



Ciencia de Redes (Humanas y Sociales)

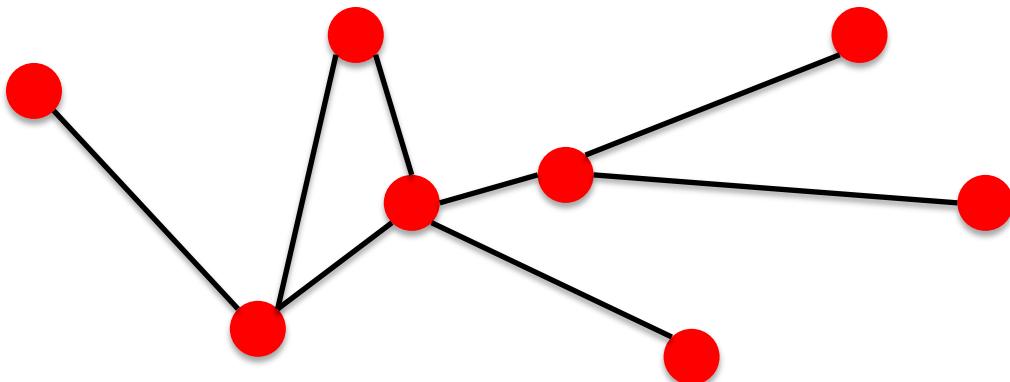
#2

Carlos Sarraute

Instituto de Cálculo, Abril-Junio 2019

Conceptos fundamentales de Teoría de Grafos

COMPONENTES DE UN SISTEMA COMPLEJO



■ **componentes:** nodos, vértices

N

■ **interacciones:** vínculos, enlaces, aristas

L

■ **sistema:** red, grafo

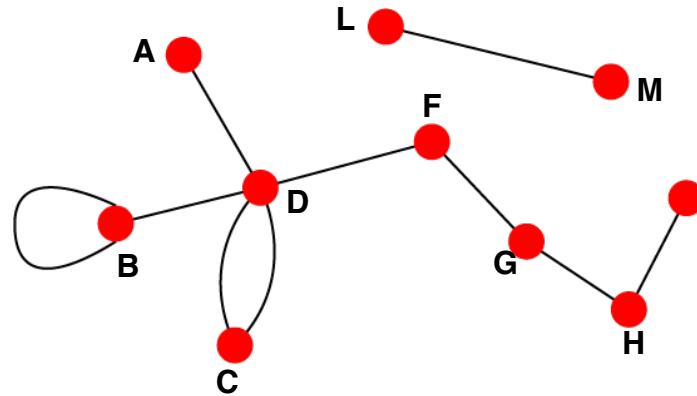
(N,L)

REDES DIRIGIDAS VS. NO DIRIGIDAS

No dirigido

Enlaces: no dirigidos (simétricos)

Grafo:



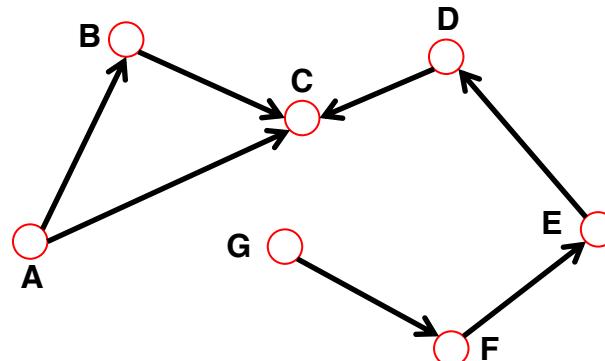
Enlaces no dirigidos:

Vínculo de coautor
Red de actores
Interacciones entre proteínas

Dirigido

Links: directed (*arcs*).

Digrafo = directed graph:



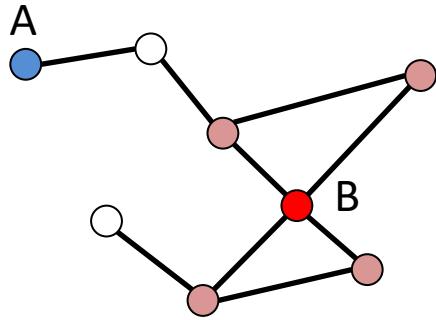
Enlaces dirigidos :

URLs en la web
Llamados telefónicos
Reacciones metabólicas

Distribución de grados

GRADO DE UN NODO

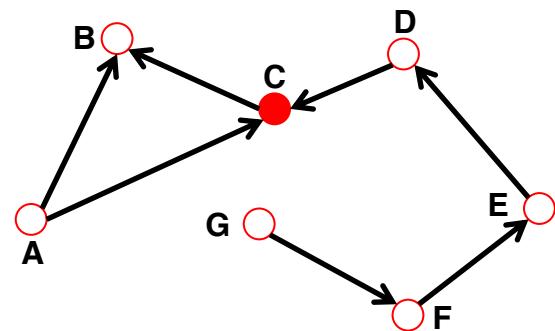
No dirigido



Grado del nodo: cantidad de enlaces que conectan con el nodo

$$k_A = 1 \quad k_B = 4$$

Dirigido



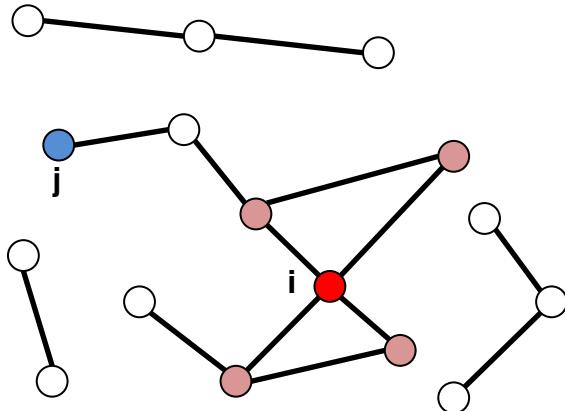
En los *grafos dirigidos* se puede definir un **in-degree** y **out-degree**. El grado (total) es la suma de in- y out-degree.

$$k_C^{in} = 2 \quad k_C^{out} = 1 \quad k_C = 3$$

Source: nodo con $k^{in}= 0$; Sink: nodo con $k^{out}= 0$.

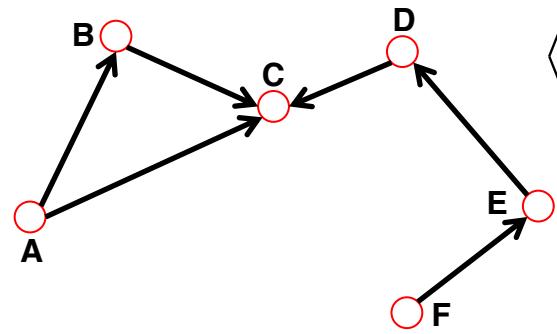
GRADO PROMEDIO

No dirigido



$$\langle k \rangle \equiv \frac{1}{N} \sum_{i=1}^N k_i \quad \langle k \rangle \equiv \frac{2L}{N}$$

Dirigido



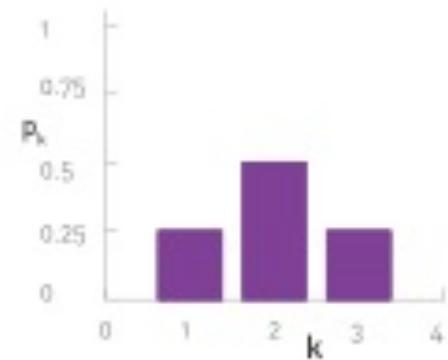
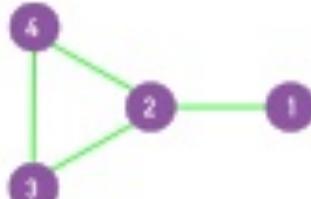
$$\langle k^{in} \rangle \equiv \frac{1}{N} \sum_{i=1}^N k_i^{in}, \quad \langle k^{out} \rangle \equiv \frac{1}{N} \sum_{i=1}^N k_i^{out}, \quad \langle k^{in} \rangle = \langle k^{out} \rangle$$

$$\langle k \rangle \equiv \frac{L}{N}$$

DISTRIBUCIÓN DE GRADOS

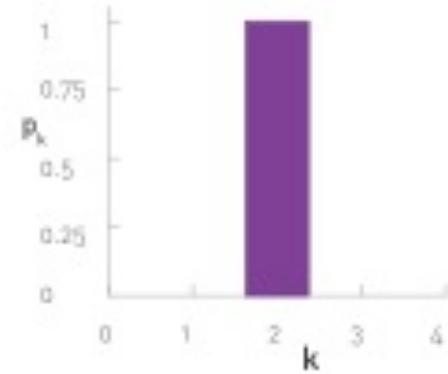
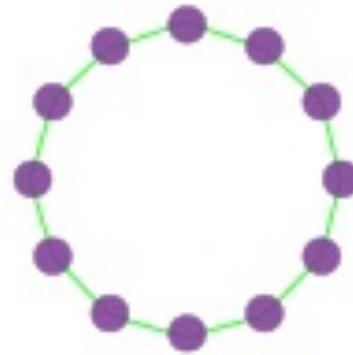
Distribución de grados

$P(k)$: probabilidad de que un nodo al azar tenga grado k



$N_k = \# \text{ nodos con grado } k$

$P(k) = N_k / N \rightarrow \text{plot}$



DISTRIBUCIÓN DE GRADOS

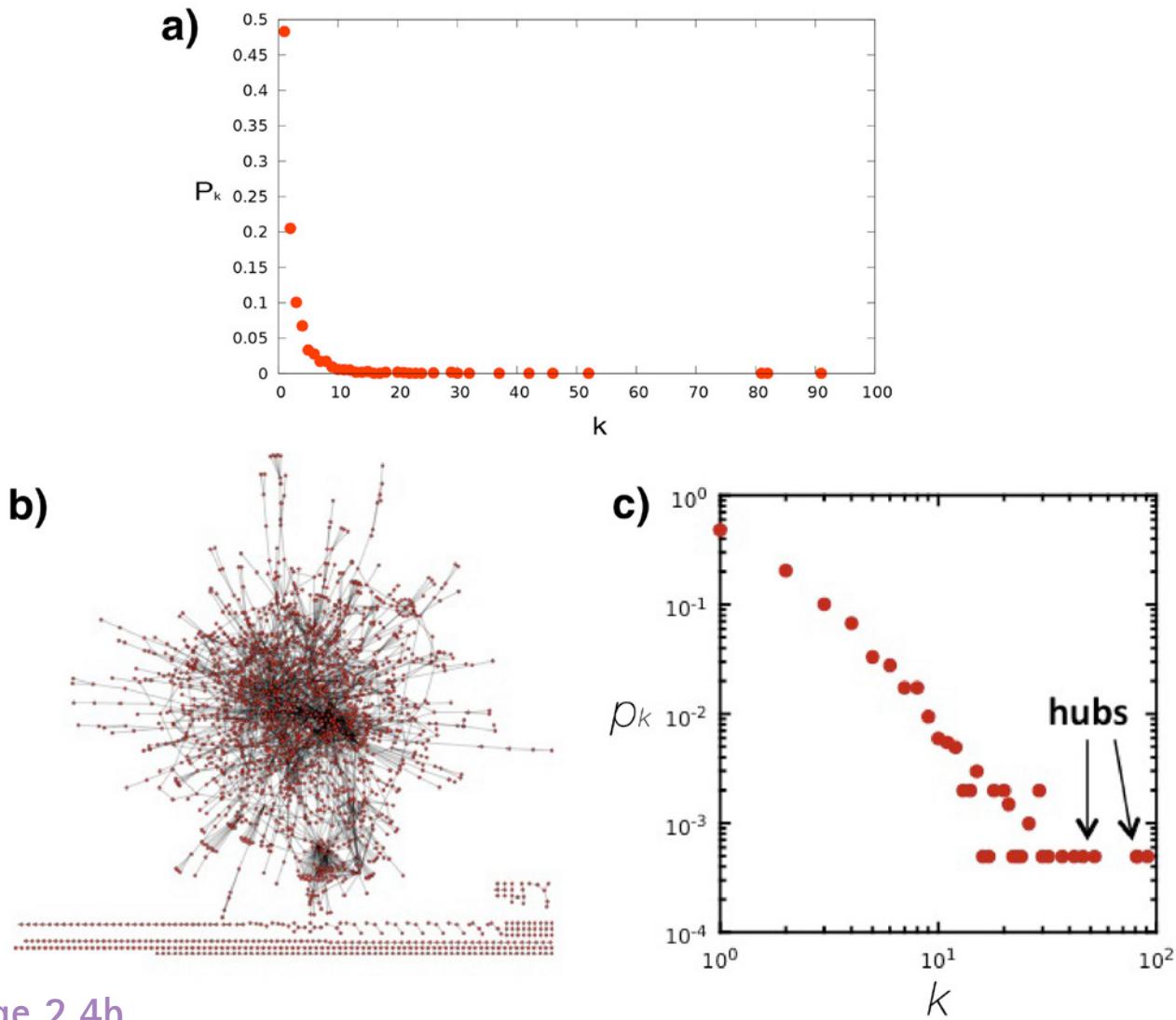


Image 2.4b

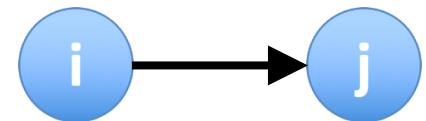
Sección 2

Matriz de adyacencia

Matriz de adyacencia

- Representa enlaces como matriz

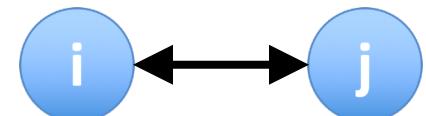
- $A_{ij} = 1$ si nodo i tiene enlace hacia nodo j
 $= 0$ sino



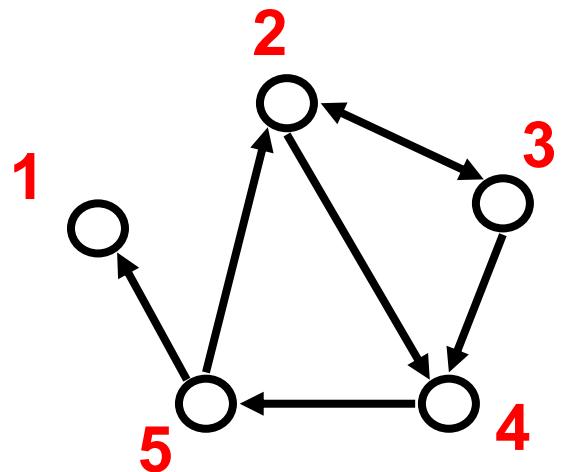
- $A_{ii} = 0$ salvo que el grafo tenga “self-loops”



- $A_{ij} = A_{ji}$ si el grafo es no dirigido,
o si i y j tienen un enlace recíproco

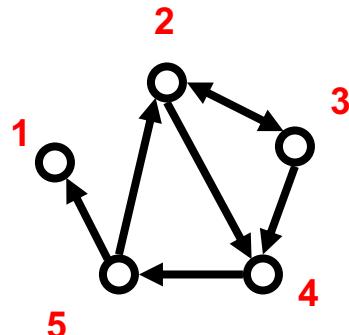


Ejemplo de matriz de adyacencia



$$A = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 & 0 \end{bmatrix}$$

Grados de nodos usando matriz



$$\text{Outdegree} = \sum_{j=1}^n A_{ij}$$

ejemplo: outdegree para nodo 3
sumamos la 3^{er} fila

$$\sum_{j=1}^n A_{3j}$$

$$\text{Indegree} = \sum_{i=1}^n A_{ij}$$

ejemplo: indegree para nodo 3
Sumamos la 3^{er} columna

$$\sum_{i=1}^n A_{i3}$$

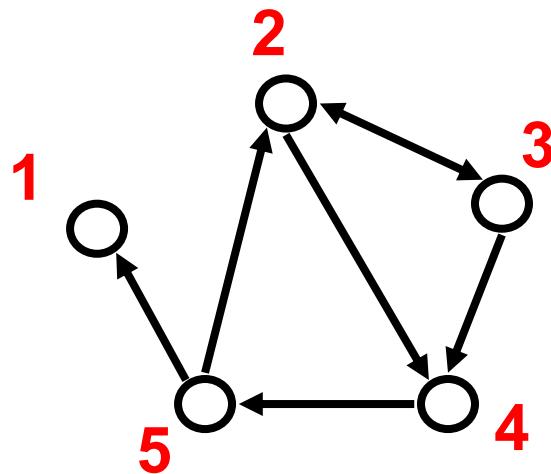
$$A = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 & 0 \end{bmatrix}$$

$$A = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 & 0 \end{bmatrix}$$

Lista de aristas

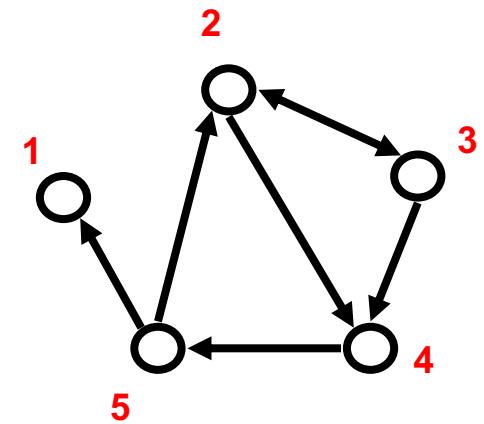
- Lista de aristas

- 2, 3
- 2, 4
- 3, 2
- 3, 4
- 4, 5
- 5, 2
- 5, 1

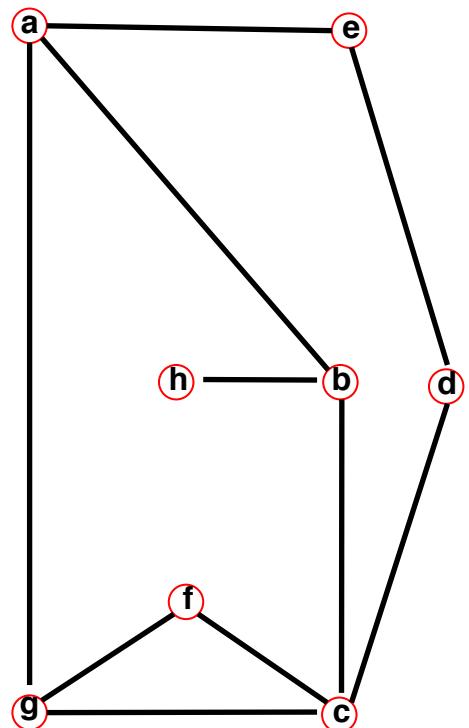


Lista de adyacencia

- Lista de adyacencia
 - Mas facil de usar para redes
 - grandes
 - ralas (sparse)
 - Recuperar facilmente los vecinos de un nodo
 - 1:
 - 2: 3 4
 - 3: 2 4
 - 4: 5
 - 5: 1 2



Ejemplo de matriz de adyacencia



	a	b	c	d	e	f	g	h
a	0	1	0	0	1	0	1	0
b	1	0	1	0	0	0	0	1
c	0	1	0	1	0	1	1	0
d	0	0	1	0	1	0	0	0
e	1	0	0	1	0	0	0	0
f	0	0	1	0	0	0	1	0
g	1	0	1	0	0	1	0	0
h	0	1	0	0	0	0	0	0

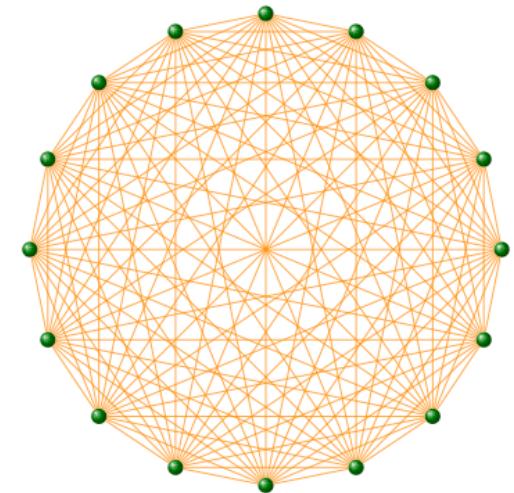
Sección 3

Las redes reales son ralas (sparse)

Grafo completo

La cantidad máxima de vínculos en una red con N nodos:

$$L_{\max} = \binom{N}{2} = \frac{N(N-1)}{2}$$



Un grafo con vínculos $L = L_{\max}$ se llama **grafo completo**, su grado promedio es $\langle k \rangle = N-1$

LAS REDES REALES SON RALAS

La mayoría de las redes observadas en sistemas reales son ralas (sparse):

$$L \ll L_{\max}$$

$$\langle k \rangle \ll N-1.$$

WWW (ND Sample):	$N=325,729;$	$L=1.4 \cdot 10^6$	$L_{\max}=10^{12}$	$\langle k \rangle=4.51$
Protein (<i>S. Cerevisiae</i>):	$N= 1,870;$	$L=4,470$	$L_{\max}=10^7$	$\langle k \rangle=2.39$
Coauthorship (Math):	$N= 70,975;$	$L=2 \cdot 10^5$	$L_{\max}=3 \cdot 10^{10}$	$\langle k \rangle=3.9$
Movie Actors:	$N=212,250;$	$L=6 \cdot 10^6$	$L_{\max}=1.8 \cdot 10^{13}$	$\langle k \rangle=28.78$

(Source: Albert, Barabasi, RMP2002)

MATRICES DE ADYACENCIA SON RALAS

Las matrices de adyacencia de grafos ralas tienen la siguiente forma:

$$\begin{pmatrix} 0 & 1 & 0 & \dots & 0 \\ 1 & 0 & 1 & \dots & 0 \\ 0 & 1 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \end{pmatrix}$$

En donde el orden de la matriz es igual al número de vértices del grafo.

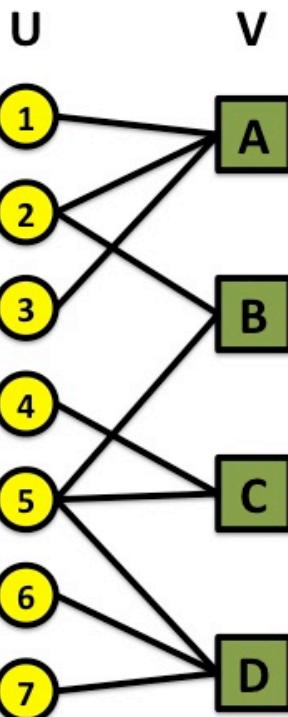
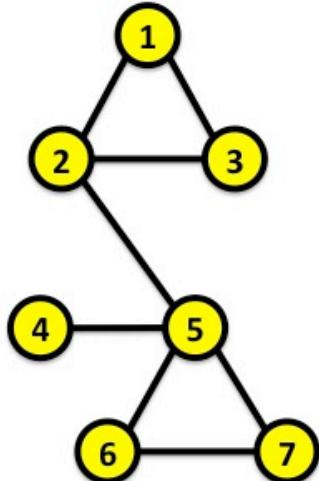
Sección 4

REDES BIPARTITAS

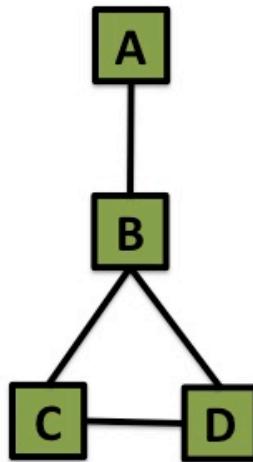
GRAFO BIPARTITO

grafo bipartito es un grafo cuyos nodos se pueden dividir en dos conjuntos separados U y V, de manera que cada enlace conecta un nodo en U con uno en V; es decir, U y V son **conjuntos independientes**.

Projection U



Projection V

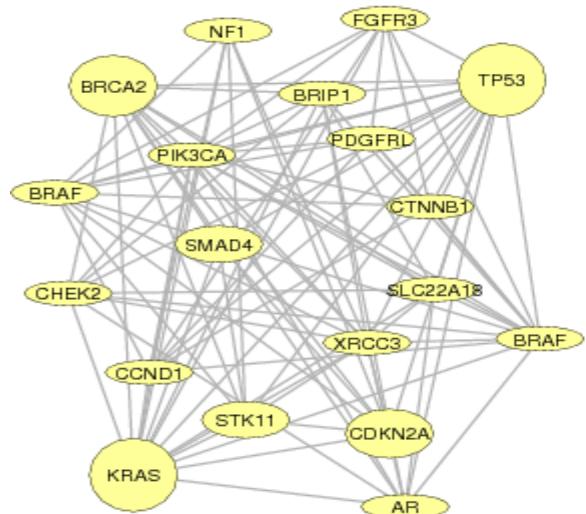


Ejemplos:

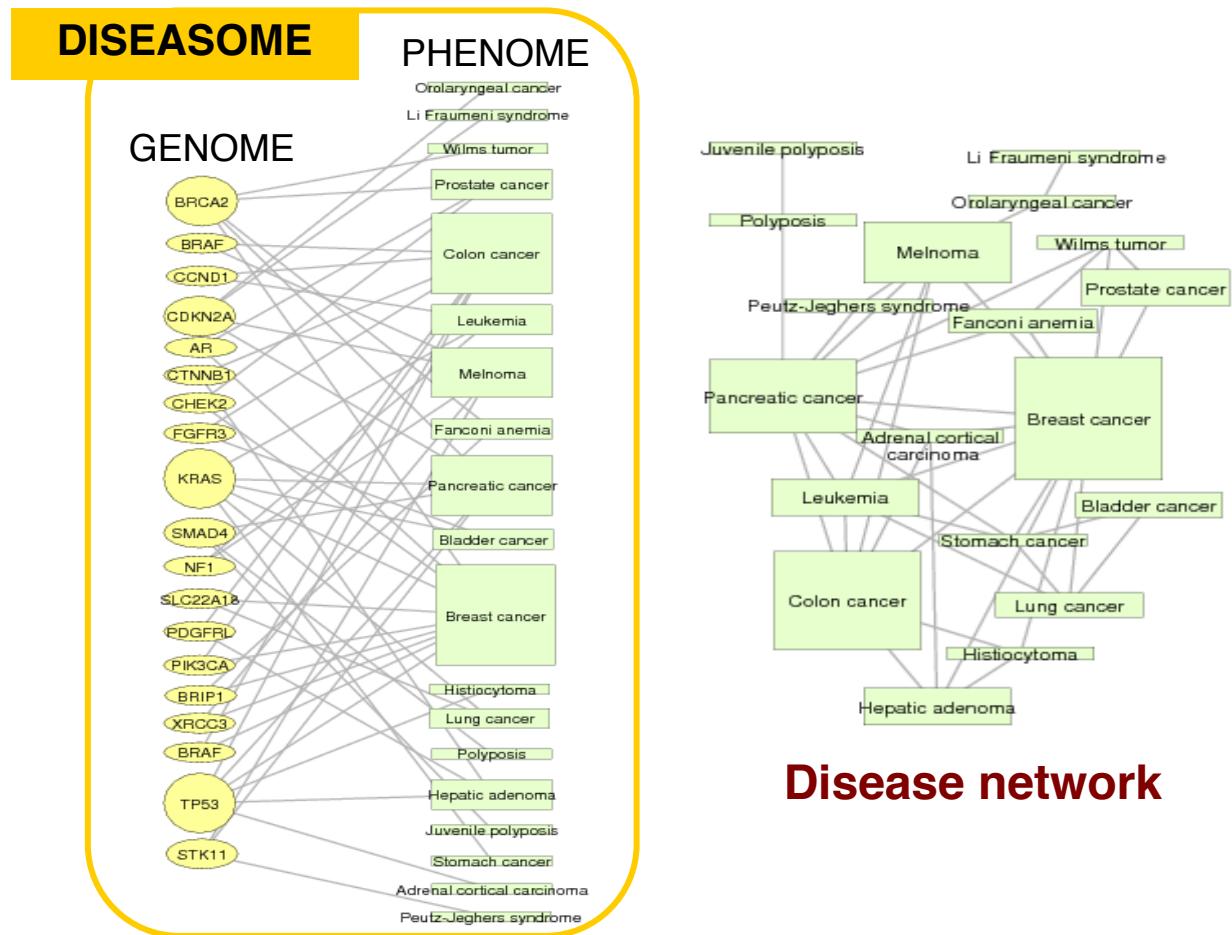
Red de actores del cine argentino

Red de enfermedades

RED DE GENES Y RED DE ENFERMEDADES



Gene network



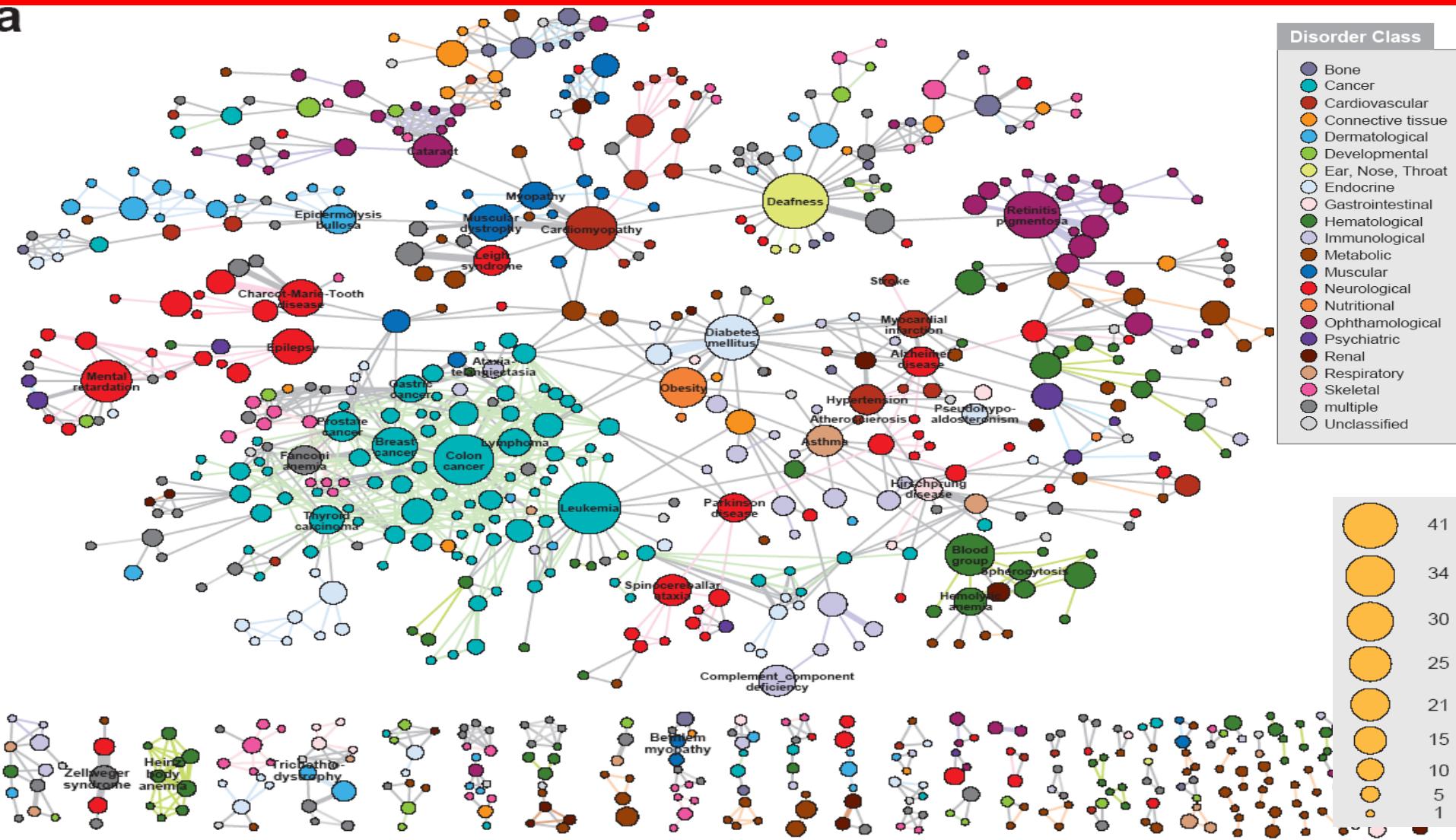
Goh, Cusick, Valle, Childs, Vidal & Barabási, PNAS (2007)



GRANDATA

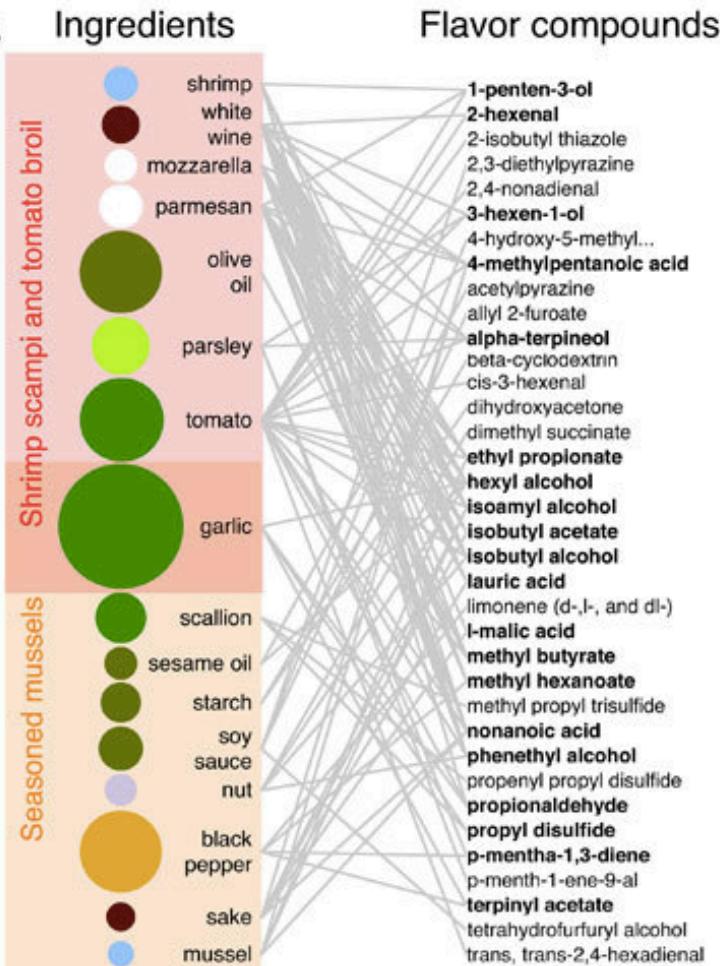
RED DE ENFERMEDADES HUMANAS

a

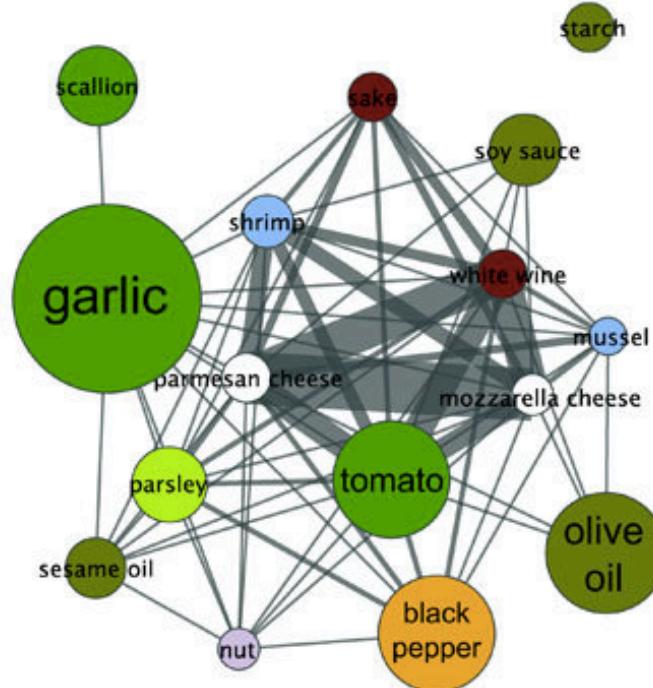


RED BIPARTITA DE INGREDIENTES Y SABORES

A



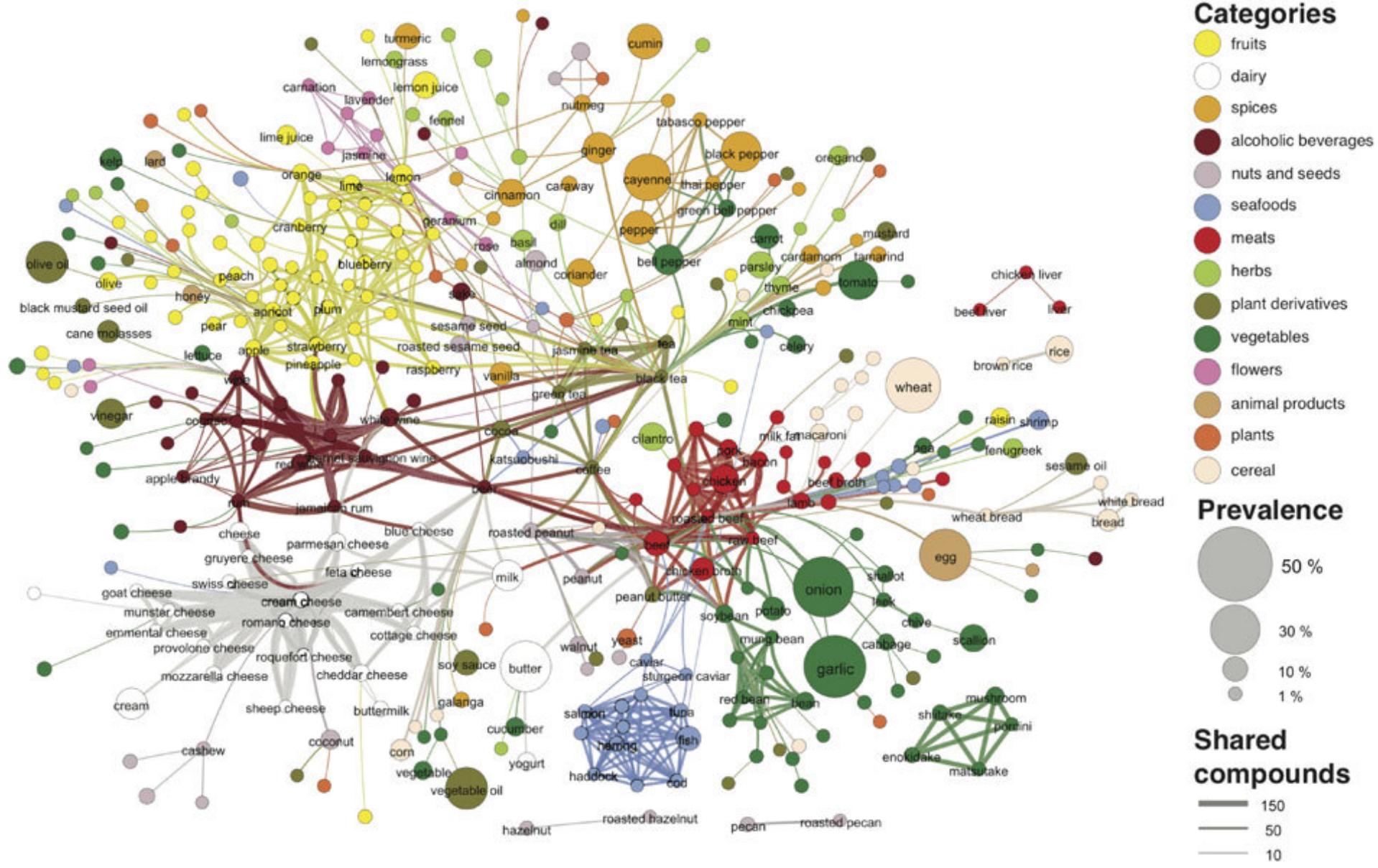
B Flavor network



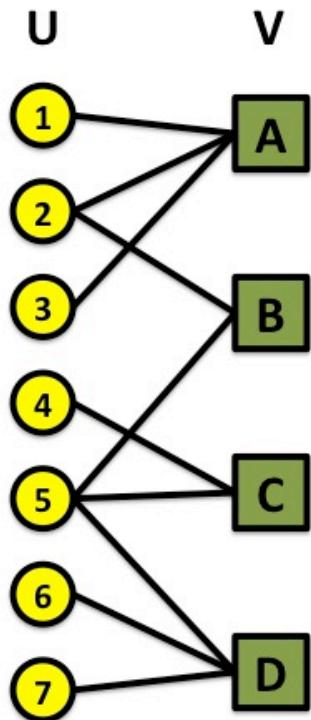
Y.-Y. Ahn, S. E. Ahnert, J. P. Bagrow, A.-L. Barabási *Flavor network and the principles of food pairing*, *Scientific Reports* 196, (2011).



GRANDATA

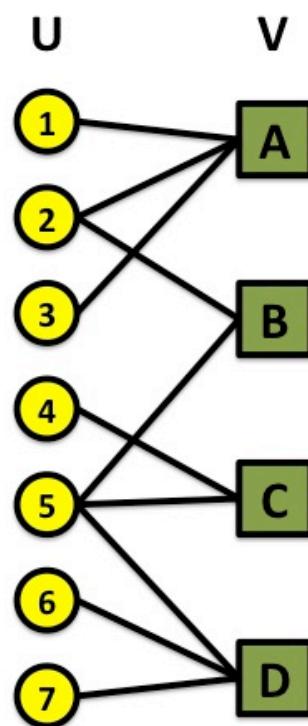


Ejemplos de grafos bipartitos



Ejemplos de grafos bipartitos

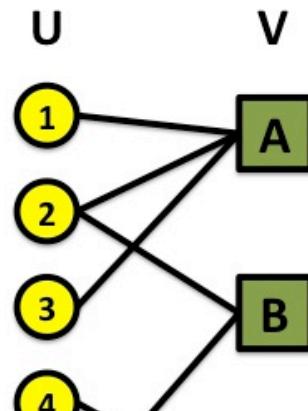
- Científicos
- Actores
- Músicos



- Papers
- Películas
- Bandas,
conciertos

Ejemplos de grafos bipartitos

Legisladores



Leyes



Nahue @nahuelhds · 20 abr.

Comparto todas las votaciones desde 1993 de la Cámara de Diputados de Argentina [@HCDNArgentina](#) en formato SQL y CSV para quien quiera utilizarla



github.com/nahuelhds/vota...

Si te interesa saber qué hice y por qué 👏👏👏



nahuelhds/votaciones-diputados-argentina

Información sistematizada y normalizada de las votaciones de la Cámara de Diputados de Argentina - nahuelhds/votaciones-diputados-argentina

[github.com](https://github.com/nahuelhds/vota...)



57



349



1,0K

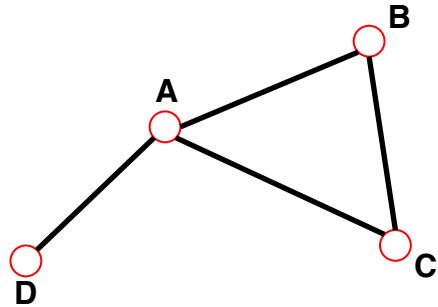


Sección 5

CAMINOS

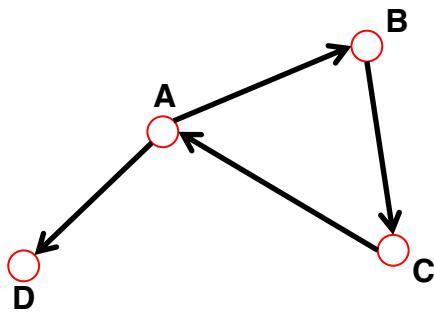
DISTANCIA EN UN GRAFO

Caminos más corto, camino geodésico



The *distance (shortest path, geodesic path)* between two nodes is defined as the number of edges along the shortest path connecting them.

*If the two nodes are disconnected, the distance is infinity.



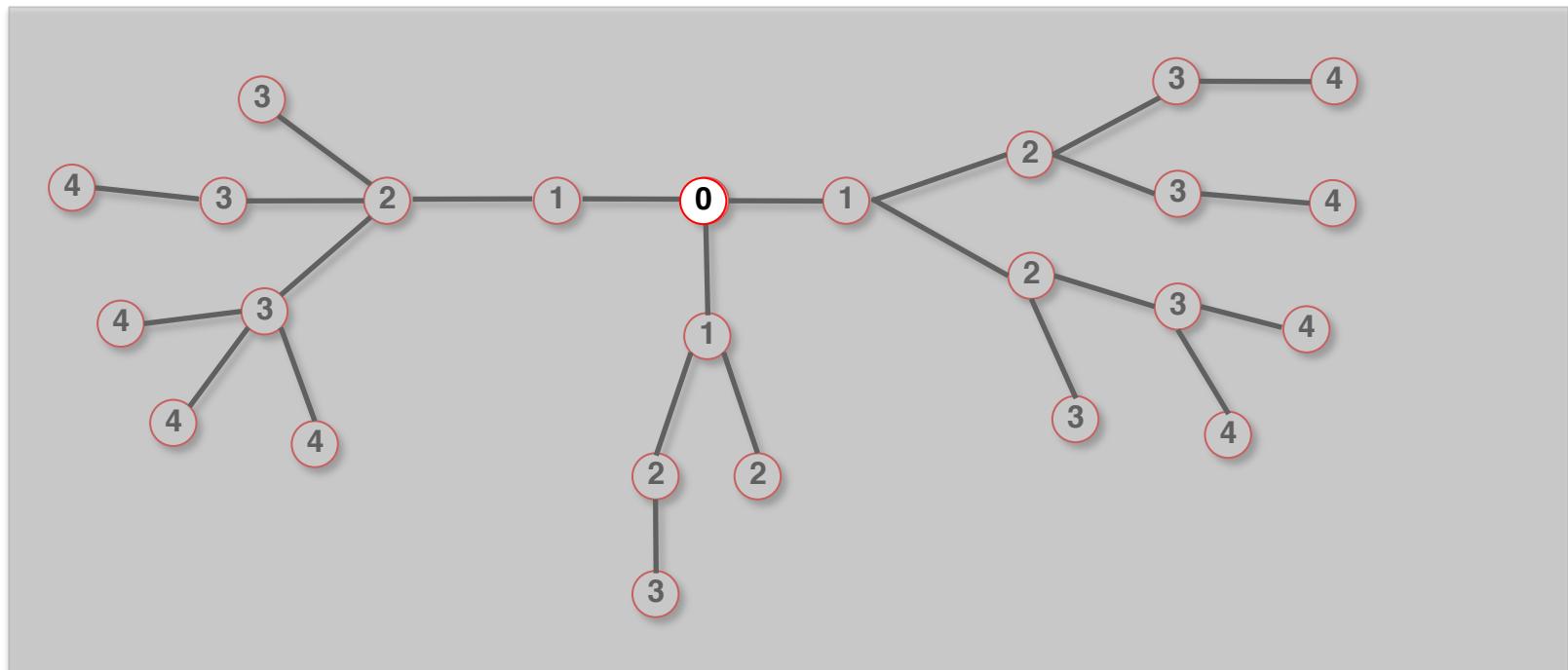
In *directed graphs* each path needs to follow the direction of the arrows.

Thus in a digraph the distance from node A to B (on an AB path) is generally different from the distance from node B to A (on a BCA path).

CALCULANDO DISTANCIAS: BREADTH FIRST SEARCH

Distance between node 0 and node 4:

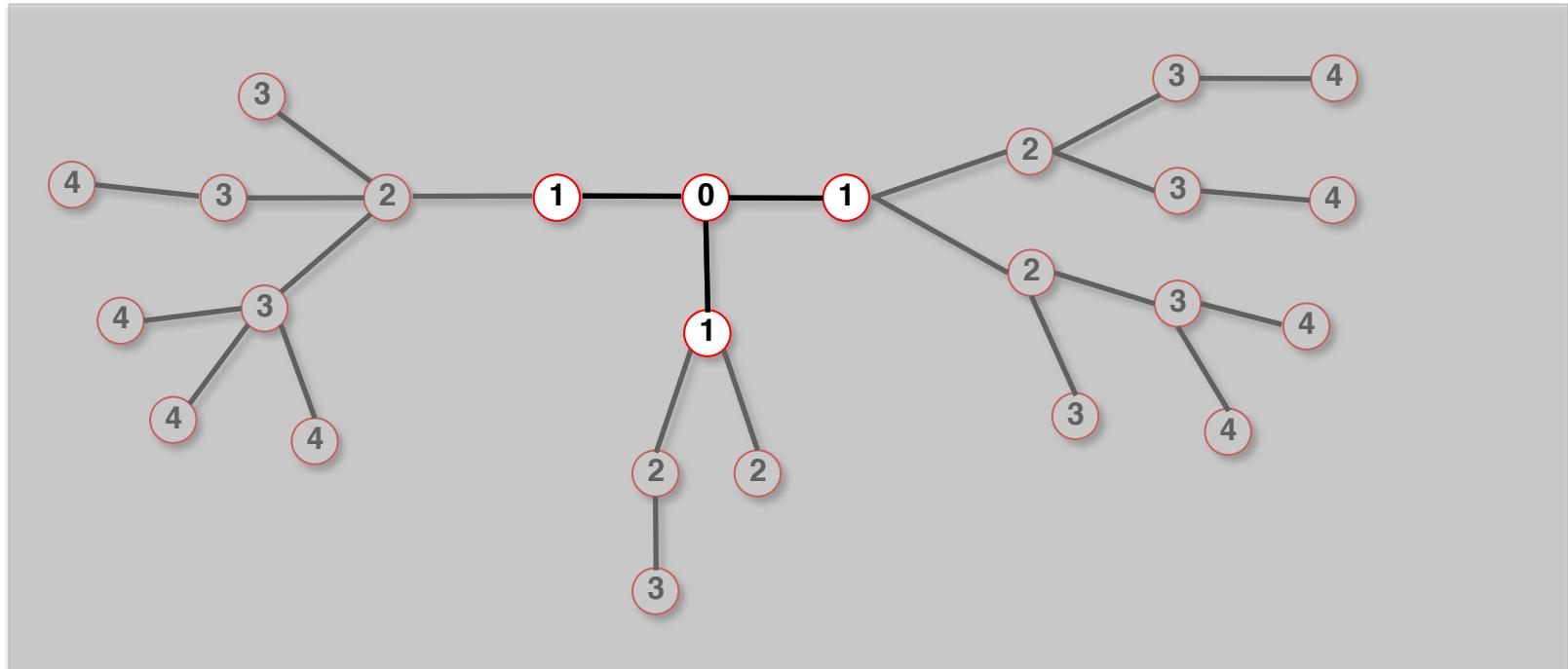
1. Start at 0.



CALCULANDO DISTANCIAS: BREADTH FIRST SEARCH

Distance between node 0 and node 4:

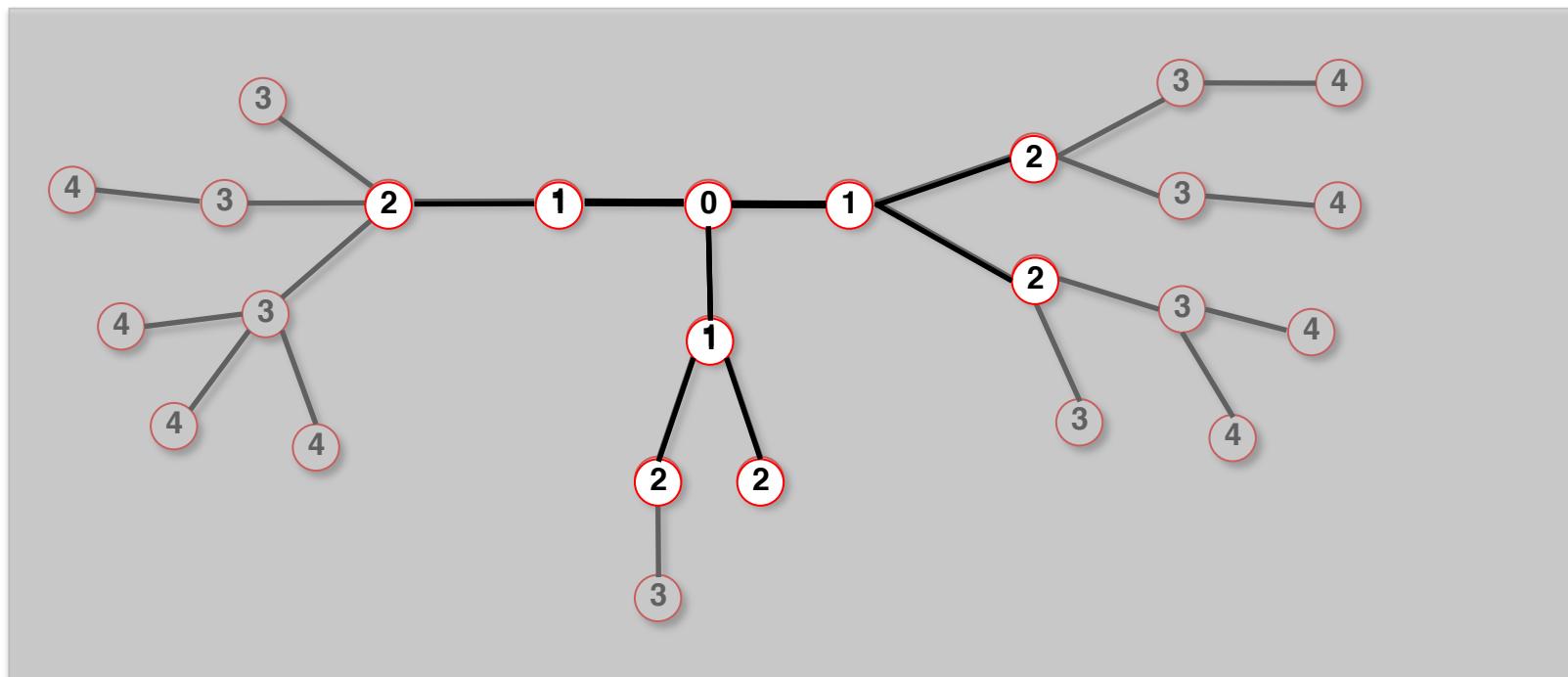
1. Start at 0.
2. Find the nodes adjacent to 1. Mark them as at distance 1. Put them in a queue.



CALCULANDO DISTANCIAS: BREADTH FIRST SEARCH

Distance between node 0 and node 4:

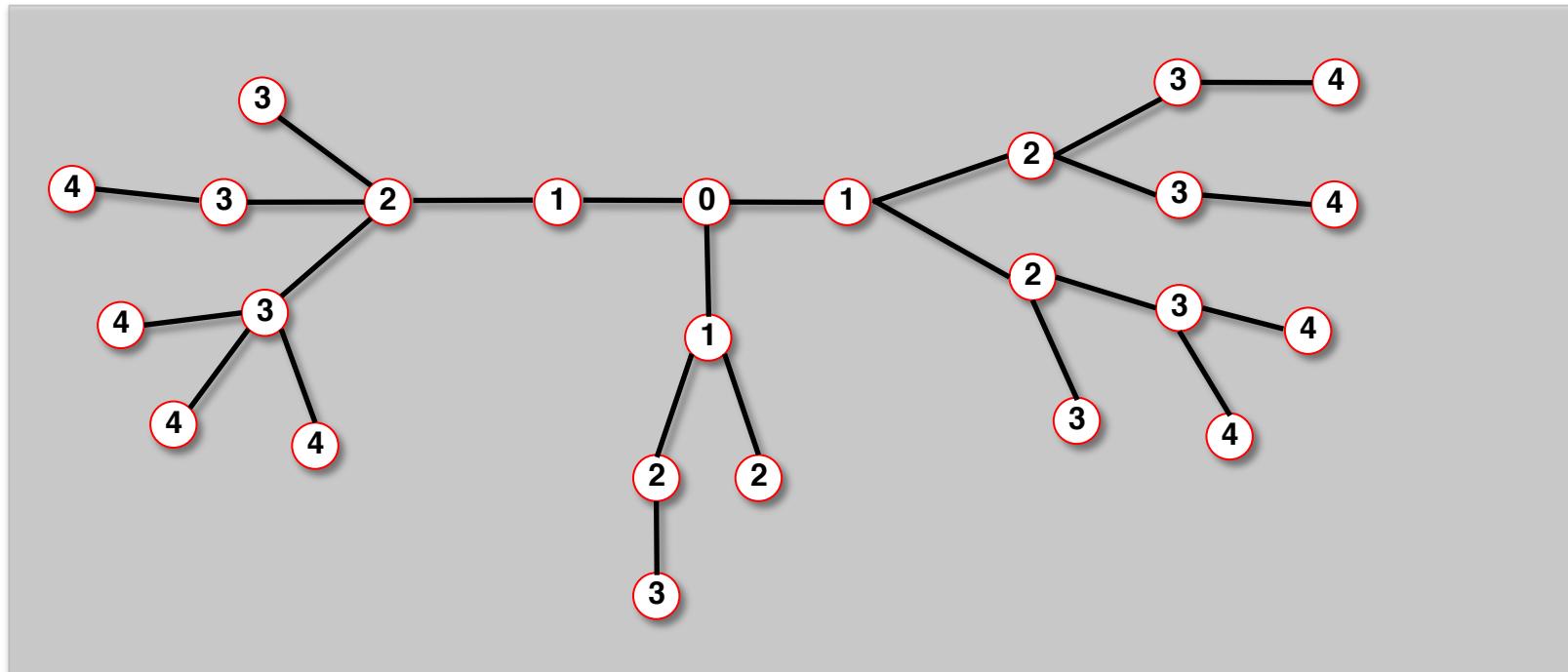
1. Start at 0.
2. Find the nodes adjacent to 0. Mark them as at distance 1. Put them in a queue.
3. Take the first node out of the queue. Find the unmarked nodes adjacent to it in the graph. Mark them with the label of 2. Put them in the queue.



CALCULANDO DISTANCIAS: BREADTH FIRST SEARCH

Distance between node 0 and node 4:

1. Repeat until you find node 4 or there are no more nodes in the queue.
2. The distance between 0 and 4 is the label of 4 or, if 4 does not have a label, infinity.



DIAMETRO DE LA RED Y DISTANCIA PROMEDIO

Diameter: d_{max} the maximum distance between any pair of nodes in the graph.

Average path length/distance, $\langle d \rangle$, for a **connected graph**:

where d_{ij} is the distance from node i to node j

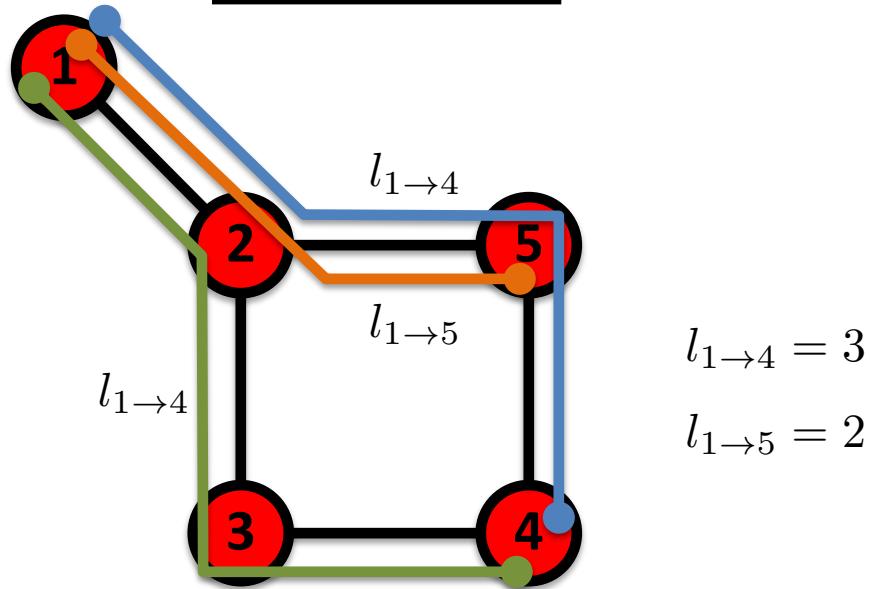
$$\langle d \rangle \equiv \frac{1}{2L_{\max}} \sum_{i,j \neq i} d_{ij}$$

In an *undirected graph* $d_{ij} = d_{ji}$, so we only need to count them once:

$$\langle d \rangle \equiv \frac{1}{L_{\max}} \sum_{i,j > i} d_{ij}$$

CAMINOS: RESUMEN

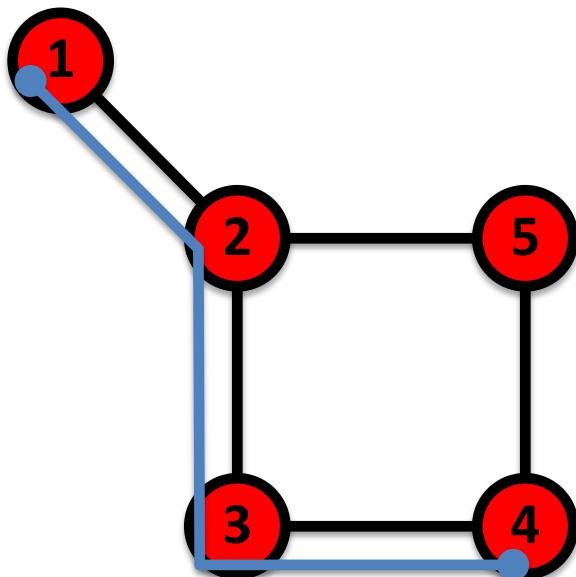
Shortest Path



The path with the shortest length between two nodes
(distance).

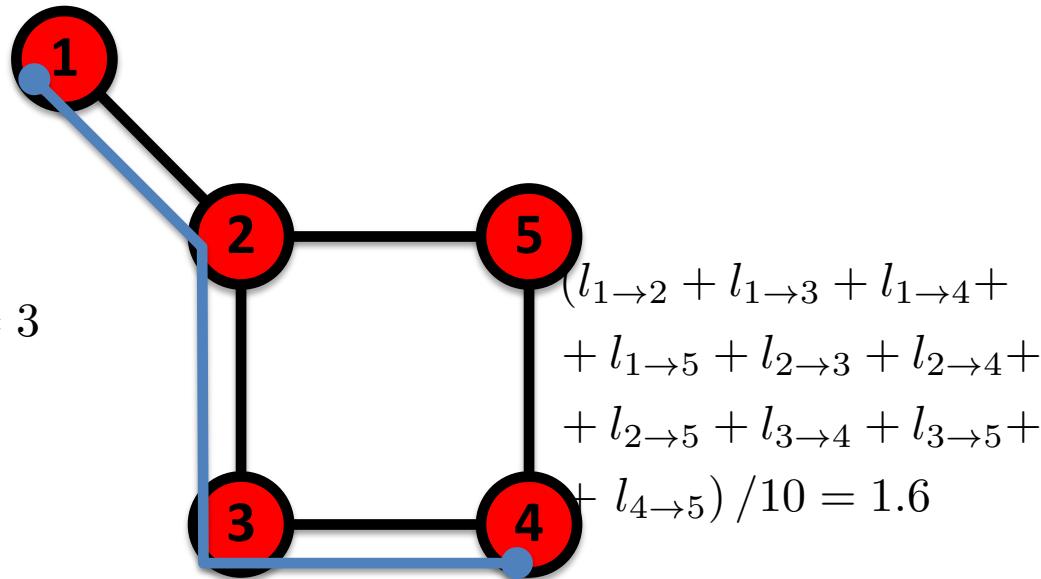
CAMINOS: RESUMEN

Diameter



$$l_{1 \rightarrow 4} = 3$$

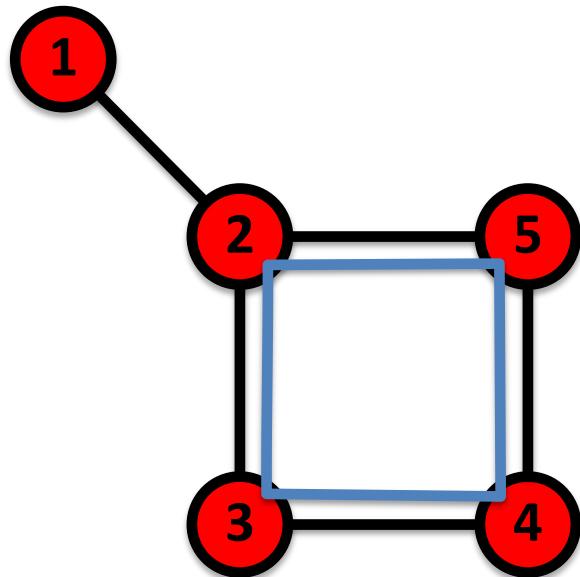
Average Path Length



The average of the shortest paths for all pairs of nodes.

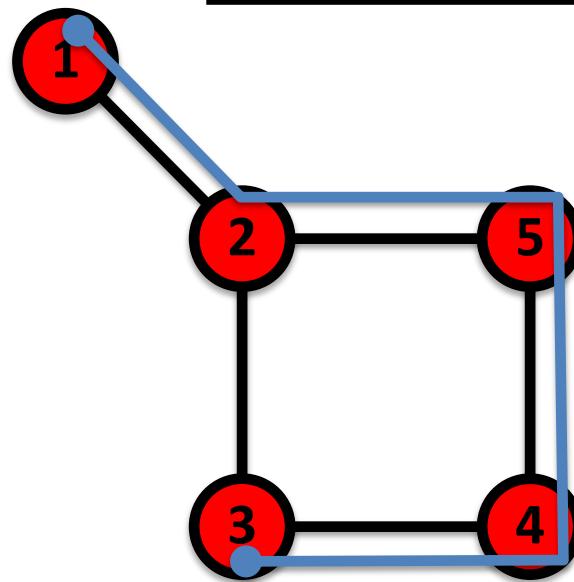
CAMINOS: RESUMEN

Cycle



A path with the same start
and end node.

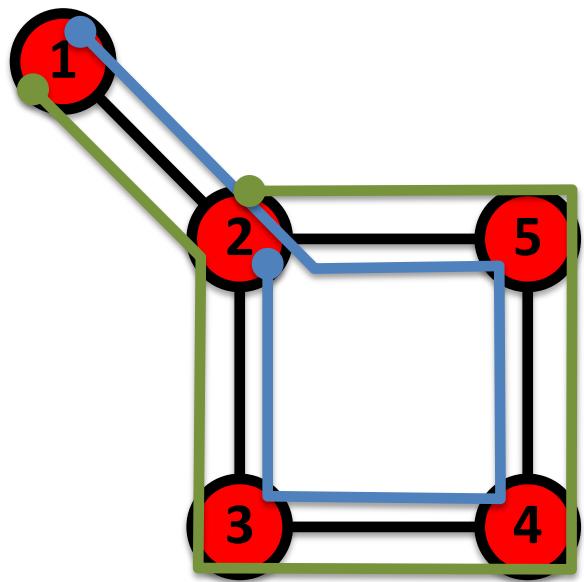
Self-avoiding Path



A path that does not intersect
itself.

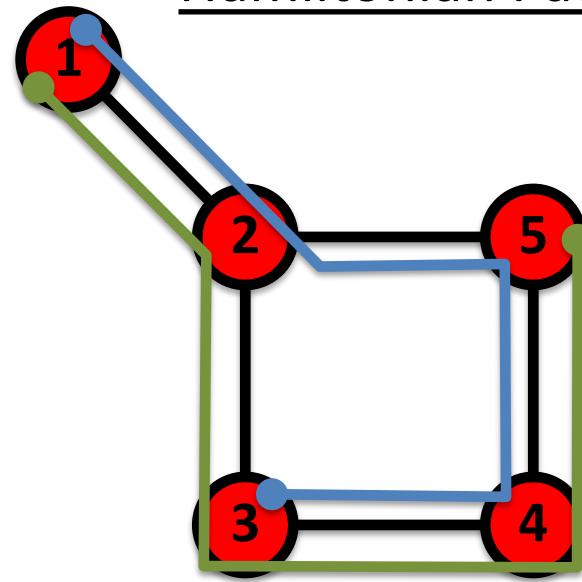
CAMINOS: RESUMEN

Eulerian Path



A path that traverses each link exactly once.

Hamiltonian Path



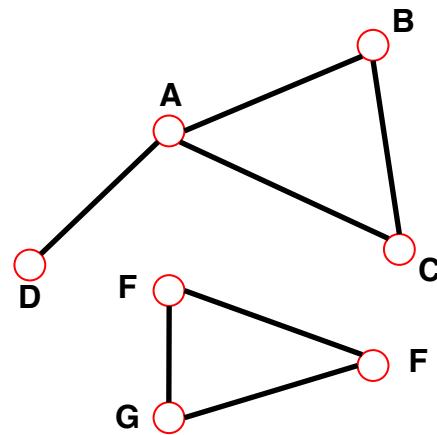
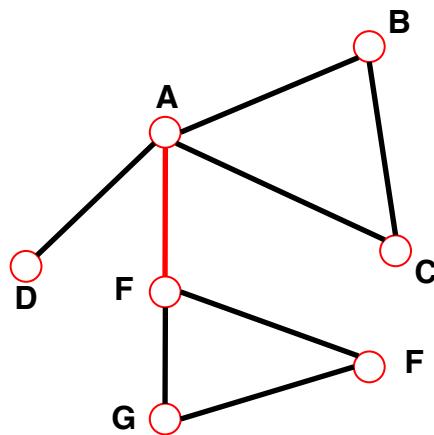
A path that visits each node exactly once.

Sección 6

CONECTIVIDAD

CONECTIVIDAD EN GRAFOS NO DIRIGIDOS

Connected (undirected) graph: any two vertices can be joined by a path.
A disconnected graph is made up by two or more connected components.



Largest Component:
Giant Connected Component

The rest: **Isolates**

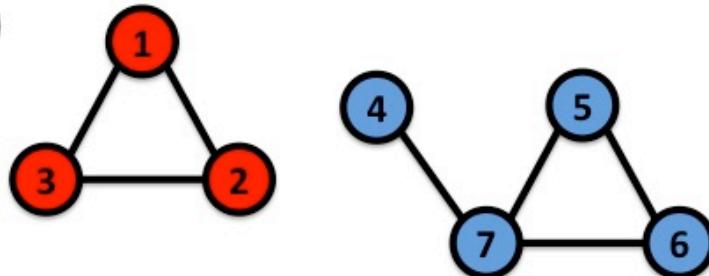
Bridge: if we erase it, the graph becomes disconnected.

CONECTIVIDAD EN GRAFOS NO DIRIGIDOS

Matriz de Adyacencia

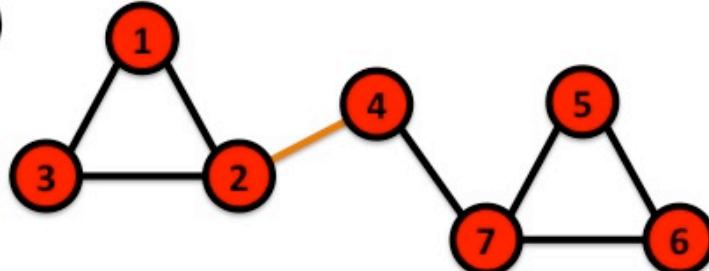
The adjacency matrix of a network with several components can be written in a block-diagonal form, so that nonzero elements are confined to squares, with all other elements being zero:

(a)



$$\begin{pmatrix} 0 & 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 0 \end{pmatrix}$$

(b)



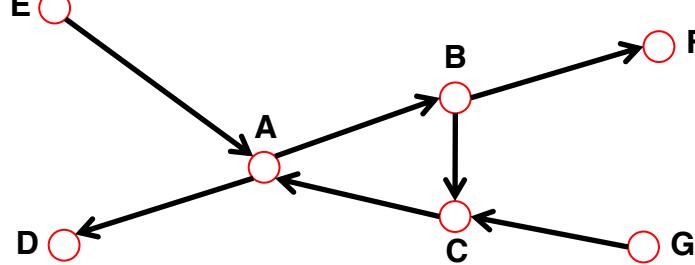
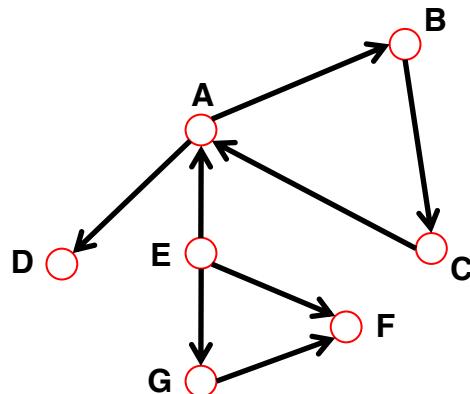
$$\begin{pmatrix} 0 & 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 0 \end{pmatrix}$$

CONECTIVIDAD EN GRAFOS DIRIGIDOS

Strongly connected directed graph: has a path from each node to every other node and vice versa (e.g. AB path and BA path).

Weakly connected directed graph: it is connected if we disregard the edge directions.

Strongly connected components can be identified, but not every node is part of a nontrivial strongly connected component.



Coeficiente de Clustering

COEFICIENTE DE CLUSTERING

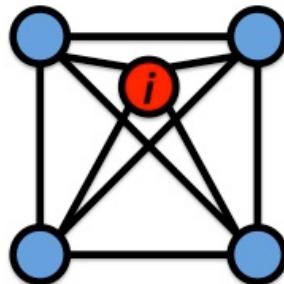
* Coeficiente de Clustering:

qué fracción de tus vecinos están conectados?

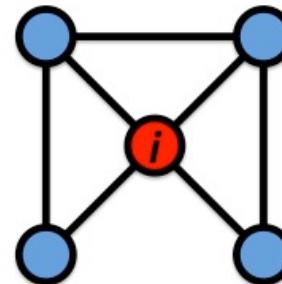
* Nodo i con grado k_i

* C_i en $[0,1]$

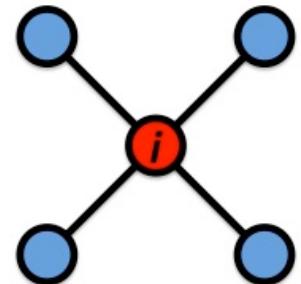
$$C_i = \frac{2e_i}{k_i(k_i - 1)}$$



$$C_i = 1$$



$$C_i = 1/2$$



$$C_i = 0$$

Watts & Strogatz, Nature 1998.

COEFICIENTE DE CLUSTERING

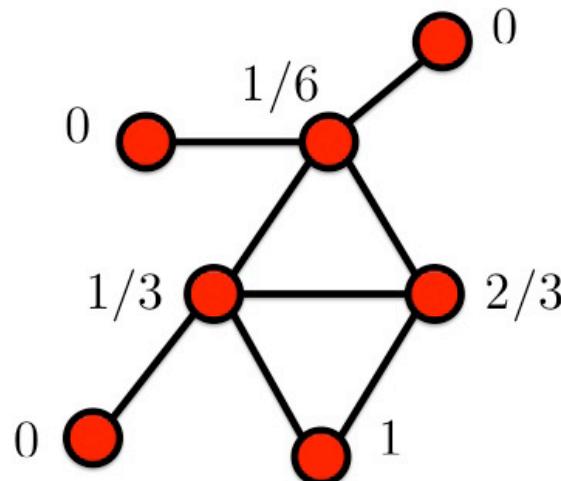
* Coeficiente de Clustering:

qué fracción de tus vecinos están conectados?

* Nodo i con grado k_i

* C_i en $[0,1]$

$$C_i = \frac{2e_i}{k_i(k_i - 1)}$$



$$\langle C \rangle = \frac{13}{42} \approx 0.310$$

$$C = \frac{3}{8} = 0.375$$

Watts & Strogatz, Nature 1998.

Sección 8

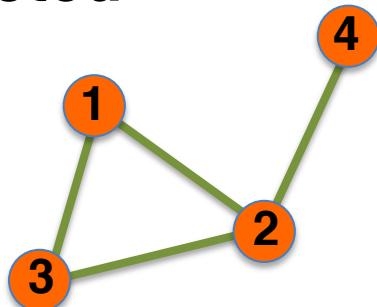
RESUMEN

TRES MÉTRICAS CENTRALES EN CIENCIA DE REDES

Distribución de grados: $P(k)$

Longitud de caminos: $\langle d \rangle$

Coeficiente de Clustering:
$$C_i = \frac{2e_i}{k_i(k_i - 1)}$$

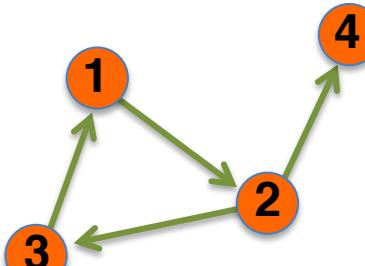
Undirected

$$A_{ij} = \begin{pmatrix} 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix}$$

$$A_{ii} = 0$$

$$L = \frac{1}{2} \sum_{i,j=1}^N A_{ij} \quad \langle k \rangle = \frac{2L}{N}$$

Actor network, protein-protein interactions

Directed

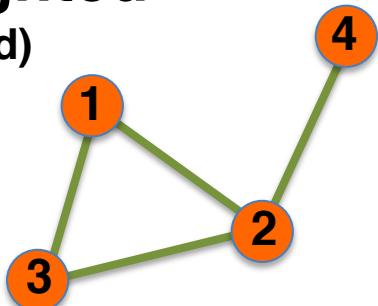
$$A_{ij} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$A_{ii} = 0$$

$$L = \sum_{i,j=1}^N A_{ij} \quad \langle k \rangle = \frac{L}{N}$$

WWW, citation networks

Unweighted (undirected)



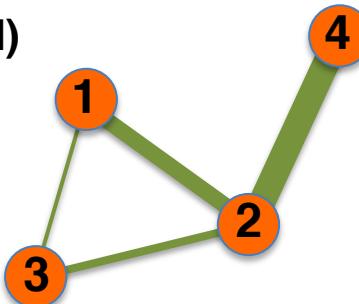
$$A_{ij} = \begin{pmatrix} 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix}$$

$$A_{ii} = 0$$

$$L = \frac{1}{2} \sum_{i,j=1}^N A_{ij} \quad \langle k \rangle = \frac{2L}{N}$$

protein-protein interactions, www

Weighted (undirected)



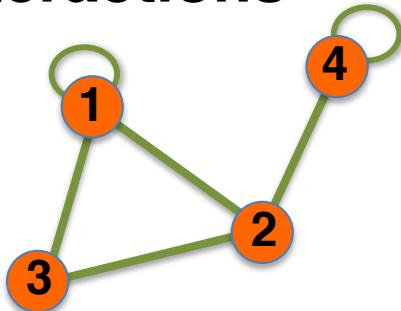
$$A_{ij} = \begin{pmatrix} 0 & 2 & 0.5 & 0 \\ 2 & 0 & 1 & 4 \\ 0.5 & 1 & 0 & 0 \\ 0 & 4 & 0 & 0 \end{pmatrix}$$

$$A_{ii} = 0$$

$$L = \frac{1}{2} \sum_{i,j=1}^N \text{nonzero}(A_{ij}) \quad \langle k \rangle = \frac{2L}{N}$$

Call Graph, metabolic networks

Self-interactions



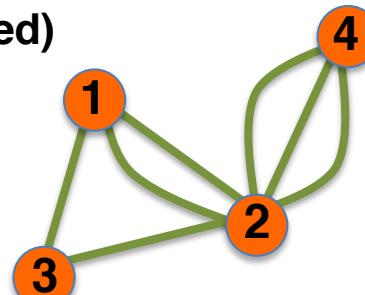
$$A_{ij} = \begin{pmatrix} 1 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 \end{pmatrix}$$

$$A_{ii} \neq 0 \quad A_{ij} = A_{ji}$$

$$L = \frac{1}{2} \sum_{i,j=1, i \neq j}^N A_{ij} + \sum_{i=1}^N A_{ii} \quad ?$$

Protein interaction network, www

Multigraph (undirected)



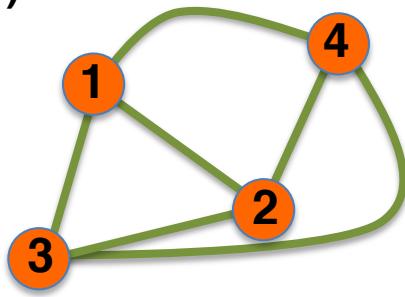
$$A_{ij} = \begin{pmatrix} 0 & 2 & 1 & 0 \\ 2 & 0 & 1 & 3 \\ 1 & 1 & 0 & 0 \\ 0 & 3 & 0 & 0 \end{pmatrix}$$

$$A_{ii} = 0 \quad A_{ij} = A_{ji}$$

$$L = \frac{1}{2} \sum_{i,j=1}^N \text{nonzero}(A_{ij}) \quad \langle k \rangle = \frac{2L}{N}$$

Social networks, collaboration networks

Complete Graph (undirected)



$$A_{ij} = \begin{pmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{pmatrix}$$

$$A_{ii} = 0$$

$$A_{i \neq j} = 1$$

$$L = L_{\max} = \frac{N(N-1)}{2} \quad \langle k \rangle = N - 1$$

Actor network, protein-protein interactions

