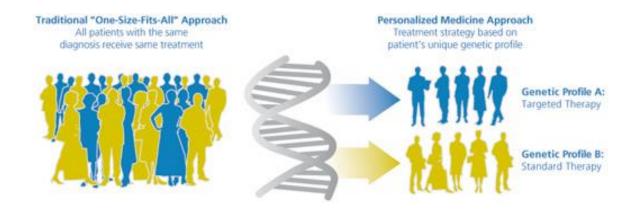


Instrumenting the Health Care Enterprise for Discovery in the Course of Clinical Care

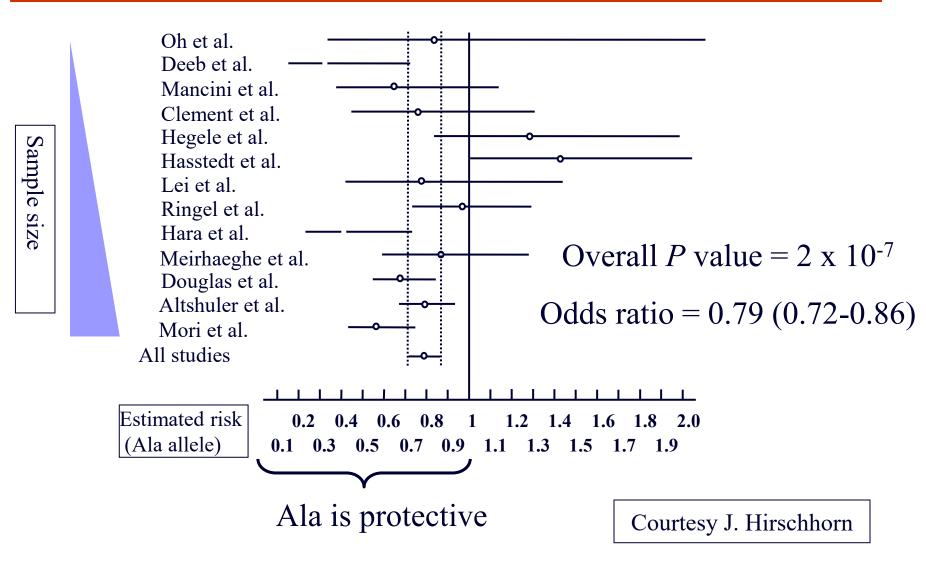
Shawn Murphy MD, Ph.D. Chief Research Information Officer Harvard Medical School / Mass General Brigham

# Personalized Medicine and Genomic technology are critical to managing populations



- Managing a population involves improving health outcomes of the group as a whole by identifying, monitoring and addressing health needs of individuals through:
  - Subpopulation stratification
  - Targeted, evidence-based treatment protocols
  - Predictive analytics

#### **Example: PPARy Pro12Ala and Diabetes**



## High Throughput Methods for supporting Translational Research

- Set of patients is selected from medical record data in a high throughput fashion
- Investigators explore phenotypes of these patients using Machine Learning tools and a translational team developed to work specifically with medical record data
- Distributed networks cross institutional boundaries for phenotype selection, public health, and hypothesis testing
- Digital medicine is delivered into clinical care through Digital Twin

# Data problems that make working with Electronic Healthcare Data to conduct research difficult

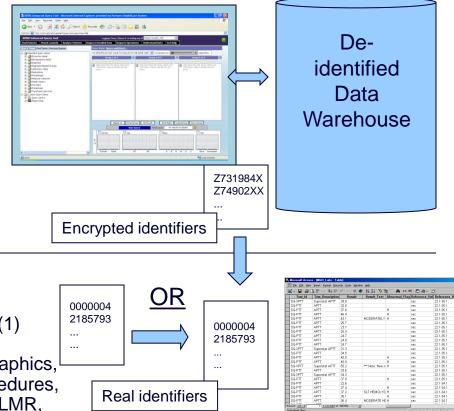
- 1) There are significant risks of a data breach which will result in very large fines and loss of confidence in the hospitals where the breach occurred.
- 2) The data are not collected for research purposes, and therefore the data can be poorly structured with significant omissions, biases, and inaccuracies.

## Research Patient Data Registry (RPDR) at Mass General Brigham to find patient cohorts and distribute data

Query construction in web tool

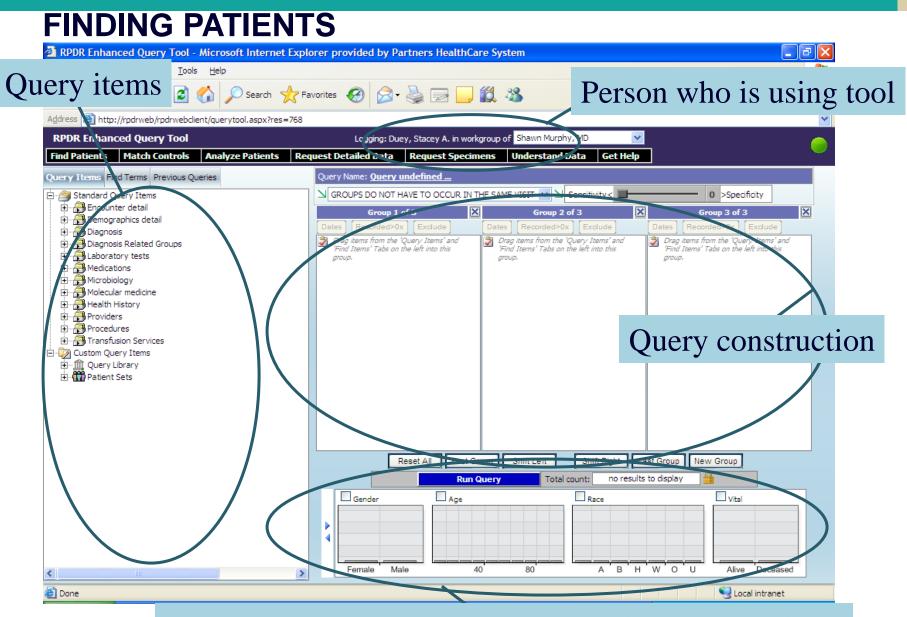
#### 1) Queries for aggregate patient numbers

- Warehouse of in & outpatient clinical data
- 6.7 million Mass General Brigham patients
- 2.6 billion diagnoses, medications, genomics, procedures, laboratories, & physical findings coupled to demographic & visit data
- Authorized use by faculty status
- Clinicians can construct complex queries
- Queries cannot identify individuals, internally can produce identifiers for (2)



#### 2) Returns detailed patient data

- Start with list of specific patients, usually from (1)
- Authorized use by IRB Protocol
- Returns contact and PCP information, demographics,<sup>L</sup> providers, visits, diagnoses, medications, procedures, laboratories, microbiology, reports (discharge, LMR, operative, radiology, pathology, cardiology, pulmonary, endoscopy), and images into a Microsoft Access database and text files.

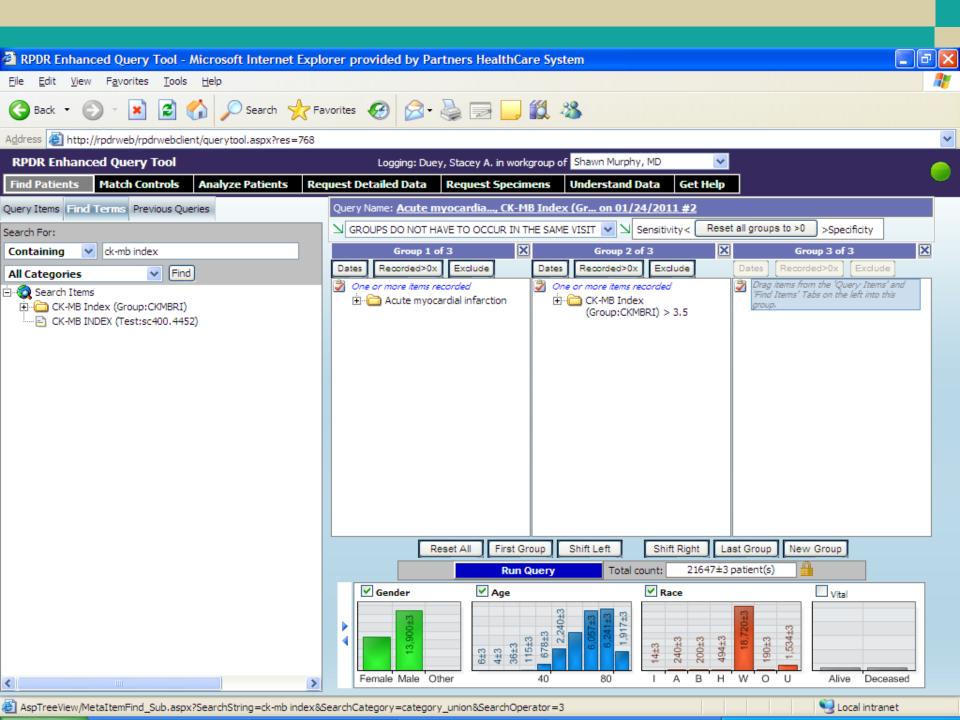


Results - broken down by number distinct of patients

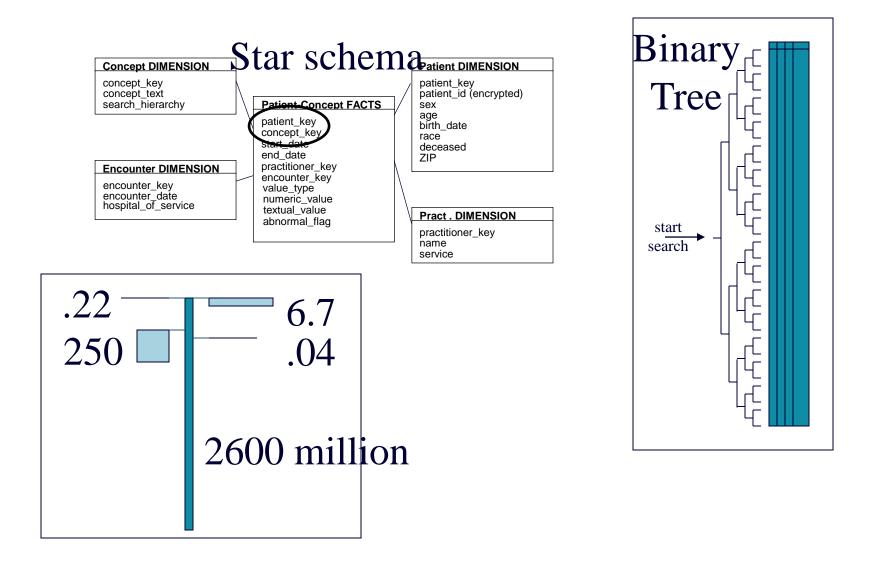
| Participation of the second | prer provided by Partners HealthCare System                      | _ 7 🗙    |
|---|--|----------|
| <u>File E</u> dit <u>V</u> iew F <u>a</u> vorites <u>T</u> ools <u>H</u> elp  |  | <b>.</b> |
| 🚱 Back 🔹 🐑 💌 😰 🏠 🔎 Search 👷 Fa  | avorites 🚱 🛜 🍡 🎇 🛄 🎇   |          |
| Address ahttp://rpdrweb/rpdrwebclient/querytool.aspx?res=768  |  | ~        |
| RPDR Enhanced Query Tool  | Logging: Duey, Stacey A. in workgroup of Shawn Murphy, MD        |          |
| Find Patients Match Controls Analyze Patients Rec   | quest Detailed Data Request Specimens Understand Data Get Help   |          |
| Query Items Find Terms Previous Queries   | Query Name: Jsut Diagnos AMI                                     |          |
| Standard Query Items   Encounter detail   Demographics detail   Diagnosis   Circulatory system   Acute Rheumatic fever   Acute Rheumatic fever   Cardiac problem-Oncall   Cardiac risk factors-Oncall   Cardiac risk stratification-Oncall   Cardiac risk stratification   Chronic Rheumatic heart disease   Condition appetoris   Standard Query   Conditions in the perinatal period   Congenital anomalies  | GROUPS DO NOT HAVE TO OCCUR IN THE SAME VISIT       Sensitivity< | ns' and  |
| ⊡ Congenital anomalies ⊡ Digestive system   | Click the image or check box to request an aggregated count by   |          |
| Endocrine disorders      Events of pregnancy  | patient gender for this query.                                   |          |
|   | Female Male 40 80 A B H W O U Alive Dec                          | ceased   |
| AspTreeView/MetaDataTree_Sub_aspx2ParentEolderId=DGNCircula   | tory system (390-459)0x5CTschemic heart disease (410-414)        | Inet     |

AspTreeView/MetaDataTree\_Sub.aspx?ParentFolderId=DGNCirculatory system (390-459)0x5CIschemic heart disease (410-414)

🧐 Local intranet



#### **Theory of Kimball translated to Healthcare Data**



| 🚰 RPDR Detailed Data Request Wizard Web Page I | Dialog   |
|--|--|
| RPDR [   | DETAILED DATA REQUEST WIZARD   |
|  | d in the RPDR Identified database) to obtain data from the RPDR                  |
| You are logged in a                            | is Murphy, Shawn N. in workgroup Shawn Murphy, MD                                |
|  | Select protocol number(s)  |
| Partners IRB (required):                       | mgh-demo-1   |
|  | Title: RPDR protocol - Demonstration IRB number for Dr. Murphy<br>Status: Active |
| Newton Wellesley Hospital IRB:                 | NWH Demo 1   |
|  | Title: test  |
|  | Status: Active   |
| Spaulding Rehabilitation Hospital IRB:         |  |
| Options for returned set of patien             | ts:  |
| Create a static set of patient                 | nts from this query that can be used in other RPDR queries                       |
| Rerun the base query sho                       | wh above to obtain a fresh set of patients                                       |
|  |  |
|  |  |
|  |  |
|  |  |
|  |  |
|  |  |
|  |  |
| Help <   | Back STEP 3 Next > Cancel  |

| 🚰 RPDR Detailed Data Request Wizard Web Page Dialog   | × |
|---|---|
| RPDR DETAILED DATA REQUEST WIZARD<br>Using IRB#mgh-demo-1 (found in the RPDR Identified database) to obtain data from the RPDR<br>You are logged in as Murphy, Shawn N. in workgroup Shawn Murphy, MD   |   |
| Please select if you would like a HIPAA-defined (deidentified)<br>limited data set or an identified data set  |   |
| VVhat's a limited data set?   |   |
| <ul> <li>C Limited Data Set         <ul> <li>The files that result from this request will be available in a protected file share with no special encryption.</li> <li>Identified Data Set             <ul> <li>The text files that result from this request will be encrypted and the Microsoft Access file will be password</li> <li>State of the state of the microsoft Access file will be password</li> <li>The text files that result from this request will be encrypted and the Microsoft Access file will be password</li> <li>The text files that result from this request file will be password</li> <li>The text files that result from the microsoft Access file will be password</li> <li>The text files that result for the microsoft Access file will be password</li> <li>The text files that result for the microsoft Access file will be password</li></ul></li></ul></li></ul> |   |
| protected. In <u>order to access the data, a password will</u><br>be provided.  |   |
|   |   |
| Help < Back STEP 8 Next > Cancel  |   |

\_

| 🚰 RPDR Detailed Data Request Wizard Web Page Dialog   | × |
|---|---|
| <b>RPDR DETAILED DATA REQUEST WIZARD</b><br>Using IRB#mgh-demo-1 (found in the RPDR Identified database) to obtain data from the RPDR<br>You are logged in as Murphy, Shawn N. in workgroup Shawn Murphy, MD  |   |
| Select the types of data that should be returned from the RPDR<br>Only data allowed by your protocol should be chosen<br>(Identified data sets will always return a set of identified patient medical numbers)  |   |
| <ul> <li>Detail Data Items</li> <li>Demographic Data</li> <li>Identifying Patient Information - not available for Limited Data Sets</li> <li>LMR (Longitudinal Medical Record)</li> <li>Medications, Diagnoses and Procedures</li> <li>Medications, Diagnoses and Procedures from Billing Data - only visits where query criteria occur all in the same visit</li> <li>Patient Clinical Reports - not available for Limited Data Sets</li> <li>Cardiology Reports</li> <li>Cardiology Data</li> <li>Microbiology Data</li> <li>Operative Notes</li> <li>Pathology Reports</li> <li>Radiology Reports</li> <li>Radiology Reports</li> <li>Transfusion Data, Blood Bank Data</li> </ul> |   |
| Help < Back STEP 9 Next > Cancel  |   |

### Detailed data is gathered for request and distributed

- 🗆 ×

| Process IRB files                                  |  |                               |                        |
|--|--|-------------------------------|------------------------|
| Environment Record O                               | ptions Help  |                               |                        |
| File: SNM0_022                                     | 502164303648842.×ML  |                               | <b>▼</b>               |
| File Type: Control File                            | - Cum  | ent Production Database: 🕐    | RPDR_12                |
| Update Statu<br>IRB Information<br>IRB Number: 200 | 0P000000   | After 9:00:00 PM              | Status ProcesIRBFiles: |
| Date from: 017<br>Primarv User: snm                |  | /1900                         | Status Detail          |
|  | ners\snm0,Partners\zzp,Partner<br>ners\kra1,Partners\snm0,Partne | Contact/Lab: 🔲                |                        |
| - Data Requested                                   |  |                               | Demographics:          |
| Demographics Encounters                            | Medical Record Numbers 🔽<br>Contact Information                  | Chemistry 🔽<br>Radiology 🗖    | Encounter:             |
| Hematology   |  | Pathology                     | MBN:                   |
| Discharge Summaries 🗹<br>Medications 🔽             | Immunology 🗖<br>Operative Notes 🗍                                | LMR Notes 🗖<br>LMR Problems 🗖 | Medications:           |
|  | LMR Medications  | Build Access Database 🔽       | Access DB:             |
| Groups: BUN  |  |                               |                        |
|  |  | Run Close                     | Clear                  |

#### Data is gathered from RPDR and other MGB sources

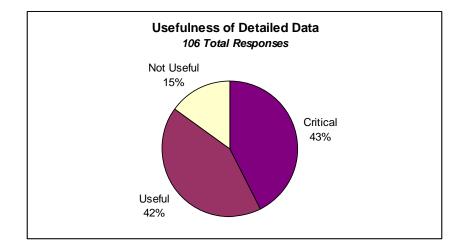
# Output files placed in special directory

|    | Microsoft Access | - [MGH_Labs : Tab                       | lel    |                 |               |                 | - 🗆 ×           |
|----|------------------|---|--------|-----------------|---------------|-----------------|-----------------|
|    |                  | Insert Format Recor                     | -      | Help            |               |                 | _ 문 ×           |
|    |                  |   |        |                 |               |                 |                 |
|    |                  | <b>). 🖤   X 📭 🖻</b> :                   |        | 2↓ X↓ 🍕 ¥       |               |                 | ·               |
|    | Test_Id          | Test_Description                        | Result | Result_Text     | Abnormal_Flag | g Reference_Uni | Reference_Range |
| L  | SQ-XPTT          | Superstat APTT                          | 29.8   |                 |               | sec             | 22.1-35.1       |
|    | SQ-PTT           | APTT                                    | 32.8   |                 |               | Sec             | 22.1-35.1       |
|    | SQ-PTT           | APTT                                    | 37.8   |                 | н             | Sec             | 22.1-35.1       |
|    | SQ-PTT           | APTT                                    | 46.4   |                 | Н             | sec             | 22.1-35.1       |
|    | SQ-PTT           | APTT                                    | 43.1   | MODERATELY      | Н             | sec             | 22.1-35.1       |
|    | SQ-PTT           | APTT                                    | 25.7   |                 |               | sec             | 22.1-35.1       |
|    | SQ-PTT           | APTT                                    | 23.7   |                 |               | sec             | 22.1-35.1       |
|    | SQ-PTT           | APTT                                    | 25.4   |                 |               | sec             | 22.1-35.1       |
|    | SQ-PTT           | APTT                                    | 24.7   |                 |               | sec             | 22.1-35.1       |
|    | SQ-PTT           | APTT                                    | 24.0   |                 |               | sec             | 22.1-35.1       |
|    | SQ-PTT           | APTT                                    | 24.7   |                 |               | sec             | 22.1-35.1       |
|    | SQ-XPTT          | Superstat APTT                          | 31.3   |                 |               | sec             | 22.1-35.1       |
|    | SQ-PTT           | APTT                                    | 34.5   |                 |               | sec             | 22.1-35.1       |
|    | SQ-PTT           | APTT                                    | 40.0   |                 | н             | sec             | 22.1-35.1       |
|    | SQ-PTT           | APTT                                    | 45.0   |                 | н             | sec             | 22.1-35.1       |
|    | SQ-XPTT          | Superstat APTT                          | 55.2   | *** Note: New n | Н             | sec             | 22.1-35.1       |
|    | SQ-PTT           | APTT                                    | 33.6   |                 |               | sec             | 22.1-35.1       |
|    | SQ-XPTT          | Superstat APTT                          | 34.3   |                 |               | sec             | 22.1-35.1       |
|    | SQ-PTT           | APTT                                    | 37.9   |                 | Н             | sec             | 22.1-35.1       |
|    | SQ-PTT           | APTT                                    | 22.6   |                 |               | sec             | 22.1-34.1       |
|    | SQ-PTT           | APTT                                    | 37.4   |                 | н             | sec             | 22.1-34.1       |
|    | SQ-PTT           | APTT                                    | 37.2   | SLT HEMOLYS     | н             | sec             | 22.1-34.1       |
|    | SQ-PTT           | APTT                                    | 35.1   |                 | Н             | sec             | 22.1-34.1       |
|    | SQ-PTT           | APTT                                    | 36.4   | MODERATE HE     | Н             | sec             | 22.1-34.1       |
| R  | ecord:           | 1 • • • • • • • • • • • • • • • • • • • | 108164 |                 |               | 1               | ·····           |
| -  | atasheet View    |   |        |                 |               |                 |                 |
| 12 |                  |   |        |                 |               |                 | //              |

#### Files include Small Database

#### One year's usage of RPDR

- 4526 registered users, 1113 new in just 2019
- 834 teams/year gathering data for research studies
- 4472 detailed patient data sets returned to these teams in 2019, containing data of 24.7 million patient records.
- From a survey of 153 teams
  - Importance of the data received from the RPDR was evaluated in relation to the study it was supporting.
  - Calculated over 4 years (FY15-FY19) the total agreement amounts were \$2.27 Billion, making per year consumption critically dependent on RPDR \$244 Million.



#### Rapid investigation of QTc prolongation

#### FDA warning 2011 for Celexa

Safety Announcement: [8-24-2011] "should no longer be used at doses greater than 40 mg per day because it can cause abnormal changes in the electrical activity of the heart."

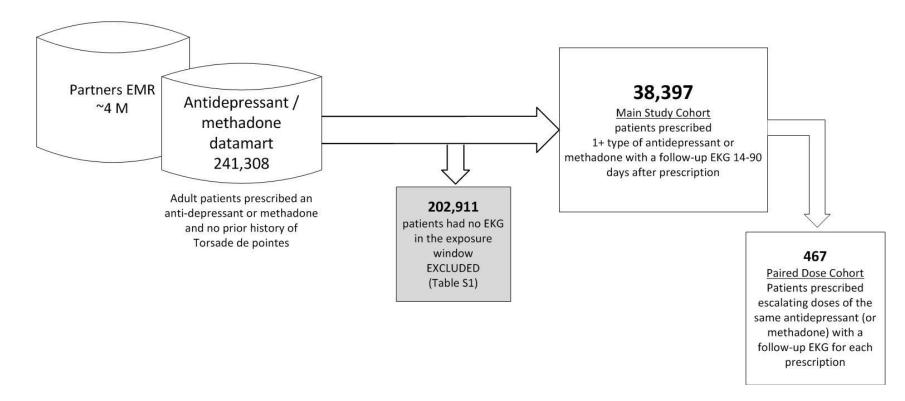
- But, did NOT include Lexapro (which is active ingredient of Celexa [s-enantiomer])
- Shown to be true with RPDRderived data set with >38,000 EKGs obtained within 14 – 90 day window after medication initiated

|   | Adjusted model <sup>+</sup> |         |  |  |  |  |
|---|-----------------------------|---------|--|--|--|--|
|   | prolongatio                 | p-value |  |  |  |  |
| Anti-depressant   | n                           |         |  |  |  |  |
| SSRI  |                             |         |  |  |  |  |
| Citalopram (Celexa)   | 2.85                        | 0.004   |  |  |  |  |
| Escitalopram (Lexapro)  | 3.80                        | < 0.001 |  |  |  |  |
| Fluoxetine (Prozac)   | 1.44                        | 0.150   |  |  |  |  |
| Paroxetine (Paxil)  | 0.07                        | 0.943   |  |  |  |  |
| Sertraline (Zoloft)   | 0.87                        | 0.383   |  |  |  |  |
| Other anti-depressants  |                             |         |  |  |  |  |
| Amitriptyline   | 4.10                        | < 0.001 |  |  |  |  |
| Bupropion   | -2.15                       | 0.032   |  |  |  |  |
| Duloxetine  | 0.60                        | 0.547   |  |  |  |  |
| Mirtazapine   | -1.46                       | 0.145   |  |  |  |  |
| Nortriptyline   | 1.23                        | 0.219   |  |  |  |  |
| Venlafaxine   | 1.15                        | 0.251   |  |  |  |  |
| previously known prolonger  |                             |         |  |  |  |  |
| Methadone   | 5.32                        | < 0.001 |  |  |  |  |
| Adjusted for age, gender, race, type of insurance,<br>history of major depression, history of myocardial<br>infarction and Charlson comorbidity score |                             |         |  |  |  |  |

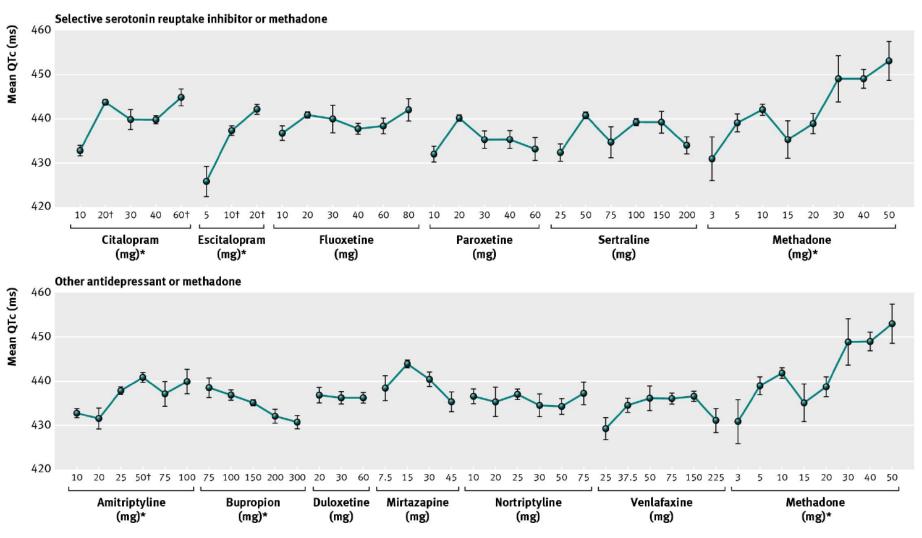
Roy Perlis MD, MSc and team

#### Relevant Cohorts of Patients are Gathered through RPDR and Detailed Data Obtained

- Medication use by individual patients over time
- Patient EKG QTc values at various time points



#### **Results: QTc interval and medication use**



<sup>\*</sup> Dose a significant predictor of QTc in fully adjusted linear models at  $\alpha$ =0.05

† QTc at specified dose is significantly different from that at prior dose in fully adjusted linear models at α=0.05

Mean (SD) corrected QT (QTc) interval recorded on electrocardiogram 14–90 days after prescription of antidepressant or methadone, by drug dose

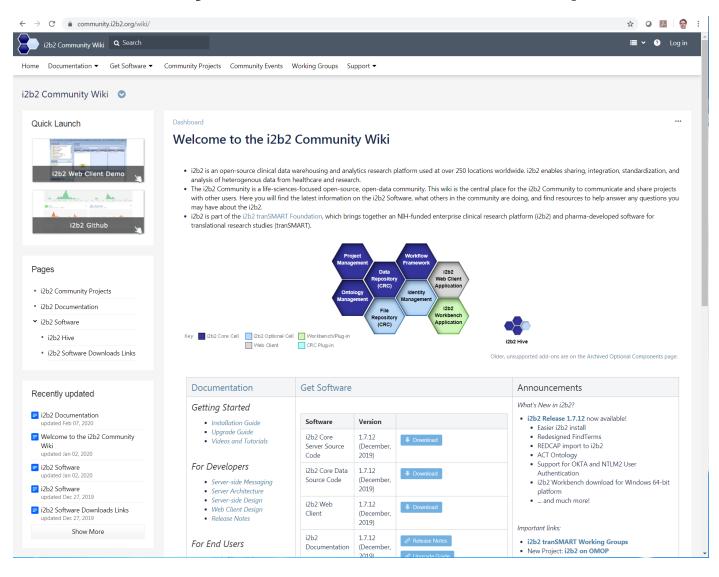
## High Throughput Methods for supporting Translational Research

- Set of patients is selected from medical record data in a high throughput fashion
- Investigators explore phenotypes of these patients using Machine Learning tools and a translational team developed to work specifically with medical record data
- Distributed networks cross institutional boundaries for phenotype selection, public health, and hypothesis testing
- Digital medicine is delivered into clinical care through Digital Twin

#### RPDR Evolved into international "Informatics for Integrating Biology and the Bedside (i2b2)" sponsored by the National Institutes of Health, what is it?

- Software for explicitly organizing and transforming personoriented clinical data to a way that is optimized for clinical genomics research
  - Allows integration of clinical data, trials data, and genotypic data
- A portable and extensible application framework
  - Software is built in a modular pattern that allows additions without disturbing core parts
  - Available as open source at <u>https://www.i2b2.org</u>

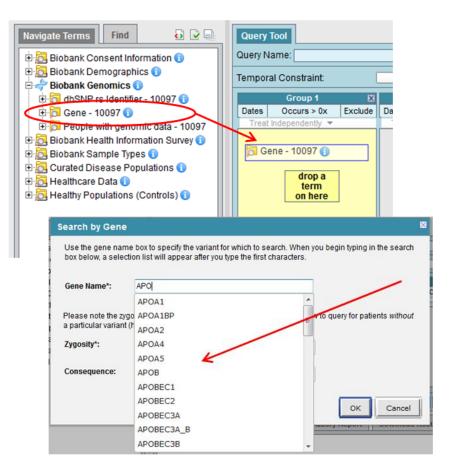
#### I2b2 Community Software distributed as open source

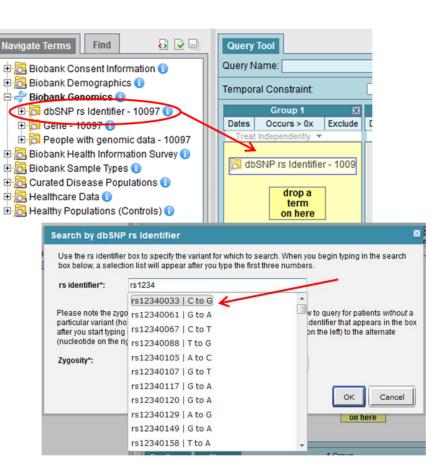


#### I2b2 Software adapts through new plugins

| Contraction - i2b2 Wiki - Windows Internet Explorer  |                      |
|--|----------------------|
| 🕢 💽 🗢 🔣 https://community.i2b2.org/wiki/dashboard.action 🛛 🔹 🔒 🍫 🗙 🚼 Googl   | e 🔎 🔽                |
| File Edit View Favorites Tools Help  |                      |
| Favorites Dashboard - i2b2 Wiki  |                      |
| <ul> <li>i2b2 Sponsored Project - i2b2 Web Client</li> <li>The i2b2 Web Client is one of several core projects that are directly sponsored by the i2b2 team.</li> </ul>  |                      |
| Image: system of the image: |                      |
| Related Project - Clinical Trender       @         The Clinical Trender aims to allow researchers to track and visualize certain clinical variables related to a selected p       @  | B)                   |
| Related Project - CRC Tester     A Workbench plugin that tests the CRC web services by Mike Mendis     A   | E                    |
| Image: Related Project - Crimson         Image: Related Project - Crimson           A project to make specimens available through i2b2 infrastructure, organized by Lynn Bry   |                      |
| Related Project - ExportXLS The i2b2 Web Client Plugin that tabulates patient data & applicable specified concepts; and facilitates export to spread   |                      |
| Image: Related Project - FACE caGrid CQL2 Data Source         Image: Related Project - FACE caGrid CQL2 Data Source           A caGrid/TRIAD data service that runs i2b2 queries via the RESTful interface to CRC         Image: Related Project - FACE caGrid CQL2 Data Source  |                      |
| <ul> <li>Related Project - Familial, Associational, &amp; Incidental Relationships (FAIR)</li> <li>Initiative.</li> <li>A collection of DBA tools and webclient plugins to facilitate the identification of related concepts amongst related pat</li> </ul>  |                      |
| <ul> <li>Related Project - Federated Query Simulations</li> <li>Simulations of federated query tools that return aggregate counts, such as SHRINE, by<br/>Griffin Weber</li> </ul>   | -                    |
| See Internet   Protected Mode: C   | off 🛛 🖓 👻 🔍 100% 👻 🚲 |

#### **Genotype Data**



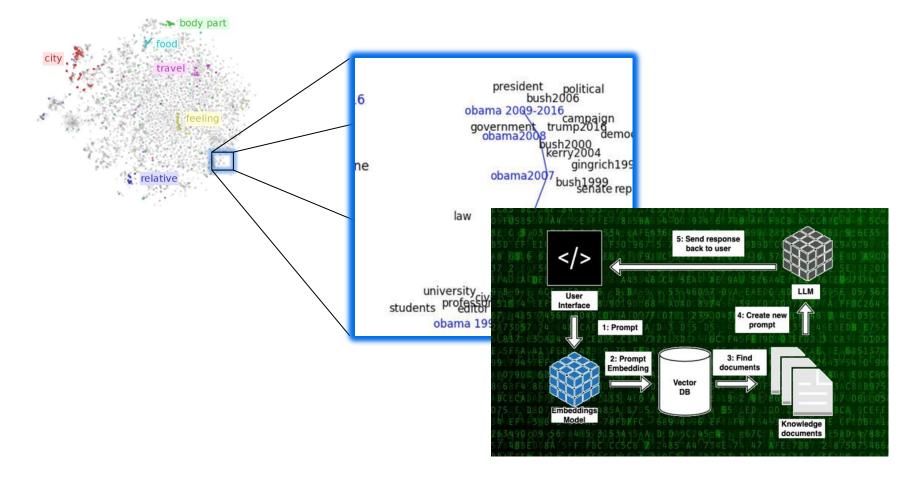


https://community.i2b2.org/wiki/display/IGD/Loading+Genomic+VCF+Files+into+i2b2

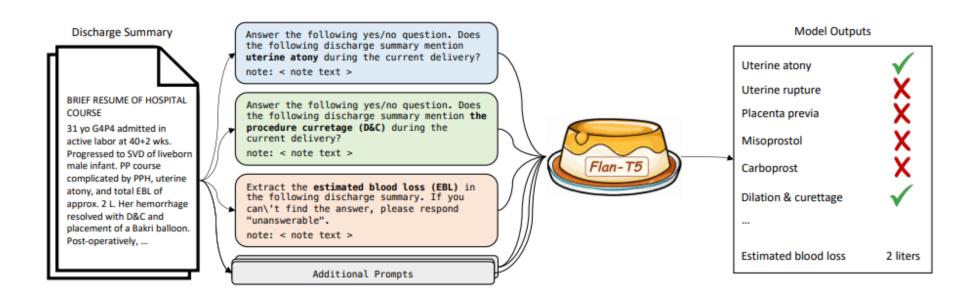
# Use NLP to extract the relevant features from the set of patient notes.



### LMM Enhanced interaction with Patient Representation



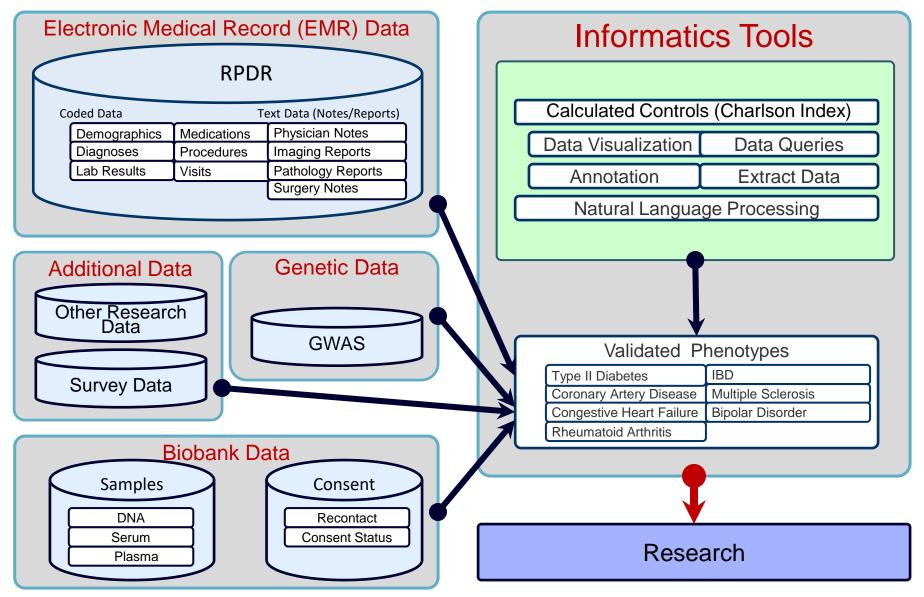
#### Medical conditions supported by description in chart



#### Emily Alsentzer et al

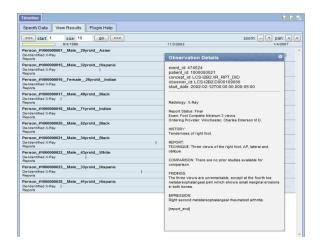
Zero-shot Interpretable Phenotyping of Postpartum Hemorrhage Using Large Language Models medRxiv preprint doi: https://doi.org/10.1101/2023.05.31.23290753

### **Data Integration in Big Data Commons**

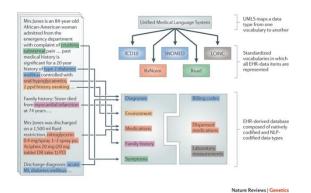


#### Curating a Disease Algorithm

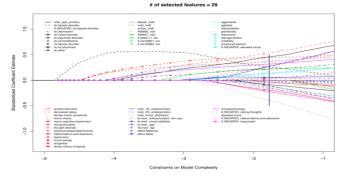
**1.** Create a gold standard training set.



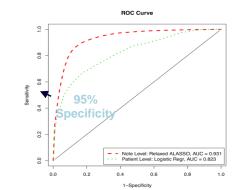
2. Create a comprehensive list of features from patient's electronic data that describe the disease of interest



3. Develop the classification algorithm. Using the data analysis file and the training set from step 1, assess the frequency of each variable. Remove variables with low prevalence. Apply adaptive LASSO penalized logistic regression to identify highly predictive variables for the algorithm



4. Apply the algorithm to all subjects in the superset and assign each subject a probability of having the phenotype



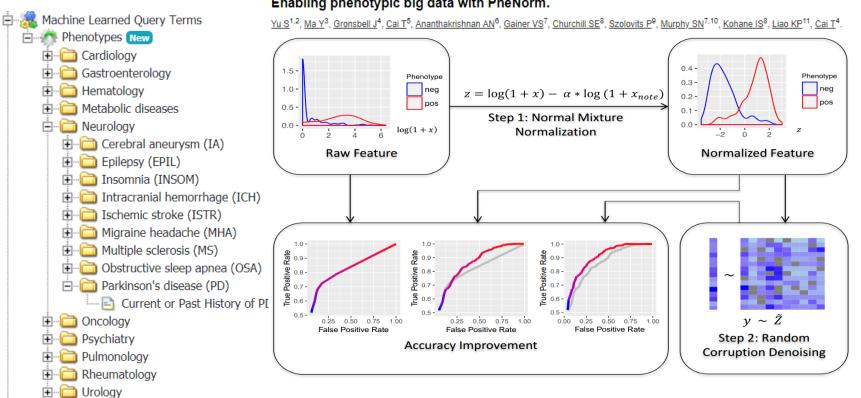
#### **Biobank Portal | Curated Diseases**

| Validated Phenotype      | Count* | Predictive Positive Value |
|--------------------------|--------|---------------------------|
| Bipolar Disease          | 71     | 89%                       |
| Congestive Heart Failure | 387    | 90%                       |
| Coronary Artery Disease  | 2,420  | 97%                       |
| Crohn's Disease          | 453    | 90%                       |
| Multiple Sclerosis       | 94     | 90%                       |
| Rheumatoid Arthritis     | 550    | 90%                       |
| Type 2 Diabetes Mellitus | 1,887  | 97%                       |
| Ulcerative Colitis       | 330    | 90%                       |

| Healthy Controls based on Charlson Index    | Count** |
|---|---------|
| 0 – 10-year survival probability is >98.3%  | 2,206   |
| 1 – 10-year survival probability is >95.87% | 4,343   |
| 2 – 10-year survival probability is >90.15% | 6,545   |

\* Based on 15,880 patients \*\* Based on 21,300 patients

#### Automated Learning Algorithms enabled in RPDR such as PheNorm Algorithm



J Am Med Inform Assoc. 2018 Jan 1;25(1):54-60. doi: 10.1093/jamia/ocx111.

#### Enabling phenotypic big data with PheNorm.

#### **Machine Learned Phenotypes**

- Abdominal hernia
- Acute bronchitis and bronchiolitis
- Acute pancreatitis ٠
- Alcoholism ٠
- Alzheimer's disease .
- Aortic aneurysm ٠
- Aplastic anemia
- Atrial fibrillation
- Atrioventricular block ٠
- Autism spectrum disorders ٠
- Basal cell carcinoma .
- **Bipolar** Disease
- Bladder cancer
- Brain cancer .
- Breast cancer ٠
- Cerebral aneurysm ٠
- Cholelithiasis ٠
- Chronic pancreatitis ٠
- Chronic sinusitis ٠
- Coronary atherosclerosis ٠
- Crohn's disease ٠
- Deep vein thrombosis ٠
- Depression
- Diverticulosis and diverticulitis
- Eating disorder
- Epilepsy
- Gastroesophageal reflux disease
- Gout
- Heart valve disorders

- Hyperlipidemia
- Hyperparathyroidism
- Hypertension
- Hypothyroidism •
- Insomnia •
- Intracranial hemorrhage .
- Ischemic stroke .
- Leukemia
- Lung cancer
- Melanoma
- Migraine headache
- Multiple sclerosis
- Myocardial infarction
- Neutropenia ٠
- Non-Hodgkin lymphoma
- Obesity ٠
- Obsessive compulsive disorder .
- Obstructive sleep apnea •
- Ovarian cancer
- Pancreatic cancer
- Parkinson's disease
- Peripheral vascular disease ٠
- Polycystic ovaries
- Prostate cancer
- Pulmonary heart disease
- Renal cancer .
- Renal failure .
- Schizophrenia
- Substance addiction

- Suicidal ideation
- Suicide attempt or self-inflicted injury
- Thyroid cancer ٠
- Tobacco use disorder ٠
- Type 1 diabetes
- Type 2 diabetes
- Ulcerative colitis
- Urinary calculus ٠
- Uterine cancer

- Pneumonia

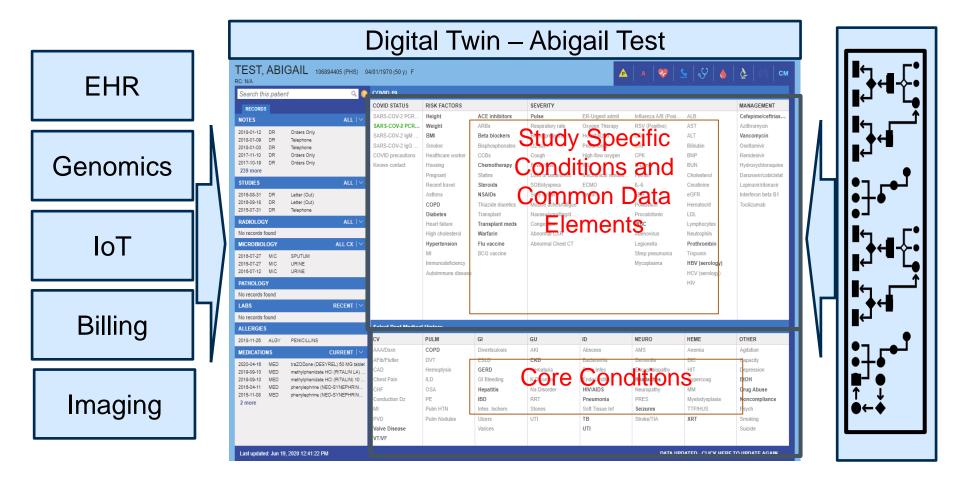
#### Phenotype Automation: Phenotype Quality Dashboard

|    |            | enotype Dashboa |        |                                       |                | Phe       | enotypes M | ethods | About us |  |                        |
|----|------------|-----------------|--------|---------------------------------------|----------------|-----------|------------|--------|----------|--|------------------------|
|    | category 🔷 | PheWAS_code 🔶   | abbr 🍦 | PheWAS_name 🍦                         | model 🔶        | ICD_PPV 🔶 | ICD_AUC 🔶  | AUC 🔶  | PPV 🔶    | TPR 🜩  |                        |
| 1  | ONC        | PheWAS:189.21   | BLCA   | Bladder cancer                        | PheNorm_ICD    | 0.80      | 0.903      | 1.000  | 1.00     | 0.42   |                        |
| 2  | ONC        | PheWAS:204      | LEUK   | <u>Leukemia</u>                       | PheNorm_ICD    | 0.73      | 1.000      | 1.000  | 1.00     | 0.91   |                        |
| 3  | PSYCH      | PheWAS:297.1    | SI     | Suicidal ideation                     | PheNorm_ICDNLP | 0.93      | 0.786      | 1.000  | 1.00     | 0.43   |                        |
| 4  | PSYCH      | PheWAS:305.2    | EATD   | Eating disorder                       | PheNorm_ICDNLP | 0.53      | 0.482      | 1.000  | 1.00     | 1.00   |                        |
| 5  | NEURO      | PheWAS:327.4    | INSOM  | Insomnia                              | PheNorm_ICDNLP | 0.93      | 0.821      | 1.000  | 1.00     | 0.50   |                        |
| 6  | CARDIO     | PheWAS:452.2    | DVT    | <u>Deep vein</u><br><u>thrombosis</u> | PheNorm_ICDNLP | 0.87      | 0.692      | 1.000  | 1.00     | 1.00   |                        |
| 7  | NEURO      | PheWAS:817      | CONC   | Concussion                            | PheNorm_NLP    | 0.73      | 0.682      | 1.00   |          | Phenotypes New   |                        |
| 3  | METAB      | PheWAS:250.1    | T1DM   | <u>Type 1 diabetes</u>                | PheNorm_ICD    | 0.17      | 0.882      | 0.9    |          | Aortic aneurysm (A   |                        |
| 9  | ONC        | PheWAS:184.11   | OVCA   | Ovarian cancer                        | PheNorm_ICDNLP | 0.60      | 0.926      | 0.9    |          | Atrial fibrillation (AF     Atrioventricular bloc  |                        |
| 10 | ONC        | PheWAS:182      | UTCA   | Uterine cancer                        | PheNorm_ICD    | 0.50      | 0.867      | 0.9    |          | Coronary atheroscle     Deep vein thrombos   |                        |
| 11 | GI         | PheWAS:555.1    | CD     | Crohn's disease                       | PheNorm_mean   | 0.54      | 0.961      | 0.9    |          | 🗄 🫅 Heart valve disorde  | rs (HVD)               |
|    |            |                 |        |                                       |                |           |            |        |          | Hypertension (HTN)     Hypertension (HT | n (MI)<br>disease (PVD |

#### **High Quality Phenotypes for Research Studies**

| <u>File Edit V</u> iew Hi <u>s</u> tory <u>B</u> ookmarks <u>T</u> ools <u>H</u> elp  |   | _                  |   |             |   |  | _ 0                     | x        |
|---|---|--------------------|---|-------------|---|--|-------------------------|----------|
| Partners Biobank Portal × +   |   |                    |   |             |   |  |                         |          |
| Employed Antropy A-0/?user=snm0   | C   | <b>Q</b> Search    | +   |             | 1 🖻 🛡   | 1 🐖  | -                       | ≡        |
| PARTNERS BIOBANK<br>PORTAL Biobank Portal Genomic Pilot [Logout]  |   | 🚨 Find Patients    | S Make Reques   | at 🔛 H      | elp & Support   | Shawr  | n Murphy, M             | ID       |
| Navigate Terms       Find         Biobank Consent Information ()       Biobank Consent Information ()         Biobank Demographics ()       Biobank Genomics ()         All people with genomic data - 9358         Carbon dbSNP rs Identifier - 4930 ()         Biobank Health Information Survey ()         Biobank Sample Types ()         Curated Disease Populations ()         Asthma (AST) ()         Bioplar Disorder (BD) ()         Bioplar Disorder (BRCA) ()  | Query Tool<br>Query Name: Prima-CHFGen<br>Temporal Constraint:<br>Group 1<br>Dates Occurs > 0x<br>Treat Independently ▼<br>Primary dilated cardiomy<br>4002 | X<br>Exclude Dates | Treat all group<br>Group 2<br>Occurs > 0x<br>Independently *<br>F - current or past hist<br>700 | Exclude     | Dates<br>Treat Indepe<br>Concerning<br>Homozygous<br>missense O | ✓<br>Group 3<br>Occurs > 0x<br>andently ✓<br>Intains "TTN ANI<br>s AND (Framesl<br>R nonsense OF<br>R stop_loss)"]<br>One or | Exclude<br>D<br>Dift OR |          |
| Corporative Heart Failure (CHF) Congestive Heart Failure (CHF) Congestive Heart Failure (CHF) CHF - current or past history (PPV 0.90) - 700 CHF - no history (NPV 0.99) - 36024 Coronary Artery Disease (CAD)  | Run Query Clear   |                    | more of<br>these  |             |   | more of<br>these   |                         |          |
|   |   |                    | ·   |             |   |  |                         | <u> </u> |
| ⊕ 🔂 Epilepsy (EPIL) 🕦   | Show Query Status Gra   | aph Results Que    | ery Report Downl  | oad Results | S   |  | Ę                       | 1        |
| <ul> <li>Gout (GOUT) 1</li> <li>Hypertension (HTN) 1</li> <li>Multiple Sclerosis (MS) 1</li> <li>Obesity (OBES) 1</li> <li>Rheumatoid Arthritis (RA) 1</li> </ul>   |   |                    | Number of patients  | 5           |   |  |                         |          |
| Schizophrenia (SCZ) (1)  Type 1 Diabetes Mellitus (T1DM) (1)  Type 2 Diabetes Mellitus (T2DM) (1)  Type 2 Diabetes (1)  Type 2 D |   | For Query "        | "Prima-CHFGene(   | @14:22:40   | )"  |  | -                       | -        |

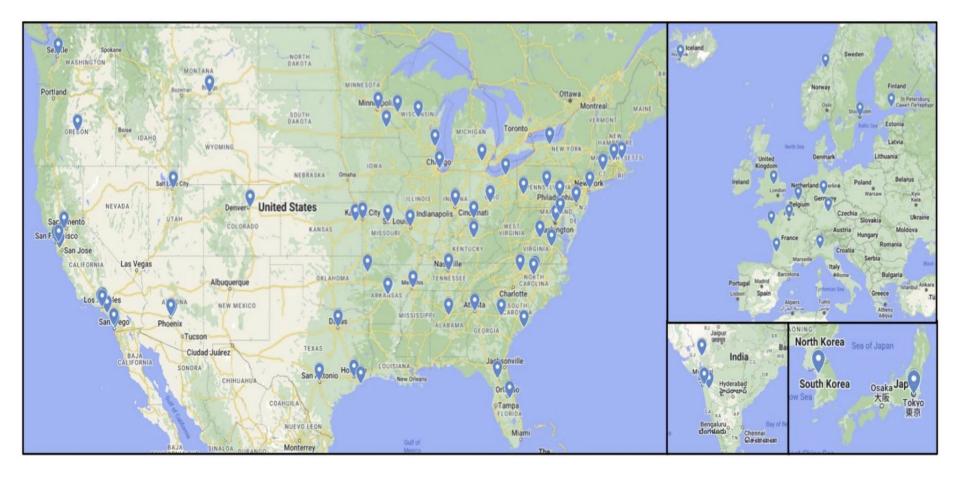
# Combined with Generative AI can produce \_ Digital Twin of Patient



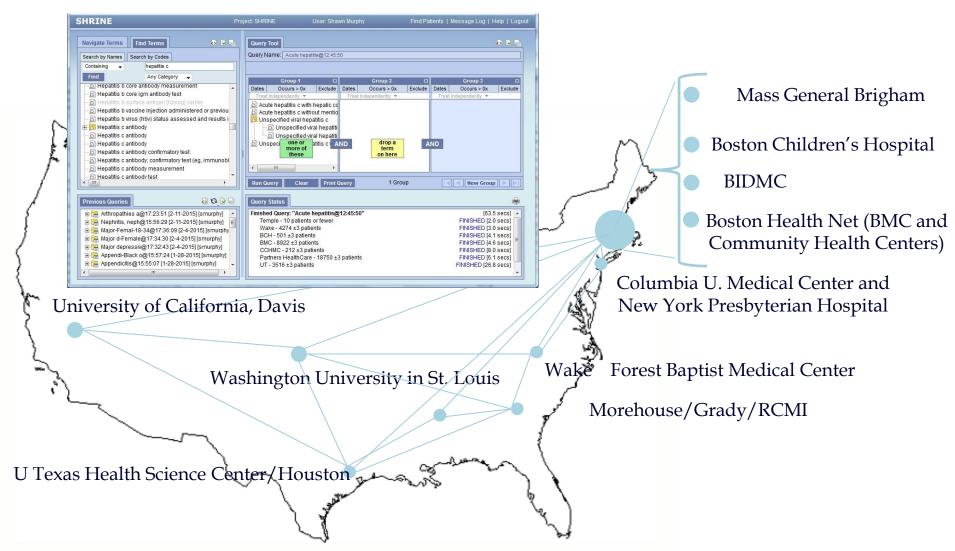
## High Throughput Methods for supporting Translational Research

- Set of patients is selected from medical record data in a high throughput fashion
- Investigators explore phenotypes of these patients using Machine Learning tools and a translational team developed to work specifically with medical record data
- Distributed networks cross institutional boundaries for phenotype selection, public health, and hypothesis testing
- Digital medicine is delivered into clinical care through Digital Twin

#### **I2b2 Implementations** >250 across the USA and Internationally, some illustrated below:



# **Federated Queries**



### **Drive Pragmatic Clinical Studies**

| 🗅 i2b2 Web Client 🗙 🔽                            |              |           | _             |           |                | -   | L             | - 0 <mark>-</mark> × |            | -  |           |          |           |          |               |              |              |          |             |          |            | <u> </u>  |          |          |
|--|--------------|-----------|---------------|-----------|----------------|---|---------------|----------------------|------------|--|-----------|----------|-----------|----------|---------------|--------------|--------------|----------|-------------|----------|------------|-----------|----------|----------|
| → C 🛈 Not secure   i2b2act.dipr.partners.org/AC  | TFullDemo/   | #         |               |           |                |   |               | ፃ 🏠 🚺 (              | <b>O</b> I | s.org/ACTFull                                    | Demo/#    |          |           |          |               |              |              |          |             |          |            |           | 키 ☆      |          |
| b2 Ouery & Analysis Tool Project ACT Develop     | nentUser: Cr | oniob Sei | vice Find Pat | ients   A | Analysis Tools | I Message Log   Heli  | p   Change Pa | ssword   Logo        | ut :1      | F Development l                                  | Jser: Cro | niob Se  | ervice Fi | nd Patie | ents I A      | nalvsis      | Tools        | 🚽 I Me   | ssage Lo    | oa I He  | lp   Ch    | nange Pa  | asswor   | rd   Loc |
|  |              |           |               |           |                |   |               |                      |            |  |           |          |           |          |               |              |              |          |             |          |            |           |          |          |
| Navigate Terms Find 💦 🔯 🖳                        | Patient Se   | t Viewer  | 1             |           |                |   |               | ð 🕑 🕠                |            | - Patient Data A                                 | nalysis   |          | _         |          |               |              |              |          |             |          | d          | 🔗 Gliaph  | I        | ∄ Table  |
| E Cardiovascular agents                          | Specify D    | ata Vi    | ew Patients   |           |                |   |               |                      | -          |  |           |          |           |          |               |              |              |          |             |          |            |           |          |          |
| 🖻 🔂 Central nervous system agents                |              |           |               |           |                |   |               |                      |            | Vitals   |           |          | _         | Me 🗹     | dications     |              |              |          | _           | 🖊 Labs   |            |           |          |          |
| ⊕- C Analgesics                                  |              | 11016     | 4 F           | 65        | WHITE          |   | Yes           | Nc                   |            | Weight, Blood Pres                               | sure      |          | •         | Me       | toprolol, Spi | ironolactor  | ne, Valsarta | an 🔻     |             | Creatini | ne, Potass | ium, eGFR |          | •        |
| Anorexiants     Anticonvulsants                  |              |           |               |           |                |   | 100           |                      |            |  |           |          |           |          |               |              |              |          |             |          |            |           |          |          |
| Antiemetic antivertigo agents                    |              | 12310     | 14 F          | 63        | WHITE          |   | Yes           | NC                   |            |  |           |          |           |          |               |              |              |          |             |          |            |           |          |          |
|  |              | 12319     | 2 M           | 64        | WHITE          |   | Yes           | Nc                   |            |  |           |          |           | Viewinį  | g 2 weeks st  | tarting from | n 01/0       | 07/17 🗎  |             |          |            |           |          |          |
| 🕀 🔂 General anesthetics                          |              | 12010     | 12 111        | 04        |                |   | 163           | THE .                |            |  |           |          | ER Visit  | s —      | ICU Visits    | Fic          | oor Visits   | CI       | inic Visits | _        |            |           |          |          |
| 🕀 🔂 Miscellaneous central nervous system         |              | 12441     | 5 M           | 62        | HISPANIC       |   | Yes           | Nc                   |            |  |           |          | 14        |          | 2             | 5            | 6            | 3        | 3           |          |            |           |          |          |
| agents 🗸   |              | 12667     | '3 F          | 76        | WHITE          |   | Yes           | Nc                   |            |  | 01/07/17  | 01/08/17 | 01/09/17  | 01/10/17 | 01/11/17      | 01/12/17     | 01/13/17     | 01/14/17 | 01/15/17    | 01/16/17 | 01/17/17   | 01/18/17  | 01/19/17 | 01/20/1  |
| orkplace 🔂 🖏                                     |              | 13070     | 12 M          | 81        | WHITE          |   | Yes           | Nc                   |            | Events   |           |          |           |          | ICD           |              |              |          |             |          |            |           |          |          |
| Cronjob  |              | 13077     | '3 M          | 80        | UNKNOWN        | [Sinemet 25/100 -<br>LMR 691](15)   | Yes           | Nc                   |            | View.All   |           | Echo     |           |          | Implant       |              |              |          |             |          |            |           |          |          |
|  |              | 13992     | 6 M           | 68        | WHITE          | [Sinemet 25/100 -<br>LMR 691](5)  | Yes           | Nc                   |            | Weight<br>(Ibs)<br>View All                      | 200       | 202      | 230       | 150      | 400           | 430          | 500          | 506      | 100         |          |            |           |          |          |
| - <b>§</b> 160803<br>- <b>§</b> 24339            |              | 15143     | 0 M           | 62        | WHITE          |   | Yes           | Nc                   |            | Blood Pressure                                   |           |          |           |          |               |              |              |          |             |          |            |           |          |          |
| - 60677<br>- 91326                               |              | 15849     | 2 F           | 78        | WHITE          |   | Yes           | N                    |            | (mmHg)<br>View All                               |           |          |           |          |               |              |              |          |             |          |            |           |          |          |
| evious Queries Find Q 🗘 🖓 🖓                      | V            | 16080     | 13 F          | 63        | WHITE          | [Sinemet 25/100 -<br>LMR 691](2)<br>[Pramipexole<br>(mirapex) - LMR<br>2639](2) | Yes           |                      |            | Metoprolol<br>(mg)<br>View All<br>Spironolactone |           |          |           | 25       | 25            | 25           | 25           | 50       | 50          | 50       | 50         | 50        | 50       | 50       |
| Parkins-Tobacco@12:12:29 [4-12-2017] [cronjob]   |              | 16612     | 9 M           | 62        | WHITE          | [Sinemet 25/100 cr -<br>LMR 1069](8)  | Yes           | N                    | 1          | : (mg)<br><u>View All</u>                        | 50        | 50       | 50        |          |               |              |              |          |             |          |            |           |          |          |
| - Parkins-Tobacco@12:10:33 [4-12-2017] [cronjob] |              | 17179     | 4 M           | 62        | WHITE          |   | Yes           | Nc                   |            | Valsartan<br>(mg)                                |           |          |           |          |               |              |              |          |             |          |            |           |          |          |
| Fetus or newbor@12:14:33 [4-11-2017] [cronjob]   |              |           |               |           |                | [Sinemet 25/100 -   |               |                      |            | View All   |           |          |           |          |               |              |              |          |             |          |            |           |          |          |
| Parkins-Tobacco@11:59:18 [4-11-2017] [cronjob]   |              | 17776     | i2 M          | 74        | WHITE          | [Sinemet 25/100 -   | Yes           | Nc                   |            |  |           |          |           |          |               |              |              |          |             |          |            |           |          |          |
| Parkins-Tohacco@11:58:24 [4-11-2017] [cronioh]   |              |           |               |           |                |   |               | •                    |            | Creatinine                                       |           |          |           |          |               |              |              |          |             |          |            |           |          |          |

### **RECOVER Study Data Harmonization**



### RECOVER: Researching COVID to Enhance Recovery

The National Institutes of Health (NIH) created the RECOVER Initiative to learn about the long-term effects of COVID.

The goal of RECOVER is to rapidly improve our understanding of and ability to predict, treat, and prevent PASC (post-acute sequelae of SARS-CoV-2), including Long COVID.

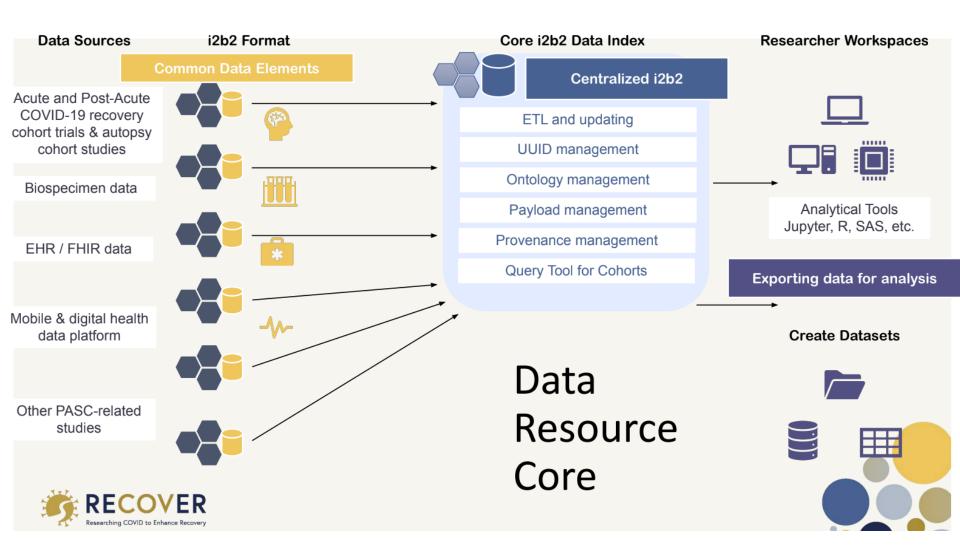
LEARN MORE ABOUT LONG COVID (



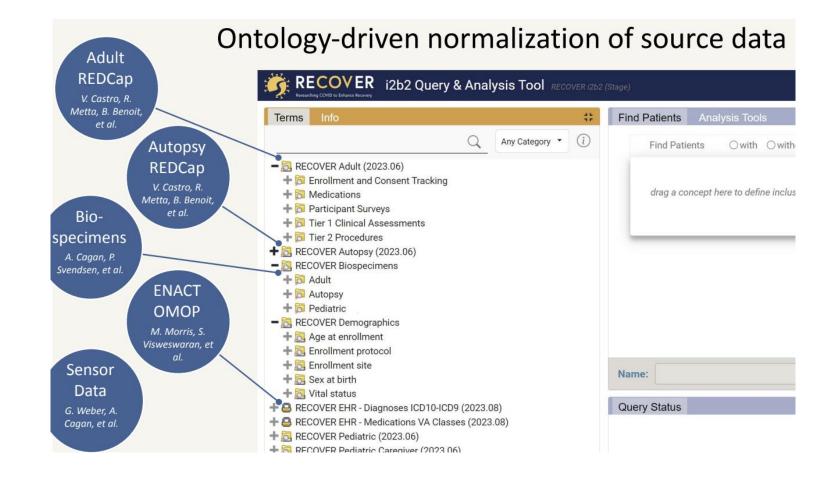


https://recovercovid.org

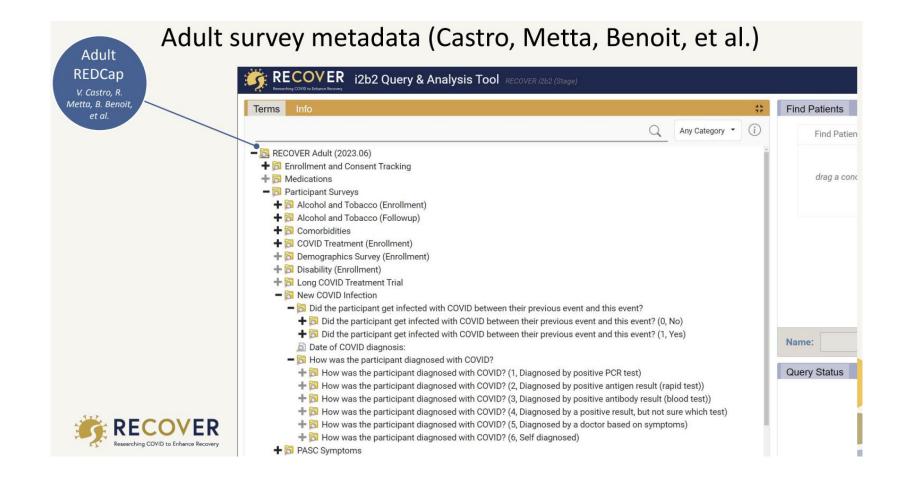
### Data harmonized within i2b2 star schema



# Concepts in database available in harmonized ontology



### Includes study participant survey data



### New i2b2 Query Tool to be released:

| 2b2 Query & Analysis Tool   |              |  |                    | ?      |
|---|--------------|--|--------------------|--------|
| Queries   |              | Find Patients Analysis Tools   |                    | 0      |
| Q Any Ca  | tegory •     | Find Patients Owith Owithout Owhen   | 0                  |        |
| + 🔚 Diabetes mellitus-Disorders of parathyroid gland@10:15:45 [9-22   | 2-2023] [dem | Event 1  |                    |        |
| 🕂 🔚 Query 8 - Timeline Notes [9-22-2023] [demo]   |              | Endocrine, nutritional and metabolic diseases (e00-e89)  |                    |        |
| ┿ Image Operations on lymphatic system-Anti-infectives@8:21:45 [9-22-2<br>+ Image Operations on lymphatic system-Anti-infectives@8:21:45 [9-22-2  |              | or drag additional concepts  |                    |        |
| ┿ Image Operations on lymphatic system@8:20:4 [9-22-2023] [demo] ┿ Image Operations on lymphatic system@8:20:4 [9-22-2023] [demo]   |              | Set date range this entire panel $$  |                    |        |
| ┿ Image Provide the Image Provided HTML HTML HTML HTML Image Provided HTML HTML HTML HTML HTML HTML HTML HTML   |              | The start of the first occurrence of Event 1 occurs before the start of the first occurrence of Event 2 $\sim$ |                    |        |
| <ul> <li>Query 5 -Capoten [9-22-2023] [demo]</li> <li>The Control of the Control of the</li></ul> |              | Event 2  |                    |        |
|   |              | Hormones   |                    | 1.1    |
|   |              |  |                    |        |
|   |              | or drag additional concepts  |                    |        |
|   |              | Set date range this entire panel $\sim$  |                    |        |
|   |              |  |                    | _      |
|   |              | Name: (t) Query 4- Female-Age Temporal - Endocrine Hormones  | Find Patients Clea | ar All |
| Tigestive syste@05:36:20 [9-22-2023] [demo]   |              |  |                    |        |
| here mild dementia@10:53:51 [9-22-2023] [demo]  |              | Query Status   |                    | 5      |
| 🔚 Certain-Disease@13:48:54 [9-22-2023] [demo]<br>🔚 Livin- 18Acute@20:29:26 [9-22-2023] [demo]   |              | Finished Query: (t) Query 4- Female-Age Temporal - Endocrine Hormones  | [0.5 secs] View Re | aport  |
|   |              |  | VIEW RE            | pon    |
| Liv- 1-Pri-Acu@20:26:08 [9-22-2023] [demo]  |              | Compute Time: 0.4 secs   |                    |        |
| 🕂 🔯 Results of Liv- 1-Pri-Acu@20:26:08 [9-22-2023] [demo] - FINI  | SHED         | Number of patients   |                    |        |
| 🕂 🔚 L5-5-5-5-5-6-A@20:22:37 [9-22-2023] [demo]  |              | 6  |                    |        |
| 🕂 🔚 Livin- 18Acute@20:19:41 [9-22-2023] [demo]  |              | 6  |                    |        |
| + 🔚 Livin- 18Acute@20:18:47 [9-22-2023] [demo]  |              |  |                    |        |
| 🕂 🔚 Femal-Essen-Tvpe @20:14:21 [9-22-2023] [demo]   |              |  |                    |        |

#### https://i2b2transmart.org/2023-i2b2-symposium/2023-symposium-recordings-slides/

### **I2B2 AI**

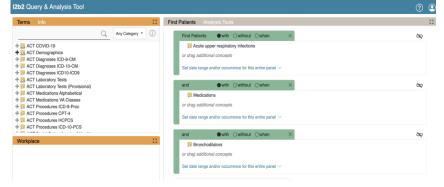
#### User asks a question

i2b2 Query & Analysis Tool

? (2)

I want to find all patients with a diagnosis of acute respiratory

#### Result rendered in web client



#### Al returns response in i2b2 format

I want to find all patients with a diagnosis of acute respiratory infections who have also been prescribed bronchodilators.

#### <query\_definition>

<query\_name>Acute respiratory i@19:45:36</query\_name>

<query\_timing>ANY</query\_timing>

<specificity\_scale>0</specificity\_scale>

#### <panel>

<panel\_number>1</panel\_number>

<panel\_timing>ANY</panel\_timing>

<panel\_accuracy\_scale>100</panel\_accuracy\_scale>

<invert>0</invert>

<total\_item\_occurrences>1</total\_item\_occurrences>

```
<item>
```

<hlevel>3</hlevel>

<item\_name>Acute respiratory infections</item\_name>

<item\_key>\\iSyph\Diagnoses\Respiratory system (460-519)\Acute respiratory infections (460-466)\</item\_key>

<item\_icon>FA</item\_icon>

<tooltip>Diagnoses \ Respiratory system \ Acute respiratory infections</tooltip>

<class>ENC</class>

<constrain\_by\_date>

<date\_from/>

</constrain\_by\_date>

<item\_is\_synonym>false</item\_is\_synonym>

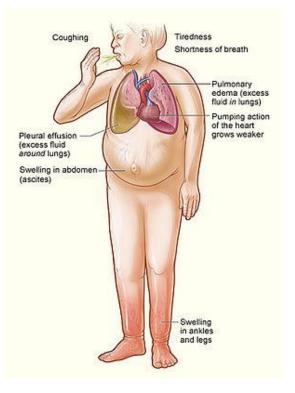
</item>

</panel>

# High Throughput Methods for supporting Translational Research

- Set of patients is selected from medical record data in a high throughput fashion
- Investigators explore phenotypes of these patients using Machine Learning tools and a translational team developed to work specifically with medical record data
- Distributed networks cross institutional boundaries for phenotype selection, public health, and hypothesis testing
- Digital medicine is delivered into clinical care through Digital Twin

### **Congestive Heart Failure**



### Affects 2% of the adult population

Risk of death first year after diagnosis: 35%

 In patient hospital costs in 2011: \$10.5B which is a small fraction of all heart failure related care

### Early Detection of Worsening or Improving Anemia

#### **Background and Methods**

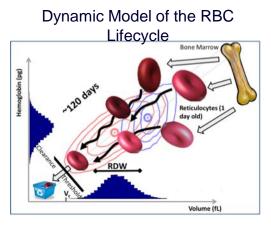
- Anemia is one of the strongest predictors of morbidity and mortality in CHF.
- Increasing or decreasing HGB is a further strong predictor, but there is no good way to determine whether a patient's HGB is on its way up or down (*Circulation. 2005;112:1121-1127*)

#### **Results and Conclusions**

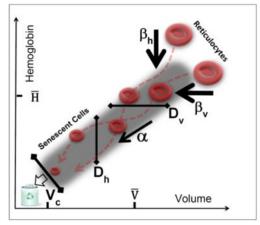
 A novel mathematical model of the RBC lifecycle enables estimation of patient-specific rates of RBC maturation and turnover from a routine CBC.

#### **Applications**

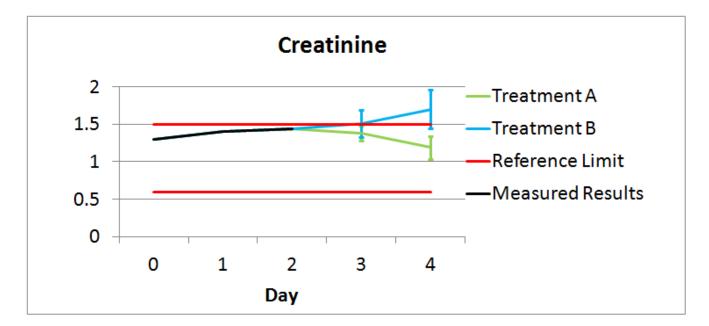
- 1. CHF patients most likely to have decreasing HGB may benefit from altered treatment or longer hospitalization to avoid readmission.
- 2. CHF patients most likely to have increasing HGB may be responding well to treatment and benefit from earlier discharge or maintenance of current therapy.



#### **Quantify Maturation and Clearance Rates**

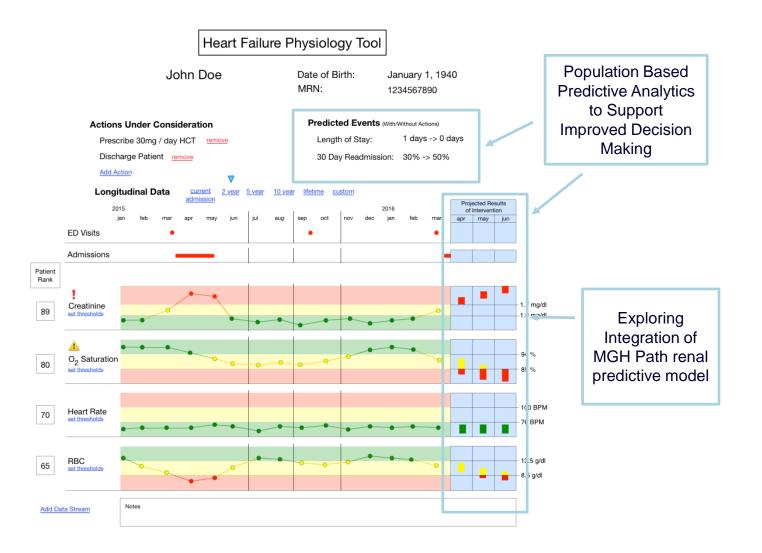


### **Creatinine Prediction: Hypothetical Application**

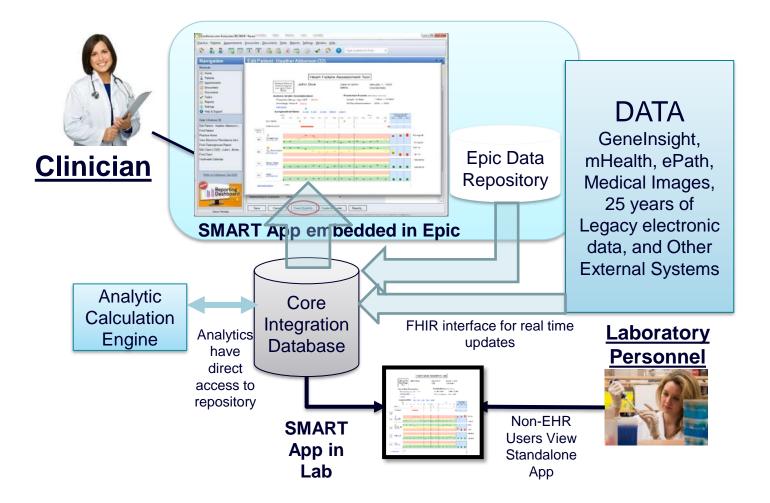


•Hypothetical analysis of creatinine times series where possible treatments are introduced into the model

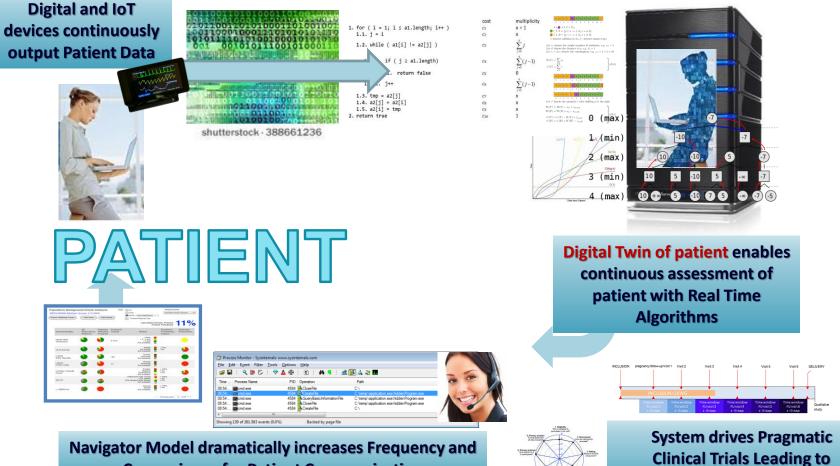
•The model hypothetically provide a future trajectory conditioned on each treatment



### Bringing Big Data into Clinical Care with Open App Development



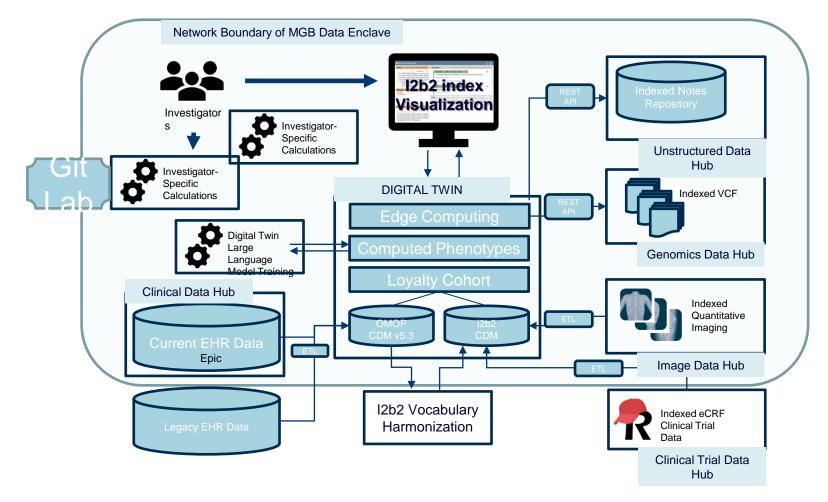
### **Transforming Care in the Digital Age**



**Continuous Process Improvement** 

Convenience for Patient Communication

#### **MGB** Data Enclave Overview







## I2b2 tranSMART Software

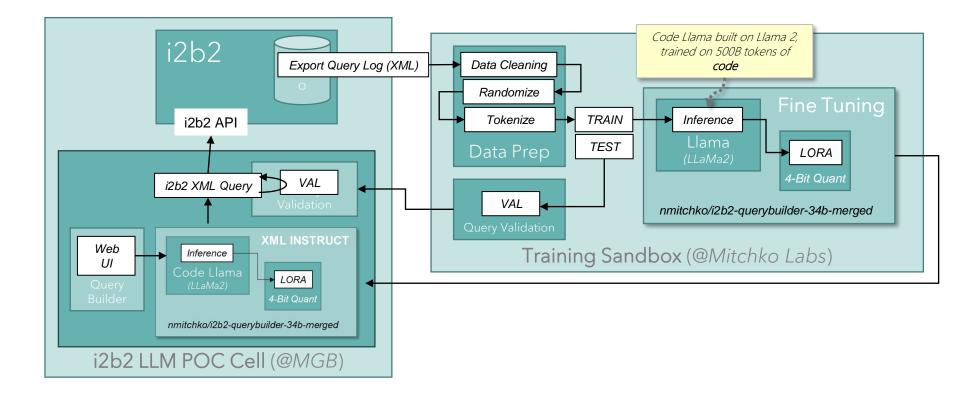
i2b2 Homepage (<u>https://www.i2b2.org</u>) i2b2 Software (<u>https://www.i2b2.org/software</u>) i2b2 Community Site (<u>https://community.i2b2.org</u>) https://i2b2transmart.org/2023-i2b2-symposium/2023symposium-recordings-slides/ A National Center for Biomedical Computing





### **APPENDIX**

### **AI-ENABLED QUERY BUILDER:** (I.E. INSTRUCTION-TUNED POC)



# **12B2 TRAINING ENVIRONMENT**

#### Training Requirements

| GPUs (48GB of VRAM or More) |
|-----------------------------|
| Patience                    |
| Cooling                     |

#### Training Run

3x Nvidia RTX 6000 3 days time 271 Quadrillion Floating Point Operations 10.05 epochs (runs through the dataset)

#### Software

Qlora – State of the art mixed precision training library Transformers, Pytorch, Accelerate – ML Libraries

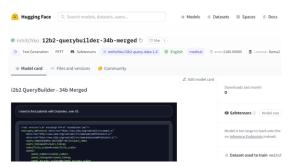
#### Experience:

Nicholai Mitchko background in ML and other model finetunes

#### All Open source

#### Found at

#### https://huggingface.co/nmitchko



## **I2B2 AI DEVELOPMENT SANDBOX**

#### WebUI oobabooga

#### https://github.com/oobabooga/text-generation-webui

| Download LLM   |               |
|--|---------------|
| Download model or LoRA<br>Enter the Hugging Face username/model path, for instance: fac<br>after a ":" character like this: facebook/galactica-125m:main. To |               |
| hmitchko/i2b2-querybuilder-34b-merged  |               |
| File name (for GGUF/GGML)  |               |
| Download   | Get file list |

nmitchko/i2b2-querybuilder-34b-merged

#### Load Model

| Model                                |   |   |   |              |        |        |                  |
|--------------------------------------|---|---|---|--------------|--------|--------|------------------|
| nmitchko_12b2-querybuilder-34b-merge | • | 5 |   | Load         | Unload | Reload | Save<br>settings |
| Model loader                         |   |   |   |              |        |        |                  |
| Transformers                         |   |   |   |              |        |        |                  |
| 6                                    |   |   |   |              |        |        |                  |
| gpu-memory in MiB for device :0      | 0 |   |   |              |        |        |                  |
|                                      |   |   |   | load-in-8bit |        |        |                  |
| cpu-memory in MiB                    | 0 |   |   | bf16         |        |        |                  |
|                                      |   |   |   | auto-devices |        |        |                  |
| load-in-4bit params:                 |   |   |   | disk         |        |        |                  |
| compute_dtype                        |   |   |   | load-in-4bit |        |        |                  |
| bfloat16                             |   |   |   |              |        |        |                  |
|                                      |   |   | 2 | use_double,  | _quant |        |                  |

Compute\_dtype: bfloat16 Check: bf16 Check: load-in-4bit Check: use\_double\_quant

#### Parameters



Divine Intellect

#### Enable API

| Apply flags/extensions and restart  |   |  |  |  |  |  |  |
|---|---|--|--|--|--|--|--|
| Toggle 🕈  | Save UI defaults to settings.yam  |  |  |  |  |  |  |
| Valiable extensions Valiable extensions Valiable extensions Valiable extensions Valiable Val | Boolean command-line flags<br>api<br>auto_launch<br>deepapeed<br>Giston<br>model_menu<br>monkey_patch<br>mult_user<br>no_cache<br>no_steam<br>public_api<br>mvkv_cuda_on<br>sdp_attention<br>share<br>verbose<br>xformers |  |  |  |  |  |  |

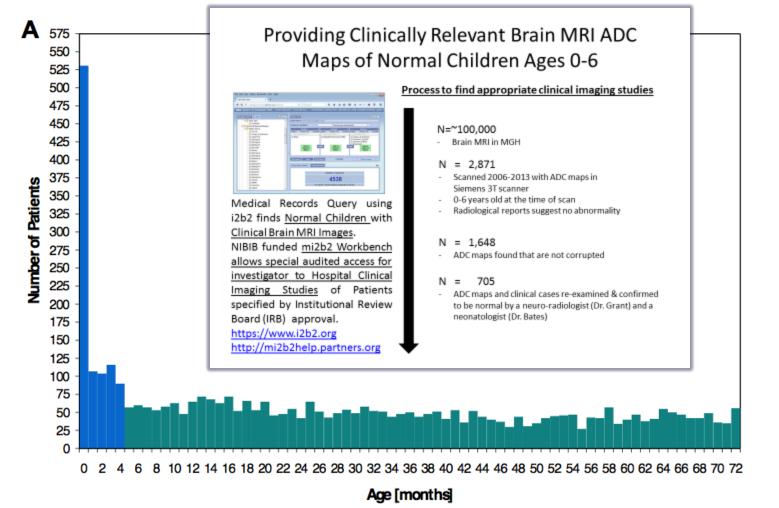
#### I2b2 AI Cell

https://github.com/i2b2/i 2b2-ai

#### I2b2 Web Client

https://github.com/i2b2/i 2b2-webclientprototype/tree/i2b2\_ai/

# Find Normal Brain MRI's of Children



Number of patients who had a brain MRI scan at a particular age in months from 0 to 6 years (A) and in weeks from 0 to 4 months (B)

# Atlases provide a visual guide for Radiology Decision Support, such as determining Perinatal Hypoxic Ischemic Encephalopathy

ADC map from 4 infants: Each statistically compared to age matched atlas yields visual guide to pathology

Quantitative analysis tools + large data sets = Great insights for practicing doctors

