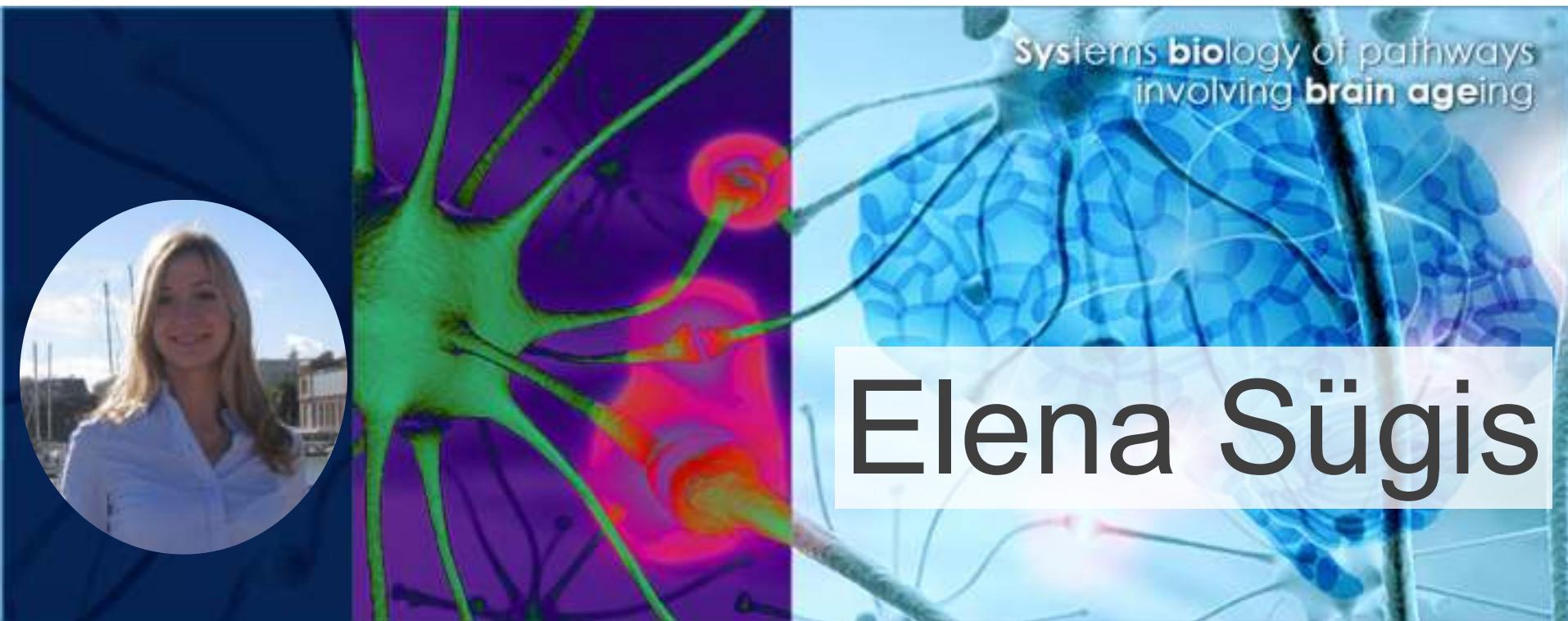


Decoding ageing brain

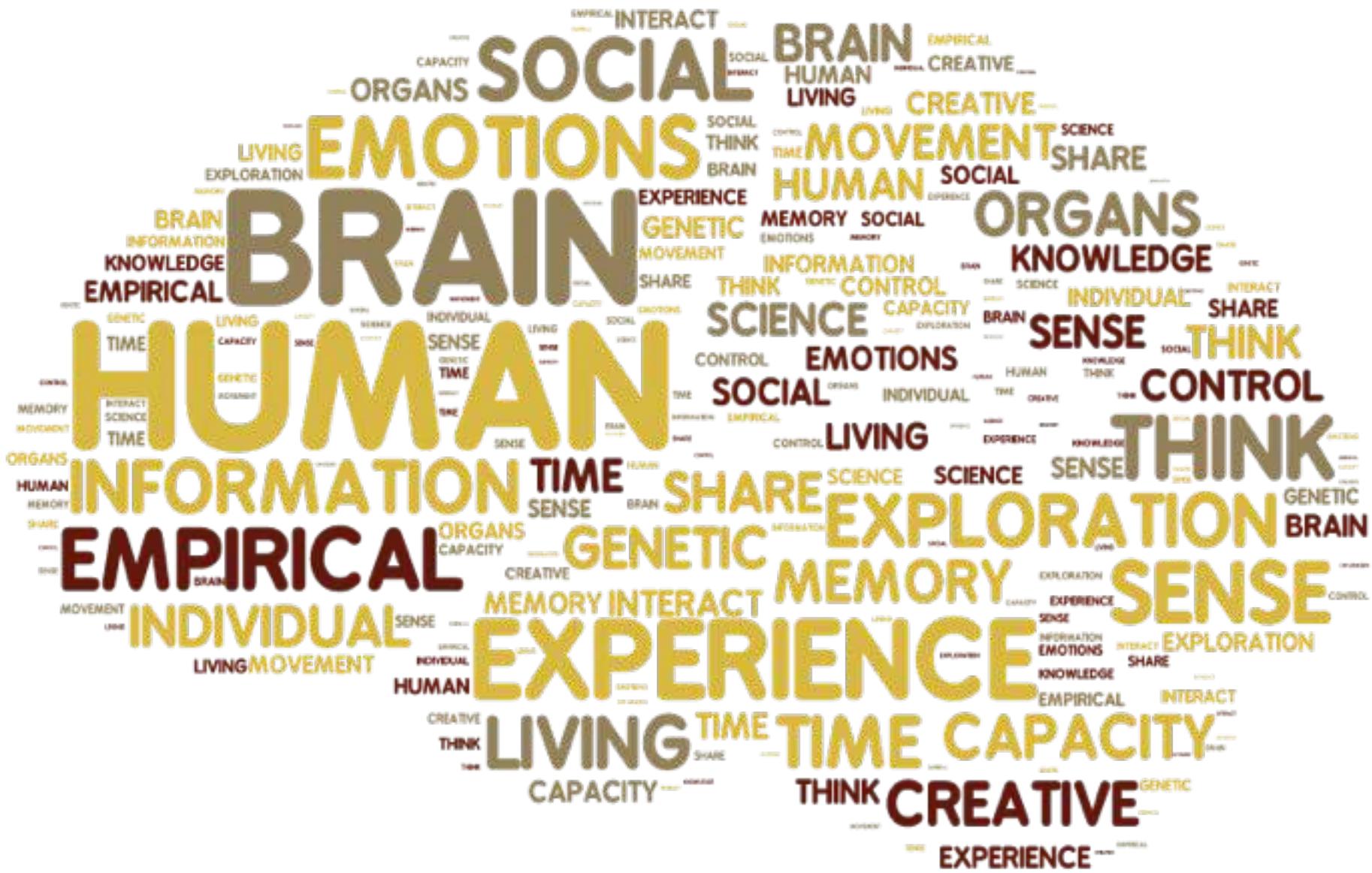


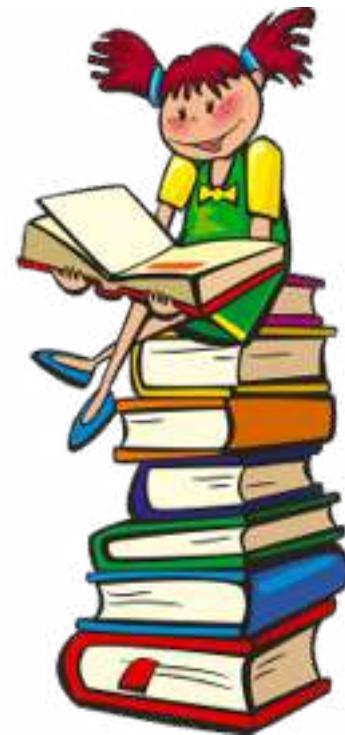
Elena Sügis



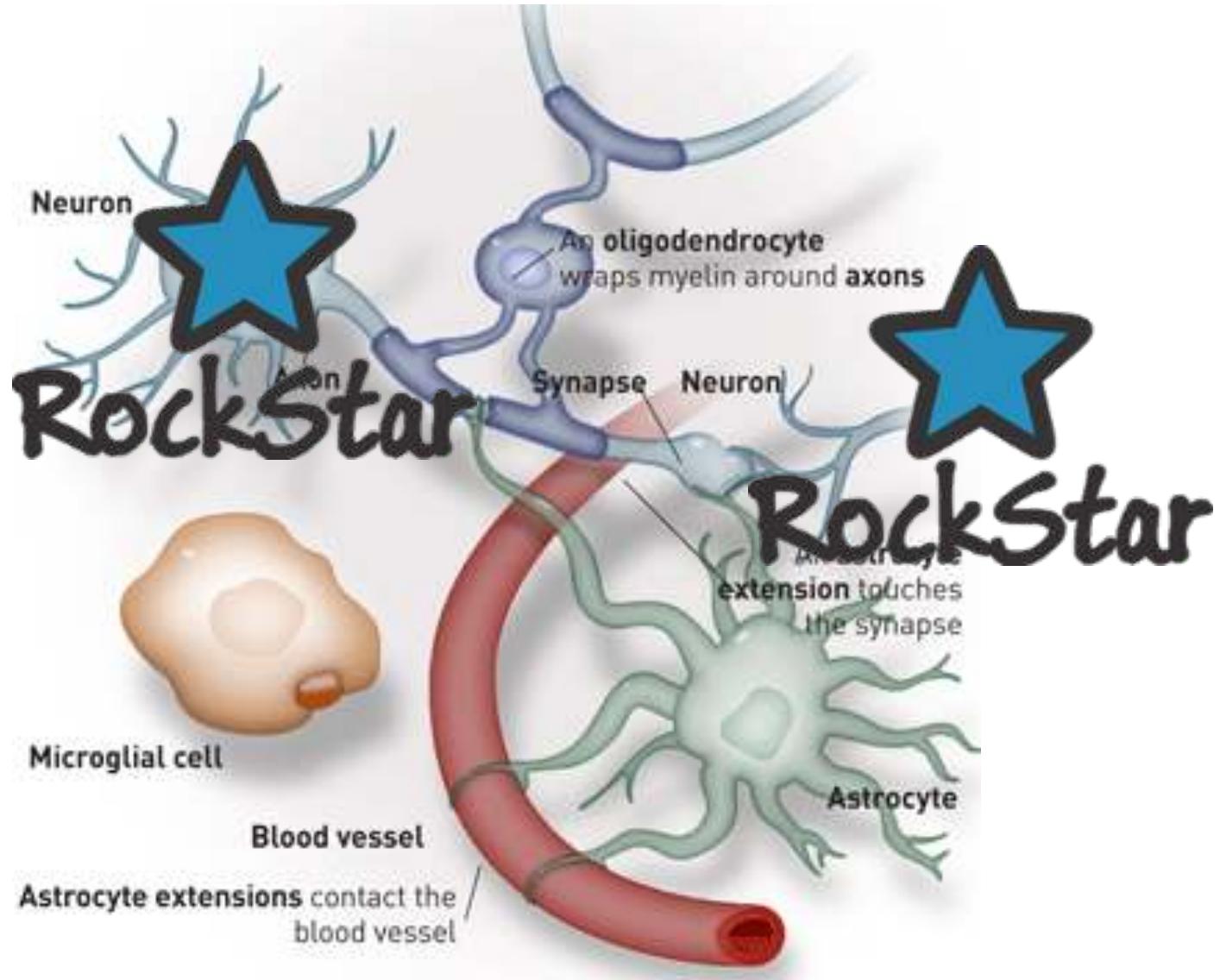
UNIVERSITY OF TARTU
Institute of Computer Science

Brain words



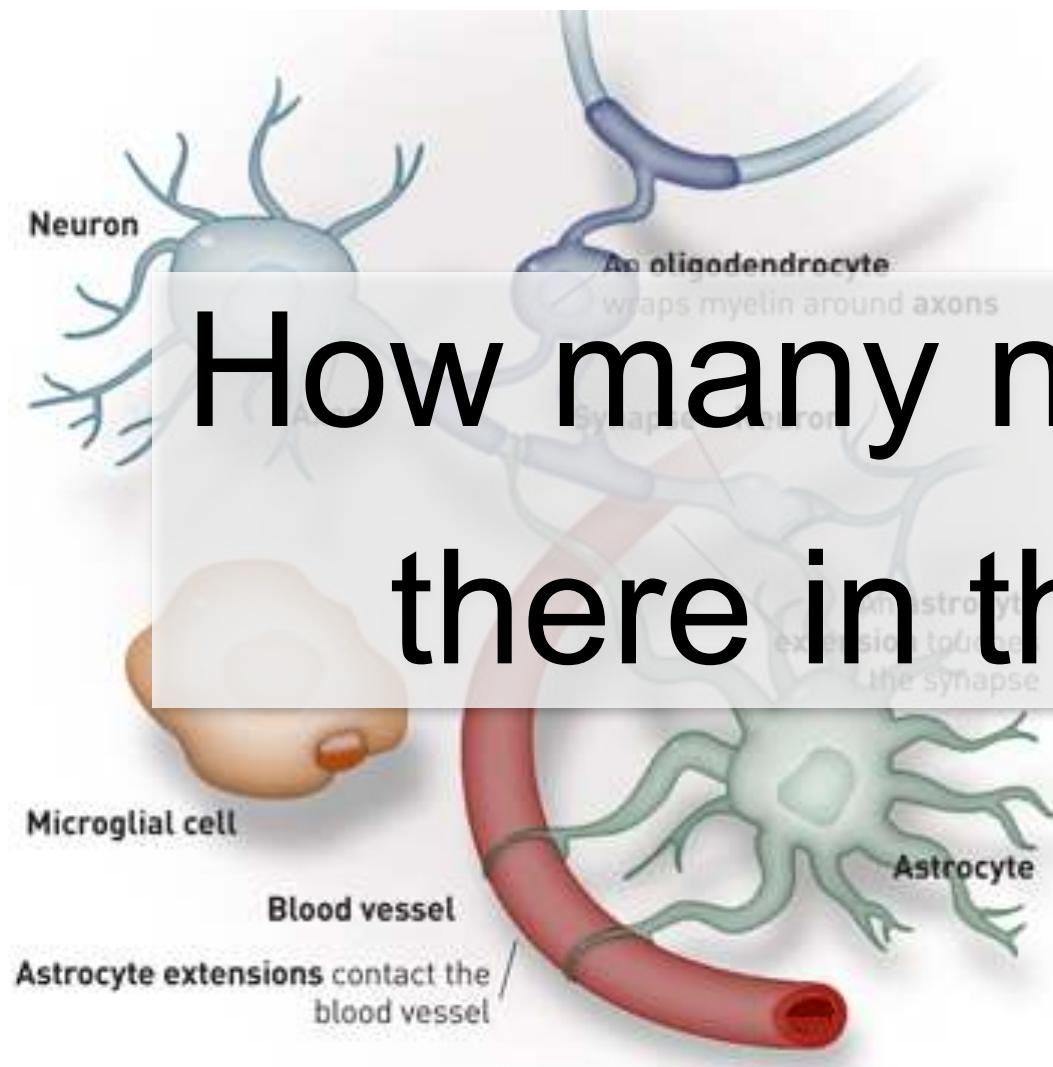


Inside the brain



Brain is extremely complex structure

How many neurons are there in the brain?



majority of nerve cells in our brain and spinal cord are born before birth and will last a lifetime



All might kill neurons

Stress



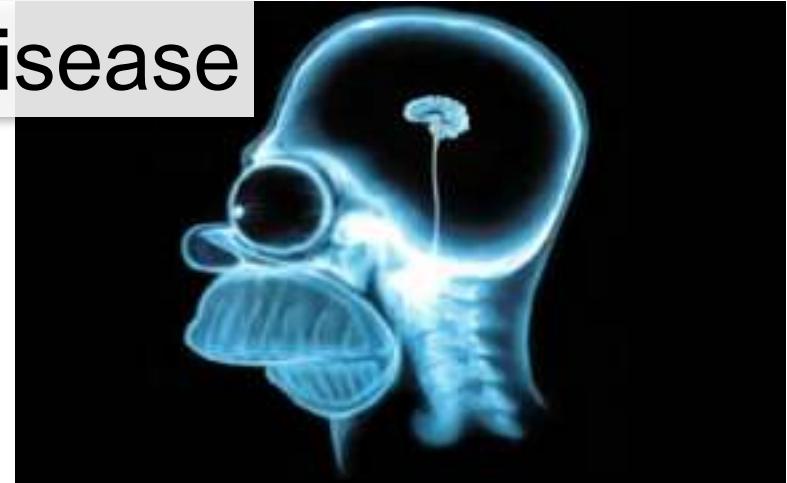
Sleep deprivation



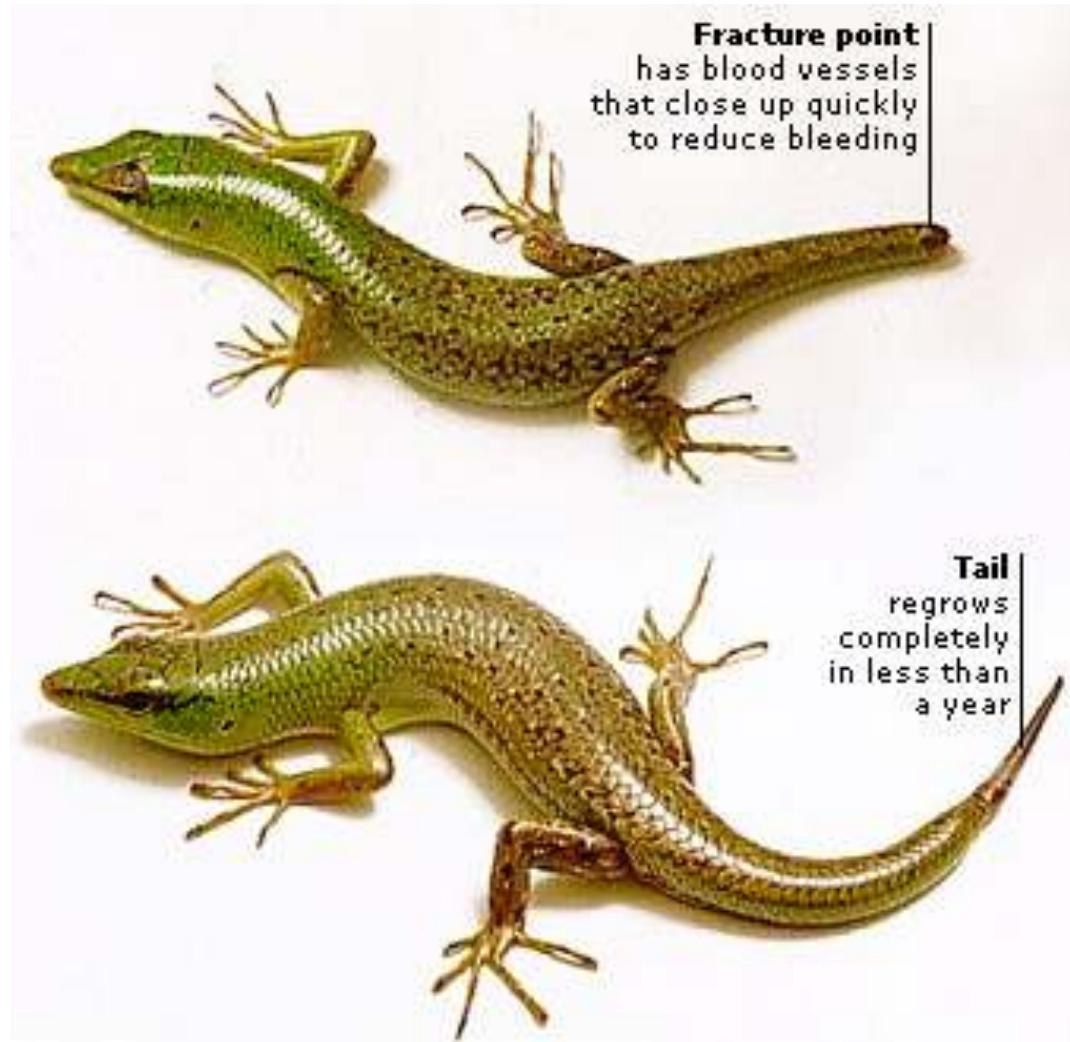
Ageing



Disease



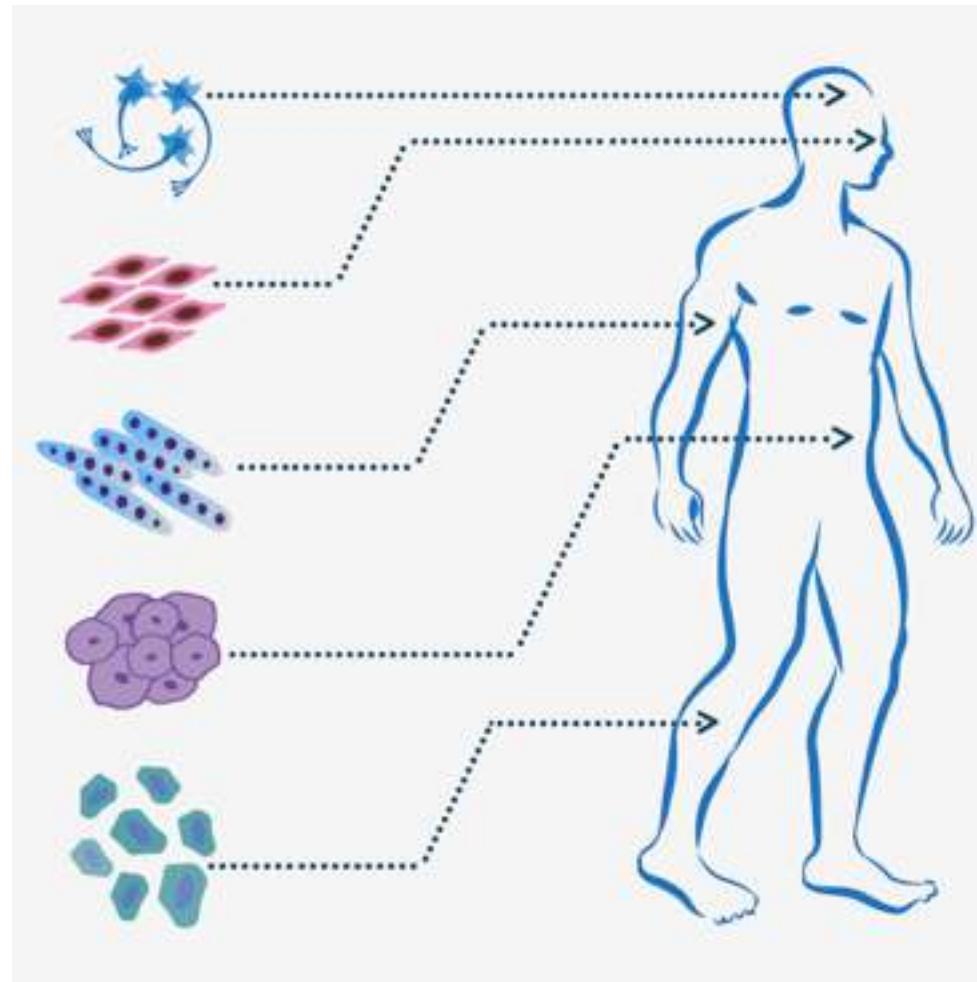
Regeneration in animals



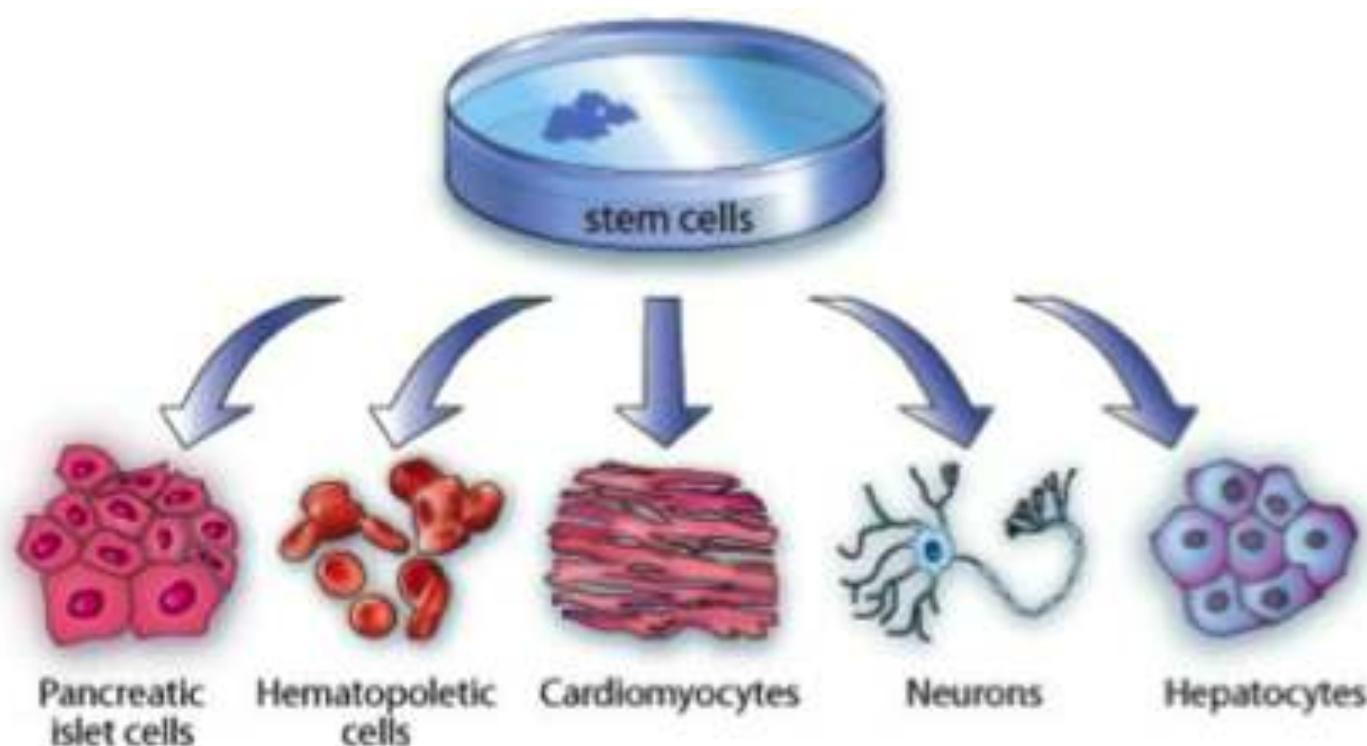
Ability to regenerate

Ability of our body to repair the damaged parts

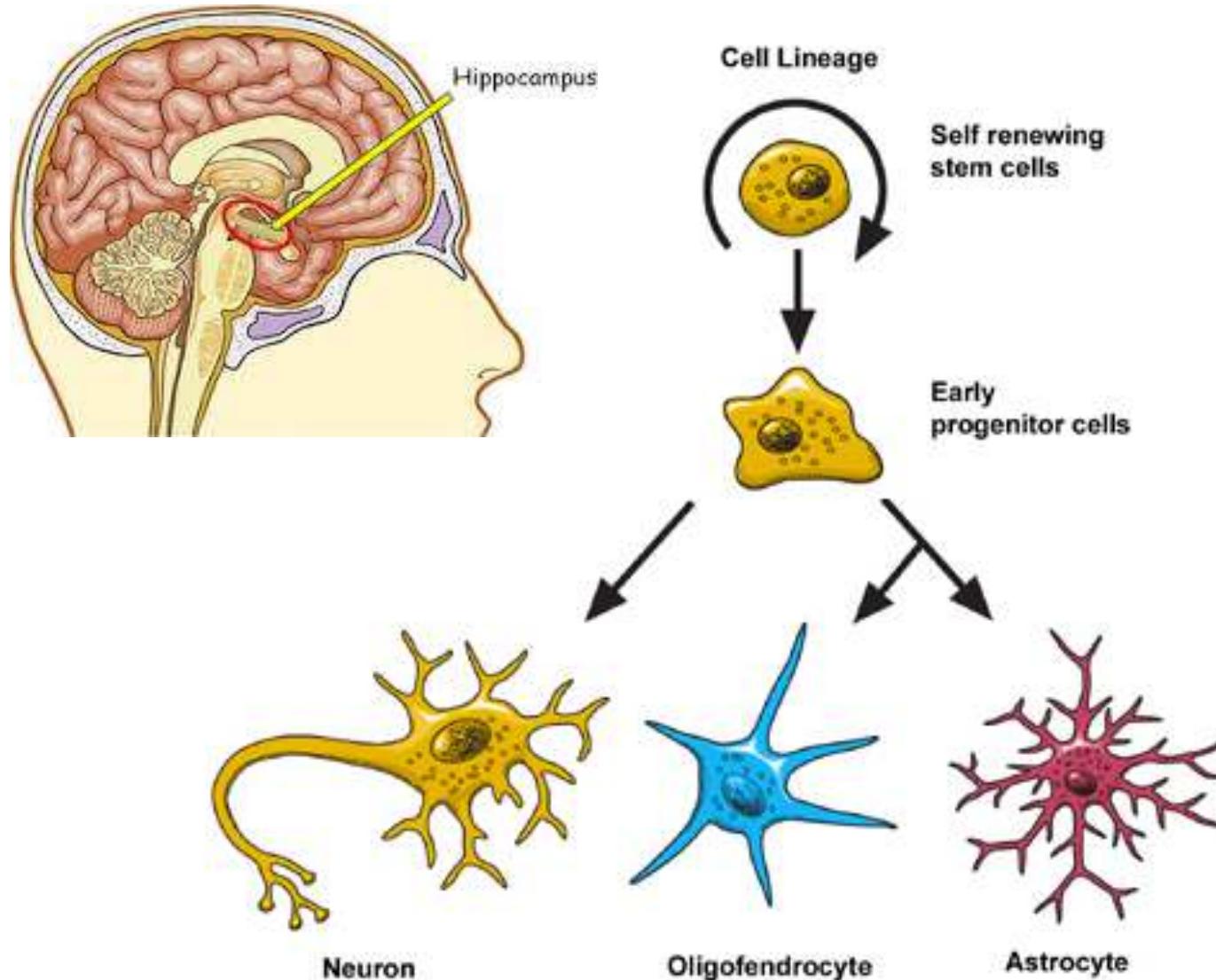
- Skin
- Bones
- Liver
- Etc.



Stem Cells



Neural Stem Cells



Why are they important? Curing with stem cells

- Brain injury
- Neurodegenerative disorders (Parkinson's, Huntington's, Alzheimer's)
- Other brain diseases (multiple sclerosis, etc.)

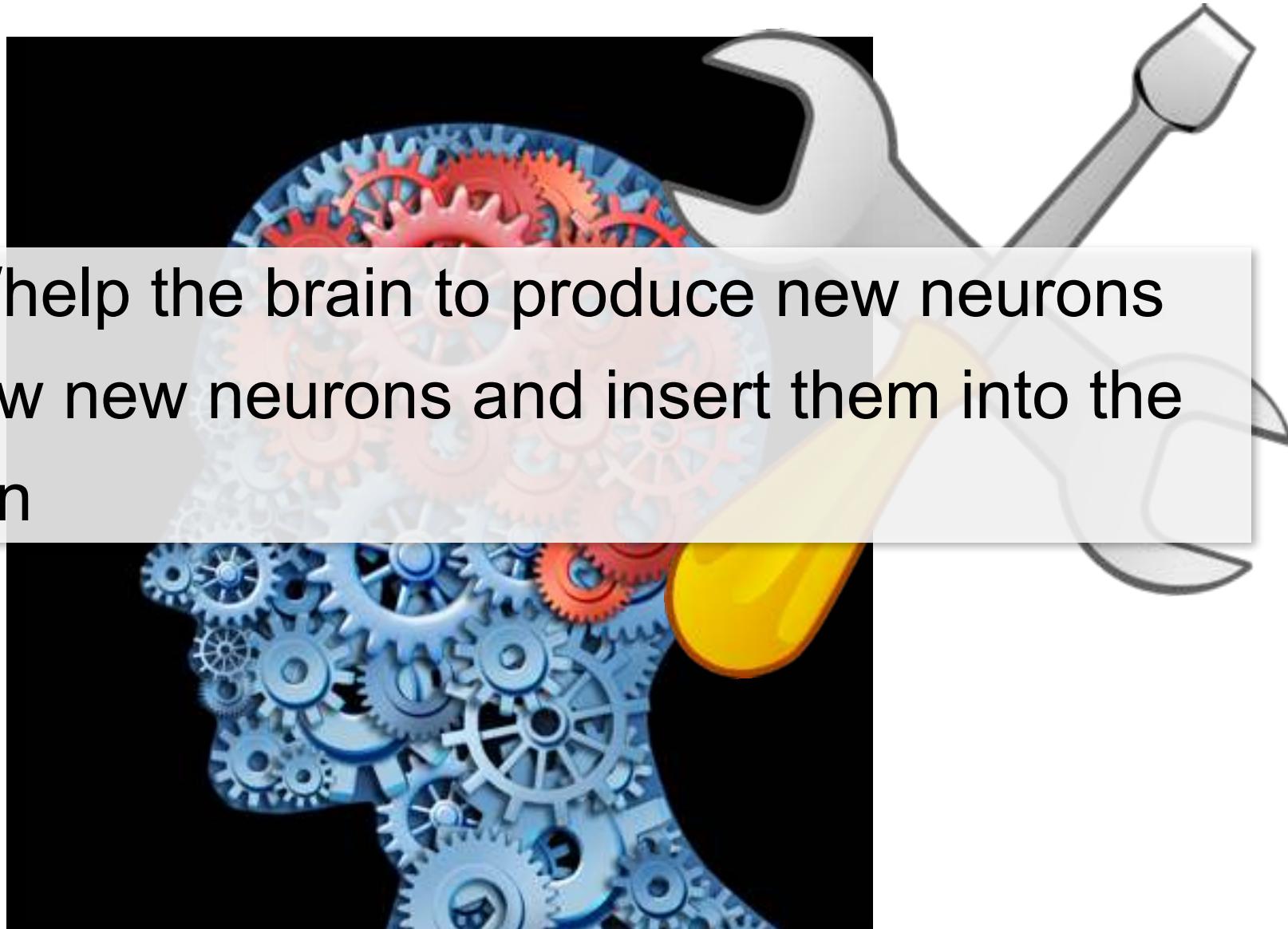


Brain own capacity is not enough



What can we do?

- Tell/help the brain to produce new neurons
- Grow new neurons and insert them into the brain



Reprogramming body cells



Conferences Calendar | Cell Symposia | Jobs | Cell DNA |
Iphone App

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Cell, Volume 126, Issue 4, 663-676, 25 August 2006

doi:10.1016/j.cell.2006.07.024

Article

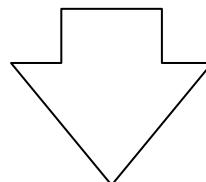
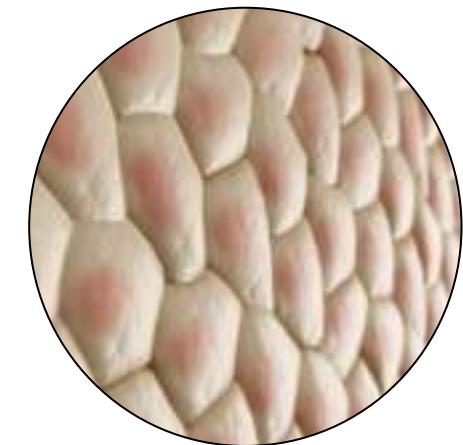
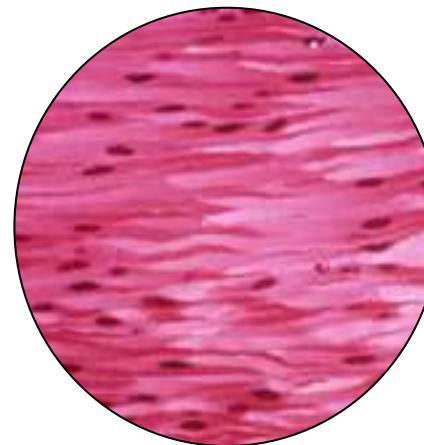
Induction of Pluripotent Stem Cells from Mouse Embryonic and Adult Fibroblast Cultures by Defined Factors

Kazutoshi Takahashi¹ and Shinya Yamanaka^{1, 2,}  

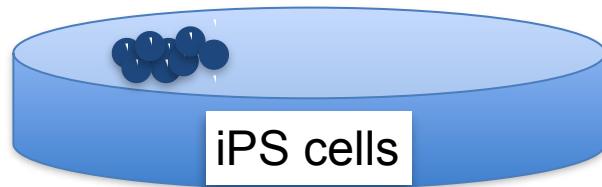
¹ Department of Stem Cell Biology, Institute for Frontier Medical Sciences, Kyoto University, Kyoto 606-8507, Japan

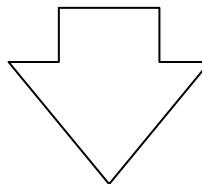
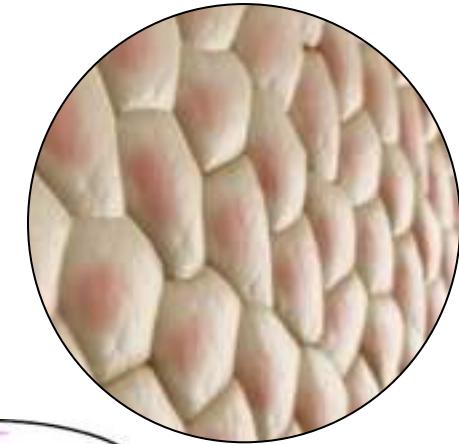
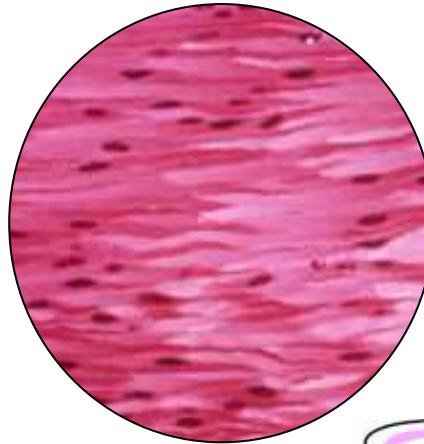
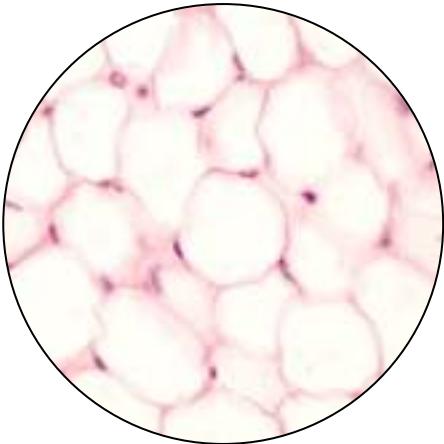
² CREST, Japan Science and Technology Agency, Kawaguchi 332-0012, Japan

Cell reprogramming



+ Oct4
+ Sox2
+ Klf4
+ c-Myc

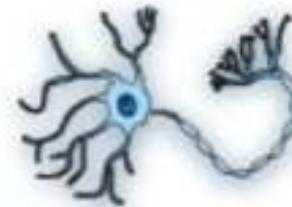




+ Oct4
+ Sox2
+ Klf4
+ c-Myc



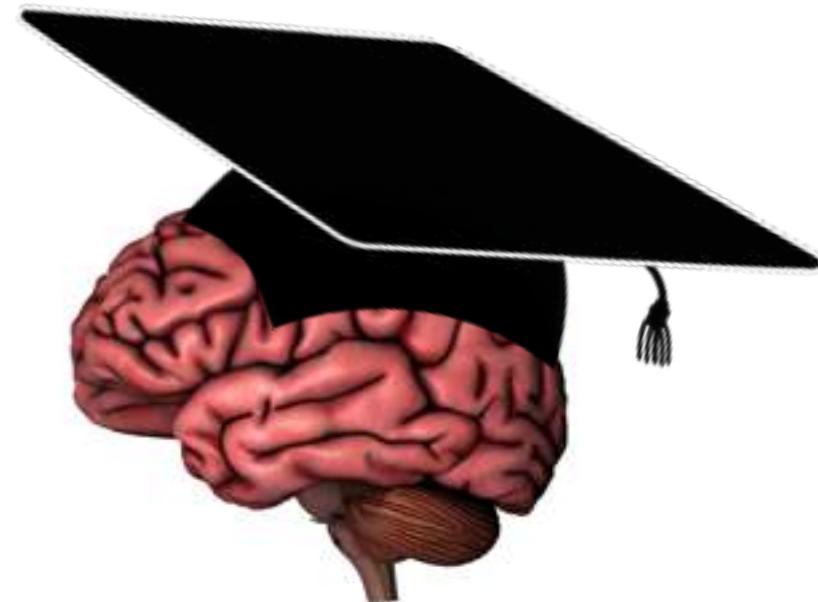
NEURONS



Silly brain



Smart brain



Neuroplasticity

A)



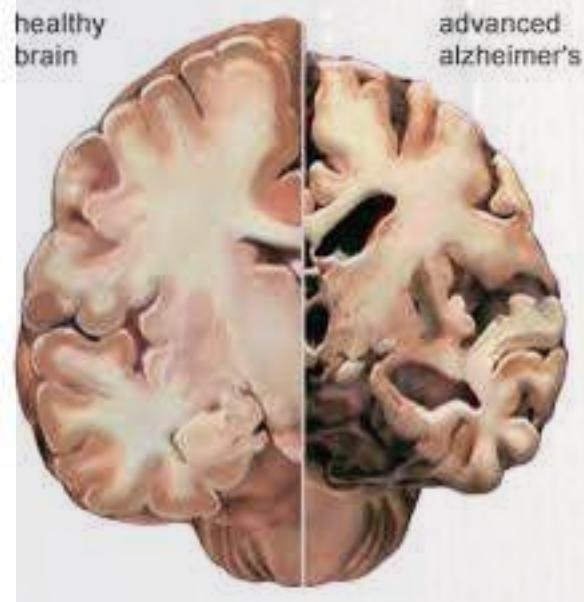
B)



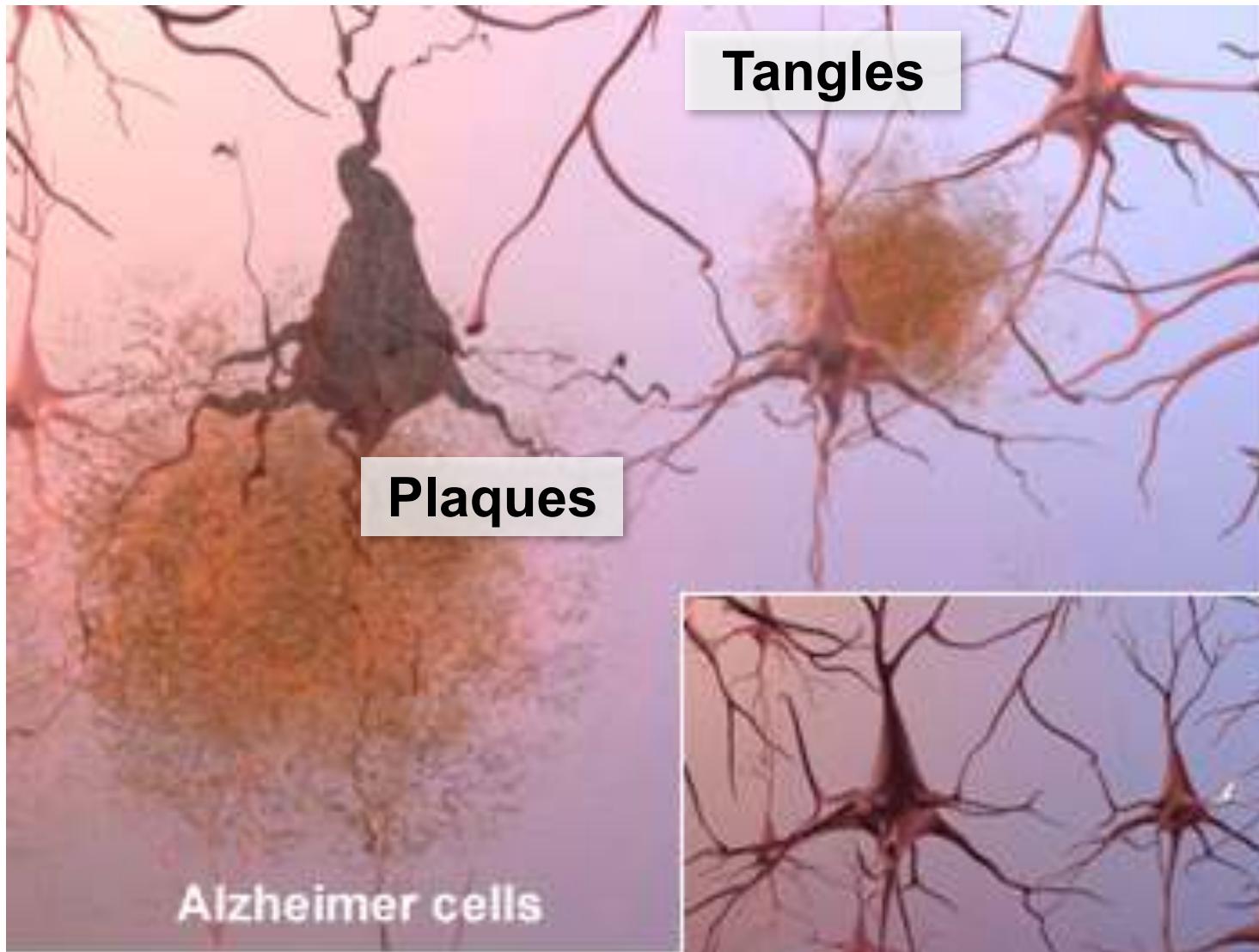
- A) A segment of one brain hemisphere is damaged (shown in red), resulting in a loss of a particular function.
- B) Over time, the opposite hemisphere can take over the lost function in the damaged hemisphere (shown in green).

Ageing & Alzheimer's

Nearly 50% of adults over 85 afflicted with Alzheimer



Look from inside



Alzheimer cells



healthy cells



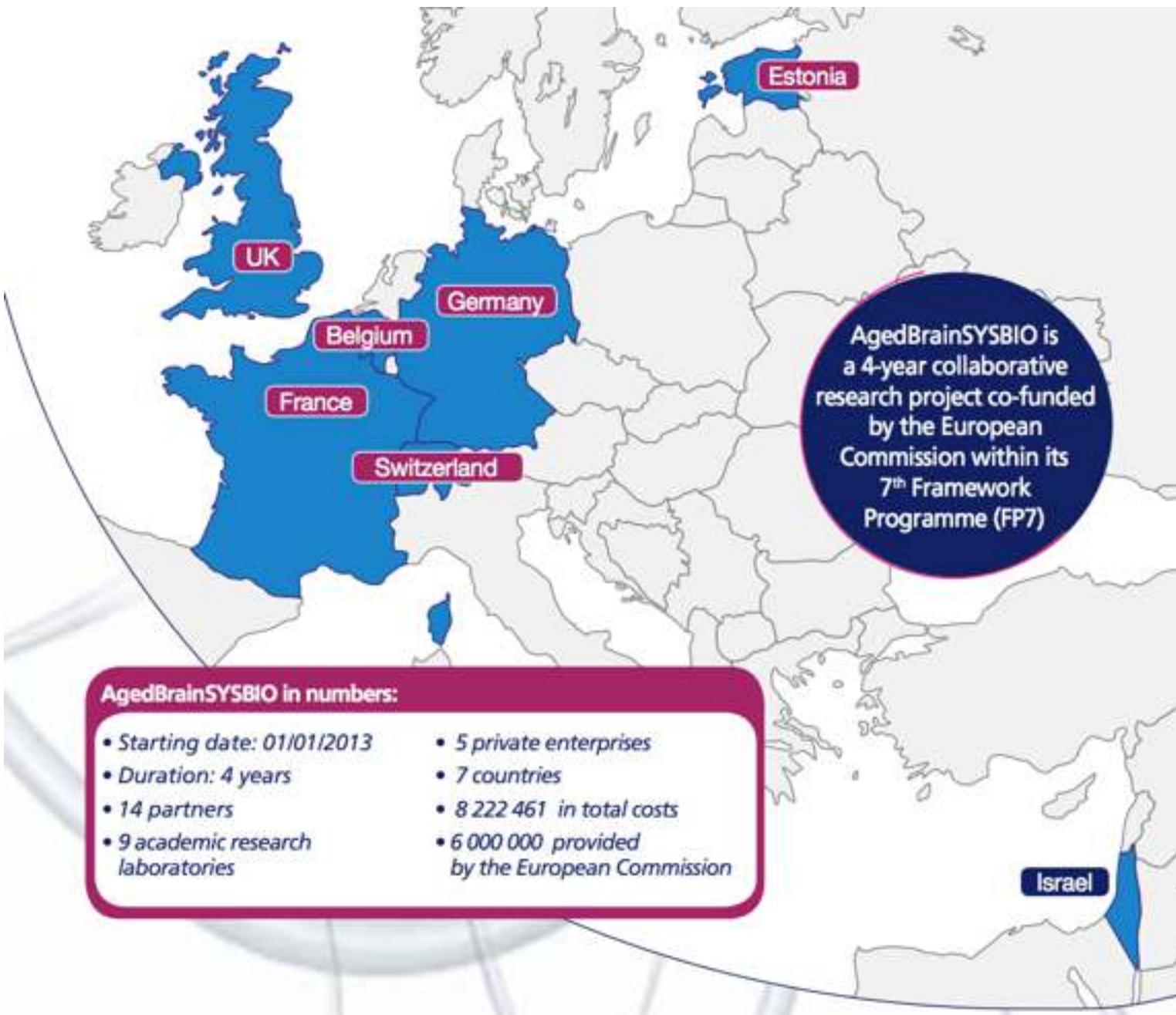
What do we do to understand?

- Address the basis of brain ageing in normal and in disease conditions
- Identify and validate new drug targets for Alzheimer's Disease

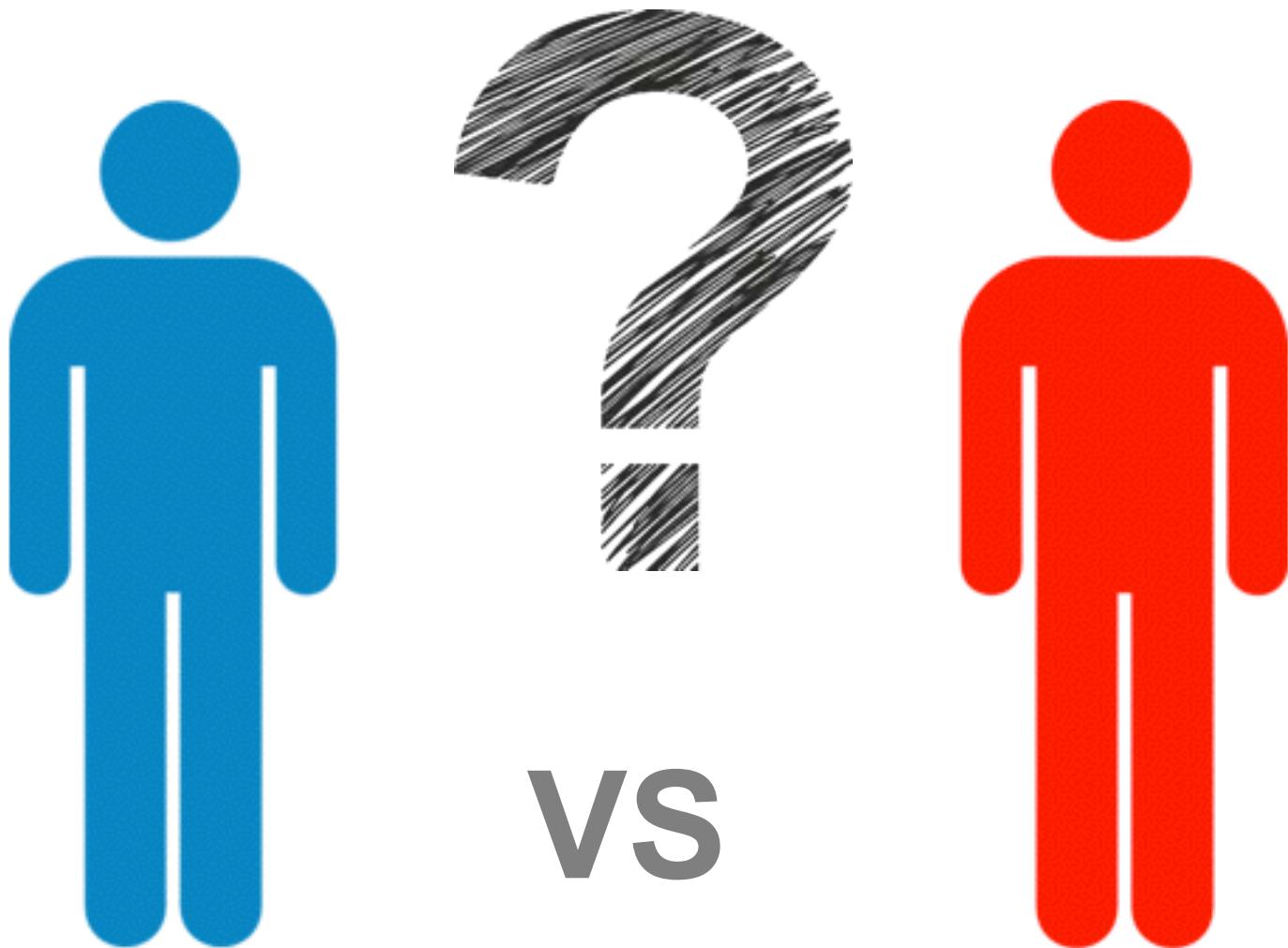


QureTEC

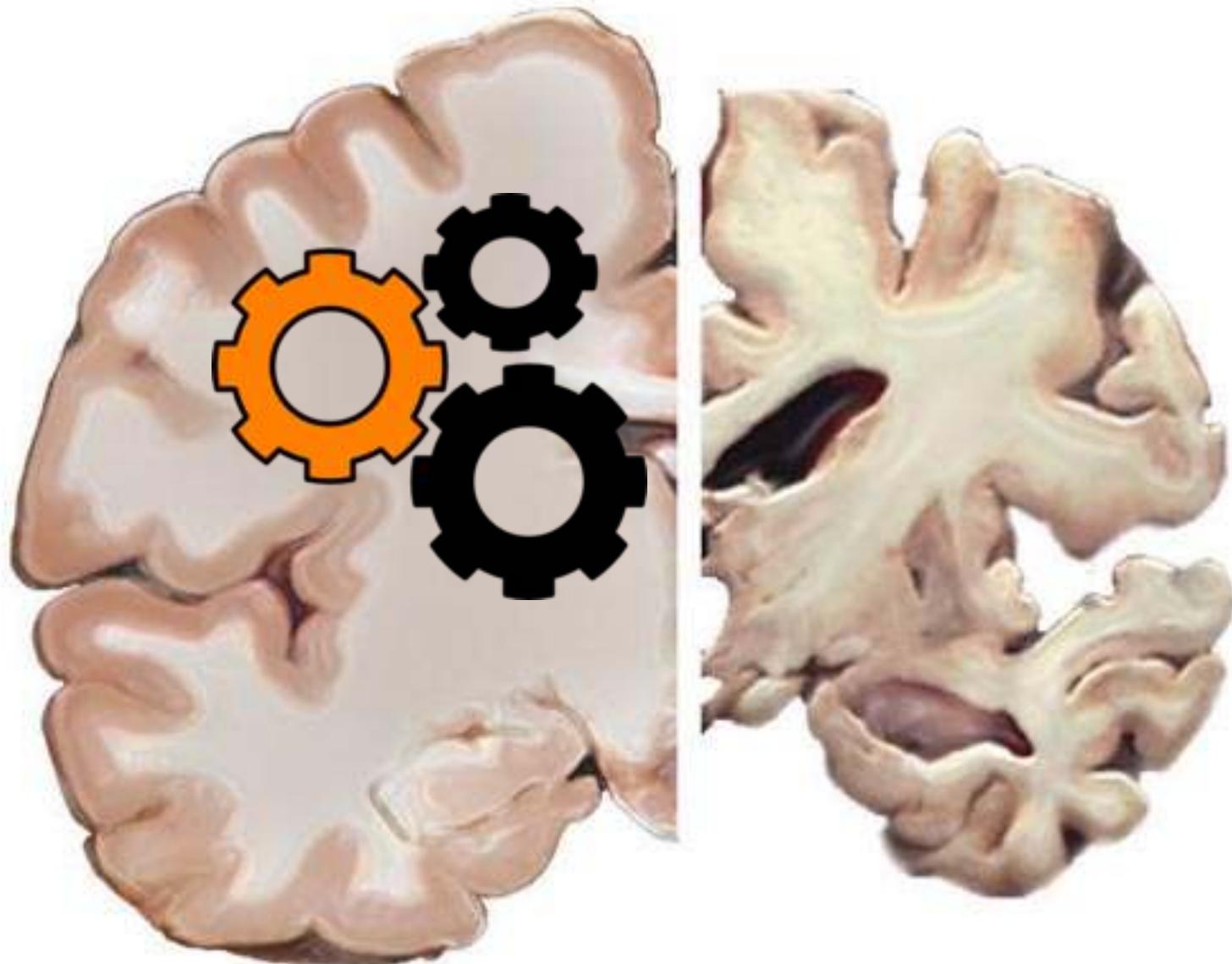




We want to understand



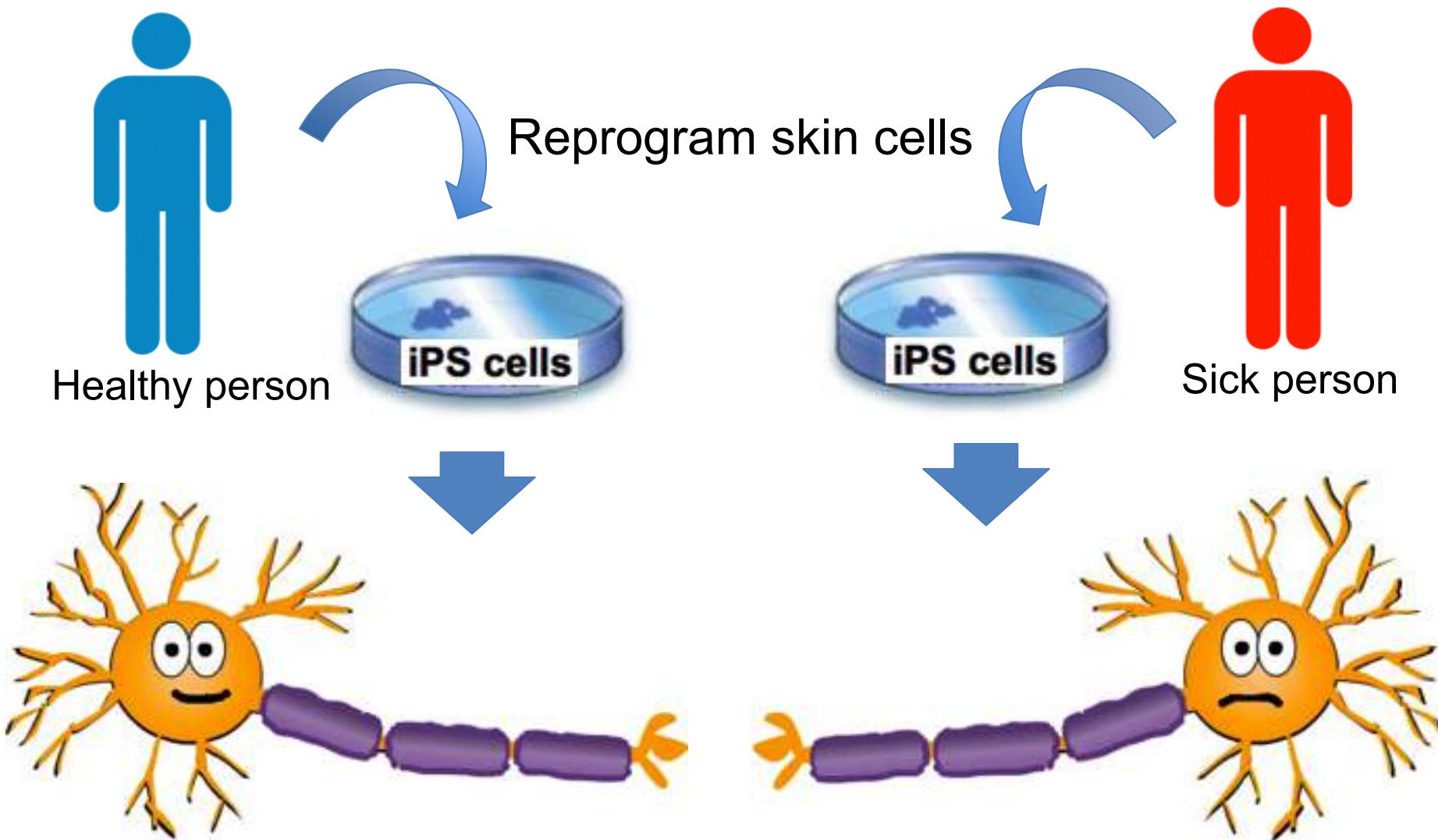
We want to understand



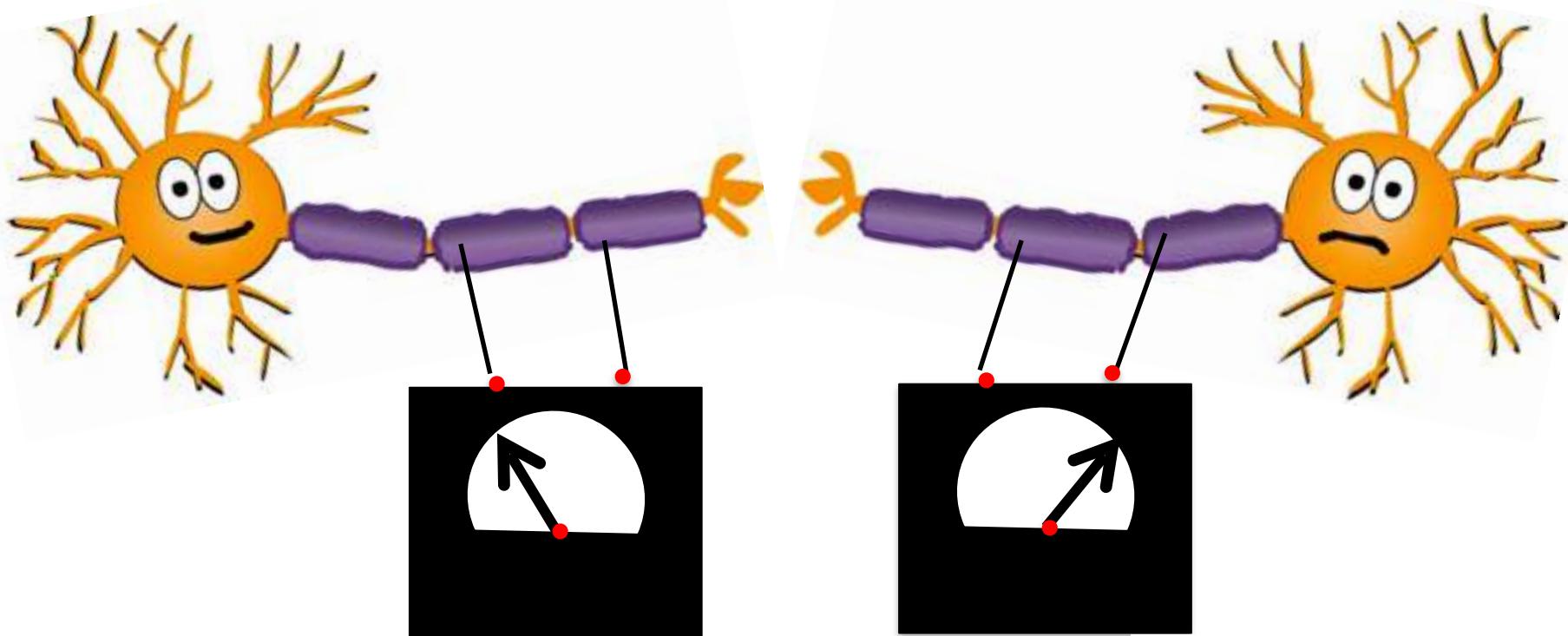
Experiments



Reprogram cells to study disease



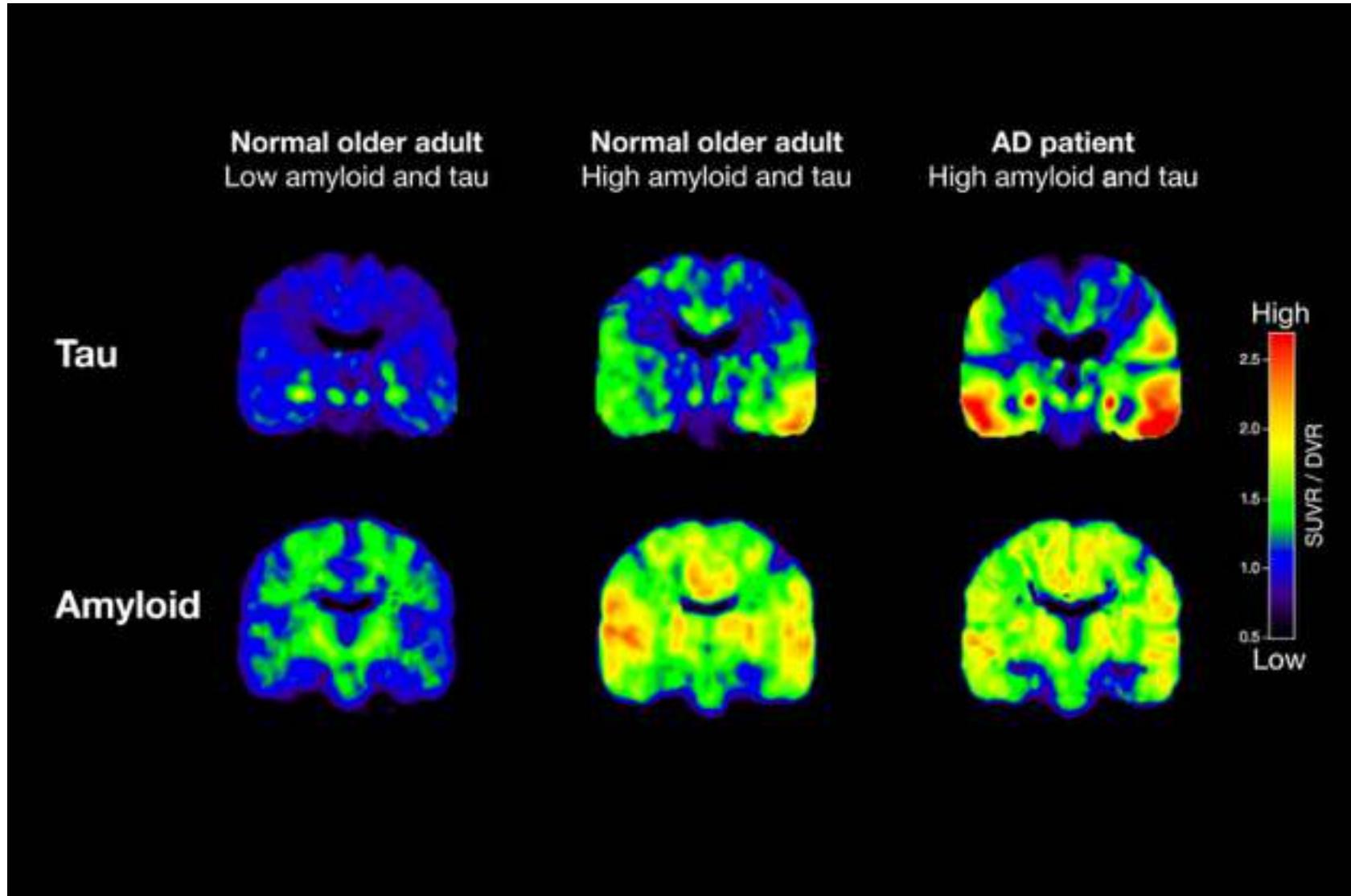
Understand the difference



Statistical power
Computational tools



Neuroimaging



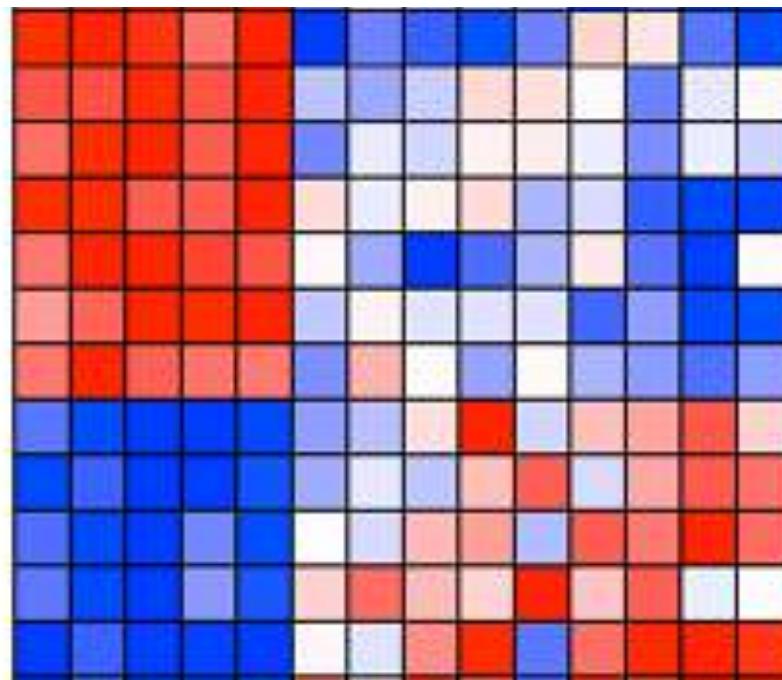
Important genes



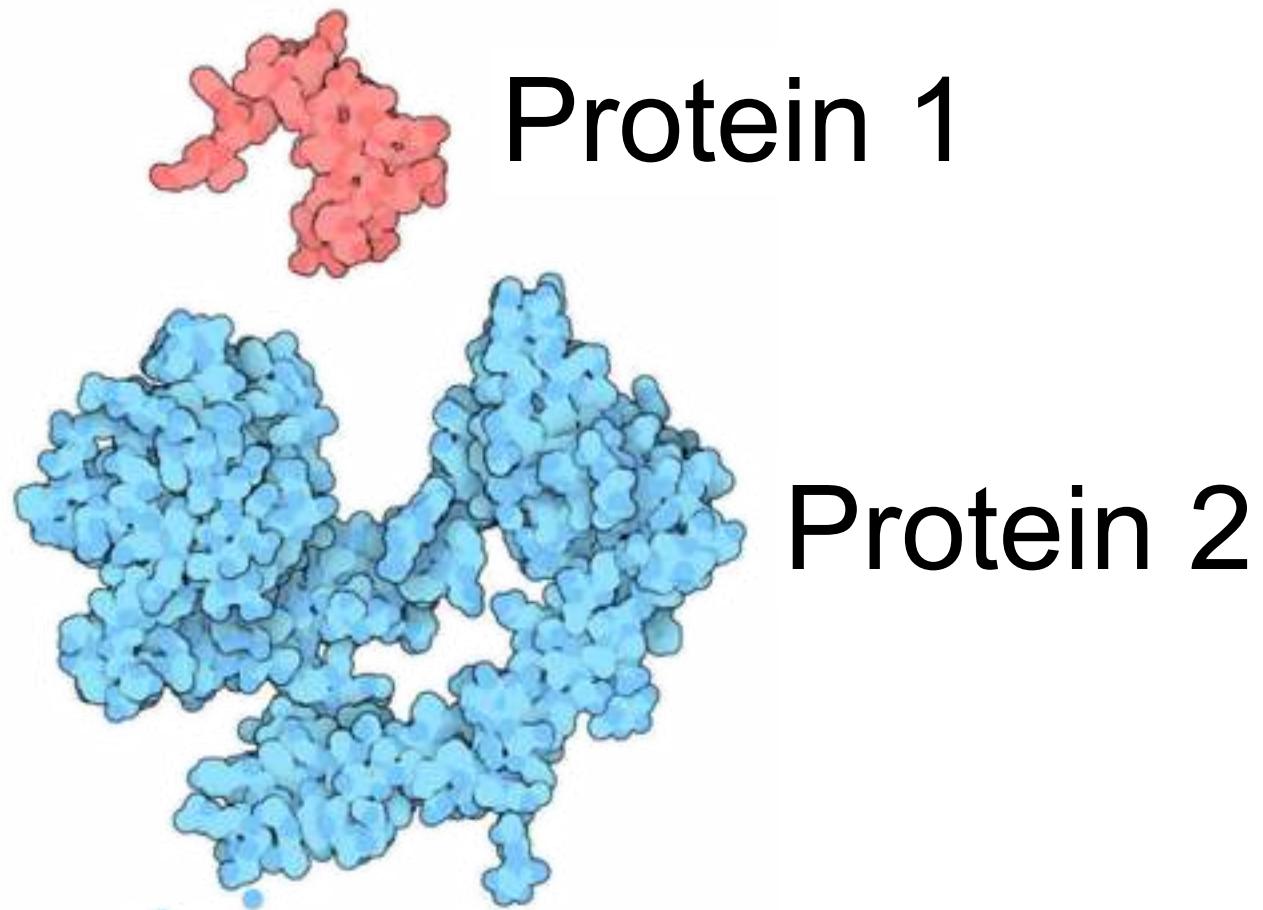
Gene 1

:

Gene n



Protein-protein interactions



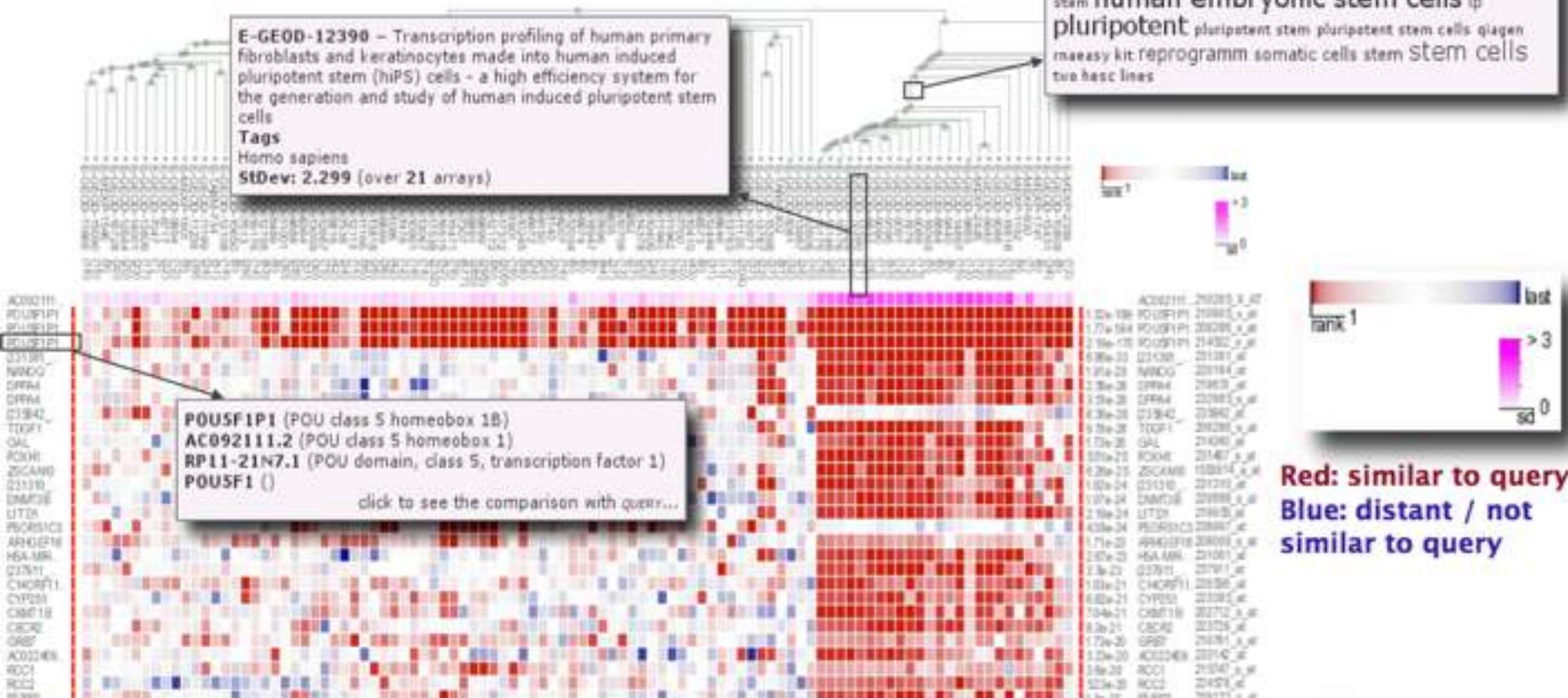
Co-expression

Results

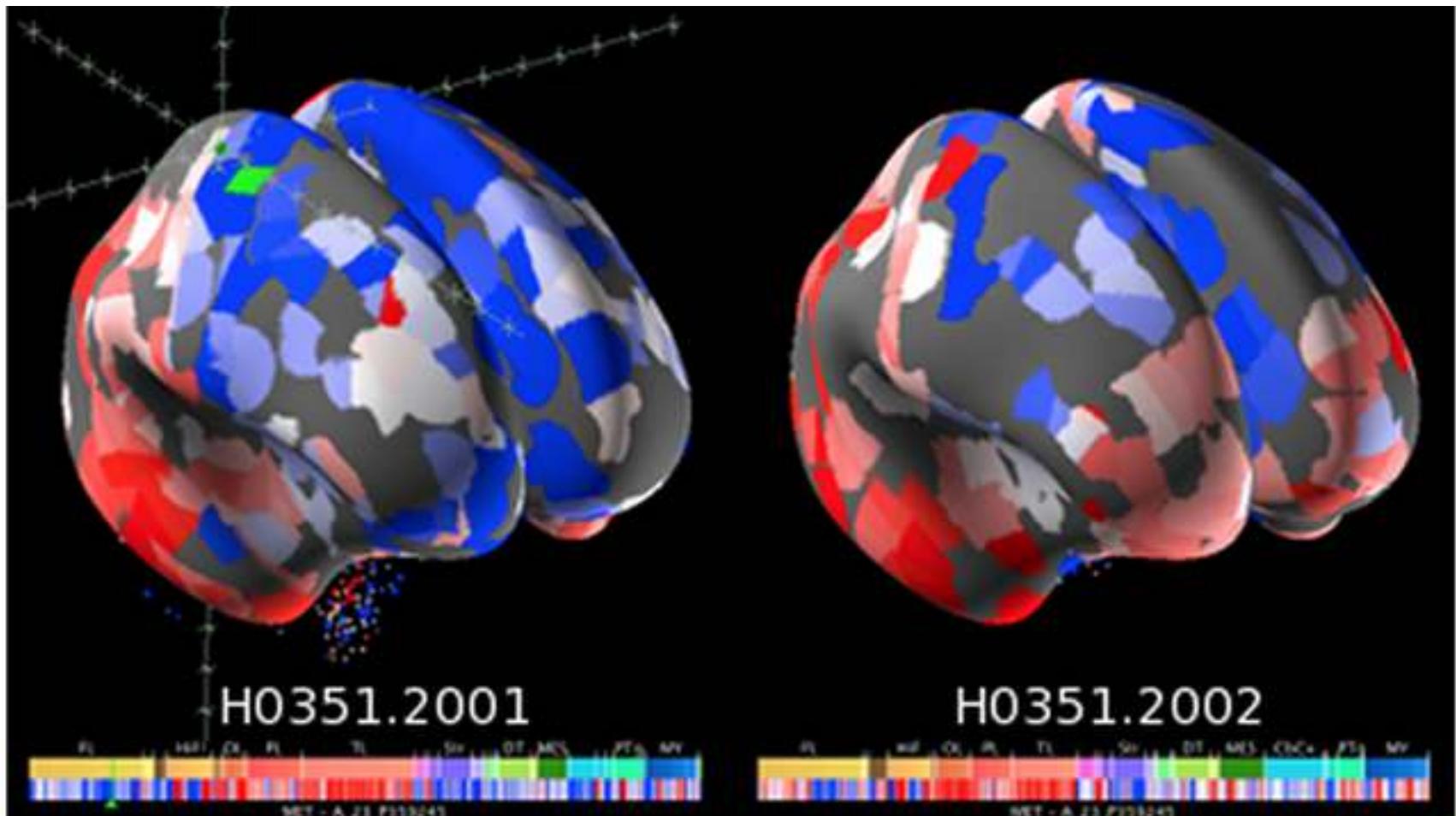
[?] | g:Profiler(GOSt) | ExpressView | URLMap link | Static URL | Query details |

Handpicked datasets : ✓=+ ; ✗=- ; +~✓ ; ~-✗ ; reset all | 100 datasets used in query (661 excluded by filters)

MEM - Multi Experiment Matrix



Expression in brain regions



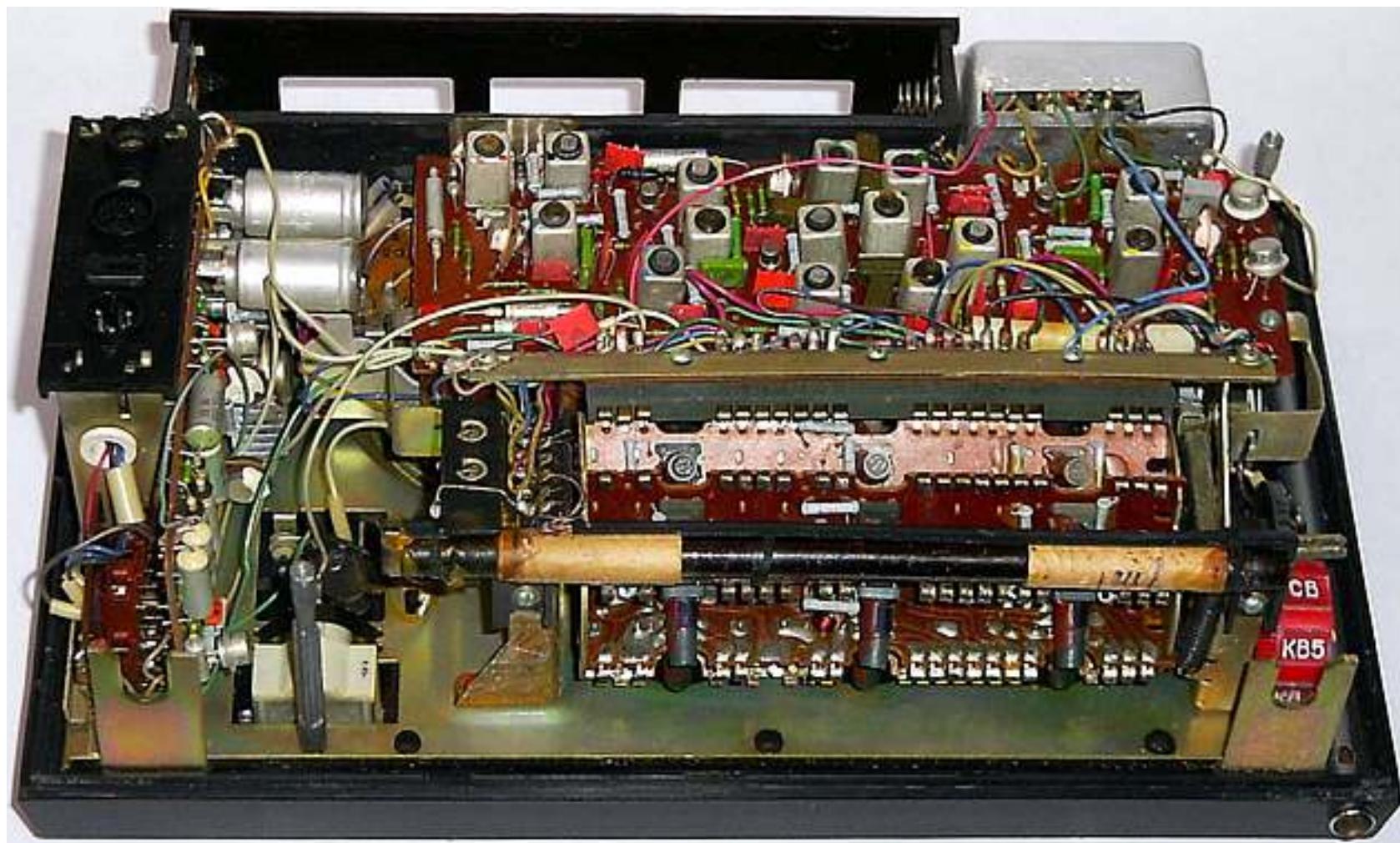
We solved it! Right?



How does it work?



How does work?

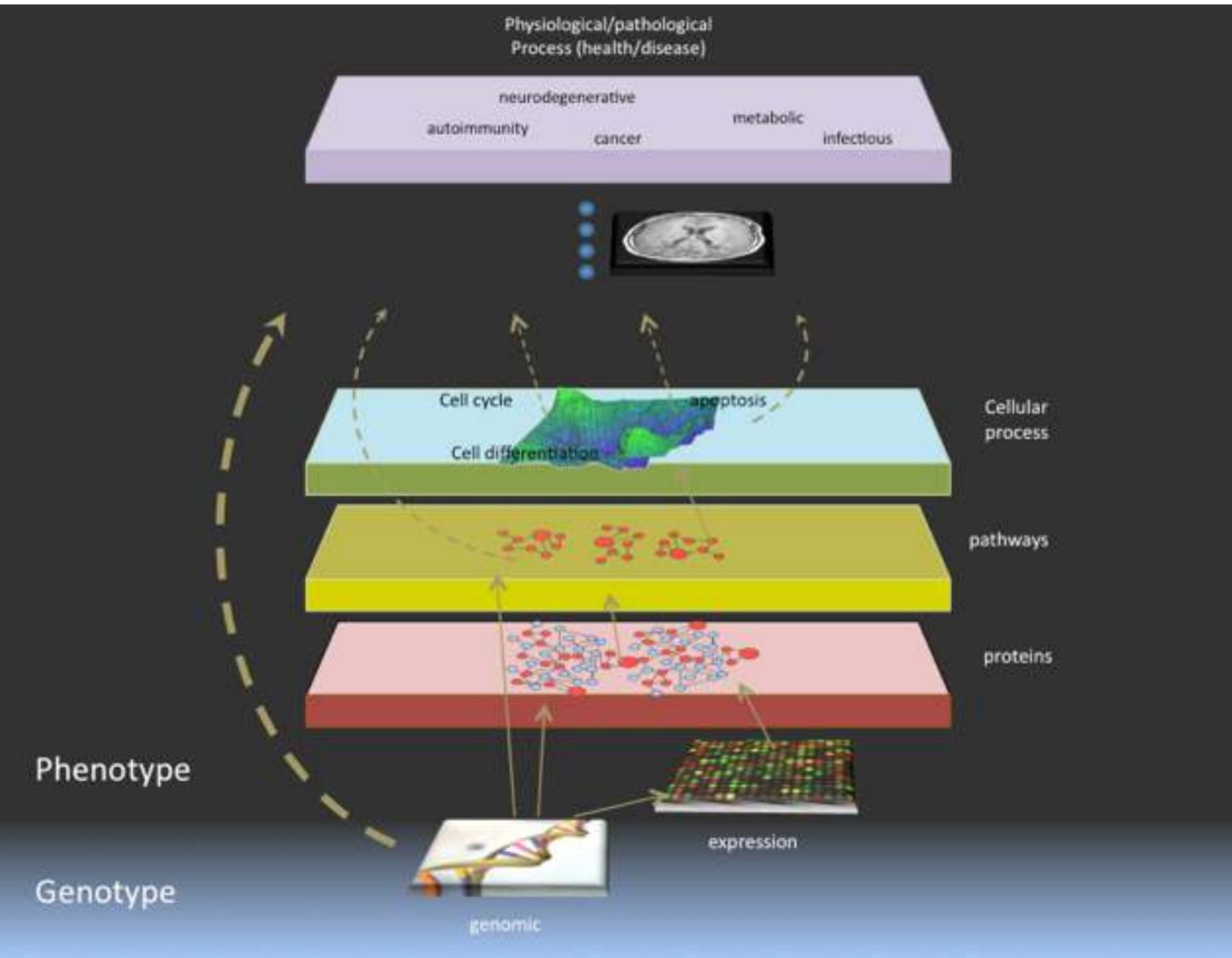


How does it work?





- Each **dataset** can be considered as a **layer**
- **Combinations** of different layers brings us to **new knowledge**



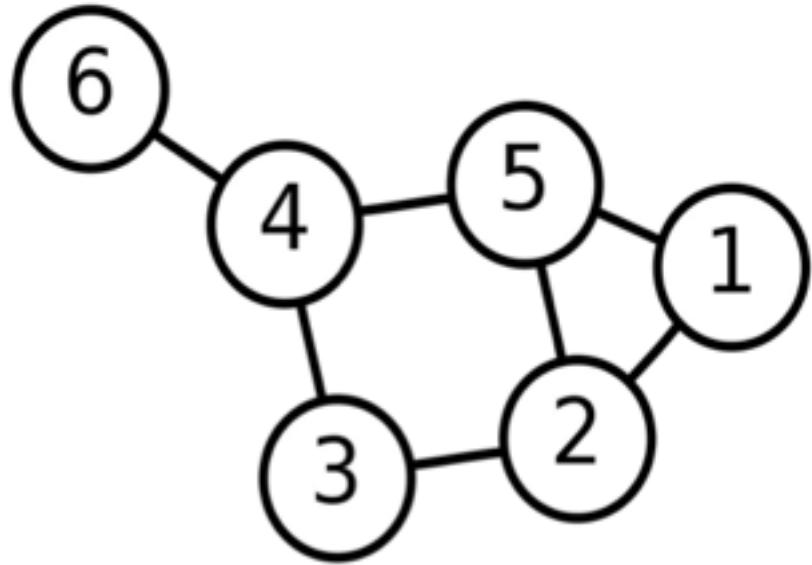
Q:

How to combine such
heterogeneous information and
to make sense of it?

A:

Networks!

Nodes



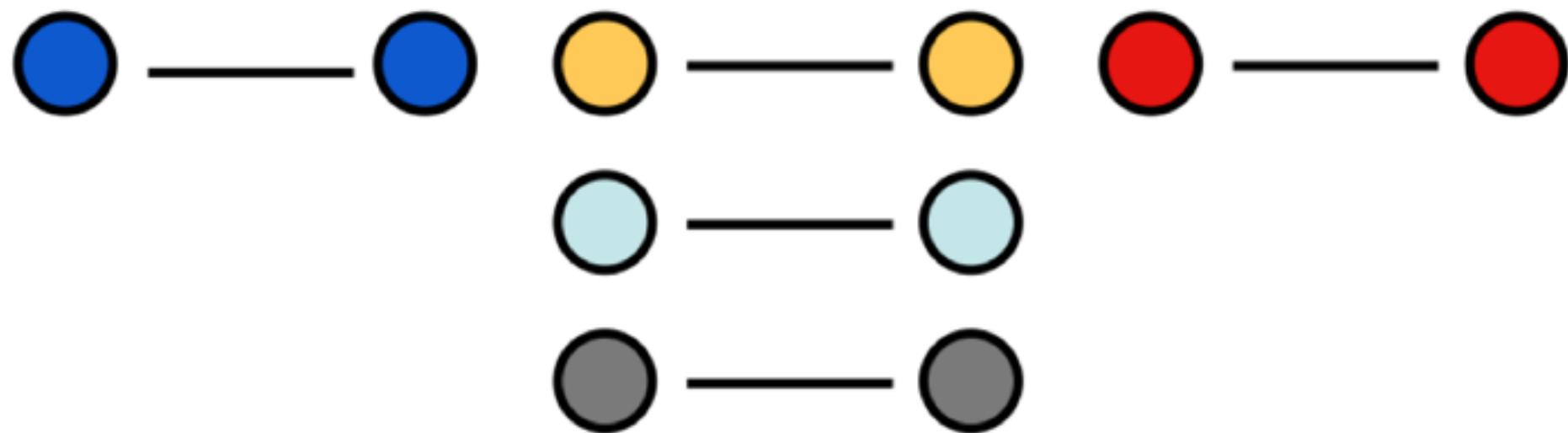
- Genes
- Proteins
- Metabolites
- Enzymes
- etc.

Edges

Co-expression

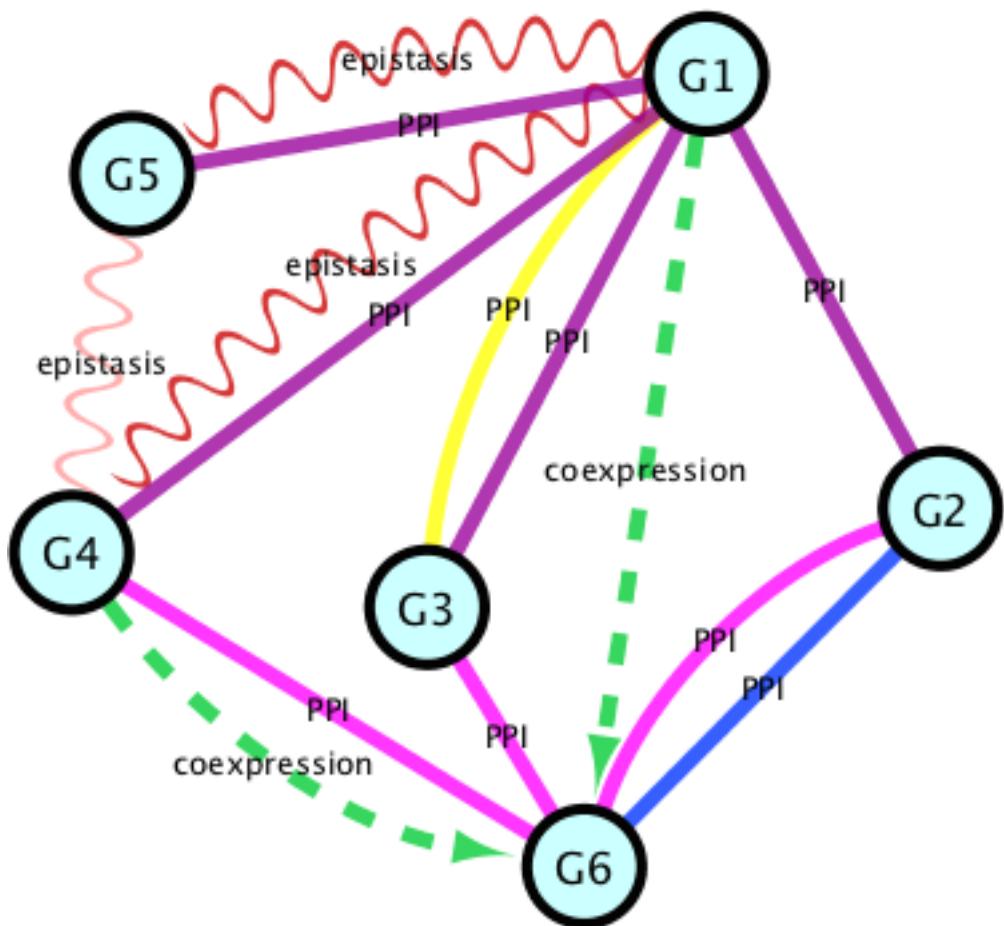
PPI

Epistasis



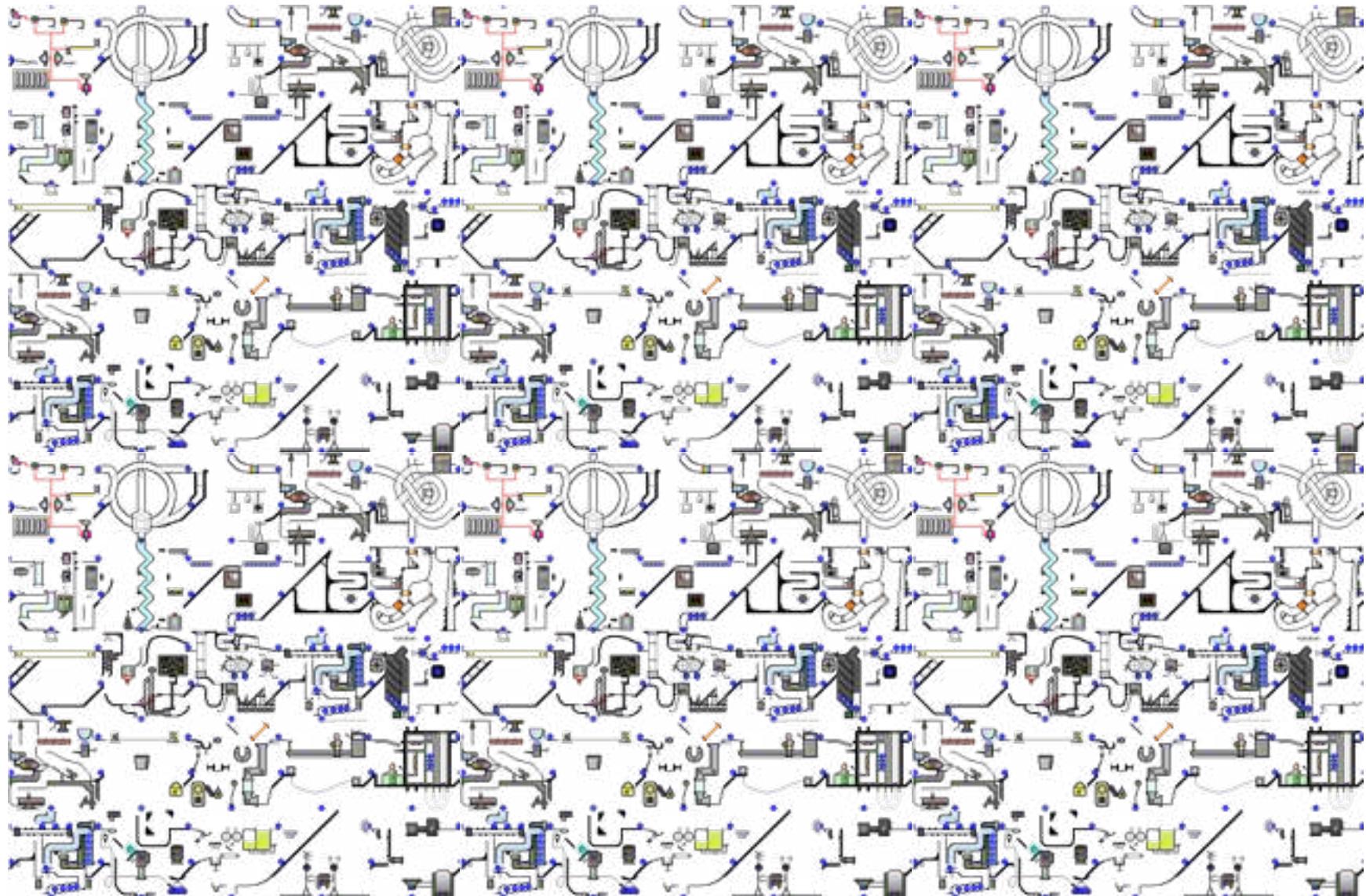
Biological relationships, interactions, regulations, reactions, transformations, activations, inhibitions, etc.

Edges in the network

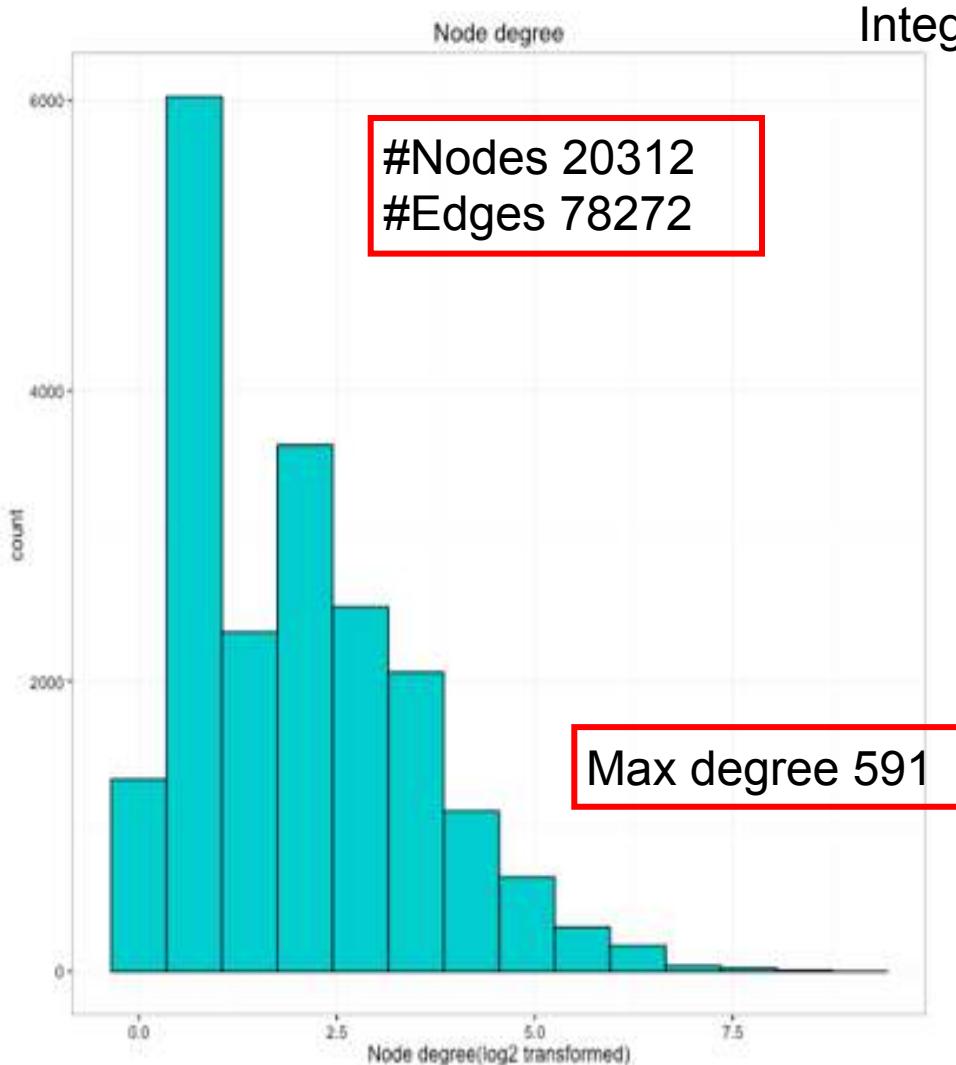


Edge Line Type	interaction
—	PPI
----	coexpression
wavy	epistasis
Edge Color (Unselected)	evidence_code
Blue	ADIA
Green	ADN
Red	ADNI_VER
Magenta	HEI
Cyan	HLC
Yellow	IAHS
Orange	TGEN
Edge Target Arrow Shape	direction
→	directed

Simple representation of the integrated data



Network analysis. Node degree.



Integrated dataset is represented as graph

Gene name	Degree
LRRK2	278
APP	192
MAPT	76
KHDRBS1	74
CLU	63
FYN	52
BIN1	41
RIN3	38
EPHA1	38
APOE	32
CD2AP	31
CASS4	30
INPP5D	22
MACF1	22

Topological network analysis



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public web server for graph-based analysis of biological networks

<http://biit.cs.ut.ee/graphweb/index.cgi>

1. Extract the largest connected component.
2. Markov Cluster Algorithm
3. Hub-based Modules

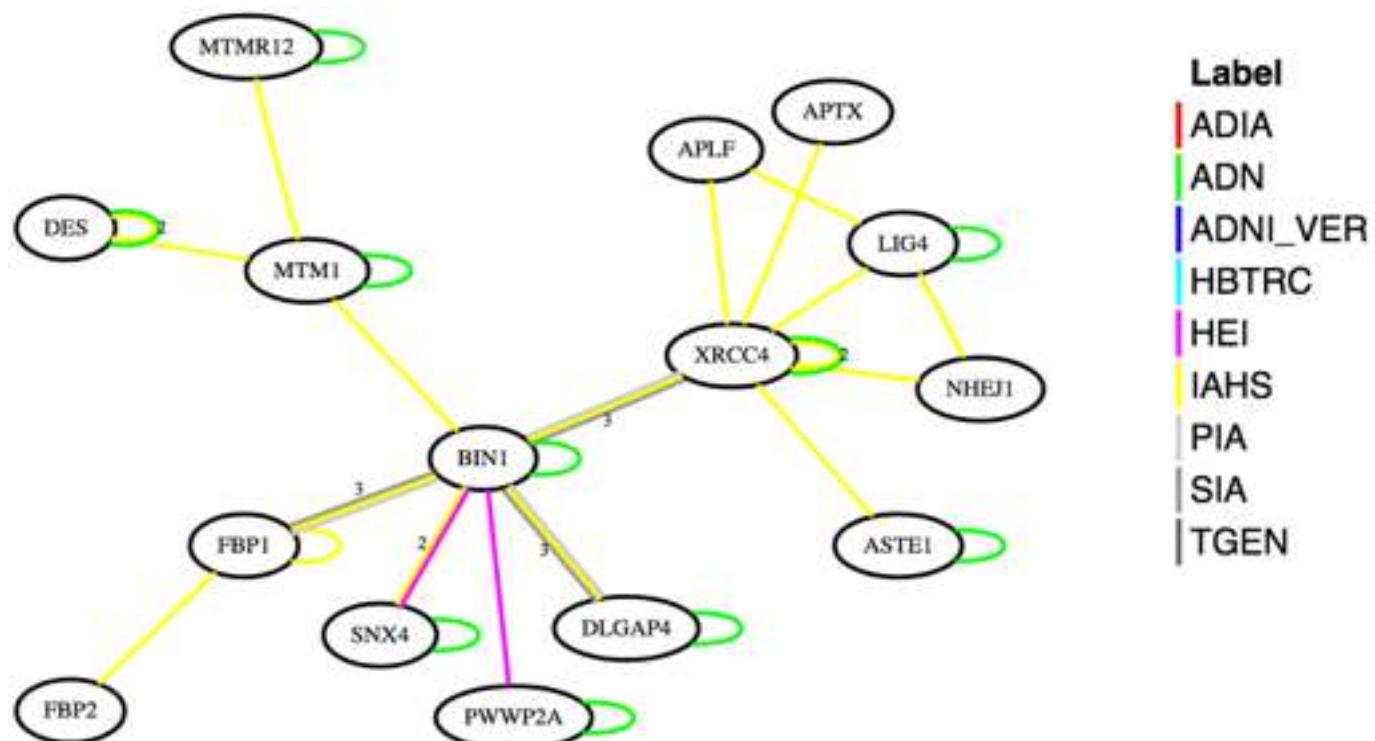
#Nodes 11322
#Edges 34244

Markov Cluster Algorithm



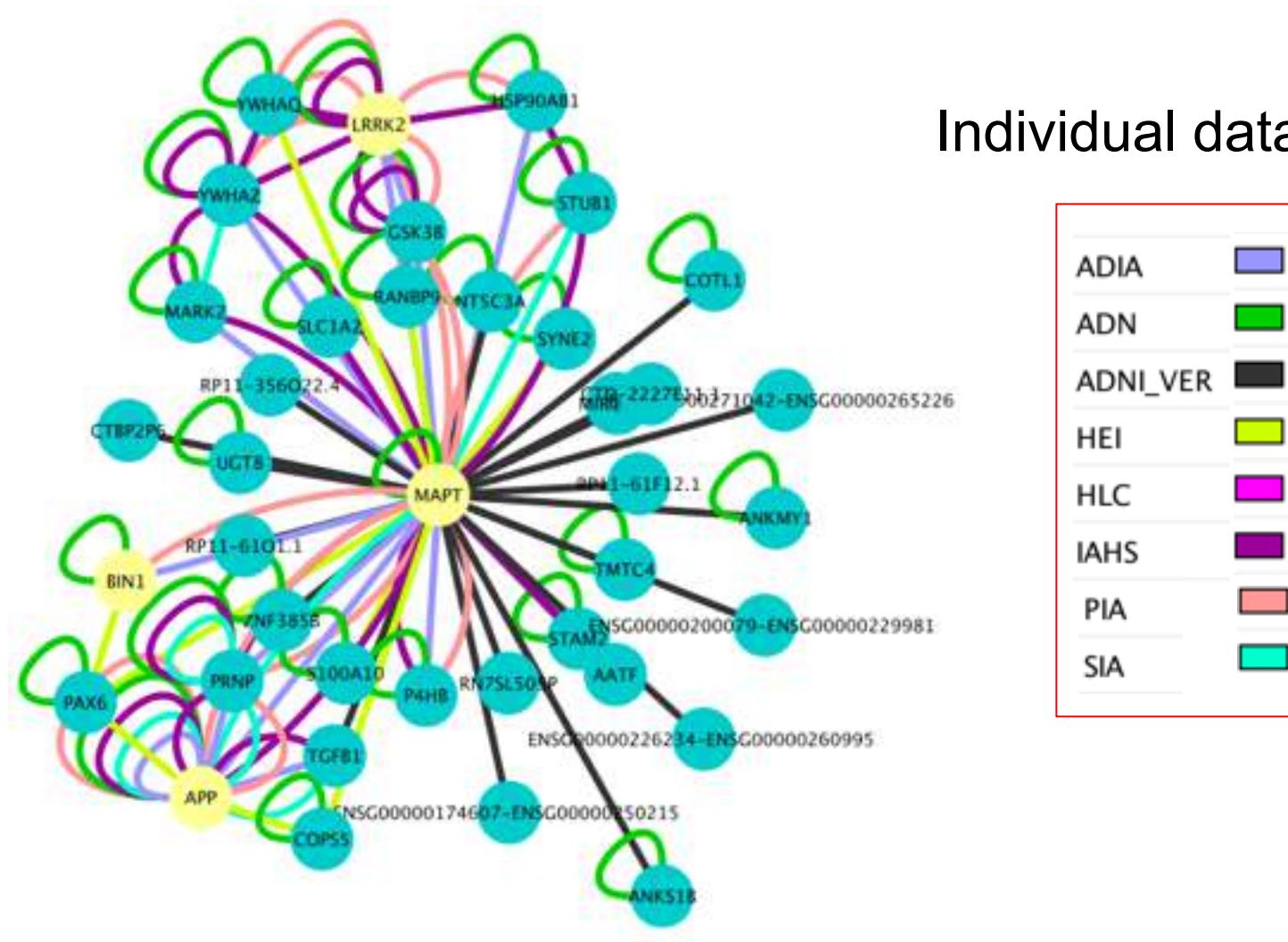
2175 modules found

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Hub-based modules

Individual datasets



g:Profiler toolset

<http://biit.cs.ut.ee/gprofiler>

The screenshot shows the g:Profiler website. On the left, there's a large "g:Profiler" logo with a blue and orange design. Below it is a navigation bar with links: Welcome!, Contact, FAQ, R / APIs, Beta, and Archive. On the right, there's a vertical list of tools with blue headers:

- g:GOST Gene Group Functional Profiling
- g:Cocoa Compact Compare of Annotations
- g:Convert Gene ID Converter
- g:Sorter Expression Similarity Search
- g:Orth Orthology search
- g:SNPense Convert rsID

J. Reimand, M. Kull, H. Peterson, J. Hansen, J. Vilo: *g:Profiler - a web-based toolset for functional profiling of gene lists from large-scale experiments* (2007) NAR 35 W193-W200
Jüri Reimand, Tambet Arak, Priit Adler, Liis Kolberg, Sulev Reisberg, Hedi Peterson, Jaak Vilo:
g:Profiler -- a web server for functional interpretation of gene lists (2016 update) Nucleic Acids Research 2016; doi: 10.1093/nar/gkw199

g:GOSt

Tool for functional profiling of gene lists from **large-scale** experiments

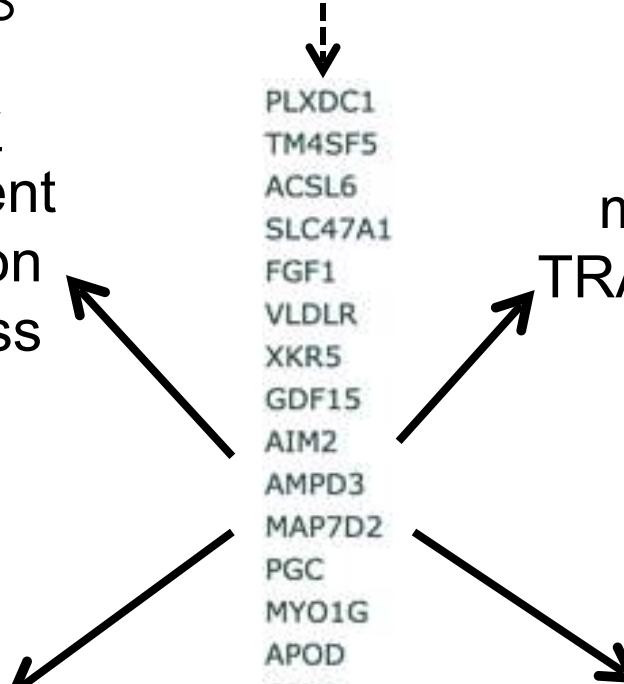
Gene Ontology
Cellular component
Molecular function
Biological process

Pathways
KEGG
Reactome

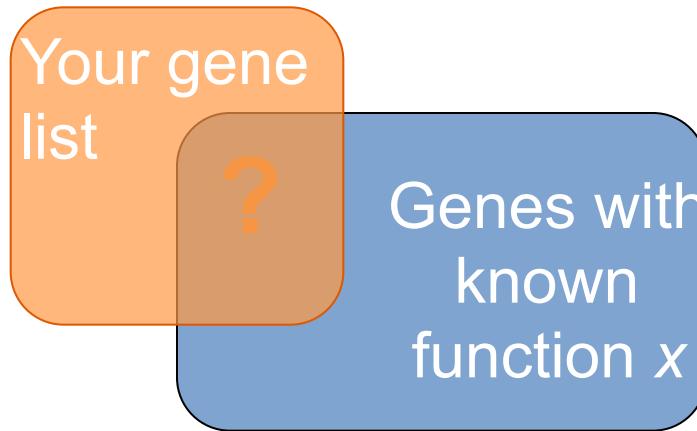
PLXDC1
TM4SF5
ACSL6
SLC47A1
FGF1
VLDLR
XKR5
GDF15
AIM2
AMPD3
MAP7D2
PGC
MYO1G
APOD
CPN1
CLGN
PLCD3
D11WSU47E
FBXO27
ESYT3

Regulation
miRBase miRNAs
TRANSFAC TF targets

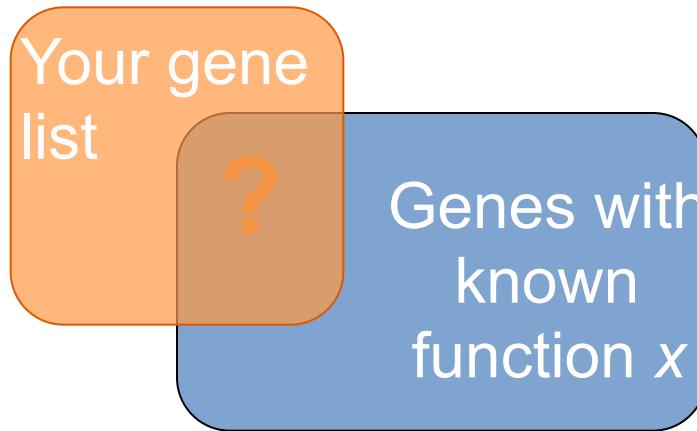
Varia
Biogrid PPIs
CORUM protein complexes
Human Phenotype Ontology



Functional enrichment statistics

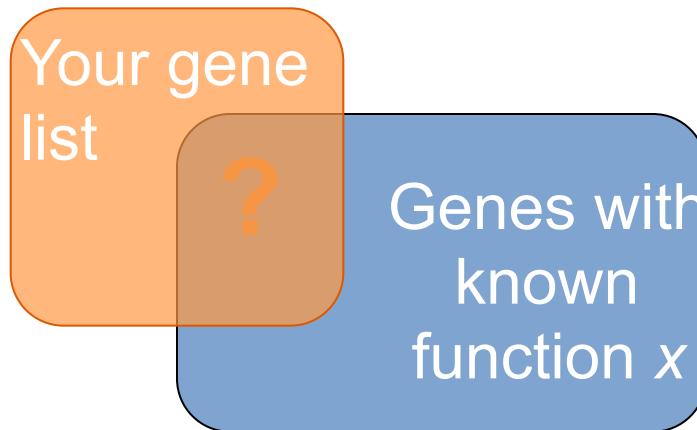


Functional enrichment statistics



Does your gene list includes **more** genes with function x than expected by random chance?

Functional enrichment statistics



Does your gene list includes **more** genes with function x than expected by random chance?

$$p = \sum_{i=k_\pi}^{\min(n, K_\pi)} \frac{\binom{K_\pi}{i} \binom{N-K_\pi}{n-i}}{\binom{N}{n}}$$

Reading the output

>> g:Convert
Gene ID Converter

>> g:Orth
Orthology Search

>> g:Sorter
Expression Similarity Search

>> g:Cocoa
Compact Compare of Annotations

>> Static URL
Come back later

[?] Significantly enriched network of BioGRID interactions found. >> Show network

source	term name	term ID	n. of term genes	n. of query genes	n. of common genes	corrected p-value
	Gene Ontology (Biological process)					
BP	-72 negative regulation of biological process	GO:0048519	4572	25	13	4.72e-02
BP	-negative regulation of cellular process	GO:0048523	4299	25	13	1.68e-02
BP	-70 embryo development	GO:0009930	1914	22	9	4.20e-02
BP	-embryonic morphogenesis	GO:00048598	593	8	4	2.95e-02

source term name

Gene Ontology (Biological process)

BP ncRNA metabolic process

BP tRNA metabolic process

source	term name	term ID	n. of term genes	n. of query genes	n. of common genes	corrected p-value
	Gene Ontology (Biological process)					
BP	ncRNA metabolic process	GO:0034660	475	50	10	8.05e-04
BP	tRNA metabolic process	GO:0006399	199	50	6	3.02e-02

Your genes
50
10

GO:0034660
ncRNA metabolic process
475 genes

Statistics

Markov Cluster Algorithm

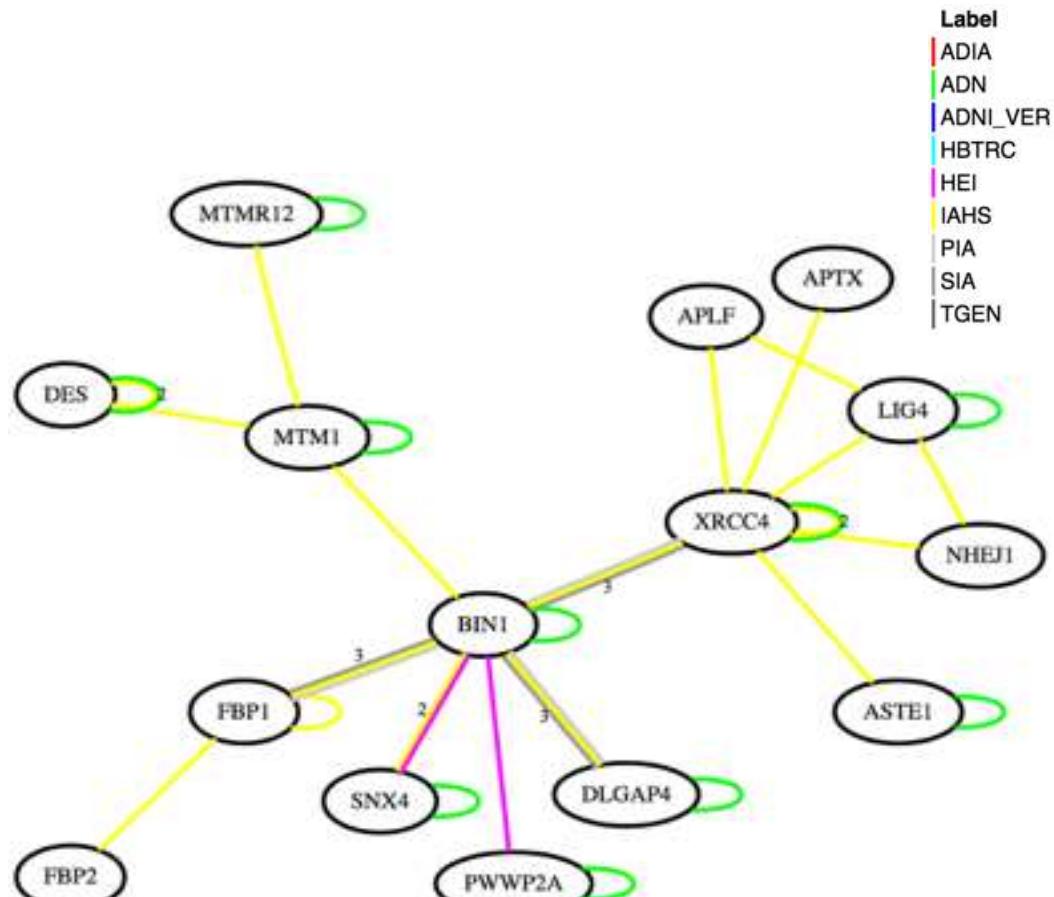


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2175 modules found

Enrichment results for example module

4.33e-07	GO:BP	DNA ligation
2.40e-05	GO:CC	nonhomologous end joining complex ...
1.00e-03	GO:MF	fructose 1,6-bisphosphate ...
4.68e-08		DNA ligase IV-XRCC4-XLF complex ...
8.75e-06	KEGG	Non-homologous end-joining
1.86e-03	REACTOME	Nonhomologous End-joining ...



Sharing with the community



AgedBrainSYSBIO

DATASETS

Project browser

PPI DATA OVERVIEW

Data Tables

Explore protein-protein interactions.

TOOLS

ExpressView

DiffExp

MEM

g:Profiler

Explore collection of datasets. Currently contains gene expression datasets from ArrayExpress related to Alzheimer's disease in human (15 datasets) and mouse (8 datasets). Datasets originate from different microarray platforms.

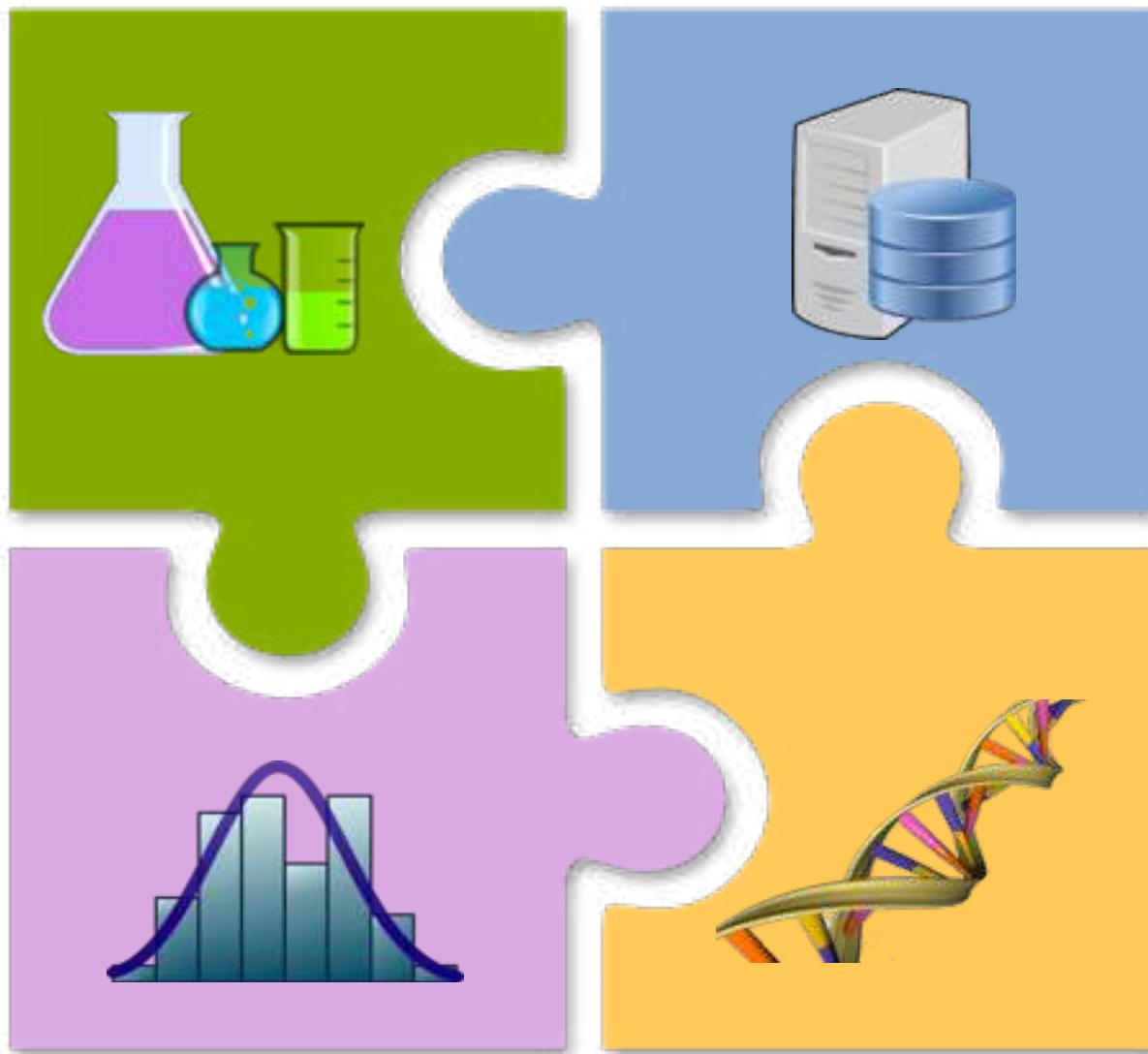
AgedBrainSYSBIO data collection

ID	Initial self interacting PPI	Genomic interactions	Isolated	Protein summarized pathways	Date	Version
SHC1	SHC1	2.05e-03	0.22	219,0	2019-06-05	0
POVA	POVA	2.05e-03	0.00	219,0	2019-06-05	0
WIF1	WIF1	1.45e-11	12.20	219,0	2019-06-05	0
WIF1	WIF1	1.45e-11	12.20	219,0	2019-06-05	0
WIF1	WIF1	1.45e-11	12.20	219,0	2019-06-05	0
PRKARIA	PRKARIA	1.27e-03	19.69	219,0	2019-06-05	0
PRKARIA	PRKARIA	1.27e-03	19.69	219,0	2019-06-05	0
PRKARIA	PRKARIA	1.27e-03	19.69	219,0	2019-06-05	0
PRKARIA	PRKARIA	1.27e-03	19.69	219,0	2019-06-05	0
PRKARIA	PRKARIA	1.27e-03	19.69	219,0	2019-06-05	0
PRKARIA	PRKARIA	1.27e-03	19.69	219,0	2019-06-05	0
MDRD	MDRD	2.41e-05	0.00	219,0	2019-06-05	0

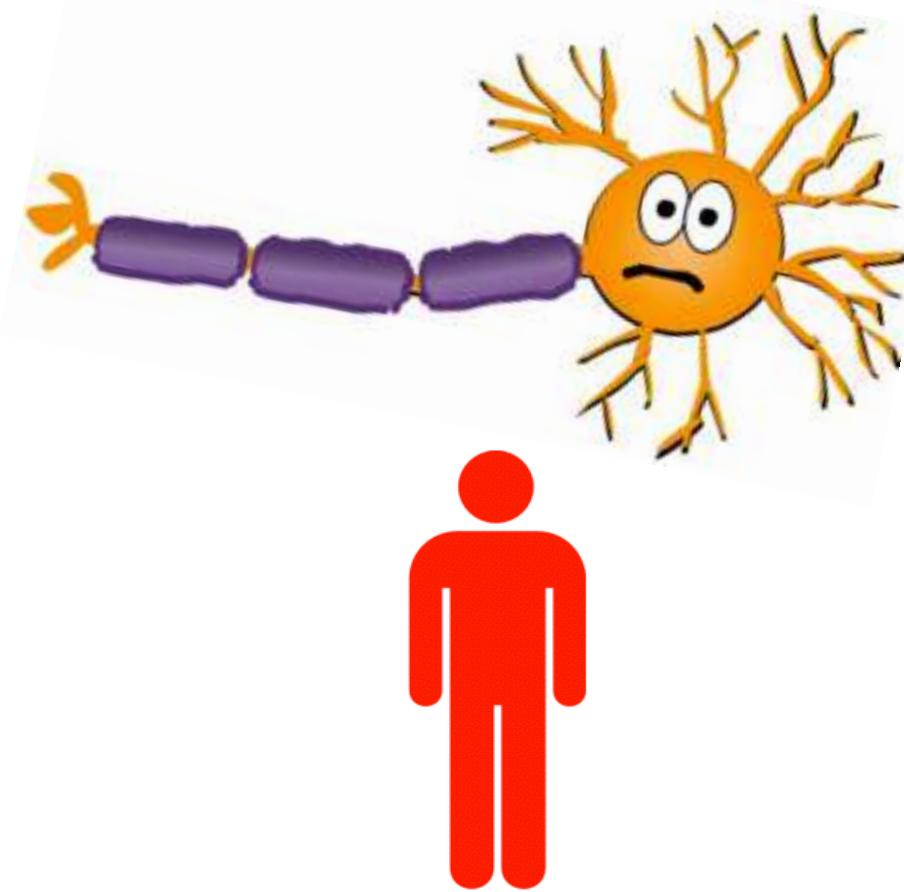
Showing 1 to 10 of 96,000 entries

Previous 1 2 3 4 5 ... Next Last

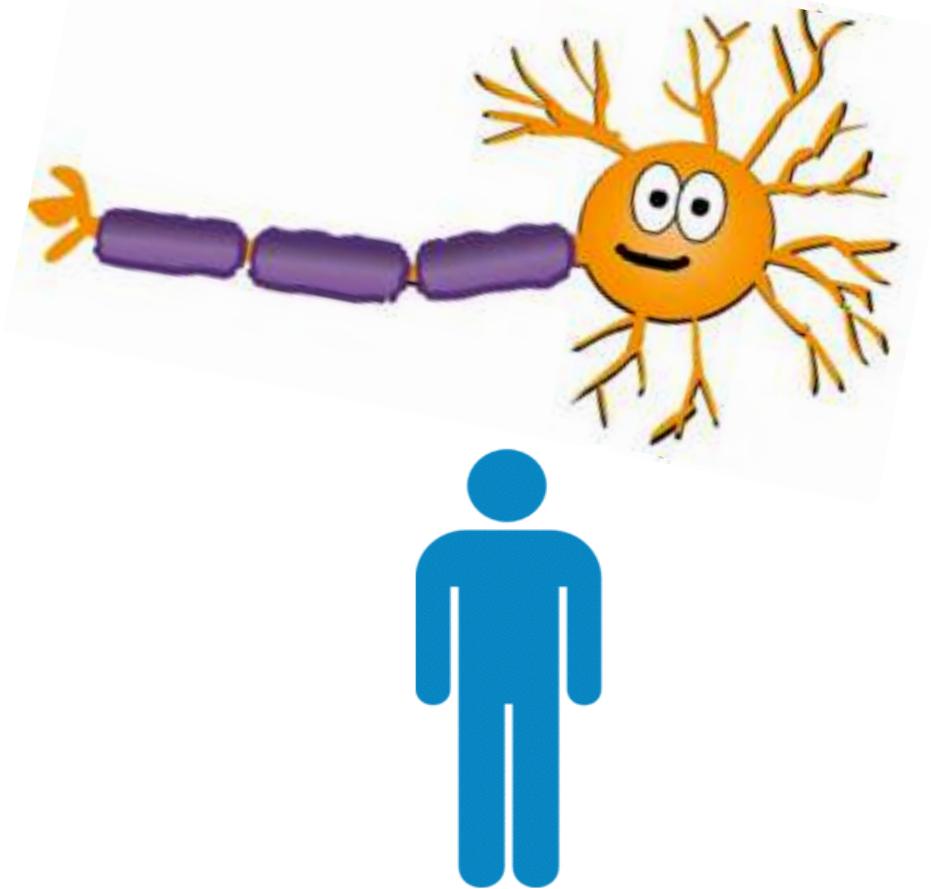
Data integration = understanding the disease better



Make unhappy neurons



Make neurons happy



I will study human brain
I will study human brain



**“The brain is a commodity used
to fertilize ideas.”**



Elbert Hubbard



THANK YOU!

Questions

