

# Statistical Analysis of Networks

## Lecture 1 - Course Introduction

---

DOMENICO DE STEFANO [ddestefano@units.it](mailto:ddestefano@units.it)

SUSANNA ZACCARIN [susanna.zaccarin@deams.units.it](mailto:susanna.zaccarin@deams.units.it)

MARCH 6 2024



# Course schedule

---

- Wednesday 9 - 11 am
- Thursday 13 – 15 pm
- Friday 15 – 17 pm

## Office hours:


D. De Stefano, **Tuesday 14.30 – 16.30 pm**, Building A, room 432

S. Zaccarin, **Thursday 11 am - 13 pm**, Building D, room 2.15

(write an-email for other days/hours)

## Next lecture: **March 8**

Course organization combines basic theoretical concepts with software and practical applications with R



# COURSE SYLLABUS

---

- Basic analytical concepts in network analysis
- Network data collection
- Network visualization
- Descriptive analysis and network indices
- Network decomposition
- Modeling network structure

The course will include practical examples and hands-on computer laboratories based on the analysis of real-life relational data.

In the laboratories, the emphasis will be on the analysis of social networks in structured social and economic settings such as, for example, business companies, and organizations.

Software tools in R



# TEXTBOOKS

---

1. Kolaczyk E.D. (2009) Statistical Analysis of Network Data. Methods and Models, Springer, New York.
2. Lusher D., Koskinen J. and Robins G. (eds.) (2013) Exponential random graph models for social networks: Theory, methods, and applications. Cambridge University Press (selected chapters).

## *Recommended readings:*

3. Hanneman R.A. and Riddle M. (2005) Introduction to social network methods. Riverside, CA: University of California, Riverside (published in digital form at <http://faculty.ucr.edu/~hanneman/>).

Additional materials, lecture notes and information will be available at the course web page and via moodle2 e-learning platform.



# COURSE PREREQUISITES

---

Students are required to have basic knowledge in inferential statistics and should be familiar with linear and logistic regression models.

Basic knowledge of software R will be also required.



# EXAM


---

The final exam will consist in the presentation and discussion, in groups of 2/3 students, on the analysis (description and model fitting) of a real network dataset, explaining the working steps and the obtained results.

The writing of a short report is also requested, with the commented R code.

During the presentations, few questions will be asked to assess the individual contributions and preparation on the topics of the course.

During the course, homeworks and discussion on specific topics will be proposed.




## SOFTWARE TO BE INSTALLED FOR NEXT WEEK CLASSES

---

R packages:

- statnet (suite including several useful NA packages)
- igraph

Other packages will be included in the R script provided for the upcoming R labs



# WHAT IS A NETWORK?

---


*a collection of interconnected things*

**things:** individuals, blogs, digital devices, genes, organizations, academic papers, words, ...

**interconnected:** one (or more) relations among them

Relations can be of any kind (in general, to be defined):

contacts - acquaintance, friendship, contagion - blogger interactions, wired or wireless communications, regulatory behaviour of genes, collaboration in management, co-citations, semantic links, ...





# WHAT IS A NETWORK?

---

A (conceptual and formal) model to represent a set of connections, tangible or ideal among objects (entities/ units)

that can be applied in many fields: physics, chemistry, computer science, sociology, economics, etc.

- phenomena exhibiting a relational nature can be properly formalized by network(s)

Examples in the next slides

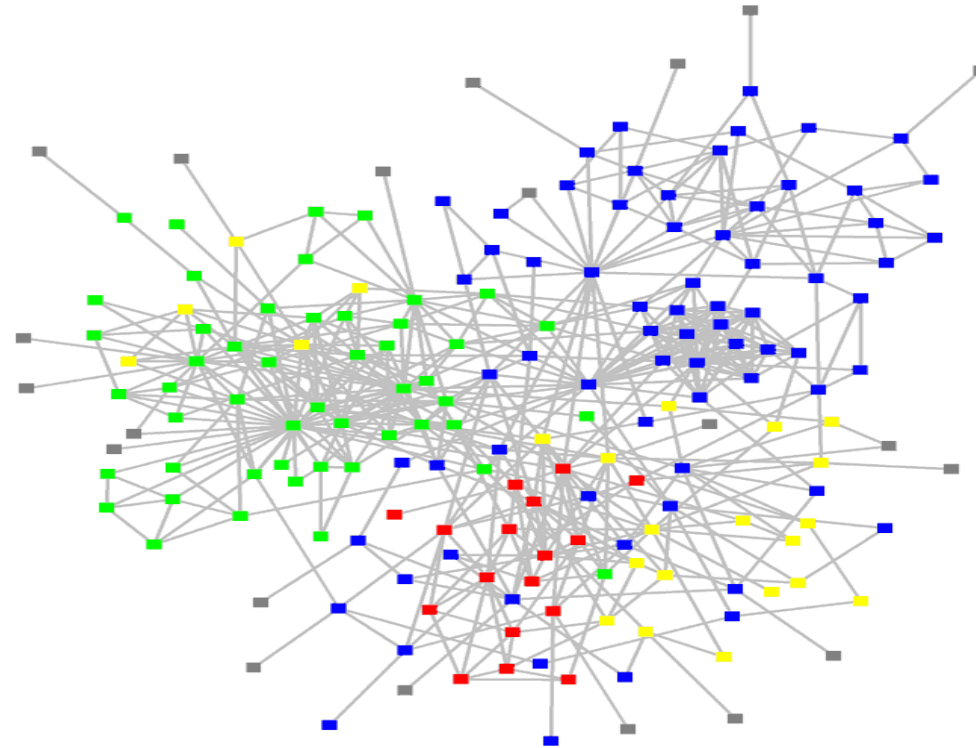


The "Social Graph" behind Facebook (2017)



# Structure of an organization

(note graph visualization!)



- ■ ■ : departments
- : consultants
- : external experts

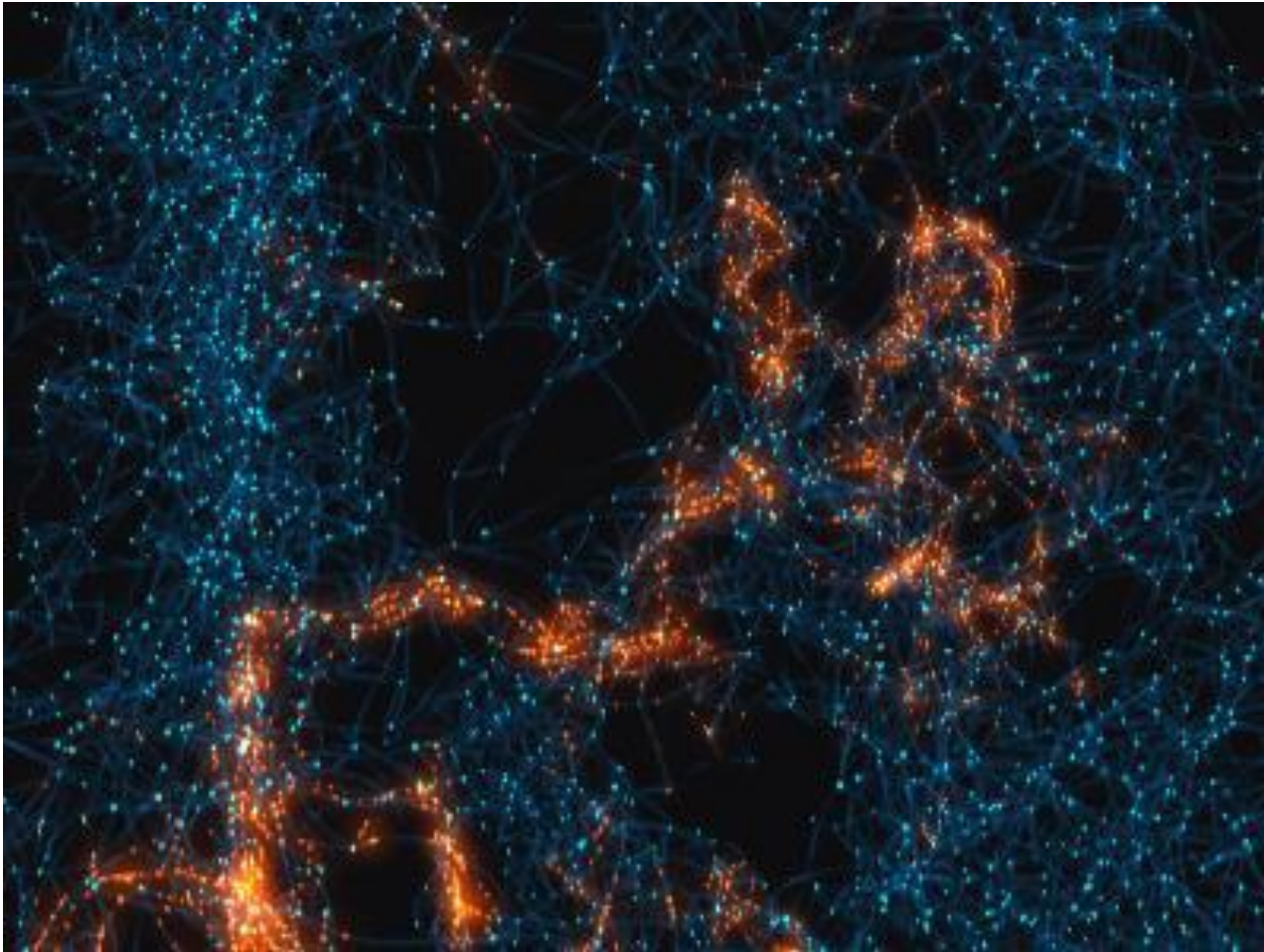
[www.orgnet.com](http://www.orgnet.com)







## Neuron network of brain



<http://nxxcxx.github.io/Neural-Network/>

Red areas (nodes and links) indicate brain activity (simulation)

# US biotech-industry – business relations in 1988 - 1991

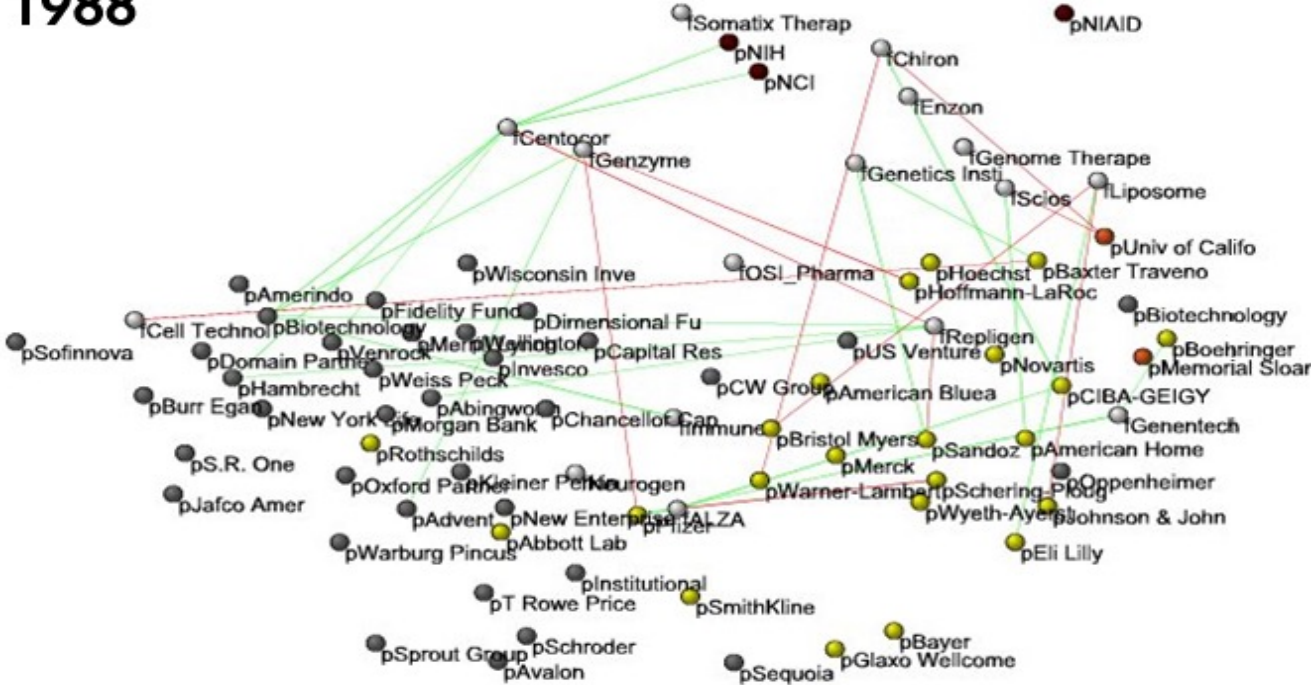
1988

**Nodes:**

- Companies
- Investment
- Pharma
- Research Labs
- Public
- Biotechnology

**Links:**

- Collaborations
- Financial
- R&D



# US biotech-industry – business relations in 1988 - 1991

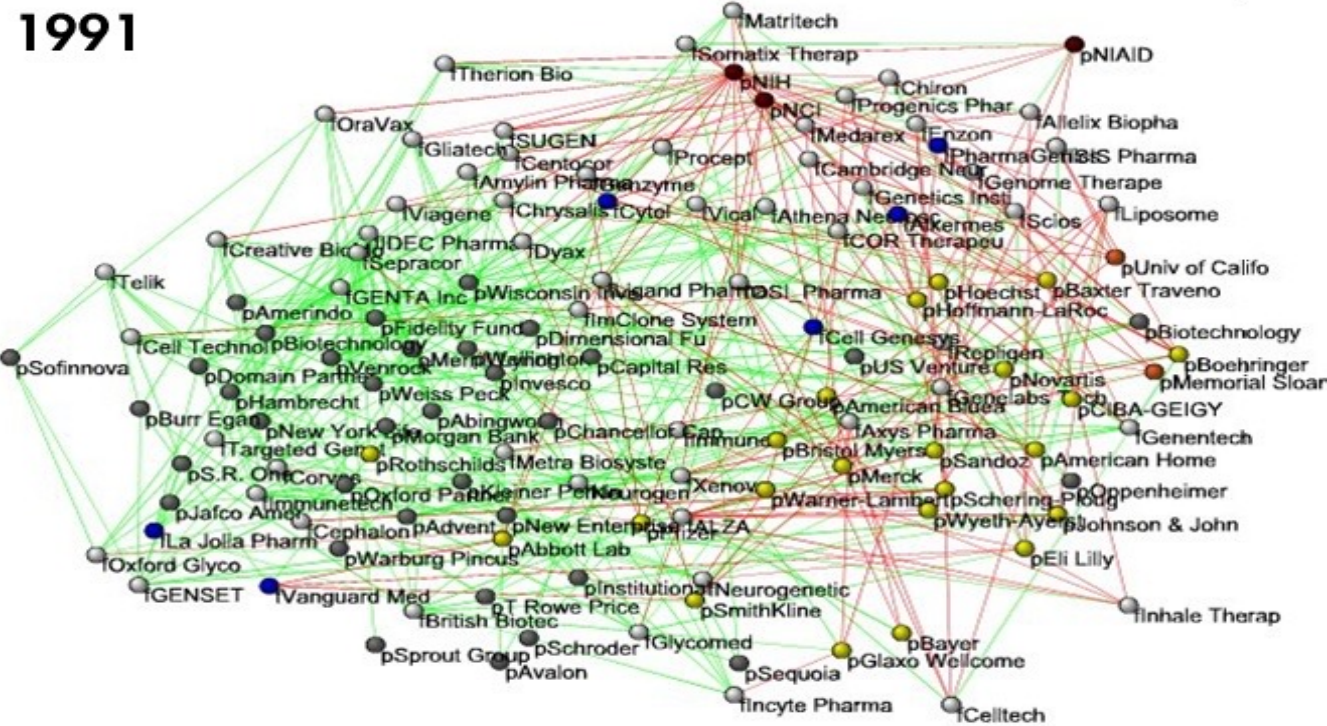
1991

**Nodes:**

- Companies
- Investment
- Pharma
- Research Labs
- Public
- Biotechnology

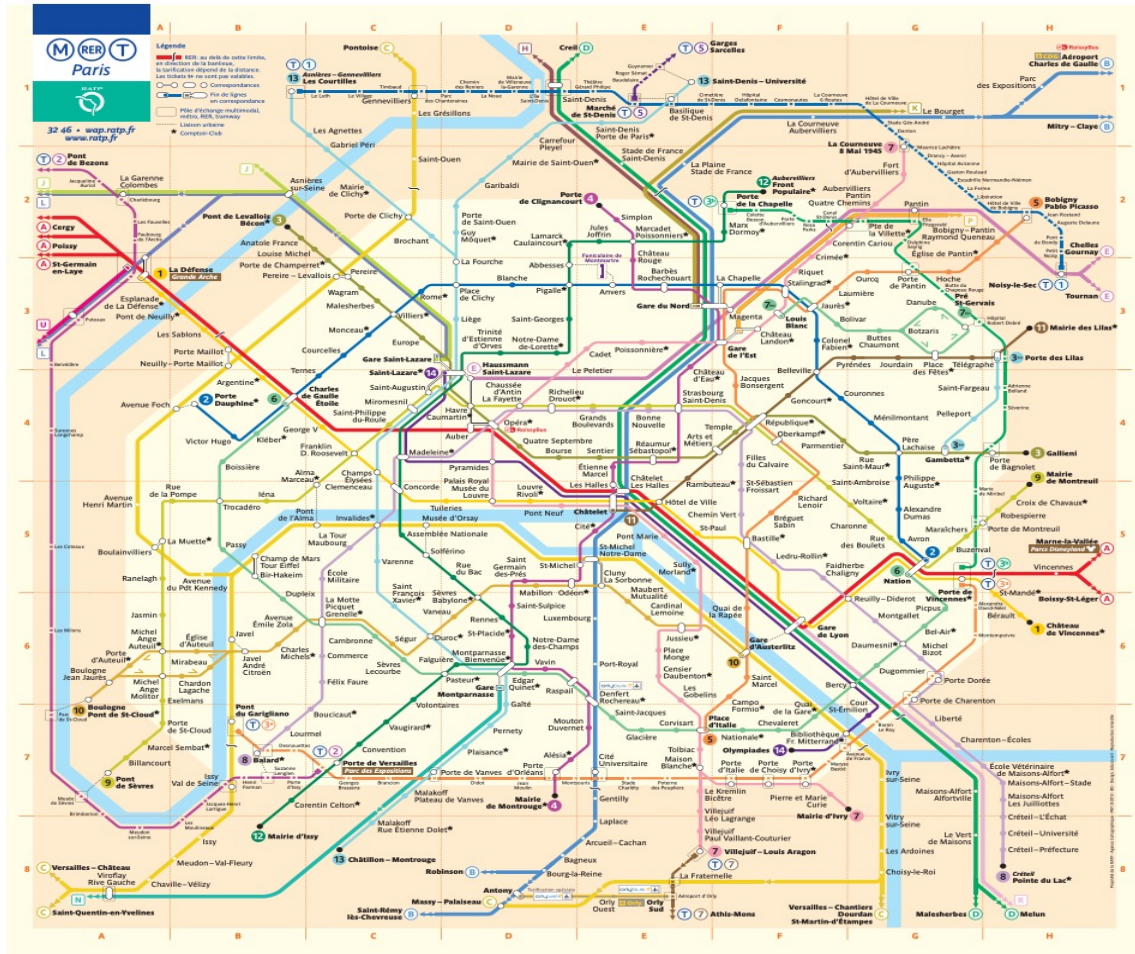
**Links:**

- Collaborations
- Financial
- R&D





# Paris metro

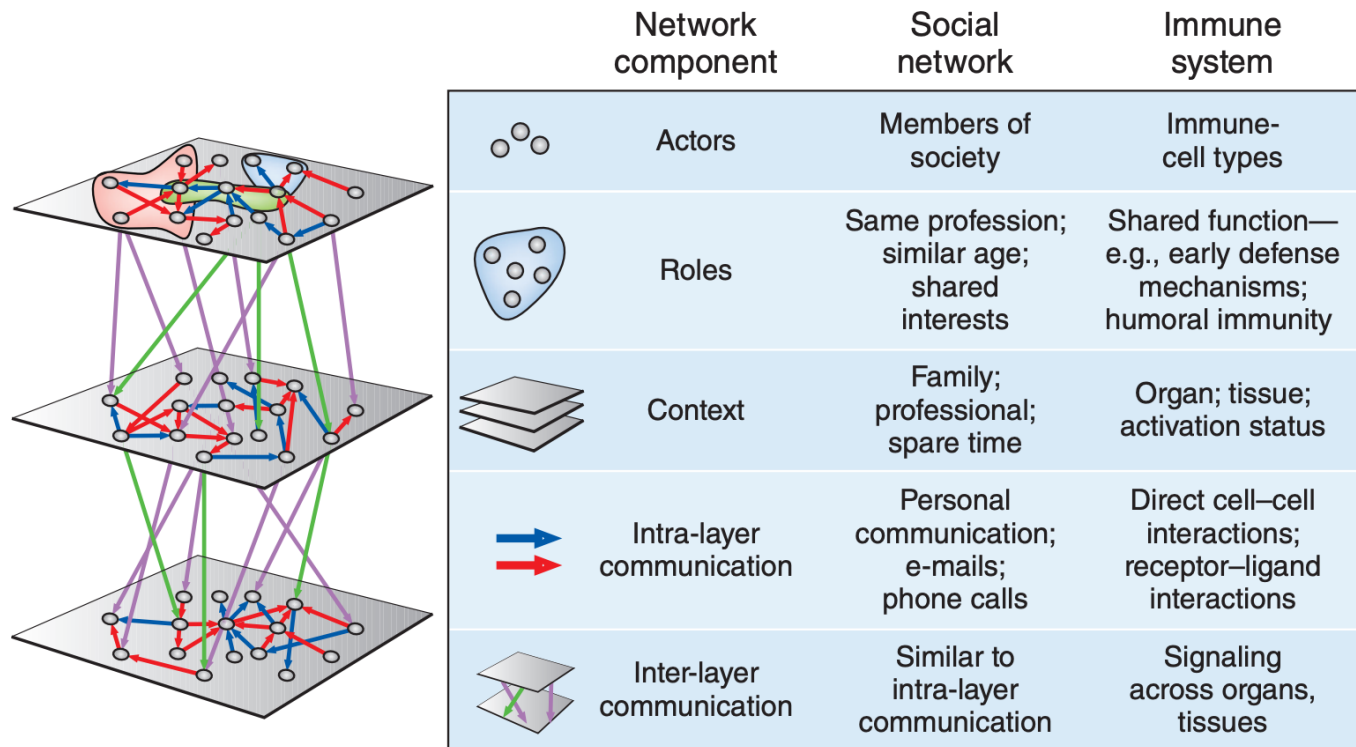


14 (+ 2) lines  
303 stations



# The immune system as a social network (MULTILAYER Network)

Andreas Bergthaler & Jörg Menche *Nature Immunology*, 18, 5, 2017

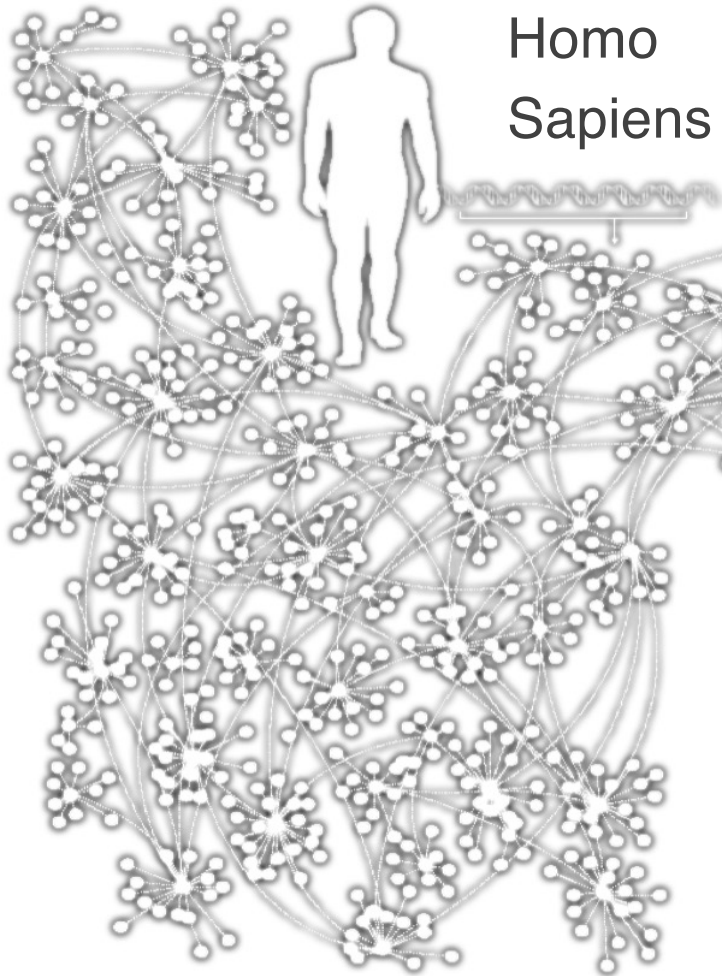


**Figure 1** The immune system as a multi-layered social network. The complex interplay among immune cells resembles in many ways a social network. Immune cells need to act in a highly cooperative and coordinated fashion when they respond to diverse external and internal threats. Communication

The immune system employs a multitude of molecules, cells and organs that act together throughout the entire body to guard human health.

Much like in a social network, immune cells can exert full functionality only through effective collaboration and communication.

Humans genes



Homo  
Sapiens

Drosophila  
Melanogaster



Points: elements of each organism's genetic network  
Dotted lines: interactions between elements

**Complex systems**

Made of many non-identical **elements** connected by diverse **interactions**.



# ON-LINE SOURCES

Stanford Large Network Dataset Collection

<https://snap.stanford.edu/data/>

Cambridge Networks Network

<https://www.cnn.group.cam.ac.uk/Resources/resources>

Covert networks

<https://sites.google.com/site/ucinetsoftware/datasets/covert-networks>

# WHAT IS A NETWORK?

---

Behind each complex system there is a network, that defines the interactions between the components

# WHY TO DO NETWORK ANALYSIS ?

---

Some good reasons:

- to study *whether the environment/system (network) affects unit outcomes*
- to study *whether occupying specific position (central, intermediary, peripheral) in the system (network) have different outcomes*
- to study *how single units affect the system*
- to study processes/mechanisms sustaining the system (network) structure
- to study intertwining of unit outcomes and the system (network) structure
- to study the global system (network) effectiveness

*(visualization is important but by itself it can not be sufficient to provide a definitive answer)*



# WHY NETWORK SCIENCE?

(TERM USED TO ADDRESS NETWORK RESEARCH)


---

## **Interdisciplinary nature:**

Network research crossed many disciplinary boundaries.

Network science offers a language through which different disciplines can seamlessly interact with each other.

Indeed, cell biologists and computer scientists alike are faced with the task of characterizing the wiring diagram behind their system, extracting information from incomplete and noisy datasets, and the need to understand their systems' robustness to failures or deliberate attacks.



# WHY NETWORK SCIENCE?


(TERM USED TO ADDRESS NETWORK RESEARCH)

---

## **Empirical, data driven nature:**

The tools of network science have their roots in **graph theory**, a fertile field of mathematics.

What distinguishes network science from graph theory is its empirical nature, i.e. its focus on data (real networks) and utility (insights mathematical tools can offer about a system's structure or evolution).



# WHY NETWORK SCIENCE?

(TERM USED TO ADDRESS NETWORK RESEARCH)


---

## **Quantitative and mathematical nature:**

To contribute to the development of network science, it is essential to master the mathematical tools behind it.

The tools of network science borrowed the formalism to deal with graphs from graph theory and the conceptual framework to deal with randomness.

The field also is benefiting from concepts borrowed from engineering, control and information theory, statistics and data mining, helping us extract information from incomplete and noisy datasets.





# WHY NETWORK SCIENCE?

(TERM USED TO ADDRESS NETWORK RESEARCH)


---

## **Computational nature:**

Given the size of many networks, and the exceptional amount of data behind them, network science offers a series of computational challenges.

Hence, the field has a strong computational character, actively borrowing from algorithms, database management and data mining.

A series of software tools have been developed in the last years to analyze networks (also for practitioners).




# INDIVIDUAL-BASED VS NETWORK-BASED RESEARCH

---

**Individual-based** (usual approach): interest in individual outcomes

- definition of the (statistical) population
- selection of a (random) sample and data collection
- inference (from the sample to population) based on (conditional) independency condition among units

**Network-based**: interest in individual and system outcomes

- network *boundary*: units to be considered in the network
  - unit as well as relation *dependency* (outcomes affected by a structure of relations)
  - network synthesis: *adaptation* of usual statistical measures
- 

# HOMWORKS – 1<sup>^</sup> WEEK

---

## Readings:

Barabasi A. (2012) The network takeover, *Nature Physics*, 8, pp. 14-16.

Brandes U., Robins G., McCrannie A., Wassermann S. (2013) What is network science? *Network Science*, 1, pp.1-15.

Carrington P.J., Scott J. (2011) *Introduction*, in Carrington P.J., Scott J. (Eds.) *The Sage Handbook of Social Network Analysis*, Sage Publications.

(Second Edition: McLevey J., Scott J., Carrington P.J. (2023) *The Sage Handbook of Social Network Analysis SECOND EDITION*, Sage Publications)

