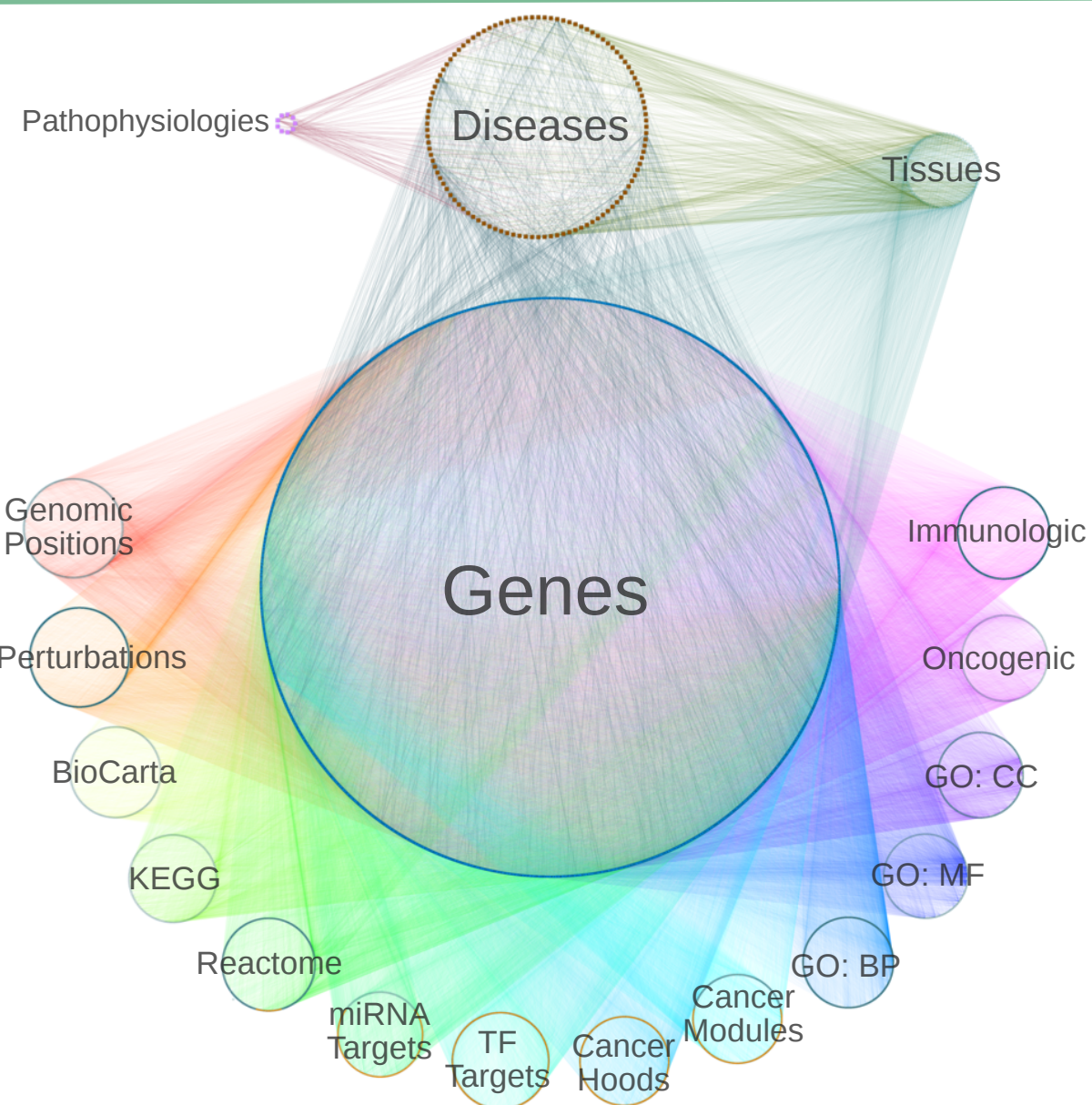


Daniel Himmelstein, Leo Brueggeman & Sergio Baranzini present *Repurposing drugs on a heterogeneous network*

Last year, we introduced *heterogeneous network edge prediction* (HNEP) to predict disease-associated genes.

Heterogeneous networks contain multiple node and edge types.

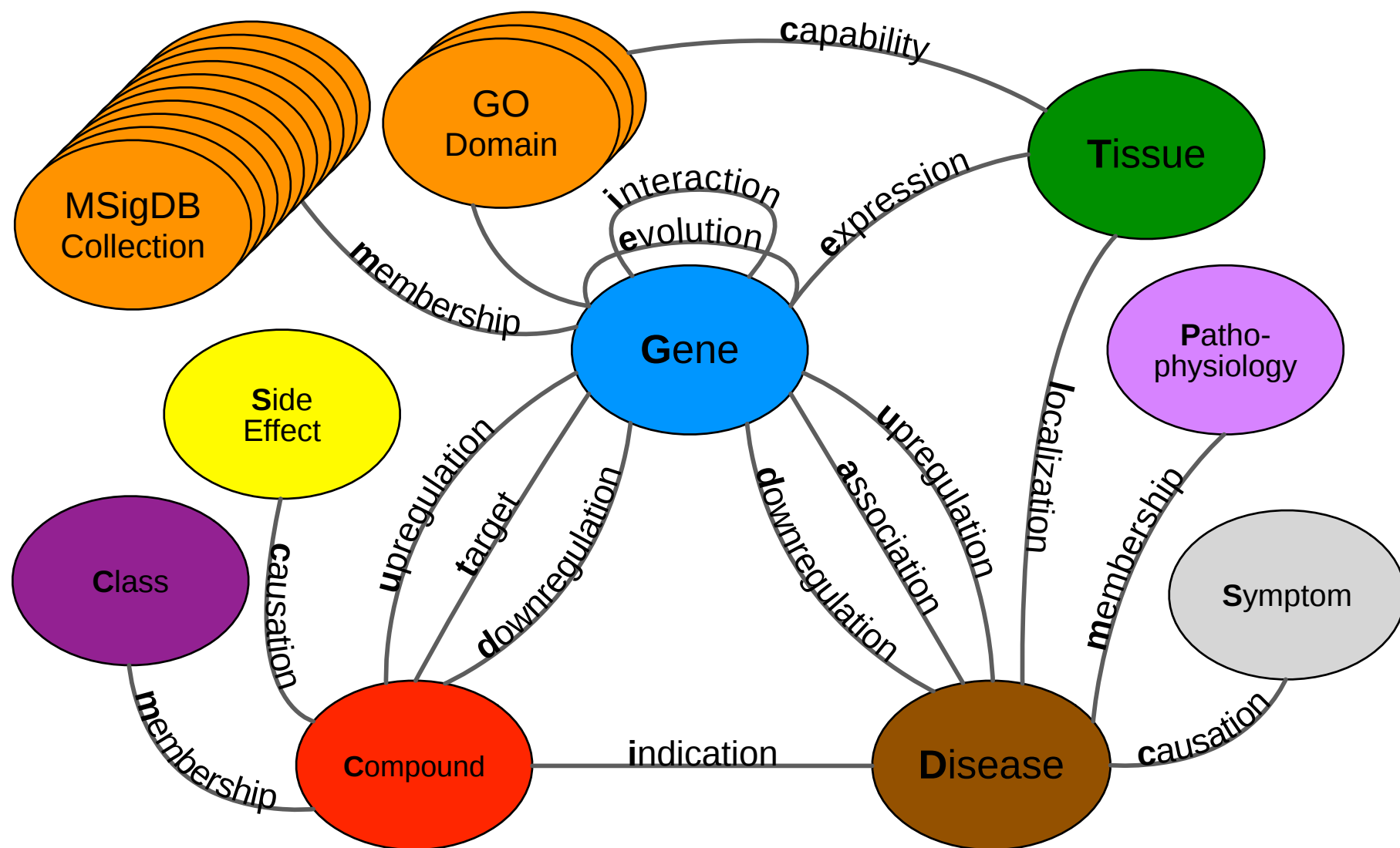
Our network contained 40,343 nodes (of 18 types) and 1,608,168 edges (of 19 types).



Now in 2015, we will use this data integration approach to repurpose drugs on a heterogeneous network.

Planning the Network Construction

Metagraph



Nodes

Type	Resource
Compound	DrugBank
Disease	Disease Ontology
Gene	Entrez Gene
Tissue	Uberon
Gene Set	MSigDB
Side Effect	UMLS
Pathophysiologies	Manual
Symptom	MeSH

Standardized terminologies:

- provide a scalable framework for data integration
- prevent redundancy
- enable semantic data

Edges

Source	Target	Type	Resource
Compound	Disease	Indication	MEDI
Compound	Disease	Indication	LabeledIn
Compound	Gene	Expression	LINCS
Compound	Side Effect	Causation	SIDER 2
Compound	Side Effect	Causation	OFFSIDES
Disease	Gene	Target	ChEMBL
Disease	Gene	Association	GWAS Catalog
Disease	Gene	Expression	STAR-GEO
Disease	Pathophysiologies	Membership	Manual
Disease	Symptom	Causation	Human symptoms--disease network
Gene	Gene	Interaction	Human Interactome Project
Gene	Gene	Interaction	The Incomplete Interactome
Gene	Gene	Evolution	Evolutionary Rate Covariation
Gene	Gene Set	Membership	MSigDB
Gene	Tissue	Expression	GNF Gene Expression Atlas

Ideal resources are:

- high-throughput
- systematic
- unbiased
- aggregately diverse

And you can follow in realtime and get paid to participate.

ThinkLab

thinklab.com

doi:10.15363/thinklab.4

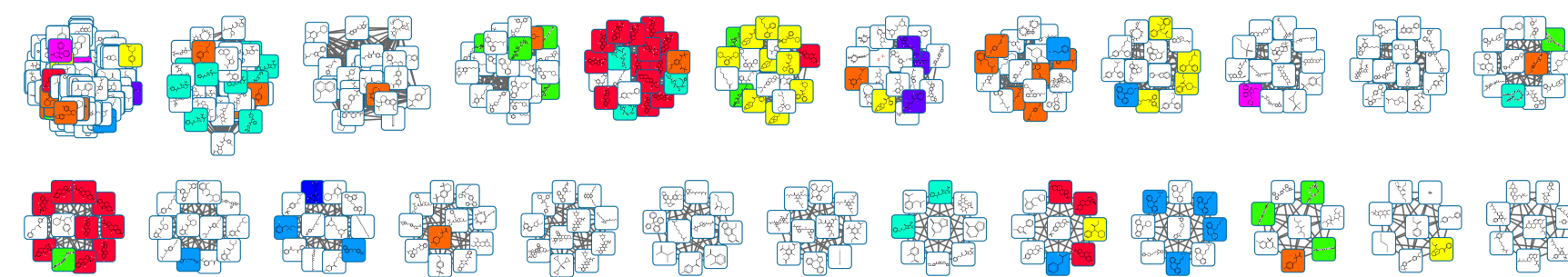


ThinkLab is:

- massively collaborative — all are welcome
- open science — all content is CC-BY
- incentivized — contributions are rewarded
- productive — scientific markdown editor
- efficient — code and results public upon commit

Results (as of March 2015)

We analyzed **SIDER 2** and investigated its strengths and weaknesses as well as pharmacological utility.



Side-effect similarity modules were concordant with structural similarity modules (colored).

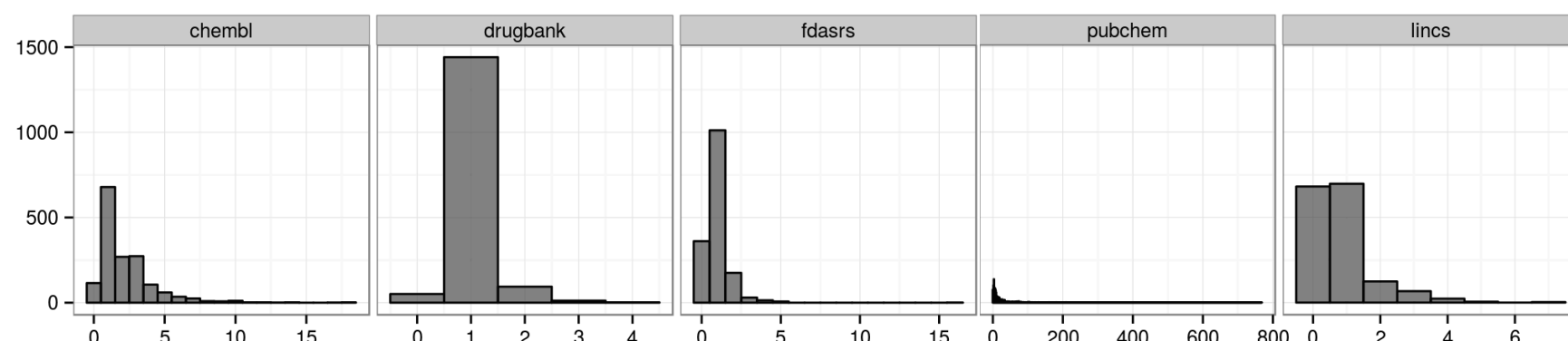
git.dhimmel.com/SIDER2

We created a user-friendly service to retrieve **Gene Ontology annotations** with optional propagation.

Propagated	Unpropagated
Entrez	Symbol
All Genes	Protein-coding Genes

git.dhimmel.com/gene-ontology

We mapped compound vocabularies to DrugBank using **UniChem** to enable fuzzy matching.



Number of matches to each approved small molecule in DrugBank

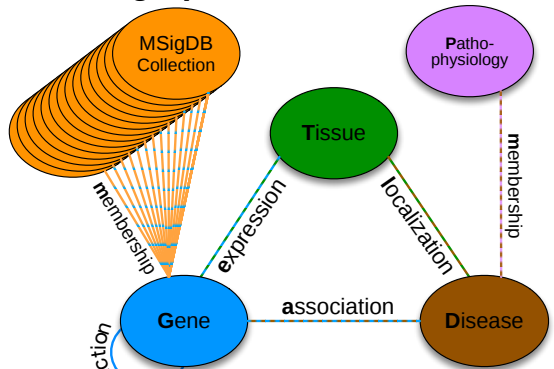
git.dhimmel.com/drugbank/unichem-map.html

Acknowledgements

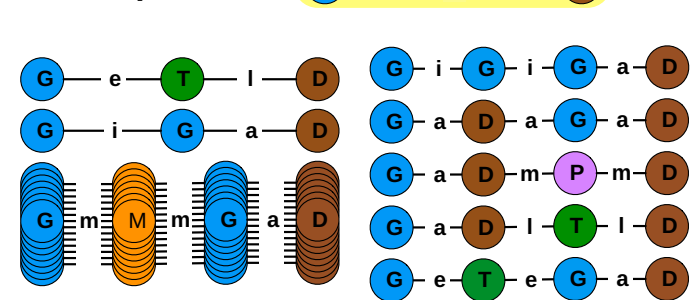
We would like to thank our ThinkLab contributors (thinklab.com/p/repheto/leaderboard) and Alex Pico for the SIDER visualization. This material is based upon work supported by the National Science Foundation under Grant No. 1144247 to DSH. SEB is a Harry Weaver Neuroscience fellow from the National Multiple Sclerosis Society.

HNEP Method

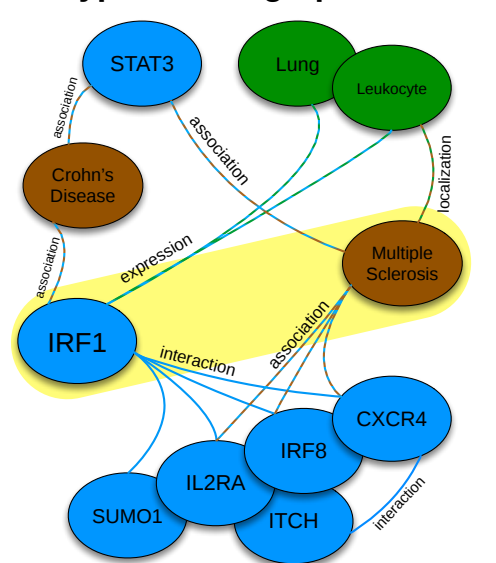
A. Metagraph:



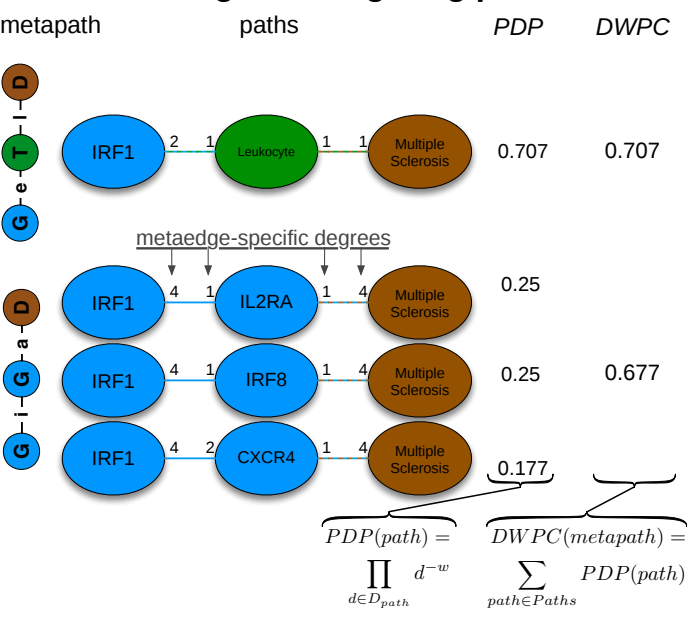
B. Metapaths for G-a-D:



C. Hypothetical graph:



D. Calculating and weighting path counts:



Forthcoming in *PLOS Computational Biology* preprint on *bioRxiv* [doi:10.1101/011569]



Predictions online at het.io

Predictions for multiple sclerosis					
Disease Ontology	BPID	UMLS Concepts	mean prediction	prediction	
IL10A	MONDO:0000000	IL10A	2	42.711%	42.711%
STAT4	MONDO:0000000	STAT4	5	12.537%	42.602%
IL10B	MONDO:0000000	IL10B	5	24.665%	41.679%
PTPN22	MONDO:0000000	PTPN22	4	6.699%	38.571%
IL10	MONDO:0000000	IL10	4	8.114%	38.349%
IL2RA	MONDO:0000000	IL2RA	5	18.881%	38.309%
IRF5	MONDO:0000000	IRF5	5	8.140%	31.477%
ICOSL1	MONDO:0000000	ICOSL1	3	3.847%	29.452%
HLA-DQA1	MONDO:0000000	HLA-DQA1	2	4.357%	28.301%
TNFRSF25	MONDO:0000000	TNFRSF25	4	7.097%	25.865%