

APAC Scientific Forum

November 2, 2023



Agenda

- Medicinal Cannabis Study Overview by Christine Hallinan
- Data Quality of OHDSI APAC: CDM Inspection Study Update & Deep Learning Comparison Study Overview by Chungsoo Kim
- December Meeting Preview
- Support Areas Follow-up Survey Extension

Medicinal Cannabis in Australia

Capturing Evidence from Australia's Medicinal Cannabis Natural Experiment

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Acknowledgment:

Prof Yvonne Bonomo MBBS FRACP PhD FAChAM Department of Addiction Medicine, St Vincent's Hospital Melbourne Departments of General Practice and Medicine, The University of Melbourne



26-29 October 2023 Sydney, Anstralia



Prescribers–What do GP's and physicians say?...

'...it doesn't work for everybody and for some people it has no benefit whatsoever... for some people, it has terrible side effects, but I believe that users are best able to work with their doctors if they think it is a benefit to them.

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'That's our challenge now—to re-think our legislative structure and how we manage problems so that we can reduce the induced indirect harm... the legal harms... (associated with) increasing access, availability, advertising, promotion, and cost incentives to increase consumption... That's our challenge. But who's going to lead this? I seem to be—not a lone voice, but I feel alone in that message—I am sending'

'The problem—I think that people—general public will have their views about it being useful for x and y, because that's already out there. I think the medical profession, hopefully if the data gets better, will have a better idea about what it actually is useful for and what combination of different compounds are...'

V.1.1.1.1.3.1.3.1

'I'm a strong advocate for this (medicinal cannabis) being treated the same as any other medicine. In that way, ideally cannabinoid trials would continue, just like for any other medicine...' Most of us—people are generating trial data but really in very specific...(conditions).'

*Boyle, Douglas et al. (2019) Patron Primary Care Research Data Repository. University of Melbourne. Accessed at <u>https://medicine.unimelb.edu.au/school-structure/general-practice/engagement/data-for-decisions</u> There is no robust comprehensive national monitoring system in place, to ascertain Medicinal Cannabis effects and detect 'signals' that indicate the presence of side effects and adverse events, that could be attributed to Medicinal Cannabis use.

GP's and Physicians—Key Informant Interviews

Pharmacists—Narrative Review Stakeholders—Qualitative Analysis TGA Approvals—Quantitive Analysis Electronic Medical Record—Analysis Social Media—Reviews

/Electronic Medical Records (EMR) as a Data Source Background:

• Medicinal cannabis prescribing can be monitored using data from the general practice EMR.

Method:

- Rule-based digital phenotyping
- 1,164,846 active patients from 109 practices
- September 2017 to September 2020 **Results:**
- 80 patients with 170 prescriptions of medicinal cannabis were identified in the PATRON* database.
- Reasons for prescription included anxiety, multiple-sclerosis, cancer, nausea, and Crohn's disease.
- 9 patients had possible adverse effects that included depression, motor vehicle accidents, and gastrointestinal symptoms.

Discussion:

• EMR's enable the monitoring of community use of medicinal cannabis.

What is 'REAL-WORLD' Evidence? Multiple sources of REAL-WORLD DATA relating to the safety and benefits of Medicinal Cannabis enable PHARMACOVIGILANCE



- We face many Challenges in Healthcare today, with rising numbers of difficult to treat problems including mental illhealth, chronic pain, addiction, insomnia, and treatment resistant conditions.
- Consumers are increasingly driving demand for new therapeutics such as medicinal cannabis; this changes the paradigm through which Evidence Based Medicine is practiced.
- Greater Consumer Involvement in healthcare is valuable but should not be at the cost of rigorous and reproducible approaches that underpin Scientific Methodologies.
- The way forward is to <u>INTEGRATE</u> all available Real-World Evidence including Digital Health Data, Clinical Trial Results, Consumer Discourse, and Provider and Patient Experience to grow the knowledge and understanding about New and Emerging Therapeutics

Internal use

The Need for Evidence – WHY?



*SAS-B - Prescriptions Applications to TGA by Medical Practitioners for therapeutics that are not included in ARTG for a single patient under their care

**Authorised Prescribers – Medical Practitioners approved by a HREC or endorsed by a specialist college to prescribe medicinal cannabis

Internal use

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Current status of OMOP-CDM in OHDSI APAC regions : Lessons for Data Quality Assessment

Updates

2023-11-02

Chungsoo Kim & Sujin Gan



What is this study for?

• Collecting CDM Inspection reports from OHDSI APAC community

Why is this study needed?

• To check the current status of OMOP-CDMs, to get insights from the our CDMs, and to seek quality improvement point.

What is the final goal?

- It could provide a basic reference of statistics which can be used for future CDM conversion.
- Disclosure of current status of conversion, contents, and data distribution of CDMs of the OHDSI APAC community.



- Data sources: CDM databases from OHDSI APAC community
- Collecting inspection reports from each site.
- R package for automatically creating inspection reports.
- Collectibles
 - Number of record, person
 - Number of unique concepts per person
 - Source-CDM mapping ratio
 - Proportion of standard concepts in mapped codes
 - Drug mapping level (granularity)
 - Frequent concept list in each domain
 - Achilles heel result (error / notification / warnings)



- New report from NUHS (Thx Singapore team)
- Study close
- Inquiries for missing data and some issues in reports to data partners
- e.g., Number of the source code

condition	13946	217545	3208016	10266	40,481	27,250	25072	28,120
device	6333	716	0	0	0	0	3935	0
drug	18526	16683	200910	6868	2,367	20,893	15832	21,141
measurement	8473	2964	21625	1921	0	31,477,976	5988	1,013
measurement-unit	92	129	9	73	0	6	137	
measurement-value	5122	327	1561938	224701	0	1	0	15,464
observation	29431	2147	33595	294	40,481	0	13	1,208
observation-unit	1	13	0	1	1	0	3	0
observation-value	0	2	1	2	1	0	1	2
procedure	16652	1133	14147	0	0	0	26184	5,275
visit_occurrence	6	1	5	5	33,327,559	7,188,545	1	3

visit_source_value	This field houses the verbatim value from the source data representing the kind of visit that took place (inpatient, outpatient, emergency, etc.)
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Some institutions store visit ids than visit type in visit_source_value column for their needs.

- Writing a draft manuscript
- Making a poster for the Annual conference of the American Medical Informatics Association



- 2023 AMIA @ New Orleans ٠
- Poster session (Nov 13th 17:00-18:30) ٠
- [Action Needed] Community feedback ٠ before printing the poster (~ Nov 6th)

OMOP-CDM in Asia-Pacific regions and Lessons for Data Quality Assessment Sujin Gan, RN¹, Chungsoo Kim, PharmD¹, Seongwon Lee, PhD², Jing Li³, Jiawei Qian³, Gyeol Song¹, Clair Blacketer⁴, Anthony Molinaro⁴, Dinuja Willigoda Liyanage⁵, Zhang jingyi⁶, Li Chao⁶, Roger Ward⁷, Mengling Feng⁸, PhD, Mui Van Zandt³, Rae Woong Park, MD, PhD12

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Introduction

The Observational Medical Outcome Partnership-Common Data Model (OMOP-CDM), an open community data standard, is being implemented globally, but data quality control for CDM adoption is challenging. The data quality assessment tools including the Achilles Heel1 and Data Quality Dashboard2 have been performed only individually at each institution. Therefore, European Health Data and Evidence Network (EHDEN) has developed the CDM Inspection report, which writes a report on data statistics, mapping, and quality checks, to provide insight into the completeness, transparency, and quality of the data.

Methods

The CDM Inspection report was collected on the OHDSI Asian Pacific (APAC) community, using the R package (https://github.com/ABMI/CdmInspection). A total of 22 databases from Korea, 2 from Japan, and 1 each from Australia, China, and Singapore were included, and they consisted of 25 EMRs and 2 claims. The report describes an analysis result of the number of records or patients, the ratio of records per person (RPP), the ratio of records per observation period, the mapping ratio between source and transformed data, the mapping level of drug vocabulary and list of frequent concepts for each domain table of the OMOP-CDM.

The overall database contained the data of 49,567,744 persons and 37,846,583,033 records, and the majority part was measurement data with 29.8%. For the RPP ratio, drugs had a mean ± SD of 152.9 ± 124.5, and measurements showed 610.0 ± 586.0 . The observation record ratio per observation period was only $45.0 \pm 21.7\%$ of total patients. Among all records, 96.32 [95.01, 98.1] % in the drug exposure table were mapped, while 34.57 [20.90, 92.78] % of the procedure and 7.74 [4.12, 39.73] % of measurement values were mapped. The most frequently used vocabulary for drug mapping was the Branded Drug (Mean \pm SD, 31.8 \pm 26.6%) and the Clinical Drug (21.4 \pm 25.5) of the RxNorm.

Data in the CDM databases for the APAC region showed a heterogeneous distribution. Continued collection of CDM inspection reports over the APAC is necessary to improve healthcare data quality.

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The growing adoption of OMOP-CDM in Asia Pacific region requires continuous data quality measure & management

Title: Current status of OMOP-CDM in Asia-Pacific and Needs for Data Quality Assessment

Background: The Observational Medical Outcome Partnership Common Data Model (OMOP-CDM), an open community data standard, is being implemented globally, especially Asia Pacific region (APAC). However, data quality control for CDM adoption is challenging. This study collected the CDM inspection reports from OHDSI APAC community and conducted descriptive analyses.



Mean (SD) Median (IQR) + Mapping rate by data Most of Condition Drug, Observation and Visit occurrence to mapped (190%) to the OMOP standardised vocabulary However, the Measurement (especially value of me Procedure showed low rate of mapping. Our next priority I 96.5 (6.1) 99.8 (4.9) 76.7 (20.1) 77.8 (28.3) 96.3 (3.7) 54.2 (10.1) domain.
 domain.
 79.4 (29.8) 96.1 (30.6) ▼ Detailed RaNorm class for drug records mappin Among total drug records (n to the RoNorm class containi "Quant Branched Drug", "Marks
 Bits
 Bits
 Clinical
 Guart
 Marketed
 Guart
 Ingredient

 55.4
 (48.4)
 80.2
 (100.0)
 Drug
 Dru

 50.1 (36.8)
 34.6 (71.9)
 243.%
 180.%
 13.5 %
 12.7 %
 8.1 %
 4.5 %

Result 3: Some checkpoints for the data quality

Warning
 Result Type
 Error
 Warning
 An mean of 7 and a median of 3 Achilles' heel quality

 Site, n
 Mean±SD
 Median(IQR)
 Median(IQR)
 Median(IQR)
 errors occurred in the EMR database. 35% of the 7.2 ± 7.9 3 (14.25) 17.8 ± 5.7 20 (4.5) o Observation table (2) Distribution of Records Per Person ratio





Limitation: The APAC region is not fully represented as most of the databases are from South Korea. We need more information on more diverse databases including national claims, clinical registry.

Conclusion: By collecting CDM inspection reports, we were able to identify ready-to-analyze databases in the APAC region. However, some database showed a quality check failure and a heterogeneous distribution. Continued collection of CDM inspection reports over the APAC is necessary to improve healthcare data quality



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Deep Learning Comparison An OHDSI Network Study

This slide was presented in 2023 OHDSI European Symposium.

2023-11-02

LH John, Chungsoo Kim, JM Reps, EA Fridgerisson





Yang 2022 - Figure 1 - J Am Med Inform Assoc, Volume 29, Issue 5, May 2022, Pages 983–989, https://doi.org/10.1093/jamia/ocac002

Observational healthcare data limit efficacy of deep learning:

- highly sparse
- high-dimensional
- heterogenous



Study design

Aims and objectives

Assess the added value of massive _ observational healthcare data for the development of deep learning models

Prediction methods

- Logistic regression L1 -
- Gradient Boosting
- ResNet (Gorishniy, 2021) -
- FT-Transformer (Gorishniy, 2021) -

Prediction problems

- Dementia in persons aged 55 and above
- Lung cancer in persons aged 45 and above
- Bipolar in persons diagnosed with major depressive disorder

Confirmed databases

- **Optum SES**
- **Optum EHR**
- **MDCR**
- **IQGER**
- IPCI
- AUSOM





Cohort	Database	Method	AUROC
Dementia	IPCI	Logistic Regression	83.28
		Gradient Boosting	82.86
		ResNet	82.50
		Transformer	82.36
	AUSOM	Logistic Regression	77.98
		Gradient Boosting	76.59
		ResNet	58.58
		Transformer	63.08

Cohort	Database	Method	AUROC
Lung cancer	IPCI	Logistic Regression	71.04
		Gradient Boosting	70.84
		ResNet	67.26
		Transformer	TBD
	AUSOM	Logistic Regression	74.31
		Gradient Boosting	TBD
		ResNet	50.43
		Transformer	68.75



Join The Network Study!

Help us assess the added value of observational data for the development of deep learning models.

Head over to GitHub <u>https://github.com/ohdsi-studies/DeepLearningComparison</u> Docker Hub : <u>https://hub.docker.com/r/egillax/deeplearningcomparison</u>

Watch out!

This study requires a recent **Nvidia graphics card** to execute.

DeepLearningComp generated from ohdsi-studies/EmptySt	Darison Public	⊙ Unwatch 10	▼ V Fork 1 ▼ ★ Starred 2 ▼		
				docker hub Q egillax/deeplearningcomparison	X Explore Pricing Sign In Sign up
P master - P 5 branch	es 🔊 0 tags Go to file	Add file - <> Code -	About	Explore / egillax/deeplearningcomparison	
🕂 Ihjohn Add Table 1 script	32cedlc 2	days ago 🕚 54 commits	Investigating different deep learning approaches		
cohorts	Add Table 1 script	2 days ago	🖽 Readme	egillax/deeplearningcomparison 🕸	💆 Pulls 2
extras	Remove torch dependency from JSON	2 months ago	Ar Activity	By egillax • Updated 2 days ago	
C README.md	Update README.md	10 months ago	 10 watching 	Image	
StudyProtocol.pdf	Add protocol	3 months ago	😵 1 fork		
🗋 codeToRun.R	generic strings for user inputs	6 months ago	Report repository	Overview Tags	
CodeToRunTable1.R	Add Table 1 script	2 days ago		Sort by Newest - Filter Tags Q	
CodeToRunValidation.R	Add code to run dementia validation study	3 months ago	Releases	710	
deep_comp_bipolar_val	Turn off sampling	2 months ago	No releases published Create a new release	latest	docker pull egillax/deeplearning
deep_comp_cuda0_stud	Update JSONs	last month		Last pushed 2 days ago by egillax	
deep_comp_cuda1_stud	Update JSONs	last month	Packages	2a2da011140f linux/amd64	3.47 GB
🗋 deep_comp_dementia_v	Turn off sampling	2 months ago	No packages published		
deep_comp_lungcancer	Turn off sampling	2 months ago	Publish your first package		



Thank you for listening!



December Meeting Preview

- Anna Ostropolets from the OHDSI Vocabulary Team will join us again to go over community contribution guidelines for drug vocabularies
 - <u>https://github.com/OHDSI/Vocabulary-v5.0/wiki/Community-contribution-guidelines:-drug-vocabularies</u>
- Previous sessions handled non-drug vocabularies only
 - <u>https://github.com/OHDSI/Vocabulary-v5.0/wiki/Community-contribution-guidelines:-non%E2%80%90drug-vocabularies</u>
 - Recordings of the sessions are available at https://www.ohdsi.org/apac/



Support Areas Follow-up Survey Extension

- Follow-up survey on the APAC community's support areas of interest have been extended until <u>end of November</u>
- Results will be used to plan out topics for 2024 so please make sure to submit your responses if you want your interests to be reflected!
- Survey is focused on understanding specific needs of the community around two areas that received the most votes from the initial survey – data analytics and uniform data representation
- Direct link to survey: <u>https://forms.office.com/r/sGXSnXV4G3</u>



Thank you!