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Shahid Beheshti University

Higher-order genetic interactions and their contribution to cancerous cells

Reza Jafari

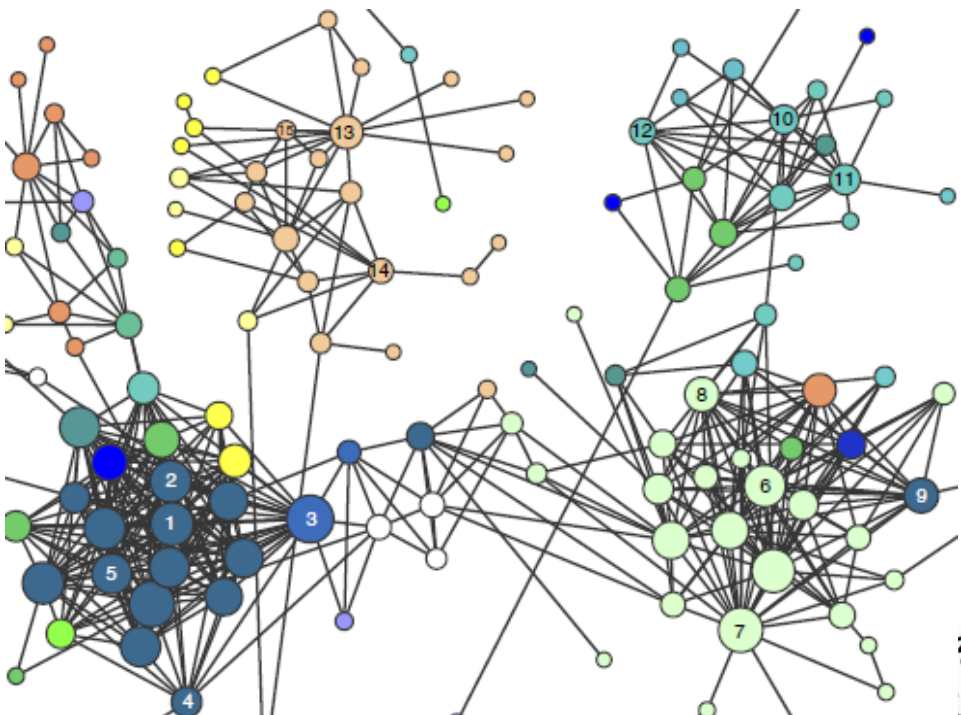
Evolution of Complexity from the Statistical Physics Perspective

June 29 - July 2, 2022

Yerevan Physics Institute, Yerevan - Armenia



Cancer & Social science



ARTICLE

Received 7 Nov 2013 | Accepted 27 May 2014 | Published 26 Jun 2014

DOI: 10.1038/ncomms5212

Human symptoms-disease network

XueZhong Zhou^{1,2,3,*}, Jörg Menche^{2,3,4,*}, Albert-László Barabási^{2,3,4,5,6} & Amitabh Sharma

nanor & Mid Arroz Alum
Network biology approach for identifying key
cancer
open access
Discovery of the interface of physical and biological sciences
www.bioinformatics.net
Volume 8(23)
Hypothesis



Nature is more complex than we previously thought



NAUTILUS



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BIOLOGY | GENETICS

Aging Is a Communication Breakdown

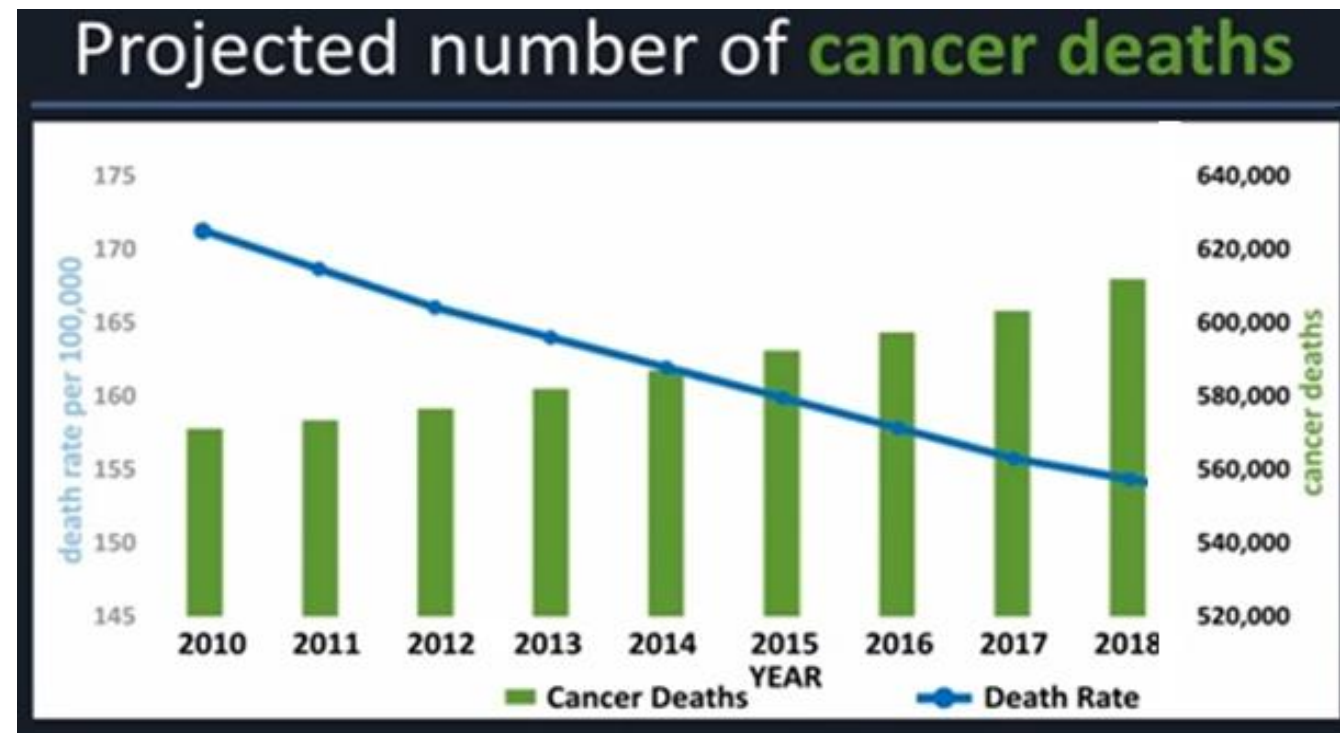
Genes that can't express themselves may be hallmarks of cancer.

BY JIM KOZUBEK

MARCH 21, 2019

Cancer

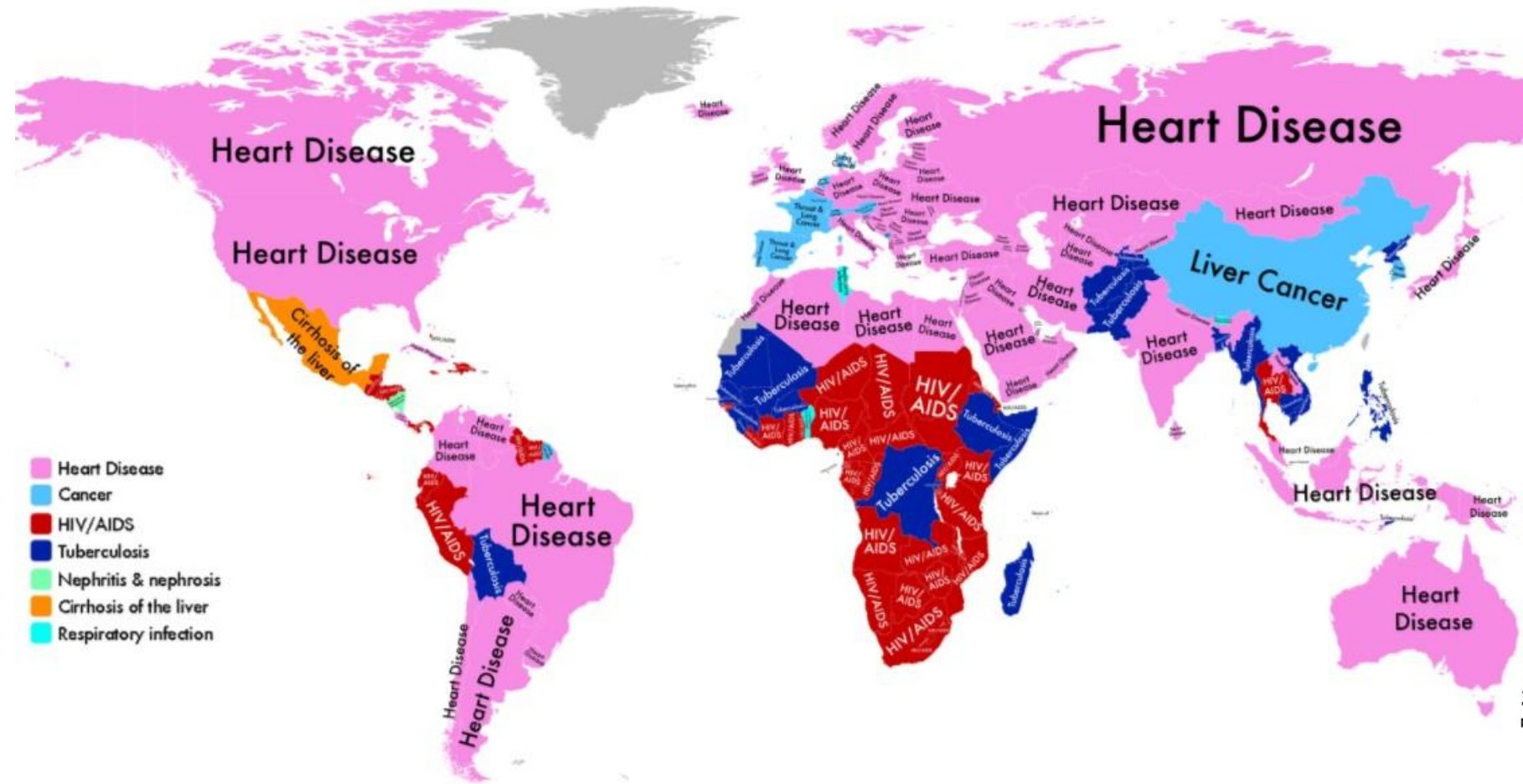
- Cancer is a name that refers to a set of diseases that occurs as a result of uncontrolled cell proliferation.
- Contrary to the many researches in medicine over the past years, we are still faced with many challenges in controlling and treating cancer.



Weir HK, Thompson TD, Soman A, Møller B, Leadbetter S. [The past, present, and future of cancer incidence in the United States: 1975 through 2020.External](#), *Cancer* 2015;121(11):1827–1837.

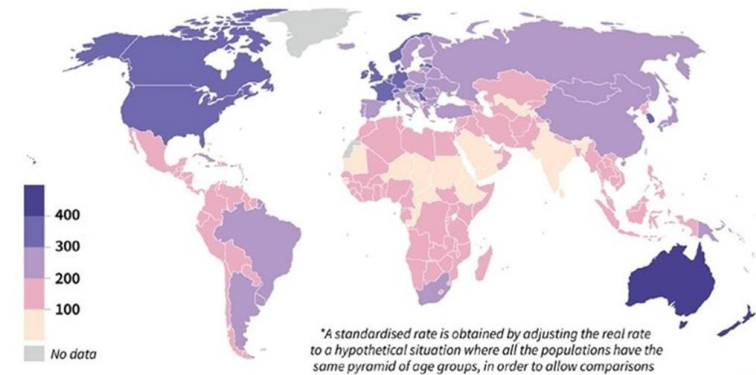
https://www.cdc.gov/cancer/dcpc/research/articles/cancer_2020.htm

Cancer is the second leading cause of death among diseases

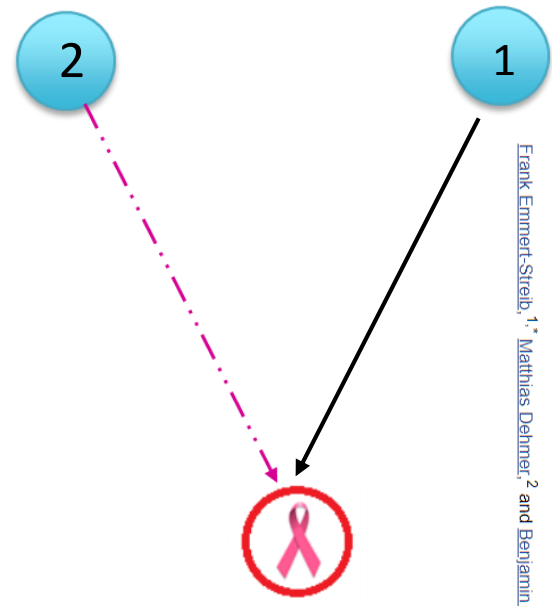
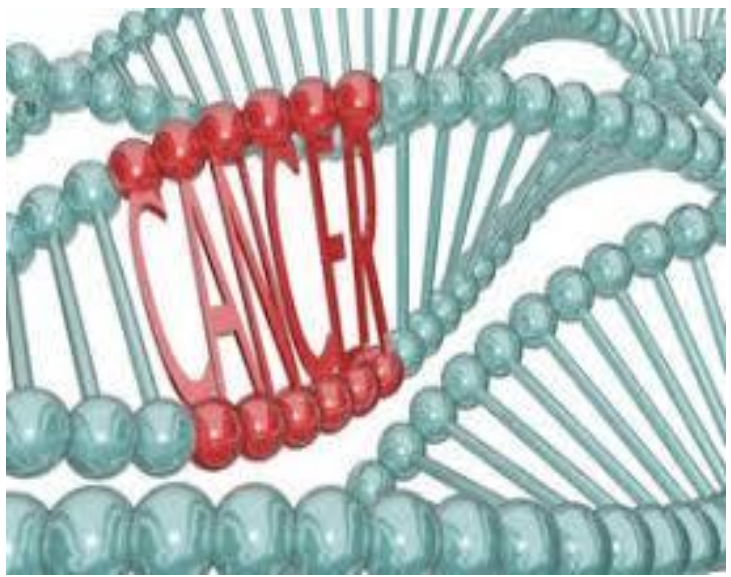
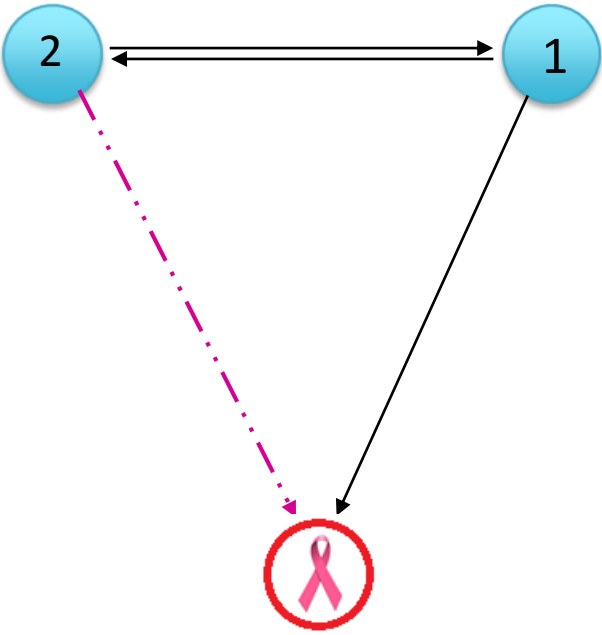


man Khosla / GlobalPost

18.1 million new cancer cases predicted for 2018
Number of estimated cases per 100,000 people, rate standardised* for age



Cancer is known as a Gene disease



Gene regulatory networks and their applications: understanding biological and medical problems in terms of networks
 Frank Emmert-Streib,^{1*} Matthias Dehmer,² and Benjamin Helber-Kamins^{3,*}

Front Cell Dev Biol. 2014; 2: 38.
 Published online 2014 Aug 19; doi: 10.3389/fcell.2014.00038

PMCID: PMC4207011
 PMID: 25364745

Science Home News Journals
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SCIENTIFIC REPORTS
 Article | OPEN | Published: 21 March 2019

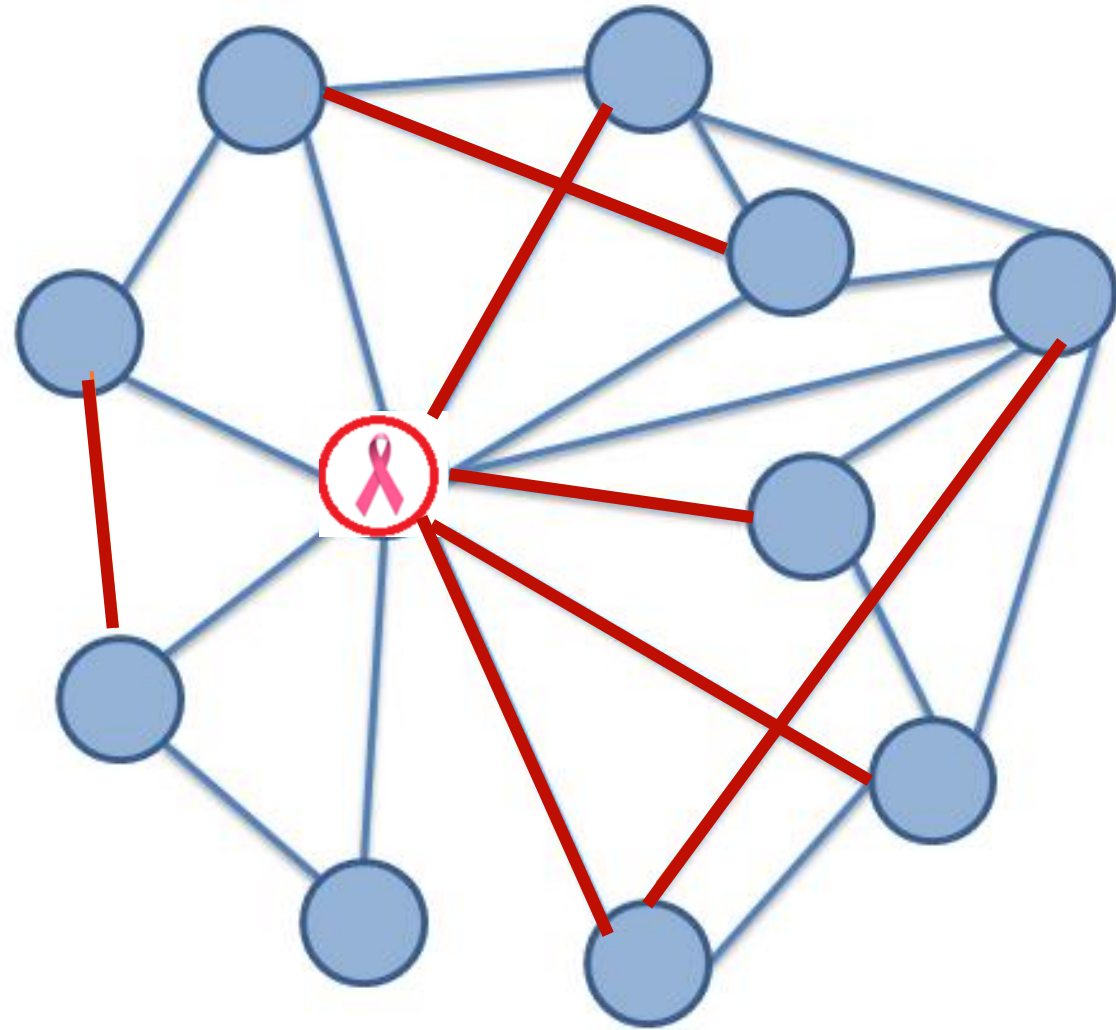
SHARE YEAST GENETICS
 A global genetic interaction network
 Laura M. Zahn
 See all authors and affiliations

SHARE RESEARCH ARTICLE
Global Mapping of the Yeast Genetic Interaction Network
 Amy Hin Yan Tong^{1,2,*}, Guillaume Lesage^{3,*}, Gary D. Bader⁴, Huiming Ding¹, Hong Xu^{1,2}, Xiaofeng Xin^{1,2}, James Young⁶, Gabriel
 Science 06 Feb 2004.
 Vol. 303, Issue 5659, pp. 808-813
 DOI: 10.1126/science.1091317

Network biology approach for regulatory genes by expression breast cancer
 amini Chand* & Md Afroz Alam

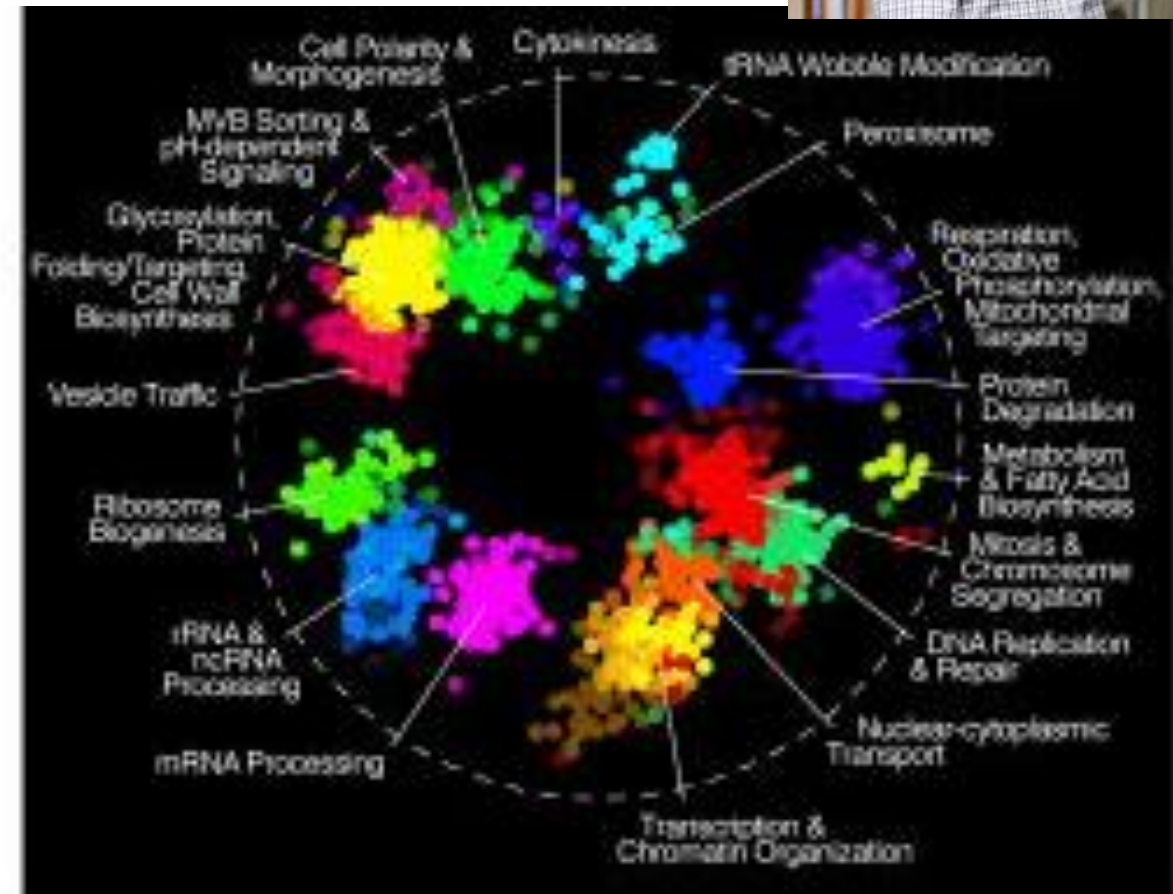
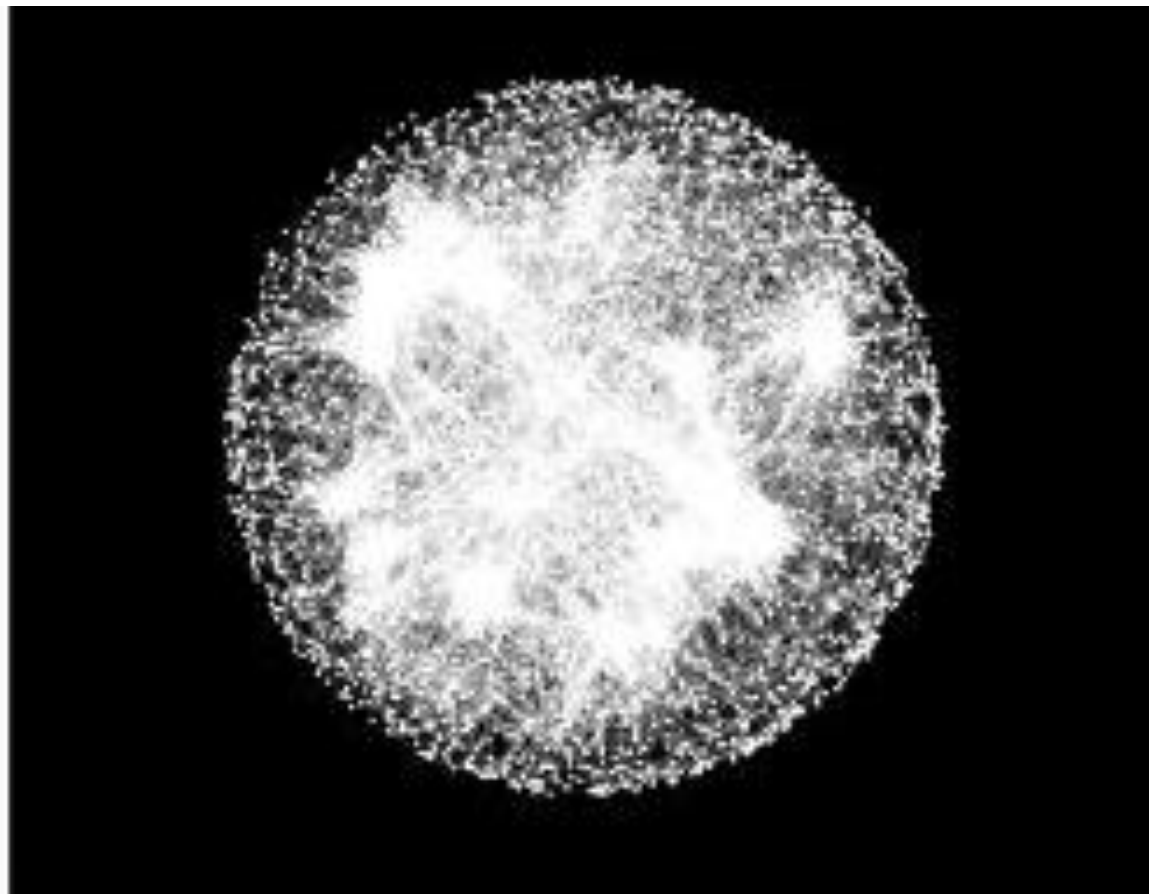
Gene Saturation: An Approach to As Exploration Stage of Gene Interaction Networks
 Ziqiao Yin, Binghui Guo, Zhilong Mi, Jiahui Li & Zhiming Zheng

Gene – Gene Interactions



Gene – Gene interaction networks (*Saccharomyces* is a genus of fungi)

Saccharomyces 6000 genes, identifying about one million interactions, which 550,000 are negative, and 350,000 positives.



An idea from Social and Psychology sciences

The relationships between two persons is not depends on them.



frontiers
in Physiology

ORIGINAL RESEARCH
published: 20 January 2021
doi: 10.3389/fphys.2020.573732



Stability of Imbalanced Triangles in
Gene Regulatory Networks of
Cancerous and Normal Cells

PLOS ONE



The structure balance of gene-gene networks beyond pairwise interactions

Frustration Cognitive Balance Theories



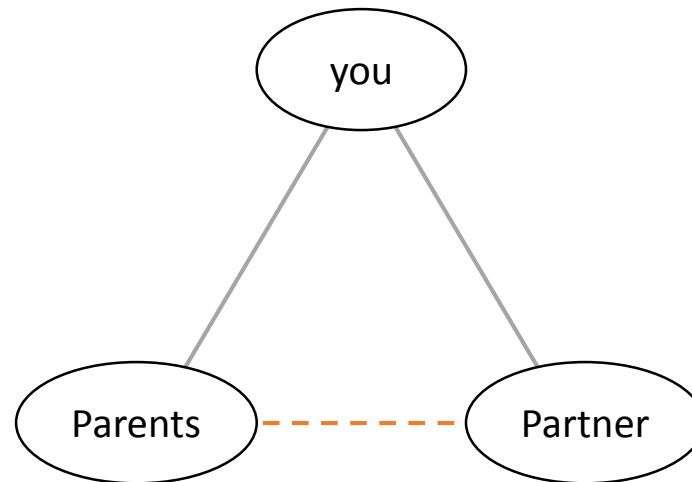
(Fritz Heider 1946)

**We adjust our relationship based on reducing
the psychological stresses**

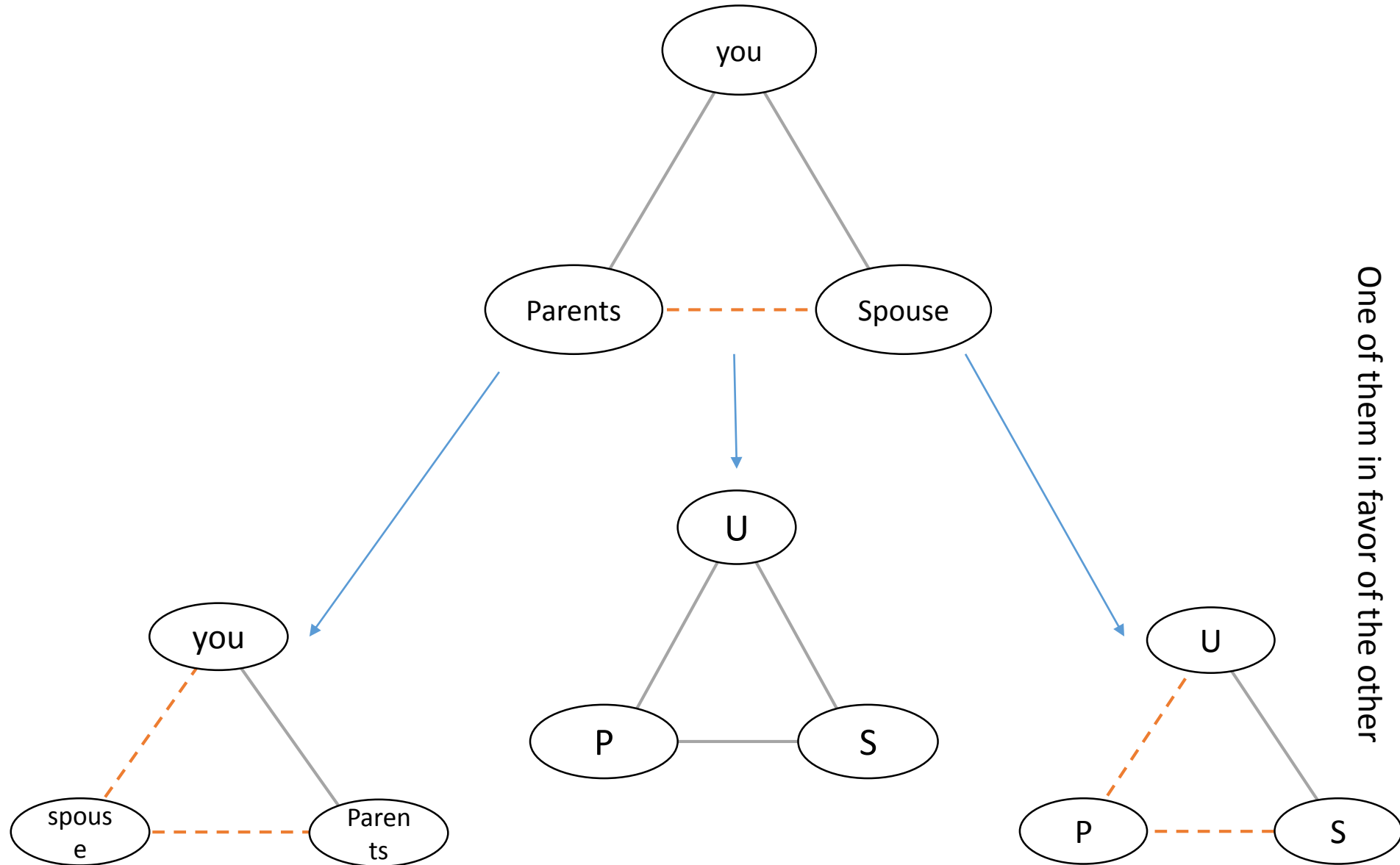
If a person's beliefs are unbalanced, psychological stresses will generate internal pressures to change either some of the sentiments (liking, disliking) or some relationships (proximity, membership) into a more congruent pattern.

The hard situation

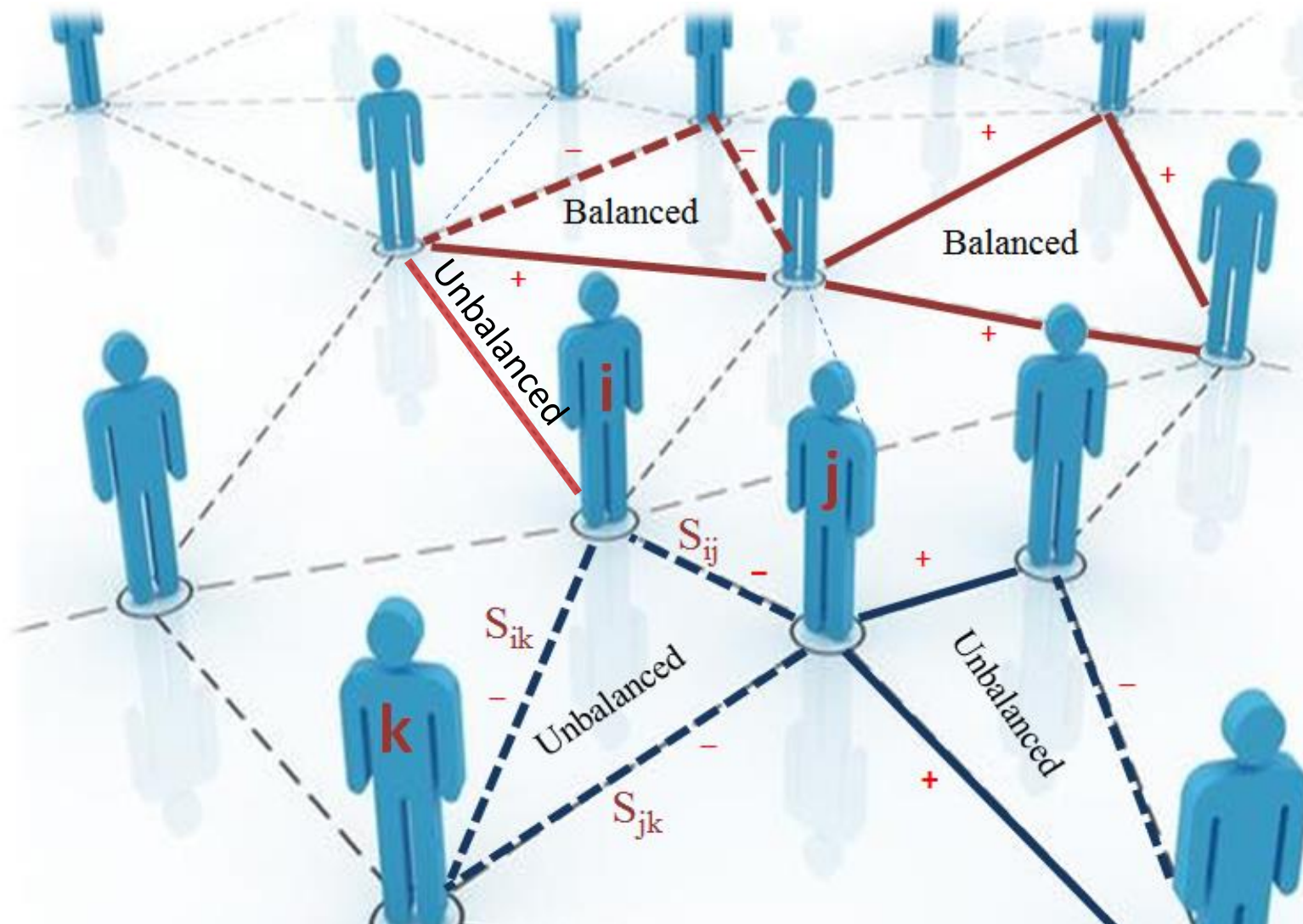
Cartwright and Harary (1950)



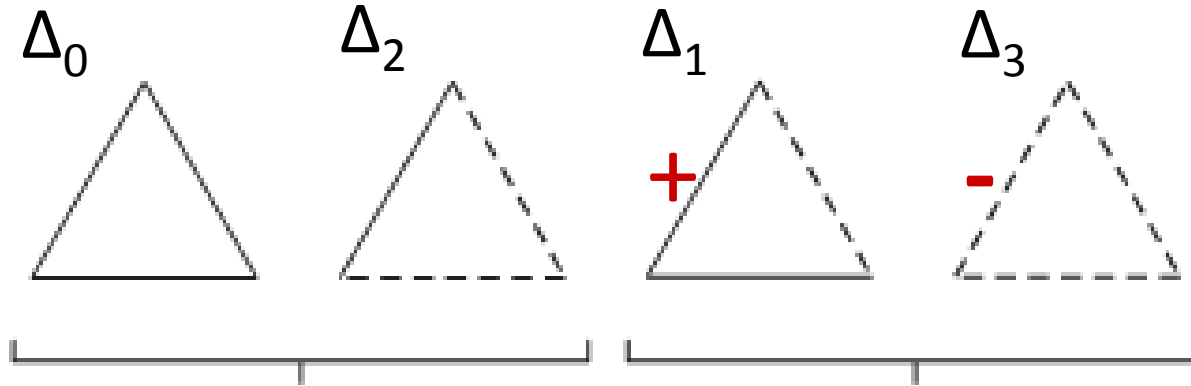
Reducing the stress



The influence of relationship



Mathematical model



balanced triangles

$$S_{ij} S_{jk} S_{ki} = 1$$

Even number of dashed edges

unbalanced triangles

$$S_{ij} S_{jk} S_{ki} = -1$$

Odd number of dashed edges

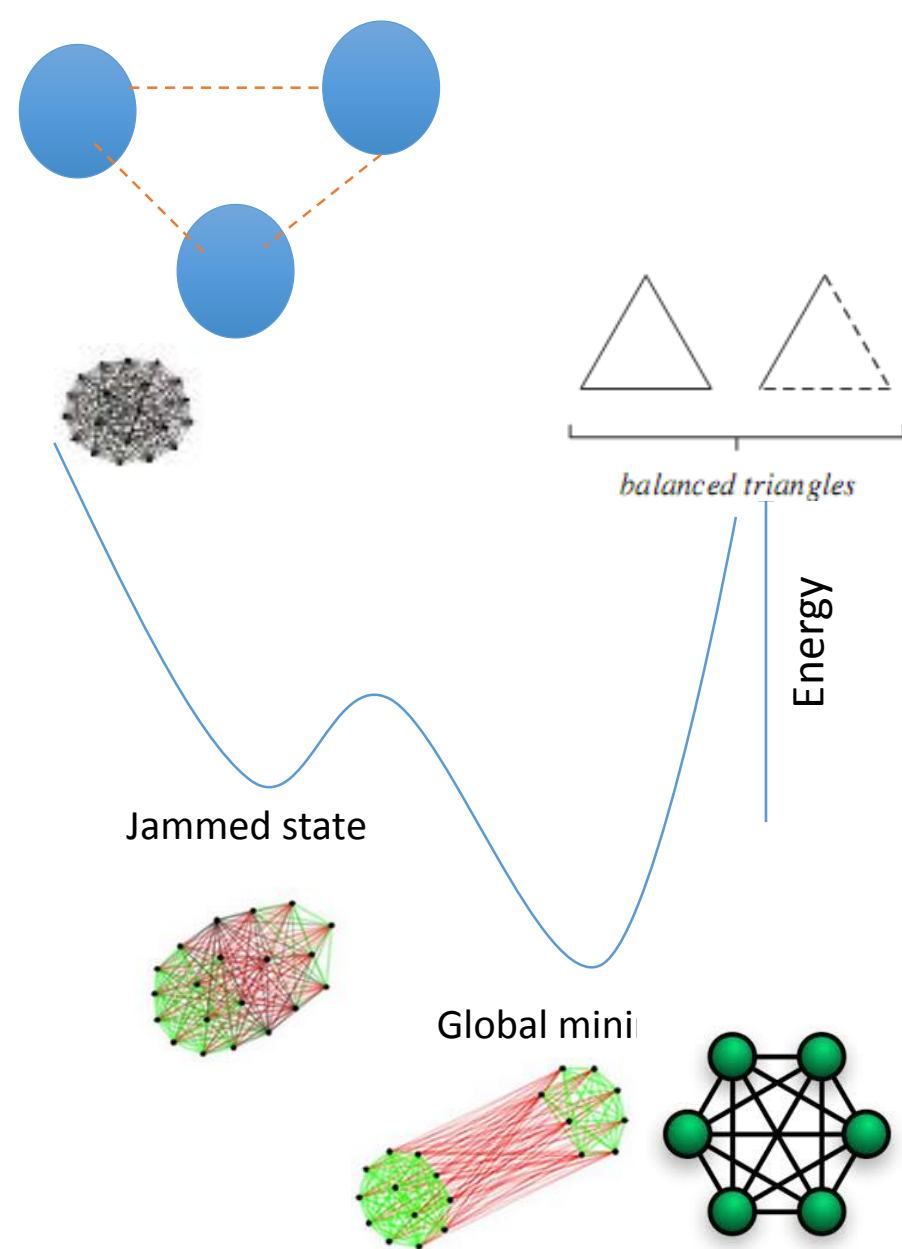
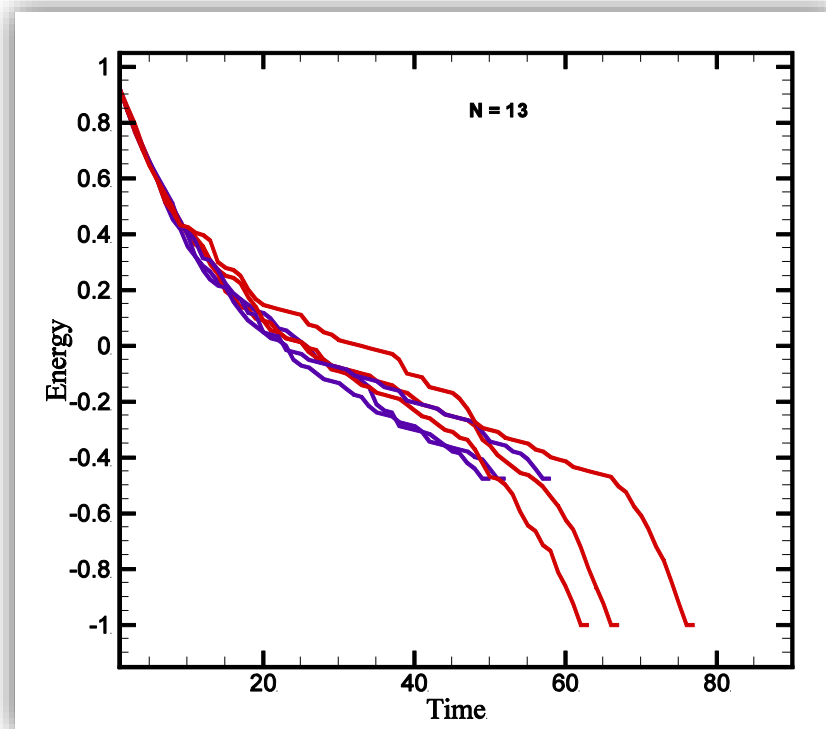
$$H = -\frac{1}{\binom{n}{3}} \sum S_{ij} S_{jk} S_{ik}$$

$$\frac{dx_{ki}}{dt} = \sum_{l=1} X_{kl}(t') X_{li}(t')$$



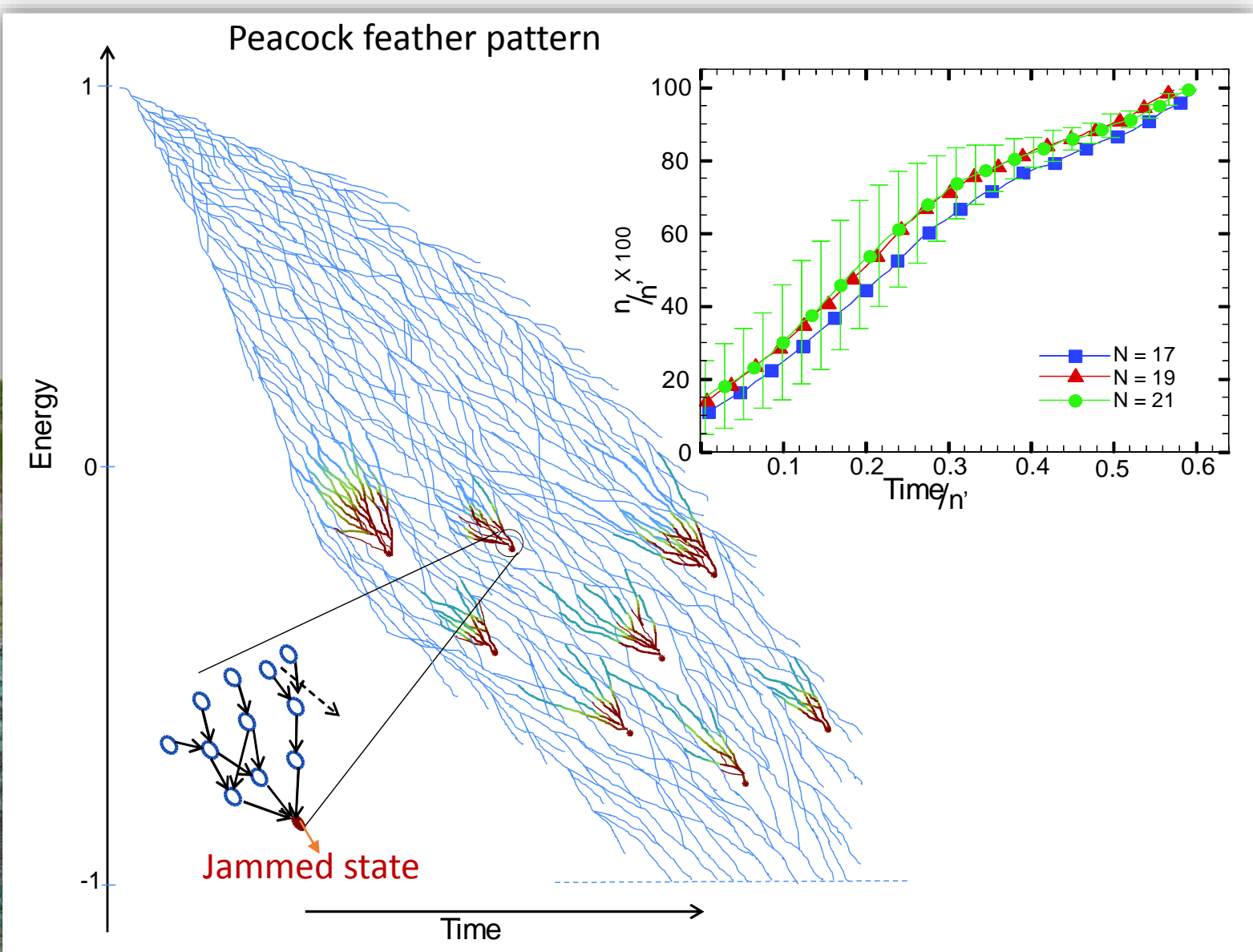
Jammed states

$$H = -\frac{1}{\binom{n}{3}} \sum S_{ij} S_{jk} S_{ik}$$



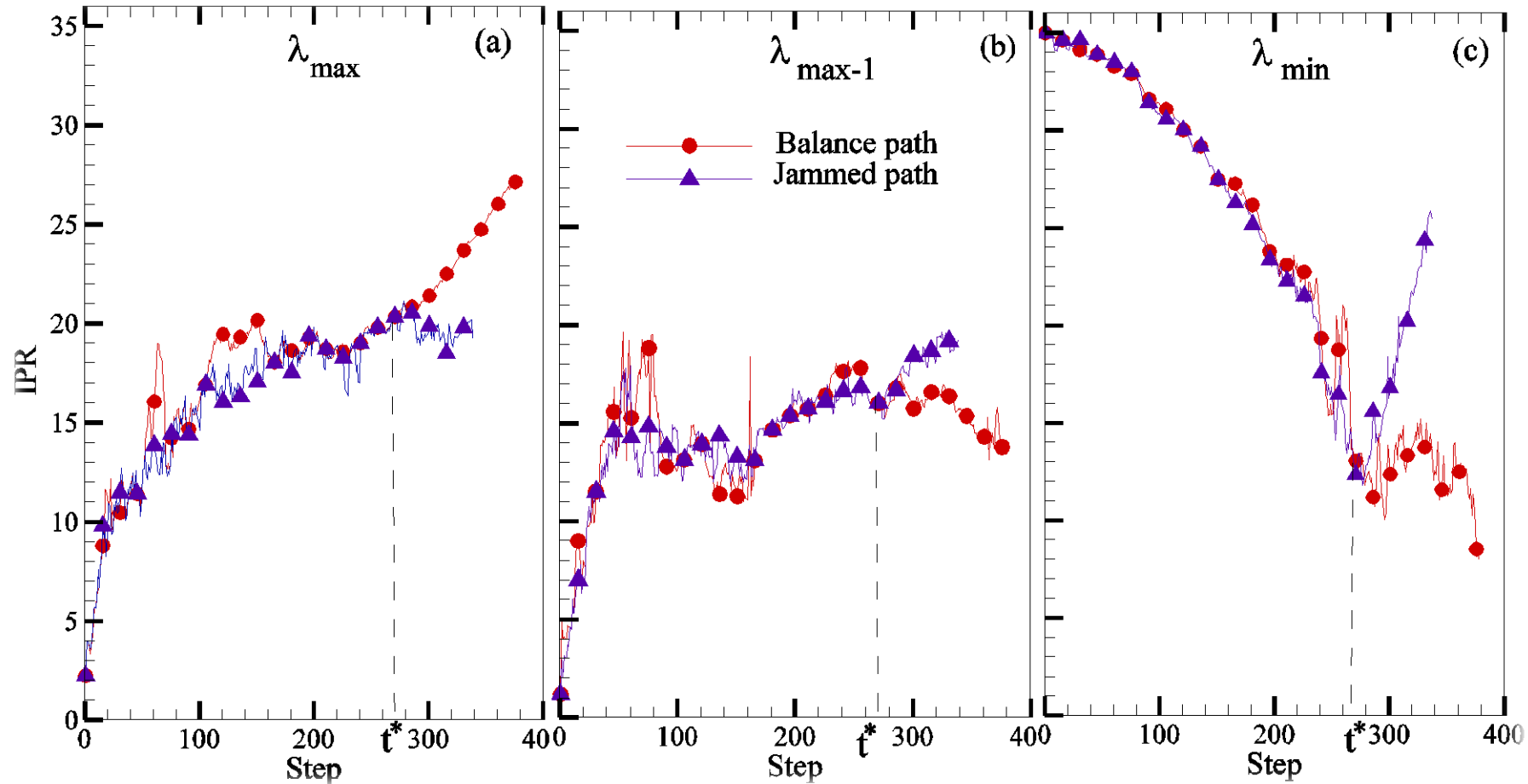
Jammed states are possible if and only if the network size is $N = 9$ or $N \geq 11$

Pseudo Paths



Pseudo paths towards minimum energy states in network dynamics, L. Hedayatifar, F. Hassanibesheli, A.H. Shirazi, S.V. Farahani, G.R. Jafari, Physica A, 483, 109-116 (2017)

Inverse participation ratio (IPR)



$$|\psi\rangle = \begin{pmatrix} \psi_1 \\ \psi_2 \\ \psi_3 \end{pmatrix}$$

$$\text{IPR} = \frac{1}{\sum_i |\psi_n(i)|^4}$$

$$1 \leq \text{IPR} \leq N$$

Pseudo paths towards minimum energy states in network dynamics, L. Hedayatifar, F. Hassanibesheli, A.H. Shirazi, S.V. Farahani, G.R. Jafari, Physica A, 483, 109-116 (2017)

The data: 20532 genes in the case of Breast Cancer

(BRCA: Breast invasive carcinoma)

It has been downloaded from data bank:

The Cancer Genome Atlas (TCGA) project.

The expression levels have done with the technique of RNA sequencing.



	Normal	Cancerous samples
#Samples	114	764

nature
genetics

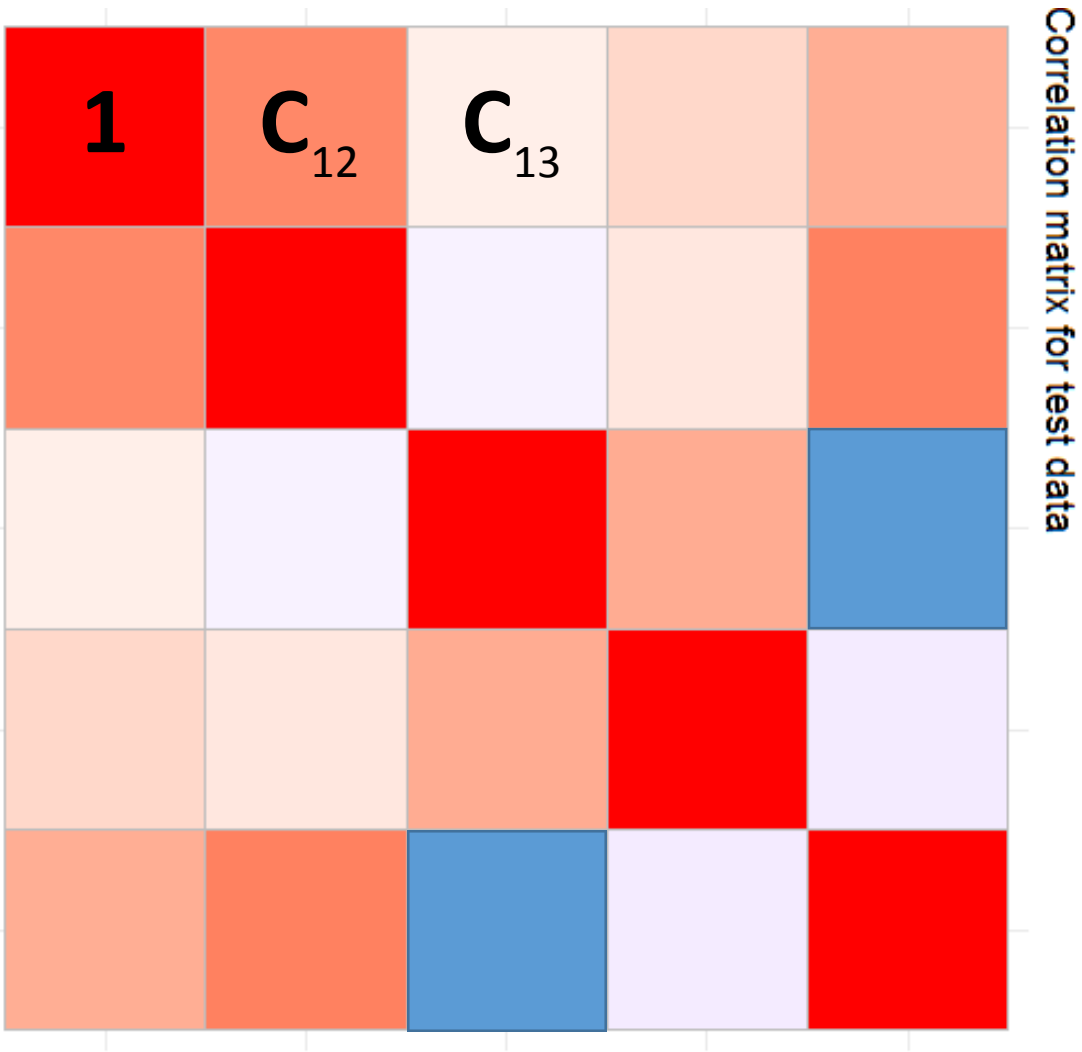
Commentary | [OPEN](#) | Published: 26 September 2013

The Cancer Genome Atlas Pan-Cancer analysis project

The Cancer Genome Atlas Research Network, John N Weinstein, Eric A Collisson, Gordon B Mills, Kenna R Mills Shaw, Brad A Ozenberger, Kyle Ellrott, Ilya Shmulevich, Chris Sander & Joshua M Stuart [✉](#)

Nature Genetics **45**, 1113–1120 (2013) | [Download Citation](#) ↓

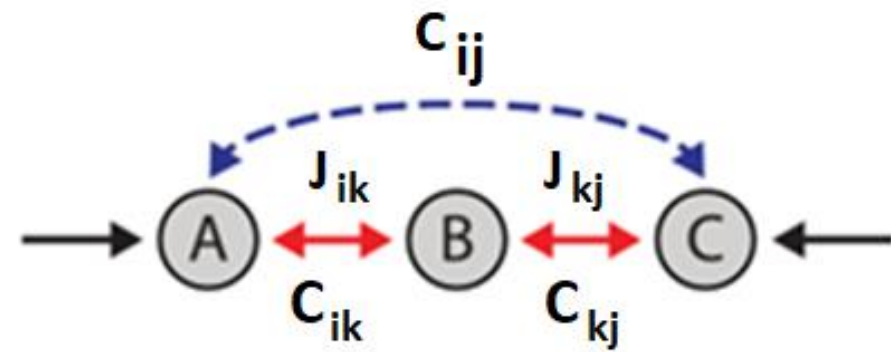
Correlation Matrix



$$C_{ij} = \frac{\sum_{l=1}^n (X_l - \langle X \rangle)(Y_l - \langle Y \rangle)}{\sqrt{\sum_{l=1}^n (X_l - \langle X \rangle)^2} \sqrt{\sum_{l=1}^n (Y_l - \langle Y \rangle)^2}}$$

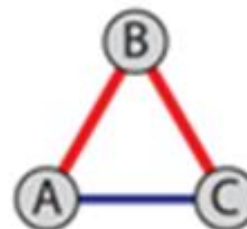
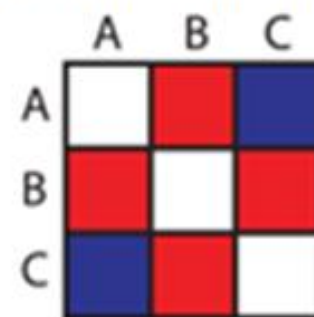
We had to reduce the number of genes because it is a difficult task to handle a 20532 in 20532 matrix computationally.

- Fortunately, most of the correlation matrix is zero
- There are many Genes that their expressions are the same in normal and cancerous cells. So for better comparison, we removed from our data sets. Now, we have only 450 genes to study.

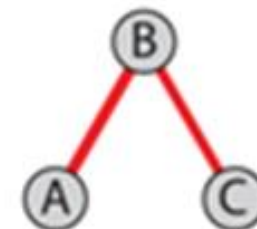
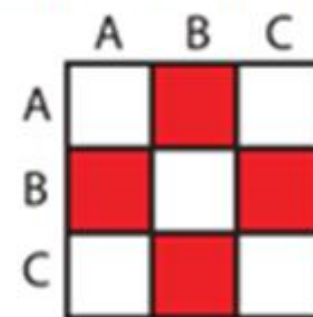


Partial Correlation

Correlation Matrix



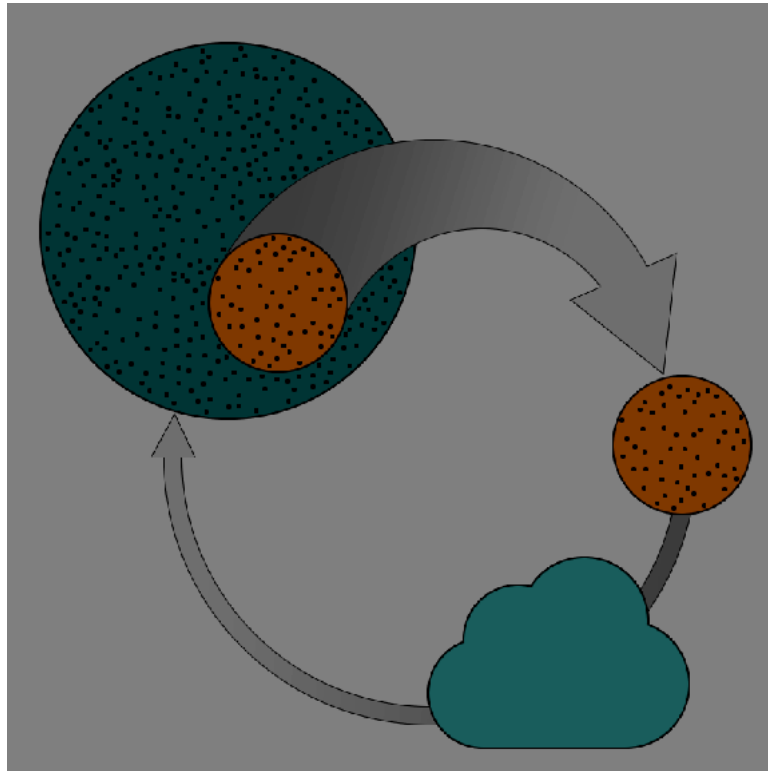
Interaction Matrix



Inverse Spin Glass and Related Maximum Entropy Problems,
Michele Castellana, and William Bialek, 2013 PRL 113(11)

Lezon TR, Banavar JR, Cieplak M, Maritan A, Fedoroff NV, [PNAS](#), 12;103(50):19033-8 (2006)

Invers problem



$$H = - \sum_{ij} J_{ij} s_i s_j - \sum_i h_i s_i$$

$$\int_x P(x) dx = 1$$

$$\langle x_i \rangle = \int_x P(x) x_i dx = \frac{1}{M} \sum_{m=1}^M x_i^m = \bar{x}_i$$

$$\langle x_i x_j \rangle = \int_x P(x) x_i x_j dx = \frac{1}{M} \sum_{m=1}^M x_i^m x_j^m = \overline{x_i x_j}$$

$$\text{maximize } S = - \int_x P(x) \ln P(x) dx$$

Max Entropy

With some Conjunctions

$$L = L(P(x); \alpha, \beta, \gamma)$$

$$\frac{\delta L}{\delta P(x)} = 0$$

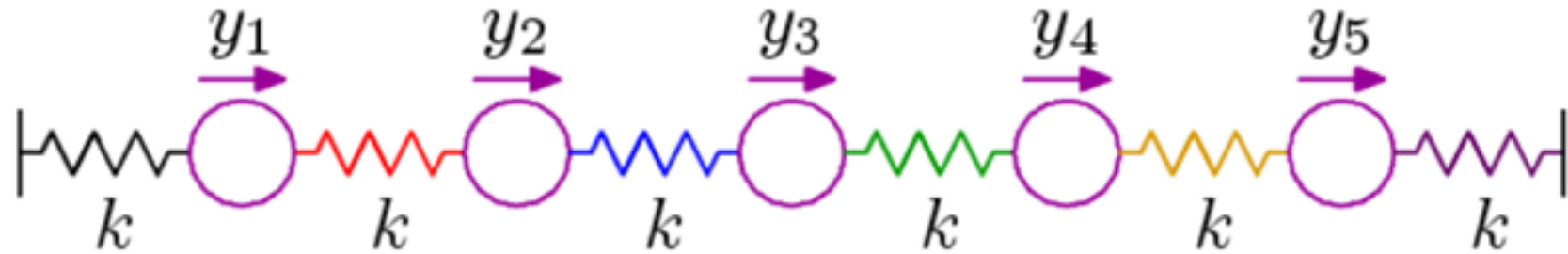
$$P(x, \beta, \gamma) = \exp(-1 + \alpha + \sum_i^L \beta_i x_i + \sum_{i,j}^L \gamma_{ij} x_i x_j) = \frac{1}{Z} e^{-H(x, \beta, \gamma)}$$

$$P(x; \langle x \rangle, C) = (\pi)^{\frac{-L}{r}} \det(C)^{\frac{-1}{r}} \exp\left(-\frac{1}{r} (x - \langle x \rangle)^T C^{-1} (x - \langle x \rangle)\right)$$

Lezon TR, Banavar JR, Cieplak M, Maritan A, Fedoroff NV, [PNAS](#), 12;103(50):19033-8 (2006)

$$J = -\frac{1}{2} C^{-1}$$

Inverse-covariance matrix



inverse-covariance matrix

or

covariance matrix?

$$\mathbf{K}^{-1} = \frac{k}{T} \begin{bmatrix} 2 & -1 & 0 & 0 & 0 \\ -1 & 2 & -1 & 0 & 0 \\ 0 & -1 & 2 & -1 & 0 \\ 0 & 0 & -1 & 2 & -1 \\ 0 & 0 & 0 & -1 & 2 \end{bmatrix}$$

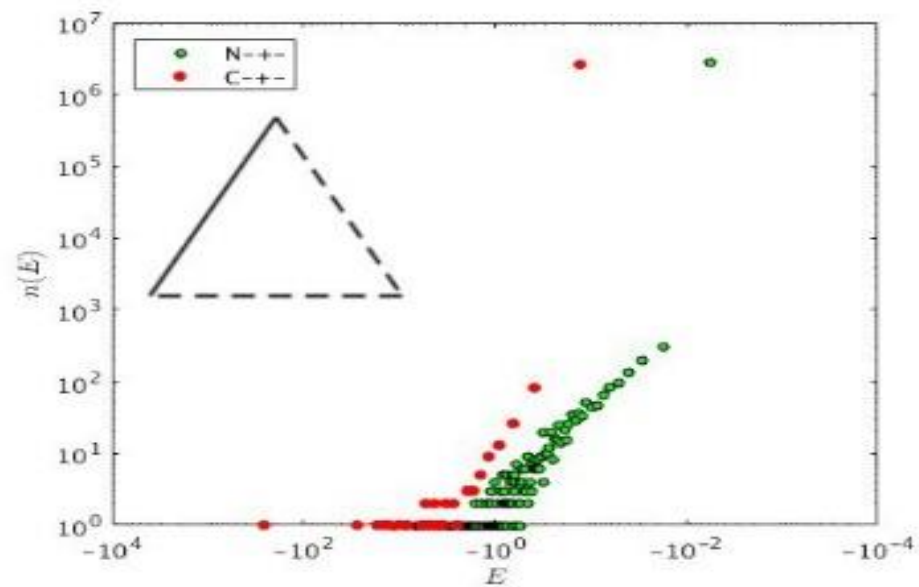
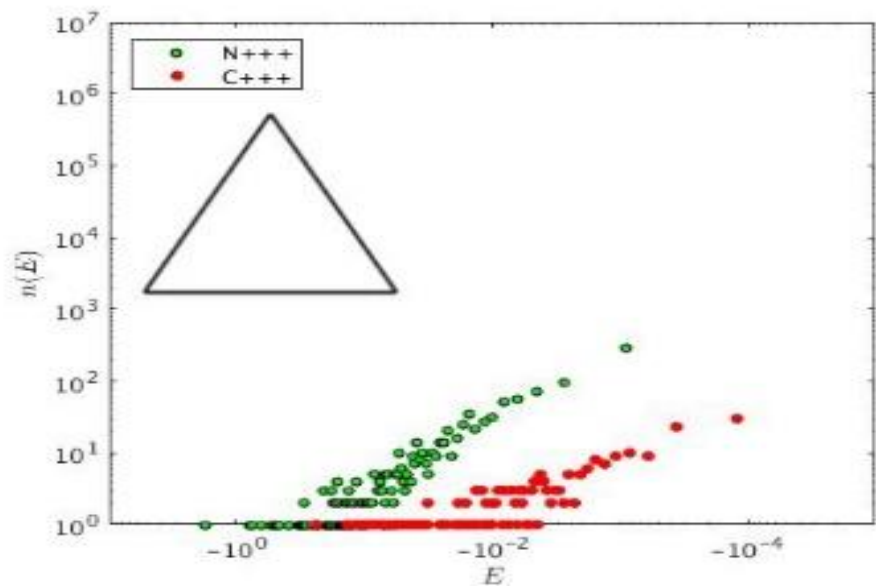
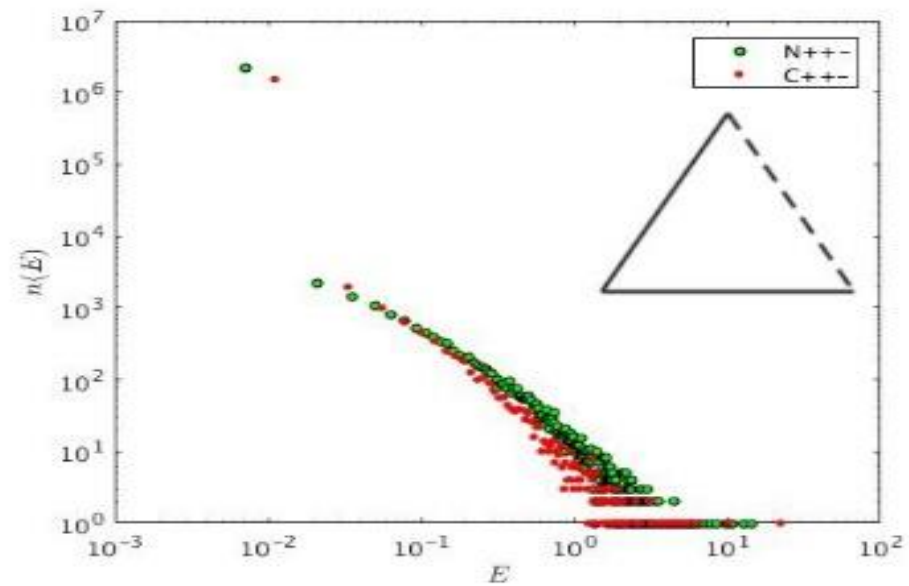
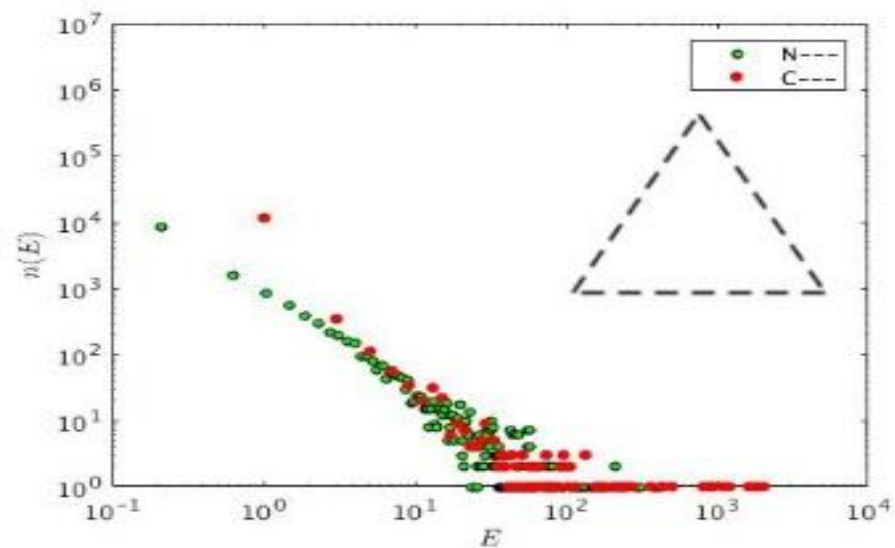
$$\mathbf{K} = \frac{T}{k} \begin{bmatrix} 1.00 & 0.67 & 0.50 & 0.33 & 0.17 \\ 0.67 & 1.00 & 1.00 & 0.67 & 0.33 \\ 0.50 & 1.00 & 1.00 & 1.00 & 0.50 \\ 0.33 & 0.67 & 1.00 & 1.00 & 0.67 \\ 0.17 & 0.33 & 0.50 & 0.67 & 1.00 \end{bmatrix}$$

Correlation matrix to Interaction matrix

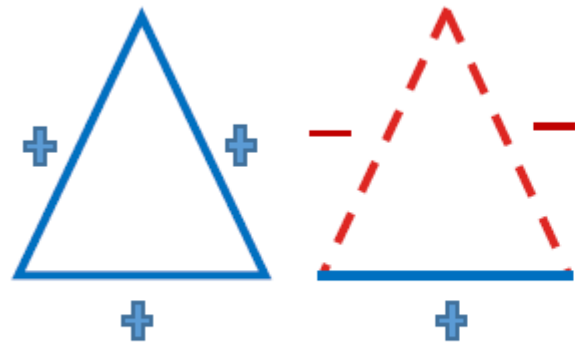
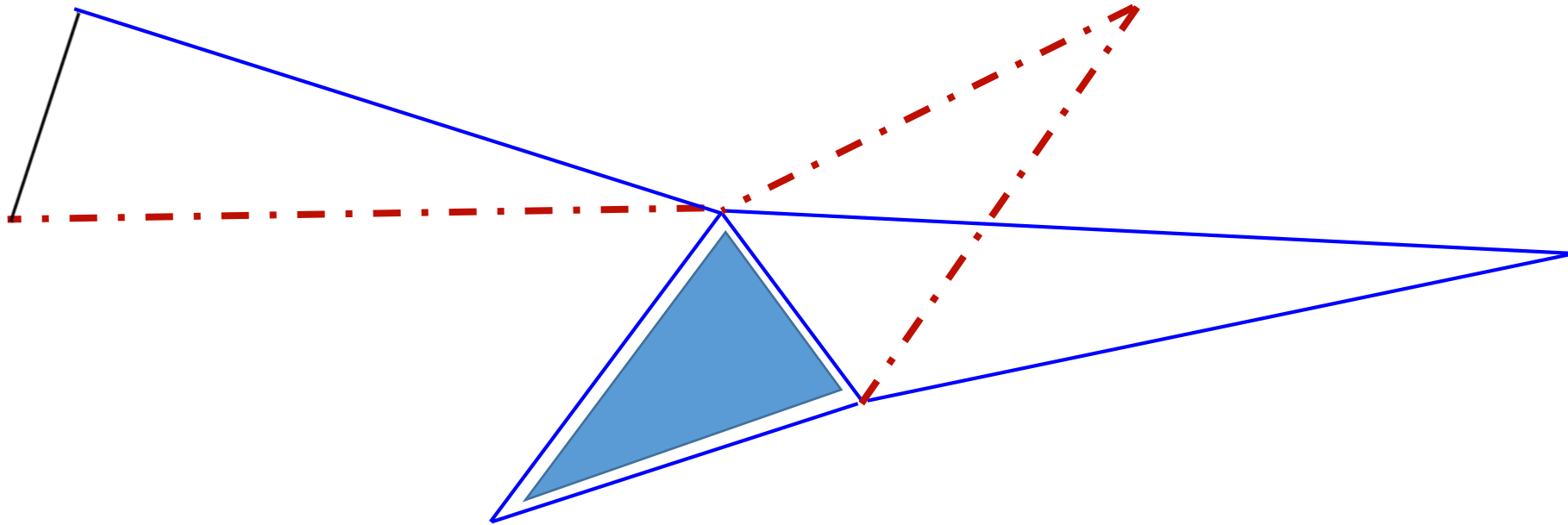
$$C = \begin{pmatrix} C_{11} & C_{12} & \dots \\ C_{21} & C_{22} & \dots \\ \vdots & & \ddots \end{pmatrix} \rightarrow \frac{-1}{2} C^{-1} = \begin{pmatrix} 0 & + & \dots \\ + & 0 & \dots \\ + & - & 0 & \dots \\ \vdots & & & \ddots \end{pmatrix} = J$$

$E = -J_{12} J_{13} J_{23}$

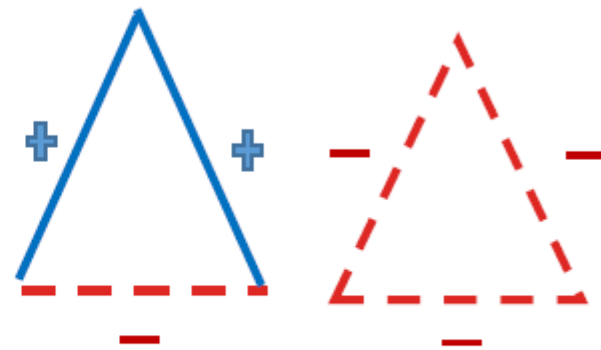
Energy distribution



Energy-Energy Correlation



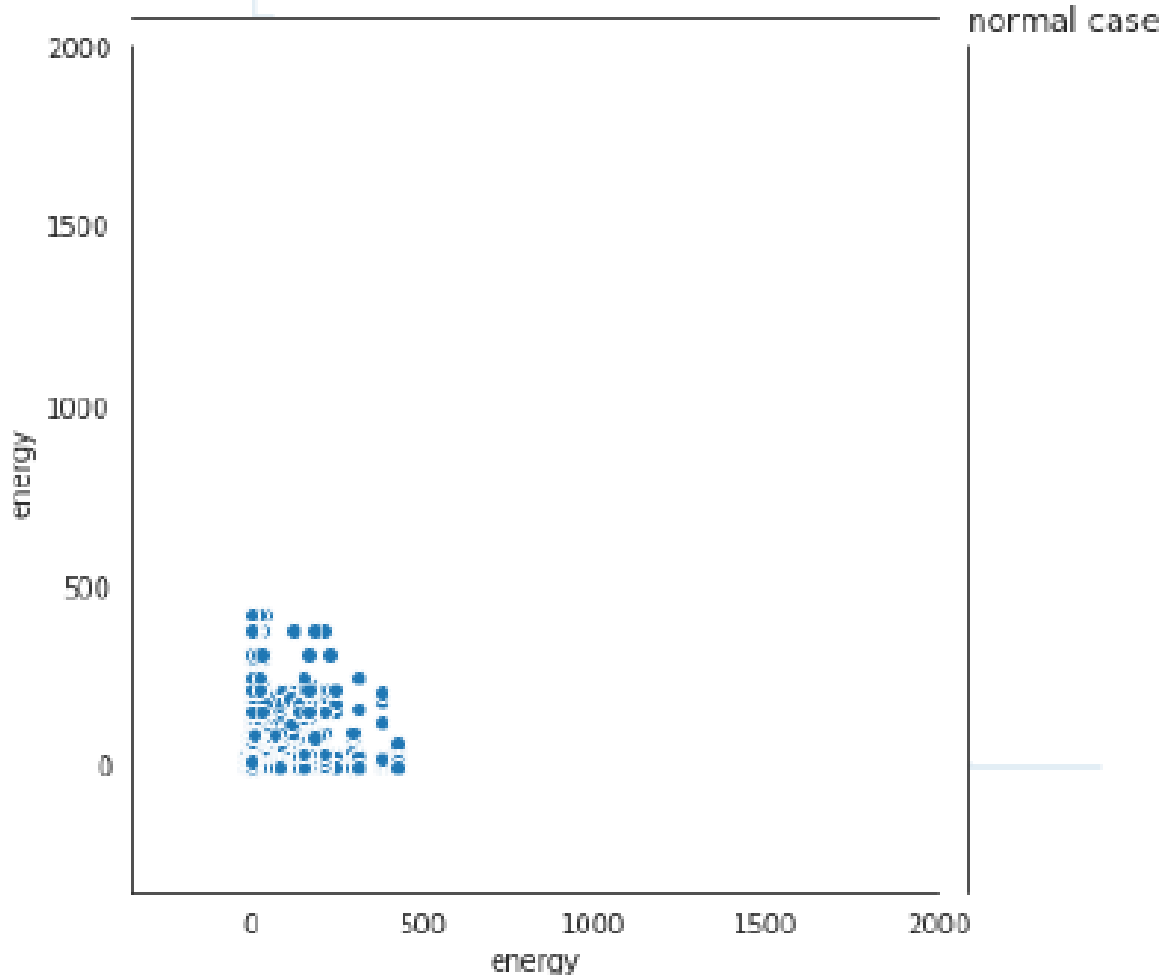
(a) Balanced



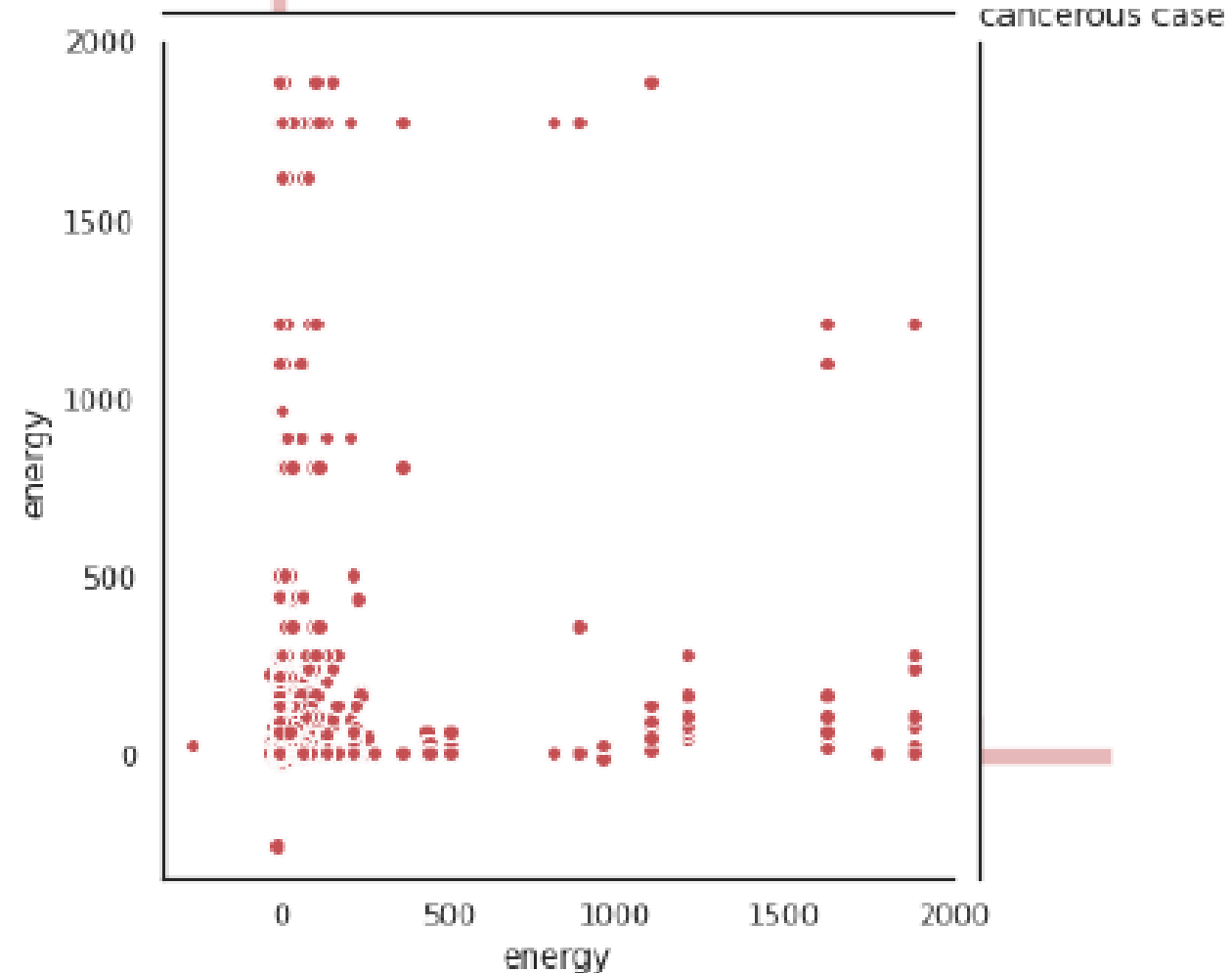
(b) Imbalanced

How triads with different energies are connected to each other?

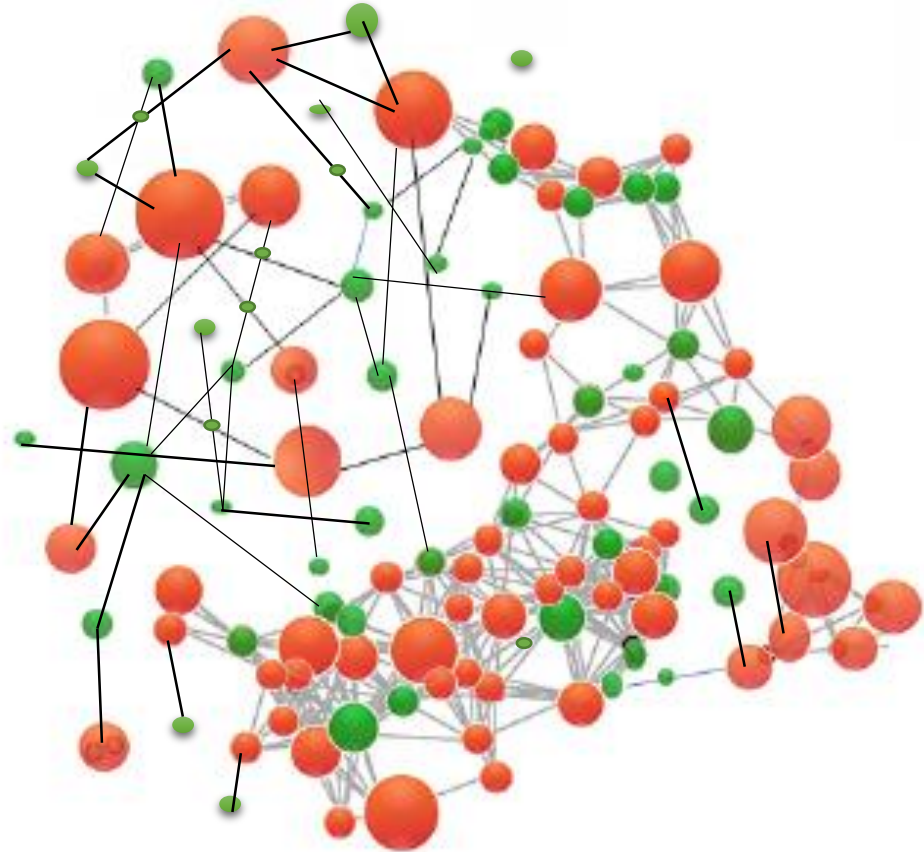
Normal Cells



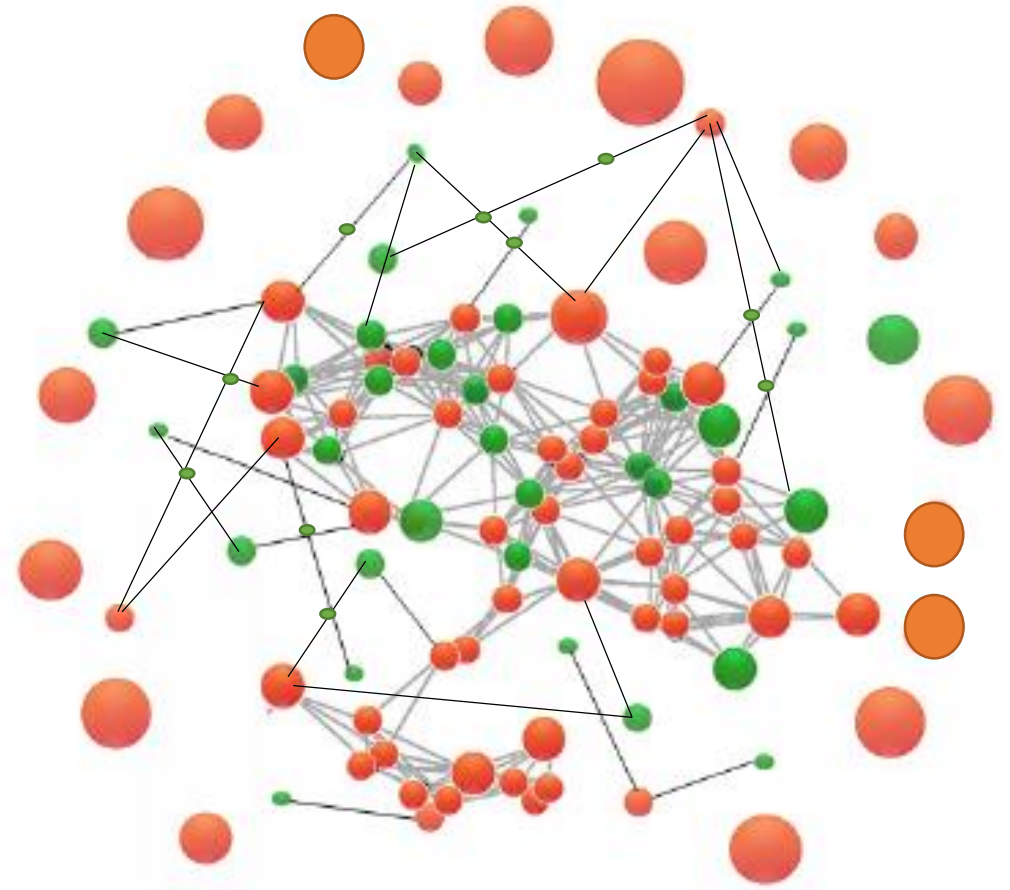
Cancerous cell



Network of balance and imbalance triangles



Cancerous cell



Normal Cells

λ_1 λ_2 λ_N

$|\psi\rangle = \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_N \end{pmatrix}, |\psi\rangle = \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_N \end{pmatrix} \dots$

$NPR_i = \sum_{i=1}^N |\alpha_i|^4$

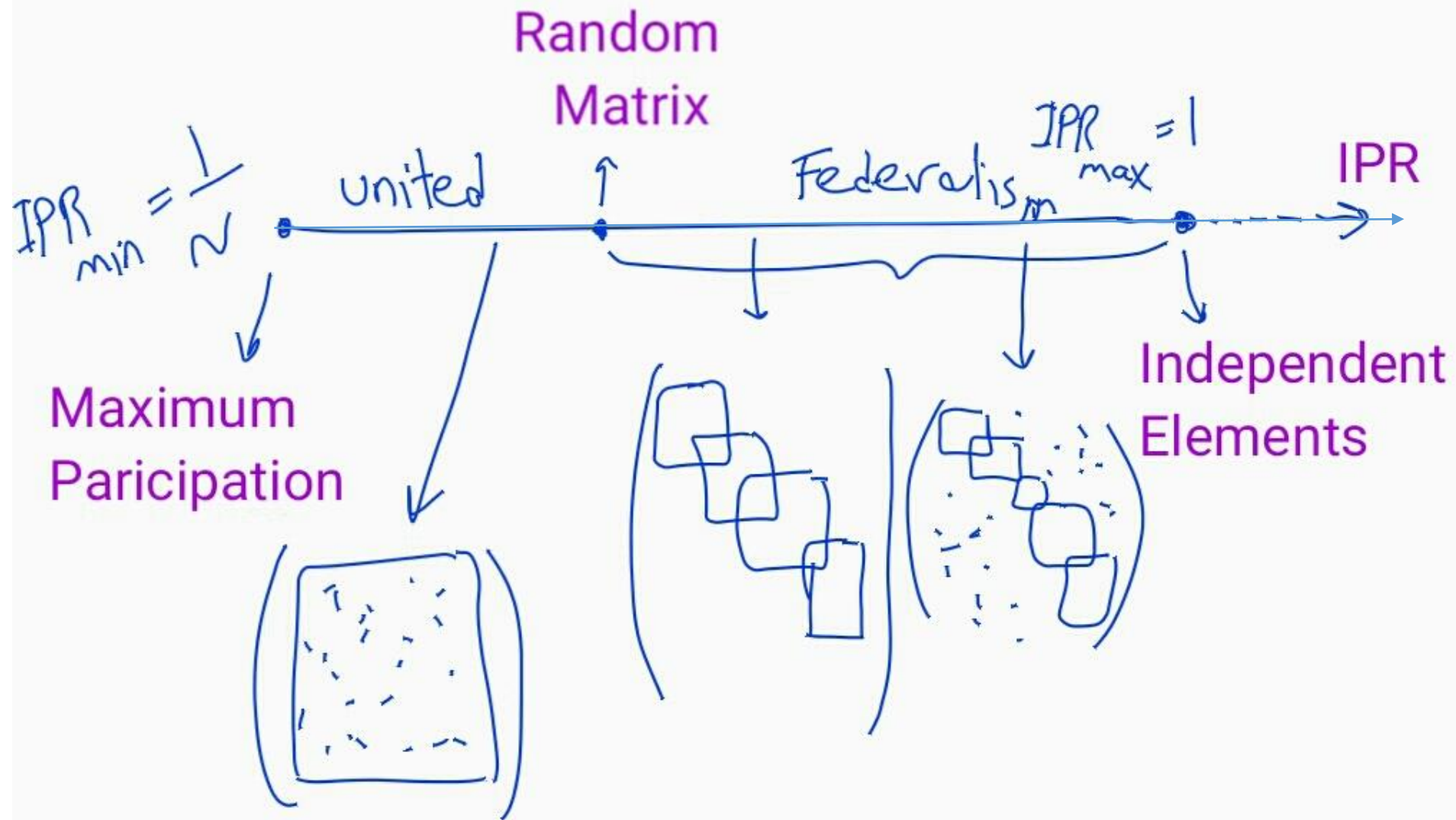
$IPR_i = \frac{1}{\sum_{i=1}^N |\alpha_i|^4}$

$PR_i = \sum_{i=1}^N |\alpha_i|^4$

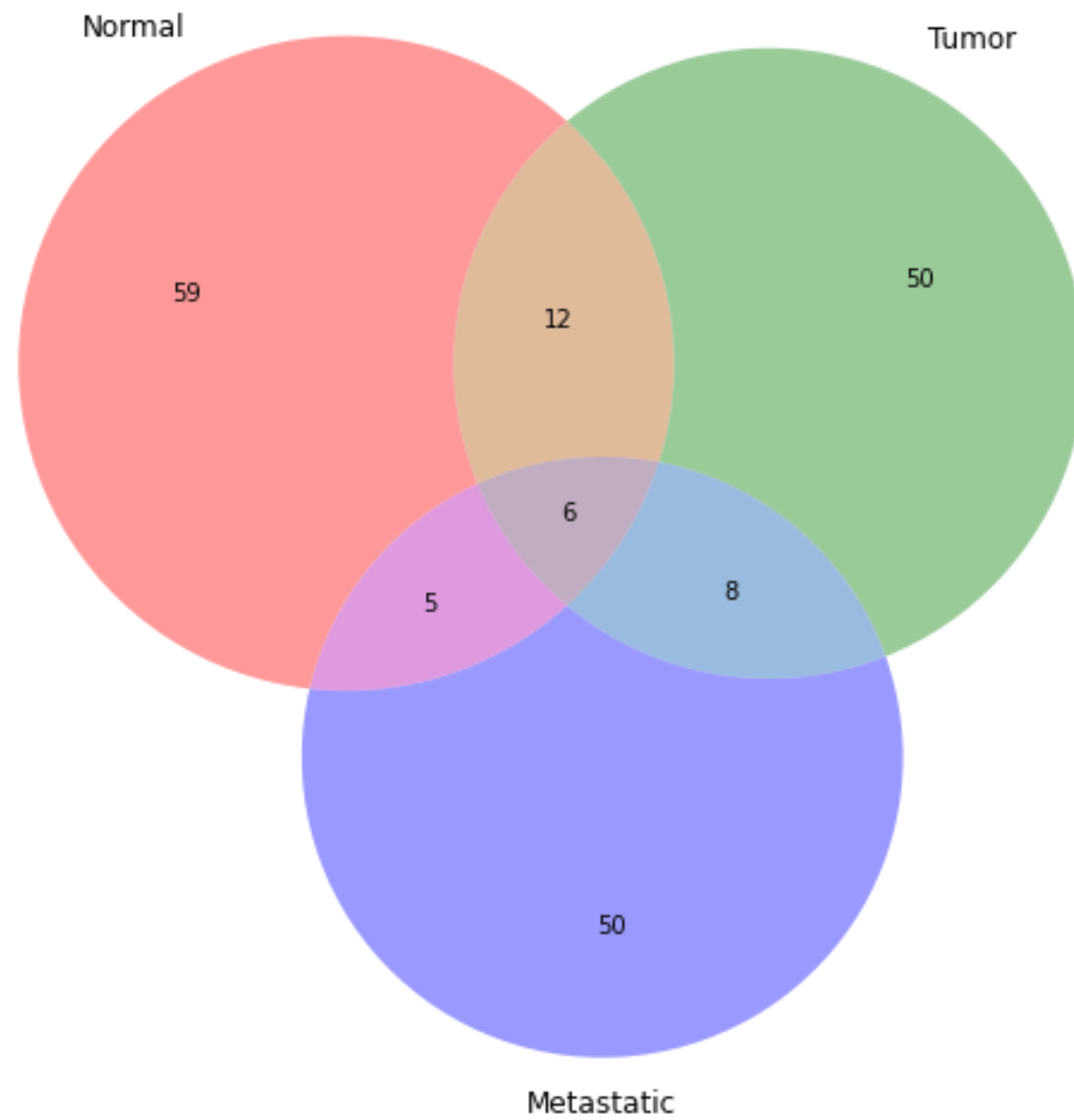
IPR

$|\psi\rangle = \begin{pmatrix} 1 \\ \vdots \\ 0 \end{pmatrix}$

$IPR_{\min} = \frac{1}{\sum_{i=1}^N \left(\frac{1}{\sqrt{N}}\right)^4} = \frac{1}{N}$



Top genes according to W

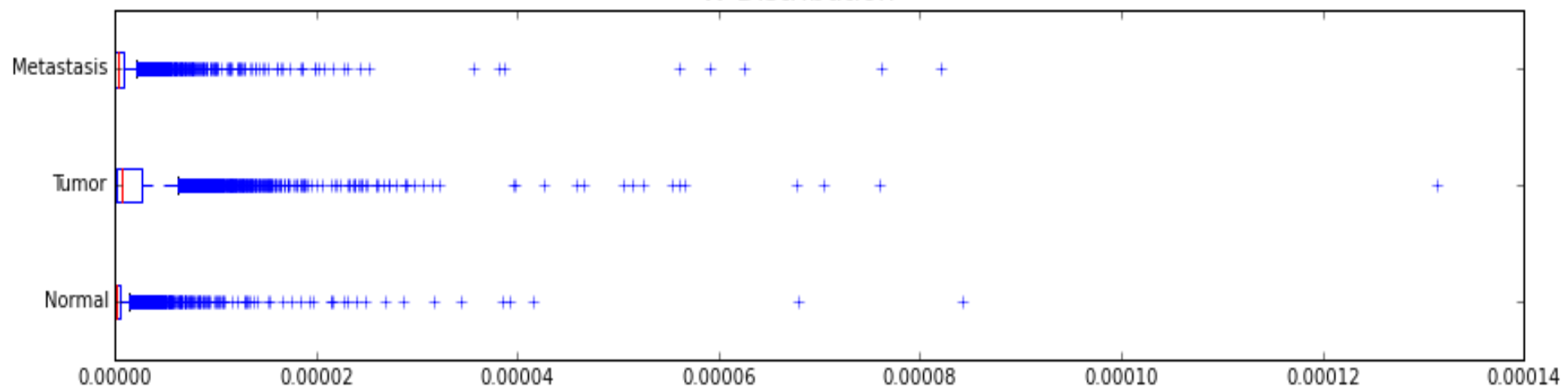
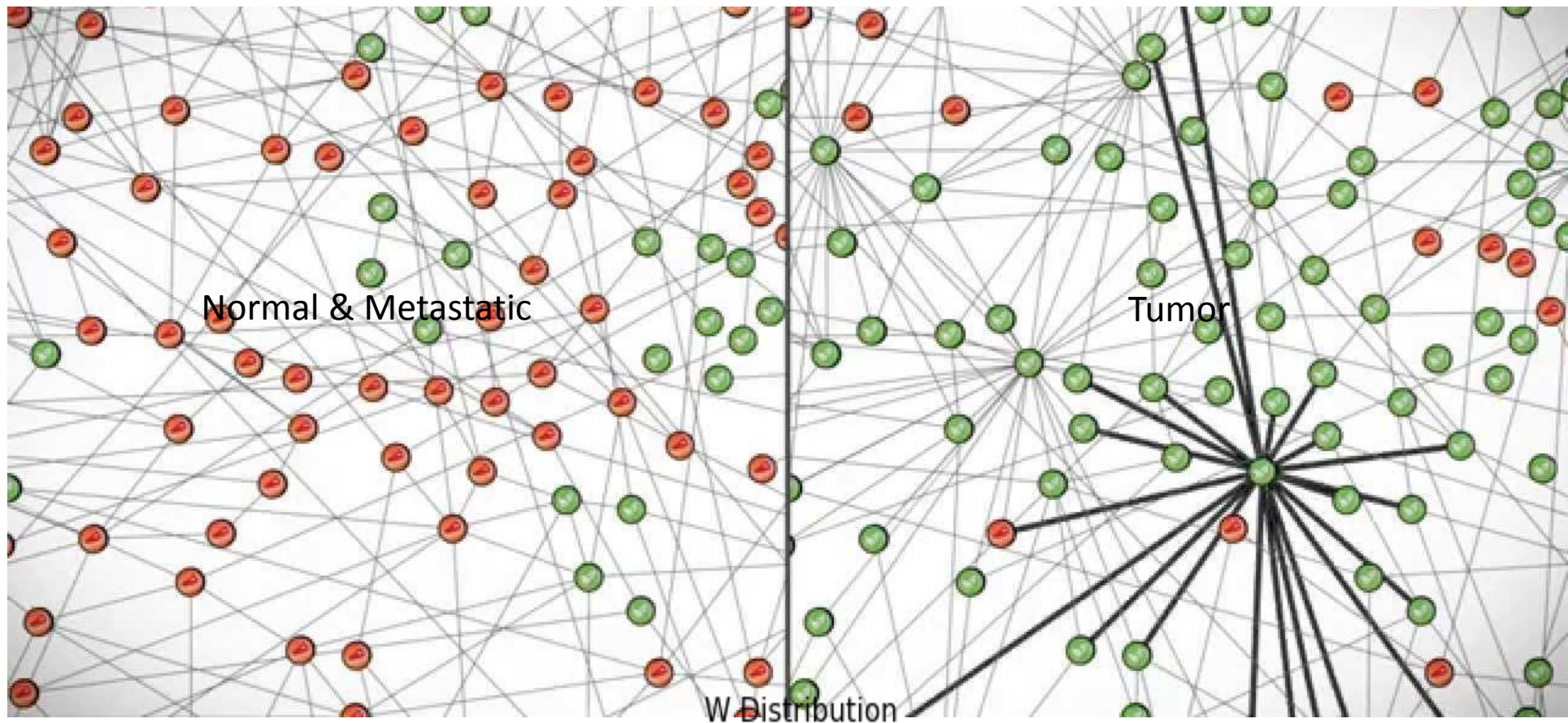


Top driver genes in metastatic cells

ALB	M64936	CSTA	IGL@	APOA1	DST	LOC10012658 3	D26561	ESM1	NF1
PCK1	RBP4	MUC1	FN1	KLK3	CYP3A7	HPR	ACTG2	PDE5A	ALDOB
CRP	IGL@	CHI3L1	FN1	FABP4	ALB	KLK2	IRS1	MAPK8	TGM4
MYH11	IGHV4-31	MCF2	UGT2B15	PSPH	FGG	HPGD	APOB	HPGD	STAC
KLK3	GATM	SULT1C2	PRKG2	MAGEB1	PRG4	SSX2B	KLK3	FGB	SLC25A13
FGA	KLK2	FABP1	DDC	C5	MHY11	CPB1	PLAT	IL2	PF4V1
IGK@	SLC25A24	MSMB	AF070543	ORM2	IGH@	KNG1	PCK1	SERPINA1	

Cancer related
 Immune system
 related
 Growth related

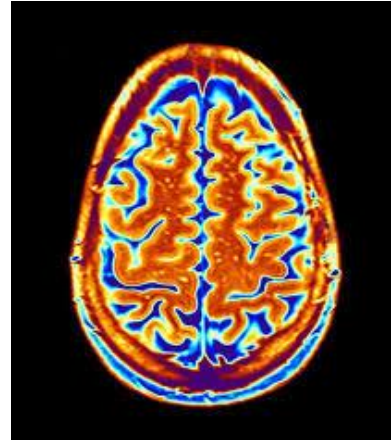
Adhesion and migration
 related
 Proliferation and
 differentiation





Autism Brain Imaging
Data Exchange

Summary of data:



- ❖ The data used to build the brain networks is borrowed from the **Autism Brain Imaging Data Exchange (ABIDE II)** which is available in:
http://fcon_1000.projects.nitrc.org/indi/abide/abide_II.html
- ❖ We chose **the longitudinal collections** in ABIDE II, which are from the university of California Los Angeles and the university of Pittsburgh.
- ❖ We had **23** and **15** participants in the Autism Spectrum Disorder and the control groups, respectively. Each participant underwent **two resting-state fMRI sessions**, collected one to three years apart.

Longitudinal Collections

University of California Los Angeles:
Longitudinal Sample



University of Pittsburgh
School of Medicine: Longitudinal Sample





Langone Medical Center

Hyperactive



Summary of data:

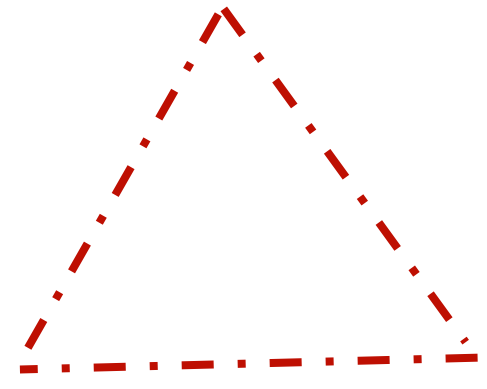
❖ The data used to build the following brain networks is from the **ADHD-200** dataset and is available in:

http://fcon_1000.projects.nitrc.org/indi/adhd200/



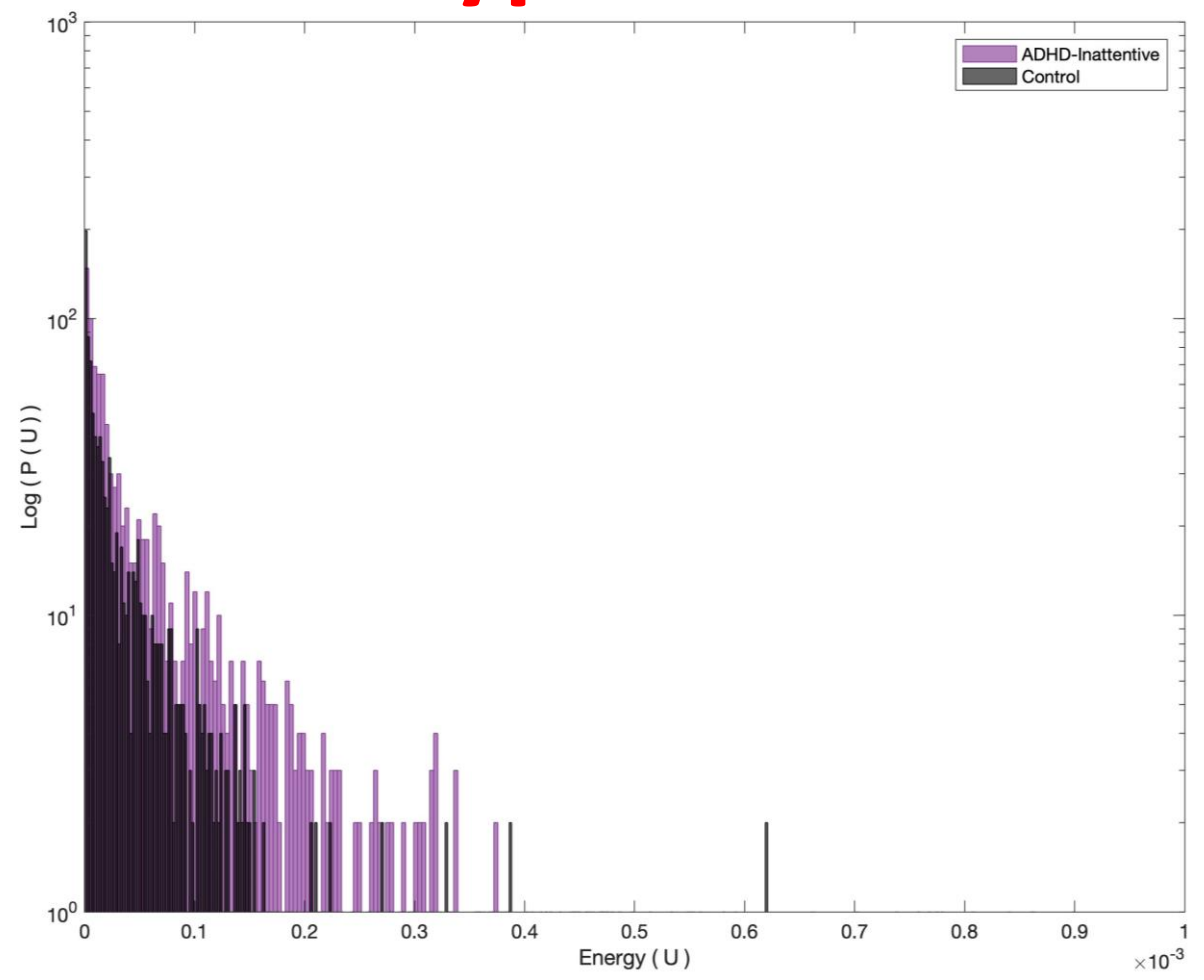
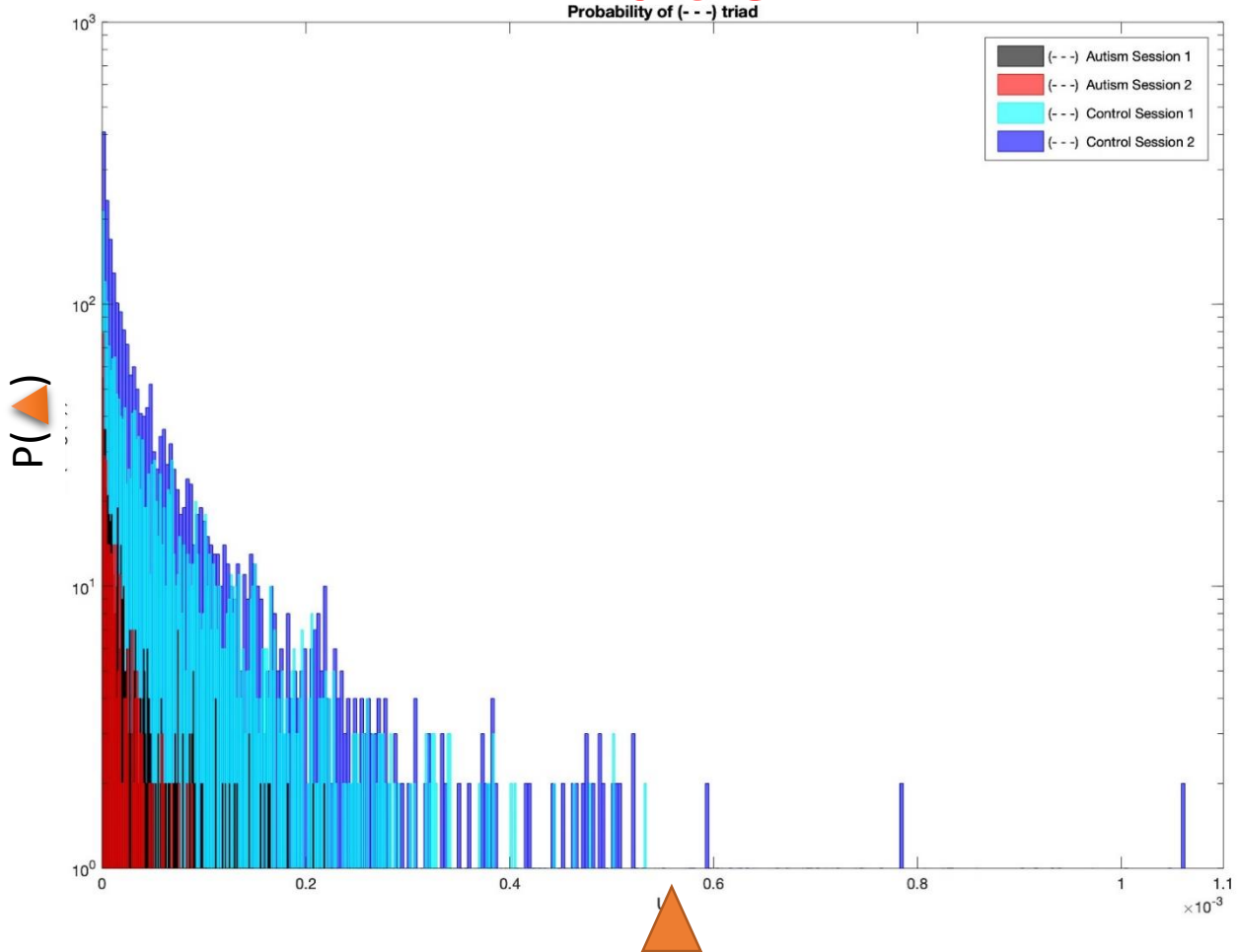
- ❖ We chose the **New York University Child Study Center** site, from the ADHD-200.
- ❖ After the preprocessing steps, the sample size was:
 - ❖ **98** individuals in the **Control group**
 - ❖ **115** individuals in the **ADHD (Attention Deficit **Hyperactivity** Disorder)**

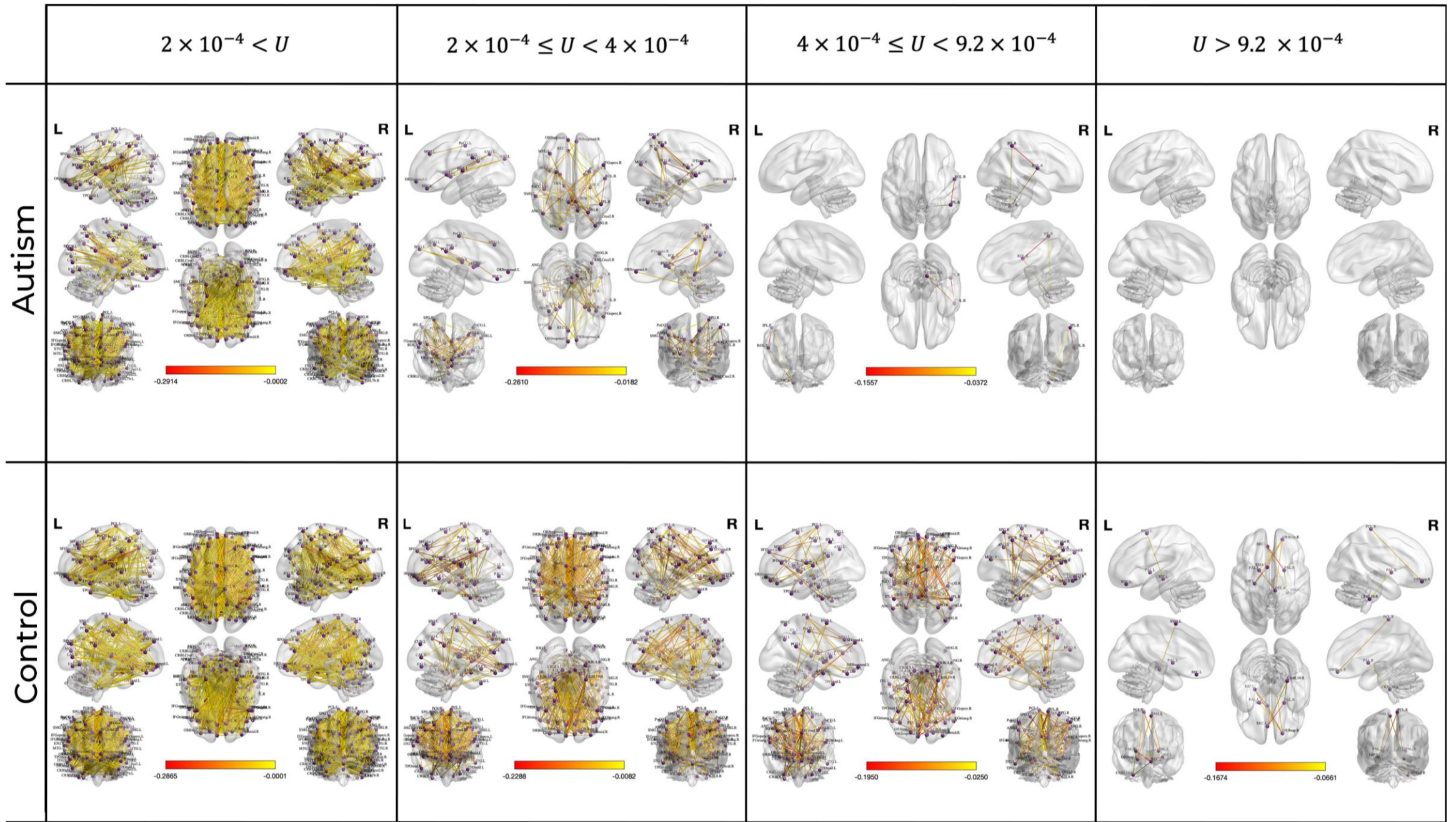
PDF of



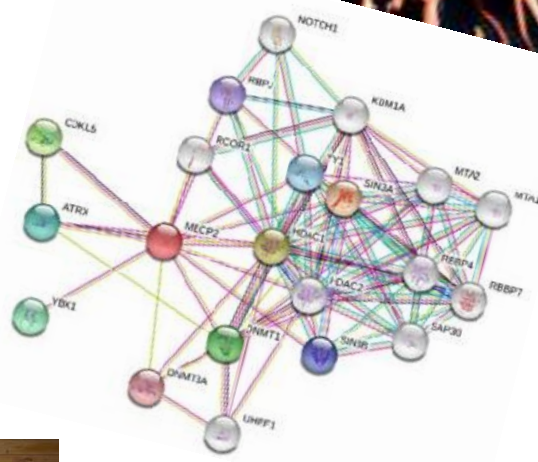
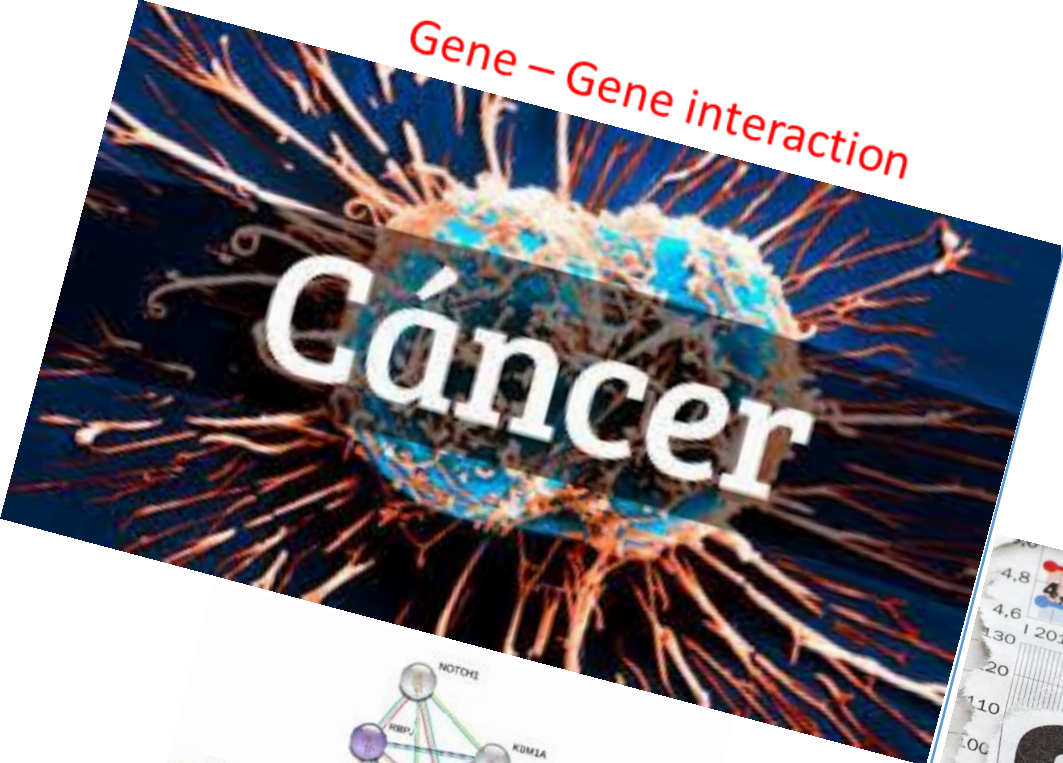
Autism

Hyperactive





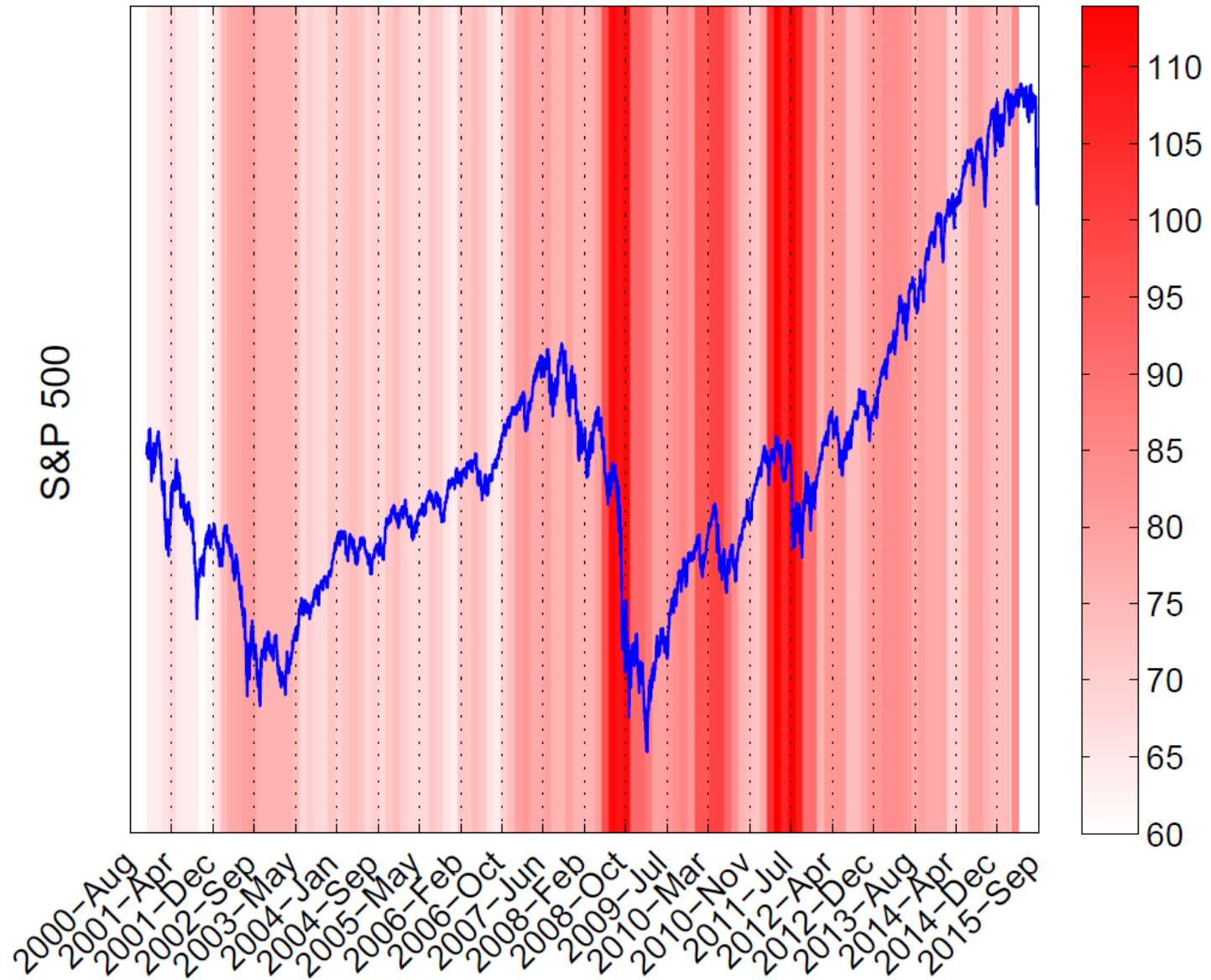
Gene – Gene interaction



Market – Market interaction



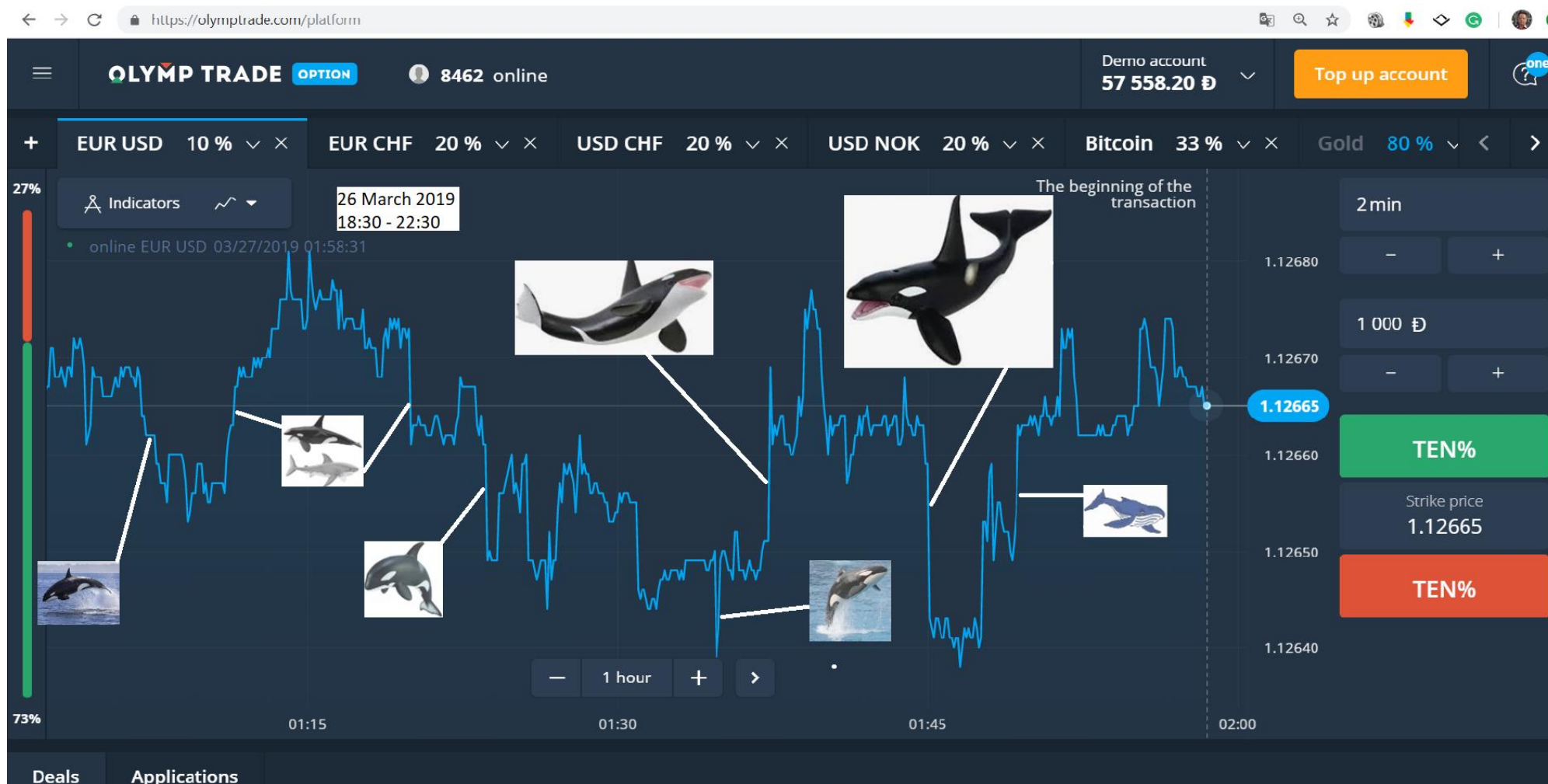
Heat map of S&P500

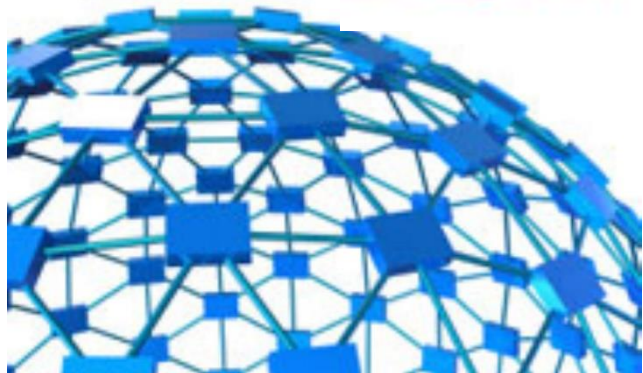
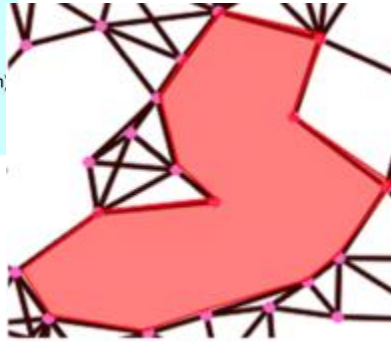
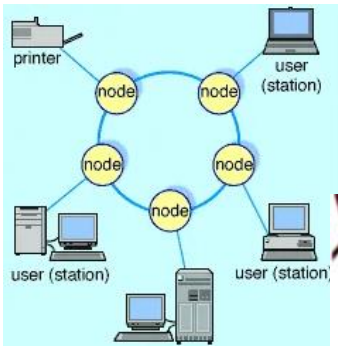
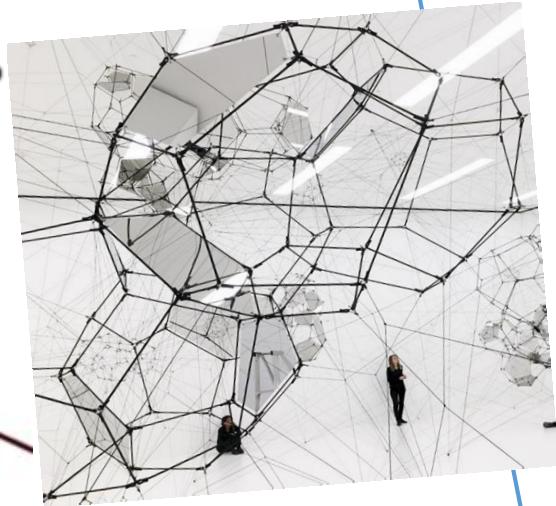
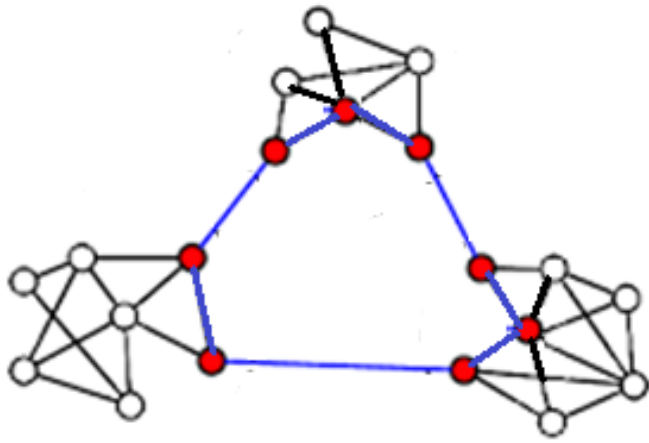




The Whales dance

Quanta Trading Signals





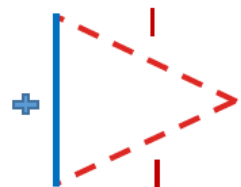
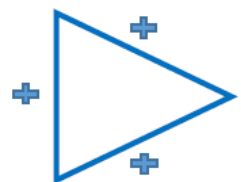
Thanks

Compare the Balance states between 3th and 4th order interactions

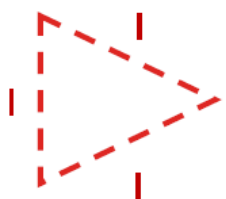
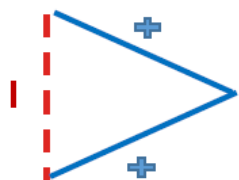
3th order interactions

$$\mathcal{H} = - \sum_{i < i < k} \Delta_{ijk}$$

(a) Balanced

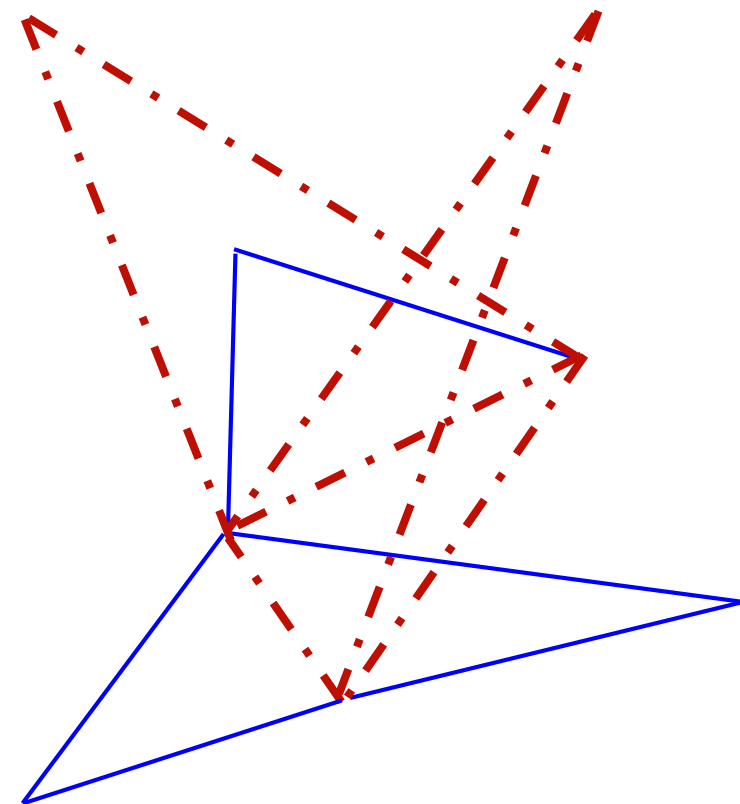
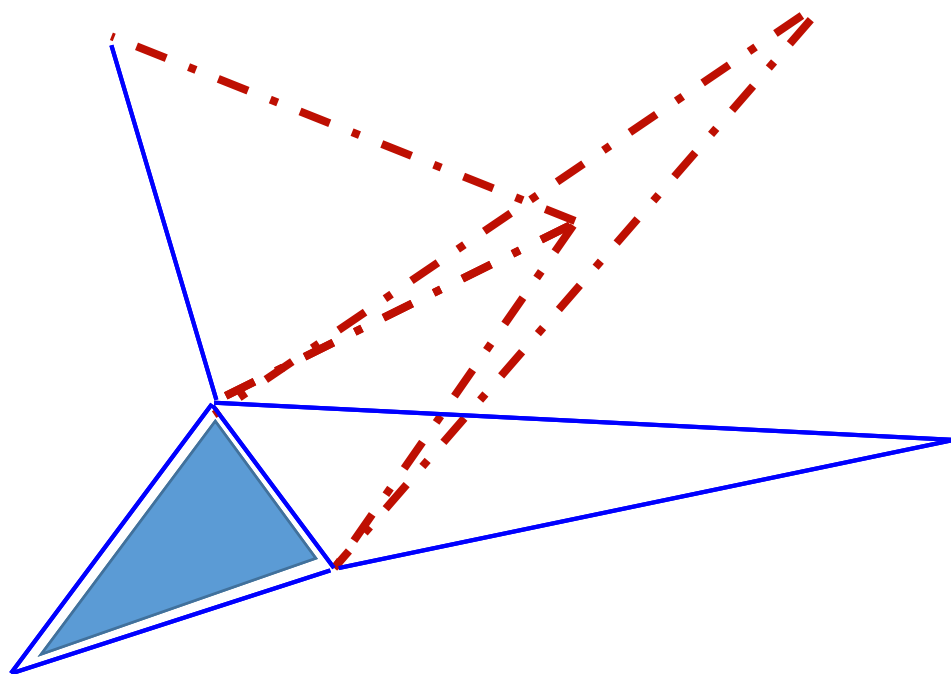


(b) Imbalanced

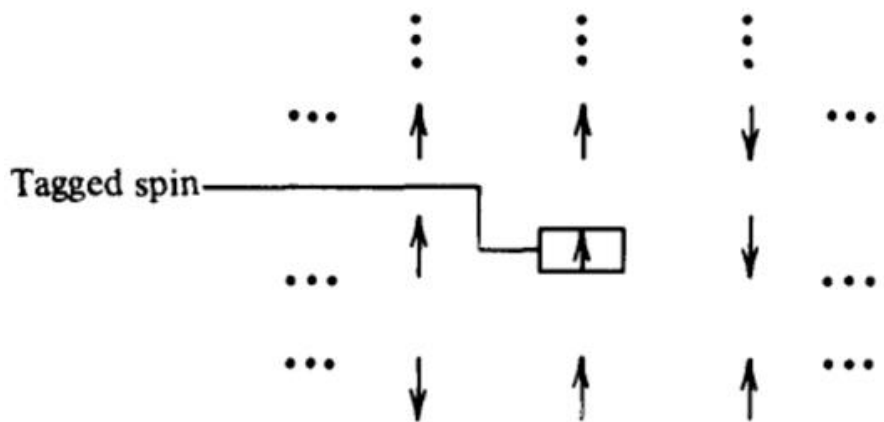


$$\mathcal{H} = - \sum_{i < i < k < l} \Delta_{ijk} \Delta_{ijl}$$

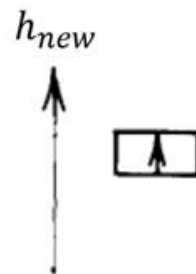
$$= - \sum_{i < i < k < l} \Delta_{ijk} \Delta_{ijl}$$



Schematic mean field theory



Mean field approximation



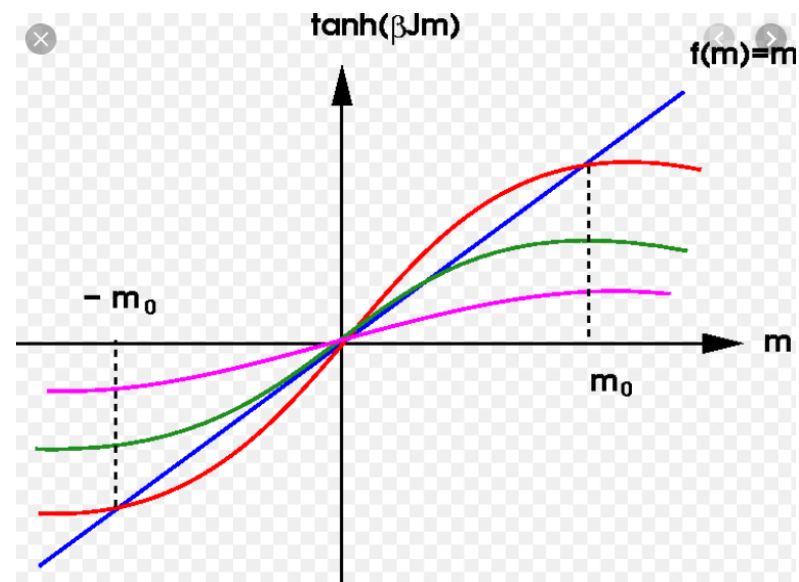
$$H = -h \sum_i^N S_i - J \sum_{ij}^N S_i S_j$$

Mean field

$$H_i = -(h + J \langle S \rangle) S_i$$

$$h_{new} = (h - J \langle S \rangle)$$

$$H = \sum_i^N H_i$$



$$H = -h \sum_i^N S_i - J \sum_{ij}^N S_i S_j$$

$$Z = \sum e^{-\beta H}$$

$$m = \langle S \rangle = (1/N) \sum_i^N S_i$$

$$m = \tanh(\beta J m)$$

Mean-field solution of structural balance dynamics in nonzero temperature

$$H = -J \sum_{ij} S_{ij} S_{jk} S_{ki}$$

$$H_{ij} = -S_{ij} J \sum_k S_{jk} S_{ki}$$

$$P\{S_{ij}\} = P(S_{ij} = 1) * (1) + P(S_{ij} = -1) * (-1)$$

$$= \left\langle \frac{e^{-\beta H_{ij}(S_{ij}=1)} - e^{-\beta H_{ij}(S_{ij}=-1)}}{e^{-\beta H_{ij}(S_{ij}=1)} + e^{-\beta H_{ij}(S_{ij}=-1)}} \right\rangle$$

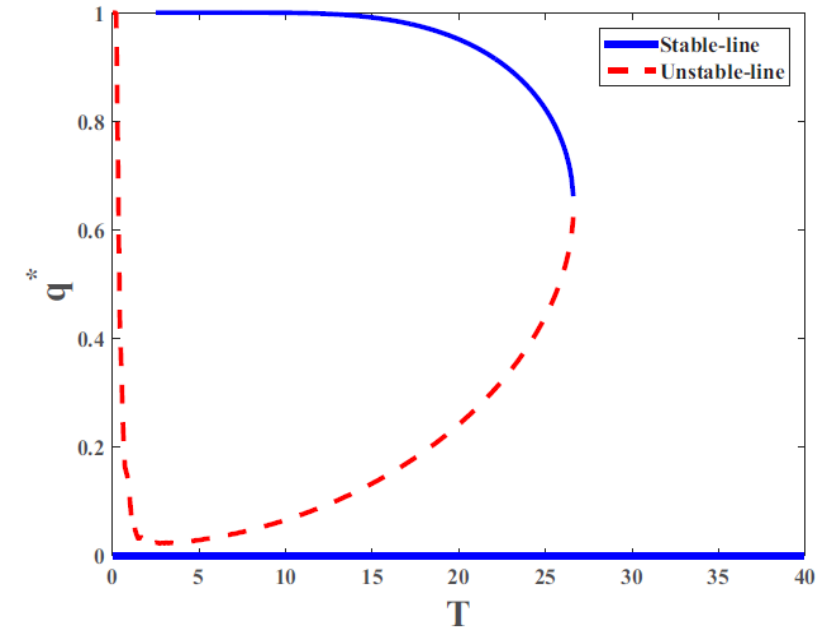
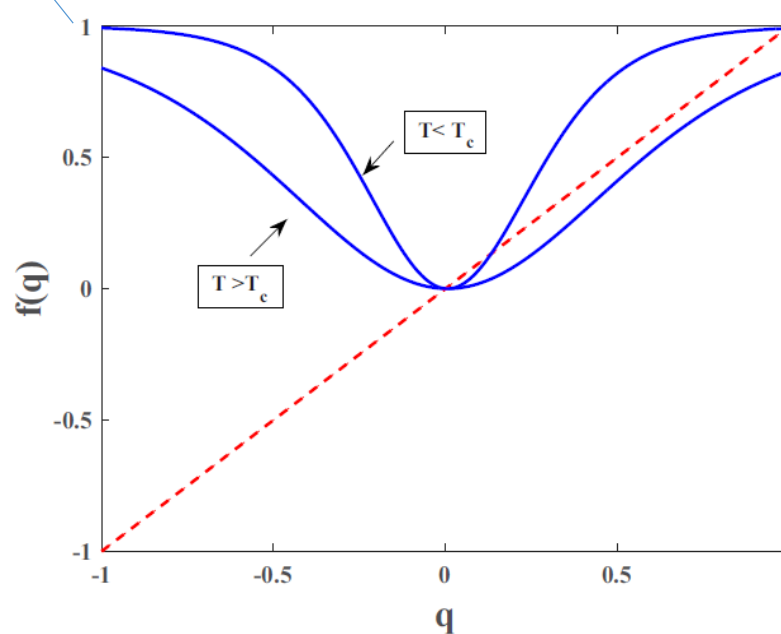
$$= \left\langle \frac{e^{\beta \sum_{k \neq i,j} S_{jk} S_{ki}} - e^{-\beta \sum_{k \neq i,j} S_{jk} S_{ki}}}{e^{\beta \sum_{k \neq i,j} S_{jk} S_{ki}} + e^{-\beta \sum_{k \neq i,j} S_{jk} S_{ki}}} \right\rangle$$

$$= \left\langle \tanh \left(\beta \sum_{k \neq i,j} S_{jk} S_{ki} \right) \right\rangle$$

$$p = \tanh(\beta(N-2)q), \quad q = \left[\frac{e^{-\beta(N-3)(-2q)} - 2e^{-\beta(2p)} + e^{-\beta(N-3)(2q)}}{e^{-\beta(N-3)(-2q)} + 2e^{-\beta(2p)} + e^{-\beta(N-3)(2q)}} \right]$$

To solve these two equations with two unknowns, we use a self-consistency condition on q by substituting p into q :

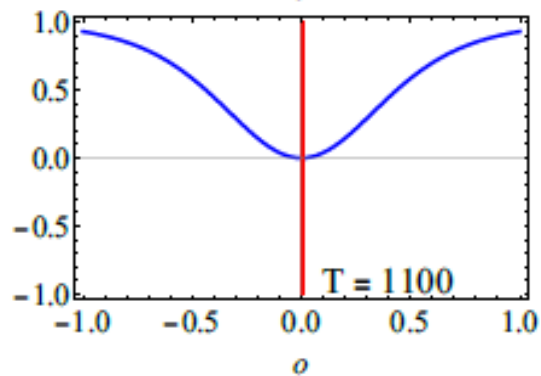
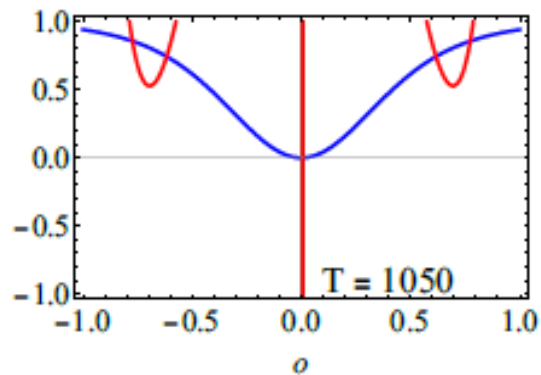
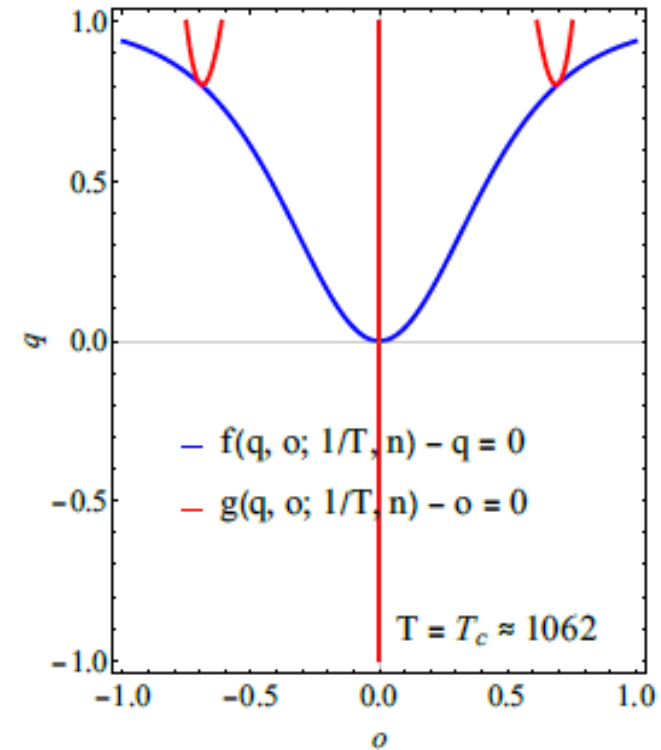
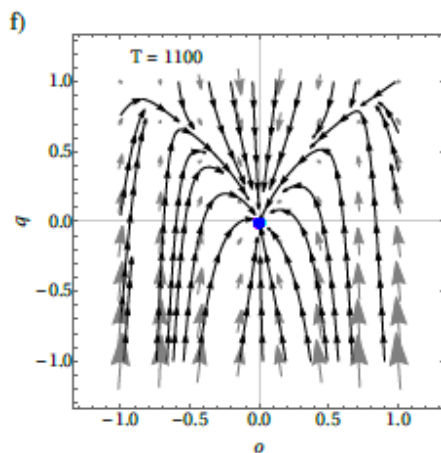
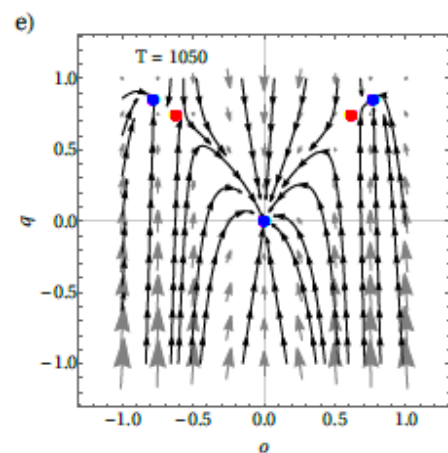
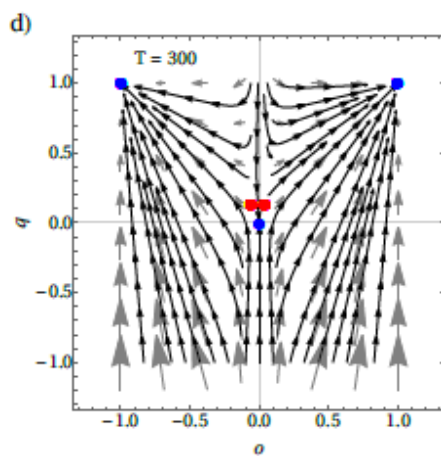
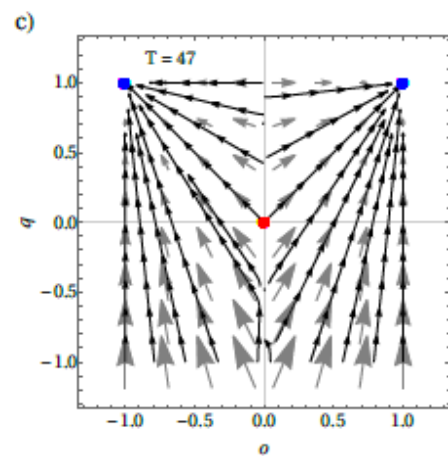
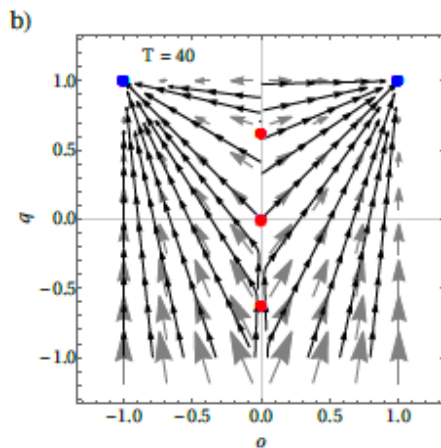
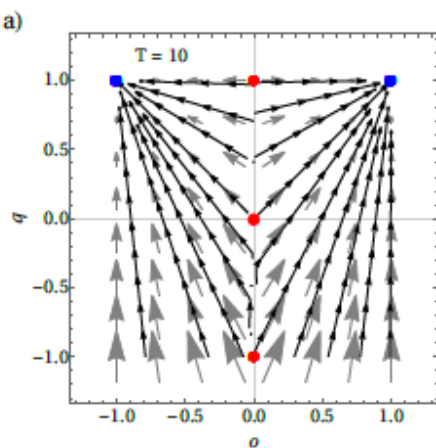
$$q \equiv \langle S_{ik} S_{kj} \rangle = \left[\frac{e^{-\beta(N-3)(-2q)} - 2e^{-2\beta \tanh(\beta(N-2)q)} + e^{-\beta(N-3)(2q)}}{e^{-\beta(N-3)(-2q)} + 2e^{-2\beta \tanh(\beta(N-2)q)} + e^{-\beta(N-3)(2q)}} \right] = f(q).$$



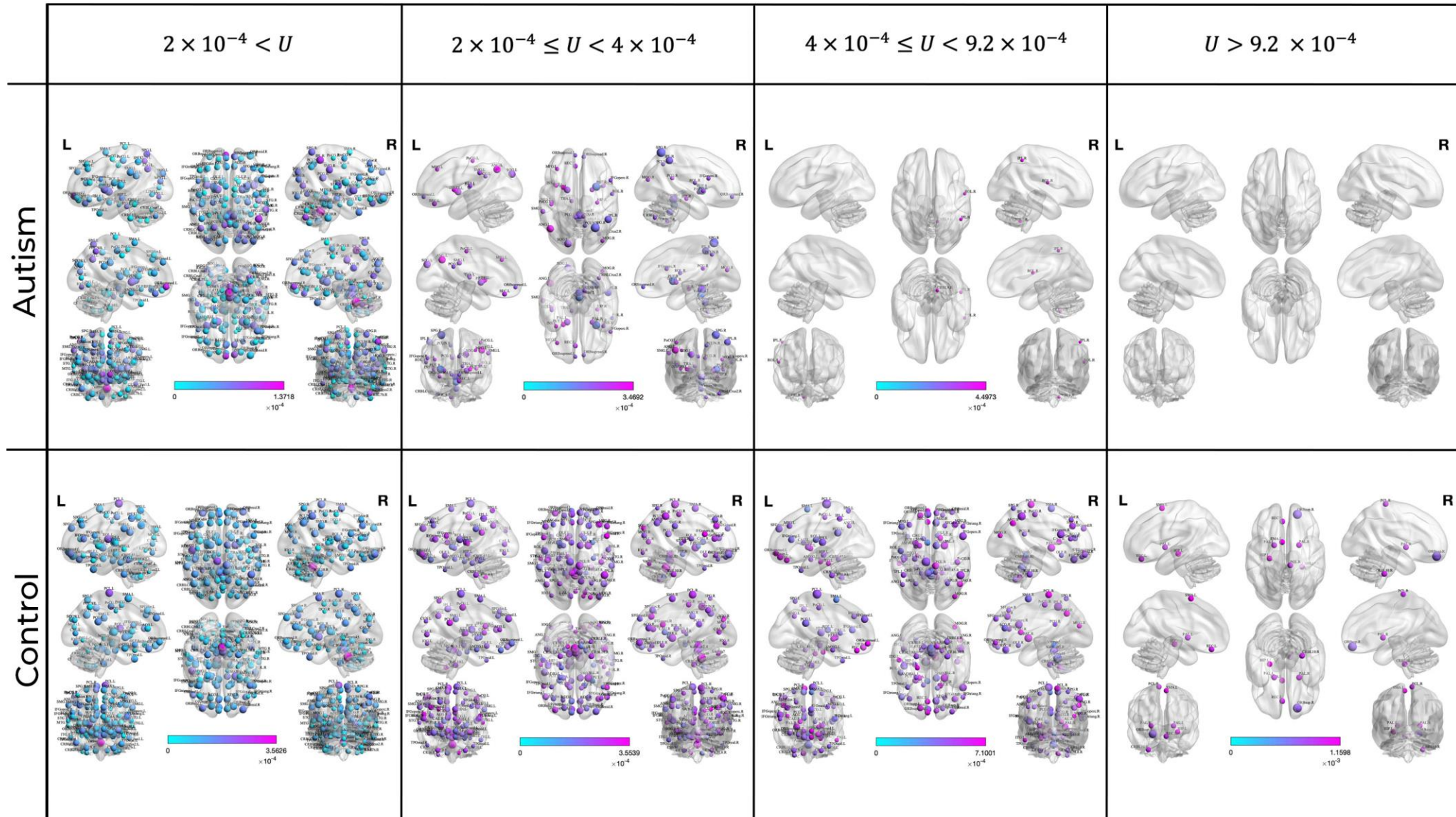
$$H = - \sum_{j < k < l} S_{jk} S_{kj} S_{jl} S_{lj}$$

$$q \equiv \langle S_{ik} S_{kj} \rangle$$

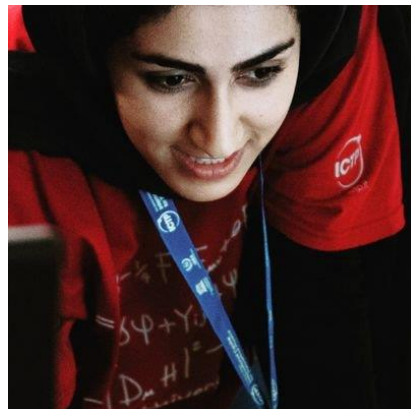
$$O = \langle S_{ij} S_{jk} S_{kl} \rangle$$



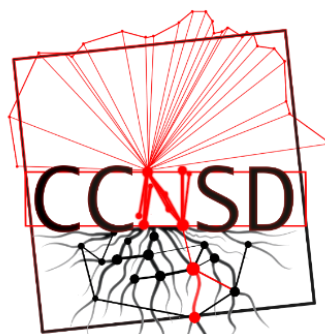
Session One



Thanks



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