

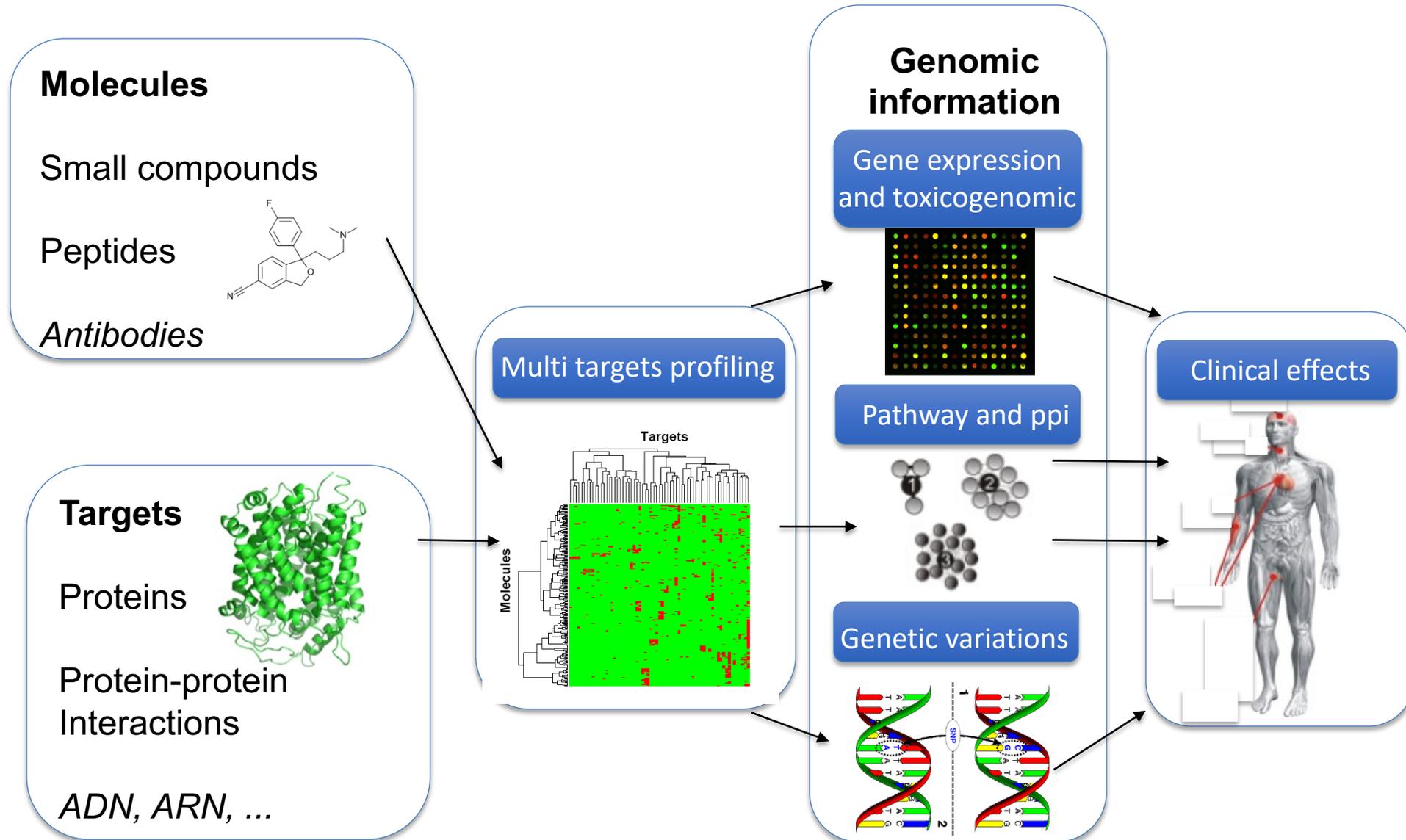
Chemoinformatics and network science

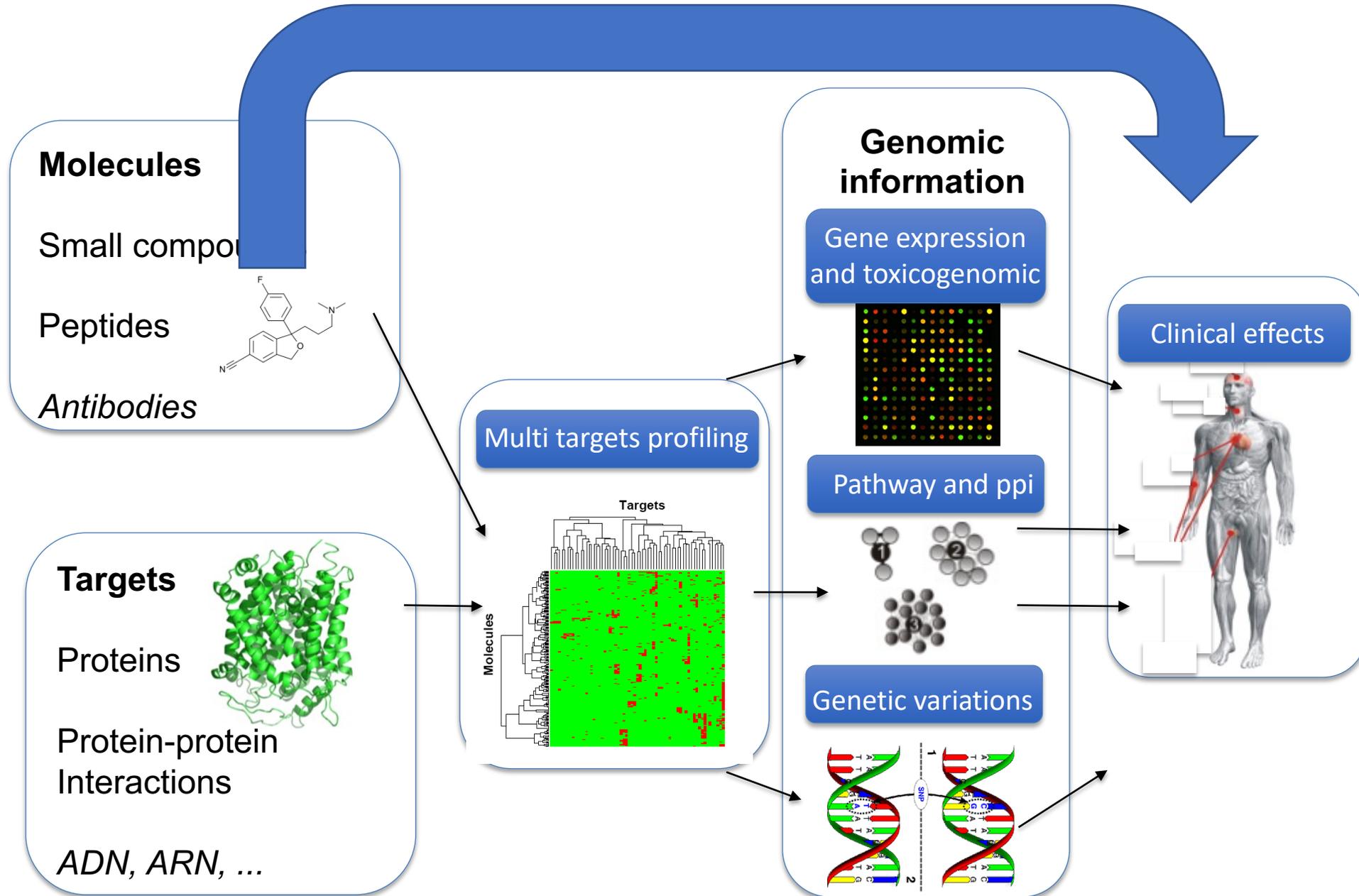
Pr. Olivier TABOUREAU

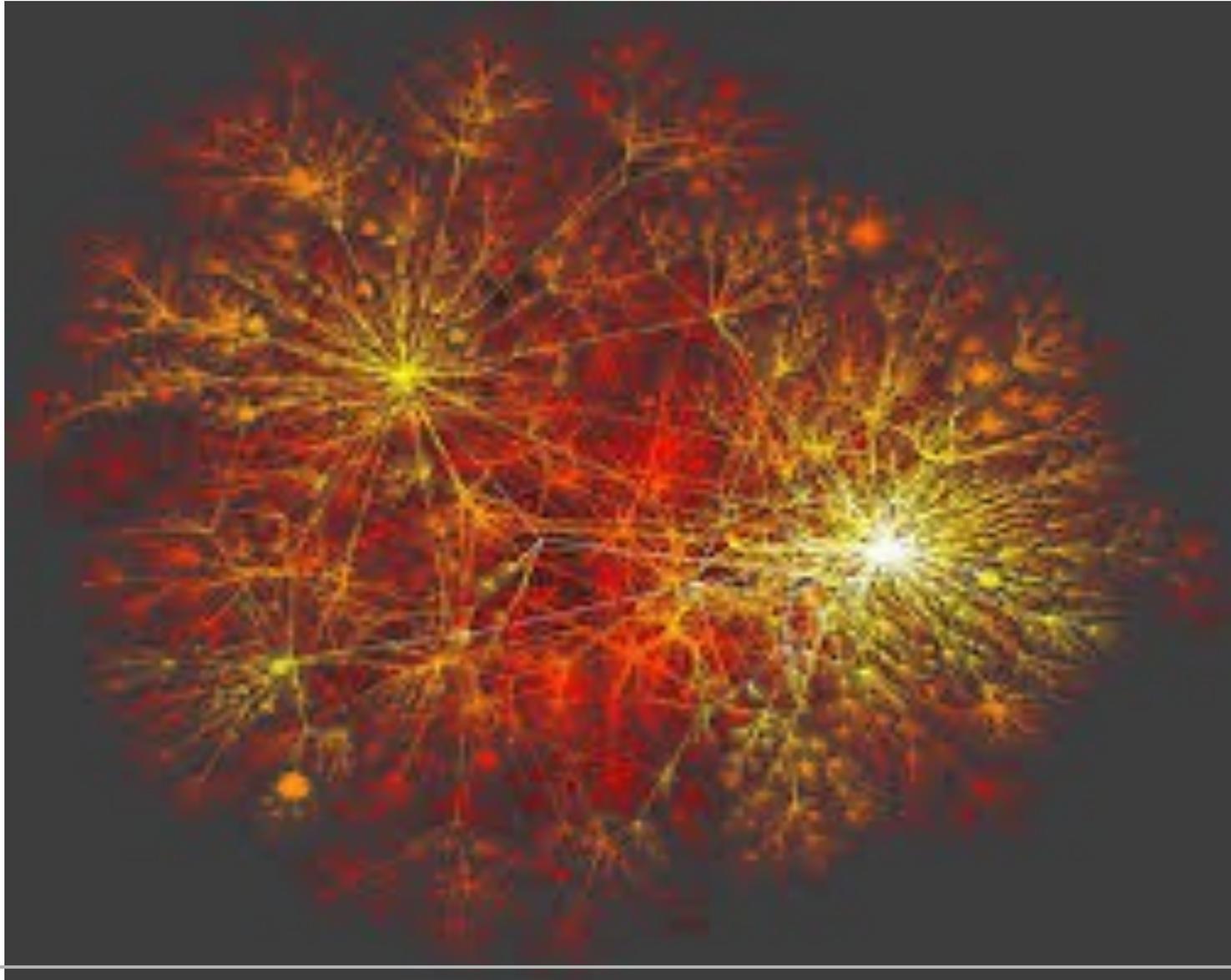


Computational modeling of protein-ligand interactions (Inserm U1133)

Objective: Understand the relationship between molecules (environmental chemicals, drugs, natural products, peptides) and proteins (or genes) from the molecular to more complex levels and in relation to diseases.



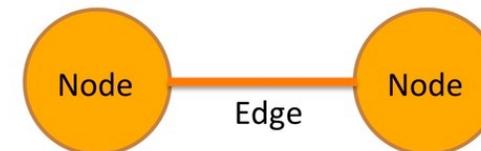




Network science is an academic field which studies complex networks such as telecommunication networks, computer networks, [biological networks](#), cognitive and semantic networks, and social networks, **considering distinct elements or actors represented by *nodes* (or *vertices*) and the connections between the elements or actors as *links* (or *edges*)**. The field draws on theories and methods including [graph theory](#) from mathematics, [statistical mechanics](#) from physics, [data mining](#) and [information visualization](#) from computer science, [inferential modeling](#) from statistics, and social structure from sociology.

Network = nodes + edges

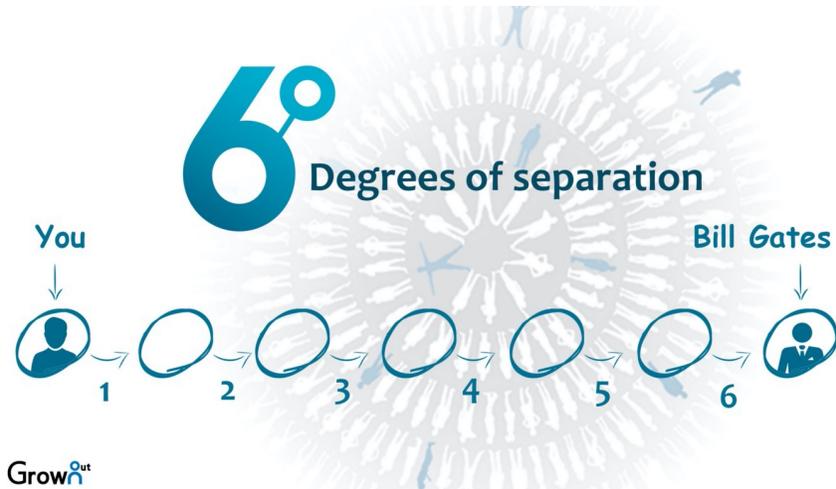
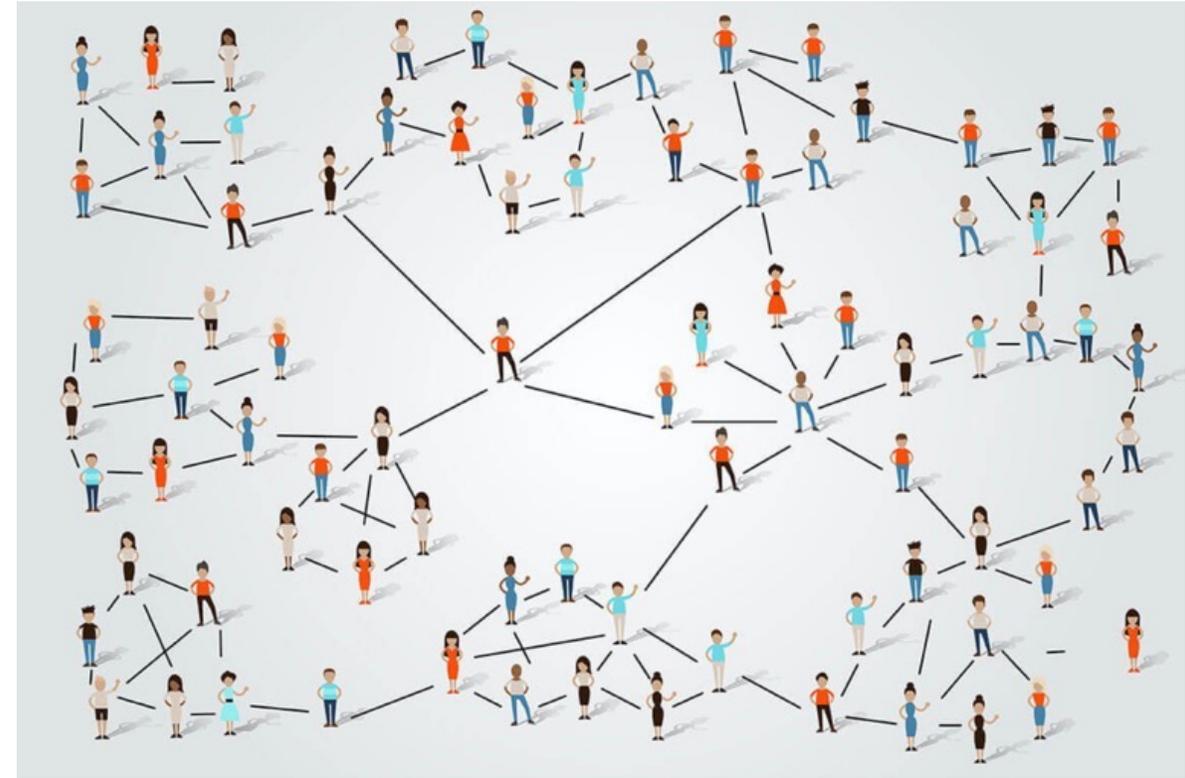
- **Nodes** (vertices) are the objects in the network
- **Edges** are the links/interaction in the network



Social Network



6 degrees of separation

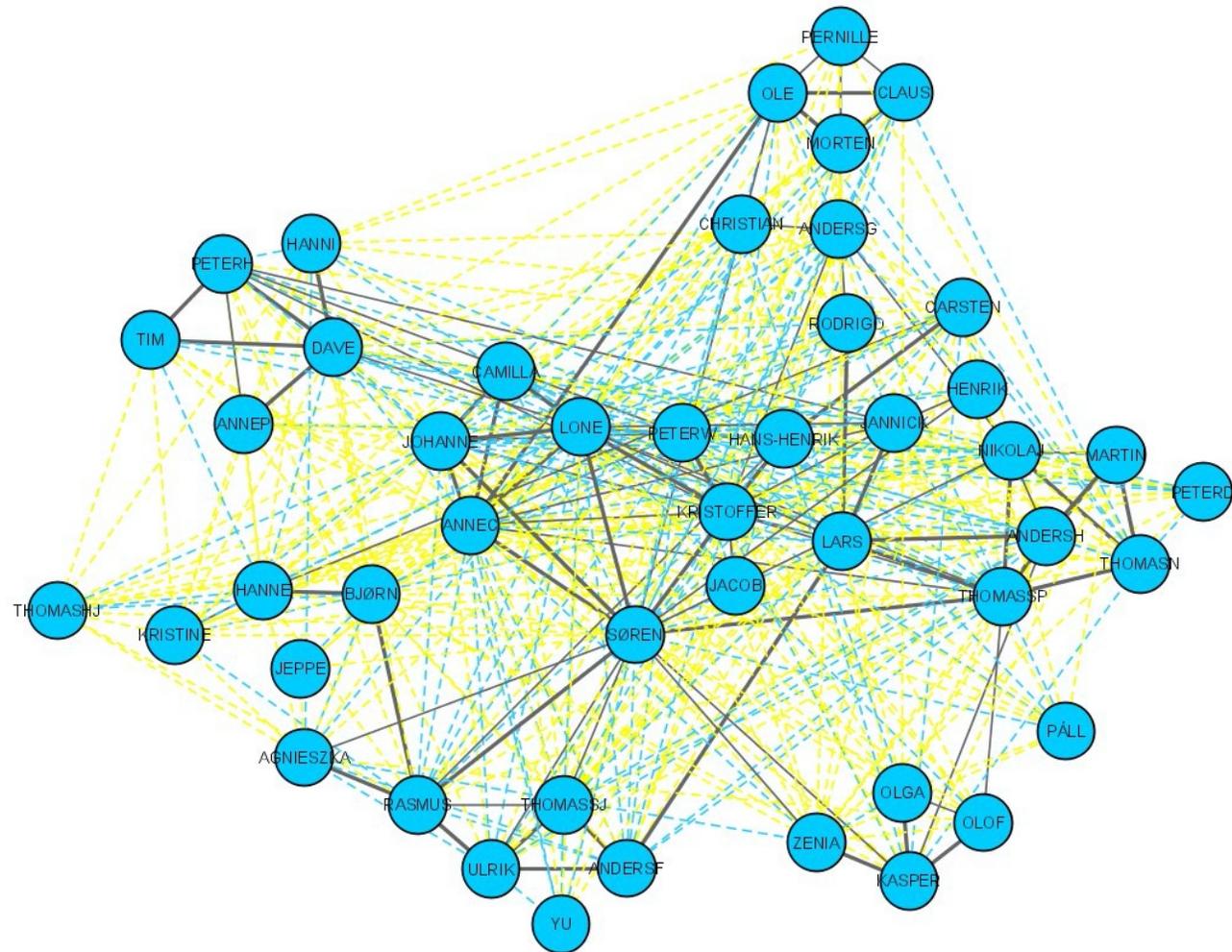


All people are six or fewer social connections away from each other

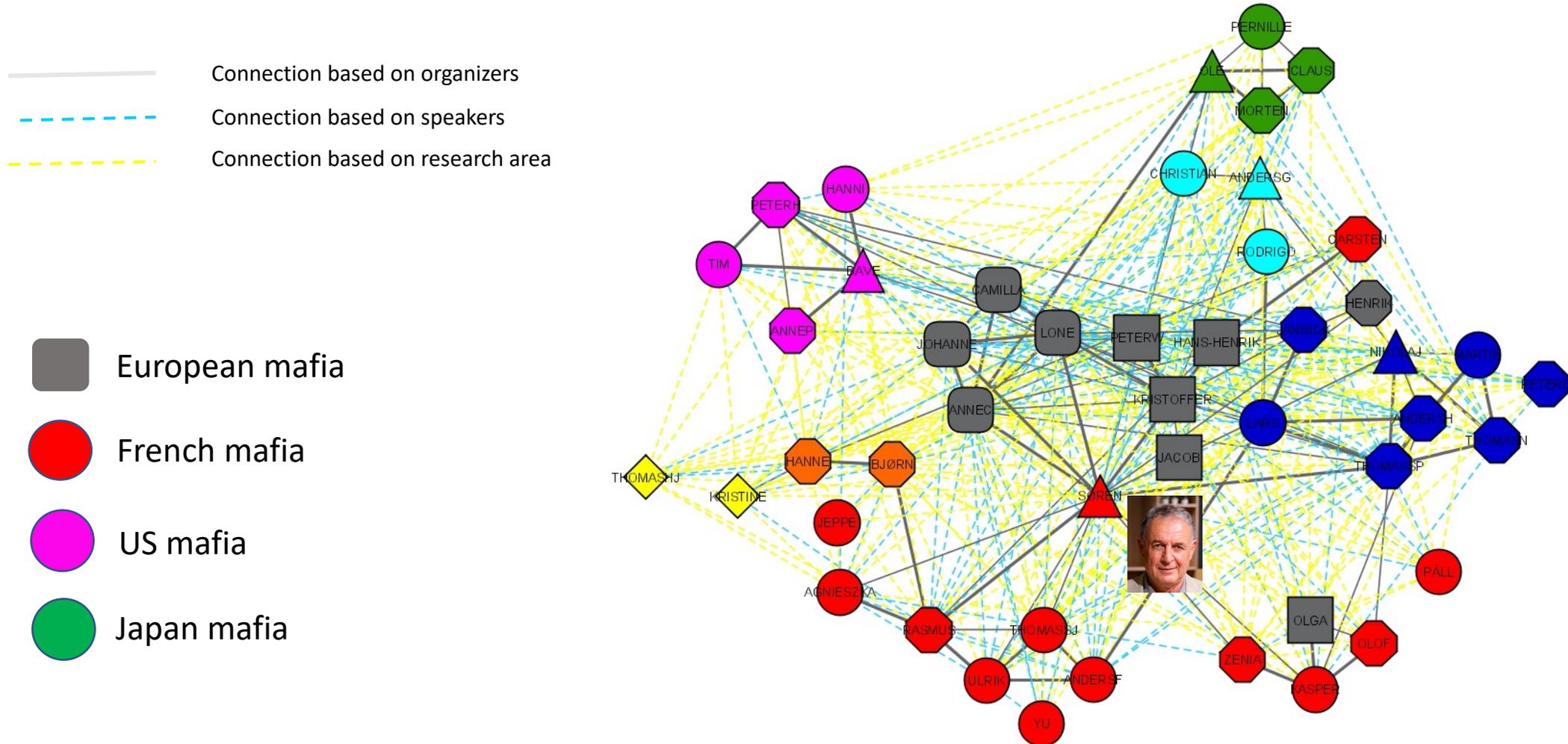
The chemoinformatics summer school network

- Connection based on organizers
- - - Connection based on speakers
- - - Connection based on research area

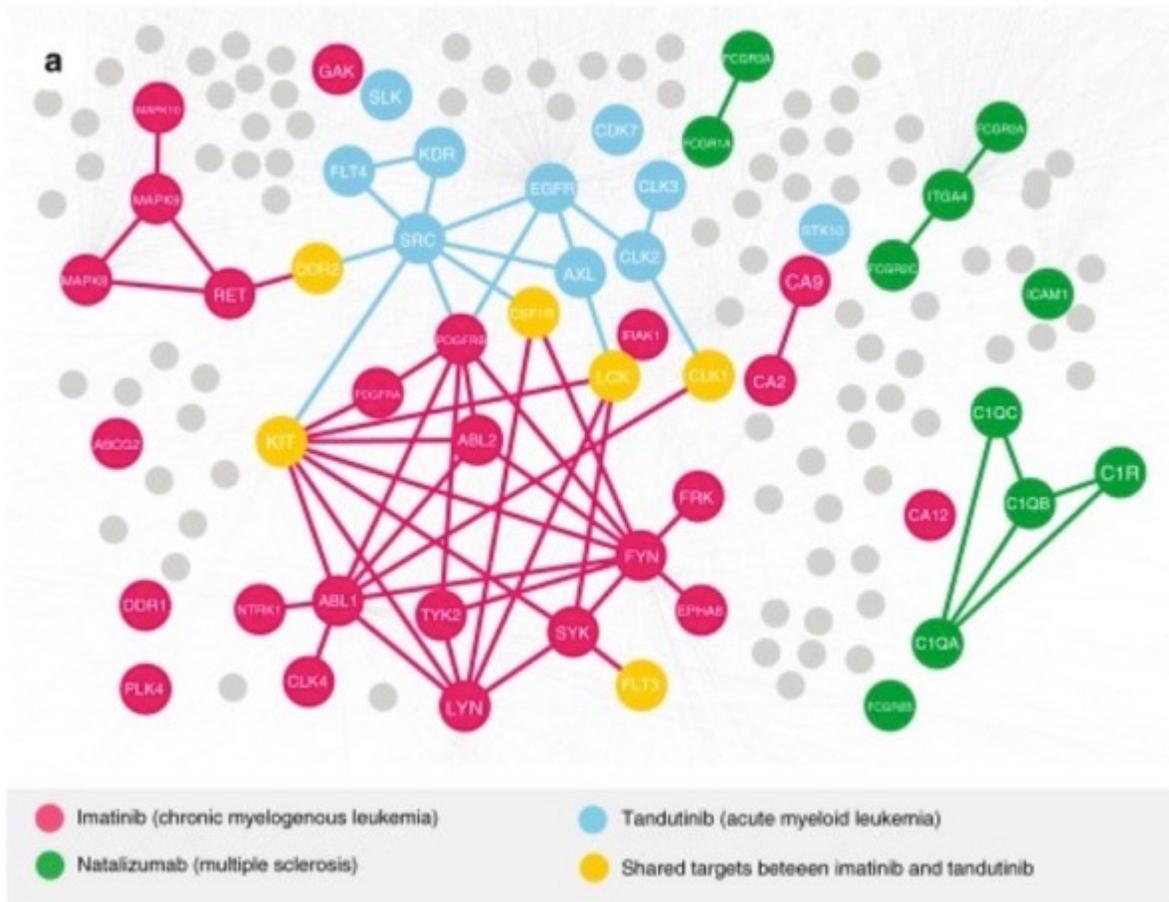
- **Can you cluster them?**
- **Who is the central connection?**



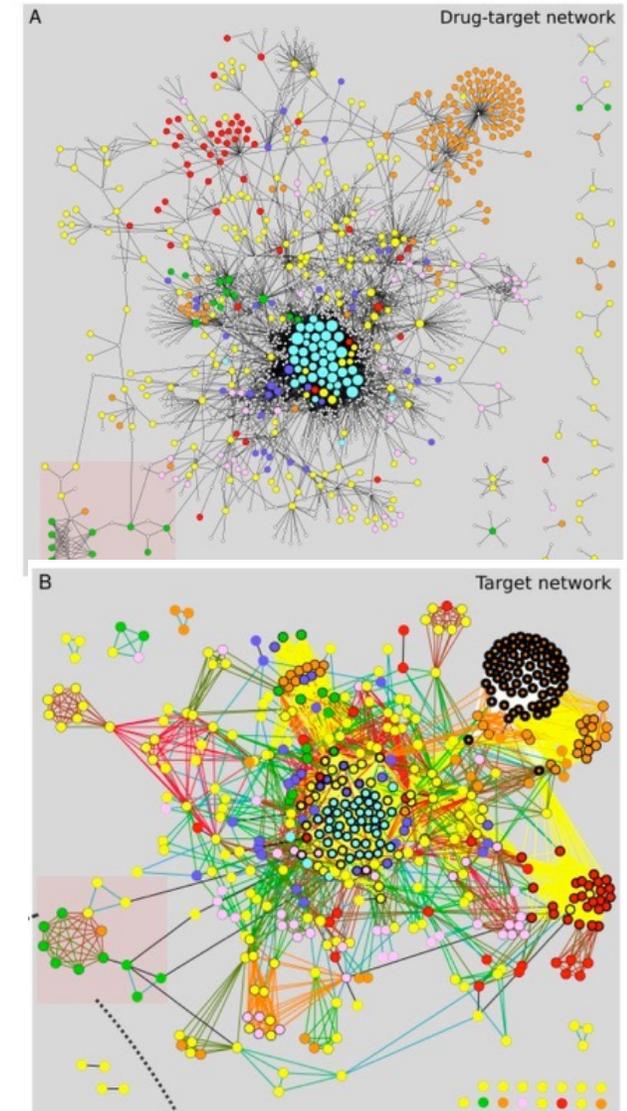
The chemoinformatics summer school Network



Network pharmacology

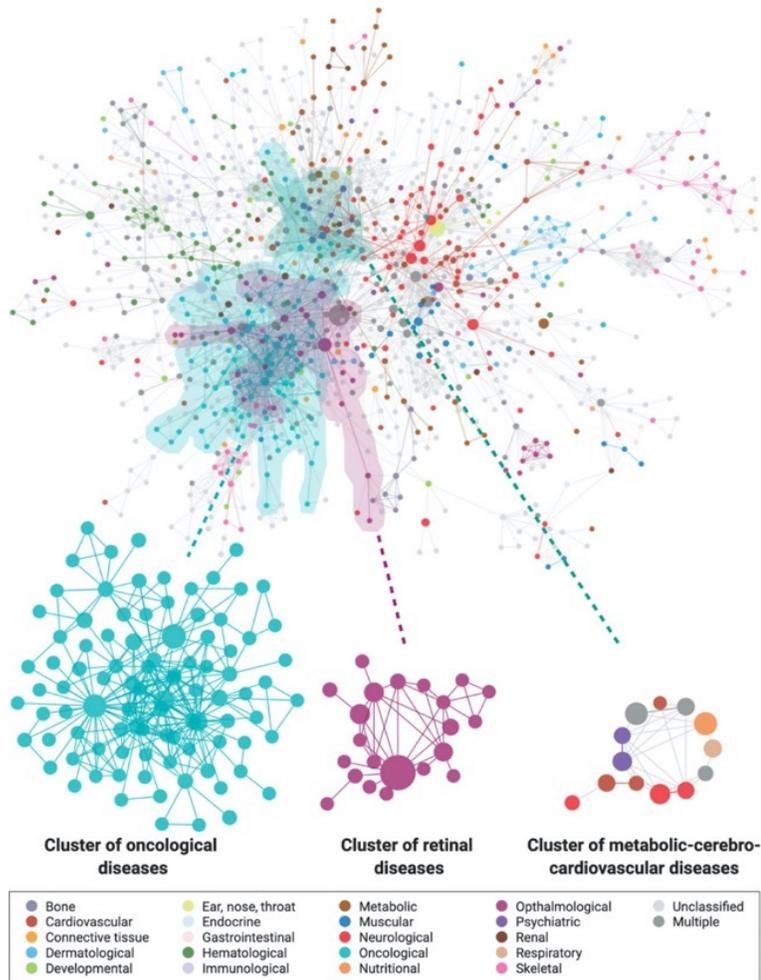


Cheng F. et al. Nat. Commun. 2019



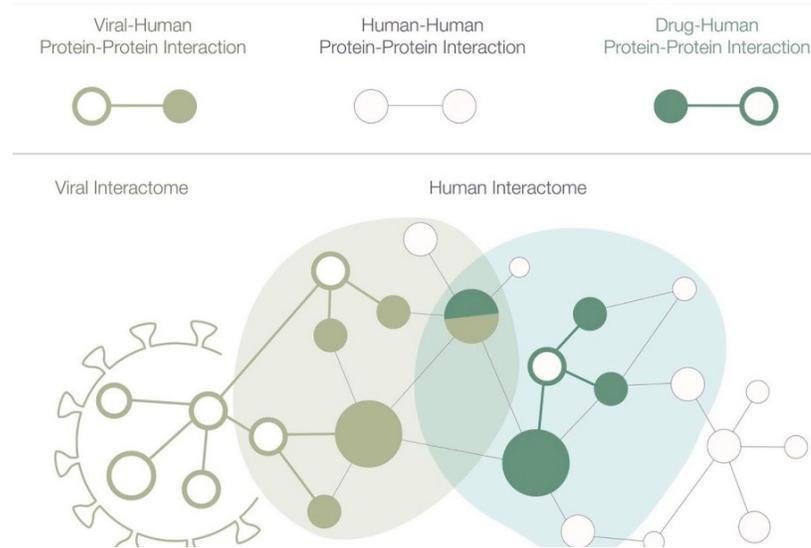
Vogt I. et al. Mol. Inf 2019

Human disease network



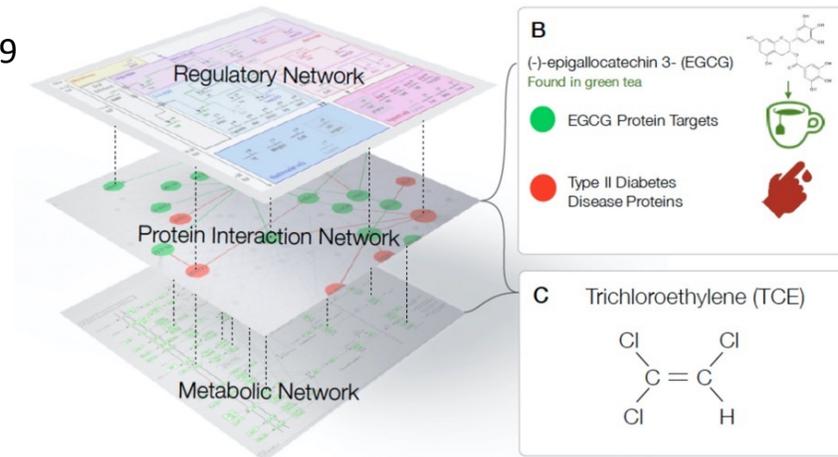
Nogales C. et al. Trends Pharmacol Sci. 2022

Covid-19 research



<https://www.networkscienceinstitute.org/covid-19>

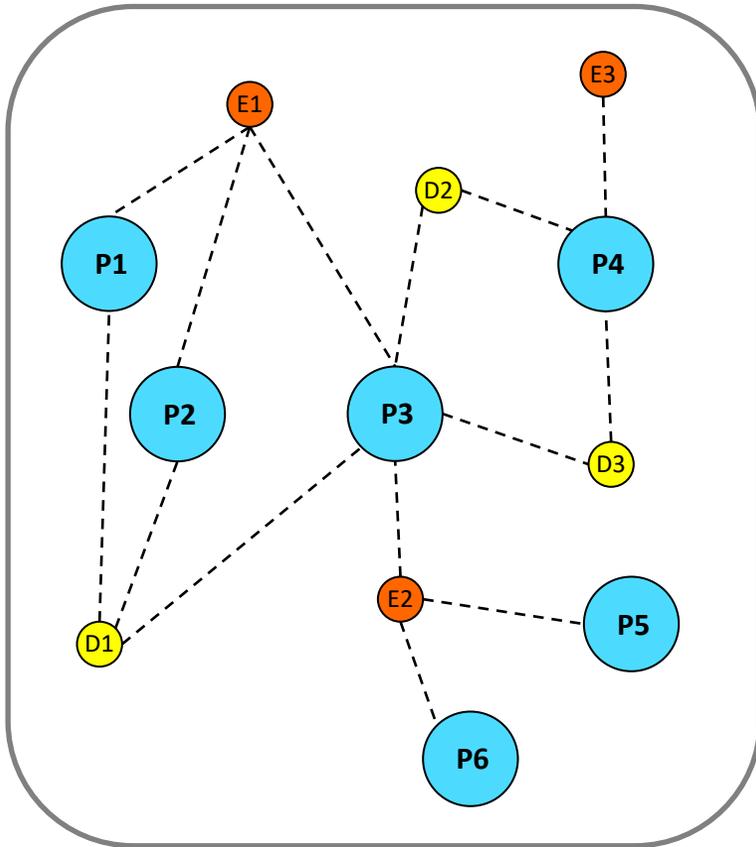
Exposome research



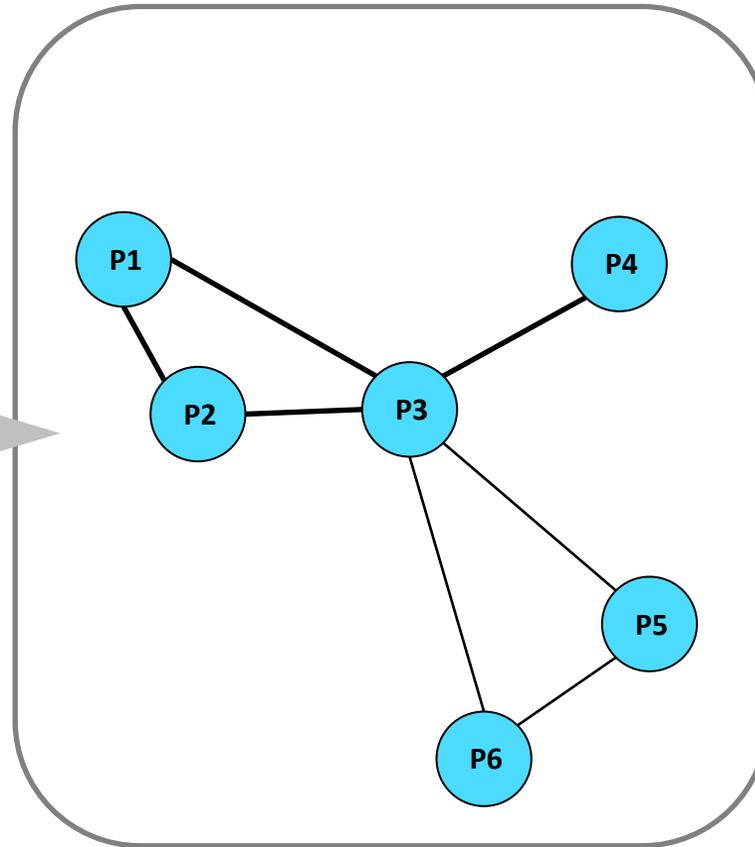
Vermeulen R. et al. Sciences. 2020

Network Science : Protein-Protein association network -> inferred network

Protein-chemical associations



Protein-protein associations



reliability
scores



Guilt by association theory

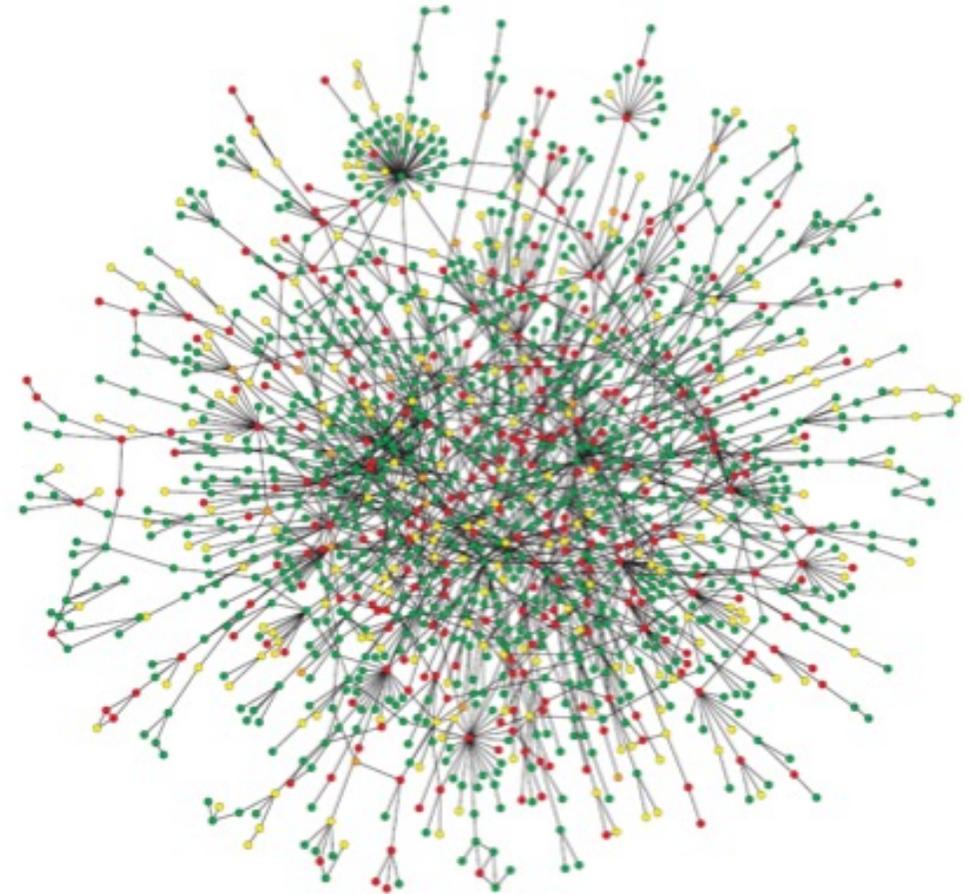


Based on the guilt by association theory

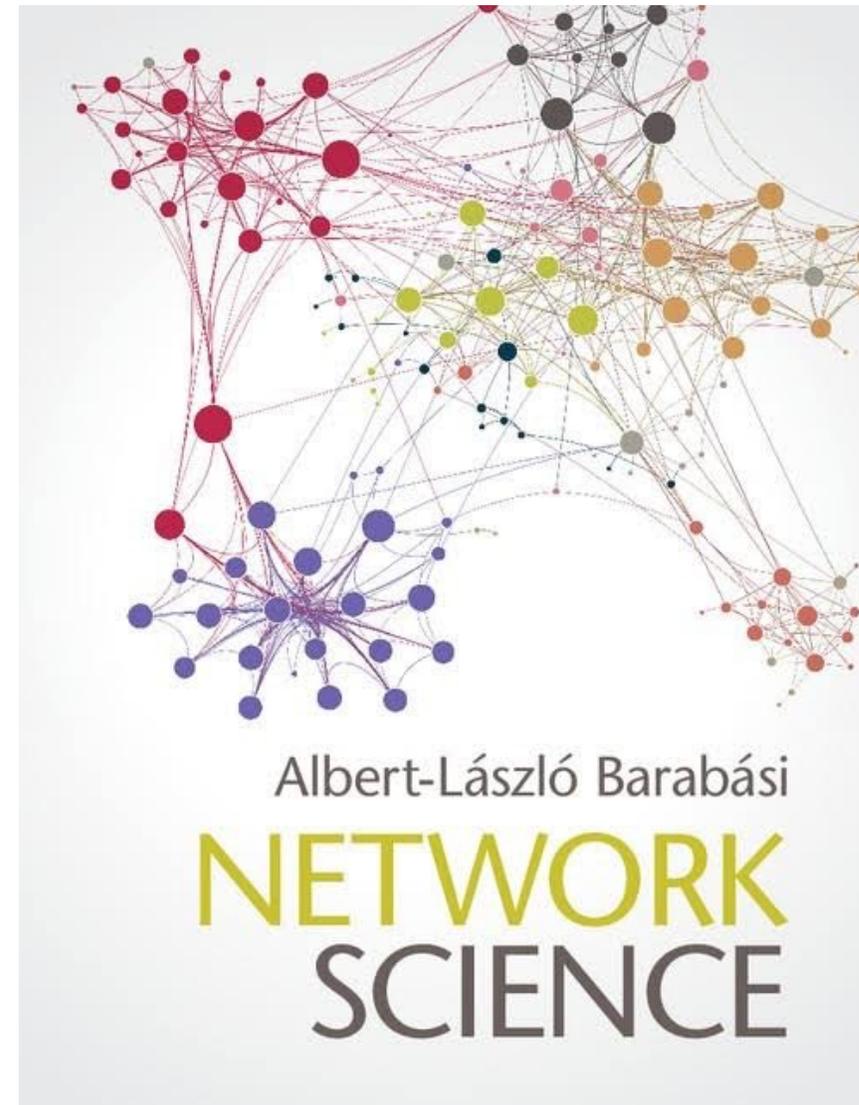
Many type of biological interactions

- Protein-drug interactions
- Protein-protein interactions
- Protein-DNA interactions
- Protein-disease interactions
- Co-expressions interactions
- Metabolic interactions

...



**More information can be found
about network science theory**



Network Science in toxicology and adverse drug reactions



Environmental exposure in Europe
430,000 /year [1]

Adverse Drug Reactions in Europe
197,000 /year [2]

Toxicology profiling is failing

Disability-adjusted life years (environmental exposure)
Worldwide
86m /year [3]

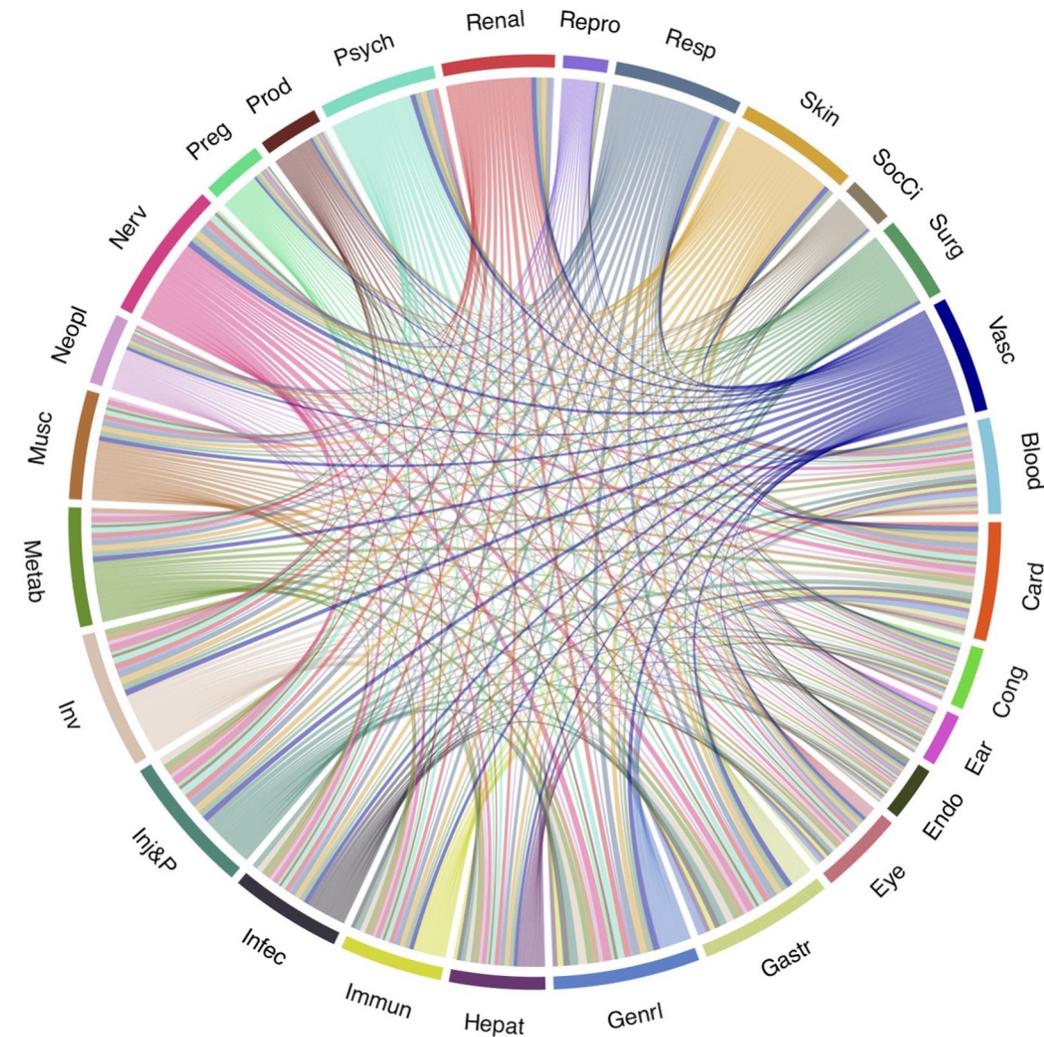
Environmental exposure deaths
Worldwide
4,9m /year

[1] <https://www.eea.europa.eu/media/newsreleases/many-europeans-still-exposed-to-air-pollution-2015/premature-deaths-attributable-to-air-pollution>
[2] Jacoline C. Bouvy, et al. 2015, Drug Saf.
[3] Annette Prüss-Ustün, et al. 2011, Environ. Health

Network Science : An example with adverse drug reactions

Drugs – Adverse Drug Reactions (ADR) - System Organ Classification (SOC) network

- 1000 Drugs and 6 164 ADRs classified in 27 SOC
 - Drugs have ADRs belonging to different SOC
- Most common ADRs are : ‘Nausea’, ‘Vomiting’
 - resp. 41% and 40% of drugs respectively
- Specific ADRs : 1 626 ADRs are only related to one drug
- Drugs with the highest number of ADRs : methotrexate, alendronic acid and prednisone
 - resp 1 561, 1 422 and 1 377 ADRs
- 155 drugs are associated to only one ADR
- **Knowing drug-target and drug-ADR -> Target-ADR can be suggested**

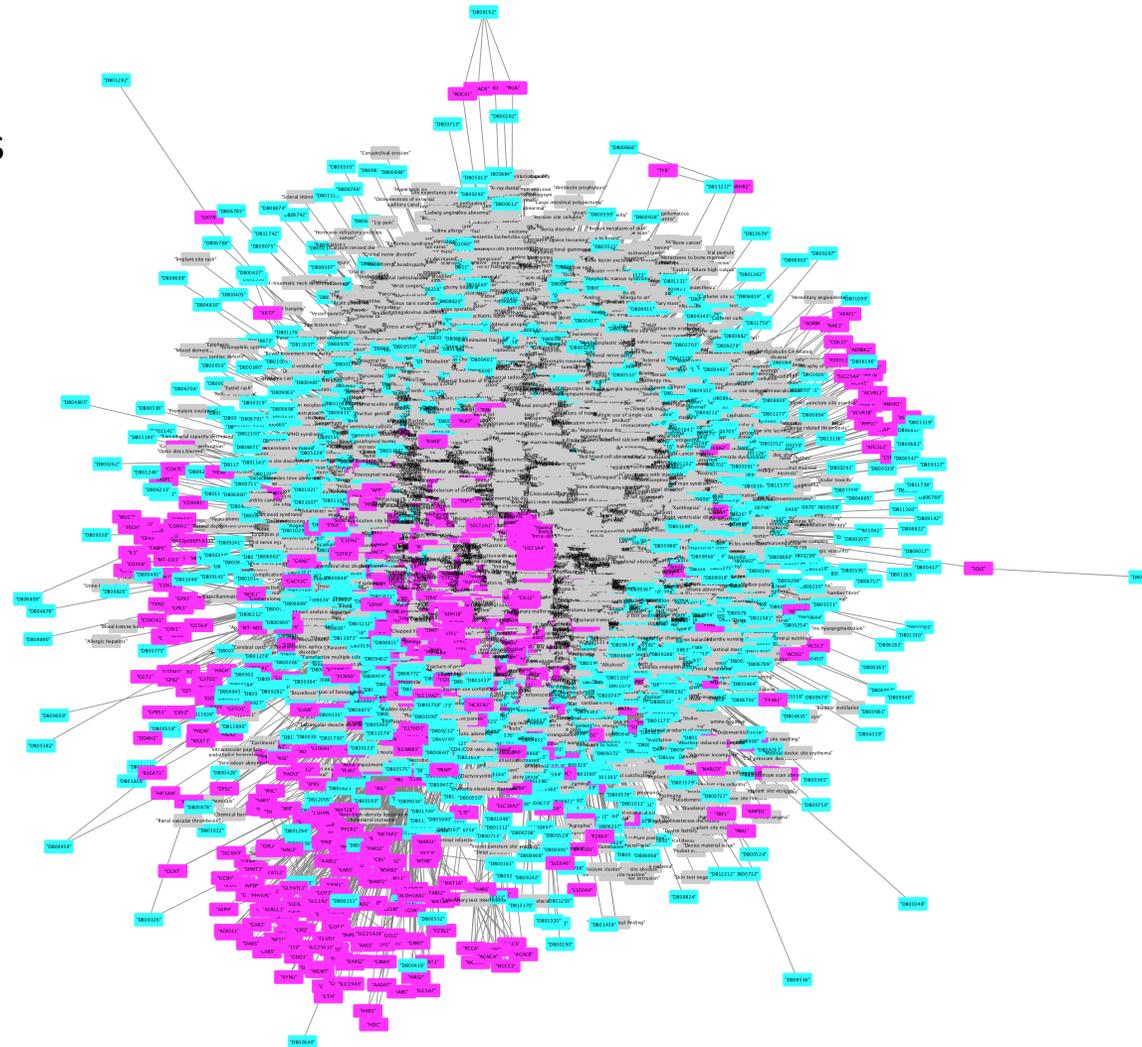


Network Science : An example with adverse drug reactions

Drugs – Adverse Drug Reactions (ADR) - System Organ Classification (SOC) network

Identification and prioritization of drug-target involved in specific ADRs, as well as in more general terms using SOC

Network of 515 959 interactions



Network Science : An example with adverse drug reactions

Drugs – Adverse Drug Reactions (ADR) - System Organ Classification (SOC) network

- Targets associated with the most ADRs are cytochromes and transporters.
- Possible to assess the role of proteins in specific ADRs
 - HPD (hydroxyphenyl pyruvate dioxygenase) might be highly related to 'Hepatocellular carcinoma'
 - SQLE (Squalene epoxidase) is connected to the ADR 'Blister'

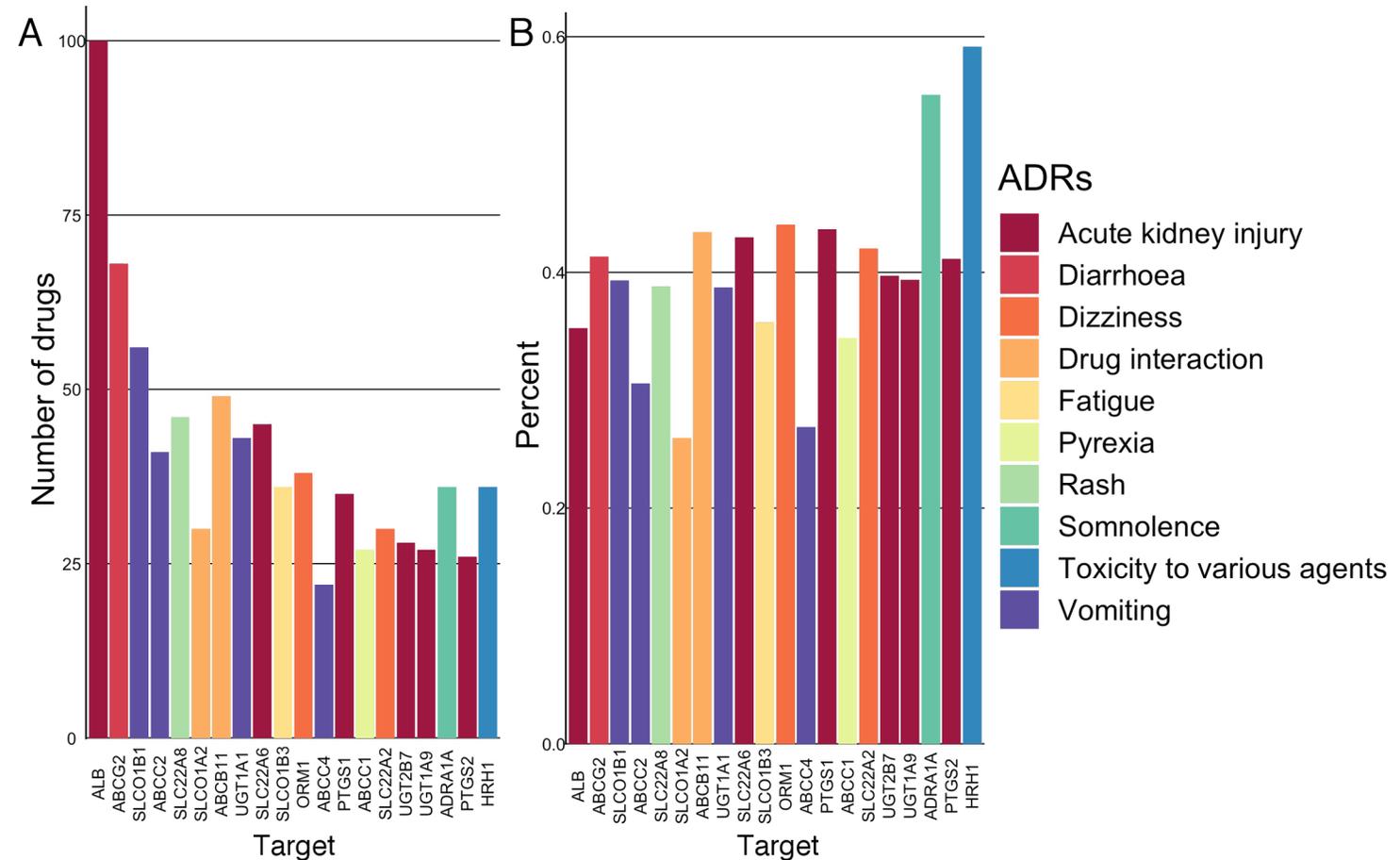


Figure : A) The 20 first proteins associated the most to drugs and ADRs. B) The proportion of these 20 proteins associated with an ADR among the others proteins linked to this ADR.

Network Science : An example with adverse drug reactions

Drugs – Adverse Drug Reactions (ADR) - System Organ Classification (SOC) network

- The majority of proteins do not seem to be specific to one SOC
- Development of a scoring function to estimate the contribution of each protein to an ADR

Scoring function

$$ADR_x = \left(\sum_{g=1}^h \frac{1}{T_{Dg}} D_g \right) \text{Protein}_1 + \left(\sum_{i=1}^j \frac{1}{T_{Di}} D_i \right) \text{Protein}_2 + \dots + \left(\sum_{k=1}^l \frac{1}{T_{Dk}} D_k \right) \text{Protein}_n$$

$D_{(g,i,k)}$: Drugs known to cause ADR_x , that also interact with the protein associated with ADR_x

$T_{D(g,i,k)}$: Total number of proteins interacting with D_i

$\text{Protein}_{(1\dots n)}$: Protein associated with ADR_x

g,i,k can represent the same drug if it targets multiple proteins for ADR_x

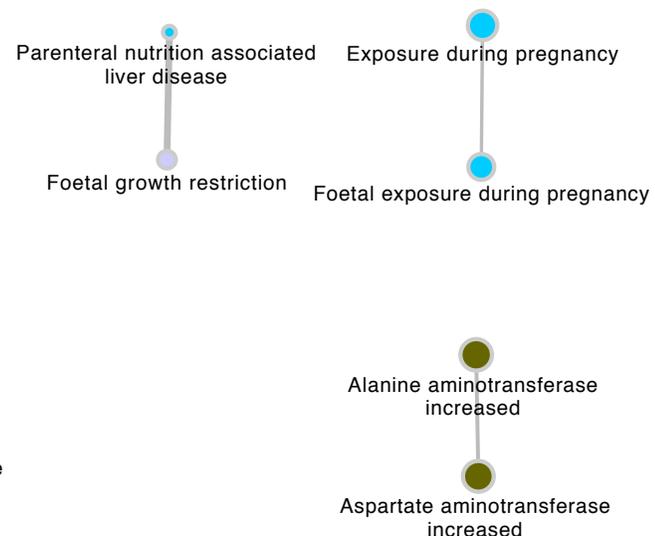
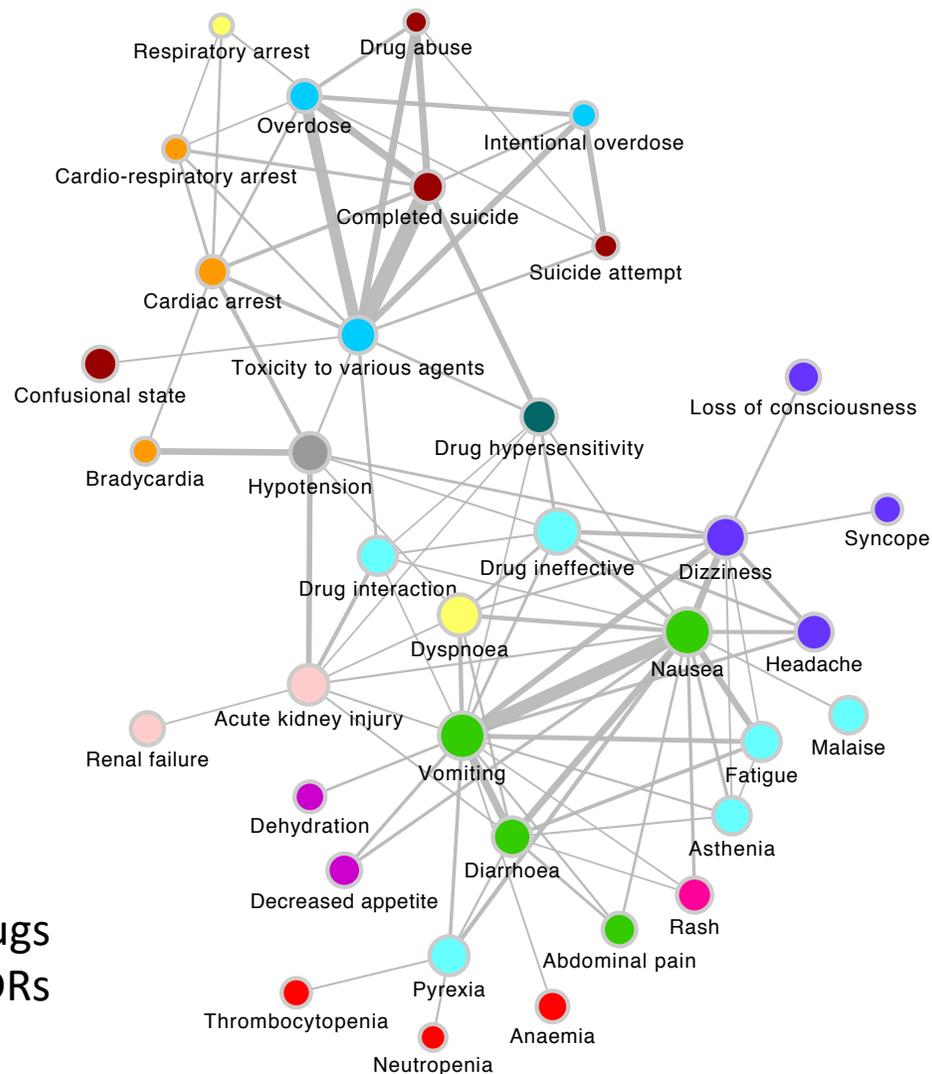
- 'Prostate cancer stage IV' = 0.5xGNRHR + 0.5xCYP3A4 (1 drug = leuprolide)

- 'Bone fragmentation' = 2.033xFDPS + 0.5xPTGS2 + 0.333xABCC1 + 0.333xGGPS1 + 0.2xATP6V1A + 0.2xPTPN4 + 0.2xPTPRE + 0.2xPTPRS (4 drugs = alendronic, ibandronate, risedronic, zoledronic)

 **Such network can suggest the role and the importance of targets involved in an ADR**

Network Science : An example with adverse drug reactions

ADR – ADR network



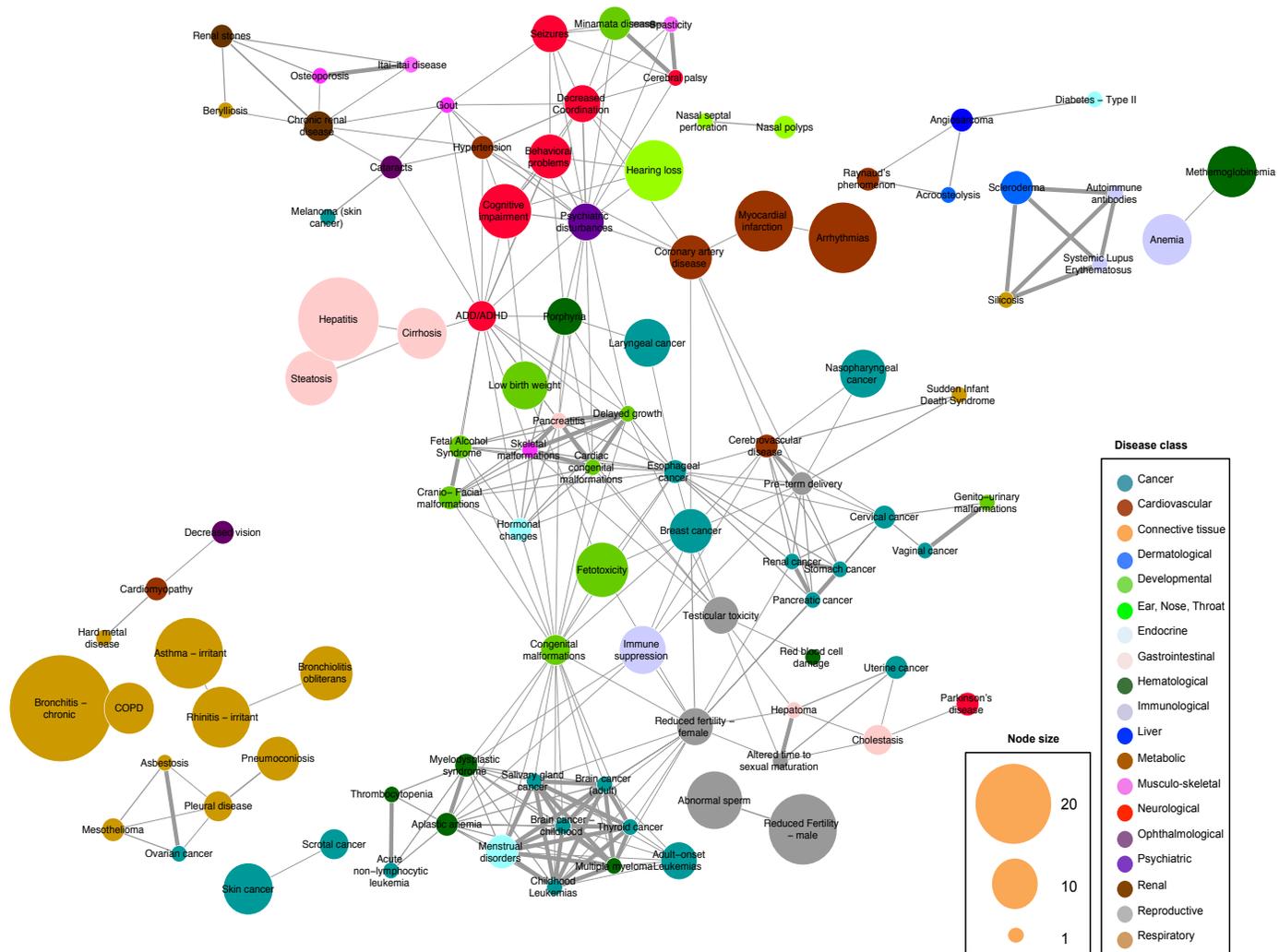
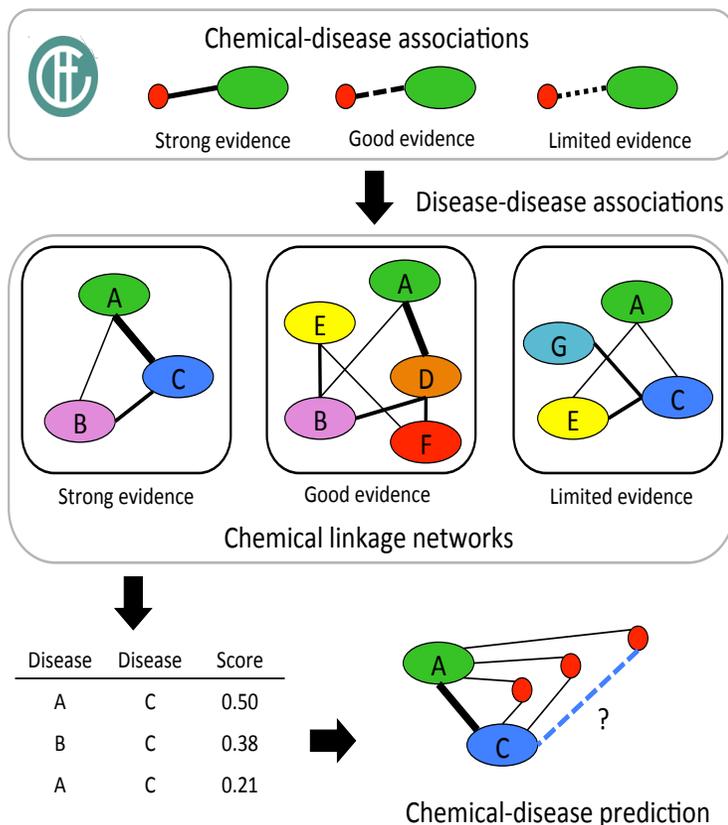
System organ class (SOC)	
●	Blood and lymphatic system disorders
●	Cardiac disorders
●	Gastrointestinal disorders
●	General disorders and administration site conditions
●	Immune system disorders
●	Injury, poisoning and procedural complications
●	Investigations
●	Metabolism and nutrition disorders
●	Nervous system disorders
●	Pregnancy, puerperium and perinatal conditions
●	Psychiatric disorders
●	Renal and urinary disorders
●	Respiratory, thoracic and mediastinal disorders
●	Skin and subcutaneous tissue disorders
●	Vascular disorders

Node size	
● (Large)	409 Nausea
● (Medium)	255 Abdominal pain
● (Small)	18 Parenteral nutrition associated liver disease

- 1000 Drugs
- 6164 ADRs

Network Science : Environmental disease network (EDN)

A semi-quantitative network-based analysis to predict disease-disease associations based on chemical-disease data



Network Science : Environmental disease network (EDN)

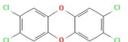
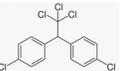
Example with Type 2 Diabetes and hormonal change

Disease-disease associations:

SE	GE	LE	Disease name	Disease class
			Adult-onset leukemias	cancer
			Bladder cancer	cancer
			Brain cancer (adult)	cancer
			Breast cancer	cancer
			Colo-rectal cancer	cancer
			Gallbladder cancer	cancer
			Hepatocellular cancer	cancer
			Lung cancer	cancer
			Prostate cancer	cancer
			Skin cancer	cancer
			Stomach cancer	cancer
			Uterine cancer	cancer
			Arrhythmias	cardiovascular
			Coronary artery disease	cardiovascular
			Dyslipidemia	cardiovascular
			Hypertension	cardiovascular
			Contact dermatitis- Irritant	dermatological
			Hyperkeratosis	dermatological
			Skin ulceration	dermatological
			Cardiac congenital malformations	developmental
			Congenital malformations	developmental
			Genito-urinary malformations	developmental
			Oral clefts	developmental
			Hearing loss	Ear, nose, throat

SE	GE	LE	Disease name	Disease class
			Early onset menopause	endocrine
			Hormonal changes	endocrine
			Menstrual disorders	endocrine
			Thyroid disorders	endocrine
			Aplastic anemia	hematological
			Hodgkin's disease	hematological
			Multiple myeloma	hematological
			Porphyria	hematological
			Thrombocytopenia	hematological
			Angiosarcoma	liver
			Hepatitis	liver
			ADD/ADHD	neurological
			Alzheimer	neurological
			Parkinson	neurological
			Peripheral neuropathy	neurological
			Cataracts	ophthalmological
			Macular degeneration	ophthalmological
			Abnormal sperm	reproductive
			Altered sex ratio	reproductive
			Altered time to sexual maturation	reproductive
			Reduced fertility-female	reproductive
			Reduced fertility-male	reproductive

Chemical-Target-Disease associations:

Chemical	Biological enrichment	Nb. gene	P-val.	Gene list	
TCDD 	Disease	diabetes mellitus	26	1.90 ^{e-19}	CPT1A;EDN1;AKT2;GCK;HMOX1;HNF4A;HP;IRS1;KCNJ11;LEP;LEPR;NFKB1;ENPP1;PPARA;PPARG;RETN;PTPN1;SLC2A1;SLC2A2;SLC2A4;TNFRSF1A;C3;UCP2;WFS1;CAT;ADIPOQ
	GO process	regulation of hormone secretion (go:0046883)	14	3.36 ^{e-11}	CPT1A;EDN1;GCK;HNF4A;IL6;IRS1;KCNJ11;LEP;SLC2A1;SLC2A2;HNF1A;TCF7L2;IRS2;ADIPOQ
		response to steroid hormone stimulus (go:0048545)	9	0.032	EDN1;HMOX1;FAS;IL6;KCNJ11;PPARA;PPARG;TNFRSF1A;ADIPOQ
DDT 	Disease	diabetes mellitus	16	0.799	ADRB3;ESR2;ALB;G6PD;GGT1;GPT;IL1R1;NFKB1;PON1;MAPK8;RBP4;SHBG;SOD1;TNFRSF1A;CAT;PDE5A
	GO function	steroid hormone receptor activity (go:0003707)	10	1.59 ^{e-05}	NR0B1;ESR1;ESR2;HNF4G;AR;PGR;RXRB;NR0B2;NR1I2;NR1I3
		hormone activity (go:0005179)	4	n.s.*	FSHB;GNRH1;POMC;TSHB

* n.s. indicates corrected p-values non significant (>1).

➔ Chemical-protein-disease network

Network Science : Integrative systems chemical toxicology

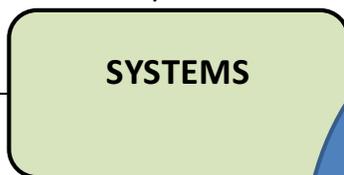
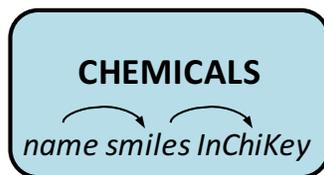
1. DATA

2. MODEL

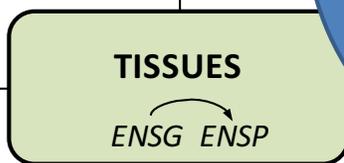
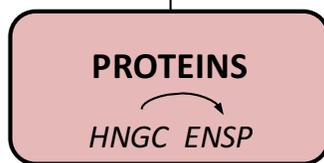
3. PREDICTION

25000 chemicals with toxicity data

24 systems



RTECS

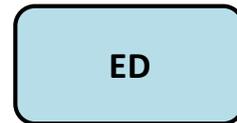
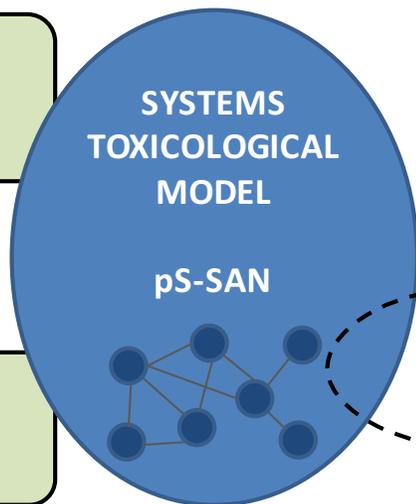


HPA

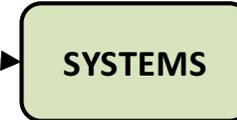
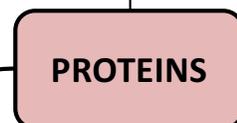
STITCH

15000 proteins

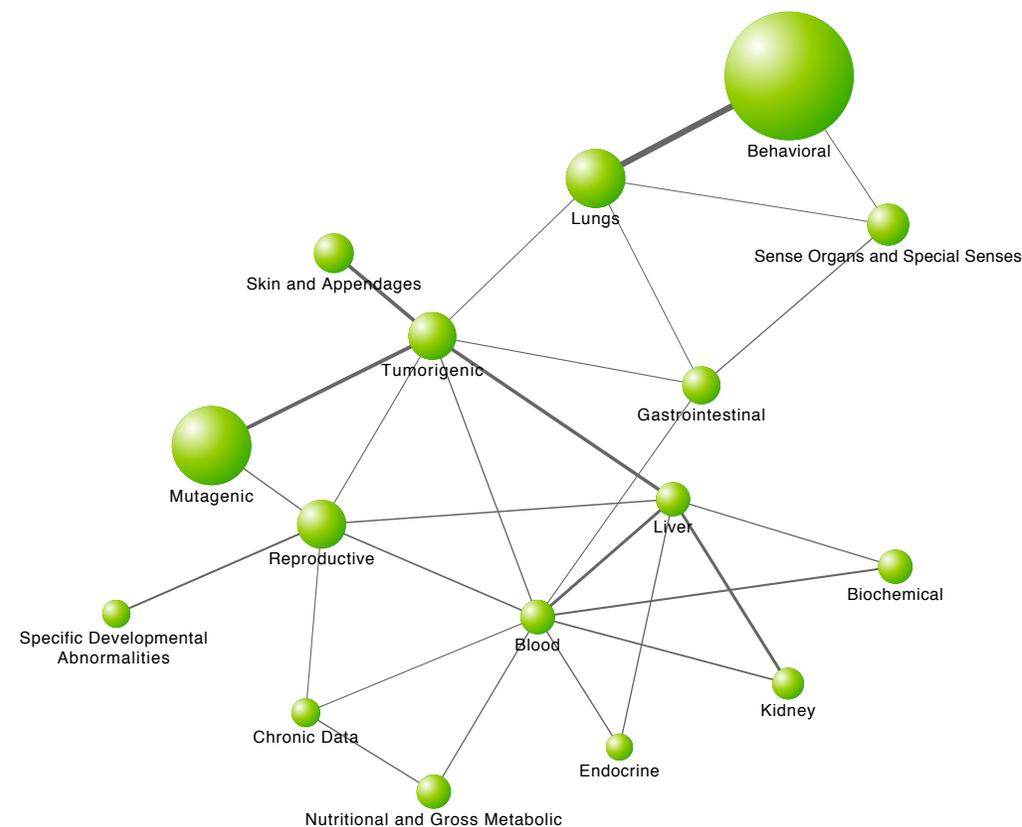
55 tissues



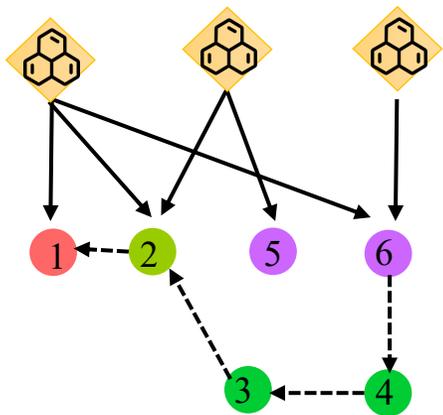
CTD



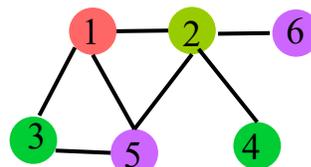
Chemical-protein-system toxicology network



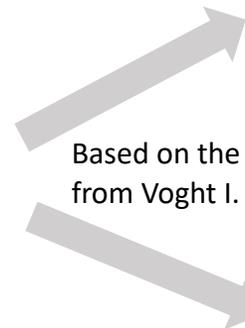
Network Science : Measuring the loss of coverage in network biology



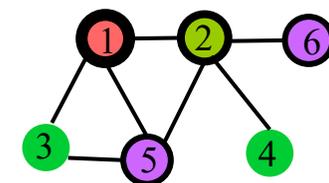
drug-event bipartite network (from AOP)



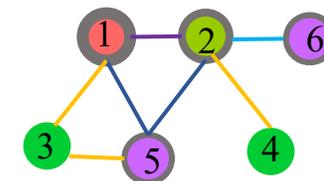
event- event network



Based on the work from Voght I. 2019



Increase in uncertainty



loss of coverage

Data compilation & Bipartite network development

Monopartite network development

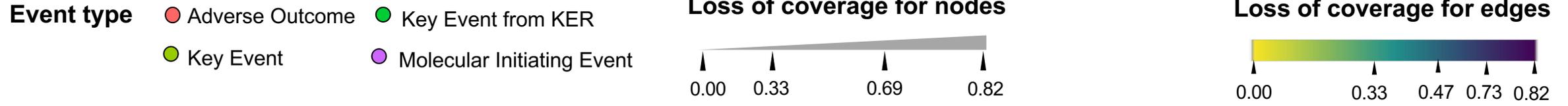
Network analysis

- Molecular initiating event
- Key event
- Key event from KER

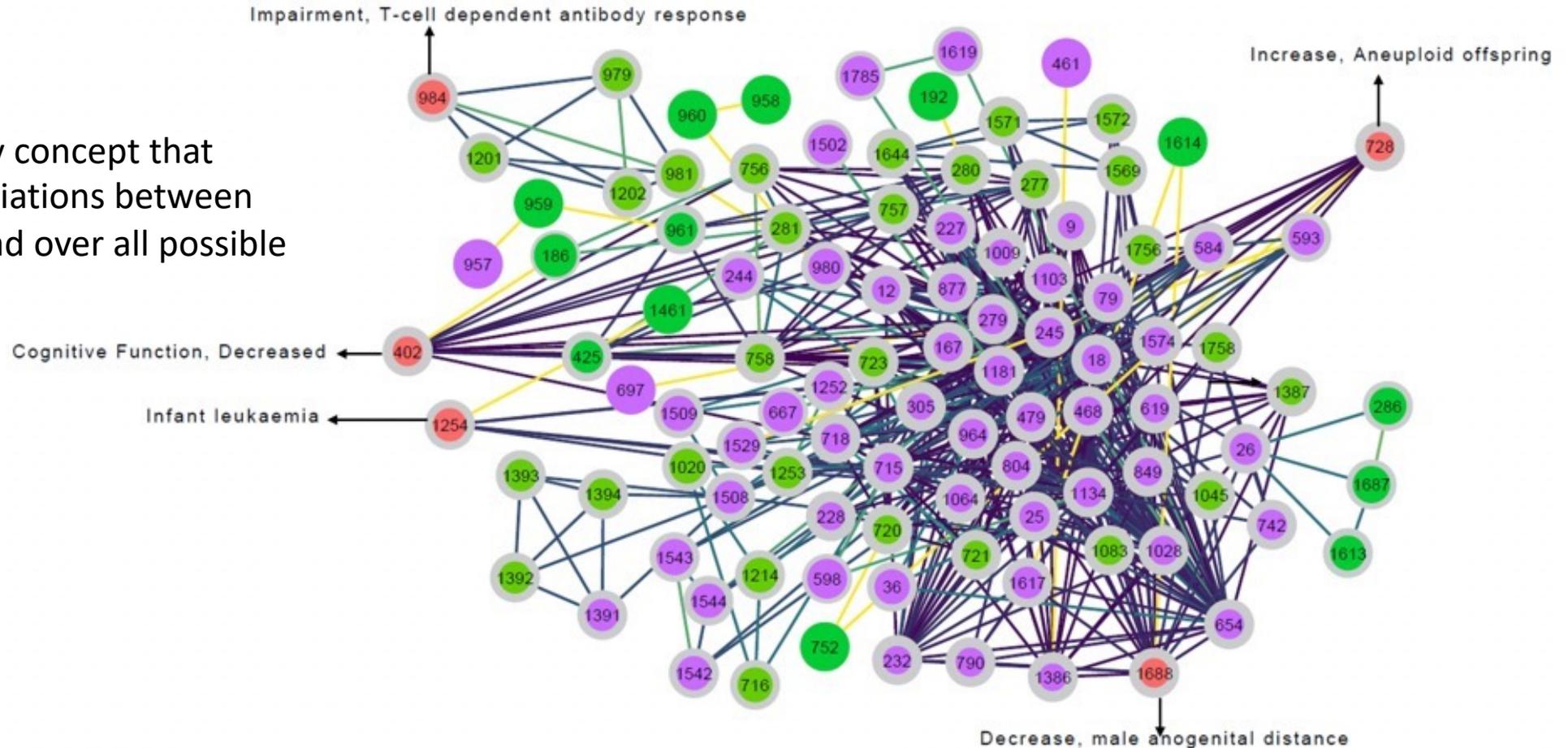
- Adverse outcome
- Drug

- Drug-event associations
- Key event relationships (KER)
- Event-event relationships

Network Science : Measuring the loss of coverage in network biology



Based on Shannon entropy concept that measures how many associations between drugs and events are spread over all possible connections in a network



Network Science : Measuring the loss of coverage in network biology

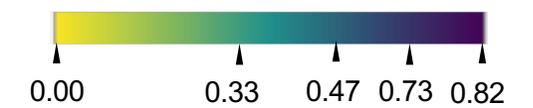
Event type

- Adverse Outcome
- Key Event from KER
- Key Event
- Molecular Initiating Event

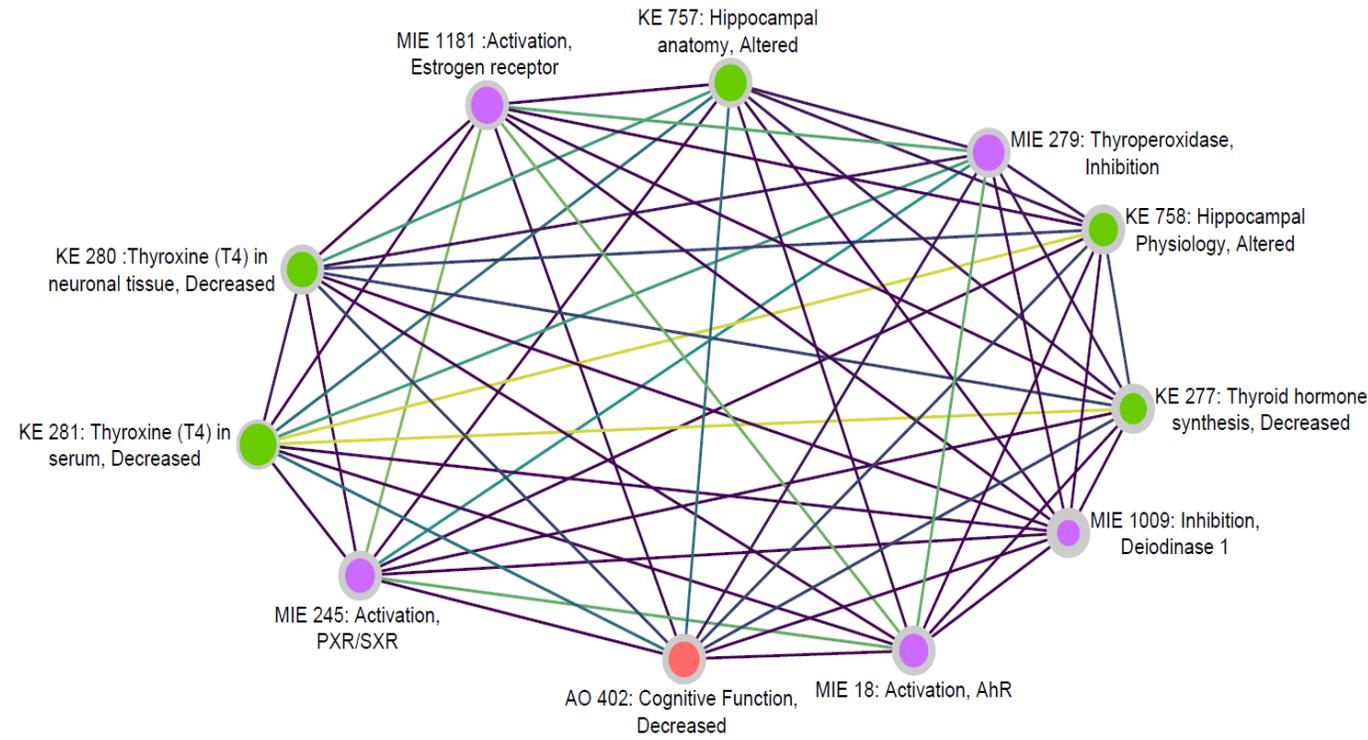
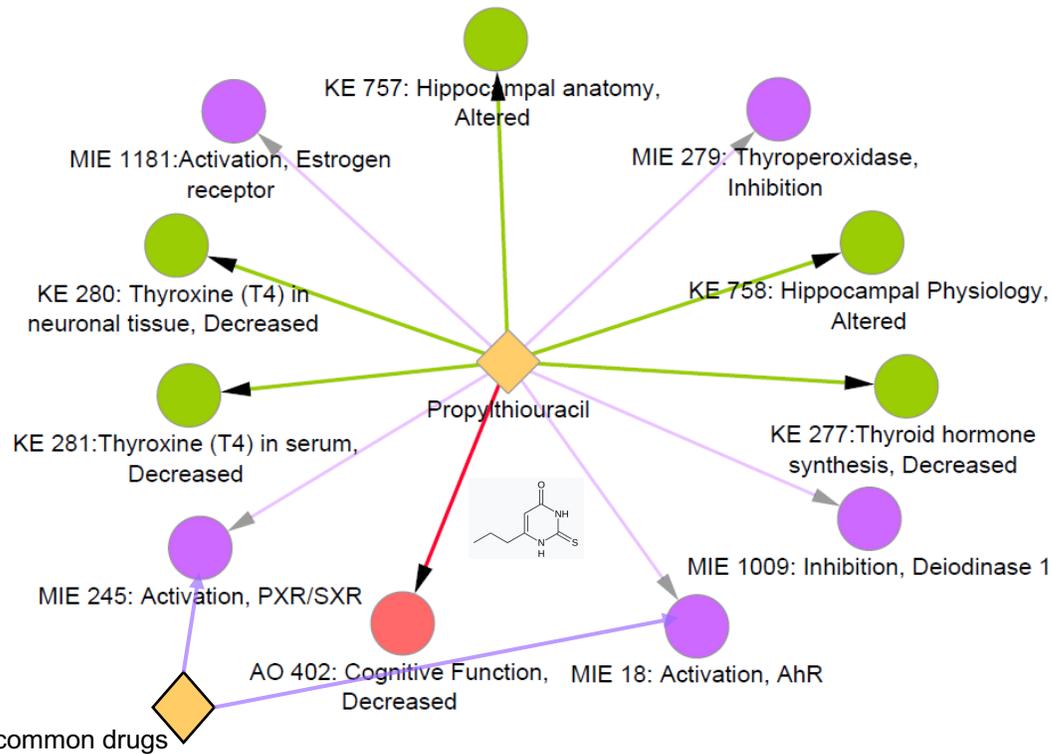
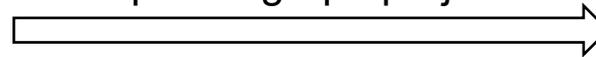
Loss of coverage for nodes



Loss of coverage for edges

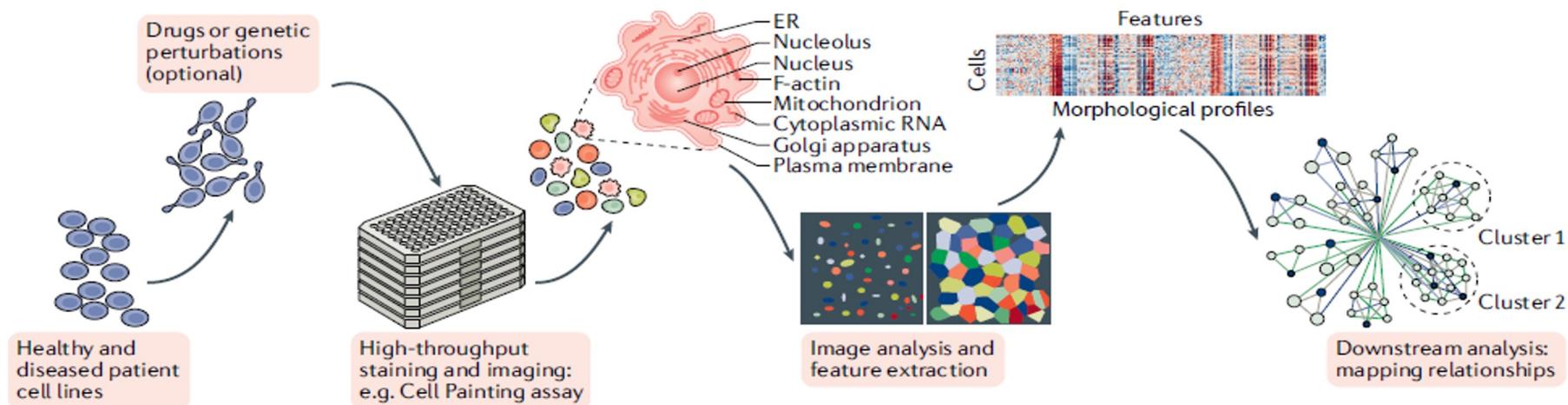


Monopartite graph projection



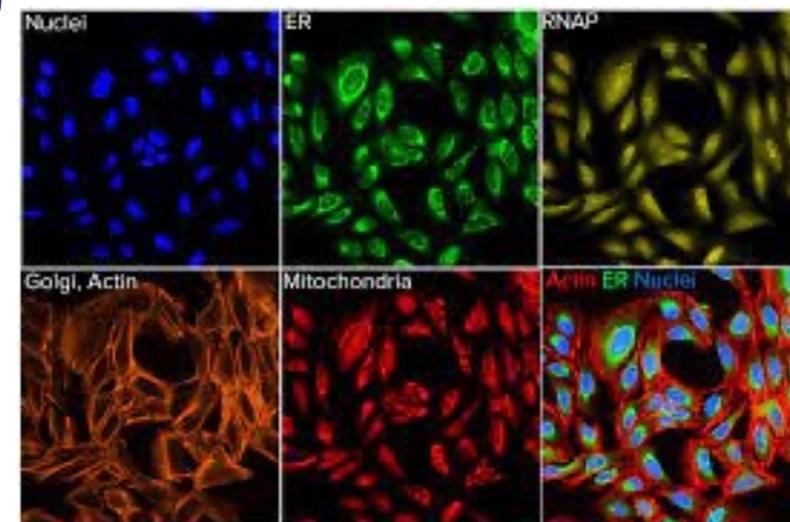
Prioritize the relation between 2 events

Link the proteome perturbation by a compound leading to morphological cell changes and diseases.



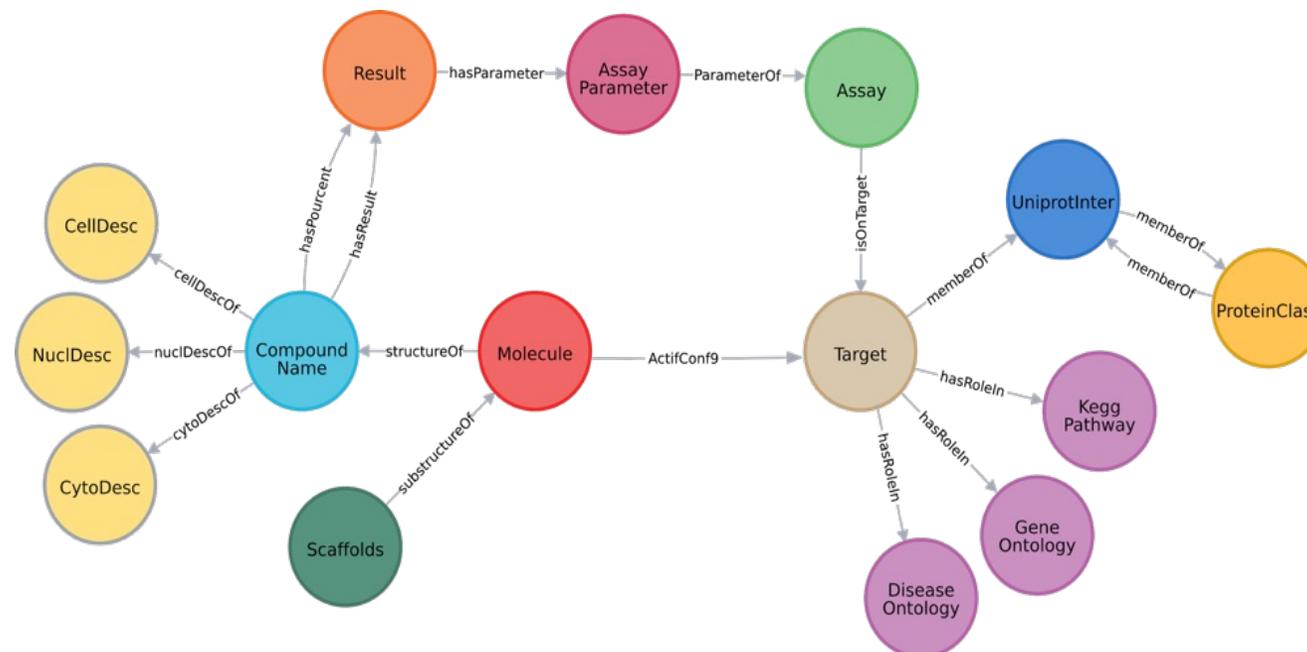
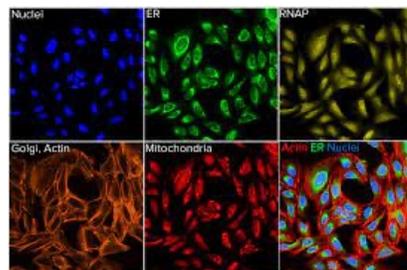
High Content imaging => Morphological cell perturbation (Cell painting)

- 29935 compounds tested on osteosarcoma cells (U2OS)
- Aim to colour the 7 major cell compartments with fluorescent dyes
- Fluorescence capture and automatic image analysis
- Extraction of the cell parameters: shape, texture, intensity, *etc.* using CellProfiler (<http://cellprofiler.org/>)
- Cell Painting morphological profiling assay [Bray MA, *et al.* Nat Protoc 2016]



Link the proteome perturbation by a compound leading to morphological cell changes and diseases.

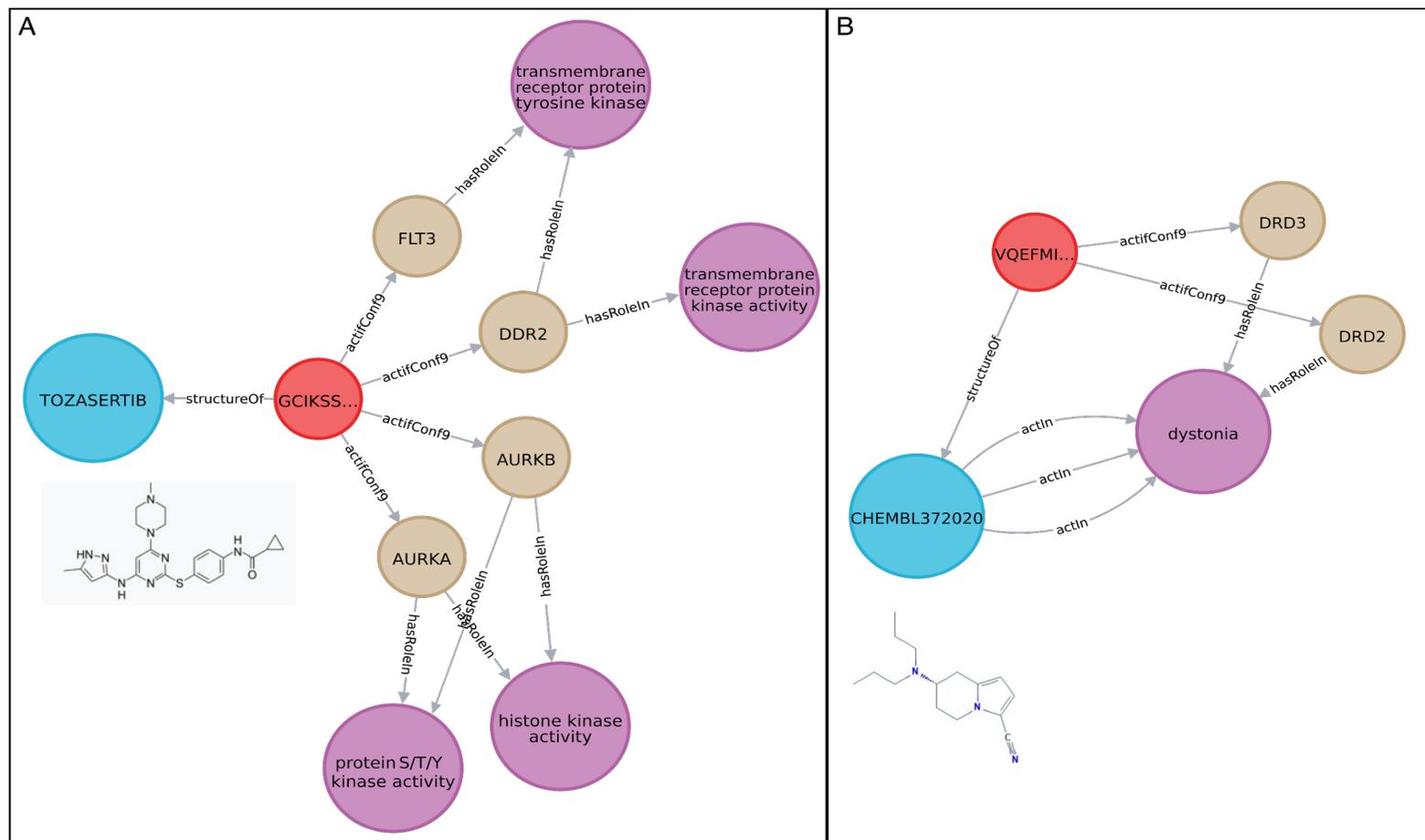
Cell Morphological features



Example 1: Ensemble of protein's target with $IC_{50} < 1 \mu M$

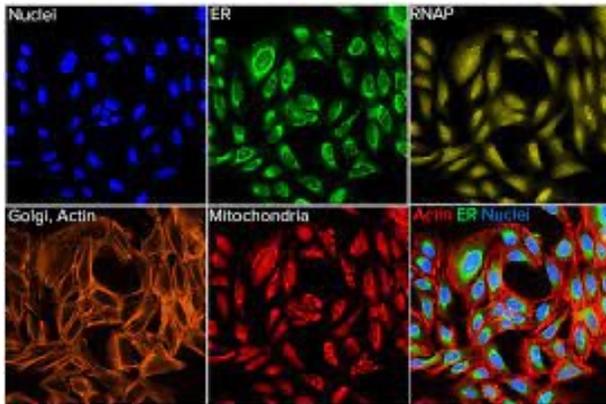


Example 2: Pathways and diseases enrichment associated to a chemical

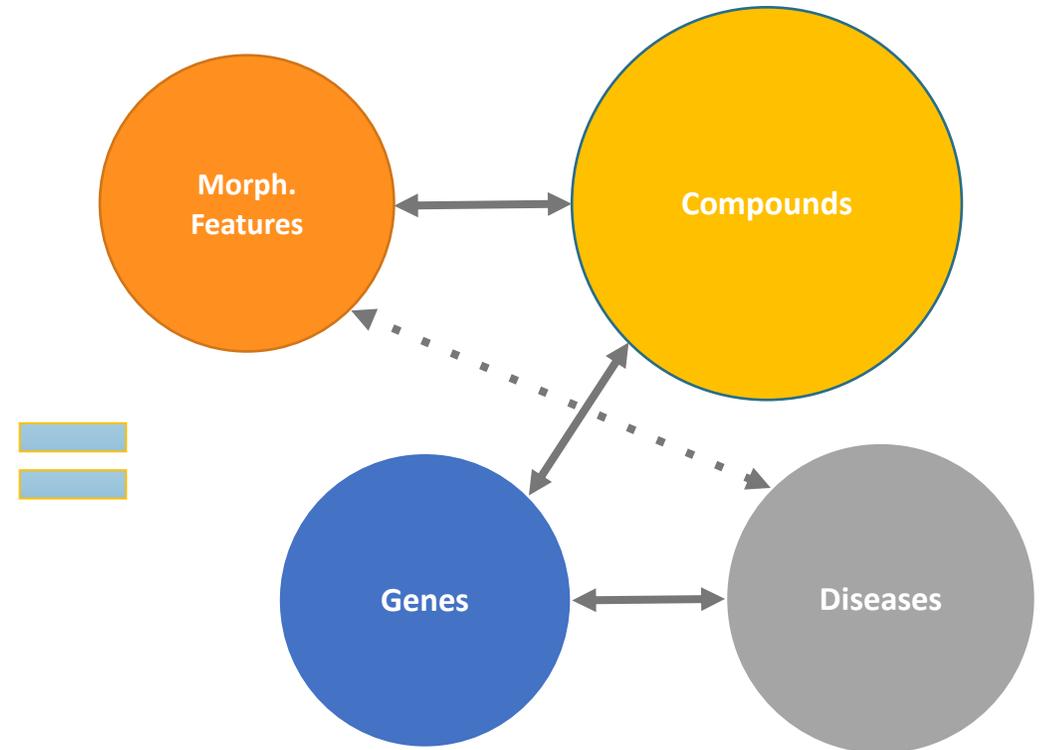
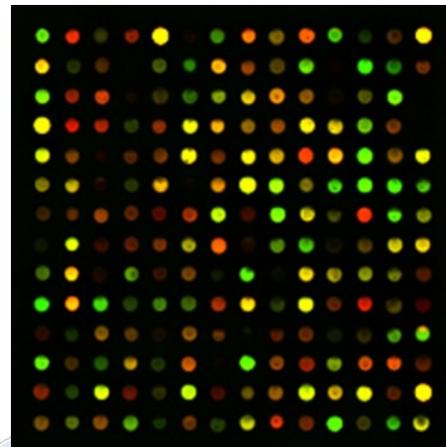


Link the gene deregulation by a compound to morphological cell changes and diseases.

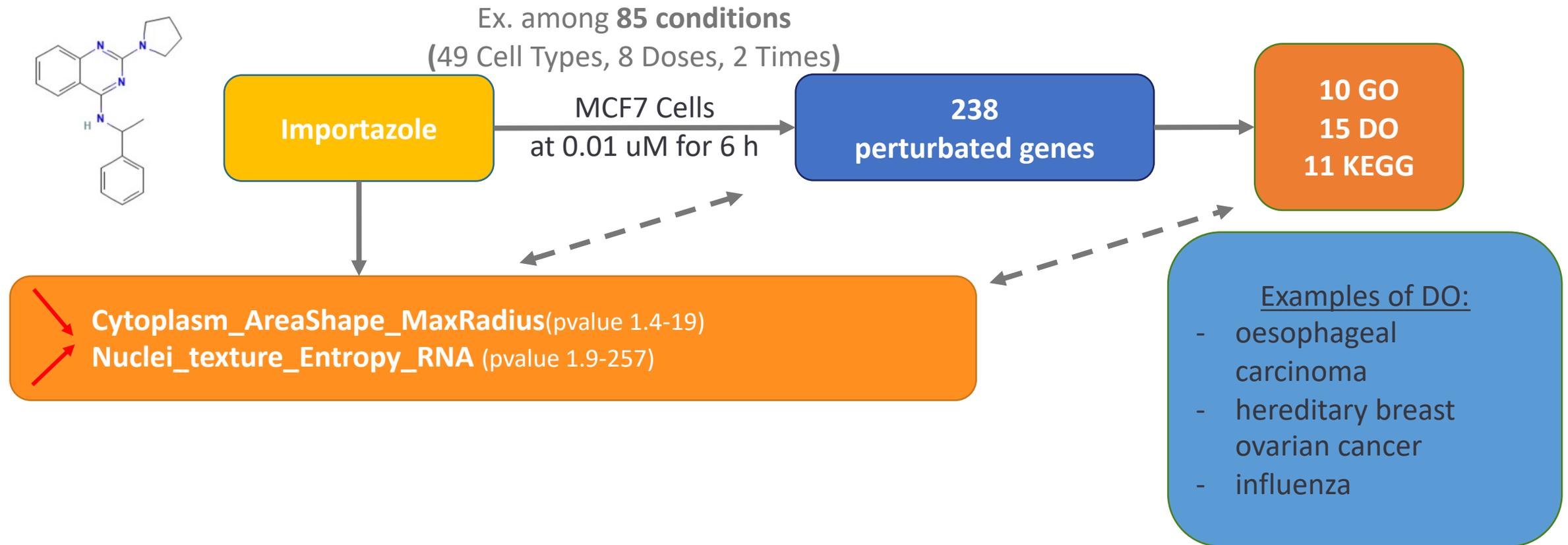
Morphological features



Transcriptomics



Link the gene deregulation by a compound to morphological cell changes and diseases.

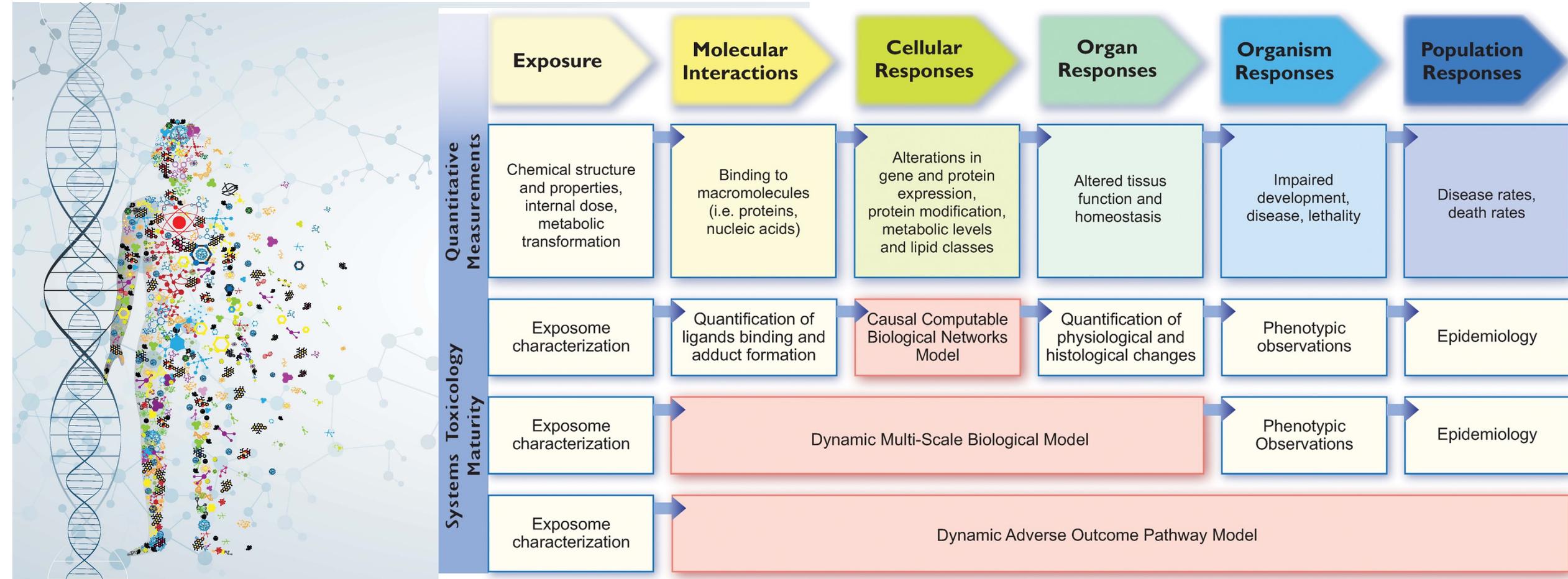


- Suggestion of a relation between the morphological perturbation of some compartments in cells and diseases

Work in progress

Conclusion

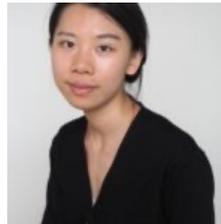
Data integration of biological information related to chemical is possible at different level of complexity (from molecular to population).



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Questions?