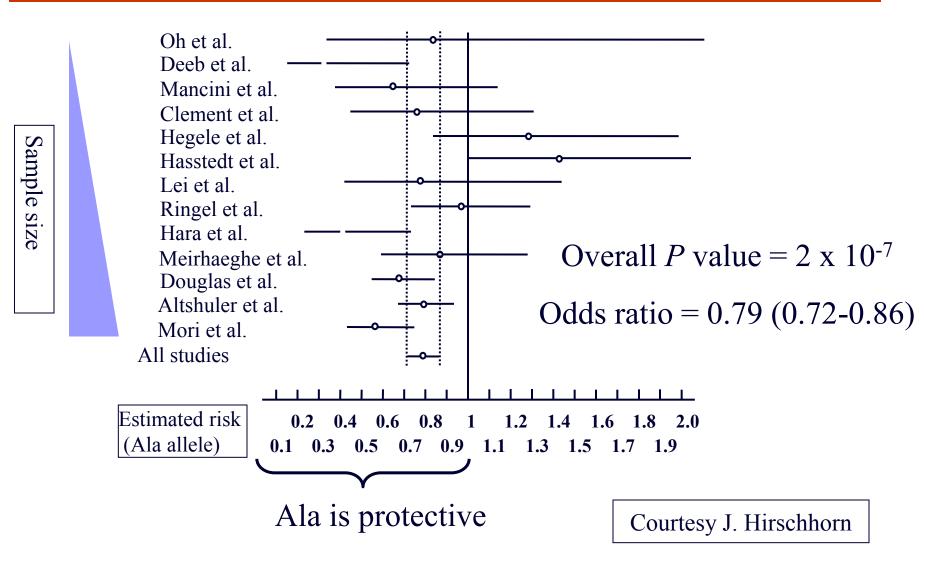




Computing our Patient's Future Using Data from our Healthcare Institutions

Shawn Murphy MD, Ph.D. NETTAB 2011 Workshop on Clinical Bioinformatics

Example: PPARy Pro12Ala and Diabetes



The Power of Numbers: Efficiently Reaching a Large N

- High throughput genotyping
- High throughput phenotyping
- High throughput sample acquisition

DHHS Secretary's Advisory Committee on Genetics, Health, and Society (SACGHS) argues for the health value of a 500,000 to 1M subject study. Estimated cost: \$3,000,000,000

Cost of the pediatric 100,000 study recently launched >> \$1B + decades.

High Throughput Methods for supporting Research at Partners Healthcare

- Set of patients is selected from medical record data in a high throughput fashion
- Investigators work with the data of these patients using new i2b2 tools and a specialized team, both developed to work specifically with medical record data
- Using the Crimson system, tissues of these patients can be made available for genomic and biochemical analysis
- Automated discovery can be created from these projects to support further hypothesis-driven research

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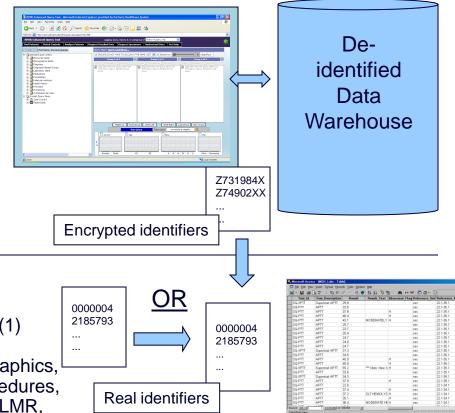
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Research Patient Data Registry exists at Partners Healthcare to find patient cohorts for clinical research

1) Queries for aggregate patient numbers

- Warehouse of in & outpatient clinical data
- 5.0 million Partners Healthcare patients
- 1.3 billion diagnoses, medications, 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)



Query construction in web tool

2) Returns identified patient data

- Start with list of specific patients, usually from (1)
- Authorized use by IRB Protocol
- Returns contact and PCP information, demographics, ^L 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.

Security and Patient Confidentiality of Step 1

- All patients at Partners are added
 - HIPAA notification that their data may be used for research upon registration.
- RPDR data is anonymized at the Query Tool.
 - Aggregated numbers are obfuscated to prevent identification of individuals; automatic lock out occurs if pattern suggests identification of an individual is being attempted.

A Security Architecture for Query Tools used to Access Large Biomedical Databases Shawn N. Murphy, MD, Ph.D. and Henry C. Chueh, MD, M.S. Laboratory of Computer Science, Massachusetts General Hospital, Boston, MA.

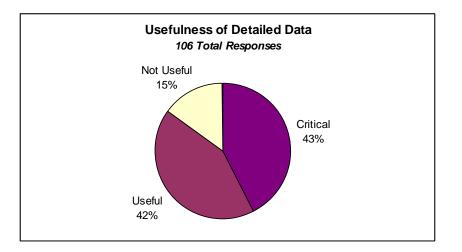
- Queries done in Query Tool available for review by RPDR team, a user lock out will specifically direct a review.
- De-identified data warehouse is a "Limited Data Set" by HIPAA
 - Medical record numbers are encrypted and obvious identifiers are removed from data.
- Concept of "established medical investigator" is promoted by classification as a faculty sponsor.

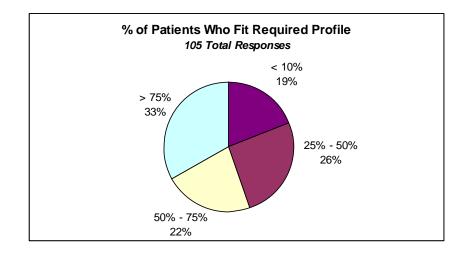
Security and Patient Confidentiality of Step 2

- Only studies approved by the Institutional Review Board (IRB) are allowed to receive identified data.
- Queries may be set up by workgroup member, but faculty sponsor on IRB protocol must directly approve all queries that return identified data.
- Special controls exist when distributing data regarding HIV antibody and antigen test results, substance abuse rehab programs, and genetic data, due to specific state and federal laws.
- Queries that return identified data are reviewed (retrospectively) by the IRB.

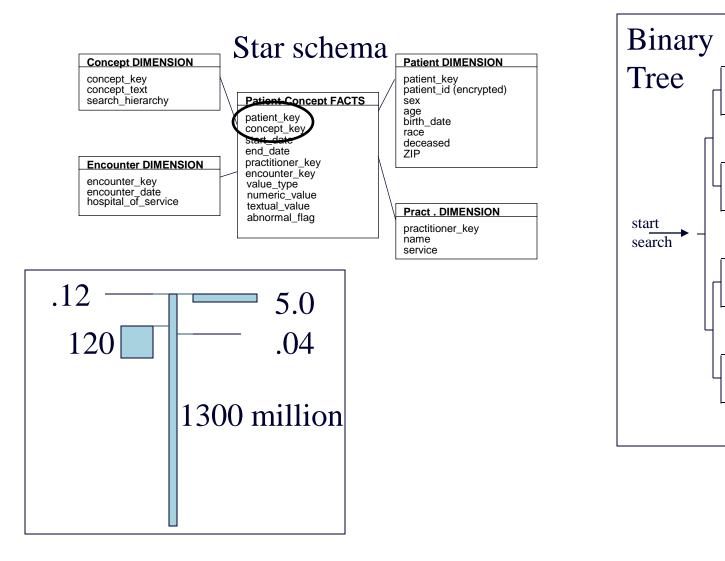
2009's usage of RPDR

- 2,227 registered users, 457 new in 2008
- 338 teams gathering data for research studies
- 1286 identified patient data sets returned to these teams, containing data of 7.8 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.
 - The adequacy of the match of a patient profile that could be obtained through the RPDR query tool was estimated.
- \$94-136 million total research support critically dependent on RPDR from patient data received throughout life of funding.
- ~300 data marts were created to support hospital operations, representing about 80 million patient records

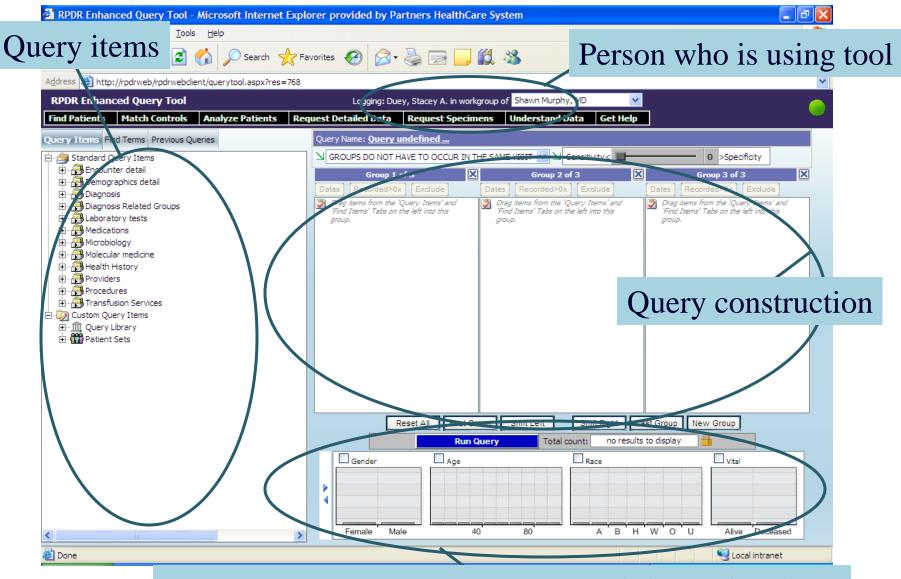




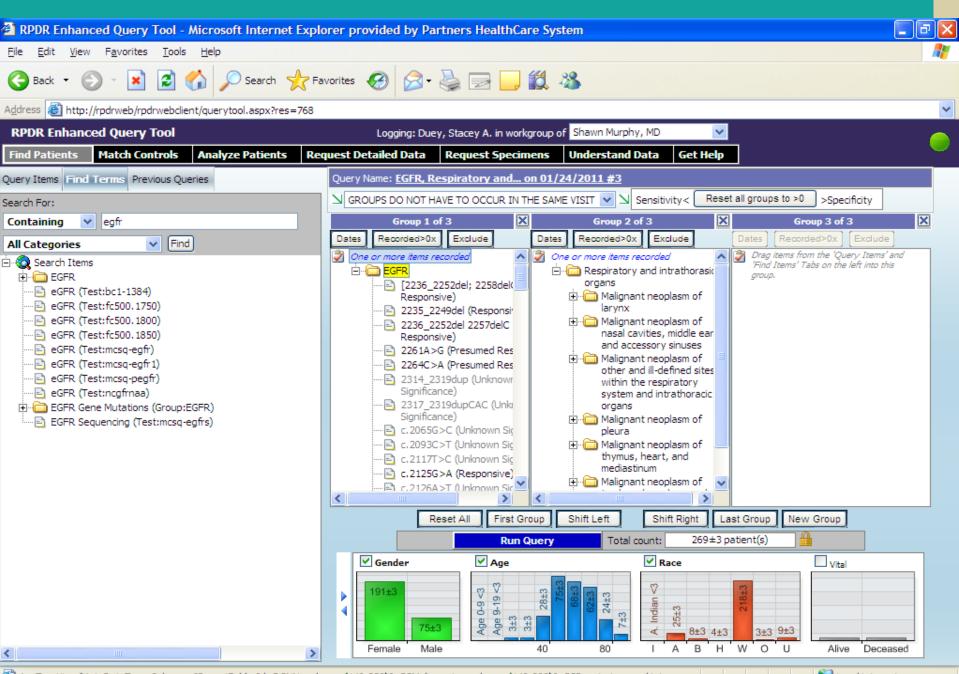
Organizing data in the Clinical Data Warehouse



FINDING PATIENTS



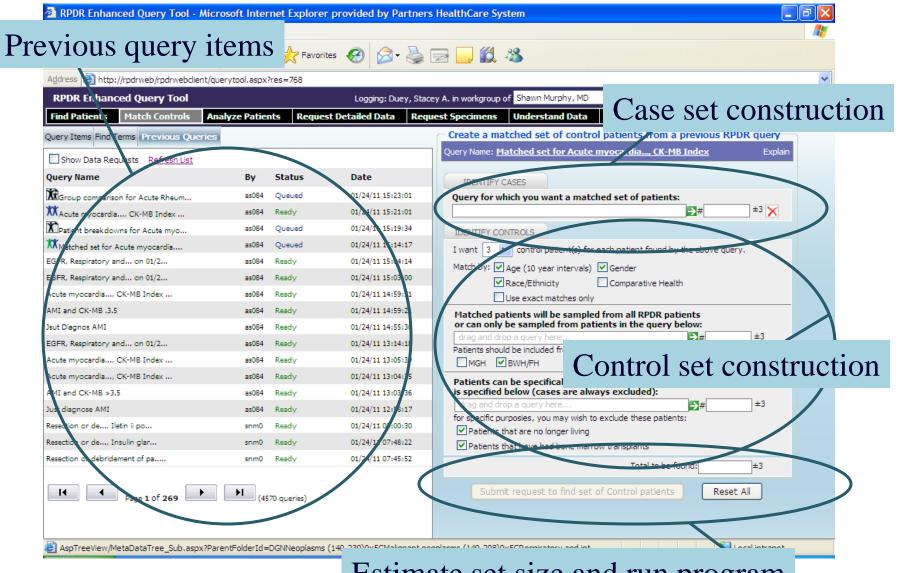
Results - broken down by number distinct of patients



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Secol intranet

MATCHING PATIENTS



Estimate set size and run program

RPDR Detailed Data Request Wizard -- Web Page Dialog

Using Partners IRB#2002P000381 (Research Patient Data Registry (RPDR)) to obtain data from the RPDR

×

You are logged in as Duey, Stacey A. in workgroup Shawn Murphy, MD

Please enter your IRB protocol.

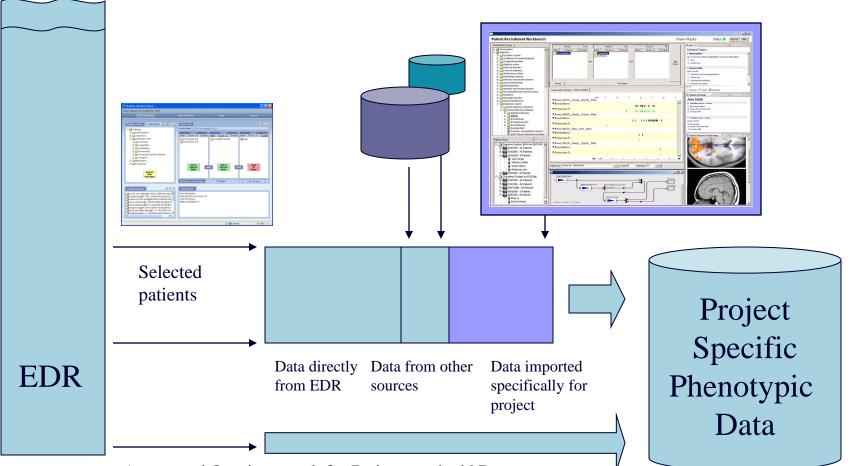
Partners IRB (required):	2002P000381
Paralelis IRD (required).	
	Title: Research Patient Data Registry (RPDR) Status: Active - Ongoing
	Status: Active - Origonig
Newton Wellesley Hospital IRB:	~
Spaulding Rehabilitation Hospital IRB:	▼
North Shore Medical Center IRB:	NSM 2008-786 demo 🗸
	Title:
	Status:
Options for returned set	of patients:
🗹 Exclude Partners Healt	ncare employees
Create a static set of p	atients from this query that can be used in other RPDR queries
Rerun the base query	shown above to obtain a fresh set of patients
Help < Back	Step 3 Next > Cancel

🚰 RPDR Detailed Data Request Wizard Web Page Dialog	
Using Partners IRB#2002P000381 (Research Patient Data Registry (RPDR)) to obtain data from the RPDR	
You are logged in as Duey, Stacey A. in workgroup Shawn Murphy, MD	
Select the types of data that should be returned from the RPDR Only data allowed by your protocol should be chosen (Identified data sets will always return a set of identified patient medical numbers)	
 Allergy Data from PEAR (Partners Enterprise Allergy Repository) Demographic Data Identifying Patient Information - not available for Limited Data Sets Identifying Patient Information - not available for Limited Data Sets Identifying Patient Clinical Record) Medications, Diagnoses and Procedures Patient Clinical Reports - not available for Limited Data Sets Cardiology Reports Discharge Summaries Endoscopy Reports Microbiology Data Operative Notes Pathology Reports Radiology Reports Transfusion Data, Blood Bank Data Top three providers for each patient 	
Help < Back Step 9 Next > Cancel	

High Throughput Methods for supporting Research at Partners Healthcare

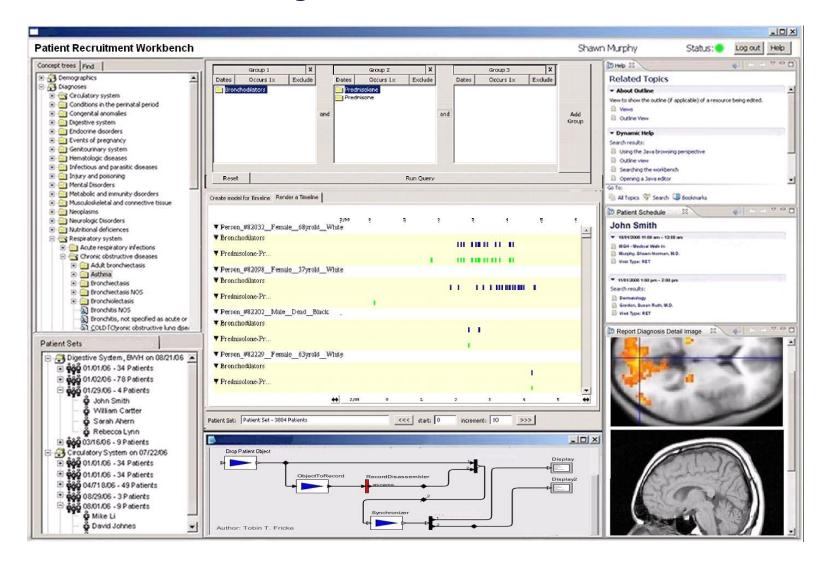
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Set of patients is selected through Enterprise Repository and data is gathered into a data mart

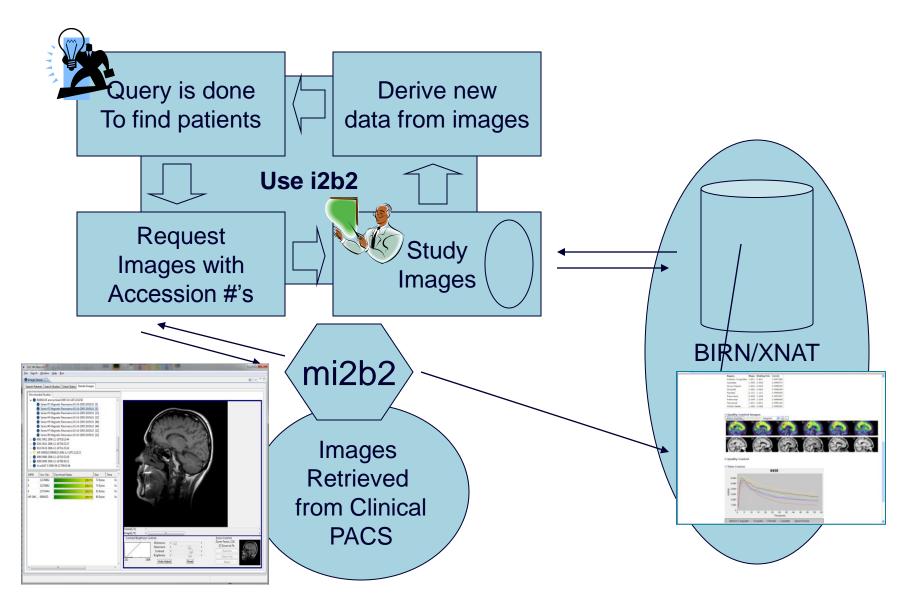


Automated Queries search for Patients and add Data

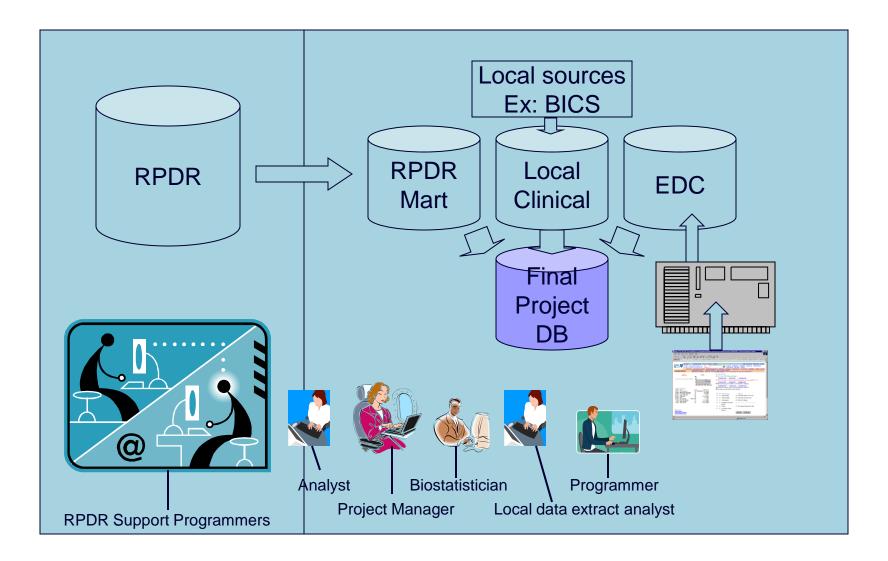
Data is available through the i2b2 Workbench



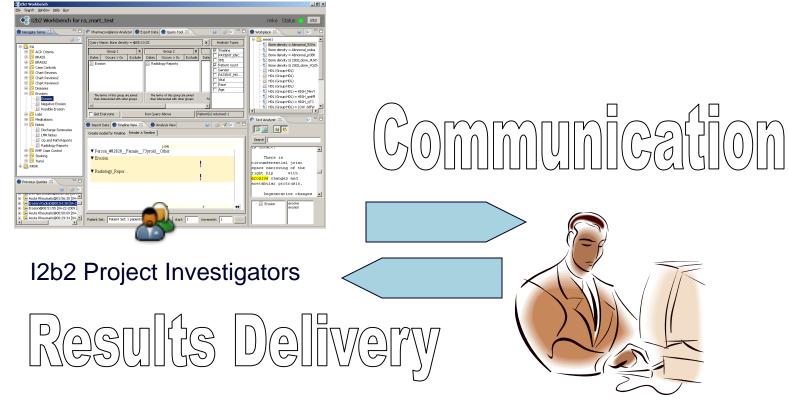
Research Investigator Workflow enabled by mi2b2



Team support for Projects

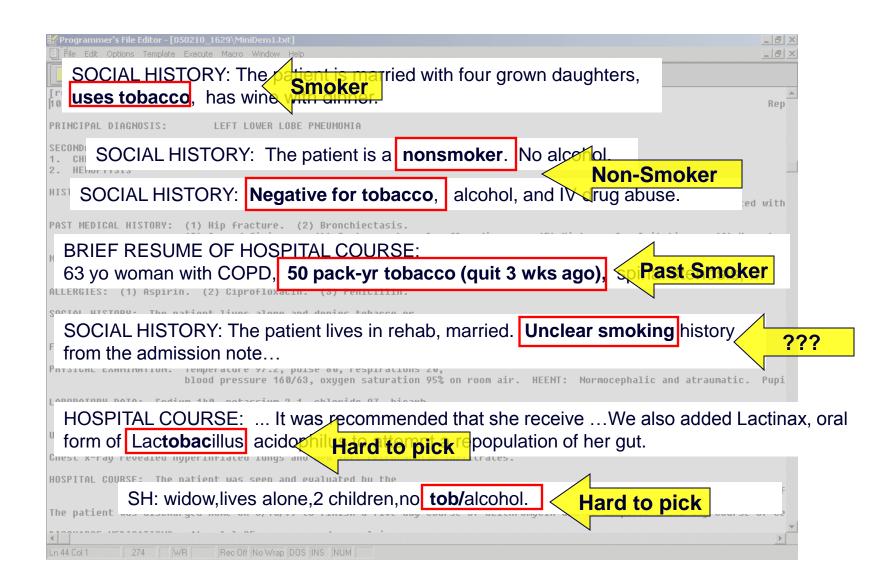


NLP Workflow



NLP Specialists

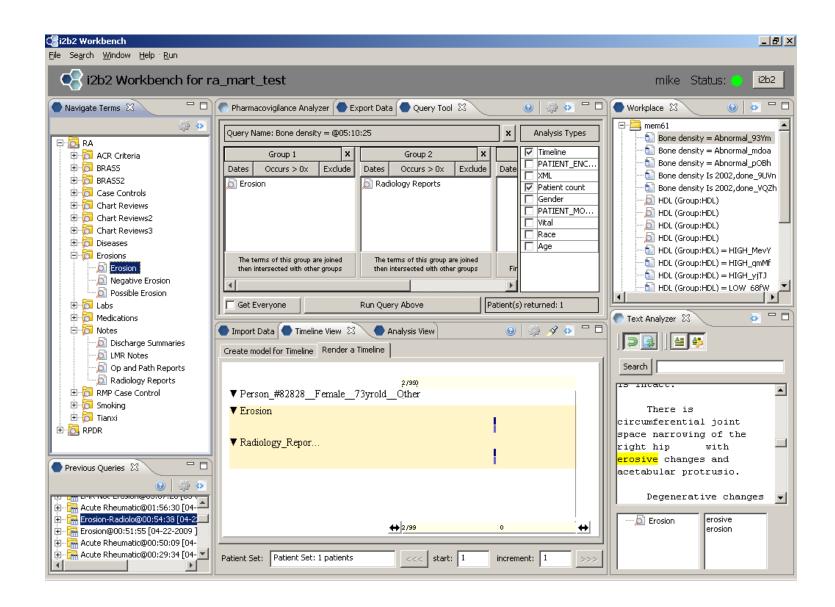
NLP (and comedy) is not pretty



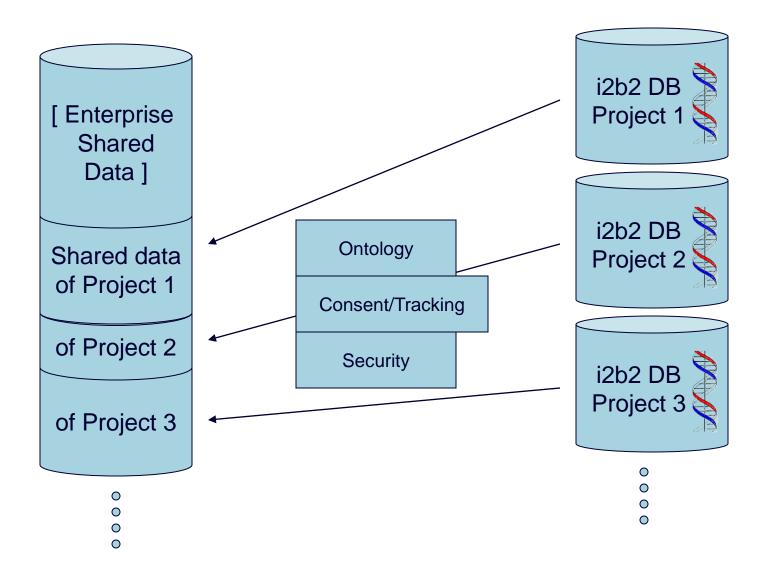
NLP Specialists Workstation

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Investigator Review



Project data can be added back to Enterprise Repository



Community

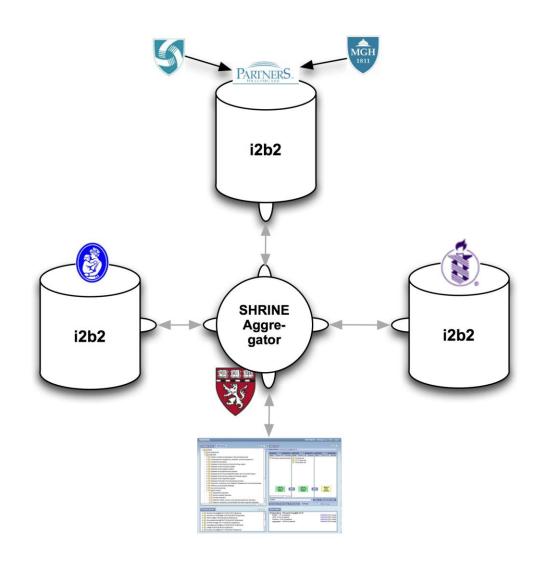
United States

- Arizona State University
- Beth Israel Deaconness Hospital, Boston, MA
- Boston University School of Medicine, Boston, MA
- Brigham and Women's Hospital, Boston, MA
- Case Western Reserve Hospital
- Children's Hospital, Boston, MA
- (Denver) Children's Hospital, Denver, CO
- Children's Hospital of Philadelphia, PA
- Childrens's National Medical Center (GWU)
- Cincinnati Children's Hospital, Cincinnati, OH
- Cleveland Clinic, Cleveland, OH
- (Weil Medical College of) Cornell, NYC, NY
- Duke Medical College
- Group Health Cooperative
- Harvard Pilgrim Healthcare
- Harvard Medical School, Boston, MA
- Health Sciences South Carolina
- Kaiser Permanente Health
- Kimmel Cancer Center (Thomas Jefferson University)
- Massachusetts General Hospital, Boston, MA
- Maine Medical Center, Portland, ME
- Marshfield Clinic, Wisconsin
- Morehouse School of Medicine, Atlanta, GA
- Ohio State University Medical Center, Columbus, OH
- Oregon Health & Science University, Portland, OR
- Renaissance Computing Institute, Chapel Hill, NC
- South Carolina Clinical and Translational Research Institute
- Tufts Medical Center, Boston, MA
- University of Alabama
- University of Arkansas Medical School
- University of California Davis, Davis, CA
- University of California San Francisco, SF, CA
- University of Chicago
- University of Massachusetts Medical School, Worcester, MA
- University of Michigan Medical Center, Ann Arbor, MI
- University of Pennsylvania School of Medicine, Philadelphia, PA
- University of Rochester Medical Center, Rochester, NY
- University of Texas Health Sciences Center at Houston, Houston, TX
- University of Texas Health Sciences Center at San Antonio, SA, TX
- University of Texas Health Sciences Center Southwestern, Dallas, TX
- Utah Health Science Center, Salt Lake City, UT
- University of Washington, Seattle, WA
- University of Wisconsin Madison
- Veterans Administration Boston and Utah

International

- Georges Pompidous Hospital, Paris, France
- Institute for Data Technology and Informatics (IDI), NTNU, Norway
- Karolinska Institute, Sweden
- University of Erlangen-Nuremberg, Germany
- University of Goettingen, Goettingen, Germany
- University of Leicester and Hospitals, England (Biomed. Res. Informatics Ctr. for Clin. Sci)
- University of Pavia, Pavia, Italy
- University of Seoul, Seoul, Korea

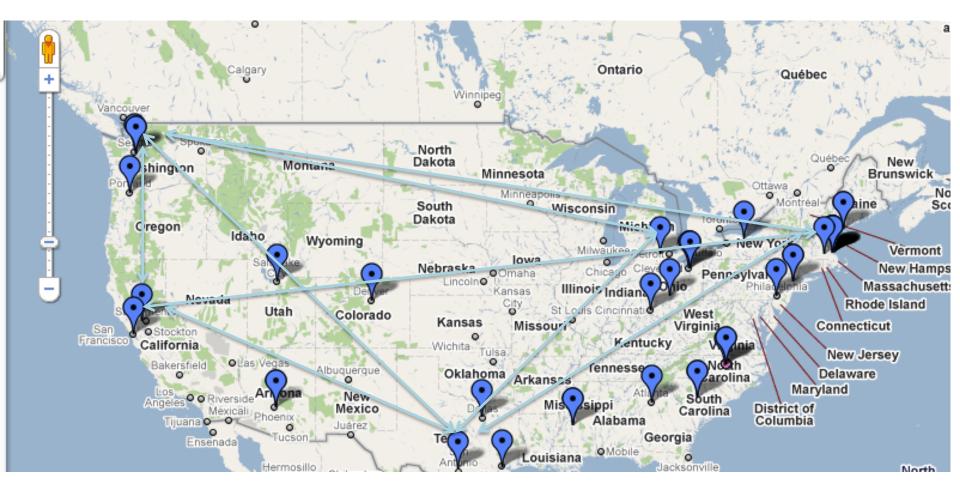
Aggregating across 4 hospitals, 3 i2b2 instances SHRINE (Shared Research Informatics Network) = Distributed Queries



Clinical data in SHRINE

- 10 years (2001-2011)
- 4 hospitals
- 6 million total patients
- >1 billion medical observations
 - Demographics
 - Diagnoses (ICD9-CM)
 - Medications (RxNorm)
 - Labs

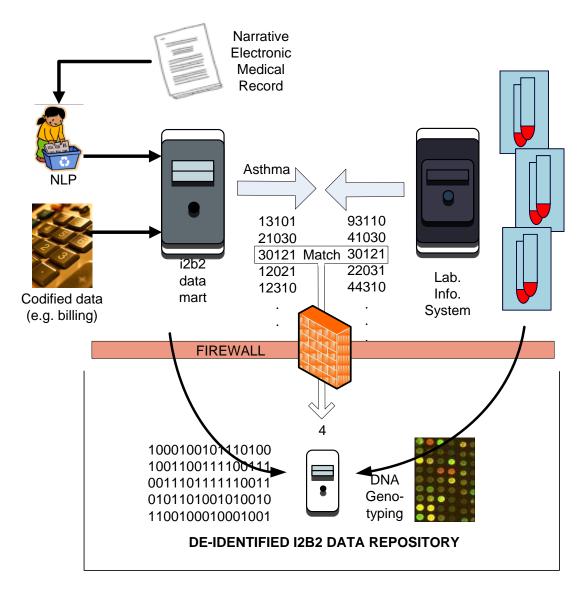
orm) (LOINC) SHRINE Find Patients | Message Log | Help | Logout **Navigate Terms** Find Terms 🚯 🗟 🗉 Query Tool 🐻 🗟 🗖 Query Name: Pervasi-0-9 yea@00:44:10 🗄 🔂 Demographics Group 1 Group 2 Group 3 🗄 🔂 Diagnoses Dates Occurs > 0x Exclude Occurs > 0x Occurs > 0x Dates Exclude Dates Exclude E Certain conditions originating in the perinatal period Pervasive developmental 0-9 vears old E Complications of pregnancy, childbirth, and the puerperium 10-17 years old E Congenital anomalies 🔂 18-34 years old Diseases of the blood and blood-forming organs Diseases of the circulatory system E Diseases of the digestive system 🗄 🔂 Diseases of the genitourinary system 🗄 🔂 Diseases of the musculoskeletal system and connective tissue Diseases of the nervous system and sense organs Diseases of the respiratory system Diseases of the skin and subcutaneous tissue Endocrine, nutritional, and metabolic diseases and immunity disorders Infectious and parasitic diseases one or one or drag a 🗄 🔂 Injury and poisoning more of AND more of AND term 🗄 👩 Mental Illness these to here these Adjustment disorders E Alcohol-related disorders 4 + Anxiety disorders Autism \$ Info Request New Topic E C Attention deficit, conduct, and disruptive behavior disorders 🗄 🔂 Delirium, dementia, and amnestic and other cognitive disorders Run Query New Query Print Query 2 Groups New Group 4 1 🕢 🕞 🗔 Query Status **Previous Queries** Finished Query: "Pervasi-0-9 yea@00:44:10" E Rervasi-0-9 yea@00:44:10 [9-27-2010] [kohane] BIDMC - 141 ±3 patients FINISHED [78.7 secs] E Rerva-0-9 y-PHENO@17:57:23 [9-26-2010] [kohane] CHB - 9103 ±3 patients FINISHED [78.7 secs] E R PDD-0-34@17:40:42 [9-26-2010] [kohane] Partners - 5134 ±3 patients FINISHED [78.7 secs] E Rerva-Male-Schiz@16:27:52 [9-26-2010] [kohane] Heine Al+PDD-0-34@17:47:17 [9-26-2010] [kohane] medicated-ppd-34@16:41:28 [9-26-2010] [kohane] . E Rervasi-0-9 yea@16:37:13 [9-26-2010] [kohane] ÷



High Throughput Methods for supporting Research at Partners Healthcare

- Set of patients is selected from medical record data in a high throughput fashion
- Investigators work with the data of these patients using new i2b2 tools and a specialized team, both developed to work specifically with medical record data
- Using the BETR/Crimson system, tissues of these patients can be made available for genomic and biochemical analysis
- Automated discovery can be created from these projects to support further hypothesis-driven research

Genotype samples and compare to controls

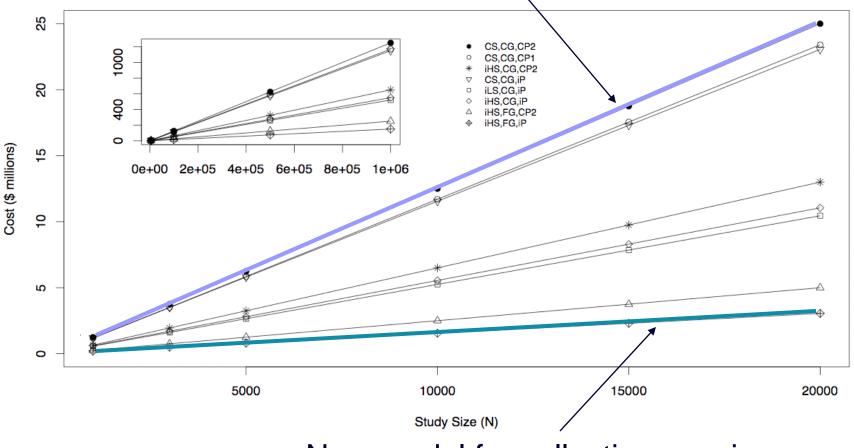


Cost and time benefit of Instrumenting with Sample Collection for Modest-size Study with 10,000 subjects (cases + controls)

Old vs. New	Cost (\$)	Time
1 chart review per patient (CP1)	\$20	15 minutes/subject
High-throughput phenotyping (iP) through RPDR and i2b2	\$50K Total	1 month total (conservative high estimate)
Sample acquisition through primary care provider (CP)	\$650	3-5 subjects/week ¹
High-throughput sample acquisition through RPDR and BETR/Crimson.	\$20	50-200 subjects /week ²

= \$6.7 million/study vs. \$250 thousand/study

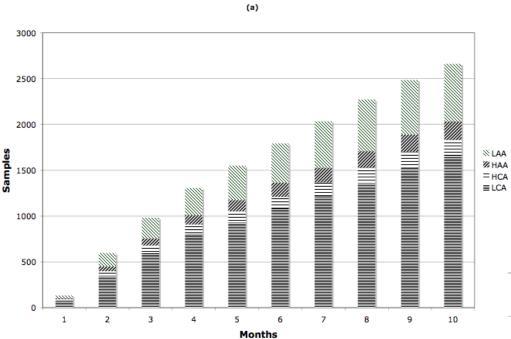
Escalating cost and time benefit of Instrumenting with Sample Collection

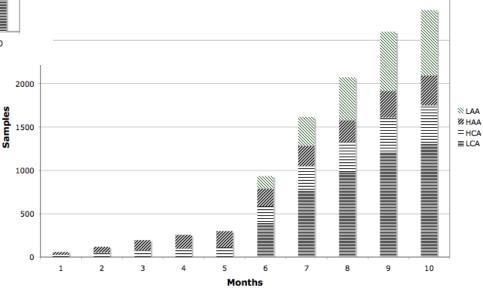


Previous model for collecting specimens

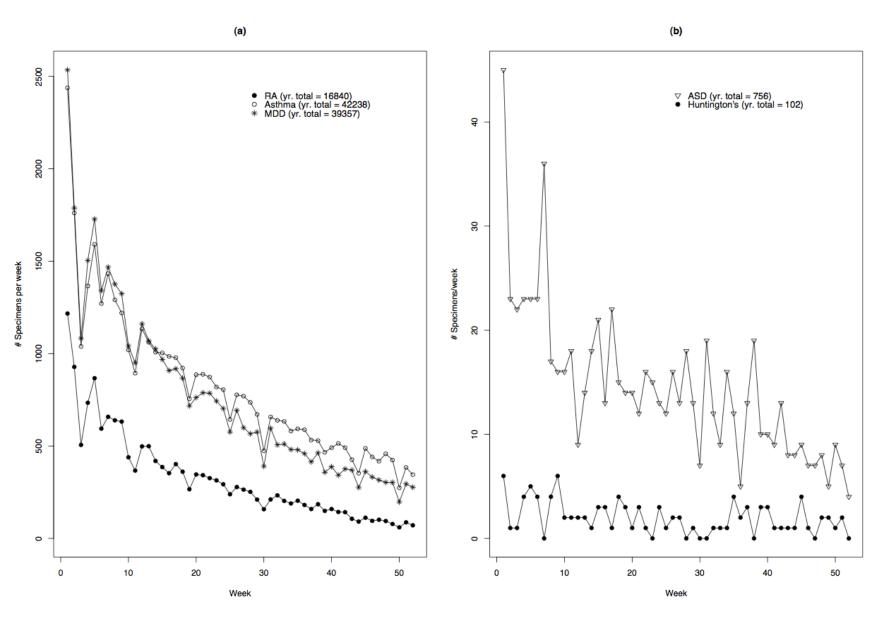
New model for collecting specimens

Meeting Expectations





Accrual Rates



High Throughput Methods for supporting Research at Partners Healthcare

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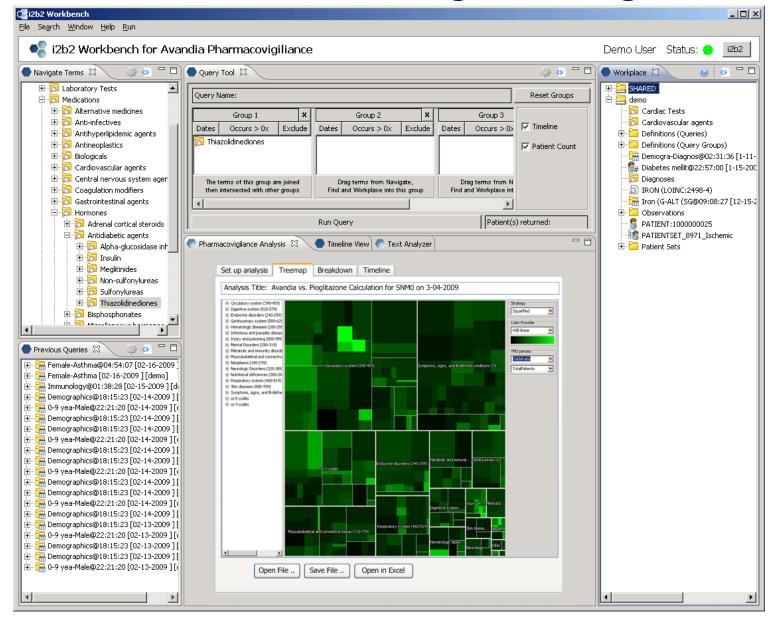
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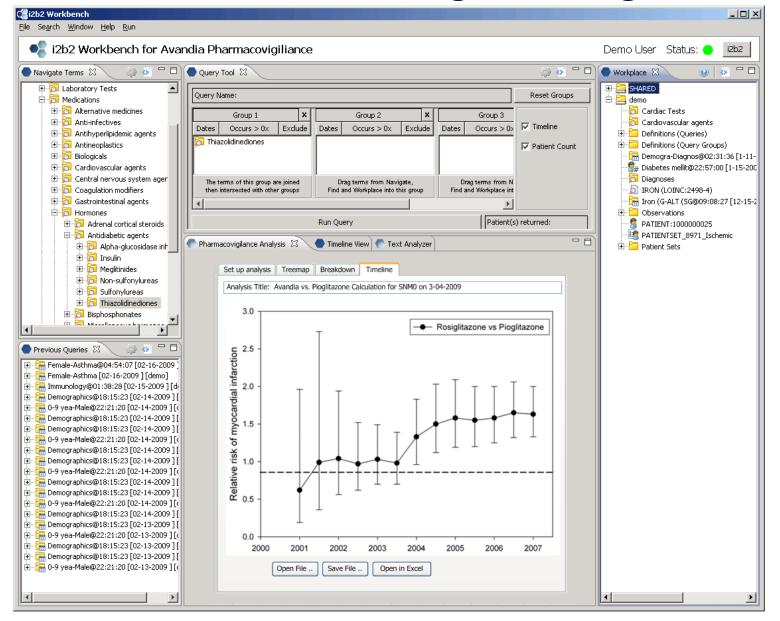
Performing Clinical trials "in-silico"

- Performing an observational, phase IV study is an expensive and complex process that can be potentially modeled in a retrospective database using groups of patients available with large amounts of well organized medical data.
- Fundamental problems complicate this approach:
 - Patients drift in and out of the healthcare system. Sophisticated statistical models using adequate control populations are necessary to compensate for the drift.
 - Confounding variables may not be found in the database. Natural language processing may be needed to extract the confounders from textual reports to allow confounders to be exposed.
 - Unknown missing data disrupts typical statistical approaches.
 - Biases in the data can easily mislead the investigator to false conclusions; data exploration and visualization tools are needed to expose these kinds of potential problems.

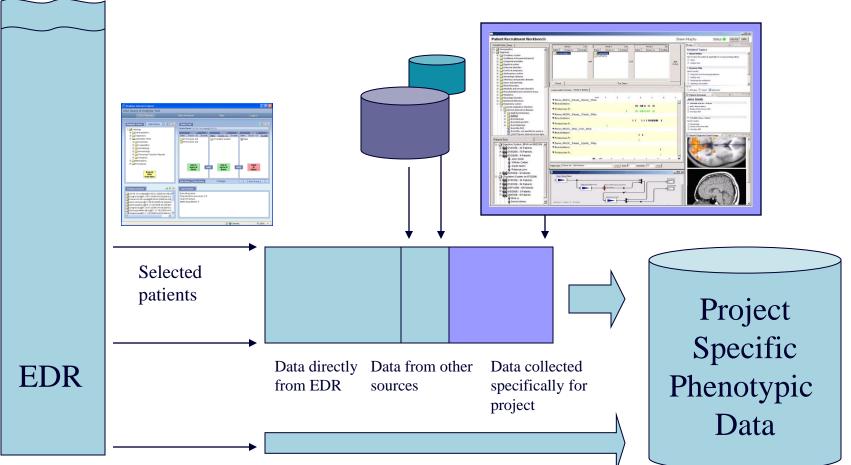
Dashboard used to observe high-level signals



Dashboard used to observe high-level signals



Set of patients is selected through Enterprise Repository and data is gathered into a data mart



Daily Automated Queries search for Patients and add Data

Builds complex "Custom Study" displays

<u>File Search Window H</u> elp <u>R</u> un			
🔹 i2b2 Workbench for Avandia I	Pharmacovigiliance	Sh	awn Murphy, MD, PhD 🛛 Status: 🔶 🗾 i2b2
🔵 Navigate Terms 🛛 🧼 💿 🗖 🗖	Query Tool X	÷ 🖓 🖗 🗖	
Avandia_NLP	Query Name:	Reset Groups	⊕ SHARED
tire avandia			Avandia-Cardiov@09:20:31 [11-24-2008] [snml
🗈 📟 Cardiovascular Event	Group 1 Group 2 Group 3	Timeline	
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No Rosi And No CV		IV Patient Count	
	Add		
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Health History			
E laboratory Tests	Run Query Pa	atient(s) returned:	
🖻 🖷 Medications		🌼 🔗 💿 🗖 🗖	
Alternative medicines	Timeline View 🛛 🕒 Export Data		-
Anti-infectives	Create model for Timeline Render a Timeline		
Antihyperlipidemic agents Antineoplastics		1	
Andreeplastics Biologicals			
Cardiovascular agents	▼ Person_#240407_Female_Dead_White		
E- Central nervous system agents	▼ Enrolled	-	
庄 📲 Coagulation modifiers			
CPT Devices	▼ Rosiglitazone_e		
CPT Medications	▼ Cardiovascular		
Devices Gastrointestinal agents	♥ Carubvascular		
Hormones	▼ No Rosi And CVN		
Adrenal cortical steroids			
Antidiabetic agents	Vandiarosiglit		
😟 💷 Alpha-glucosidase inhibitors	▼ Person #240419 Female_53yrold_Black		
Byetta (exenatide) - LMR 5176			
😟 🖷 Insulin	▼ Enrolled		
I Meglitinides I Mon-sulfonylureas	▼ Rosiglitazone e		
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😟 💷 Metformin-rosiglitazone			
🗄 🖷 Pioglitazone	▼ No_Rosi_And_CVN		
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Previous Queries 22 33 40 "	Avanuar osign	1 m m	
	▼ Person_#240606_Female_73yrold_White		
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Circulatory sys@11:05:19 [07-02-2008] [snm0] Circulatory sys@11:05:19 [07-02-2008] [snm0]	Patient Set: Patient Set: 954 patients <<< start: 1	increment: 10 >>>	
Circulatory sys@11:05:19 [07-02-2008] [snm0] Circulatory sys@11:00:50 [07-02-2008] [snm0] Circulatory sys@11:00:31 [07-02-2008] [snm0] Circulatory sys@11:00:31 [07-02-2008] [snm0]			
Medications@11:00:31 [07-02-2008] [som0]			

Builds complex "Custom Study" displays

Timeline View 🙁 🔵 Export Data			i 🖉 🖉 👘 🗖
eate model for Timeline Render a Timeline			-01
Person_#241696_Female_80yrold_White			
Enrolled			
Rosiglitazone_e			
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No_Rosi_And_CVN			
Avandiarosiglit			
Person_#241699Male69yroldWhite			
/ Enrolled			
Rosiglitazone_e			
Cardiovascular			
No_Rosi_And_CVN			
Person_#241700Female59yroldBlack			
'Enrolled			
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Cardiovascular			
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Person_#241813Male53yroldHispanic			
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No_Rosi_And_CVN			
Avandiarosiglit	and the second se	and a management of the second	1 A A
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ient Set: Patient Set: 954 patients		<<< start: 11	increment: 10 >>

Seven important factors enabled by i2b2 platform

- 1) Enables enterprise-wide repurposing of health care data for research
- 2) Enables extensible software architecture for developers
- 3) Extends EHR research so that data may be shared among sites
- 4) Enables natural language processing
- 5) Provides method for materializing scientific method for EHRbased investigations
- 6) Extends EHR research so that data may be shared among sites and samples may be obtained
- 7) Provides platform for Clinical Trials "in silico"

Collaborators

RPDR

- Eugene Braunwald
- John Glaser
- Diane Keogh
- Henry Chueh

i2b2

- Isaac Kohane
- Susanne Churchill
- Griffin Weber
- Michael Mendis
- Vivian Gainer
- Lori Phillips
- Rajesh Kuttan
- Wensong Pan
- Janice Donahue
- William Simons (SHRINE)
- Andy McMurry (SHRINE)
- Doug McFadden (SHRINE)

- Medical Imaging (mi2b2)
 - Christopher Herrick
 - David Wang
 - Bill Wang
- Sample Acquisition
 - Lynn Bry
 - Natalie Boutin
- i2b2 Driving Biology Projects
 - Vivian Gainer
 - Victor Castro
 - Raul Guzman
 - Robert Plenge
 - Scott Weiss
 - Stan Shaw
 - John Brownstein
 - Qing Zeng
 - Guergana Savova