

Innovations in Research Approaches: Data Aggregation and Linkage

*NASEM Prepregnancy BMI and Gestational Weight Gain
Workshop*

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Stephanie A. Leonard, PhD, MS
Assistant Professor
Dunlevie Center for Maternal-Fetal Medicine
Stanford University School of Medicine

Stanford University

Disclosures

- No conflicts of interest

Outline

- Promises & pitfalls of studying prepregnancy BMI and gestational weight gain using big data sources:
 - Vital statistics data
 - Medical claims data
 - Electronic health record (EHR) data
- Conclusions & research gaps

Vital statistics data

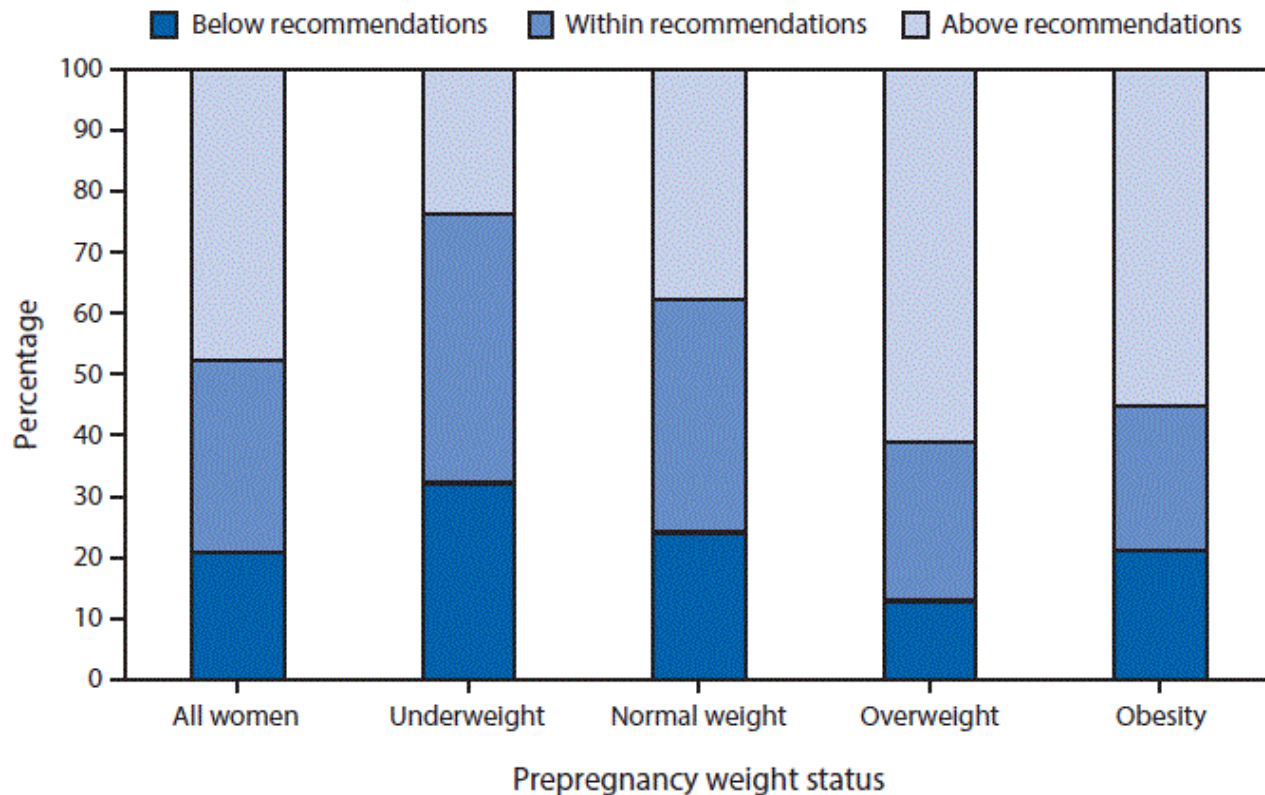
- Data are collected on Certificates of Live Birth and Fetal Death
- Typically, forms are completed during the birth hospitalization then reported to state health departments and the U.S. National Vital Statistics System of the CDC
- The CDC revised forms in 2003 to include fields for height and for weight prepregnancy and at birth
 - Fully implemented in the U.S. in 2016

31. MOTHER'S HEIGHT _____ (feet/inches)	32. MOTHER'S PREPREGNANCY WEIGHT _____ (pounds)	33. MOTHER'S WEIGHT AT DELIVERY _____ (pounds)
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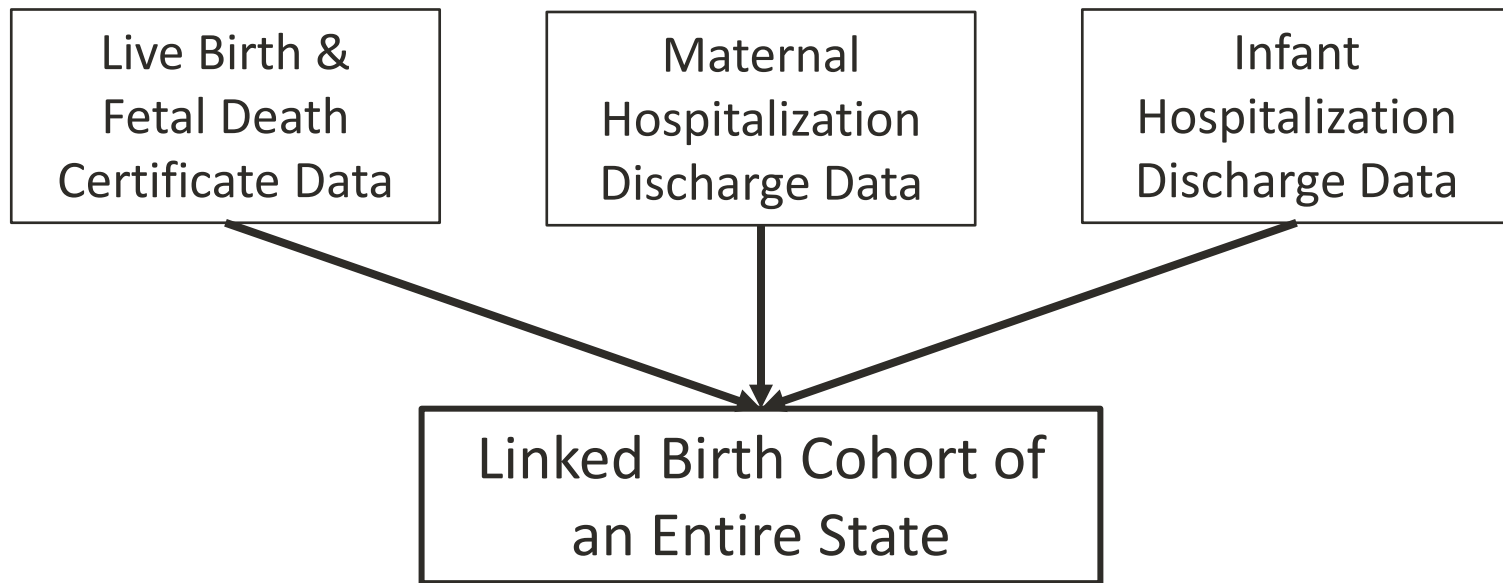
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Promises of vital statistics data

Large sample size and high generalizability

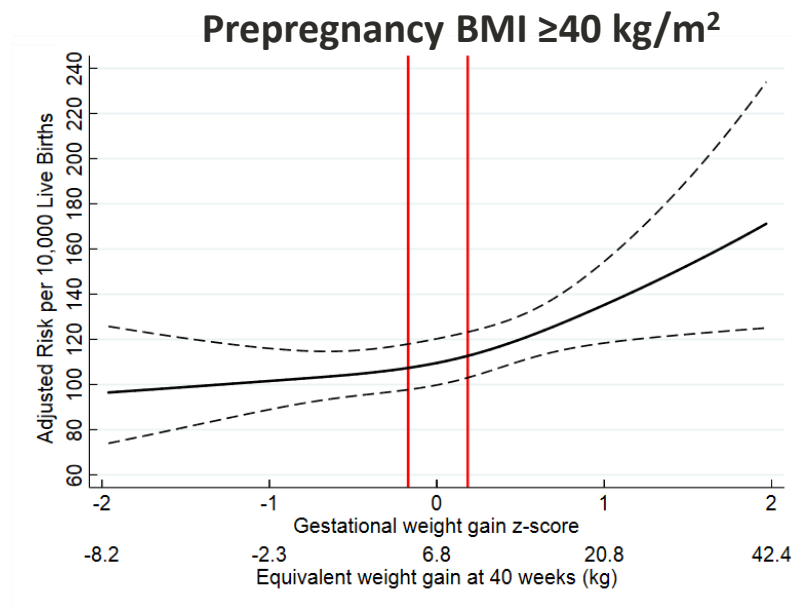
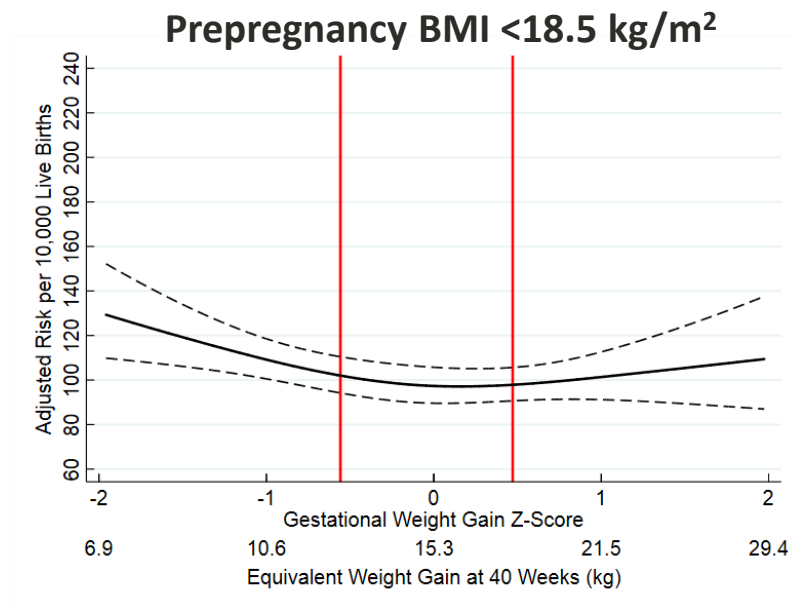


Ability to link to other data sources

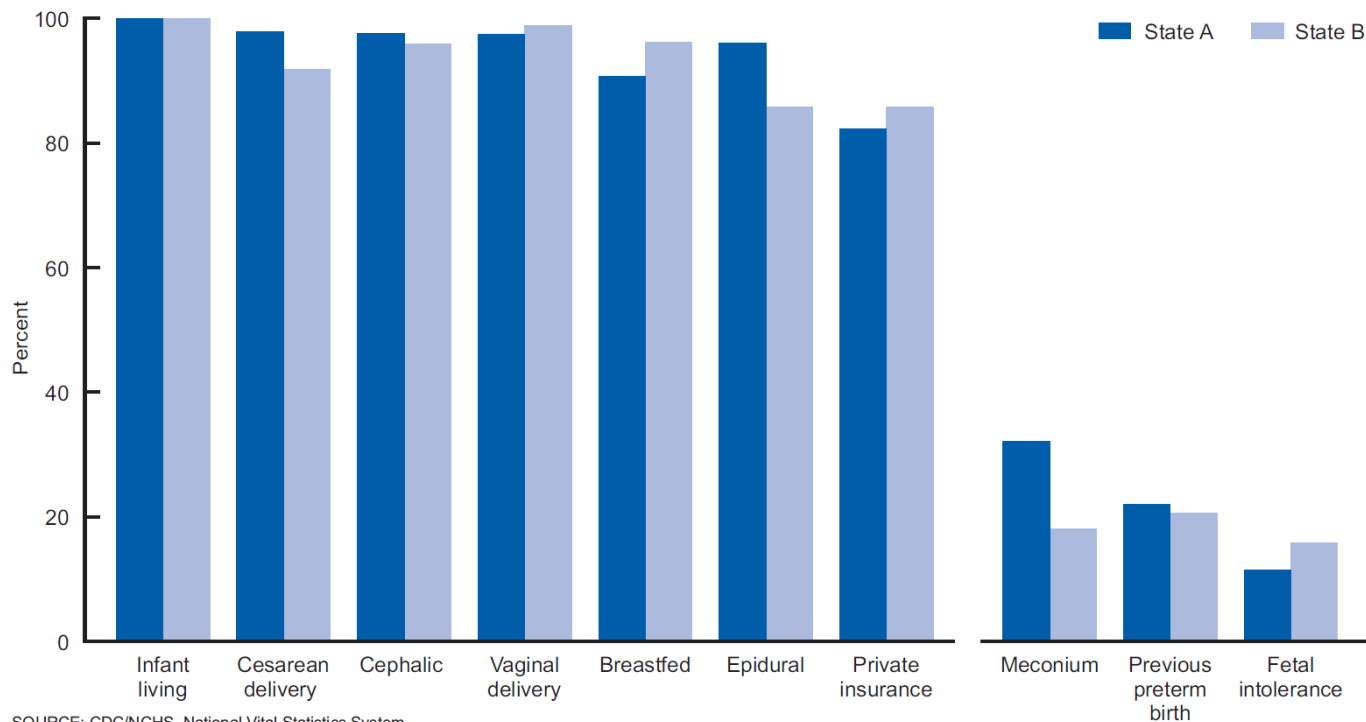


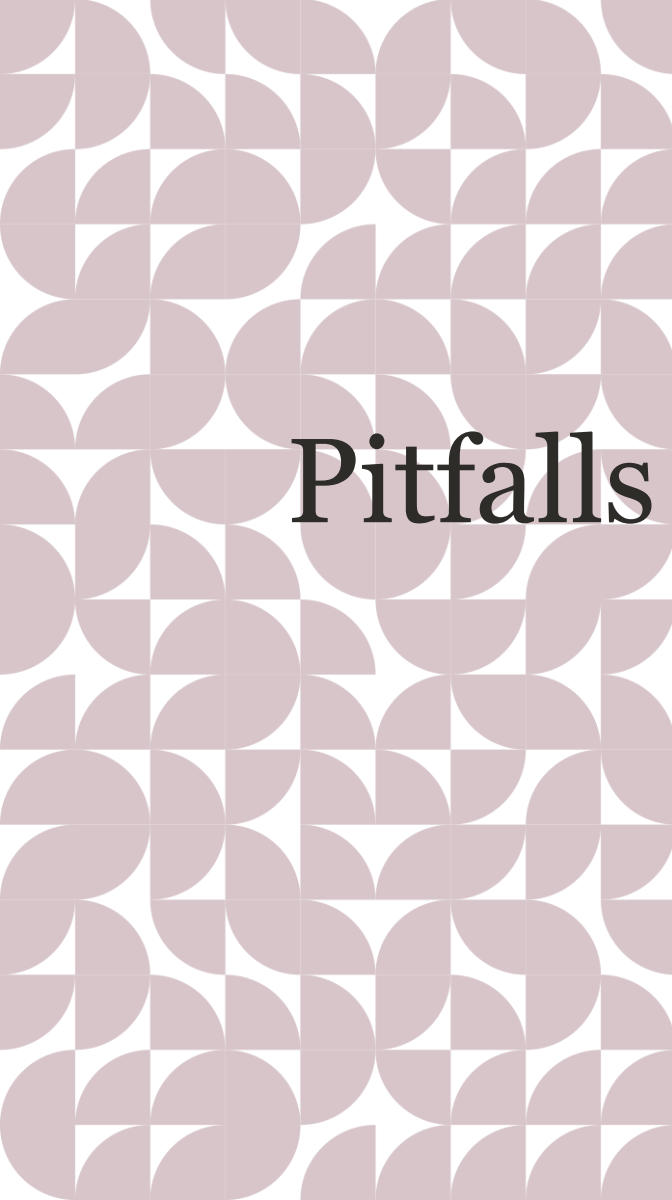
Strength in size, generalizability, and linkage

Example of assessing the risk of severe maternal morbidity across gestational weight gain values and stratified by prepregnancy BMI group in 2.5 million people



Multiple useful data fields have high validity





Pitfalls of vital statistics data

Validity of weight measurements

- Prepregnancy BMI and gestational weight gain reported in vital statistics data have been found to be overall consistent with medical records^{1,2}
- However, validity found to vary by patient characteristics^{1,2}
 - Separate studies have reported lower validity for Black patients than White patients
- Systematic review found people tend to underreport prepregnancy and delivery weight and overreport gestational weight gain → moderate misclassification of BMI/GWG categories that largely does not bias associations³

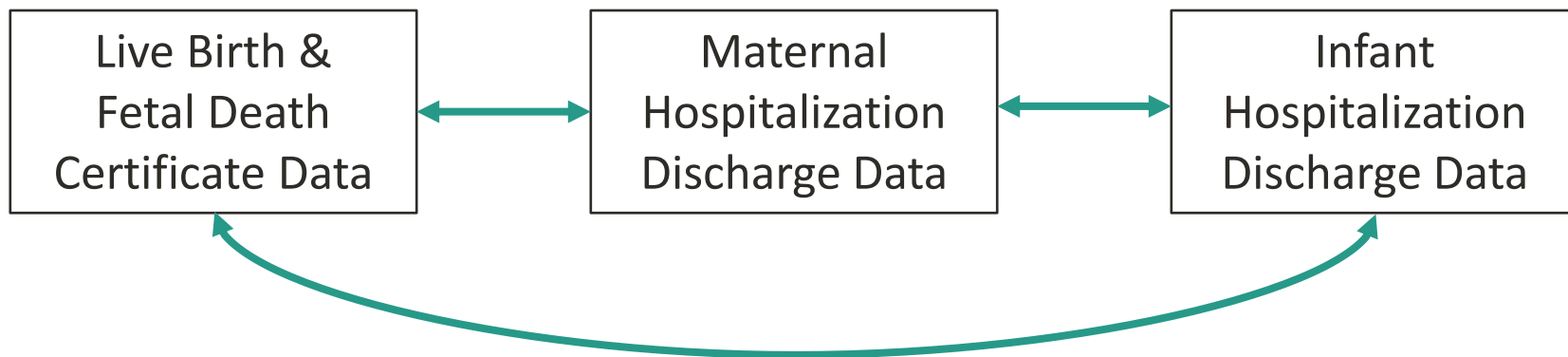
¹Bodnar LM, Abrams B, Bertolet M, et al. *Paediatr Perinat Epidemiol* 2014;28:203-212

²Deputy NP, Sharma AJ, Bombard JM, et al. *Epidemiology* 2019;30(1):154-159

³Headen I, Cohen AK, Mujahid M, Abrams B. *Obesity Rev* 2017;18(3):350-369

Challenges in linking to other datasets

- Valid clinical information in vital statistics is limited, but linkage to other datasets is typically probabilistic and often prohibitive



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Medical claims data

Medical claims data

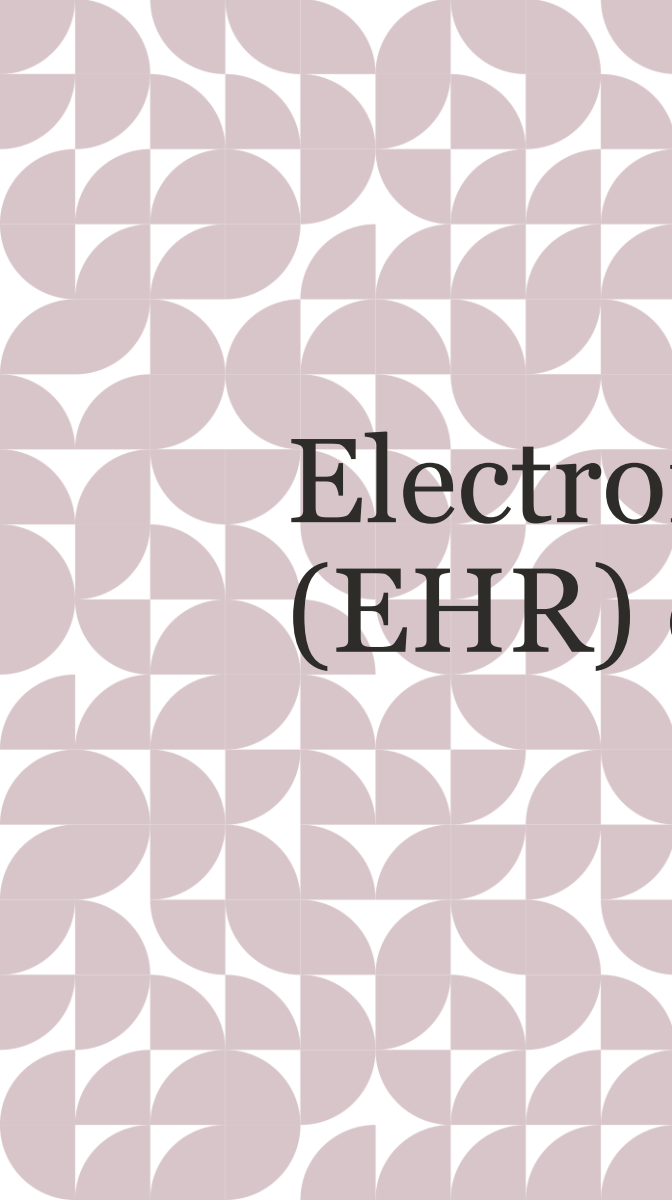
- Claims data include patient encounter information, including diagnoses and procedures
- Claims data may also include information on:
 - Billed and paid amounts
 - Demographics
 - Enrollment in an insurance program
 - Medications
 - Laboratory values
- Commonly used nationwide claims datasets include Medicaid research files, Merative™ MarketScan®, Optum Clinformatics, Veterans Affairs, National Inpatient Sample, others

Promises of claims data

- Large, generalizable patient groups
- Data often longitudinal
- Detailed information on diagnoses, procedures, and often medications
- Multiple types of pregnancy outcomes (not limited to live births)
- Can be linked to other datasets: vital statistics, EHR data

Pitfalls of claims data

- Limited information on weight, height, and demographic details
 - Often reliant on ICD-10-CM codes for BMI group
- Barriers to linking maternal-child pairs
- Lag time from data collection to research
- Often costly or difficult to obtain access
- Data management, storage, and analysis can be challenging

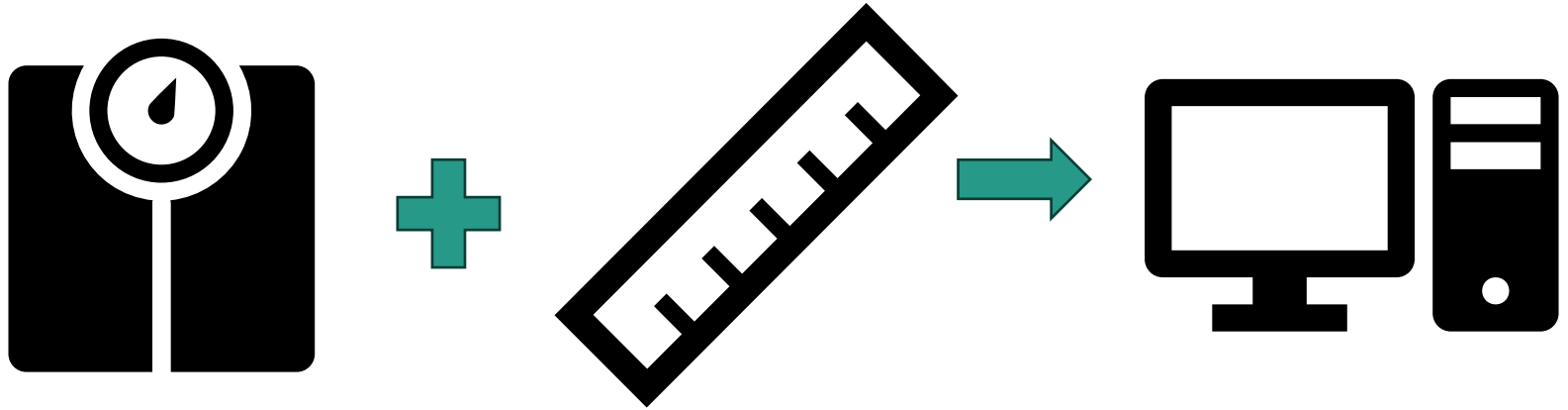


Electronic health record (EHR) data

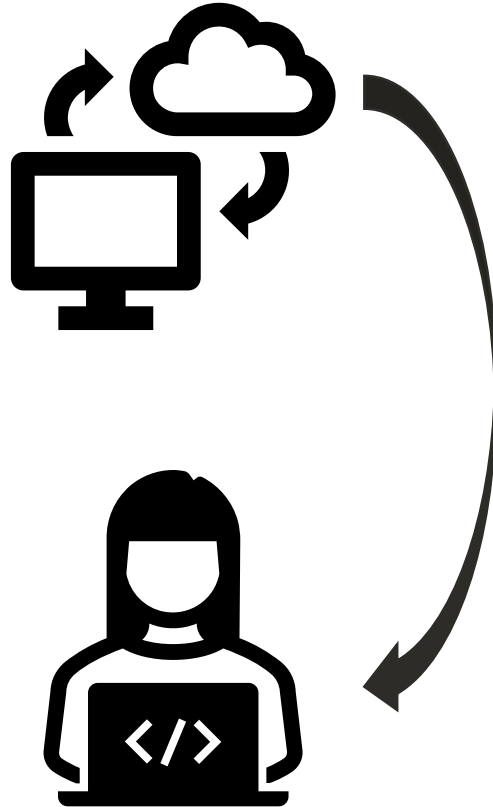
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Promises of EHR data

Weight & height measured by healthcare professionals & recorded in EHR system



Data aggregated and shared for research purposes



Stanford MEDICINE Research IT Technology & Digital Solutions

Cohort Discovery ▼ scweber ?

Searches Count Analyze Save for Review Upload Ids Go to Chart Review

DEMOGRAPHICS

- Current Age
- Gender
- Race
- Ethnicity
- Language

CLINICAL EVENTS

- Diagnosis
- Procedure
- Laboratory Test
- Drug Ingredient/Brand
- Drug Class
- Clinical Documents
- Encounter
- Admission
- Flowsheets
- Vital Status

TEMPORAL CONSTRAINTS

- Pair of Events

BIOSPECIMENS

- Precision Health Biobank 20

PATIENT LIST

- Saved Patient List

Untitled Cohort Query (modified) ~ 905 patients

Diagnosis of E84 Cystic fibrosis (1-10k) 1960-2023

from any source

age at event is any age

event date is any date

occurrence frequency is one or more

~ 1,755 patients

Lab result Hemoglobin [HGB] (>1M) 1990-today has numeric value low

age at event is any age

event date is any date

occurrence frequency is one or more

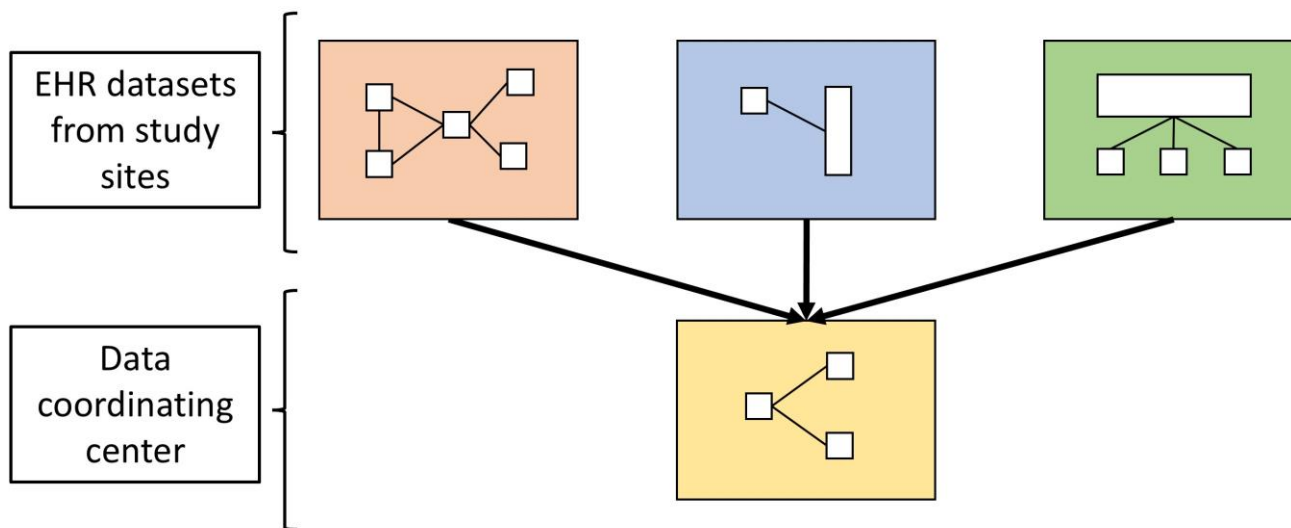
All Lab Orders Weekly Accumulation

Count (thousands)

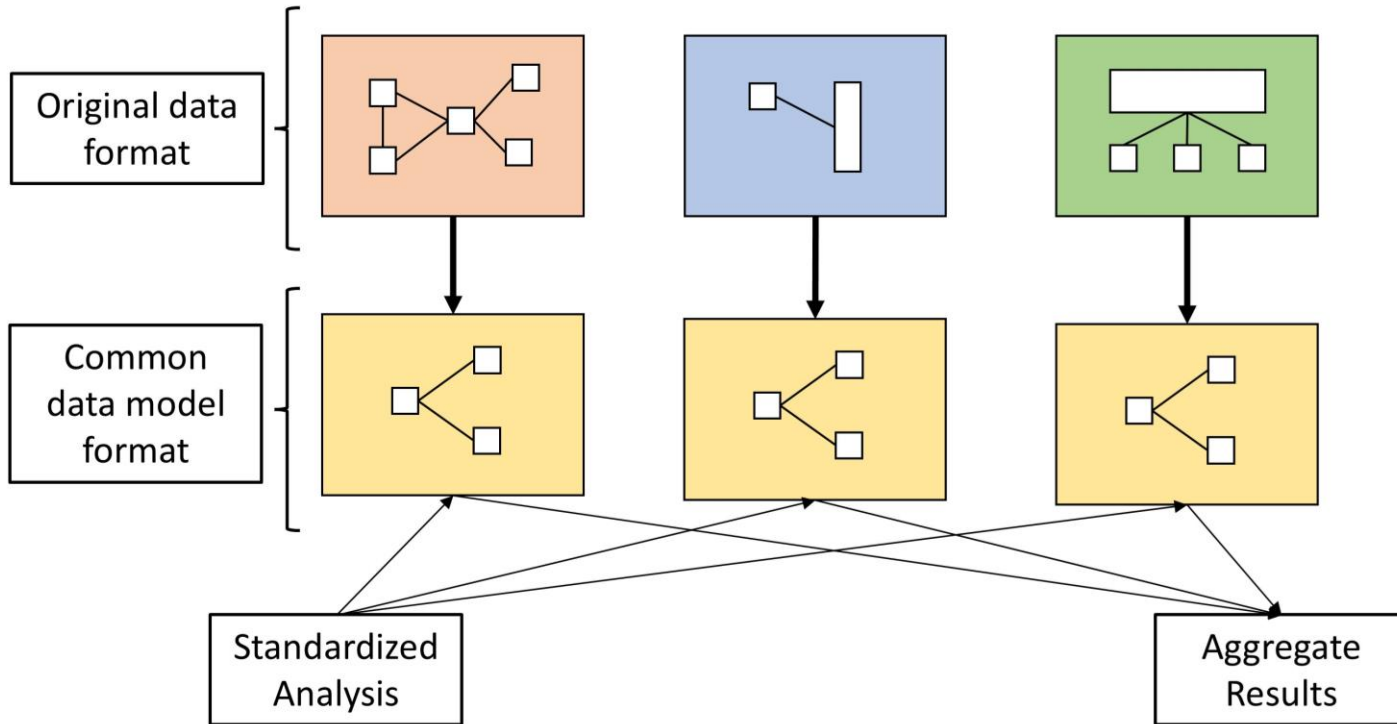
1990 1995 2000 2005 2010 2015 2020

There are approximately 688,734,754 labs since 2000.

Multi-site EHR cohorts can have higher generalizability and sample size



Distributed data network studies leverage common data models for reproducibility & rigor: OHDSI



Centralized EHR datasets have massive sample sizes: Epic Cosmos



Cosmos Community



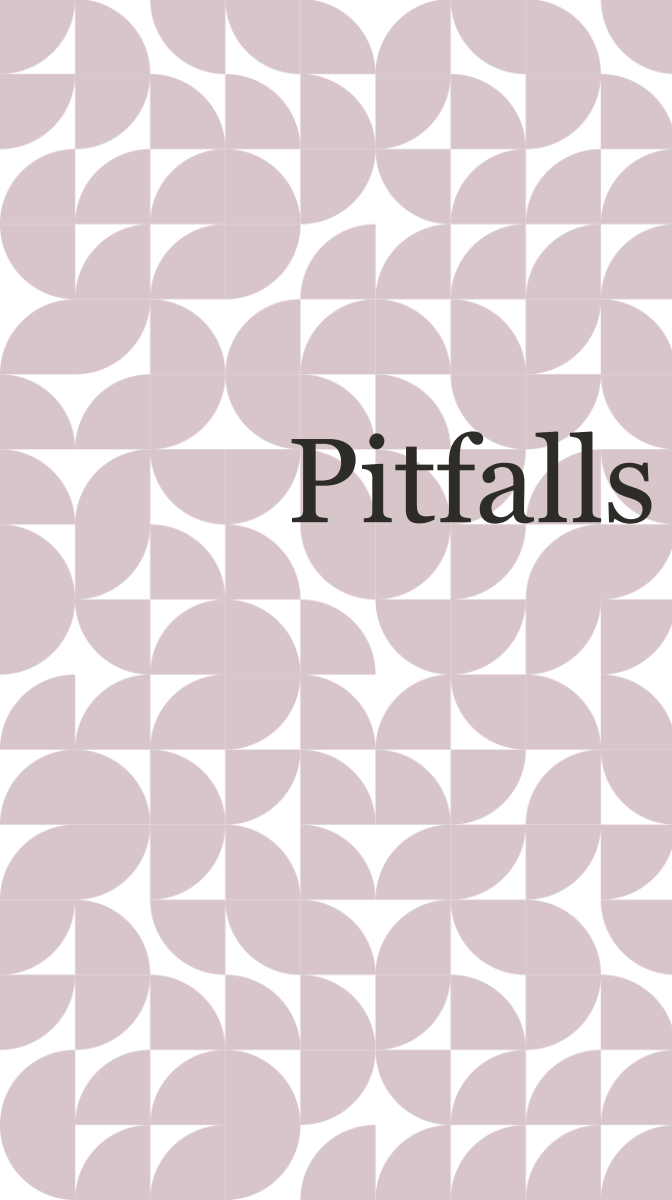
1,762
Hospitals



300 Million
Patients



40.7K
Clinics



Pitfalls of EHR data

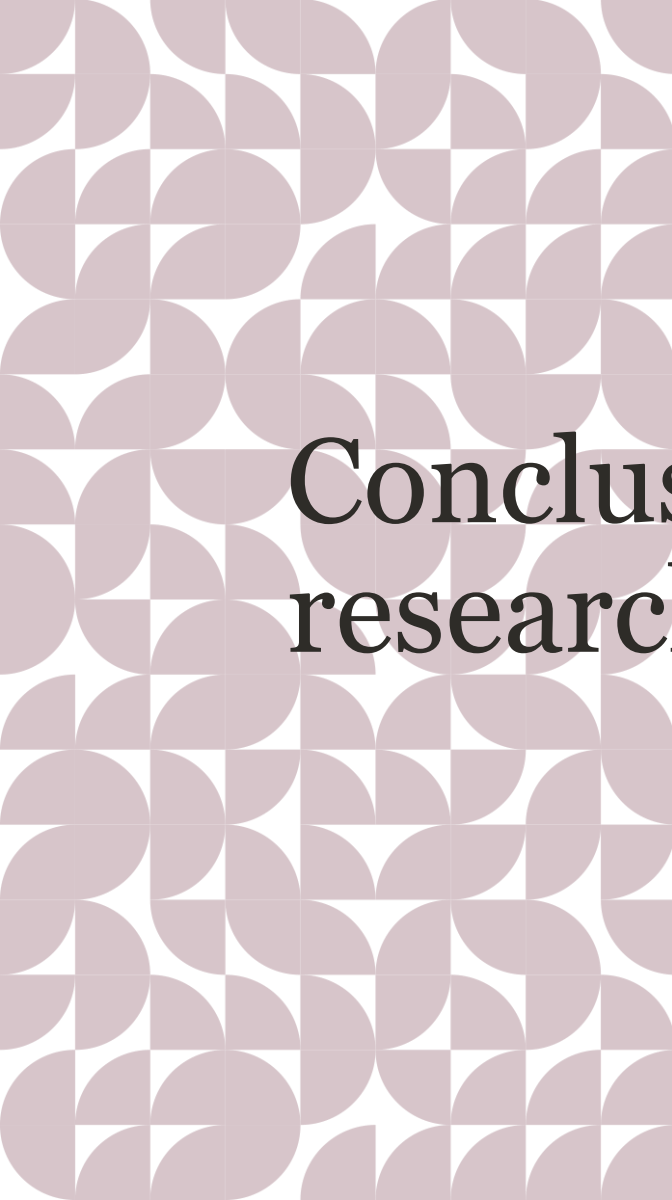
Multiple potential sources of bias

- Missingness of weight & height data can be high and differentially higher in healthier people^{1,2}
- Challenging to identify and correct erroneous values³
- Weight is not universally measured at start and end of pregnancy
 - Weight at first prenatal care visit is often used as proxy measure of prepregnancy weight, and weight at last prenatal care visit as proxy measure of delivery weight
 - Depending on visit timing, this could result in biasing prepregnancy BMI up and gestational weight gain down

1. Rea S, et al. *AMIA Jt Summits Transl Sci Proc* 2013;214-218

2. Baer HJ, et al. *JAMA Internal Med* 2013;173(17):1648-1652

3. Guide A, et al. *J Biomed Informatics* 2024; 104660



Conclusions and research gaps

Conclusions

- Big data sources have advanced in recent years and have important strengths of large size and generalizability
- Researchers and policy makers should consider the strengths and weaknesses of big data sources for prepregnancy BMI and gestational weight gain

Research gaps

- Updated validation studies for trajectory of weight from before to the end of pregnancy in different big data sources
- Methods to improve the internal validity of research using maternal weight, height, and BMI values in big data sources



Thank you

stephanie.leonard@stanford.edu
<https://med.stanford.edu/leonard>