



# Collaboration Update (It's Good News!)

**OHDSI Community Call**  
**Feb. 10, 2026 • 11 am ET**



# Three Stages of The Journey

**Where Have We Been?**

**Where Are We Now?**

**Where Are We Going?**



# OHDSI Shoutouts!



Congratulations to the team of **Charlotte Vercammen, Antje Heinrich, Christophe Lesimple, Alessia Paglialonga, Jan-Willem A Wasmann, and Mareike Buhl** on the recent publication of **Data standards in audiology: a mixed-methods exploration of community perspectives and implementation consideration** in the *International Journal of Audiology*.

INTERNATIONAL JOURNAL OF AUDIOLOGY  
<https://doi.org/10.1080/14992027.2026.2619921>



ORIGINAL ARTICLE

OPEN ACCESS

## Data standards in audiology: a mixed-methods exploration of community perspectives and implementation considerations

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### ABSTRACT

**Objective:** This study addresses conceptual issues around data standardisation in audiology, and outlines steps towards achieving it. It reports a survey of the computational audiology community on their current understanding, needs, and preferences concerning data standards. Based on survey findings and a panel discussion, recommendations are made concerning moving forward with standardisation in audiology.

**Design:** Mixed-methods: (1) review of existing standardisation efforts; (2) a survey of the computational audiology community; (3) expert panel discussion in a dedicated session at the 2024 Virtual Conference of Computational Audiology.

**Sample:** Survey: 82 members of the global community; Panel discussion: five experts.

**Results:** A prerequisite for any global audiology database are agreed data standards. Although many are familiar with the general idea, few know of existing initiatives, or have actively participated in them. Ninety percent of respondents expressed willingness to follow or contribute to standardisation efforts. The panel discussed relevant initiatives (e.g. OMOP, openEHR, Noah) and explored both challenges (around harmonisation) and opportunities (alignment with other medical fields and conversion among approaches).

**Conclusions:** Combining conceptual discussion with stakeholder views, the study offers guidance for implementing interoperable data standards in audiology. It highlights community support, key issues to address, and suggests paths for future work.

### ARTICLE HISTORY

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### KEYWORDS

Data standards;  
interoperability; online  
survey; audiological data;  
hearing loss management



# OHDSI Shoutouts!






Congratulations to the team of **Hanieh Razzaghi, Kimberley Dickinson, Kaleigh Wieand, Samuel Boss, Hunter Weidlich, Yungui Huang, Keith Morse, Sujan Kumar Mutyala, Jyothi Priya Alekapatti Nandagopal, Karthik Viswanathan, Christopher B Forrest, and L Charles Bailey** on the recent publication of **A multifaceted approach to advancing data quality and fitness standards in multi-institutional networks** in *JAMIA*.

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Research and Applications



## Research and Applications

### A multifaceted approach to advancing data quality and fitness standards in multi-institutional networks

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#### Abstract

**Objective:** To construct a data quality (DQ) system that incorporates combinations of methods to evaluate data characteristics and analytic fitness across research questions for multiple uses.

**Materials and Methods:** Drawing from experience of other data quality programs, network data extraction needs, and recurring study requirements, we developed 5 standards to guide development of a modular, multifaceted data quality system. These included annotation and documentation, ability to measure research readiness, reproducibility across networks, flexibility for the user, and interpretability to research and project teams. Implementation of checks based on these principles focused on reusability and interactive visualization of results.

**Results:** We identified 10 check types producing over 444 check applications and deployed them in 2 multi-institutional networks. Check types span structural conformance to a data model, utility for common research needs, and study-specific customization. All check types are customizable without dependencies between them. A dashboard visualizes results, permitting adjustments based on number of data sources, need for source masking, and the user's focus. All components can be applied as written to any data source using OMOP and are readily modified for other data models.

**Discussion:** We have extended previous work through our novel and multifaceted approach to data quality assessment, addressing needs in both network data improvement and research usage. We developed a capable and deployable system rather than tailoring to specific use cases.

**Conclusion:** Our novel DQ assessment system provides essential components for future standardization and collaboration to improve fitness of clinical data for intended use.

**Key words:** data quality; electronic health records; clinical research networks; multi-institutional research; learning health systems.



# OHDSI Shoutouts!



Congratulations to the team of **Meredith C B Adams, Robert W Hurley, Karsten Bartels, Matthew L Perkins, Cody Hudson, Umit Topaloglu, J Perren Cobb, Karin Reuter-Rice, Jacqueline C Stocking, Ashish K Khanna** on the recent publication of **Extending the Observational Medical Outcomes Partnership (OMOP) Common Data Model for Critical Care Medicine: A Framework for Standardizing Complex ICU Data Using the Society of Critical Care Medicine's Critical Care Data Dictionary (C2D2)** in *Critical Care Medicine*.

## Critical Care Medicine

February 2026 • Volume 54 • Number 2 • Pages 270-279

### CLINICAL INVESTIGATION

OPEN

## Extending the Observational Medical Outcomes Partnership (OMOP) Common Data Model for Critical Care Medicine: A Framework for Standardizing Complex ICU Data Using the Society of Critical Care Medicine's Critical Care Data Dictionary (C2D2)

**OBJECTIVES:** To evaluate the compatibility of the Society of Critical Care Medicine's (SCCM) Critical Care Data Dictionary (C2D2) with the Observational Medical Outcomes Partnership (OMOP) Common Data Model (CDM) and initiate a set of steps extending OMOP to accommodate specialized critical care data elements.

**DESIGN:** Systematic analysis and mapping study using a three-tiered semantic matching approach to demonstrate technical feasibility and identify fundamental challenges in critical care data standardization.

**SETTING:** Critical care medicine informatics research environment.

**SUBJECTS:** The SCCM's C2D2 elements.

**INTERVENTIONS:** None.

**MEASUREMENTS AND MAIN RESULTS:** We evaluated the compatibility of C2D2 clinical variables with the OMOP CDM using a three-tier classification system (full match, partial match, and no match). Our analysis of 226 C2D2 elements revealed that 49.6% of concepts had full OMOP equivalents, 46.4% required modification, and 4.0% had no suitable mapping. Key incompatibilities were identified in ventilator parameters, composite scoring systems, and advanced organ support documentation. A large language model-based semantic matching system yielded a precision of 59.5%, recall of 87.0%, and F1 score of 70.7% at an optimized similarity threshold of 0.90. These

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Ashish K. Khanna, MD, MS<sup>12,13</sup>





# OHDSI Shoutouts!



Congratulations to the team of **Md Fantacher Islam, Molly Douglas, Jarrod Mosier and Vignesh Subbian** on the recent publication of **Standardizing Data Elements for Implementation of ICU Liberation Bundle** in *Applied Clinical Informatics*.

Accepted Manuscript

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Accepted Date: 2026-02-02  
Accepted Manuscript online: 2026-02-03

## Applied Clinical Informatics

### Standardizing Data Elements for Implementation of ICU Liberation Bundle

Md Fantacher Islam, Molly Douglas, Jarrod Mosier, Vignesh Subbian.

Affiliations below.

DOI: 10.1055/a-2802-7458

**Please cite this article as:** Islam M, Douglas M, Mosier J et al. Standardizing Data Elements for Implementation of ICU Liberation Bundle. ACI 2026. doi: 10.1055/a-2802-7458

**Conflict of Interest:** The authors declare that they have no conflict of interest.

#### Abstract:

**Background and Significance:** Getting patients out of intensive care units (ICUs) is a major goal for acute care clinicians, as prolonged stays increase the risk of complications and strain critical resources such as staff, equipment, and beds. The ICU Liberation bundle or the ABCDEF (A-F) care bundle is an evidence-based framework for improving outcomes in critically ill patients by addressing pain, sedation, delirium, mobility, and family engagement. However, variability in documentation and lack of standardized data elements hinder effective implementation and evaluation of adherence to bundle components.

**Objectives:** This study aims to characterize data elements of the A-F liberation bundle using a large, single-center critical care database and to develop standardized bundle cards that map bundle components to controlled vocabularies.

**Methods:** We conducted a retrospective analysis of data elements related to A-F bundle using the MIMIC-IV database. Clinical concepts were mapped to standardized vocabularies and aligned with the OMOP common data model. Bundle cards were developed for each component to provide structured, accessible documentation of assessment tools, adherence criteria, and terminology mappings.

**Results:** Pain assessments were documented in over 11,000 patients, with a median of 23 assessments per day. Sedation levels for nearly 59,000 patients were evaluated, with 37.7% meeting Society of Critical Care Medicine (SCCM) adherence criteria. Delirium assessments followed standardized protocols incorporating RASS and CAM-ICU scores. Components E and F lacked formal compliance specifications; bundle cards for these components identified key activities and highlighted gaps in standardized vocabularies. Adherence analyses revealed variability likely due to non-standardized documentation practices.

**Conclusion:** We developed and validated six ICU Liberation Bundle cards that map bundle components to standardized vocabularies and common data models, enabling retrospective adherence evaluation in real-world data. These information resources promote consistent documentation, support interoperability, and provide a foundation for prospective monitoring to enhance bundle implementation in critical care.



# Three Stages of The Journey

**Where Have We Been?**

**Where Are We Now?**

**Where Are We Going?**



# 2026 Global Symposium

## 2026 OHDSI Global Symposium Call for Plenary Sessions

Symposium plenaries provide opportunities to share innovative, community-developed content to empower researchers to generate reliable real-world evidence. The community is currently seeking proposals for our #OHDSI2026 plenaries. These sessions will be 60 minutes in duration and must touch on at least two of following pillars of our community:

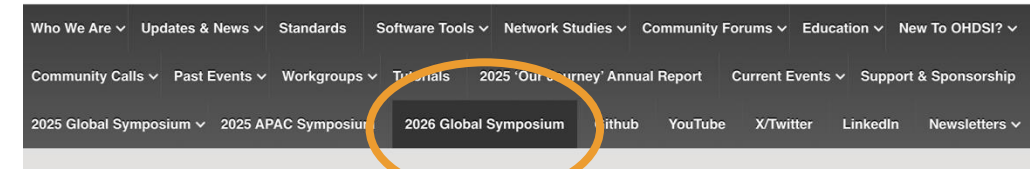
- Open community data standards
- Methodological research
- Open-source development
- Clinical applications

Plenary sessions must also involve three or more on-stage participants across at least two organizations. Sessions may include a combination of keynote talks, panel discussions, interactive activities, and more. We strongly encourage using multiple formats and synthesizing completed research, current perspectives and future calls-to-action to maximize community engagement.

The deadline for proposal submissions is January 30, 2026. Please use the link below to submit your proposal by answering the following questions:

- Name(s) of plenary session organizers:
- Your email address(es):
- Short (2,500 character max) description / abstract of your proposed session:
- Which pillars are you targeting:
- One sentence "pitch" of your session to excite the community:
- Names and roles of individuals who have tentatively agreed to participate in your session:

**Deadline to submit  
proposals for #OHDSI2026  
plenaries or tutorials is  
Feb. 20, 2026!**



## 2026 OHDSI Global Symposium

**Oct. 20-22 • New Brunswick, N.J. • Hyatt Regency Hotel**

## 2026 OHDSI Global Symposium Call for Tutorials

Tutorial sessions aim to deliver educational content, led by community members who wish to train our global collaborators on scientific, technical, and other skills that can support advancing OHDSI's mission and the effective use of real-world data and the generation and dissemination of reliable real-world evidence. Examples of prior tutorials offered are provided here: <https://www.ohdsi.org/tutorials>.

Tutorial sessions are 4 hours in duration. Registrants for your tutorial will be requested to pay a registration fee. The fees will be used to offset the costs of the symposium and other OHDSI expenses. Sessions may include a combination of talks, interactive activities, and more. We strongly encourage using multiple formats to maximize community engagement. Your session must include at least three people from at least two different organizations.

The deadline for tutorial proposal submissions is January 30, 2026. Please use the link below to submit your proposal by answering the following questions:

- Name(s) of tutorial session organizers:
- Your email address(es):
- Short (2,500 character) description / abstract of your proposed session:
- Names and roles of individuals who have tentatively agreed to participate in your session:





# 2026 Europe Symposium

The 2026 OHDSI Europe Symposium returns to Rotterdam next year and will be held **April 18-20.**

Registration is open on the **OHDSI & OHDSI Europe** web sites.

Symposium Agenda - Monday April 20, 2026		Location
8:00	<b>Registration and Coffee</b>	Queen's Lounge
9:00	<b>Welcome to OHDSI Europe</b> <u>Dr. Renske Los</u> , Department of Medical Informatics, Erasmus MC <u>Dr. Aniek Markus</u> , Department of Medical Informatics, Erasmus MC	Theatre
9:05	<b>Journey of OHDSI</b> <u>Prof. Peter Rijnbeek</u> , Chair Department of Medical Informatics, Erasmus MC	Theatre
9:30	<b>Collaborator Showcase - part 1</b> Moderated by <u>Dr. Egill Fridgeirsson</u> , Department of Medical Informatics, Erasmus MC	Theatre
10:00	<b>Speed networking</b>	Theatre
10:15	<b>Coffee Break &amp; posters National Nodes</b>	Queen's Lounge
11:15	<b>Collaborator Showcase - part 2</b> Moderated by <u>Dr. Egill Fridgeirsson</u> , Department of Medical Informatics, Erasmus MC	Theatre
11:45	<b>Dreaming about the OHDSI journey ahead</b> <u>Dr. Patrick Ryan</u> , Vice President, Observational Health Data Analytics, Johnson & Johnson <u>Dr. Renske Los</u> , Department of Medical Informatics, Erasmus MC	Theatre

12:15	<b>Lunch break &amp; networking &amp; posters/demo's</b> (Early investigator meeting - 13:00-13:45 Queen's Lounge)	La Fontaine & Odyssee Room
13:45	<b>From dreams to reality</b> <u>OHDSI Titan Award winners</u>	Theatre
14:30	<b>Propositions for collaboration from the National Nodes</b> <u>National Node leads</u>	Theatre
14:45	<b>Coffee break &amp; posters/demo's</b>	La Fontaine & Odyssee Room
16:15	<b>The OH Factor</b> <u>To be announced</u>	Theatre
17:00	<b>Closing</b>	Theatre
17:15	<b>Networking reception</b>	Queen's Lounge



# Columbia DBMI Summer School

## The 2026 Summer School in Observational Health Data Science & Informatics, AI, and Real World Evidence

June 22–26, 2026, Columbia Biomedical Informatics



The Columbia OHDSI Summer School provides health professionals, researchers, and industry practitioners with an immersive, hands-on training to working with real-world health data and generating real-world evidence (RWE). Participants will explore the types of healthcare data captured during routine clinical care—such as electronic health records and administrative claims—and learn how to standardize these data using the OMOP Common Data Model to support collaborative, distributed research as part of a data network.

Over the course of the week, participants will engage with three real-world analytic use cases:

- **Clinical characterization** – using descriptive epidemiology to study disease natural history and treatment patterns
- **Population-level estimation** – applying causal inference to assess drug safety and comparative effectiveness
- **Patient-level prediction** – leveraging machine learning for early disease detection and precision medicine

Participants will be guided through the full RWE study lifecycle: from designing observational studies tailored to each use case, to applying open-source tools from the [OHDSI community](https://www.ohdsi.org), and executing analyses across real-world data sources.

The curriculum combines foundational lectures on analytical methods with hands-on, interactive, faculty-led group exercises. In addition, participants will have dedicated time to develop and advance their own study concepts with personalized feedback and mentoring.







# #OHDSISocialShowcase This Week

## Monday

## Australian Health Data Evidence Network (AHDEN): Building a National Data Infrastructure for Standardised, Federated Health Data Research

(**Roger Ward**, Nicole Pratt, Graeme Hart, Ilan Meyers, Clair Sullivan, Blanca Gallego Luxan, Georgina Kennedy)

*The Australian Health Data Evidence Network (AHDEN): Building a National Data Infrastructure for Standardised, Federated Health Data Research*

PRESENTER: Nicole Pratt,  
Clinical and Health Sciences,  
University of South Australia

Co-authors: Roger Ward, Nicole Pratt,  
Graeme Hart, Ilan Mears, Clair Sullivan,  
Blanca Gallego Luxan, Georgina Kennedy

- Australia's healthcare system generates a vast amount of data, however, data systems are highly fragmented, with information captured across diverse and often incompatible systems (Figure 1).
- This lack of interoperability creates major barriers to the integration and analysis of health data at scale, limiting the nation's ability to conduct efficient, multi-centre research and generate timely, actionable evidence for health policy and clinical care.
- To address this critical need in the Australian context, The University of South Australia (UniSA), with co-investment from the Australian Research Data Commons (ARDC), has established the Australian Health Data Evidence Network (AHDEN) to support the implementation of the OMOP CDM across jurisdictional nodes in Australia (Figure 2).

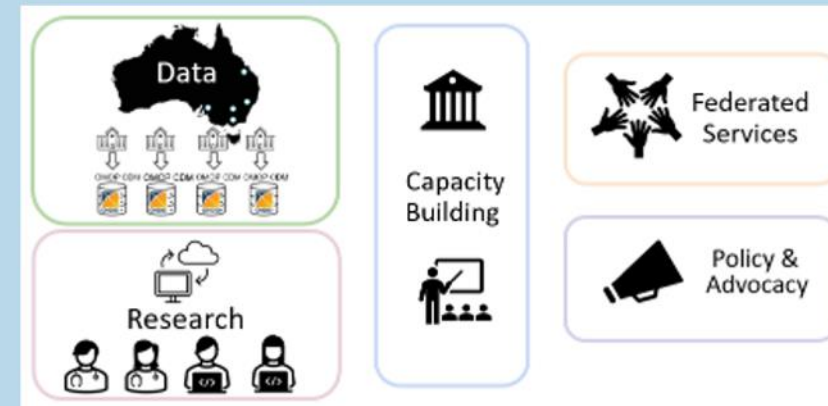
Figure 1: Public Hospital Electronic Medical Records System in Australia



<https://ardc.edu.au/project/australian-health-data-evidence-network-ahden/>

## AHDEN

### The Australia Health Data Evidence Network



Supporting the transformation of hospital-based Electronic Medical Record (EMR) data to the OMOP CDM to enable researchers to generate insights efficiently without compromising data security or privacy



ARDC is enabled by NCRIS



Figure 2: AHDEN strategy

DATA	Support the implementation of the OMOP CDM across jurisdictional nodes in Australia
RESEARCH	Enable enhanced collaboration across jurisdictional nodes to facilitate a coordinated approach to network research that delivers high priority and high-quality insights from existing EMR data
CAPACITY	Build capacity of researchers and health services in the harmonization, management and use of healthcare data to generate robust and reliable evidence
SERVICES	Facilitate the creation of shared resources, tools and reusable solutions to create efficiency for leveraging health data and to reduce research "waste"
POLICY	Generate policy consistency and coherence in approach to data harmonization and data governance strategies to ensure data security and "trust" in research





# #OHDSISocialShowcase This Week

## Tuesday

# Standardized use of PNGs/JPEGs for AI-Based Detection of Thyroid Eye Disease via Federated Learning

(**Michael Lau**, Vishwanath Prathikanti, Angela McCarthy, Ye Tian, Christopher Nielsen, Sina Gholami, Andrea Kossler, Eric Brown, Minhaj Alam, Lora Dagi Glass, Kaveri A. Thakoor)

### Standardized use of PNGs/JPEGs for AI-Based Detection of Thyroid Eye Disease via Federated Learning

**COLUMBIA ENGINEERING**  
The Fu Foundation School of Engineering and Applied Science

Wai Tak Lau<sup>1</sup>, Vishwanath Prathikanti<sup>2</sup>, Angela McCarthy<sup>3</sup>, Ye Tian<sup>3</sup>, Christopher Nielsen<sup>4</sup>, Sina Gholami<sup>5</sup>, Andrea Kossler<sup>4</sup>, Eric Brown<sup>6</sup>, Minhaj Alam<sup>6</sup>, Lora Dagi Glass<sup>7</sup>, Kaveri A. Thakoor<sup>1,3,7</sup>

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<sup>5</sup>Stanford University Department of Ophthalmology, Nashville, TN, USA  
<sup>6</sup>Department of Electrical and Computer Engineering, University of North Carolina, Charlotte, NC, USA

#### Motivation & Contribution

**Background:** Thyroid eye disease (TED), is an autoimmune disorder caused by the same antibody as autoimmune thyroid disorder, causing inflammation, swelling, excess and scarring<sup>1</sup>. Diagnosis based on clinical appearance, and radiologic imaging can be used for confirmation<sup>1</sup> preparing for surgery. Early diagnosis via Artificial Intelligence (AI) can help quickly identify TED.

**Motivation:** Hard to obtain a large and diverse dataset of facial images for training deep learning models. Federated training (FL) enables collaboration across different institutions while preserving data privacy.

Reduces inhomogeneity and manual errors arising from variations in data collection methods, formats, and quality that can threaten the development of robust AI models.

The Observational Health Data Sciences and Informatics (ODHSI) initiative is working to address this issue by promoting the use of the Observational Medical Outcomes Partnership (OMOP) Common Data Model (CDM) to standardize data across institutions.

Park and colleagues<sup>2</sup> developed robust image feature tables for the use of DICOM metadata, but it remains untested. With the development of new medical technology such as portable OCT devices, it is necessary to have robust data standardization for image file types outside of DICOM, such as PNGs and JPEGs.

**Contribution:** Our work represents the first effort to integrate ophthalmic imaging into the OMOP CDM and address the sensitivity of eye images. To test the tables and their use for PNGs and JPEGs, we leveraged the OMOP CDM and deep learning to predict Thyroid eye disease (TED) in a federated learning setting.

**In this work:**

- We propose FedTED, real-world federated learning (FL) with masked autoencoder (MAE) for TED
- We conduct extensive experimentation across sites to show utility of personalized vs global FL
- We systematically compare different widely adopted pretraining strategies, including on an open-source facial dataset to identify the most effective training regimes
- We developed an OMOP-like schema to generate sample queries across multiple institutions and build cohorts for site comparisons

#### Dataset

- Our study includes two clinical sites, **Columbia** and **Stanford**
- **Columbia** dataset contains more diverse control population, with more eyelid lesions and epiphora
- **Stanford** dataset contains a higher proportion of eyelid aging/malposition and asian population

Category	Attribute	TED		CONTROL	
		Columbia (N=316)	Stanford (N=106)	Columbia (N=316)	Stanford (N=106)
Sex	Female	131 (41%)	86 (80%)	123 (39%)	86 (80%)
	Male	22 (7%)	14 (13%)	39 (12%)	14 (13%)
Race	American Indian or Alaska Native	0 (0%)	0 (0%)	0 (0%)	0 (0%)
	Asian	8 (3%)	40 (38%)	12 (4%)	32 (30%)
	Black or African American	18 (6%)	2 (2%)	10 (3%)	0 (0%)
	Native Hawaiian or Other Pacific Islander	0 (0%)	4 (4%)	0 (0%)	0 (0%)
	White	39 (12%)	22 (21%)	86 (28%)	43 (40%)
	Other	27 (9%)	34 (32%)	23 (7%)	17 (16%)
	Unknown	15 (5%)	3 (3%)	23 (7%)	8 (8%)
Ethnicity	Hispanic or Latino	13 (4%)	16 (15%)	20 (6%)	12 (11%)
	Not Hispanic or Latino	85 (27%)	79 (74%)	107 (34%)	79 (74%)
Decoded	Decoded	131 (41%)	86 (80%)	123 (39%)	86 (80%)
	Not Decoded	0 (0%)	0 (0%)	0 (0%)	0 (0%)
Thyroid eye disease	Yes	131 (41%)	86 (80%)	123 (39%)	86 (80%)
	No	0 (0%)	0 (0%)	0 (0%)	0 (0%)
	Unknown	0 (0%)	0 (0%)	0 (0%)	0 (0%)
	Not Decoded	0 (0%)	0 (0%)	0 (0%)	0 (0%)
	Other	0 (0%)	0 (0%)	0 (0%)	0 (0%)

#### Methods & Results

##### Methods

**FL + AI:** Our FL model takes facial images as input, and predicts TED vs. control. We combine FL with masked autoencoder (MAE) pretraining to enhance model representations and robustness of our model. There are two client institutions, Columbia and Stanford, each with 291 and 200 images, respectively, with different data distributions. Federated Averaging is used as the main FL algorithm; both clients train for 10 epochs in each round, and we perform 10 rounds of FL training in total. ResNet-18 and ViT-B were used as image encoders for MAE and SimCLR pre-training, respectively.

**OMOP:** We extended the OMOP image feature and image occurrence tables to store relevant metadata extracted from PNG and JPEG files such as the image occurrence date, path, and features. In an effort to remain accessible to investigators at multiple sites with limited SQL knowledge, our data model was simplified from the full OMOP CDM. However, the relationships between the image extension tables were designed to mirror tables in the full CDM. Each image was linked to clinical features in the OMOP CDM via person id and visit occurrence id, enabling multimodal analyses. To simulate a federated environment, we deployed our data model at two institutions and used a federated learning framework (FLARE via Rhino Cloud Platform).

##### Results

**Average ROC Curves Across Folders - Columbia**

Model	AUC	95% CI
Global FL MAE	0.8547	0.8090 - 0.9004
Person FL MAE	0.8528	0.8090 - 0.8966
Global FL ResNet-18	0.8278	0.8090 - 0.8466
Person FL ResNet-18	0.8305	0.8131 - 0.8479
Global FL ViT	0.8482	0.8077 - 0.8887
Person FL ViT	0.8884	0.8577 - 0.9191
ResNet-18	0.8338	0.8187 - 0.8489
ViT-B	0.8524	0.8350 - 0.8698
MAE FL + Cu	0.8825	0.8537 - 0.9113
MAE FL + Cu	0.8825	0.8537 - 0.9113
SimCLR FL + Cu	0.8830	0.8530 - 0.9130
SimCLR FL + Cu	0.8830	0.8530 - 0.9130
Exemplar ResNet-18	0.8370	0.8090 - 0.8650

**Average ROC Curves Across Folders - Stanford**

Model	AUC	95% CI
Global FL MAE	0.9750	0.9329 - 1.0000
Person FL MAE	0.9870	0.9329 - 1.0000
Global FL ResNet-18	0.9330	0.9330 - 0.9330
Person FL ResNet-18	0.9600	0.9330 - 0.9870
Global FL ViT	0.9330	0.9330 - 0.9330
Person FL ViT	0.9750	0.9330 - 1.0000
ResNet-18	0.9330	0.9330 - 0.9330
ViT-B	0.9670	0.9330 - 1.0000
MAE FL + Cu	0.9660	0.9488 - 0.9832
MAE FL + Cu	0.9710	0.9388 - 1.0000
SimCLR FL + Cu	0.9580	0.9453 - 0.9707
SimCLR FL + Cu	0.9490	0.9300 - 0.9680
Exemplar ResNet-18	0.9670	0.9330 - 1.0000

**FL + AI:** We divided our experiment into two stages: local and FL. From our local experiments we found that the same methods applied to both sites yielded different results, with highest AUC across methods being 88.35% for the Columbia dataset and 97.17% for the Stanford dataset with personalized FL MAE. The difference in model performance shows that datasets have different difficulties, with the Columbia dataset having more diverse control conditions, further motivating the need of FL. Due to this reason, we focused on personalization in FL, fine tuning FL models to better fit the needs of each institution while benefiting from collaborative training. MAE combined with FL yielded the best results with 89.26% AUC and 98.70% AUC for Columbia and Stanford, respectively.

**OMOP:** To assess the practical utility of the schema, we simulated the process of building cohorts using the data collected. Cohorts built on various demographic data such as race, ethnicity, age and gender allowed us to understand generalizability based on parameters from different sites. Furthermore, by connecting diagnosis data from non-TED patients, we are able to understand how different diagnoses may affect parameters. The image extension tables afforded convenient storage of metadata and accessibility for retrieval.

#### Conclusion/Future directions

**FL + AI:** We developed and evaluated a framework for FL in TED, integrated self-supervised pretraining to improve model representation and robustness. The future vision for further improving not only TED detection but also models for detection of other diseases would require multimodal data, where data standardization would play a key role in enabling cross-institution collaboration. To that end, standardized image data table integration for multiple file types is essential.

**OMOP:** Through a full deployment of the OMOP schema, we can gain access to the full vocabulary list OHDSI provides. This is mainly beneficial for research questions pertaining to the use of certain medications and ingredients they may contain. Furthermore, the use of more specialized codes such as SNOMED, allows a new suite of research questions to be asked due to increased specificity compared to ICD billing codes.

**Future Directions:** For image feature tables, a new push in ophthalmology is in the definition of new vocabularies in imaging. The Radiological Society of North America developed a standardized vocabulary specific to radiology titled RadLex. RadLex vocabulary allows anatomy, imaging modalities, pathologies, and procedures to be coded and then mapped to vocabularies like SNOMED<sup>3</sup>. Park et al. designated certain fields in the image occurrence and feature tables to be populated with RadLex vocabulary specifically. With this and future projects involving eye images and AI analysis, the development of a lexicon for ophthalmology would not only serve to enable more structured reporting and interoperability, but there is already an existing infrastructure to map them into the OMOP database. This would represent a massive new step in image analysis and documentation.

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# #OHDSISocialShowcase This Week

## Wednesday

### Transforming Breast and Cervical Cancer Screening Data into the OMOP CDM: Early Implementation Insights from Senegal

(**Ousmane Diop**, Rachel Odhiambo, Abdoulaye Samba Diallo, Ousmane Diouf, Bakary Dembo Diatta, Mamadou Lamine Cissé, Fatou Mbaye, Yacine Amet Dia, Mame Sokhna Gueye, Aminata Dia, Abdou Padane, Nafissatou Leye, Seyni Ndiaye, Abdoulaye Leye Sarr, Maryline Aza-Gnandji, Mamadou Ndao, Astou Guèye, Steve Bicko Cygu, Samuel Iddi, Miranda Barasa, Agnes Kiragga, Moussa Sarr, Souleymane Mboup, Aminata Mboup)



#### Transforming Breast and Cervical Cancer Screening Data into the OMOP CDM: Integrating Clinical and Genomic Insights from Senegal

Ousmane Diop<sup>1</sup>, Rachel Odhiambo<sup>2</sup>, Mamadou Lamine Cissé<sup>2</sup>, Fatou Mbaye<sup>2</sup>, Yacine Amet Dia<sup>1</sup>, Abdou Padane<sup>1</sup>, Nafissatou Leye<sup>1</sup>, Steve Bicko Cygu<sup>3</sup>, Samuel Iddi<sup>3</sup>, Agnes Kiragga<sup>3</sup>, Moussa Sarr<sup>1</sup>, Souleymane Mboup<sup>1</sup>, Aminata Mboup<sup>1</sup>

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<sup>3</sup> African Population and Health Research Center (APHRC), Nairobi, Kenya

#### Background

- Breast and cervical cancers remain major health burdens in LMICs, where screening coverage is limited and late-stage diagnoses are common.
- In Senegal alone, recent estimates reported about 2 000 new cervical cancer cases and nearly 1 800 breast cancer cases in 2022, making them the two most frequent cancers among women.
- Annual Pink October campaigns have become essential for raising awareness and providing community-based screening, yet the large volumes of data generated during these initiatives are rarely exploited for research or long-term follow-up.
- The OMOP Common Data Model (CDM), developed by the OHDSI community, offers a standardized framework to harmonize such data and enable reproducible, collaborative research.
- This study represents the first implementation in West Africa mapping cancer screening data into OMOP CDM while also integrating genomic and microbiome results, demonstrating feasibility in a low-resource setting and opening opportunities for future multi-omic cancer research in the region.

#### Methods

- This study analyzed data from 491 women who participated in a community-based breast and cervical cancer screening campaign jointly organized by IRESSEF and Diamniadio Children's Hospital.
- Dataset included demographics, clinical outcomes, HPV PCR results, lesion status, comorbidities, and metagenomic profiles of the vaginal microbiome and virome.
- Data transformation followed the OHDSI ETL framework using White Rabbit (profiling), Rabbit-in-a-Hat (ETL design), and Usagi/Athena (concept mapping), with implementation in PostgreSQL.
- Variables originally collected in French were translated into English to ensure OMOP compatibility.
- Data were mapped to core OMOP domains (person, observation, condition\_occurrence, measurement, procedure\_occurrence...), and microbial/viral abundances were captured as quantitative measurements—representing one of the first attempts to integrate multi-omic data into OMOP CDM in West Africa.

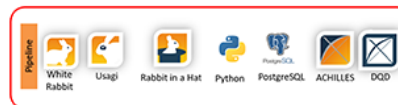


Figure 1: ETL pipeline

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#### Results

- The transformation achieved ~80% overall mapping to OMOP domains, with full coverage of demographic variables, 85% of clinical conditions, and 75% of laboratory and measurement data.
- Some specific variables, such as PCR-specific tests and genomic data required manual curation due to missing OMOP equivalents.
- Study population (n=491) ranged in age from 14–71 years (mean 35), with frequent comorbidities including diabetes and hypertension.
- Integration of microbiome and genomic data revealed distinct patterns: Gardnerella vaginalis, G. plovitii, and G. swidsinskii were enriched among women with precancerous lesions and positive HPV tests, while Lactobacillus variants were more common in women without lesions, suggesting a potential protective role.
- Oncogenic HPV types (e.g., Alphapapillomavirus 7) and co-infections such as Human gammaherpesvirus 4 were also detected, demonstrating OMOP CDM's ability to support integrative multi-omic analyses in this setting.

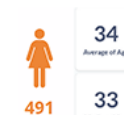


Figure 2: Summary statistics

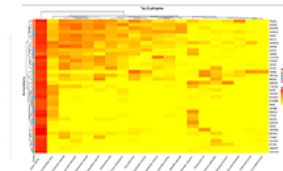


Figure 3: Heatmap pathogens

#### Conclusions

This initiative represents the first implementation in West Africa that integrates clinical, demographic, and genomic cancer screening data into the OMOP CDM. By aligning local data with international standards, it lays the foundation for reproducible and comparable analyses across settings and fosters opportunities for collaborative research.

Early insights from the microbiome analysis, particularly the associations between Gardnerella species, Lactobacillus variants, and precancerous cervical lesions, point to potential biomarkers that warrant further investigation. Ultimately, this work illustrates how data standardization can bridge clinical research and bioinformatics in LMICs, enabling the development of more precise, context-specific strategies for cancer prevention and control. By sharing ETL scripts and mappings, this work can serve as a reusable template for other African institutions adopting OMOP CDM.

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# #OHDSISocialShowcase This Week

## Thursday

# Thematic Classification of Articles Using Graph Representations

(Robert Barrett, Haeun Lee, Paul Nagy)

### Thematic Classification of Articles Using Graph Representations

PRESENTER: Robert Barrett

#### INTRO:

- Tracking community-driven impact on scientific literature is difficult
- Keyword-based retrieval of articles are often insufficient for topic identification
- Article retrieval through standard methods (APIs) are frequently limited to abstracts and metadata
- Our team aimed to identify scientific articles influenced by the OHDSI community, allowing for more flexible search criteria to capture related articles within a broader reach.

#### METHODS

- 704 known OHDSI pre-determined articles and 877 non-OHDSI-related articles retrieved from PubMed, spanning 2010 to 2025 were used for training/testing. An 80:20 test/train split was used for model development
- Preprocessing strategies evaluated:
  - Numeric
  - Removal of numeric tokens
  - Stemming (Porter algorithm)
  - Lemmatization (WordNet-based)
  - Stop-word removal (standard NLTK stopwords list)
- A graph was constructed using the DOI, author name, and journal name as nodes. Directed edges were created for cited articles, authorship, and the journal they were published
- PageRank, degree, related-author count, and citation overlap were calculated as features, in addition to extracted TF-IDF features from the abstract, title, and keywords
- XGBoost/Logistic Regression were compared for discriminatory power



## A system for classifying topic-related articles

Preprocessor	Pipeline	Precision	Recall	F1	Accuracy
Raw	TF-IDF	0.944	0.912	0.928	0.937
Raw	SciBERT	0.918	0.926	0.922	0.930
No numbers	TF-IDF	0.944	0.912	0.928	0.937
No numbers	SciBERT	0.915	0.918	0.916	0.925
Stemming	TF-IDF	0.950	0.909	0.929	0.938
Stemming	SciBERT	0.908	0.910	0.909	0.919
Lemma	TF-IDF	0.947	0.908	0.927	0.936
Lemma	SciBERT	0.913	0.922	0.917	0.926
Stopwords	TF-IDF	0.949	0.915	0.931	0.940
Stopwords	SciBERT	0.934	0.925	0.929	0.937

Table 1: Performance comparison of abstract only preprocessing strategies and pipelines

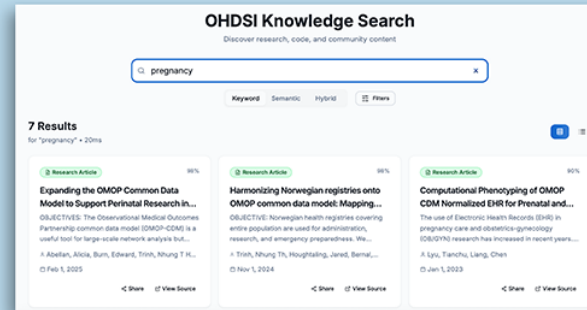


Figure 1: Visual demonstration of article retrieval post-classification



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#### RESULTS

- In evaluating preprocessing strategies, TFIDF + LogisticRegression pipeline yielded the strongest overall scores:
  - Precision: 0.949
  - Recall: 0.915
  - F1 score: 0.931
  - Accuracy: 0.940
- XGBoost model achieved the best performance:
  - Accuracy: 0.953
  - Precision (PPV): 0.986
  - Recall (Sensitivity): 0.920
  - F1-score: 0.952
  - Specificity: 0.987
  - (TN = 148, TP = 138, FP = 2, FN = 12)

Feature importance analysis (mean decrease in impurity) was primarily driven by:

- OHDSI-related-author count
- Citation-overlap
- TF-IDF-derived signals such as "OMOP" and "common data model"

#### CONCLUSION

- Demonstrates high specificity/sensitivity classification of OHDSI-related articles
- Improved value in classification through graph-derived features
- Vocabulary coverage and preprocessing methods are critical for model success
- Potential for topic agnostic process for classification

#### ACKNOWLEDGEMENT

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Barrett, Robert; Lee, Haeun; Nagy, Paul





# #OHDSISocialShowcase This Week

## Friday

# Replicating Alzheimer's Research using standardized phenotyping with the OMOP common data model imaging extension

(Gabriel Lucca de Oliveira Salvador, Jen Park, Teri Sippel Schmidt, Blake Dewey, Paul Nagy)

Replicating Alzheimer's Research using standardized phenotyping with the OMOP common data model imaging extension

PRESENTER: Jen Park

### INTRODUCTION:

- We implemented the OMOP Medical Imaging extension (MI-CDM) on the public ADNI research database (phase 4), harmonizing MRI metadata and clinical measures to support imaging-enhanced computable phenotypes.
- This is the first in our knowledge to replicate the published AD study to test reproducibility of imaging research.

### METHODS

- We extracted Digital Imaging and Communications in Medicine (DICOM) metadata around the imaging acquisition parameters for magnetic resonance imaging.
- This metadata was added to the OMOP CDM for DICOM vocabulary concepts. Demographic and neuropsychiatric data were transformed from the ADNI dataset and organized into OMOP CDM tables.
- Inclusion/exclusion criteria were established for the patients with T1-weighted brain MRI scans and neuropsychiatric inventory scores. The images meeting this criteria was used to obtain volumetric values of brain regions using the OpenMap-T1 ML segmentation algorithm.
- We replicated a published ADNI study correlating hippocampal volumes with varying degrees of dementia and AD.

## MI-CDM Enables Imaging-enhanced Computable Phenotypes, Advancing Reproducible Alzheimer's Disease Research.

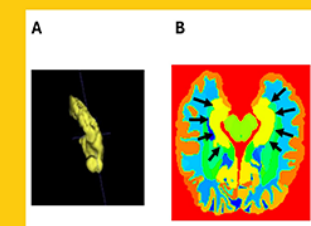


### CORRELATION ANALYSIS RESULTS

- We stratified by age, sex, and neurological condition to compare the right and left hippocampal volume from our algorithm and published work.
- Both models found reduced brain volume in AD groups compared to brain volumes with other neurological conditions.



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### BRAIN SEGMENTATION ALGORITHM COMPARISON

The figures illustrates the differences in segmentation patterns between the original study and OpenMap-T1.

A - Image from the study replicated showing the authors' semi-manual segmentation of one of the hippocampal formations.

B - Segmentation performed by OpenMap-T1 illustrating the head, body, and tail of the hippocampus (black arrows).

### RESULTS

- We extracted and loaded 545 studies (4,756 series, 14,816 images) from 289 patients, plus 4,152 demographic records and 88,819 neuropsychiatric scores mapped to the Measurement table.
- By using the DICOM metadata and neuropsychiatric inventory (NPI) on ATLAS, we found patients meeting the criteria of having done volumetric T1 series with NPI survey completed.
- 100% of DICOM series identified from the cohort definition were able to run OpenMAP algorithm.
- The results of brain segmentation algorithm was loaded to the MI-CDM Measurement table, so that the correlation analysis could be done on ATLAS characterization tab by neurological condition.
- Results showed hippocampal volume decline with dementia and age.
- Visual review revealed that OpenMap-T1 segments the entire hippocampus, unlike the original study's focus on the hippocampal head and limited brain areas.

### DISCUSSION

- We replicated a published imaging study that used the same research database (ADNI) with our internal brain segmentation algorithm.
- By organizing file-based database into standardized common data model, MI-CDM, we developed a system to replicate imaging research in computable and reproducible manner.
- Future studies should further evaluate reproducibility of imaging algorithms using MI-CDM.

Gabriel Salvador, Jen Park, Teri Schmidt, Blake Dewey, Paul Nagy





# 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?**





**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-2025](https://www.ohdsi.org/community-calls-2025)**