model (OMOP), supporting research within a federated network, the BD4BO IMI2 programme, and outcomesbased healthcare Outcomes



- Standards development within OHDSI
- Medium-sized Enterprises (SME<sub>p</sub>fti<sup>astructure</sup> the EU Certified & qualified Small to
- Harmonisation fund to support mapping to CDM



### THE EHDEN OPEN CALL PROCESS





### **SME CERTIFICATION PROCEDURE: PILOT CALL**



Certification and training of selected SMEs in all necessary competencies.

efpia

### **SME CERTIFICATION PROCEDURE**

#### Training

EHDEN's aim is that all selected SMEs as companies will obtain all the necessary competencies to work with a data source within the EHDEN federated network following the established quality standards.

Training topics include, but are not limited to:

- The EHDEN project and its objectives
- OMOP-CDM and the Standardized Vocabularies
- ETL steps and their implementation with OHDSI tools and approaches developed in EHDEN.
- Fundamentals of proper documentation of the ETL process to assure transparency and reproducibility.
- Expertise in the installation and use of the OHDSI Tools for federated data analyses

The course material will be made available through our online learning platform, the EHDEN academy (https://academy.ehden.eu).



The final step in the certification program is a face-to-face training and assessment.

Once the SME receives the certificate it obtains the following privileges:

- Member of the EHDEN SME Community
- Listed in the Certified SME Catalogue
- Use of the EHDEN Certification Badge

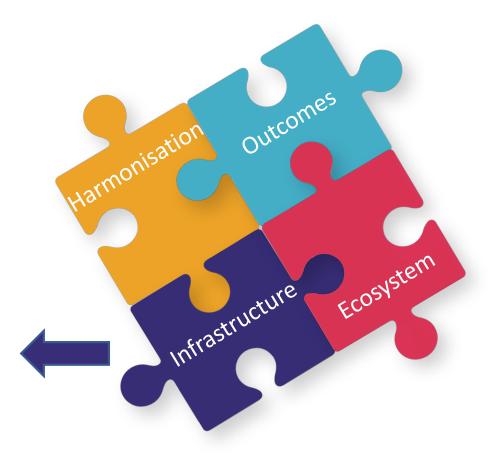
Following certification, SMEs can support and map data sources which have received an EHDEN grant.



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#### **BUILDING THE ECO-SYSTEM: INFRASTRUCTURE**

- Build the infrastructure to enable federated studies on an unprecedented scale in Europe
- Leverage and further develop OHDSI and other tools for high quality analyses
- Use Case driven development
- Ensure interoperability on a global level



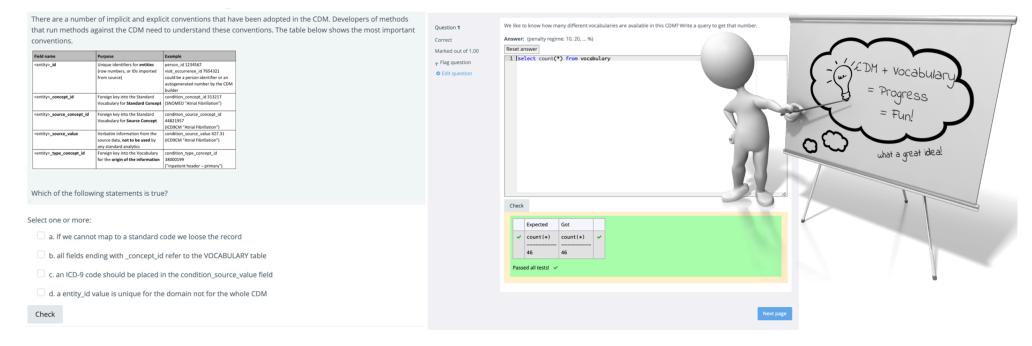
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EHDEN is developing the EHDEN Academy (<u>academy.ehden.eu</u>): an online learning platform that will host courses from OHDSI and EHDEN.

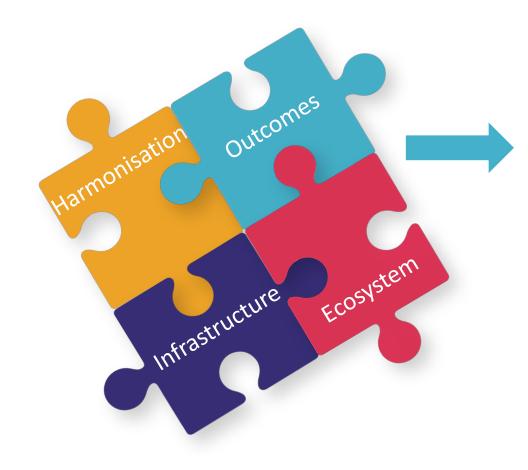
#### ➢ EHDEN Academy

It will contain: Video Lectures, Quizzes, SQL Questions, OHDSI-IN-A-BOX VM integration, and more..





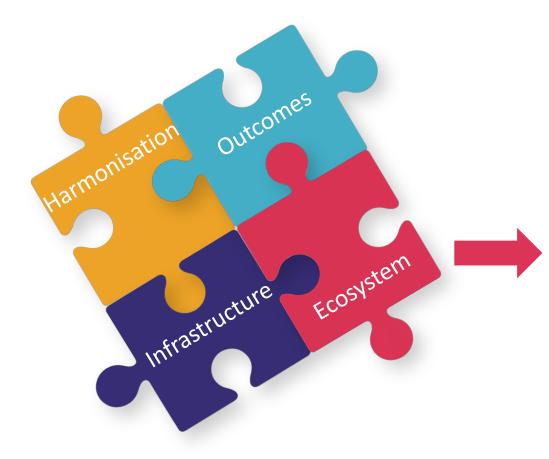




- Incorporation of outcome standards (ICHOM)
- Supporting outcomes-based research and medicine
- Demonstrate the additive value of EHDEN through use cases

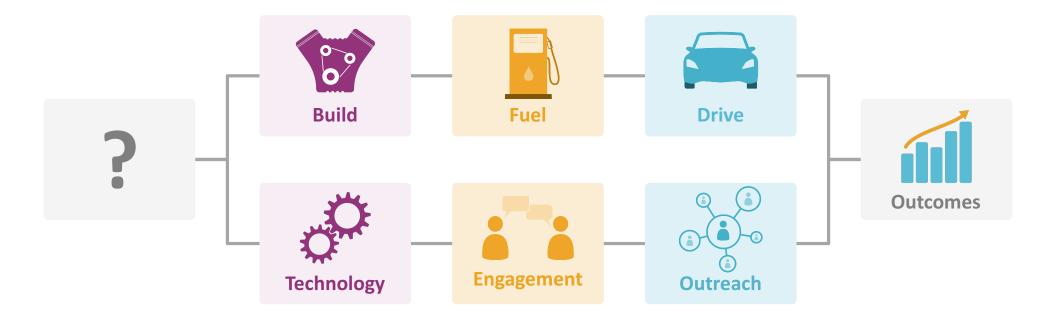
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BUILDING THE ECO-SYSTEM: ECOSYSTEM



- A trusted federated data network expanding beyond EHDEN project lifetime
- European-wide network of certified SMEs
- Enable new and augmented health services during and after EHDEN
- Stimulate active collaboration within the community
- Sustainability is a responsibility of all stakeholders









#### WP1: Evidence Workflow Development

Incorporating the use cases for supporting development and validation of the EHDEN sociotechnical approach, inclusive of BD4BO projects

#### **WP2: Outcome Driven Healthcare**

Related to all activities specific to e.g. BD4BO projects outcome focus, and ICHOM standards incorporation

#### **WP3: Personalized Medicine**

Focusing on the support of outcomes/value based healthcare, inclusive of clinical prediction models, with the incorporation of 'novel' patient data

#### **WP4: Technical Implementation**

Key priority is socio-technical development of the EHDEN federated framework and relevant services

#### WP5: Data Workflow Implementation & Service Deployment

Development, oversight and evaluation of the ecosystem development from SME qualification/ certification to data source engagement, OMOP CDM mapping and evaluation

#### WP6: Outreach and Sustainability

Ensuring the development of value propositions for key stakeholders, and developing the sustainable operational model for EHDEN during and post IMI phase

#### WP7: Project Management and Dissemination

Concentrating on intra-project project management, internal communications and external dissemination, and responding to IMI deliverables







EHDEN and OHDSI will collaboratively work on the implementation of an ecosystem for federated analysis in Europe at an unprecedented scale.

We hope that we can further intensify our collaboration with SNOMED to speak the same language. We are convinced this will empower our research community to generate the reliable evidence our patients need!

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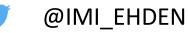
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in

github.com/EHDEN



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