Who's this guy?

Knowledge Discovery in Databases through Complex Networks: application to phylodynamics

Knowledge Discovery in Databases through Complex Networks: application to phylodynamics

Luiz Max F. de Carvalho Scientific Computing Programme (PROCC), Fiocruz Pan American Center for Foot-and-Mouth Disease (PAHO/WHO)

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Outline

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- **2** Complex Networks
- 3 Example 1: Chitin pathway phylogeny
- 4 Example 2: Foot-and-mouth disease virus in South America
- WaFiS 2012
- Knowledge Discovery i Databases

Knowledge Discovery in Databases (KDD)

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Knowledge Discovery in Databases (KDD) Lots of data

- human brain very limited processing capacity
- Information \rightarrow Knowledge
- Increasing number of molecular data (sequences, 3D structures, antigenicity,...)
- Is it possible to explore these databases to discover useful stuff?

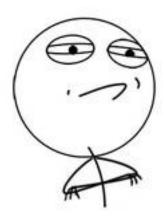
Well...Let's see

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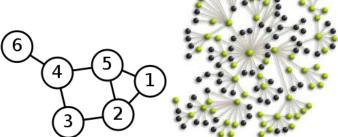
Knowledge Discovery in Databases (KDD)



We may use Complex Networks

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• Graphs $\rightarrow G = (V, E)$



Yeah. but how?

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We can explore the "dynamic signature" of these Complex Networks, i.e., study and compare their structural properties. Some useful formulas:

- Clustering Coefficient < c >: $\frac{3 \times \# triangles}{\# triples}$
- Degree distribution $P_K = \sum_{K'-K}^{\infty} p_{K'}$
- Diameter: max(d(i, j))

Ok, Let's work then

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Knowledge Discovery i Databases (KDD) Grab n sequences;

- 2 Create an $n \times n$ matrix using some kind of (normalized) distance (say, S);
- **3** For each $\sigma \in [0,1]$ build $M(\sigma)$ such that:

$$m_{ij}(\sigma) = \begin{cases} 1 & \text{if } Sij > \sigma, \\ 0 & \text{if } S_{ij} < \sigma. \end{cases}$$

In a sense, we are transforming a single network in a family of networks.

Analysis

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Knowledge Discovery in Databases (KDD) We shall explore the relationships between these networks: First, define a higher-order neighborhood indicator function, such that you binarize the adjacency matrix with regard the path length ℓ , obtaining a matrix $\hat{M} = \sum_{\ell=1}^D \ell M(\ell)$. Then

$$\delta(\alpha,\beta) = \frac{1}{N^2} \sum_{i=1}^{N} \sum_{j=1}^{N} \left(\frac{\hat{m}_{ij}(\alpha)}{D(\alpha)} - \frac{\hat{m}_{ij}(\beta)}{D(\beta)} \right) \tag{1}$$

Evaluating $\delta(\sigma, \sigma + \Delta\sigma)$ can give some interesting insights.

Example 1: Chitin pathway phylogeny

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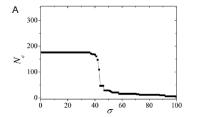
Detecting Network Communities: An Application to Phylogenetic Analysis

Roberto F. S. Andrade¹, Ivan C. Rocha-Neto², Leonardo B. L. Santos^{1,3}, Charles N. de Santana⁴, Marcelo V. C. Diniz⁵, Thierry Petit Lobão², Aristóteles Goés-Neto⁵, Suani T. R. Pinho¹, Charbel N. El-Hani⁶⁸

1 Institute of Physics, Federal University of Bahia, Campus Universitário de Ondina, Salvador, Bahia, Brazil, 2 Institute of Mathematics, Federal University of Bahia, Campus Universitàrio de Ondina, Salvador, Bahia, Brazil, 3 National Institute for Space Research, São José dos Campos, São Paulo, Brazil, 4 Mediterranean Institute of Advanced Studies, IMEDEA (CSIC-UIB), Espodes (Islas Baleared), Spain, 5 Department of Biological Sciences, State University of Feira de Santana, Feira de Santana, Bahia, Brazil, 6 Institute of Biology, Federal University of Bahia, Campus Universitário de Ondina, Salvador, Bahia, Brazil

- Proteins related to the chitin metabolic pathway from 1605 complete genomes;
- BLAST distances (which are asymmetric);
- Search for phylogenetic relationships

Example 1: Some results



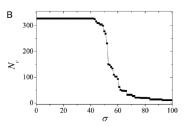


Figure 1. The size of the largest connected component (N) versus the threshold similarity σ : a) Acetyl; b) UDP. doi:10.1371/journal.pcbi.1001131.g001

Example 1: Some more results

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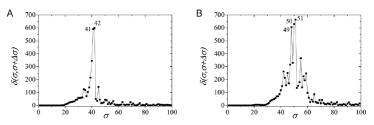


Figure 2. The distance $\delta(\sigma,\sigma+M\sigma)$ between networks for successive similarities at the maximal value, with $M\sigma=1$, in the case of: a) Acetyl at $\sigma=\sigma_{max}=42\%$; b) UDP at $\sigma=\sigma_{mb}=51\%$. doi:10.1371/journal.pcb.it.001131.q002

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Knowledge Discovery Databases

Example 1: The expected Network(s)

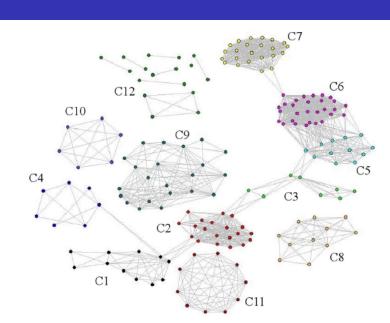
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Knowledge Discovery in Databases

Complex



Example 2: Foot-and-mouth disease virus in South America

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Knowledge Discovery in Databases (KDD)

PHYLODYNAMICS OF FOOT-AND-MOUTH DISEASE VIRUS: A COMPLEX NETWORK APPROACH

Luiz Max Fagundes de Carvalho^{1,3}, <u>Leonardo Bacelar Lima Santos</u>², Pedro Jeovah Pereira³, Waldemir de Castro Silveira³

¹Sector of Infectious Diseases Epidemiology – Federal University of Rio de Janeiro, Rio de Janeiro, Brazil, luizepidemiologia@gmail.com

²National Institute os Space Research – INPE, São José dos Campos – SP, Brazil, santoslbl@gmail.com

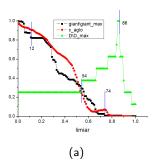
"National institute os Space Research = INPE, sau Jose dos Campos - SF, Frazil, santosio @ginal.com

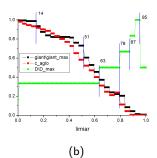
Laboratory of Bioresources - Pan American Foot-and-mouth Disease Center (PANAFTOSA) - Pan American Health Organization

(PAHO), Duque de Caxias - RJ, Brazil, silveiraw @paho.ore, poercira@paho.ore

- S was built with phylogenetic (TN93) distances for NT and JTT distances for AA;
- Try to make sense of a somewhat big data set (167 seqs);
- Extract some nice patterns;

Indexes $\times \sigma$





A nice network

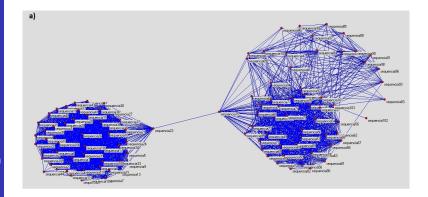
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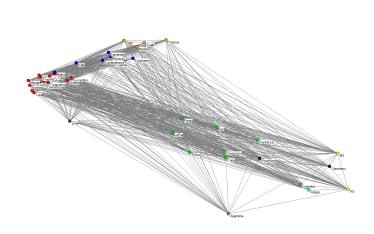
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Knowledge Discovery in Databases (KDD)



Some more developments



Related Work

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Knowledge Discovery in Databases (KDD) Identify transmission clusters (HIV, HCV) (Lewis et al, 2008,Plos Medicine)

 Explore scale-free behavior in phylodynamics (Shiino, 2012, Frontiers in Microbiology)

Future Directions

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Knowledge Discovery in Databases (KDD) Explore the spatial aspect in the construction of SMaybe $\hat{S} = \mu + S(G)^{\alpha}$

- Power law analysis
- Implement assortativity
- Suggestions. . .

Thank You!



