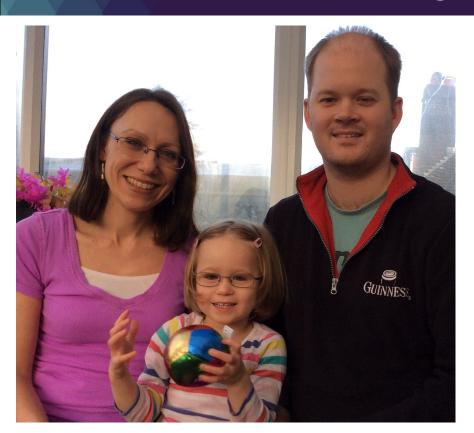
Bringing Phenotype to the Genotype: Semantics for Maximizing Disease Discovery

Melissa Haendel, PhD March 31, 2020





Jessica's story: Evidence-based diagnosis requires evidence

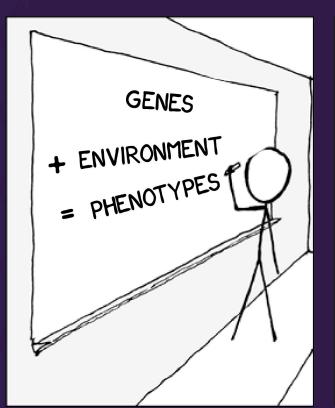


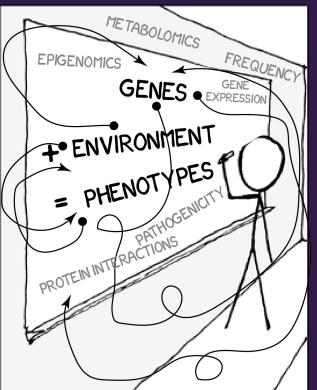
Jessica (aged 4) has a rare condition which causes epilepsy, affects her movement and developmental delay. Standard genetics tests negative.

To solve her case requires the ability to compare Jessica to a multitude of other available data, both from humans and from other animals.

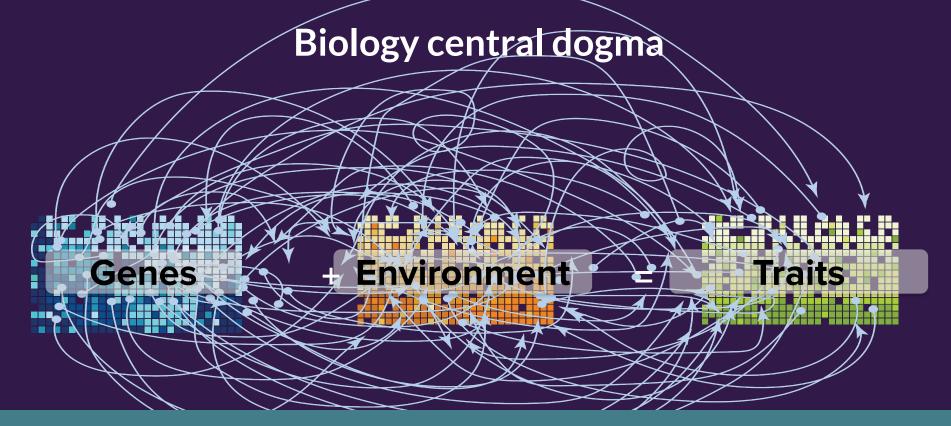
Genomics

Biology central dogma





ADAPTED FROM xkcd.com/295



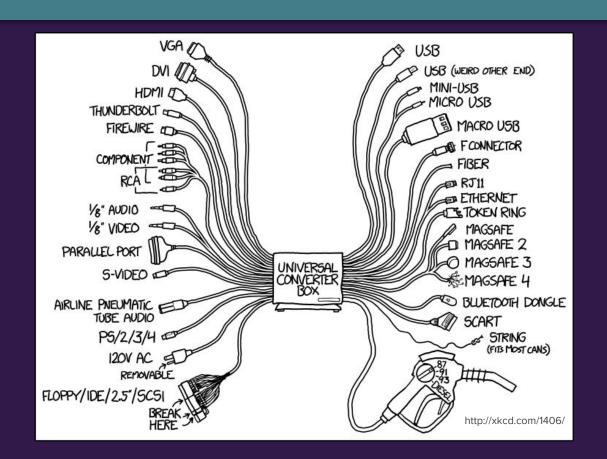
Standards for encoding and exchanging data must be up to these challenges.

It is not just the bits... G-G (kind of) E-P G-P or D (disease) regulates contributes to (E->P) causes negatively regulates (inhibits) influences (E->P) contributes to positively regulates (activates)exacerbates (E->P) is risk factor for manifest in (P->E) directly regulates protects against interacts with correlates with co-localizes with is marker for G-E (kind of) co-expressed with modulates expressed in involved in expressed during

The relationships too must be captured

correlates with hallmark of (P->D)

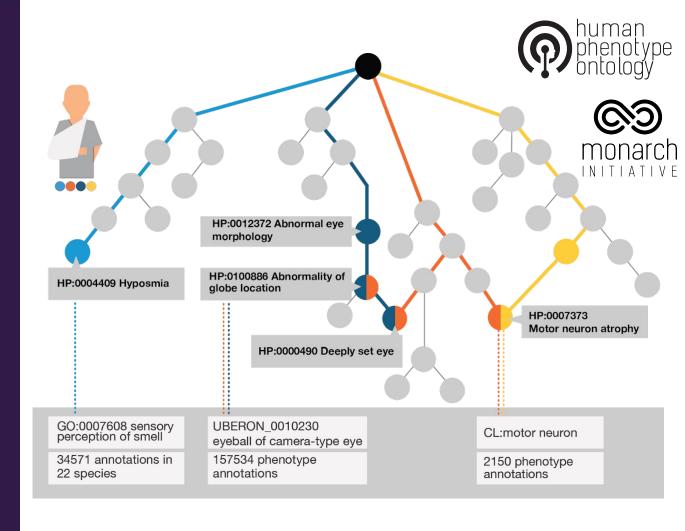
Semantics are the ultimate universal converter



Human Phenotype Ontology (HPO)

A standardized, machine readable vocabulary of phenotypic abnormalities encountered in human disease.

- Over 14,000 phenotype terms
- It is used to create computational models of disease



hpo.jax.org

Fuzzy Phenotype Matching

Not same variant, but same disease and gene, KMT2A.

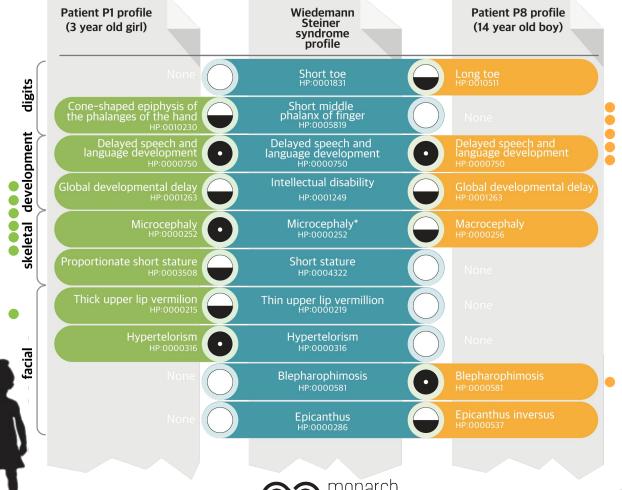
DOI: 10.1126/scitransImed.3009262

Legend

Perfect Match

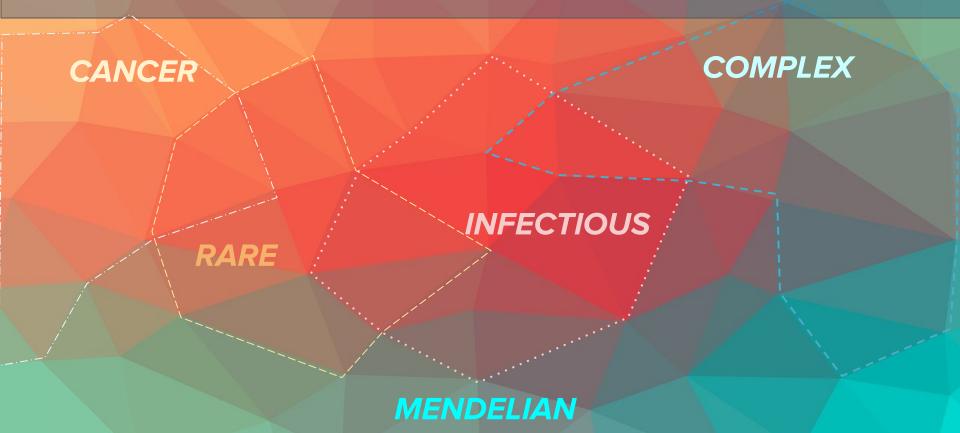
Fuzzy Match

No Match





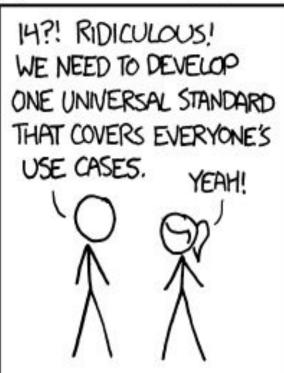
What is the most clinically useful way to define and group diseases?



Standards proliferation: how do you know you need a new one?

HOW STANDARDS PROLIFERATE: (SEE: A/C CHARGERS, CHARACTER ENCODINGS, INSTANT MESSAGING, ETC.)

SITUATION: THERE ARE 14 COMPETING STANDARDS.



500N:

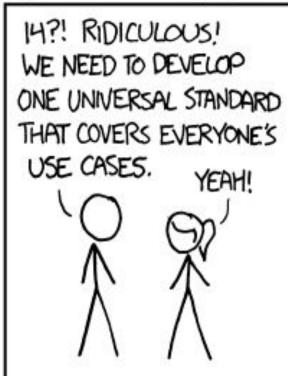
SITUATION: THERE ARE 15 COMPETING STANDARDS

xkcd.com/927

Standards proliferation: how do you know you need a new one?

HOW STANDARDS PROLIFERATE:
(SEE: A/C CHARGERS, CHARACTER ENCODINGS, INSTANT MESSAGING, ETC.)

SITUATION: THERE ARE 14 COMPETING STANDARDS.



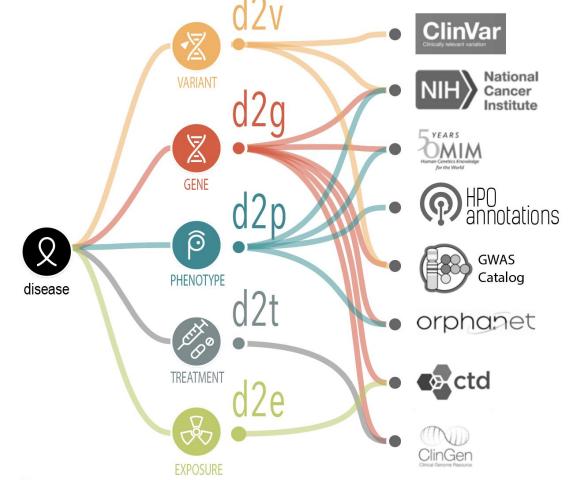


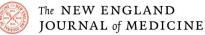
SITUATION: THERE ARE 15 COMPETING STANDARDS.



SITUATION:

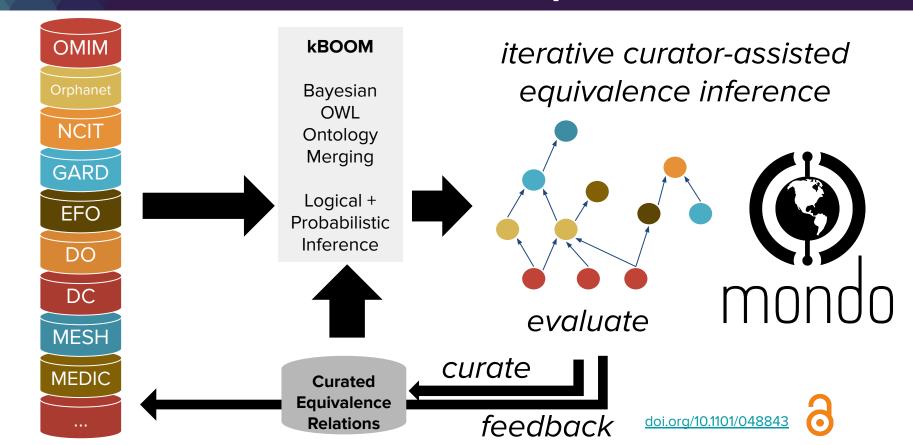
THERE ARE 15*14=210 SETS OF MAPPINGS. Different communities define different aspects of diseases differently



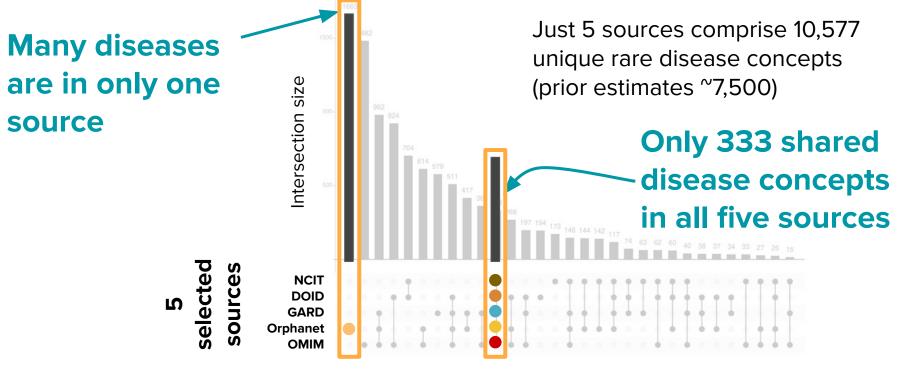


Classification, Ontology, and Precision Medicine

Evidence-based merging of equivalent disease concepts



If rare diseases are not counted, rare disease patients will not count

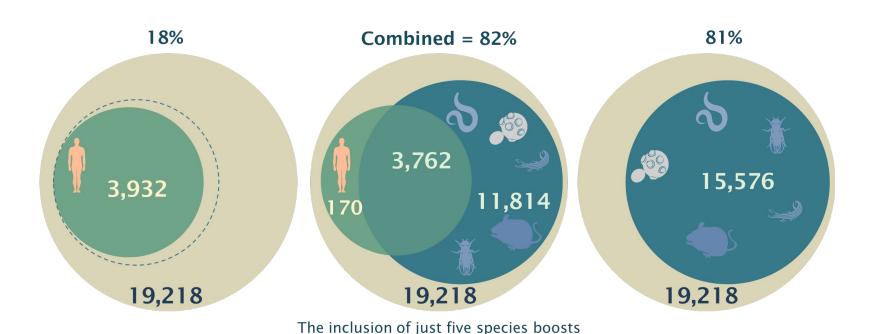






Why model organisms matter to patients

More species = more coverage



phenotypic coverage of genes by 64%

Fuzzy matching across species improves diagnostics



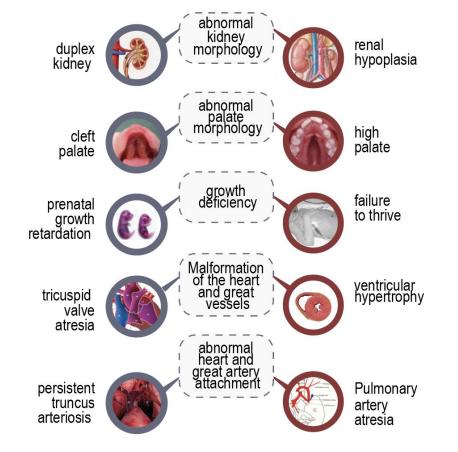


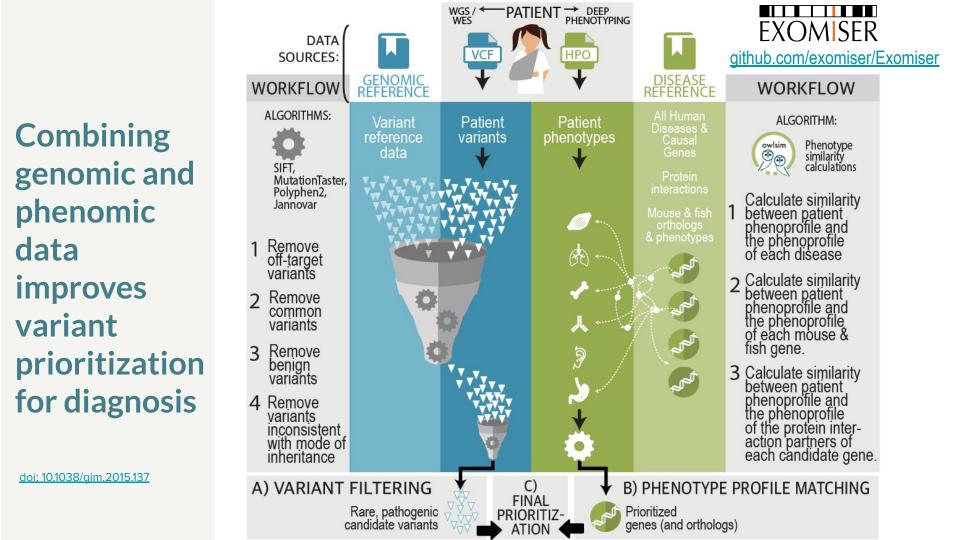


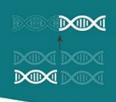












6,414,934 variants in Jessica's genome

677,556 are rare

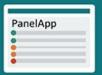




2,826 predicted to cause change in a protein

67 different to her parents





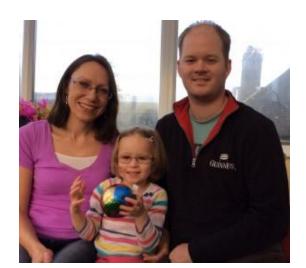
was in a gene listed in PanelApp

Jessica

- Jessica (age 4) has a rare condition which causes epilepsy, affects her movement and developmental delay. Standard genetics tests negative.
- De novo deletion in SLC2A1 identified as the cause of her Glut 1 deficiency syndrome
- Exomiser ranked this variant first
- Now being successfully treated with a ketogenic, low-carb diet
- Low risk for future pregnancies



Exomiser ranking 94% in top 3 candidates using human + model organism data



Phenotyping is not free ... or easy; So how much is enough?



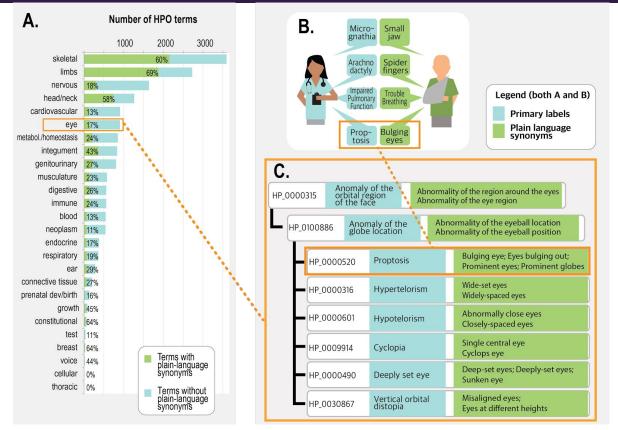


- The more phenotype data we have, the better able we are in answering that question
- We can help inform users whether their phenotyping is sufficient for analysis, and which new phenotypes to examine
- We need to improve case-level information sharing to better understand the heterogeneity of presentation and progression





Plain language synonyms for patients to use

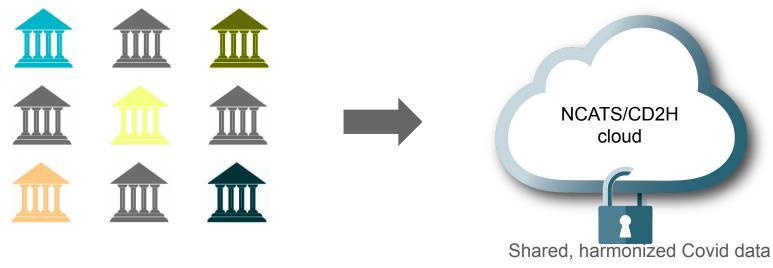


4887 of 13823 HPO terms have lay synonyms



National Covid-19 data and analytics platform





Centralized model advantages

- Large dataset
- Consistency
- Improved machine learning applications & analytics over patient-level data
- Shared compute infrastructure and application deployment
- Purpose-driven curation/data modeling for covid-19





National Covid-19 data and analytics platform



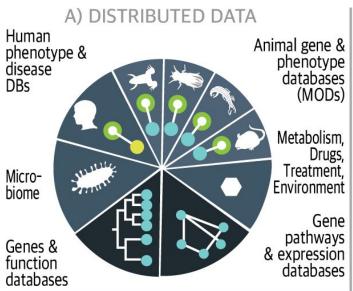
Qualified Researcher, e.g. **Data Contributors CTSA Institutions Data Access Committee** Data access request Authorized access Data contribution Created by shashank singh from Noun Project Data Harmonization & Transfer Method **Central Data Pool** Data access registration **Synthetic Data**

Public Data Scientist

How we can all help improve disease diagnosis and care

- We need deeper phenotyping of a patient's condition beyond billing
- Ontologies can support discovery by aiding data integration and analytics across domains
- Use non-human data to support diagnostics, drug discovery, and mechanism discovery
- Use semantics to combine genomics with phenomics (and other data)!
- Richer and more widespread sharing of patient-level data is needed to fully understand disease heterogeneity
- There is an urgent need to improve data collection and flow between all stakeholders
 across full lifecycle of disease care -- from patients and families themselves through
 clinicians and hospitalists => this is true in all fields !!

Think about downstream data reuse



B) AGGREGATED DATA

THE THE PARTY OF T

INTEGRATED DATA

We learn different things from different species and different data sources. However, building a bigpicture view is non-trivial.

DISEASE
PHENOTYPE
GENE

Often in the aggregation process, many of the the original connections are lost. Moreover, it is difficult to make new connections without a common conceptual model.

Computable phenotypes
Exact matches
Fuzzy matches
IIII Inference

PMC4981258 NIH: R24 OD011883

Thank you

Chris Mungall Peter Robinson Ken Gersing Chris Chute Julie McMurry **Monica Munoz-Torres Matt Brush** Jules Jacobsen Nico Matentzoglu Nicole Vasilevsky Kent Shefchek Tom Conlin Nomi Harris Marcin Joachimiak Seth Carbon Justin Reese Deepak Unni Sebastian Koeller

Tudor Groza



Monarch - Linking diseases to model organism resources 5 R24 OD011883 Biomedical Data Translator: OT3 TR002019-01S2 Center for Data To Health (CD2H) U24TR002306