

Married 6-year Olds and Other Diseases of Data

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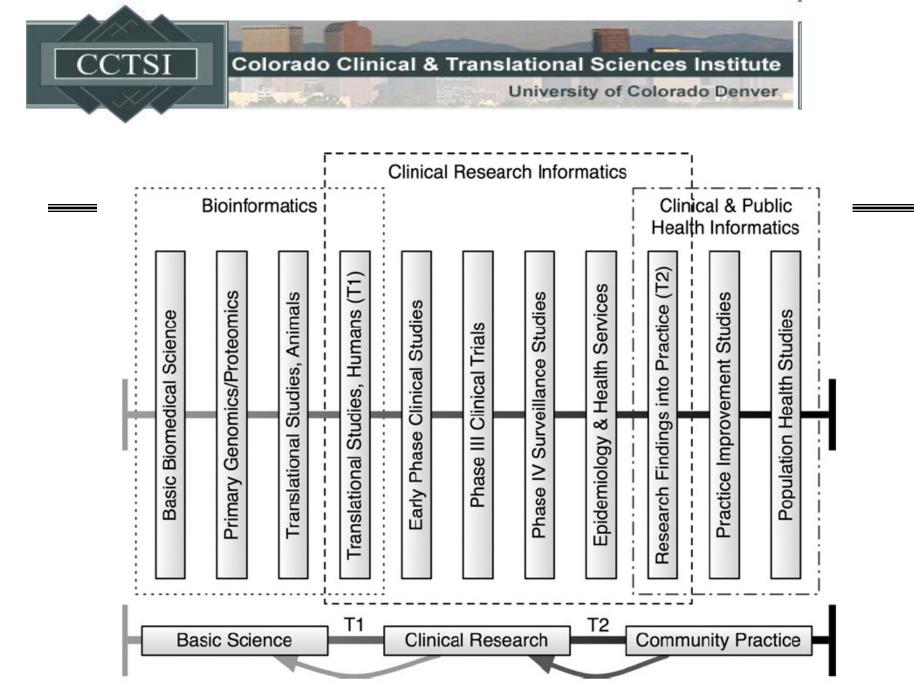
> National Data Integrity Conference Enabling Research :New Challenges & Opportunities 8 May 2015 Michael.Kahn@ucdenver.edu

Funding was provided by a contract from AcademyHealth. Additional support was provided by AHRQ 1R01HS019912-01 (Scalable PArtnering Network for CER: Across Lifespan, Conditions, and Settings), AHRQ 1R01HS019908 (Scalable Architecture for Federated Translational Inquiries Network), and NIH/NCRR Colorado CTSI Grant Number UL1 RR025780 (Colorado Clinical and Translational Sciences Institute).



Guide to the Presentation

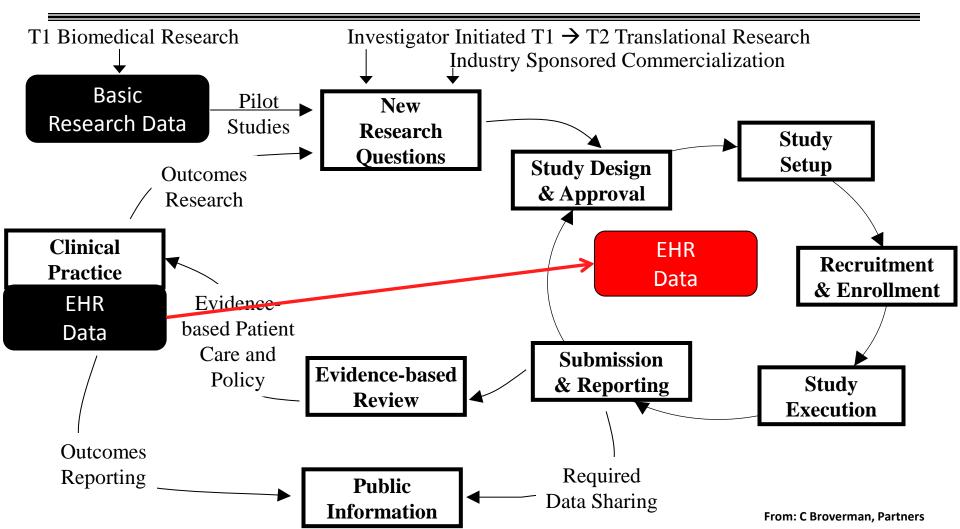
- The fun stuff
 - What is "clinical and translational" data management?
 - The changing landscape of clinical research
 - Learning health systems
 - National data networks
- The grunt work to do the fun stuff
 - Data harmonization
 - Data quality
 - My database can't count



Embi, Payne: J. Am Med Inform Assoc 16(3) 2009

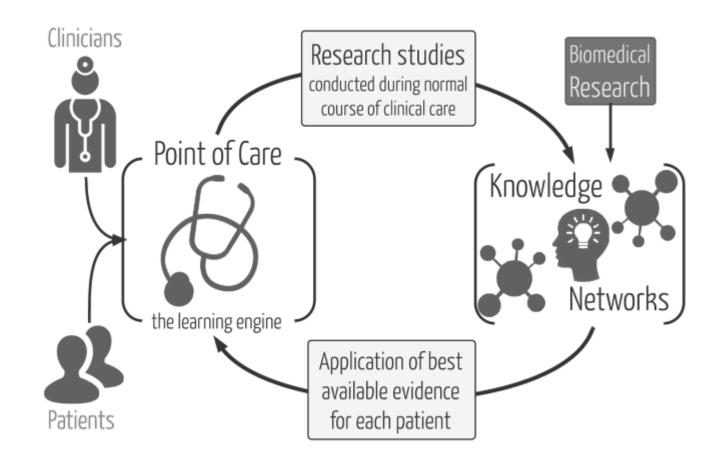


The Changing View of Clinical Research





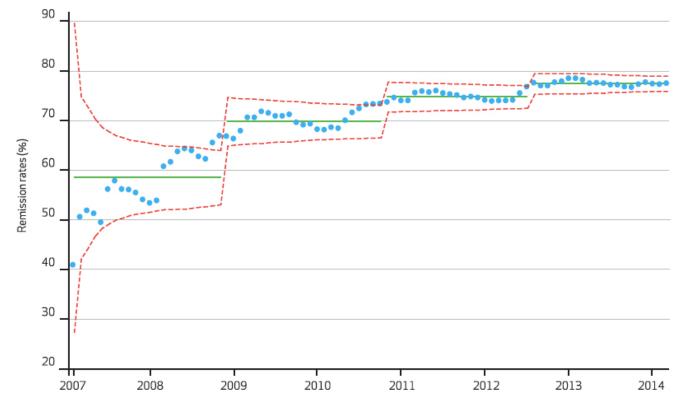
Learning Health Systems: Every patient contributes knowledge





ImproveCareNow: A Pediatric LHS

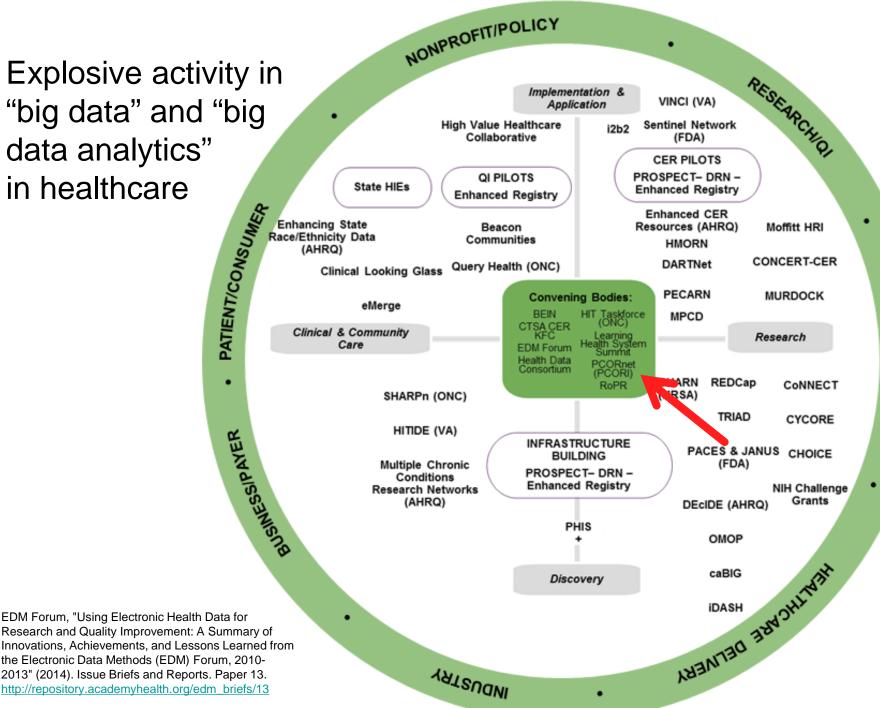




Forrest CB, Margolis P, Seid M, *et al.* PEDSnet: How A Prototype Pediatric Learning Health System Is Being Expanded Into A National Network. *Health Aff* 2014;**33**:1171– 7. doi:10.1377/hlthaff.2014.0127

SOURCE Data are from the ImproveCareNow pediatric inflammatory bowel disease registry for 2007–14. **NOTES** Each blue dot represents the percentage of patients in remission among care centers with more than 75 percent of their patients enrolled in Improve CareNow in a given month. The figure shows the upper and lower confidence limits (dashed red lines in red) and the mean (green solid lines).

Explosive activity in "big data" and "big data analytics" in healthcare



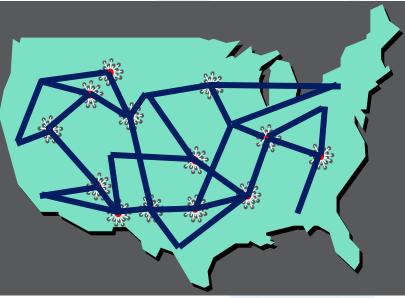
GOVERNMENT

PCORnet: The National Patient-Centered Clinical Research Network

The National Patient-Centered Clinical Research Network

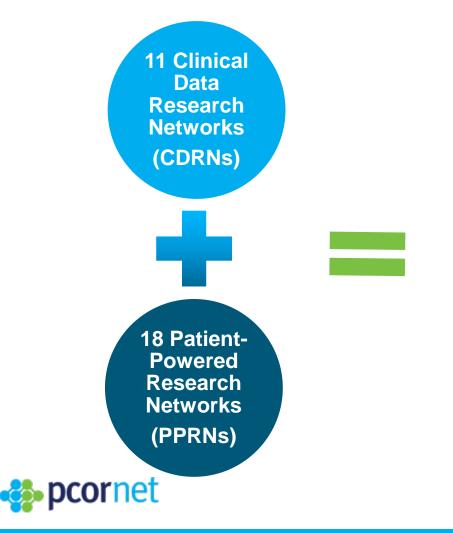
Both researchers and funders now recognize the value in integrating clinical research networks

- Linking existing networks means clinical research can be conducted more effectively
- Ensures that patients, providers, and scientists form true "communities of research"
- Creates "interoperability" networks can share sites and data





PCORnet embodies a "community of research" by uniting systems, patients & clinicians



PCORnet: A national infrastructure for patient-centered clinical research

11 CDRN and 18 PPRN awards



pcornet

This map depicts the number of PCORI-funded Patient-Powered or Clinical Data Research Networks that have coverage in each state.



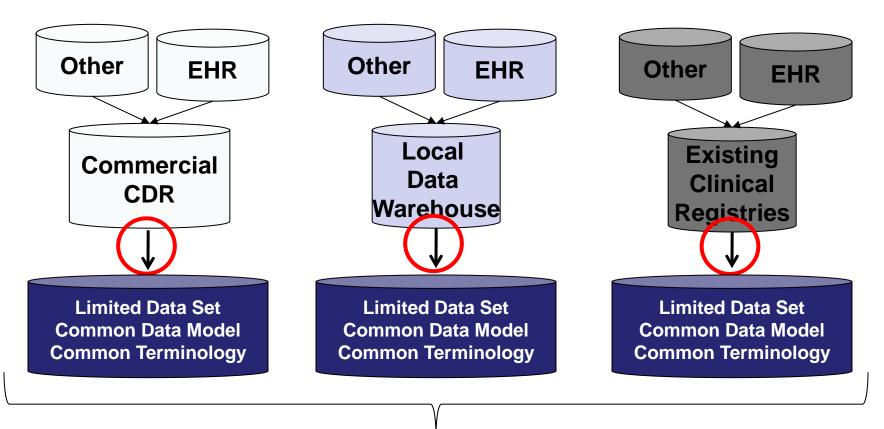


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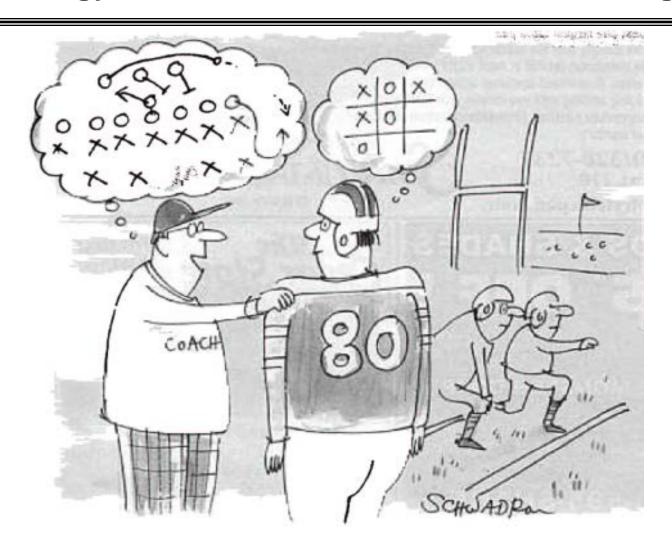
Harmonizing data into a common structure



Common Query Interface



Terminology Harmonization – What are we talking about?



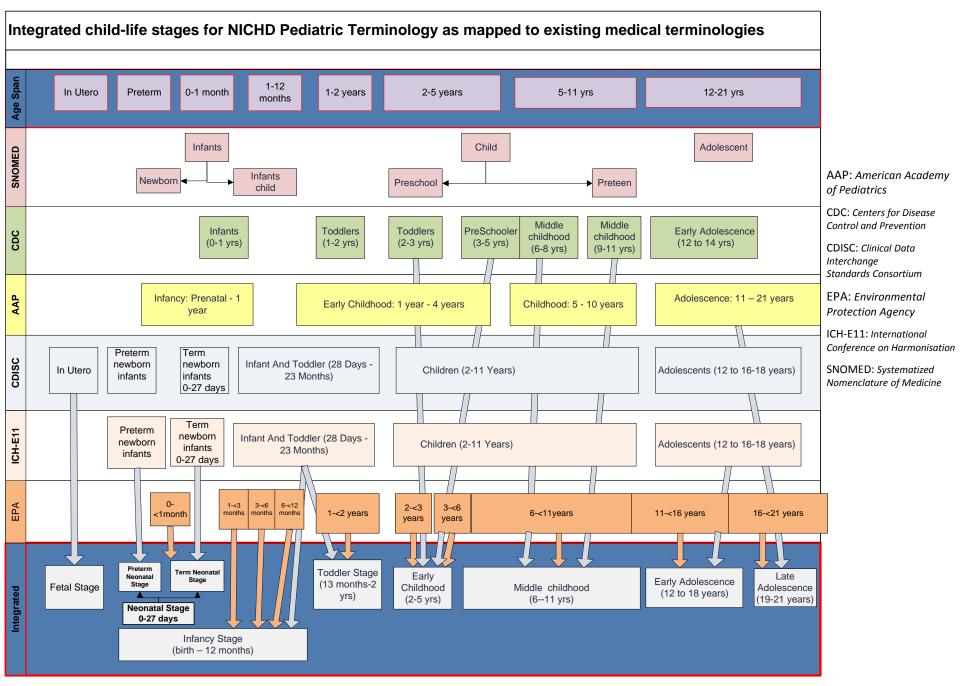
Examples of Variations in Platelet (Quantitative) Result Units in Source Data

Platelet count original result units[‡]

Blank	FL	TH/UL	X10(3)
%	K/CMM	THOU/CMM	1000/UL
/100 W	k/cmm	thou/cmm	X10(3)/MCL
/CMM	K/CU MM	thou/mm3	X10(3)/UL
CMM	K/CUMM	THOU/UL	X10(6)/MCL
10 3L	K/MCL	THOUS/CU.MM	X10*9/L
10X3UL	K/mcL	THOUS/MCL	X10E3/UL
10^3/UL	K/UL	THOU/mcL	X1000
10*3/uL	k/uL	THOUS/UL	X10X3
10?3/uL	KU/L	Thou/uL	X10^3/UL
10E3/uL	K/MM3	THOUSA	x10
10e3/uL	K/mm3	THOUSAND	X10?3/ul
10e9/L	LB	THOUSAND/UL	X10E3/UL
E9/L	PLATELET CO	U	X10E3
BIL/L	T/CMM	X 10-3/UL	K/A?L
bil/L	TH/MM3	X 10(3)/UL	K/B5L
CU MM	th/mm3	X10 3	

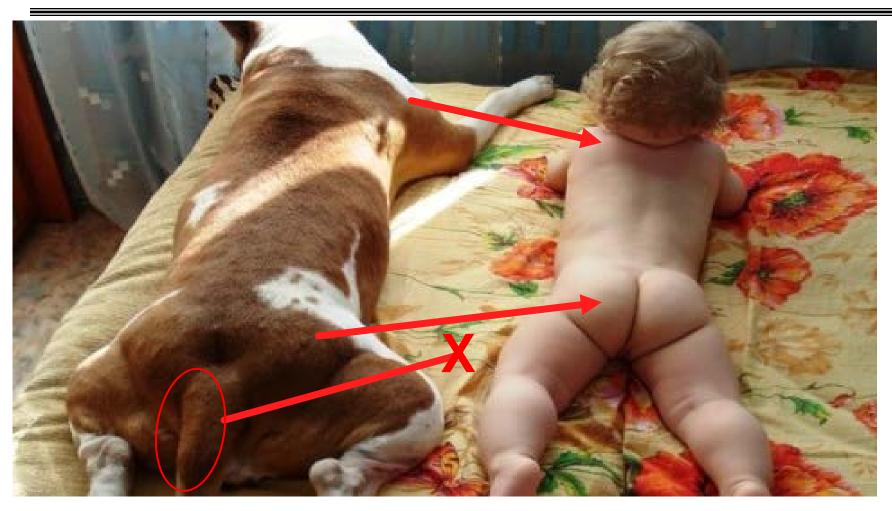
Examples of Variations in (Qualitative) **Pregnancy Result Units in Source** Data (aka, how many ways can you spell negative?)

NEGATIVE	
PUSITIVE	820
UNDETERMINED	840
BORDERLINE	1615
ROPDERLI	ABNORMAL
NEG	BOARDERL
NONE DET	BODERLIN
POS	CANCELLE
COMMENT:	DUPLICAT
160.8	EQIVOCAL
0.5	EQUIVOCA
1.2	HIRABAYA
1000	NE-CHECK
122	NEAGTIVE
14	NEG (-)
140	NEGA
15	NEGA T I
2	NEGA TIV
2	NEGAT IV
2.1	NEGATAIV
203	NEGATIAV
252.3	NEGATIBE
278	NEGATIE
28	NEGATRIV
3178.2	NEGATTVE
345	NEGATVIE
38.1	NEGAVTIV
400	NEGITIVE
5 Int	NEGTIVE
5272.4	NETGATIV
642.2	NORM
670	NORMAL
697.7	POA
DETECTED	POPSITIV
INDETERM	POSIITIV
N	POSITIFV
NOT DETE	POSITTVE
Neg	POSITVE
Negative	POSOTIVE
Negatvie	POSTIVE
P	PSOITIVE
Positive	REPEAT
SPRCS	STAT
	URINE
n	
neg	KAISER PERMANENTE
negative	





Aligning Terminologies





SNOMED CT: A "mandated" clinical standard

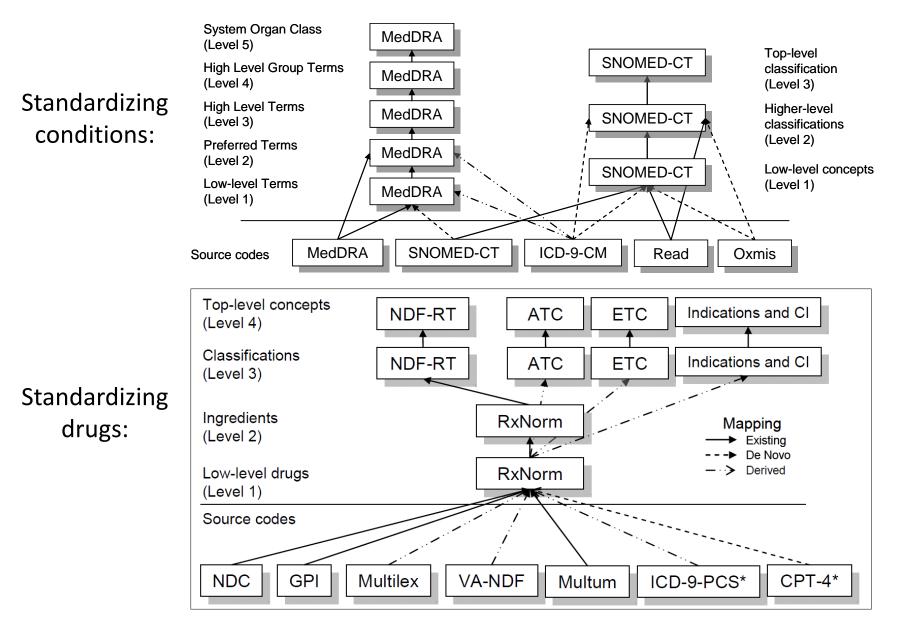
- Sign and symptoms of attention deficit hyperactivity disorder Attention deficit hyperactivity ADHD disorder (disorder) attention deficit hyperactivity ADD loss of scalp hair **Diffuse loss of** scalp hair loss scalp hair (finding)
 - Epistaxis
 - nosebleeds

Epistaxis (disorder)

From: Rachel Richesson PhD

OBSERVATIONAL MEDICAL OUTCOMES PARTNERSHIP

Standardizing terminologies to accommodate disparate observational data sources





Data Quality in Electronic Health Records

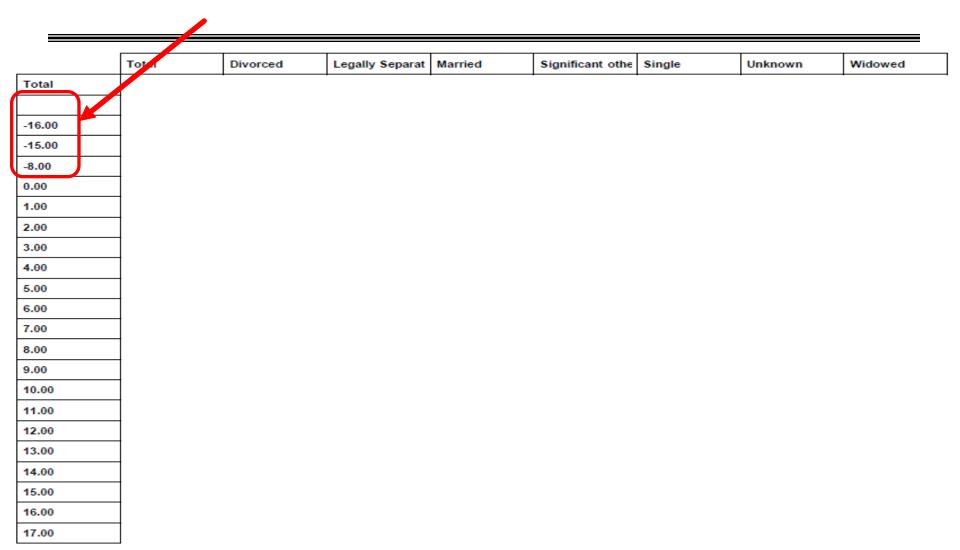
- Data collection tools optimized for efficiency
 - Text templates
 - Copy/paste
- Minimal data validation checks
 - Min/Max limits
 - Pick lists
 - Required fields



Even "simple" stuff has problems



A trivial example: Martial Status by Age





A trivial example: Martial Status by Age It's tough being 6 years old in Denver..... Would these results be worrisome?

	Total	Divorced	Legally Separat	Married	Significant othe	Single	Unknown	Widowed
Total	423,508	33	3	1,606	81	420,944	830	11
	70	0	0	0	0	70	0	0
-16.00	2		1.10	1000		2	0	0
-15.00	1		A CARCEN	NO TO		1	0	0
-8.00	1	1		Carl Land De La Carl Carl	A DEST	1	0	0
0.00	768		alley and	a start	N R EE	768	0	0
1.00	13,660		That A 11	Part I	Lea Cal	652	5	0
2.00	21,350	6		and and	COM ST .	290	25	0
3.00	24,960		ARRAY CO		T COL	885	31	0
4.00	27,861		C.B.L	This day		806	32	0
5.00	29,933		And a		- Aller	889	24	0
6.00	30,932		and the second	AN	and a	810	40	0
7.00	27,381			1-1	140	268	46	0
8.00	24,198		A contract of the second	The second second	4	124	31	0
9.00	22,522					448	35	0
10.00	20,283				Participa_	231	22	0
11.00	18,705				Contra Contra	659	16	0
12.00	17,340				A	296	19	0
13.00	16,510				all and a	470	17	0
14.00	15,792		and the			761	15	0
15.00	15,354					302	21	0
16.00	15,474	2	0	19	1	15,439	13	0
17.00	15,208	1	0	9	0	15,181	17	0



Should we be worried?

- No
 - Large numbers will swamp out effect of anomalous data or use trimmed data
 - Simulation techniques are insensitive to small errors
- Yes
 - Observed site variation may be driven by differences in data quality, not clinical practices
 - Genomic associations look for small signals (small differences in risks) amongst populations

"Big Data" and "Big Data Analytics"

Data Scientist: The Sexiest people of the 21st Century



Meet the people who can coax treasure out of messy, unstructured data. by Thomas H. Davenport and D.J. Patil



rived for work in June 2006 at LinkedIn, the business networking site, the place still felt like a start-up. The company had just under 8 million accounts, and the number was growing quickly as existing members invited their friends and colleagues to join. But users weren't

hen Jonathan Goldman ar-

seeking out connections with the people who were already on the site at the rate executives had expected. Something was apparently missing in the social experience. As one LinkedIn manager put it, "It was like arriving at a conference reception and realizing you don't know anyone. So you just stand in the corner sipping your drink—and you probably leave early."

OBSERVATIONAL MEDICAL OUTCOMES PARTNERSHIP

Cough

Respiratory, thoracic and mediastinal disorders

Chest pain

Acute upper

respiratory infection

Exploring prevalence of all diseases

Treemap displays 3 dimensions: **Size of rectangle** : Standardized prevalence **Color** : 'Prevalence ratio', comparing source prevalence with the overall community average **Hierarchy** : Aggregate SNOMED-CT concepts by MedDRA System Organ Class

					Obes	sity Ed		EUDRA	4 SYSL	emo	rgan	CIdSS	
Acute pharyngitis	Acute bronchitis	Allergic rhinitis	Dyspnea	Diabetes mellitus type 2	Type I diabete mellitu - poor Swellin	defici s enc s Dehy	Abnor Hypo mal kalem gluc ia Morbid Pu obesity re	Long-term dr therapy Psy	rug ned immuni zatio ychiatric disore	mu eli ders Attentio		Laboratory procedure subcutaneous tis. Verru	xamin Anten atal Blood hemis Abnor mal Po- ve Nervous system dis c
	Chronic obstructiv Brone	chitis Chronic		Va	of limb	n	Abnor Hy	Tobacco dependence	Anxiety	n deficit hypera ctivity	Eruption	n Acne a vulga s	ri Headache
Acute sinusitis	e lung disease Zing Pneumon	maxill gr Exac Acut				Corona arterioscle is		syndrome	Insom Chil nia Alco hol.	er e	sm of uncer	Sebo Epid Sen rrheic ermo e der id hy Disord Ingro Urti er of wi aria	Skin sensation Seizure
Asthma	ia Allerg c rhin Streptoco ccal sore media	Hype Ac rso	Pl P	Essential hypertension		Migraine	Perip Old heral myoc vas ard	Depressive disorder	Dysthy Bip	Psy Alt R cho er ec Adj Cli Si ust nic ng	due t Actinic keratosi	Benig n ne Cellulit s an De Di.	
Musculosk	throat ted xe Co Ac co a Musculoskeletal and connective tissue disorders					e Low Acu bl te	Reproductive	system an qiniti Female	Renal and u	inary dis	Cardiac disord		
Pain in limb	Arthralgia of the lower leg	Neck pain	Shoulder joint pain	Benign essenti hypertension		Chro di nic gi Preinf arct	n Tr Be	lump vulv Benign Irregu	u Dys Prim	K	ase dne Findi	tion disorde r of t Conges Mitral tive valv	
	of the sa	Icral Osteop	Nonal Sprai lopat n of	Gastro	intestinal	disorders	, Nausea	Excess Diso	Uter Pa	st	y ngof one fre cu Uri Bl	heart failure Card	A Fever Ear and Ia Social c
Low back pain	and/or f Osteoarth	on mall Joint M athi pain o	hic ankle Nonall Cervic opat al s	Abdominal pain	Diarrhe	a Consti on	pati and	Cyst Leuk orr Eye dis	m Fib	a syndr ^{te} Infections an	r nar oo d infesta	Routine	Impacted Dy cerumen of the state of the stat
LOW DACK Pain	Arthralgia Cer	spla sprain m Rhe Joir	Spas Locali m zed, nt Deg Lum		Vomitin g	Nausea	Gastr Right o- lower eso qu	y Myopia	Presb yopia Visual Acute	Otitis media	Viral disease		Otalgia Cti on Menop njury, Neoplasms
Backache	Disorder of bone Art	thy Han Art d j hra hral Gou Cur	Sp E	reflux disease	Divertic ular disease of colon	Epigastri c pain Right upper	Gene Left ralize lower d a qu De Es Ga ntal op Ac	d	visual Acute listurb conju ance ncti Pri Te Gla na ar y fil Op	Acute suppurati ve otitis	fectiv Chr otit oni tes Acu	Ly m Anemia ph ad	Acciden Benign Pla neoplas Pla Injury of Immu He All Ch Di C open erg se hr

Regenst

Нуро

Metabolism and nutrition disorders

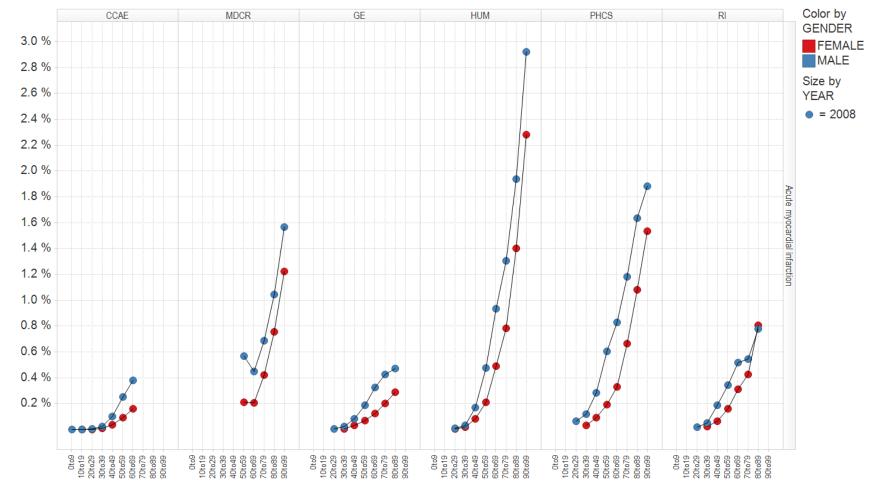
Hyperlipidemia

Pure

hypercholester

olemia

Exploring prevalence of disease with standardize databases: **OBSERVATIONAL** MEDICAL **O**UTCOMES ex: Acute Myocardial Infarction **P**ARTNERSHIP



PERSON_PCT

AGE GROUP



The Tale of A Trivial Data Request

• The original data request:

"For an upcoming grant application, how many patients were seen recently with neurofibromatosis-1 (NF-1) and scoliosis?"



The Tale of A Trivial Data Query

- Getting more specificity:
 - "Recently seen" = an encounter of any type since
 1/1/2012
 - NF-1: ICD-9 code starts with "237.7"
 - Scoliosis: ICD-9 code starts with "737.3"



The Tale of A Simple Data Query

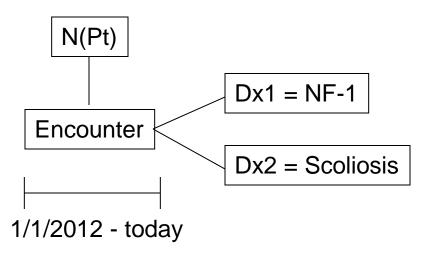
• First query result: N = 15

Clinical investigator did not believe this result even though we used her definitions.



The Tale of A Simple Data Query

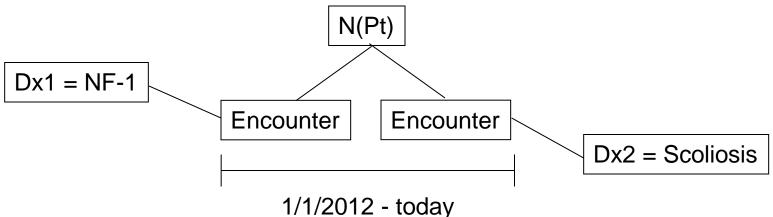
- Drilling down:
 - This query required both diagnoses to be coded on the same encounter (event).





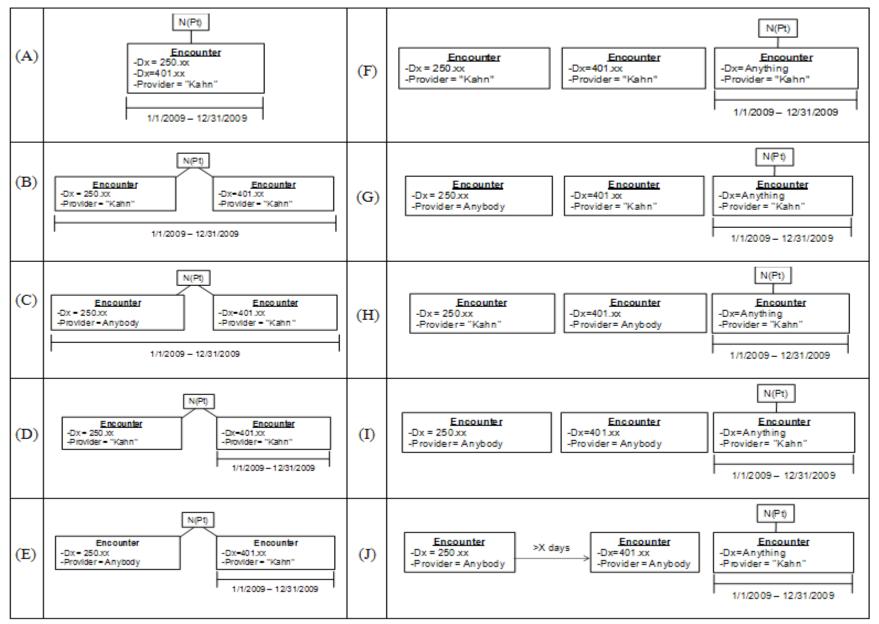
The Tale of A Simple Data Query

- Second query:
 - NF-1 and Scoliosis diagnoses can be coded on different encounters, both within time window
 - N= 28



Investigator still did not like the answer!

Table 1: Ten graphical diagrams representing the question: "How many ambulatory patients did I ("Provider = Kahn") see with diabetes mellitus (ICD-9 = 250.xx) and essential hypertension (ICD-9 = 401.xx) between January 1, 2009 and December 31, 2009?" Each diagram, when converted into a database query, returns a different result. N(Pt) = number of patients.





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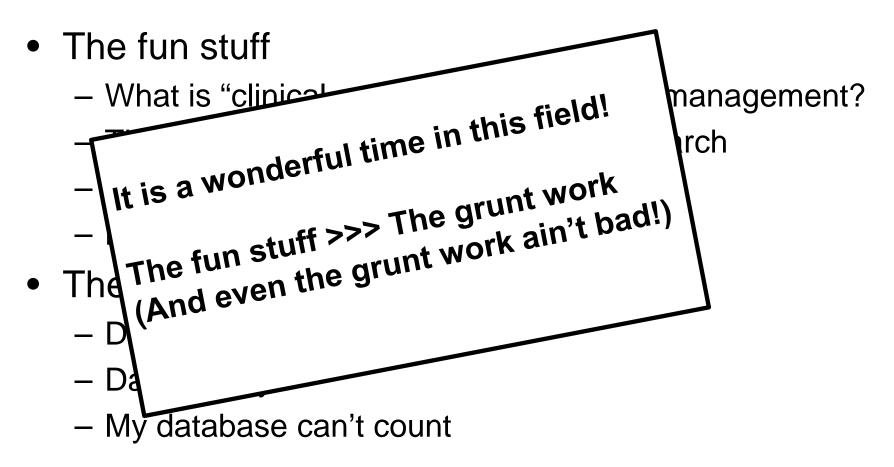
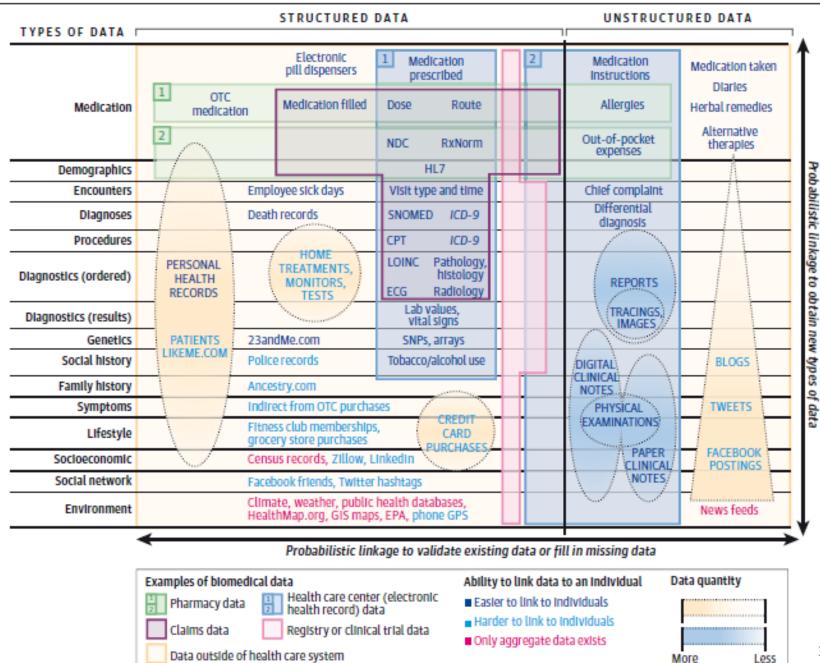
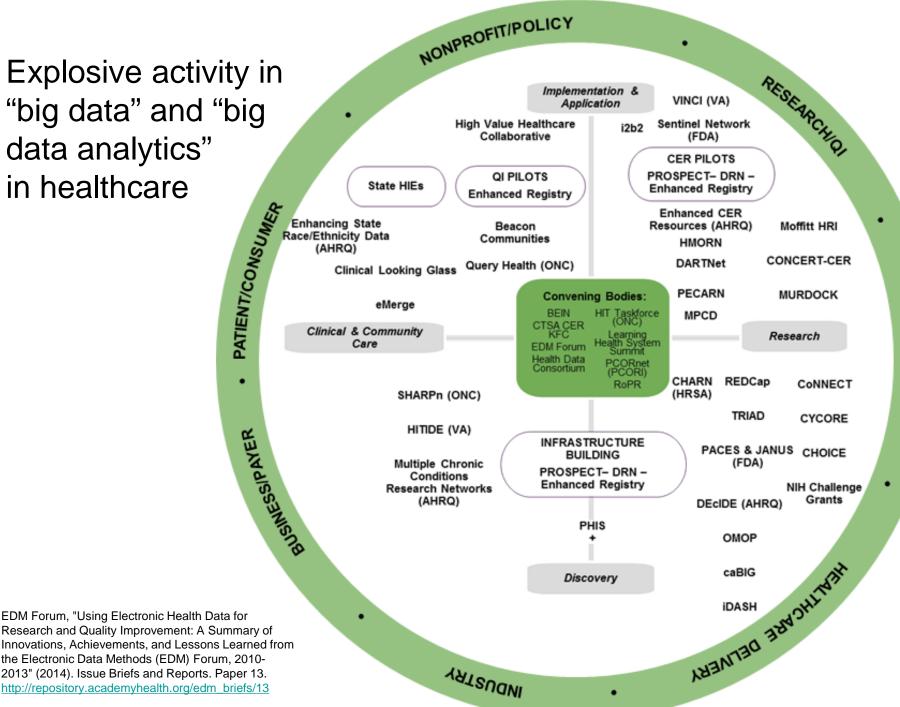


Figure. The Tapestry of Potentially High-Value Information Sources That May be Linked to an Individual for Use in Health Care



Weber, G. M., Mandl, K. D. & Kohane, I. S. Finding the missing link for big biomedical data. JAMA 311, 2479–2480 (2014).

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