



March 13th

Computational approaches to Assessing Radiation Exposure Across the Lifecourse: Risks, Insights, and Innovations.

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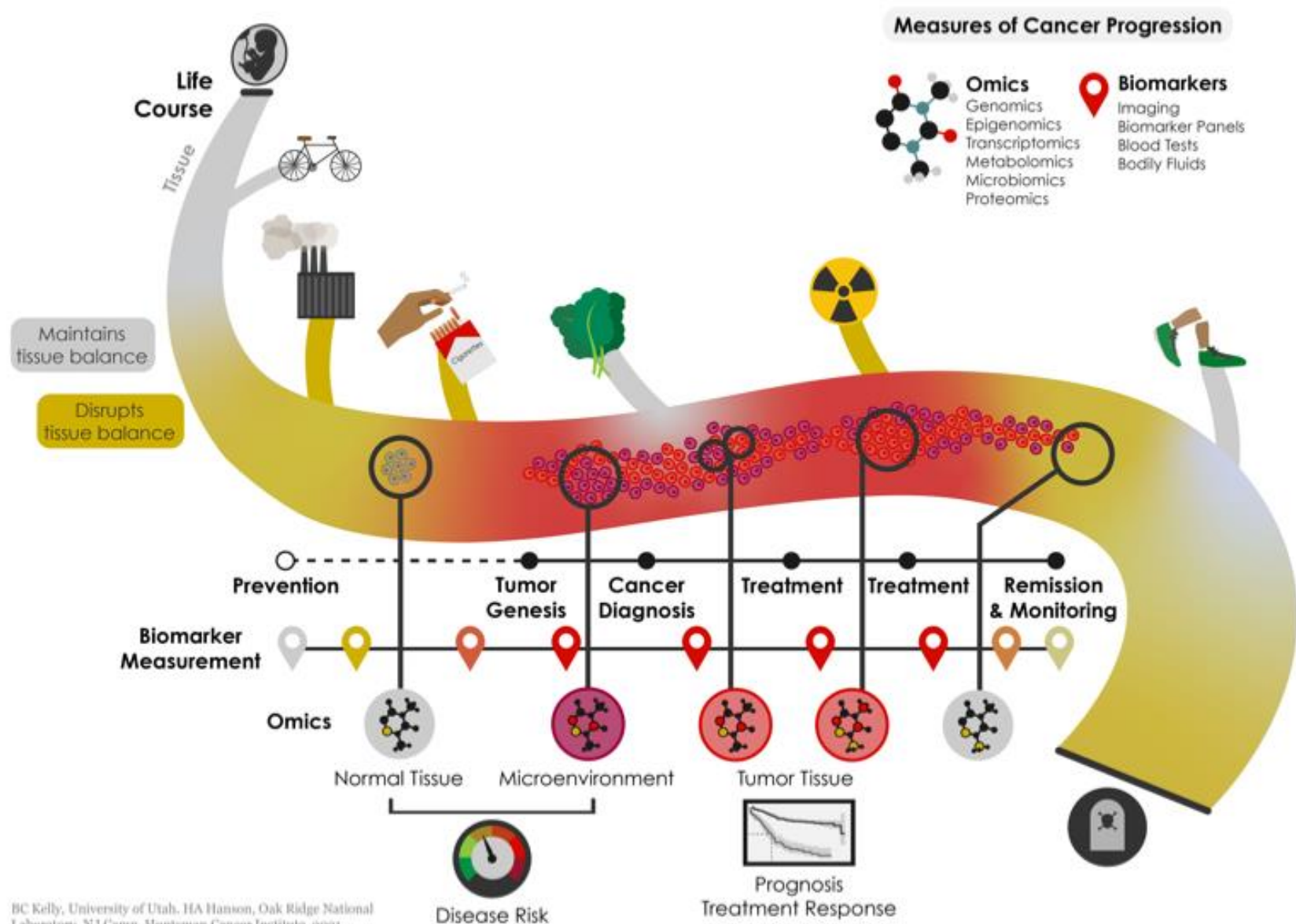


U.S. DEPARTMENT OF
ENERGY

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Life Course Approach to Radiation Exposure



Charting the Life Course: An Interdisciplinary Blueprint

5 Pathways Forward



- 1** Integrate datasets across the lifespan.

Use transdisciplinary principles and methods, borrowing from:



Engineering



Statistics



Communication



Mathematics



Epidemiology

- 2** Data science cores focus on curation, cleaning, analysis, and modeling of data.



- 3** New ways to model high-dimensional data.

- 4** Expand traditional epidemiological methods to include systems and network modeling.



- 5** Improve reproducibility with standard operationalizing measures.



Charting the Life Course: An Interdisciplinary Blueprint

5 Pathways Forward

Workflows for AI Ready Data and AI for Data Harmonization



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Integrated Electronic Health, Administrative, and Population Data

Computational tools enabling decision making for risk assessment, prevention, and treatment

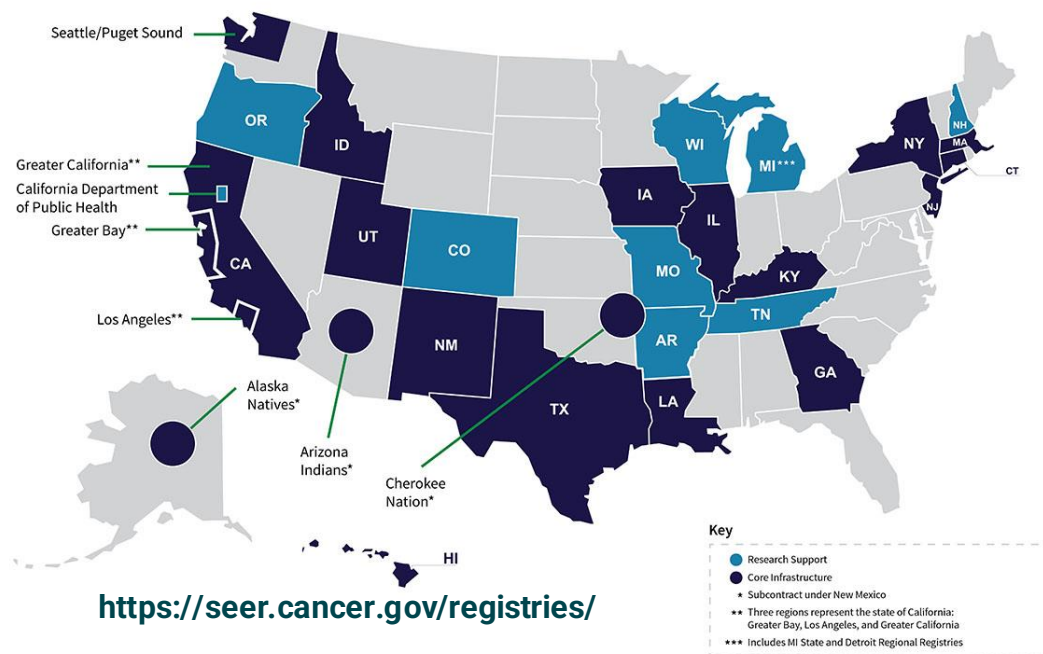
- Easy management of large amounts of heterogenous and siloed sensitive information
- Assessment of longitudinal trajectories of radiation exposure from residential radon, medical treatments, and occupational exposures.
- Identification of geographical areas at risk for high residential radon exposure
- Data-driven tools to guide decision making in the clinic and community





Modeling Outcomes Using Surveillance Data & Scalable Artificial Intelligence for Cancer

Surveillance Epidemiology End-Results (SEER) Registries
> 850,000 Diagnoses Annually



Decision Making Tool Built into SEER Data Management System (2021)

Auto-Extraction from Pathology Reports:

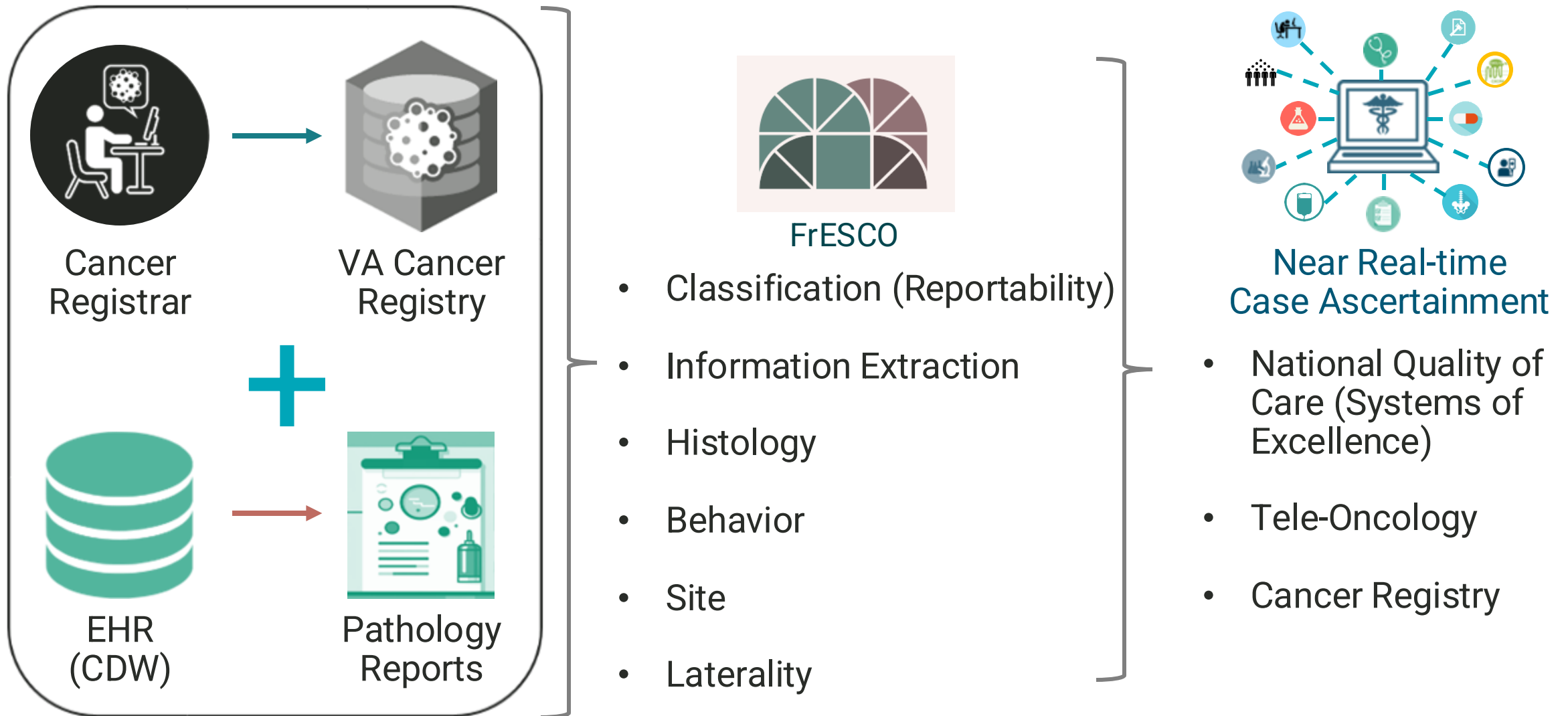
Accuracy: **Auto-coding of 23-27% of path reports (N ~ 230,000)** with > 98% accuracy across all data elements.

- Phenotype classifications:
 - Site = 70 categories
 - Sub-site = 324 categories
 - Histology = 626 categories
 - Laterality = 7 categories
 - Behavior = 4 categories

Production implementation Hierarchical Self Attention Model (HiSAN) with Deep Abstention:

- **Total 16 registries** (additional 4 since March 2022; ~31% of US population)
- **Default** as part of any new DMS installation, regardless of SEER affiliation
- **5+ new registries** anticipated in 2023/2024
- Testing phase with the Veteran's Health Administration

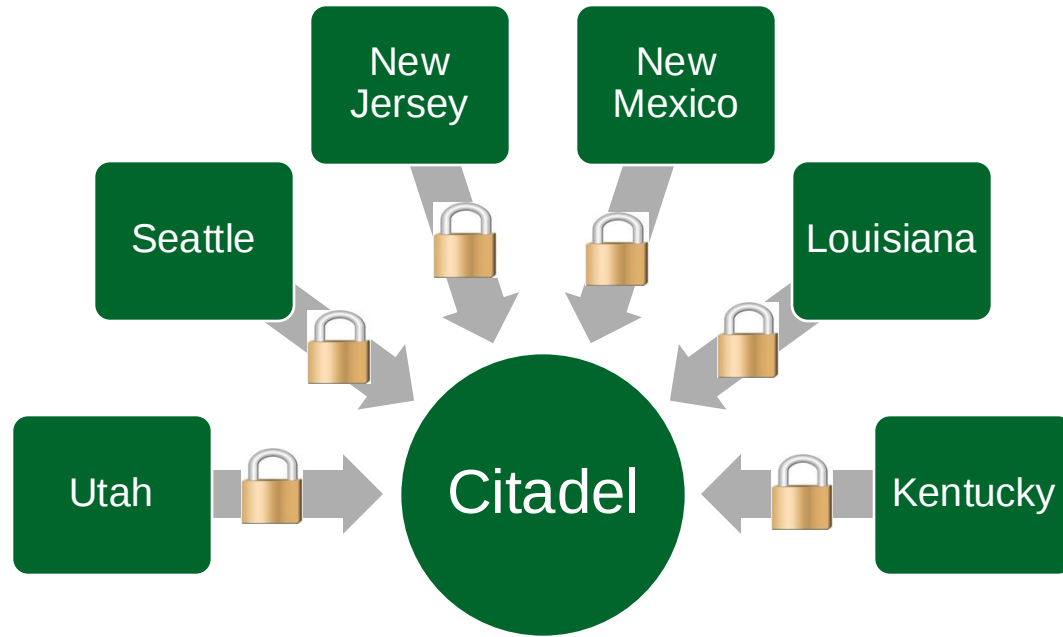
Real World Evidence: Clinical Decision Support with EHR Data



Interoperable AI Systems at Scale:

Bringing Compute Closer to the Source

Rapid Data Integration while Protecting Private Information



The National AI Research Resource (NAAIR) Task Force
National Childhood Cancer Registry
DOE ASCR Biopreparedness (BRaVE) Funding

Our current implementation

- Cross Silo
- Horizontal
- Model Centric – Need for Cooperative Agreement between institutions/ participants

Trusted host

Differential privacy

- “Epsilon Indistinguishability”

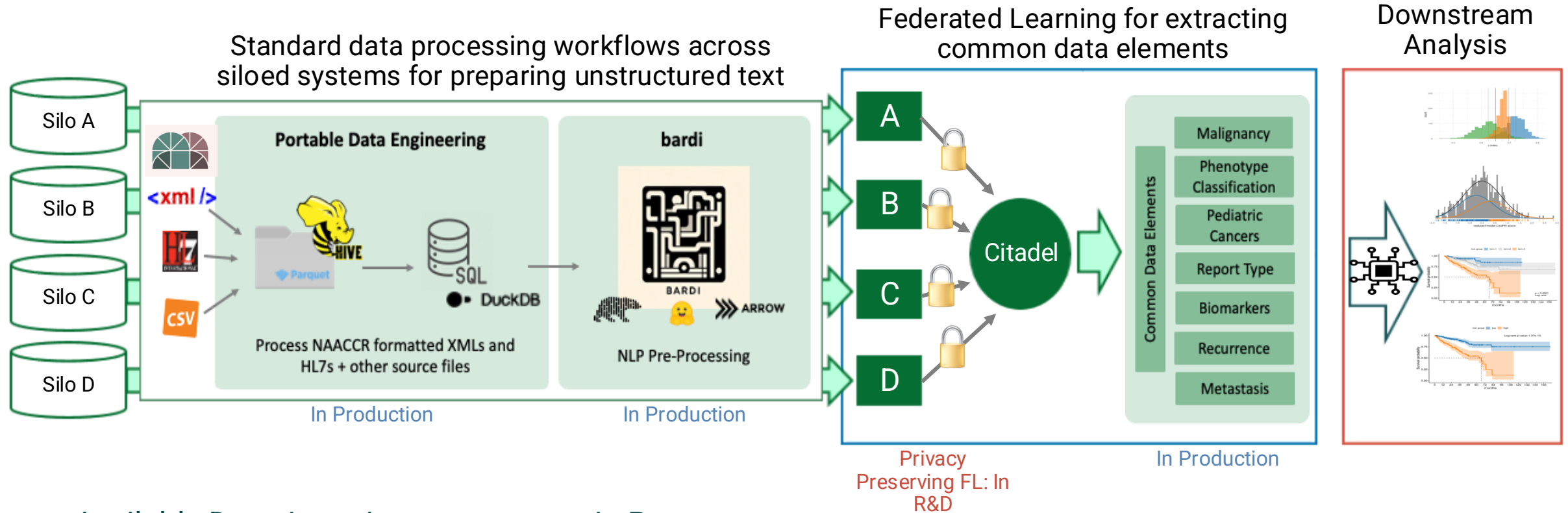
Privacy/accuracy tradeoff



Innovation: SEER data make us completely innovative in this space. We can design and test solutions for real-world application at population scale.

MOSSAIC Computational Workflows

Rapid Response for Health Surveillance



Available Deep Learning Tools

MT-CNN, MT-HiSAN,
Clinical BigBird,
PathBigBird, Deep
Abstaining Classifier

In Progress:

Multimodal/longitudinal
DNN, Phrase Level Attention
DNN, Tools for Bias
Detection, Exposomic Data
Linkage, Privacy Preserving
FL, Synthetic Clinical
Reports



FrESCO

<https://github.com/DOE-NCI-MOSSAIC>

Raster Data

- Satellite images, aerial photographs, and derived raster products

Vector Data

- Geographic features such as boundaries, roads, and points of interest

Tabular Data

- Meta data, user information, analytical results, modeled data

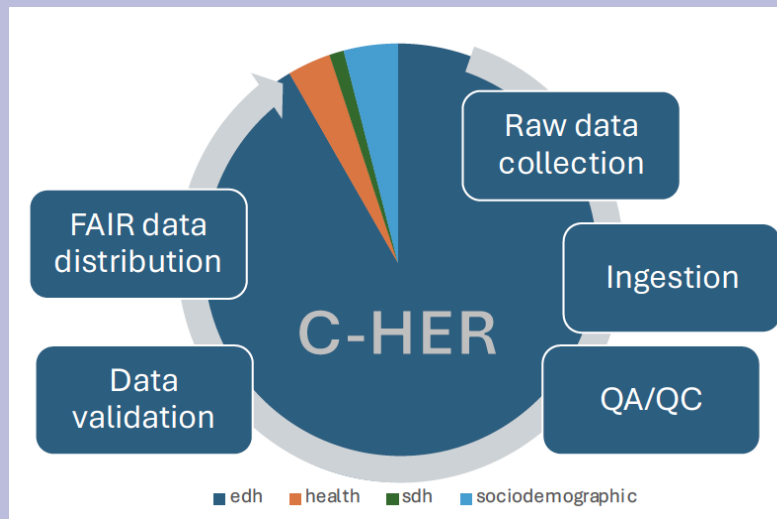
Large Source Files

- Raw data files stored in MinIO

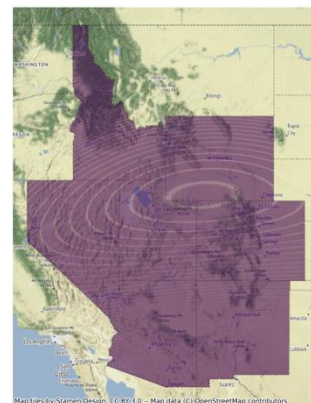
Data Ingestion Code

- The code used to ingest the data is also stored in the database, allowing us to quickly refresh data and identify issues

Metadata and data storage: Metadata is captured with Prefect, which has been integrated with the Centralized Health and Environmental Repository (C-HER) tools. This provides dynamic metadata extraction and ensures robust error handling, making metadata documentation reliable and scalable. Data security is ensured with encryption for data at rest and in transit. We implement strict access controls and authentication measures to safeguard our data.



The C-HER Ecosystem utilizes Open-Source software such as MinIO, PostgreSQL, Prefect, and Python. The goal of the project is to optimize the reuse of data, thus the C-HER ecosystem adheres to principles of findability, accessibility, interoperability, and reusability (FAIR).



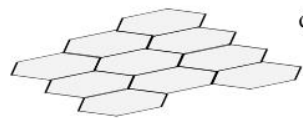
Data Quality Assurance: Regularly validate data using automated tools and manual checks to ensure accuracy and reliability. Document all QA-QC procedures and maintain logs of all data checks.

Ethical Concerns: We consider and address privacy concerns for data that is potentially identifiable. We ensure that data usage complies with all applicable laws and ethical guidelines.

C-HER Serves as the foundation for modeling exposures

Example: Indoor Radon

C-HER Protocol for AI Ready Data

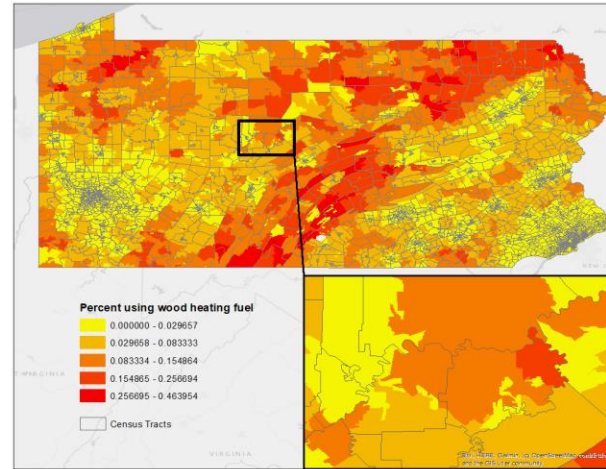


data harmonized at H3
Level 8 grid cells ~
0.737 km²

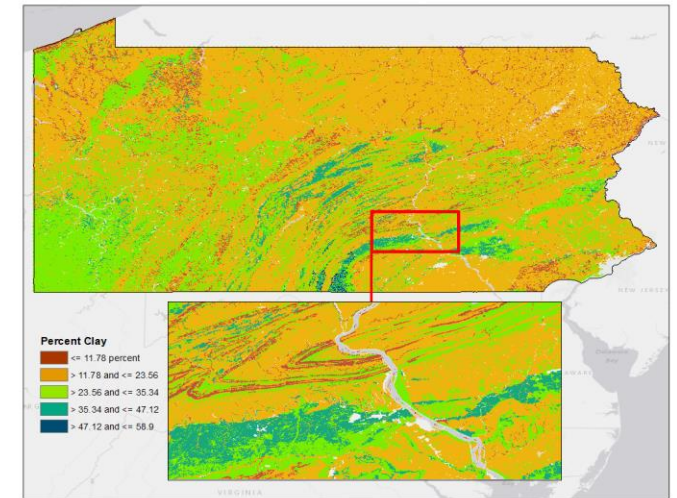
Radon data

- | | |
|---|--|
|  Population density |  Temperature |
|  Soil characteristics |  Precipitation |
|  Uranium, thorium, potassium content |  Elevation |
| |  Depth to groundwater |

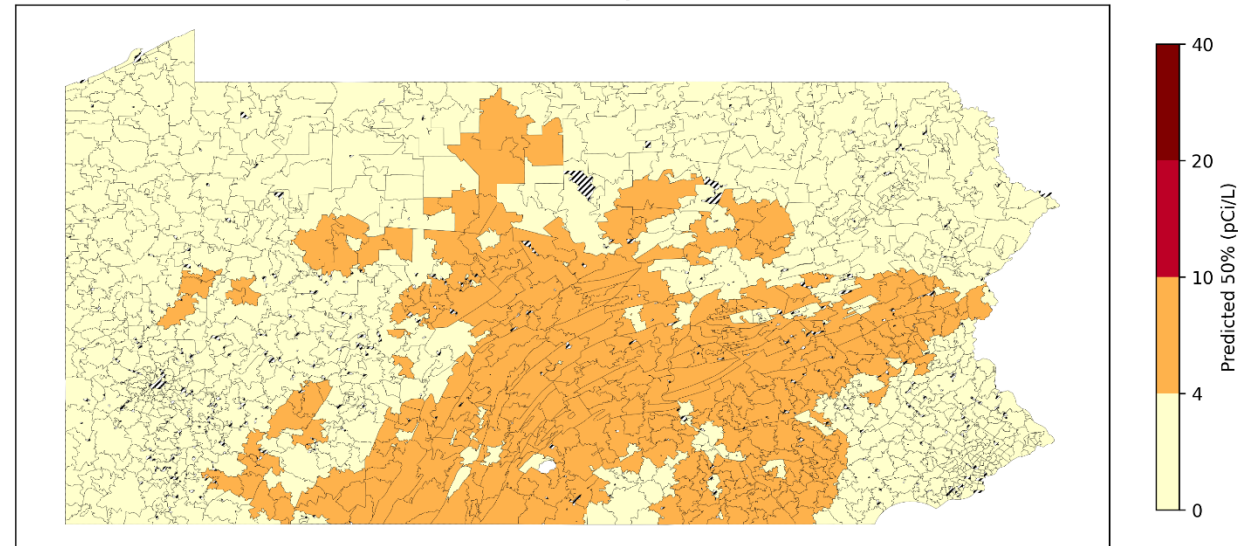
ZCTA level housing Characteristics
Individual test results with ZCTA information



Low spatial resolution geologic, soil, and other data



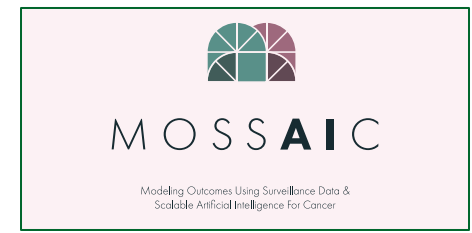
Predicted 50% by ZCTA



25+ years of multimodal exposure data linked to cancer records

LexisNexis Residential History data

- 11 SEER registries have been linked (3.2 million individuals diagnosed from 2005 - 2022)
- 15 SEER registries should be linked by the end of 2023
- Residential history constructed for Louisiana and expected to be expanded to all linked registries this year
- High quality data from 1995 – 2020
- 83% are geocoded to the point location



Combined PM2.5 & Wildfire Smoke AQI: September, 2020

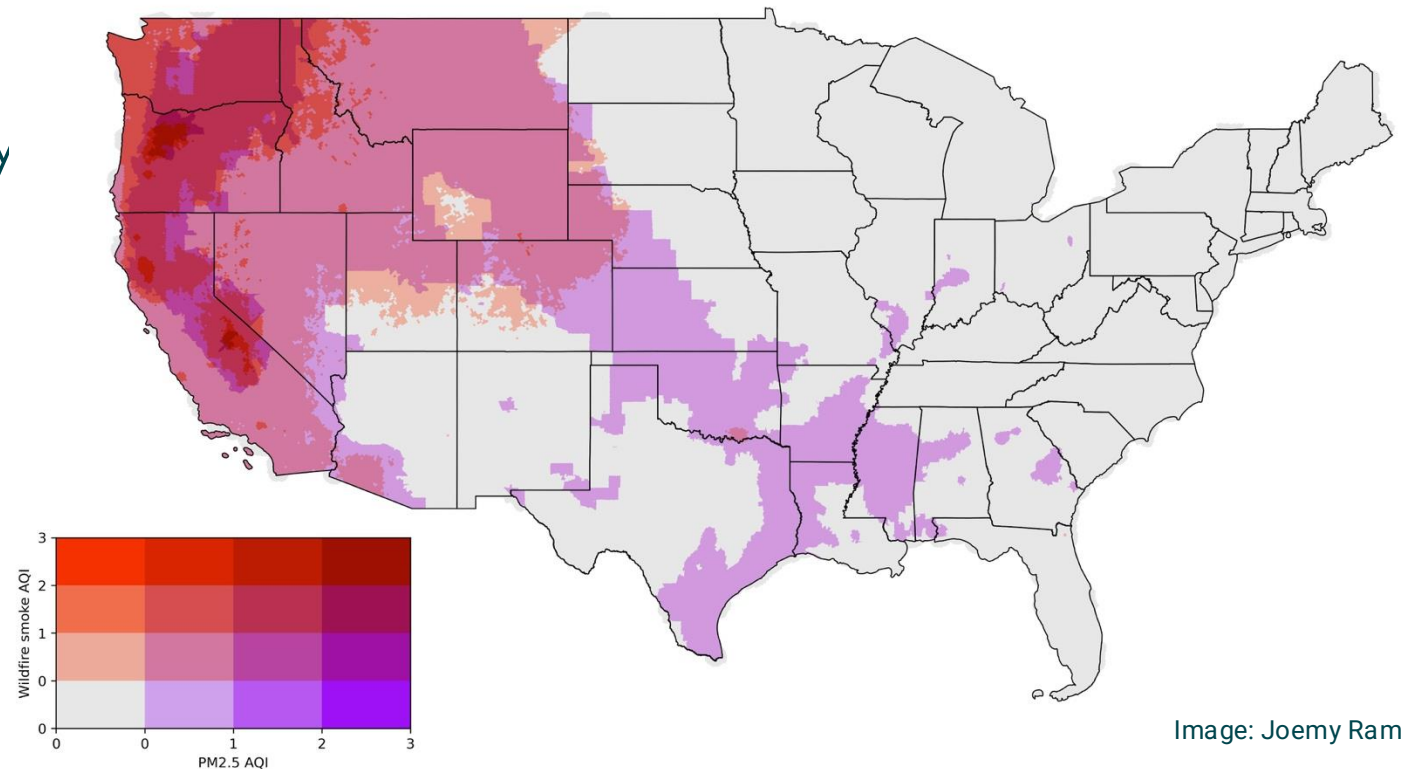
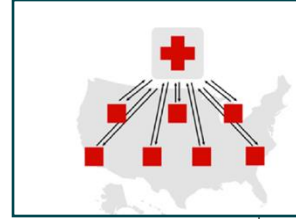


Image: Joemy Ramsay

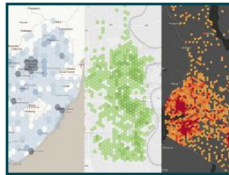
“Electronic Health Record informed Lagrangian method for precision public Health (EHLICH)” will be a set of new computational capabilities developed to enable the rapid assimilation of real-world data for digital twins of population level biological threats.



Scalable, Trustworthy and Responsible AI



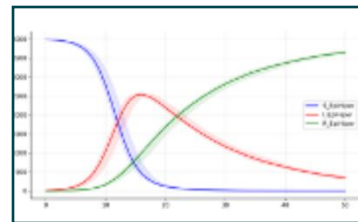
Privacy-Aware Simulation for Clinical AI and Learning (PASCAL)



AI Ready Social and Environmental Determinants of Health Data



High-spatial resolution and high-fidelity synthetic populations that include demographic and biologic characteristics



ENABLE: An agent-based simulation framework

Interoperable AI Systems at Scale:

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Rapid Data Integration while Protecting Private Information

Alternative approach:
synthetic data generation

Our Solution:

- Leverage generative AI
- Minimize leakage of PHI & other identifiable info
- Evaluate for **utility**, **similarity**, and **privacy**

Initial demonstration on
SEER data

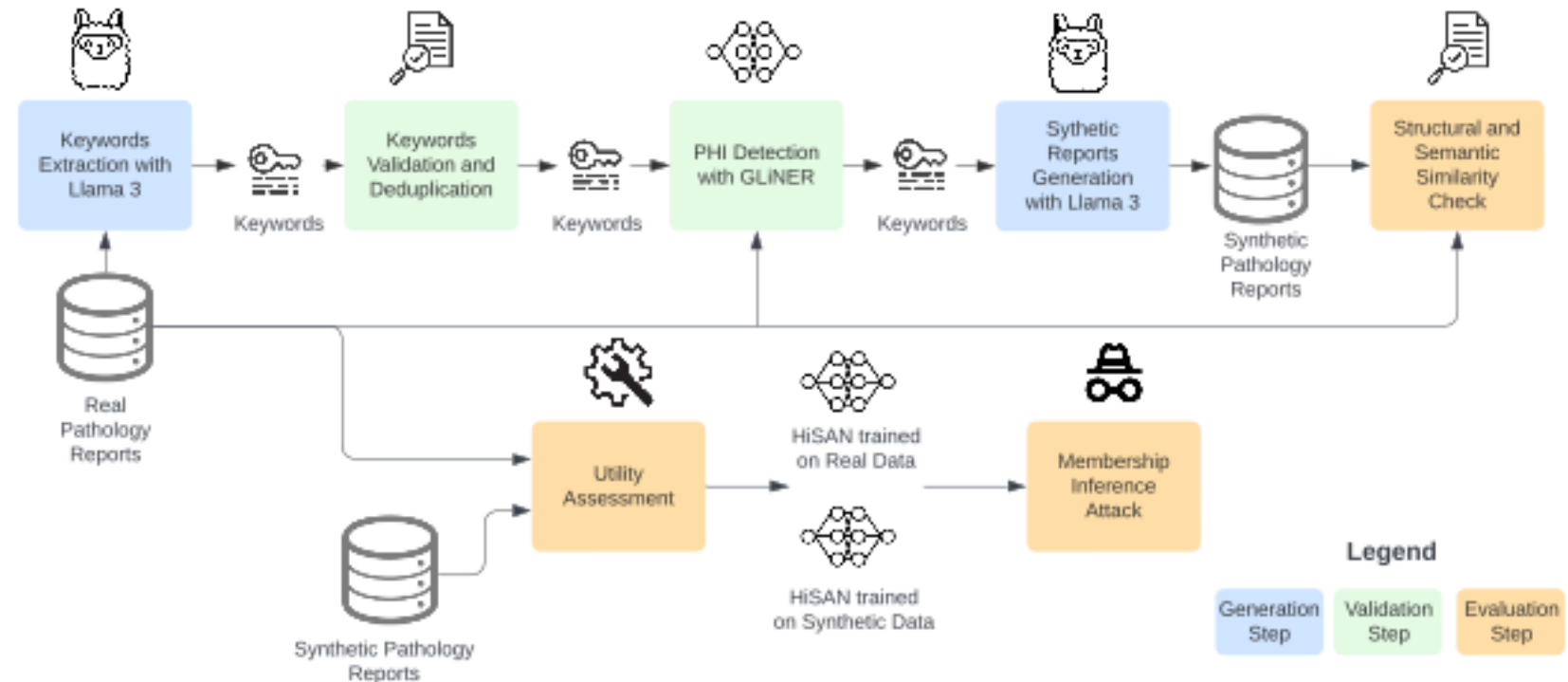
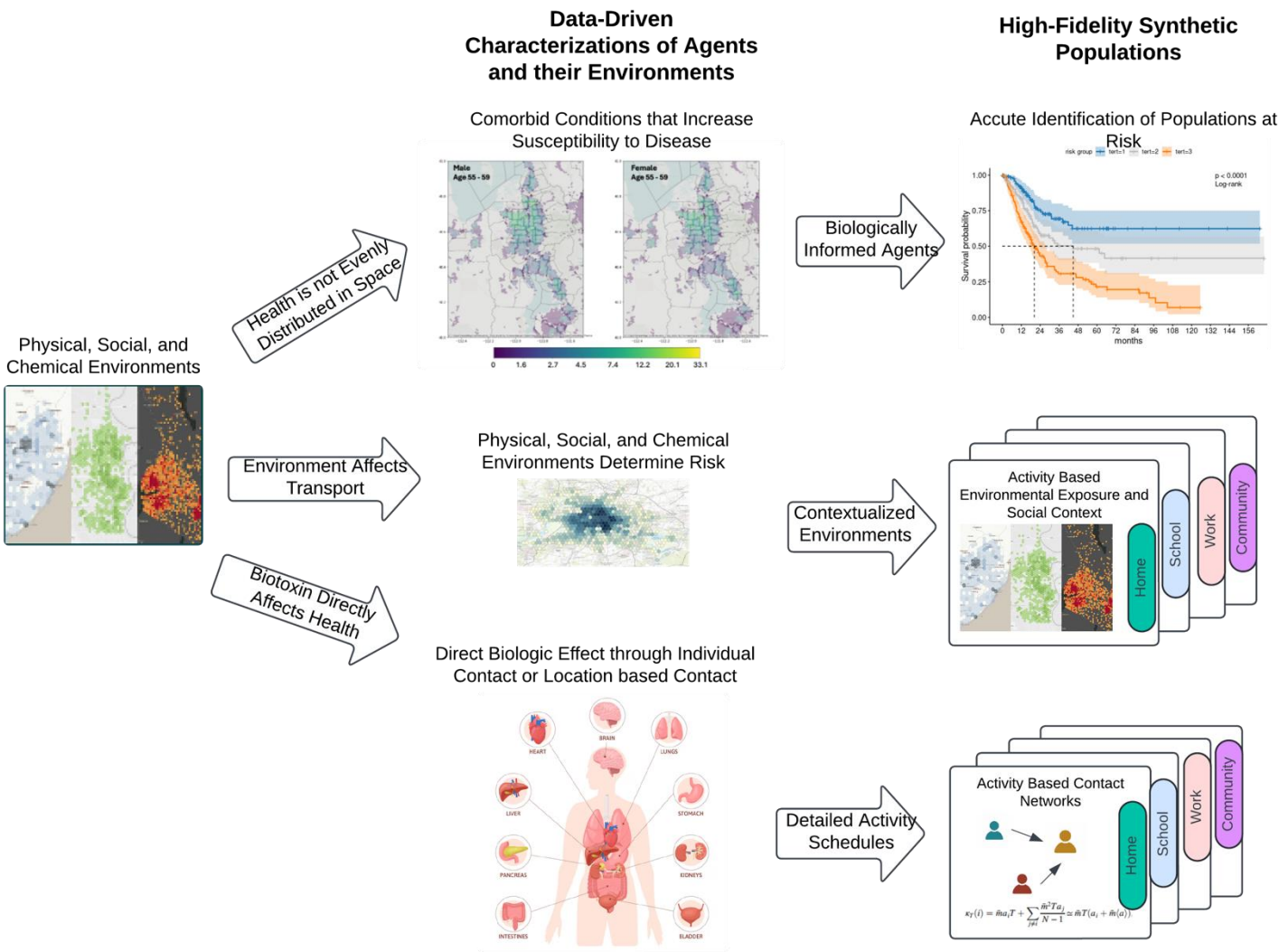


Image credit: Patrycja Krawczuk

EHRlich tools are designed to be agile and adaptable



PLACING A PULSE ON POPULATION HEALTH



**Federated
Learning**



**Synthetic Data
Generation**



**Privacy
Preserving AI**



**Identify existing
patterns and
anomaly detection**



**Predict patterns
relating to
environmental
exposures**



**High-performance
computing for
modeling
and simulating
health outcomes**



C-HER

CENTRALIZED HEALTH AND
ENVIRONMENTAL REPOSITORY

MOSSAIC

National Cancer Institute

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Sylkk Ansah
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Sajal Dash
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Abhishek Shivanna
Robert Bridges
Christopher Stanley
Vandy Tombs
Christoph Metzner



EHRlich



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