



**FESC** 

## A Complex Network Approach for Phylogenetic Analysis

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## The Team

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- Biologists
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## Mathematical and Computer Modelling of Biosystems

- Nonlinear differential equations (population-based models)
  - population dynamics models (Malthus, Verhulst)
  - ecological models (Lotka-Volterra)
  - epidemics models (Kermack and Mc. Kendrick)
- Cellular automata (individual-based models) Von Neumann and Ulam
- Complex networks (based on graph theory) Barabasi-Albert, Watts-Strogatz models
- Data mining (genome project) Watson and Crick

## **Interdisciplinary Approach**

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- Biomathematics
- Biostatistics
- Physical Biology
- Computational Biology

System Biology

## Outline

- Introduction complex networks
- Fundamental concepts
  - Neighborhood matrices
  - Distance between networks
  - Identifying community structure
- Application in Biology
  - Protein similarity networks
  - The computational methodology
- Concluding remarks

## **Complex Networks**

- The regular and random networks are not suitable to describe some biological and social networks. In the last 20 years, complex networks, neither regular nor random, were proposed and applied to real systems.
- The concepts of Graph Theory and, more recently, Statistical Physics are used to study complex networks, comparing with regular and random networks.
- Identification of community structure is a hot topic in the context of complex networks. It is particularly relevant for biological networks.
- In this talk, we present and analyse a neighborhood matrix representation of network and its computational powerful to its characterization. We apply this concept to phylogentic analysis based on protein similarity networks. 6

## **Basic Concepts**

#### **Neighborhood Matrix**



8

## **Higher order neighborhood**

- Define neighborhood of order  $\ell$
- $\ell$  steps to walk from node to another in the  $\ell$  -neighborhood
- Boolean matrix  $M_{\ell}$  describes  $\ell$ -neighborhood
- $M_0 = I$ : each node is in *0*-neighborhood of itself
- *M*<sub>1</sub> = A
- $M_2 = [(M_0 \oplus M_1) \otimes M_1] (M_0 \oplus M_1) = [(I+A) \otimes A] (I \oplus A)$
- In general:

$$M_{\ell} = \left( \bigoplus_{g=0}^{\ell-1} M_g \right) \otimes M_1 - \left( \bigoplus_{g=0}^{\ell-1} M_g \right)$$

#### **Color Representation of Neighborhood Matrix**

+

+



$$\hat{M} = \sum_{\ell=1}^{D} \ell M_{\ell}$$

Andrade, Miranda, Petit Lobao, PRE 73, 046101 (2006).

## Using neighborhood matrix to characterize the network

- ✓ It carries out the same information of the adjacency matrix but this information is processed.
- $\checkmark$  Diameter  $\Rightarrow$  the maximum value of its elements
- ✓ Average shortest path  $\langle d \rangle \Rightarrow$  mean value of its elements
- ✓ Edge betweenness ⇒ a simple algorithm to calculate the sum of fractions of shortest paths between the nodes that pass through an edge
- Color code plots to visualize the neighborhood structure of network
- Andrade, Miranda, Pinho, Petit Lobao, Eur. Phys. J. B 61, 2470-256 (2008)

## **Distance between Networks**

 Define distance (or "numbering energy") E between networks P and Q :

$$E = \sum_{i,j} \left[ (\hat{M}_P)_{i,j} - (\hat{M}_Q)_{i,j} \right]^2$$

- Project P over Q, i.e., minimize E by Monte-Carlo procedure
- Changing randomly lines and columns of  $M_P$ , the minimization of **E** leads to  $\hat{M}_P$  with properties of  $\hat{M}_Q$

Andrade, Miranda, Pinho, Petit Lobao; Phys. Let. A 372, 5265-5269 (2008).

## **Deterministic networks**

Bethe lattice (Cayley tree)



Appollonian network





## **Projection based on the distance**







Cayley (C)



C→A

## **Modular structure of a network**

Newman-Girvan procedure – elimination of links with maximum edge betweenness using the modularity to indicate branching in dendrogram

✓ Distance between successive eliminations of links in NG procedure is more precise than modularity.

Andrade, Pinho, Petit Lobao, Int. J. Bif. Chaos App. Sci. Eng. 19, 2677-2685 (2009)



2: a) Dendrogram produced by the sequence

## Protein similarity networks

#### **Phylogenetic analysis using networks**

#### Methodology:

#### Database and network construction

- Select the protein sequences corresponding to the enzymes of a metabolic pathway of genomes of organisms. Select also the relevant information to set up the similarity level between the sequences (Genebank - NCBI – www.ncbi.nlm.nih.gov).
- ✓ The comparison between the sequences is performed by the program BLAST . As a result a similarity matrix N x N is set up, associated with each enzyme of the metabolic pathway, with N as the number of protein sequences.
- ✓ Generate 101 networks associated with the similarity threshold ( $\sigma$ ) from 0 to 100 : the nodes are the protein sequences and there is a link between nodes if the similarity is greater or equal to  $\sigma$ .

## Metabolic pathway: chitin

# There are 1695 protein sequences corresponding to 13 enzymes

Enzyme	Enzymatic classification	Domain (#)
UDP-acetylglucosamine pyrophosphorylase	2.7.7.23	E(2), B(324), A(2)
Acetylglucosamine phosphate deacetylase	3.5.1.25	B(170), A(6)
Glucosaminephosphate isomerase	2.6.1.16	E(23), B(285), A(5)
Hexosaminidase	3.2.1.52	E(3), B(235)
Phosphoglucoisomerase	5.3.1.9	E(16), B(472), A(12)

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Displa	ay Summary	Show	20 🗘	Sort By	\$ Sen	d to 🛛 🗘	)	
AII:	176 Bacteria: 174 RefSeq: 0	Related Structures	: 171 🛠					
Item	ns 1 - 20 of 176					Page	1	of 9 Next
 1.	N-acetylglucosamine-6-phos 382 aa protein ACA31257.1 GI:168825886	phate deacetylase	[Haemopł	nilus somnus :	2336]			
 2.	N-acetylglucosamine-6-phos 387 aa protein ACA15057.1 GI:168193110	phate deacetylase	[Methylob	acterium sp.	<u>4-46]</u>	(17	Acet 6 no	yl des)
□ 3.	N-acetylglucosamine-6-phos 385 aa protein ACA11796.1 GI:167964786	phate deacetylase	[Xylella fa	stidiosa M12]	1			
□ 4.	<u>N-acetylglucosamine-6-phos</u> 383 aa protein	phate deacetylase	[Haemoph	<u>nilus parasuis</u>	29755]			

#### **Characterizing the networks**

- To set up the neighborhood matrix from the adjacency matrix.
- To calculate the size of the largest cluster for different values of  $\sigma$ .
- For low values of σ, there is a large cluster; for high values of σ, there are many sub-networks. For intermediate values of σ, the modular structure is revealed.





## Size of the largest cluster (UDP e Acetyl)



Goes-Neto, Diniz, Santos, Pinho, Miranda, Andrade, Petit Lobao, Borges, El-Hani. Biosystems 101: 59-66 (2010).

## **Critical network**

 Distance between the networks (based on the neighborhood matrices) before and after the removing of the links: the maximum value corresponds to the critical network (acetyl: 42%).



#### **Edge Betweenness**

- Edge Betweenness: sum of fractions of shortest paths connecting the pairs of nodes through a certain edge.
- Assuming the NG procedure (Newman and Girvan, PRE 69, 026113 (2004):
  - Remove the edge with the maximal value of edge betweenness.
  - Repete the process until there is no link.
- In order to set up the modular structure, we set up the dendrogram for the critical network as well as the color representation of the neighborhood matrix.

## Dendrograms (acetyl)



Vértices Arestas removidas Dendrogram (30%)

#### **Identifying the modular structure**





#### **Modular network of Acetyl**



Com 1.dat: 7 Actinobacteria; 4 Crenarchaeota; Com 2.dat: 14 Betaproteobacteria; 2 Gammaproteobacteria; 2 Deinococcus-Thermus; 2 Alphaproteobacteria; Com 3.dat: 8 Firmicutes; Com 4.dat: 7 Actinobacteria: Com\_5.dat: 12 Alphaproteobacteria; Com 6.dat: 27 Gammaproteobacteria; 2 Alphaproteobacteria; Com 7.dat: 23 Gammaproteobacteria; Com 8.dat: 12 Cyanobacteria; Com 9.dat: 16 Firmicutes; 6 Gammaproteobacteria; Com 10.dat: 6 Gammaproteobacteria; Com 11.dat: 9 Firmicutes; Outros.dat: 6 Actinobacteria; 4 Planctomycetes; 2 Crenarchaeota; 2 Bacteroidetes; 2 Acidobacteria; 1 Betaproteobacteria;

## **Summary of Results**

Enzyme	S <sub>th</sub> <sup>(crit)</sup>	# nodes	# communities
UDP-acetylglucosamine pyrophosphorylase	51	327	6
Acetylglucosamine phosphate deacetylase	42	176	11
Glucosaminephosphate isomerase	40	313	4
Hexosaminidase	37	238	9
Phosphoglucoisomerase	37	501	5

## **Concluding Remarks**

✓ The interdisciplinary character of this research project reveals the importance of getting together the knowledge of biologists with physicists, mathematicians and computer scientists. However the team was concerned about the computational features of it.

✓The neighborhhod matrix present the processed information about the network. It leads to the concept of distance between networks.

✓ The distance is more efficient in revealing the critical network in which the modular structure of the network, if it is the case, is revealed.

✓The above concepts are applied in protein similarity proteins revealing the classification of organisms that present these protein sequences.

✓The methodology may be applied to other networks that presents a modular character.

## Thank you!

#### Congruence

Let p and q the critical networks associated with two enzymes;  $N_{pq}$  the number of common organisms; the number of matching organisms  $M_{pq'}$  i.e., number of organisms that are placed in the same communities in the two networks. If the number of communities in networks p and q are different, it is necessary to make a correspondence of two or more communities of network p to the same community in network q. The value  $G_{pq}$  is just the ratio  $M_{pq}/N_{pq}$ .

Ex: UDP (Norg=245) e Acetyl (	Norg=88) : N <sub>nn</sub> =44	; M <sub>nn</sub> =40; G <sub>nn</sub> = 91%
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UDP
(327)

Acetyl (176)

		u1	u2	u3	u4	u5	u6
	a1	0	0	0	0	2	0
	a2	0	0	3	1	0	0
	a3	0	0	0	0	0	0
	a4	0	0	0	0	0	0
	a5	0	0	0	3	0	0
L	a6	0	0	8	1	0	0
•	a7	0	0	8	0	0	0
	a8	2	0	0	0	0	0
	a9	0	10	2	0	0	0
	a10	0	0	0	0	0	0
	a11	0	4	0	0	0	0

nterseção	44	Não-congruentes	4
		u3a9	2
Associação	40	u4a2	1
u1a8	2	u4a6	1
u2a9a11	14		
u3a2a6a7	19		
u4a5	3	Congruência	40/44
u5a1	2	Congruência	90,9%

#### References

[1] Andrade, R. F. S.; Miranda, J. G. V.; Lobão, T. P.; *Neighborhood concepts in complex networks.* Phys. Rev. E **73**, 046101 (2006).

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