

Electronic medical records (EMR) for discovery genomics research in immune-mediated disease

Robert Plenge, M.D., Ph.D.

i2b2 Annual Academic Users' Group Meeting

June 28, 2011



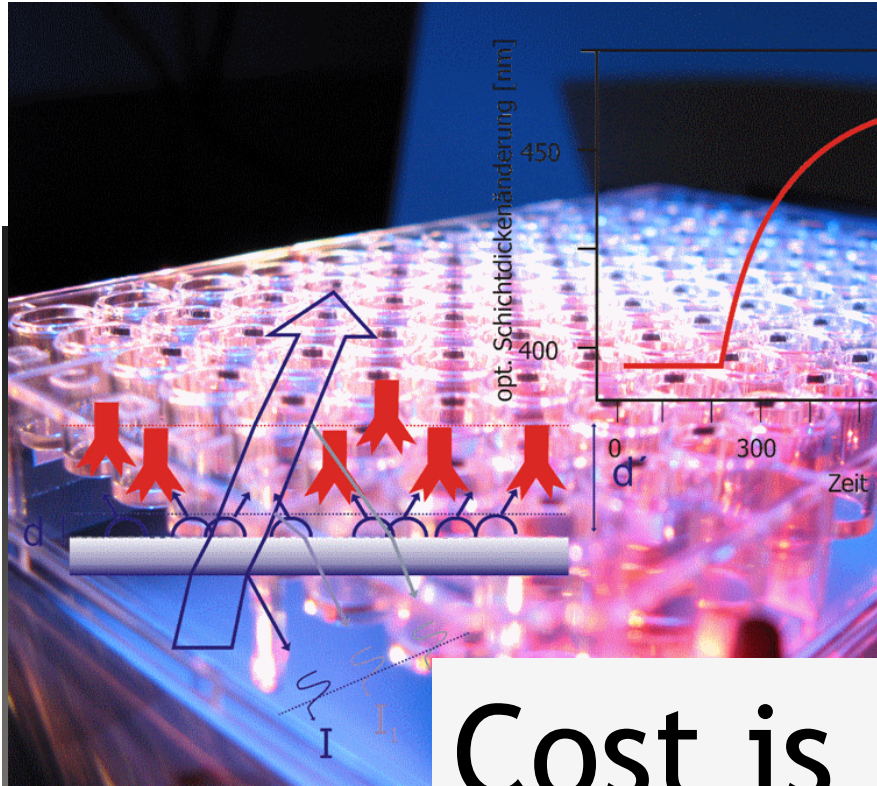
BRIGHAM AND
WOMEN'S HOSPITAL
A Teaching Affiliate of Harvard Medical School



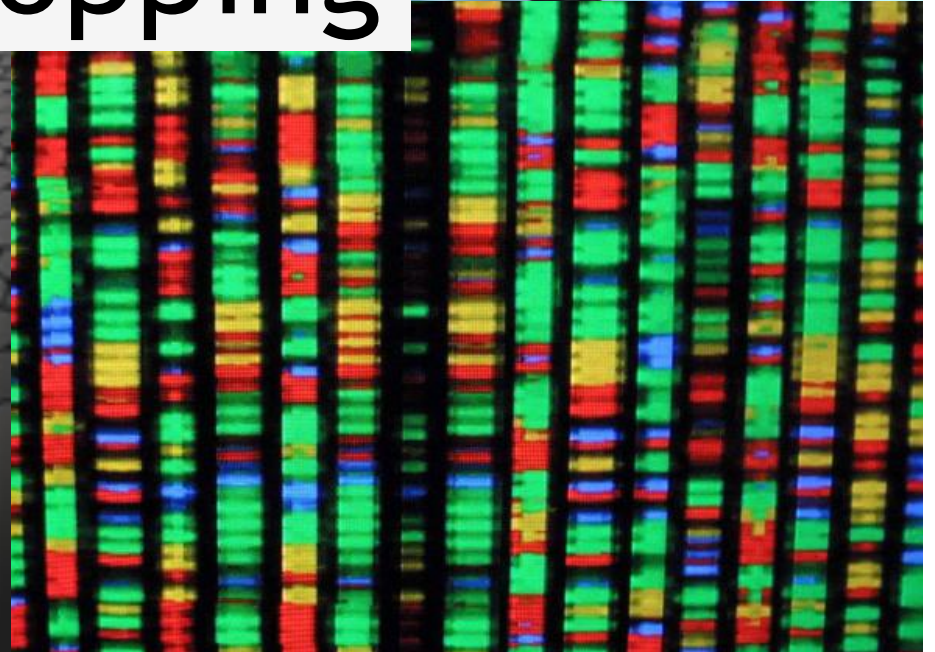
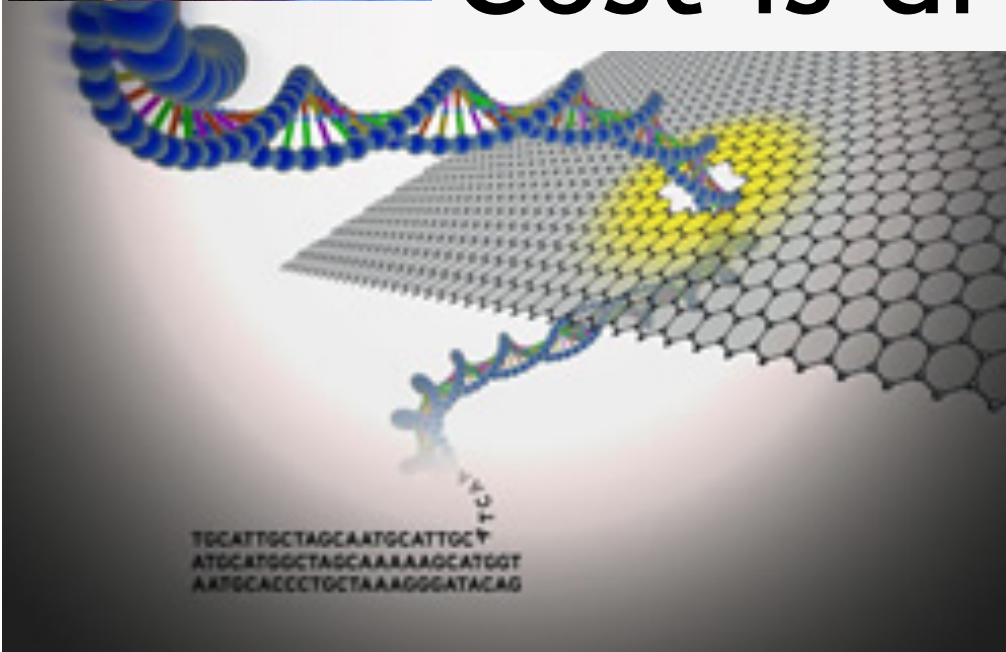
HARVARD
MEDICAL SCHOOL



BROAD
INSTITUTE

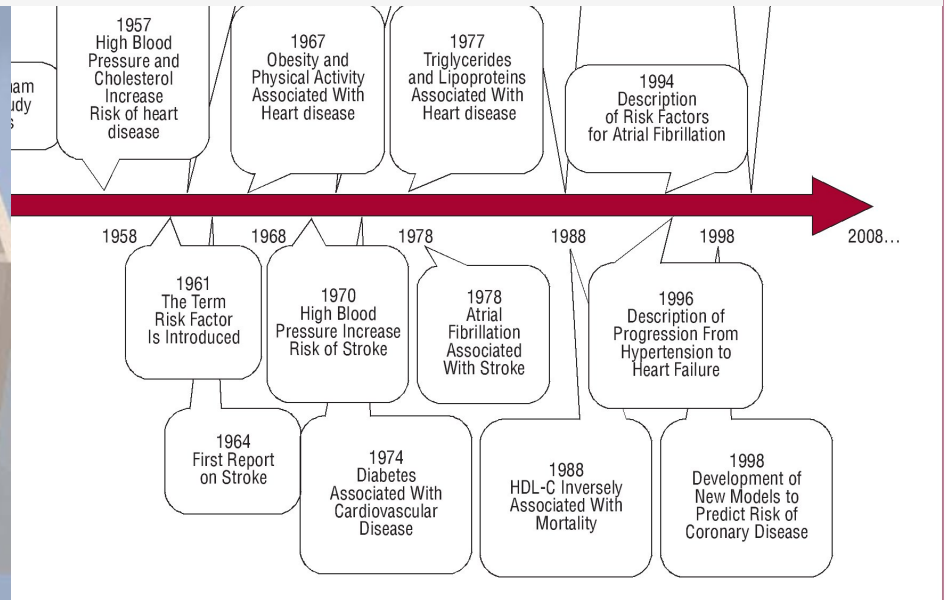


Cost is dropping



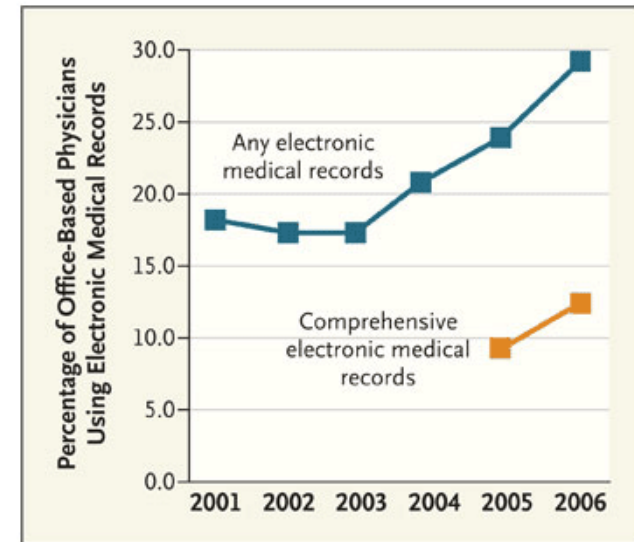


Phenotyping remains expensive



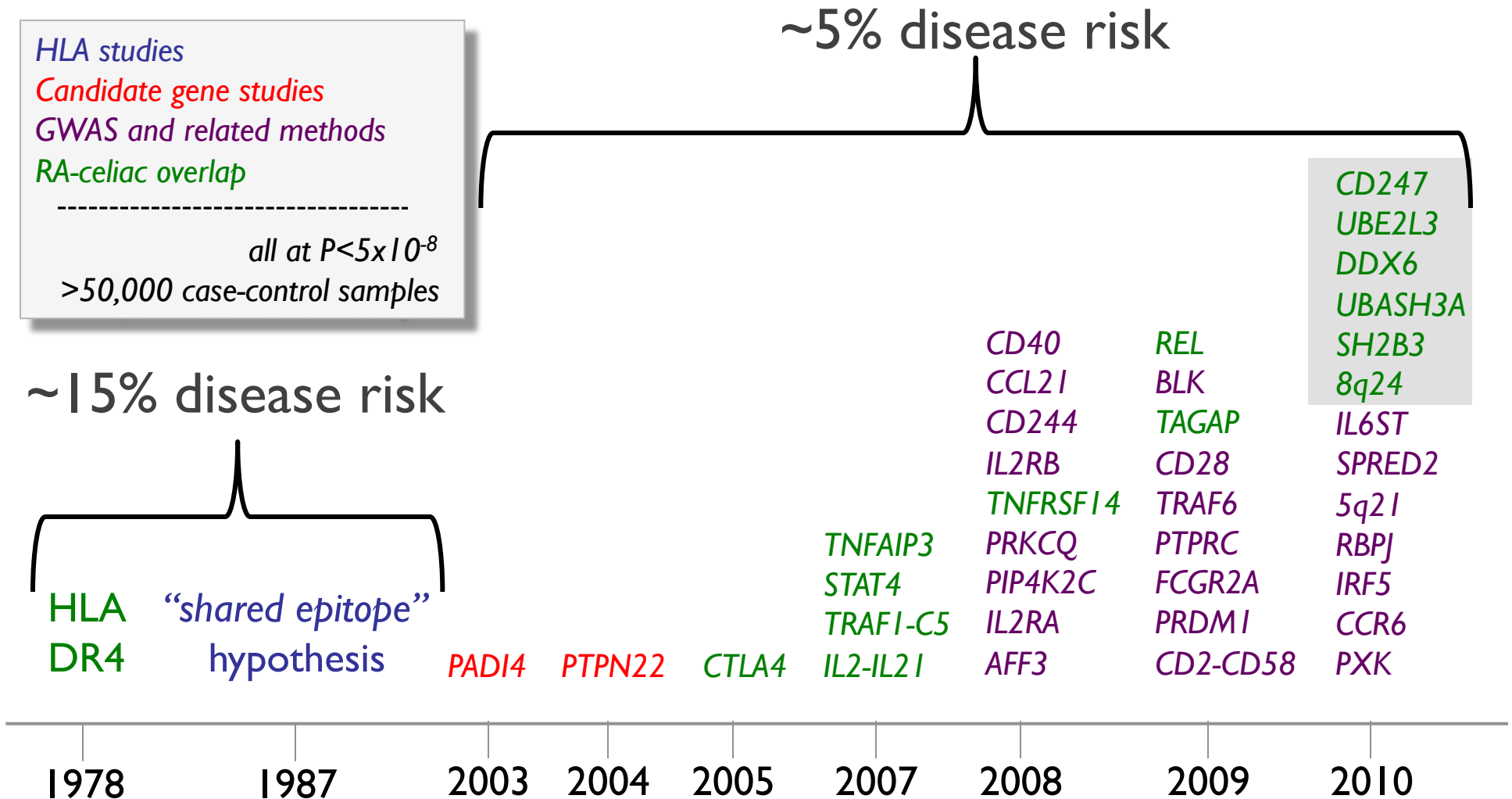
How will we realize the
ultimate potential of
genomics if phenotyping
is rate-limiting?

Can electronic medical records help?





Many risk loci remain “hidden”



Clinically relevant subsets of RA



Lung and cardiovascular diseases, response to therapy



What are the options for
collecting clinical data *and*
DNA for genetic studies?

Options for clinical + DNA

design	Clinical data	DNA	Sample size	cost
clinical trial	+++	+++	+	\$\$\$
registry	++	+++	++	\$\$
claims data	+	n/a	+++	\$
EMR	++	+++	+++	\$

i2b2

Informatics for Integrating Biology & the Bedside

A National Center for Biomedical Computing

About Us | Driving Biology Projects | Software | Resources | Events | Training | News | Collaborations | Publications

MISSION

i2b2 (Informatics for Integrating Biology and the Bedside) is an NIH-funded National Center for Biomedical Computing based at Partners HealthCare System. The i2b2 Center is developing a scalable informatics framework that will enable clinical researchers to use existing clinical data for discovery research and, when combined with IRB-approved genomic data, facilitate the design of targeted therapies for individual patients with diseases having genetic origins. This platform currently enjoys wide international adoption by the CTSA network, academic health centers, and industry. i2b2 is funded as a cooperative agreement with the National Institutes of Health.

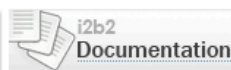
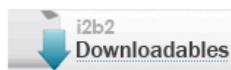
DRIVING BIOLOGY PROJECTS

- :: Overview
- :: Airways Diseases
- :: Hypertension
- :: Type 2 Diabetes Mellitus
- :: Huntington's Disease
- :: Major Depressive Disorder
- :: Rheumatoid Arthritis
- :: Obesity

RESOURCES

- :: Overview
- :: Computational Tools
- :: De-Identification Demo
- :: Documentation
- :: NLP Research Data Sets
- :: NLP Shared Tasks

SOFTWARE



HIGHLIGHTS

**** i2b2 NLP DATA SETS #2 AND #3 NOW AVAILABLE FOR RESEARCH PURPOSES ****

A complete set of annotated and unannotated, deidentified patient discharge summaries from the First, Second (Obesity) and Third (Medication) Shared Tasks for Challenges in NLP for Clinical Data are now available to the community for research purposes. Check it out at our [NLP Data Sets page](#). Please note you must register AND submit a DUA for access.

****FALL AUG MEETING**** (In conjunction with CTSA IT Annual Meeting) Slides now available on our [AUG Page](#).

UC Davis Team Wins Gold Award for Cohort Discovery Project
see [details](#) on our AUG page



Kohane



Murphy



Churchill

...and many others!

Outline of talk today

- Demonstration: developing an algorithm to define an RA cohort, proof-of-concept genomic studies
- Portability: implementing the EMR classification algorithm at other institutions
- Application: defining subsets of patients with clinically-relevant outcomes – and cardiovascular disease in particular

This is not a new idea...

THE SENSITIVITY AND SPECIFICITY OF COMPUTERIZED DATABASES FOR THE DIAGNOSIS OF RHEUMATOID ARTHRITIS

SHERINE E. GABRIEL

Objective. To examine the accuracy of a computerized medical database for the diagnosis of rheumatoid arthritis (RA).

Methods. The complete medical records of all prevalent cases of RA (according to the 1987 American College of Rheumatology diagnostic criteria) on January 1, 1987 were reviewed to determine the sensitivity, specificity, and predictive value of database diagnoses compared with those obtained by medical record review. Agreement between database and medical record diagnoses was calculated using the kappa statistic.

Results. Computerized database diagnoses of RA had a sensitivity of 89%, a specificity of 74%, a positive predictive value of 57%, and a negative predictive value of 94% compared with diagnoses based on clinical information abstracted from the complete medical record. Agreement between database and medical record diagnoses was poor ($\kappa = 0.54$).

Conclusion. The sole reliance on such databases for the diagnosis of RA can result in substantial misdiagnosis.

ters (dating back graphically circ limited number been computeriz original, complet all inpatient, outp home encounters unique data reso was undertaken record-linked da determining the s derived diagnosi medical record c cases of RA.

The population itself to epidemiologic vided primarily by the M hospitals (Rochester M smaller group practice affiliated Olmsted Com

Table 1. Comparison of database-derived versus medical record-derived diagnoses of rheumatoid arthritis (RA)*

	Medical record diagnosis of RA†		Total
	Yes	No	
Database diagnosis of RA			
Yes	399	300	699
No	50	853	903
Total	449	1,153	1,602

* Sensitivity = $399/449 = 89\%$; specificity = $853/1,153 = 74\%$; positive predictive value = $399/699 = 57\%$; negative predictive value = $853/903 = 94\%$.

† Based on the American College of Rheumatology diagnostic criteria (9).

Sens: 89%
PPV: 57%

...but EMR data are “dirty”

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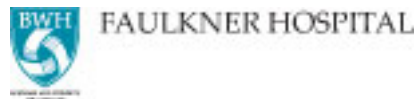
Conclusion. The sole reliance on computerized database diagnoses for the diagnosis of RA can result in substantial misdiagnosis.

ters (dating back to 1910) among residents of a geographically circumscribed area, which is served by a limited number of providers (6). This database has been computerized since 1950, and in addition, the original, complete medical records (including data on all inpatient, outpatient, emergency room, and nursing home encounters) are available for review. Using this unique data resource as an example, the present study

Conclusion: The sole reliance on such databases for the diagnosis of RA can result in substantial misdiagnosis.

hospitals (Rochester Methodist and Saint Mary's) and one smaller group practice (the Olmsted Medical Group and its affiliated Olmsted Community Hospital) (7). Any diagnosis

Partners HealthCare: *4 million patients*



Partners HealthCare: *linked by EMR*



Reminders

- No known documented smoking status. Click to enter status.
- Patient 65 yrs or older, due for Pneumovax.
- Patient with DM overdue for HbA1c (next 4-6 months).

Notes

Date	Subject
04/28/2009	Diabetes Intake and Follow-up Patient Consult Note
03/20/2009	Treatment Plan
02/25/2009	Patient Note
02/25/2009	Treatment Plan
02/20/2009	Treatment Plan
02/19/2009	Patient Note
02/19/2009	Treatment Plan
02/02/2009	Patient Note
06/09/2008	Patient Note
06/05/2008	Patient Note
02/28/2008	Patient Note
01/23/2008	Need approval
01/23/2008	Supervisor Note
01/10/2008	Patient Note
06/26/2007	Patient Note
06/14/2007	Patient Note
06/14/2007	Consult Letter
06/14/2007	Patient Note

Medications

Item Name	03/09/2009	02/04/2009	07/03/2007	07/03/2007
Abilify (ARIPRAZOLE) 10 MG (10MG TABLET Take 1) PO QD x 30				
Olanzapine (OLANZAPINE) 5 MG (5MG CAPSULE Take 1) PO QD				
Prednisone 5 MG (5MG TABLET Take 1) PO QD x 90 days				

Allergies

Allergen	Reaction
Bee Stings	Anaphylaxis
Penicillin	Hives
Flovabeets	Anaphylaxis

Flovabeets

Item Name	03/09/2009	02/04/2009	07/03/2007	07/03/2007
BLOOD PRESSURE	120/80			
TEMPERATURE	97.2			
PULSE	72			
RESPIRATORY RATE				
O2 SAT				
HEIGHT	54 in	52 in	52 in	52 in
WEIGHT	179 lb		200 lb	200 lb
BMI	43.2	52.1	52.1	52.1
PAIN LEVEL	0			

Problems

Problem	03/09/2009	02/04/2009	07/03/2007	07/03/2007
Restless leg syndrome - Motor				
Head trauma				
Headache				
Growth hormone deficiency				
Celiac disease				
Diabetes mellitus type 2 - customized description				
Bantouan positive				
Bloating - bloating				

Family History

Family History Problem	Relative
diabetes mellitus type 2	Father

Last Known Values

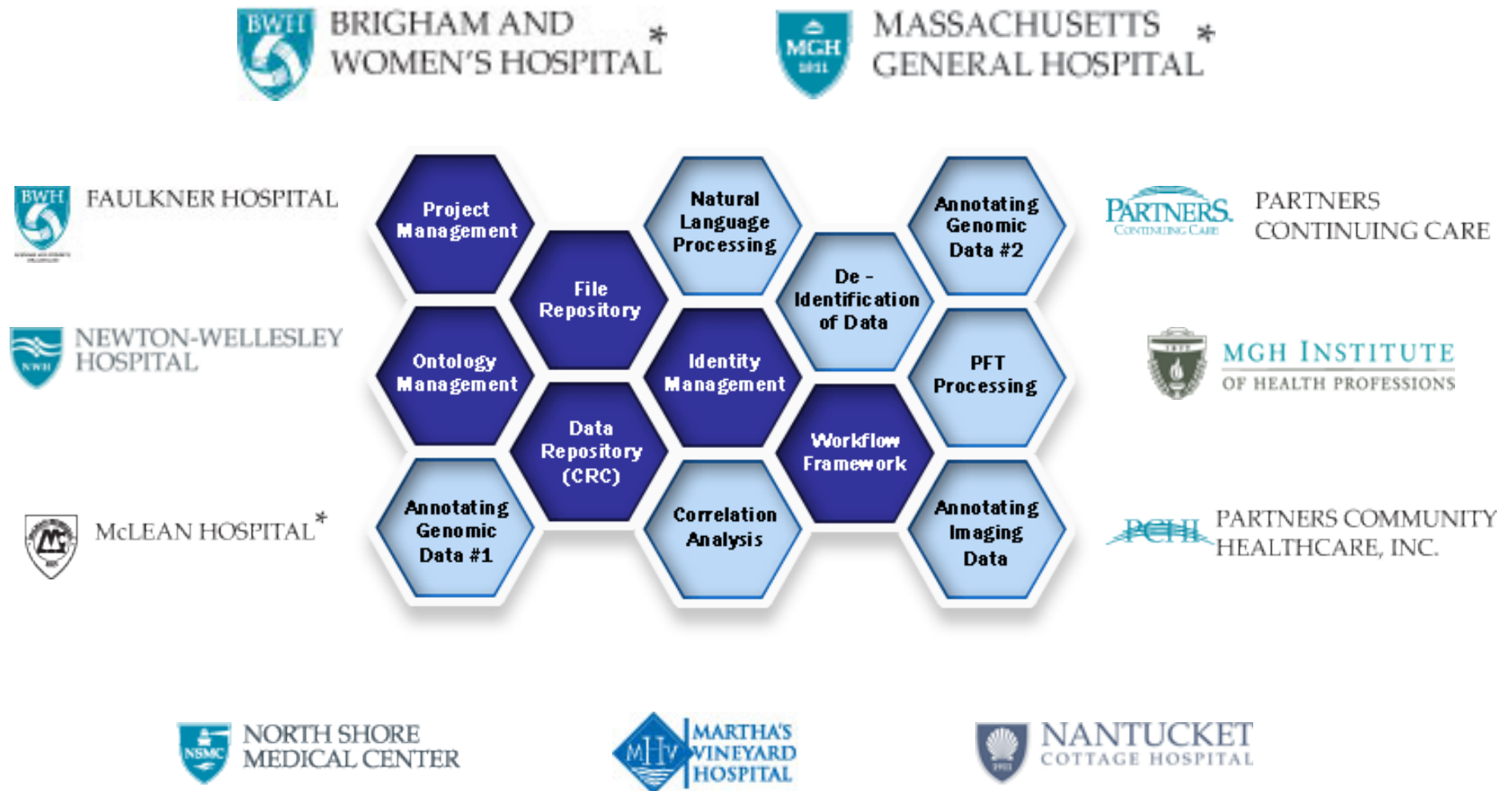
Item Name	03/09/2009	02/04/2009	07/03/2007	07/03/2007
BLOOD PRESSURE	120/80			
TEMPERATURE	97.2			
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RESPIRATORY RATE				
O2 SAT				
HEIGHT	54 in	52 in	52 in	52 in
WEIGHT	179 lb		200 lb	200 lb
BMI	43.2	52.1	52.1	52.1
PAIN LEVEL	0			

Immunization

Item Name	Dose 1	Dose 2	Dose 3	Dose 4	Dose 5	Dose 6
DTP	12/01/95	02/01/96	04/01/96			
Hib	12/01/95	02/01/96	04/01/96			
Hep B	10/10/95	11/01/95	07/01/96			
IPV	10/10/95	02/01/96	04/01/96			
H1N1	12/01/96	04/05/2000				
PPD	09/01/96	04/19/2005				
Tdap	04/04/2007					
Tetanus	08/29/2008					
Varicella	09/03/98	04/04/2007				



Partners HealthCare: *organized by i2b2*

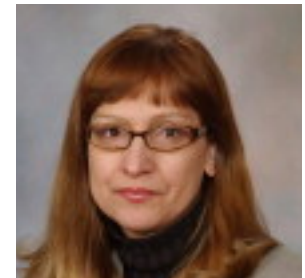


Our library of RA phenotypes

- Natural language processing (NLP)
 - *disease terms (e.g., RA, lupus)*
 - *medications (e.g., methotrexate)*
 - *autoantibodies (e.g., CCP, RF)*
 - *radiographic erosions*



Qing Zeng



Guergana Savova

Concept/term	Accuracy of concept
presence of erosion	88%
seropositive	96%
CCP positive	98.7%
RF positive	99.3%
etanercept	100%
methotrexate	100%

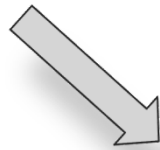
Our library of RA phenotypes

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 - *autoantibodies (e.g., CCP, RF)*
 - *radiographic erosions*
- Codified data
 - *ICD9 disease codes*
 - *prescription medications*
 - *laboratory autoantibodies*



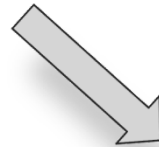
Shawn Murphy

4 million patients



*ICD9 RA and/or CCP checked
(goal = high sensitivity)*

31,171 patients



*Classification algorithm
(goal = high PPV)*

3,585

RA patients

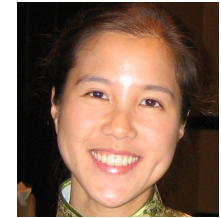


High PPV with adequate sensitivity

Model	PPV (SE)	Sensitivity (SE)
Codified + NLP	0.93 (0.02) ☆	0.63 (0.06)
NLP only	0.89 (0.02)	0.56 (0.05)
Codified only	0.88 (0.02)	0.51 (0.05)

☆392 out of 400 (98%) had definite or possible RA!

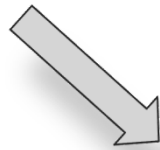
Clinical features of patients



Characteristics	<i>i2b2</i> RA	CORRONA
total number	3,585	7,971
Mean age (SD)	57.5 (17.5)	58.9 (13.4)
Female (%)	79.9	74.5
Anti-CCP(%)	63	N/A
RF (%)	74.4	72.1
Erosions (%)	59.2	59.7
MTX (%)	59.5	52.8
Anti-TNF (%)	32.6	22.6

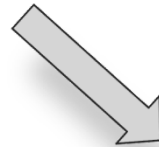
CCP has an OR = 1.5 for predicting erosions

4 million patients



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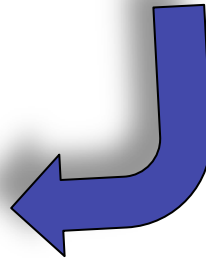


*Classification algorithm
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3,585

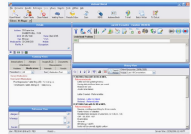
RA patients

*Discarded blood
for DNA*



“On demand” biorepository

NLP data



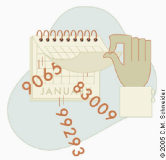
Narrative electronic medical record



Natural Language Processing (NLP)

Codified data

codified data (e.g., billing codes)



i2b2 informatics infrastructure

Algorithm to define patients with RA

i2b2 RA-DataMart

30,655 patients

NLP queries

autoantibody status
medication history

codified data

billing codes
laboratory values

Within 1 year (at \$30/sample):

1,800 RA cases

2,400 matched controls

RA patients



IDs

13100
65773
23001
12543

IDs

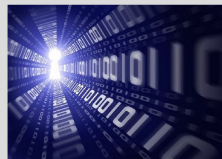
87443
61103
49011
12543

Match!

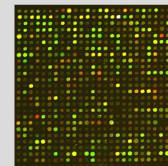


FIREWALL

anonymous clinical data

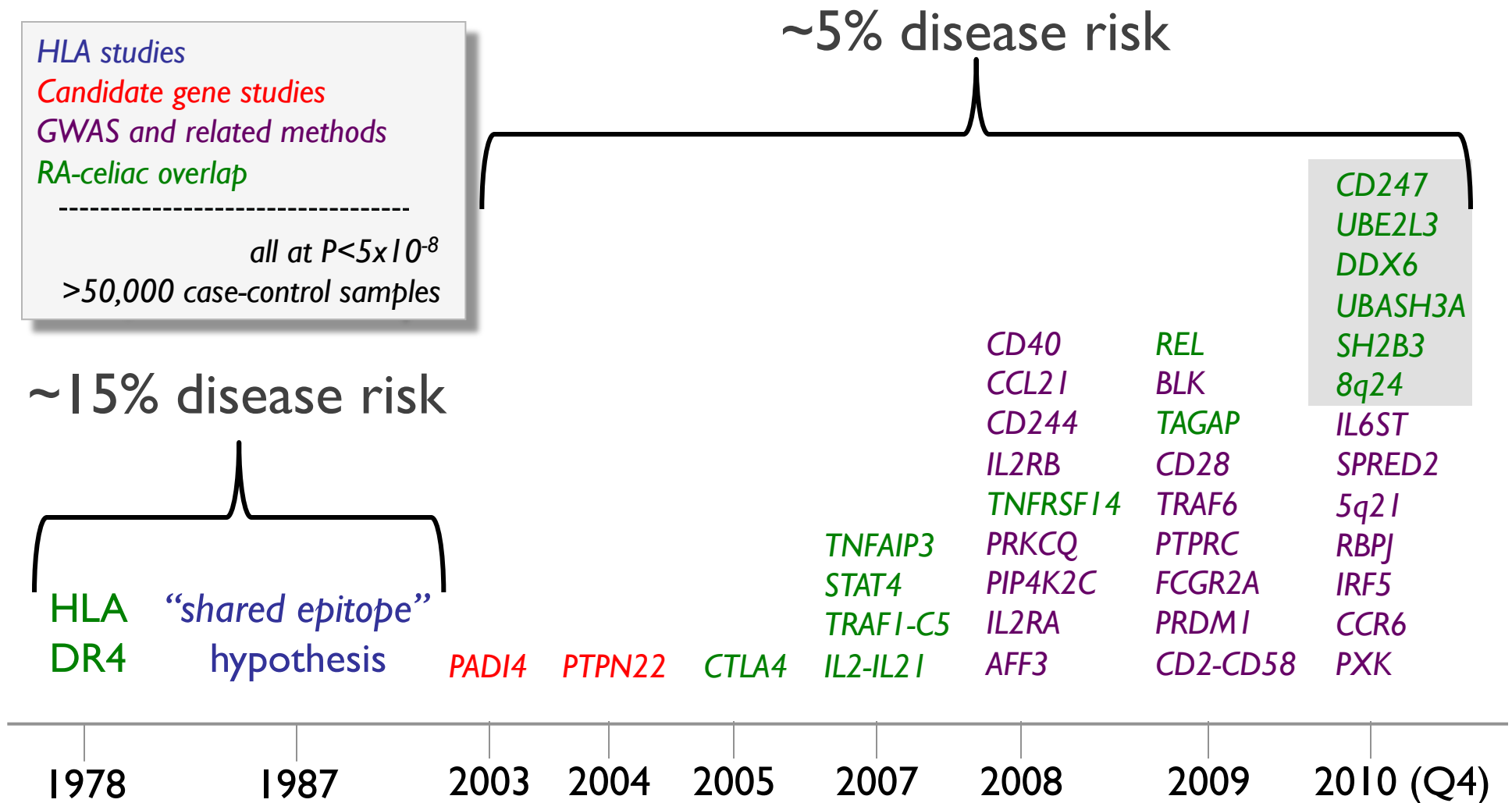


discarded blood sample for DNA



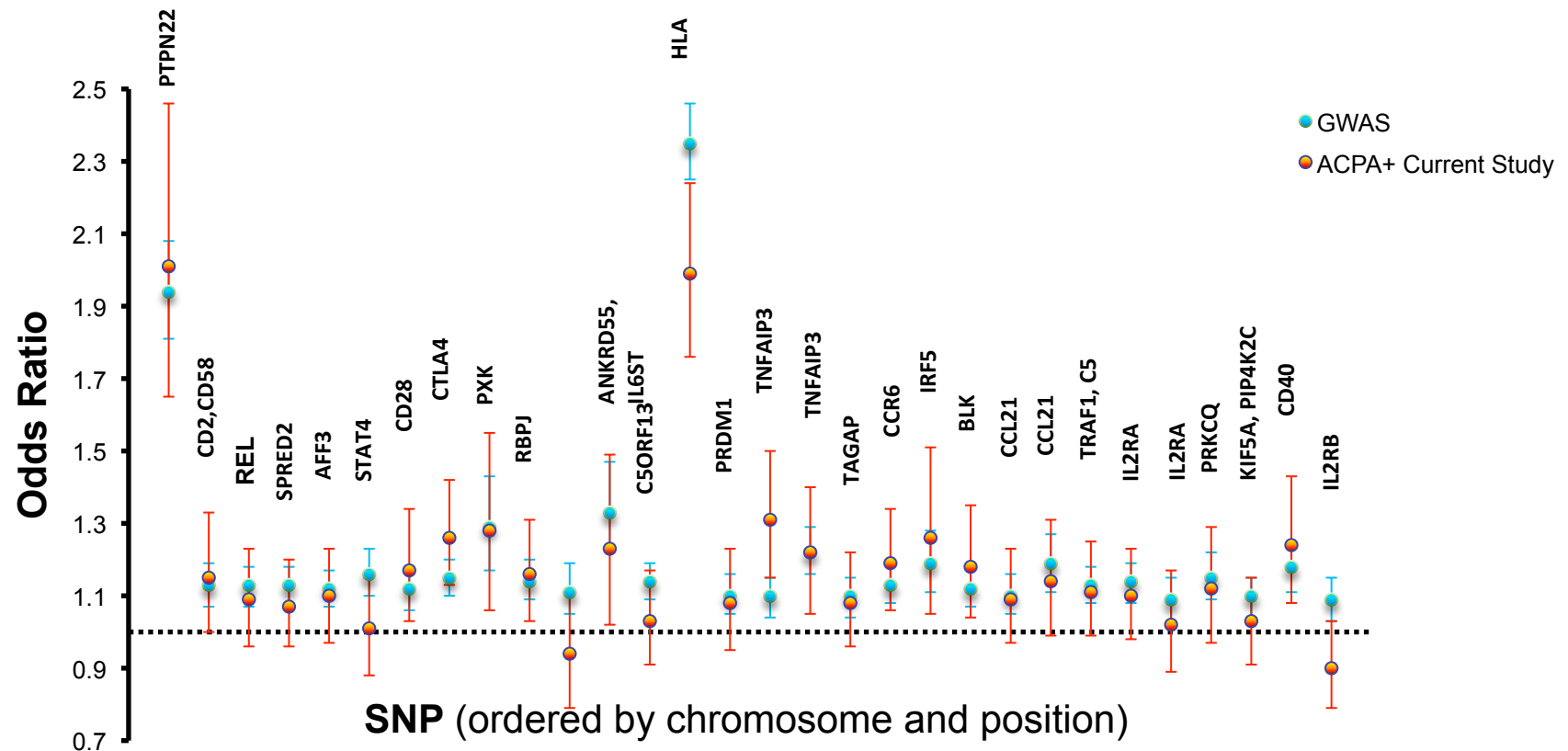
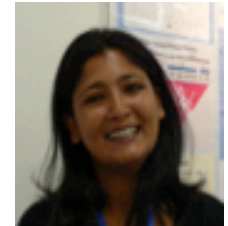
FINAL ANONYMOUS PATIENT CLINICAL DATA
WITH DNA FOR GENETIC STUDIES

June 2011: >35 RA risk loci



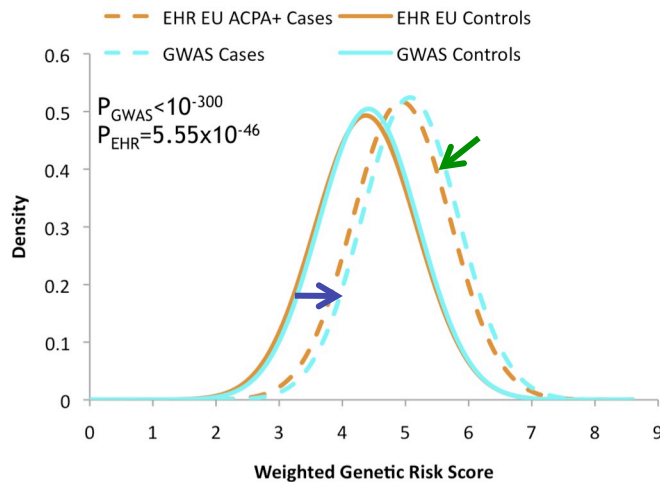
Zhernakova et al *PLoS Genetics* 2011 (in press)

OR similar in EMR cohort



~1,500 multi-ethnic RA cases and 1,500 matched controls

Genetic risk score similar...



European ancestry

→ RA case vs control

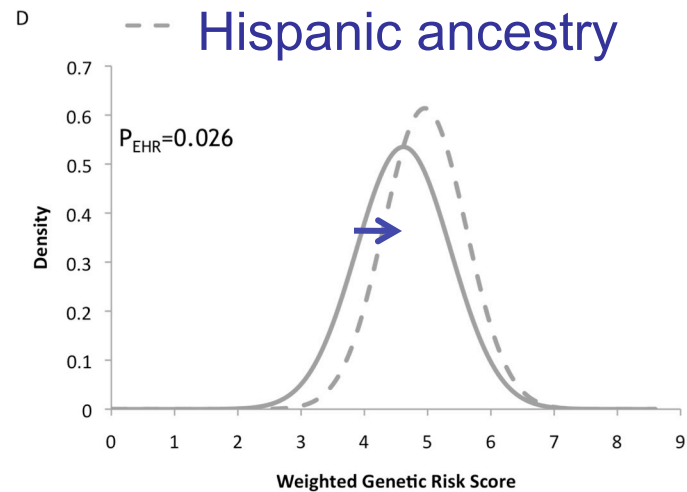
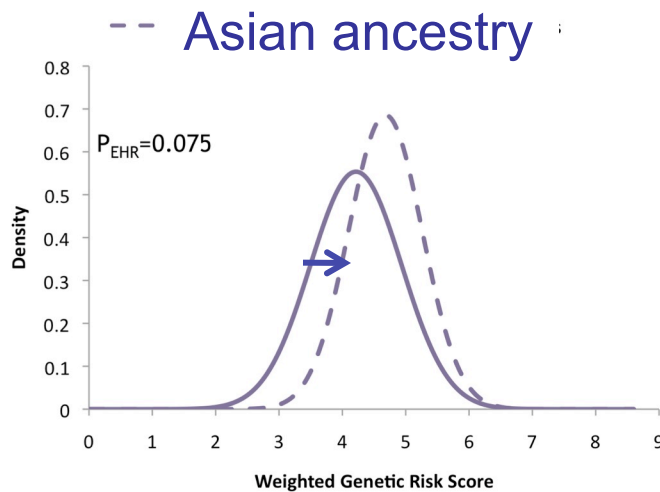
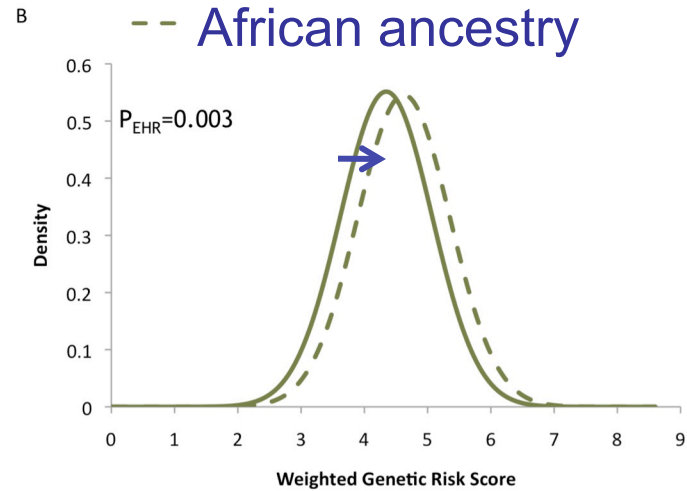
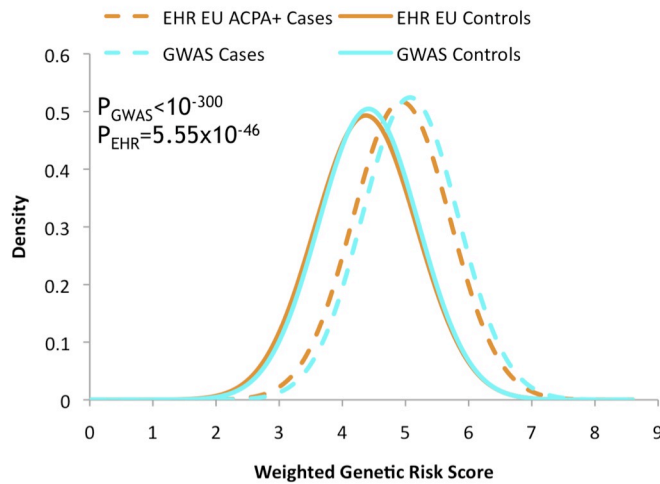
→ GWAS vs EMR (no difference!)

Low GRS

High GRS

1. Assign each risk allele a weight based on OR
2. Sum weights across all risk alleles per person (= “genetic risk score”)
3. Compare distribution of weighted GRS in cases vs controls
4. Compare GWAS GRS vs EMR GRS

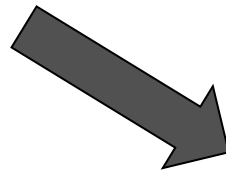
... across all ethnic groups



Outline of talk today

- Demonstration: developing an algorithm to define an RA cohort, proof-of-concept genomic studies
- Portability: implementing the EMR classification algorithm at other institutions
- Application: defining subsets of patients with clinically-relevant outcomes – and cardiovascular disease in particular

Portability to other institutions



Pharmacogenomics
Research Network

Good portability to other institutions

Institution	PPV (SE)	Sensitivity (SE)
Partners	0.93 (0.02)	0.63 (0.06)
Northwestern	0.80 (0.02)	0.50 (0.05)
Vanderbilt	0.92 (0.02)	0.54 (0.05)

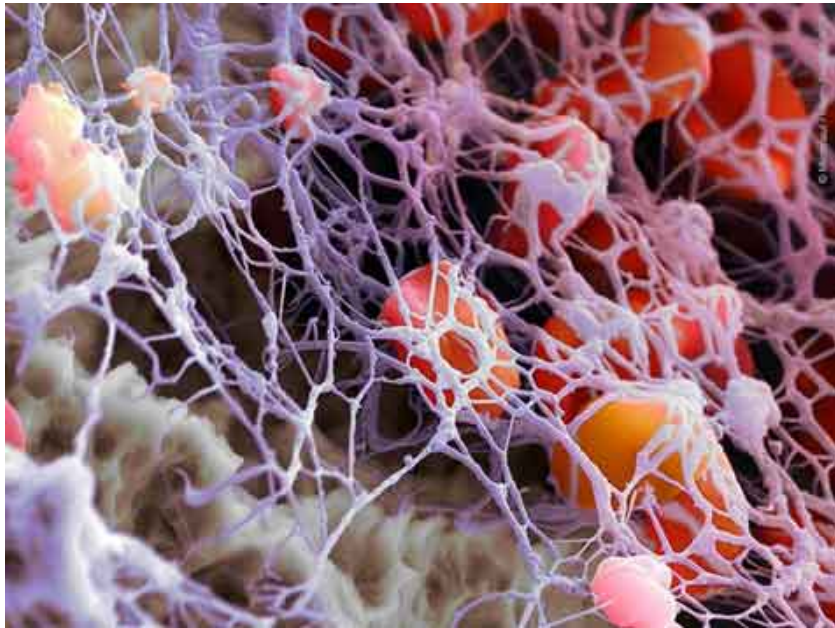
Note: it took us 2 years to develop the algorithm at Partners, but ~2 months to apply it at Northwestern/Vanderbilt. *Still, this needs to be faster (e.g., 2 minutes!)*

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Clinically relevant subsets of RA

cardiovascular disease



response to therapy



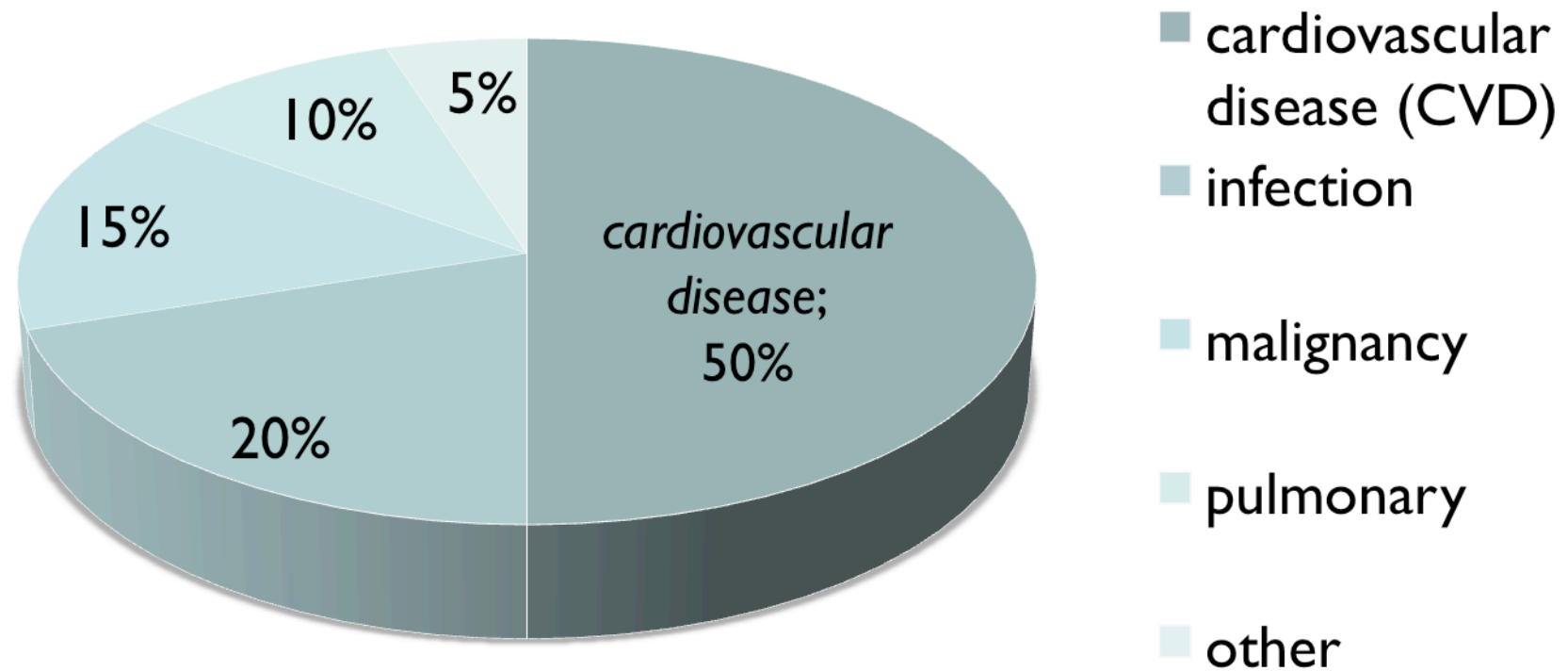
i2b2

Informatics for Integrating Biology & the Bedside



Pharmacogenomics
Research Network

Subset patients in clinically meaningful ways: *causes of mortality*



There is a 2-fold increased risk of CVD in RA patients...is this due to inflammation?

Link between CVD and inflammation

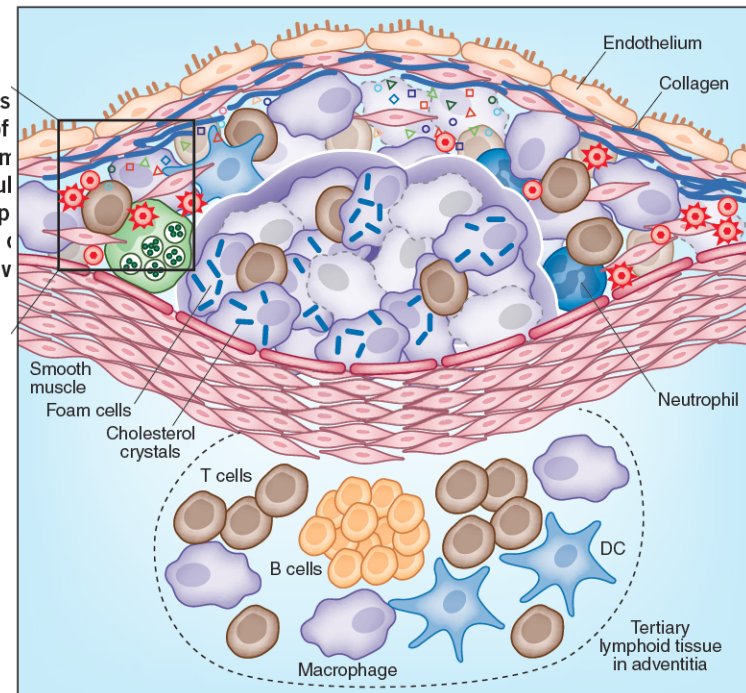
REVIEW

nature
immunology

The immune system in atherosclerosis

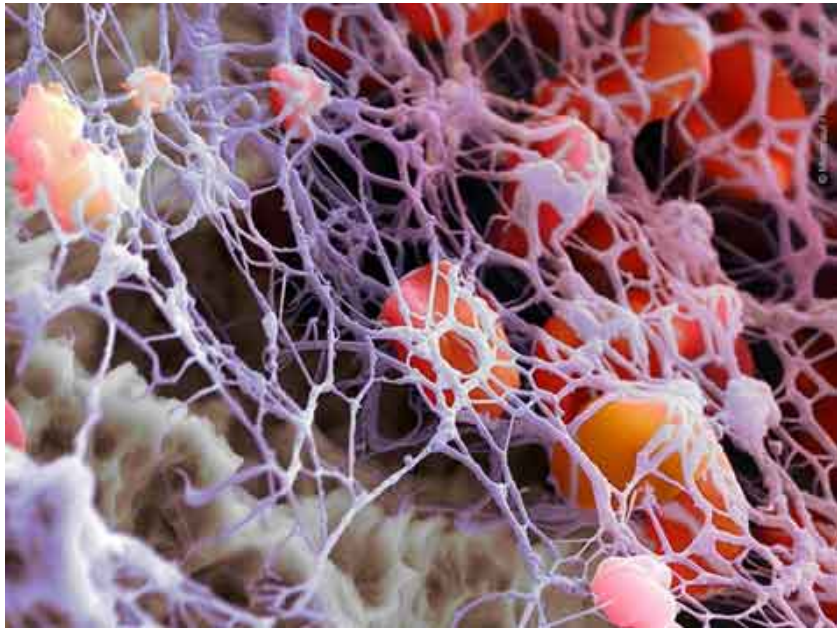
Göran K Hansson & Andreas Hermansson

Cardiovascular disease, a leading cause of mortality worldwide, is caused mainly by atherosclerosis disease of blood vessels. Lesions of atherosclerosis contain macrophages, T cells and other cells of together with cholesterol that infiltrates from the blood. Targeted deletion of genes encoding costimulatory proinflammatory cytokines results in less disease in mouse models, whereas interference with regulation of innate as well as adaptive immune responses have been identified in atherosclerosis, with components carrying low-density lipoprotein triggering inflammation, T cell activation and antibody production in disease. Studies are now under way to develop new therapies based on these concepts of the involvement of the immune system in atherosclerosis.

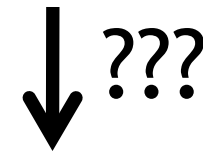


Work in progress: *model of CVD in RA*

cardiovascular disease



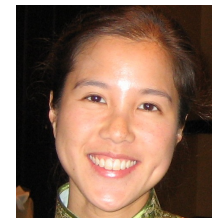
genetics + autoAbs



CVD

i2b2

Informatics for Integrating Biology & the Bedside



Clinical characteristics of CVD in our EMR RA cohort

Characteristic	CAD yes, n=335 (7.5%)	CAD no, n=4118 (92.5%)	P-value
Age, years, mean (SD)	72.9 (10.2)	60.0 (14.7)	<0.0001
Female gender, n (%)	207 (5.9)	3316 (94.1)	<0.0001
Male gender, n (%)	128 (13.8)	802 (68.2)	
Race- white, n (%)	265 (79.0)	2714 (91.1)	
Seropositive, n (%)	87 (67.4)	1099 (60.2)	0.10
MTX, n (%)	158 (47.2)	1851 (45.0)	0.45
TNFi, n (%)	96 (28.7)	1189 (28.8)	1.0
Plaquenil, n (%)	101 (30.2)	1200 (29.1)	0.71
CRP mean, median (mg/L)	10.2, 4.2	7.9, 2.0	<0.0001
ESR_mean (mm/hr)	36.5	26.2	<0.0001
Erosions, n (%)	206 (61.5)	2168 (52.6)	0.0021
HTN, n (%)	252 (75.2)	1160 (28.2)	<0.0001
DM, n (%)	108 (32.2)	375 (9.1)	<0.0001
Hyperlipidemia, n (%)	214 (63.9)	817 (19.8)	<0.0001

Clinical characteristics of CVD in our EMR RA cohort

Characteristics	OR (95% CI)
Age	1.06 (1.05, 1.08)
Female gender	0.35 (0.27, 0.46)
HTN	2.64 (1.88, 3.72)
DM	1.64 (1.20, 2.23)
Hyperlipidemia	2.86 (2.10, 3.90)
Ever smoker	2.30 (1.73, 3.04)

Conclusions

EMRs for discovery research

design	Clinical data	DNA	Sample size	cost
clinical trial	+++	+++	+	\$\$\$
registry	++	+++	++	\$\$
claims data	+	n/a	+++	\$
EMR	++	+++	+++	\$

Conclusion: Informatics methods can yield accurate clinical data.

EMRs for discovery research

design	Clinical data	DNA	Sample size	cost
clinical trial	+++	+++	+	\$\$\$
registry	++	+++	++	\$\$
claims data	++	++	++	\$
EMR	++	+++	+++	\$

Conclusion: EMR-based biorepositories for genetic studies yield effect sizes similar to traditional cohorts.

EMRs for discovery research

design	Clinical data	DNA	Sample size	cost
clinical trial	+++	+++	+	\$\$\$
registry	++	+++	++	\$\$
claims data	++	++	++	\$
EMR	++	+++	+++	\$

Conclusion: It should be possible to extend this same framework to a multitude of other phenotypes across multiple institutions, **but**...

Of course, this is not the only way

- This approach will be good for some applications, and not good for others.
- This may serve as an effective way to generate hypotheses.
- There will always be a role for traditional registries.

i2b2 and PGRN acknowledgments



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