

# A basic introduction to Complex Networks, using python

Marcelo Ponce



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# Outline

## 1 Introduction

- Motivation & Examples

## 2 Basic Elements

- Python Packages
- Building your own CN: **NetworkX**
- Networks
- Several types of graphs/networks in **NetworkX**
- Basic Network Analysis

## 3 Real-life applications

- Twitter relationships
- SciNet GPC cluster

## 4 Simulating CNs

- PyCX

## 5 References

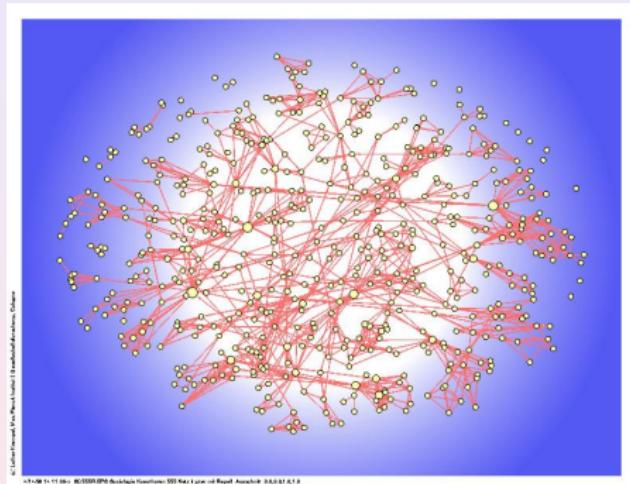


# Examples of Complex Networks

A system composed of many nonlinear interacting units often forms a complex system with new emergent properties that are not held by the individual units.

Examples:

- circadian rhythms,
- electrochemical reactions,
- laser arrays,
- neuron networks,
- populations of fireflies,
- Josephson junction arrays...,  
etc.



▲ Co-Authorship Network By Lothar Krempel

<http://www.mpi-fg->

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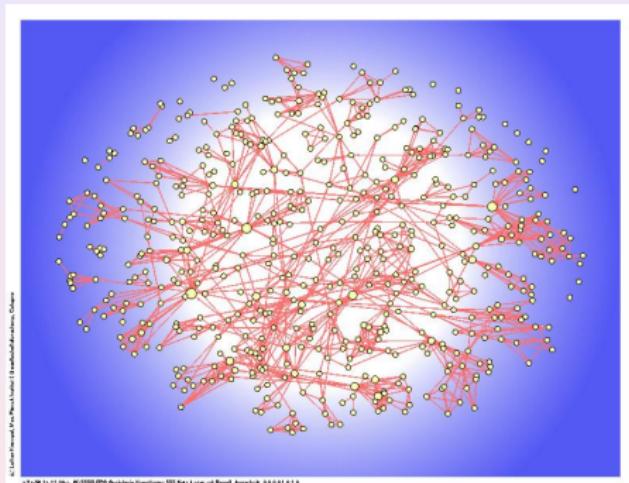


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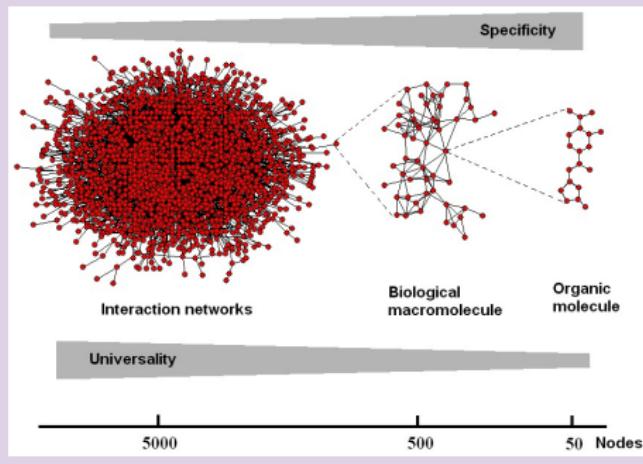
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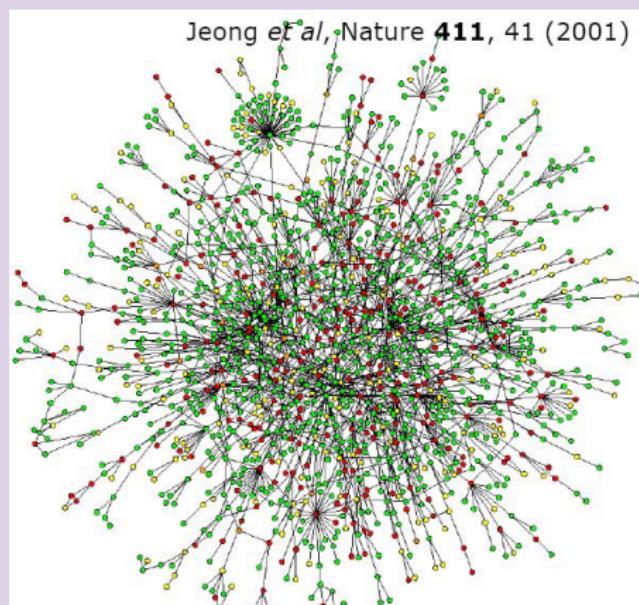
# Examples of Complex Networks

Biological networks...

- ▼ Proteins, molecules, chemical reactions...
- ▼ Genes, proteins, ...



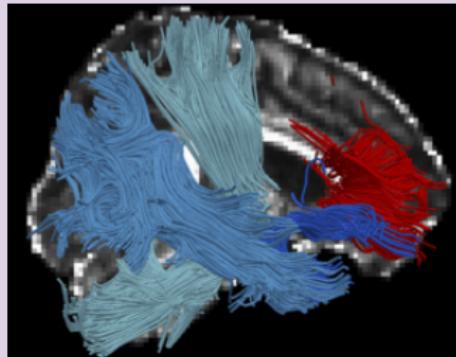
Map of protein-protein interactions. The colour of a node signifies the phenotypic effect of removing the corresponding protein (lethal; non-lethal; slow growth; unknown). ▶



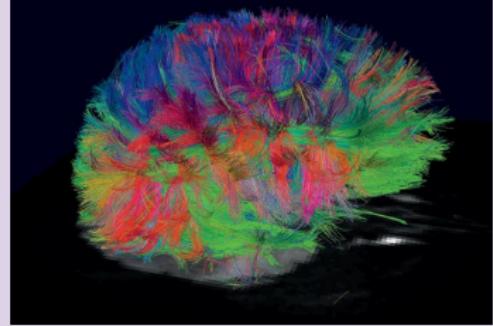
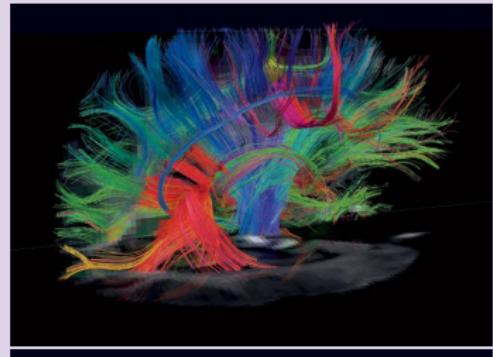
# Examples of Complex Networks

Brain networks...

## ▼ Physiological networks



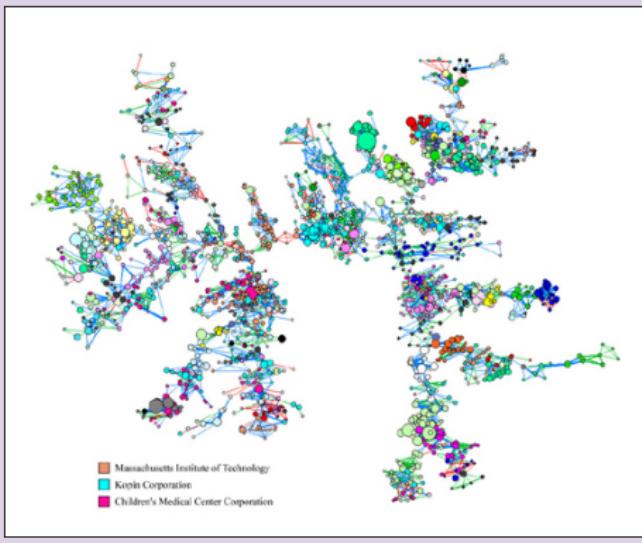
## ▼ Neurological pathways



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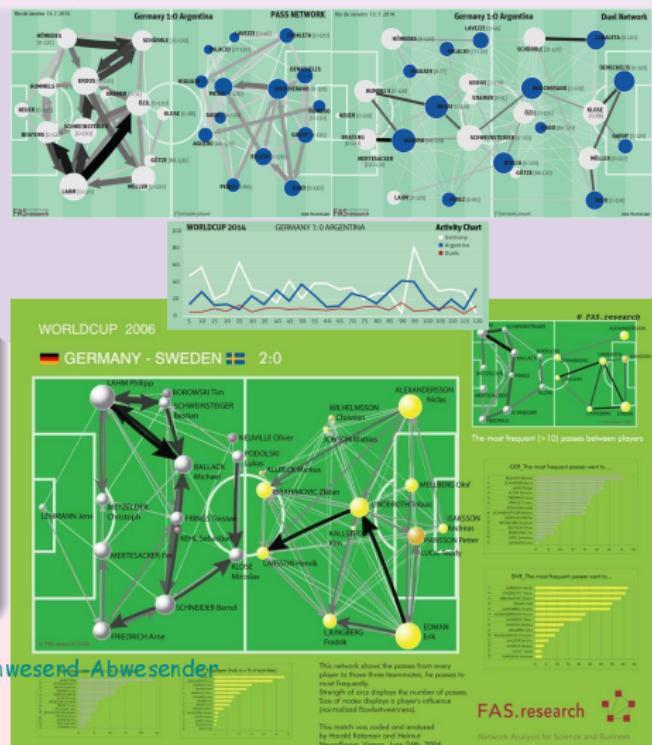
Social networks...

- ▼ Boston Inventor Networks mid-1990s



<http://derstandard.at/2000002589338/Messi-glaenzt-als-anwesend-Abwesender>

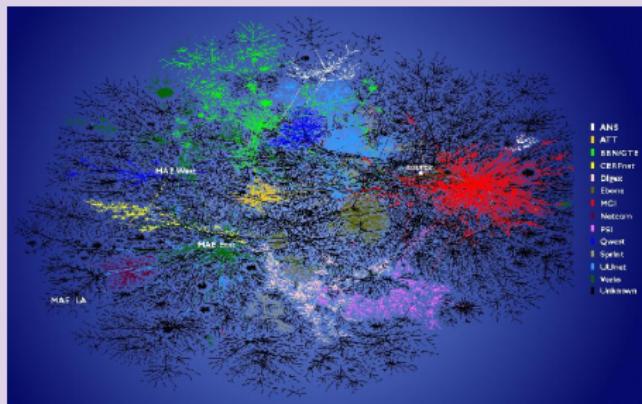
- ▼ Sports: Futbol - world cup...



# Examples of Complex Networks

## Internet

- ▼ Internet connectivity,



with selected backbone ISPs (Internet Service Provider)

colored separately

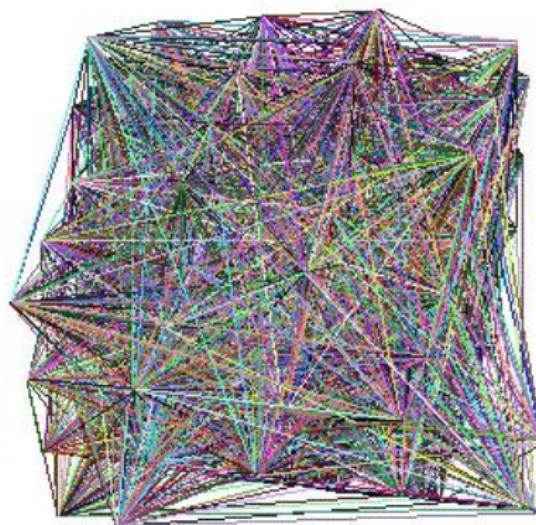
- ▼ Arc map showing the world-wide internet traffic



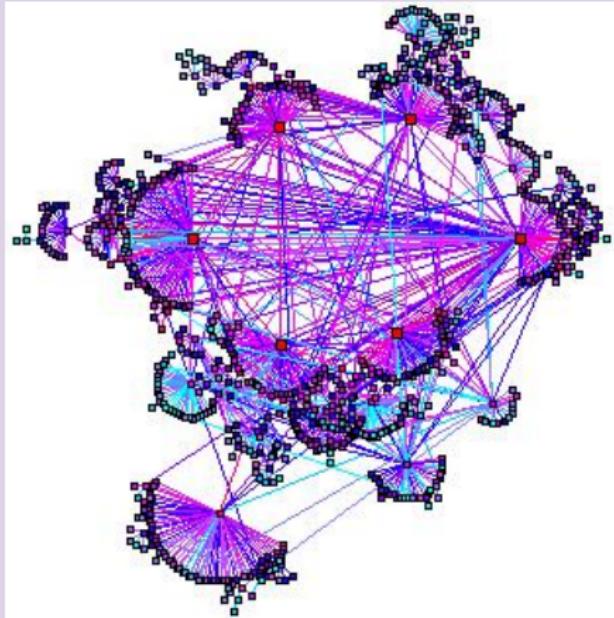
# Examples of Complex Networks

## Internet

- Mapping the internet: tracing domain routes



- Hierarchical topology of the international web cache

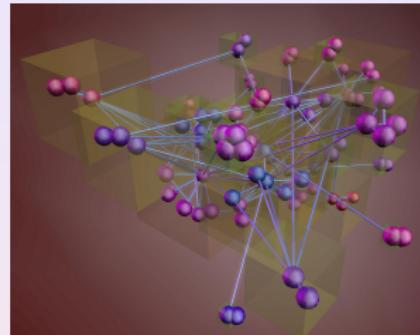


# nodes, network, maps...

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- ➊ [nodes]: ensemble of **N** oscillators
- ➋ [dynamics+network]: each oscillator evolves according to its own **intrinsic dynamics** [map (fn)] + a **coupling term** [network].
- ➌ The **coupling term** is defined by:
  - **coupling strength**
  - **coupling topology** (all-to-all, n-n, etc)

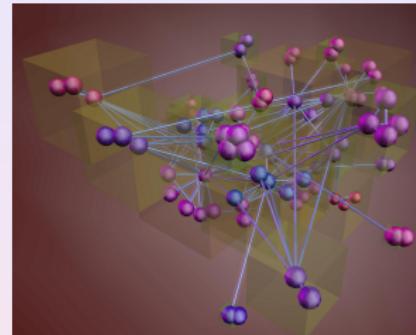
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- ▼ from the previous examples:
- Internet: Nodes  $\sim$  routers & Links  $\sim$  optical fibers.
  - WWW: Nodes  $\sim$  document files & Links  $\sim$  hyperlinks.
  - The Scientific Citation Network: Nodes  $\sim$  papers & Links  $\sim$  citations.
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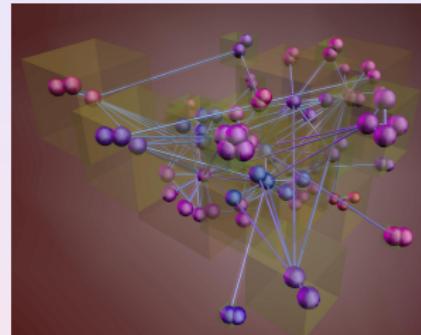
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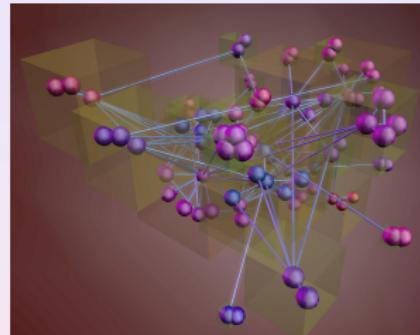
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# CN: several packages & tools available in Python

- **networkX:** <http://networkx.lanl.gov/>
- **PyCX Project:** <http://pycx.sf.net/>
- ComplexNetworkSim
- SimPy
- graph-tool [<http://graph-tool.skewed.de/>]
- pyGraphViz (GraphViz) [<http://www.graphviz.org/>]
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# Required Packages



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



[matplotlib.org](http://matplotlib.org)

► NetworkX

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# Defining a graph

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import networkx as nx

g = nx.Graph()

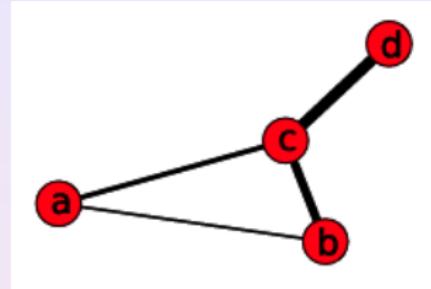
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```

## Some basic props. of the graph

```
g.number_of_nodes() # also g.order()
4
g.order() # also g.size()
4
```

```
g.nodes()
['a', 'c', 'b', 'd']
g.edges()
[('a', 'c'), ('a', 'b'), ('c', 'b'), ('c', 'd')]
g.edges('b')
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```

```
g.degree('a')
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g.neighbors('c')
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```



```
print nx.shortest_path(g, 'b', 'd')
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d = nx.draw(g)
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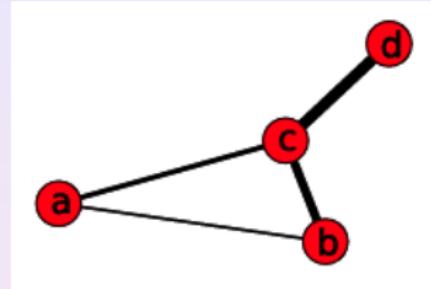
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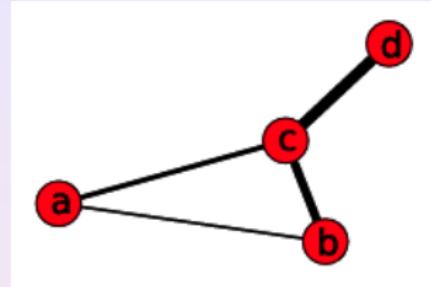
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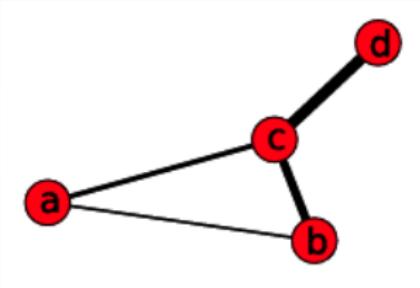
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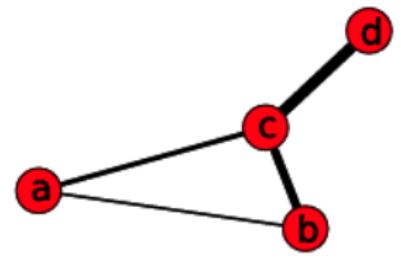
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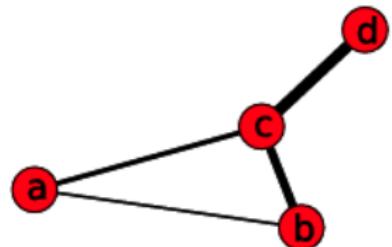
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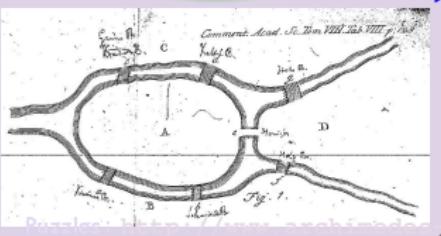
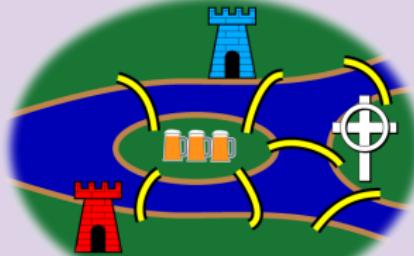
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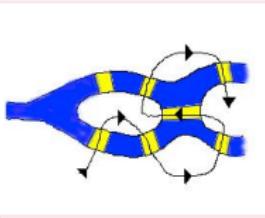
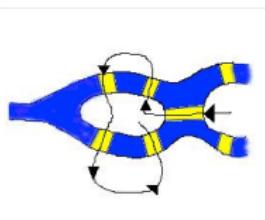
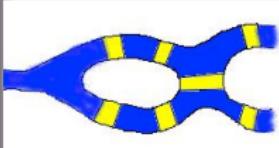
# Network Topology

## Graph Theory

- 7 bridges of Königsberg [Euler, 1736]

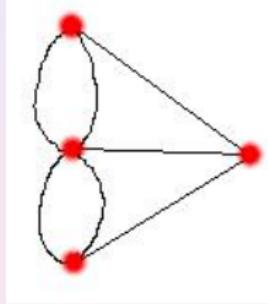
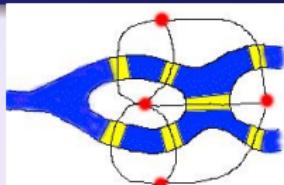
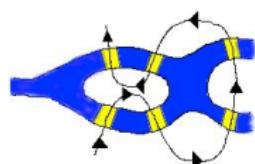
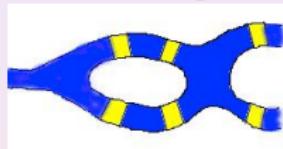


- 7 bridges



▶ NO soln!!!

- 6 bridges

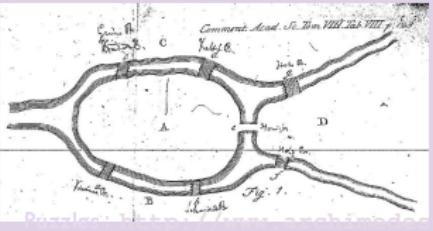
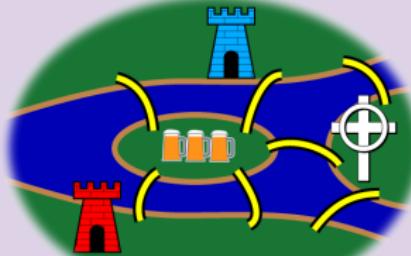


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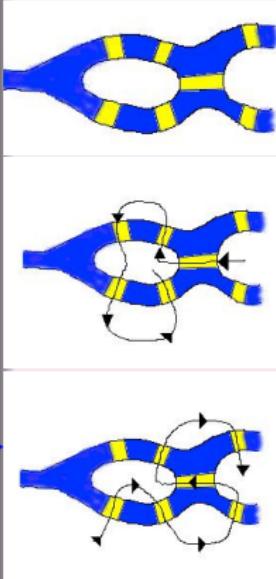
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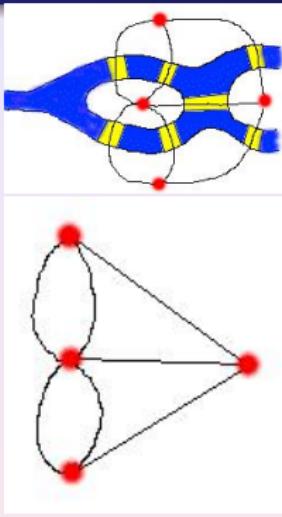
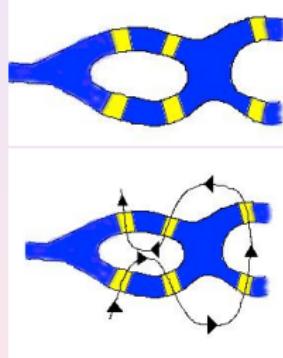


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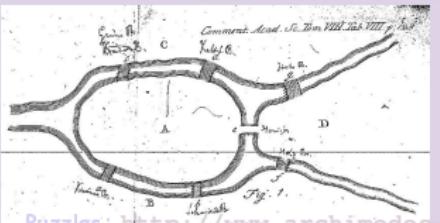
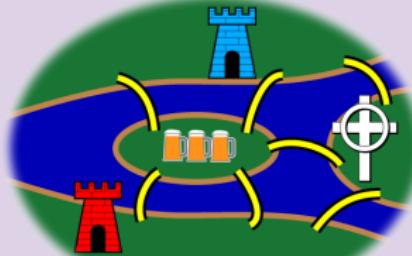


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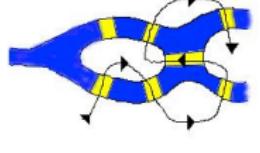
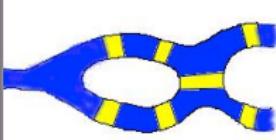
# Network Topology

## Graph Theory

- ▼ 7 bridges of Königsberg [Euler, 1736]

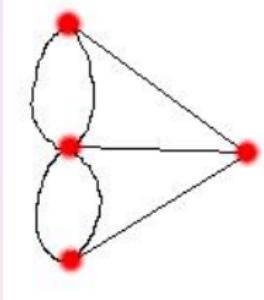
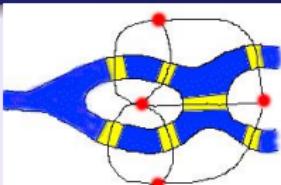
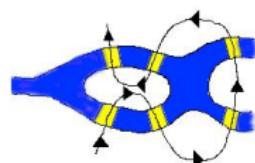


- ▼ 7 bridges



► NO soln!!!

- ▼ 6 bridges

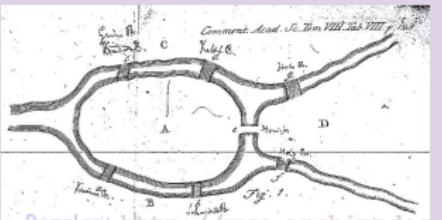
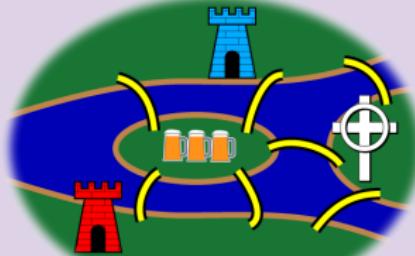


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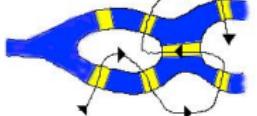
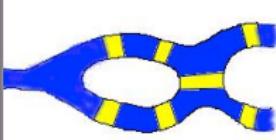
# Network Topology

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- ▼ 7 bridges of Königsberg [Euler, 1736]

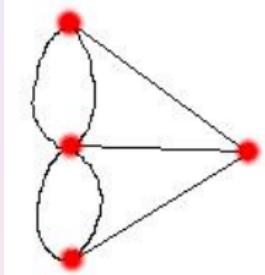
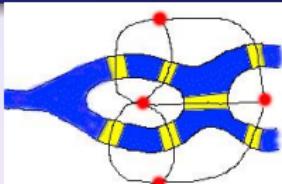
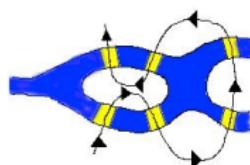


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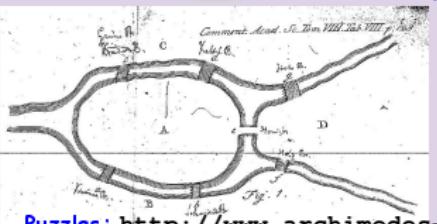
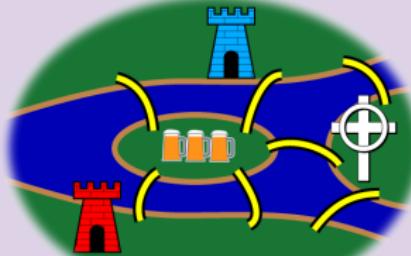


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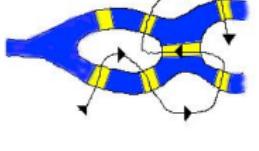
# Network Topology

## Graph Theory

- ▼ 7 bridges of Königsberg [Euler, 1736]

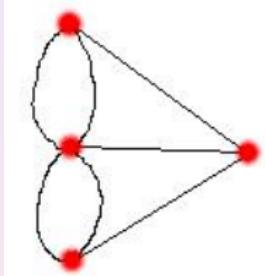
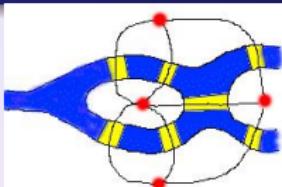
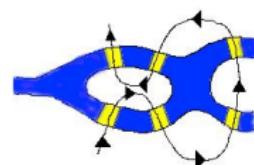


- ▼ 7 bridges



► NO soln!!!

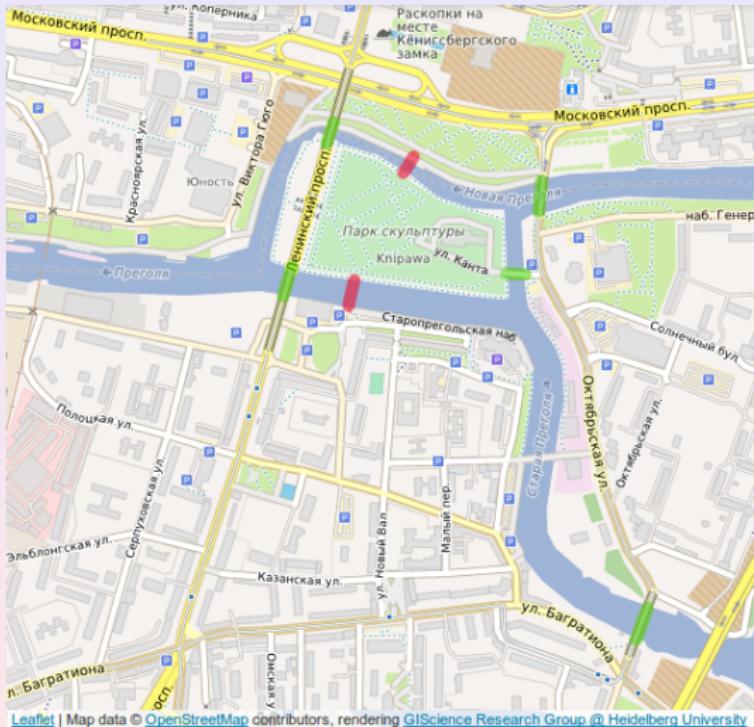
- ▼ 6 bridges



If a network has more than two odd vertices, it does NOT have an Euler path.

Puzzles: [http://www.archimedes-lab.org/How\\_to\\_Solve/Water\\_gas.html](http://www.archimedes-lab.org/How_to_Solve/Water_gas.html)  
<http://www.mathsisfun.com/activity/seven-bridges-konigsberg.html>

# Present state of the seven bridges of Königsberg



Src: [http://en.wikipedia.org/wiki/Seven\\_Bridges\\_of\\_K%C3%B6nigsberg](http://en.wikipedia.org/wiki/Seven_Bridges_of_K%C3%B6nigsberg)

# Network Topology

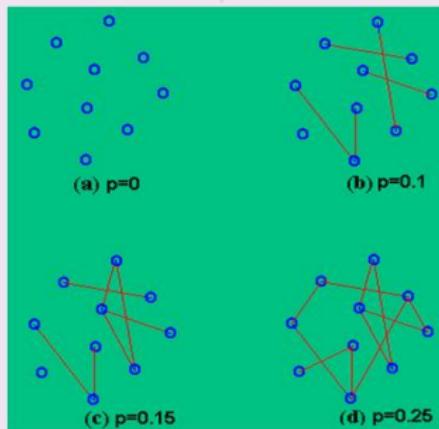
how nodes are connected

- According to the topology is possible to "classify" the networks.

Distribution of links  $P(k)$  (degree distrib), avg. nbr. of links  $\langle k \rangle$ , clustering coeff, avg. length  $\langle \ell \rangle$  (diam), ...

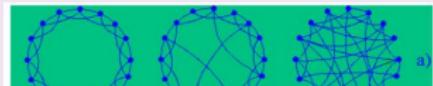
▼ random networks

► Erdős-Rényi (1960)



▼ regular or uniform

► globally coupling, nearest-neighbors, ...



rewiring of links

▼ irregular or heterog.

► non-uniform distrib. (star), ...



e.g. airlines routes

# Network Topology

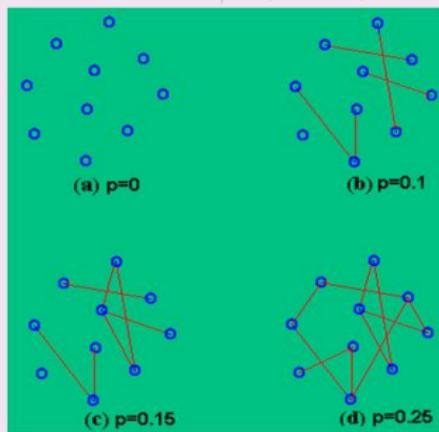
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# Network Topology

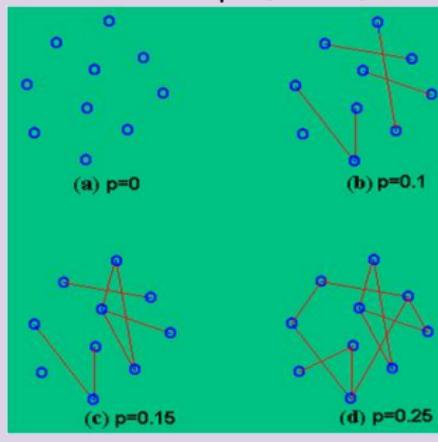
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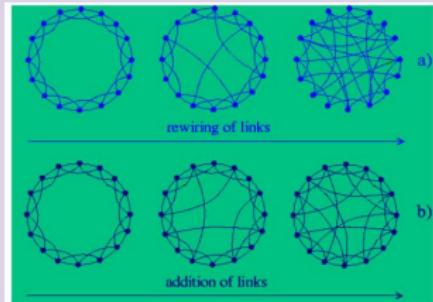
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# Complex Networks I

Properties of «REAL» networks

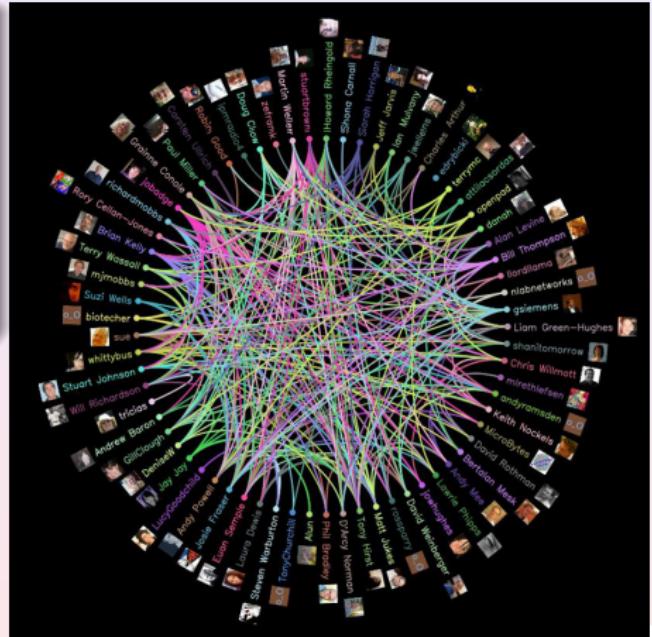
## ► Small-World [Watts and Strogatz, Nature, 1998]

most nodes can be reached by a small number of steps («6 dof»!)

- **Connectivity distribution:** uniform but decays exponentially
- **Homogeneous nature:** each node has roughly the same number of links



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# Complex Networks II

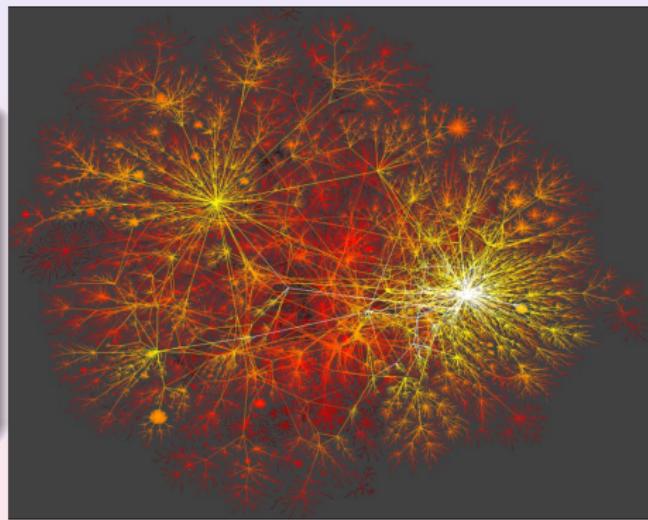
## Properties of «REAL» networks

► Scale-Free [Barabási and Albert, Science, 1999]

some nodes act as “highly connected hubs”, and most nodes are of low degree distribution:

$$P(k) = k^{-\gamma}$$

$2 \leq \gamma \leq 3$  (power law)



## MultiGraph: multiple edges

```
mg = nx.MultiGraph()
mg.add_weighted_edges_from([(1,2, .5), (1,2, .75), (2,3, .5)])
mg.degree()
{1: 2, 2: 3, 3: 1}
mg.degree(weight='weight')
{1: 1.25, 2: 1.75, 3: 0.5}
```

## DiGraph: directed graphs

```
dg = nx.DiGraph()
dg.add_weighted_edges_from([(1,4,0.5), (3,1,0.75)])
dg.degree(1,weight='weight')
1.25
dg.successors(1)
[4]
dg.predecessors(1)
[3]
```

## More graphs generators

```
K_5=nx.complete_graph(5)
nx.draw(K_5)

K_3_5=nx.complete_bipartite_graph(3,5)
nx.draw(K_3_5)

barbell=nx.barbell_graph(10,10)
nx.draw(barbell)

lollipop=nx.lollipop_graph(10,20)
nx.draw(lollipop)
```

## Random graphs generators

```
er=nx.erdos_renyi_graph(100,0.15)
nx.draw(er)

ws=nx.watts_strogatz_graph(30,3,0.1)
nx.draw(ws)

ba=nx.barabasi_albert_graph(100,5)
nx.draw(ba)

red=nx.random_lobster(100,0.9,0.9)
nx.draw(red)
```



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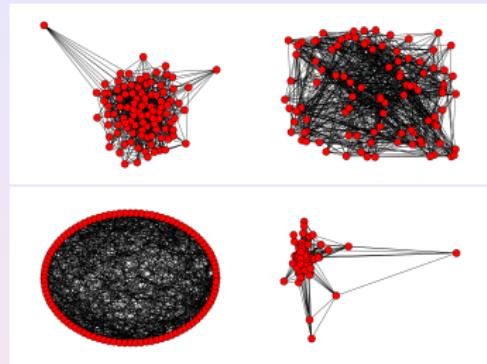
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# Visualization & Graph properties...

## Visualization

```
import networkx as nx
import pylab as plt
g = nx.erdos_renyi_graph(100, 0.15)
nx.draw(g)
nx.draw_random(g)
nx.draw_circular(g)
nx.draw_spectral(g)
plt.savefig('graph.png')
```



## Graph properties

```
N,K = g.order(), g.size()
avg_deg = float(K)/N
print "Nodes:", N
print "Edges:", K
print "Average_degree:", avg_deg
```

## Degree distribution

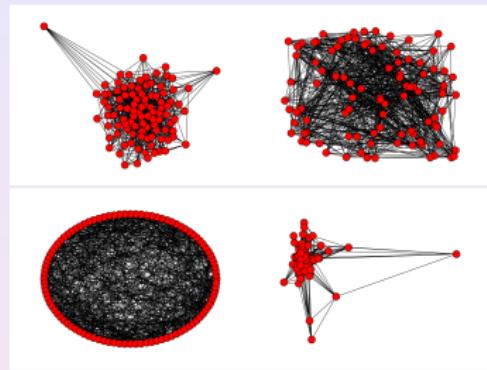
```
g.in_degrees()
```

```
Traceback (most recent call last):
  File "<stdin>", line 1, in <module>
AttributeError: 'Graph' object has no attribute 'in_degrees'
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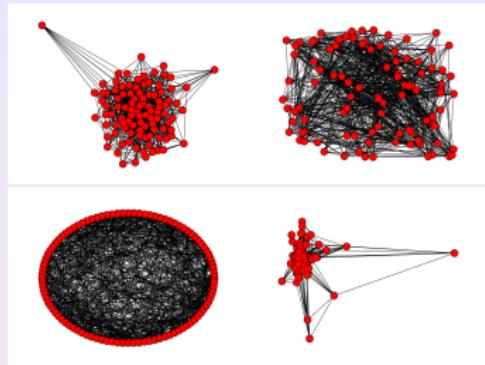
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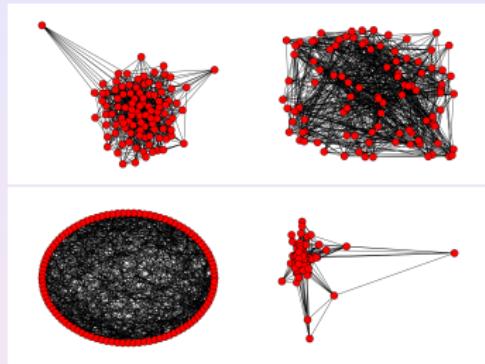
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```
Traceback (most recent call last):
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AttributeError: 'Graph' object has no attribute 'in_degrees'
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# Graph properties: degree distribution ...

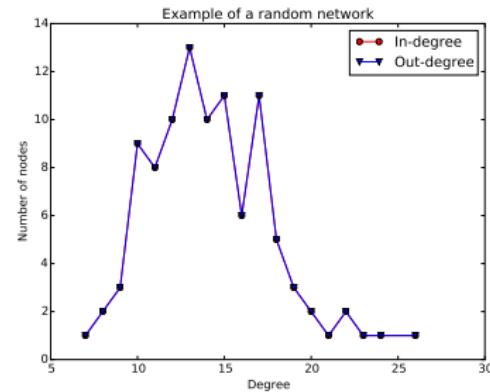
## Degree distribution

```

diG=nx.DiGraph(g) #create a directed graph
# account for in/out-going connections...
in_degrees = diG.in_degree() # dictionary node:degree
out_degrees = diG.out_degree() # dictionary node:degree
in_values = sorted(set(in_degrees.values()))
out_values = sorted(set(out_degrees.values()))
# put data in histogram-form
in_hist = [in_degrees.values().count(x) for x in in_values]
out_hist = [out_degrees.values().count(x) for x in out_values]

# plot
plt.figure()
plt.plot(in_values,in_hist,'ro-') # in-degree
plt.plot(out_values,out_hist,'bv-') # out-degree
plt.legend(['In-degree','Out-degree'])
plt.xlabel('Degree')
plt.ylabel('Number_of_nodes')
plt.title('Example_of_a_random_network')
plt.savefig('network_degree_distribution.pdf')
plt.close()

```

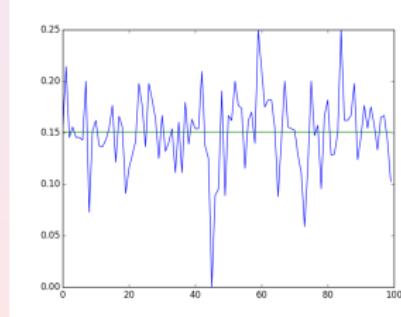
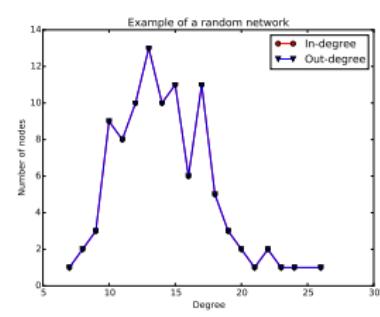
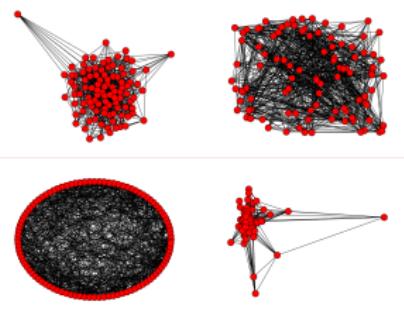


# Graph properties: connectivity properties ...

## Clustering Coefficient

```
import numpy as np
g_ud = diG.to_undirected()
# Clustering coefficient of node 0
print nx.clustering(g_ud, 0)
# Clustering coefficient of all nodes (in a dictionary)
clust_coefficients = nx.clustering(g_ud)
# Average clustering coefficient
ccs = nx.clustering(g_ud)
avg_clust = sum(ccs.values()) / len(ccs)

plt.plot(ccs.keys(),ccs.values())
plt.plot(ccs.keys(),np.ones(len(ccs))*avg_clust)
```



# Modeling Twitter Relationships I

## "Following"

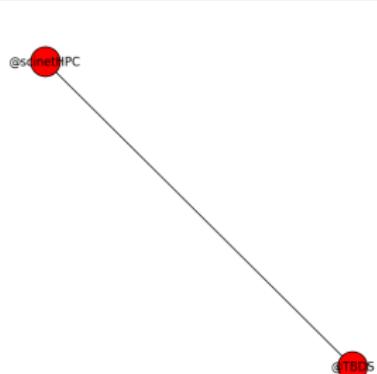
```
import networkx as nx
import pylab as plt

g = nx.Graph() #nx.DiGraph()
g.add_node(0, {"handle": "@scinetHPC"})
g.add_node(1, {"handle": "@TBDS"})
g.add_edge(0, 1)

node_labels = nx.get_node_attributes(g, 'handle')

plt.figure(figsize=(6, 6))
plt.ion()

pos = nx.spring_layout(g)
nx.draw(g, pos, arrows=True, node_size=900)
nx.draw_networkx_labels(g, pos, labels = node_labels)
```



# Modeling Twitter Relationships II

## "Re-Tweets"

```
g = nx.DiGraph()
g.add_node(0, {"id": "@scinetHPC", "color": "r"})
g.add_node(1, {"id": "@TBDS", "color": "r"})
g.add_node(2, {"id": "Tweet_#530038648860065793", "color": "orange"})
g.add_edge(0, 1, {"type": "follows"})
g.add_edge(0, 2, {"type": "retweeted"})

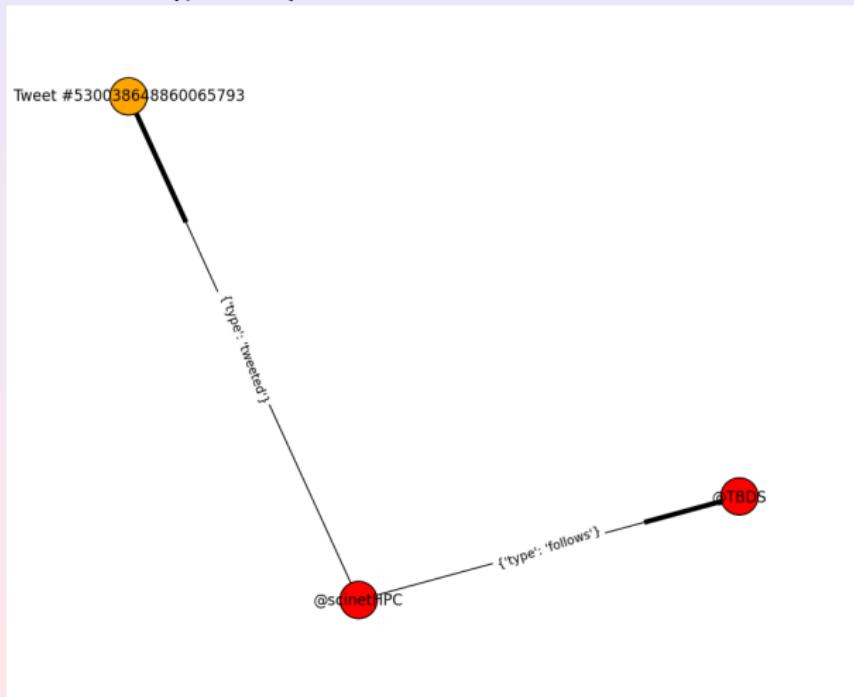
node_labels = nx.get_node_attributes(g, 'id')
edge_labels = nx.get_edge_attributes(g, 'type')
colors = nx.get_node_attributes(g, 'color')

plt.figure(figsize=(6, 6))

pos = nx.spring_layout(g)
nx.draw(g, pos, arrows=True, node_size=900, node_color=colors.values())
nx.draw_networkx_labels(g, pos, labels = node_labels)
nx.draw_networkx_edge_labels(g, pos, labels = edge_labels)
```

# Modeling Twitter Relationships III

## A model for the Twitter network



# Modeling the I/O-ops @ SciNet I

## Data

```
tcs-f06n01  tcs-f06n01  0.013554078
tcs-f06n01  tcs-f06n01  0.004655432
tcs-f07n02  gpc-f119n017  0.002597279
tcs-f07n02  tcs-f07n02  0.000283613
tcs-f06n02  tcs-f06n02  0.000703586
tcs-f06n02  tcs-f06n02  0.001286946
tcs-f07n01  tcs-f07n01  0.025607990
tcs-f07n01  tcs-f07n01  0.018086633
tcs-f07n01  tcs-f07n01  0.003623529
tcs-f06n04  tcs-f06n04  0.012281074
tcs-f06n04  tcs-f06n04  0.012600201
tcs-f06n03  tcs-f06n03  0.006405814
tcs-f06n03  tcs-f06n03  0.024002783
tcs-f06n03  tcs-f06n03  0.019571064
tcs-f07n04  tcs-f07n04  0.034180323
tcs-f07n04  tcs-f07n04  0.100195957
```

449 records

## ▼ networkX.read\_edgelist(...)

```
g =
nx.read_format("path/to/file.txt",...options...
nx.write_format(g,"path/to/file.txt",...options...
```

## ► Read and write edge lists

```
g = nx.read_edgelist(path,comments='#',
                     create_using=None,
                     delimiter=' ',nodetype=None,
                     data=True,edgetype=None,
                     encoding='utf-8')

nx.write_edgelist(g,path,comments='#',
                  delimiter=' ',data=True,
                  encoding='utf-8')
```



# Modeling the I/O-ops @ SciNet II

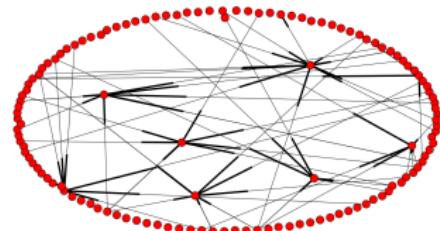
## Importing and plotting the data

```
import networkx as nx
import pylab as plt

scinetG = nx.read_edgelist("scinet_data.txt",
    comments='#',
    create_using=nx.MultiDiGraph(),
    delimiter=' ',
    nodetype=str,data=((('weight',float),) ) )

plt.ion()
nx.draw(scinetG)

plt.figure()
```



# Modeling the I/O-ops @ SciNet III

## "Massaging" the network (data)

```
plt.figure()
plt.axis('off')

cutOff=0.001
elarge=[(u,v) for (u,v,d) in scinetG.edges(data=True) if d['weight'] > cutOff]
esmall=[(u,v) for (u,v,d) in scinetG.edges(data=True) if d['weight'] <= cutOff]

#try different ones from the ones below...
#pos=nx.spring_layout(scinetG) # positions for all nodes
pos=nx.graphviz_layout(scinetG)
#pos=nx.shell_layout(scinetG)
#pos=nx.random_layout(scinetG)
#pos=nx.spectral_layout(scinetG)

# nodes
nx.draw_networkx_nodes(scinetG,pos,
#                         node_color=nodeType.values(),
#                         node_size=tmW.values(),
#                         cmap=plt.cm.Reds, alpha=0.7)

# edges
color_edges=range(scinetG.number_of_edges())
nx.draw_networkx_edges(scinetG,pos,edgelist=elarge,
                      width=3,alpha=0.7,edge_color='r',
                      with_labels='True',edge_cmap=plt.cm.Blues)
nx.draw_networkx_edges(scinetG,pos,edgelist=esmall,
                      width=2,alpha=0.4,edge_color='b',style='dotted')
```

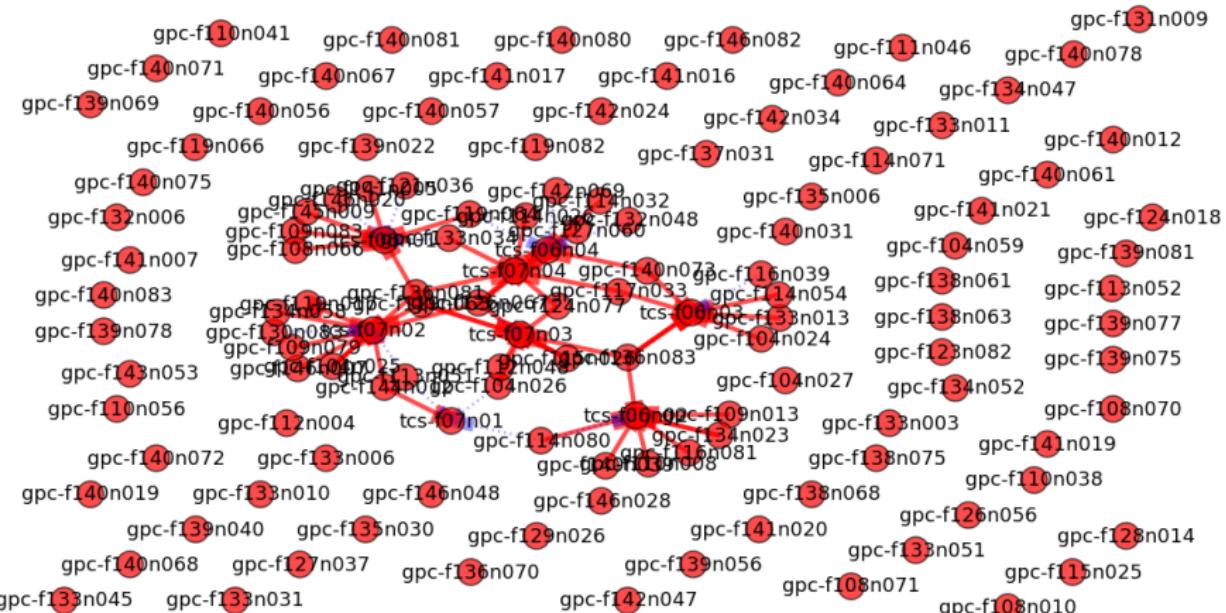


# Modeling the I/O-ops @ SciNet IV

```
# labels
try:
    nx.draw_networkx_labels(scinetG, pos, labels, font_size=13,
                           font_family='sans-serif', font_color='Indigo')
except:
    nx.draw_networkx_labels(scinetG, pos, font_size=13, font_family='sans-serif')
```

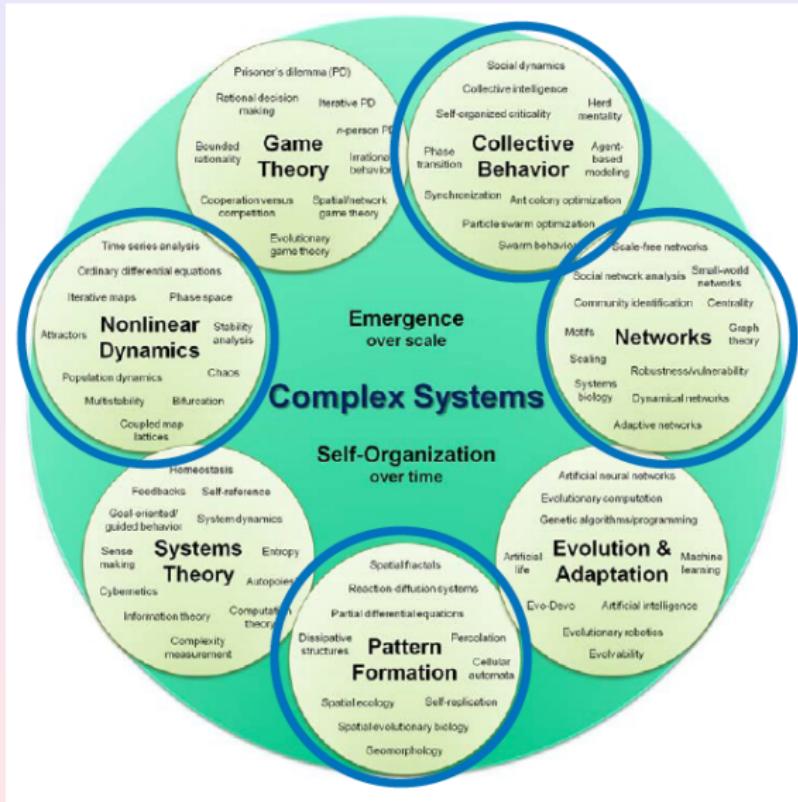


# Modeling the I/O-ops @ SciNet V



# PyCX: Python-based Complex systems simulations

Online repository: <http://pycx.sf.net/>



# Further Resources & Examples

- ▼ NetworkX: Network Analysis with Python, Salvatore Scellato  
<http://www.cl.cam.ac.uk/~cm542/teaching/2011/stna-pdfs/stna-lecture11.pdf>
- ▼ Graph Theory, including modelling of relationships, tweets, how to build a NetworkX graph from Twitter crawl results ...  
[http://nbviewer.ipython.org/github/jquacina/IS609\\_Group/blob/master/GraphTheory/Graph%20Theory%20-%20Social%20Networking.ipynb](http://nbviewer.ipython.org/github/jquacina/IS609_Group/blob/master/GraphTheory/Graph%20Theory%20-%20Social%20Networking.ipynb)
- ▶ Graph and Network Analysis  
<http://mlg.ucd.ie/files/summer/tutorial.pdf>
- ▶ Social Network Analysis in Python  
<https://ep2013.europython.eu/media/conference/slides/social-network-analysis-in-python.pdf>
- ▶ Tutorial on PyCX  
<http://pycx.sourceforge.net/PyCX-ECAL2013-tutorial.pdf>

# More about Complex Networks

## NECSI:

<http://necsi.edu/> <http://www.necsi.edu/publications/dcs/index.html>

<http://www.necsi.edu/events/vidlib/> <http://www.necsi.edu/research/overview.php>

## Some other good references:

<http://www-personal.umich.edu/~mejn/courses/2004/cscs535/review.pdf>

[http://www.ifr.ac.uk/netsci08/Download/Invited/ws1\\_Caldarelli.pdf](http://www.ifr.ac.uk/netsci08/Download/Invited/ws1_Caldarelli.pdf)

<http://barabasilab.neu.edu/courses/phys5116/>

## SantaFe Institute:

[http:](http://www.santafe.edu/education/schools/complex-systems-summer-schools/2015-program-info/)

[//www.santafe.edu/education/schools/complex-systems-summer-schools/2015-program-info/](http://www.santafe.edu/education/schools/complex-systems-summer-schools/2015-program-info/)

