



Published Studies

OHDSI Community Call
Sept. 6, 2022 • 11 am ET



Upcoming OHDSI Community Calls

Date	Topic
Sept. 13	Clinical Registry Efforts In OHDSI
Sept. 20	2022 OHDSI Symposium Preview
Sept. 27	HTA Challenge
Oct. 4	OHDSI Debates
Oct. 11	Final OHDSI2021 Logistics
Oct. 18	Welcome To OHDSI
Oct. 25	Future Directions For OHDSI



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Sept. 13: Clinical Registry Efforts in OHDSI

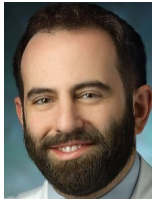
How clinical registries and OHDSI can benefit from each other

Presenter: Paul Nagy • Program Director for Graduate Training in Biomedical Informatics and Data Science, Deputy Director of the Johns Hopkins Medicine Technology Innovation Center



How to adapt a manual clinical registry to OMOP

Presenter: Matt Robinson • Assistant Professor, The Johns Hopkins University School of Medicine



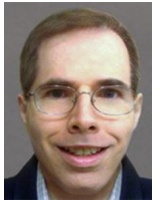
How to lower the ETL barrier going to OMOP with Perseus (Demonstration)

Presenter: Zachary Wang • Graduate Student, Johns Hopkins (2022 Kheiron Cohort member)



Lowering the deployment burden with the cloud

Presenter: Lee Evans • Owner, LTS Computing LLC



How to take harmonization to the next step with unit level harmonization; How to create the governance for robust community developed phenotypes

Presenter: Emily Pfaff • Research Assistant Professor, University of North Carolina at Chapel Hill





Three Stages of The Journey

Where Have We Been?

Where Are We Now?

Where Are We Going?





OHDSI Shoutouts!



Congratulations to the 53 individuals or teams who were nominated for a 2022 Titan Award.

The Titan Award recipients will be announced during the closing talk at the 2022 OHDSI Symposium.

Thamir Alshammary | Juan Banda | Adam Black | Fan Bu | Montse Camprubi | Yong Chen | Marcel de Wilde | Frank DeFalco | Egill Fridgeirsson | Jamie Gilbert | Jake Gillberg | Jason Hsu | Nigel Hughes | Yu-Chuan Jack Li | Mik Kallfelz | Andy Kanter | Elisse Katzman | Chungsoo Kim | Greg Klebanov | Chris Knoll | Kristin Kostka | Manlik Kwong | Christophe Lambert | Martin Lavallee | Jing Li | Xintong Li | Star Liu | Ajit Londhe | Aniek Markus | Evan Minty | Paul Nagy | Karthik Natarajan | Aki Nishimura | Anna Ostropolets | Melanie Philofsky | Gowtham Rao | Berta Raventos | Craig Sachson | Martijn Schuemie | Azza Shoaibi | Marc Suchard | Cynthia Sung | Joel Swerdel | May Terry | Don Torok | Cynthia Yang | Jacob Zelko | Center for Surgical Science Prediction study team | LEGEND-T2DM | N3C | Thrombosis w Thrombocytopenia phenotype project team | Vaccine Evidence Workgroup



OHDSI Data Partners

Who has already joined the journey and adopted the OMOP CDM? There are currently 331 databases, including 284 electronic health records and 28 administrative claims sources, that come from 34 different countries. Together, these databases represent more than 810 million unique patient records, approximately 11% of the world's population.

Jiangxi Province People's Hospital (JPH) (China)	South China Hanyang University Hospital (SCH) (China)
Jinhua Hospital (JH) (China)	STARflood Medicine Research Data Repository (STARflood MCR) (USA)
Kangbuk Samsung Hospital (KSH) (South Korea)	Stony Brook (SHR) (USA)
Kangbuk Sacred Heart Hospital (KSH) (South Korea)	Surveillance, Epidemiology, and End Results Program (SEER) (Claims, Registry, Netherlands)
Kyungju National University Hospital (KNUH) (South Korea)	Surveillance, Epidemiology, and End Results Program (SEER) - G-Cell (SHR) (USA)
Keck Medicine of University of Southern California (KMC) (USA)	Sydney Local Health District (SLHD) (Australia)
Koo Teck Poo Hospital (KTPH) (Singapore)	Tripoli Hospital (TH) (Libya, Tunisia)
Koo Teck Poo Hospital (KTPH) (Singapore)	TCC - Los Angeles (SHR) (USA)

Data Partners Survey 2022

*** Required**

2021 Data Partners Listing

OHDSI Data Partners

The collage consists of four images arranged in a 2x2 grid. The top-left image is a screenshot of Python code from a GitHub repository, showing a class definition for 'User' and a method 'get_user'. The top-right image is a screenshot of JavaScript code from a GitHub repository, showing a function 'get_user' and a 'User' object. The bottom-left image is a screenshot of Java code from a GitHub repository, showing a 'User' class and a 'get_user' method. The bottom-right image is a group photo of the team members, with a caption 'Team Photo' below it.

Name of Data Partner *

Your answer

Database Name (if different from organization name)

Your answer

Country *

Your answer

bit.ly/OHDSI-Data-Survey

Type of Data *

- ☐ EHR
- ☐ Claims
- ☐ Registry
- ☐ Hospital Billing
- ☐ Clinical Data Warehouse
- ☐ Other: _____

Number of Unique Patients ★

Your answer

What Database Platform Do You Use? *

Oracle

- ☐ Oracle
- ☐ PostgreSQL
- ☐ Amazon RedShift
- ☐ Impala
- ☐ IBM Netezza
- ☐ Google BigQuery
- ☐ Microsoft PDW
- ☐ Snowflake
- ☐ Azure Synapse
- ☐ Apache Spark
- ☐ SQLite
- ☐ Other: _____

Contact Name ★

Your answer

Contact Email

Your answer

Submit

[Clear form](#)

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#JoinTheJourney





OHDSI Shoutouts!



Any shoutouts from the community? Please share and help promote and celebrate OHDSI work!

Have a study published? Please send to sachson@ohdsi.org so we can share during this call and on our social channels.

Let's work together to promote the collaborative work happening in OHDSI!





Three Stages of The Journey

Where Have We Been?

Where Are We Now?

Where Are We Going?





Upcoming Workgroup Calls



Date	Time (ET)	Meeting
Wednesday	2 am	Patient-Level Prediction/Population-Level Estimation
Wednesday	7 am	Medical Imaging
Wednesday	8 am	Psychiatry
Wednesday	9 am	ATLAS
Wednesday	11 am	Open-Source Community
Wednesday	12 pm	Health Equity
Wednesday	12 pm	FHIR and OMOP Terminologies Subgroup (ZOOM)
Wednesday	4 pm	FHIR and OMOP Data Model Harmonization Subgroup (ZOOM)
Thursday	10 am	Data Quality Dashboard Development
Thursday	12 pm	FHIR and OMOP Oncology Subgroup
Friday	9 am	Education
Friday	9 am	GIS – Geographic Information System General
Friday	9 am	Phenotype Development and Evaluation
Friday	10 pm	China Chapter
Monday	10 am	Healthcare Systems Interest Group
Monday	11 am	Early-Stage Researchers
Tuesday	9 am	OMOP CDM Oncology Genomic Subgroup



CBER BEST Seminar Series Returns Tomorrow!



Webinar Registration



Matthew Fox, DSc, MPH

Professor, Departments of Epidemiology and Global Health @Boston University

Matthew Fox, DSc, MPH, is a Professor in the Departments of Epidemiology and Global Health at Boston University. Dr. Fox joined Boston University in 2001. His research interests include treatment outcomes in HIV-treatment programs, infectious disease epidemiology (with specific interests in HIV and pneumonia), and epidemiologic methods. Dr. Fox works on ways to improve retention in HIV-care programs in South Africa from the time of testing HIV-positive through long-term treatment. As part of this work, he is involved in analyses to assess the impact of changes in South Africa's National Treatment Guidelines for HIV. Dr. Fox also does research on quantitative bias analysis and co-authored a book on these methods, *Applying Quantitative Bias Analysis to Epidemiologic Data*. He is also the host of a public health journal club podcast called *Free Associations* designed to help people stay current in the public health literature and think critically about the quality of research studies.

Topic: Quantitative Bias Analysis Methods to Improve Inferences

Speaker: Matthew P. Fox, DSc, MPH
Professor, BUSPH

Description: Observational epidemiologic research around vaccine efficacy and safety can provide important insights into causal relationships, but key sources of bias often impair the inferences we draw from these studies. Uncontrolled confounding, selection bias and information bias are common in epidemiologic research and failure to account for their impacts in a quantitative manner (rather than qualitative assessments in discussion sections of manuscripts after conclusions have been drawn) can lead to poor inferences. This talk will give an overview of quantitative bias analysis methods to demonstrate how they can be implemented on both summary and record level data to account for the impact of systematic error on study results and provide some examples using data from the literature on how to apply these methods to vaccines safety data.

OHDSI Newsletter Is Available



Publications

Vorisek CN, Lehne M, Klopfenstein SAI, Mayer PJ, Bartschke A, Haese T, Thun S. [Fast Healthcare Interoperability Resources \(FHIR\) for Interoperability in Health Research: Systematic Review](#). JMIR Med Inform. 2022;10(7):e35724. Epub 20220719. doi: 10.2196/35724. PubMed PMID: 35852842; PubMed Central PMCID: PMCPCMC9346559.

Kim S, Bang JI, Boo D, Kim B, Choi IY, Ko S, Yoo IR, Kim K, Kim J, Joo Y, Ryoo HG, Paeng JC, Park JM, Jang W, Kim B, Chung Y, Yang D, Yoo S, Lee HY. [Second primary malignancy risk in thyroid cancer and matched patients with and without radioiodine therapy analysis from the observational health data sciences and informatics](#). Eur J Nucl Med Mol Imaging. 2022;49(10):3547-56. Epub 20220401. doi: 10.1007/s00259-022-05779-9. PubMed PMID: 35362796.

Lin V, Tsuchnik A, Allakhverdiev E, Rosen AW, Gögenur M, Clausen JSR, Bräuner KB, Walbech JS, Rijnbeek P, Drakos I, Gögenur I. [Training prediction models for individual risk assessment of postoperative complications after surgery for colorectal cancer](#). Tech Coloproctol. 2022;26(8):665-75. Epub 20220520. doi: 10.1007/s10151-022-02624-x. PubMed PMID: 35593971.

Bräuner KB, Rosen AW, Tsuchnik A, Walbech JS, Gögenur M, Lin VA, Clausen JSR, Gögenur I. [Developing prediction models for short-term mortality after surgery for colorectal cancer using a Danish national quality assurance database](#). Int J Colorectal Dis. 2022;37(8):1835-43. Epub 20220718. doi: 10.1007/s00384-022-04207-6. PubMed PMID: 35849195.

Lamer A, Moussa MD, Marcilly R, Logier R, Vallet B, Tavernier B. [Development and usage of an anesthesia data warehouse: lessons learnt from a 10-year project](#). J Clin Monit Comput. 2022. Epub 20220806. doi: 10.1007/s10877-022-00898-y. PubMed PMID: 35933465.

Shah SC, Canakis A, Halvorson AE, Dorn C, Wilson O, Denton J, Hauger R, Hunt C, Suzuki A, Matheny ME, Siew E, Hung A, Greevy RA, Jr., Roumie CL. [Associations Between Gastrointestinal Symptoms and COVID-19 Severity Outcomes, Based on a Propensity Score-Weighted Analysis of a Nationwide Cohort](#). Gastro Hep Adv. 2022. Epub 20220807. doi: 10.1016/j.gastha.2022.06.015. PubMed PMID: 35966642; PubMed Central PMCID: PMCPCMC9357443.

September Update Podcast



Community Updates

Where Have We Been?

- Paul Nagy led an informative session on what it takes to build organizational support for adopting the OMOP CDM and OHDSI tools, as well as building organizational capacity for conducting observational research. Video from this panel, which included Keran Moll, Greg Klebanov and Ajit Londhe, [is available here](#).
- The third workshop leadership summit of 2022 was held in August and focused on building connections between workgroups and maximizing time together at the OHDSI symposium. [There are several activities planned](#) during the symposium weekend to strengthen bonds both within and between the workgroups.
- The first session of the Early-Stage Researchers Speaker Series, Asieh Golozar discussed her career path, how OHDSI has influenced her journey, and shares some tips for moving forward in health data sciences. [You can watch it here](#), and be on the lookout for similar sessions in the future.

Where Are We Now?

- The [agenda for the OHDSI 2022 Symposium](#) was recently announced. This document lists the full agenda for the main conference, including all planned presentations, as well as the lightning talks and software demos for the Collaborator Showcase. A later version will include the 100+ posters that will also be presented at the showcase. If you haven't already registered for any of the events, [please visit our symposium homepage](#) and assure your spot at our biggest event of the year.

The Journey Newsletter (September 2022)

The #OHDSI2022 Symposium is less than seven weeks away, and the weekend agenda is available in this newsletter. There will be presentations, the collaborator showcase, a full-day tutorial, workgroup activities and plenty more. You can also learn more about our new data partner survey, a new collaborator spotlight, 13 recent publications, several presentations from the past month and plenty more in the latest edition of The Journey! [#JoinTheJourney](#)

Objective Diagnostics Plenary, OHDSI Support for Regulatory Agencies Highlight Agenda For 2022 Symposium

OHDSI 2022 Symposium Oct. 14-16, 2022 Baltimore, North Howard Street & Conference Center	
Main Conference Agenda - Oct. 14	
7:30 am - 8:30 am Baltimore DE	Registration and Welcome
8:30 am - 10:30 am Baltimore DE	State of the Community George Hens, Columbia University + presentation: at 9:30, Jeff Tuckman
10:30 am - 11:30 am Baltimore DE	Workgroup and Chapter presentations + presentation: at 10:30, Jeff Tuckman
11:30 am - 12:30 pm Baltimore DE	Plenary: Objective Diagnostics: A pathway to provably reliable evidence Mark Schumacher, JPMorgan & Chase
12:30 pm - 1:30 pm Baltimore DE	Breakfast
1:30 pm - 2:30 pm Baltimore DE	Presentations: OHDSI support for regulatory agencies + presentation: at 1:30, Jeff Tuckman
2:30 pm - 3:30 pm Baltimore DE	Collaborator Showcase: Round 1 + presentation: at 2:30, Jeff Tuckman
3:30 pm - 4:30 pm Baltimore DE	Collaborator Showcase: Lightning Talks + presentation: at 3:30, Jeff Tuckman

OHDSI 2022 Symposium Oct. 15-16, 2022 Baltimore, North Howard Street & Conference Center	
Main Conference Agenda - Oct. 15	
7:30 am - 8:30 am Baltimore DE	Registration and Welcome
8:30 am - 10:30 am Baltimore DE	State of the Community George Hens, Columbia University + presentation: at 9:30, Jeff Tuckman
10:30 am - 11:30 am Baltimore DE	Workgroup and Chapter presentations + presentation: at 10:30, Jeff Tuckman
11:30 am - 12:30 pm Baltimore DE	Plenary: Objective Diagnostics: A pathway to provably reliable evidence Mark Schumacher, JPMorgan & Chase
12:30 pm - 1:30 pm Baltimore DE	Breakfast
1:30 pm - 2:30 pm Baltimore DE	Presentations: OHDSI support for regulatory agencies + presentation: at 1:30, Jeff Tuckman
2:30 pm - 3:30 pm Baltimore DE	Collaborator Showcase: Round 2 + presentation: at 2:30, Jeff Tuckman
3:30 pm - 4:30 pm Baltimore DE	Collaborator Showcase: Lightning Talks + presentation: at 3:30, Jeff Tuckman

Simply returning together for the first time since 2019 would be enough reason to look forward to the 2022 OHDSI Symposium, but there will also be several impactful presentations, the largest collaborator showcase in community history and plenty more activities that will make the Oct. 14-16 weekend a special one.



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OHDSI Newsletter Is Available





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Welcome to OHDSI!

The Observational Health Data Sciences and Informatics (or OHDSI, pronounced "Odyssey") program is a multi-stakeholder, interdisciplinary collaborative to bring out the value of health data through large-scale analytics. All our solutions are open-source.

2022 OHDSI Symposium

The 2022 OHDSI Symposium will be held Oct. 14-16 at the [venue] hotel & [venue] is opened.

We will hold [event] day, Oct. 14, which will [event] showcase.

On Saturday [event] day

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#OHDSI2022 Agenda



OHDSI 2022 Symposium
Oct. 14-16, 2022
Bethesda North Marriott Hotel &
Conference Center

Main Conference Agenda • Oct. 14

7:30 am - 8:30 am Ballroom AE Foyer	Registration and Lite Breakfast
9:00 am - 10:00 am Ballroom DE	State of the Community George Hripcsak, Columbia University • presentation of 2020, 2021 Titan Awards
10:00 am - 10:45 am Ballroom DE	Workgroup and Chapter connections • workgroup/chapter leads will be distributed across the venue and available for networking to share activities and progress and connect for future collaborations OHDSI Speed Dating
10:45 am - 12:15 pm Ballroom DE	Plenary: Objective Diagnostics: A pathway to provably reliable evidence Martijn Schuemie, Johnson & Johnson
12:15 pm - 1:00 pm Ballroom Foyer	Buffet Lunch • buffet in exhibitor space
1:00 pm - 2:00 pm Ballroom DE	Presentations: OHDSI support for regulatory authorities moderator: Jody-Ann McLeggon, Columbia University • "US FDA/CBER: Performance of vaccine safety surveillance methods" Fan Bu, UCLA • "Korea Ministry of Food and Drug Safety: Replication of clinical trials in electronic health records" Seng Chan You, Yonsei University • "European Medicines Agency: DARWIN-EU" Peter Rijnbeek, Erasmus MC
2:00 pm - 3:00 pm Ballroom ABC	Collaborator Showcase, Round 1 • Poster presentations with poster walks • Software demonstrations • Exhibitors
3:00 pm - 4:00 pm Ballroom DE	Collaborator Showcase Lightning Talks moderator: Kristin Kostka, Roux Institute at Northeastern University • "Disambiguation of ICPC codes using free-text and active learning to improve concept mappings" Tom Seinen, Erasmus MC • "OHDSI Phenotype Phebruary: lessons learned" Azza Shoaibi, Johnson & Johnson



OHDSI 2022 Symposium
Oct. 14-16, 2022
Bethesda North Marriott Hotel &
Conference Center

Main Conference Agenda • Oct. 14

3:00 pm - 4:00 pm Ballroom DE (continued)	<ul style="list-style-type: none"> • "Reduce, Reuse, & Recycle: Going Green with Atlas Reusables" Ajit Londhe, Amgen • "Best practices for prognostic model development using observational health data: a scoping review" Cynthia Yang, Erasmus MC • "Machine Learning for Predicting Patients at Risk of Prolonged Opioid Use Following Surgery" Behzad Naderalvajoud, Stanford University • "When does statistical equality meet health equity: developing analytical pipelines to compare associational and causal fairness in their application to EHR data" Linying Zhang, Columbia University • "Analyzing the Effect of Hypertension on Retinal Thickness Using Radiology Common Data Model (R-CDM)" Chul Hyoun Park, Ajou University • "Multinational Patterns of Second-line Anti-hyperglycemic Drug Initiation: A LEGEND-T2DM Study" Lovedeep Dhingra, Yale University
4:00 pm - 5:00 pm Ballroom ABC	Collaborator Showcase, Round 2 <ul style="list-style-type: none"> • Poster presentations with poster walks • Software demonstrations • Exhibitors
5:00 pm - 6:00 pm Ballroom DE	Closing Talk: Building A Healthier World Together Patrick Ryan, Johnson & Johnson, Columbia University <ul style="list-style-type: none"> • 2022 Titan Awards • Group photo at conclusion
6:00 pm - 7:00 pm Ballroom ABC	Networking Reception



Register Here:
ohdsi.org/ohdsi2022symposium/

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bit.ly/OHDSI2022-Agenda



#OHDSI2022 Agenda



OHDSI 2022 Symposium
Oct. 14-16, 2022
Bethesda North Marriott Hotel &
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Full-Day Tutorial • Oct. 15

An Introductory Journey From Data To Evidence

In this tutorial, we will introduce participants to steps along the journey from data to evidence using the OMOP Common Data Model, OHDSI tools and scientific best practices. In each 50-minute segment, the class will learn the conceptual framing of the problem and approach to the solution. Then, the class will have the opportunity to have hands-on exposure to design and implementation of analyses and interpretation of results. The course will be motivated by a real use case: using observational data to generate evidence about the relationship between an exposure and outcome, and will highlight how the suite of OHDSI tools and practices can enable such learning.

This class is designed for newcomers to the OHDSI community who are looking for a high-level summary across a wide range of topics covered within the OHDSI community. It's also designed for those in the OHDSI community who may be focused in one particular area of the journey who want exposure to the other areas, so they can better understand how their work contributes to be 'big picture,' and advances the mission to improve health by empowering a community to collaboratively generate the evidence that promotes better health decisions and better care

The tutorial will be held in White Oak A.

Time	Title	Faculty
7:30 am - 8:30 am	Registration/Lite Breakfast (White Oak Foyer)	
8:30 am - 9:00 am	Overview of the OHDSI Journey: where are we going?	Patrick Ryan
9:00 am - 9:50 am	OMOP Common Data Model and vocabulary	Clair Blacketer
9:50 am - 10:00 am	Energy Break	
10:00 am - 10:50 am	ETL a source database into OMOP CDM	Melanie Philofsky
10:50 am - 11:00 am	Energy Break	
11:00 am - 11:50 am	Creating Cohort Definitions	Asieh Golozar
11:50 am - 12:30 pm	Buffet Lunch	
12:30 pm - 1:20 pm	Phenotype Evaluation	Gowtham Rao
1:20 pm - 1:30 pm	Energy Break	
1:30 pm - 2:20 pm	Characterization	Kristin Kostka
2:20 pm - 2:30 pm	Energy Break	
2:30 pm - 3:20 pm	Estimation	Martijn Schuemie
3:20 pm - 3:30 pm	Energy Break	
3:30 pm - 4:20 pm	Prediction	Jenna Reps
4:20 pm - 5:00 pm	Recap of the OHDSI Journey: Where do we go from here?	George Hripcsak



OHDSI 2022 Symposium
Oct. 14-16, 2022
Bethesda North Marriott Hotel &
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Collaborator Showcase

Poster Presentations

The 2022 OHDSI Symposium will host two sessions featuring a total of over 100 posters that highlight the breadth of global research happening within our community. Closer to the symposium weekend, we will announce all of the posters and presenters.

Software Demos

The 2022 OHDSI Symposium will feature 17 software demonstrations during the Collaborator Showcase sessions, listed below:

A demonstration of the EnsemblePatientLevelPrediction package (Jenna M. Reps, Jenna Wong, and Ross Williams)
CohortIncidence: A Software Demonstration (Christopher Knoll)
Criteria2Query 2.0: Combining Human and Machine Intelligence for Cohort Identification (Yilu Fang, Betina Idnay, Yingcheng Sun, Hao Liu, Zhehuan Chen, Karen Marder, Hua Xu, Rebecca Schnall, Chunhua Weng)
Data Network Feasibility Tool - Software Demonstration (Frank DeFalco, Clair Blacketer)
Data Quality Dashboard v2.0 (Clair Blacketer, Frank DeFalco, Anthony Molinaro, Dmitry Ilyin, Luis Alaniz, Maxim Moinat)
Einstein-ATLAS: Leveraging OHDSI/ATLAS and Open-Source Development to Support Translational Research, Data Science, and Regulatory Compliance (Parsa Mirhaji, Selvin Soby, Erin Henninger, Chandra Nelapattai, Manuel Wahle, Boudewijn Aasman, Eran Belin)
ohdsitargets - An R package for building OHDSI study pipelines using targets (Adam Black, Martin Lavallee, Asieh Golozar, Gregory Klebanov)
OmopPopEpi: An R package to compute population-level incidence and prevalence using the OMOP common data model (Marta Catala, Berta Raventas, Mike Du, Yuchen Guo, Xintong Li, Ross Williams, Talita Duarte Sales, Daniel Prieto Alhambra, Edward Burn)
PHOEBE 2.0: selecting the right concept sets for the right patients using lexical, semantic, and data-driven recommendations (Anna Ostropolets, George Hripcsak, Patrick Ryan)
Real World Assessment and Research of Drugs (REWARD): presenting an open-source package for Population-level effect estimation at the scale of all outcomes by all exposures (James Gilbert)
Simple and practical EMR to OMOP CDM ETL tool (Pieter-Jan Lammertyn, Stijn Dupulthys, Louise Berteloot, Peter De Jaeger, Kim Denturck, Nathalie Mertens)
Standardizing Knowledge of Drug Effects: An Application of PheKnowLator for Drug Safety (Tiffany J. Callahan, Patrick B. Ryan, George Hripcsak)
Strategus: Marching towards transparent, reproducible research (Anthony G. Sena, Christopher Knoll, James Gilbert, Jenna Reps, Frank DeFalco, Clair Blacketer, Anthony Molinaro, Joshua Ide, Patrick Ryan, Martijn Schuemie)
The OHDSI Community Dashboard: Tracking the Health and Impact of the Open Science Observational Health Data Sciences and Informatics Community (Star Liu, Asieh Golozar, Jody-Ann McLeggon, Adam Black, Paul Nagy)
Understanding circe-be logic through Capr for generating complex cohort definitions (Martin Lavallee, Adam Black and Asieh Golozar)
Using dbt - a free and open-source software - to transform data into OMOP CDM in the ETL process (Thanapat Pitchayarat, Gun Pinyo, Watcharaporn Tanchotsrinon, Somkid Khamsimuang, Chalita Issarasitiphip, Chaiyanun Bootnumpech, Noppon Siangchin, Kanphitcha Promma, Nattachai Bovormmongkolsak, Prapat Suriyaphol, Natthawut Adulyanukosol)
Vocabulary Versioning: Tracking Concepts over Time Software Demonstration (Tom Seinen, Peter Rijnbeek)

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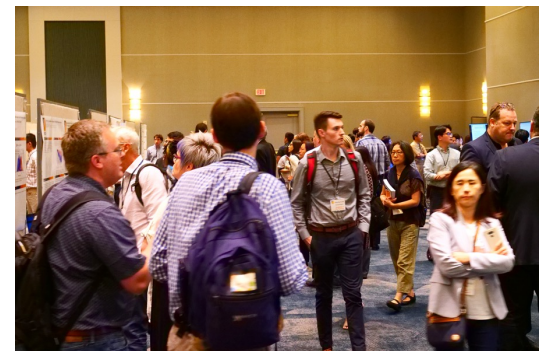


2022 OHDSI Symposium

Registration is OPEN for
#OHDSI2022!

The 2022 OHDSI Symposium
will be held Oct. 14-16 at the
Bethesda North Marriott Hotel
& Conference Center.

www.ohdsi.org/ohdsi2022symposium





2022 OHDSI Symposium

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2022 OHDSI Symposium

Oct. 14-16 • Bethesda North Marriott Hotel & Conference Center



We are thrilled to announce that registration for the 2022 OHDSI Symposium, which will be held Oct. 14-16 at the Bethesda North Marriott Hotel & Conference Center, is now open!

It is so exciting to bring our community back together this fall. [Our collaborator showcase will return](#); please click the link to see how you can take part in our poster presentations, software demos and lightning talks. The full agenda for our conference is still being developed, so please continue to check the OHDSI website (www.ohdsi.org) and our social platforms for updates as we plan for the 2022 Symposium.

The main conference will be held Friday, Oct. 14. A full-day tutorial will be held Saturday, Oct. 15, while other community activities will be held both Oct. 15 and Oct. 16.

Symposium Registration Details

Friday, Oct. 14 – Main Conference

Registration Fee: \$500*

** this is an open and inclusive event; if the registration fee represents a burden to you, please contact symposium@ohdsi.org.*

Should you need to make changes or cancel your registration ticket, please follow the instructions you will receive on your Eventbrite confirmation upon registration completion. Please note that tickets can be refunded up until 7 days prior to the event; Eventbrite fees are not refundable.

[Register For The Main Conference • Friday, Oct. 14](#)

Saturday, Oct. 15 – Full-Day Tutorial: An Introductory Journey From Data To Evidence

Registration Fee: \$300*

** this is an open and inclusive event; if the registration fee represents a burden to you, please contact symposium@ohdsi.org.*

Should you need to make changes or cancel your registration ticket, please follow the instructions you will receive on your Eventbrite confirmation upon registration completion. Please note that tickets can be refunded up until 7 days prior to the event; Eventbrite fees are not refundable.

[Register For The Full-Day Tutorial • Saturday, Oct. 15](#)

[What Will Be Taught At This Tutorial?](#)

Saturday, Oct. 15 and Sunday, Oct. 16 – Community Activities

A highlight of the OHDSI Symposium will be a full weekend of workgroup activities and meetings within the Bethesda North Marriott Hotel & Conference Center. You are now able to [register for any workgroup sessions as long as there is no overlap between any two sessions](#); registration is free, but please do so early as this will be first-come, first-served due to room capacity.

[See The Schedule & Agenda For Workgroup Activities • Weekend of Oct. 15-16](#)

[Register For Workgroup Activities • Weekend of Oct. 15-16](#)

Hotel Information and Sleeping Room Block

Hotel: [Bethesda North Marriott Hotel & Conference Center](#)

Address: 5701 Marinelli Road, Rockville, Maryland, 20852

Hotel Main Number: (301) 822-9200

Reservations Toll Free: (877) 212-5752


Reservations Local Phone: (301) 822-9200

This year, OHDSI is holding a sleeping room block for the nights of Oct. 13 and 14 with a special room rate of \$179 plus taxes. Please note that all sleeping rooms are on a first-come, first-served basis. To help us in the planning process, we ask that you do not cancel your hotel room ordered through the OHDSI Room Block. If you must cancel, please let us know prior to Thursday, Sept. 1, so that we can offer the room to others who may need one. Once the room block is full, or if specific nights are sold out, you may make additional room reservations [on the hotel's website](#) or by calling the hotel phone number above. Please note that OHDSI is not holding any sleeping rooms on Saturday, Oct. 15. Therefore, please call the hotel phone number or make this reservation online should you plan to stay Saturday night.

ohdsi.org/ohdsi2022symposium



#OHDSISocialShowcase This Week



EUROPEAN HEALTH DATA & EVIDENCE NETWORK

INTRODUCTION

Dementia is an umbrella term to describe various illnesses that affect cognition and may lead to mental degradation. Early diagnosis of individuals at high risk of dementia allows for improved care and risk-factor targeted intervention. In recent years models have increasingly been developed on observational health data. These routinely collected data from administrative claims and electronic health records are considered to enhance a model's applicability at the point of care.

However, the systematic reviews of Hou et al. and Goerdtzen et al. conclude that although many dementia risk prediction models have been developed, only a handful of them have been externally validated [1, 2]. External validation assesses a model's reliability for clinical use in external data sources that have not been used for model development. A lack of external validation can lead to a plethora of proposed models with little evidence about which are reliable and under what circumstances.

In this study, we aim to externally validate existing dementia prediction models. To that end, we define replicability criteria, review published models, and externally validate three selected models using routinely collected health data from administrative claims and electronic health records.

MATERIALS AND METHODS

The replicability criteria that a study must report are presented in the following table and were directly inferred from the prediction approach in OHDSI, where among a population at risk, we predict which patients at a defined moment in time (the index) will experience some outcome during a time-at-risk.

Category	Replicability criteria	Description
Population settings	Target population definition	Definition or description of the population for which predictions are made.
	Index date	Date at which a patient qualifies for inclusion in the target population.
	Time-at-risk	Time window in which a model's predictions are valid relative to the index date.
Statistical analysis settings	Outcome definition	Definition or description of the outcome to be predicted during the time-at-risk.
	Prediction method	Prediction methods in this study are limited to logistic regression and Cox proportional hazard for predicting a binary outcome.
	Predictor definitions	Predictor descriptions or definitions in terms of data source codes.
	Predictor time window	Time window in which the predictor is assessed.
Model specifications		The prediction model, e.g., parameters to construct the model given a prediction method.
		We also distinguish here between fully and partially specified models.

Included dementia prediction studies were reviewed for these criteria to obtain the current state of reporting in the literature. Moreover, we selected three well reported models for replication and external validation in a network of observational databases, with the aim to investigate factors beyond our criteria that may impact successful external validation. These three models will in the remainder of this poster be referred to based on their first author names: Walters, Mehta, and Nori, respectively [3-5].

EXTERNAL VALIDATION OF EXISTING DEMENTIA PREDICTION MODELS

HENRIK JOHN¹, JAN KORS¹, EIGILL FRIDGEIRSSON¹, JENNA REPS², PETER RIJNBEEK¹

(1) Department of Medical Informatics, Erasmus University Medical Center, Rotterdam, The Netherlands (2) Janssen Research and Development, Raritan, NJ, United States

The following databases were selected for external validations of these models as they contain an adequate number of elderly patients.

Database	Acronym	No. of patients (million)	Country	Data type
IBM MarketScan Medicare Suppl.	MDCR	10	US	Claims
Iqvia Germany Disease Analyzer	IQGER	30	DE	GP, EHR
Optum Socioeconomic Status	OPSES	85	US	Claims
Optum Electronic Health Records	OPEHR	94	US	EHR
Clinical Practice Research Datalink	CPRD	13	UK	GP
Integrated Primary Care Information	IPCI	2.5	NL	GP
Iqvia Medical Research Database	IMRD	18	UK	GP

RESULTS AND DISCUSSION

The inclusion criteria of our literature search were met by 35 studies, which described a total of 59 prediction models. The following table summarizes the reporting of our replicability criteria in the included articles.

Category	Replicability criteria	Reported by no. of models (%)
Population settings	Target population definition	59 (100)
	Index date	23 (39)
	Time-at-risk	39 (66)
Statistical analysis settings	Outcome definition	59 (100)
	Prediction method	59 (100)
	Predictor definitions	46 (78)
	Predictor time window	21 (36)
	Model specifications: Full model	8 (14)
	Model specifications: Partial model	19 (32)

Our results showed that while reporting was complete for some criteria such as target and outcome definitions, reporting of statistical analysis criteria are mostly insufficient to fully replicate the dementia prediction models.

Moreover, our external validation of three selected models (Walters, Mehta, and Nori) showed that even if reporting was sufficient for replication, it did not guarantee that replication and external validation becomes non-trivial, because predictors had to be present, and inclusion and exclusion criteria of target and outcome had to be generalizable to other data sources. Specific problems that we encountered were the following:

- Walters: Uses a "social deprivation score", which ranges from 1 to 5 indicating social deprivation. The information in this variable has been established through a linkage, which is no longer available, or unlikely to exist in other databases across the world.
- Mehta: Does not report a time-at-risk, which was estimated to be 5 years. Also does not provide the baseline hazard so that only a risk stratification model could be replicated rather than the original Cox proportional hazard model.
- Nori: Does not report a time-at-risk, which was estimated to be 5 years.

Performance across external data sources showed substantial differences in discrimination performance measured as the area under the receiver operating characteristic curve (AUROC) and 95% CI as presented in the following table.

Model	Internal	MDCR	IQGER	OPSES	OPEHR	CPRD	IPCI	IMRD
Walters	0.84	0.69	0.75	0.74	0.73	0.67	0.76	0.68
	THIN	(0.69 – 0.69)*	(0.75 – 0.75)*	(0.74 – 0.74)*	(0.73 – 0.73)*	(0.66 – 0.66)	(0.75 – 0.75)	(0.68 – 0.68)*
Mehta	0.81	0.69	0.72	0.71	0.73	0.79	0.78	0.79
	CPRD	(0.69 – 0.70)	(0.71 – 0.72)	(0.70 – 0.71)	(0.73 – 0.73)	(0.78 – 0.80)	(0.76 – 0.80)	(0.78 – 0.80)
Nori	0.69	0.66	0.67	0.67	0.62	0.68	0.64	0.68
	Optum	(0.66 – 0.67)	(0.66 – 0.68)	(0.66 – 0.68)	(0.62 – 0.63)	(0.67 – 0.69)	(0.62 – 0.67)	(0.68 – 0.69)

We believe that the lack of external validation in dementia prediction literature can to some extent be attributed to the insufficient reporting of models. Models should be developed with external validation in mind. This could for example mean to report all aspects of the model explicitly. Such transparency is best achieved programmatically through code lists and underlying logic rather than literal descriptions, for example by providing a full description of the model (development) in code, ideally against a common data model. This approach will likely eliminate ambiguity as a source of error.

Development choices should not rely on properties unique to the development database, e.g., the Walters model contained criteria to define the target population and predictors that did not exist in the external data sources, for example the cohort entry event "one year following new registration with a THIN practice".

In general, authors should avoid uncommon predictors during model development to guarantee replicability, if the model is meant to be applied in external healthcare settings. Instead of building a single model with multiple, complex cohort entry events, it can be beneficial to build a model for each entry event, which may be easier to interpret and replicate. The Nori model suffered from this problem as it had a complex target population definition with multiple entry events. Defining the time-at-risk window is crucial to indicate in which time window a model's predictions are valid. Using the full follow-up of a population is not a valid approach, as follow-up can vary per person.

CONCLUSION

We reviewed 35 studies that proposed a total of 59 dementia risk models. We observed that reporting is mostly insufficient to fully replicate and externally validate published dementia prediction models, and therefore, it is uncertain how well these models would work in other clinical settings. In addition, we replicated and externally validated three existing dementia prediction models and encountered difficulties beyond our replicability criteria, such as ambiguous cohort or predictor definitions. We recommend that reporting should be more explicit and have external validation in mind if the model is meant to be applied in different settings.

References:

- Hou XH et al. Models for predicting risk of dementia: a systematic review. *Journal of Neurology, Neurosurgery and Psychiatry*. 2019;90(4):373-9.
- Goerdtzen J et al. Statistical methods for dementia risk prediction and recommendations for future work: A systematic review. *Alzheimer's and Dementia: Translational Research and Clinical Interventions*. 2019;5:363-9.
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- Nori VS et al. Identifying incident dementia by applying machine learning to a very large administrative claims dataset. *PLoS ONE*. 2019;14(7).


CONTACT: LJOHN@ERASMUSMC.NL

www.ehden.eu [@IMI_EHDEN](https://twitter.com/IMI_EHDEN) [IMI_EHDEN](https://www.linkedin.com/company/IMI_EHDEN) github.com/EHDEN [epia](https://www.epia.eu)

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MONDAY

External validation of existing dementia prediction models on observational data
Lead: Henrik John

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#JoinTheJourney

 ohdsi



#OHDSISocialShowcase This Week

Characteristics and outcomes of inflammatory bowel disease patients: an open, multinational OHDSI network study

PRESENTER: **Chen Yanover**
KI Research Institute

INTRO

- Crohn's disease (CD) and ulcerative colitis (UC) are chronic inflammatory bowel diseases with consistently increasing incidence rates
- These conditions significantly impact the quality of life of patients and families

METHODS

Study design. A multinational cohort study using routinely collected healthcare data.

Data sources.

- IQVIA Medical Research Data, primary care electronic health records (EHRs) from the United Kingdom (IMRD-UK, version: 2019-03); over 12.5 million patients, approximately 5% of the UK population
- The Ajou University School of Medicine (AUSOM) data, an EHR database from Ajou university hospital in South Korea, containing >2.8 million patients, from 1995 to 2022

Study population. IBD cohorts include individuals with at least two diagnoses of IBD or with an IBD diagnosis and a prescription for an IBD medication; CD and UC cohorts further require at least one diagnosis of the corresponding disease and none of the other

Characteristics and outcomes.

- Baseline characteristics: during patients' entire history, 1Y, 1M before index date
- Outcomes and treatments: during the 1M, 1Y, 3Y, 5Y, 10Y, following index date and full follow-up time window
- Attributes: OHDSI predefined features (demographics, condition groups, drug era groups), +100 IBD-specific features

A unified, global characterization of INFLAMMATORY BOWEL DISEASE patient cohorts



Setting the stage for IBD-related predictive and estimation OHDSI network studies

... download analysis package



... read study protocol



... explore full results



COLLABORATORS



Ajou University

Jimyung Park, Rae Woong Park, Kwang Jae Lee, Sung Jae Shin



Ben-Gurion University

Nadav Rappoport



Rambam Medical Center

Ramit Magen-Rimon, Yehuda Chowers, Roni Weissshof



KI Research Institute

Tal El-Hay, Maytal Bivas-Benita, Pinchas Akiva



	Crohn's disease		Ulcerative colitis	
	AUSOM	IMRD-UK	AUSOM	IMRD-UK
n	402	6,936	635	13,924
Females	36.6%	54.4%	36.2%	49.1%
Age				
<20 years	25.9%	16.3%	7.9%	5.8%
20-65 years	67.3%	69.8%	83.2%	71.3%
>65 years	4.7%	13.8%	8.0%	22.8%
BMI category ^a				
Underweight	4.0%	5.6%	1.9%	2.7%
Normal	12.7%	31.5%	15.7%	29.3%
Overweight	4.0%	25.8%	5.0%	30.2%
Obese	-	17.4%	-	18.5%
Medications (index date - 30 days)				
5-ASA	65.9%	43.2%	84.4%	67.0%
Immunomodulator	15.4%	8.2%	4.7%	2.0%
Corticosteroids ^b	7.0%	6.2%	5.0%	4.2%
Antibiotics	23.1%	1.1%	9.3%	0.6%
Outcomes (index date - 3 years)				
Anxiety	10.4%	6.7%	13.2%	6.7%
Depression	4.7%	9.3%	5.8%	8.2%
Anemia	5.5%	7.6%	1.7%	6.6%
Procedures (index date - 3 years)				
S bowel resection	4.7%	10.5%	1.1%	4.7%

^a Underweight: BMI <18.5 kg/m², normal weight: 18.5-25 kg/m², overweight: 25-30 kg/m², obese: >30 kg/m².
^b Systemic (prednisone, methylprednisolone) and intestinal locally acting (prednisolone, hydrocortisone, prednisone, betamethasone, triamcinolone, budesonide, beclomethasone) steroids.

INTERESTED?
Join us!

TUESDAY

Characteristics and outcomes of inflammatory bowel disease patients: an open, multinational OHDSI network study
Lead: Chen Yanover



#OHDSISocialShowcase This Week

Mapping PROMs data from the Dutch PROFILES registry to the OMOP CDM - experiences and challenges

PRESENTER: Peter Prinsen

INTRO

- Data from the Netherlands Cancer Registry (NCR) is being converted to the OMOP-CDM.
- PROFILES contains patient-reported outcome measures (PROMs) data that is linked to patients in the NCR.
- How do we add this PROMs data to the OMOP-CDM to make the data set even more interesting? OHDSI is not very clear on that.

METHODS

Use cases we foresee:

- Specific studies on PROMs data.
- General studies where PROMs data provides additional information about the patient.

A possible solution for adding PROMs data to the OMOP-CDM is outlined in Fig. 1.

RESULTS

We mapped the EORTC QLQ-C30 and the Hospital Anxiety and Depression Scale (HADS, see Fig. 2), and added those to our OMOP-CDM. They both contain question/answer (Q/A) pairs and scores. The latter are calculated from subsets of Q/A pairs.

Examples of mappings are shown in Figs. 3 & 4.

DISCUSSION

There are several open questions, to be answered by the OHDSI community:

- Should there be an overarching "PROMs questionnaire filled out" concept that Q/A pairs and scores can be linked to?

Yes, that is a good way to group data from a single filled out questionnaire.

No, irrelevant for studies.

- Should Q/A pairs be added to the OMOP-CDM? Should scores?

Yes, they are relevant for specific studies. No, only clinical facts and events are relevant.

- Should Q/A pairs be mapped to clinical concepts?

Yes, always.

Yes, but only if there are no scores (questions are just a tool to determine scores, the latter are the only relevant outcomes).

Guideline for adding PROMs to OMOP-CDM nonexistent PROMs vocabulary incomplete

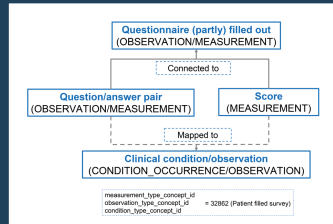


Figure 1. A solution for adding PROMs data to the OMOP-CDM.

Score	Value	Concept ID	Name	Vocabulary
Anxiety score	0-7 (Normal)	do not map		
	8-10 (Borderline abnormal)	4322025	Mild anxiety	SNOMED
	11-21 (Abnormal)	441542	Anxiety	SNOMED
Depression score	0-7 (Normal)	do not map		
	8-10 (Borderline abnormal)	40540087	Depressed mood	SNOMED
	11-21 (Abnormal)	40540087	Depressed mood	SNOMED

Figure 3. Example of score mapping: HADS scores.

Figure 2. Example of a PROMs questionnaire: HADS.

Question	Concept ID*	Answer	Concept ID**	Mapped question/answer	Concept ID***
Do you have any trouble doing strenuous activities, like carrying a heavy shopping bag or a suitcase?	701340	not at all	45883172	Capable of heavy physical activity	763659
		a little	45876949	?	?
		quite a bit	45884456	?	?
		very much	45882566	?	?
Do you have any trouble taking a long walk?	701341	not at all	45883172	Able to walk	4086870
		a little	45876949	Unable to walk long distances	44792042
		quite a bit	45884456	Unable to walk long distances	44792042
		very much	45882566	Unable to walk long distances	44792042

Figure 4. Example of question and answer concepts, and mapping of Q/A pairs: the first two questions of QLQ-C30.

- Should scores be mapped to clinical concepts?
Yes, that is the clinically most important part.
No?

- Should we add negative concepts (no pain)?
Yes, they are relevant observations.
No, this information is not useful in studies.

- Can we use ..._type_concept to indicate patient-reported (vs physician-reported)?
Yes, we do not want to create a whole bunch of new concepts.
No, nobody uses the ...type_concept field in analyses.

- What observation or event date do we associate with the PROMs data ("Did you experience pain in the past month")?

CONCLUSIONS

- A guideline for adding PROMs to the OMOP-CDM does not exist: the OHDSI community should develop conventions for capturing PROMs data in OMOP.
- The vocabulary is very incomplete when it comes to representing PROMs data.

JOIN US!

Do you want to collaborate on harmonizing PROMs data in OMOP? Then contact Sebastiaan and join the EHDEN PROMS WG (sebastiaan.van.sandijk@odysseusinc.com)

Peter Prinsen^{1*},
Chiara Attanasio¹,
Corina van den Hurk¹,
Nicole Horevoorts¹,
Sebastiaan van Sandijk²

¹IKNL,
²Odysseus Data Services,
*p.prinsen@iknl.nl



Mapping PROMs data from the Dutch PROFILES registry to the OMOP CDM - experiences and challenges

WEDNESDAY

Lead: Peter Prinsen



#OHDSISocialShowcase This Week

OHDSI Germany
A recap after one year
📍 PRESENTER: Michele Zoch

INTRO

- Establishing OHDSI Germany as a multi-stakeholder interest group in Spring 2021
- Goals:
 - Establishing a German research network of hospitals on OMOP for OHDSI
 - Creating and sharing best practices as well as deliverables
 - Involving German stakeholder and providing assistance with getting started in the OHDSI community
 - Collaborating for administrative work in terms of German data security laws and ethics

METHODS

- Monthly community meetings
- Presentations and discussions
- Sharing best practices in plenary

RESULTS

- Formation of an interdisciplinary interest group
- Identifying common topics and synergy effects
- Offering workshops and tutorials

OUTLOOK

- Onboarding of further participants in Germany
- Exploiting further data sources and terminologies
- Extending of ETL processes
- Providing a technology stack (esp. German OMOP Stack, ETL jobs)
- Participating in DARWIN
- Collaborating in newly applied projects of the German Medical Informatics Initiative
- Working together to overcome administrative hurdles
- Intensify joint work

OHDSI Germany is one of the first contact points for
sharing best practices and
experiences with a focus on
German patient data.



Networking



Sharing



Adapting



Collaborating

PREVIOUS TOPICS

- Development of a German specific OMOP Stack and ETL jobs
- Strategies for using OMOP for rare diseases
- Integration of federally-mandated Medical Information Objects
- Effects of pre- and post-coordination on mapping

RELATED LINKS



📍 Michele ZOCH¹, Elisa HENKE¹,
Yuan PENG¹, Najia AHMADI¹,
Joshua WIEDEKOPF², Mareike
PRZYSUCHA³, Josef SCHEPERS⁴,
Martin SEDLMAYR¹, Ines
REINECKE¹

¹Institute for Medical Informatics and
Biometry at Carl Gustav Carus Faculty
of Medicine, Technische Universität
Dresden, Germany

²IT Center for Clinical Research,
University of Lübeck, Germany

³Health Informatics Research Group,
University AS Osnabrück, Germany

⁴Berlin Institute of Health, Charité –
Universitätsmedizin Berlin, Germany



THURSDAY

OHDSI Germany: A recap after one year
Lead: Michele Zoch



#OHDSISocialShowcase This Week

RCTrep: An R package for the validation of methods for treatment effect estimation using real-world data

PRESENTER: **Lingjie Shen**^{1*}

INTRO:

- Who cares? – policy makers; regulators; real world evidence (RWE) evaluators.
- Why? There is an increasing attention for the leverage of large real-world data (RWD) in treatment effect estimation to drive fast and precise decision making.
- Challenge: Since we do not observe the true treatment effect for each individual- which is the fundamental problem of causal inference - validation of treatment effect estimation methods using RWD is challenging.
- Aim: In the absence of a ground truth, how can we validate different methods using RWD to select the most reasonable method for the data at hand, driving fast regulatory and clinical decision making?

METHODS:

- We identify under which conditions the estimate from randomized control trial (RCT) can be regarded as the ground truth for methods validation using RWD. We illustrate differences between RCT and RWD in Figure 1. We assume the RWD and RCT data are two random samples from a, potentially different, population, and hence allow for a fair comparison of estimates of treatment effect between two samples after population composition is controlled for.
- We consider a set of candidate treatment effect estimators $\mathcal{F} = \{f_1, \dots, f_m\}$, where $f(x): \mathcal{X} \mapsto \mathbb{R}$ is an estimator of conditional average treatment effect of population with characteristics $X = x$. We select the best one using the following evaluation metric:

$$f^* = \underset{f \in \mathcal{F}}{\operatorname{argmin}} L(f; \hat{f}) = \underset{f \in \mathcal{F}}{\operatorname{argmin}} \left(\hat{f} - \sum_{x \in \mathcal{X}} w(x) f(x) \right)^2$$
$$\text{s.t. } p(x) = q(x)w(x)$$

where \hat{f} is an unbiased estimate of average treatment effect of a population that a RCT represents, $p(x)$ and $q(x)$ are the empirical density of x in RCT data and RWD, $w(x)$ is a weight for individuals in RWD with characteristics $X = x$.

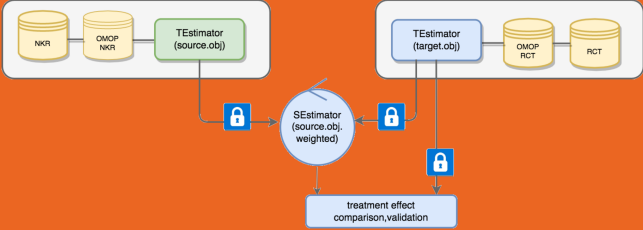


Figure 2: Diagram of RCTrep basic structure

- TEstimator:** R6 class TEstimator is responsible for estimating population- and subpopulation-level treatment effects, and diagnosing assumptions.
- SEstimator:** R6 class SEstimator is responsible for computing weights, so that the weighted covariates in source.obj and covariates in target.obj are balanced. The two objects communicate within the object of the class SEstimator, sharing either unit-level data or aggregated data for computing the weights.

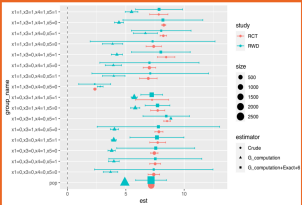


Figure 3: A working example of RCTrep. We use the **G-computation** method to adjust the treatment assignment mechanism, and use **exact matching** to adjust the sampling mechanism. Results show that **only** correcting for **both** mechanisms can allow for comparison of treatment effect estimation between RWD and RCT data.

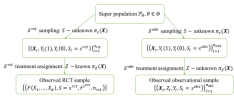


Figure 1: The mechanisms of RWD and RCT data generations

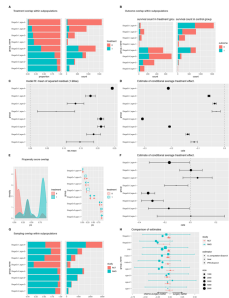


Figure 4: Estimates comparison between NKR and QUASAR trial using RCTrep. Subfigure (a) diagnosis overlap of treatment within subgroups in NKR data and survival in treatment and control groups in NKR data. Figure (b) diagnosis: G-computation model fit and estimates of treatment effect in subgroups. Figure (c) diagnosis: propensity score overlap between treatment and control groups and estimates of treatment effect using inverse propensity score weighting. Figure (d) diagnosis: covariates balance between NKR and QUASAR trial and comparison of estimates from QUASAR and estimates from NKR with and without weighting.

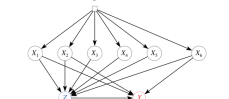


Figure 5: Illustration of adjustment sets in TEstimator and SEstimator. S is an indicator of selection into RCT and Z is an indicator of selection into treatment group.

Lingjie Shen^{1*},
Gijs Geleijnse²,
Maurits Kaptein³

¹Department of Methodology and Statistics,
Tilburg University,
²KNL,
³Jheronimus Academy of Data Science,
*L.Shen@uvt.nl



FRIDAY

A dashboard for visual comparison of OMOP CDM databases
Leads: João Almeida, José Oliveira



Where Are We Going?

**Any other announcements
of upcoming work, events,
deadlines, etc?**





Three Stages of The Journey

Where Have We Been?

Where Are We Now?

Where Are We Going?





Sept. 6: OHDSI Publications

A database of pediatric drug effects to evaluate ontogenic mechanisms from child growth and development

Presenter: Nick Giangreco



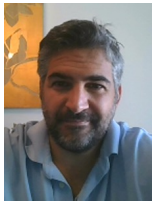
Development and external validation of prediction models for adverse health outcomes in rheumatoid arthritis: A multinational real-world cohort analysis

Presenter: Cynthia Yang



Empirical assessment of alternative methods for identifying seasonality in observational healthcare data

Presenter: Anthony Molinaro



Phenotype Algorithms for the Identification and Characterization of Vaccine-Induced Thrombotic Thrombocytopenia in Real World Data: A Multinational Network Cohort Study

Presenter: Azza Shoaibi



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

Presenter: Aniek Markus



Figure 4 A

A database of pediatric drug effects using ATC and MedDRA

Giangreco NP, Tatonetti NP. A database of pediatric drug effects to evaluate ontogenic mechanisms from child growth and development. Med (N Y). 2022 Aug 12;3(8):579-595.e7. doi: 10.1016/j.medj.2022.06.001. Epub 2022 Jun 24. PMID: 35752163; PMCID: PMC9378670.

- We generated 500,000 drug safety signals for every drug (ATC5) and adverse event (PT) identified during childhood
- We compared drug safety signals to a null distribution to estimate statistical significance for each pediatric ADE
- We *grouped* ADEs into the ATC1-5 and PT-SOC categories
- We asked if the drug or adverse event class was enriched for significant ADEs at a stage, and the matrix shows how many
- The graphs show select drug ingredients enriched in systemic adverse events across childhood

	ATC1 N=14	ATC2 N=77	ATC3 N=166	ATC4 N=358	ATC5 N=926	Number of associated adverse event classes
SOC N=27	43	66	60	38	36	243
HLGT N=334	23	15	9	3	2	52
HLT N=1,472	3	3	2	2	1	11
PT N=6,841	0	0	0	0		0
Number of associated drug classes	69	84	71	43	39	306
						Number of associated ADE classes

B

