

Redes Biológicas y Biología de Sistemas

21-29 Abril 2008

Coordinación: Florencio Pazos (CNB-CSIC)

Redes Metabólicas

Redes Biológicas

Protein-protein physical interaction networks. Here nodes represent proteins, and links represent direct physical contacts between proteins. In addition to direct interaction, two proteins can interact indirectly through other proteins when they belong to the same complex.

Protein-protein genetic interaction networks. In general, two genes are said to interact genetically if a mutation in one gene either suppresses or enhances the phenotype of a mutation in its partner gene.

Expression networks. Large-scale microarray experiments probing mRNA expression levels yield vast quantities of data useful for constructing expression networks. In an expression network, genes that are coexpressed are considered connected.

Regulatory networks. Protein-DNA interactions are an important and common class of interactions. Most DNA-binding proteins are transcription factors that regulate the expression of target genes.

Combinatorial use of transcription factors further complicates simple interactions of target genes for a given transcription factor. A regulatory network consists of transcription factors and their targets with a specific directionality to the connection between a transcription factor and its target.

Metabolic networks. These networks describe the biochemical reactions within different metabolic pathways in the cell. Nodes represent metabolic substrates and products, and links represent metabolic reactions.

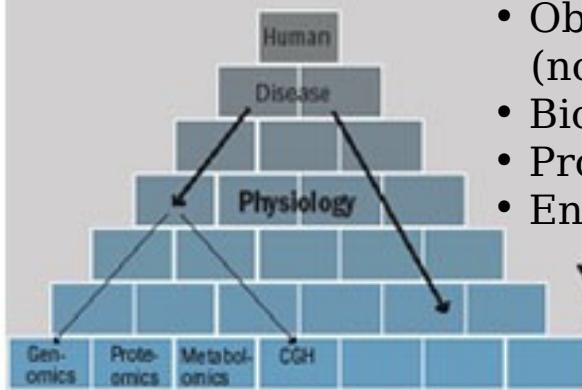
Signaling networks. These networks represent signal transduction pathways through protein-protein and protein-small molecule interactions. Nodes represent proteins or small molecules, and links represent signal transduction events.

Other... Phosforilation netowrk, ...

Bioinformática y Sistemas Complejos en Biología

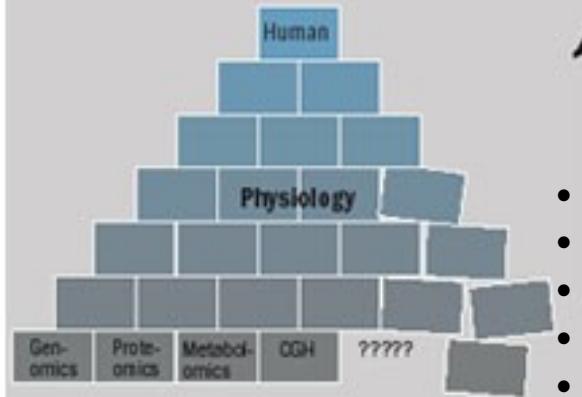
Biología de Sistemas vs. Biología Molecular

Top-Down Approach

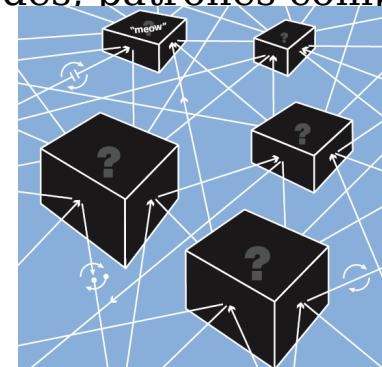


- Visión desde el punto de vista de sistemas complejos
- Objeto de estudio: **redes, relaciones, propiedades emergentes** (no propiedades de componentes individuales).
- Biología de Sistemas
- Propiedades “globales” (emergentes, etc.) -> conocimiento biológico
- Enfermedades (dianas/marcadores = redes, patrones complejos)

Bottom-Up Approach



- Visión reduccionista
- Objeto de estudio: **componentes (genes proteínas)**.
- Biología Molecular
- **Propiedades de los componentes** -> conocimiento biológico
- Enfermedades (dianas/marcadores = proteínas/genes).



“genomics” / “post-genomics”

Multi-level high-throughput characterization of components

Full-genome sequencing (“genome”).

Characterization of transcripts (mRNA) (“transcriptome”)

Characterization of the protein repertory (“proteome”)

Cellular localization of the components (“localizome”)

Genetic regulation networks (“regulome”)

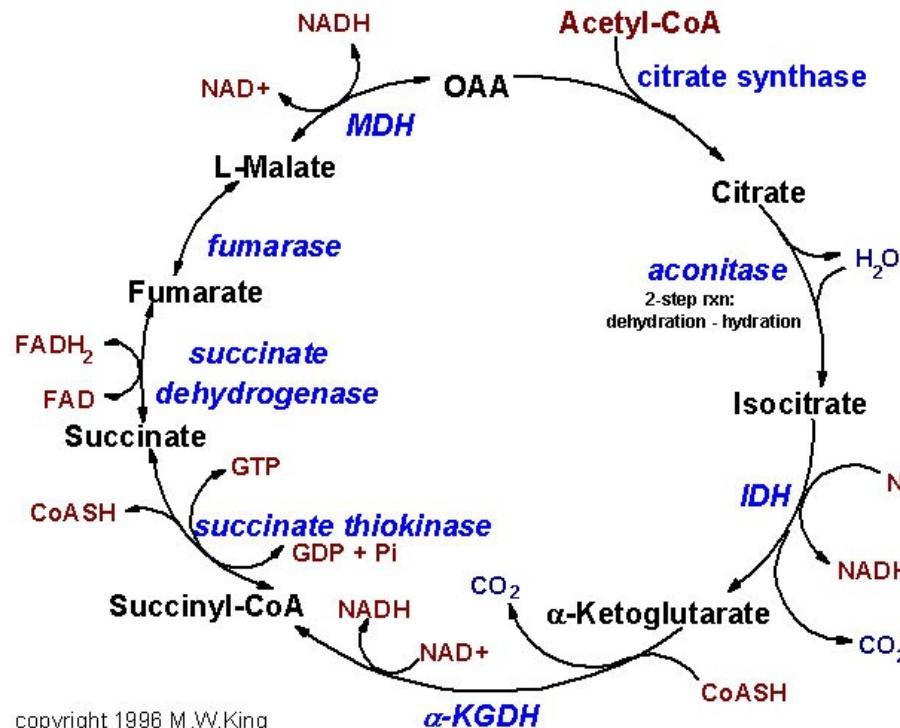
Protein interaction networks (“interactome”)

High throughput characterization of gene-phenotype relationships (“phenome”)

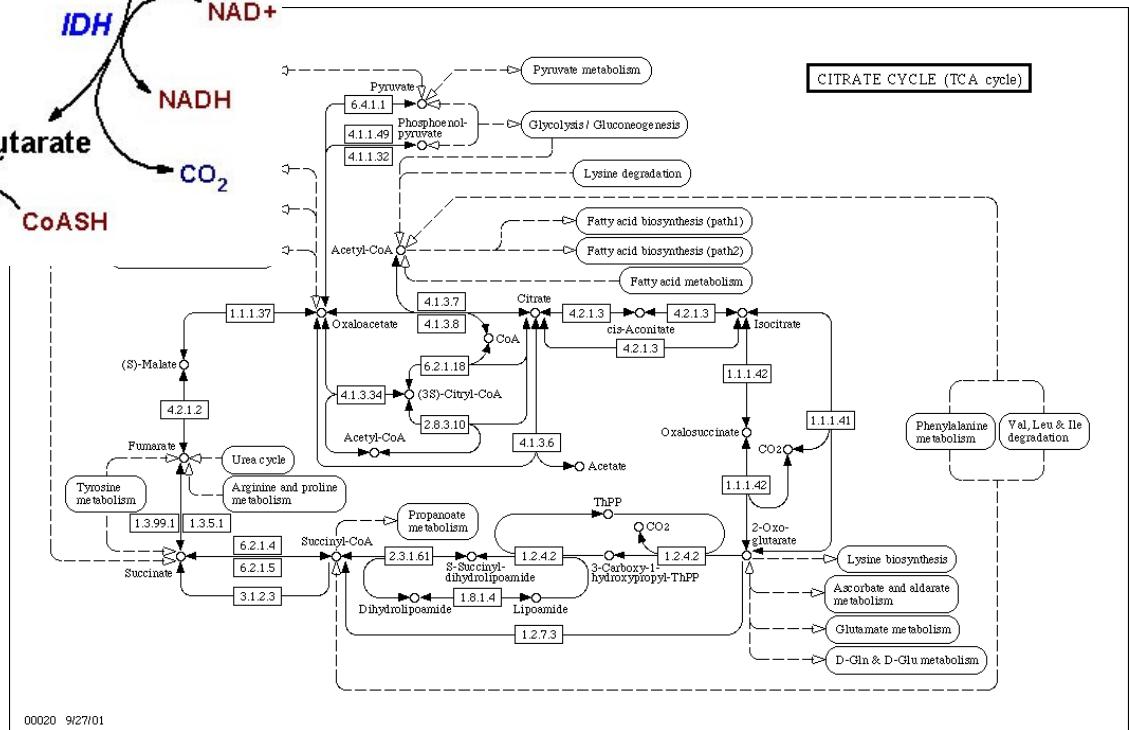
Metabolic networks (“metabolome”)

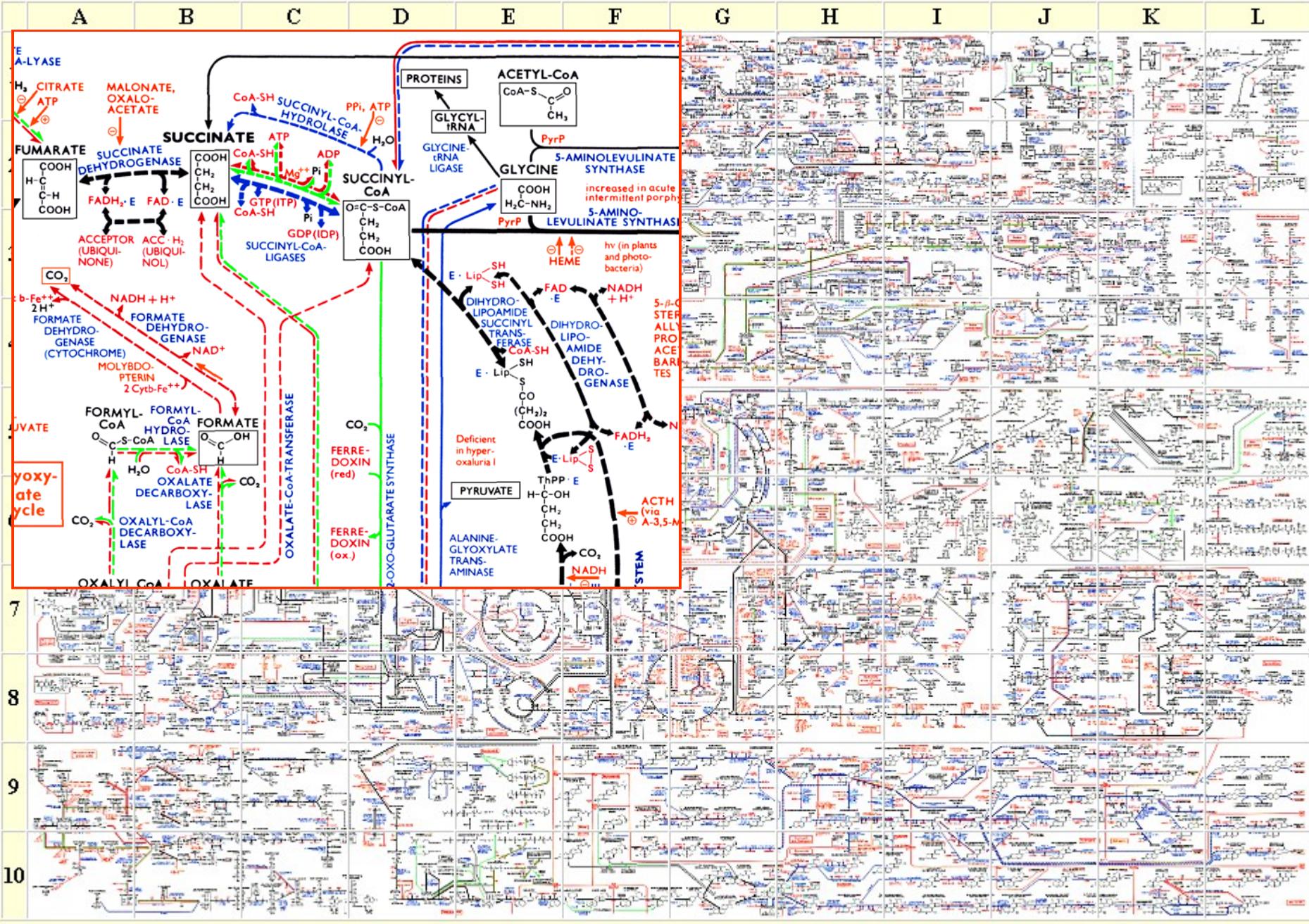
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Representación del Metaboloma como Red

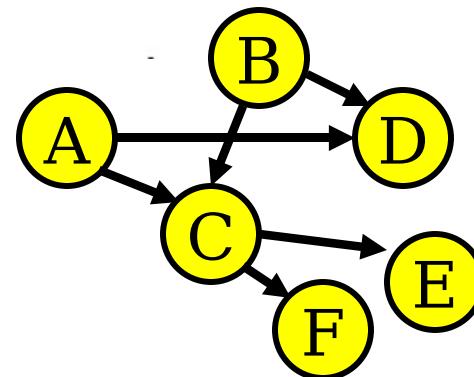
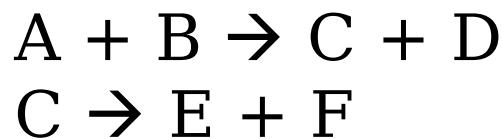
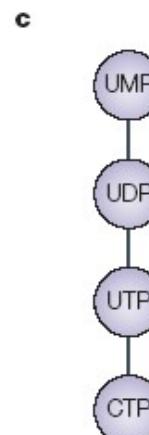
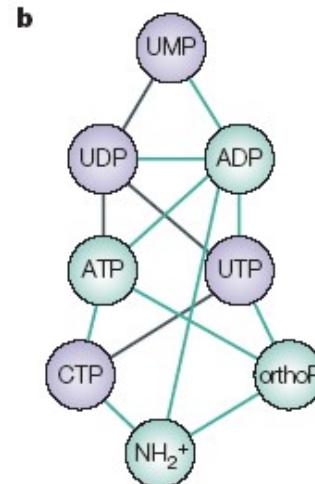
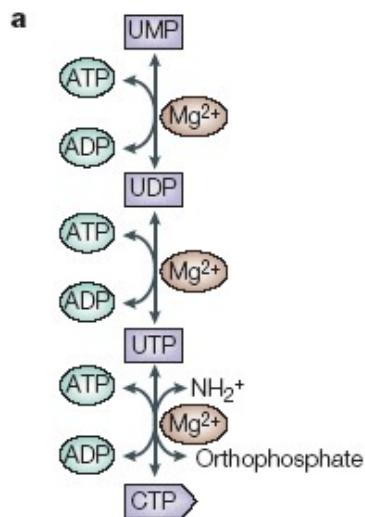


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Conversión de Redes Metabólicas a Grafos



$\forall A, B, C, \dots \neq H_2O, \text{ions, cofactors, or Nr} > X$

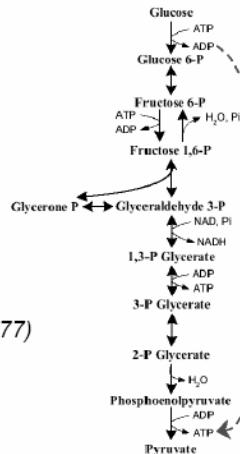
Conversión de Redes Metabólicas a Grafos

Exclusión de metabolitos “no informativos”

1) Definición a priori (metabolitos “moneda”)

However, many authors found it necessary to remove the most highly connected metabolites, because they tend to distort the statistics.

(Ma & Zeng 2003,
Bioinformatics 17:270-277)



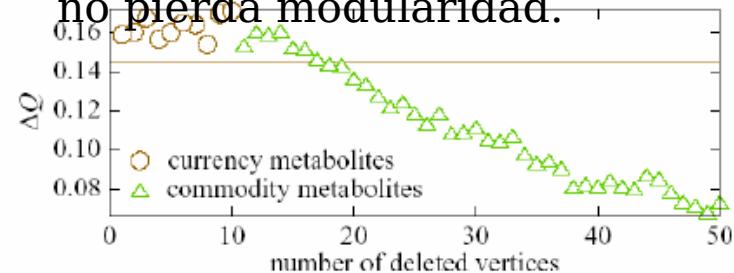
These metabolites are often called **currency metabolites**, since they act as a kind of molecular currency in the cell (for example, ATP). Some investigators call them “current metabolites”, emphasizing their flow through the network. We call the currency metabolites, and by analogy to economics, we call the other metabolites **commodity metabolites**.

Wagner & Fell 2001	Schuster et al. 2002	Ma & Zeng 2003
ATP	ATP	ATP
ADP	ADP	ADP
NADP	NADP	NADP
NADPH	NADPH	
NAD		NAD
NADH		
P _i	P _i	
H ₂ O	H ₂ O	
H ⁺		
PP _i		
CMP		
	CO ₂	
	O ₂	
	NH ₃	

2) Por criterios de conectividad (k>10, 15, ...)

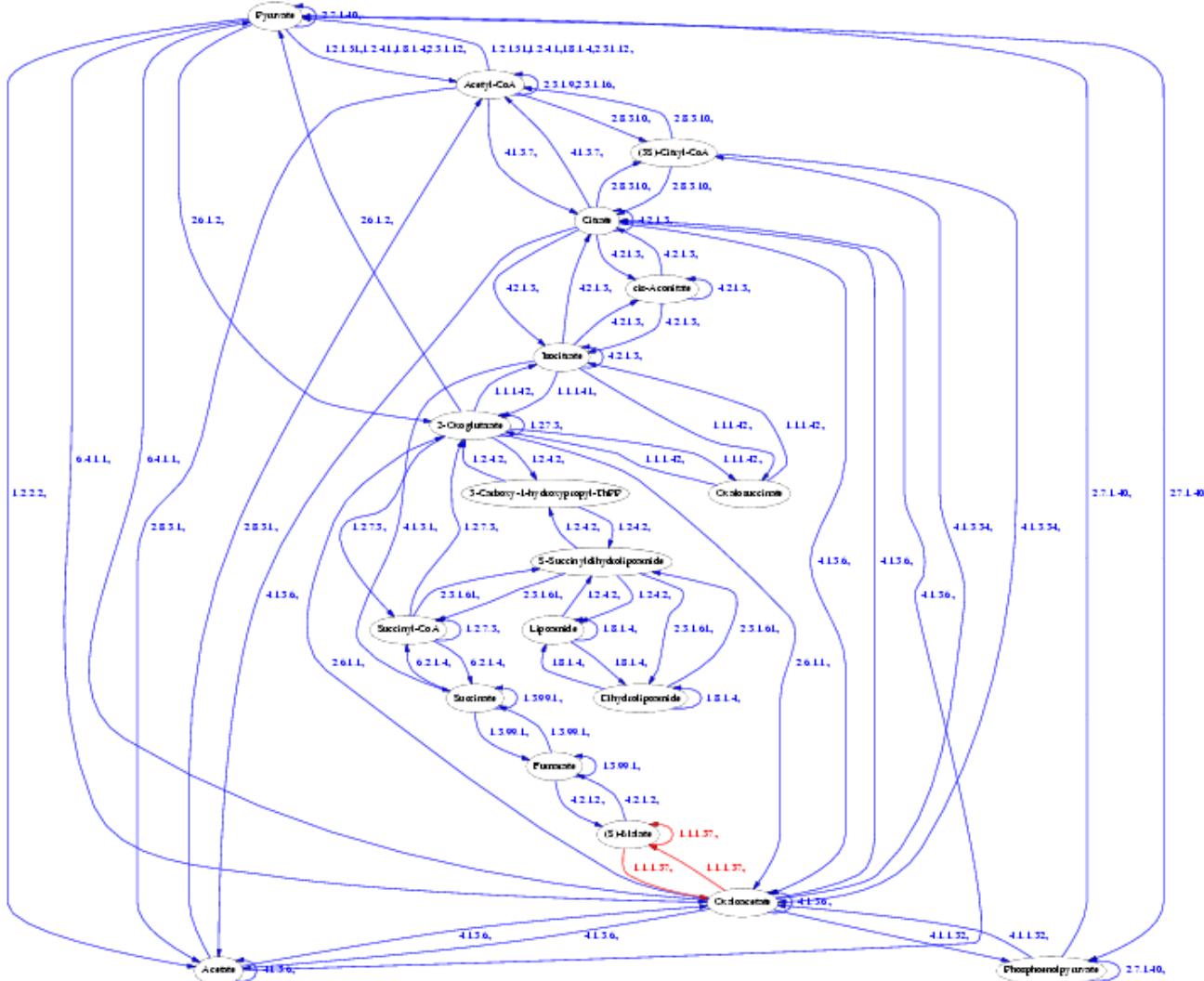
3) Otros criterios topológicos.

Ej. Metabolitos que conectan muchas veces diferentes módulos: Ir quitando metabolitos muy conectados mientras la red no pierda modularidad.



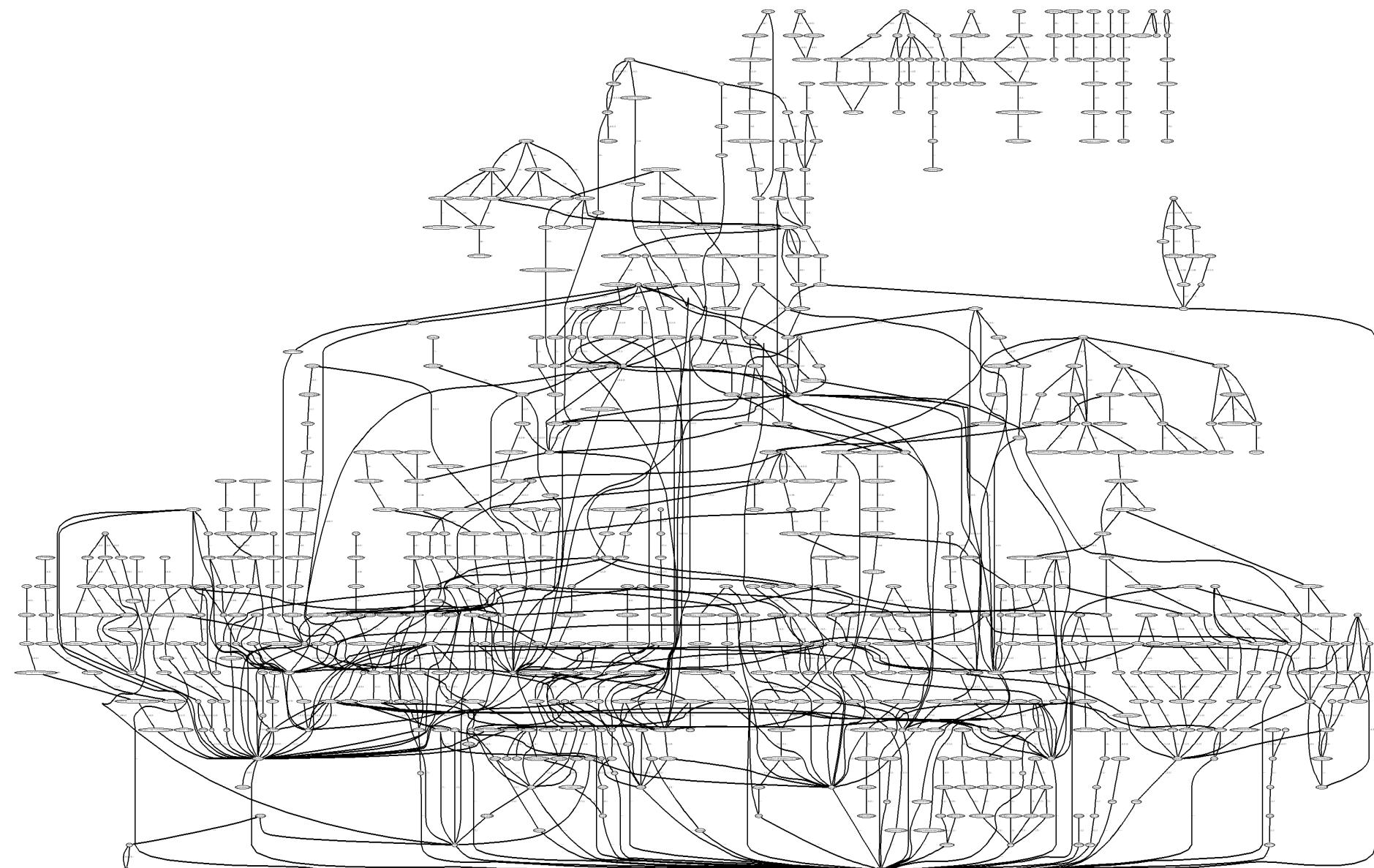
Wagner & Fell 2001	Schuster et al. 2002	Ma & Zeng 2003
ATP	ATP	ATP
ADP	ADP	ADP
NADP	NADP	NADP
NADPH	NADPH	
NAD		NAD
NADH		
P _i	P _i	
H ₂ O	H ₂ O	
H ⁺		
PP _i		
CMP		
	CO ₂	
	O ₂	
	NH ₃	

Representación del Metaboloma como Red



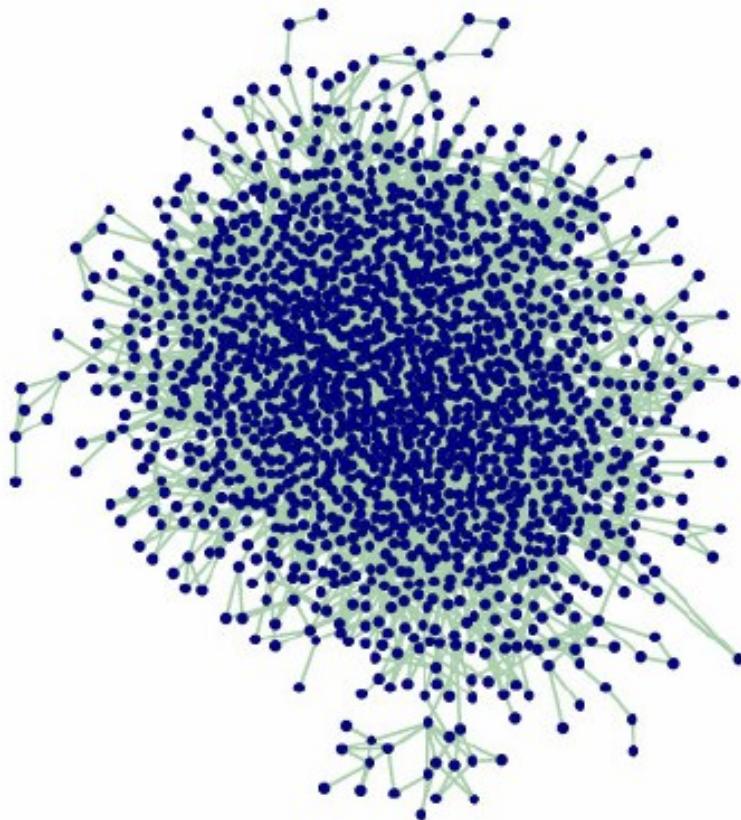
R. Chaleil (IC, London)

Representación del Metaboloma como Red

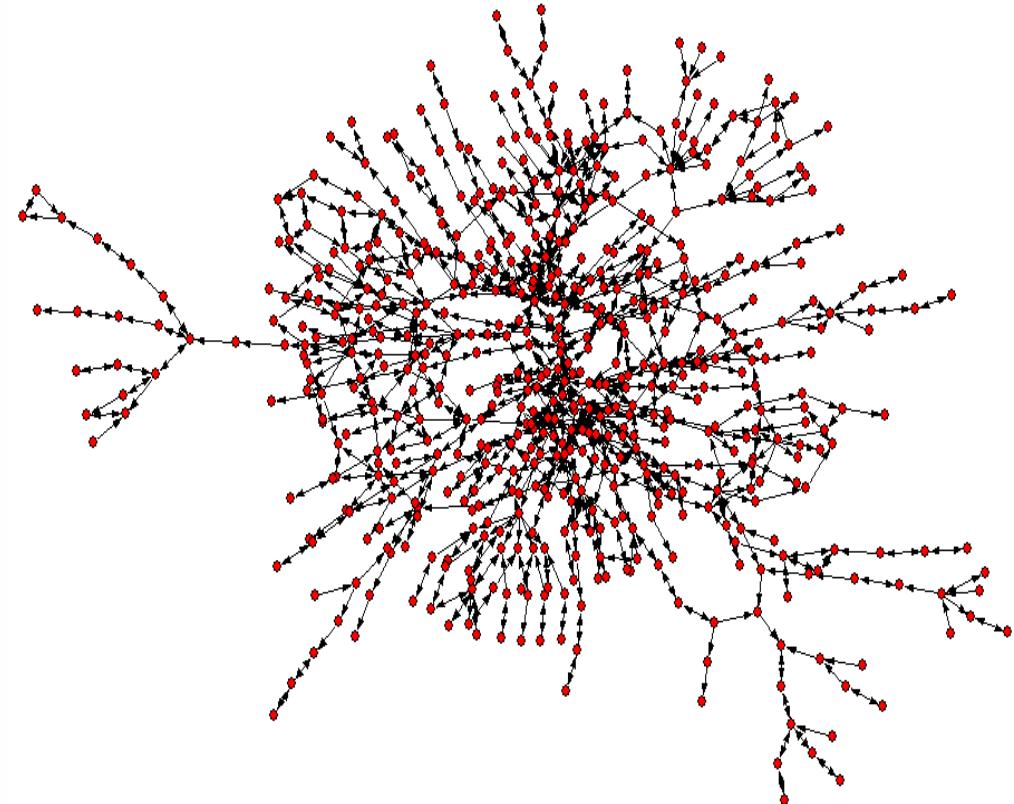


Florencio Pazos, Victor De Lorenzo & Alfonso Valencia. (2003). The organization of the Microbial Biodegradation Network from

Representación del Metaboloma como Red



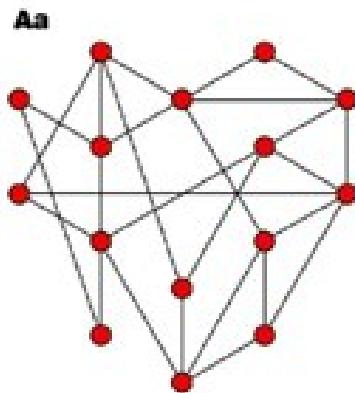
Yeast metabolic network



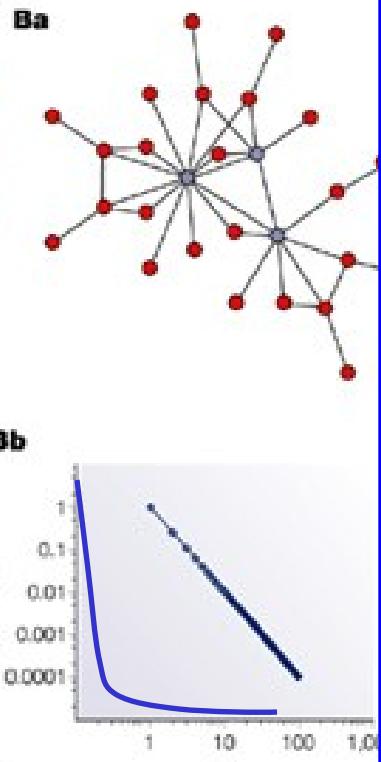
E. coli metabolic network

Topological Characteristics of the Metabolic Network

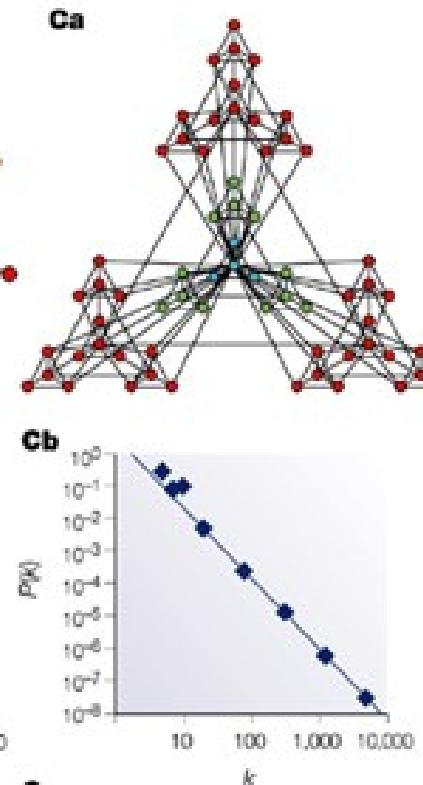
A Random network



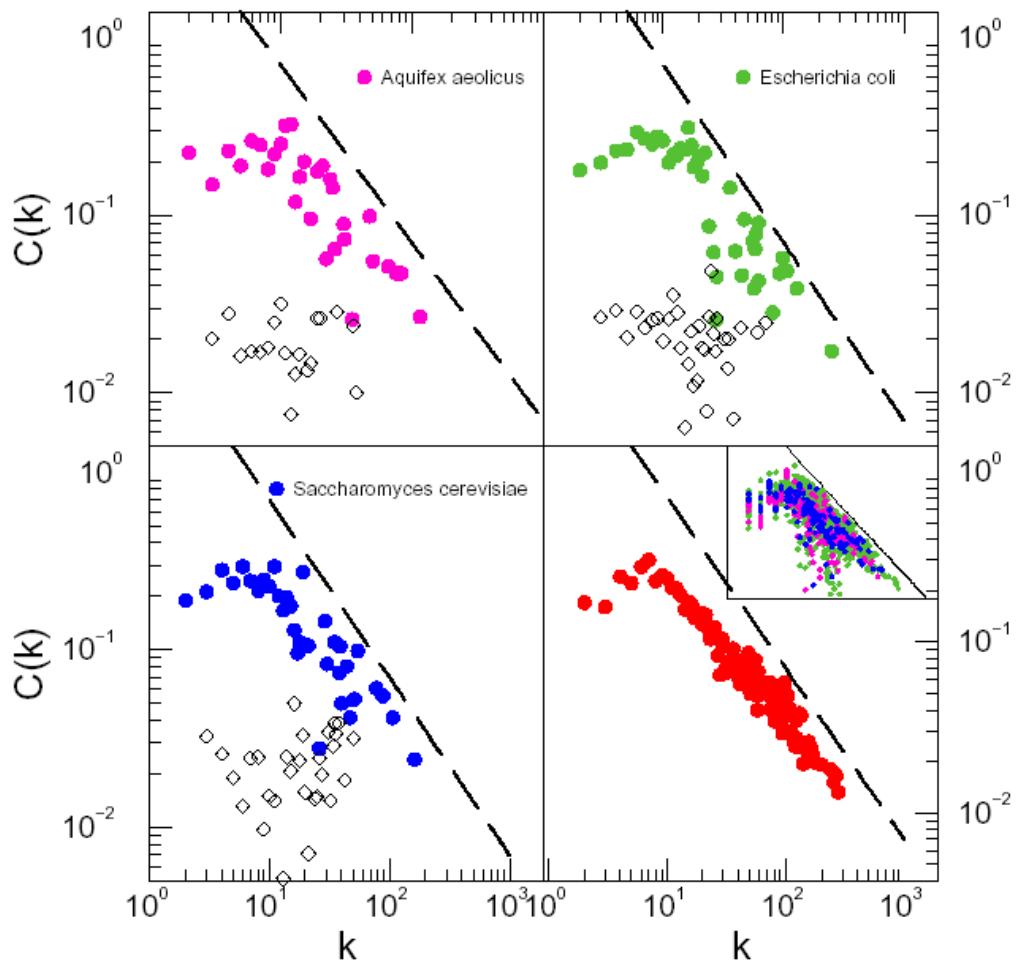
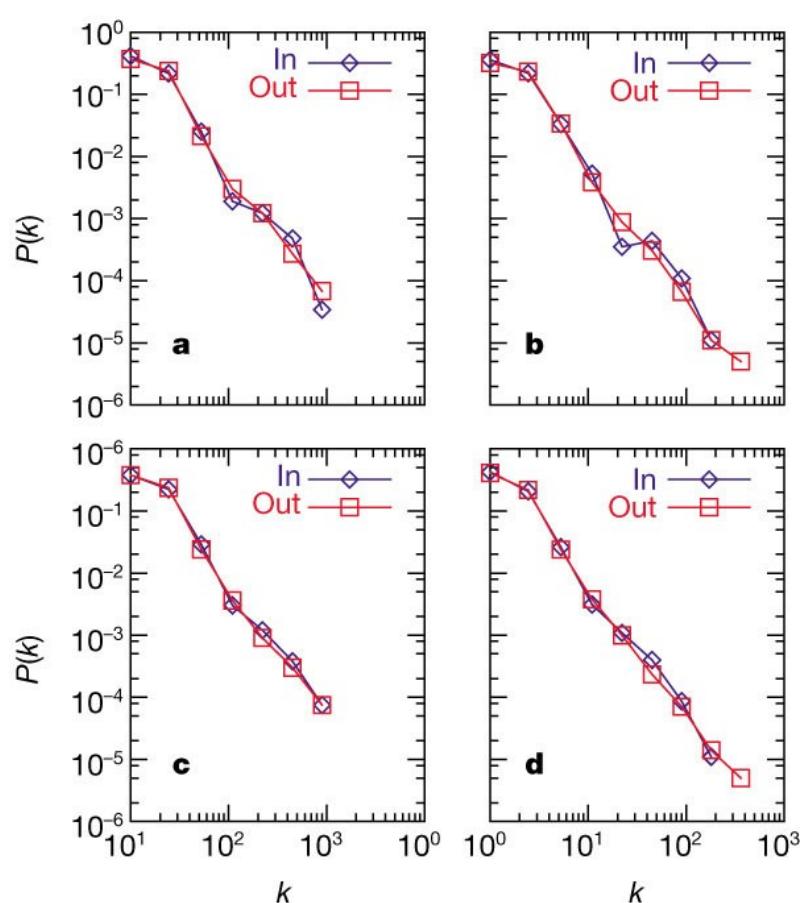
B Scale-free network



C Hierarchical network



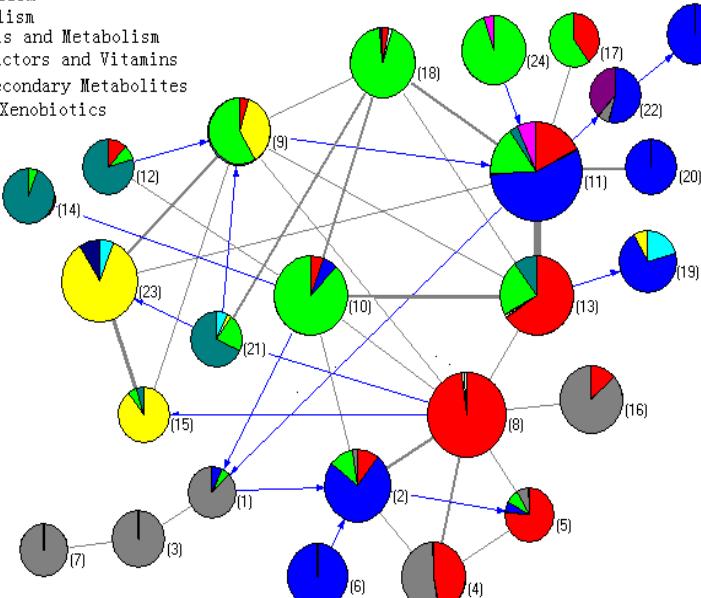
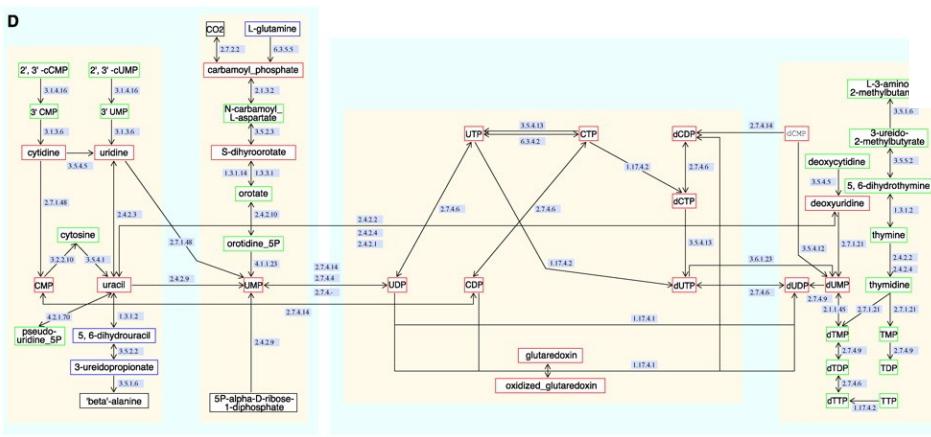
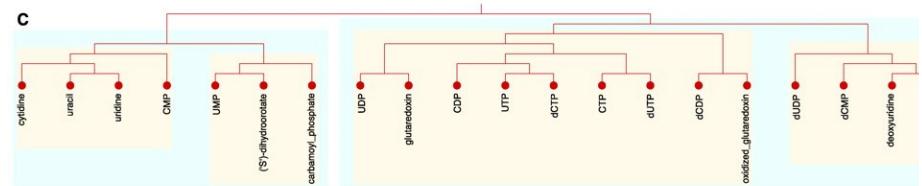
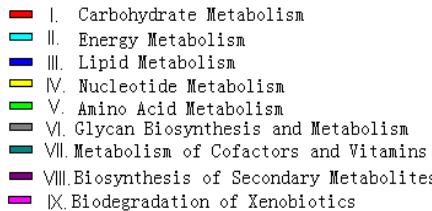
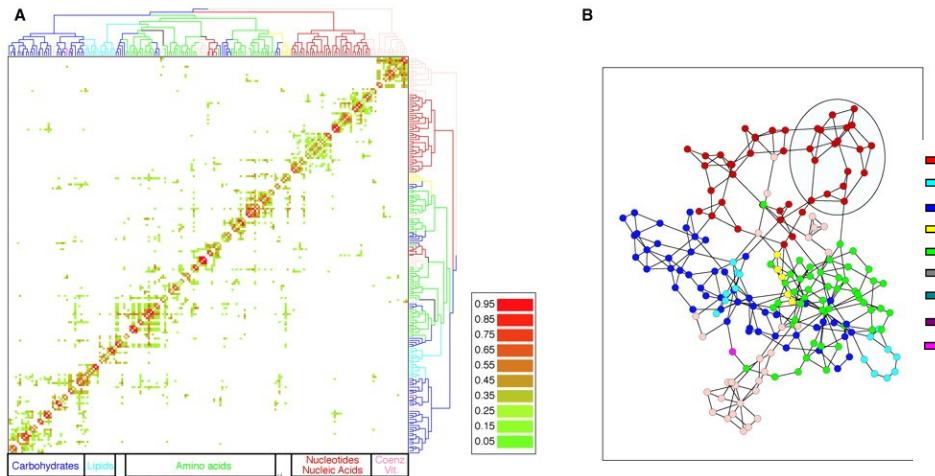
Topological Characteristics of the Metabolic Network



Jeong, H., Tombor, B., Albert, R., Oltvai, Z. N. & Barabasi, A. L. (2000). The large scale organisation of metabolic networks. *Nature* **407**, 651-653

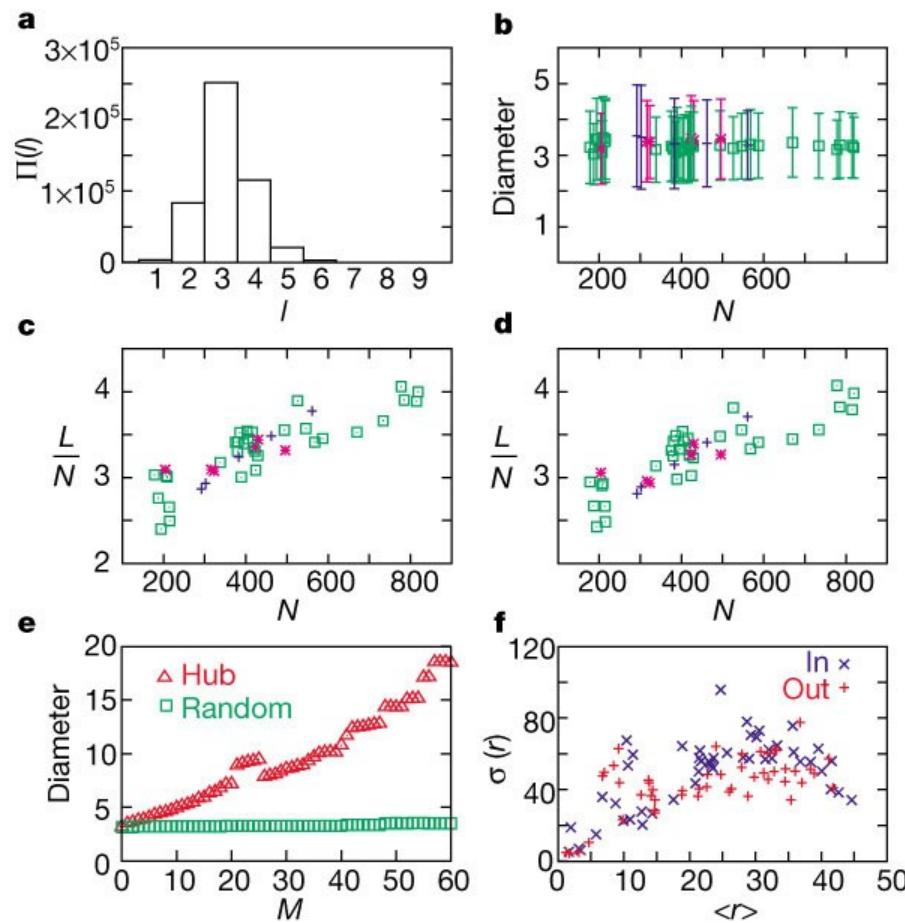
Barabasi, A.L. and Oltvai, Z.N. (2004) Network biology: understanding the cell's functional organization. *Nat Rev Genet*, **5**, 101-113.

Topological Characteristics of the Metabolic Network

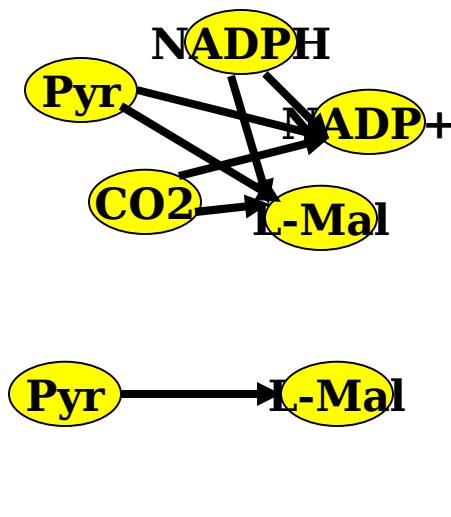
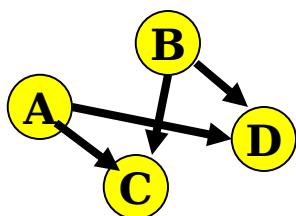


Ravasz, E., Somera, L., Mongru, D.A., Oltvai, Z.N. and Barabási, A.L. (2002) Hierarchical organization of modularity in metabolic networks. *Science*, **297**, 1551-1555.

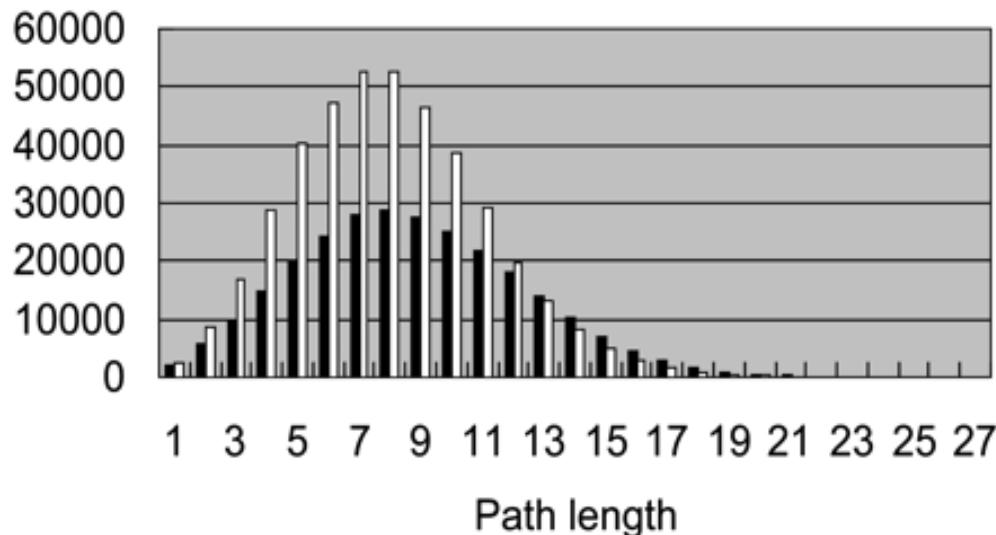
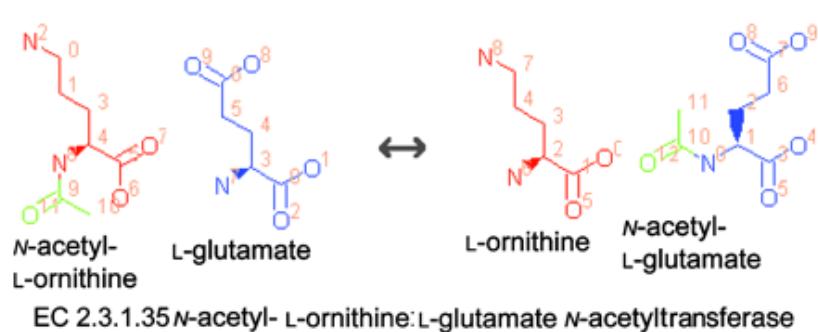
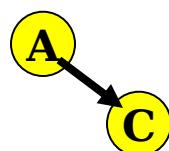
Topological Characteristics of the Metabolic Network



Small world?



B D

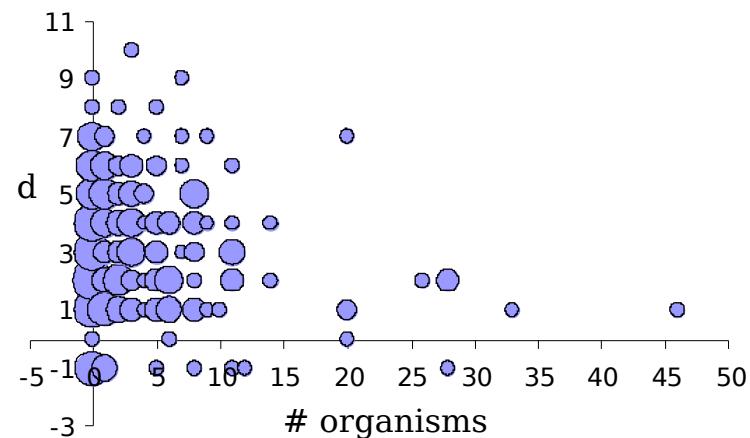
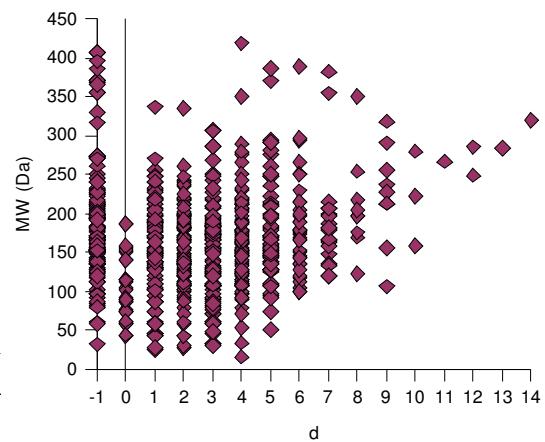
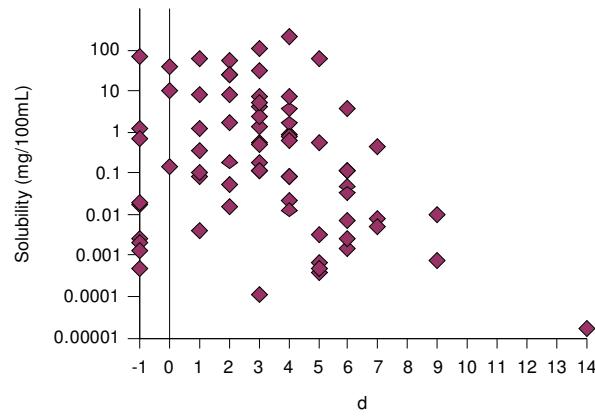
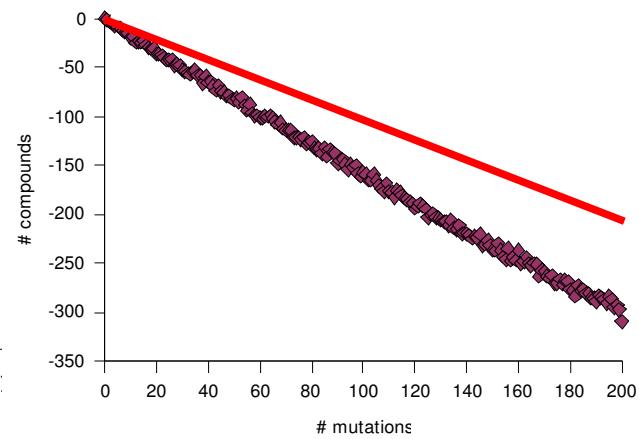
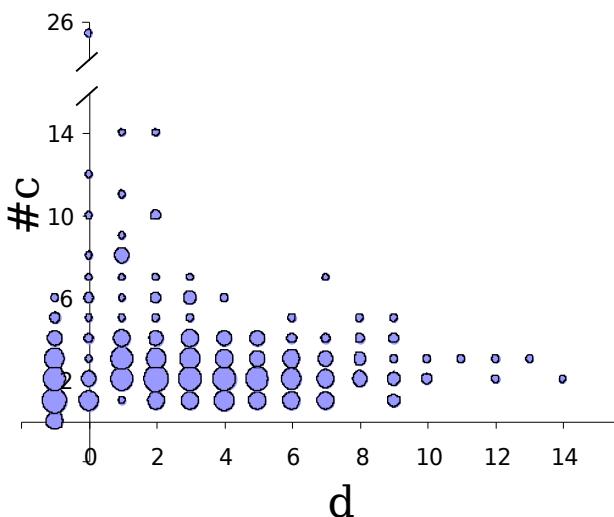
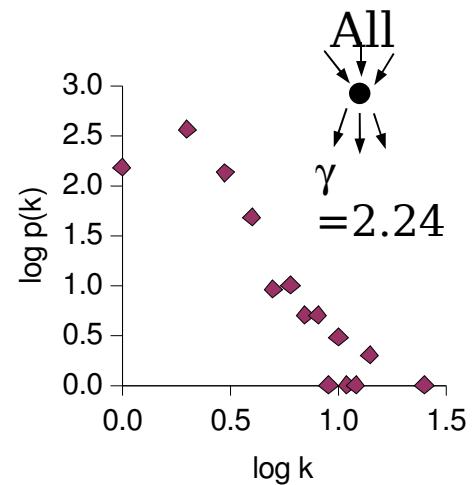


The Biodegradation Network



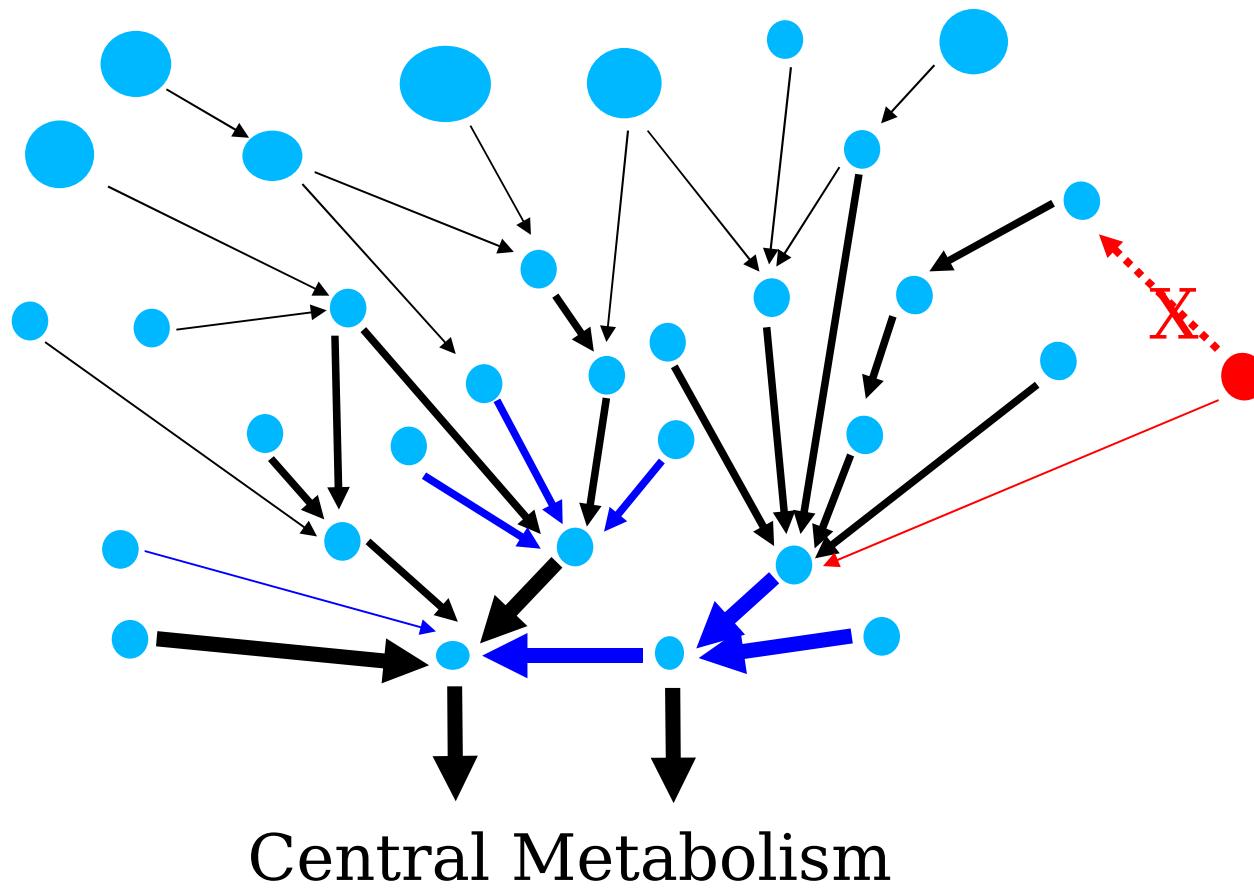
- Florencio Pazos, Victor De Lorenzo & Alfonso Valencia. (2003). The organization of the Microbial Biodegradation Network from a Systems-Biology perspective. *EMBO Rep.* **4(10)**:994-999.
http://pdg.cnb.uam.es/biodeg_net

Properties of the Biodegradation Network



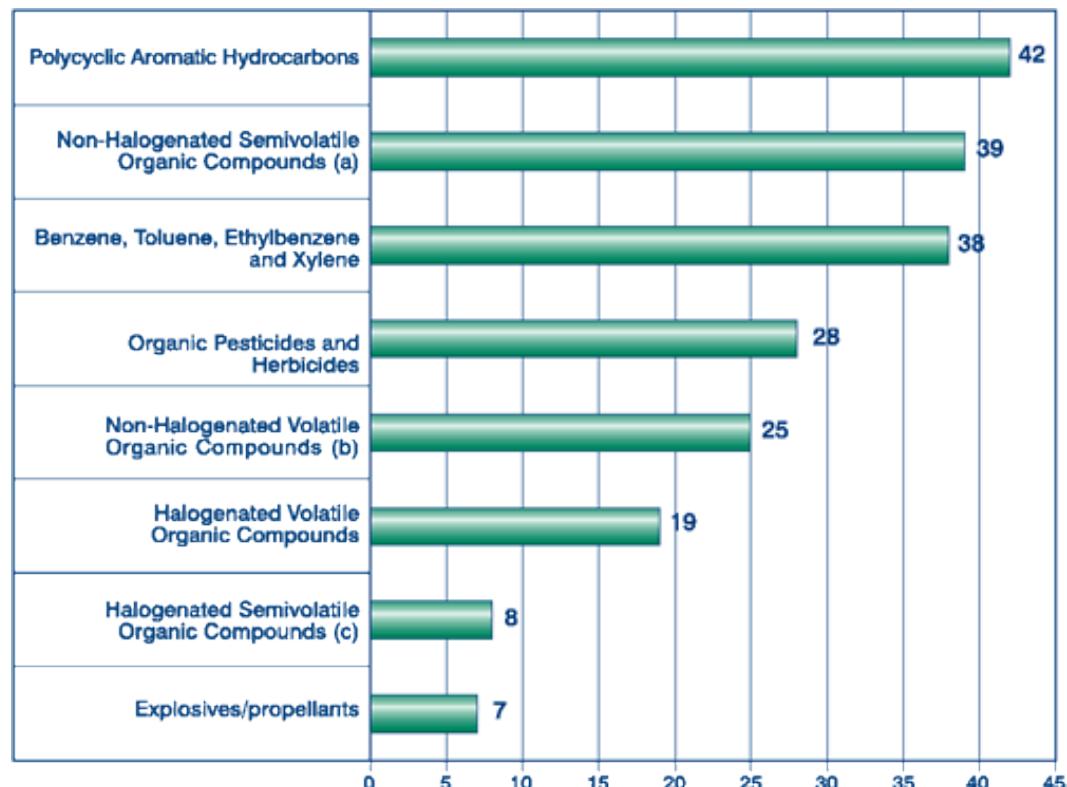
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Properties of the Biodegradation Network



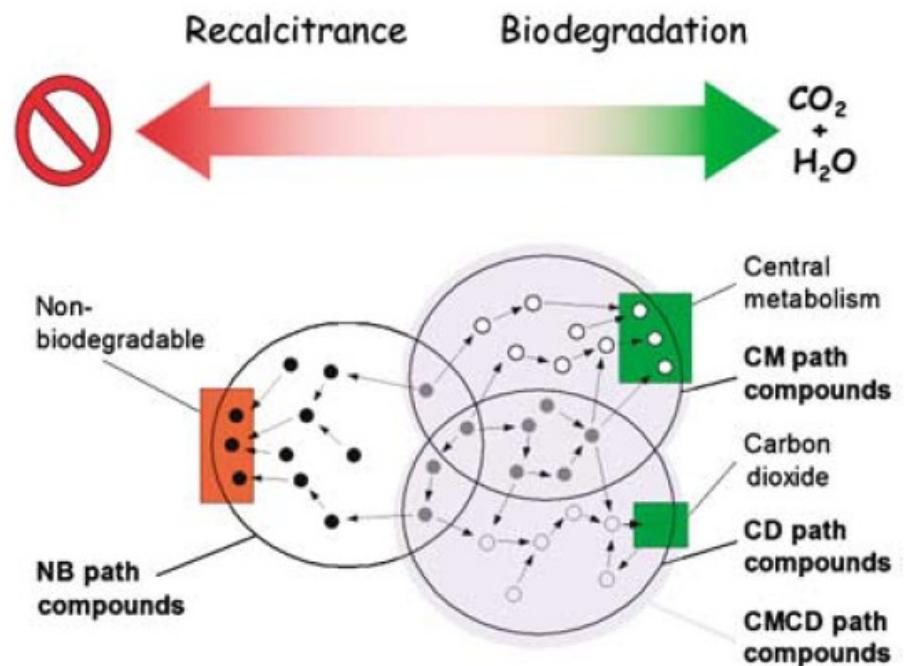
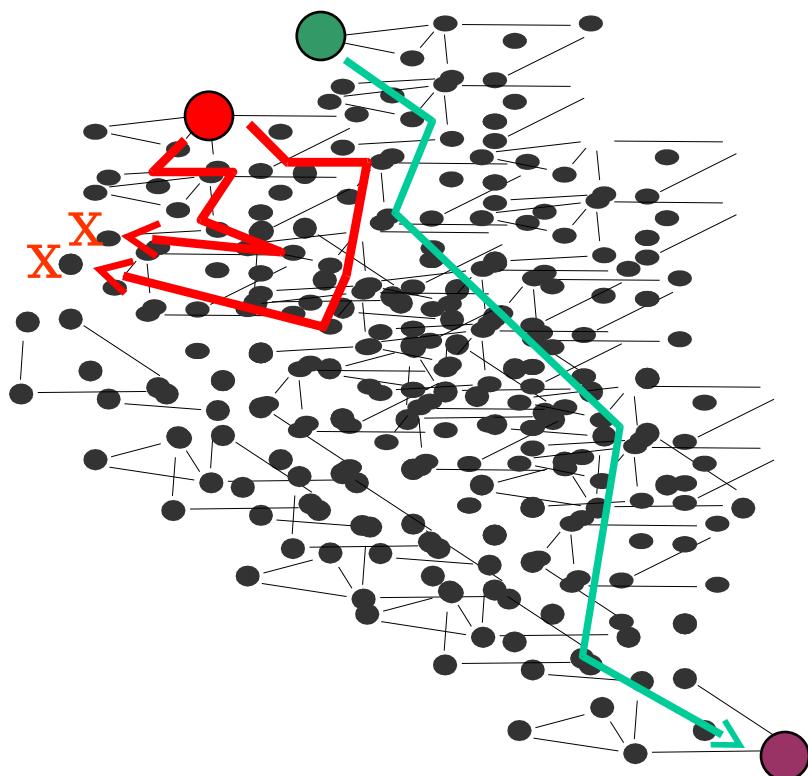
- Florencio Pazos, Victor De Lorenzo & Alfonso Valencia. (2003). The organization of the Microbial Biodegradation Network from a Systems-Biology perspective. *EMBO Rep.* **4(10)**:994-999.

Biodegradación/Bioremediaci n



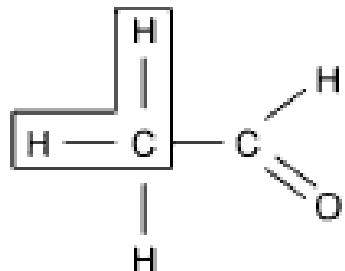
Obtaining examples of biodegradable/recalcitrant compounds

umption: biodegradable compounds are the ones which have a pathway to CM.
Otherwise they are **recalcitrant**.



Description of the chemical structure

Chemical structure



SMILES

CC=O

Atomic triads

C-C-H	4
C-C=O	1
H-C-H	3
H-C=O	1

MW

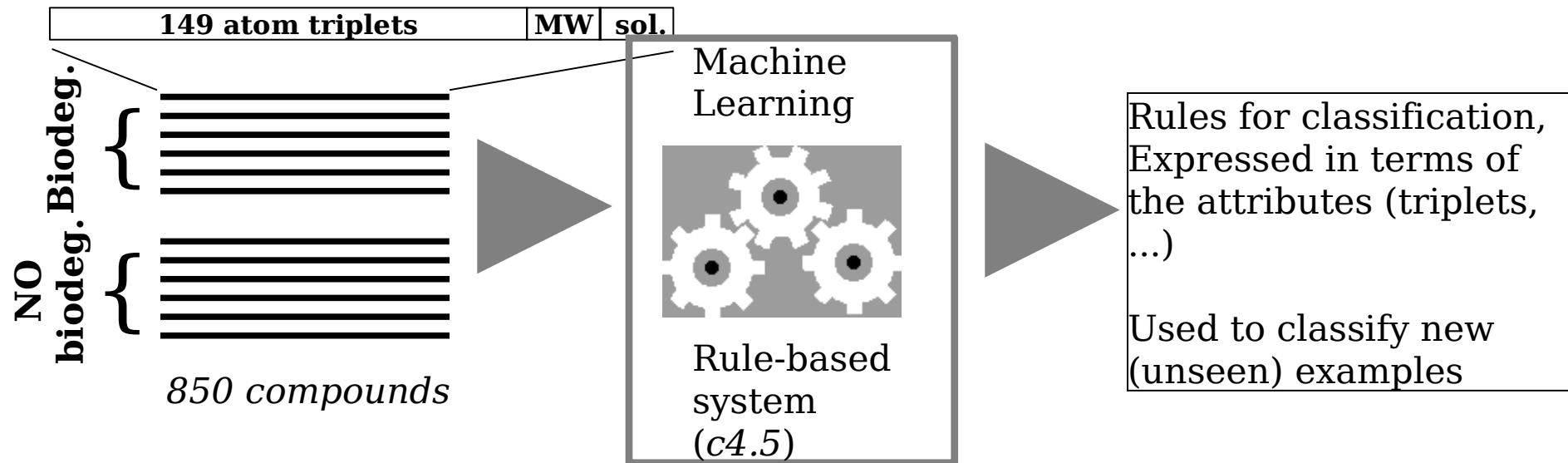
44 Da

Solubility

0.1 g/100 ml

Compound vector = (0.1, 44, 4, 1, 3, 1, 0, 0, ..) Vector len= 152 (149 triads + MW + ws)

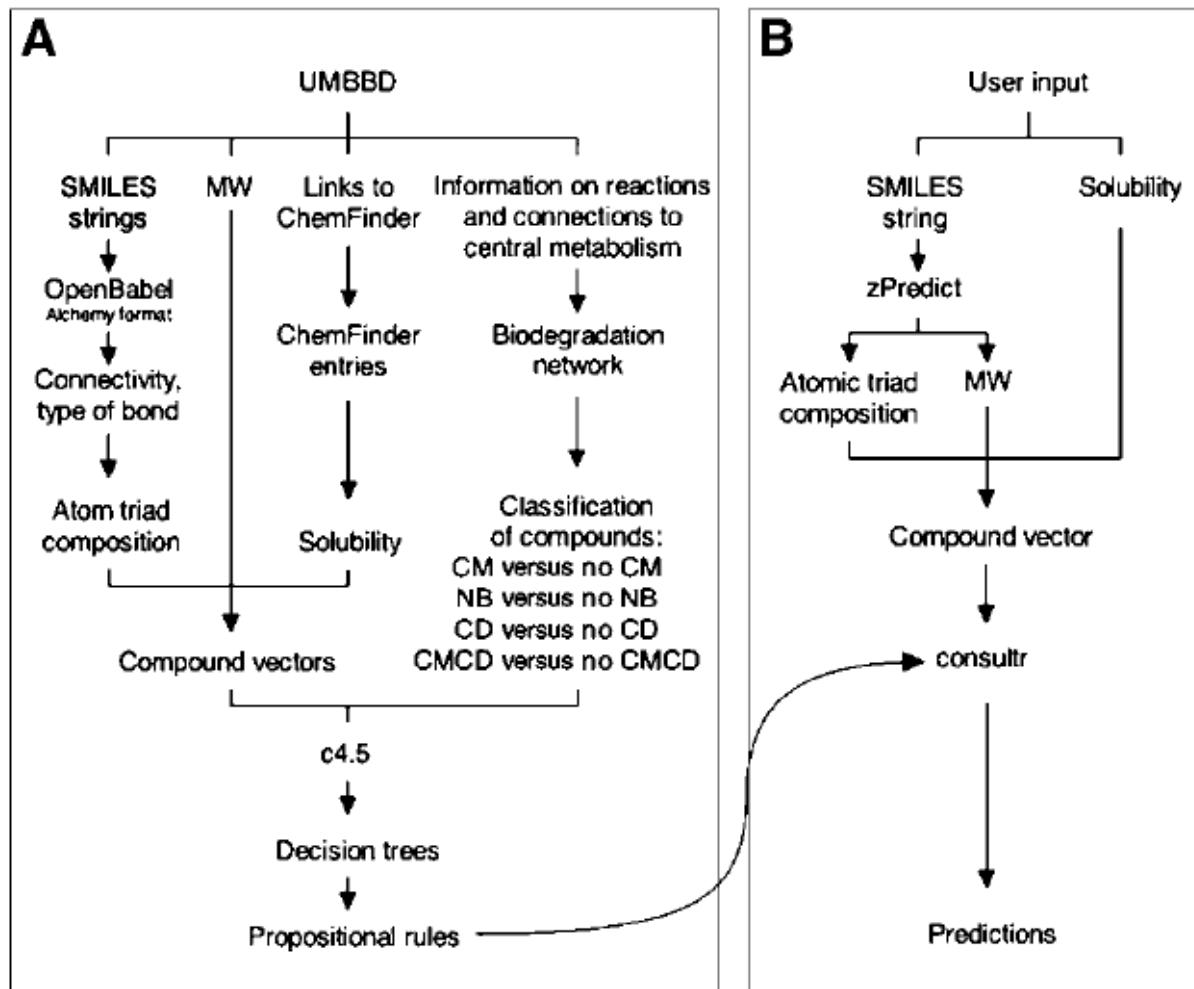
Training the system



Advantages of rule-based systems compared with other ML, i.e. NN:

- Explicit rules in human-readable format
- Can handle missing values (solubility)

Schema



Results

Table I Example of propositional rule generated for the classification compounds in the scheme NB or No NB

Rule 55: IF
 $-C-C-C > 19$
 $-O-C-C > 1$
 $-O-C-C \leq 3$
 THEN, the compound belongs to the NB class (Confidence 90.6%)
 Examples (14 cases)

52 out of the 152 attributes involved in rules

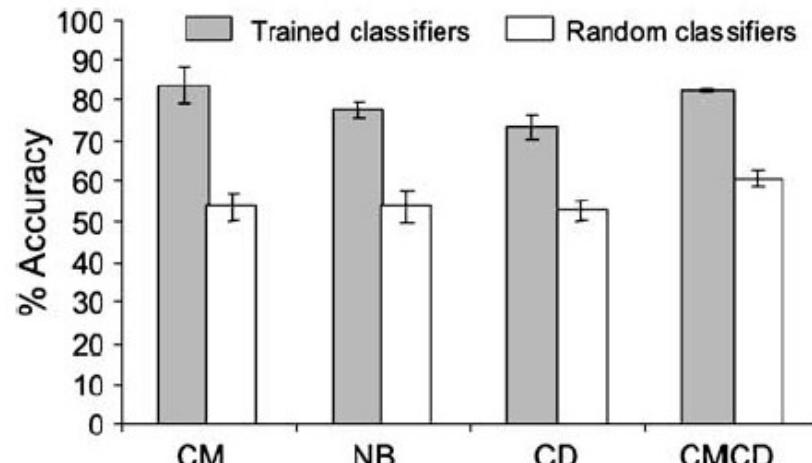


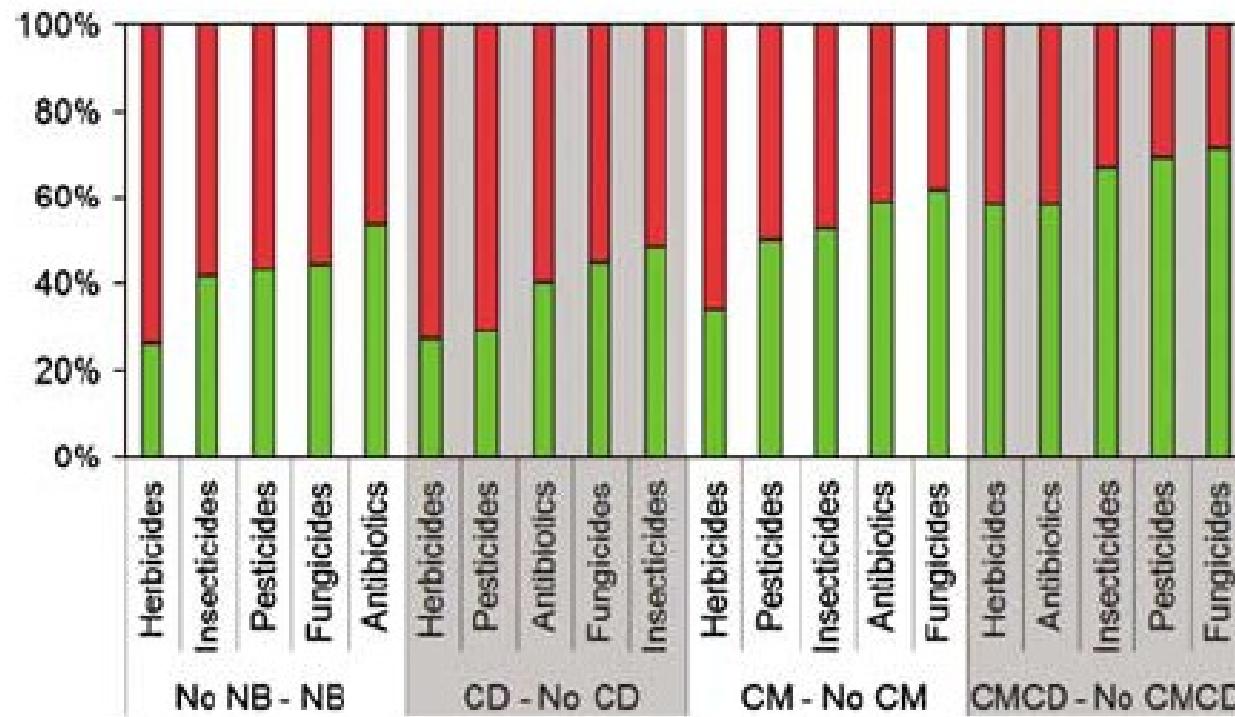
Table II Predictive performance in fivefold cross-validation experiments

Classification scheme ^a	CM or No CM	NB or No NB	CD or No CD	CMCD or No CMCD
Accuracy (%)	87 ± 4	77 ± 4	73 ± 5	82 ± 3
Significance respect to random—P(N)	1.1×10^{-39}	1.2×10^{-24}	2.4×10^{-16}	6.4×10^{-26}
Default class	CM	No NB	No CD	CMCD
Majority class	CM	No NB	No CD	CMCD
No. of cases	533	496	513	634
Sensitivity (%)	93 ± 4	86 ± 4	78 ± 7	92 ± 3
Specificity (%)	83 ± 3	78 ± 2	78 ± 3	85 ± 2
Minority class				
No. of cases	308	345	328	207
Sensitivity (%)	67 ± 5	64 ± 5	66 ± 9	50 ± 11
Specificity (%)	85 ± 8	77 ± 4	66 ± 5	71 ± 5

coli CM (733 comps.): 90.31% CMCD; 81.44% No-NB

Blind predictions

- Chemical compounds regulated by the ECB (HPVCs, LPVCs & Annex-I): 70% CMCD
- PubChem (USA). 3600 comps. aprox.



Using the predictor - BDPServer

The image shows two browser windows side-by-side. The left window is titled 'BDPServer - Mozilla Firefox' and displays the BDPServer homepage. It features a logo with the letters 'PDG' and the words 'Protein Design Group'. Below the logo, the text 'BDPServer: prediction of environmental fate for chemical compounds' is displayed. A green arrow points from the SMILES input field on the BDPServer page to the JME drawing area on the right. The right window is titled 'http://www.pdg.cnb.uam.es - JME Molecular' and shows a chemical structure of 1,2-dichlorobenzene (a benzene ring with two chlorine atoms) drawn using the JME applet. The JME interface includes a toolbar with various chemical symbols and a panel for drawing atoms.

BDPServer

BDPServer predicts whether chemical compounds can be biodegraded or not.

Chemical compound descriptions can be typed directly in SMILES format or drawn with the JME applet. Solubility information is optional. The JME applet has been provided by Peter Ertl, from Novartis.

SMILES: CC(=C)C1CCC(=CC1)C

Solubility (g/100 ml):

JME

For prediction

Results

Optional predictions for known compounds: Yes

Smi: CC(=C)C1CCC(=CC1)C Sol: ?

>> The compound has been found in the BDPServer database, with ID: c0626.

BDPServer classification:	NB:	Yes
BDPServer classification:	CM:	No
BDPServer classification:	CD:	No
BDPServer classification:	CMCD:	No

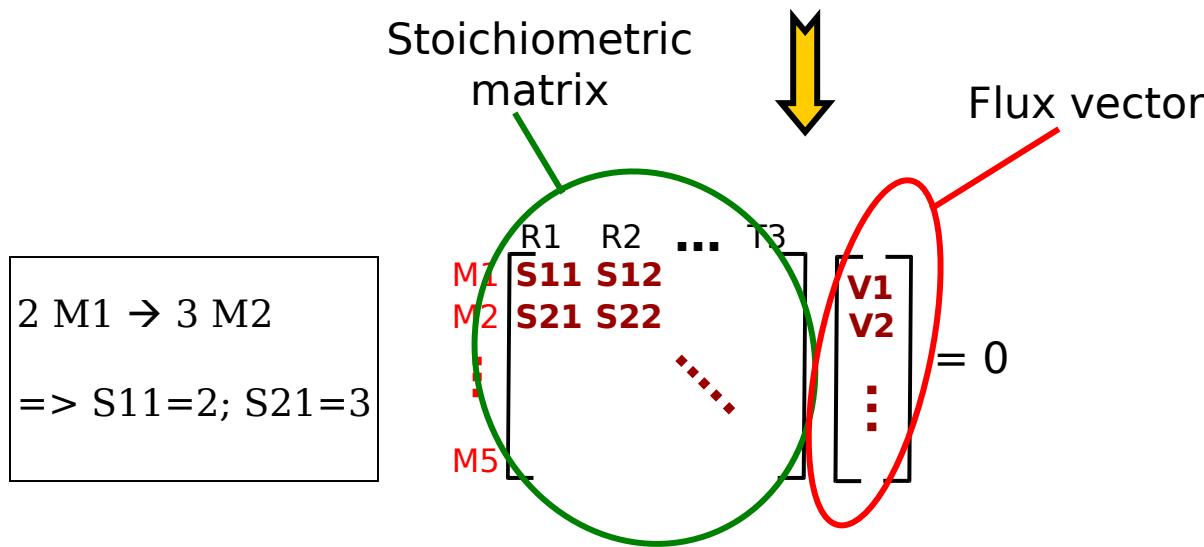
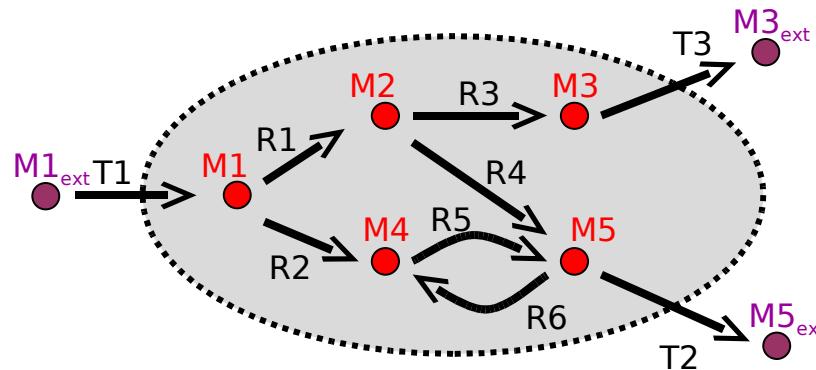
>> Adding predictions for known compound.

BDPServer prediction:	NB	Yes	CF = 0.95
BDPServer prediction:	CM	No	CF = 0.94
BDPServer prediction:	CD	No	CF = 0.86
BDPServer prediction:	CMCD	No	CF = 0.91

Done

- Gómez, MJ, Pazos F, Guijarro FJ, de Lorenzo V, Valencia A. (2007). The environmental fate of organic pollutants through the global microbial metabolism. *Mol Syst Biol.* 3:114.
- <http://pdg.cnb.csic.es/DBPSERVER/>

Análisis de Flujo

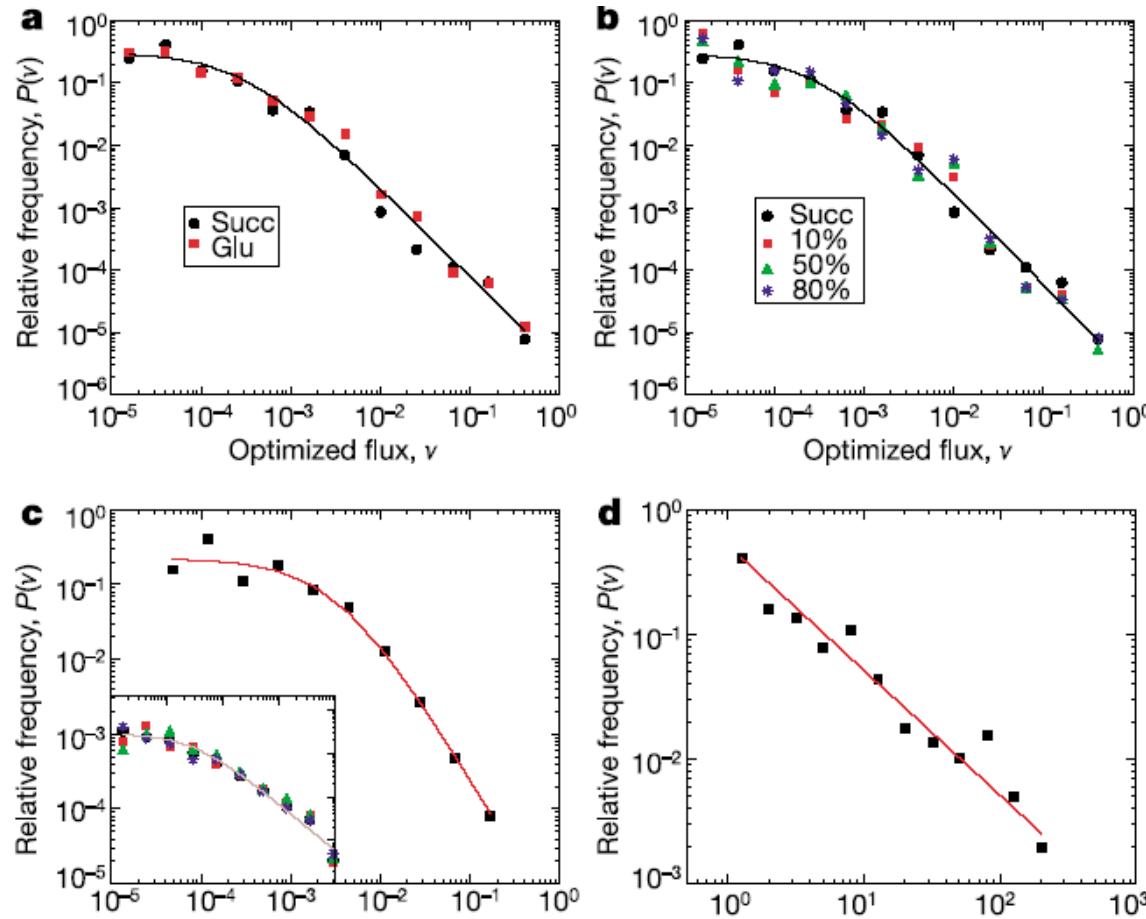


J.S. Edwards & B.O. Palsson, *Proc. Natl. Acad. Sci. USA* **97**, 5528 (2000)

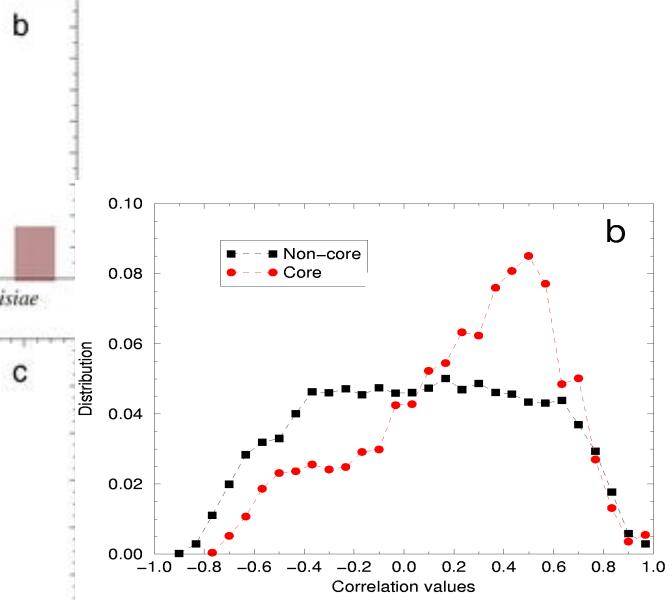
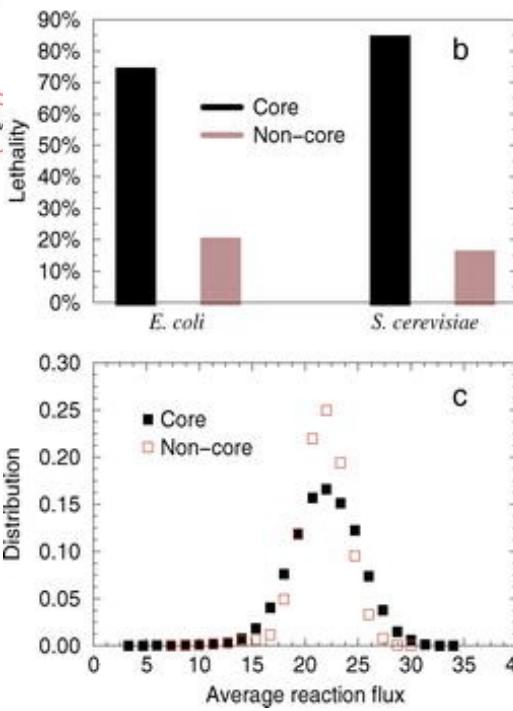
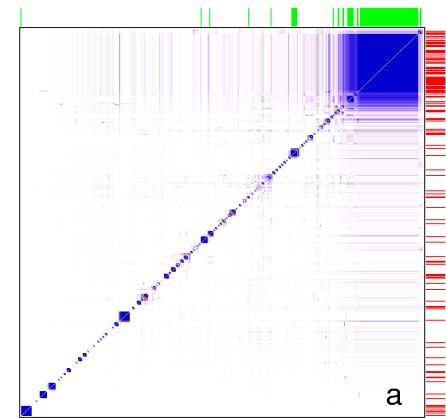
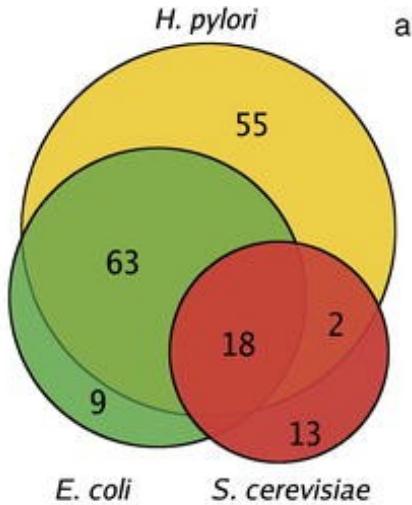
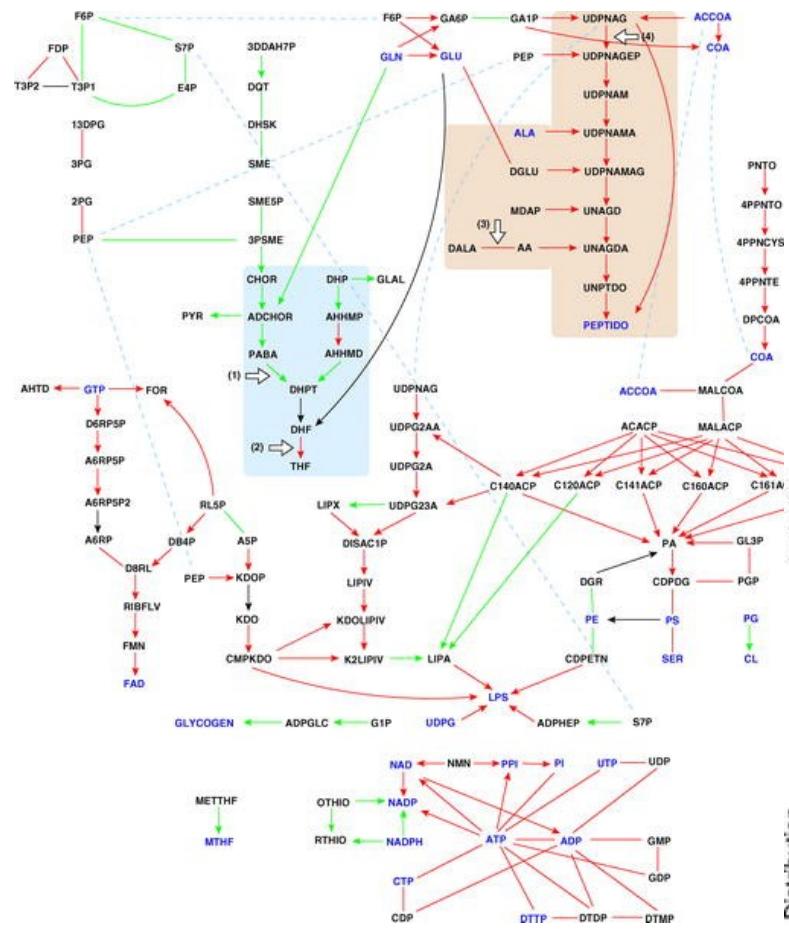
R.U. Ibarra, J.S. Edwards & B.O. Palsson, *Nature* **420**, 186 (2002)

D. Segre, D. Vitkup & G.M. Church, *Proc. Natl. Acad. Sci. USA* **99**, 15112 (2002)

scale-free más alla de la topología

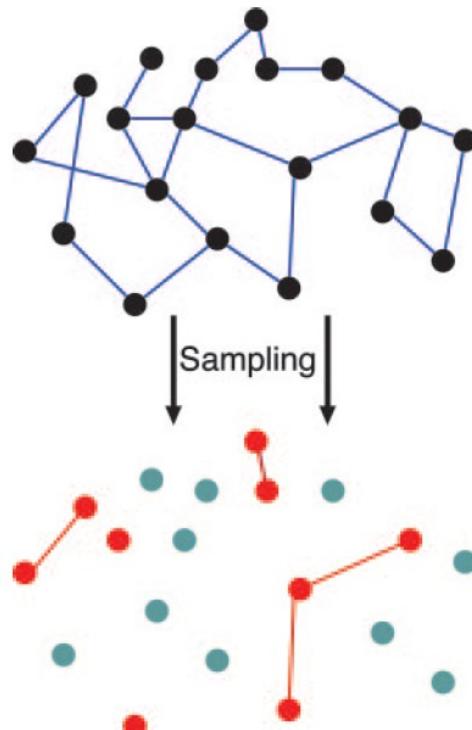


Características del core de flujos metabólicos



¿Como apareció la estructura scale-free?

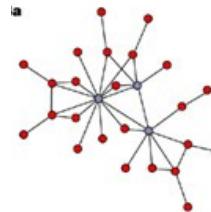
Muestreo (=>artefacto) Evolución



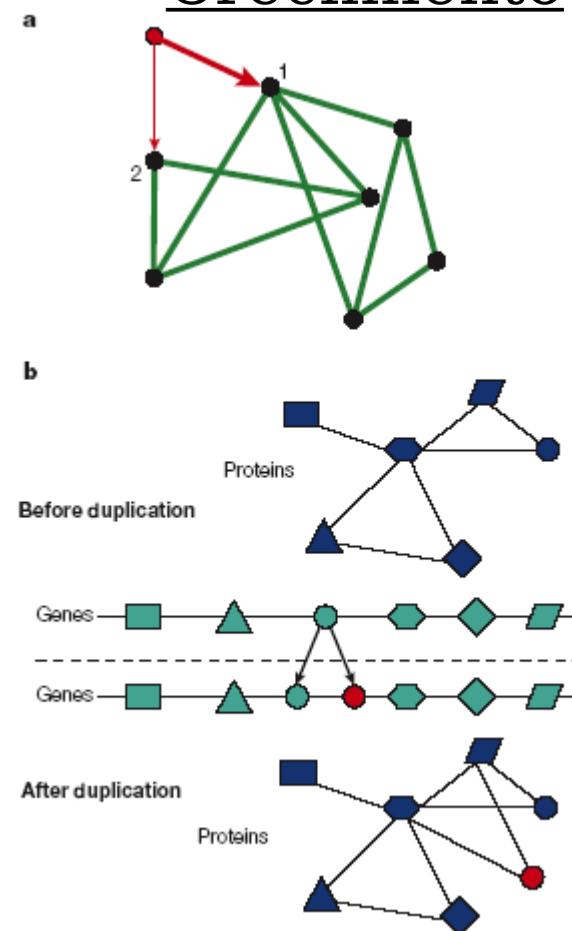
Random network

evolution

“scale-free”



Crecimiento

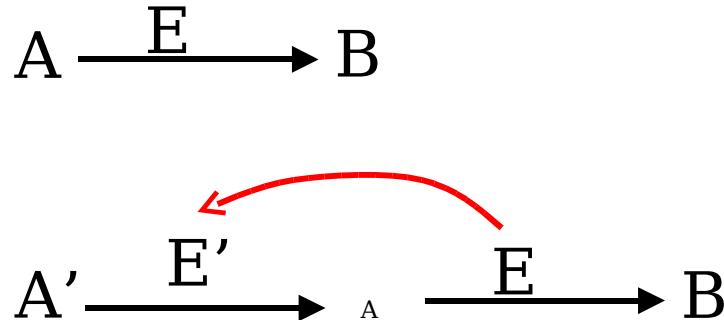


Stumpf, M.P., Wiuf, C. and May, R.M. (2005) Subnets of scale-free networks are not scale-free: Sampling properties of networks. *Proc Natl Acad Sci U S A*, **102**, 4221-4224.

Barabasi, A.L. and Oltvai, Z.N. (2004) Network biology: understanding the cell's functional organization. *Nat Rev Genet*, **5**, 101-113.

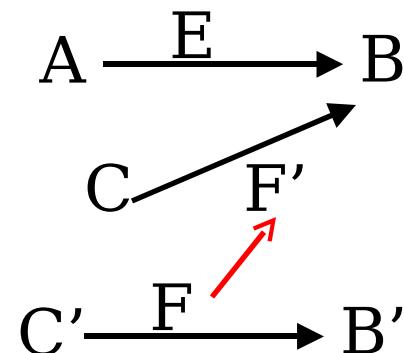
Evolución de la Red Metabólica

Retroevolution



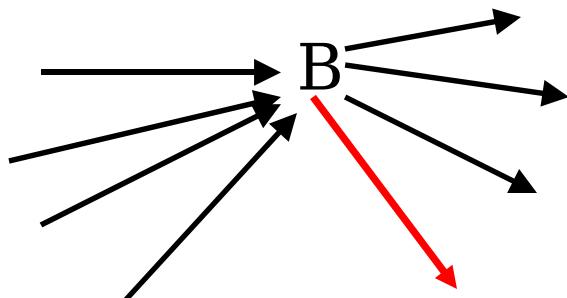
Horowitz, N.H. (1945) On the evolution of biochemical syntheses. *Proc. Natl Acad. Sci. USA*, **31**, 153-157.

Recruitment



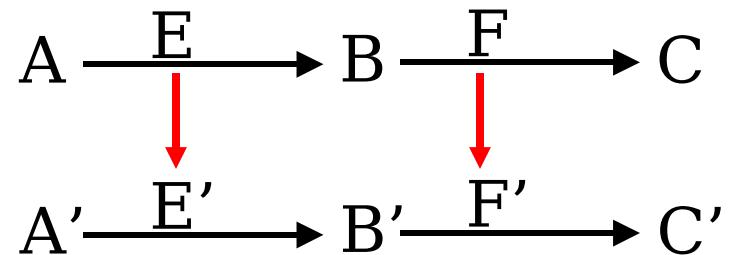
Jensen, R.A. (1976) Enzyme recruitment in evolution of new function. *Annu Rev Microbiol*, **30**, 409-425.

Preferential attachment



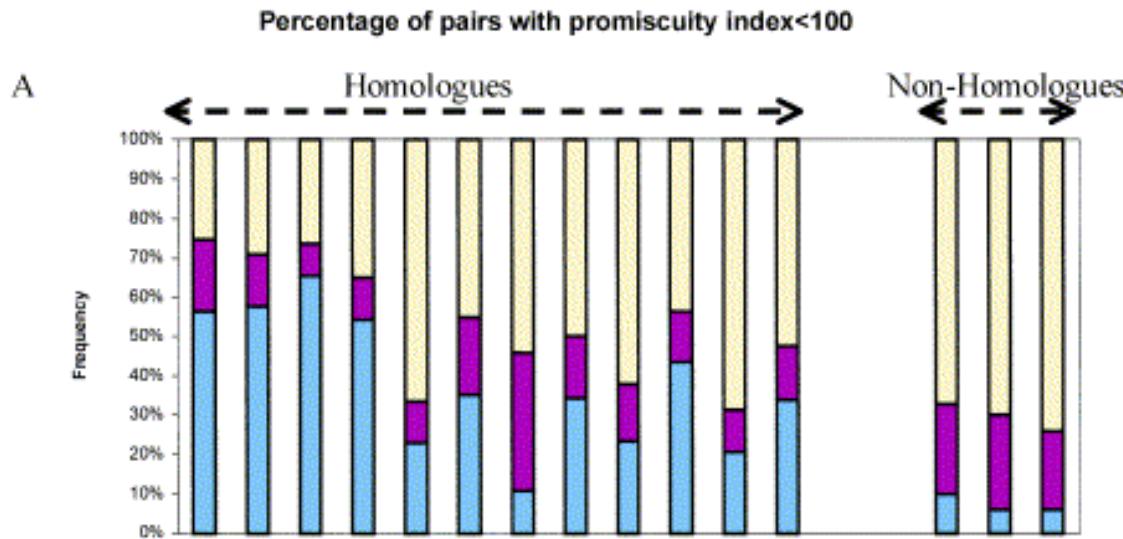
Schmidt, S., Sunyaev, S., Bork, P. and Dandekar, T. (2003) Metabolites: a helping hand for pathway evolution? *Trends Biochem Sci*, **28**, 336-341.

Pathway duplication

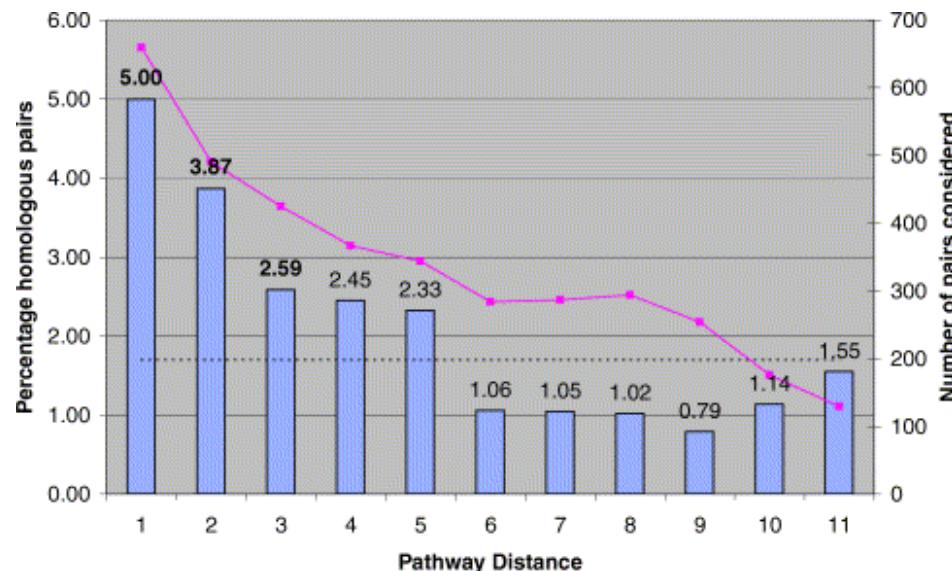


Rison, S.G.C. and Thornton, J.M. (2002) Pathway evolution, structurally speaking. *Curr Opin Struct Biol*, **12**, 374-382.

Evolución de la Red Metabólica

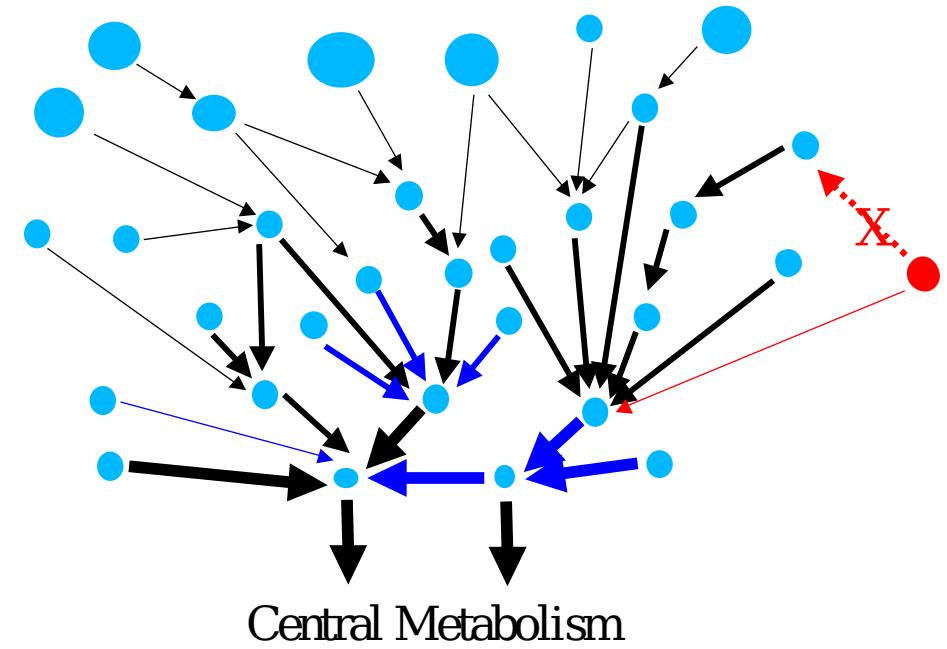
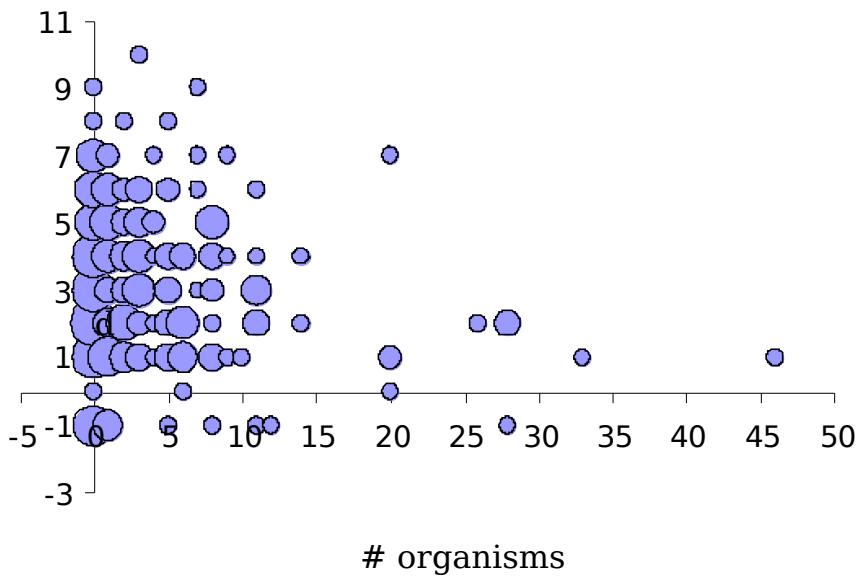


Alves, R., Chaleil, R.A.G. and Sternberg, M.J.E. (2002) Evolution of enzymes in metabolism: a network perspective. *J Mol Biol*, **320**, 751-770.

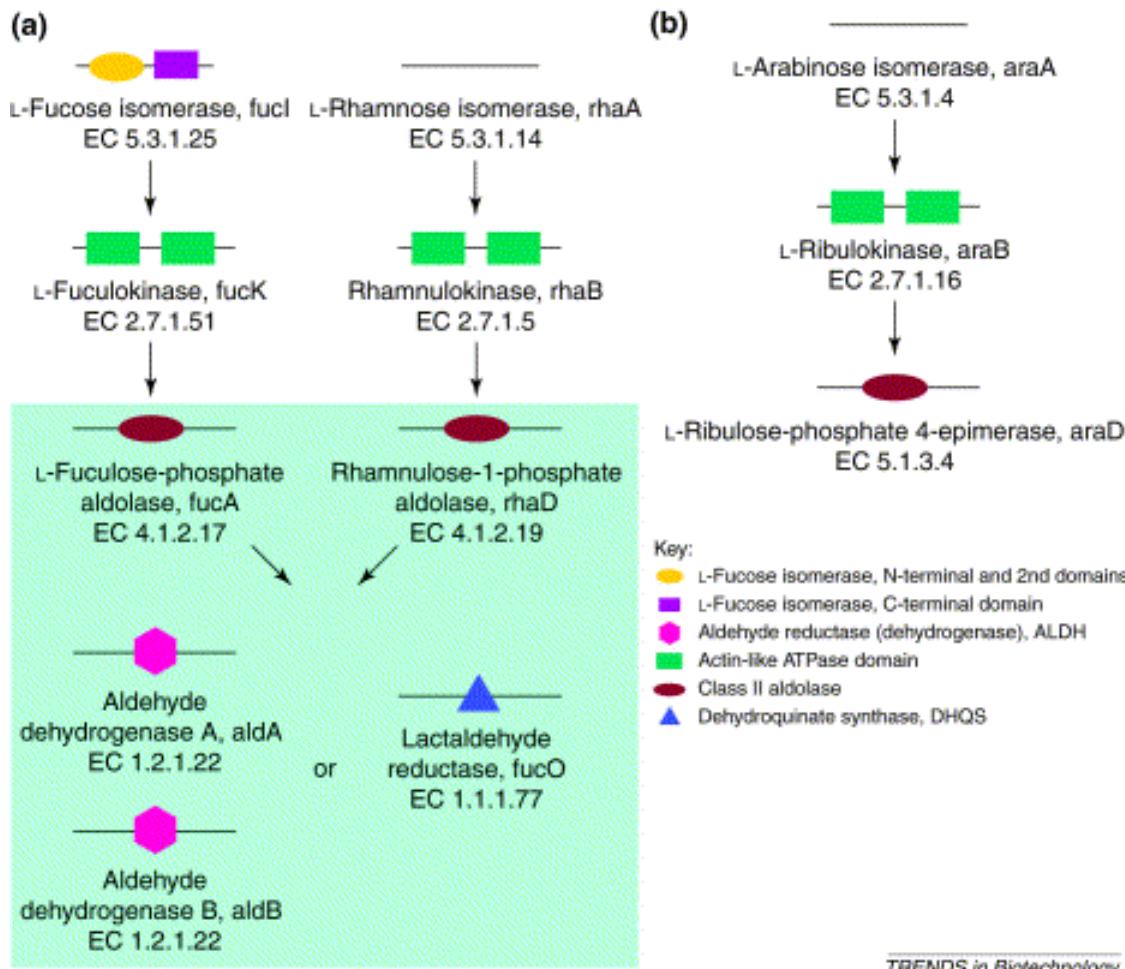


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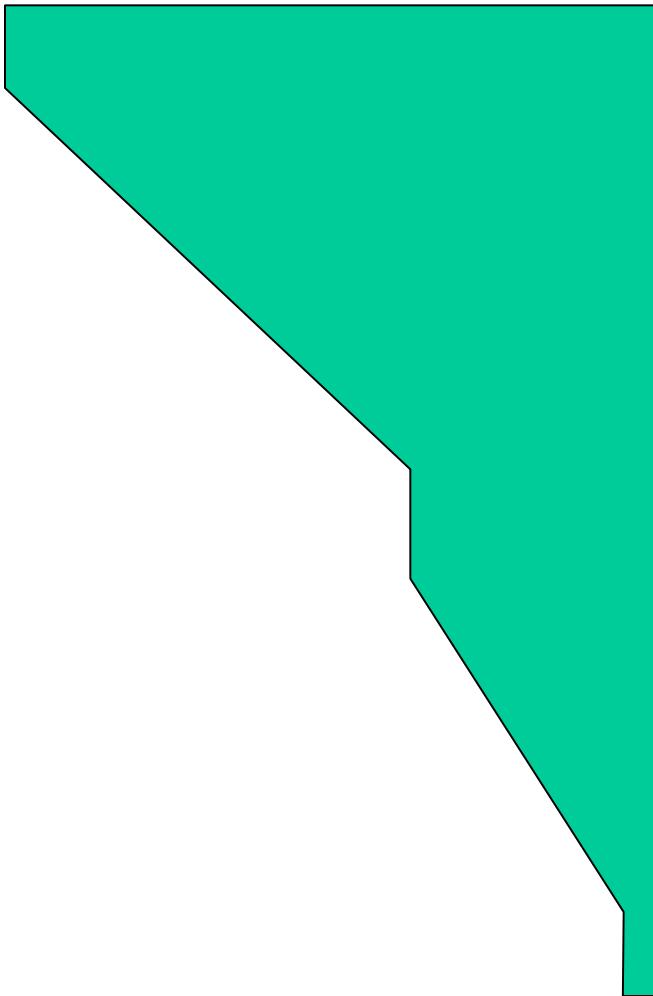
Evolución de la Red Metabólica



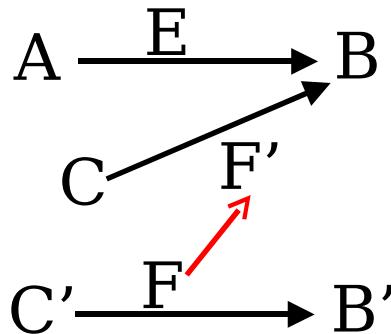
Evolución de la Red Metabólica



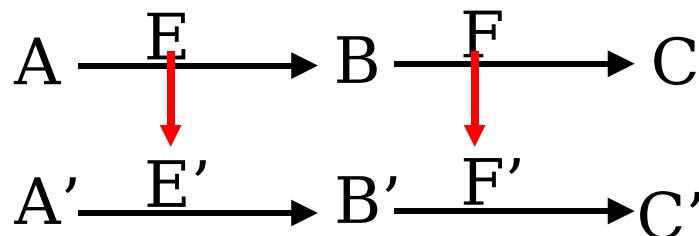
TRENDS in Biotechnology



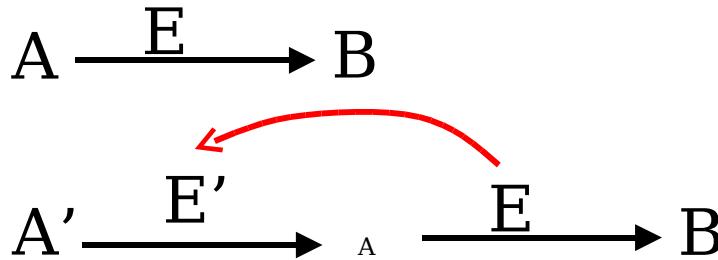
Recruitment



Pathway duplication



Retroevolution



Principales bases de datos con información metabólica

KEGG – *Metabolismo en general*

<http://www.genome.jp/kegg/>

Kanehisa, M., Goto, S., Kawashima, S., Okuno, Y. and Hattori, M. (2004) The KEGG resource for deciphering the genome. *Nucleic Acids Res.* **32**, D277-280.

EcoCyc – *Metabolismo de E. coli*

<http://ecocyc.org>

Keseler, I. M., Collado-Vides, J., Gama-Castro, S., Ingraham, J., Paley, S., Paulsen, I. T., Peralta-Gil, M. and Karp, P. D. (2005). EcoCyc: a comprehensive database resource for Escherichia coli. *Nucleic Acids Res.* **33**: D334-337

BRENDA – *Centrada en enzimas (incluyendo parámetros cinéticos y termodinámicos)*

<http://www.brenda.uni-koeln.de/>

Schomburg, I., Chang, A., Ebeling, C., Gremse, M., Heldt, C., Huhn, G. and Schomburg, D. (2004) BRENDA, the enzyme database: updates and major new developments. *Nucleic Acids Res.* **32**, D431-433.

MBBD – *Centrada en biodegradación y bioremediación*

<http://umbbd.msi.umn.edu/>

Ellis, L.B., Hou, B.K., Kang, W. and Wackett, L.P. (2003) The University of Minnesota Biocatalysis/Biodegradation Database: post-genomic data mining. *Nucleic Acids Res.* **31**, 262-265.