

Integration of Data Streams to Augment Genomic Data

Accelerating the Use of Pathogen Genomics: a workshop

National Academies

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On Behalf of the N3C Consortium



Why N3C?

- Urgent need for observational data at scale
- In the US, there is no centralized healthcare, and therefore no centralized healthcare data
- Data from a single person is spread across multiple providers across time and geography

Fragmentation of COVID-19 Patient Clinical Data





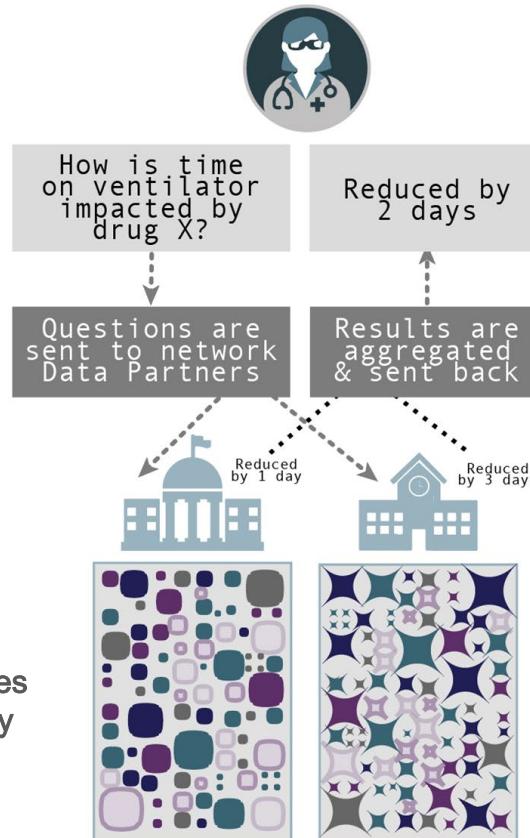
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COVID
Cohort
Collaborative

Federated and centralized approaches are synergistic



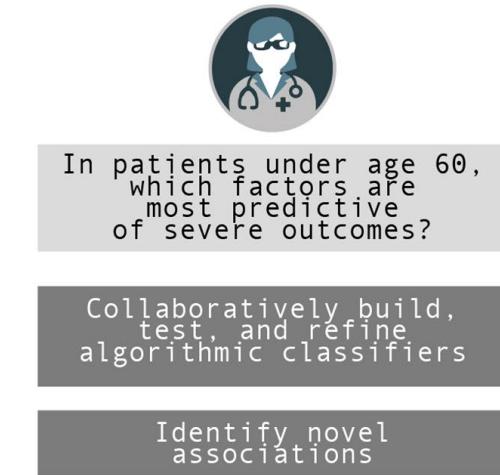
<https://bit.ly/3eMIO1j>

Federation is good for specific queries



For the new disease that is COVID-19, we needed centralized analytics

Centralized analytics are good for discovery





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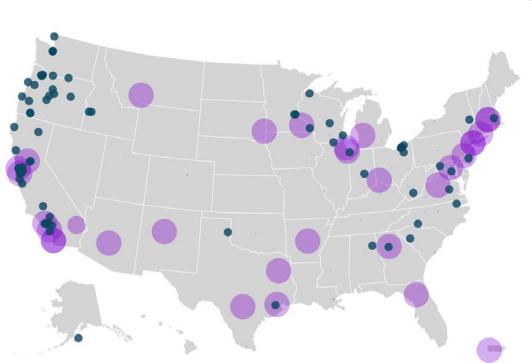
What is N3C?

National COVID Cohort Collaborative

N3C: largest national public HIPAA-limited data set in US history

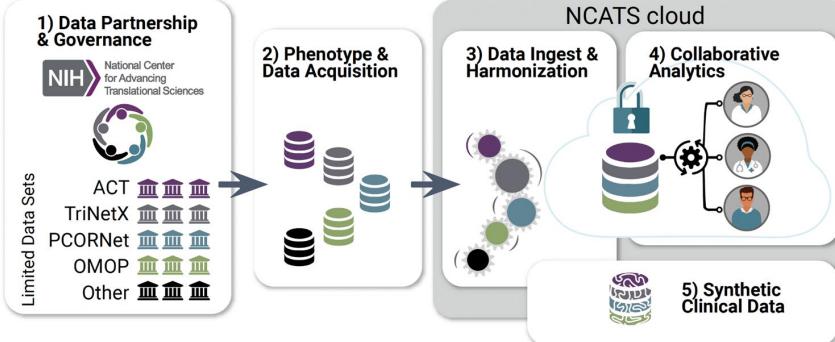
22.8 M records from >230 institutions

covid.cd2h.org/dashboard



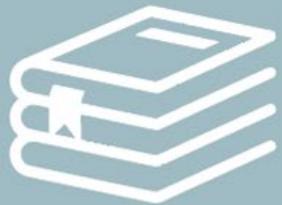
- ✓ **Representative pan-US cohort:** race, ethnicity, gender, geography, socio-economic status, health background
- ✓ **Harmonized:** overcomes source data heterogeneity
- ✓ **Linked:** Patient records, viral variants, vaccine data, CMS, environment, SDoH, etc.
- ✓ **Public Health Surveillance:** new variants, comparative effectiveness of drugs

N3C's data-model agnostic, data harmonization and QC pipeline



>450 organizations participating, >4,600 users

Diverse impact of N3C collaborative analytics



RESULTED IN
SIGNIFICANT
SCHOLARLY
PRODUCTIVITY

ATTRIBUTED AT
SCALE AND
INCENTIVIZED
COLLABORATION

TRANSFORMED
CARE
GUIDELINES

DEVELOPED
EVIDENCE-
BASED
DISEASE
DEFINITIONS

DEVELOPED
COMPLEX RISK
PREDICTION
MODELS

bit.ly/n3c-google-scholar >4300 citations, H index of 30



Ensuring harmonization rigor with scalable, continuous monitoring and provenance



Complete transparency into lineage/provenance of harmonization pipelines for >232 sites (77 DTAs) and >50,000 transforms



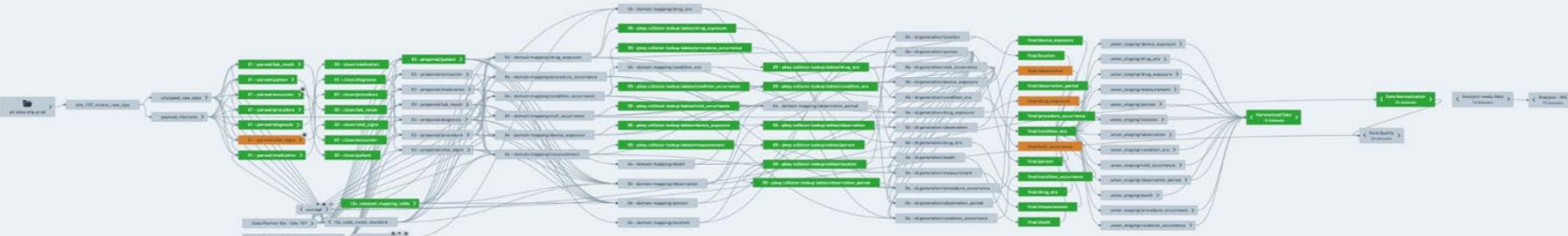
Pipeline versioning, deployment, upgrades, and automated data quality checks of new and existing sites



Curators and developers can quickly identify and address issues



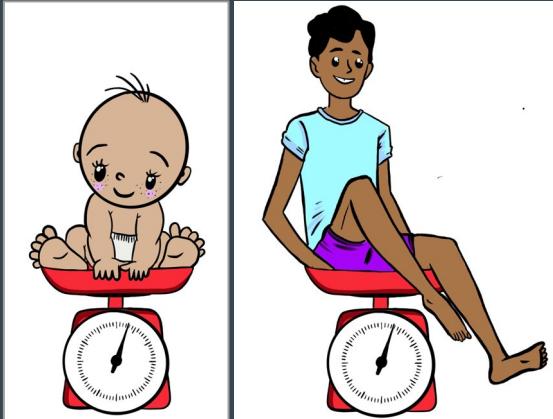
Scalability of compute resources; pipelines can be refreshed in <20 mins



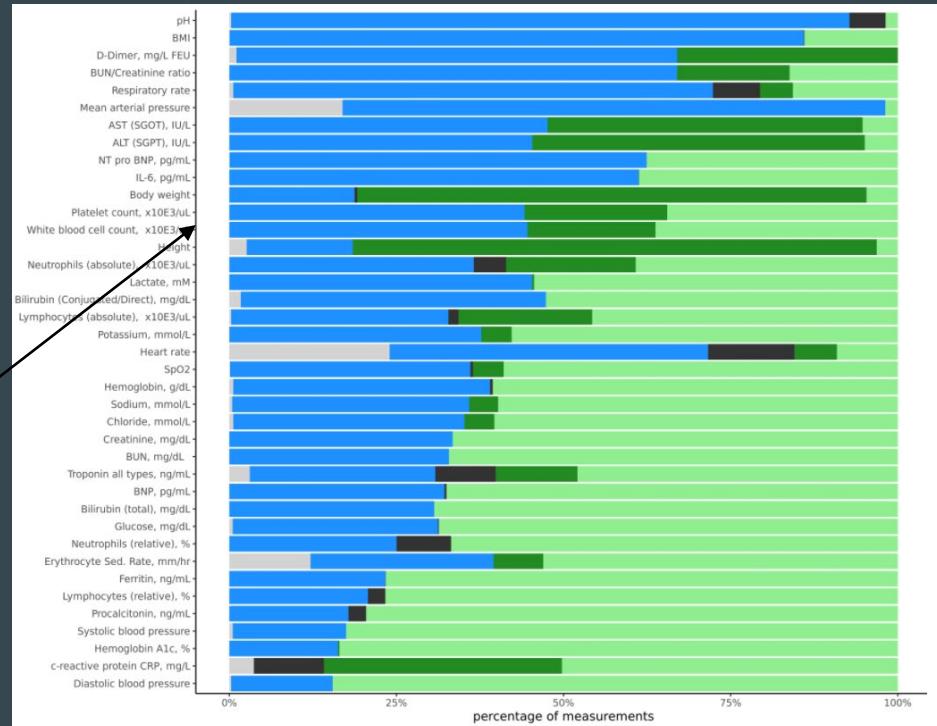


Algorithmic data repair is made possible by having many sites

Humans measured in **grams** do not look the same as humans measured in **kilograms**!



Canonical unit
Uses a known conversion
Unit not plausible
Missing unit inferred
Unit still missing

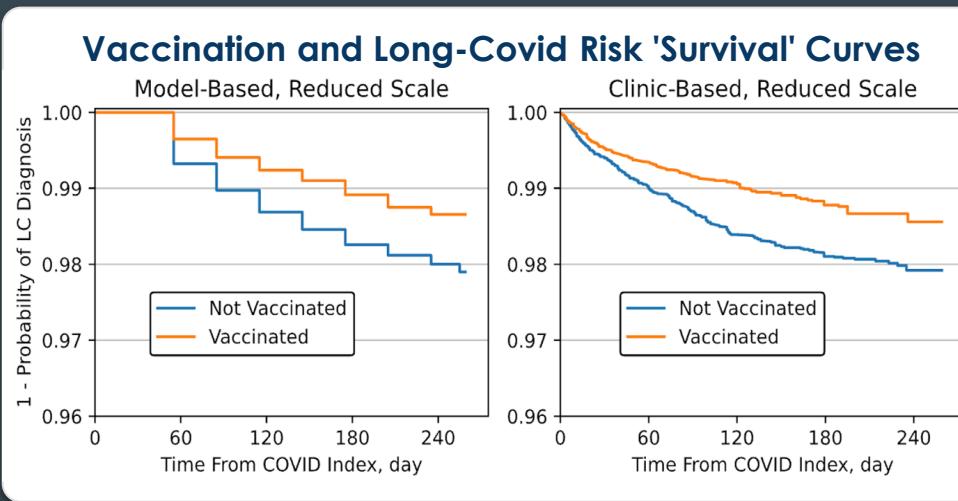


~2x increase in usable data!

N3C provided the earliest and most representative data to predict risk and inform health policy

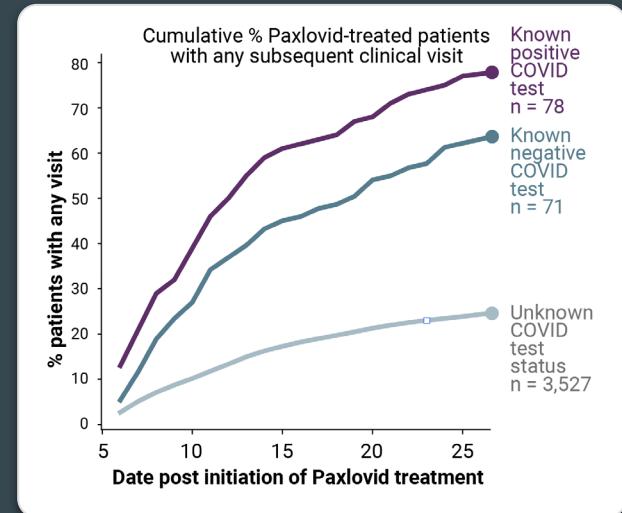
Problem:

- Conflicting research on effects of vaccination on long-covid
- Determining who is vaccinated is challenging in the US



Problem:

- To plan pandemic response White House needed real-world evidence that Paxlovid was effective



Solution: N3C reconciled vaccination data and demonstrated using multiple methods that vaccination lowers risk of Long-COVID

Nature Comms: <https://doi.org/10.1038/s41467-023-38388-7>

Solution: N3C showed that few patients require care or are hospitalized post-COVID following Paxlovid treatment

<https://doi.org/10.1101/2024.01.20.243>

1) EHRs for patients diagnosed with PASC*

*U09.9 code or long-covid clinic



Machine learning

2) Learned patterns of clinical features of PASC

- Dyspnea
- Fatigue
- No vax on record
- New albuterol Rx
- Many outpatient visits
- New corticosteroid Rx
- ...

Search EHRs for similar patients

3) Identify previously unknown cases using learned patterns



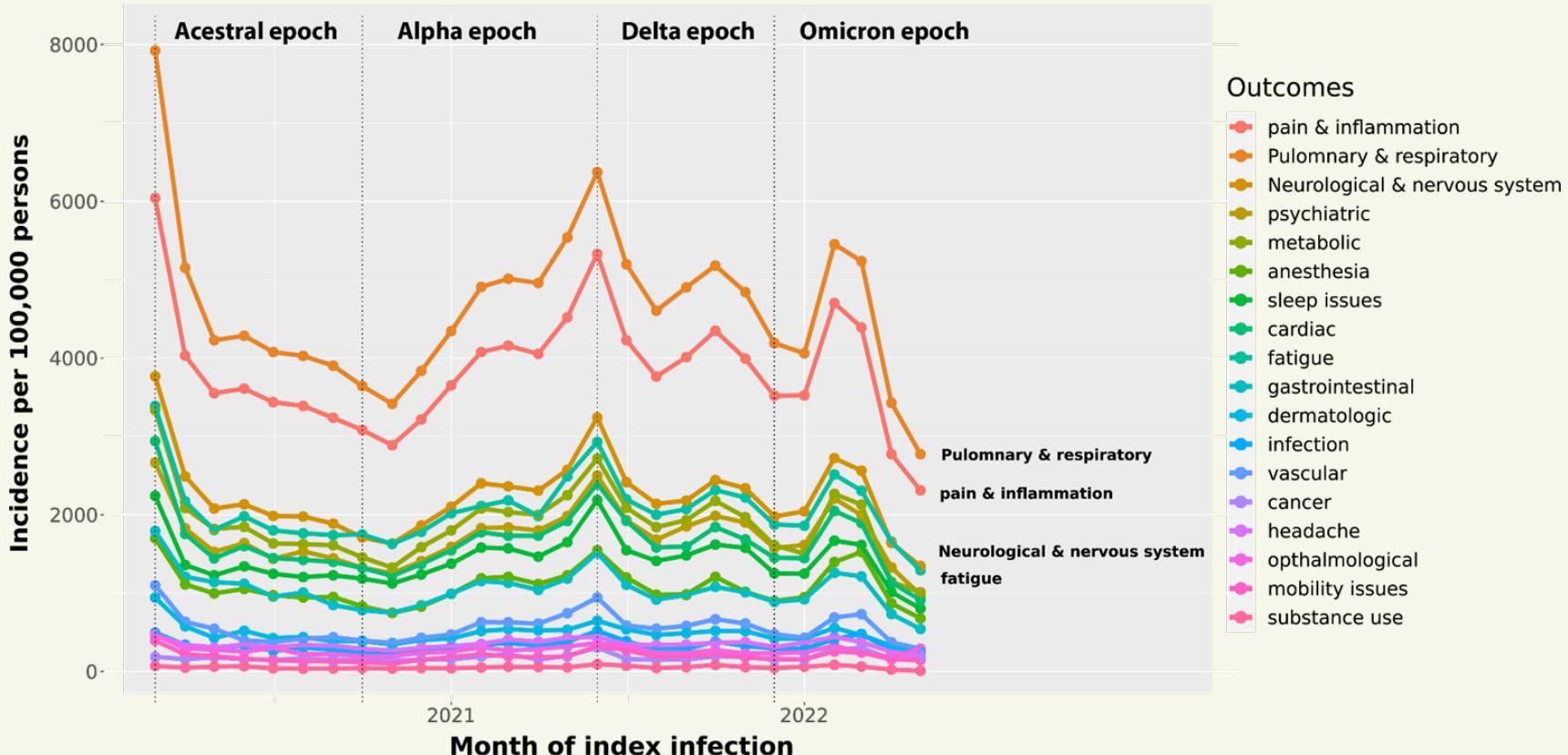
The N3C algorithm can be used to identify cohorts for study recruitment and treatment considerations nationwide, in EHRs beyond N3C

 **Lancet Digital Health**; May 16, 2022; [https://doi.org/10.1016/S2589-7500\(22\)00048-6](https://doi.org/10.1016/S2589-7500(22)00048-6)

NIH Director's blog: <https://directorsblog.nih.gov/tag/national-covid-cohort-collaborative/>



Important features for classifying patients as having Long-Covid during different viral epochs

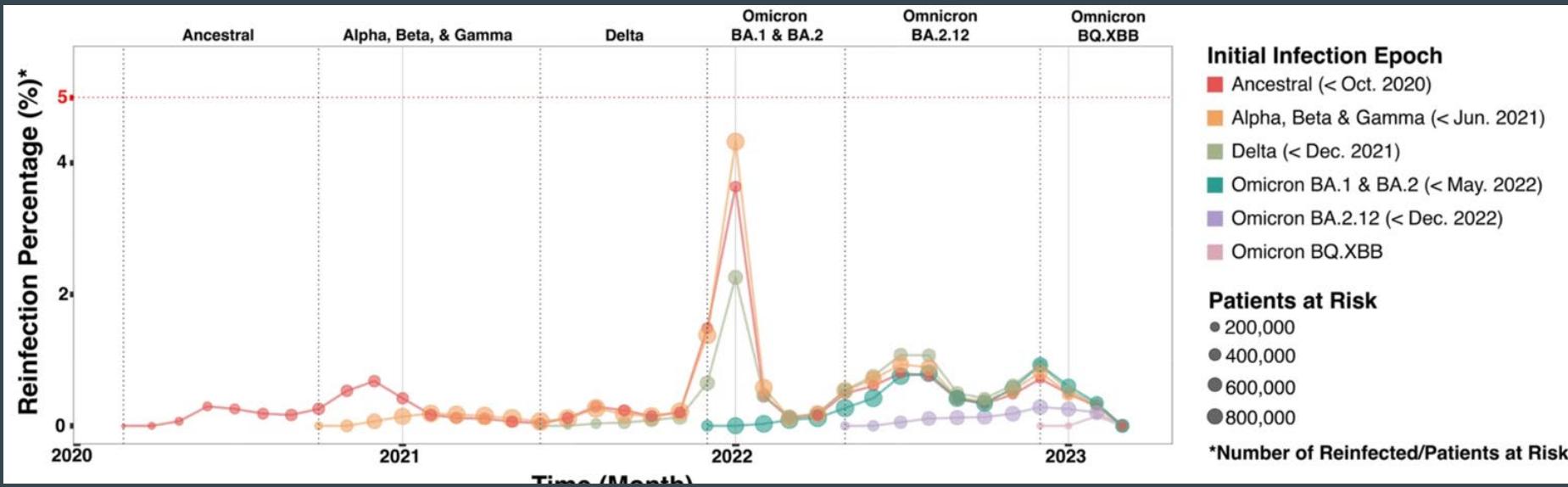




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Reinfections during different variant epochs

RECOVER
Researching COVID to Enhance Recovery
An Initiative Funded by the National Institutes of Health



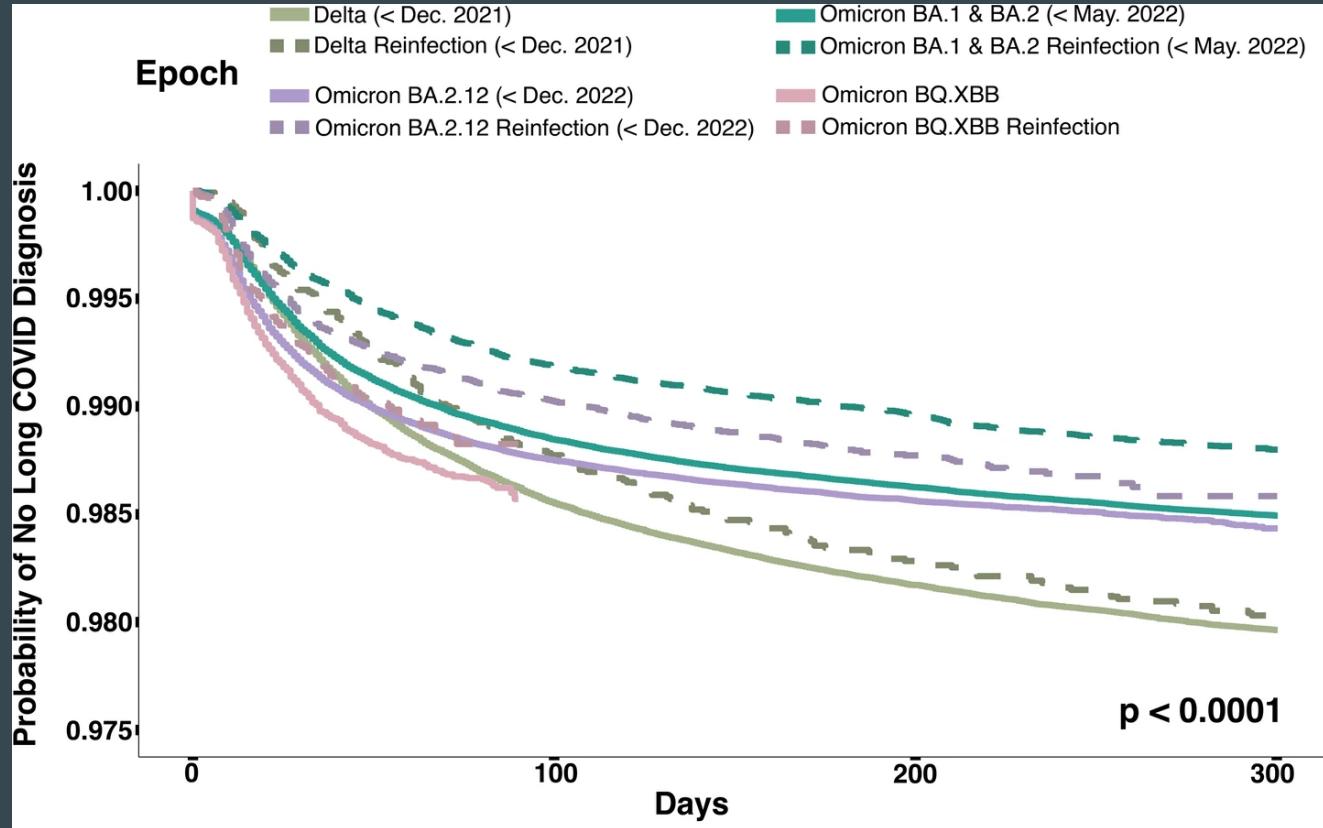
Majority of reinfections occurred during the Omicron epoch



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Probability of Long-Covid after initial versus reinfection during different variant epochs

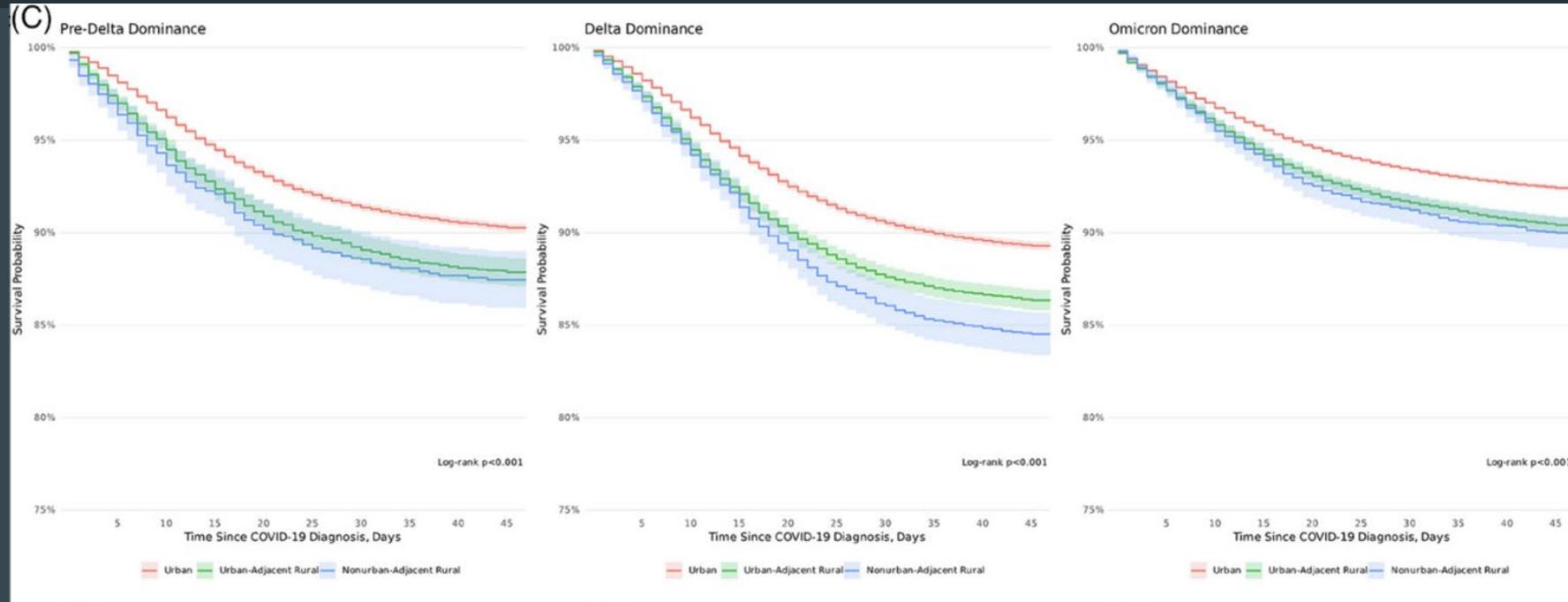
 **RECOVER**
Researching COVID to Enhance Recovery
An Initiative Funded by the National Institutes of Health



- Severity of reinfection associated with severity of initial infection
- Long-COVID diagnoses occur more often following initial infection than reinfection in the same epoch



Rural disparities persisted throughout all variant epochs



Rural patients had higher hospitalization odds, greater inpatient death hazard, and greater risk of other adverse events compared to urban dwellers. Effectiveness of some therapeutics varied based on rurality.

Putting the Patient back together again: EHR data is necessary, but not sufficient

PPRL Deduplication
Multimodal Data
Cohort
Discovery



CMS Data



SDOH Data



Viral
Variant
Mortality Data



Clinical Data



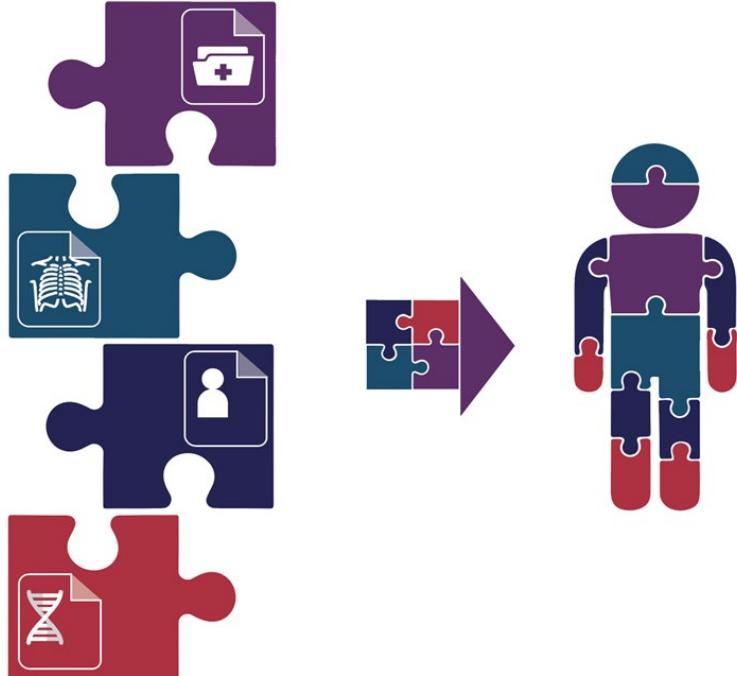
Vaccine data



Imaging



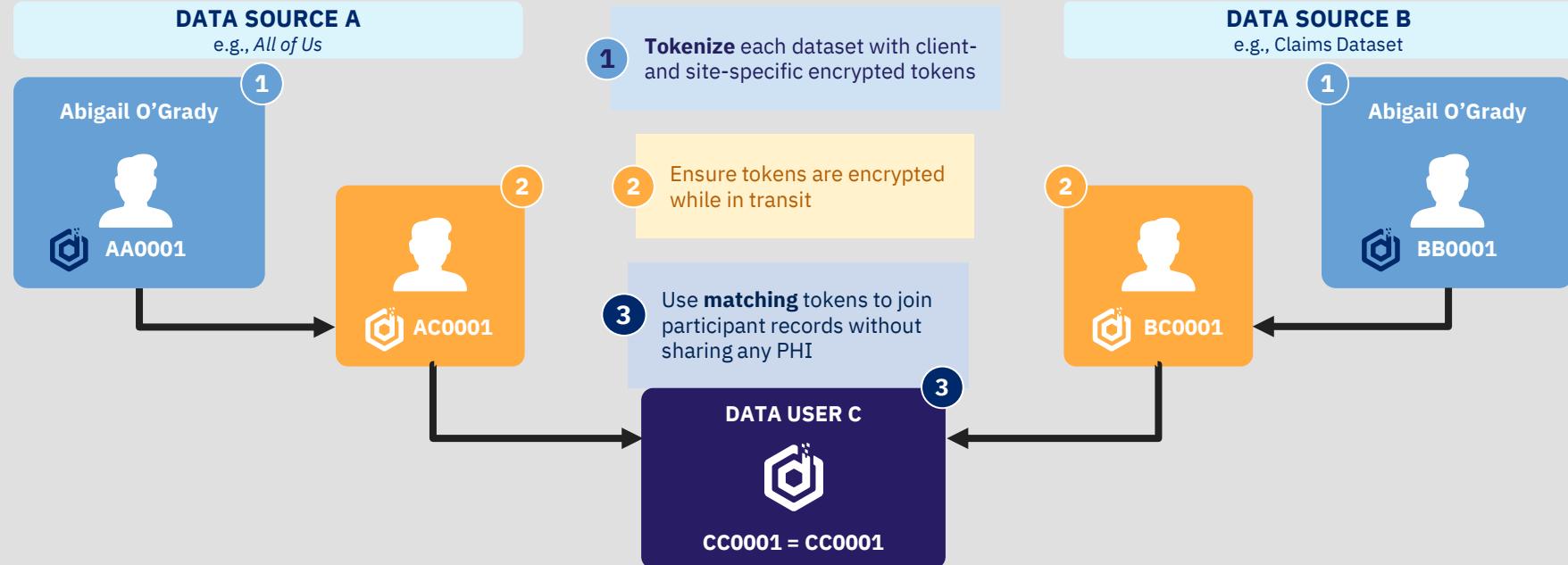
Registries



Privacy Preserving Record Linkage (PPRL) securely connects records across different data sources while maintaining privacy



PPRL meets the definition of de-identified data under HIPAA



What does a token actually look like?

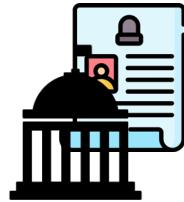
This is just one! We use multiple of these

j05G69K7BH0ugG3VbSusHLbeL7sNm6H3xpiwM6DHfjl=



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Secure PPRL from EHR data to other data



*Non-EHR
Mortality Data*



*Patient Viral
Variants*



*Medicare/Medicaid
Billing Data*



*Cancer
Registry Data*

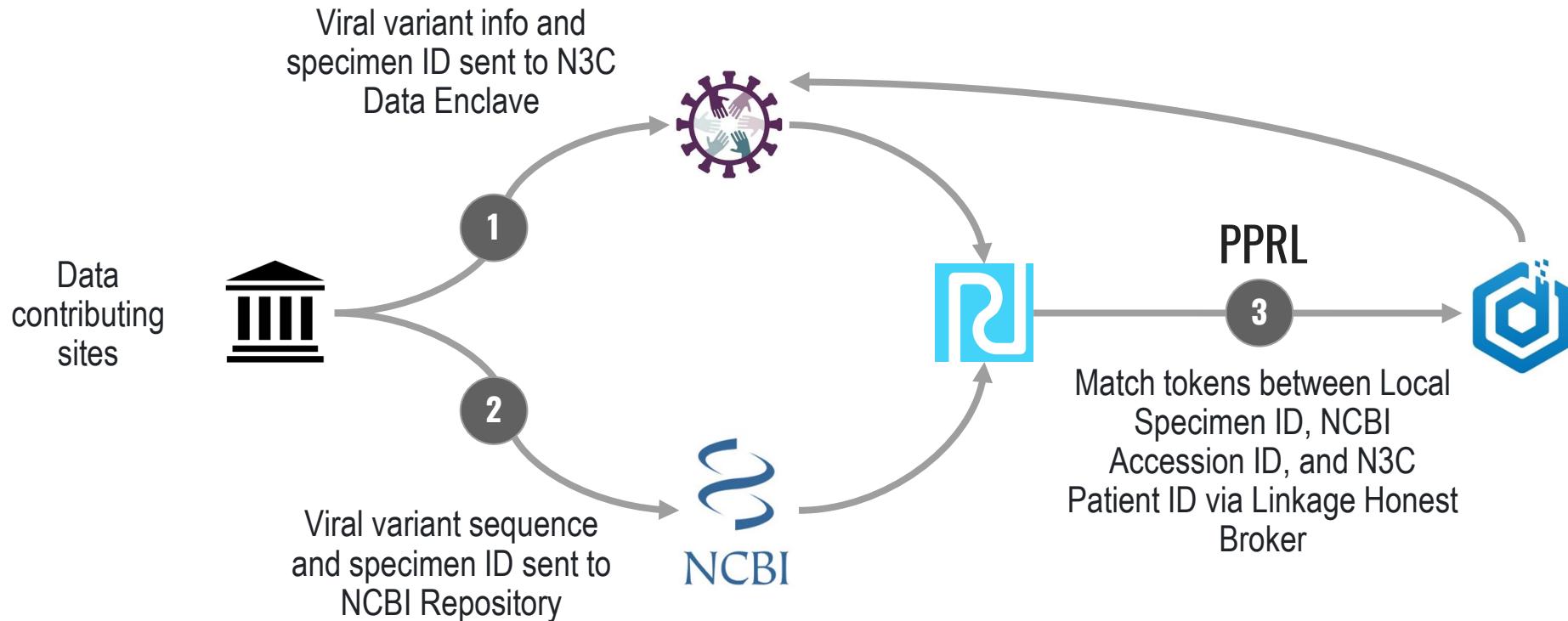
<https://national-covid-cohort-collaborative.github.io/guide-to-n3c-v1/>

The Researcher's Guide to N3C

A National Resource for Analyzing Real-World Health Data



Sending viral variants collected from patients to N3C should be easier



Roads not taken: challenges in linking clinical outcomes



- **EHR data from rural and small clinics very hard to obtain**
 - Lack of representativeness
- **No patient genomic or other 'omic data to link to**
 - Lack of understanding patient response to infection and modifier variants
- **Vaccination data very hard to obtain and trust**
 - Incomplete vaccination record makes it hard to understand clinical impacts
- **Viral variant data incomplete and challenging to link to patient records**
 - Temporality used more than actual known variants to infer correlations

Thank you!

NIH National Center
for Advancing
Translational Sciences

CTSA Clinical & Translational
Science Awards Program



The National COVID Cohort Collaborative (N3C): Rationale, Design, Infrastructure, and Deployment

Journal of the American Medical Informatics Association, ocaa196,

<https://doi.org/10.1093/jamia/ocaa196>

Published: 17 August 2020 Article history ▾

[bit.ly/n3c-methods-jamia](https://doi.org/10.1093/jamia/ocaa196)

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¹CREDIT_00000001 (Conceptualization)⁴CREDIT_00000004 (Funding acquisition)⁷CRO_00000007 (Marketing and Communications)⁸CREDIT_00000008 (Resources)⁹CREDIT_00000009 (Software role)¹⁰CREDIT_00000010 (Supervision role)¹³CREDIT_00000013 (Original draft)¹⁴CREDIT_00000014 (Review and editing)¹⁵CRO_00000015 (Data role)⁵²CRO_0000052 (Standards role)⁷⁸CRO_0000078 (Infrastructure role)¹⁰⁰Clinical Use Cases¹⁰¹Governance

Open science and team science at an unprecedented scale in clinical informatics!