

Introduction to data networks analysis

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Introduction

- Graphs & networks

- Network data model

Real-world networks

- Types

- Examples

- Properties

- Complexity

Network analysis

- PageRank

- HITS

- Infomap

Applications

- Fraud detection

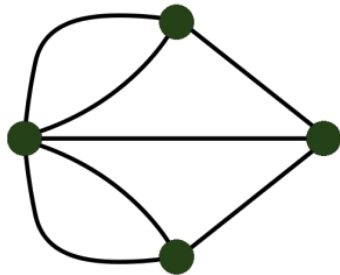
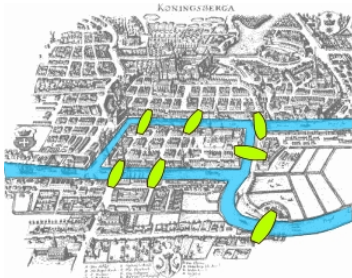
- Telecommunications

- Biology

- WWW

Motivation

Seven bridges of Königsberg (Euler, 1735).

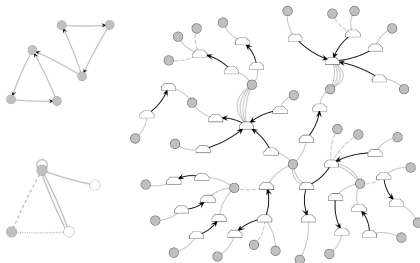


Graphs

Graph G is a discrete mathematical object that consists of:

- ▶ a set of nodes $V = \{v_1, v_2 \dots v_n\}$ and
- ▶ a set of edges among nodes $E \subseteq \{\{v_i, v_j\} \mid v_i, v_j \in V\}$.

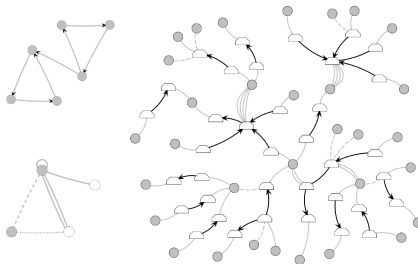
Furthermore, graph can be directed, labeled, weighted, multi-, hyper-, etc.



Networks

Networks are graphs with some additional information on the nodes and edges. Network N consists of:

- ▶ a set of nodes V and edges E ,
- ▶ node features $F_V : V \rightarrow \Sigma_V$ and
- ▶ edge features $F_E : E \rightarrow \Sigma_E$.



Network data model

Why networks?

- ▶ Natural representation of many domains.
- ▶ Strong mathematical foundations.
- ▶ Useful whenever we are interested in relations among entities.

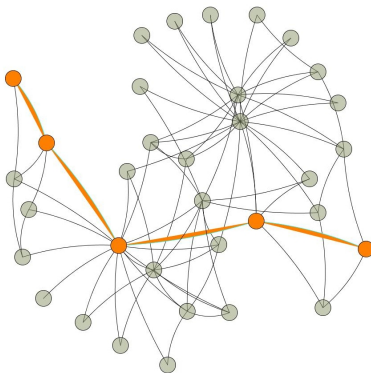
Network data model has to store:

- ▶ nodes, edges and features,
- ▶ their incidence.

Standard data models (e.g. relational, hierarchical, etc.) are **not appropriate** for storing networks.

Querying

Toy example: find the shortest path from node v_i to v_j in a simple graph G .



Querying, cont.

Toy example: find the shortest path from node v_i to v_j in a simple graph G .

Relational data model (e.g. *Oracle 11g*)

- ▶ Schema *Edge*(#node1, #node2).
- ▶ Use Dijkstra's algorithm with v_i being the source node.
- ▶ Algorithm does $\mathcal{O}(|E|)$ queries over the database.

Network data model (e.g. *Oracle 11g Spatial*)

- ▶ Schemas *Node*, *Link*, *Path*, etc.
- ▶ Use `manager.shortestPath( $G, v_i, v_j$ )` (Java API).
- ▶ Algorithm does a single (?) query over the database!

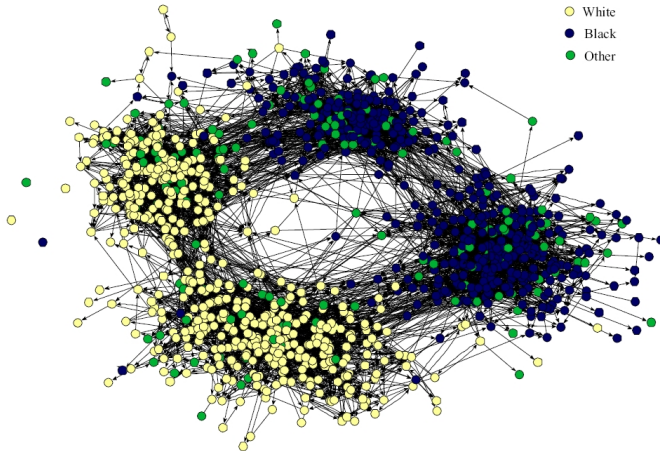
Types of networks

Real-world networks are of different types:

- ▶ social (e.g. network of Twitter, friendship network),
- ▶ information (e.g. WWW, citation networks),
- ▶ biological (e.g. gene regulatory networks, food webs),
- ▶ technological (e.g. Internet, power grid),
- ▶ etc.

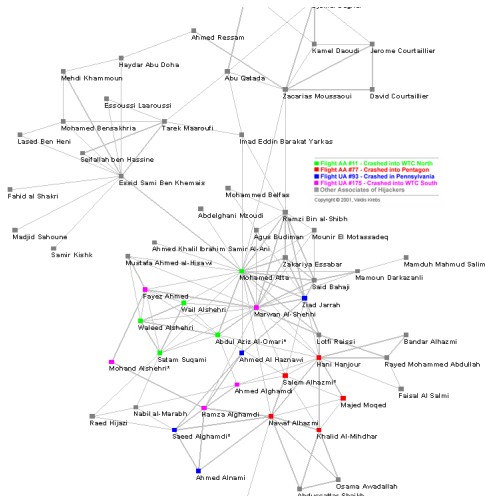
Examples - social networks

Friendship network of children in U.S. school.



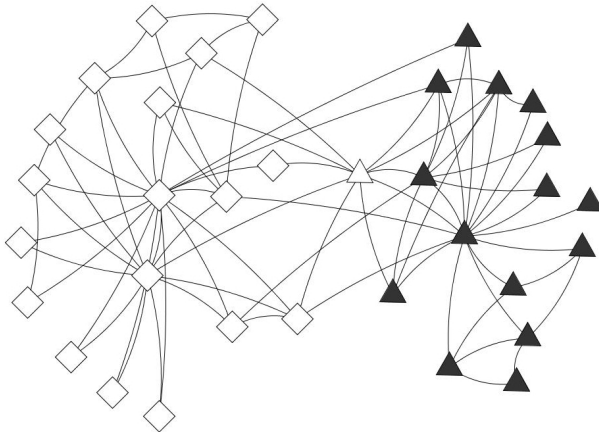
Examples - social networks

Krebs's terrorist network of 9-11.



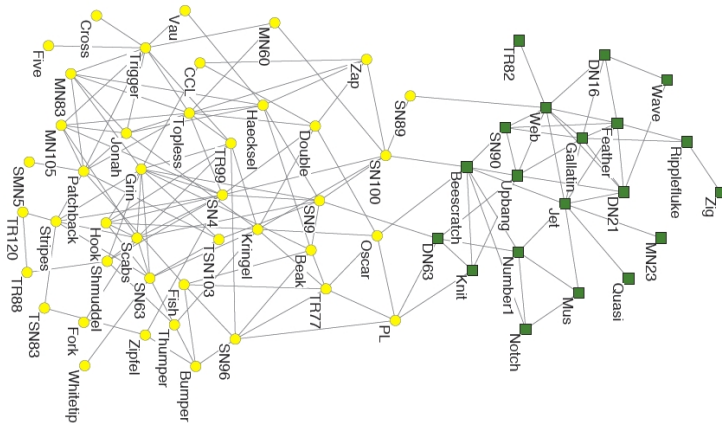
Examples - social networks

Zachary's karate club network.

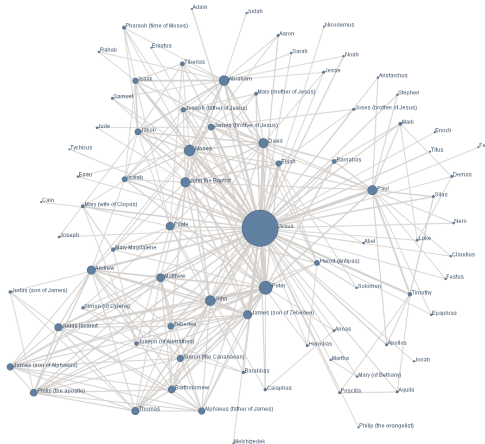


Examples - social networks

Lusseau's bottlenose dolphins network.

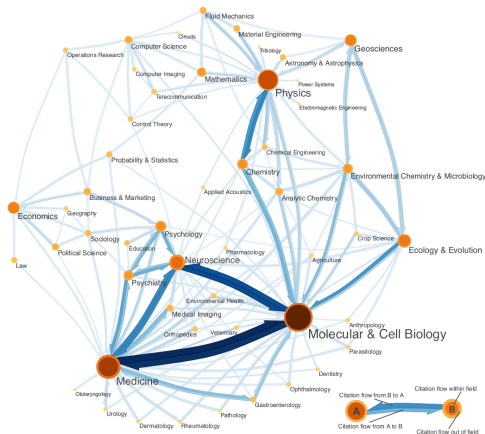


Social network of Jesus.



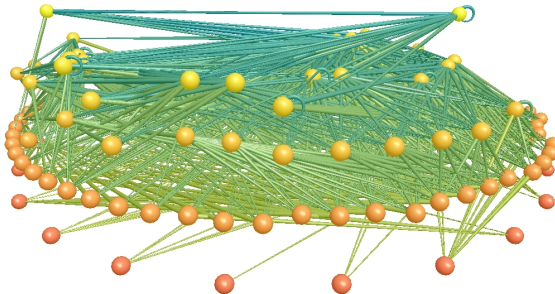
Examples - information networks

Map of science.



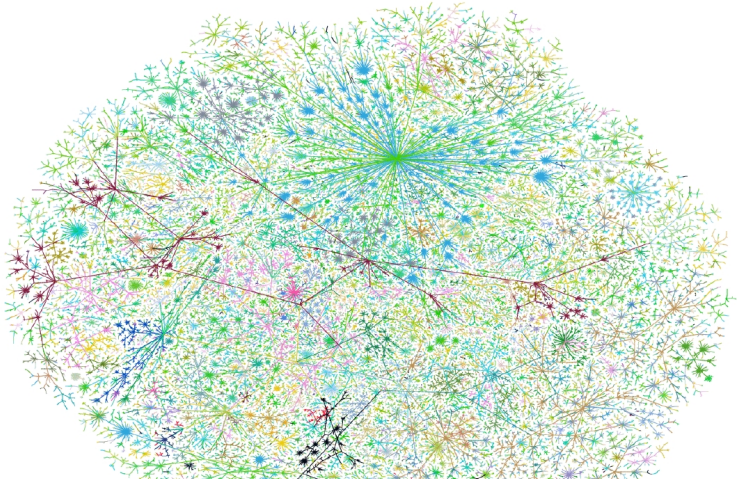
Examples - biological networks

Food web of predator-prey interactions.



Examples - technological networks

Internet at the level of "autonomous systems".



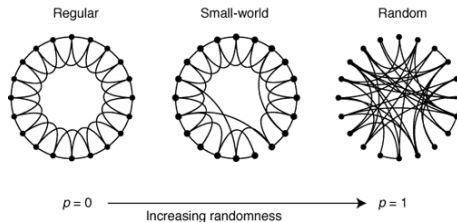
Properties of networks

Properties of many real-world networks:

- ▶ *small-world effect*,
- ▶ *power law*,
- ▶ transitivity,
- ▶ community structure,
- ▶ network resilience,
- ▶ network navigation,
- ▶ etc.

Small-world effect

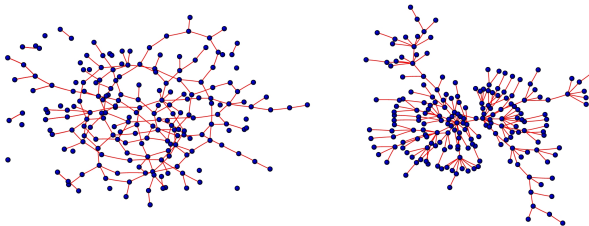
- ▶ Famous Milgram's passing letters experiment.
- ▶ *Small-world effect (6 degrees of separation)*: the average distance among nodes is small, even when the network is very large.



- ▶ The effect is actually mathematically obvious.
- ▶ Applications for dynamics of processes on networks.

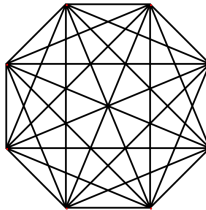
Power law

- *Power law (scale-free networks)*: the degree distribution of nodes is power law in the tail, i.e., there exist nodes with extremely high degree.
- Possible explanation is *preferential attachment* (i.e. *rich-get-richer*).



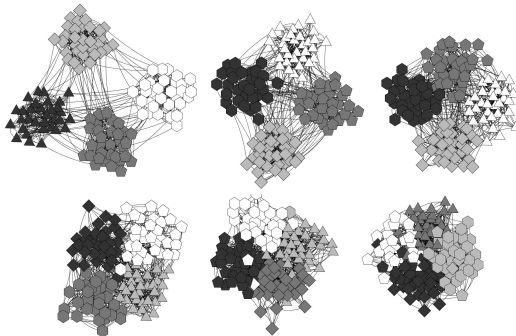
Transitivity

- ▶ Transitivity (clustering): "friend of a friend is often also a friend".
- ▶ Transitivity means presence of a high number of triangles.



Community structure

- Community structure: existence of groups of nodes (i.e. communities) with many edges within communities and only a few edges between communities.



Complexity

Networks can be classified as follows:

small tens of nodes,

medium several hundreds of nodes,

large several thousands or millions of nodes,

huge several millions of nodes and more.

- ▶ Even medium networks are commonly hard to visualize.
- ▶ Large networks are those that can still be stored in computer's memory, whereas huge networks cannot – a problem as nodes cannot be analyzed independently!

Many of problems over networks are NP-hard.

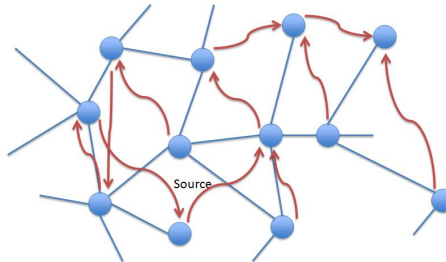
Network analysis

Network analysis (from AI perspective) is a broad field:

- ▶ (social) network analysis (SNA),
- ▶ link analysis (LA),
- ▶ graph-based data mining (GBDM),
- ▶ (statistical) relational learning (SRL),
- ▶ community detection,
- ▶ etc.

PageRank (Brin and Page, 1998)

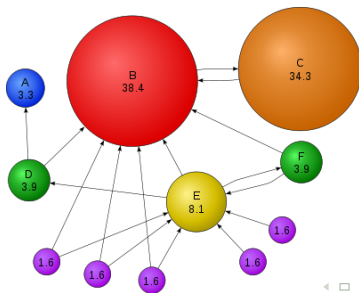
- ▶ A link analysis algorithm that exploits hyperlinks among web pages.
- ▶ Result is a ranking of web pages (due to their importance).
- ▶ Main hypothesis: random walker on a (WWW) network would spend most of the time on "important" web pages, and less on "unimportant" ones.



PageRank, cont.

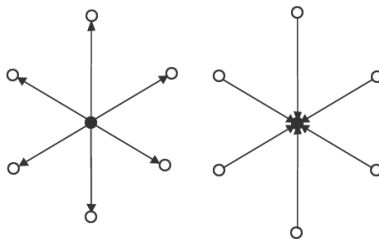
- ▶ Result of a random walk corresponds to the first eigenvector of the adjacency matrix of the network.
- ▶ *PageRank* algorithm (d is a damping factor, e.g., 0.85):

$$PR(v_i) = \frac{1-d}{|V|} + d \sum_{v_j \in \mathcal{N}(v_i)} \frac{PR(v_j)}{\deg^+(v_j)}$$



HITS (Kleinberg, 1999)

- ▶ A link analysis algorithm that exploits hyperlinks among web pages (*Hyperlink-Induced Topic Search*).
- ▶ Result is a ranking of web pages (due to their *authority* and *hub* importance).
- ▶ Main hypothesis: page is a good hub if it links to many good authorities (and vice versa).



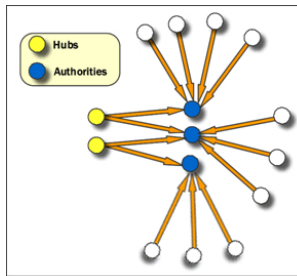
Hubs

Authorities

► *HITS* algorithm:

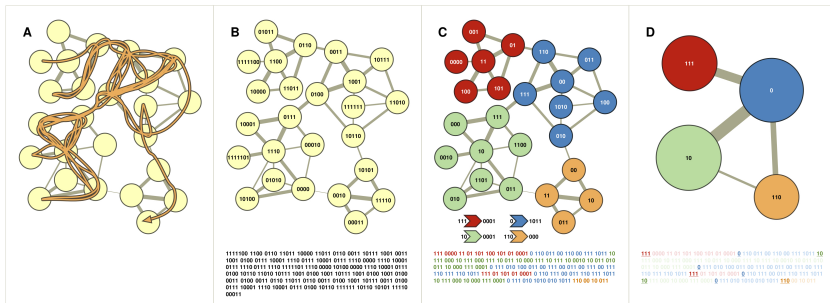
$$a(v_i) = \sum_{(v_j, v_i) \in E} h(v_j)$$

$$h(v_i) = \sum_{(v_i, v_j) \in E} a(v_j)$$



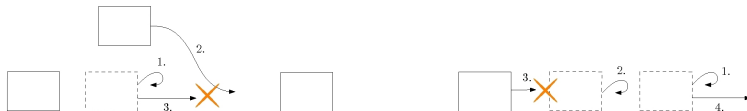
Infomap (Rosvall and Bergstrom, 2008)

- ▶ A community detection algorithm that again uses random walks on the network and Huffman coding.
- ▶ Result is a clustering of the nodes of the graph.



Automobile insurance fraud

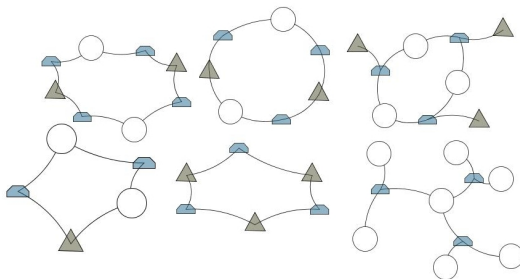
- Focus on groups of collaborating fraudsters staging traffic "accidents". Groups consist of drivers, chiropractics, police officers, lawyers, insurance workers, etc.
- Common characteristics of such accidents (e.g. at night, no children, no narcotics, etc.).
- Standard schemes for staging traffic accidents like *Drive Down* and *Swoop and Squat*.



- Such accidents are very expensive and dangerous.

Suspicious patterns detection

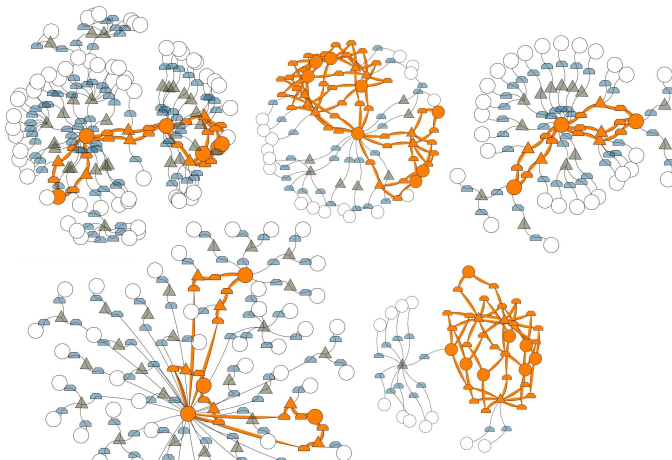
- There are many suspicious patterns:



- Patterns can be detected using simple subgraph isomorphism.

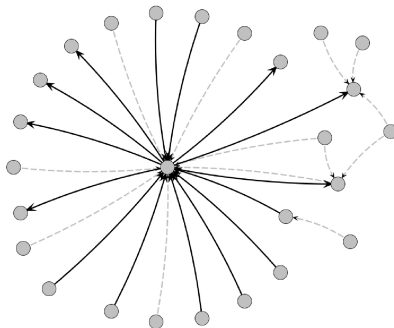
Suspicious patterns detection, cont.

Results of suspicious patterns detection.



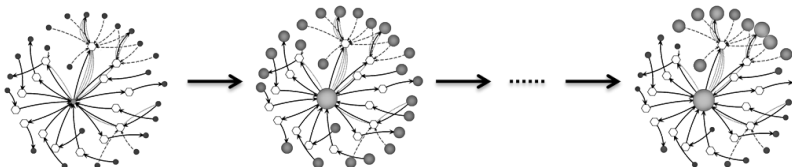
Suspicious components detection

- Detection of suspicious components using their common (topological) characteristics (e.g. cycles, diameter, central nodes, etc.).
- *PRIDIT* analysis of indicators of suspicious components.



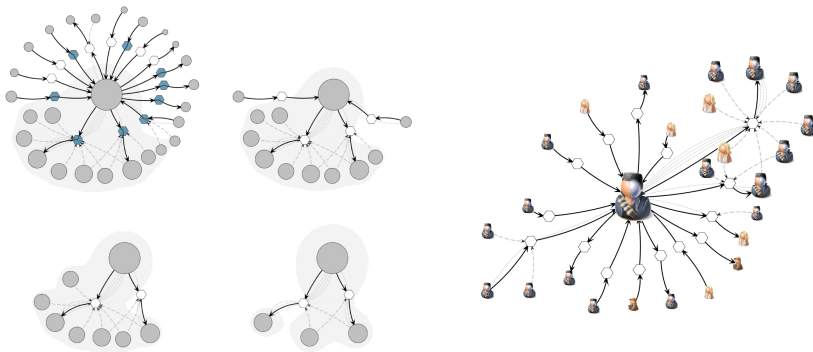
Suspicious entities detection

- ▶ Main hypothesis: suspicion of every entity is well defined with the suspicions of its related entities (participant-accident).
- ▶ Suspicion of every entity is inferred from the network and propagated onward using *PageRank*-like algorithm (i.e. *IAA* algorithm).
- ▶ Iterative assessment reduces locality of the approach.



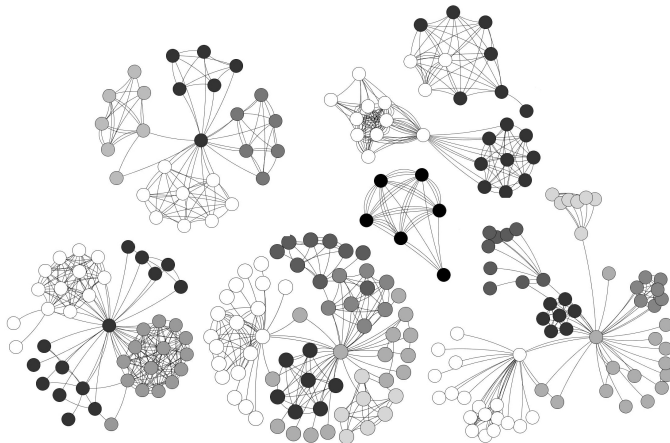
Suspicious entities detection, cont.

Results of suspicious entities detection.



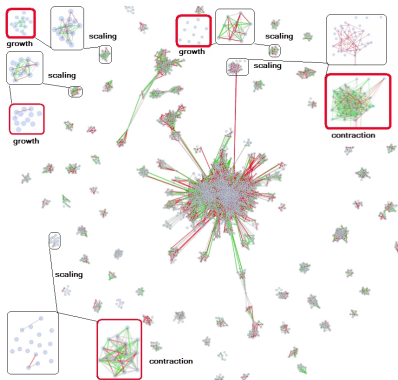
Suspicious communities detection

Results of “suspicious” communities detection.



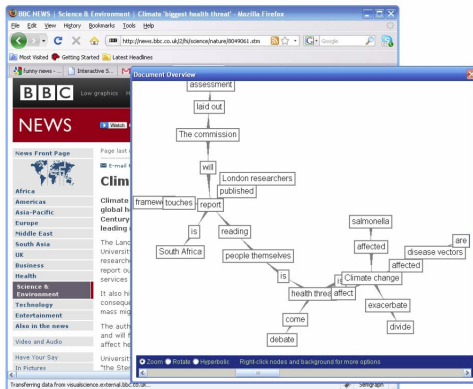
TeleComVis project (Beijing University)

- ▶ Visual analysis of telecom communities in massive call graphs.
- ▶ Main objective is to adjust market strategies to different social communities.



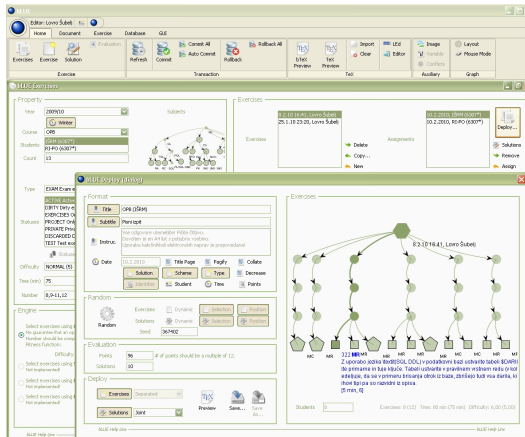
SemGraph project (Jožef Stefan Institute)

- ▶ Enhanced web page visualization using semantic graphs and text mining.
- ▶ Works as a Firefox plug-in.



bBLUE project

- ▶ Automatic generation of exams from the database of exercises.
- ▶ Exams and exercises visualized with networks.



Questions?