

# Supporting Researchers with Electronic Patient Data

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#### **Weill Cornell Medicine**

## Overview

- Problem
- Approach
  - Research Computing at Weill Cornell Medicine
  - Architecture for Research Computing in Health (ARCH)
- Discussion

## Problem

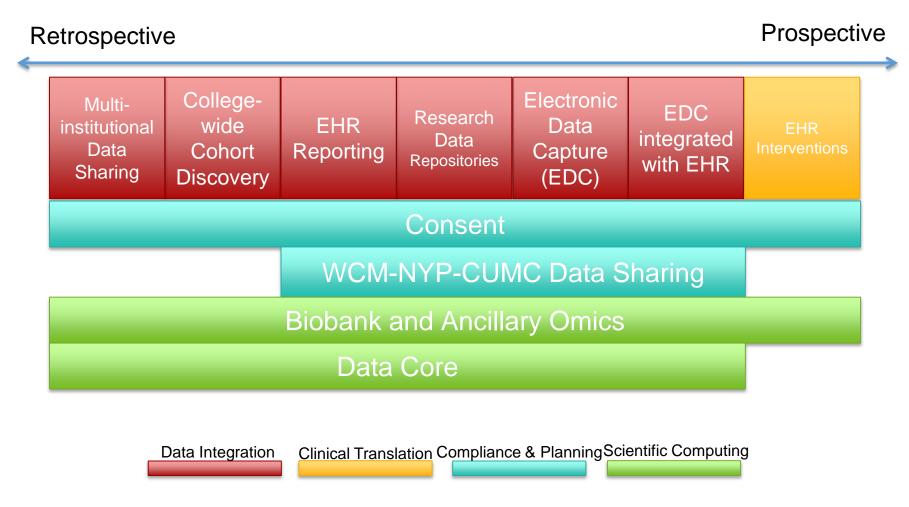
- Obtaining electronic health record (EHR) data for clinical and translational research is difficult
  - Repurpose transactions for research
  - Use one or more tools
  - Understand strengths and limitations
  - Obtain approval
- Optimal approaches are unknown
- Research Informatics can help investigators
  - Obtain EHR data
  - Collect novel measures
  - Integrate data

# Approach: Research Computing at WCM

Administration Conduct **Scientific** Research Research **Informatics** Computing **Administrative** Computing

Information Technology Infrastructure





Retrospective

College-Electronic Multi-EDC Research wide **EHR** Data institutional integrated Data Data Cohort Reporting Capture Repositories with EHR Sharing (EDC) Discovery

Data Integration

Data

Sharing

#### Architecture for Research Computing in Health (ARCH)

Retrospective

Multiinstitutional Wide EHR Research Data EDC

Data

Repositories

integrated

with EHR

Capture

(EDC)

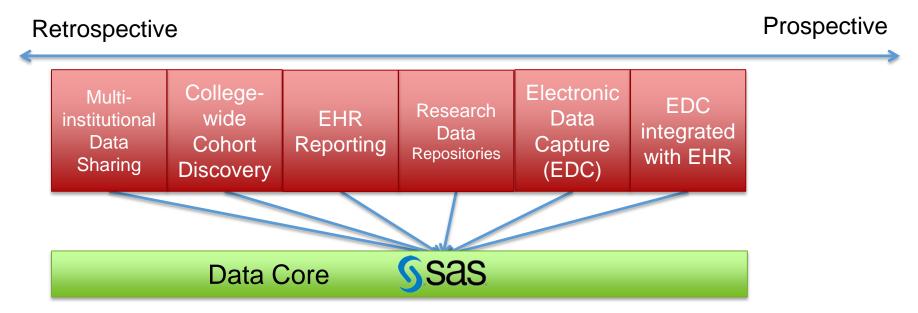


Data Integration

Cohort

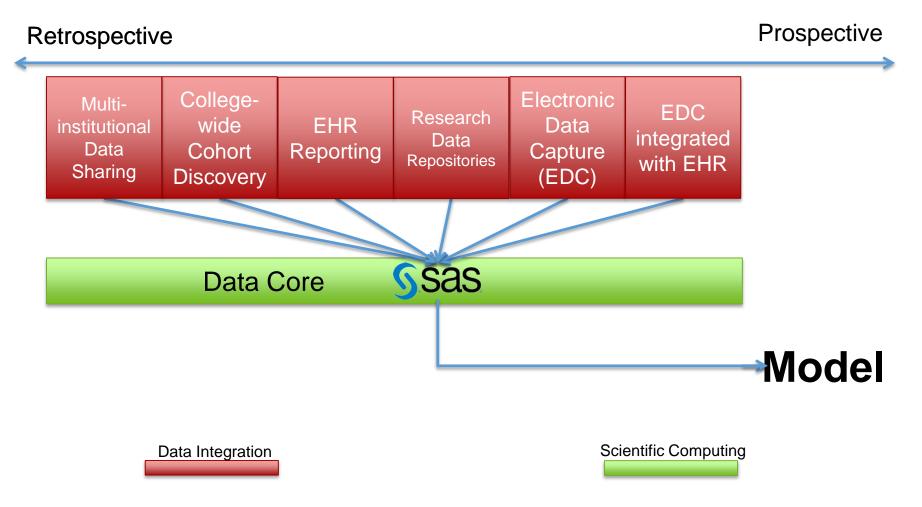
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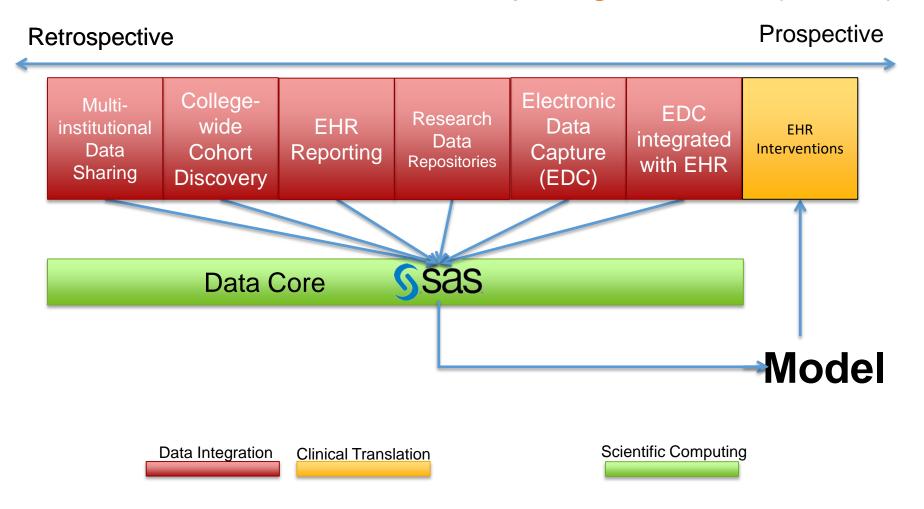
Reporting

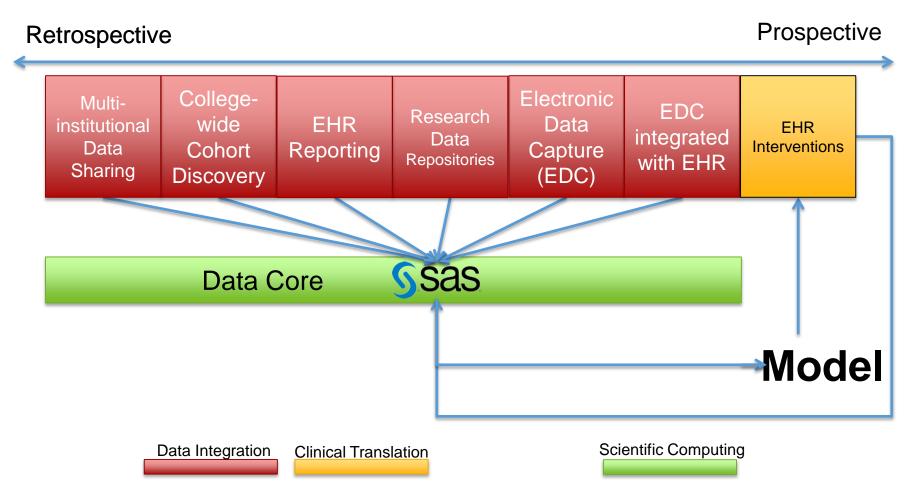


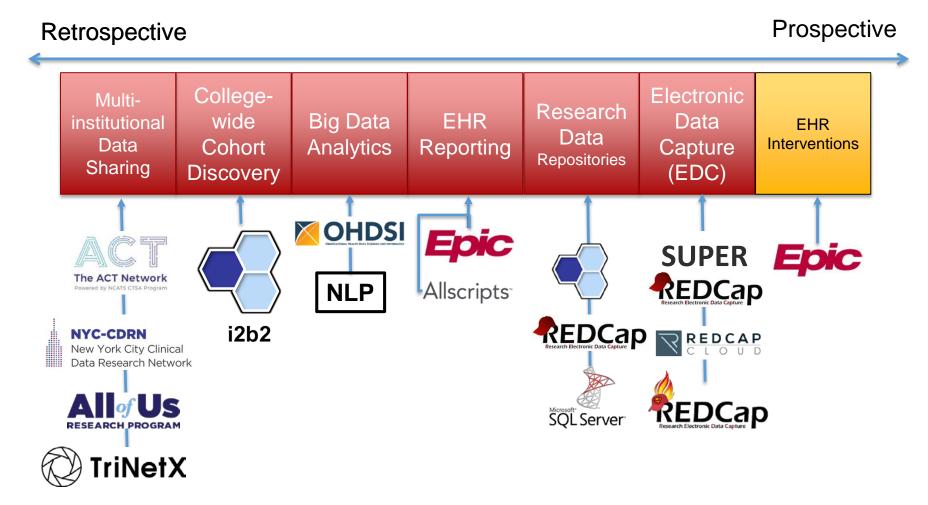
Data Integration

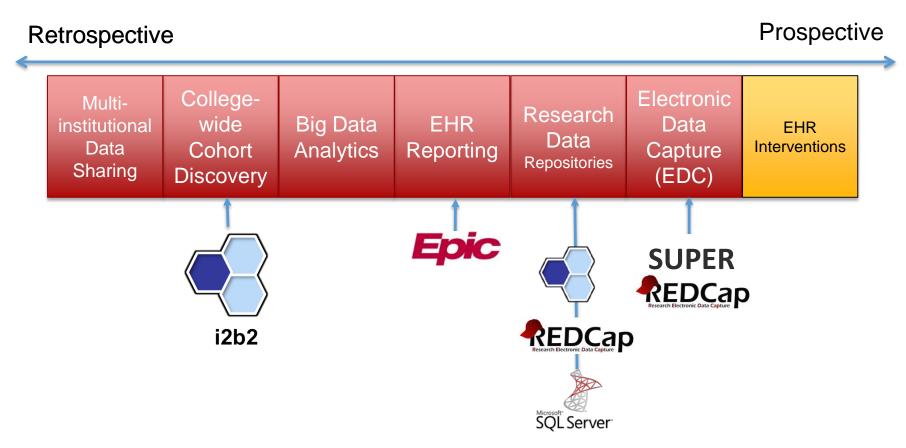
Scientific Computing











Retrospective Prospective





Step 1: obtain raw EHR data describing diagnoses assigned in patients encounters

	patient_n	enc_num	line	dx_id	primary_dx	comme	dx_chronic	dx_unique	dx_ed_yn
828	30147	1644452	3	935415	N	NULL	N	4	N
829	69154	1644417	1	12866	Υ	NULL	N	1	N
830	42675	1644384	1	56131	N	NULL	N	2	N
831	42675	1644384	2	135365	N	NULL	N	3	N
832	63397	1644381	1	813625	N	NULL	N	1	N
833	63397	1644381	2	807290	Υ	NULL	N	2	N
834	42839	1644380	1	15598	Υ	NULL	N	1	N
835	105950	1644350	1	8059885	Υ	NULL	N	1	N
836	44243	1644293	1	24458	Υ	NULL	N	1	N
837	13016	1644287	1	8028848	Υ	NULL	N	1	N
838	32482	1644244	1	1005381	Υ	NULL	N	1	N
839	79305	1644243	1	8078416	N	NULL	N	2	N
840	79305	1644243	2	6111	N	NULL	N	3	N
841	79305	1644243	3	8005450	N	NULL	N	4	N
842	79305	1644243	4	9834	N	NULL	N	5	N
843	79305	1644243	5	8059219	N	NULL	N	6	N
844	79305	1644243	6	8082109	N	NULL	N	7	N
845	79305	1644243	7	8091886	N	NULL	N	8	N

#### Step 2: match raw EHR data describing diagnoses to ICD-9 codes

	patient_num	enc_num	line	dx_id	enc_icd_code	primary_dx	comme	dx_chronic_yn	dx_uniq	dx_ed_yn
828	30147	1644452	3	935415	250.60	N	NULL	N	4	N
829	69154	1644417	1	12866	996.82	Υ	NULL	N	1	N
830	42675	1644384	1	56131	389.18	N	NULL	N	2	N
831	42675	1644384	2	135365	388.30	N	NULL	N	3	N
832	63397	1644381	1	813625	555.1	N	NULL	N	1	N
833	63397	1644381	2	807290	NULL	Υ	NULL	N	2	N
834	42839	1644380	1	15598	239.6	Υ	NULL	N	1	N
835	105950	1644350	1	8059885	782.1	Υ	NULL	N	1	N
836	44243	1644293	1	24458	281.1	Υ	NULL	N	1	N
837	13016	1644287	1	8028848	415.19	Υ	NULL	N	1	N
838	32482	1644244	1	1005381	715.96	Υ	NULL	N	1	N
839	79305	1644243	1	8078416	185	N	NULL	N	2	N
840	79305	1644243	2	6111	581.9	N	NULL	N	3	N
841	79305	1644243	3	8005450	198.5	N	NULL	N	4	N
842	79305	1644243	4	9834	787.91	N	NULL	N	5	N
843	79305	1644243	5	8059219	780.79	N	NULL	N	6	N
844	79305	1644243	6	8082109	279.00	N	NULL	N	7	N
845	79305	1644243	7	8091886	593.9	N	NULL	N	8	N

Step 3: transform raw EHR data describing ICD-9 codes into dichotomous scientific variable

	1_									
	Race	Height	Weight	BMI	Diabetes	Hypertension	Obesity	NSAID	Antidiabetic	Current_Cigarette_Smoker
1	WHITE	5' 4"	2096	22.84	N	N	N	N	N	N
2	OTHER COMBINATIONS NOT DESCRIBED	6' 0"	4704	39.47	Υ	Υ	Υ	N	Υ	N
3	WHITE	5' 5"	3168	34.72	N	N	Υ	N	N	Υ
4	WHITE	5' 4.5"	2352	24.85	N	N	N	N	N	N
5	ASIAN	5' 9"	2656	25.60	N	Υ	N	N	N	N
6	DECLINED	6' 0"	3648	30.92	N	N	N	N	N	N
7	DECLINED	5' 6"	1827.2	18.44	N	N	N	N	N	N
8	OTHER COMBINATIONS NOT DESCRIBED	4' 9.5"	1872	25.11	N	N	N	N	N	N
9	OTHER COMBINATIONS NOT DESCRIBED	5' 2"	2448	28.92	N	N	N	N	N	N
10	ASIAN	5' 9"	3072	28.34	N	N	N	N	N	N
11	NULL	5' 0"	3609.6	44.06	N	N	N	N	N	N
12	DECLINED	5' 1.024"	2704	31.95	Υ	N	Υ	Υ	Y	N
13	WHITE	6' 1"	3056	26.08	N	Υ	N	Υ	N	N
14	NULL	6' 0"	3408	3827.82	Υ	Υ	N	N	Y	Υ
15	NULL	6' 2"	3284.8	25.99	N	N	N	N	N	N
16	DECLINED	5' 8"	2560	26.63	N	N	N	N	N	N
17	DECLINED	5' 8"	2640	25.09	N	N	N	N	N	Υ
18	NULL	5' 1"	2769.6	32.72	N	N	Υ	N	N	N

Step 4: repeat process to define additional diagnosis dichotomous variables

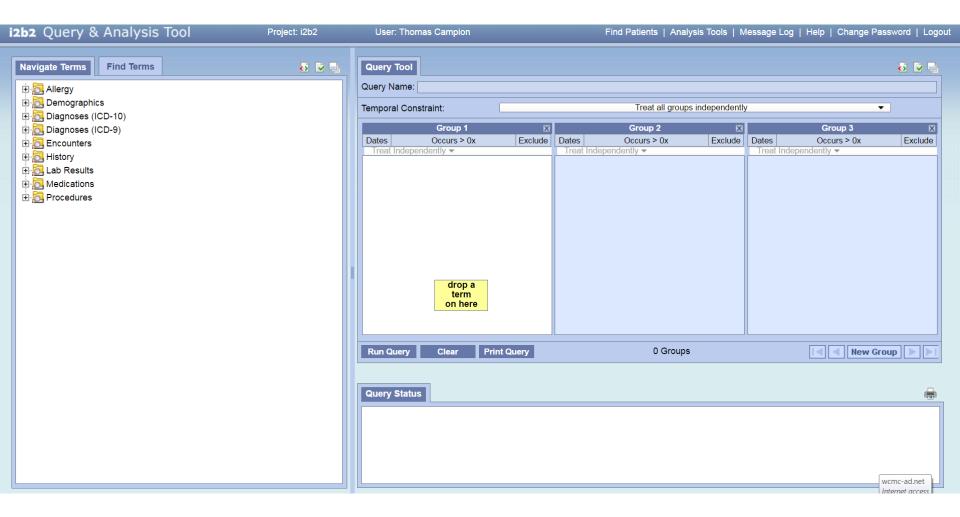
	Race	Height	Weight	BMI	Diabetes	Hypertension	Obesity	NSAID	Antidiabetic	Current Cigarette Smoker
1	WHITE	5' 4"	2096	22.84	N	N	N	N	N	N
^		6' 0"			Y	Y	Y	N	Y	N
2	OTHER COMBINATIONS NOT DESCRIBED		4704	39.47						
3	WHITE	5' 5"	3168	34.72	N	N	Y	N	N	Υ
4	WHITE	5' 4.5"	2352	24.85	N	N	N	N	N	N
5	ASIAN	5' 9"	2656	25.60	N	Υ	N	N	N	N
6	DECLINED	6' 0"	3648	30.92	N	N	N	N	N	N
7	DECLINED	5' 6"	1827.2	18.44	N	N	N	N	N	N
8	OTHER COMBINATIONS NOT DESCRIBED	4' 9.5"	1872	25.11	N	N	N	N	N	N
9	OTHER COMBINATIONS NOT DESCRIBED	5' 2"	2448	28.92	N	N	N	N	N	N
10	ASIAN	5' 9"	3072	28.34	N	N	N	N	N	N
11	NULL	5' 0"	3609.6	44.06	N	N	N	N	N	N
12	DECLINED	5' 1.024"	2704	31.95	Υ	N	Υ	Υ	Y	N
13	WHITE	6' 1"	3056	26.08	N	Υ	N	Υ	N	N
14	NULL	6' 0"	3408	3827.82	Υ	Υ	N	N	Y	Υ
15	NULL	6' 2"	3284.8	25.99	N	N	N	N	N	N
16	DECLINED	5' 8"	2560	26.63	N	N	N	N	N	N
17	DECLINED	5' 8"	2640	25.09	N	N	N	N	N	Υ
18	NULL	5' 1"	2769.6	32.72	N	N	Υ	N	N	N

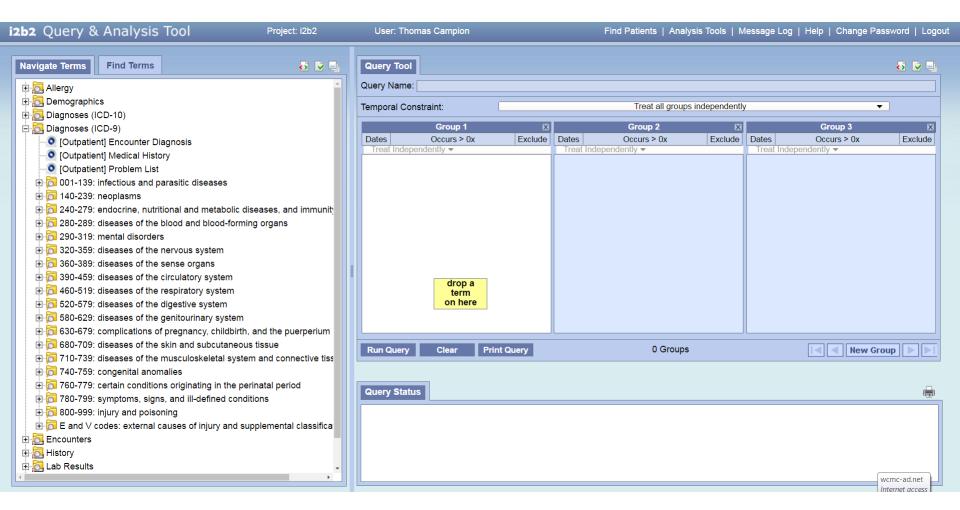
Step 4: repeat process to transform raw EHR data for medications and other domains (raw data for medications and other domains not pictured)

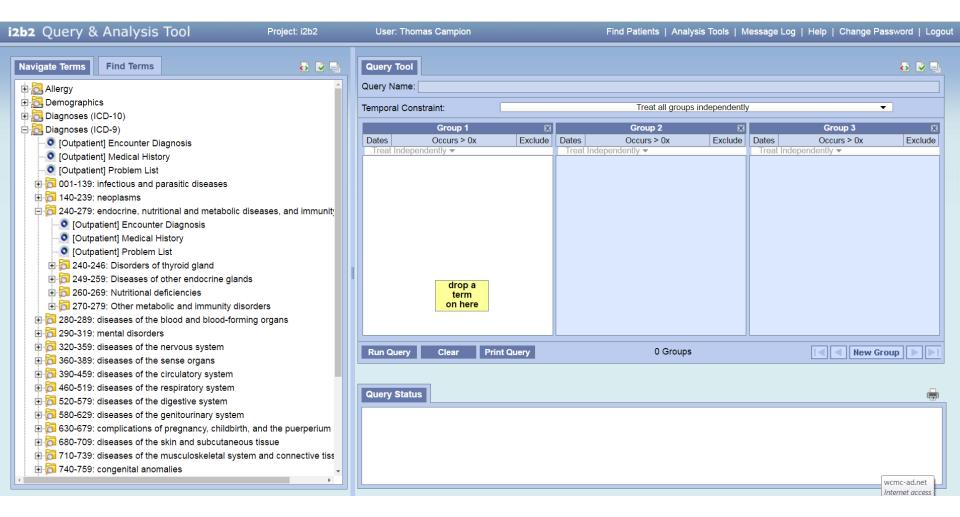
	Race	Height	Weight	BMI	Diabetes	Hypertension	Obesity	NSAID	Antidiabetic	Current_Cigarett
1	WHITE	5' 4"	2096	22.84	N	N	N	N	N	N
2	OTHER COMBINATIONS NOT DESCRIBED	6' 0"	4704	39.47	Υ	Υ	Υ	N	Y	N
3	WHITE	5' 5"	3168	34.72	N	N	Υ	N	N	Υ
4	WHITE	5' 4.5"	2352	24.85	N	N	N	N	N	N
5	ASIAN	5' 9"	2656	25.60	N	Υ	N	N	N	N
6	DECLINED	6' 0"	3648	30.92	N	N	N	N	N	N
7	DECLINED	5' 6"	1827.2	18.44	N	N	N	N	N	N
8	OTHER COMBINATIONS NOT DESCRIBED	4' 9.5"	1872	25.11	N	N	N	N	N	N
9	OTHER COMBINATIONS NOT DESCRIBED	5' 2"	2448	28.92	N	N	N	N	N	N
10	ASIAN	5' 9"	3072	28.34	N	N	N	N	N	N
11	NULL	5' 0"	3609.6	44.06	N	N	N	N	N	N
12	DECLINED	5' 1.024"	2704	31.95	Υ	N	Υ	Υ	Y	N
13	WHITE	6' 1"	3056	26.08	N	Υ	N	Υ	N	N
14	NULL	6' 0"	3408	3827.82	Υ	Υ	N	N	Y	Υ
15	NULL	6' 2"	3284.8	25.99	N	N	N	N	N	N
16	DECLINED	5' 8"	2560	26.63	N	N	N	N	N	N
17	DECLINED	5' 8"	2640	25.09	N	N	N	N	N	Υ
18	NULL	5' 1"	2769.6	32.72	N	N	Υ	N	N	N

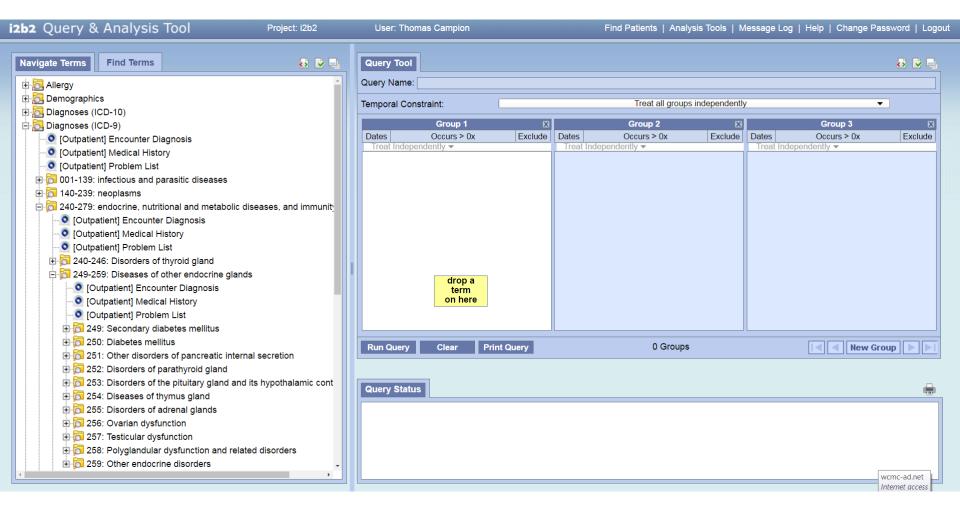
Prospective Retrospective College-Electronic Multi-Research wide Big Data **EHR** Data institutional **EHR** Data Data Interventions Cohort Analytics Reporting Capture Repositories Sharing (EDC) Discovery

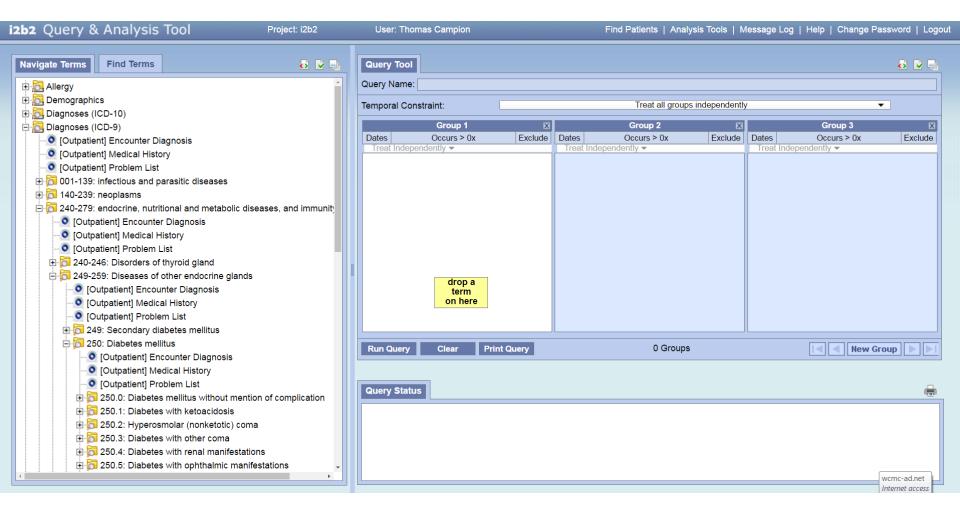


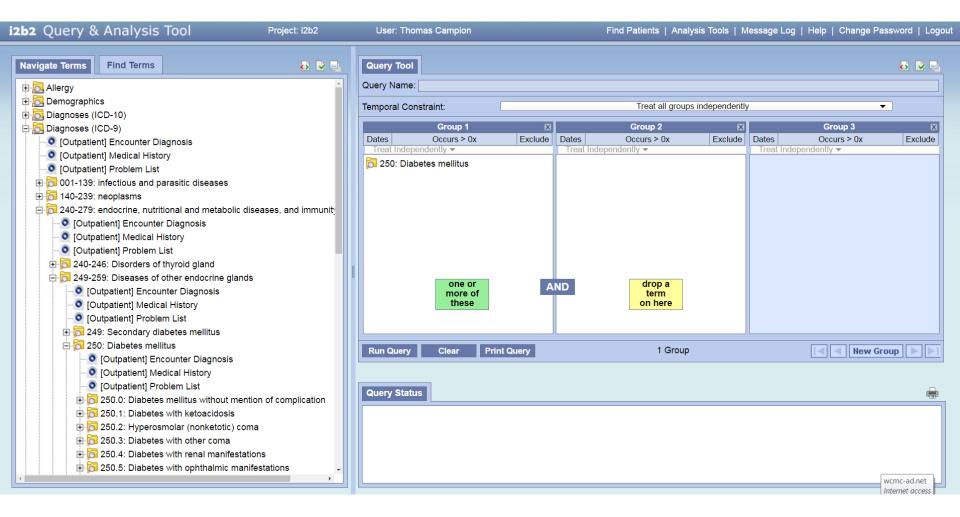


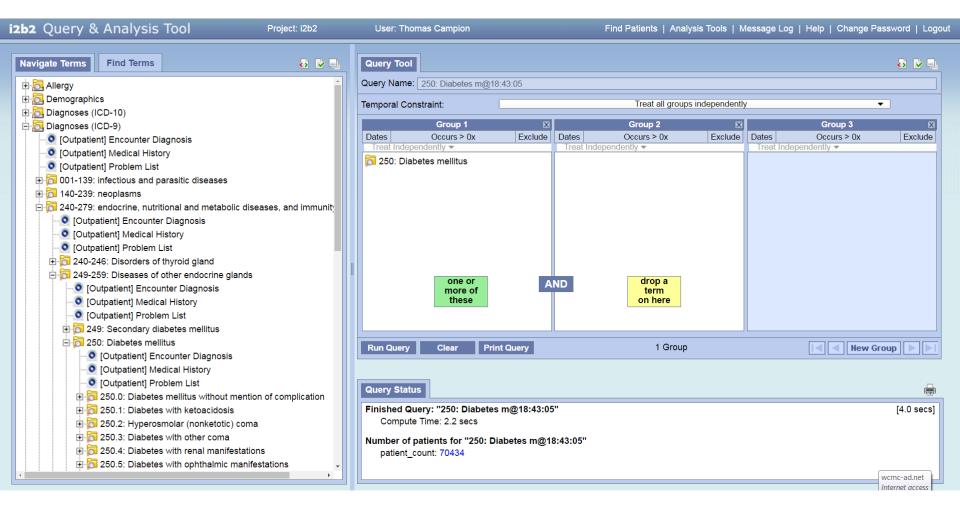


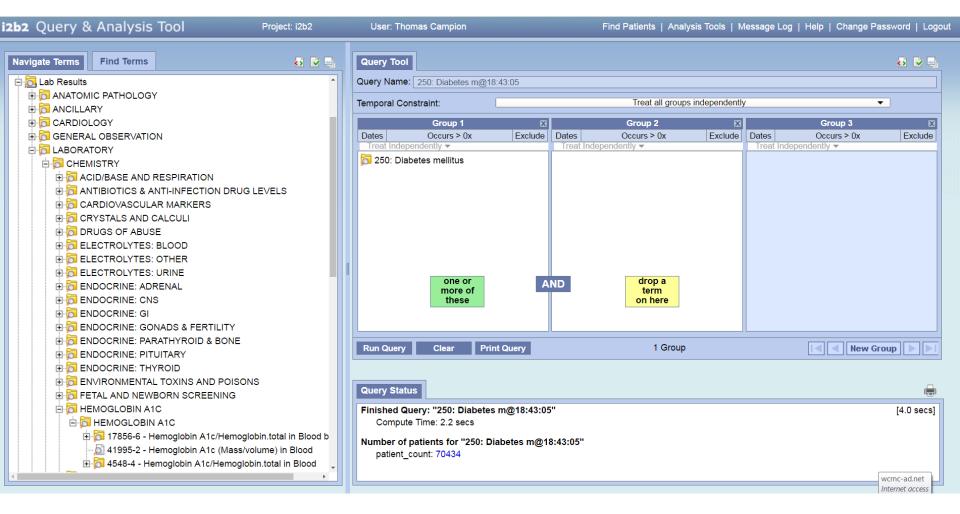


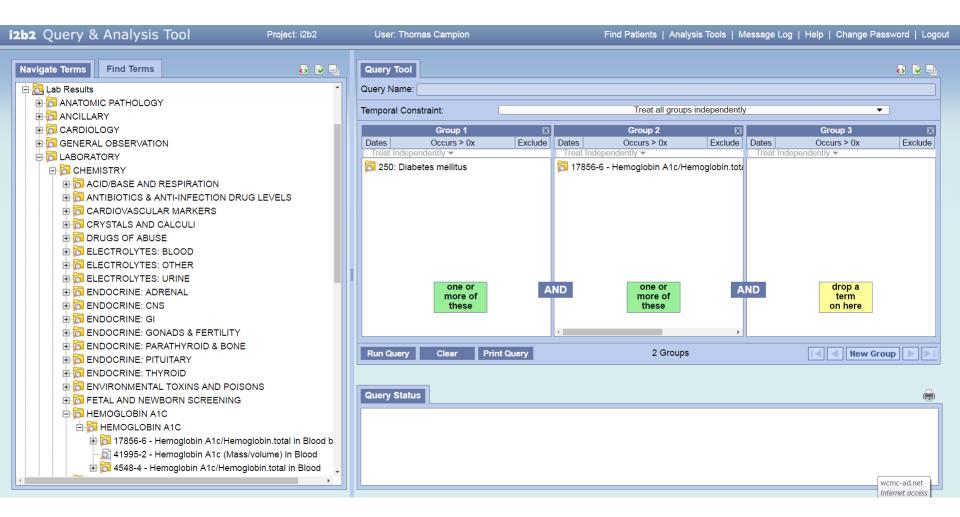


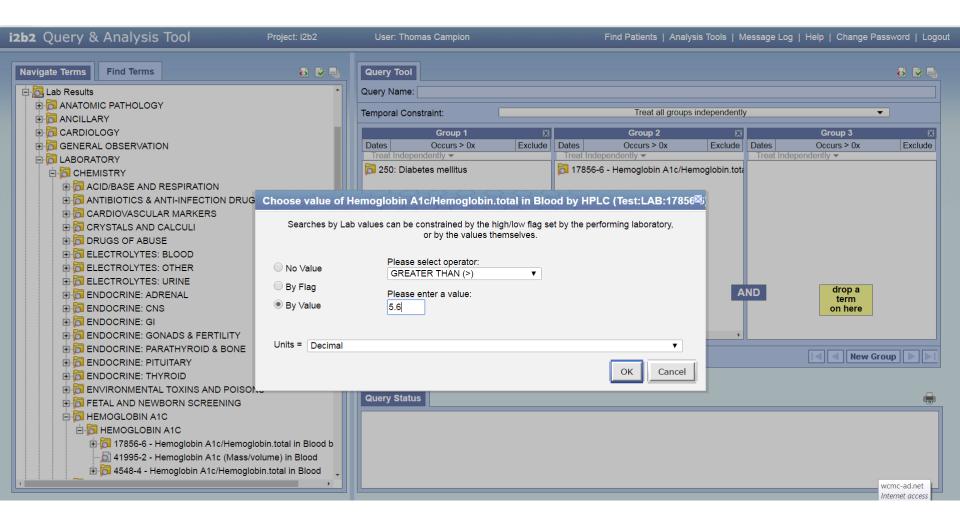


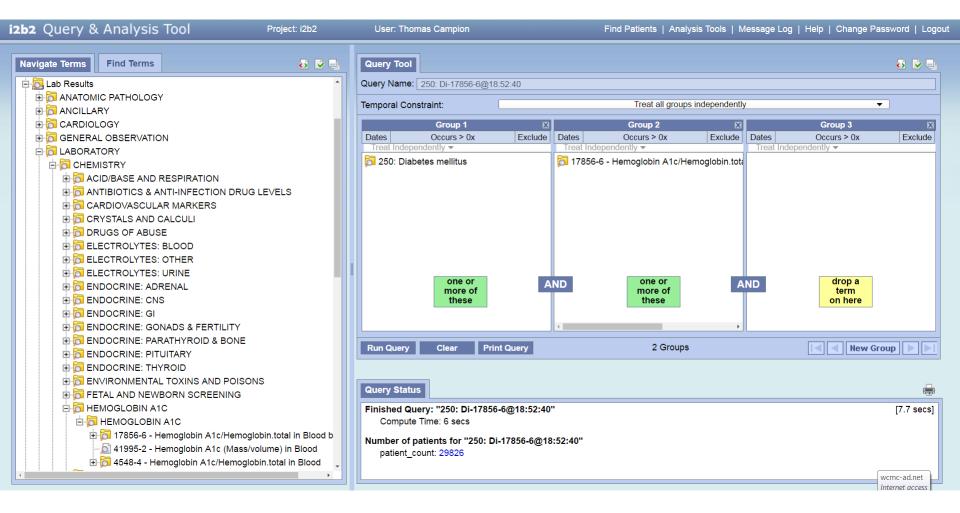






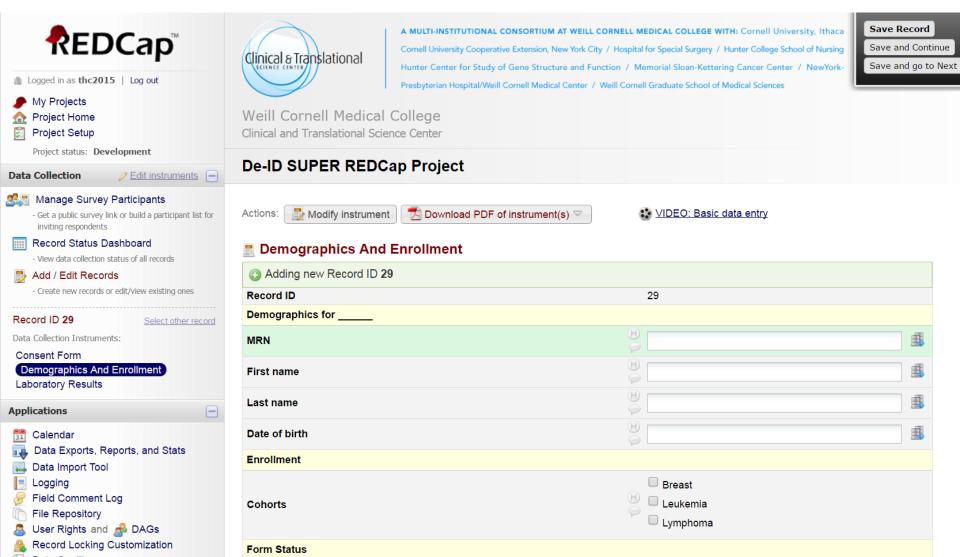


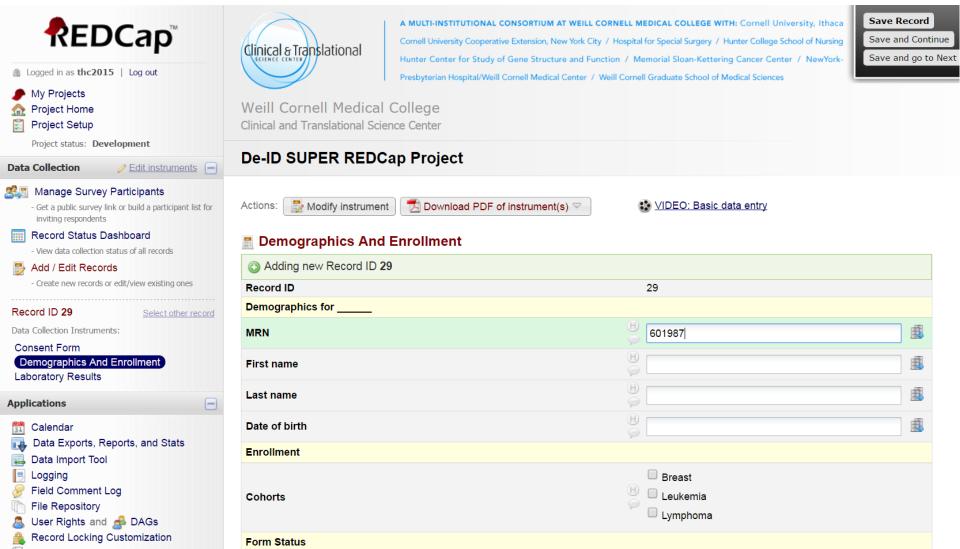


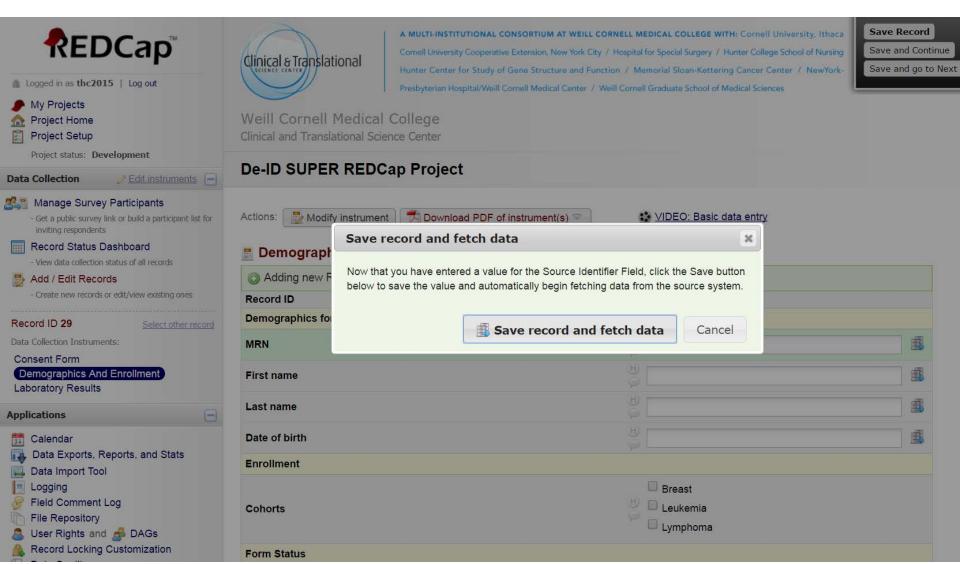


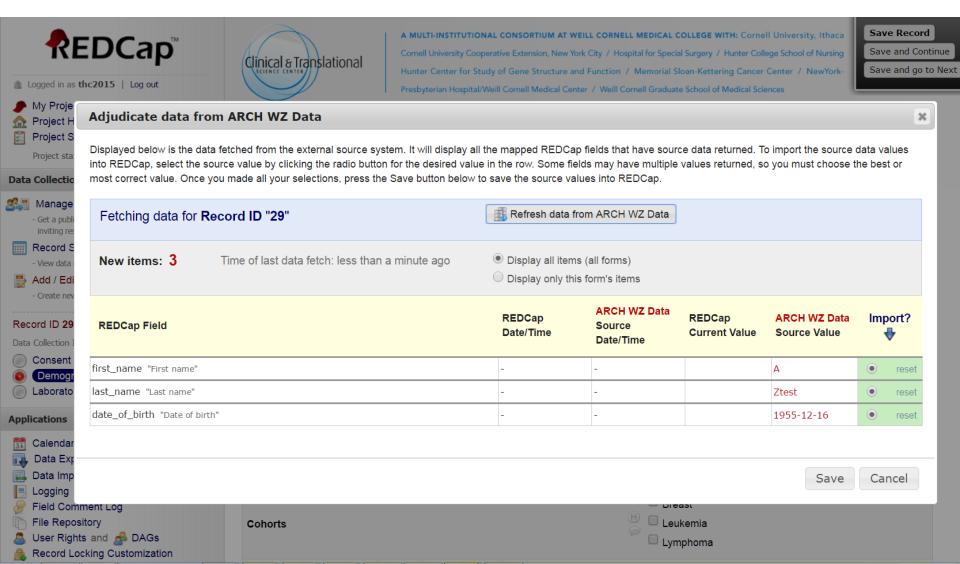


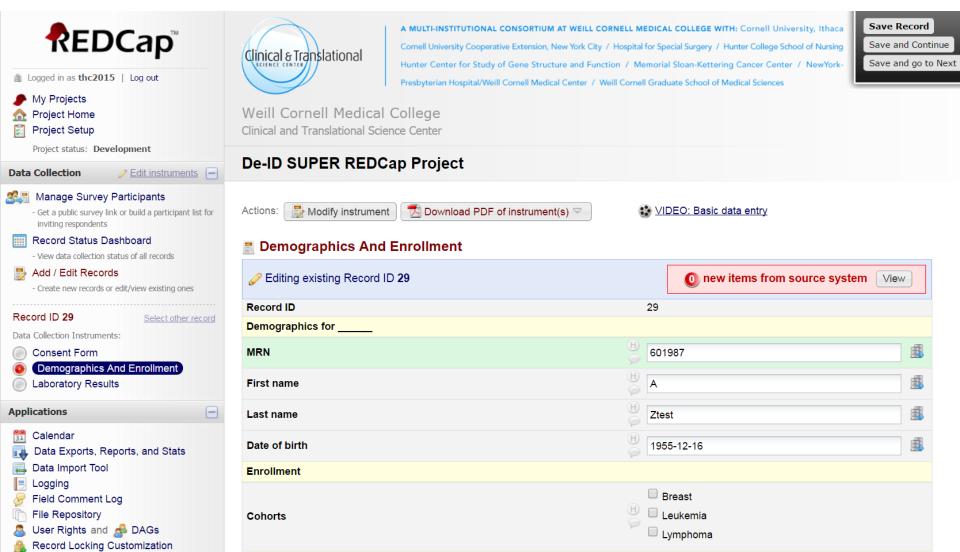


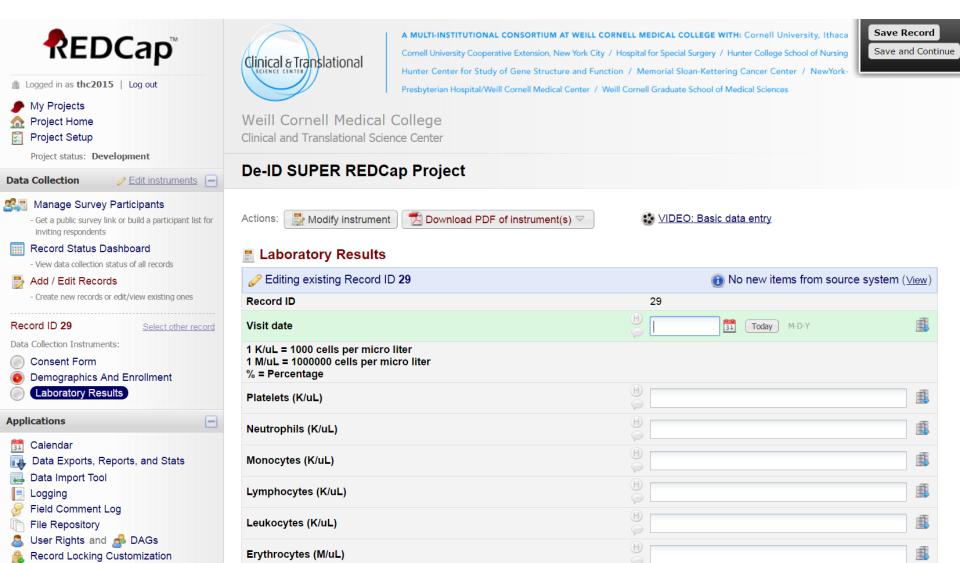


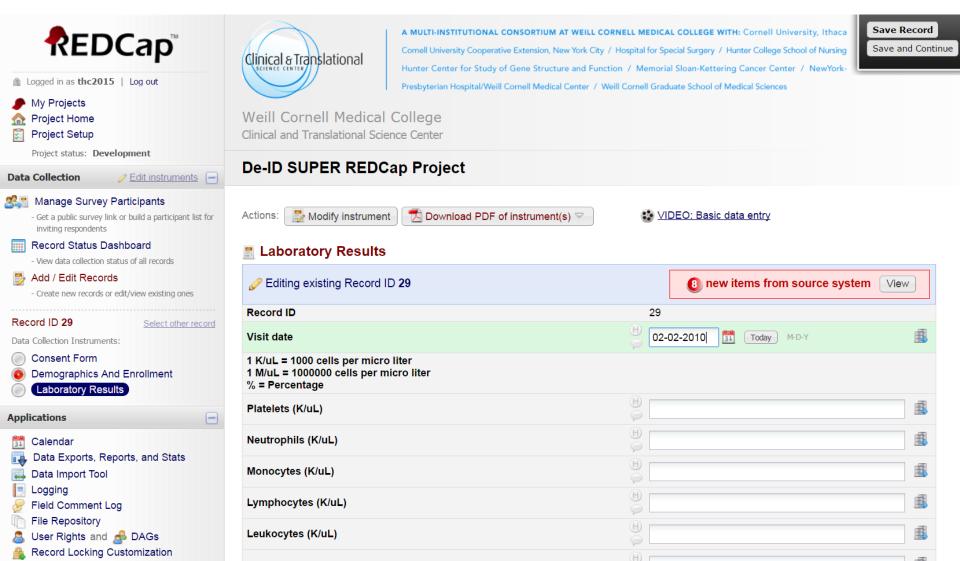


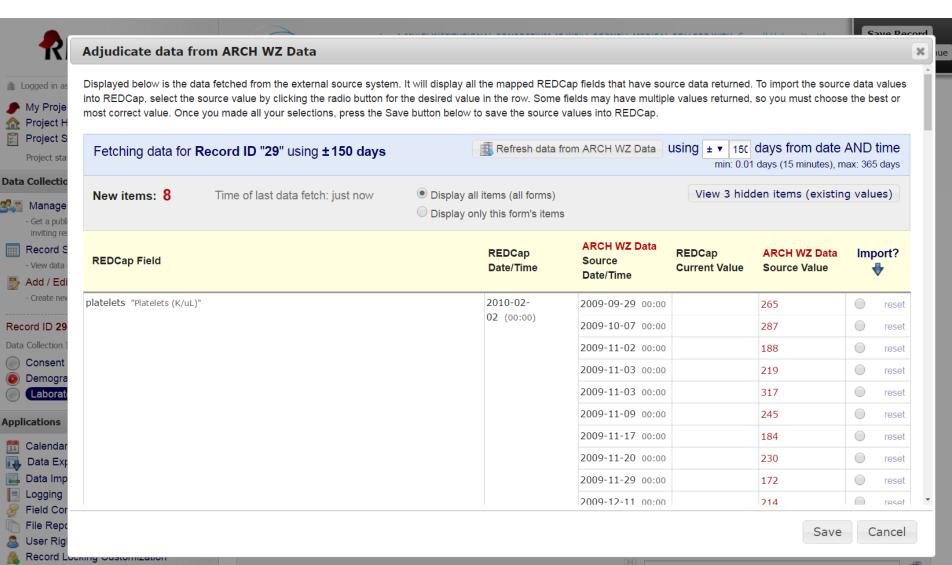


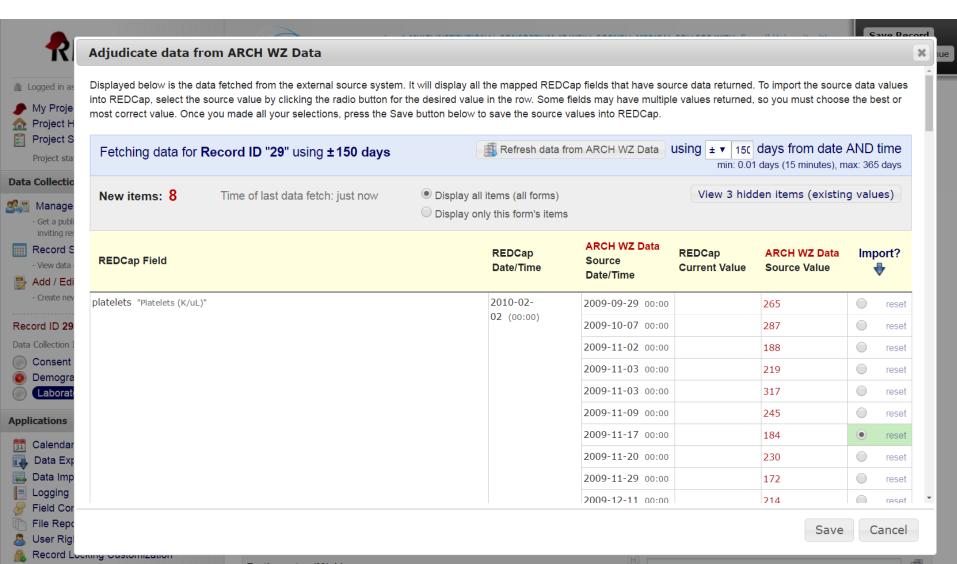


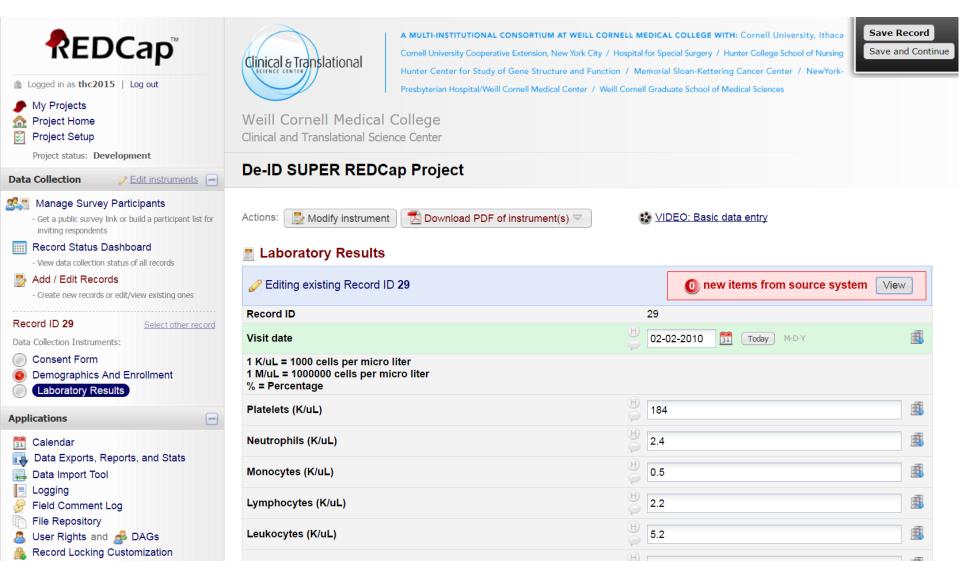




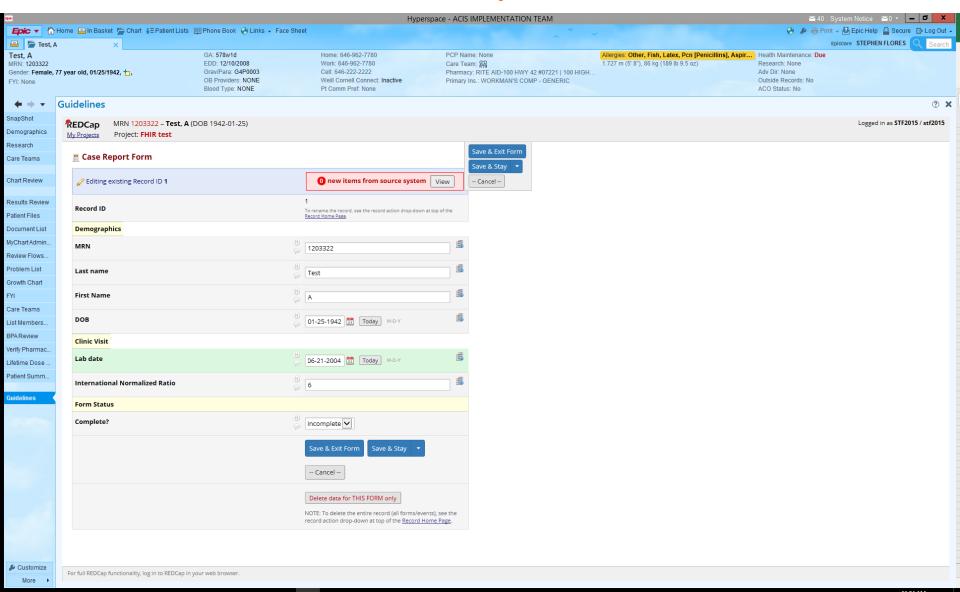








#### ARCH: SUPER REDCap on FHIR



Retrospective





Retrospective







Retrospective

College-Electronic Multi-**EDC** Research wide **EHR** Data institutional integrated Data Data Cohort Reporting Capture Repositories with EHR Sharing (EDC) Discovery







Retrospective

College-Electronic Multi-**EDC** Research wide **EHR** Data institutional integrated Data Data Cohort Reporting Capture Repositories with EHR Sharing (EDC) Discovery





Allscripts

CompuRecord

Retrospective Prospective

College-Electronic Multi-**EDC** Research wide **EHR** Data institutional integrated Data Data Cohort Reporting Capture Repositories with EHR Sharing (EDC) Discovery



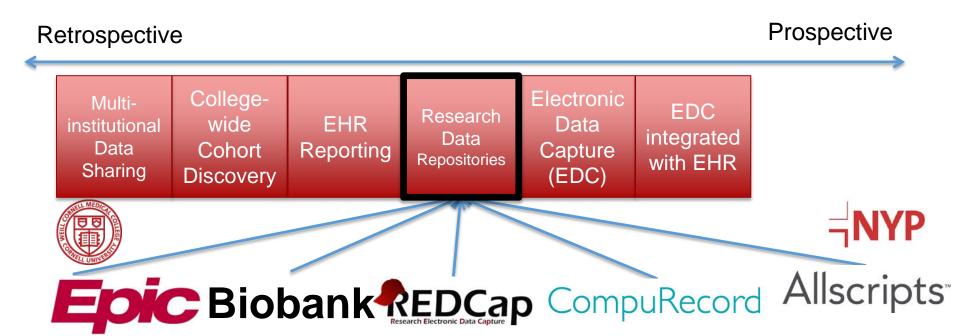


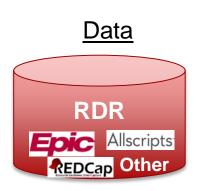
Allscripts<sup>\*</sup>

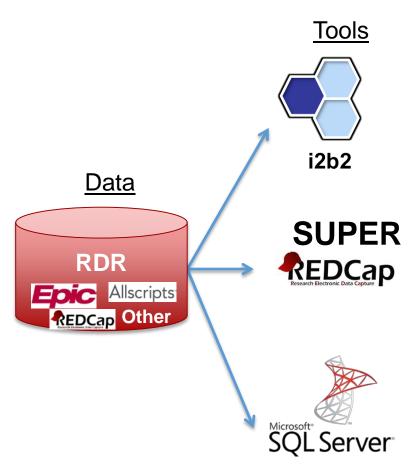
CompuRecord

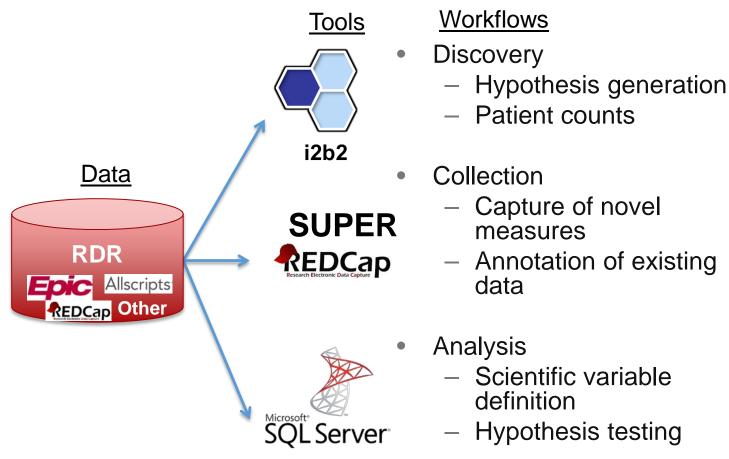
**Biobank** 

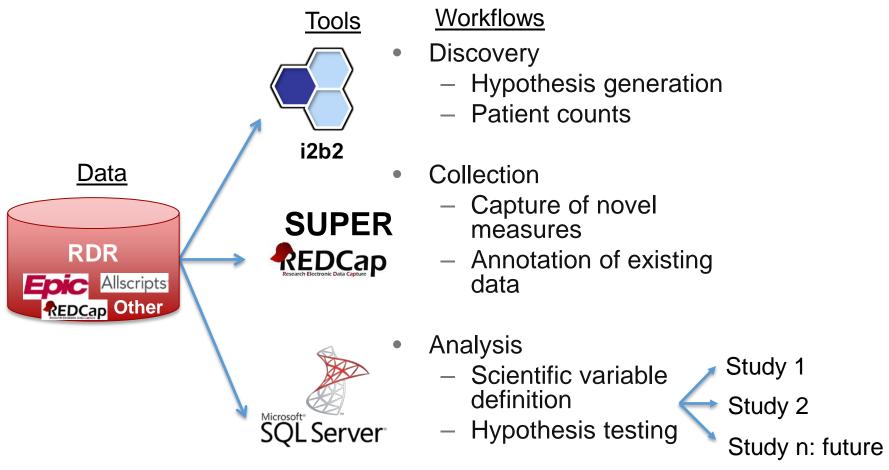






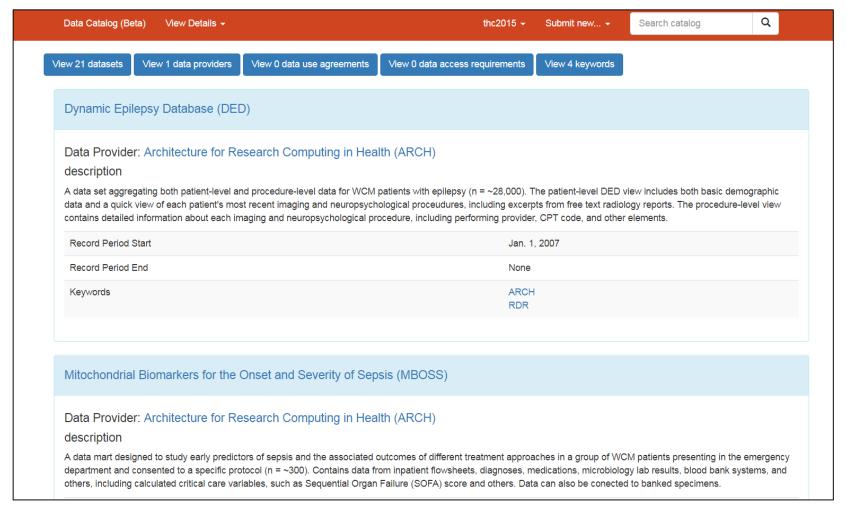






- All of Us Research Program (Kaushal)
- Anesthesiology (Turnbull)
- Center for Advanced Digestive Care (Shah)
- Clinical and Translational Neuroscience Unit (Kamel)
- Dalio Institute (Min)
- Health Informatics (Pathak/Zhang)
- Leukemia Program (Roboz)
- Myeloproliferative Neoplasms (Scandura)
- Neurological Surgery (Hoffman)
- Ophthalmology (D'Amico)
- Pediatric Epilepsy (Grinspan)
- Pediatric Pulmonology (Ono)
- Pulmonary and Critical Care (Schenck)
- Urology (Hu)
- Trauma and Surgical Critical Care (Winchell)

## **ARCH RDRs: Data Catalog**



#### Discussion

- Research Informatics can help investigators
  - Obtain EHR data
  - Collect novel research data
  - Integrate data
- ARCH matches investigators with right tools and services with respect to
  - Study design
  - Data sources
  - Cost

# Acknowledgments

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- Steven Flores
- Joseph Kabariti
- Ryan McGregor
- Sean Pompea
- Julian Schwartz
- Evan Sholle
- Scott Turner
- Jacob Weiser

#### Questions

- Access and inquiries
  - ARCH inquiries: arch-support@med.cornell.edu
  - i2b2 access: i2b2-support@med.cornell.edu
- Web resources
  - ARCH: http://arch.weill.cornell.edu
  - TRAC: https://webapps.nyp.org/trac
  - Data Catalog: http://datacatalog.weill.cornell.edu
- Tom Campion
  - thc2015@med.cornell.edu
  - **646-962-2345**
  - http://vivo.med.cornell.edu/display/cwid-thc2015