

On Growth and Form Networks

Ernesto Estrada

Department of Mathematics & Statistics University of Strathclyde Glasgow, UK

> ernesto.estrada@strath.ac.uk www.estradalab.org







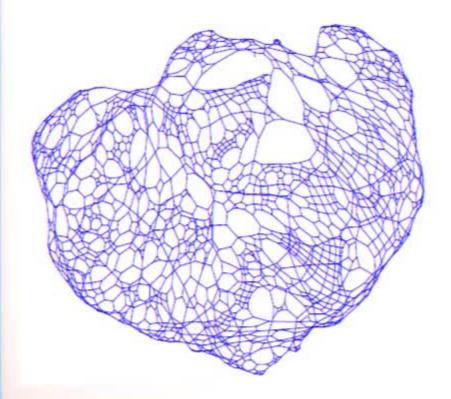
Motivation



How efficiently geometrically-embedded networks use the available space?

Urban street network

Protein residue network







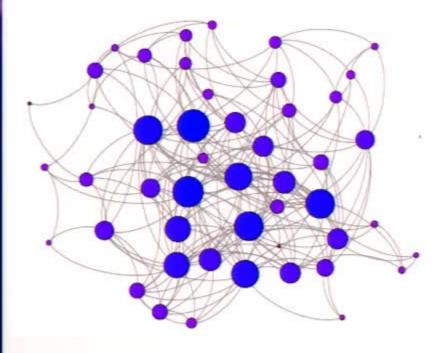
Motivation



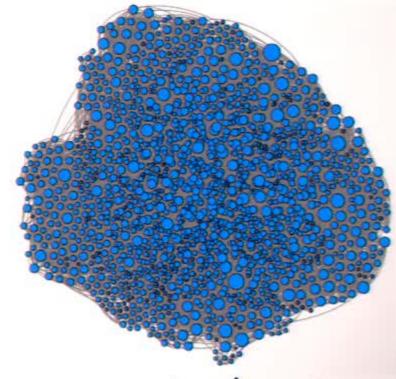
What about networks not embedded geometrically?

Food web

Social network







Social space?







"describes how much time, effort and cost a given arrangement produces (...) as compared to alternative arrangements".

"refers to the organization of physical assets (...) which structure the transportation, communication, (...) within the region and beyond".

Sarzynski & Levy, Spatial Efficiency and Regional Prosperity: A Literature Review and Policy. http://www.gwu.edu/~gwipp/SpatialEfficiencyWPAug16.pdf.



Spatial Efficiency Ingredients

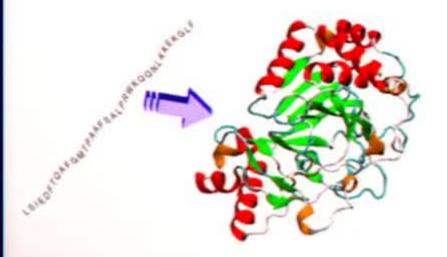


1) Planarity

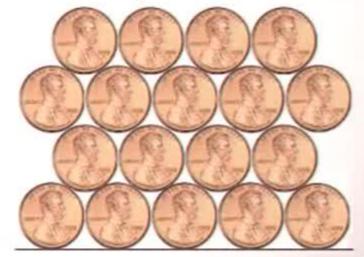




2) Foldeness



3) Packing







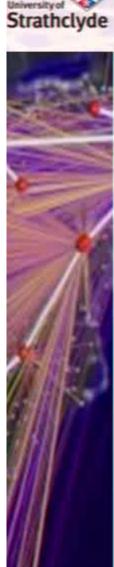


Definition 1. The spatial efficiency of a network is the efficiency in the communication between the nodes produced by a given spatial embedding of the network.

Definition 2. The communication efficiency of a network is the ratio of the amount of information 'successfully transferred' between every pair of nodes in the network to the amount of information 'lost' in the communication process.



Network Communicability



Definition 3. The communicability between two nodes in a network is defined as a function of the total number of walks connecting them, giving more weights to the shorter than to the longer ones.

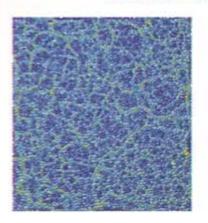
$$G_{pq} = \sum_{l=0}^{\infty} \frac{(A^l)_{pq}}{l!} = (e^A)_{pq} = \sum_{j=1}^{n} \psi_{j,p} \psi_{j,q} e^{\lambda_j}$$

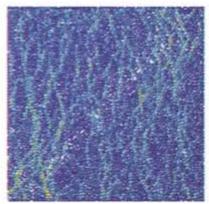


Communicability Applications

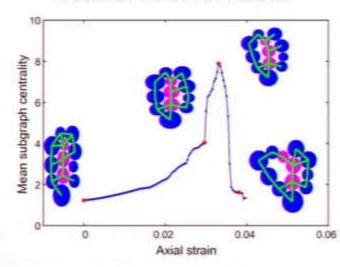


Granular materials



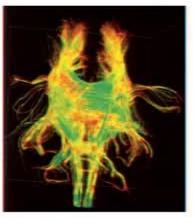


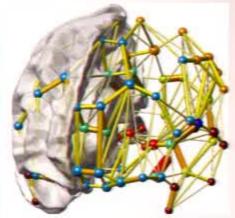
Effects of axial strain



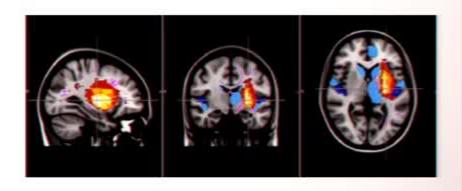
Walker, Tordesillas .: Int. J. Sol. Struct. 47 (2010) 624-629.

Brain networks





Effects of brain strokes



Crofts et al.: NeuroImage 54 (2011) 161-169.



Communication Efficiency



Consider that two nodes p and q are trying to communicate with each other by sending information both ways through the network.

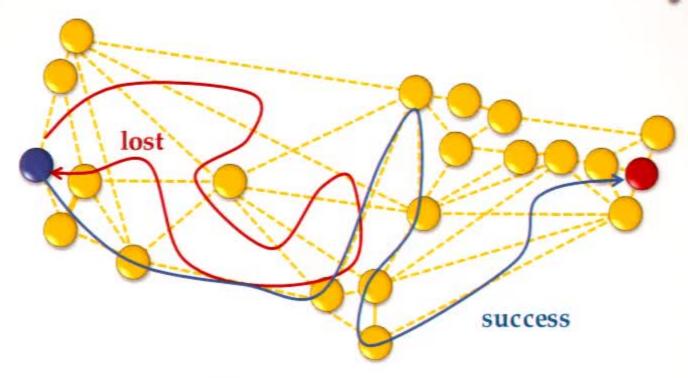
We know that:

Quantifies the amount of information that is sent by p, wanders G_{pp} around the network, and returns to the origin, the node p. i.e., it is the amount of lost information.

Quantifies the amount of information that is sent by p, wanders G_{pq} around the network, and arrives at its destination, the node q, i.e., it is the amount of successfully delivered information.



Communication Efficiency



$$\xi_{pq}^{def} = G_{pp} + G_{qq} - 2G_{pq}$$

$$\gamma_{pq} = \frac{G_{pq}}{\sqrt{G_{pp}G_{qq}}}$$







Theorem 1. The function ξ_{pq} is a squared Euclidean distance between the nodes p and q in the network.

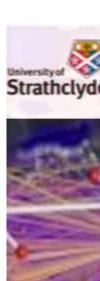
Theorem 2. The function γ_{pq} is the cosine of the Euclidean angle spanned by the position vectors of the nodes p and q.

Estrada: Lin. Algebra Appl. 436 2012 4317-4328

Estrada & Hatano: SIAM Rev 2015 accepted, Arxiv (2014) 1412.7388.



Communicability Geometry



Theorem 3. The communicability distance induces an embedding of a network into an (n-1)-dimensional Euclidean sphere of radius:

$$R^2 = \frac{1}{4} \left(c - \frac{(2-b)^2}{a} \right)$$

$$a = \vec{1}^T e^{-A} \vec{1}$$

$$b = \vec{s}^T e^{-A} \vec{1}$$

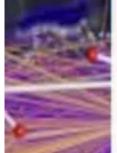
$$c = \vec{s}^T e^{-A} \vec{s}$$

Remark. The communicability distance matrix C is circum-Euclidean:

$$C = \vec{s} \vec{1}^T + \vec{1} \vec{s}^T - 2e^A \qquad \vec{s} = diag(e^A)$$







Example:

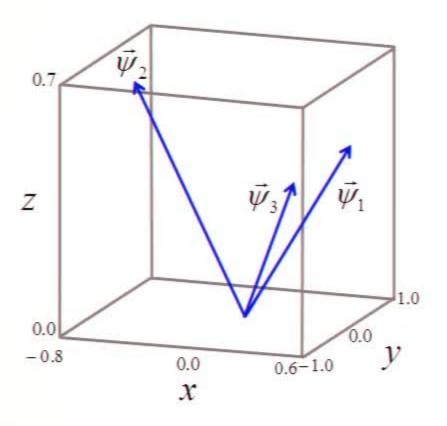


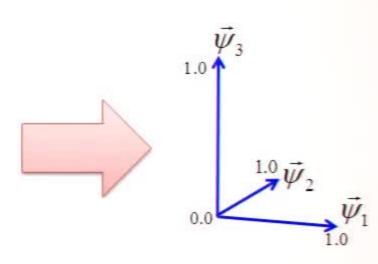
$$A\vec{\psi}_j = \lambda_j \vec{\psi}_j$$

$$\vec{\psi}_1 = \begin{bmatrix} 1/2 \\ 1/\sqrt{2} \\ 1/2 \end{bmatrix} \qquad \vec{\psi}_2 = \begin{bmatrix} -1/\sqrt{2} \\ 0 \\ 1/\sqrt{2} \end{bmatrix} \qquad \vec{\psi}_3 = \begin{bmatrix} 1/2 \\ -1/\sqrt{2} \\ 1/2 \end{bmatrix}$$













$$U = \begin{bmatrix} \psi_{1,1} & \psi_{2,1} & \cdots & \psi_{n,1} \\ \psi_{1,2} & \psi_{2,2} & \cdots & \psi_{n,2} \\ \vdots & \vdots & \vdots & \vdots \\ \psi_{1,n} & \psi_{2,n} & \cdots & \psi_{n,n} \end{bmatrix} \qquad \Lambda = \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{bmatrix}$$

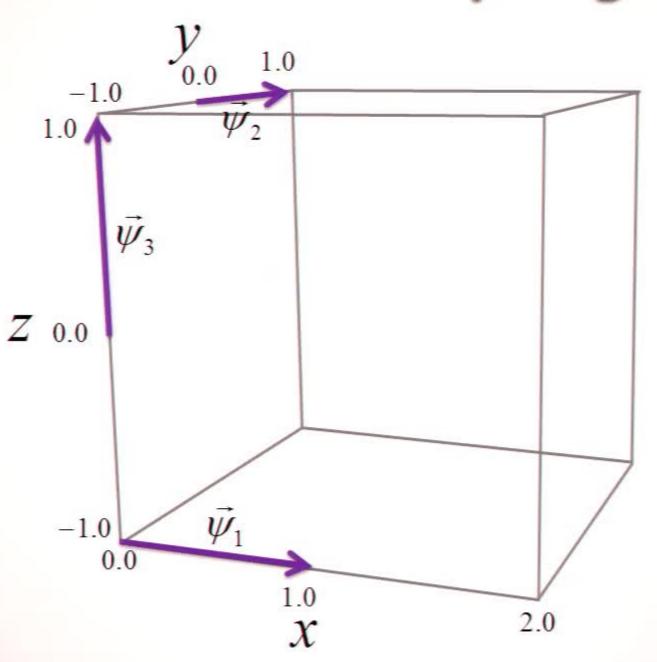
$$\Lambda = \begin{pmatrix}
\lambda_1 & 0 & \cdots & 0 \\
0 & \lambda_2 & \cdots & 0 \\
\vdots & \vdots & \ddots & \vdots \\
0 & 0 & \cdots & \lambda_n
\end{pmatrix}$$

$$\vec{\phi}_p = \left[\psi_{1,p} \cdots \psi_{\mu,p} \cdots \psi_{n,p} \right]^T$$

$$\vec{x}_p = \exp(\Lambda/2)\vec{\phi}_p$$







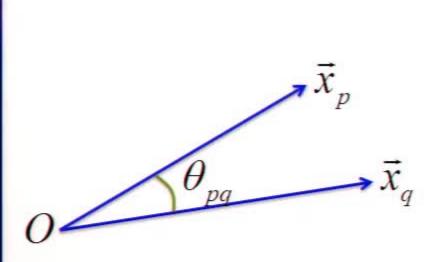




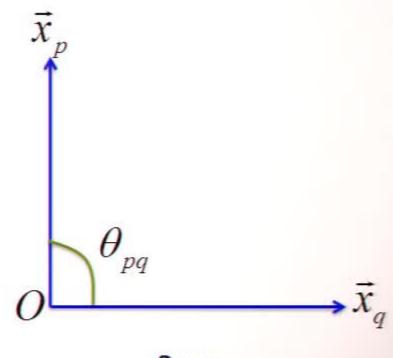
 $\langle heta
angle$: average communicability angle

For simple, unweighted, undirected networks:

$$0^{\circ} \le \theta_{pq} \le 90^{\circ}$$



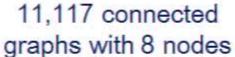
Good spatial efficiency

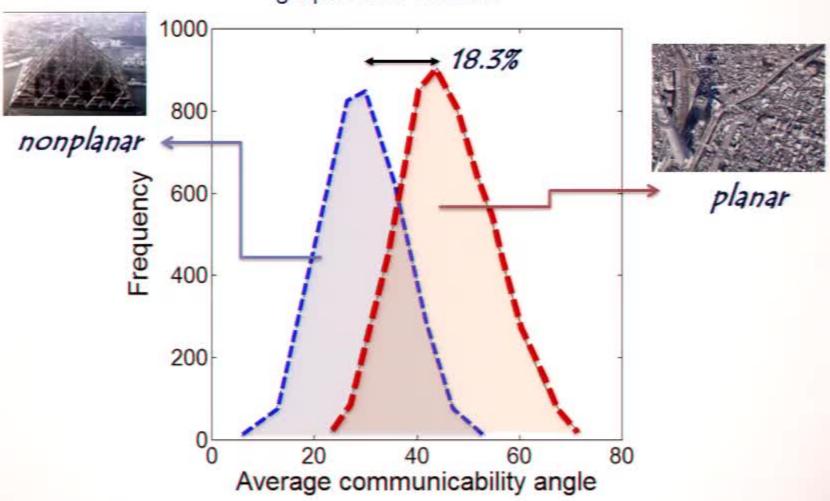


Poor spatial efficiency



$\langle \theta \rangle$ and Planarity



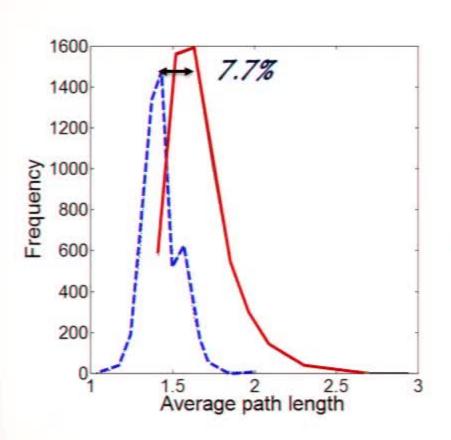


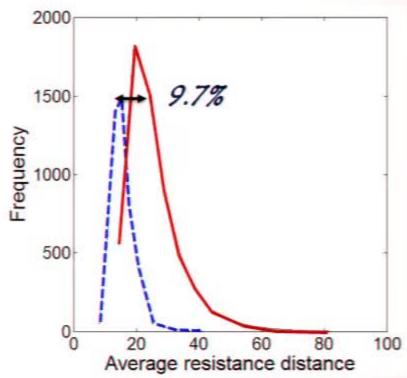
Estrada & Hatano: SIAM Rev. (2015), accepted, Arxiv (2014) 1412.7388.





11,117 connected graphs with 8 nodes



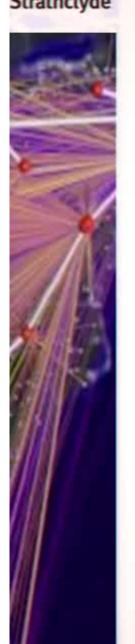


For the 261,080 connected graphs with 9 nodes the percentages of variation are 23.2% for the angles, 9.4% for path length and 10.7% for resistance.

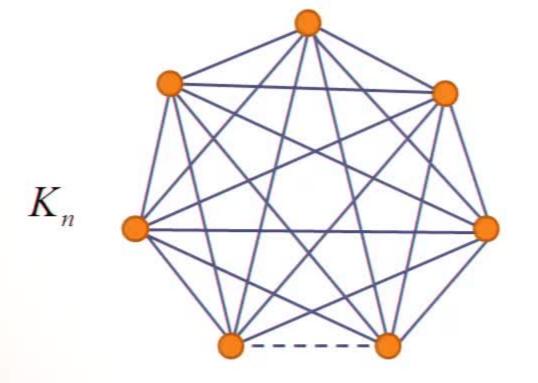


$\langle heta angle$ and Spatial Efficiency





$$\lim_{n\to\infty} \langle \theta(P_n) \rangle = 90^{\circ}$$



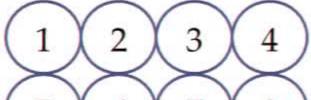
$$\lim_{n\to\infty} \langle \theta(K_n) \rangle = 0^\circ$$

Estrada & Hatano: SIAM Rev. (2015), accepted, Arxiv (2014) 1412.7388.



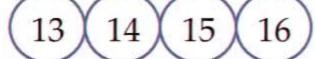
$\langle heta angle$ and Spatial Efficiency

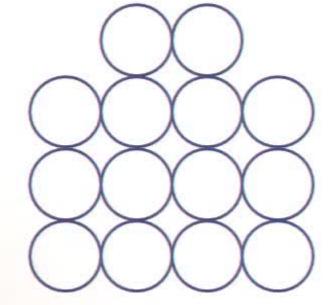
3) Packing

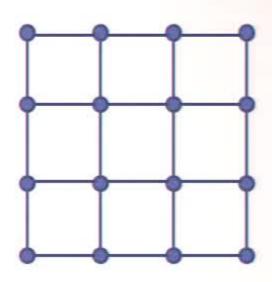


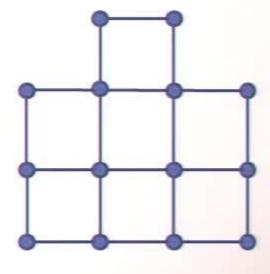








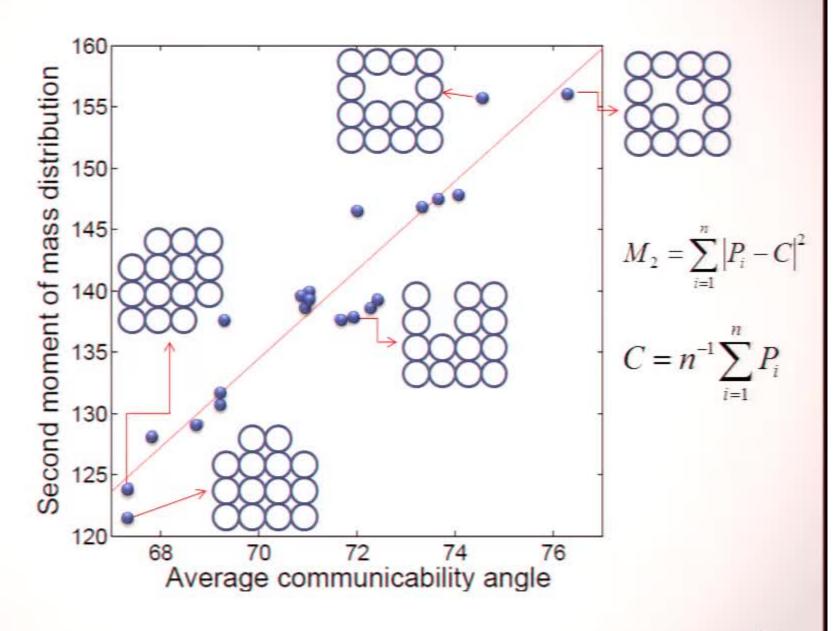






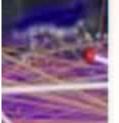
$\langle heta angle$ and Spatial Efficiency



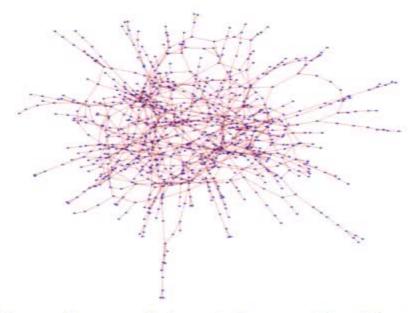








Erdős-Rényi Random Graphs



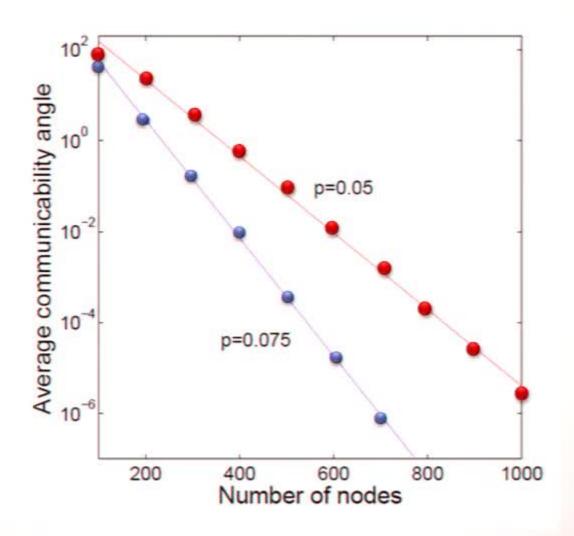
Theorem 4. Let $G_{n,p} = G(n,p)$ be an Erdős-Rényi graph. Let p and q be a pair of nodes in G(n,p). Then,

$$\lim_{n\to\infty} \left\langle \theta_{pq} \left(G_{n,p} \right) \right\rangle = 0^{\circ} \quad \forall p,q \in V$$



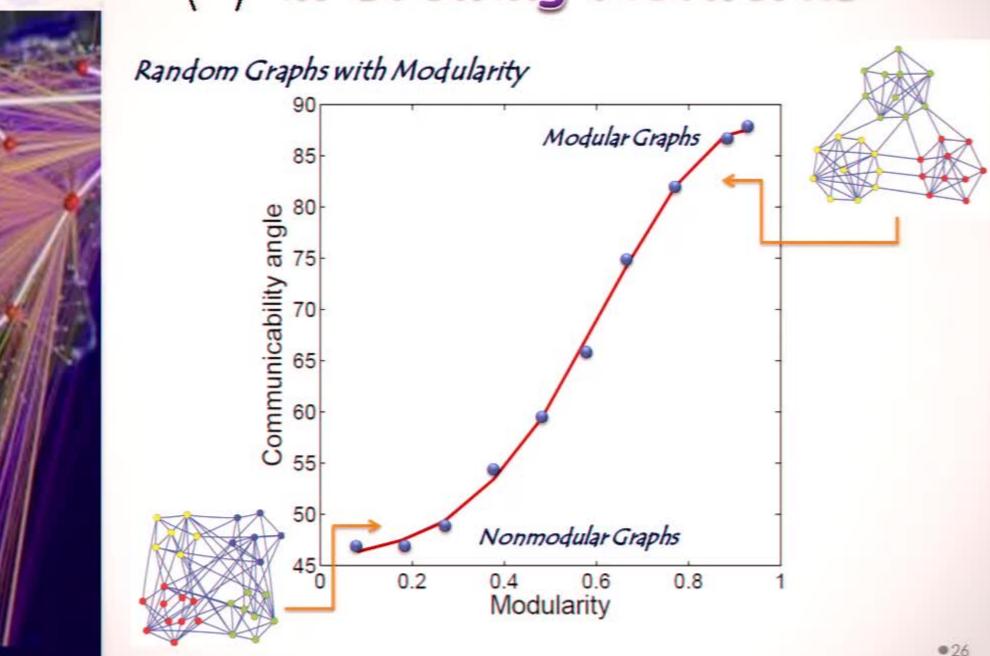
$\langle heta angle$ in Growing Networks

Erdős-Rényi Random Graphs





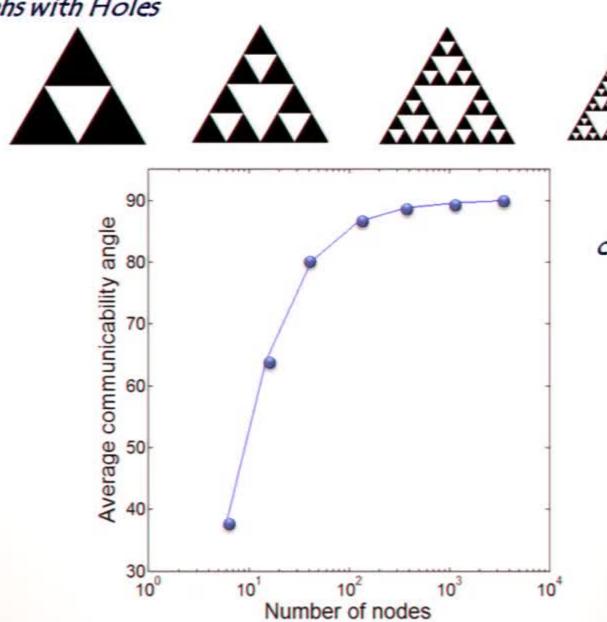
$\langle heta angle$ in Growing Networks





$\langle \theta \rangle$ in Growing Networks

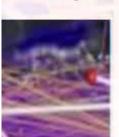
Graphs with Holes



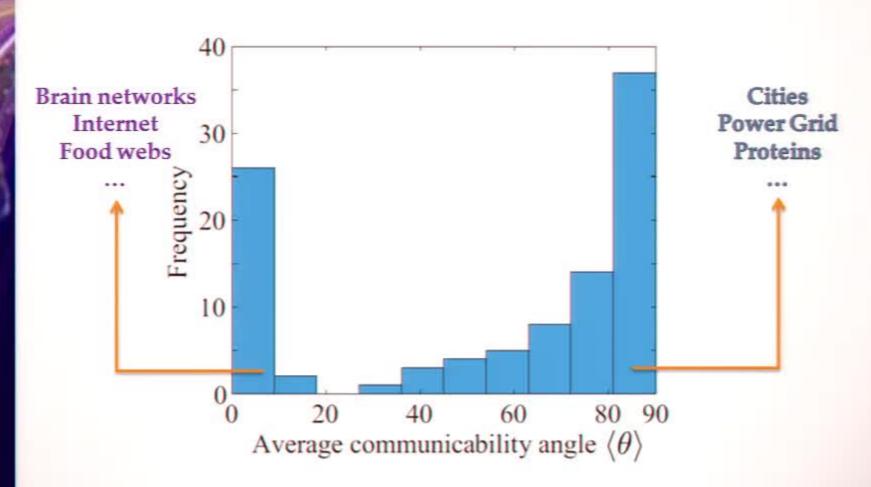


chordless cycle



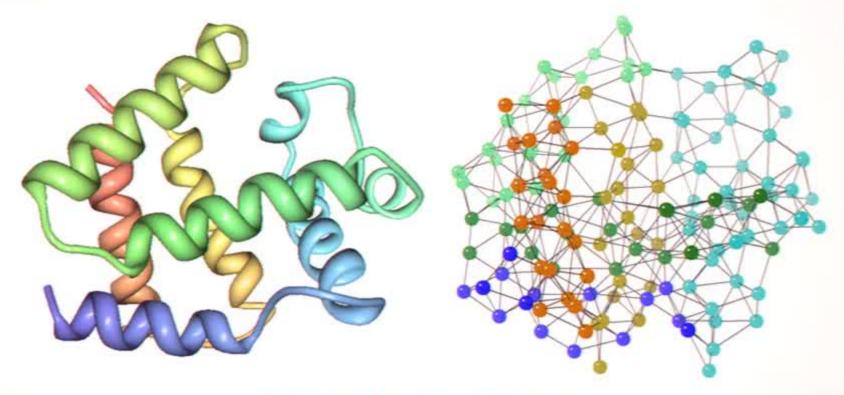


We studied 120 real-world networks representing biological, ecological, infrastructural, social, and technological systems.







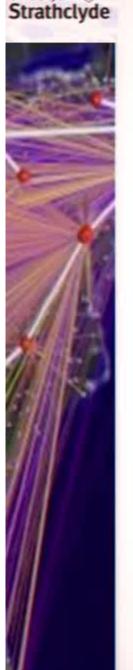


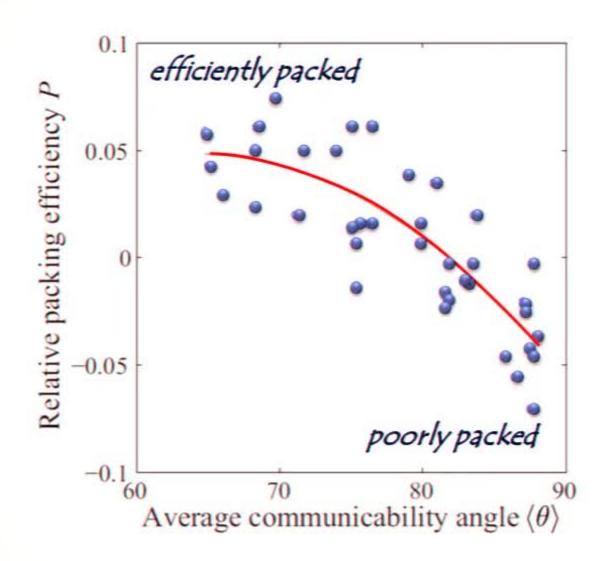
Relative Packing Efficiency

$$P = \frac{V_e - V_o}{V_e}$$

 V_e expected volume from ideal 3D structure of the protein V_o observed volume from X-rays crystal of the protein

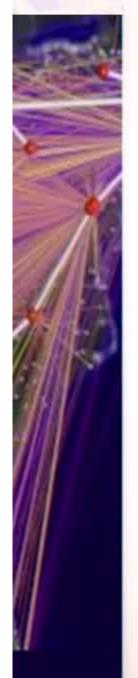


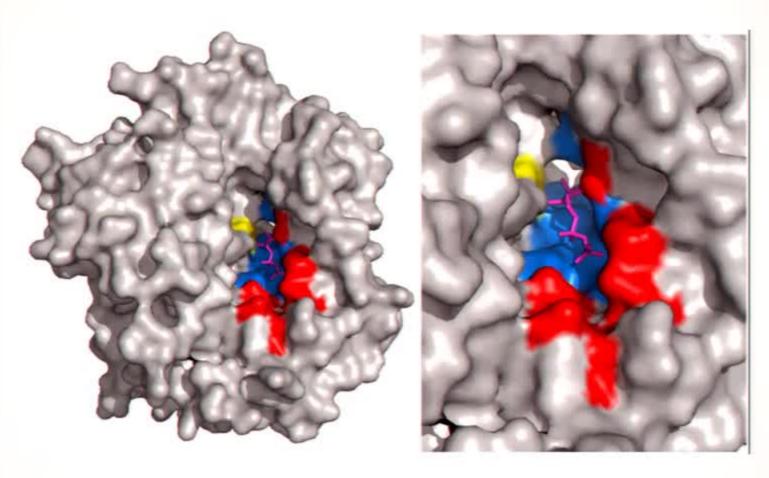






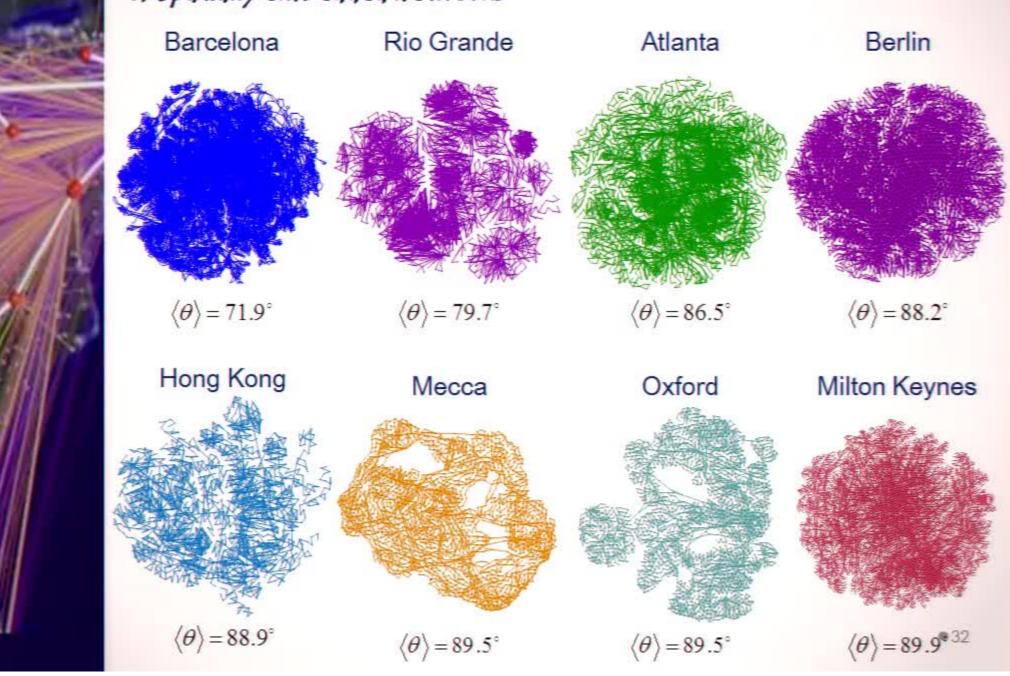
Proteins are Spongy







1) Spatially embedded networks







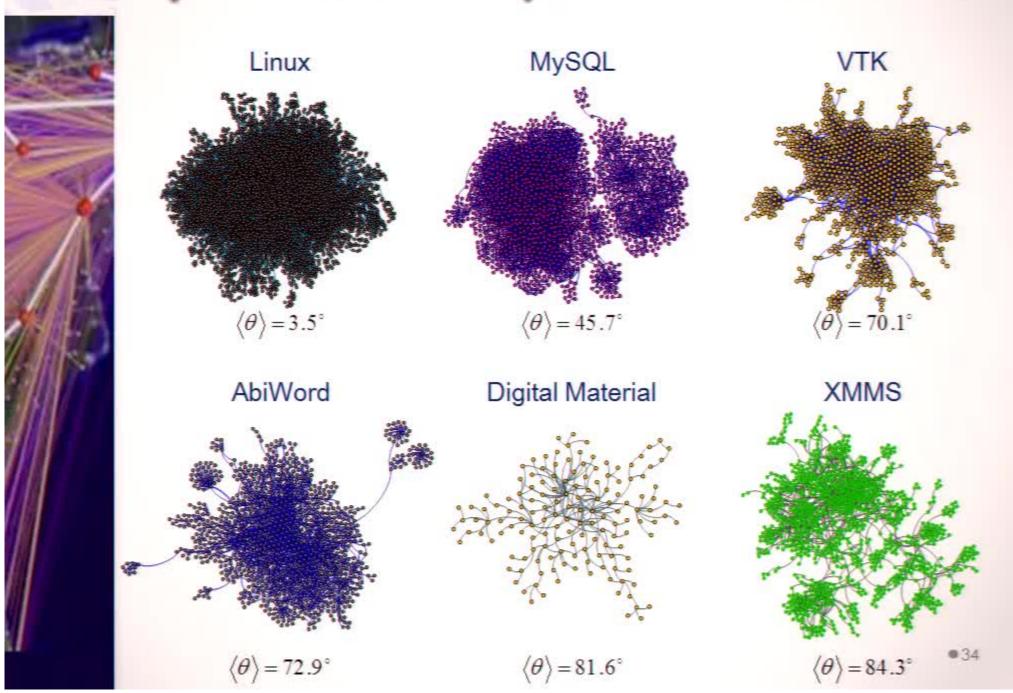
2) Non geographically embedded networks

Class Collaboration Graph

```
class A {
    // definition of class A
class B {
    A* ab;
    // rest of definition of class B
class C {
    A* ac;
     B* bc:
    // rest of definition of class C
                                       D-A
class D: public C {
    A* ad;
                                          D
    // rest of definition of class D
};
```



Strathclyde Spatial Efficiency in the Real-World





Conclusions



- Spatial efficiency is a fundamental (multifactorial) property of complex networks.
- Network spatial efficiency refers to the average quality of communication among the nodes.
- The communication goodness is quantified as the ratio of the amount of information successfully delivered to its destination to the one which is frustrated in its delivery and returned to their originators.
- We have provided analytical and empirical pieces of evidence which reaffirm the idea that the spatial efficiency accounts for fundamental structural characteristics of complex networks.