

Update from the Phenotypes, Data Standards, Data Quality Core of the NIH HCS Research Collaboratory

NIH Collaboratory Grand Rounds

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Outline

- PSQ Core and Charter
- Background and Landscape
- Phenotype-related activities
- Standards approach
- Data Quality Assessment
- Impact of PSQ core
- Future directions

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Duke members: Rachel Richesson, Michelle Smerek, Ed Hammond, Monique Anderson

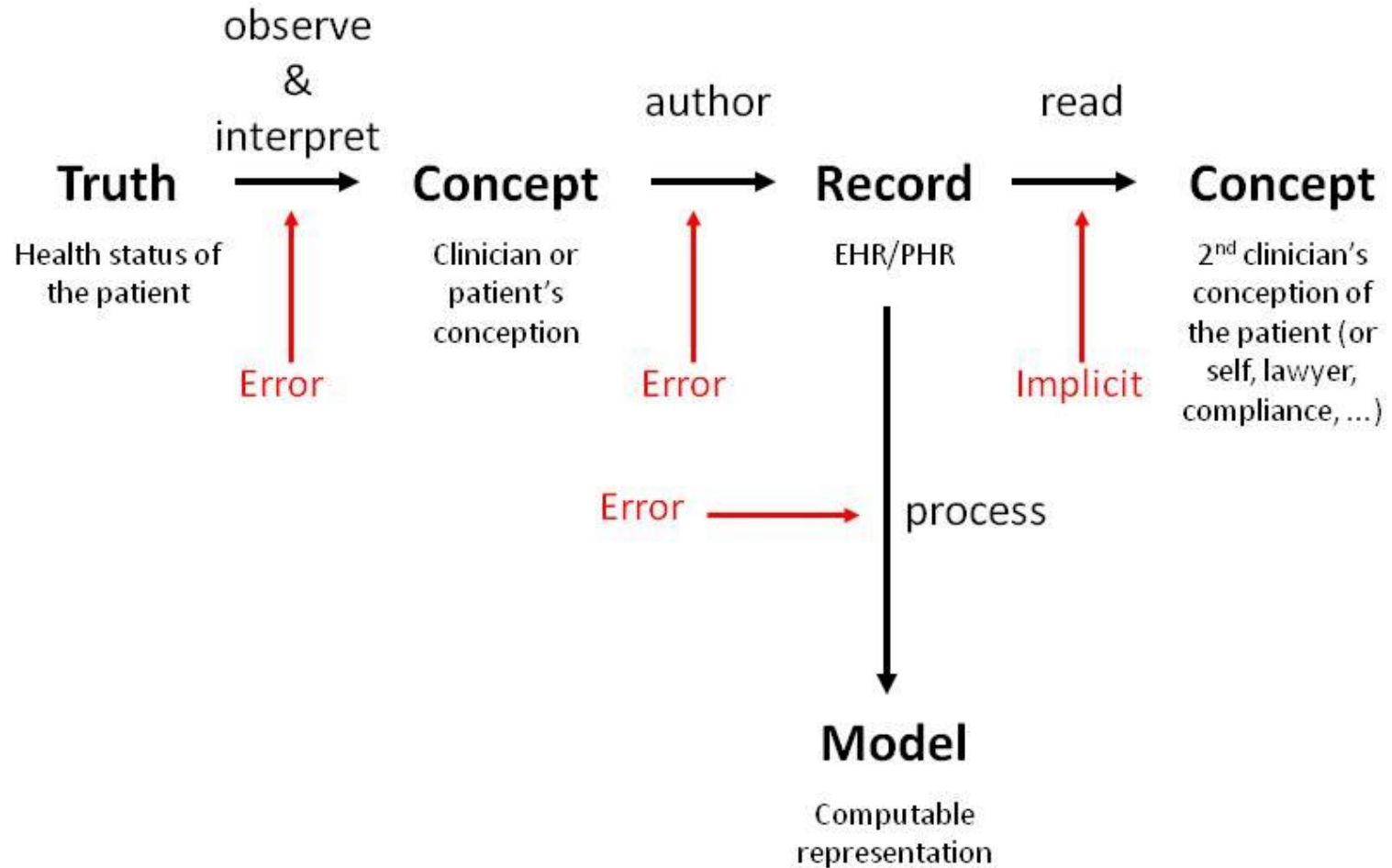
Charter – Phenotype, Data Standards, and Data Quality Core (PSQ Core)

- Share experiences using EHR to support research in various disease domains and for various purposes.
- Identify generalizable approaches and best practices to promote the consistent use of practical methods to use clinical data to advance healthcare research.
- Suggest where tools are needed.
- Explore and advocate for cultural and policy changes related to the use of EHRs for identifying populations for research, including measures of quality and sufficiency.

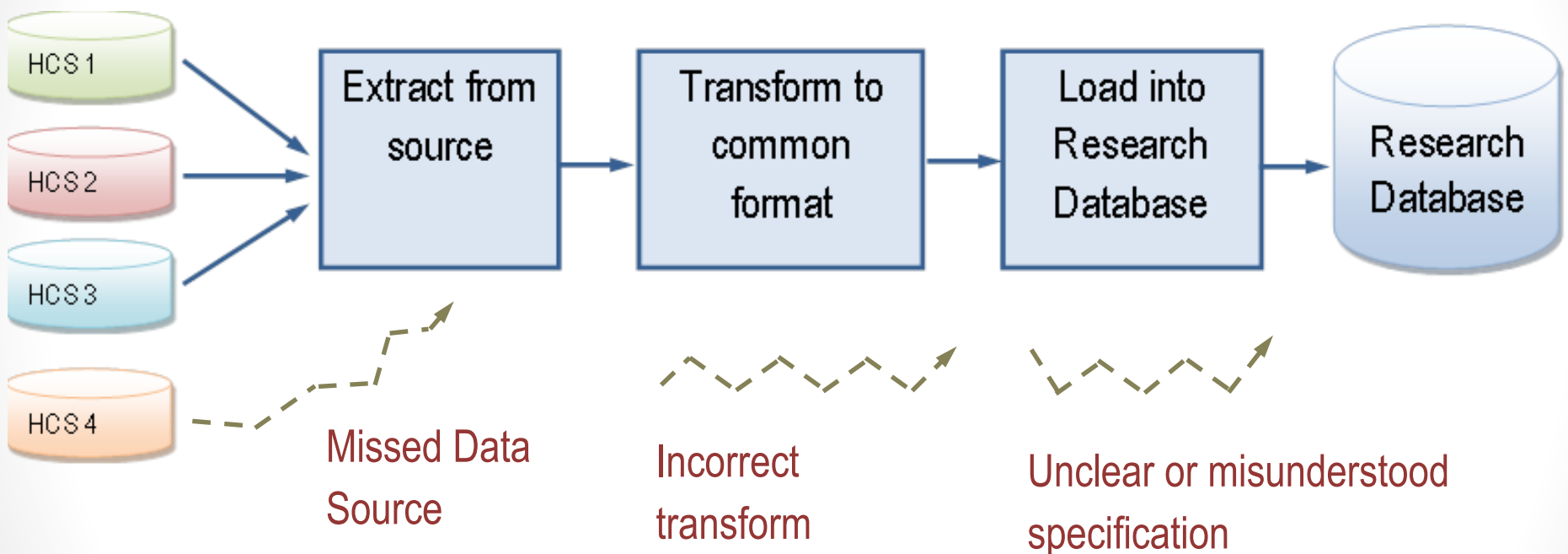
The Landscape

- Little standardized data representation in EHRs
- What appears standard is not always so
 - Multiple sources of ICD-9-CM codes, lab values, and medication data
 - Use of codes varies by institution
- Coding systems change
- No standard representation or approach for phenotype definitions
- Reproducibility is a concern
- Data reflect patient and clinician/organizational factors
- Data quality is a concern

Imperfection of Clinical Data



Additional Challenges with Clinical Data from Multiple Healthcare Systems



Questions for PCT:

**Are data from different
sites comparable?**

Valid? Reliable?

*Graphic courtesy of Alan Bauck, Kaiser Permanente
Center for Health Research, 2011. (adapted)*

Use of EHRs in Collaboratory PCTs

- PPACT needs to identify patients with chronic pain for the intervention. This is done in different EHR systems using a number of “phenotypes” for inclusion – e.g., neck pain, fibromyalgia, arthritis; long term opioid use .
- STOP CRC needs to continually identify screenings for colorectal cancer from each site, so must maintain master list of codes (CPT and local codes) related to fecal immunochemical test orders across multiple organizations.
- The TSOS trial needs to screen patients for PTSD on ED admission. How can different EHRs systems and patient data be leveraged to ensure consistency and efficiency of screening?

Use of EHRs in Collaboratory PCTs

- LIRE trial uses EHR data to identify cohorts (dynamically as radiology reports are produced), insertions based on rules in the EHR processing), and as primary source of outcome variables.
- The SPOT trial needs to identify possible suicide attempts (as study outcome measure) from different populations and information systems using a set of injury codes (in ICD-9-CM and ICD-10-CM).

Transparency and Reproducibility of PCTs

Patient characteristics:

Table 1. Patient Demographics and Baseline Characteristics

Characteristic	No. (%) of Patients ^a	
	Gentamicin-Collagen Sponge (n = 753)	Control (n = 749)
Patient demographics		
Age, median (IQR), y	64.2 (58.0-71.5)	64.9 (57.2-72.1)
White race	688 (91.4)	683 (91.2)
Weight, median (IQR), kg	98.0 (86.1-113.0)	98.8 (85.0-111.1)
Body mass index, median (IQR)	33.1 (30.2-37.2)	32.8 (30.0-36.2)
Body mass index >30	574 (76.2)	563 (75.2)
Male sex	530 (70.4)	530 (70.8)
Medical history		
History of hypertension	659 (87.5)	659 (88.0)
History of diabetes	493 (65.5)	513 (68.5)
Current or history of smoking	458 (60.8)	450 (60.1)
Current smoking	136 (29.7)	123 (27.3)
History of chronic obstructive pulmonary disease	117 (15.5)	107 (14.3)
History of peripheral vascular disease	105 (13.9)	89 (11.9)
Previous median sternotomy	52 (6.9)	42 (5.6)
History of TIA or stroke	77 (10.2)	81 (10.8)
History of myocardial infarction	233 (31.0)	245 (32.7)
History of congestive heart failure	89 (11.8)	90 (12.0)
History of hyperlipidemia	619 (82.2)	607 (81.0)
Steroid use ≤1 mo prior to surgery	28 (3.7)	33 (4.4)
Receiving dialysis preoperatively	4 (0.5)	2 (0.3)
Preoperative diagnostic values		
Left ventricular ejection fraction, median (IQR), %	55 (45-60)	55 (45-60)
Serum glucose, median (IQR), mg/dL	125 (101-160)	124 (103-167)
Serum hemoglobin A _{1c} , median (IQR), %	6.5 (5.9-7.6)	6.6 (5.9-7.7)
Hematocrit, median (IQR), %	39 (36-42)	39 (36-42)
Serum creatinine, median (IQR), mg/dL	1.0 (0.9-1.3)	1.0 (0.9-1.2)
Preoperative core temperature, median (IQR), °C	97.6 (97.0-98.2)	97.7 (97.0-98.2)
Preoperative hospital stay, median (IQR), d	1.0 (0-3.0)	1.0 (0-3.0)
Parsonnet risk score, median (IQR) ^b	9.0 (6.0-14.5)	9.0 (6.0-16.0)

Abbreviations: IQR, interquartile range; TIA, transient ischemic attack.
 SI conversion factors: To convert creatinine to μmol/L, multiply by 88.4; glucose to mmol/L, multiply by 0.0555.
^aUnless otherwise indicated.
^bTheoretical range is 0 to 148; 50% in Parsonnet et al¹¹ had a score between 0 and 9.

Multiple phenotype definitions:

SUPREME-DM Phenotype

Definition:

Adult Durham Population patients who meet **ONE OR MORE** of the following criteria during a DukeMed encounter between 2007-2011:

- One or more instances of the specified ICD-9-CM diagnosis codes (see table 7) on an inpatient encounter
- OR 2 or more instances of the specified ICD-9-CM diagnosis codes (see table 7) on outpatient encounters on separate days
- OR 1 or more instances of active stand-alone medication (see table 8) reported during outpatient medication reconciliation³
- OR 1 or more Oral Glucose Tolerance Test (OGTT) 2-hour 75g result ≥ 200 mg/dl where there is NO DIAGNOSIS CODE on the same encounter indicating pregnancy (V22, V23)⁴
- OR 2 or more hemoglobin A1c results ≥ 6.5% on 2 different days within 730 day span
- OR 2 or more fasting glucose results ≥ 126 mg/dl on 2 different days within 730 day span
- OR 2 or more random glucose results ≥ 200 mg on 2 different days within 730 day span
- OR within a 730 day span on 2 different days:
 - Fasting glucose results ≥ 126 mg/dl
 - AND Random glucose results ≥ 200 mg
- OR within a 730 day span (can be same day):
 - Hemoglobin A1c results ≥ 6.5%

Abnormal Lab Results

Source:

Laboratory results

Definition:

Adult Durham Population patients who meet **ONE OR MORE** of the following criteria during a DukeMed encounter between 2007-2011:

- One or more instances of hemoglobin A1c results ≥ 6.5%
- OR one or more fasting glucose results ≥ 126 mg/dl within 365 day span
- OR one or more random glucose results ≥ 200 mg/dl within 365 day span

Abnormal HbA1c (NCY A1c Registry Definition)

Source:

Glycated hemoglobin laboratory results

Definition:

Adult Durham Population patients who meet **ONE OR MORE** of the following criteria during a DukeMed encounter between 2007-2011:

- One or more instances of hemoglobin A1c results ≥ 6.5%

July 2016-

PSQ Core-suggested additions
to the proposed guidance for
reporting results from
pragmatic trials.

*(Will be posted to Living Text
site soon...)*



Reporting Pragmatic Clinical Trials

Introduction

Transparent reporting of clinical trials is essential for helping researchers, clinicians, patients, and other stakeholders understand the validity and reliability of the findings. Many have suggested that the quality of trial reporting is suboptimal and have sought consensus on the key elements of transparent reporting. To address this, a group of clinical trial methodologists and journal editors developed the [CONSORT](#) (Consolidated Standards of Reporting Trials) Statement. CONSORT is intended to improve transparency and dissemination of trial findings by providing a checklist and guidance for authors.¹ The original CONSORT statement focused on the reporting of standard, two-group randomized controlled trials (RCTs) that compare an intervention with a control. Over the years, CONSORT has been expanded for clarity and revised, most recently in 2010, and now includes several official extensions to account for variations in trial design, interventions, and data (described in Appendix A).

Pragmatic Clinical Trials

The [NIH Health Care Systems Research Collaboratory](#) supports the design, execution, and dissemination of a set of [Demonstration Projects](#), which are pragmatic clinical trials (PCTs) that address questions of major public health importance and are part of an effort to create a new infrastructure for collaborative research within healthcare systems. In contrast to RCTs, which elucidate a mechanical or biological process, PCTs are “designed for the primary purpose of informing decision makers regarding the comparative balance of benefits, burdens and risks of a biomedical or behavioral health intervention at the individual or population level.”² To be clear, PCTs are on a *continuum* with traditional RCTs, and there are aspects of PCTs that make them either more explanatory or more pragmatic (described in Appendix B). Generally, a PCT is more pragmatic if the data are collected during routine clinical care (usually through the electronic health record); if there is some flexibility in the delivery of and adherence to the [intervention](#); if a real-world population is included; and if the outcomes are relevant to patients and other decision makers.

Purpose of this Template

This template is intended to help authors with the transparent reporting of their PCT. Though we have looked to the CONSORT guidance and extensions wherever possible, new areas are emerging related to PCTs that the CONSORT checklist and guidance do not address. These include, for example, reporting around the secondary use of EHR data, wider stakeholder and health system involvement in the conduct of PCTs, and special ethical and regulatory considerations for PCTs.

Specifications regarding data from EHRs or administrative systems

- “How the population of interest was identified. Researchers should explicitly reference any specific standards, data elements, or controlled vocabularies used, and provide details of strategies for translating across coding systems where applicable.”
- “Each clinical phenotype (EHR-based condition definition) used should be clearly defined and study reports should reference a location for readers to obtain the detailed definitional logic....The use of national repository for phenotype definitions, such as PheKB or NLM VSAC is preferred. GitHub or other repository for code...”
- “Process and results from assessment of the quality of the data (should be informed by Collaboratory PSQ Core recommendations for Data Quality)”
- “Data management activities during the study, including description of different data sources or processes used at different sites. (Note that the data quality assessment recommendations are particularly relevant to monitor data quality across sites that have different information systems and data management plans for the study.)”
- “The plan for archiving or sharing the data after the study, including specific definitions for clinical phenotypes and specifications for coding system (name and version) for any coded data.....”

Collaboratory Approach to Phenotype Definitions

Review existing definitions

Selection and planning

Implementation

Definitions on Collaboratory website

link to

Justification and guidance for use in Pragmatic Trials

link to

Human readable phenotype, collaboration, versioning, public dissemination

Phenotype Definitions Used



Populations:

Patients w/ chronic pain
Patients w/ imaging studies for lower back pain
Patients who are candidates for CRC screening

....

Confounders or Risks:

Diabetes
Hypertension

...

Outcomes:

Mortality
Suicide attempt

NIH Collaboratory Rethinking Clinical Trials™
Health Care Systems Research Collaboratory

Type 2 Diabetes Mellitus Phenotype Definitions

From the NIH Collaboratory Phenotypes, Data Standards, and Data Quality Core

Available at: <https://www.nihcollaboratory.org/Pages/Knowledge-Repository.aspx>

Background: The Phenotypes, Data Standards, and Data Quality Core of the NIH Health Care Systems Research Collaboratory is developing a series of recommendations for the collection/query of data from electronic health records (EHRs) and/or ancillary systems for person characteristics and clinical features to support standardized reporting of baseline characteristics of research populations in interventional and observational studies.

Purpose of this document: This document represents our synthesis of existing phenotype definitions that have been used in diabetes research and population health activities. Using guidelines for the evaluation of existing phenotypes, our informatics and EHR phenotyping experience, and specialized clinical/research expertise, we suggest a suite of phenotype definitions, each appropriate for a particular purpose. The following is our recommendation, complete with a justification and supporting information and resources, for explicit EHR-derived phenotype definitions for diabetes. However, neither the Collaboratory nor the NIH has formally endorsed these definitions or their use in the data collection or reporting of this condition at this time (see [disclaimer](#)).

Audience: This document and supporting information is directed to clinical researchers and research sponsors who are making decisions about the data to use for studies. These documents should provide specifications and guidance that will assist researchers in making informed and deliberate choices about EHR data to use in research studies. The supporting

PheKB a knowledgebase for discovering phenotypes from electronic medical records

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Phenotypes

Group: Any Include Methods: Exclude Methods: Mine Only: Any Apply

Title	Groups	Institutions	Data and Methods	Status
Atrial Fibrillation - Demonstration Project	Vanderbilt - SCIRD Group	Vanderbilt University	CPT Codes, ICD 9 Codes, Natural Language Processing	Final
Cardiac Conduction (QRS)	eMERGE Phenotype WG	Vanderbilt University	CPT Codes, ICD 9 Codes, Laboratories, Medications, Natural Language Processing	Final
Cataracts	eMERGE Phenotype WG	Marshfield Clinic Research Foundation	CPT Codes, ICD 9 Codes, Medications, Natural Language Processing	Final
Clopidogrel Poor Metabolizers	Denny's Group at Vandy, VESPA - Vanderbilt Electronic Systems for Pharmacogenomic Assessment		CPT Codes, ICD 9 Codes, Laboratories, Medications, Natural Language Processing	Final
Crohn's Disease - Demonstration Project	Vanderbilt - SCIRD Group	Vanderbilt University	ICD 9 Codes, Medications, Natural Language Processing	Final
Dementia	eMERGE Phenotype WG	Group Health Cooperative	ICD 9 Codes, Medications	Final
Diabetic Retinopathy	eMERGE Phenotype WG	Marshfield Clinic Research Foundation	CPT Codes, ICD 9 Codes, Medications, Natural Language Processing	Final

In the future....

link to

Standard code lists (VSAC) or executable code

RESEARCH

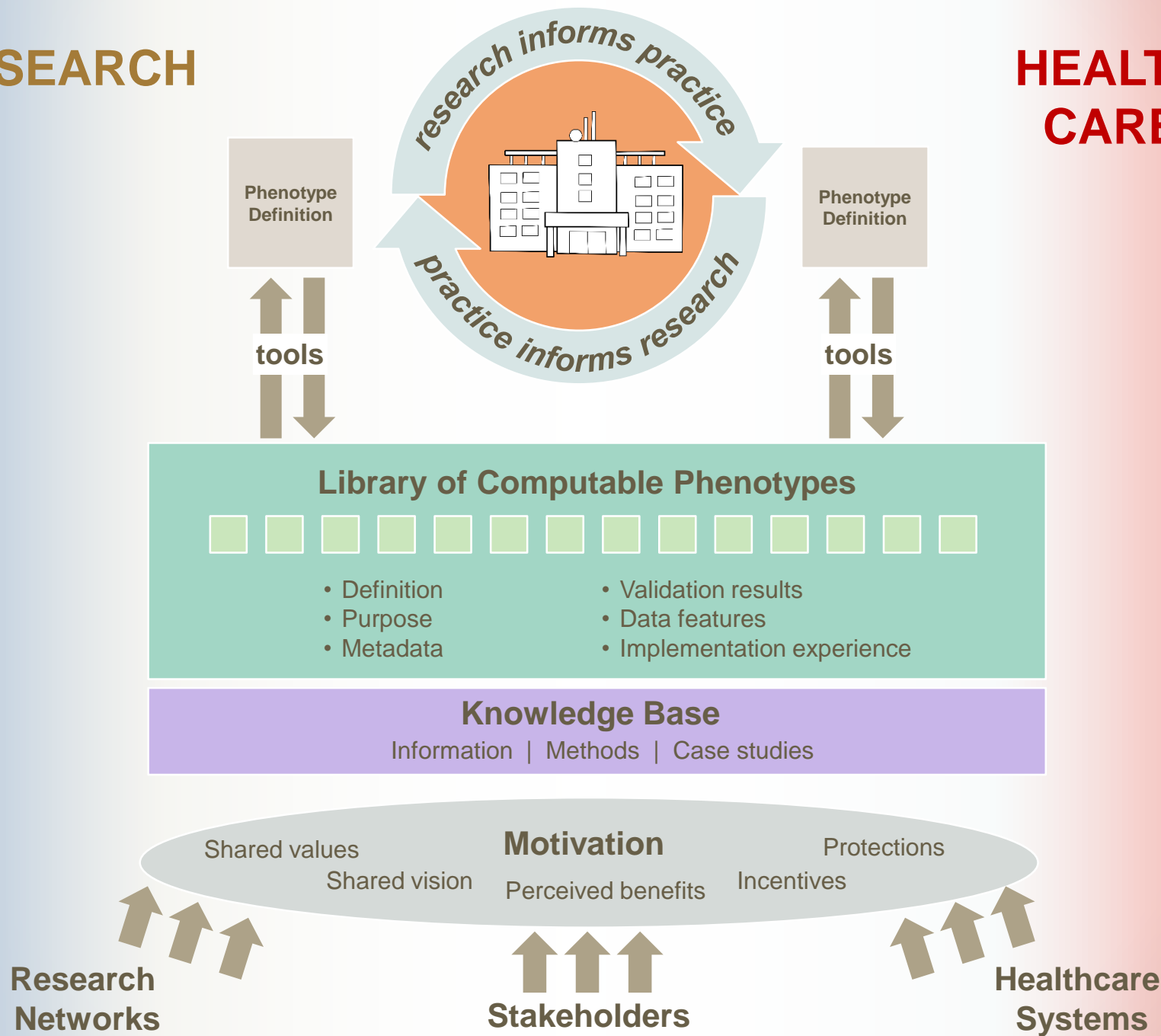
HEALTH CARE



- Ideally, *research and clinical definitions should be semantically equivalent.*
i.e., they should identify equivalent populations.

RESEARCH

HEALTH CARE



Path to Re-Usable Phenotype Definitions

- Access
- Evaluate and compare
- Facilitate use and reporting
- Explore incentives
- Engage:
 - Research sponsors
 - SDOs
 - Policy makers

“There is no real difference between work and play – it's all living.”

-Richard Branson



Blog



About



Résumé



GitHub



Archives

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RSS

The CRAP License

The [Community Research and Academic Programming License \(CRAPL\)](#), is an academic-strength open source license by the well-known professor [Matt Might](#). Its purpose is to encourage

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Terms:

- Any evidence of having been properly tested or verified is coincidental.
- You agree to hold the Author free from shame, embarrassment or ridicule for any hacks, kludges or leaps of faith found within the Program.
- You recognize that any request for support for the Program will be discarded with extreme prejudice.



Dennis Ideler

<http://dennisideler.com/blog/the-crap-license/>

Data Quality White Paper

- The use of population-level data is essential to explore, measure, and report “data quality” so that the results can be appropriately interpreted.
- Need adequate data and methods to detect the likely and genuine variation between populations at different trial sites and/or intervention groups.
- **Recommend formal assessment of accuracy, completeness, and consistency for key data elements.**
- **Should be described, reported, and informed by workflows.**



https://www.nihcollaboratory.org/Products/Assessing-data-quality_V1%200.pdf

Data Quality Recommendations: Use

- Have you read DQ recommendations and considered using?
 - 50% had read
 - 25% read upon contact for survey
 - 25% had not read/unknown
- Did you have DQ plans in place before you knew about the DQ recommendations?
 - 100% had DQA plans in place with application
- Have implemented or are in the process of implementing DQ recommendations?
 - 25% Yes
 - 75% NA or Have own plan
- Are you using a CDM?
 - 62.5% no
 - 25% yes Mini Sentinel, HMORN
 - 12.5% Project specific CDM

Data Quality Challenges

- Time-consuming
- Require population data (in addition to trial-specific data)
- Data retention requirements and related storage issues
- The cost of storage can be substantial
 - There are many storage options that impact cost, availability and completeness of data.
 - Medical record retention regulations are governed by state law and vary widely in terms of retention time requirements and the amount of information.

Areas of Impact

- Technical Challenges
 - Methods, tools, best practices
 - Measuring quality
 - Quantification of differences across populations
- Culture changes
 - Can we identify and endorse “good enough”?
 - Create culture of sharing and tools to support this

Dissemination

- “Living Textbook”
- Posters/presentations on Phenotype Template, and Methods for Development and Evaluation
- Manuscript (informatics journal) on EHR Phenotyping experience and strategies of Demonstration Projects

Future Plans

- Strategy for data standards
- ICD-9/10 (guidance for researchers)
- Cultural change/education/creativity regarding data quality
 - Getting specific about which quality dimensions are critical
 - Expecting data quality assessment
 - Comparison-based, i.e., data verification or reproducibility-based, i.e., multiple analyses on data from different sources
 - Using assessment results to answer how good is good enough?
 - Practicality versus perfection - how can we help draw some lines on the balance
- Integrate efforts and work products with other computable phenotyping initiatives (e.g., Big Data to Knowledge [BD2K], biosharing.org, CEDAR, Precision Medicine Initiative).

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