# Building a BH Data Repository: Tips and Tricks from the Trenches

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## **About Me**



- Researcher (2007-present)
  - Duke SoM, Department of Biostats & Bioinformatics
  - Informatics to enable precision medicine in mental health



- Chief Data Officer, NC DHHS (2019-present)
  - Departmental data strategy
  - Data Office 4 pillars



#### Textbook Co-editor

### A tale of 3 repositories

- COVID Data
- NC DHHS Whole Person Health
- NC Longitudinal Data System

# Data Repository #1: COVID

## **COVID Data Repository**

- Purpose and structure:
  - Streamline and automate COVID reporting
- Process used to implement data repository system
  - Cloud-based
  - Speed- stood up in 48 hours
- Management structure
  - Inter-divisional data ownership with central data custodian
- Outcome of efforts
  - BIDP: Business Intelligence Data Platform

SITUATION ROOM

#### SITUATION ROOM

# e of North Carolina GENCY OPERATIONS CE

E.

Declaration of State of Emergency *March 10, 2020* 

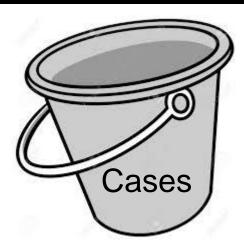
#### **Early Actionable Questions to Enable Data Driven Policy**

- How many cases will we see? When will our "epi curve" peak?
- How much (extra) PPE is needed, and where can we get it from?
- Will we run out of hospital beds? ICU beds? Ventilators?
- Should we shut down bars and restaurants? Schools? Businesses?
- How can we support families who are unable to work (either because workplace is shut down, or childcare is unavailable)?



### Buckets of NC DHHS COVID-19 Data





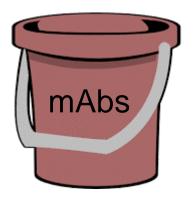




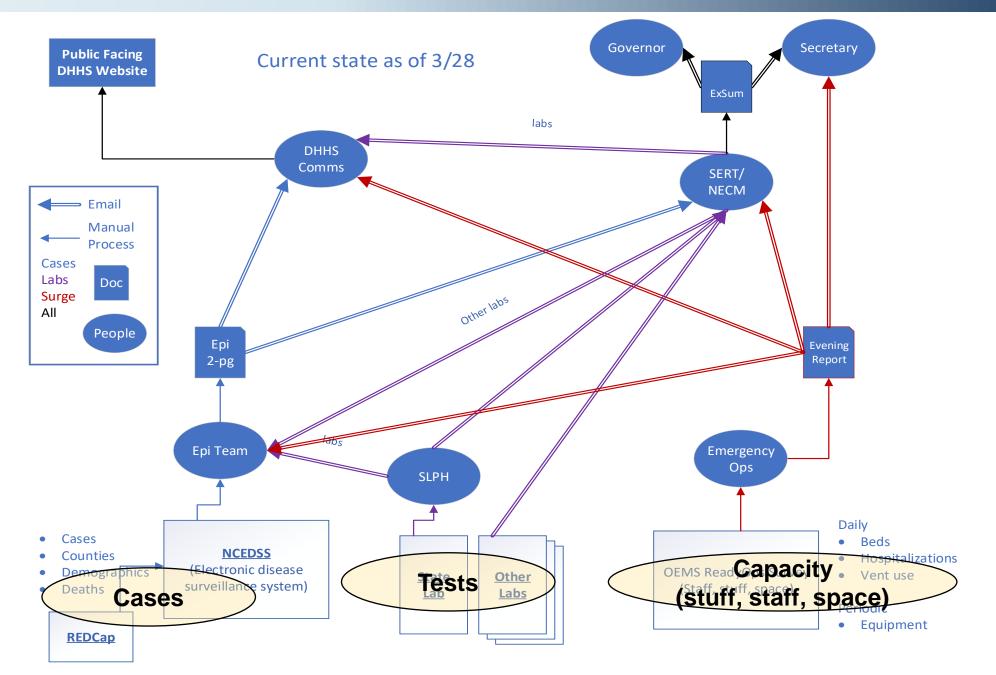




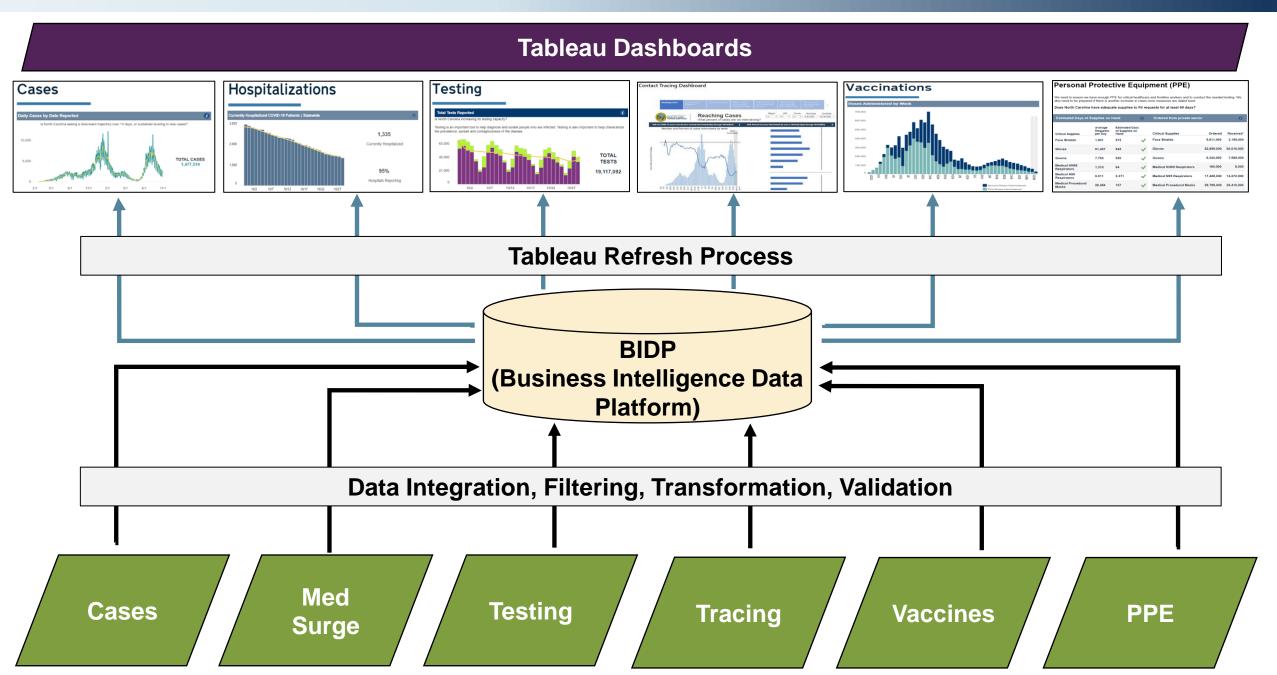


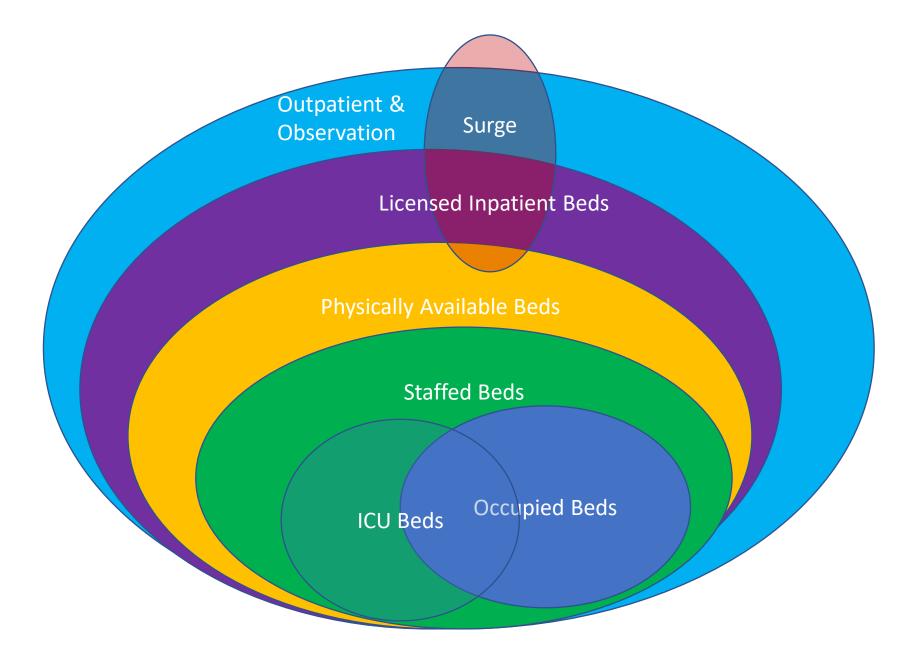


#### **Baseline Data Flow (as of March 2020)**



#### **Today's Data Flow**





Data Repository #2: Whole Person Health

### **NCDHHS Whole Person Health**

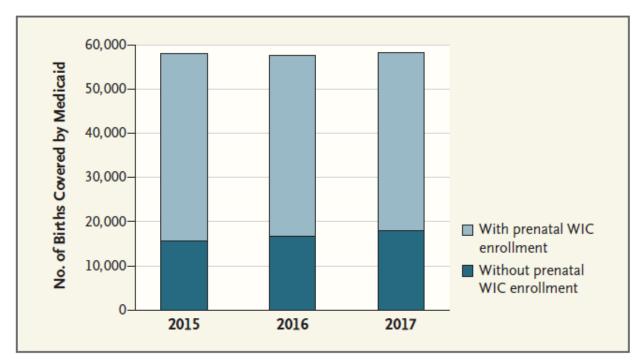
- Purpose and structure:
  - 360 Degree View of NC residents served through NCDHHS
- Process used to implement data repository system
  - Build upon BIDP from COVID
- Management structure
  - Inter-divisional data ownership with central data custodian
- Outcome of efforts
  - Enriched dataset for cross-enrollment analysis
  - More ongoing...



#### Perspective (FREE PREVIEW)

# Focusing on Population Health at Scale — Joining Policy and Technology to Improve Health

Aaron McKethan, Ph.D., Seth A. Berkowitz, M.D., M.P.H., and Mandy Cohen, M.D., M.P.H.



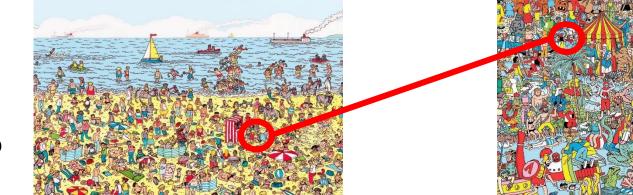
Medicaid-Covered Births with and without Concurrent Prenatal WIC Enrollment, North Carolina.

## In order to gain actionable knowledge, need 2 things:

1. The ability to integrate data between divisional silos ("Data Integration")



2. The ability to identify and link the same individual from different datasets ("Entity Resolution")

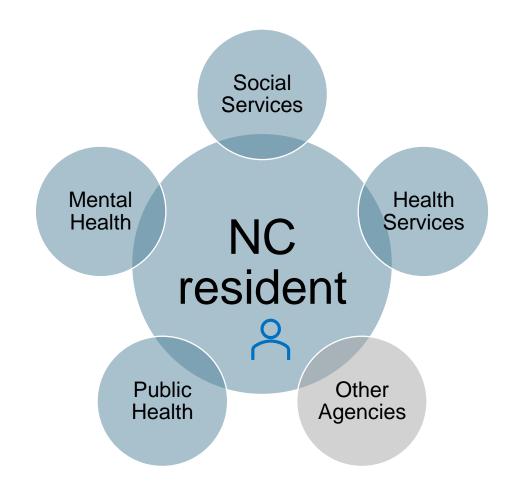


ID: 123456 Name: Waldo

ID: 123456 Name: Waldo

## **Whole Person Health**

- Goal: link data to facilitate a "Whole Person Health" view of the people we serve.
  - Real-time individual level
  - Aggregate analysis to inform policy
- Requires the ability to integrate data across divisional silos
- Which requires ability to link records between systems

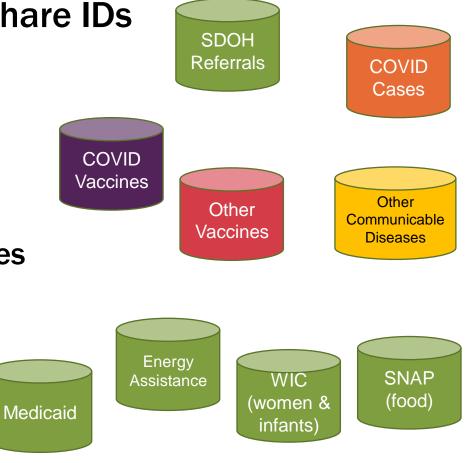


#### **Motivating Questions (Examples)**

- What % of Medicaid beneficiaries have been vaccinated?
- Which children in foster care have prescriptions for >4 psychotropic medications?
- Who receives regular food assistance referrals and therefore may benefit from SNAP but is not enrolled?
- What % of people experiencing homelessness have been vaccinated?
- What is the relationship between early grade outcomes (e.g., third grade reading) and different early childhood conditions (e.g., early learning, health, housing, child welfare)?

### Answering those questions is currently difficult at best

- Data live in silos that (mostly) do not share IDs
- Probabilistic match is possible, but
  - Labor-intensive
  - Prone to error
  - Example: initial approach for post-vax cases



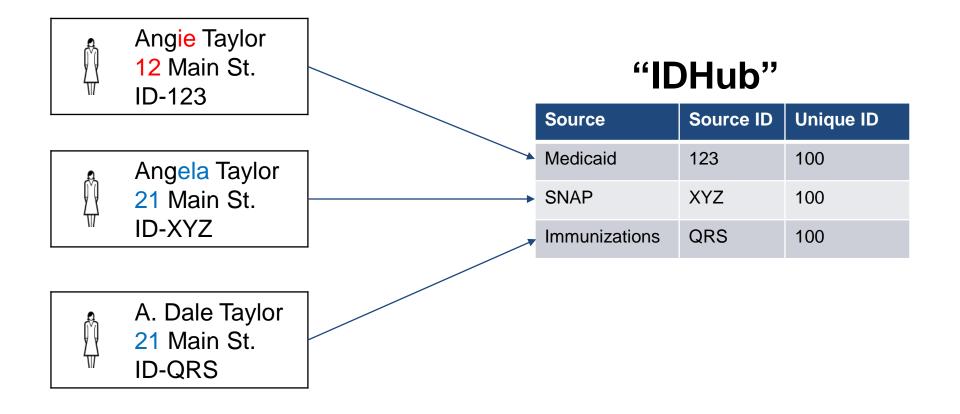
## **Options for "solving" entity resolution**

- 1. Each system uses its own ID, probabilistic "fuzzy match" between systems as needed
- 2. One universal ID, e.g. state-wide or National Health Identifier
- 3. Somewhere in between
  - a. Multiple "standard" identifiers
  - b. Map them to each other!

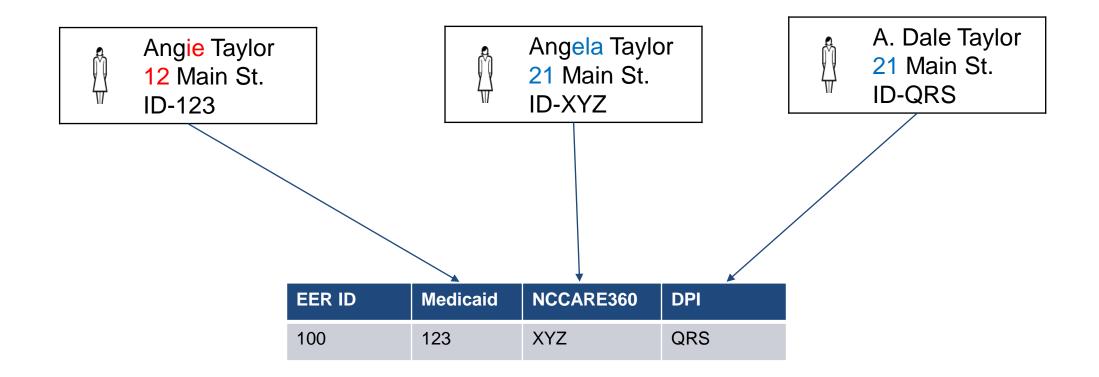


c. Refer to that mapping for efficient data integration

#### Map each ID to a universal unique identifier...



...enabling mapping each separate ID to the others



### IDs can (and do!) change over time

 MPI's (master patient index) are frequently merged as more data are incorporated

Jessie	Jessica
Address A	Address B
MPI 111	MPI <del>222 -</del> 111

Jessie Address B MPI 111

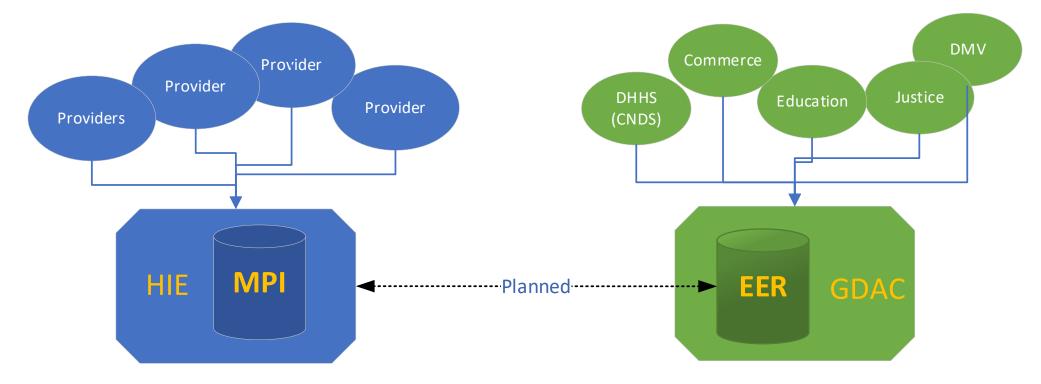
 Splits (after a "false positive" match) far rarer and more challenging to handle

## A few things about the NC Landscape

- State-run HIE- "NC HealthConnex"
- Government Data Analytics Center (GDAC)- a Division of NC's Department of Information Technology
  - Created by legislation for data management and analytics
- NC is home to SAS



### NC has 2 "Universal" Identifiers: clinical vs. non-clinical



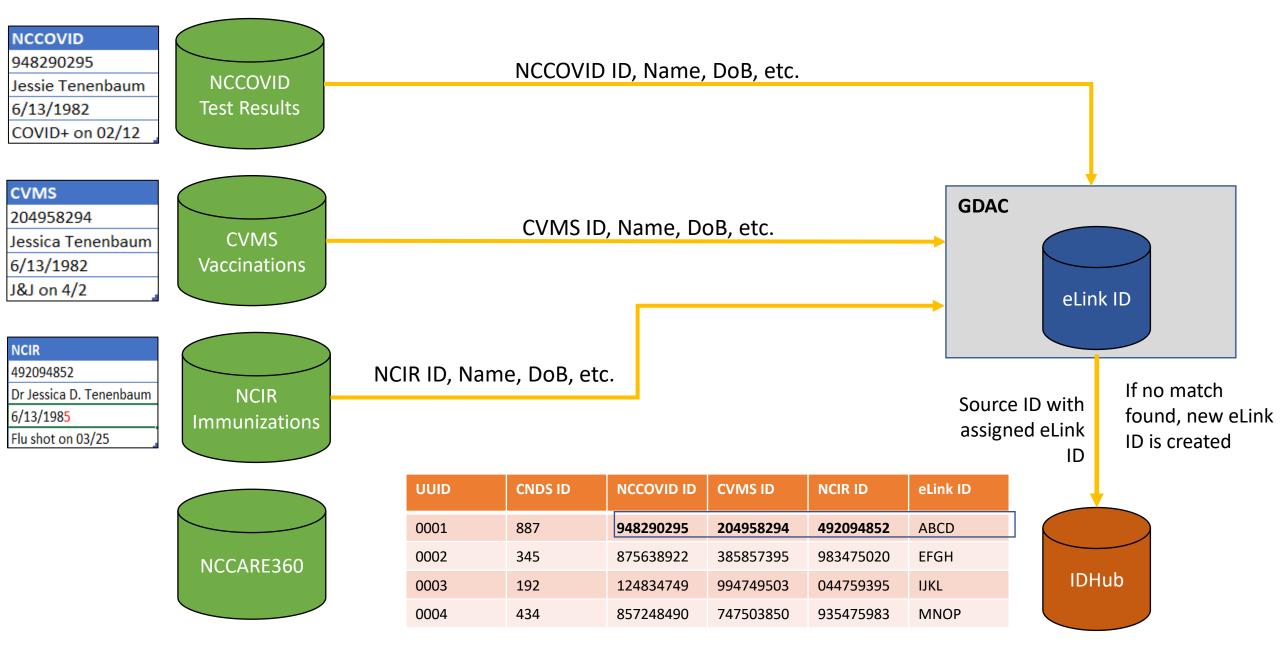
MPI ID (Master Patient Index)

- Used by Health Information Exchange (HIE) for <u>clinical data</u>
- Involves merging clinical records
- High stringency matching

**EER** ID (Enterprise Entity Resolution)

- Used by NC's Government Data Analytics Center (GDAC) for <u>non-clinical data</u>
- Slightly lower bar for matching
- Meant for analytics

#### NC DHHS Source Systems send identified attributes to ID authoritative source



# **Other options: 3rd party vendors\***

- As a service
  - Data sent externally- needs DUA
  - Leverages consumer data
- Off the shelf software
  - Subset of Master Data Management
  - Internal use
  - Increased control





Informatica

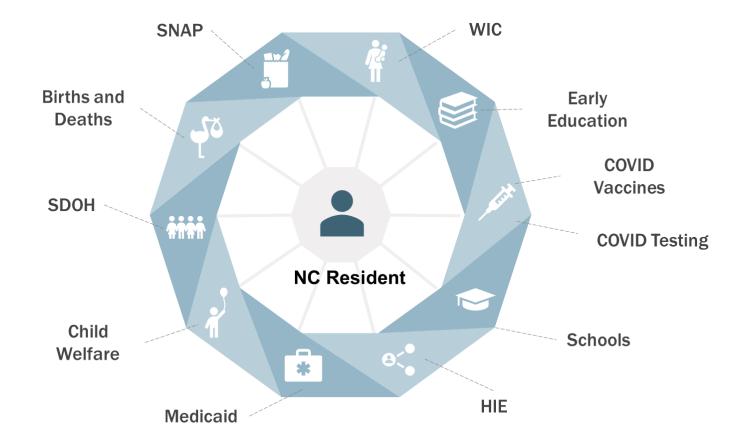






#### \*No endorsement intended or implied!

#### ...enabling Whole Person Health



Data Repository #3: NC Longitudinal Data <del>System</del> Service

# **NC Longitudinal Data Service (LDS)**

- Purpose and structure:
  - Longitudinal view from early childhood to workforce
  - System of systems
- Process used to implement data repository system
  - Roadmap built over many years
- Management structure
  - Exec Director with Governance Board (on which I serve), multiple
  - committees
- Outcome of efforts
  - Development and governance still evolving

# **Insights and Challenges**

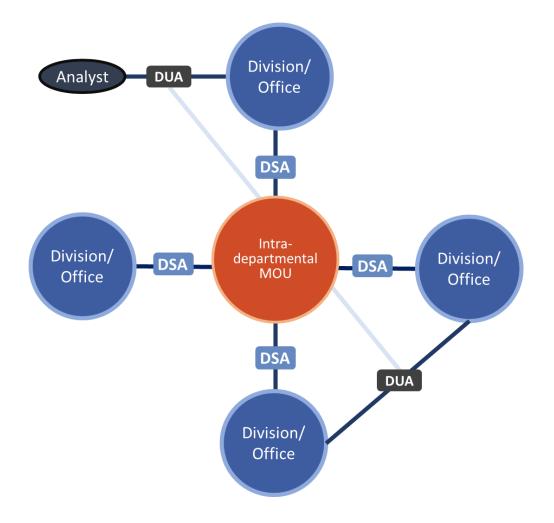
#### **Be Use Case Driven**



## **Governance is hard**

- Trust is key
- Involve legal counsel from the start
- Opt out considerations
- Data quality
  - Check!
  - Fit for use
- Legal framework

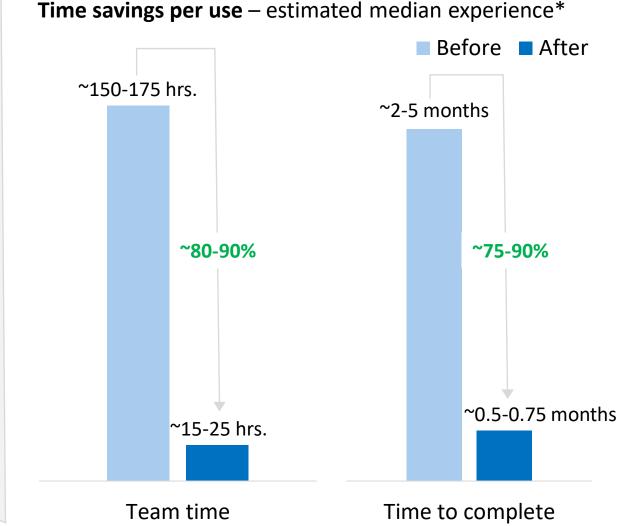
#### **Foundational legal agreements: visual representation**



#### **INTRA-DHHS DATA SHARING: BENEFITS OF NEW LEGAL / DATA GOVERNANCE FRAMEWORK**

#### **Key benefits**

**1.** Clarifies requirements and guidelines (e.g., who the permitted signatories are) 2. Provides approved language & **templates** for agreements, preventing rework while also **mitigating risk** Saves team time & effort for business 3. and legal, often by not requiring an additional Data Use Agreement Gets to data insights & program action 4. faster, given quicker time to completion



Team time Time to con

Illustrative

#### **Lessons Learned**

- This stuff is hard- the devil is in the details, and the edge cases
- Data standards are important
- Takes longer and costs more than one might expect
- Do agreements early!
- Get stakeholder buyin
- Concrete wins can help keep momentum
- Motivate business support through targeted use cases

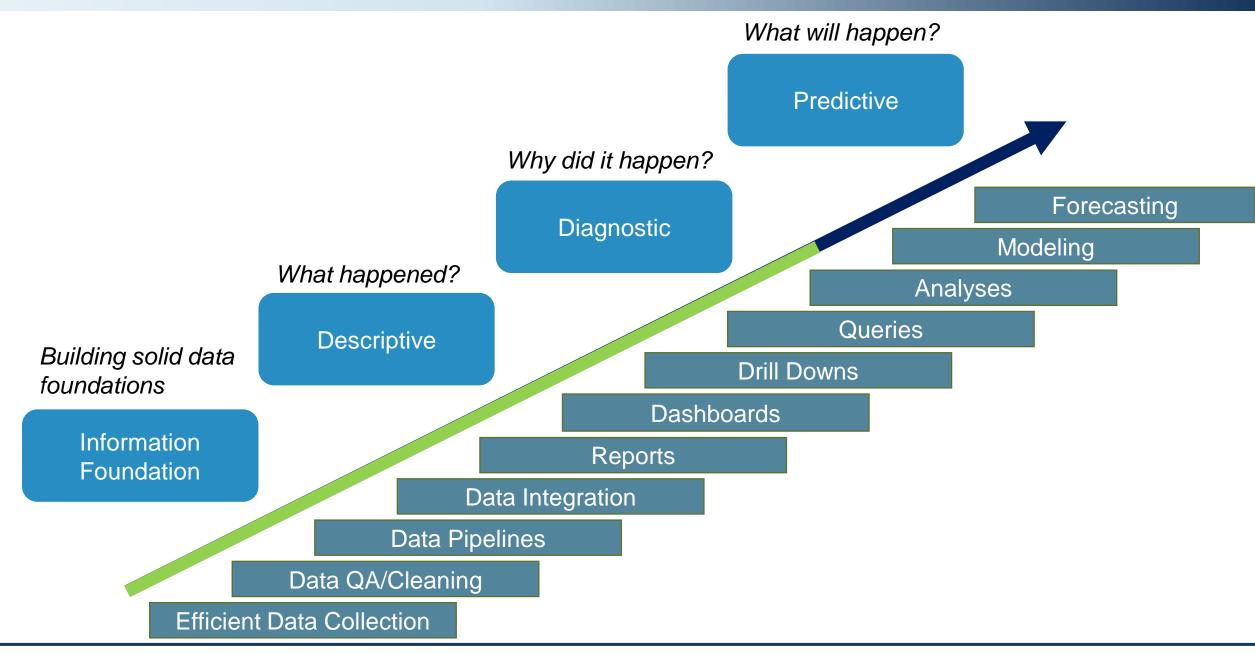
# **Thanks!**

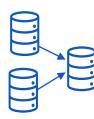
@jessiet1023

## **Data Pillars**

- Data Infrastructure: Technology used to store, exchange, and access data
- Data Governance: People, processes, and technology for data quality, security, management, and access
- Data Use: Reports, visualization, and analysis
- Data Literacy: Workforce training across all levels of baseline knowledge

#### Moving from descriptive to predictive





Limited availability of data integrated across programs for whole-person health

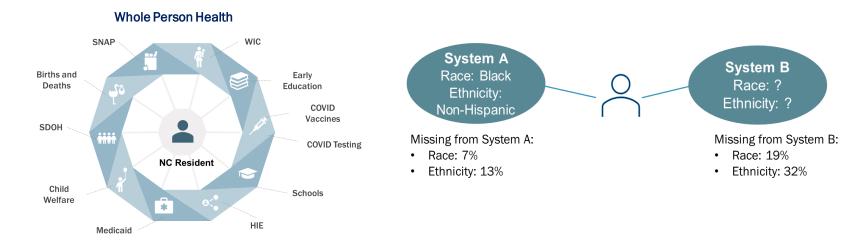


**Reusable integrated data:** Data structures have been built to integrate data across programs in a way that enables repeatable and extensible analysis of whole-person health data

#### **Objectives**

Leverage data integrated across programs for 1) enabling repeatable and extensible whole-person health use cases and 2) equity analysis and action

# Some datasets have relatively complete data on race and ethnicity, others do not.



Whole person view enables "filling in the gaps."

