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ABSTRACT

- Claims from clearinghouses are referred to as "open **claims**" while claims from insurance companies "**closed** claims"
- Open claims have the potential to be a rich source of observational data
- This poster will describe a methodology for evaluating open claims data by:
 - □ Using a cross section of patients that appear both in open/closed claims
 - □ Transforming the open/closed Observational Medical Outcomes Partnership (OMOP) Common Data Model (CDM) [1]

BACKGROUND

Clearinghouses:

- Used for "claims scrubbing" between the point of care and the insurance company
- They provide transactional support between providers and payers to facilitate fast payment by minimizing errors during the claims submission process [2]
- Claims are considered "**open**" in that they contain the same information as data available from large payers except for final adjudication (considered to be "closed")
- **Open claims have not been utilized for** observational research as regularly as closed **claims** for many reasons, one of which being because members are enrolled at the provider level rather than at the patient level
- The OMOP CDM is an effective way to standardize observational health databases however it has yet to be employed on open claims in a way that enables assessment of the data itself for use in epidemiological research [3]

OBJECTIVES

- Design metrics and cohorts meant to evaluate open claims
- 2. To enable assessment of the closed claims, translate cross section of open/closed claims into the OMOP CDM

METHODS

- Medical and pharmacy data from open claims was obtained and, to serve as a control, pharmacy data from a prescription benefits plan was linked at the patient level using protected health information (PHI)
- Non-linked commercial medical claims were also obtained from Truven Health MarketScan® Commercial Claims and Encounters Database for use as a high-level comparison

Creating a Framework for Evaluating Open Healthcare Claims Using the OMOP Common Data Model Clair Blacketer, MPH^{1,2}, Erica A Voss, MPH^{1,2}

claims to the

METHODS (cont.)

CDM CONVERSION TECHNIQUES

□ The raw dataset contained information from two separate sources (open and closed claims) which necessitated designing the tables so that each source was still identifiable after transformation. Techniques included in Table 1.

AIMS

- 1. To quantify the missing data from the open claims that would be provided by the closed claims system and viceversa
- 2. Identify drugs (by ingredient) that occur more often in one database versus the other
- 3. Assess if open claims capture the same proportion of conditions as trusted closed claims system during the time prior to drug exposure

METRICS

- 1. Measure the overlap between the **open / closed claims** for prescription drugs occurring per person per day during the same time period for a set of defined cohorts (**Figure 1**).
- 2. Compare the proportion of drug exposures at the ingredient level among persons in both the open / closed systems during the same time period for a set of defined cohorts (Figure 2)
- 3. For a set of defined new user drug cohorts, compare the baseline prevalence of conditions between the open claims and Truven Health MarketScan® Commercial Claims and Encounters Database

Figure 1: Example of drug overlap at the claim level

	DRUG_	DRUG_EXPOSURE		PE		
PERSON_ID	CONCEPT_ID	START_DATE	CONCEPT_	ID REFILLS	QUANTITY	DAYS_SUPPLY
XXXXXX	957136	5 1/10/201	l3 <mark>38000</mark>	<mark>175</mark> 5	30) 30
XXXXXX	957136	5 1/10/201	l3 <mark>38000</mark>	<mark>177</mark> 5	30) 30

Figure 2: Example of drug overlap at the ingredient/exposure level

	DRUG_	DRUG_EXPOSURE_	DRUG_TYPE_			
PERSON_ID (CONCEPT_ID	START_DATE	CONCEPT_ID REFILLS	Q	UANTITY DAYS_	SUPPLY
XXXXXX	<mark>957136</mark>	1/10/2013	38000175	5	30	30
XXXXXX	<mark>19023799</mark>	2/3/2013	38000177	4	30	28

957136 -> Furosemide 40 MG 19023799 -> Furosemide 40 MG Oral Tablet [Lasix]

Table 1. CDM conversion techniques and examples

Technique

*_TYPE_CONCEPT_ID as proxies for each database

PAYER_PLAN_PERIOD and **OBSERVATION_PERIOD** to capture the amount of tin contributed per person to each database

COHORT table to house cohorts of interest for evaluation purposes

CONCLUSIONS

- system feasible

REFERENCES

they do? 2014.

CONFLICT OF INTEREST STATEMENT

Clair Blacketer and Erica Voss are full time employees of Janssen Research and Development, a unit of Johnson and Johnson. The work on this study was part of their employment. They also hold pension rights from the company and own stock and stock options.





	Example
5	 Drug data from closed claims were given DRUG_TYPE_CONCEPT_ID = 38000177
	 Drug data from open claims were given DRUG_TYPE_CONCEPT_ID = 38000175
d o ne	 Enrollment information from closed claims was used to create PAYER_PLAN_PERIOD
	 Min and max claims dates from open claims were used to create OBSERVATION_PERIOD
	 Open claims contained a file of "stable" providers (providers that consistently sent a large number of claims to the clearinghouse). This file was used to create a cohort of patients in the COHORT table that visited only those providers.

□ The use of clearinghouse data in observational health research is still a debated topic though little is understood about how well open claims represent the true picture of healthcare utilization for a given set of patients

□ By defining a set of metrics and designing an ETL in a way that allows for the assessment of those metrics, the **OMOP** CDM can be leveraged to create a framework that makes the evaluation of such data from an open

> 1. OMOP Common Data Model [Webpage]. 2015 [cited 20 Jul 2015]. Available from: http://www.ohdsi.org/data-standardization/thecommon-data-model/.

> 2. Sculley J. What is a clearinghouse for medical claims, and what do

3. Voss E, Makadia R, Matcho A, Ma Q, Knoll C, Schuemie M, et al. Feasibility and utility of applications of the common data model to multiple, disparate observational health databases. Journal of the American Medical Informatics Association. 2015 2015 May;22(3):553-64. Epub 2015 Feb 10.