



April Olympians: What We Learned & How We Can Use It

OHDSI Community Call
April 30, 2024 • 11 am ET



Upcoming Community Calls

| Date | Topic |
|---------------|---|
| April 30 | April Olympians Update Presentation: What We Achieved & How You Can Use It |
| May 7 | DevCon 2024 Review |
| May 14 | 10-Minute Tutorials |
| May 21 | Open Studies in the OHDSI Community |
| May 28 | Collaborator Showcase Brainstorm |
| June 4 | NO CALL – EUROPEAN SYMPOSIUM |
| June 11 | European Symposium Review |
| June 18 | Application of LLMs In Evidence Generation Process |
| June 25 | Recent OHDSI Publications |



Three Stages of The Journey

Where Have We Been?

Where Are We Now?

Where Are We Going?





OHDSI Shoutouts!



Congratulations to the team of **Martin Baumgartner, Karl Kreiner, Aaron Lauschensky, Bernhard Jammerbund, Klaus Donsa, Dieter Hayn, Fabian Wiesmüller, Lea Demelius, Robert Modre-Osprian, Sabrina Neururer, Gerald Slamanig, Sarah Prantl, Luca Brunelli, Bernhard Pfeifer, Gerhard Pölzl, and Günter Schreier** on the publication of **Health data space nodes for privacy-preserving linkage of medical data to support collaborative secondary analyses in *Frontiers in Medicine*.**

frontiers | Frontiers in **Medicine**

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Check for updates

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Health data space nodes for privacy-preserving linkage of medical data to support collaborative secondary analyses

Martin Baumgartner^{1,2*}, Karl Kreiner¹, Aaron Lauschensky¹, Bernhard Jammerbund¹, Klaus Donsa¹, Dieter Hayn^{1,3}, Fabian Wiesmüller^{1,2,3}, Lea Demelius^{4,5}, Robert Modre-Osprian⁶, Sabrina Neururer^{7,8}, Gerald Slamanig⁹, Sarah Prantl⁹, Luca Brunelli¹⁰, Bernhard Pfeifer^{8,11}, Gerhard Pölzl¹⁰ and Günter Schreier^{1,2}

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Introduction: The potential for secondary use of health data to improve healthcare is currently not fully exploited. Health data is largely kept in isolated data silos and key infrastructure to aggregate these silos into standardized bodies of knowledge is underdeveloped. We describe the development, implementation, and evaluation of a federated infrastructure to facilitate versatile secondary use of health data based on Health Data Space nodes.

Materials and methods: Our proposed nodes are self-contained units that digest data through an extract-transform-load framework that pseudonymizes and links data with privacy-preserving record linkage and harmonizes into a common data model (OMOP CDM). To support collaborative analyses a multi-level feature store is also implemented. A feasibility experiment was



OHDSI Shoutouts!



Congratulations to the team of **Chungsoo Kim, Dong Han Yu, Hyeran Baek, Jaehyeong Cho, Seng Chan You, and Rae Woong Park** on the publication of **Data Resource Profile: Health Insurance Review and Assessment Service Covid-19 Observational Medical Outcomes Partnership (HIRA Covid-19 OMOP) database in South Korea** in the *International Journal of Epidemiology*.

The screenshot shows the journal's website with a yellow header and a blue navigation bar. The article title is prominently displayed in bold black text. Below the title, the authors' names are listed, followed by the journal name, volume, issue, and date. A 'Get access' button is visible. The 'Extract' section provides a summary of the article's content, mentioning the National Health Insurance (NHI) system in South Korea and the role of the Health Insurance Review and Assessment Service (HIRA).

International Journal of Epidemiology

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JOURNAL ARTICLE

Data Resource Profile: Health Insurance Review and Assessment Service Covid-19 Observational Medical Outcomes Partnership (HIRA Covid-19 OMOP) database in South Korea [Get access >](#)

Chungsoo Kim, Dong Han Yu, Hyeran Baek, Jaehyeong Cho, Seng Chan You, Rae Woong Park ✉

International Journal of Epidemiology, Volume 53, Issue 3, June 2024, dyae062, <https://doi.org/10.1093/ije/dyae062>

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Extract

Data resource basics

The health insurance system in South Korea is known as the National Health Insurance (NHI) system. It is a mandatory insurance system that covers approximately 97% of Korean citizens and residents.¹ The NHI system covers most medical expenses, including prescription drugs, diagnostic tests and medical procedures. The Health Insurance Review and Assessment Service (HIRA) in South Korea plays a crucial role in the health care system by monitoring and assessing the quality and appropriateness of medical services provided to patients under the NHI system.² Its primary mission is to ensure that the medical services provided to patients are of high quality and cost-effective, by reviewing and assessing the appropriateness of medical services, evaluating the quality of care and monitoring healthcare use and costs.



OHDSI Shoutouts!



Congratulations to the team of **Markus Falgenhauer, Aaron Lauschensky, Karl Kreiner, Stefan Beyer, Kristina Reiter, Andreas Ziegl, Robert Modre-Osprian, Bernhard Pfeifer, Sabrina Neururer, Susanne Krestan, Hanna Wagner, Andreas Huber, Sandra Plaikner, Sarah Kuppelwieser, Martin Widschwendter, and Günter Schreier** on the publication of **Towards an Electronic Health Prevention Record Based on HL7 FHIR and the OMOP Common Data Model** in *Volume 313 of Studies in Health Technology and Informatics*.

The screenshot shows the IOS Press Ebooks website interface. The top navigation bar includes 'Home', 'Ebooks', 'Open Access', 'About IOS Press', 'Contact', and 'FAQ'. A search bar is visible on the left. The main content area displays the article title, authors (Markus Falgenhauer, Aaron Lauschensky, Karl Kreiner, Stefan Beyer, Kristina Reiter, Andreas Ziegl, Robert Modre-Osprian, Bernhard Pfeifer, Sabrina Neururer, Susanne Krestan, Hanna Wagner, Andreas Huber, Sandra Plaikner, Sarah Kuppelwieser, Martin Widschwendter, Günter Schreier), page count (107 - 112), DOI (10.3233/SHTI240020), category (Research Article), series (Studies in Health Technology and Informatics), and ebook volume (Volume 313: dHealth 2024). The abstract section includes background, objectives, and methods.

Search

Browse by subject

- Computer Sciences, Mathematics & Statistics
- Environmental & Energy Sciences
- Humanities & Social Sciences
- Medicine & Health
- Natural Sciences
- Technology, Engineering & Architecture

Towards an Electronic Health Prevention Record Based on HL7 FHIR and the OMOP Common Data Model

Authors Markus Falgenhauer, Aaron Lauschensky, Karl Kreiner, Stefan Beyer, Kristina Reiter, Andreas Ziegl, Robert Modre-Osprian, Bernhard Pfeifer, Sabrina Neururer, Susanne Krestan, Hanna Wagner, Andreas Huber, Sandra Plaikner, Sarah Kuppelwieser, Martin Widschwendter, Günter Schreier

Pages 107 - 112

DOI 10.3233/SHTI240020

Category Research Article

Series [Studies in Health Technology and Informatics](#)

Ebook Volume 313: dHealth 2024

Abstract

Background: Approximately 40% of all recorded deaths in Austria are due to behavioral risks. These risks could be avoided with appropriate measures.

Objectives: Extension of the concept of EHR and EMR to an electronic prevention record, focusing on primary and secondary prevention.

Methods: The concept of a structured prevention pathway, based on the principles of P4 Medicine, was developed for a multidisciplinary prevention network. An IT infrastructure based on HL7 FHIR and the OHDSI OMOP common data model was designed.



OHDSI Shoutouts!



Congratulations to the team of **Evgeniy Krastev, Emanuil Markov, Simeon Abanos, Ralitsa Krasteva, and Dimitar Tcharaktchiev** on the publication of **Towards an Electronic Health Prevention Record Based on HL7 FHIR and the OMOP Common Data Model** in *Volume 313 of Studies in Health Technology and Informatics*.

The screenshot shows the IOS Press website interface. At the top, there is a navigation bar with the IOS Press logo and the text 'IOS Press Ebooks'. Below this is a secondary navigation bar with links for 'Home', 'Ebooks', 'Open Access', 'About IOS Press', 'Contact', and 'FAQ'. A search bar is located on the left side, with a 'SEARCH' button. Below the search bar is a 'Browse by subject' section with a list of categories: Computer Sciences, Mathematics & Statistics; Environmental & Energy Sciences; Humanities & Social Sciences; Medicine & Health; Natural Sciences; and Technology, Engineering & Architecture. The main content area displays the article 'Mapping the Bulgarian Diabetes Register to OMOP CDM: Application Results'. The article details include: Authors: Evgeniy Krastev, Emanuil Markov, Simeon Abanos, Ralitsa Krasteva, Dimitar Tcharaktchiev; Pages: 28 - 33; DOI: 10.3233/SHTI240007; Category: Research Article; Series: Studies in Health Technology and Informatics; Ebook: Volume 313: dHealth 2024. The abstract section is titled 'Abstract' and contains three sub-sections: 'Background:', 'Objectives:', and 'Methods:'. The 'Background:' section states: 'The Bulgaria Diabetes Register (BDR) contains more than 380 millions of pseudonymized outpatient records with proprietary data structures and format.' The 'Objectives:' section states: 'This paper presents the application results and experience acquired during the process of mapping such observational health data to OMOP CDM with the objective of publishing it in the European Health Data and Evidence Network (EHDEN) Portal.' The 'Methods:' section states: 'The data mapping follows the activities of the well-structured Extract-Transform-Load process. Unlike other publications, we focus on the need for preprocessing the data structures of raw data, cleaning data and procedures for assuring quality of data.'



Three Stages of The Journey

Where Have We Been?

Where Are We Now?

Where Are We Going?





Upcoming Workgroup Calls



| Date | Time (ET) | Meeting |
|-----------|-----------|--|
| Tuesday | 12 pm | Common Data Model Vocabulary Subgroup |
| Wednesday | 8 am | Psychiatry |
| Wednesday | 3 pm | Joint Vulcan/OHDSI Meeting |
| Wednesday | 7 pm | Medical Imaging |
| Thursday | 9:30 am | Themis |
| Thursday | 11 am | Industry |
| Thursday | 1 pm | OMOP CDM Vocabulary/Development Subgroup |
| Thursday | 7 pm | Dentistry |
| Friday | 10 am | GIS-Geographic Information System |
| Friday | 11:30 am | Steering Group |
| Monday | 9 am | Vaccine Vocabulary |
| Monday | 10 am | Healthcare Systems Interest Group |
| Tuesday | 9 am | Atlas/WebAPI |
| Tuesday | 10 am | Registry |
| Tuesday | 10 am | Common Data Model |



DevCon 2024 Videos Are Posted

Morning Agenda

9:00 am – Introduction ([Adam Black](#), [Paul Nagy](#))

9:15 am – Developers Panel and Lightning Talks ([Katy Sadowski](#))

- *OHDSI/OMOP – The hard way is the easy way!* ([Vishnu V Chandrabalan](#))
- *Moving OMOP to the Cloud With DBT and Snowflake* ([Roger Carlson](#))
- *Use cases for ORMs in OMOP* ([Georgina Kennedy](#))
- *Carrot: code-free OMOP ETL without full data access* ([Sam Cox](#))
- *Rabbit-in-a-blender - an ETL pipeline to transform your EMR data into OMOP* ([Pieter-jan Lammertyn](#))

10:45 am – Darwin EU[®] Developers Update ([Adam Black](#))

- *CDMConnector, PatientProfiles, CohortCharacteristics, CohortSurvival*

12:00 pm – Break

Afternoon Agenda

12:30 pm – OHDSI Ecosystem Updates

- TAB Update ([Frank DeFalco](#))
- Strategus Update ([Anthony Sena](#))
- Broadsea Update ([Lee Evans](#))
- Kheiron Updates ([Paul Nagy](#))

1:15 pm – JACKALOPE PLUS The Power of ML for Healthcare Data Mapping & Management ([Denys Kaduk](#))

2:00 pm - An Introduction to Knowledge Graphs using PheKnowLator and OMOP2OBO with Example Applications in Drug Surveillance and Computational Phenotyping ([Tiffany Callahan](#))



#OHDSI2024 Registration Is Open!

Registration is now OPEN for the 2024 OHDSI Global Symposium, which will be held Oct. 22-24 at the Hyatt Regency Hotel in New Brunswick, N.J., USA.

Tuesday: Tutorials

Wednesday: Plenary/Showcase

Thursday: Workgroup Activities



ohdsi.org/OHDSI2024



#OHDSI2024 Collaborator Showcase

Submissions are now being accepted for the 2024 Global Symposium Collaborator Showcase.

All submissions are due by 8 pm ET on Friday, June 21.

Notification of acceptance will be made by Tuesday, Aug. 20.



ohdsi.org/OHDSI2024



Maternal Health Data Science Fellowship

This program is designed to empower clinical investigators to leverage emerging technologies for improved maternal and neonatal care while reducing morbidity and mortality.

Three main components of this program:

1) Career Development (create evidence, leverage data models, build skills on network studies)

2) Practice (design effective observational research protocols, master tools, write papers/grants)

3) Networking (build relationships with mentors, learners, coordinate with global OHDSI collaborators)

Application deadline: May 15

Want to build your career?

Generate reproducible evidence by leading multi-institutional studies!

Learn more & apply!





RWE Workshop at AIME24: Call for Submissions!

Workshop: AI for Reliable and Equitable Real-World Evidence Generation in Medicine

<https://medicine.utah.edu/dbmi/aime/ai-reliable>

Organizing Committee

Linying Zhang
Adam Wilcox
Yves Lussier

Scientific Program Committee

Peter Rijnbeek Mattia Prosperi
Larry Han Xia Ning
Xiaoqian Jiang Yifan Peng

Opening Keynote

George Hripcsak

IMPORTANT DATES

May 31, 2024 | Submission Deadline

June 14, 2024 | Notice of Acceptance

July 12, 2024 | Workshop



AIME 2024
22nd International Conference on Artificial Intelligence in Medicine
Salt Lake City, Utah, USA, July 9-12
Hosted by the University of Utah



#OHDSISocialShowcase This Week

MONDAY

Opportunity and Challenge of Implementing the OHDSI System in Indonesia

(Dian Tri Wiyanti, Daniel C.A. Nugroho, Yudha Eri Saputra, Septi Melisa, Phan Thanh-Phuc, Nguyen Phung-Anh, Jason C. Hsu, Min-Huei Hsu)



Opportunity and Challenge of Implementing the OHDSI System in Indonesia

Dian Tri Wiyanti^{1,2}, Daniel C.A. Nugroho^{3,4}, Yudha Eri Saputra⁵, Septi Melisa¹, Phan Thanh-Phuc¹, Nguyen Phung-Anh^{5,6,7}, Jason C.Hsu^{1,8}, Min-Huei Hsu^{1,9,10,11*}

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Background

Indonesia, a densely populated nation with diverse ethnicities, possesses medical institutions like UKDW (Universitas Kristen Duta Wacana) that have well-established healthcare systems. These institutions publish research utilizing Electronic Health Records (EHR), making them valuable for the OHDSI (Observational Health Data Sciences and Informatics) community. This study aims to evaluate the potential for implementing OHDSI in Indonesia, with UNNES (Universitas Negeri Semarang) offering fresh insights. Despite challenges, the establishment of a new health faculty presents an opportunity to integrate OHDSI into the curriculum, foster research collaborations, and develop a strong data platform. Overcoming obstacles will necessitate careful planning, resource allocation, and active involvement of stakeholders. The study assesses healthcare preparedness, technical prerequisites, market landscape, and identifies obstacles and prospects for implementation.

Methods

Table 1. Variable Comparison across Indonesia BPJS, South Korea NHIS-NSC OMOP CDM Conversion, TMUCRD, and Indonesian Hospital Record Datasets

| Indonesia BPJS | South Korea NHIS-NSC OMOP CDM Conversion | TMUCRD | Indonesian Hospital Record |
|----------------------------------|--|-----------------------------|----------------------------|
| NHI ID | PERSON | ID_No | NHI ID |
| - | DEATH | Death | DEATH |
| PHC VISIT Hospital Visit | VISIT | OPD/IPD FEE_No | Visit |
| PHC DIAGNOSIS Hospital Diagnosis | CONDITION | ICD9_Code/ICD10_Code | Diagnosis |
| Indonesian Case Based Group | DRUG | MED_Code | Drug |
| | PROCEDURE | OPD/IPD GROUP_Code/EXPER_No | Procedure |
| | DEVICE | - | - |
| | MEASUREMENT | EXPER_Code/B_Item | - |
| | OBSERVATION | - | - |
| COST | TOT_AMT | - | - |

Results

The dataset of the Indonesian Health Social Security Agency includes primary services (registration, outpatient procedures, and discharge procedures) and administrative tasks (planning, procurement, inventory maintenance, asset management, HR management, and financial management). It encompasses 111 variables. These variables align with the OMOP CDM categories (Person, Death, Visit, Condition, Drug, Procedure, Device, Measurement, Observation, and Cost) and serve as classifiers for the Indonesian BPJS dataset.



Figure 1. Analyzing the Opportunities and Challenges: Unleashing the Potential of OHDSI Implementation in Indonesia

Conclusions

Overall, OHDSI in Indonesia brings potential and problems. The report emphasizes UKDW's infrastructure, knowledge, and stakeholder engagement in OHDSI acceptability. With UKDW's worldwide collaborations, UNNES and UKDW want to implement OHDSI. Evidence-based decision-making, innovation, and thorough awareness campaigns are effective implementation techniques. OHDSI implementation is simplified by aligning national health insurance data format with OMOP-CDM. Effective resource management and UNNES training may help adapt. For OHDSI integration, UKDW's IT and EHR competence is essential. Both UKDW and UNNES expansion activities help support OHDSI in Indonesia. Despite minimal efficacy testing, collaboration and improvement are stressed. Integration needs careful planning, resource allocation, stakeholder participation, and problem-solving. UKDW-UNNES collaboration improves OHDSI implementation in Indonesia.

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#OHDSISocialShowcase This Week

TUESDAY

HowOften: Large Scale Incidence Rate Calculation of Every Side Effect for Every Drug

(Elise Ruan, Karthik Natarajan, Ruijun Chen, Jungmi Han, Mark Velez, Taha Abdul-Basser, Edwin M. Cruz, Cindy Hsin-Yi Chen, Patrick Ryan, George Hripcsak)



HowOften: Large Scale Incidence Rate Calculation of Every Side Effect for Every Drug

Elise Ruan¹, Karthik Natarajan¹, Ruijun Chen², Jungmi Han¹, Mark Velez¹, Taha Abdul-Basser¹, Edwin M. Cruz¹, Hsin-Yi Chen¹, Patrick Ryan^{1,3}, George Hripcsak¹

1. Department of Biomedical Informatics, Columbia University Irving Medical Center, New York, NY 2. Department of Medicine, Weill Cornell Medical College, New York, NY 3. Janssen Research & Development, Titusville, NJ

Key Points

- We calculated the incidence proportion of all clinical outcomes (using SNOMED CT codes) after the first drug exposure of all drugs (using RxNorm).
- Our calculated incidence ranges had overlap with ranges found in the literature for nine out of the ten pre-selected drug-adverse outcome pairs.
- Additional work on phenotyping for clinical outcomes of interest is needed. Please join us for the extended HowOften workshop this Symposium!

Background

- Even without causality, knowing the incidence of a clinical condition after initiation of a drug can help guide clinical decision-making.
- Currently, there is no systematic evaluation of all potential side effects for all drugs.
- Using the OMOP CDM, we use real world data to calculate the incidence of every clinical condition following initiation of every drug and published results on an internal site.

Methods

- Exposure cohorts generated for each drug ingredient in RxNorm and RxNorm Extension
- Outcomes cohorts generated based on disorder concepts in SNOMED CT
- Index date: date of first exposure of drug.
- Time-at-risk: 365 days post-index date
- Analysis run October 2017 on 11 databases, (containing EHR and claims data) which had been converted to the OMOP CDM
- Incidence proportion calculated using two methods:
 - Using only patients with data present in the database for the full time-at-risk
 - Using all patients
- Pre-selected 10 known drug-adverse outcome pairs and compared calculated incidence to literature review findings

Results

13,005,797 unique drug-outcome pairs from 2,072 drug concepts and 21,433 outcome concepts

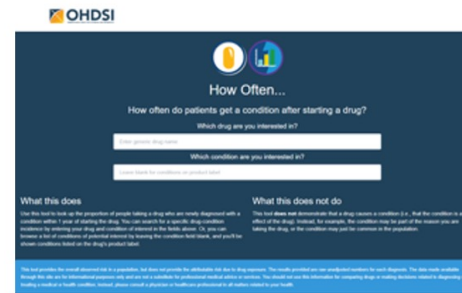


Figure 1. Entry screen for the internal site prototype

Risk of cough with Lisinopril

Amongst patients taking Lisinopril, onset of cough occurs in 3.33% to 21.67% of patients during the 1 year after starting the drug



| Source | Country | Condition | Incidence (%) | Patients at Risk | Requires Full Time at Risk Ⓢ |
|--------------------|---------|-----------|---------------|------------------|------------------------------|
| Premier | US | Cough | 21.67 | 2,598 | Yes |
| MDCD | US | Cough | 13.83 | 154,437 | Yes |
| MDCR | US | Cough | 12.89 | 472,728 | Yes |
| Optum Extended SES | US | Cough | 11.24 | 1,165,871 | Yes |
| CCAE | US | Cough | 10.42 | 1,587,428 | Yes |
| MDCR | US | Cough | 9.76 | 623,910 | No |
| MDCD | US | Cough | 9.09 | 235,013 | No |

Figure 2. Sample drug-condition pair with incidence proportions for each database that contained exposure cohort patients.

Results

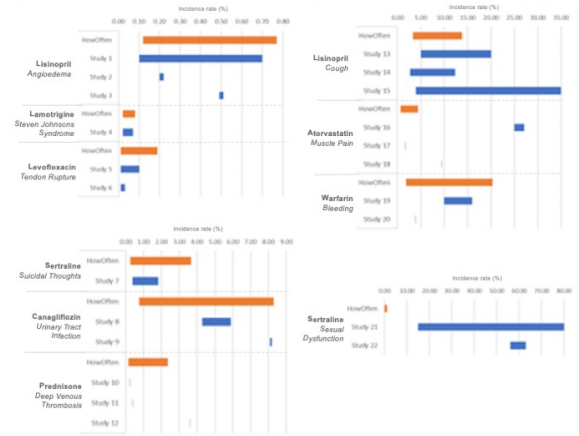


Figure 3. Comparison of calculated incidence rates with rates found in literature

Conclusions

- Large-scale incidence rate calculations may allow for evaluation of every possible ADR without manual curation and can be beneficial to clinicians in decision making.
- The calculated incidence proportions for known adverse effects for drugs were largely well-aligned with literature.
- Sensitive or subjective conditions may be underrepresented in coding data.
- Ongoing work is needed to better define outcome cohorts, explore different time-at-risks for incidence calculation, and validate results against other sources.

Elise Ruan, Karthik Natarajan, Ruijun Chen, Jungmi Han, Mark Velez, Taha Abdul-Basser, Edwin M. Cruz, Cindy Hsin-Yi Chen, Patrick Ryan, George Hripcsak. HowOften: Large Scale Incidence Rate Calculation of Every Side Effect for Every Drug. Proceedings of the 2018 ACM Conference on Health, Inference, and Learning (HIL '18), New York, NY, 2018. Copyright 2018 ACM 978-1-4503-5941-0/18/000001...\$15.00. DOI: 10.1145/3211211.3211212

Contact: ruan@ohdsi.org



#OHDSISocialShowcase This Week

WEDNESDAY

Agreement between measurement and diagnosis-based phenotype algorithms

(Azza Shoaibi, Gowtham Rao, Dmytro Dymshyts, Anna Ostropolets, Patrick Ryan)

Title: Agreement between measurement and diagnosis-based phenotype algorithms

PRESENTER: Azza Shoaibi

INTRO:

- Various types of clinical information, including diagnoses, medications, and procedures can be used to identify a specific clinical condition or event in observational data.
- Previous research indicates that the accuracy of phenotype algorithms can improve when multiple data types are incorporated.
- The aim of this paper is to compare diagnosis-based phenotype algorithms with those that are based on clinical measurements across five different clinical conditions in seven separate data sources.

METHODS:

- We selected five condition phenotypes: rhabdomyolysis, neutropenia, thrombocytopenia, pancytopenia, and end-stage renal disease
- We developed two different types of algorithms for all five conditions. Measurement-based phenotype algorithms used defined value thresholds for diagnostic markers. Diagnosis-based algorithms relied on the occurrence of at least one diagnosis.
- The cohorts were generated and evaluated across a range of data sources (listed in table 1); including 3 claims-based database, 1 general practitioner records and 2 health record (EHR). ALL data sources contained measurement data from outpatient and/or inpatient encounters with at least partial coverage

The utility of measurements vary by clinical condition and by data source. Measurements should be empirically evaluated when developing phenotype algorithms.

| Database Name | Thrombocytopenia | | | | | Neutropenia | | | | |
|-------------------------|------------------|--------|-------|---------|-----------|-------------|--------|-------|---------|-----------|
| | D Only | M Only | Both | overlap | N | D Only | M Only | Both | overlap | N |
| CCAE | 88.4% | 8.7% | 2.9% | 25.0% | 1,209,577 | 86.3% | 11.4% | 2.2% | 16.2% | 1,226,238 |
| IQVIA® Ambulatory EMR | 14.4% | 78.3% | 7.3% | 8.5% | 1,075,889 | 50.5% | 37.7% | 11.8% | 23.8% | 573,655 |
| IQVIA® LPD in Australia | 17.7% | 79.2% | 3.1% | 3.8% | 5,244 | 100.0% | 0 | 0 | NA | 1,805 |
| Medicare | 88.6% | 6.3% | 5.2% | 45.2% | 501,880 | 94.3% | 3.7% | 2.0% | 34.5% | 239,815 |
| Optum's DOD | 50.8% | 29.5% | 19.7% | 40.0% | 2,388,152 | 61.6% | 27.5% | 10.9% | 28.4% | 1,431,307 |
| Optum® EHR | 4.1% | 72.6% | 23.4% | 24.4% | 6,361,599 | 98.4% | 0.8% | 0.7% | 46.7% | 828,438 |
| Premier | 74.8% | 17.8% | 7.3% | 29.1% | 7,891,028 | 85.0% | 10.8% | 4.2% | 28.0% | 2,094,195 |

| Database Name | Pancytopenia | | | | | End-stage renal disease | | | | |
|-------------------------|--------------|--------|-------|---------|-----------|-------------------------|--------|-------|---------|-----------|
| | D Only | M Only | Both | overlap | N | D Only | M Only | Both | overlap | N |
| CCAE | 97.7% | 0.9% | 1.5% | 65.2% | 234,617 | 100.0% | 0.0% | <0.0% | NA | 250,770 |
| IQVIA® Ambulatory EMR | 47.4% | 31.1% | 21.6% | 50.0% | 83,100 | 97.6% | 0.9% | 1.5% | 62.5% | 253,075 |
| IQVIA® LPD in Australia | 100.0% | 0.0% | 0.0% | 84 | 84 | 39.6% | 57.4% | 3.0% | 5.0% | 197 |
| Medicare | 97.2% | 0.7% | 2.1% | 75.0% | 139,462 | 100.0% | 0.0% | 0 | NA | 223,704 |
| Optum's DOD | 85.4% | 5.2% | 9.4% | 64.4% | 406,264 | 99.0% | 0.1% | 0.9% | 90.0% | 527,668 |
| Optum® EHR | 24.5% | 18.5% | 57.0% | 75.5% | 433,971 | 100.0% | 0.0% | 0.0% | NA | 341,753 |
| Premier | 88.4% | 3.4% | 8.2% | 70.7% | 1,527,731 | 100.0% | 0.0% | 0.0% | NA | 2,700,277 |

| Database Name | Rhabdomyolysis | | | | | D Only: Proportion of patients identified by diagnosis-based phenotype only among total patients identified by either approach. | | | | |
|-------------------------|----------------|--------|------|---------|---------|---|--|--|--|--|
| | D Only | M Only | Both | overlap | N | M Only: Proportion of patients identified by measurement-based phenotype only among total patients identified by either approach. | | | | |
| CCAE | NA | NA | NA | NA | NA | Both: Proportion of patients identified by both approaches among total patients identified by either approach. | | | | |
| IQVIA® Ambulatory EMR | NA | NA | NA | NA | NA | Overlap: Proportion of patients identified in both approaches among those identified by the measurement-based algorithm. | | | | |
| IQVIA® LPD in Australia | NA | NA | NA | NA | NA | | | | | |
| Medicare | NA | NA | NA | NA | NA | | | | | |
| Optum's DOD | NA | NA | NA | NA | NA | | | | | |
| Optum® EHR | 99.5% | 0.2% | 0.3% | 60.0% | 161,254 | | | | | |
| Premier | 100.0% | 0.0% | 0.0% | NA | 879,951 | | | | | |

Table 1 -The overlap between the diagnosis based and measurement-based phenotype algorithms by phenotype in each data source.

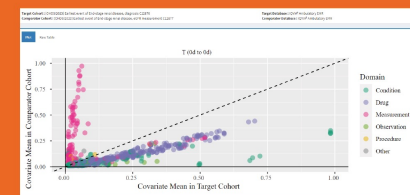


Figure 1: The covariate distribution among patients with thrombocytopenia identified through diagnosis compared to those identified through measurement on index date in IQVIA® LPD Australia

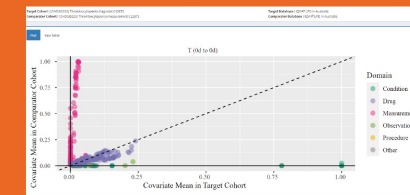


Figure 2: The covariate distribution among patients with End-stage renal disease identified through diagnosis compared to those identified through measurement on index date in IQVIA® Ambulatory EMR

METHODS, cont'd.

- For each condition we report on:
 1. the proportion of patients identified solely by each approach.
 2. the proportion identified by both approaches, and the overlap of identified patients.
- In addition, we conducted a comparison of covariate distributions among individuals who met each definition to evaluate the agreement in patient characteristics.
- We utilized the CohortDagnostic R package 3 to generate all results.

RESULTS

- A substantial heterogeneity in results was observed across data source and by condition.
- Creatine kinase (ck) and glomerular filtration rate (GFR) measurements identified a relatively small number of patients of rhabdomyolysis and end-stage renal disease respectively.
- A considerable number of patients was identified using measurement among the blood disorders.
- Among blood disorders, the characteristics of patients who met each definition were comparable (figure 1).

DISCUSSION

- Using measurements can significantly impact the sensitivity and specificity of the phenotype algorithm .
- We provide a framework for evaluating the utility of measurements for defining a phenotype within a given data source.

Azza Shoaibi, Gowtham Rao, Dmytro Dymshyts Anna Ostropolets, Patrick Ryan

Johns & Johnson Epidemiology OFFICE OF THE CHIEF MEDICAL OFFICER OHDSI





#OHDSISocialShowcase This Week

THURSDAY

Comparing Patient Self-Reported Symptoms with SNOMED/ICD-10-CM Codes at Primary Care Visits

(Victor M. Castro, Danielle M. Crookes, Vivian Gainer, Shawn N. Murphy, Justin Manjourides)

Comparing Patient Self-Reported Symptoms and SNOMED/ICD-10-CM Codes at Primary Care Visits

PRESENTER: Victor M. Castro

INTRO

- Measurement error in RWD study outcomes increases bias and compromises study validity.
- Many symptom outcomes in RWD studies rely on SNOMED and ICD-10-CM codes.
- In this study we aim to assess the sensitivity of clinician-recorded symptom SNOMED and ICD-10-CM codes compared to patient self-reports.

METHODS



- Over 400K primary care patients completed a symptom screening questionnaire prior to their visits between 2019 and 2021.
- We compared symptom self-report rates to clinician recorded symptoms in the EHR.

RESULTS

- Of the 15 symptoms evaluated, 13 were reported by patients much higher than recorded by clinicians in the EHR (Figure 1).
- Anxiety and depression were coded by physicians at a higher than reported by patients.

CONCLUSIONS

- Symptom outcomes defined by SNOMED or ICD-10-CM codes alone are likely to have poor sensitivity.

Primary care patients self-report symptoms up to twelve times more than clinicians record with structured codes in the EHR

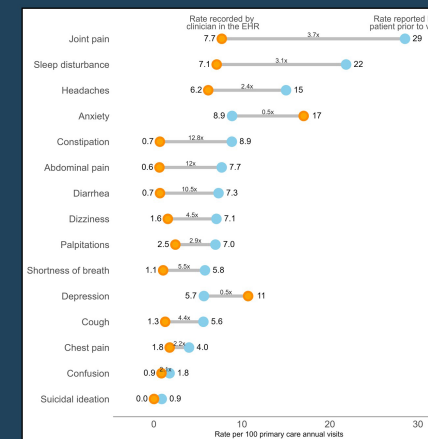


Figure 1. Case rates of symptom patient self-report (blue points) compared to clinician-recorded symptoms with ICD-10-CM or SNOMED codes (orange dots). Delta from self-report to clinician-recorded is also given above the line segment. Values below zero indicate clinicians record these symptoms more often the patients report at their visit.

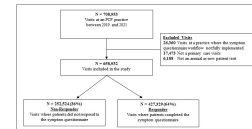


Figure 2. Study Flow Diagram

| Symptom | SNOMED code | ICD-10-CM code |
|---------------------|--|--|
| Joint Pain | 230.5000 | R52.1 |
| Sleep | 9909.0010, 9909.0020 | R57.1, R57.1A, R57.1B, R57.1C, R57.1D, R57.1E, R57.1F, R57.1G, R57.1H, R57.1I, R57.1J, R57.1K, R57.1L, R57.1M, R57.1N, R57.1O, R57.1P, R57.1Q, R57.1R, R57.1S, R57.1T, R57.1U, R57.1V, R57.1W, R57.1X, R57.1Y, R57.1Z |
| Sleep disturbance | 300.4000 | E34.1 |
| Headaches | 230.2000 | R51 |
| Constipation | 56.0000 | R54 |
| Diarrhea | 57.0000 | R56 |
| Abdominal pain | 53.0000 | R10 |
| Dizziness | 62.8200 | R52 |
| Palpitations | 62.8700 | R01 |
| Shortness of breath | 50.1000 | R05.1 |
| Depression | 300.0000, 300.0100, 300.0200, 300.0300, 300.0400, 300.0500, 300.0600, 300.0700, 300.0800, 300.0900, 300.1000, 300.1100, 300.1200, 300.1300, 300.1400, 300.1500, 300.1600, 300.1700, 300.1800, 300.1900, 300.2000, 300.2100, 300.2200, 300.2300, 300.2400, 300.2500, 300.2600, 300.2700, 300.2800, 300.2900, 300.3000, 300.3100, 300.3200, 300.3300, 300.3400, 300.3500, 300.3600, 300.3700, 300.3800, 300.3900, 300.4000, 300.4100, 300.4200, 300.4300, 300.4400, 300.4500, 300.4600, 300.4700, 300.4800, 300.4900, 300.5000 | F31.1, F31.2, F31.3, F31.4, F31.5, F31.6, F31.7, F31.8, F31.9, F31.0, F31.0A, F31.0B, F31.0C, F31.0D, F31.0E, F31.0F, F31.0G, F31.0H, F31.0I, F31.0J, F31.0K, F31.0L, F31.0M, F31.0N, F31.0O, F31.0P, F31.0Q, F31.0R, F31.0S, F31.0T, F31.0U, F31.0V, F31.0W, F31.0X, F31.0Y, F31.0Z |
| Cough | 46.0000 | R04 |
| Chest pain | 50.0000 | R01 |
| Confusion | 290.0000 | R40 |
| Suicidal ideation | 60.0000 | R62 |

Table 1. Symptom SNOMED and ICD-10-CM codes

| Symptom | SNOMED/ICD-10-CM | Prevalence | Sensitivity | Specificity | PPV | NPV |
|---------------------|--|------------|-------------|-------------|------|------|
| Joint Pain | 230.5000/R52.1 | 1.8% | 0.8% | 0.9% | 0.8% | 0.9% |
| Sleep disturbance | 300.4000/E34.1 | 0.2% | 0.1% | 0.2% | 0.1% | 0.2% |
| Headaches | 230.2000/R51 | 0.5% | 0.3% | 0.4% | 0.3% | 0.4% |
| Constipation | 56.0000/R54 | 0.3% | 0.2% | 0.3% | 0.2% | 0.3% |
| Diarrhea | 57.0000/R56 | 0.2% | 0.1% | 0.2% | 0.1% | 0.2% |
| Abdominal pain | 53.0000/R10 | 0.4% | 0.2% | 0.3% | 0.2% | 0.3% |
| Dizziness | 62.8200/R52 | 0.3% | 0.1% | 0.2% | 0.1% | 0.2% |
| Palpitations | 62.8700/R01 | 0.2% | 0.1% | 0.2% | 0.1% | 0.2% |
| Shortness of breath | 50.1000/R05.1 | 0.1% | 0.0% | 0.1% | 0.0% | 0.1% |
| Depression | 300.0000-300.5000/F31.0-300.5000/F31.9 | 1.2% | 0.5% | 0.6% | 0.5% | 0.6% |
| Cough | 46.0000/R04 | 0.2% | 0.1% | 0.2% | 0.1% | 0.2% |
| Chest pain | 50.0000/R01 | 0.3% | 0.1% | 0.2% | 0.1% | 0.2% |
| Confusion | 290.0000/R40 | 0.1% | 0.0% | 0.1% | 0.0% | 0.1% |
| Suicidal ideation | 60.0000/R62 | 0.0% | 0.0% | 0.0% | 0.0% | 0.0% |

Table 2. Concordance between patient self-reported symptoms and clinical record symptoms using SNOMED and ICD-10-CM codes

Victor M. Castro, MS (1, 2), Danielle M. Crookes, PhD (1), Vivian Gainer, MS (2), Shawn N. Murphy, MD, PhD (2), Justin Manjourides, PhD (1, 3)

(1) Bouvé College of Health Sciences, Northeastern University, Boston, MA. (2) Research Information Science and Computing, Mass General Brigham, Somerville, MA.

(3) OHDSI Center at The Roux Institute, Northeastern University, Portland, ME.



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#OHDSISocialShowcase This Week

FRIDAY

Using the Informatics
for Integrating Biology
and the Bedside
Platform to Query
OMOP Data in the
OHDSI Ecosystem

(**Jeffrey G. Klann**, Griffin M. Weber,
Michele Morris, Michael Mendis,
Diane Keogh, Shawn N. Murphy)



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environment variable driven deployment, new OHDSI apps, build from Git

```
docker-compose --profile default up -d
```

<https://github.com/OHDSI/Broadsea>



Opening: Junior Research Software Engineer, Tufts



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William Harvey, MD, MSc, FACR
Co-Director, Informatics and Tufts Medical Center CMIO

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Opening: Research Assistant, University of Oxford



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Job Details

Research Assistant in Health Data Sciences

Nuffield Department of Orthopaedics, Rheumatology and Musculoskeletal Sciences, Botnar Research Centre, Windmill Road, Oxford, OX3 7LD

We have an exciting opportunity for a Research Assistant in Health Data Sciences to join the Pharmacology and Device epidemiology research group led by Professor Daniel Prieto-Alhambra at the Botnar Research Centre, NDORMS, University of Oxford. The NDORMS Pharmacology and Device epidemiology research group is involved in a number of national and international studies exploring the conditions of use (adherence, compliance, off and on-label use) of a number of licensed drugs, devices, and vaccines for the prevention and treatment of human disease in 'real world' (routine practice) conditions.

As a Research Assistant in Health Data Sciences you will contribute to the programming of analytical pipelines for the analysis of routinely collected data mapped to the OMOP Common Data Model. You will analyse real world data to address regulatory questions related to the prevalence/incidence of disease, use of medicines/vaccines, and the risks or benefits of medicines/vaccines or devices. You will prepare analytical packages to run a number of pre-specified analyses, contribute to wider project planning, including ideas for new research projects and gather, analyse, and present scientific data from a variety of sources.

You will hold a relevant BA or MSc degree in Mathematics, Engineering, or a related field. Knowledge of medical statistics and experience analysing large datasets, experience in biostatistics and/or health data sciences and experience in the programming of R packages are essential. Experience in propensity scores, overlap weighting, inverse probability weighting and/or similar methods, expertise in pharmacology or vaccine epidemiology and experience of working with electronic medical records/routinely collected data are desirable.

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Contact Phone :
Pay Scale : STANDARD GRADE 6
Salary (£) : £32,332 - £38,205 p.a

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Closing Date & Time : 10-May-2024 12:00
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Opening: Biomedical Informatics Data Scientist at Stanford



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Clean, extract, transform and analyze various kinds of clinical data to create analysis-ready datasets that follow the FAIR (Findable, Accessible, Interoperable and Re-usable) principles. Partner with researchers and clinicians to enable effective and efficient use of Stanford Clinical data and resources for the advancement of research and the educational mission.

Postdoc/Senior Data Analyst Opening at WashU

The Zhang Lab at Washington University School of Medicine in St. Louis has **one postdoc/senior data analyst position** to work on **causal machine learning** and **responsible AI** for reliable real-world evidence generation.



PI: Linying Zhang, PhD

- More details at <https://linyingzhang.com>
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<https://linyingzhang.com/files/Postdoc.pdf>
 - Data analyst:
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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?





April 30: April Olympians Advancements



Clair Blacketer

Director, Observational Health Data Analytics
Janssen Research & Development



Melanie Philofsky

Senior Business Analyst and Project Manager
Odysseus Data Services, Inc.



The weekly OHDSI community call is held every Tuesday at 11 am ET.

Everybody is invited!

Links are sent out weekly and available at:
ohdsi.org/community-calls