

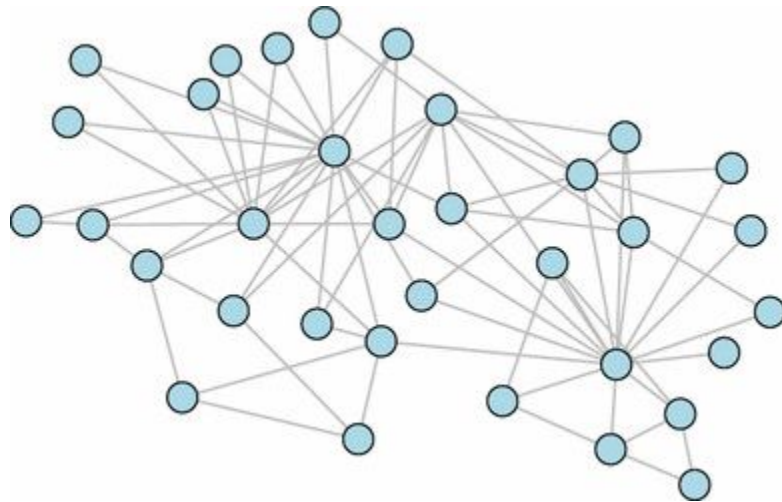
# Network biology

Practical session 12.03.2018  
Part II BBS Bioinformatics

## Lecture summary

# Network definition

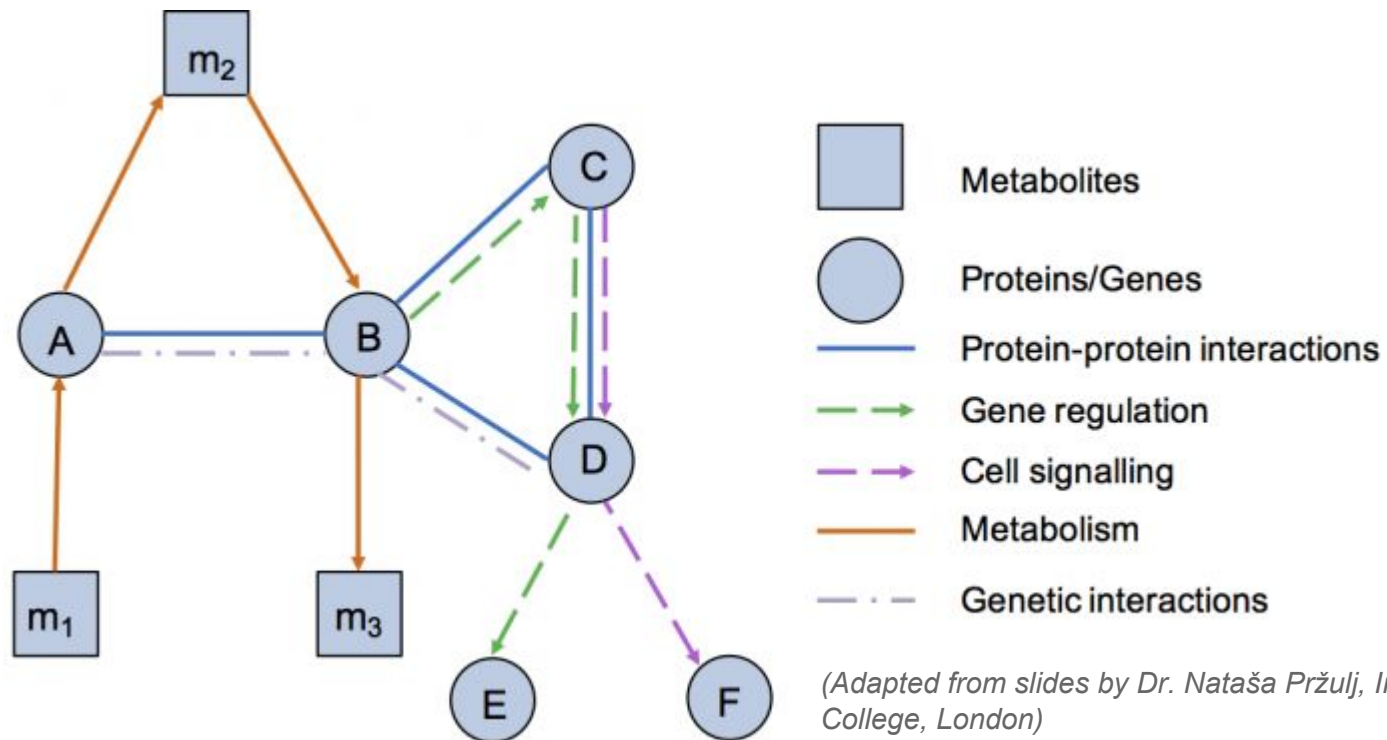
1. A system of interacting elements.
2. A map of interactions or relationships.



Can be represented mathematically as a **graph**, where vertices denote the **nodes** and **edges** (or links) denote the interactions.

# Cell = network of networks

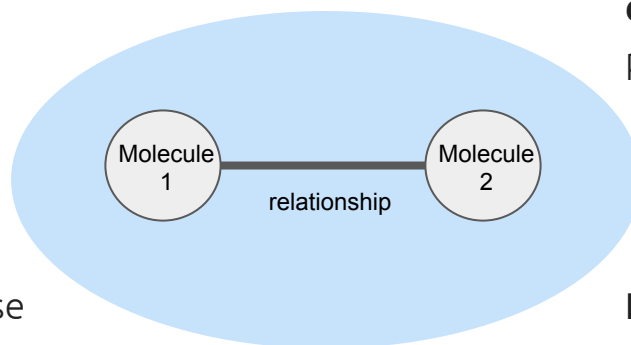
Different types of information can be represented as networks in order to model the cell.



# Sources of molecular interactions

**Expert curation** that evaluate existing published evidence.

**Computational predictions**, use existing experimental evidence to predict unexplored relationships



**Experimental: High-throughput datasets** (e.g. PPIs via affinity purification + mass spectrometry).

**Literature text-mining**, computationally extract molecular relationships from the published literature.

All of them characterized by different levels of  
**Coverage, bias & noise**

# Applications of network analysis in biology

1. Characterization of molecular networks at the systems level
  - Identify emergent patterns (structure, organization, relevant sub-networks) that can not be observed if the individual molecules were examined individually
2. Use of biological networks as a comprehensive scaffold of molecular interactions
  - To functionally characterize experimental data (high-throughput)
  - To make phenotypical predictions
3. Prediction of new interactions and networks

# Applications of network analysis in biology

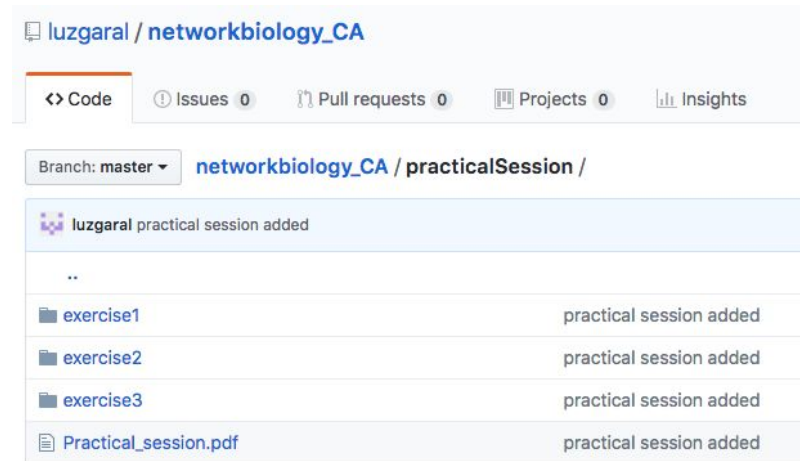
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Hands-on

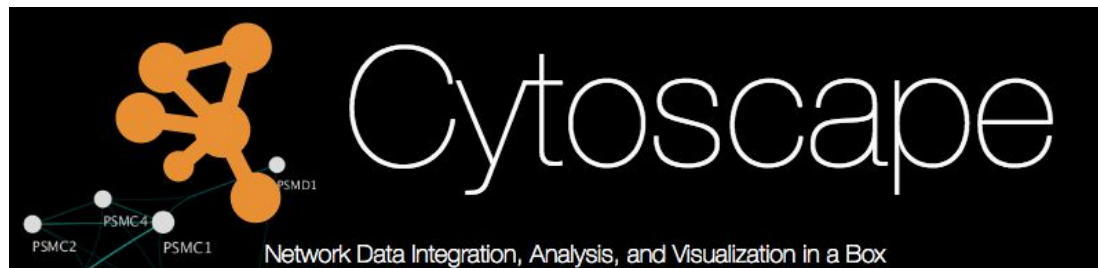


# Practical session

**DATA:** [https://github.com/luzgaral/networkbiology\\_CA](https://github.com/luzgaral/networkbiology_CA)



**SOFTWARE:** Cytoscape 3 <http://www.cytoscape.org/>



# Cytoscape

[www.cytoscape.org](http://www.cytoscape.org)

Software for Visualization, Integration & Analysis of Molecular Networks.

Free & open source software application (LGPL license)

> 10,000s users, 3000 downloads/month

Extensible through plugins/apps developed by third parties

## Cytoscape: a software environment for integrated models of biomolecular interaction networks

Google Scholar

[P Shannon](#), [A Markiel](#), [O Ozier](#), [NS Baliga](#)... - [Genome ...](#), 2003 - [genome.cshlp.org](#)

 Paperpile

Abstract **Cytoscape** is an open source software project for integrating biomolecular interaction networks with high-throughput expression data and other molecular states into a unified conceptual framework. Although applicable to any system of molecular components

☆  Cited by 10313 [Related articles](#) [All 20 versions](#) [Web of Science: 7116](#) [Import into BibTeX](#)

## Integration of biological networks and gene expression data using Cytoscape

[MS Cline](#), [M Smoot](#), [E Cerami](#), [A Kuchinsky](#)... - [Nature protocols](#), 2007 - [nature.com](#)

 Paperpile

Abstract **Cytoscape** is a free software package for visualizing, modeling and analyzing molecular and genetic interaction networks. This protocol explains how to use **Cytoscape** to analyze the results of mRNA expression profiling, and other functional genomics and

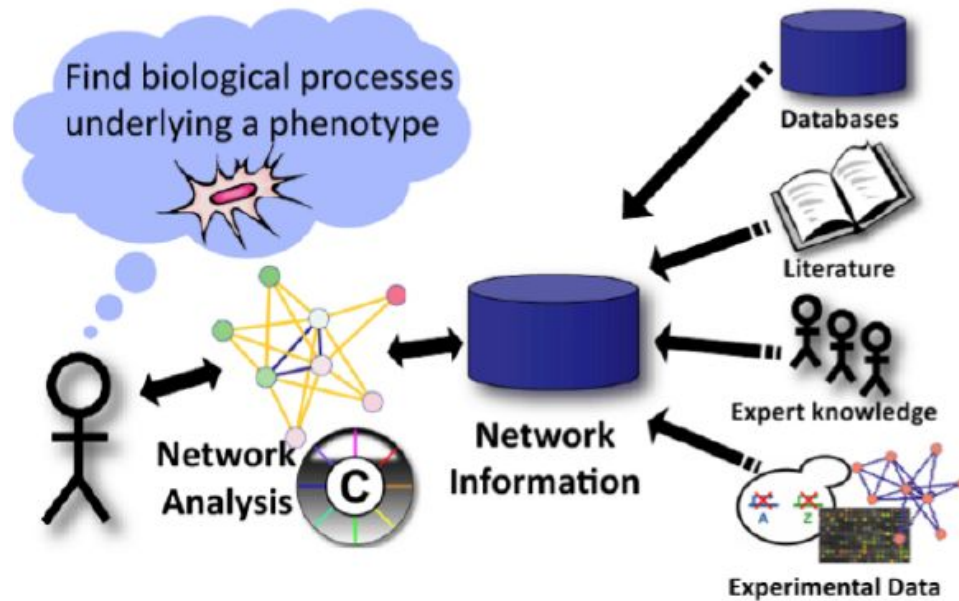
☆  Cited by 1679 [Related articles](#) [All 15 versions](#) [Web of Science: 1154](#) [Import into BibTeX](#)

# Cytoscape

## A Typical Cytoscape Workflow



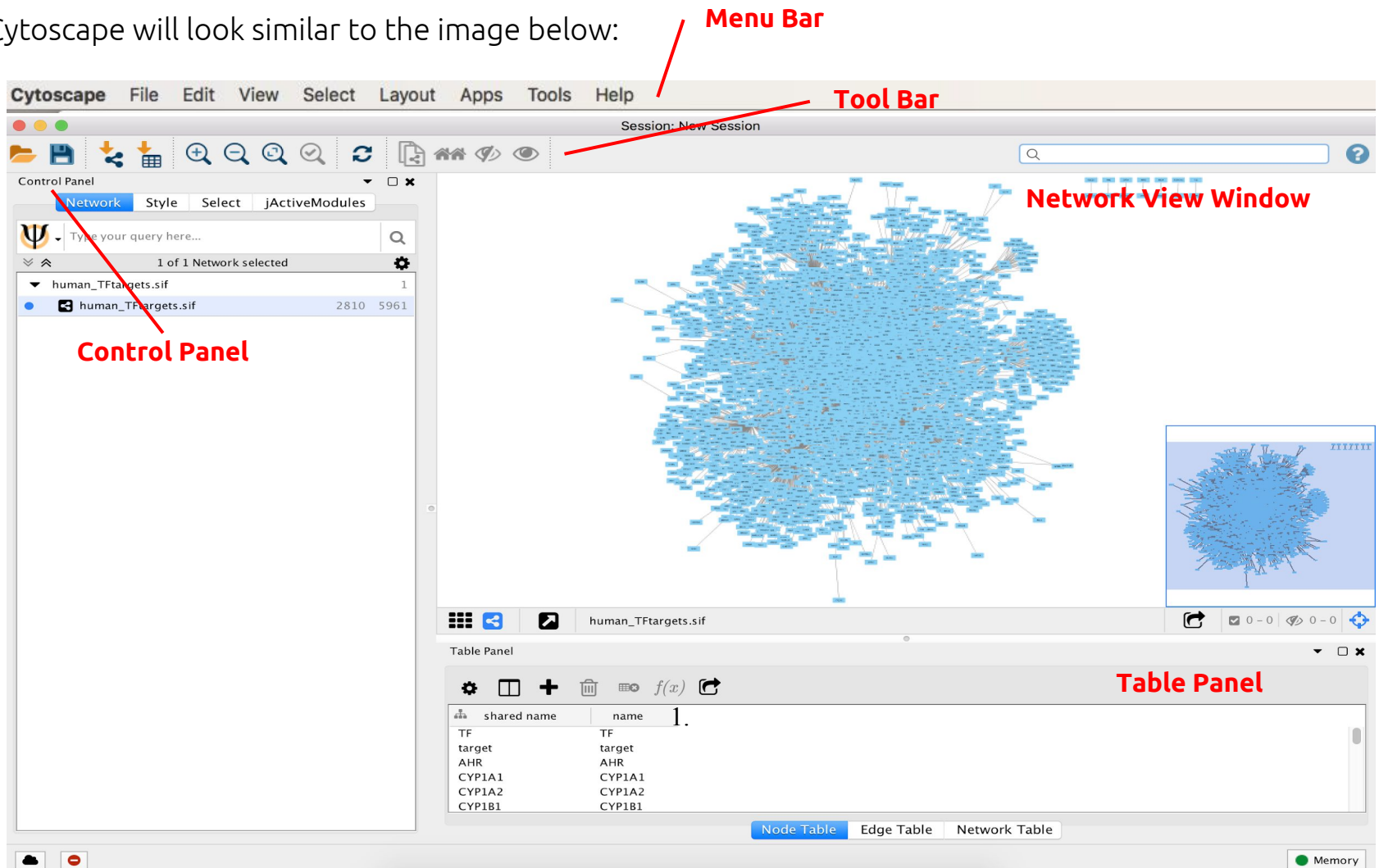
1. Load Networks (Import network data into Cytoscape)
2. Load node/edges related data or attributes (optional)
3. Analyse and Visualise Networks
4. Prepare for Publication



# Cytoscape

## Panels overview

Cytoscape will look similar to the image below:



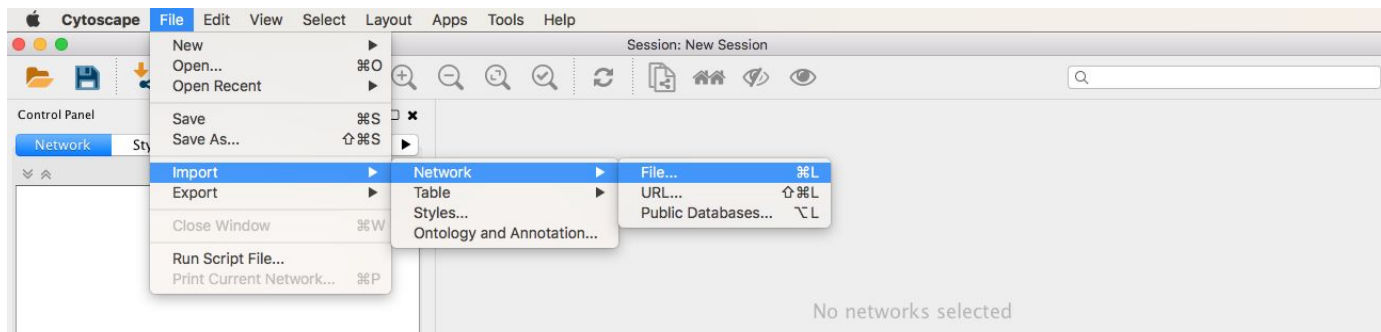
# Cytoscape

## Loading networks

Two ways:

- Use import network from File:
  - Excel file, comma or tab delimited text
  - Simple interaction file
  - Other network formats
- Use import network from Web Services
  - IntAct (EMBL EBI)
  - Reactome
  - Pathway commons (collection of sources)
  - NCBI Entrez EUtilities

In this practical session we are going to use the networks provided in the github repo as `sif` files

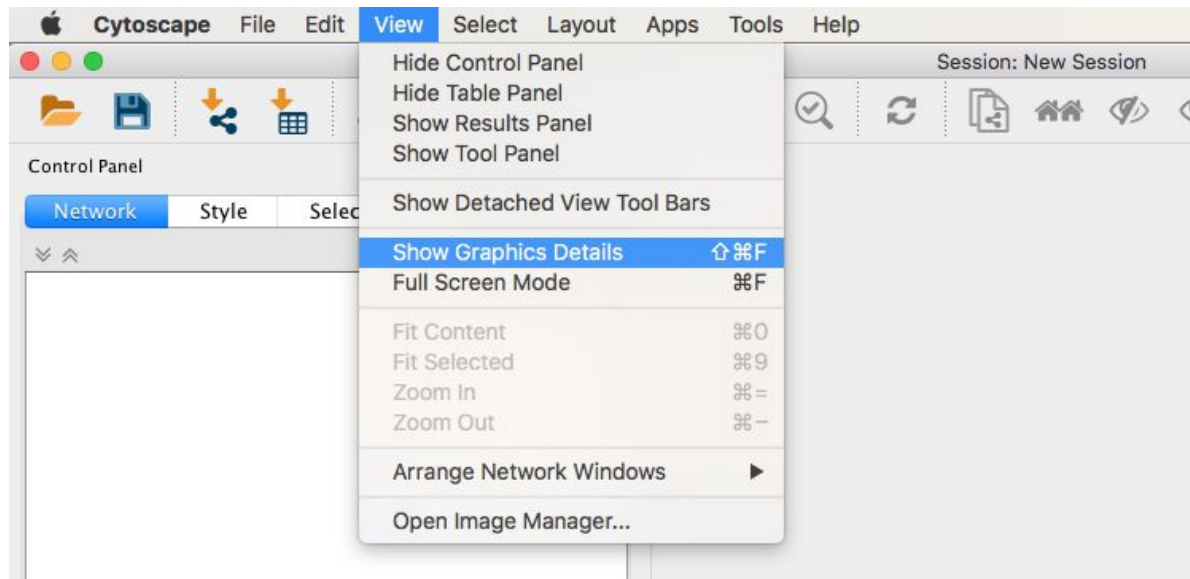


# Cytoscape

## Loading networks

### Tips & Tricks: Network View

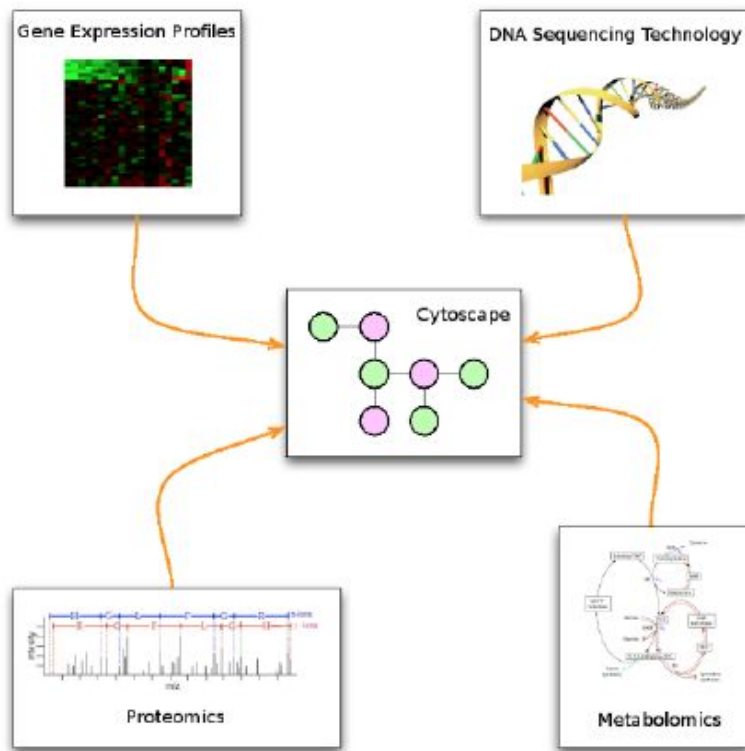
- When you open a large network, you will not get a view by default
- To see what things will look like at full detail: View → Show Graphics Details



# Cytoscape

## Loading node/edge associated data (attributes)

Nodes and edges can have data/attributes associated with them:



Cytoscape supports multiple data types: Numbers, Text, Logical, Lists.

Upload data (formats):

- excel file
- tab delimited text
- expression matrix

Import data from web services

- NCBI Entrez Gene
- Ensembl Biomart

Create attributes from the node/edge data panel

In this practical session we are going to use the node attributes provided in the github repo as **.csv** files

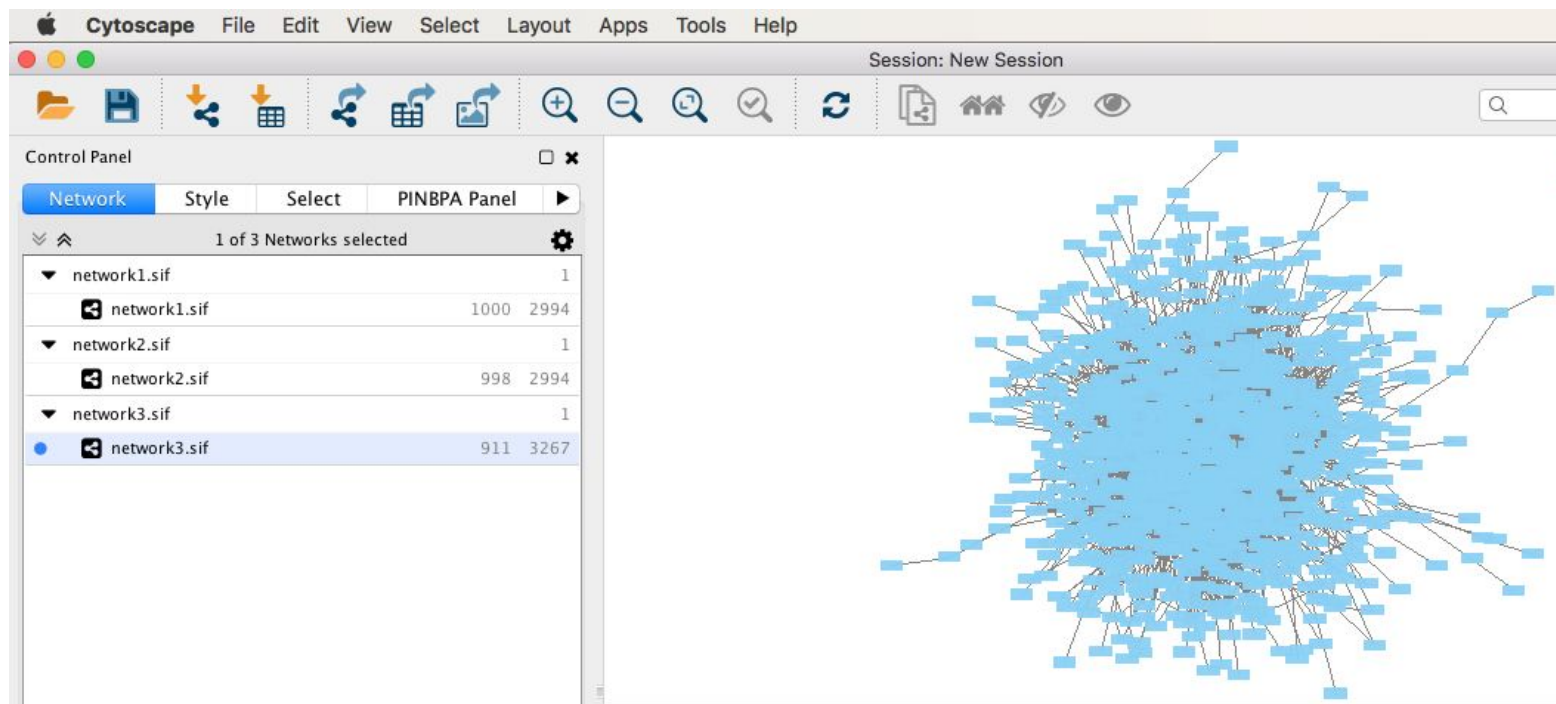


# Cytoscape

## Loading node/edge associated data (attributes)

### Tips & Tricks

- You can have multiple independent graphs in Cytoscape
- When you upload new node-related data, be sure to select the correct network

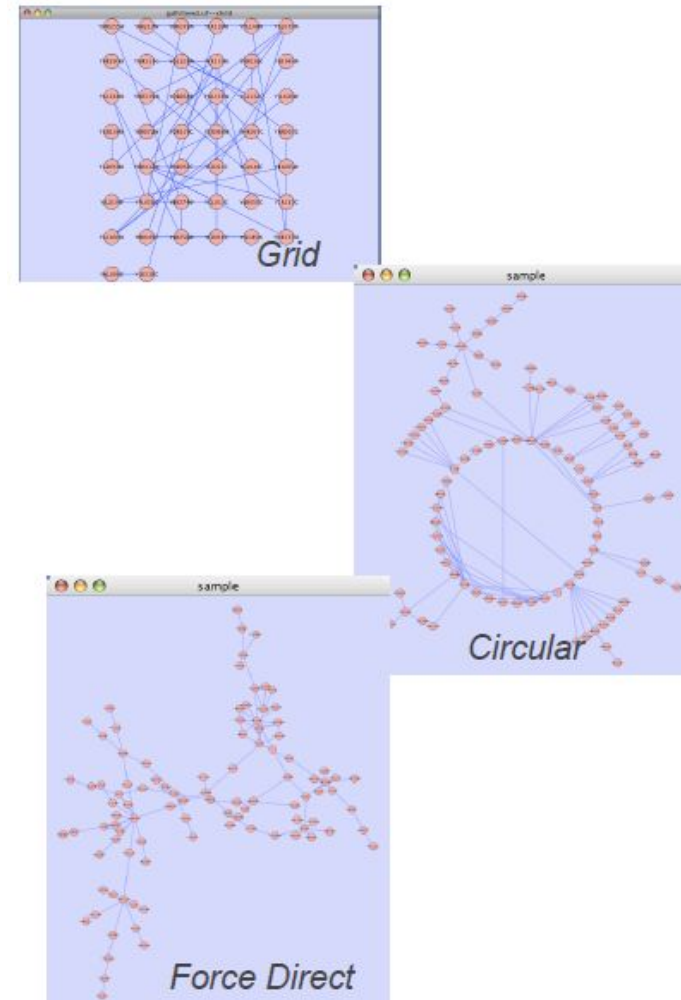
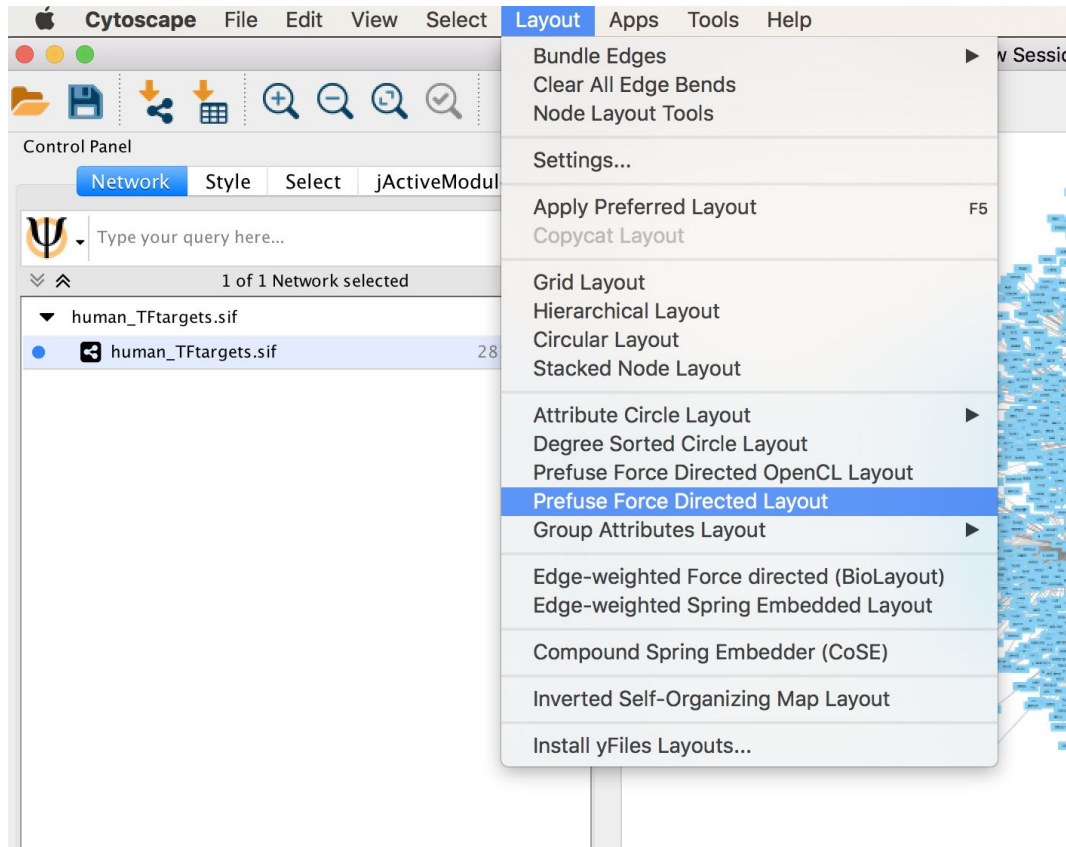




# Cytoscape

## Visualize Networks

**Layouts** several algorithms available

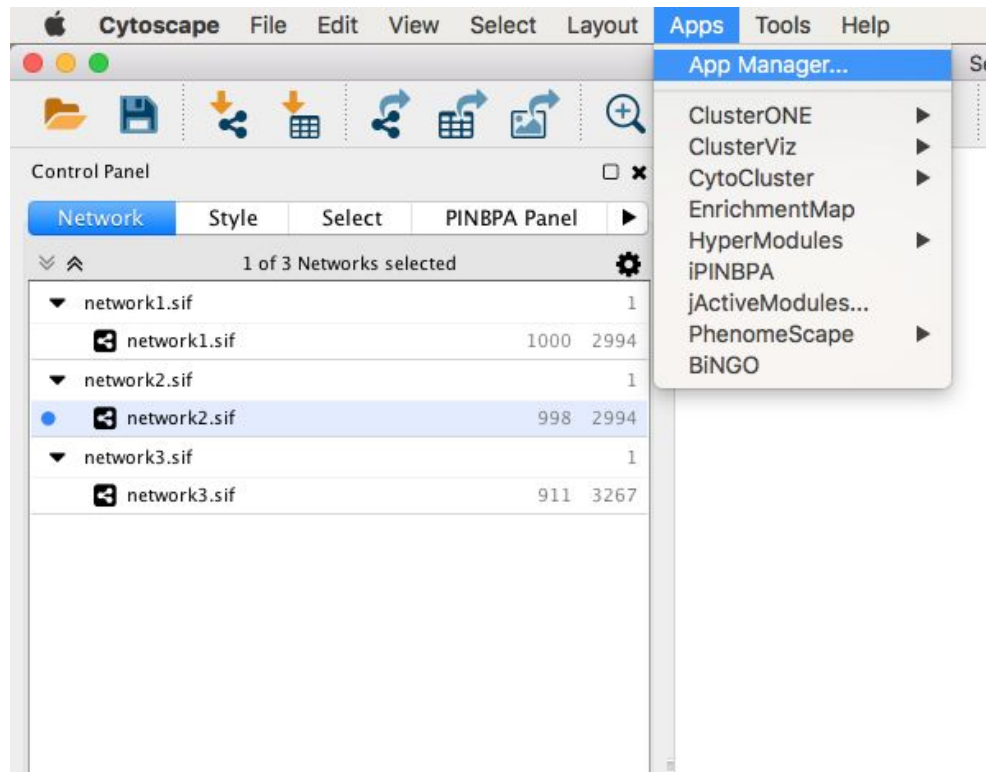


# Cytoscape

## Extended Functionality

Cytoscape extends its functionality with Apps:

- Developed by third parties
- Listed at <http://apps.cytoscape.org/>
- Major categories:
  - Online databases import
  - Graph Analysis
  - Genomics Integrated Analysis
  - Clustering
  - Visualization
  - Network Generation
- Usually available through
  - Apps Manager in the toolbar
  - Download from the Apps websites



For this practical session we are going to install the modules: jActiveModules, BiNGO, MCODE.

Install them via App Manager.

# Applications of network analysis in biology

1. Characterization of molecular networks at the systems level
  - Identify emergent patterns (structure, organization, relevant sub-networks) that can not be observed if the individual molecules were examined individually
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# Characterization of molecular networks at the systems level

Looking at different network properties can provide valuable information about:

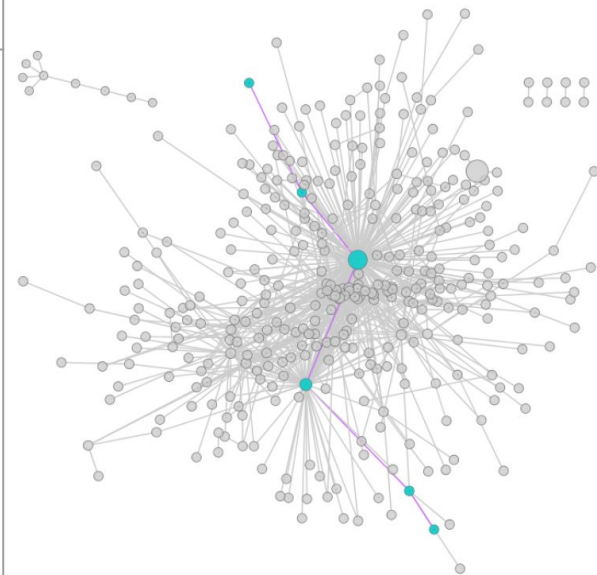
- the internal organization of a biological network (topological properties)

# Characterization of molecular networks at the systems level

Looking at different network properties can provide valuable information about:

- the internal organization of a biological network (topological properties)

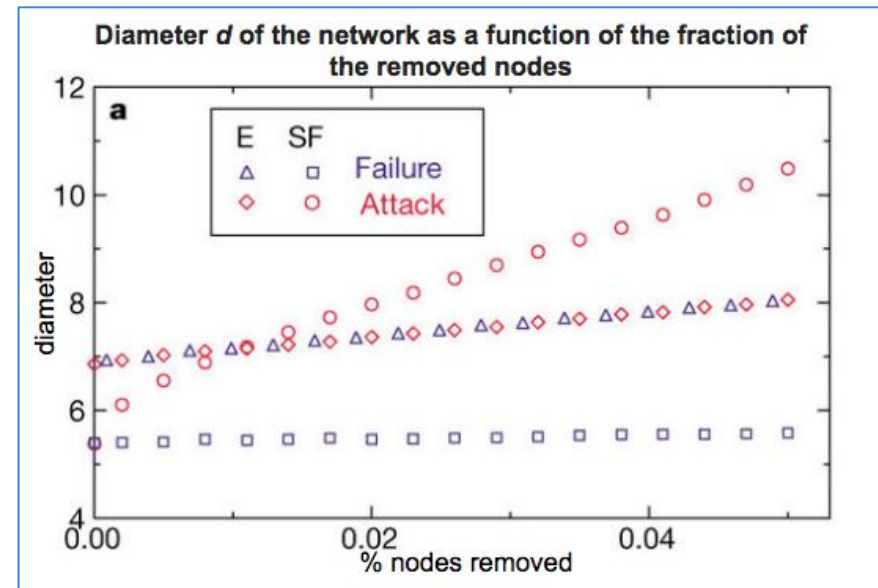
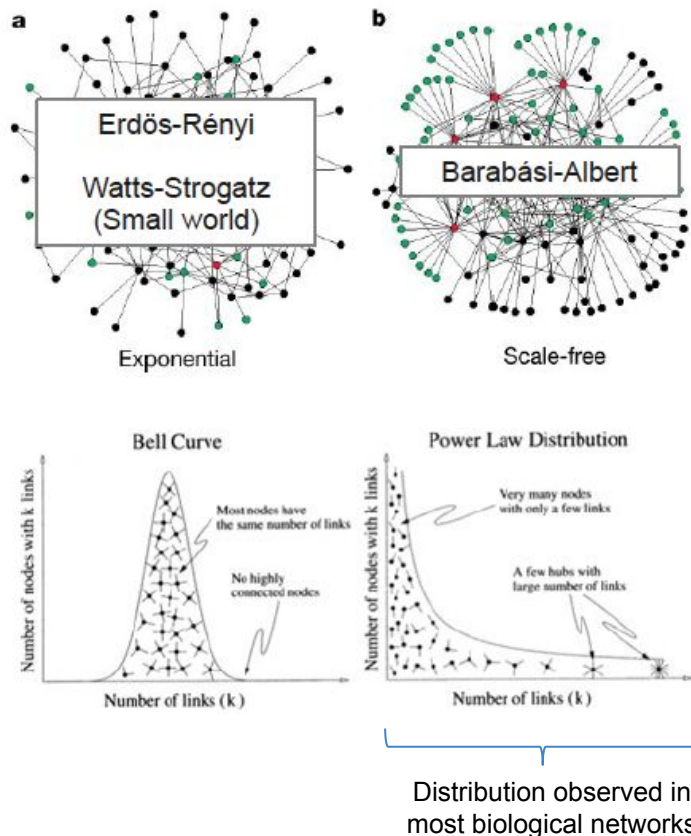
Topology Parameters	
Global	Local
<ul style="list-style-type: none"><li>- <b>Components:</b> Set of nodes connected between them but disconnected to the other nodes.</li><li>- <b>Shortest Path:</b> The path with the smallest number of edges between two nodes.</li><li>- The largest shortest path is the network <b>diameter</b>.</li><li>- <b>Degree Distribution:</b> Counts the number of nodes with degree <math>k</math> and gives the probability that a selected node has exactly <math>k</math> edges. The degree distribution allows us to distinguish between different classes of networks.</li></ul>	<ul style="list-style-type: none"><li>- <b>Degree:</b> the number of direct interactors. Hub is a node with an exceptional high degree.</li><li>- <b>Betweenness:</b> the frequency with which a node lies on the shortest path between other proteins.</li><li>- <b>Closeness:</b> the average distance to all other proteins.</li><li>- <b>Clustering Coefficient:</b> the degree to which nodes in a graph tend to cluster together.</li></ul>



# Characterization of molecular networks at the systems level

Looking at different network properties can provide valuable information about:

- the internal organization of a biological network (topological properties)
- the evolutionary constraints that have shaped an organism's molecular network



High degree of **robustness**  
against accidental node  
failures

# Characterization of molecular networks at the systems level

Looking at different network properties can provide valuable information about:

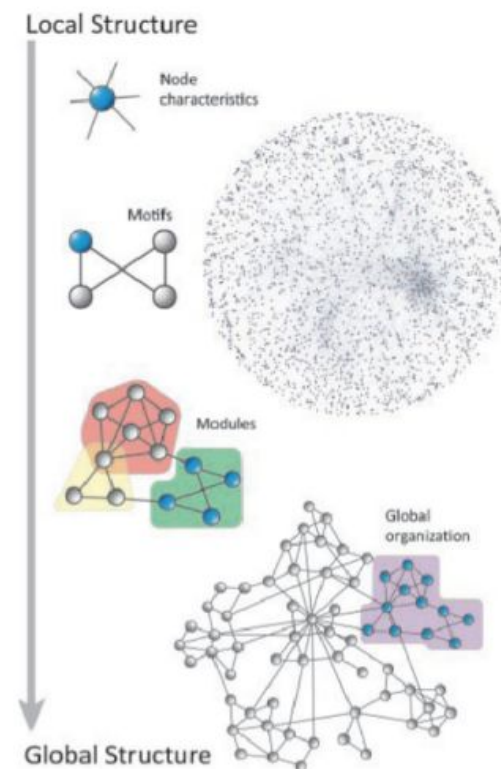
- the internal organization of a biological network (topological properties)
- the evolutionary constraints that have shaped an organism's molecular network
- the distribution of molecules among biological processes (modular properties)

**Cellular functions** are likely to be carried out in a highly modular manner.

The levels of organization of complex networks are:

- **Hierarchy** describes how the various structural elements are combined
- Groups of highly interconnected nodes are called **modules** (or communities)
- Each module can be reduced to a set of triangles (**motifs**). A high degree of triangles is reflected by the clustering coefficient (the signature of a network potential modularity)

A hierarchical architecture implies that sparsely connected nodes are part of highly clustered areas, with communication between the different highly clustered neighbourhoods being maintained by a few highly connected nodes



# Characterization of molecular networks at the systems level

Looking at different network properties can provide valuable information about:

- the internal organization of a biological network (topological properties)
- the evolutionary constraints that have shaped an organism's molecular network
- the distribution of molecules among biological processes (modular properties)
- detect coverage and bias issues



# Exercise 1

## Characterizing networks topology

**Exercise 1.** Load and study the topological properties of the 3 networks in the folder “exercise 1”.

- Upload networks (Menu: File → Import → File). Note: Create a new collection for each network. Use the Control Panel to move from one network to another.
- Customize the visualization style (Left Control panel: Style)
- Change layout (Menu: Layout → *Prefuse Force Directed Layout* (preferred))
- Could you guess the *degree distribution* of each network by looking at the network display? Would you be able to identify which network follows a random distribution of the degree (e.g. such as the Erdős–Rényi (ER) model).
- Analyze Topology (Menu: Tools → NetworkAnalyzer → Network Analysis → Analyze Network (undirected)).
  - Explore network-level properties (Menu: View → Show Results Panel):
    - **Q-1.1:** What degree distribution follows every network? Which network model describes better each network?
    - **Q-1.2:** Which is the diameter, number of components and the average path length of every network?
  - Explore node-level properties (Table panel (bottom) → Node table):
    - **Q-1.3:** Which network would be more damaged (i.e. impacting a larger proportion of shortest paths) after removing the node 24?
- Save the session as “exercise 1” and close it.

# Exercise 1

## Characterizing networks topology

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<> Code    ! Issues 0    🔑 Pull requests 0    📁 Projects 0    📊 Insights

Branch: master ▾    networkbiology\_CA / practicalSession / exercise1 /

luzgaral practical session added

..

network1.sif	practical session added
network2.sif	practical session added
network3.sif	practical session added

# Exercise 2

## Characterizing function and modularity

**Exercise 2.** In this section we are going to analyse 2 collections of human PPIs (*human\_ppis\_t1\_c.sif* and *human\_ppis\_t2\_ph.sif*) that have been identified by 2 new experimental techniques. Unfortunately, these techniques are able to identify only specific types of PPIs. Here, we are going to use GO enrichment approaches (via BiNGO app) to characterize the coverage and bias in these datasets. Also, we are going to study their organization and modularity (via MCODE).

# Exercise 2

## Characterizing function and modularity

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<> Code    ! Issues 0    🔗 Pull requests 0    📁 Projects 0    📊 Insights

Branch: master ▾    networkbiology\_CA / practicalSession / exercise2 /

luzgaral practical session added

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📄 human_ppis_t1_c.sif	practical session added
📄 human_ppis_t2_ph.sif	practical session added
📄 human_protein_census.txt	practical session added

Networks

List of censed human proteins (needed for the Gene Enrichment Analysis with BiNGO as reference list)

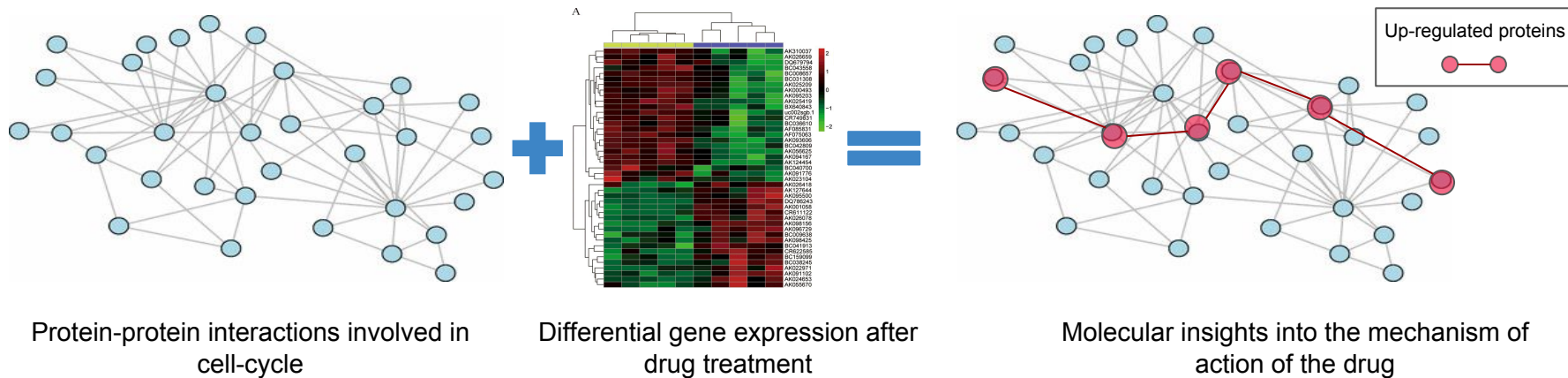
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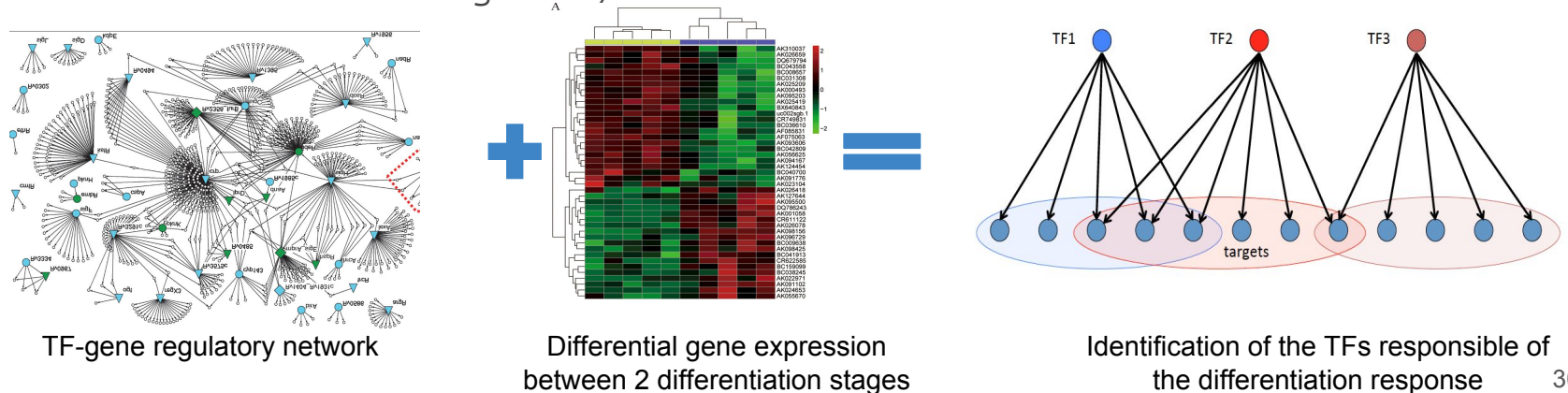
# Functional characterization of large-scale experimental data

Common examples:

1. Identification of signalling pathways with differential gene expression upon drug treatment



2. Identification of the Gene Regulatory Network involved in cell differentiation





# Exercise 3

## Network-based characterization of Differential Gene Expression data

**Exercise 3.** Imagine that you are conducting a research on Medulloblastoma (MB), a common malignant brain tumor in children, where MYC overexpression or amplification has been associated with poor clinical outcome. MYC is a well known transcriptional regulator and you would like to characterize the molecular alterations associated with aberrant MYC activity to design a therapeutic strategy. For this purpose, you conducted an *in vitro* experiment where you overexpress MYC in an MB-derived cell line and compared the gene expression profiles before and after the perturbation. Here, you are going to use two types of human biological networks (gene regulatory network: *human\_TFtargets.sif* and signalling network: *human\_signalling.sif*) to characterize the results from the differential expression analysis (file: *MYC\_GSE22139\_differential\_expression.csv*). NOTE: The expression data has been derived from Giulio et al 2011 and downloaded from GEO (accession GSE22139).

# Exercise 3

## Network-based characterization of Differential Gene Expression data

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Code Issues 0 Pull requests 0 Projects 0 Insights

Branch: master networkbiology\_CA / practicalSession / exercise3 /

luzgaral practical session added

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MYC_GSE22139_differential_expression.csv	practical session added
human_TFtargets.sif	practical session added
human_TFtargets_genelist.txt	practical session added
human_signalling.sif	practical session added
human_signalling_proteinlist.txt	practical session added

Networks

List of proteins in the networks (needed for the Gene Enrichment Analysis with BiNGO as reference list)

Differential gene expression file