

'Phenotype Phebruary' Introduction

OHDSI Community Call Feb. 1, 2022 • 11 am ET

n ohdsi



Future OHDSI Community Calls

Date	Topic
Feb. 1	Introduction to Phenotype Phebruary
Feb. 8	Phenotype Phebruary Report, Workgroup Updates (Healthcare Systems, Open Source Community)
Feb. 15	Phenotype Phebruary Report, Workgroup Updates (CDM, Data Quality, Medical Imaging)
Feb. 22	Phenotype Phebruary Report #3, Workgroup Updates (ATLAS/WebAPI, Education)
Mar. 1	Breakout Sessions (Characterization, Estimation, Prediction)
Mar. 8	CDM Workshop (Part 1)
Mar. 15	CDM Workshop (Part 2)
Mar. 22	OHDSI Vocabulary Journey
Mar. 29	Reproducibility







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February 8 OHDSI Community Call



Open Source Community Workgroup UpdateAdam Black



Healthcare Systems Interest Group Update Melanie Philofsky



Phenotype Phebruary Update #1
Patrick Ryan



Three Stages of The Journey

Where Have We Been? Where Are We Now? Where Are We Going?









Congratulations to the team of Ross D. Williams, Aniek F. Markus, Cynthia Yang, Talita Duarte-Salles, Scott L. DuVall, Thomas Falconer, Jitendra Jonnagaddala, Chungsoo Kim, Yeunsook Rho, Andrew E. Williams, Amanda Alberga Machado, Min Ho An, María Aragón, Carlos Areia, Edward Burn, Young Hwa Choi, Iannis Drakos, Maria Tereza Fernandes Abrahão, Sergio Fernández-Bertolín, George Hripcsak, Benjamin Skov Kaas-Hansen, Prasanna L. Kandukuri, Jan A. Kors, Kristin Kostka, Siaw-Teng Liaw, Kristine E. Lynch, Gerardo Machnicki, Michael E. Matheny, Daniel Morales, Fredrik Nyberg, Rae Woong Park, Albert Prats-Uribe, Nicole Pratt, Gowtham Rao, Christian G. Reich, Marcela Rivera, Tom Seinen, Azza Shoaibi, Matthew E. Spotnitz, Ewout W. Steverberg, Marc A. Suchard, Seng Chan You, Lin Zhang, Lili Zhou, Patrick B. Ryan, Daniel Prieto-Alhambra, Jenna M. Reps and Peter R. Rijnbeek on the publication of "Seek COVER: using a disease proxy to rapidly develop and validate a personalized risk calculator for COVID-19 outcomes in an international network" in BMC Medical Research Methodology.



Abstract

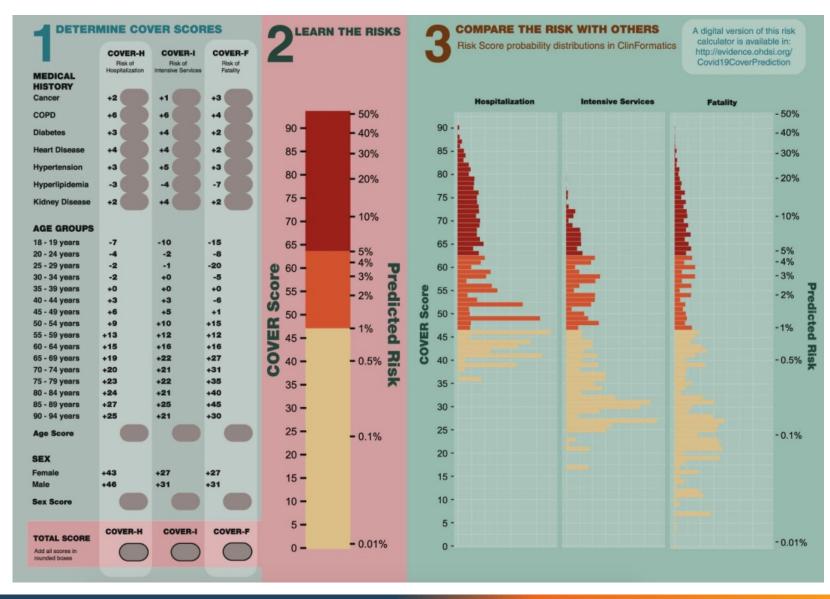
Background

We investigated whether we could use influenza data to develop prediction models for COVID-19 to increase the speed at which prediction models can reliably be developed and validated early in a pandemic. We developed COVID-19 Estimated Risk (COVER) scores that quantify a patient's risk of hospital admission with pneumonia (COVER-H), hospitalization with pneumonia requiring intensive services or death (COVER-I), or fatality (COVER-F) in the 30-days following COVID-19 diagnosis using historical data from patients with influenza or flulike symptoms and tested this in COVID-19 patients.













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Congratulations to the team of Seung-Hwa Lee, Jungchan Park, Rae Woong Park, Seo Jeong Shin, Jinseob Kim, Ji Dong Sung, Dae Jung Kim, and Kwangmo Yang on the publication of "Renin-Angiotensin-Aldosterone **System Inhibitors and Risk of Cancer: A Population-Based Cohort Study Using** a Common Data Model" in Diagnostics.





Article

Renin-Angiotensin-Aldosterone System Inhibitors and Risk of Cancer: A Population-Based Cohort Study Using a Common Data Model

Seung-Hwa Lee 1,2,† , Jungchan Park 3,4,†, Rae Woong Park 4,0, Seo Jeong Shin 4,0, Jinseob Kim 5, Ji Dong Sung 1,0, Dae Jung Kim 6,0 and Kwangmo Yang 4,7,*

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- Department of Biomedical Engineering, Seoul National University College of Medicine, Seoul 03080, Korea
- ³ Department of Anesthesiology and Pain Medicine, Samsung Medical Center, Sungkyunkwan University School of Medicine, Seoul 06351, Korea; j83.park@samsung.com
- Department of Biomedical Sciences, Ajou University Graduate School of Medicine, Suwon 16499, Korea; rwpark99@gmail.com (R.W.P.); lucid900921@naver.com (S.J.S.)
- Department of Epidemiology, School of Public Health, Seoul National University, Seoul 03080, Korea; jinseob2kim@gmail.com
- Department of Endocrinology and Metabolism, Ajou University School of Medicine, Suwon 16499, Korea; dikim@ajou.ac.kr.
- Oenter for Health Promotion, Samsung Medical Center, Sungkyunkwan University School of Medicine, Seoul 06351, Korea
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- † These authors contributed equally to this work.

Abstract: Studies have reported conflicting results on the association between the use of reninangiotensin-aldosterone system (RAAS) inhibitors and cancer development. We compared the incidence of cancer between patients using RAAS inhibitors and other antihypertensive drugs This retrospective observational cohort study used data from seven hospitals in Korea that were converted for use in the Observational Medical Outcomes Partnership Common Data Model. A total of 166,071 patients on antihypertensive therapy across the databases of the seven hospitals were divided into two groups according to the use of RAAS inhibitors. The primary outcome was the occurrence of cancer. A total of 166,071 patients across the databases of the seven hospitals was included in the final analysis; 26,650 (16%) were in the RAAS inhibitors group and 139,421 (84%) in the other antihypertensive drugs group. The meta-analysis of the whole cohort showed a lower incidence of cancer occurrence in the RAAS inhibitor group (9.90 vs. 13.28 per 1000 person years; HR, 0.81; 95% confidence interval [CI], 0.75-0.88). After propensity-score matching, the RAAS inhibitor group consistently showed a lower incidence of cancer (9.90 vs. 13.28 per 1000 person years; HR, 0.86; 95% CI, 0.81-0.91). The patients using RAAS inhibitors showed a lower incidence of cancer compared with those using other antihypertensive drugs. These findings support the association between the use of RAAS inhibitors and cancer occurrence.

Keywords: renin-angiotensin-aldosterone system inhibitors; cancer occurrence



Citation: Lee, S.-H.; Park, J.; Park, R.W.; Shin, S.J.; Kim, J.; Sung, J.D.; Kim, D.J.; Yang, K.

Renin-Angiotensin-Aldosterone System Inhibitors and Risk of Cancer: A Population-Based Cohort Study Using a Common Data Model. Diagnostics 2022, 12, 263. https:// doi.org/10.3390/diagnostics12020263

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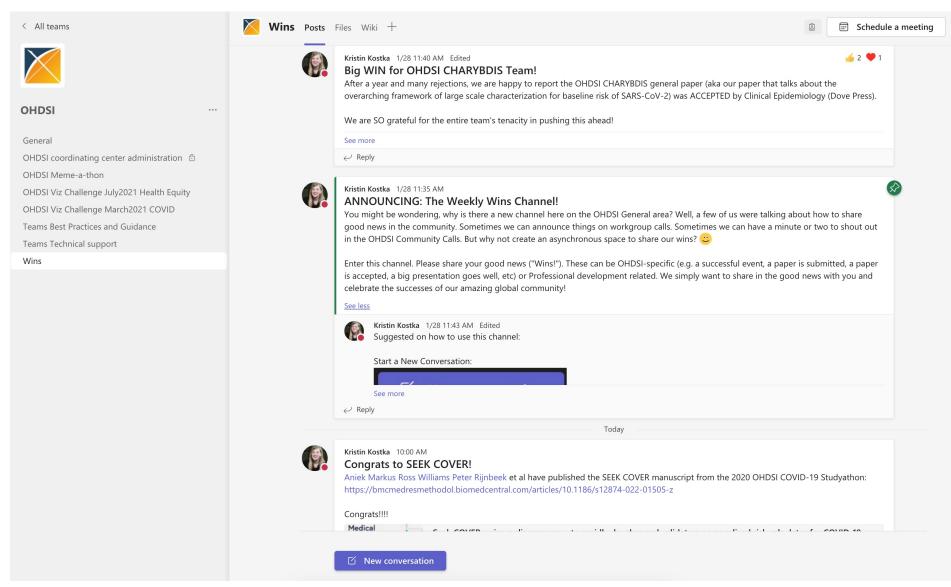
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Any shoutouts from the community? Please share and help promote and celebrate OHDSI work!

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 (Eastern Hemi)
Wednesday	9 am	ATLAS/WebAPI
Wednesday	10 am	FHIR and OMOP Data Quality Measurements Subgroup (Zoom)
Wednesday	4 pm	FHIR and OMOP Data Model Harmonization Subgroup (Zoom)
Thursday	8 am	Psychiatry
Thursday	12 pm	Patient-Level Prediction/Population-Level Estimation (Western Hemi)
Thursday	12 pm	FHIR and OMOP Oncology Subgroup
Thursday	3 pm	FHIR and OMOP Terminologies Subgroup (Zoom)
Friday	10:30 am	Clinical Trials
Monday	10 am	GIS-Geographic Information System
Tuesday	9 am	OMOP CDM Oncology Genomic Subgroup

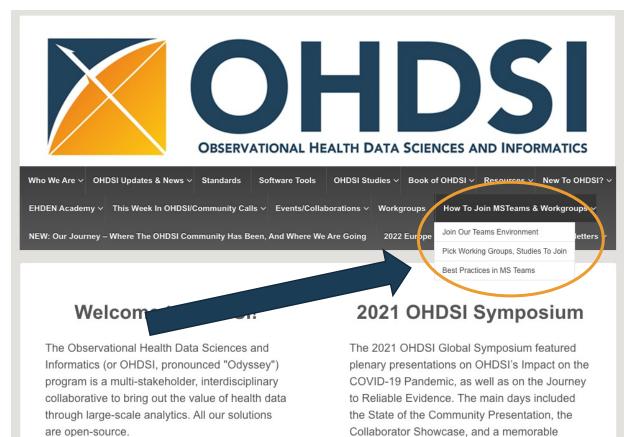
www.ohdsi.org/upcoming-working-group-calls



in ohdsi



Get Access To Different Teams/WGs/Chapters



ATLAS Psychiatry Clinical Trials Registry (formerly UK Biobank) Common Data Model Surgery and Perioperative Medicine Data Quality Dashboard Development Vaccine Evidence Early-stage Researchers Vaccine Vocabulary Education Work Group FHIR and OMOP 6. Select the chapter(s) you want to join Geographic Information System (GIS) Africa HADES Health Analytics Data-to-Evidence Suite Australia Healthcare Systems Interest Group (formerly EHR) China Europe Health Equity Japan Latin America Korea Medical Devices Singapore Medical Imaging Taiwan Natural Language Processing OHDSI APAC 7. Select the studies you want to join OHDSI APAC Steering Committee HERA-Health Equity Research Assessment OHDSI Steering Committee PIONEER for Prostate Cancer (study-a-thon ended) Oncology SCYLLA (SARS-Cov-2 Large-scale Longitudinal Analyses) Open-source Community Phenotype Development and Evaluation

Population-Level Effect Estimation / Patient-Level Prediction

5. Select the workgroups you want to join (you can refer to the WIKI for work group objectives

www.ohdsi.org/web/wiki/doku.php?id=projects:overview)

OHDSI has established an international network

of researchers and observational health

harrand at Calumbia I Inicamity

databases with a central coordinating center



Closing Ceremony that focused on OHDSI's

There were also a pair of full-day activities.

including the first OLIDCI Depreducibility

work through the perspective of a patient.



Get Access To Different Teams/WGs/Chapters



Select the workgroups you want to join (you can re www.ohdsi.org/web/wiki/doku.php?id=projects:ove	
ATLAS	
Clinical Trials	Psychiatry
Common Data Model	Registry (formerly UK Biobank)
	Surgery and Perioperative Medicine
Data Quality Dashboard Development	☐ Vaccine Evidence
Early-stage Researchers	☐ Vaccine Vocabulary
Education Work Group	
FHIR and OMOP	6. Select the chapter(s) you want to join
Geographic Information System (GIS)	Africa
HADES Health Analytics Data-to-Evidence Suite	Australia
Healthcare Systems Interest Group (formerly EHR)	China
Health Equity	Europe
Latin America	Japan
Medical Devices	☐ Korea
Medical Imaging	Singapore
Natural Language Processing	Taiwan
OHDSI APAC	
	7. Select the studies you want to join
OHDSI APAC Steering Committee	HERA-Health Equity Research Assessment
OHDSI Steering Committee	☐ PIONEER for Prostate Cancer (study-a-thon ended)
Oncology	SCYLLA (SARS-Cov-2 Large-scale Longitudinal Analyses)
Open-source Community	
Phenotype Development and Evaluation	
Population-Level Effect Estimation / Patient-Level Prediction	n





New Workgroups Page on OHDSI.org



Education

Geographic Information System (GIS)

HADES (Health Analytics Data-to-Evidence Suite)

















Latin America













OHDSI Asia-Pacific (APAC)

Oncology Lead: Mui Van Zandt



Phenotype Development & Evaluation









Population-Level Estimation

Patient-Level Prediction

Registry (formerly UK Biobank) Psychiatry









Vaccine Vocabulary

Women of OHDSI





ohdsi.org/ohdsi-workgroups



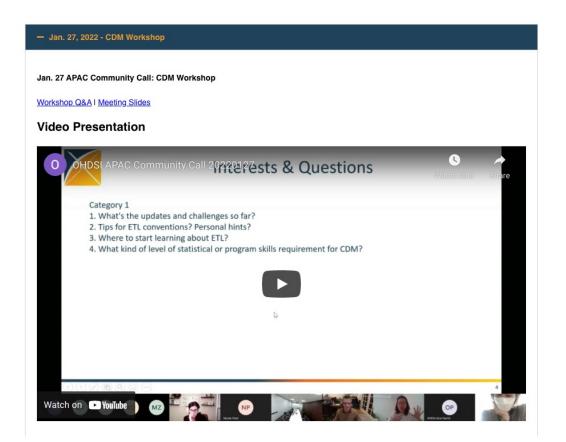
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Jan. 27 APAC Call: CDM Workshop

CDM





APAC Workshop 26 January 2022

- 1. What's the updates and challenges so far?
 - a. We recently released CDM v5.4 (http://ohdsi.github.io/CommonDataModel/cdm54.html). So far there have been many challenges but the main one we have been struggling with is how fast to update the CDM. We want to be responsive to the needs of the community but with an understanding that any change we make to the model has huge impacts down the line.
- Tips for ETL conventions? Personal hints?
 - a. Keep it as simple as possible at first. Many people want to use complex logic to make their data fit the model but I have found that the simplest approach is usually best because it makes it easier to error check later.
- Where to start learning about ETL?
 - a. The EHDEN academy is a great place to start! https://academy.ehden.eu/
- 4. What kind of level of statistical or program skills requirement for CDM?
 - a. This one depends on what you are planning to do. If you are responsible for developing the ETL you need strong SQL skills or a good understanding of the database management system you will be using for the conversion. If you are the one designing the ETL then only some SQL skills will be enough. If you plan on designing studies using existing R packages then strong R skills are required. However, if you plan on only running studies or packages then you only need some R skills, you don't need to be an expert.
- 5. Is there any best practices for CDM mapping for pediatric survey questions?
 - I am not sure about pediatric surveys but, in general, I usually use the OBSERVATION table for survey questions. You can set the

ohdsi.org/apac



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New Dates For The 2022 European Symposium



www.ohdsi-europe.org/symposium-2022









Securing OHDSI on AWS for HIPAA and Research Data Management Compliance

Michael Lubke¹, Tapati Mazumdar¹, Murat Sincan, M.D.^{1*}, Catherine Hajek, M.D.^{1*} 1. Sanford Imagenetics, Sioux Falls, South Dakota

*. Equally Contributed

Requirements for HIPAA and research protocol compliance should be at the core of a healthcare organization's implementation practices when storing and analyzing healthcare data in the cloud. Every utilized service must be thoroughly vetted to ensure security best practices and data access permissions are being established properly This process results in a secure environment where organizations can embrace utilizing the cloud for performing analysis on observational health data.

Background

With cloud services and cloud storage adoption becoming more widespread in the healthcare industry, the need for securing these environments and the health data within is of utmost importance. The OHDSI on AWS project provides an enterprise-level solution for enabling advanced analytics and outcome prediction on this observational health data with the ultimate goal of improving population and individual health. With this project comes the need for developing customized cloud solutions to process various research study data files, and for that reason it is important that every step of the data processing conforms to HIPAA requirements as well as the data management policies outlined in each research study protocol. Enforcing strict security measures in the cloud environment is not only performed to mitigate the financial risks associated with noncompliance, but it is also done to implement additional layers of security with the purpose of preventing the patient health data from being compromised in a data breach.

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The default OHDSI on AWS environment was analyzed and the following were identified as areas needing additional security:

- · Restrict access to OHDSI instance to the Sanford trusted network
- · Ensure all health data is encrypted at rest and in-transit
- Restrict access to study data to applicable persons and systems outlined in research data management plan

A site-to-site VPN solution was implemented that creates a tunnel between the on-premises trusted network at Sanford Health and the Virtual Private Cloud (VPC) on AWS. This combined with leveraging the AWS Certificate Manager to handle the generation and application of SSL certificates on the OHDSI on AWS environment enforce HIPAA compliance requirements by ensuring that the communication between all system components within the VPC are encrypted in-transit.

Data access policies were configured for each AWS S3 bucket restricting accessing the data to individual members of each research study Furthermore, in order to meet HIPAA requirements, all S3 buckets are encrypted at-rest. Policies were created for AWS System Roles that allow the components of the custom OMOP CDM ETL pipeline to decrypt the contents of the S3 buckets. The transformed research data is then inserted directly into the corresponding study schema on Amazon Redshift via a JDBC connection, which ensures that the data is once again encrypted in-transit.

Results

Figure 1 illustrates the site-to-site VPN solution that was established between the AWS VPC and a routing device at the Sanford data center. This configuration resulted in the default public-facing applications in the OHDSI on AWS environment stack (Atlas, RStudio Jupyter) being accessible via the Sanford trusted network only.



Figure 1: Site-to-Site VPN

Figure 2 shows the final state of the ETL pipeline after establishing data access permissions per HIPAA and data management policy requirements. The steps are summarized as follows:

- The research study honest broker uploads de-identified study data into an access-managed AWS S3 bucket, accessible only to those listed in the study's data management policy
- · New data event triggers the lambda function, which first decrypts the contents of the bucket, then stores the resulting processed data into a subsequent encrypted access-managed S3 bucket.
- AWS Glue Crawler automatically decrypts the processed data files and stores the contents in an AWS Glue Data Catalog
- · AWS Glue ETL scripts conform the data into the OMOP CDM and insert the records directly into the corresponding study's schema in Amazon Redshift via an encrypted JDBC connection.



Conclusions

It is imperative that every interaction with patient health data hosted in a cloud environment is thoroughly analyzed to ensure that HIPAA and research study compliance requirements are met. Each component added to the cloud environment must be carefully implemented to ensure that the health data is secured regardless of where it exists. This process can be daunting to undertake when an entire data processing pipeline has already been established without initially giving any thought to access controls and encryption practices. For that reason, employing best practices in accordance with HIPAA and data management compliance requirements at the onset of development can provide a safeguard for the cloud environment as well as the data stored within.

- AWS Site-to-Site User Guide [Internet]. Amazon Web Services; 2011 Sep 29 [updated 2020 Oct 29 cited 2021 Jun 16]. Available from:
- Internet1, Amazon Web Services: 2016 Oct [updated 2020 Oct 29: cited 2021 Jun 16], Available
- Guidance on HIPAA & Cloud Computing [Internet], U.S. Department of Health and Human Services [updated 2020 Nov 24; cited 2021 Jun 16], Available from:

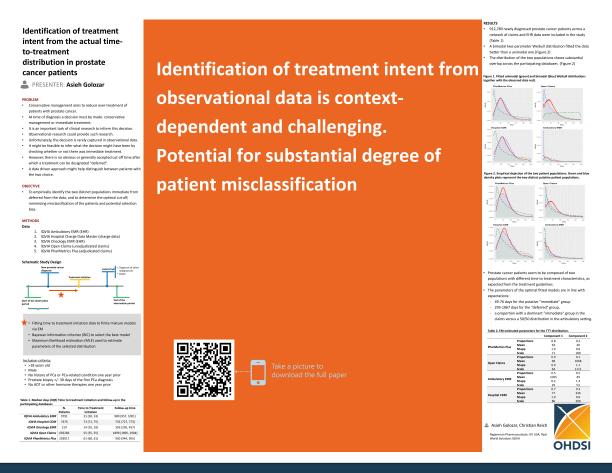
MONDAY

Securing OHDSI on AWS for HIPAA and Research Data Management Compliance

Author: Michael Lubke, Tapati Mazumdar, Murat Sincan, Catherine Hajek

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TUESDAY

Identification of treatment intent from the actual time-to-treatment distribution in prostate cancer patients

Authors: Asieh Golozar, Christian Reich







The concept of anchoring in vaccine safety studies and its influence on baseline patient characteristics and study estimates

Anna Ostropolets¹, Xintong Li², Rupa Makadia³, Gowtham Rao³, Peter R. Rijnbeek⁴, Talita Duarte-Salles⁵, Anthony G. Sena^{3,4}, Azza Shaoibi³, Marc A. Suchard^{6,7}, Patrick B. Ryan^{1,3}, Daniel Prieto-Alhambra², George Hripcsak^{1,8}

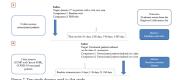


The cohort method is one of the most common methods in comparative effectiveness and safety studies. In a cohort study, we compare rates of events during time-at-risk in target and comparator groups. Such rates are, therefore, dependent on the choice of starting point for time-at-risk or, as we call it, anchoring. Choice of anchoring may influence both the rates of observed outcomes and baseline patient characteristics, which are subsequently used in propensity score models or outcome models



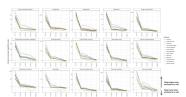
When the cohort method is applied in vaccine safety studies, a vaccinated population is compared to an unvaccinated population. The latter cohort does not have a clear index date, which is left up to researchers' judgment. In this study, we investigate how the choice of the index date (anchor) influences patien Figure 1. What is the index date for the unexposed baseline characteristics and baseline incidence

We investigated the influence of anchoring unvaccinated population on incidence rates of 15 adverse events occurring during different time-at-risk windows (A). Additionally, we investigated its impact on baseline patient characteristics in unvaccinated and vaccinated populations (B).



For the first study, we modified the study design described in the paper by Li et al. (2021). Briefly, we used 12 data sources to study incidence rates of 15 adverse events of special interest in unvaccinated population in 2017-2020 in a number of time-at-risk intervals. We calculated incidence rate ratios (IRR) of incidence rates in two pairs of cohorts with different index dates ("anchors"); a visit and a well visit, defined as a visit associated with preventive visit CPT4 codes. We performed random-effect meta-analysis across data source In the award study, we compared baseline characteristics of patients vaccinated with COVID-19 and unvaccinated patients in CUMC and Optum EHR.

b) a visit matched to the index date of one of the target group. Additionally, each target and comparator groups were matched on age and sex.



Influence of anchoring on baseline characteristics

When looking at the baseline characteristics, unva oonulation had more events (measurements, conditions, population regardless of the anchoring event (Figure 4). Similarly, the effect attenuated with increased lookback

unvaccinated population. In patients vaccinated with an

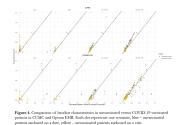
Influence of anchoring on the background incidence rate.

Incidence rates of all adverse events across all data source and all conditions were highly sensitive to the choice of anchoring. For a short time-at-risk (0-1 day) anchoring on a visit was associated with up to a 100-fold increase in incidence when compared to anchoring on January 1st (pooled IRR 26.8 (95% CI 21.9-32.8)). Acute conditions such as anaphylaxi were impacted the most (pooled IRR 47.6 (95% CI 32.8 69.1)). The effect was attenuated for longer times at risk (Figure 3) but was still present. For example, for 1 - 28 day window, pooled IRR was 1.4 (95% CI 1.3-1.5). We observed similar trends for anchoring on a well visit with the pooled IRR of 1.21 (95% CI 1.11-1.31). Overall, cates with several conditions (such as Bell's palsy or

1-42 days, 1-90 days and 1-365 days

window but was still present.

Vaccinated patients had fewer lab tests (such as body weight, blood pressure or respiratory rate), co-morbiditi (diabetes, hyperlipidemia, dyspnea etc.) and visits on day (when compared to a random date or a random visit in influenza vaccine (not shown here) compared to patient not vaccinated with influenza vaccine the opposite trend was observed: the former had more lab tests and comorbidities on day 0 than unvaccinated patients.



narcolepsy) being highly sensitive

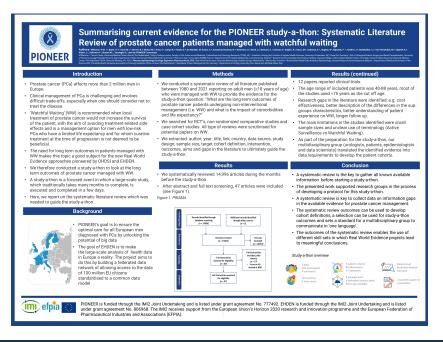
- 2. It is crucial to select an anchoring that represents the target index date best based on the knowledge of the target (e.g. vaccination settings) empirical comparison of multiple options.

Columbia University Medical Center, New York, NY, USA; Centre for Statistics in Medicine, NDORMS, University of Oxford, Oxford, United Kinodom; Janssen Research and Development, Titusville, NI, USA; Department of Medical Informatics, Enamus University Medical Center Rotterdam, The Notherlands- Production Institut University of California, Los Angeles, CA, USA; Department of Human Genetics, Devid Geffer School of Holder Health, University of California, Los Angeles, CA, USA; Department of Human Genetics, Only Geffer School of Health, University of California, Los Angeles, CA, USA; Department of Human Genetics, Devid Geffer School of Health, University of California, Los Angeles, CA, USA; Department of Human Genetics, Devid Geffer School of Health, University of California, Los Angeles, CA, USA; New York New Pools Productions Hospital, New York, NY, USA;

WEDNESDAY

The concept of anchoring in observational study design and its influence Authors: Anna Ostropolets, Talita Duarte-Salles, Xintong Li, Rupa Makadia, Daniel Prieto-Alhambra, Gowtham Rao, Peter R. Rijnbeek, Martijn Schuemie, Anthony G. Sena, Azza Shaoibi, Marc A. Suchard, Patrick B. Ryan, George Hripcsak





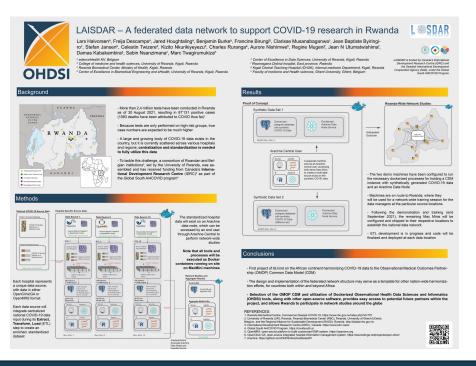
THURSDAY

Summarizing current evidence for the PIONEER study-a-thon: Systematic Literature Review of prostate cancer patients managed with watchful waiting

Authors: Peter-Paul Willemse, Katharina Beyer, Muhammed Imran Omar, Ronald Herrera, Megan Molnar, Isabella Greco, Riccardo Campi, Samuel Fatoba, Bertrand De Meulder; Susan Evans, Nazanin Zounemat Kermani, Sebastiaan Remmers, Christian Reich, Shilpa Ratwani, Asieh Golozar Robert Snijder, Mauro Gacci, Ariel Achtman, Nigel Hughes, Peter Rijnbeek, Emma Smith; Carl Steinbeirer, Mieke Van Hemelrijck; Anders Bjartell, James N'dow, Alex Asiimwe, Monique Roobol, Giorgio Gandaglia







FRIDAY

LAISDAR - A federated data network to support COVID-19 research in Rwanda Authors: Lars Halvorsen, Freija Descamps, Jared Houghtaling, Benjamin Burke, Francine Birungi, Clarisse Musanabaganwa, Jean Baptiste Byiringiro, Stefan Jansen, Celestin Twizere, Kizito Nkurikiyeyezu, Charles Ruranga, Aurore Nishimwe, Regina Mugeni, Jean N Utumatwishima, Damas Kabakambira, Sabin Nsanzimana, Marc Twagirumukizaa







Where Are We Going?

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







Welcome To OHDSI Newcomers

Are there any people new to the OHDSI community call who would like to introduce themselves?

Please raise your hand, and we will call on three people.







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