



DIAGNOSING VULNERABILITY, EMERGENT PHENOMENA,
and VOLATILITY in MANMADE NETWORKS

Modelling critical network infrastructures

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Collaborators

- Collegium Budapest
- European Commission, JRC(Ispra)
- Macedonian Academy of Sciences and Arts
- Queen Mary University of London
- Università Carlo Cattaneo
- Stakeholders
 - National Emergency Supply Agency, FINGRID (Finland)
 - UK-Electricité de France

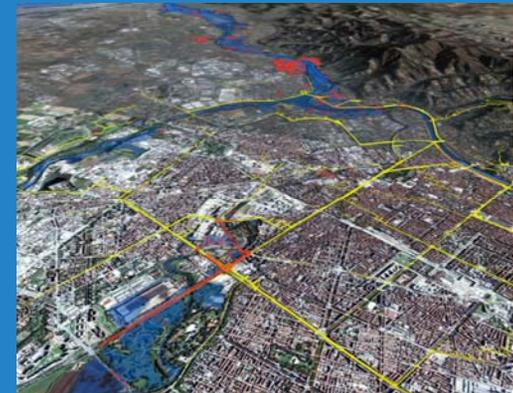
NETWORKS

- Social
- Energy
- Transport
- Communication
-compound networks

What are the correct questions to ask and the most useful analytical tools to handle them?

MANMADE – its scope

- The project concerns the compound networks that comprise Europe's critical infrastructure -
 - primarily on energy supply, emergency response systems and subsidiary key infrastructures
- Aim -
 - is to assemble, develop and apply mathematical methods to analyse large, man-made multi-element infrastructure systems



First steps - datasets

Data sets of major gas lines and exchange flows

Data sets of major gas lines between and into Western Europe

Datasets of spot price electricity

NORDPOOL time series spot price electricity in European markets

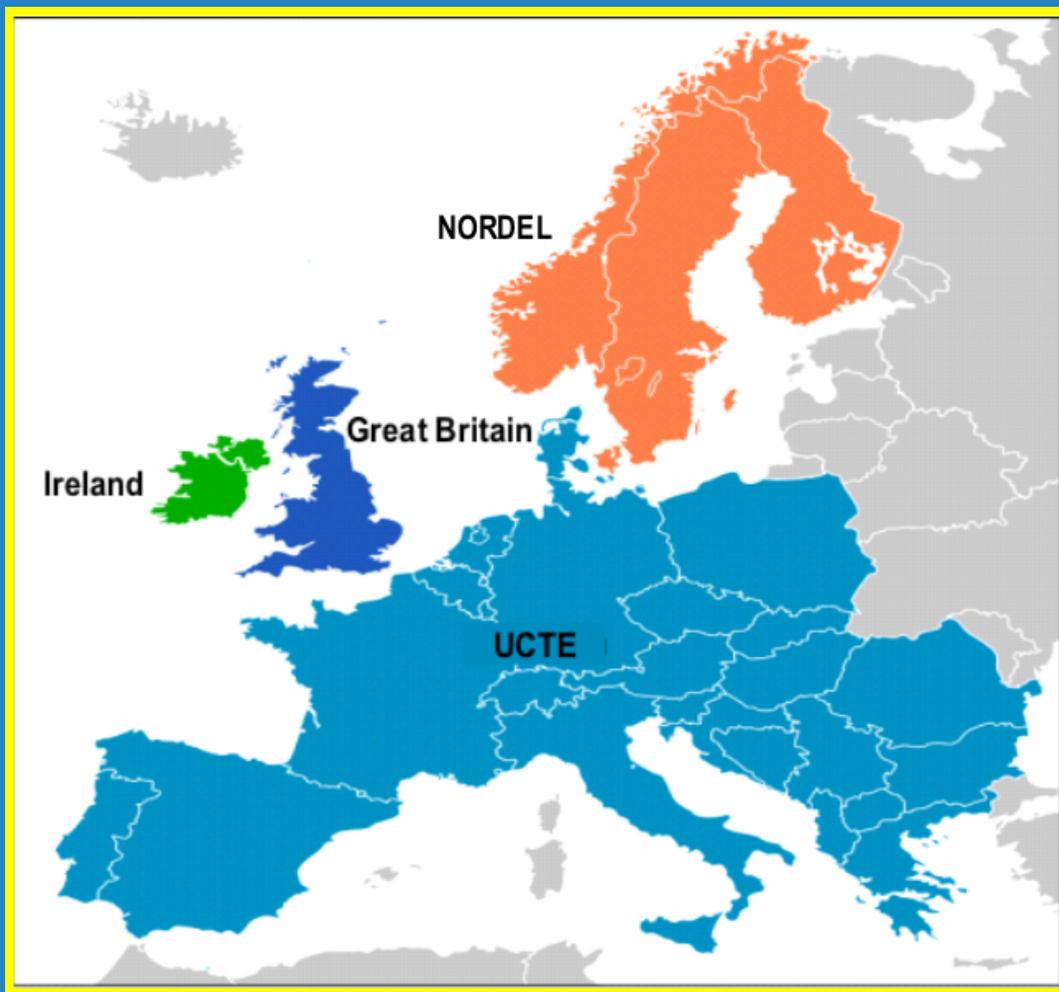
Spatial and topological maps of the road network

Urban street network of Milan, Turin and London

High voltage electricity grid

European Electricity Lines by synchronous regions
European energy interconnected network (electricity and gas)

Electricity regions



Four independent synchronous electricity regions to be investigated

Network

- Scaling
- Topology
- Weighting

Electricity networks

NORDEL:

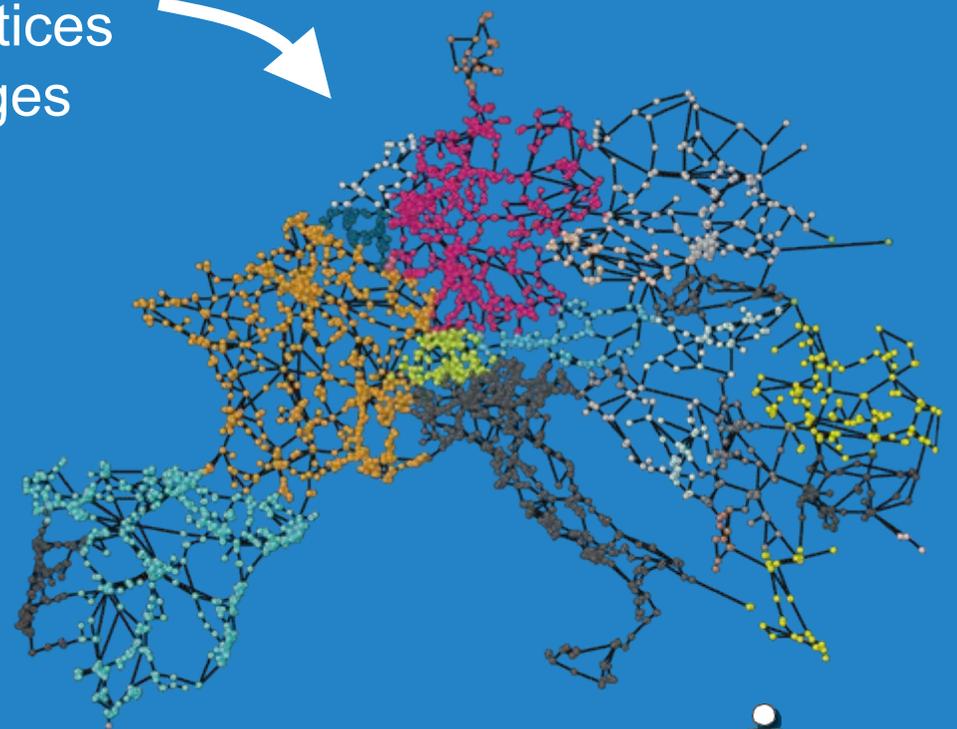
526 vertices

638 edges

UCTE:

4200 vertices

5305 edges



Vulnerability

The more critical the component the more severe is the damage to the system when it is lost

Criticality of the component



Vulnerability of the system

Critical component

=

WEAK

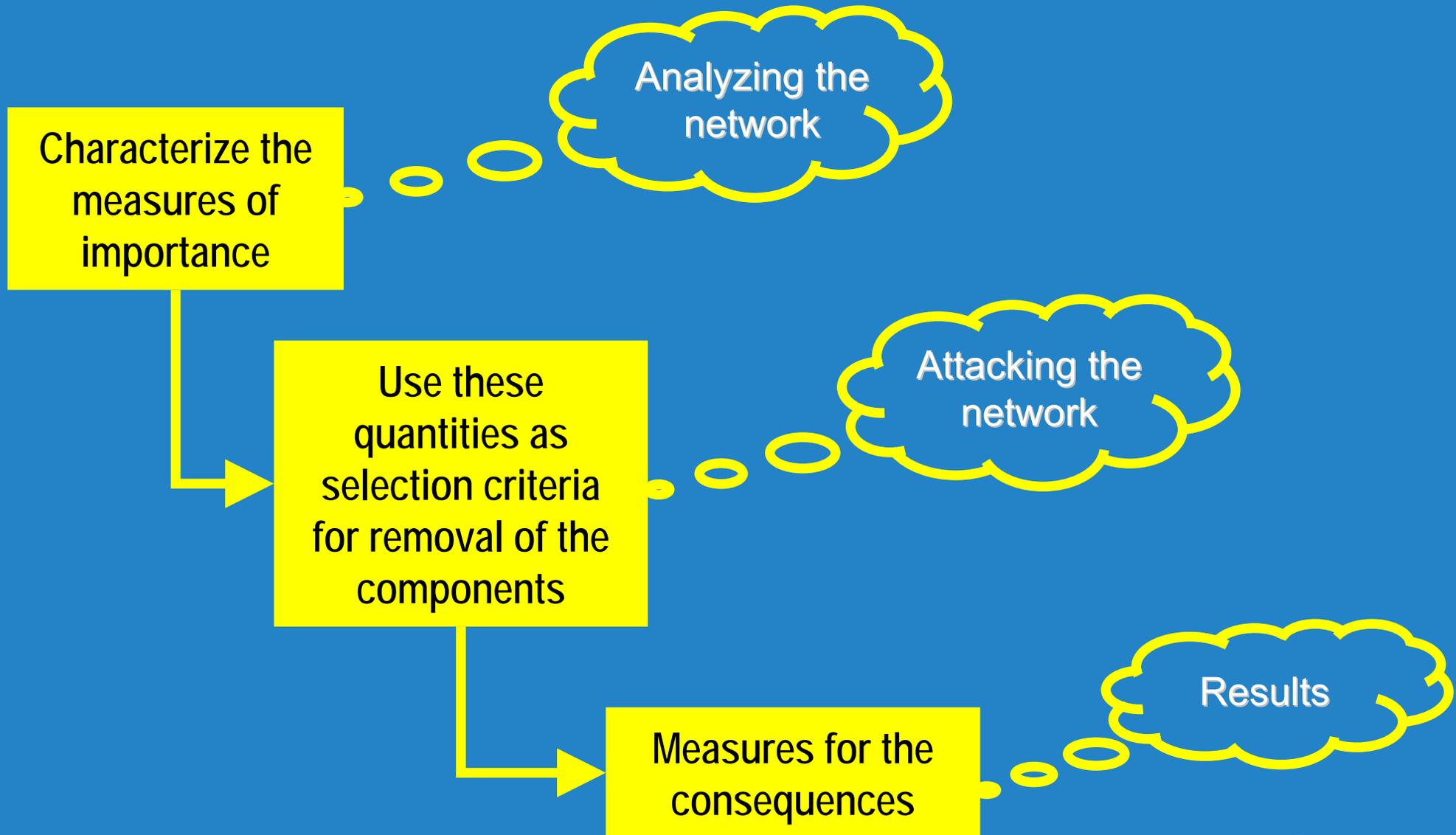
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IMPORTANT

How susceptible to malfunction?

The consequences are great

How to tackle this issue?



Network graphs and incidence matrix

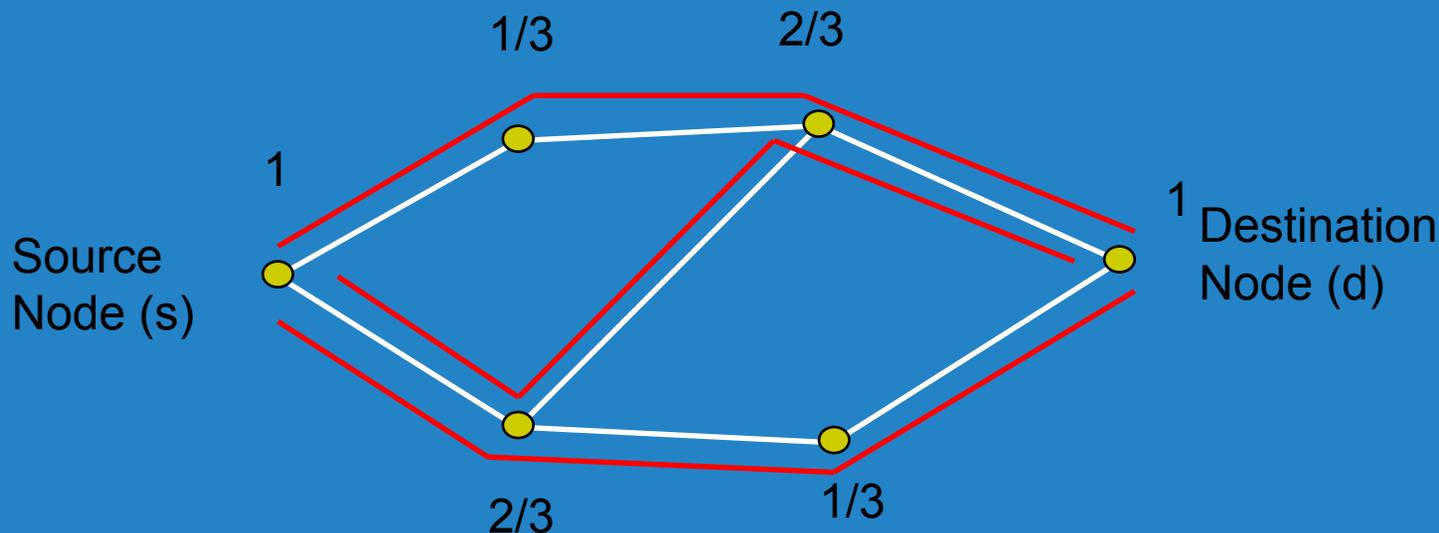
- **Graph G** consists of a set of N nodes $V=\{v_1, v_2, v_3, \dots, v_N\}$ and a set of M edges $E=\{e_1, e_2, e_3, \dots, e_M\}$
- An **edge** connects two nodes $e=(v_i, v_j)$ – there can be multi-edges
- The **incidence** between the edges and the nodes is recorded via an **adjacency matrix** $A=(a_{ij})$
- If edge j is joined to node i , then $a_{ij}=1$, otherwise 0
- The **degree** of node i is the # of '1' 's in row i of A
- The indicators $a_{ij}=1$ or 0 can be changed to **weights** w_{ij} where if $a_{ij}=0$, then $w_{ij}=0$
- The **Laplacian matrix** $L=(l_{ij})$ is an $N \times N$ matrix with $L_{ij}=\text{deg}(i)$ for $i=j$, $L_{ij}=-1$ for $i \neq j$ and v_i is adjacent by an edge to v_j , otherwise 0.

Relevance of the Laplacian matrix

- The *algebraic connectivity* of a graph G is the second-smallest eigenvalue of the Laplacian Matrix of G
- This eigenvalue is greater than 0 if and only if G is a connected graph
- This is a corollary to the fact that the number of times 0 appears as an eigenvalue in the Laplacian is the number of connected components in the graph
- The magnitude of this value reflects how well connected the overall graph is, and has implications for properties such as synchronizability and clustering

Betweenness centrality

- *Betweenness centrality* measures the importance of nodes in terms of the frequency of their appearance on shortest paths



- *Betweenness* of node i for the pair sd = relative number of shortest paths between s and d which visit node i
- *Betweenness* of node i for **graph** requires calculation for all pairs of vertices

Analysis of weighted undirected networks

Graph -> Network



$$G = (V, E)$$

Vertices (substations)

Edges (electrical lines)

Adjacency matrix

$$A(G)$$

defines which vertices are connected

$$a_{ij} = \begin{cases} 1, & \text{if } v_i \text{ and } v_j \text{ are neighbours,} \\ 0, & \text{otherwise} \end{cases}$$

Weights matrix

$$W(G)$$

defines the weights on the edges

$$w_{ij} = \begin{cases} w_{ij}, & \text{if } a_{ij} = 1, \\ 0, & \text{if } a_{ij} = 0 \end{cases}$$

Weights

=

Capacity of the electrical line (kV)

[220kV – 2500kV]

Measures of importance

Inverse "distance" from the given node to every other node in network - efficiency

$$C_i = \sum_j \frac{1}{d_{ij}}$$

Efficiency / Closeness

Degree

Betweenness

shortest paths in the graphs

d_{ij}

$$B_i = \sum_v \sum_u \frac{n_{uv}(i)}{n_{uv}}$$

The proportion of all shortest paths in the network that run through a given node

Centrality indices

Modal index

$$D_i = \sum_j a_{ij}$$

or

$$D_i = \sum_j w_{ij}$$

Spectral decomposition of the Laplacian matrix

$$L = D - W \rightarrow \lambda_i \phi_i = L \phi_i$$

$$\Gamma = L' \Phi \rightarrow M_i = \sum_j |\gamma_{ij}|$$

Sum of modal contributions (load in each mode) for each vertex

Weighted vs. unweighted node ranking

of Nordel network

Spearman ranking correlation coefficients for
Efficiency, Betweenness and Spectral analysis

NORDEL

	Eff-W	Eff-UnW	BC-W	BC-UnW	SA-W	SA-UnW
Eff-W	1.00	0.89	0.34	0.32	0.22	0.58
Eff-UnW	0.89	1.00	0.42	0.44	0.55	0.72
BC-W	0.34	0.42	1.00	0.96	0.51	0.61
BC-UnW	0.32	0.44	0.96	1.00	0.57	0.62
SA-W	0.22	0.55	0.51	0.57	1.00	0.76
SA-UnW	0.58	0.72	0.61	0.62	0.76	1.00

WEIGHTED



UNWEIGHTED

BETWEENNESS



EFFICIENCY



SPECTRAL ANALYSIS



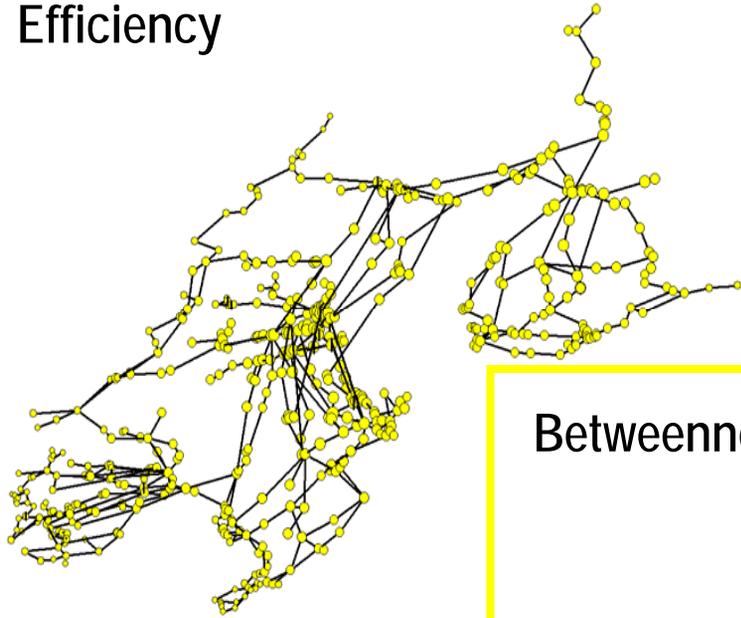
Ranking

Size of the vertices
Width of the edges



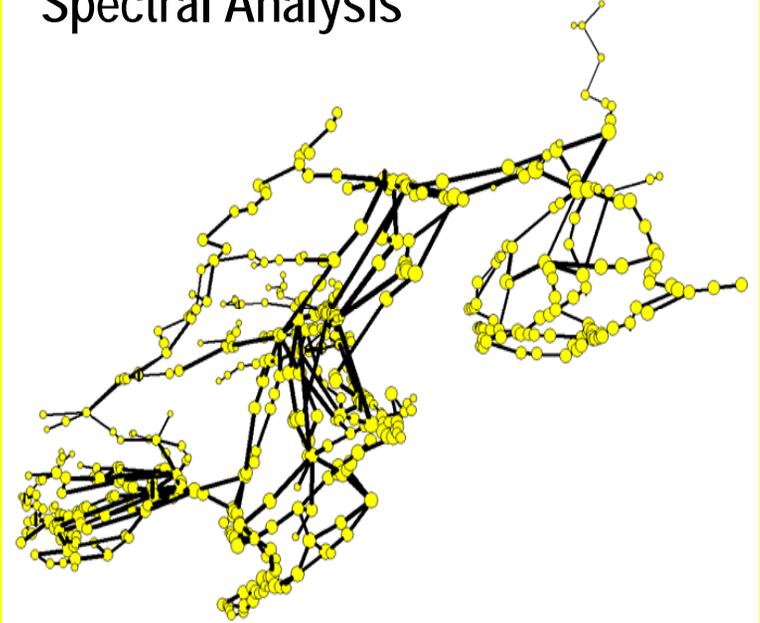
the relative value of
the importance

Efficiency



NORDEL

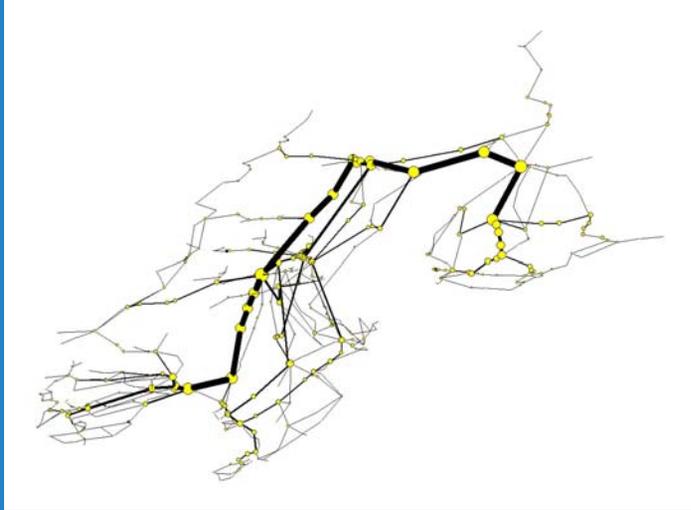
Spectral Analysis



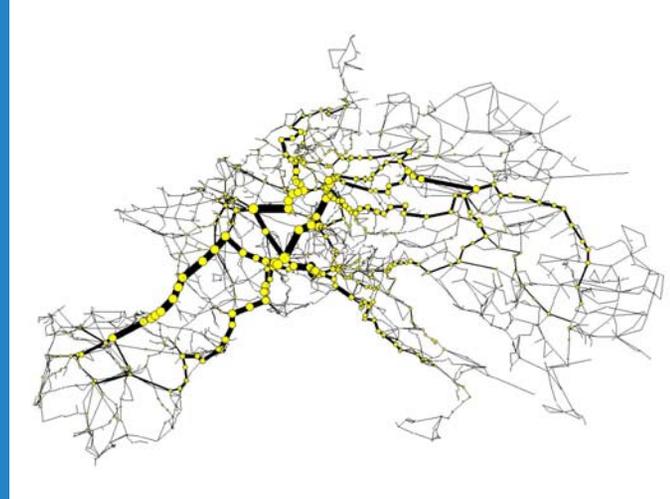
Betweenness



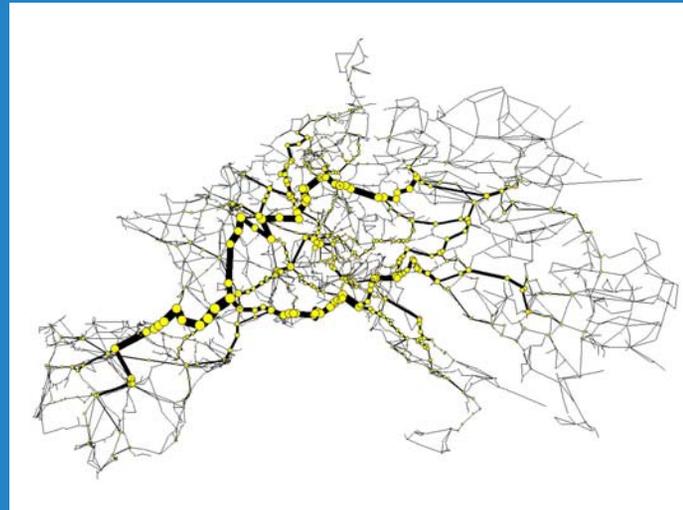
Efficiency –choice of metrics



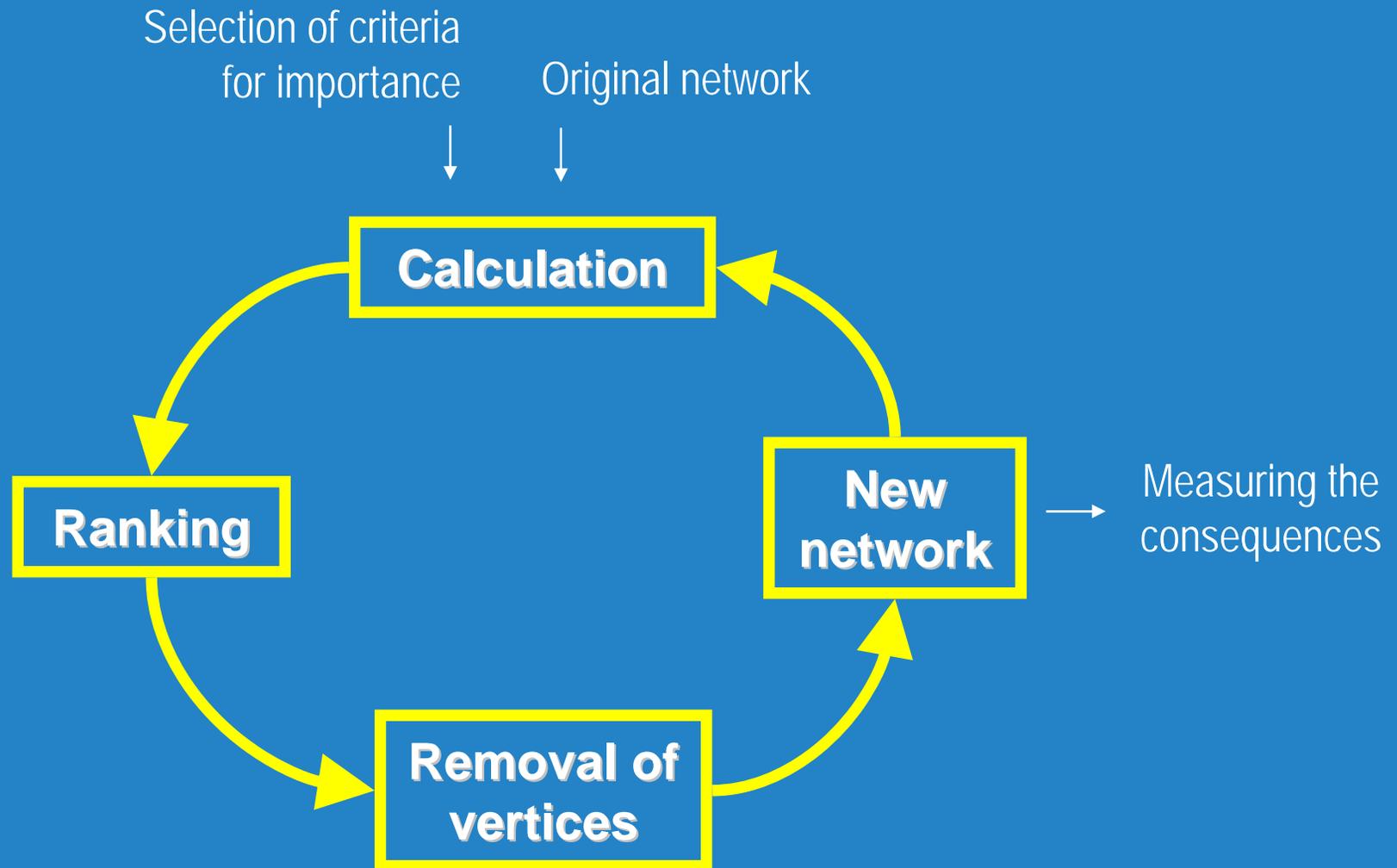
NORDEL



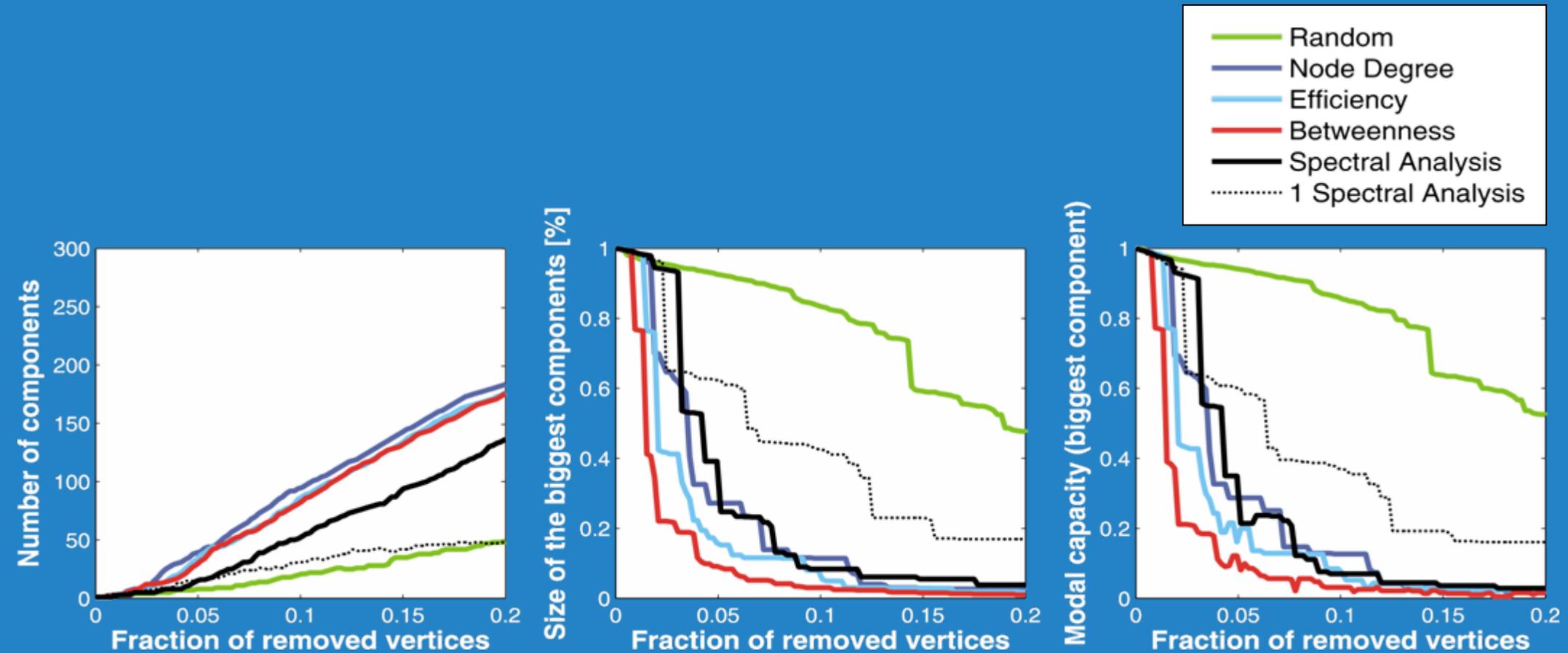
UCTE(UNION COORD TRANS ELECTRICITY)



Algorithm of attacking the network

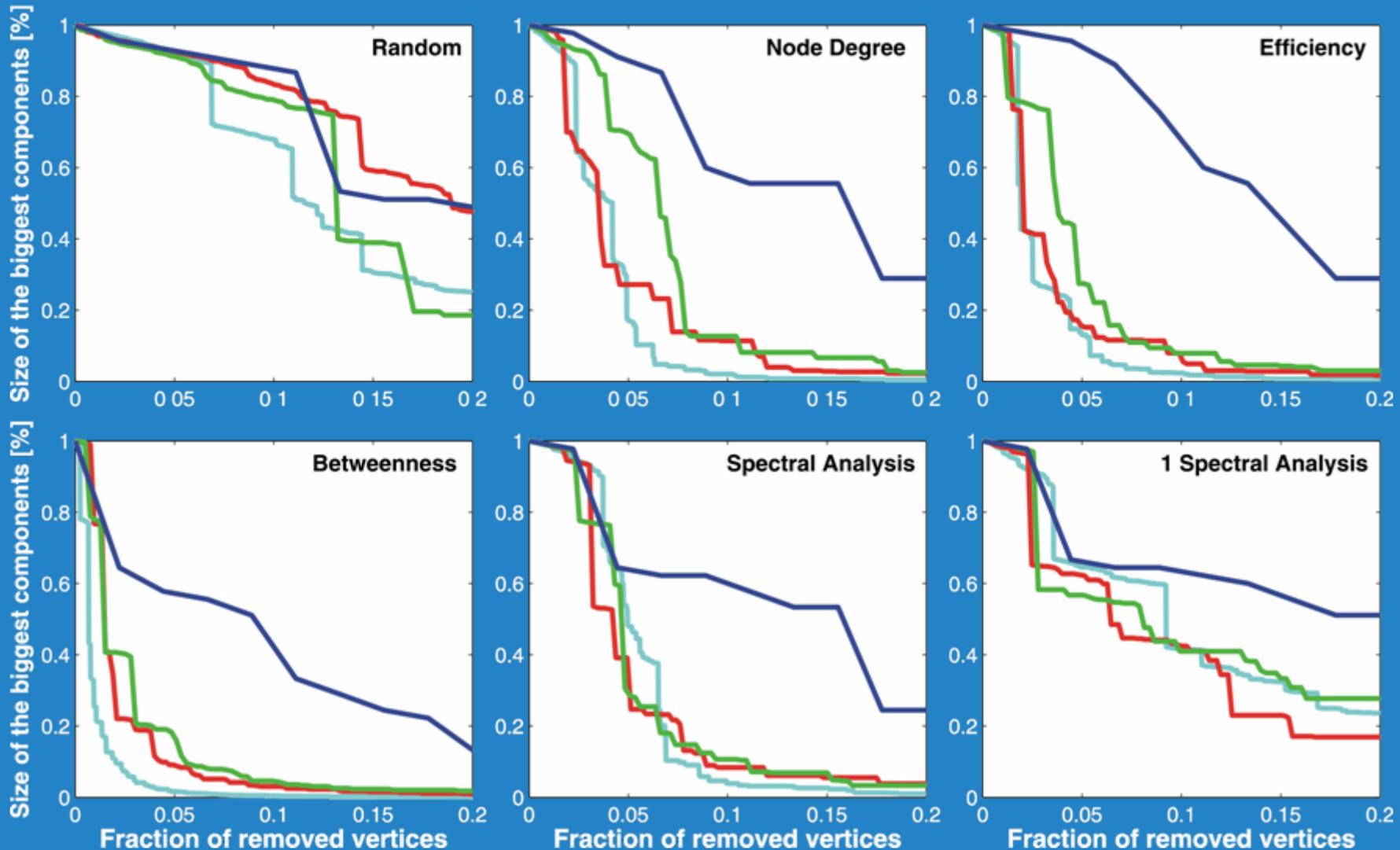


Measuring the consequences

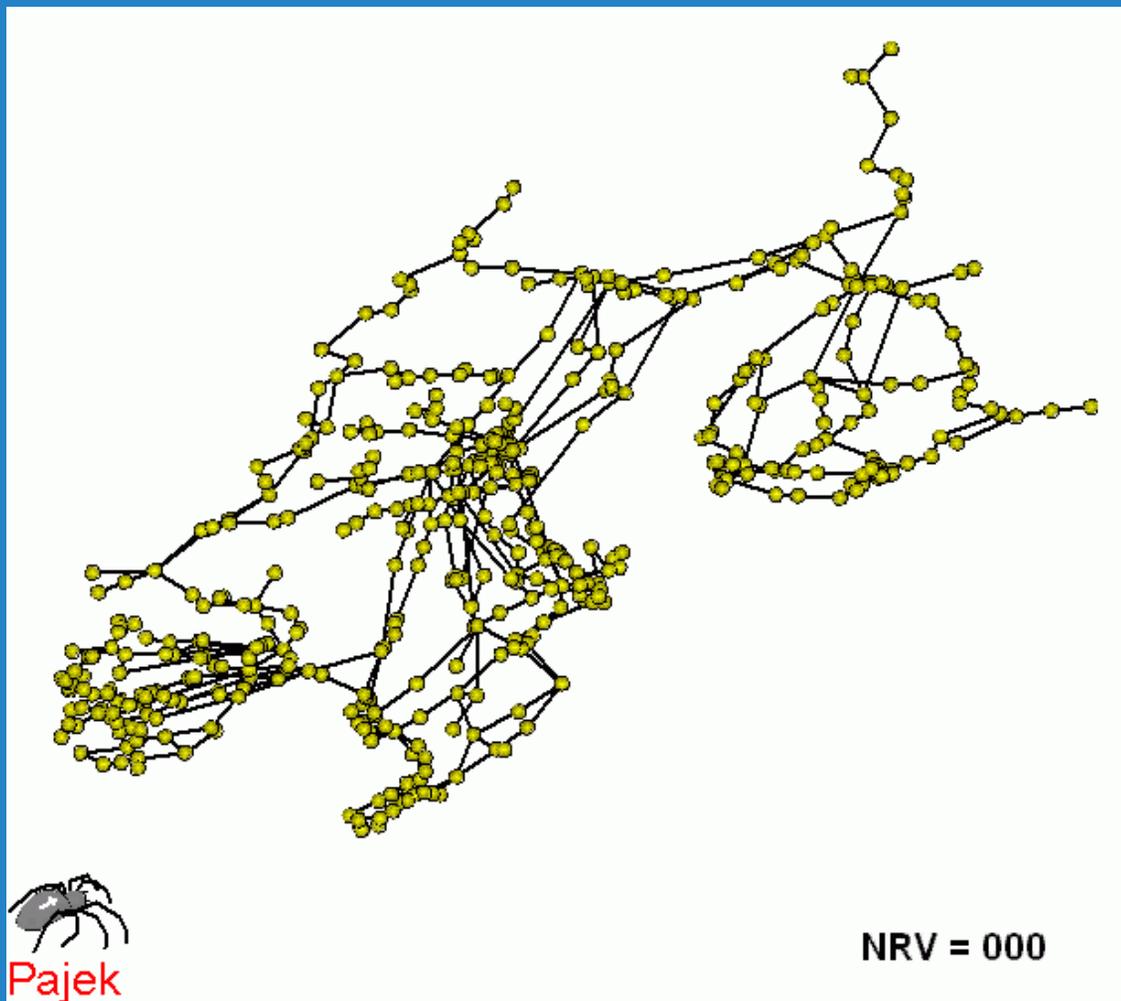


NORDEL

Measuring the consequences



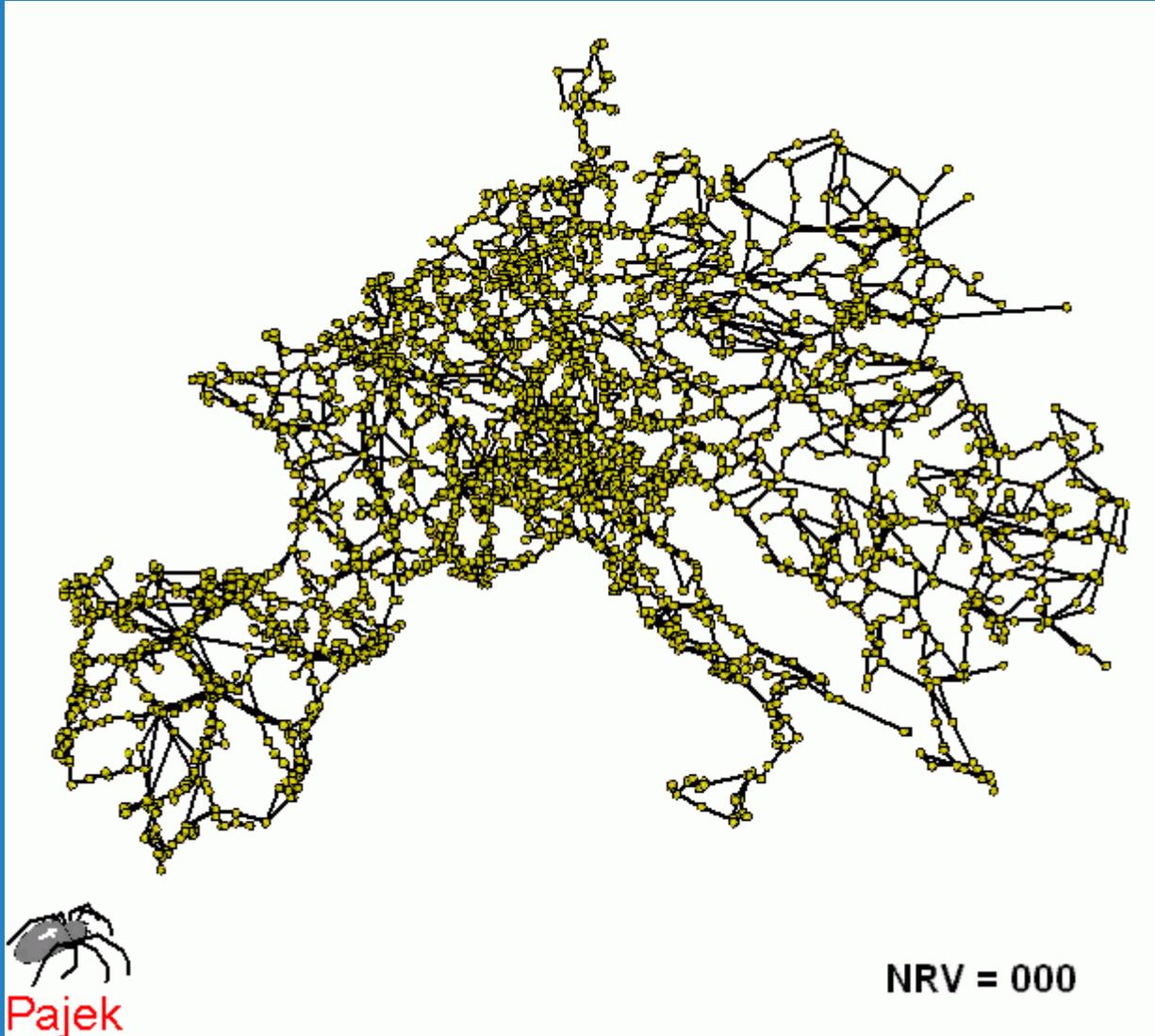
BREAK-UP OF THE NETWORK



Betweenness centrality
NORDEL

NRV – number of removed vertices

BREAK-UP OF THE NETWORK



Betweenness centrality
UCTE

NRV – number of removed vertices

FASTEST DECAY OF NETWORK

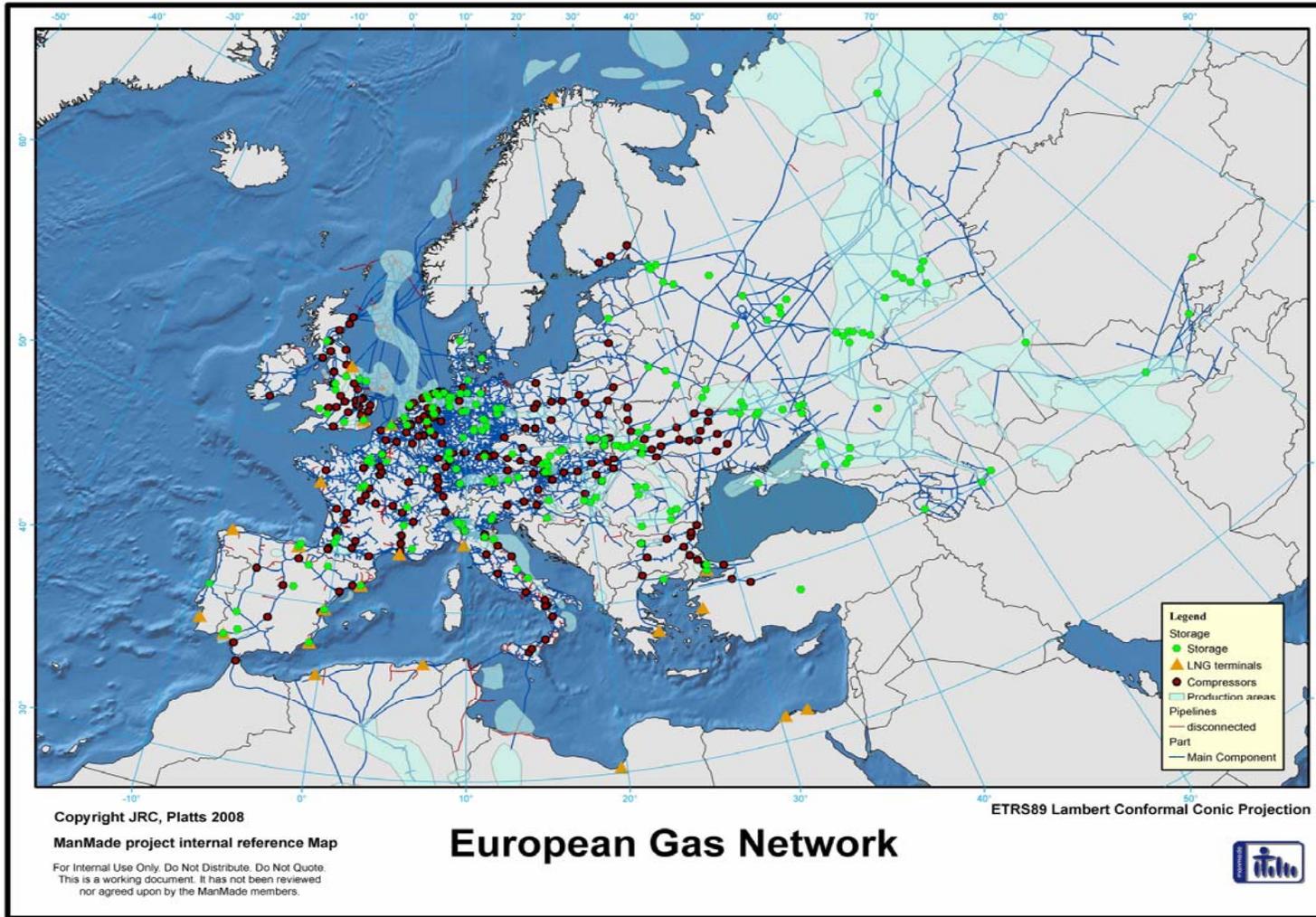
SIMULATION OF THE NETWORK ATTACK

- Removal of random nodes according to a ranking
- Decay is broadly comparable for the different types of status BC, Eff, SA
- Fastest first level fragmentation is by using the measure –

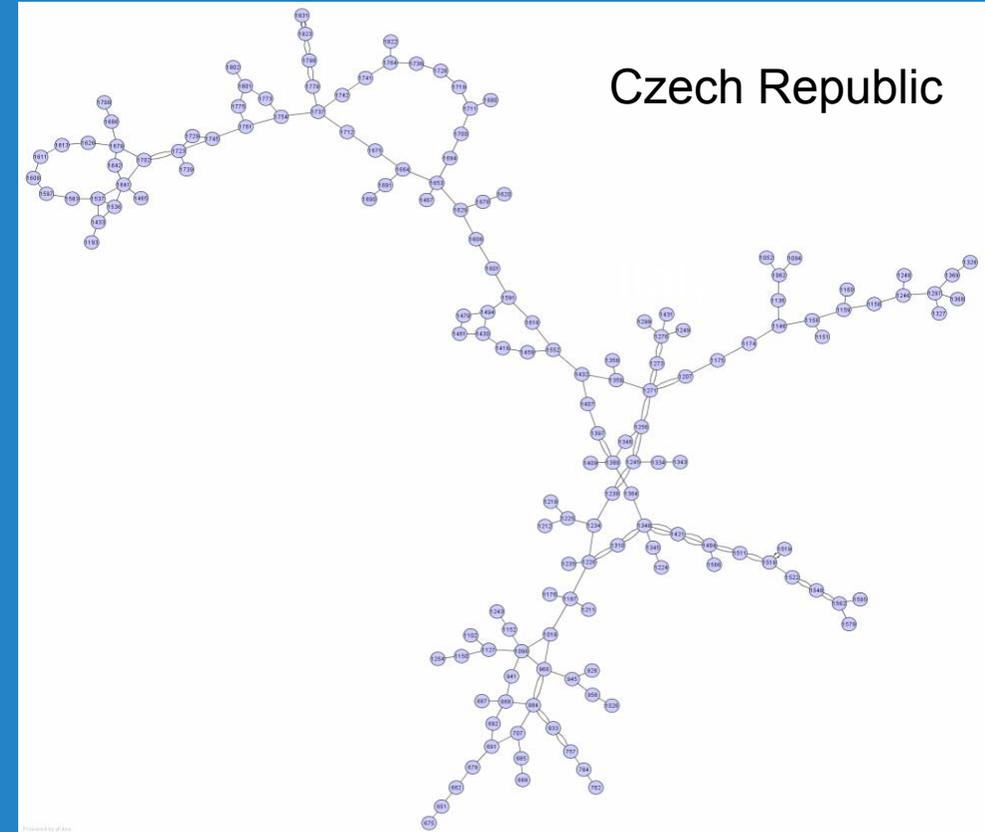
BETWEENNESS CENTRALITY

- other higher level fragmentation needs to be considered

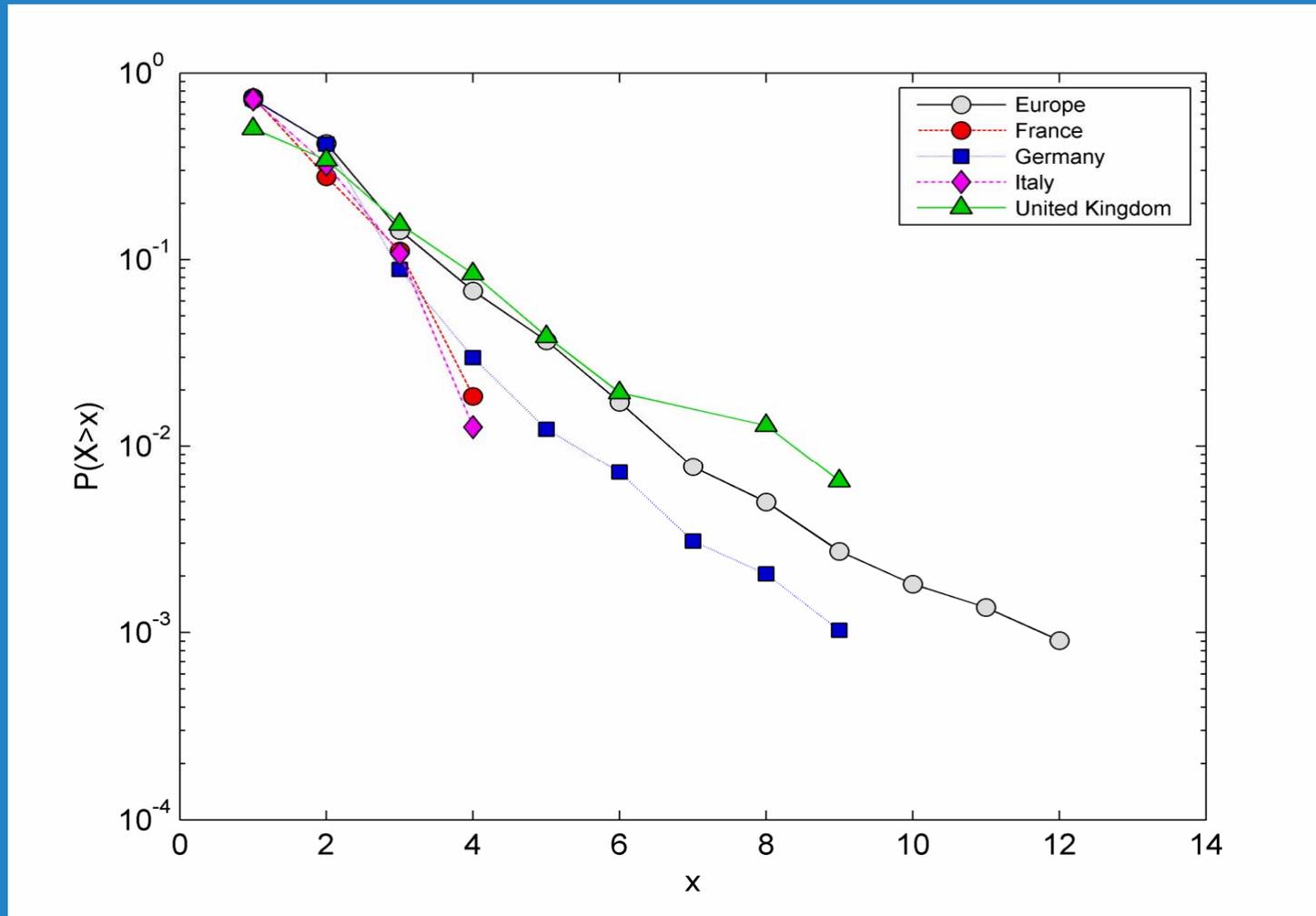
European Gas Network (JRC, Platts)



Gas Pipeline Network Layouts



Gas Pipeline Network: Cumulative Distribution of Node Degree



x : node degree
 $P(X > x)$: probability of node degree in excess of x
Exponential decay

Network Motifs: Motivation

R Carvalho, QMUL

- Basic idea: to **consider the recurring subgraphs** of interactions from which networks are built.
- **Motif** (intuitive definition): Consider a “real world” network G .
 H is a subgraph of a given graph G iff H is a graph whose vertices and edges form subsets of the vertices and edges of G .
- **A subgraph H of G is a *motif*** if the number of appearances of H in the real network exceeds the average number of appearances of H in a random network ensemble.
- **Claim: “real world” networks are organized in superfamilies according to their motifs.**

Network Motifs: Formal Definition

Milo et al. Superfamilies of Evolved and Designed Networks, *Science* 303, 1538

- First, compute the abundance of each subgraph i relative to its appearance in random networks:

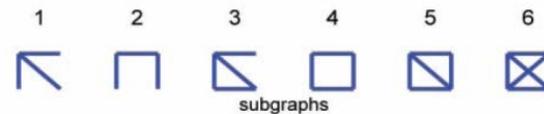
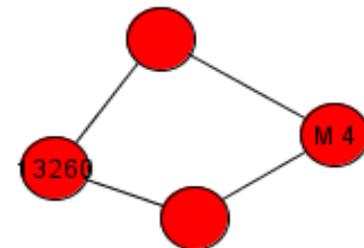
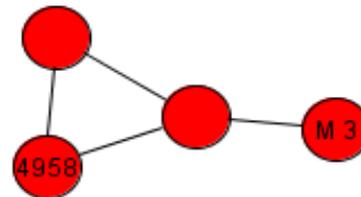
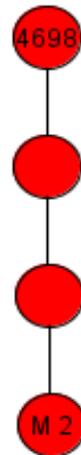
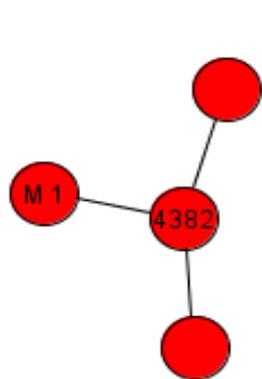
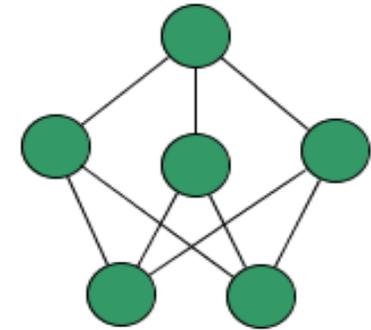
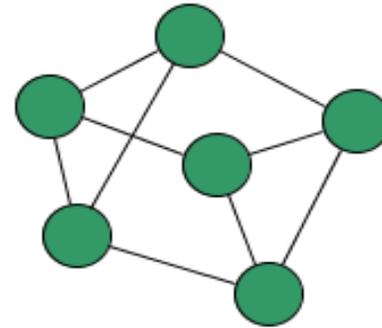
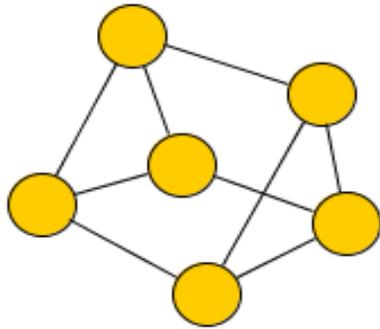
$$\Delta_i = \frac{N_{real} - \langle N_{rand} \rangle}{N_{real} + \langle N_{rand} \rangle + \varepsilon}$$

where a positive ε insures that $|\Delta|$ is not misleadingly large when the subgraph appears very few times in both the real and random networks;

- Normalizing gives *subgraph ratio profile* is defined as

$$SRP = \frac{\Delta_i}{\sum (\Delta_i^2)^{1/2}}$$

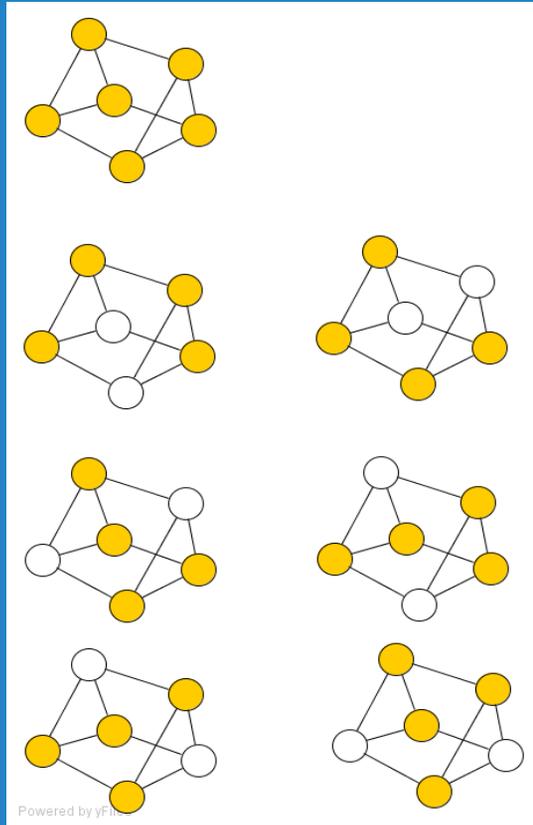
Network Motifs: Example



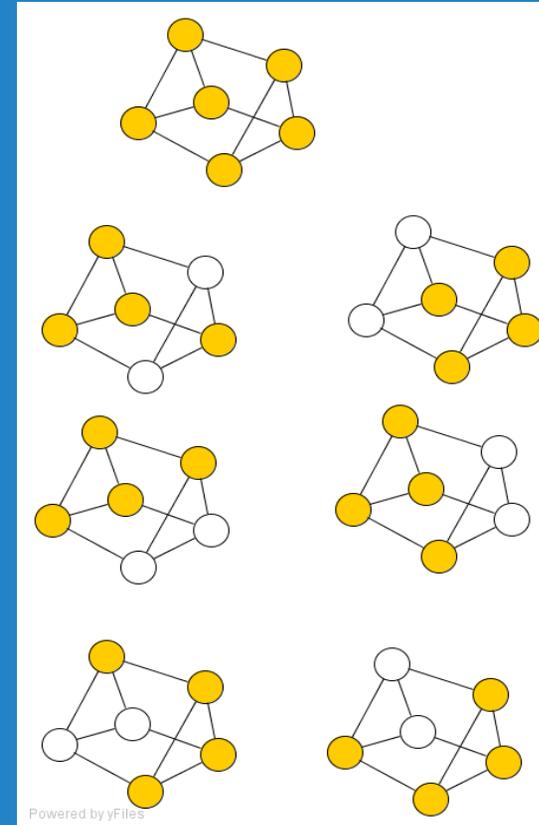
Powered by yFiles

Network Motifs: Example

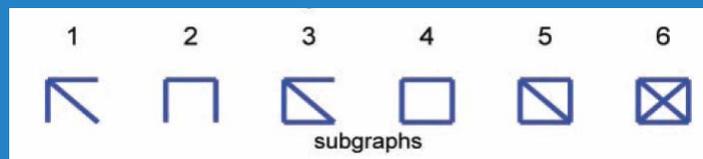
Presence of sub-graphs 2 and 3



Subgraph #2



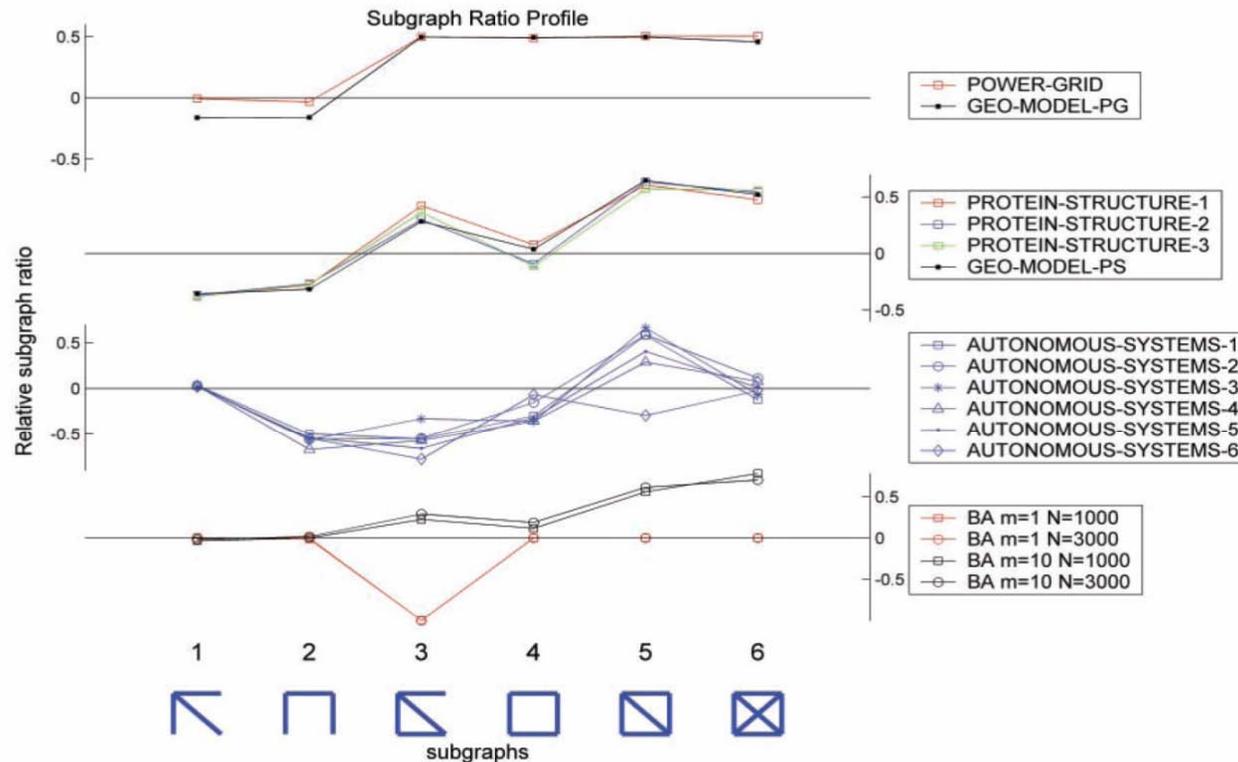
Subgraph #3



Network Motifs: Superfamilies of Networks

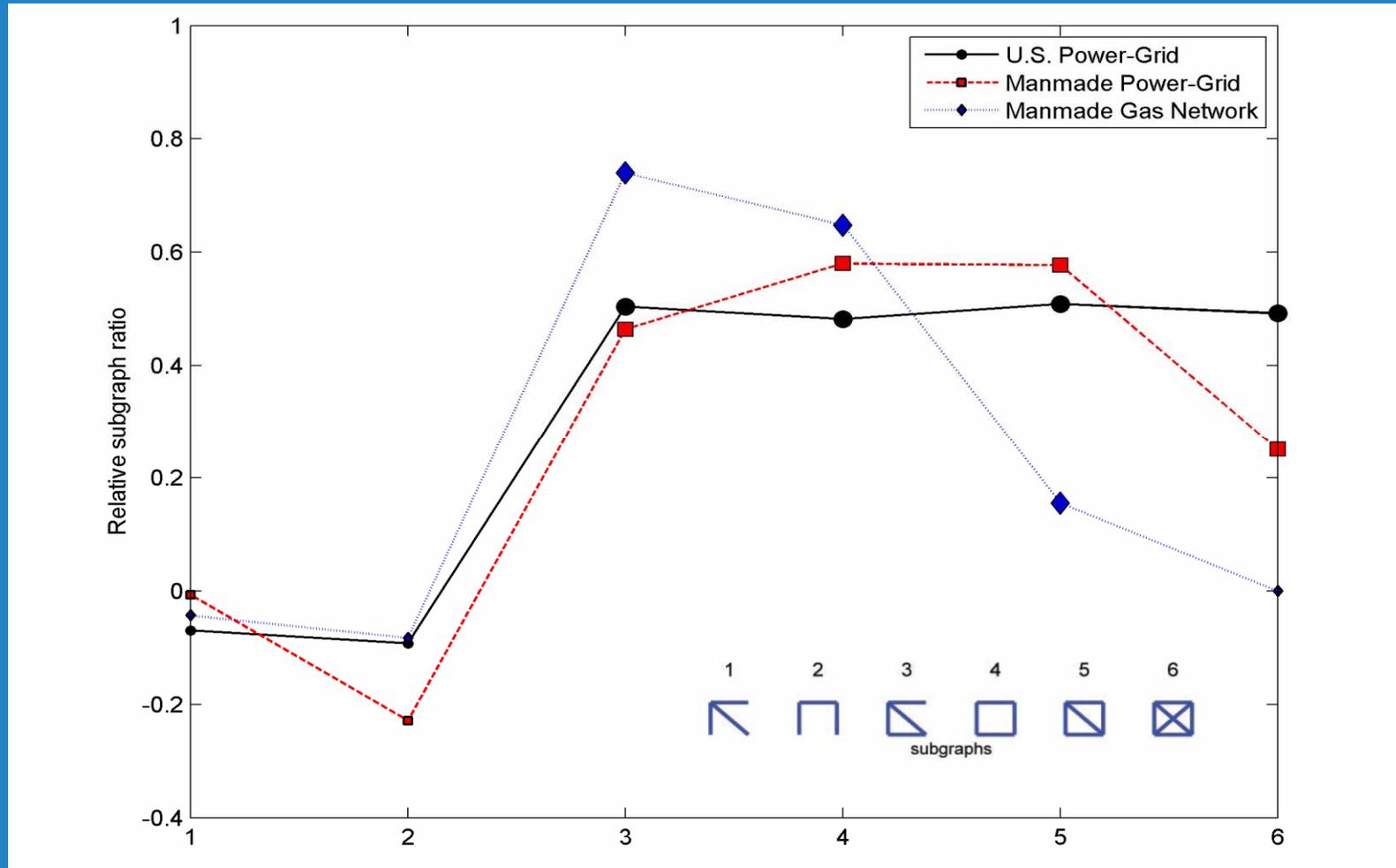
Milo et al., Superfamilies of Evolved and Designed Networks, *Science* 204, 1538

Fig. 3. The subgraph ratio profile (SRP) for various nondirected networks. The networks are as follows (12): (i) The electrical power grid of the western United States (4) (POWERGRID $N = 4941$, $E = 6594$) and a geometric model with similar clustering coefficient (GEO-MODEL-PG $N = 5000$, $E = 7499$). (ii) Networks of secondary-structure elements adjacency for several large proteins [structure based on the PDB database (www.rcsb.org/pdb/); the proteins (and their PDB ID) were 1A4J, an immunoglobulin (PROTEIN-STRUCTURE-1 $N = 95$, $E = 213$); 1EAW, a serine protease inhibitor (PROTEIN-STRUCTURE-2 $N = 53$, $E = 123$); and 1AOR, an oxidoreductase (PROTEIN-STRUCTURE-3 $N = 99$, $E = 212$)] and a geometric model with similar clustering coefficient (GEO-MODEL-PS $N = 53$, $E = 136$). (iii) The Internet at the autonomous system level (www.cosin.org) (AUTONOMOUS-SYSTEMS 1 to 6; $N = 3015$, 3522, 4517, 5357, 7956, 10515; $E = 5156$, 6324, 8376, 10328,



15943, 21455). (iv) Networks grown according to the preferential attachment BA model (3) with $m = 1$ or $m = 10$ edges per new node (BA $m = 1, 10$; $N = 1000, 3000, 1000, 3000$; $E = 1000, 3000, 9901, 29901$).

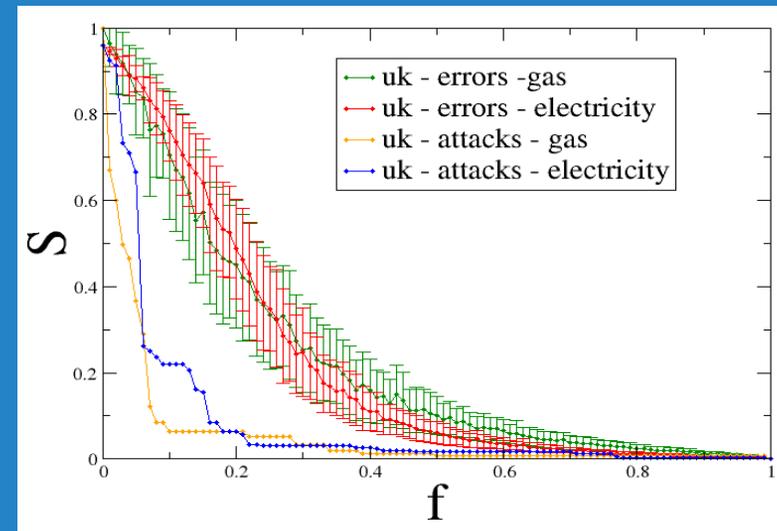
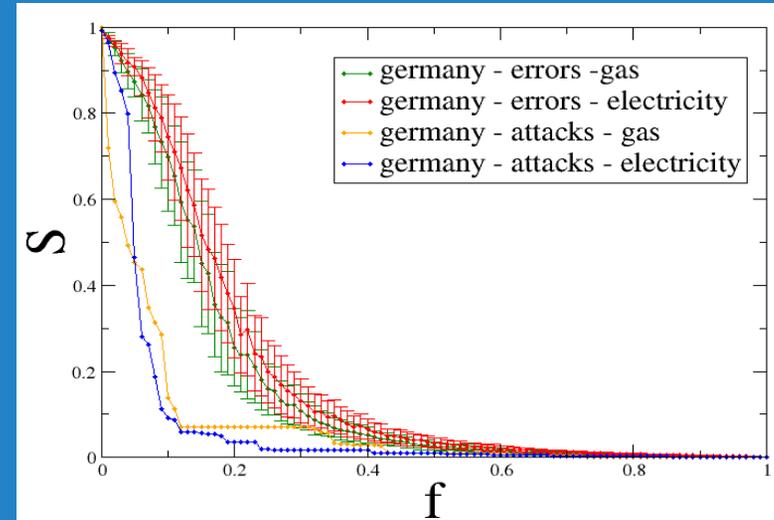
Motifs in manmade networks



Network Tolerance Against Node Removals

We study the size of the largest connected component (S) as a function of the fraction of nodes removed (f) by:

- Errors: random node removal;
- Attacks: higher degree nodes are removed first;



Wind field construction

and maps of potential wind energy production over Europe

P. Kiss, I Janosi M. Szenes, Farkas with JRC(Ispra) support

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Data and methods

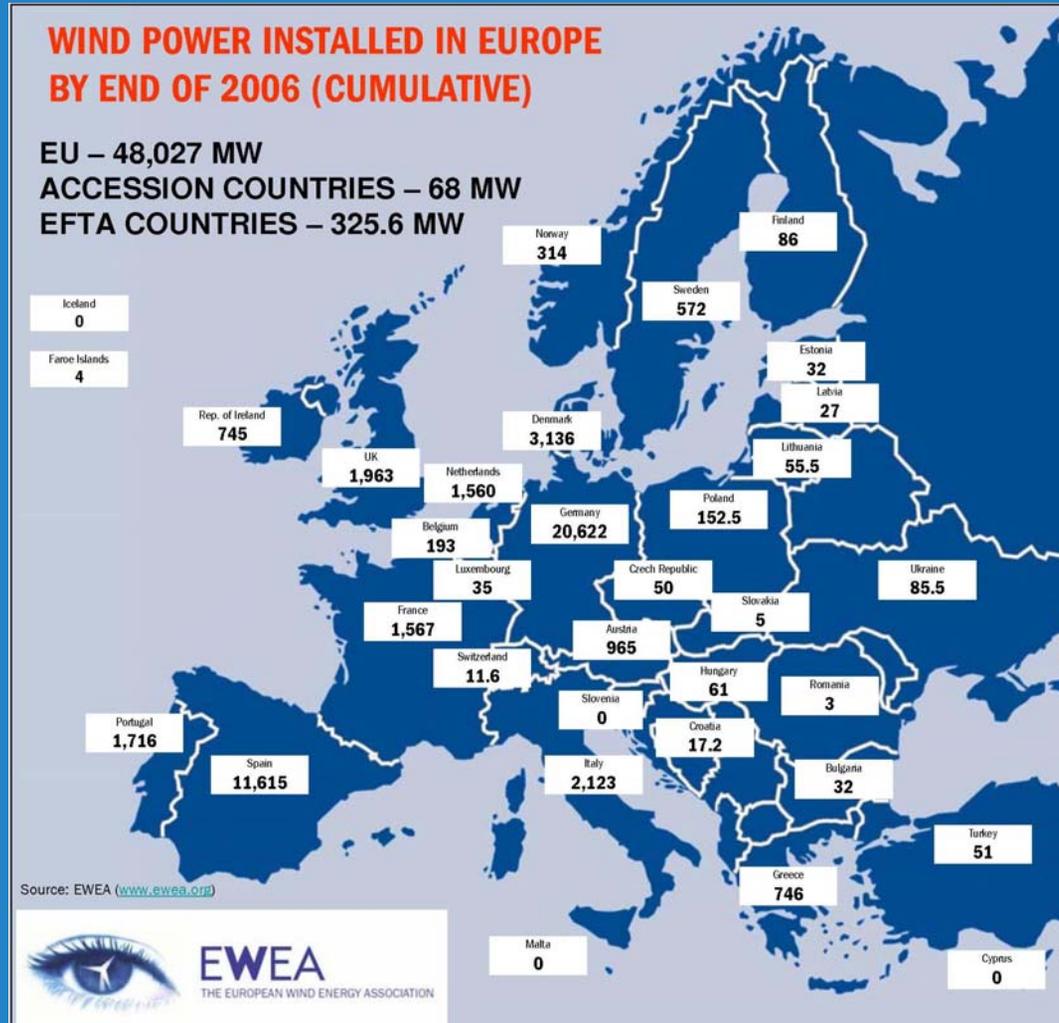
Models for wind speed histograms

Wind profiles – height dependence of wind speed

Wind power estimations

Wind power networks

Wind field construction and maps of potential wind energy production over Europe



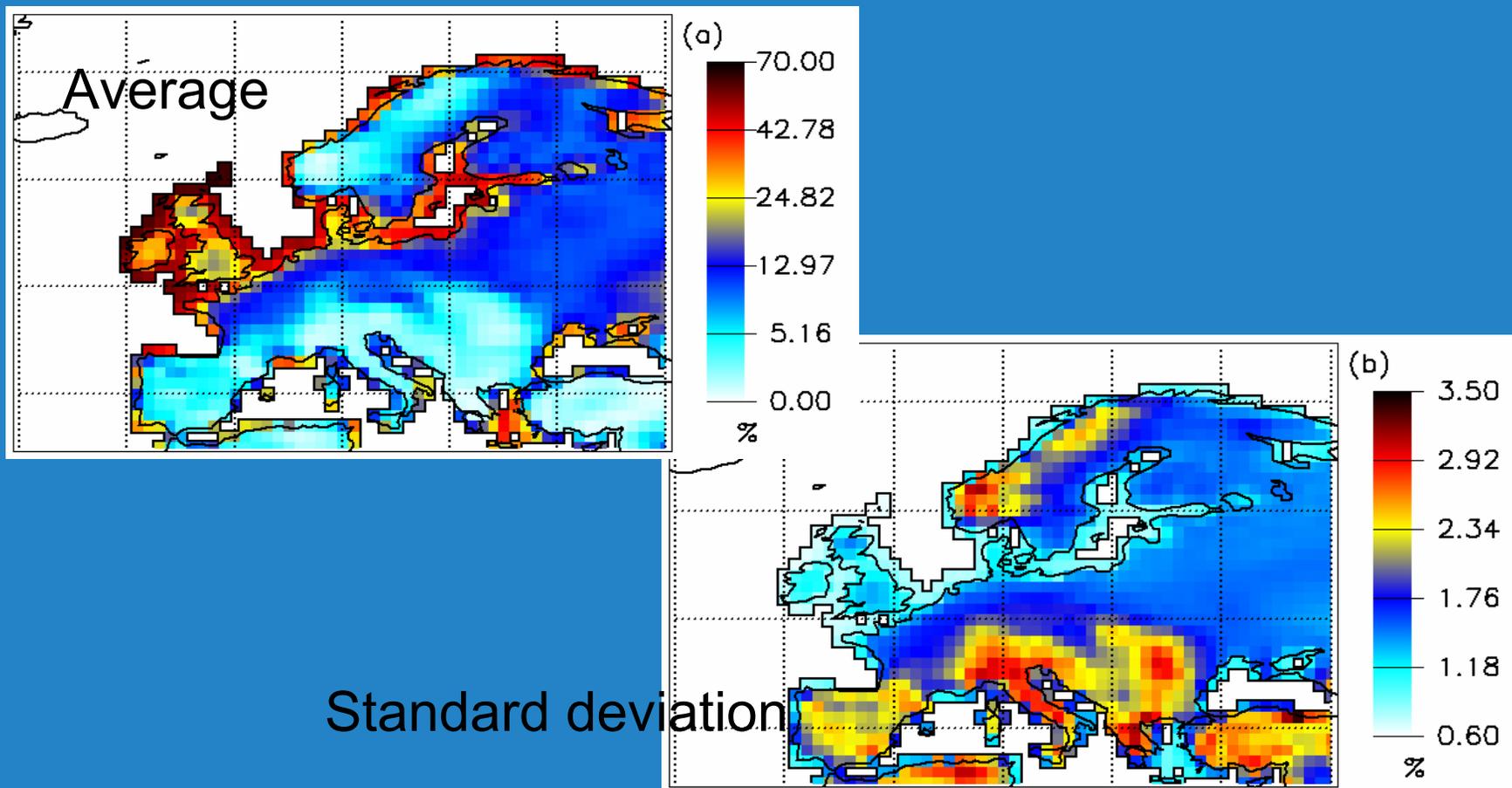
wind energy
as a %
of total energy
in EU
7-8%

Wind field construction

and maps of potential wind energy production over Europe

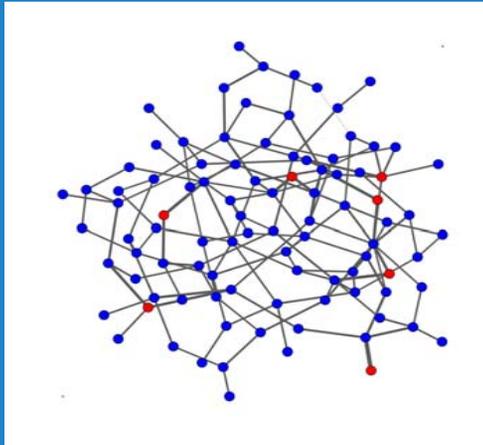
Wind power networks

Average and standard deviation of wind power

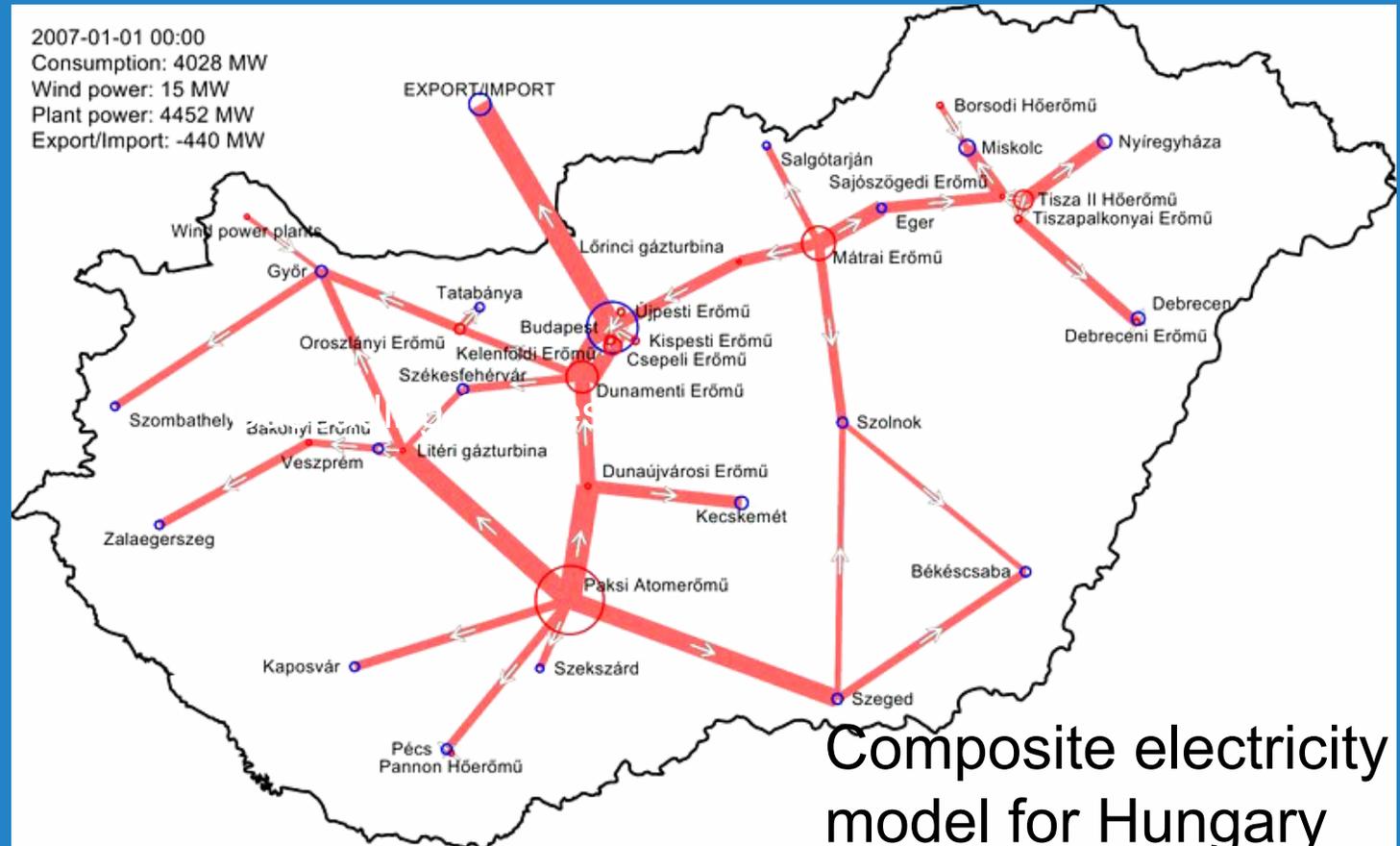


Standard deviation

Error tolerance of complex networks



Cascading breakdown
flow of a physical
quantity:
maximal load is limited
by the capacity of the
edge
edge removal leads to
redistribution of the
initial loads -overloading



Composite electricity
model for Hungary
including
wind energy

Error tolerance of complex networks

Dynamic network capacity modelling

- extend the model to the whole European power grid network using available European grid topology and REWIRING on the basis of known offshore windspeed data
 - emergence of a new network topology incorporating windpower
- identifying the most vulnerable parts of the network –
- suggesting new edges (transmission lines) which make the network more tolerant

Urban Networks

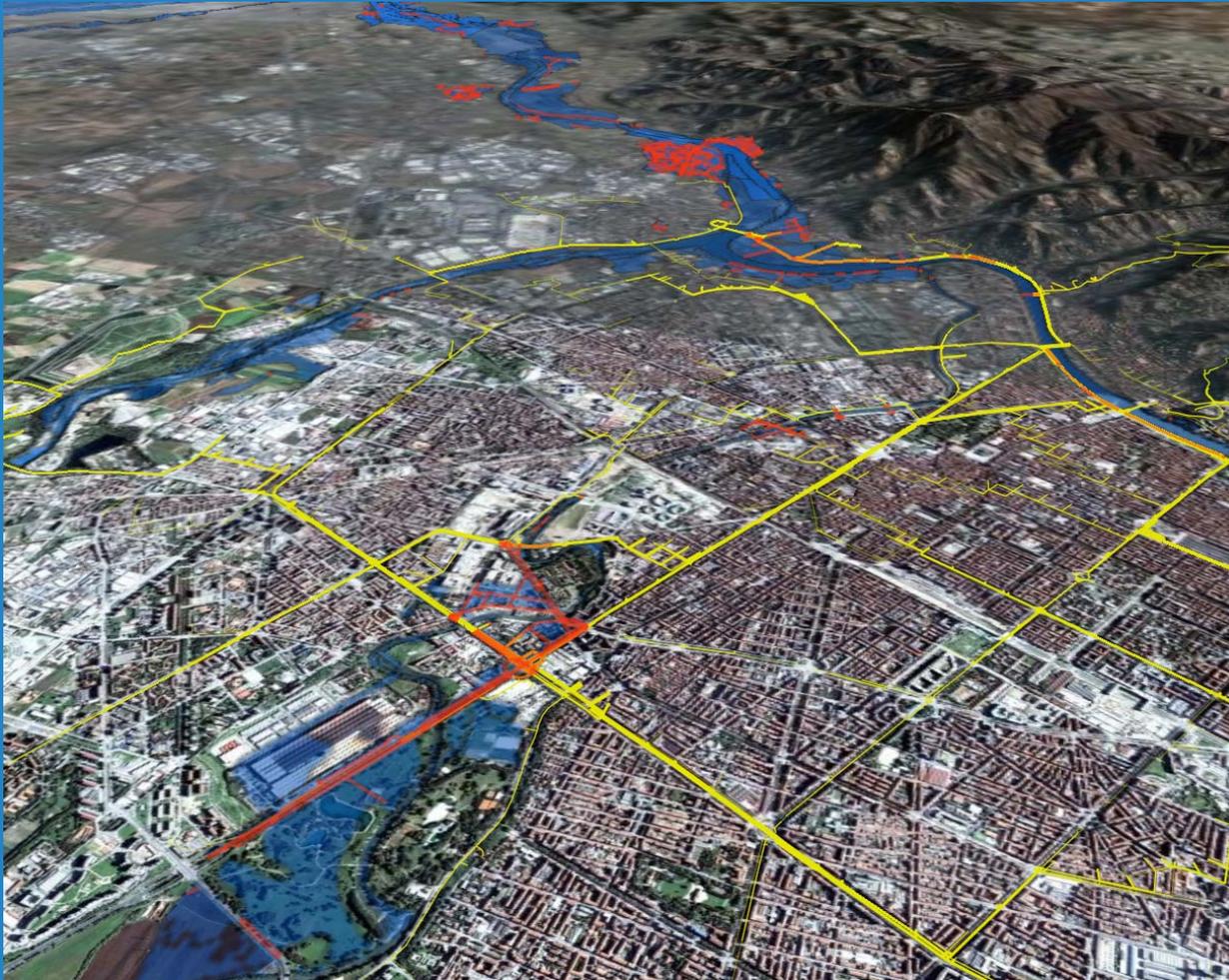


- Milan
- Turin
- London (in progress)

Data Sources:

- TeleAtlas
- UK DfT
Department for Transport
- Civil protection surveys

Identification of vulnerable transport infrastructures



Topological measures of damaged and undamaged network.

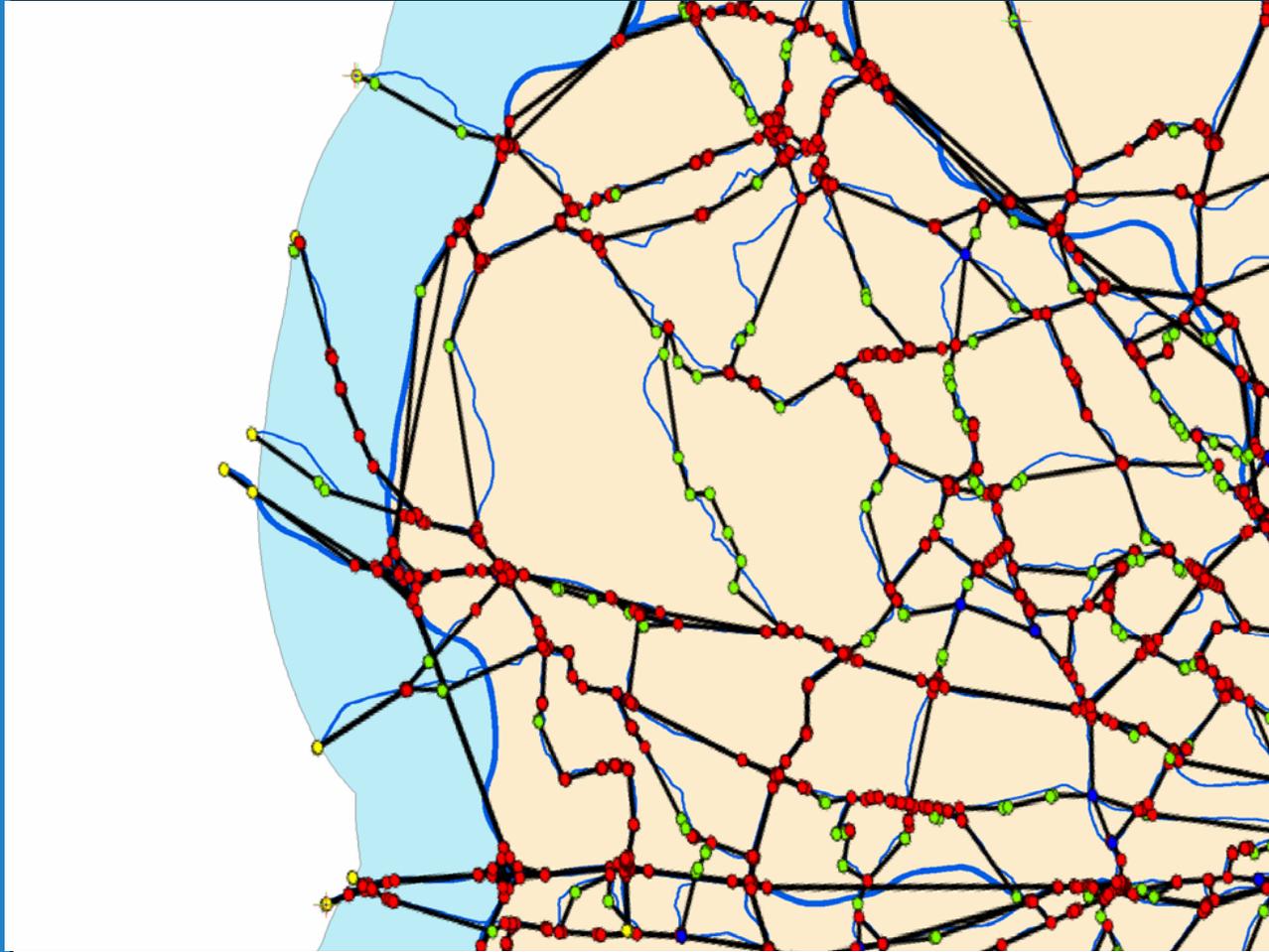
Structural vulnerability, key transport nodes, planning and **protection**.

Centrality is a topological measure of connectivity rank.

Flooded area
(Turin - October 2000).

Intersection of flood and high centrality (orange).

Urban Traffic

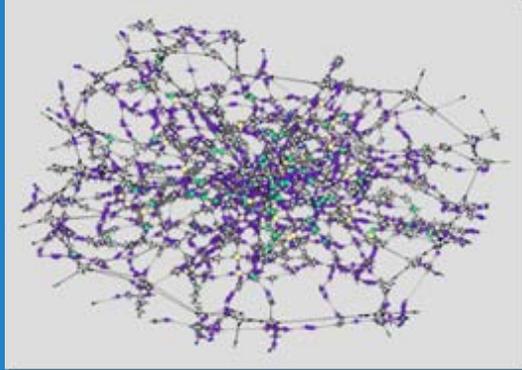


AADTF – annual average daily flow

Network simplification

Connectivity analysis

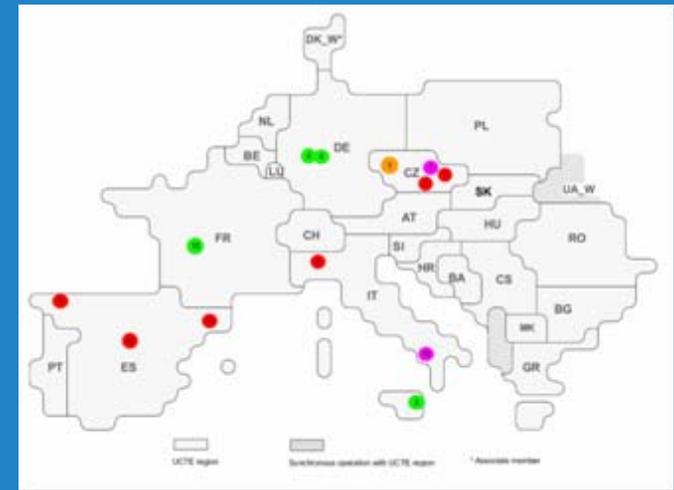
Future datasets



Urban street network of London and traffic counts

2232 counters, counts from 1999 to 2006, 8566 nodes, 15573 arcs

Electricity Network disruptions overlay on ranking networks for correlation



Commodity Flows

27 countries, 6 Major groups,
225 products, years 2005-2007



Dynamic of London Traffic (© Manmade)

