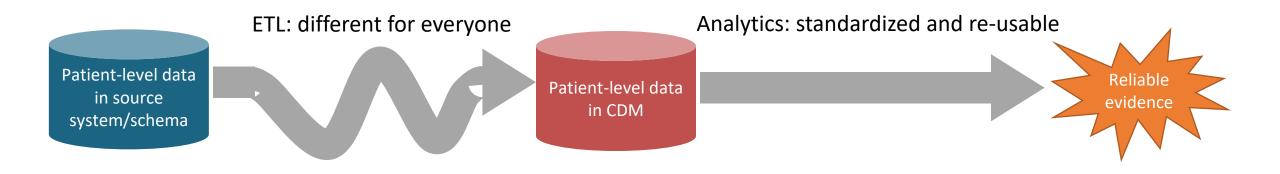


Advanced analytics with OHDSI tools



Why convert to the Common Data Model?

- Transforming data to the OMOP CDM is a large investment
- The benefits come from being able to use the same tools and analytics across many databases





OHDSI standardized analytics



- HADES is a set of open-source R package
- Developed and maintained by the community, for the community
- Can use cohort definitions created in ATLAS



966

MEDINFO 2023 — The Future Is Accessible

J. Bichel-Findlay et al. (Eds.)
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doi:10.3233/SHTI231108

Health-Analytics Data to Evidence Suite (HADES): Open-Source Software for Observational Research

Martijn SCHUEMIE^{a,b,c,1}, Jenna REPS^{a,b,d}, Adam BLACK^{a,e}, Frank DeFALCO^{a,b}, Lee EVANS^{a,f}, Egill FRIDGEIRSSON^{a,d}, James P. GILBERT^{a,b}, Chris KNOLL^{a,b}, Martin

Analytic use case	Туре	Structure
Clinical characterization	Disease Natural History	Amongst patients who are diagnosed with <insert disease="" favorite="" your="">, what are the patient's characteristics from their medical history?</insert>
	Treatment utilization	Amongst patients who have <insert disease="" favorite="" your="">, which treatments were patients exposed to amongst dist of treatments for disease> and in which sequence?</insert>
	Outcome incidence	Amongst patients who are new users of <insert drug="" favorite="" your="">, how many patients experienced <insert adverse="" drug="" event="" favorite="" from="" known="" profile="" the="" your=""> within <time exposure="" following="" horizon="" start="">?</time></insert></insert>
Population-level	Safety surveillance	Does exposure to <insert drug="" favorite="" your=""> increase the risk of experiencing <insert adverse="" an="" event=""> within <time exposure="" following="" horizon="" start="">?</time></insert></insert>
effect estimation	Comparative effectiveness	Does exposure to <insert drug="" favorite="" your=""> have a different risk of experiencing <insert (safety="" any="" benefit)="" or="" outcome=""> within <time exposure="" following="" horizon="" start="">, relative to <insert comparator="" treatment="" your="">?</insert></time></insert></insert>
	Disease onset and progression	For a given patient who is diagnosed with <insert disease="" favorite="" your="">, what is the probability that they will go on to have <another complication="" disease="" or="" related=""> within <time diagnosis="" from="" horizon="">?</time></another></insert>
Patient level prediction	Treatment response	For a given patient who is a new user of <insert chronically-used="" drug="" favorite="" your="">, what is the probability that they will <insert desired="" effect=""> in <time window="">?</time></insert></insert>
	Treatment safety	For a given patient who is a new user of <insert drug="" favorite="" your="">, what is the probability that they will experience <insert adverse="" event=""> within <time exposure="" following="" horizon="">?</time></insert></insert>

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For a given patient who is a new user of GLP-1s, what is the probability that they will experience an AMI while exposed to the drug?		
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Clinica			
		•	LP-1s have a different risk of experiencing to drug, relative to DPP-4s?
Popul			n adverse
	estimation	Comparative effectiveness	Does exposure to <insert drug="" favorite="" your=""> have a different risk of experiencing <insert (safety="" any="" benefit)="" or="" outcome=""> within <time exposure="" following="" horizon="" start="">, relative to <insert comparator="" treatment="" your="">?</insert></time></insert></insert>
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Cohorts of our examples

Cohort: a group of people who satisfy some criteria for some period of time

Indication cohorts:

Type-2 diabetes mellitus (T2DM)
 People with T2DM, while having T2DM

Exposures cohorts :

GLP-1 agonists
 People on GLP-1, while on the drug

— DPP-4 inhibitors
 People on DPP-4, while on the drug

Outcomes cohorts :

Acute myocardial infarction (AMI)
 People with AMI, at the time of AMI

These same cohorts can be re-used to answer different questions

Patrick will discuss how to build these

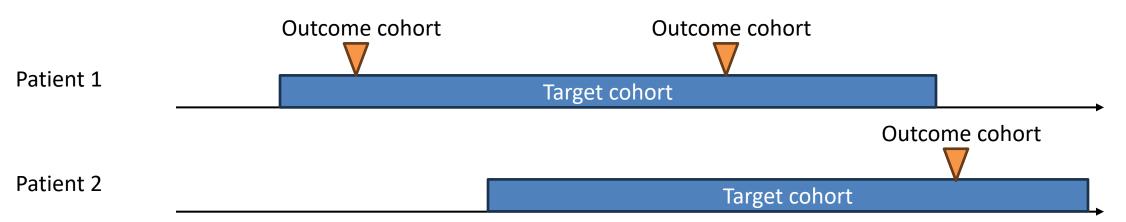


Characterization: CohortIncidence package

Amongst patients who are new users of **GLP-1s**, how many patients experienced **Acute Myocardial Infarction** within **drug exposure**?

- Target: GLP-1

Outcome: AMI



Computes the incidence rate of the Outcome cohort in some Target cohort

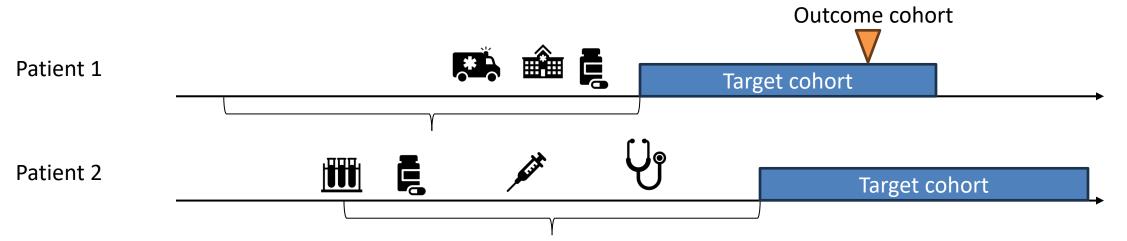
- Standardized computation of incidence rates
- Default: overall and stratified by age, sex, and calendar time



PatientLevelPrediction package

For a given patient who is a new user of **GLP-1s**, what is the probability that they will experience **an AMI** while **exposed to the drug?**

- Target: GLP-1, restricted to those with T2DM (and first use only)
- Outcome: AMI



Builds a model to predict who in the Target will have the Outcome

- Uses all observed data up to Target start
- Implements many machine learning / deep learning algorithms

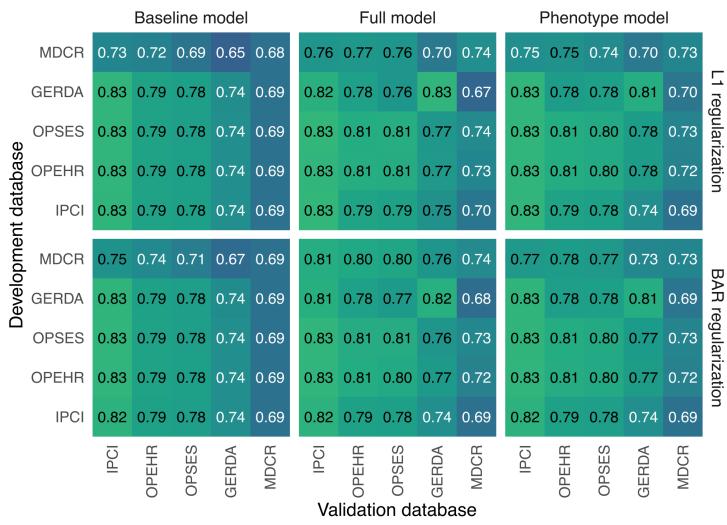


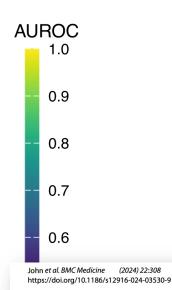
Unique feature: external validation

- When using the PatientLevelPrediction package, models fit in one database can easily be validated in other databases
 - Portability of code
 - Standardized construction of features



Example of external validation





RESEARCH ARTICLE

(2024) 22:308

Open Access

BMC Medicine

Development and validation of a patientlevel model to predict dementia across a network of observational databases

Luis H. John^{1*}, Egill A. Fridgeirsson¹, Jan A. Kors¹, Jenna M. Reps², Ross D. Williams¹, Patrick B. Ryan² and Peter R. Rijnbeek¹

Abstract

Background A prediction model can be a useful tool to quantify the risk of a patient developing dementia



CohortMethod package

Does exposure to **GLP-1s** have a different risk of experiencing **AMI** while **exposed to drug**, relative to **DPP-4s**?

- Target: GLP-1, restricted to those with T2DM (and first use only)
- Comparator: DPP-4, restricted to those with T2DM (and first use only)
- Outcome: AMI
 Outcome cohort

Patient 1

Target cohort

Outcome cohort

Patient 2

Comparator cohort

Computes the hazard of the Outcome cohort in the Target cohort compared to the Comparator



Unique feature: Large-scale propensity scores

- Treatment assignment is often non-random, which can cause confounding
 - E.g. GLP-1 may be prescribed more often to obese, who already have a higher risk of AMI
- Propensity scores are an establish way to address this
 - Fit a model to predict treatment assignment, and use to compute probability (propensity score)
 - Match subjects in Target to Comparator with similar propensity scores
- Traditionally, expert pick a few variables to use in the prediction model
- Large-scale propensity scores include all baseline covariates, and uses machine learning

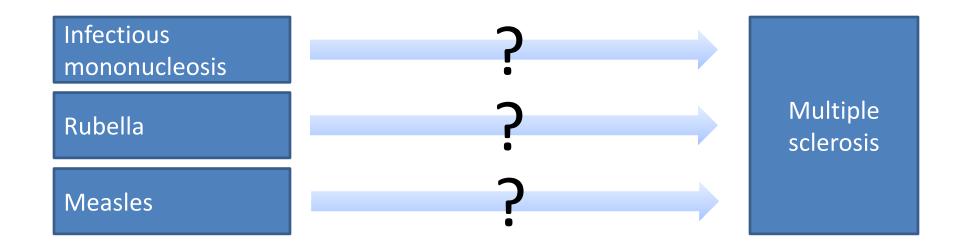


Unique feature: objective diagnostics

- Whether study results are reliable depends on whether certain assumptions have been met
 - E.g. we assume our PS adjustment makes our treatment groups comparable
- Most of these assumptions are testable through diagnostics
 - E.g. we can test whether our PS adjustment achieved balance by computing the standardized difference of means (SDM)
- By 'objective' diagnostics we mean diagnostics that are evaluated while blinded to the results of the study
 - E.g. Pre-specify that we will not look at results where max(|SDM|) > 0.1
 - Unique: negative controls



Example of a negative control



RESEARCH PAPER

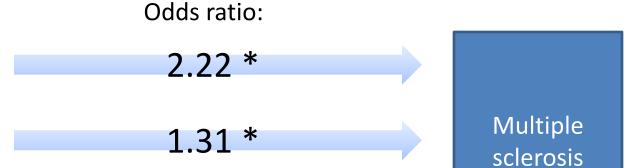
Multiple Sclerosis 2008; 14: 307-313

Selective association of multiple sclerosis with infectious mononucleosis

BM Zaadstra^{1,2}, AMJ Chorus¹, S van Buuren^{1,3}, H Kalsbeek¹ and JM van Noort⁴



Example of a negative control



Measles 1.42 *

* P < .05

RESEARCH PAPER

Infectious

Rubella

mononucleosis

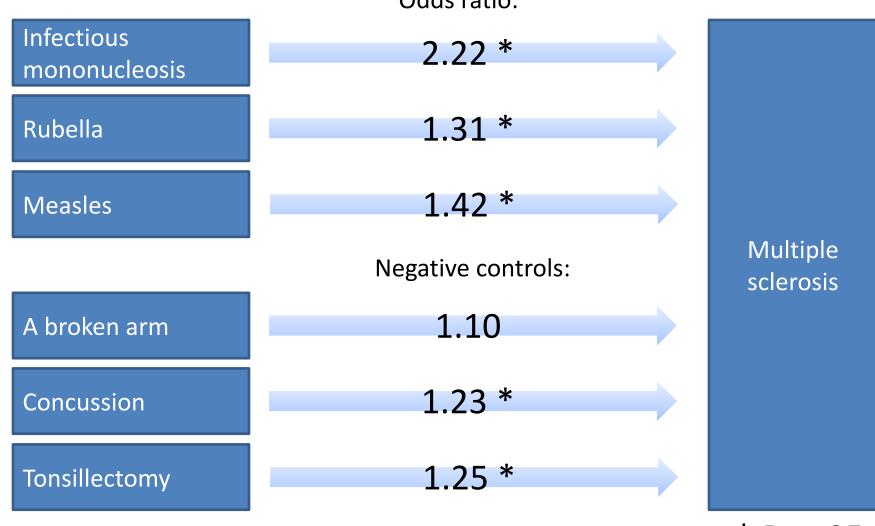
Multiple Sclerosis 2008; 14: 307-313

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Example of a negative control

Odds ratio:



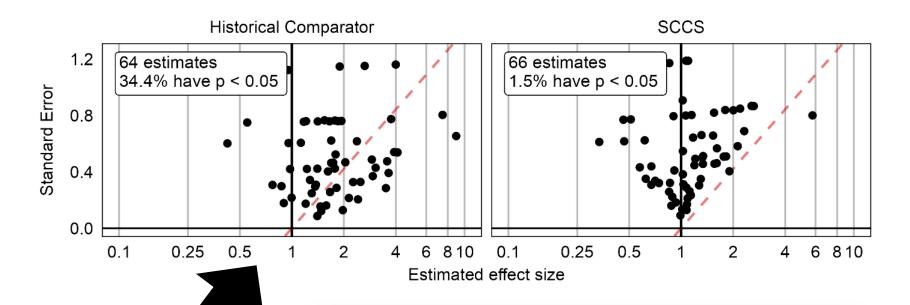


How to interpret negative control findings?

- Unique: use a sample (n > 50) of negative controls to understand distribution of bias
- Systematic error distribution can be used as
 - Diagnostic: if too much systematic error, we stop
 - Calibration: can adjust p-values and confidence intervals to take into account possible systematic error



Quantifying systematic error



Revised: 30 September 2022

Need to execute estimation studies for 66 Target-Outcome combinations.

OHDSI tools readily allow for this (simply swapping out the outcome cohort for the negative controls)

Adjusting for both sequential testing and systematic error in safety surveillance using observational data:
Empirical calibration and MaxSPRT

Martijn J. Schuemie^{1,2} | Fan Bu^{2,3} | Akihiko Nishimura⁴ | Marc A. Suchard^{2,3,5}

Accepted: 8 December 2022

¹Observational Health Data Analytics, Janssen Research & Development, Titusville, New Jersey,

²Department of Biostatistics, University of California, Los Angeles, California,

³Department of Human Genetics,
University of California Los Angeles

Received: 8 July 2022

DOI: 10.1002/sim.9631

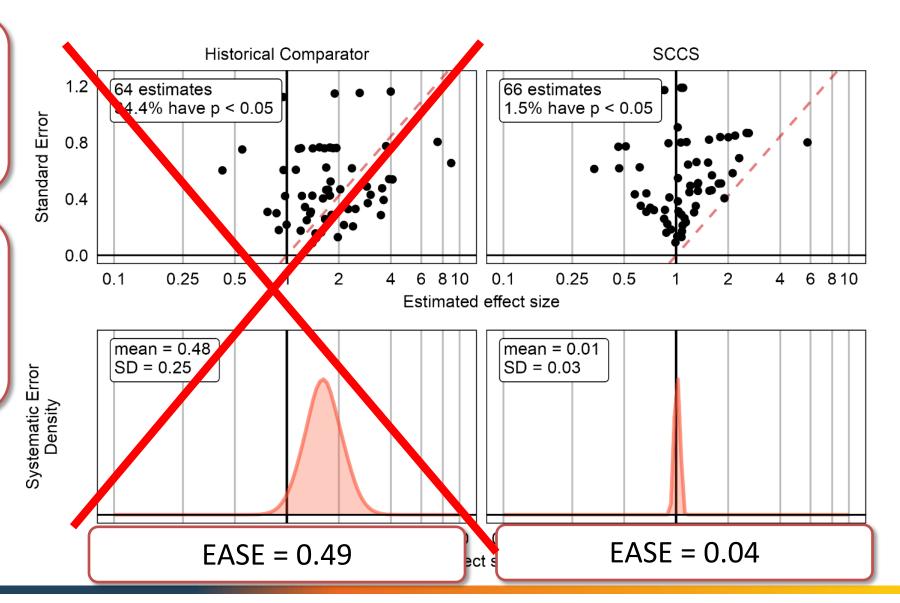
Post-approval safety surveillance of medical products using observational healthcare data can help identify safety issues beyond those found in pre-approval trials. When testing sequentially as data accrue, maximum sequential probability ratio testing (MaxSPRT) is a common approach to maintaining nominal type 1 error. However, the true type 1 error may still deviate from the

WILEY

Quantifying systematic error

Expected Absolute
Systematic Error (EASE)
summarizes this
distribution

We use a **prespecified**EASE threshold (EASE < 0.25) for go – no go
decisions for our studies





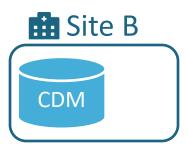
Distributed analyses

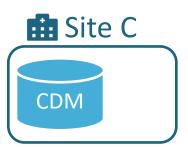
Using OHDSI tools



- Multiple sites with data
 - Hospital EHRs (Electronic Health Records)
 - Administrative Claims
- Patient-level data cannot be shared
- Each site uses the Common Data Model (CDM)







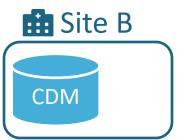




A site can lead a study







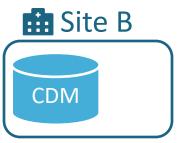






- A site can lead a study
- Analysis code is developed locally











- A site can lead a study
- Analysis code is developed locally
- Code is distributed to study participants











- A site can lead a study
- Analysis code is developed locally
- Code is distributed to study participants
- Results are generated (aggregated statistics)











- A site can lead a study
- Analysis code is developed locally
- Code is distributed to study participants
- Results are generated (aggregated statistics)
- Results are sent back to lead site





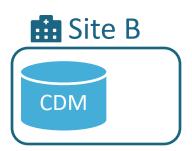






- A site can lead a study
- Analysis code is developed locally
- Code is distributed to study participants
- Results are generated (aggregated statistics)
- Results are sent back to lead site
- Evidence is synthesized



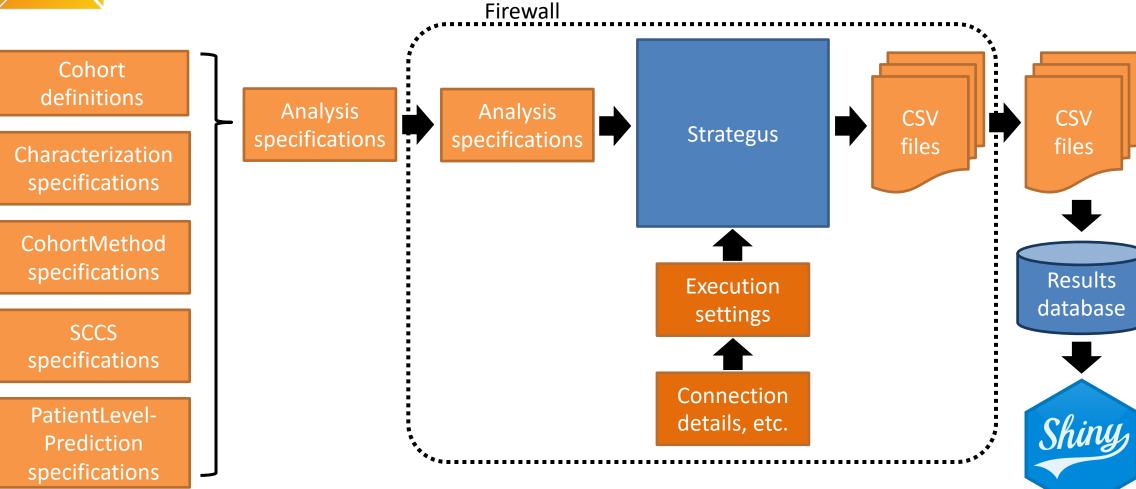








Strategus for study execution





Summary



Unique features of HADES analytics

- Re-use of cohort definitions
- Standardization of analytics in open-source software
 - Many opportunities for testing, review, fixing bugs, etc.
 - Making it hard to do the wrong thing (opinionated)
- Advanced methods to reduce bias
 - Splines for time in self-controlled case series
 - Large-scale propensity scores in cohort method
- Objective study diagnostics to improve reliability of evidence
 - Including negative controls
- Designed to run across a network of databases
 - Without sharing patient-level data