

Mark Lombardi

Network Analysis Sources of Networks

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Outline

1	Some names in the development of SNA	1
2	Some important events	2
3	Selected Books on SNA	3
4	Courses on NA	4
5	Software for SNA	5
6	Roman roads (Peutinger)	6
7	Moreno: Who shall survive?	7
9	Development of DNA (Garfield)	9
10	Organic molecule 3CRO	10
11	Hijackers (Krebs)	11
12	<i>Wall Street Follies</i>	12
13	<i>They Rule</i>	13
14	Lombardi's networks	14
18	Networks	18
19	Graph	19
21	How to get a network?	21
22	Complete and ego-centered networks	22

23	Use of existing network data	23
25	Genealogies	25
28	Molecular networks	28
29	GraphML	29
32	Approaches to computer-assisted text analysis	32
42	Neighbors	42
45	Transformations	45
46	Networks from the Internet	46
51	Random networks	51

Sociograph

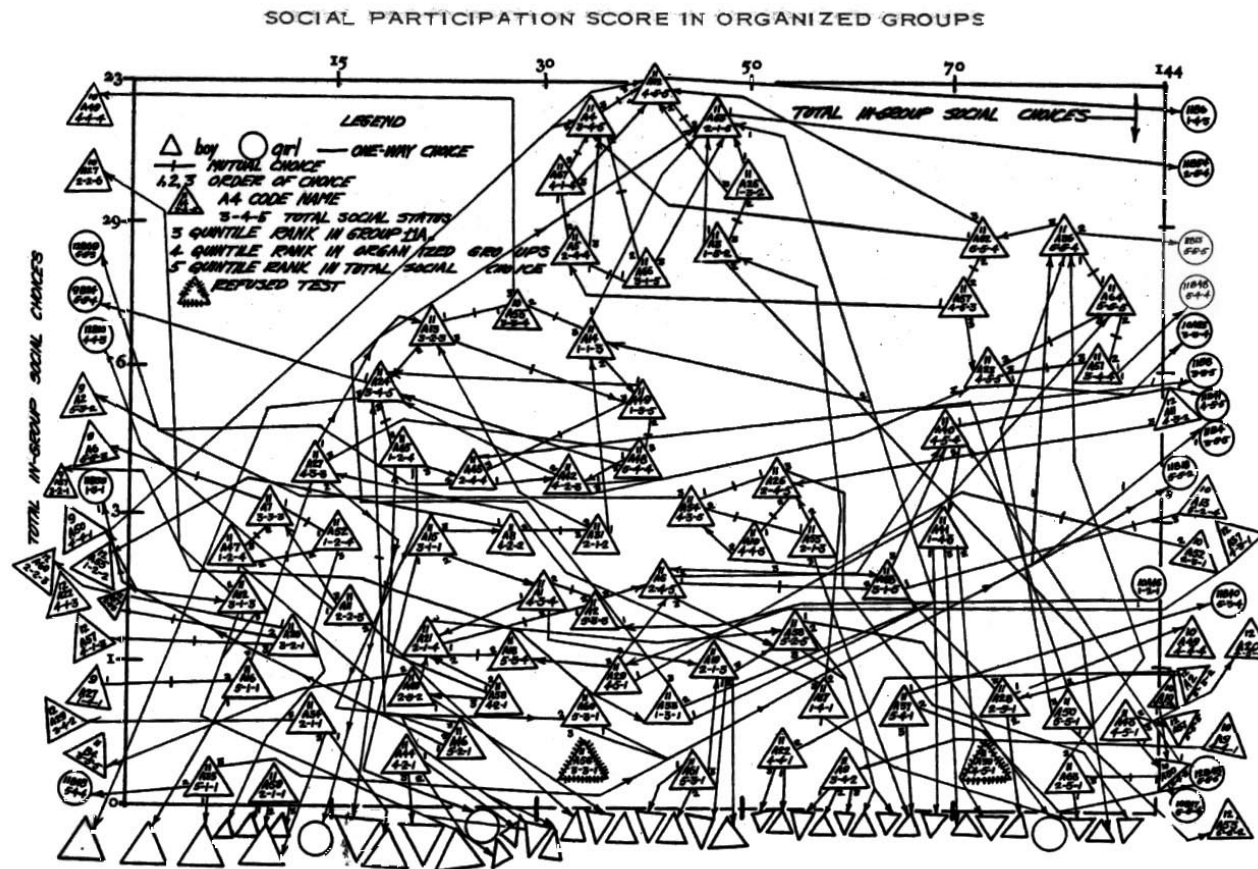


FIGURE 3. GROUP 11A. A SOCIAL STATUS SOCIOGRAPH OF THE 68 BOYS OF THE 11TH GRADE. IT SHOWS QUINTILE POSITIONS FOR: (1) GRADE AVERAGE, (2) SOCIAL PARTICIPATION SCORES IN STUDENT ORGANIZATIONS, AND (3) TOTAL FRIENDSHIP CHOICES ON A SOCIOMETRIC TEST (TOTAL CHOICES INCLUDE DIRECT AND INDIRECT CHOICES). THE STUDENT'S POSITION ON THE CHART IS THAT OF HIS CLIQUE GROUP AVERAGE FOR BOTH SOCIAL PARTICIPATION SCORES AND TOTAL FRIENDSHIP CHOICES. OUT-OF-CLASS CHOICES APPEAR IN THE SIDE MARGIN; OUT-OF-SCHOOL CHOICES IN THE LOWER MARGIN.

11

Coincident Strong Citation Connections

Coincident Weak Citation Connections

Non-Coincident Strong Citation Connections

Non-Coincident Weak Citation Connections

Asimov's Implied

Specified Historical Connections

1 Braconnot 1820

2 Mendel 1865

3 Miescher 1871

4 Flemming 1879

5 Kossel 1886

6 Fischer Piloty 1891

7 De Vries 1900

8 Fischer 1907

9 Levene Jacobs 1903

10 Muller 1926

11 Griffith 1928

12 Levene Mari London 1925

13 Alloway 1932

14 Stanley 1935

15 Levene Tipton 1935

16 Bawden Pirie 1936

17 Caspersen Schultz 1938-39

18 Beadle Tatum 1941

19 Martin Synge 1943-44

20 Avery MacLeod McCarty 1944

21 Chargaff 1947

22 Chargaff 1950

23 Pauling Corey 1950-51

24 Sanger 1951-53

25 Hershey Chase 1952

26 Wilkins 1953

27 Watson Crick 1953

28 Du Vigneaud 1953

29 Todd 1955

30 Palade 1954-56

31 Franckel Conrat 1955-57

32 Ochoa 1955-56

33 Kornberg 1956-57

34 Hoagland 1957-58

35 Jacob Monod 1960-61

36 Hurwitz 1960

37 Dintzis 1961

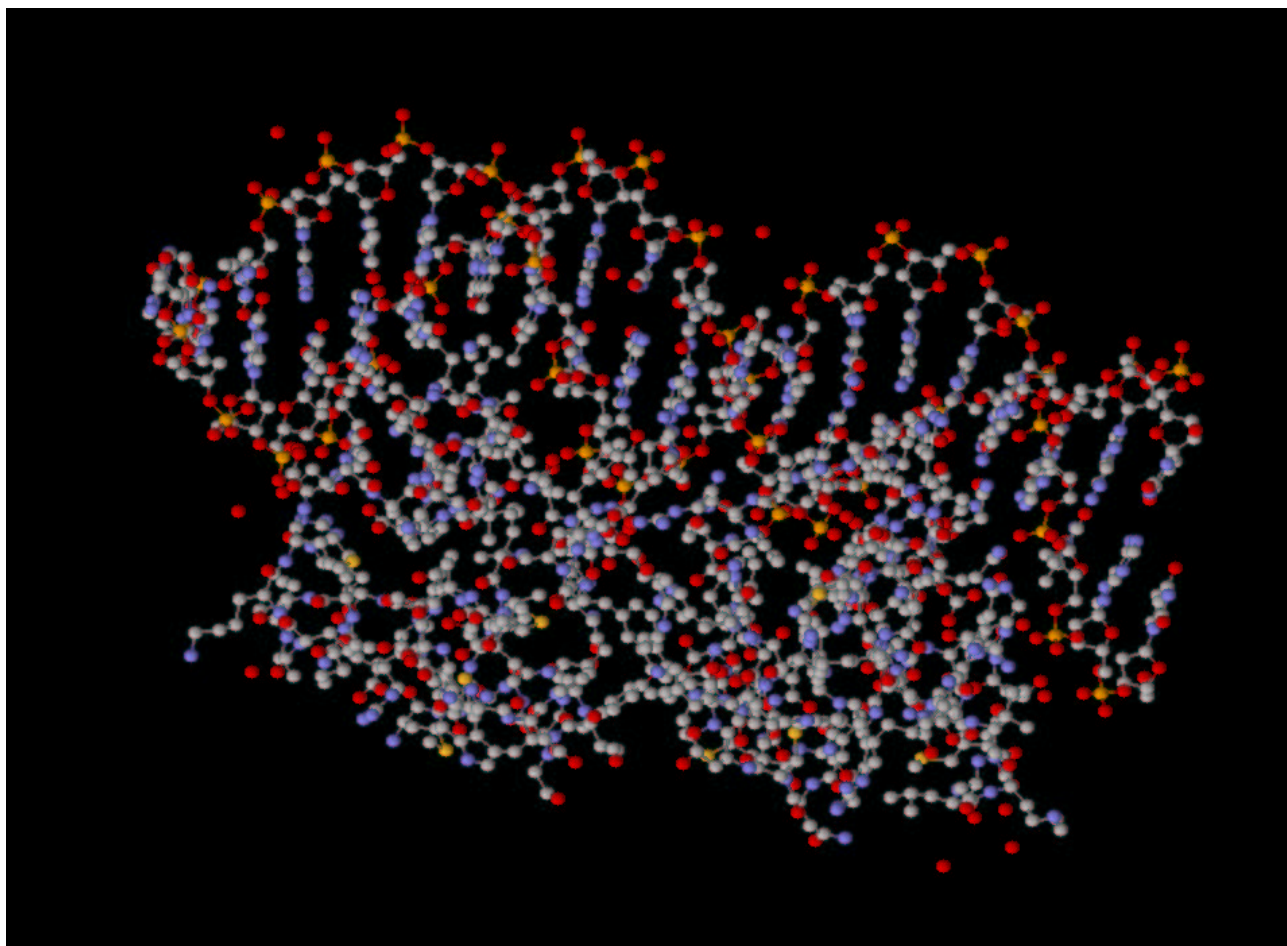
38 Novelli 1961-62

39 Allfrey / Mirsky 1962

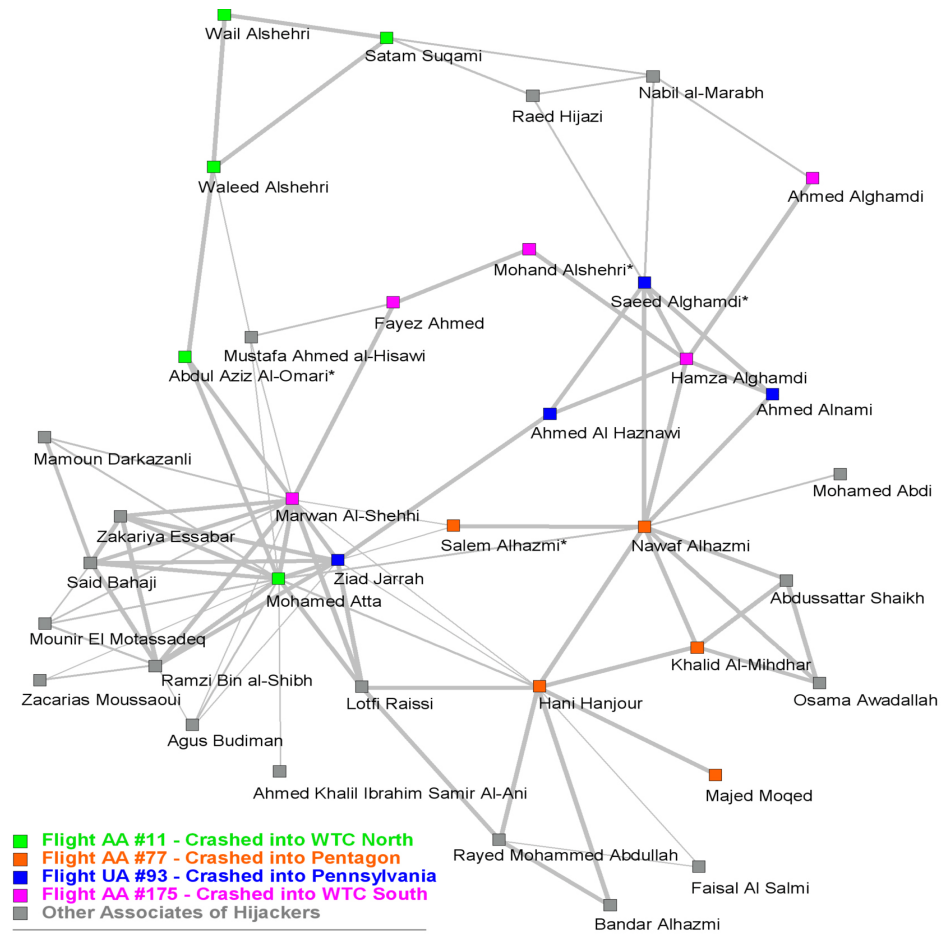
40 Nirenberg/Matthaei 1961-62

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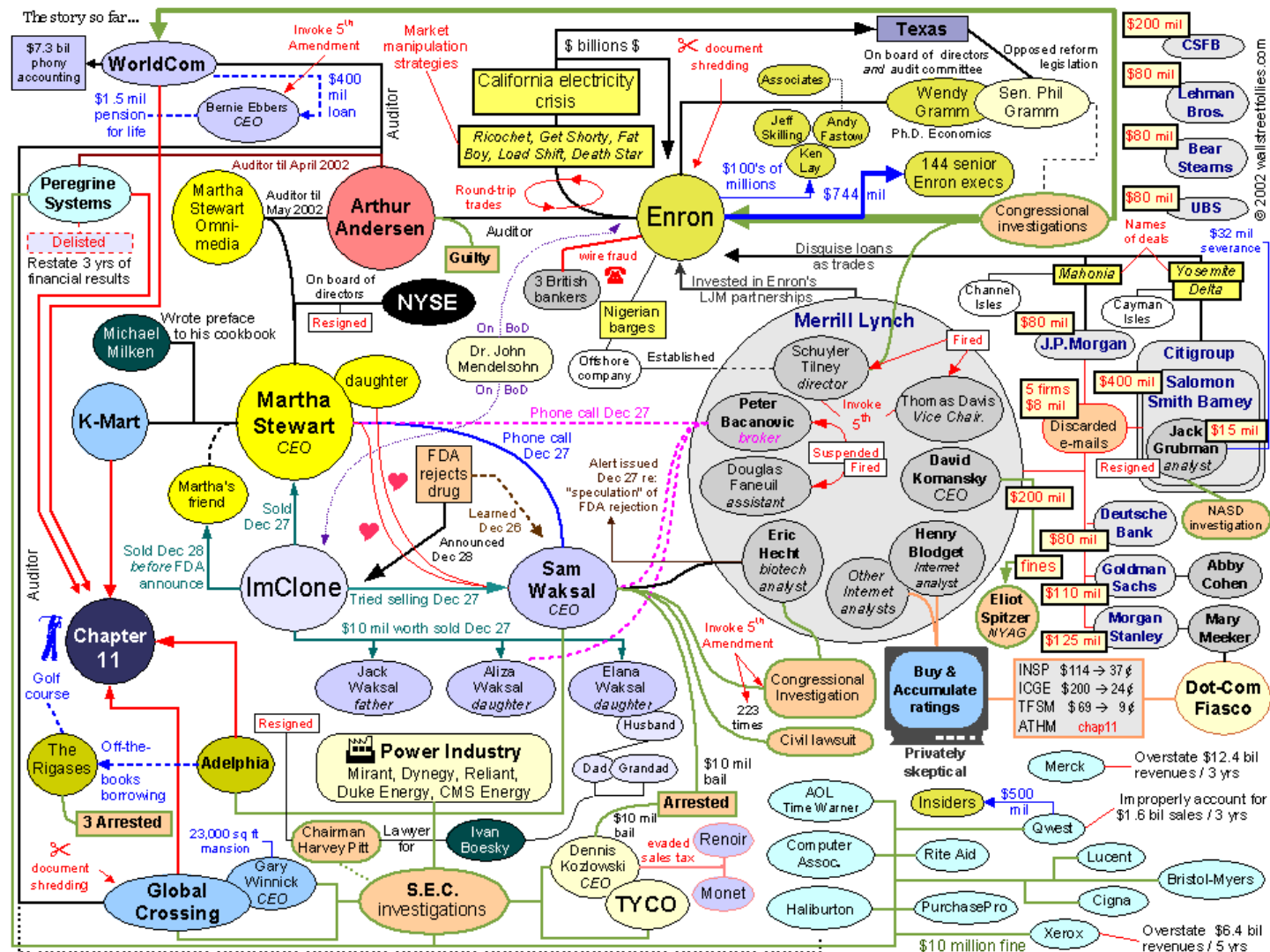
Organic molecule 3CRO



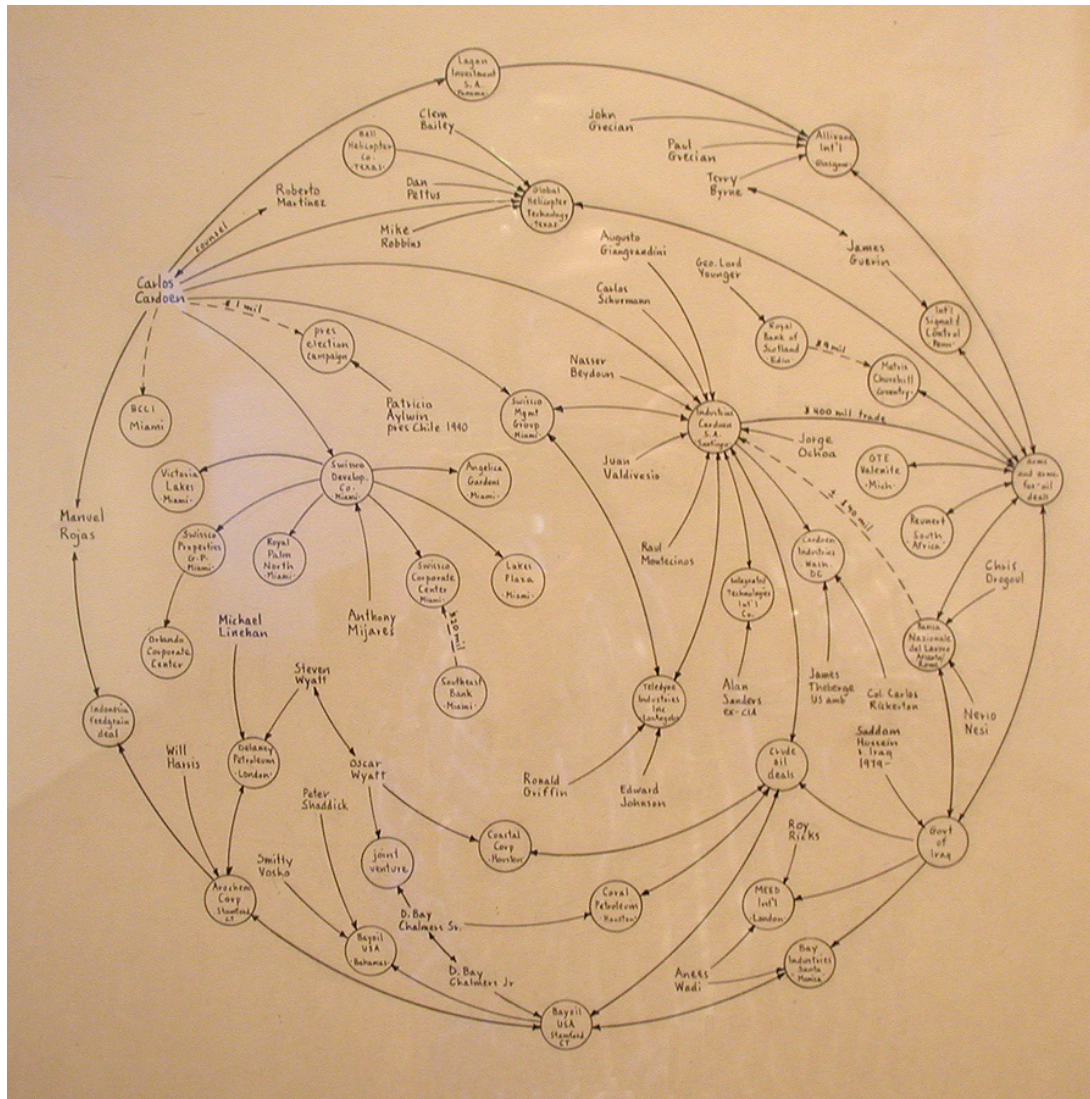
Hijackers (Krebs)



Wall Street Follies



Lombardi's networks

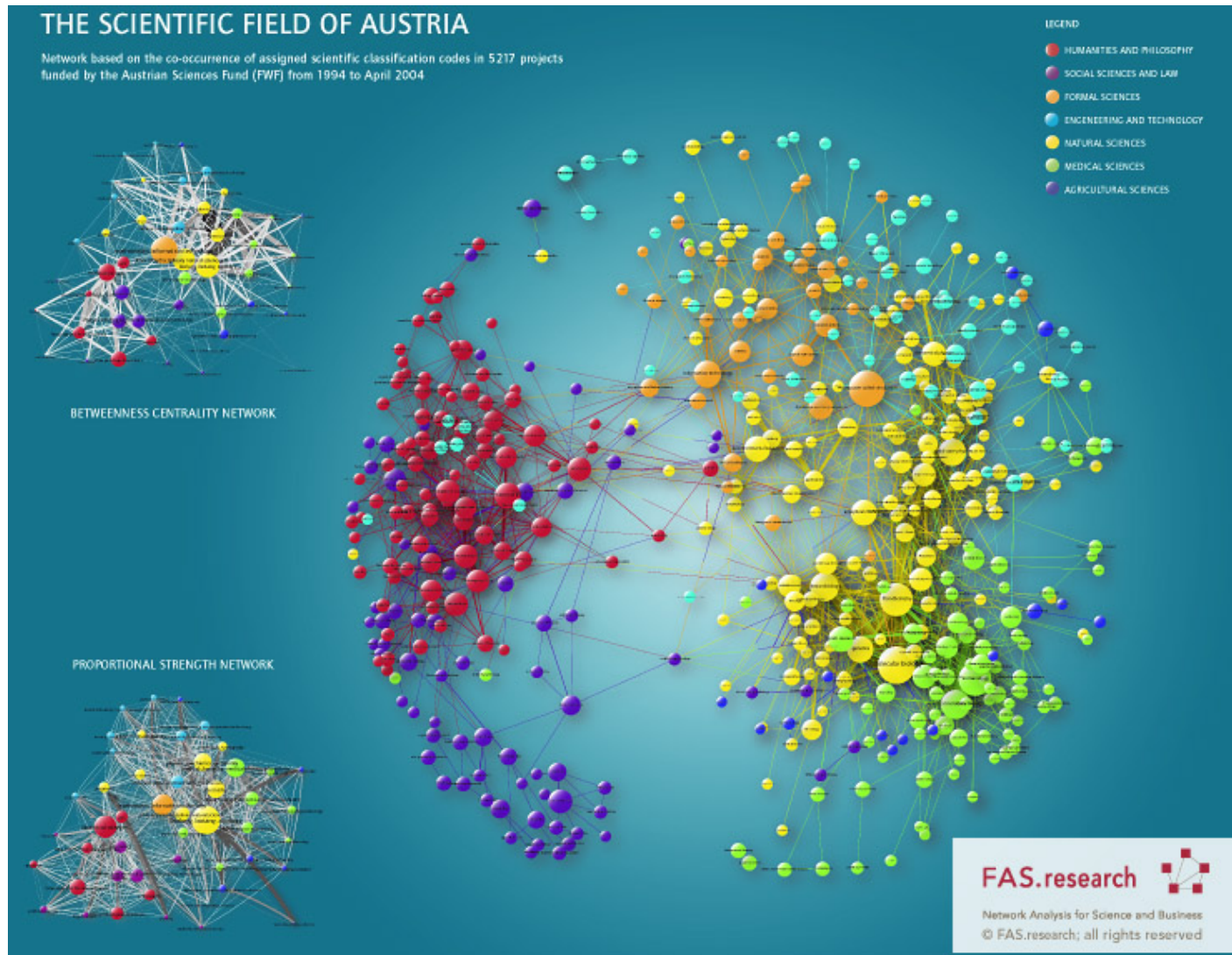


Mark Lombardi

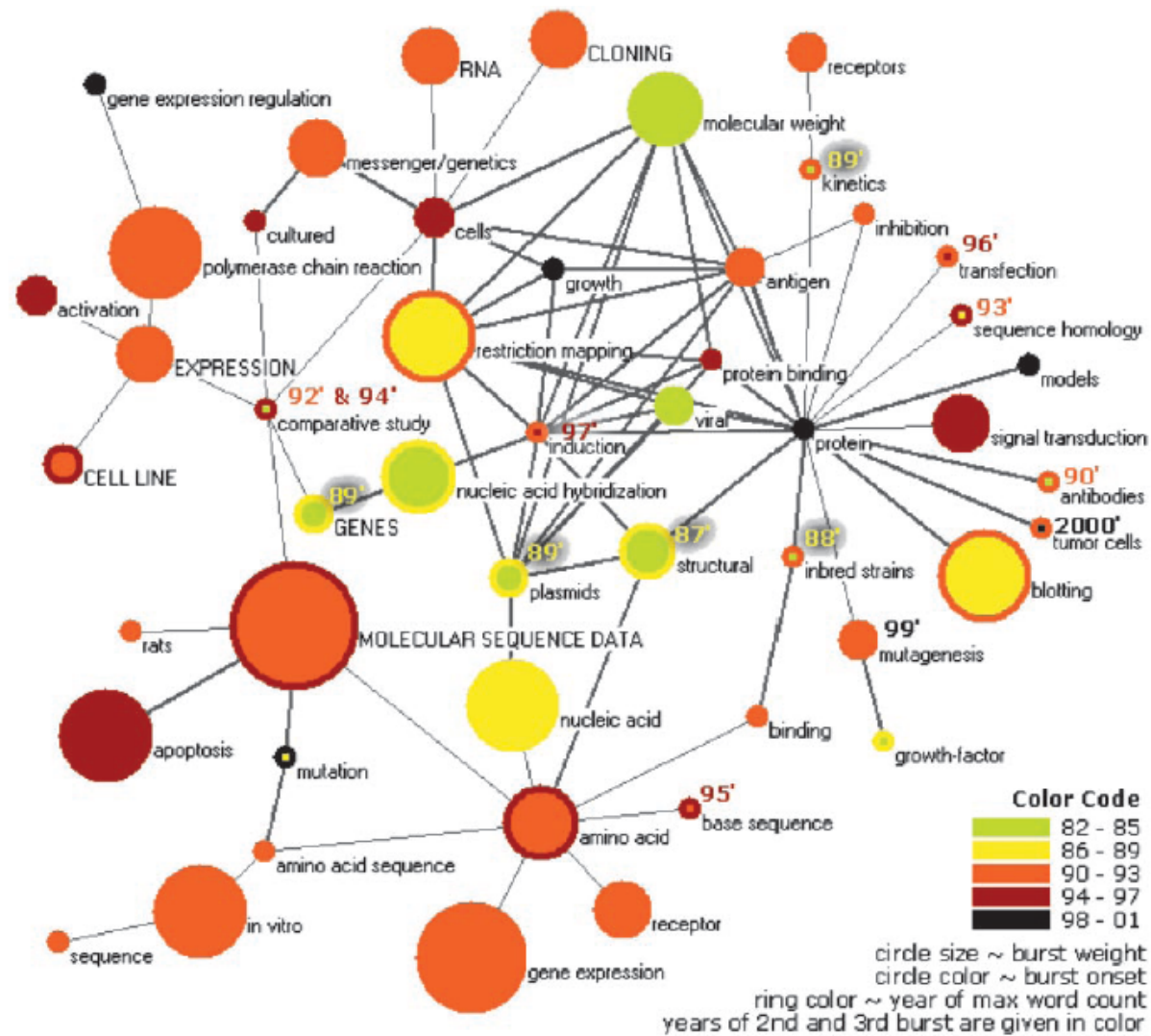
(1951-2000)

transformed business
relations into art.

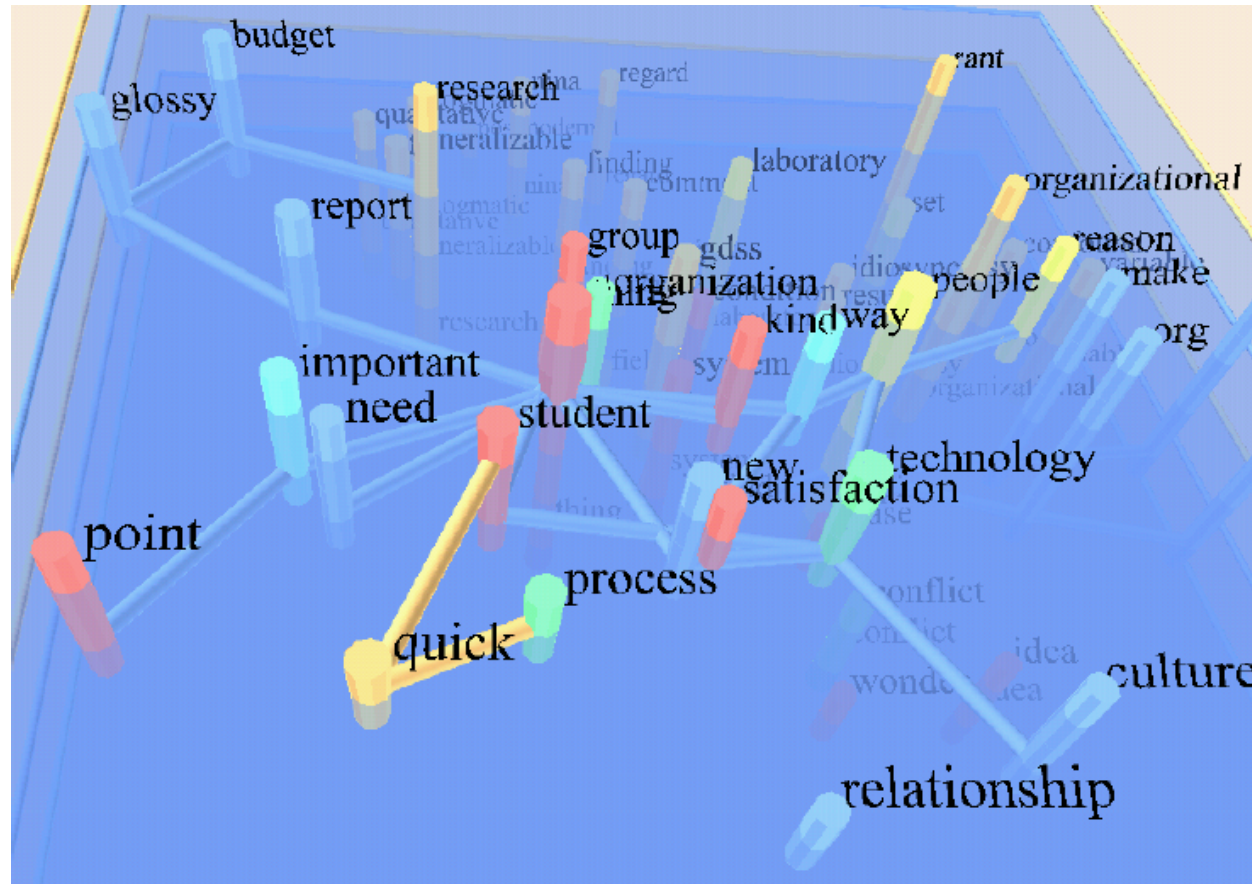
FAS: The scientific field of Austria



Katy Börner: Text analysis



Ulrik Brandes: Discourse network



How to get a network?

Collecting data about the network $\mathcal{N} = (\mathcal{V}, \mathcal{L}, \mathcal{P}, \mathcal{W})$ we have first to decide, what are the units (vertices) – *network boundaries*, when are two units related – *network completeness*, and which properties of vertices/lines we shall consider.

How to measure networks (questionnaires, interviews, observations, archive records, experiments, ...)?

What is the quality of measured networks (reliability and validity)?

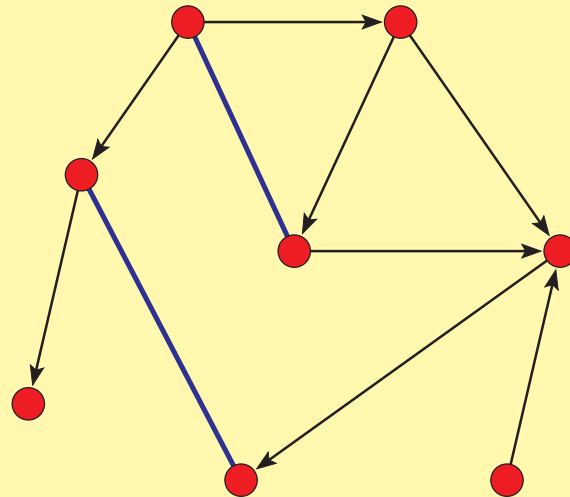
Several networks are already available in computer readable form or can be constructed from such data.

For large sets of units we often can't measure the complete network.

Therefore we limit the data collection to selected units and their neighbors.

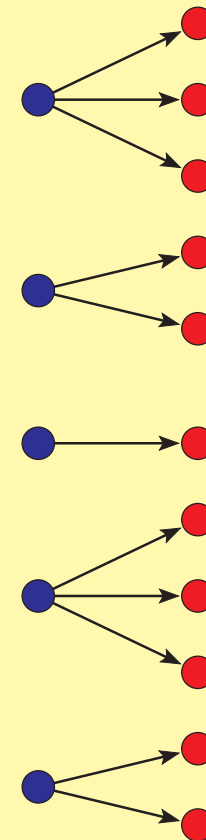
We get *ego-centered networks*.

Complete and ego-centered networks



COMPLETE NETWORK

Egos Alters



EGO-CENTERED NETWORKS

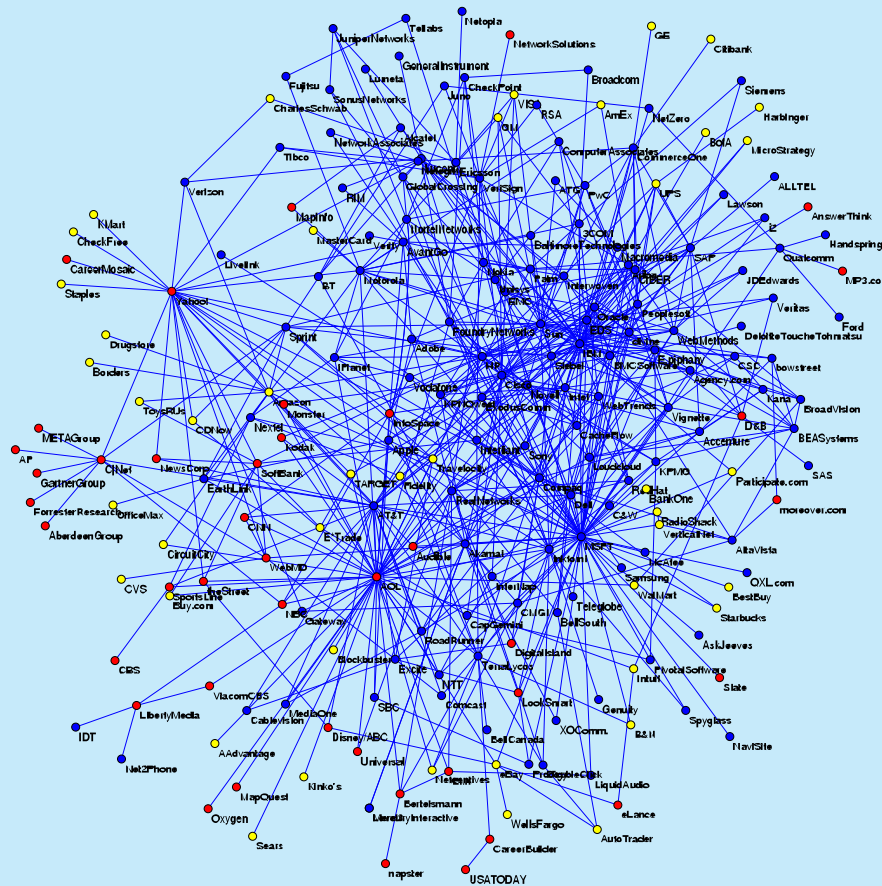
Use of existing network data

Pajek supports input of network data in several formats: UCINET's DL files, graphs from project Vega, molecules in MDLMOL, MAC, BS; genealogies in GEDCOM.

Davis.DAT, C84N24.VGR, MDL, 1CRN.BS, DNA.BS, ADF073.MAC, Bouchard.GED.

Several network data sets are already available in computer readable form and need only to be transformed into network descriptions.

Krebs Internet industries



Each node in the network represents a company that competes in the Internet industry, 1998 to 2001.

$n = 219$, $m = 631$.

red – content,

blue – infrastructure,

yellow – commerce.

Two companies are connected with an edge if they have announced a joint venture, strategic alliance or other partnership.

URL: <http://www.orgnet.com/netindustry.html>. *Recode*, *InfoRapid*.

Genealogies

For describing the genealogies on computer most often the GEDCOM format is used (*GEDCOM standard 5.5*).

Many such genealogies (files *.GED) can be found on the Web – for example *Roper's GEDCOMs* or *Isle-of-Man GEDCOMs*.

Several programs are available for preparation and maintenance of genealogies: free *GIM* and commercial *Brothers Keeper* (Slovenian version is available at *SRD*).

From the data collected in Phd. thesis:

Mahnken, Irmgard. 1960. Dubrovački patricijat u XIV veku. Beograd, Naučno delo.

the *Ragusa* network was produced.

GEDCOM

GEDCOM is a standard for storing genealogical data, which is used to interchange and combine data from different programs, which were used for entering the data.

```

0 HEAD
1 FILE ROYALS.GED
...
0 @I58@ INDI
1 NAME Charles Philip Arthur/Windsor/
1 TITL Prince
1 SEX M
1 BIRT
2 DATE 14 NOV 1948
2 PLAC Buckingham Palace, London
1 CHR
2 DATE 15 DEC 1948
2 PLAC Buckingham Palace, Music Room
1 FAMS @F16@
1 FAMC @F14@
...
0 @I65@ INDI
1 NAME Diana Frances /Spencer/
1 TITL Lady
1 SEX F
1 BIRT
2 DATE 1 JUL 1961
2 PLAC Park House, Sandringham
1 CHR
2 PLAC Sandringham, Church
1 FAMS @F16@
1 FAMC @F78@
...
...

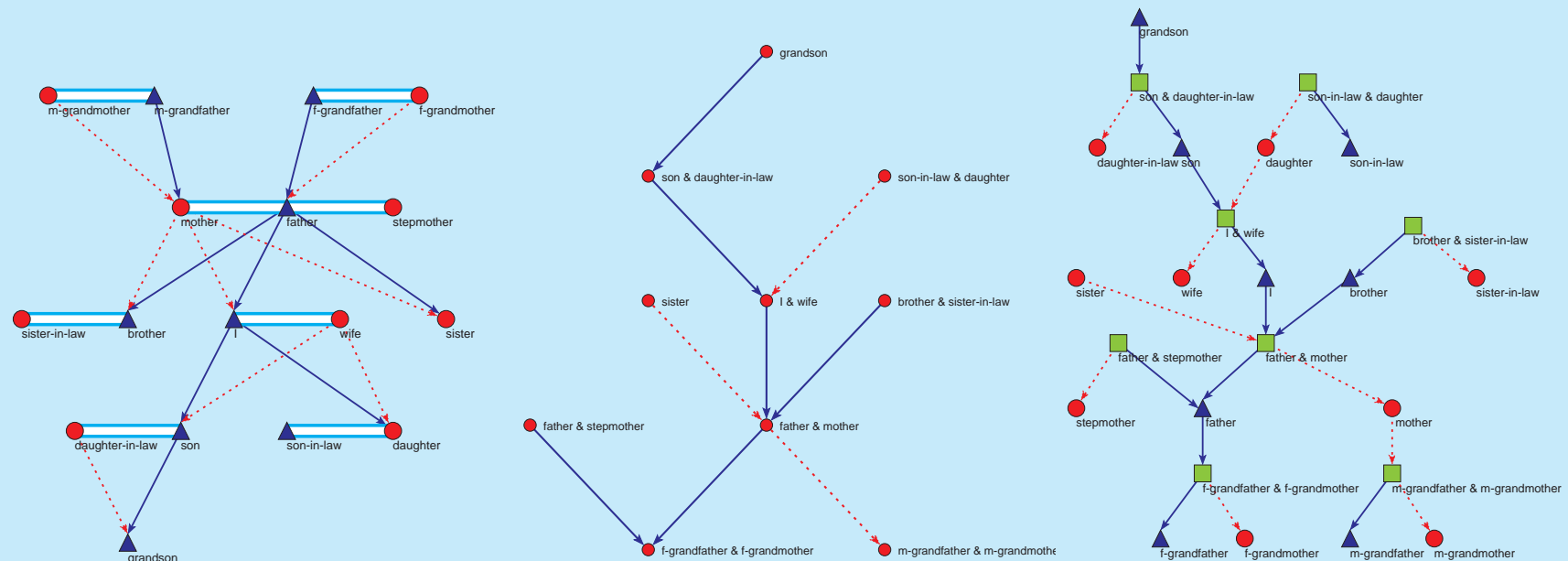
0 @I115@ INDI
1 NAME William Arthur Philip/Windsor/
1 TITL Prince
1 SEX M
1 BIRT
2 DATE 21 JUN 1982
2 PLAC St.Mary's Hospital, Paddington
1 CHR
2 DATE 4 AUG 1982
2 PLAC Music Room, Buckingham Palace
1 FAMC @F16@
...
0 @I116@ INDI
1 NAME Henry Charles Albert/Windsor/
1 TITL Prince
1 SEX M
1 BIRT
2 DATE 15 SEP 1984
2 PLAC St.Mary's Hosp., Paddington
1 FAMC @F16@
...
0 @F16@ FAM
1 HUSB @I58@
1 WIFE @I65@
1 CHIL @I115@
1 CHIL @I116@
1 DIV N
1 MARR
2 DATE 29 JUL 1981
2 PLAC St.Paul's Cathedral, London

```


Network representations of genealogies

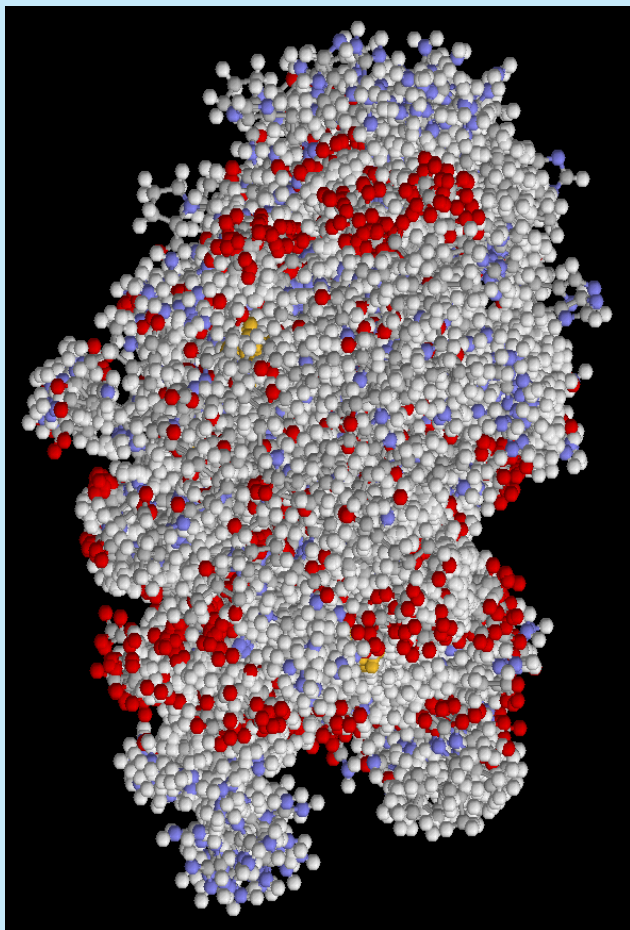
In usual *Ore* graph every person is represented with a vertex; they are linked with two relations: *are married* (blue edge) and *has child* (black arc) – partitioned into *is mother of* and *is father of*.

In *p-graph* the vertices are married couples or singles; they are linked with two relations: *is son of* (solid blue) and *is daughter of* (dotted red). More about p-graphs *D. White*.



Ore graph, p-graph, and bipartite p-graph

Molecular networks



virus 1GDY: $n = 39865$, $m = 40358$

In the **Brookhaven Protein Data Bank** we can find many large organic molecules (for example: Simian / 1AZ5.pdb) stored in PDB format.

They can be inspected in 3D using the program **Rasmol** (*RasMol*, *program*, *RasWin*) or *Protein Explorer*.

A molecule can be converted from PDB format into BS format (supported by **Pajek**) using the program *BabelWin* + *Babel16*.

GraphML

GraphML – XML format for network description.

L’Institut de Linguistique et Phonétique Générales et Appliquées (ILPGA),
Paris III; Traitement Automatique du Langage (TAL): **BaO4 : Des Textes
Aux Graphes Plurital**

LibXML, xsltproc download, XSLT, Xalan, Python, Sxslt.

```
xsltproc GraphML2Pajek.xsl graph.xml > graph.net  
java -jar saxon8.jar graph.xml GraphML2Pajek.xsl > graph.net  
java org.apache.xalan.xslt.Process -IN p.xml -XSL m.xsl -OUT p.txt
```

XSLT/Zvon

GraphML → Pajek

```
<?xml version="1.0" encoding="UTF-8"?>
<!-- Title: 1. D:\vlado\docs\Books\SKRIPTA\Nets\nets\graph.net (12) -->
<!-- Creator: Pajek: http://vlado.fmf.uni-lj.si/pub/networks/pajek/ -->
<!-- CreationDate: 11-03-2006, 17:25:13 -->
<graphml>
  <key id="a1" for="node" attr.name="Label" attr.type="string">
    <desc>Label of the node</desc> <default>NoLabel</default>
  </key>
  <key id="b1" for="edge" attr.name="Weight" attr.type="double">
    <desc>Weight (value) of the edge</desc> <default>1</default>
  </key>
  <graph id="G" edgedefault="directed" parse.nodes="12" parse.edges="23">
    <node id="v1"><data key="a1">a</data></node>
    <node id="v2"><data key="a1">b</data></node>
    <node id="v3"><data key="a1">c</data></node>
    <node id="v4"><data key="a1">d</data></node>
    <node id="v5"><data key="a1">e</data></node>
    <node id="v6"><data key="a1">f</data></node>
    <node id="v7"><data key="a1">g</data></node>
    <node id="v8"><data key="a1">h</data></node>
    <node id="v9"><data key="a1">i</data></node>
    <node id="v10"><data key="a1">j</data></node>
    <node id="v11"><data key="a1">k</data></node>
    <node id="v12"><data key="a1">l</data></node>
    <edge source="v1" target="v2"/> <edge source="v2" target="v1"/>
    <edge source="v1" target="v4"/> <edge source="v1" target="v6"/>
    <edge source="v2" target="v6"/> <edge source="v3" target="v2"/>
    <edge source="v3" target="v3"/> <edge source="v3" target="v7"/>
    <edge source="v3" target="v7"/> <edge source="v5" target="v3"/>
    <edge source="v5" target="v6"/> <edge source="v5" target="v8"/>
    <edge source="v6" target="v11"/> <edge source="v8" target="v4"/>
    <edge source="v10" target="v8"/> <edge source="v12" target="v5"/>
    <edge source="v12" target="v7"/> <edge source="v8" target="v12"/>
    <edge source="v12" target="v8"/>
    <edge directed="false" source="v2" target="v5"/>
    <edge directed="false" source="v3" target="v4"/>
    <edge directed="false" source="v5" target="v7"/>
    <edge directed="false" source="v6" target="v8"/>
  </graph>
</graphml>
```

```
*Vertices 12
1 "a"
2 "b"
3 "c"
4 "d"
5 "e"
6 "f"
7 "g"
8 "h"
9 "i"
10 "j"
11 "k"
12 "l"
*Edges
2 5
3 4
5 7
6 8
*Arcs
1 2
2 1
1 4
1 6
2 6
3 2
3 3
3 7
3 7
5 3
5 6
5 8
6 11
8 4
10 8
12 5
12 7
8 12
12 8
```

GraphML → Pajek using XSLT

```
<?xml version="1.0" encoding="iso-8859-1"?>
<xsl:stylesheet version="1.0" xmlns:xsl="http://www.w3.org/1999/XSL/Transform">
  <xsl:output method="text" encoding="iso-8859-1"/>
  <xsl:template match="/">
    <xsl:text>*Vertices </xsl:text>
    <xsl:value-of select="count(graphml/graph/node)"/>
    <xsl:text>#10;</xsl:text>
    <xsl:apply-templates select="graphml/graph/node"/>
    <xsl:text>*Edges#10;</xsl:text>
    <xsl:apply-templates select="graphml/graph/edge" mode="edge"/>
    <xsl:text>*Arcs#10;</xsl:text>
    <xsl:apply-templates select="graphml/graph/edge" mode="arc"/>
  </xsl:template>

  <xsl:template match="edge" mode="arc">
    <xsl:if test="not(./@directed='false')">
      <xsl:value-of select="substring(./@source,2)"/>
      <xsl:text> </xsl:text>
      <xsl:value-of select="substring(./@target,2)"/>
      <xsl:text> </xsl:text>
      <xsl:value-of select="./data"/>
      <xsl:text>#10;</xsl:text>
    </xsl:if>
  </xsl:template>

  <xsl:template match="edge" mode="edge">
    <xsl:if test="./@directed='false'">
      <xsl:value-of select="substring(./@source,2)"/>
      <xsl:text> </xsl:text>
      <xsl:value-of select="substring(./@target,2)"/>
      <xsl:text> </xsl:text>
      <xsl:value-of select="./data"/>
      <xsl:text>#10;</xsl:text>
    </xsl:if>
  </xsl:template>

  <xsl:template match="node">
    <xsl:value-of select="substring(./@id,2)"/>
    <xsl:text> " </xsl:text>
    <xsl:value-of select="./data"/>
    <xsl:text>"#10;</xsl:text>
  </xsl:template>
</xsl:stylesheet>
```


Approaches to computer-assisted text analysis

R. Popping: **Computer-Assisted Text Analysis** (2000) distinguishes three main approaches to CaTA: *thematic* TA, *semantic* TA, and *network* TA.

Terms considered in TA are collected in a *dictionary* (it can be fixed in advance, or built dynamically). The main two problems with terms are *equivalence* (different words representing the same term) and *ambiguity* (same word representing different terms). Because of these the *coding* – transformation of raw text data into formal *description* – is done mainly manually or semiautomatically. As *units* of TA we usually consider clauses, statements, paragraphs, news, messages, ...

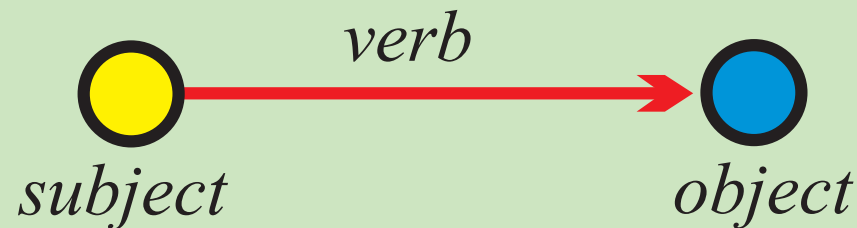
Till now the thematic and semantic TA mainly used statistical methods for analysis of the coded data.

...approaches to CaTA

In thematic TA the units are coded as rectangular matrix $\text{Text units} \times \text{Concepts}$ which can be considered as a two-mode network.

Examples: M.M. Miller: **VBPro**, H. Klein: **Text Analysis/ TextQuest**.

In semantic TA the units (often clauses) are encoded according to the S-V-O (*Subject-Verb-Object*) model or its improvements.

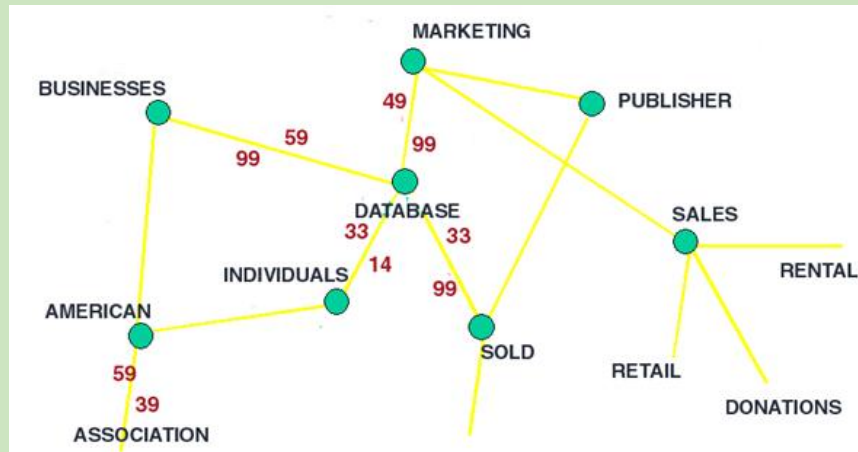


Examples: **Roberto Franzosi**; **KEDS**, **Tabari**.

This coding can be directly considered as network with $\text{Subjects} \cup \text{Objects}$ as vertices and lines labeled with *Verbs*.

See also **RDF** triples in **semantic web**.

Network CaTA



TextAnalyst's 'semantic network'

This way we already stepped into the network TA.

Examples:

Carley: **Cognitive maps**,

J.A. de Ridder: **CETA**,

Megaputer: **TextAnalyst**.

See also: W. Evans: **Computer Environments for Content Analysis**, K.A. Neuendorf: **The Content Analysis Guidebook / Online** and H.D. White: **Publications**.

There are additional ways to obtain networks from textual data.

TA – Dictionary networks

book

A collection of [leaves](#) of [paper](#), [parchment](#), [vellum](#), cloth, or other material (written, [printed](#), or [blank](#)) fastened together along one edge, with or without a protective [case](#) or [cover](#). Also refers to a literary [work](#) or one of its [volumes](#). Compare with [monograph](#).

To qualify for the special parcel post rate known in the United States as [media rate](#), a [publication](#) must consist of 24 or more [pages](#), at least 22 of which bear [printing](#) consisting primarily of reading material or scholarly [bibliography](#), with advertising limited to [book announcements](#). UNESCO defines a book as a non[periodical](#) literary publication consisting of 49 or more pages, covers excluded. The [ANSI standard](#) includes publications of less than 49 pages which have [hard covers](#). *See also:* [art book](#), [board book](#), [children's book](#), [coffee table book](#), [gift book](#), [licensed book](#), [managed book](#), [new book](#), [packaged book](#), [picture book](#), [premium book](#), [professional book](#), [promotional book](#), [rare book](#), [reference book](#), [religious book](#), and [reprint book](#).

Also, a major division of a longer [work](#) (usually of [fiction](#)) which is further subdivided into [chapters](#). Usually [numbered](#), such a division may or may not have its own [title](#). Also refers to one of the divisions of the Christian *Bible*, the first being *Genesis*.

book description in ODLIS

The Edinburgh Associative Thesaurus (*EAT*) / *net*; NASA Thesaurus.

Paper.

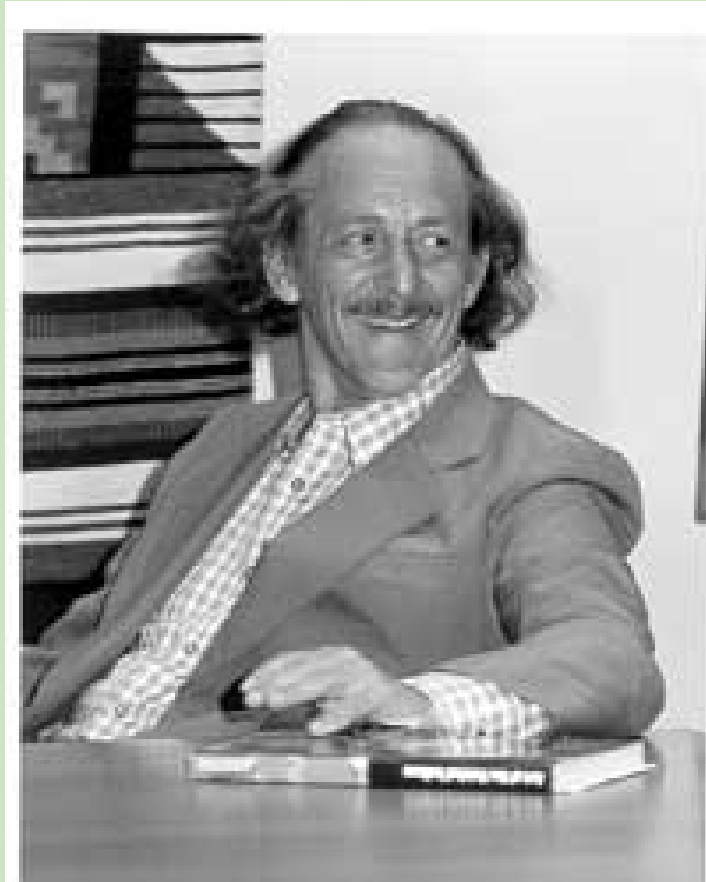
In a *dictionary graph* the terms determine the set of vertices, and there is an arc (u, v) from term u to term v iff the term v appears in the description of term u .

Online Dictionary of Library and Information Science *ODLIS*, *Odlis.net* (2909 / 18419).

Free On-line Dictionary of Computing *FOLDOC*, *Foldoc2b.net* (133356 / 120238).

Artlex, *Wordnet*, *ConceptNet*, *OpenCyc*.

TA – Citation networks

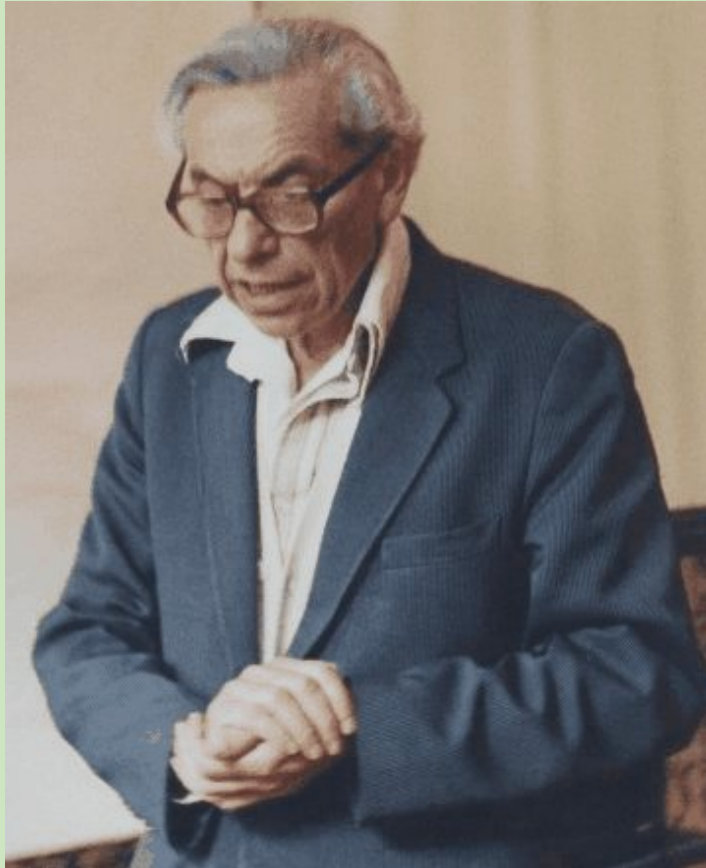


In a *citation graph* the vertices are different publications from the selected area; two publications are connected by an arc if the first is cited by the second. The citation networks are almost acyclic.

E. Garfield: *HistCite* / *Pajek*, *papers*.

An example of very large citation network is *US Patents* / *Nber*,
 $n = 3774768$, $m = 16522438$.

TA – Collaboration networks



Units in a *collaboration network* are usually individuals or institutions. Two units are related if they produced a joint work. The weight is the number of such works.

A famous example of collaboration network is *The Erdős Number Project*, *Erdos.net*.

A rich source of data for producing collaboration networks are the BibTeX bibliographies *Nelson H. F. Beebe's Bibliographies Page*.

For example B. Jones: *Computational geometry database* (2002), *FTP*, *Geom.net*.

An initial collaboration network from such data can be produced using some programming. Then follows a tedious 'cleaning' process.

Interesting datasets: *The Internet Movie Database* and *Trier DBLP*.

Both citation and collaboration networks can be obtained from *Web of Science* using *WoS2Pajek*.

TA – International Relations

Paul Hensel's International Relations Data Site,

International Conflict and Cooperation Data,

Correlates of War,

Kansas Event Data System *KEDS*,

KEDS in Pajek's format.

Recoding programs in R.

Recoding of KEDS/WEIS data in Pajek's format

```
% Recoded by WEISmonths, Sun Nov 28 21:57:00 2004
% from http://www.ku.edu/~keds/data.dir/balk.html
*vertices 325
1 "AFG" [1-*]
2 "AFR" [1-*]
3 "ALB" [1-*]
4 "ALBMED" [1-*]
5 "ALG" [1-*]

...
318 "YUGGOV" [1-*]
319 "YUGMAC" [1-*]
320 "YUGMED" [1-*]
321 "YUGMTN" [1-*]
322 "YUGSER" [1-*]
323 "ZAI" [1-*]
324 "ZAM" [1-*]
325 "ZIM" [1-*]
*arcs :0 "*** ABANDONED"
*arcs :10 "YIELD"
*arcs :11 "SURRENDER"
*arcs :12 "RETREAT"

...
*arcs :223 "MIL ENGAGEMENT"
*arcs :224 "RIOT"
*arcs :225 "ASSASSINATE TORTURE"
*arcs
224: 314 153 1 [4]          890402 YUG      KSV      224 (RIOT) RIOT-TORN
212: 314 83 1 [4]          890404 YUG      ETHALB  212 (ARREST PERSON) ALB ETHNIC JAILED IN YUG
224: 3 83 1 [4]            890407 ALB      ETHALB  224 (RIOT) RIOTS
123: 83 153 1 [4]          890408 ETHALB  KSV      123 (INVESTIGATE) PROBING

...
42: 105 63 1 [175]         030731 GER      CYP      042 (ENDORSE) GAVE SUPPORT
212: 295 35 1 [175]        030731 UNWCT    BOSSER  212 (ARREST PERSON) SENTENCED TO PRISON
43: 306 87 1 [175]         030731 VAT      EUR      043 (RALLY) RALLIED
13: 295 35 1 [175]         030731 UNWCT    BOSSER  013 (RETRACT) CLEARED
121: 295 22 1 [175]        030731 UNWCT    BAL      121 (CRITICIZE) CHARGES
122: 246 295 1 [175]       030731 SER      UNWCT    122 (DENIGRATE) TESTIFIED
121: 35 295 1 [175]        030731 BOSSER   UNWCT    121 (CRITICIZE) ACCUSED
```

... Recoding programs in R

To recode the KEDS/WEIS data we used short programs in R, such as the following one:

```
# WEISmonths
# recoding of WEIS files into Pajek's multirelational temporal files
# granularity is 1 month
# -----
# Vladimir Batagelj, 28. November 2004
# -----
# Usage:
#   WEISmonths(WEIS_file,Pajek_file)
# Examples:
#   WEISmonths('Balkan.dat','BalkanMonths.net')
# -----
# http://www.ku.edu/~keds/data.html
# -----

WEISmonths <- function(fdat,fnet){

  get.codes <- function(line){
    nlin <- nlin + 1;
    z <- unlist(strsplit(line,"\t")); z <- z[z != ""]
    if (length(z)>4) {
      t <- as.numeric(z[1]); if (t < 500000) t <- t + 1000000
      if (t<t0) t0 <- t; u <- z[2]; v <- z[3]; r <- z[4]
      if (is.na(as.numeric(r))) cat(nlin,'NA rel-code',r,'\n')
      h <- z[5]; h <- substr(h,2,nchar(h)-1)
      if (nchar(h) == 0) h <- '*** missing description'
      if (!exists(u,env=act,inherits=FALSE)){
        nver <- nver + 1; assign(u,nver,env=act) }
      if (!exists(v,env=act,inherits=FALSE)){
        nver <- nver + 1; assign(v,nver,env=act) }
      if (!exists(r,env=rel,inherits=FALSE)) assign(r,h,env=rel)
    }
  }
}
```

... Recoding programs in R

```

recode <- function(line){
  nlin <- nlin + 1;
  z <- unlist(strsplit(line,"\t")); z <- z[z != ""]
  if (length(z)>4) {
    t <- as.numeric(z[1]); if (t < 500000) t <- t + 1000000
    cat(as.numeric(z[4]),': ',get(z[2],env=act,inherits=FALSE),
        ' ',get(z[3],env=act,inherits=FALSE),' 1 [',
        12*(1900 + t %/% 10000) + (t %/% 10000) %/% 100 - t0,
        ']\n',sep='',file=net)
  }
}

cat('WEISmonths: WEIS -> Pajek\n')
ts <- strsplit(as.character(Sys.time())," ")[[1]][2]
act <- new.env(TRUE,NULL); rel <- new.env(TRUE,NULL)
dat <- file(fdat,"r"); net <- file(fnet,"w")
lst <- file('WEIS.lst',"w"); dni <- 0
nver <- 0; nlin <- 0; t0 <- 9999999
lines <- readLines(dat); close(dat)
sapply(lines,get.codes)
a <- sort(ls(envir=act)); n <- length(a)
cat(paste('% Recoded by WEISmonths,',date()),"\n",file=net)
cat("% from http://www.ku.edu/~keds/data.html\n",file=net)
cat("*vertices",n,"\n",file=net)
for(i in 1:n){ assign(a[i],i,env=act);
  cat(i,' ','',a[i],' " [1-*]\n',sep='',file=net) }
b <- sort(ls(envir=rel)); m <- length(b)
for(i in 1:m){ assign(a[i],i,env=act);
  cat("*arcs :",as.numeric(b[i]),' ','',
  get(b[i],env=rel,inherits=FALSE),' "\n',sep='',file=net) }
t0 <- 12*(1900 + t0 %/% 10000)
slice <- 0
cat("*arcs\n",file=net); nlin <- 0
sapply(lines,recode)
cat(' ',nlin,'lines processed\n'); close(net)
te <- strsplit(as.character(Sys.time())," ")[[1]][2]
cat('  start:',ts,'  finish:',te,'\n')
}

WEISmonths('Balkan.dat','BalkanMonthsR.net')

```

Note: The dictionary data structure is in R implemented as *environment*.

Neighbors

Let \mathcal{V} be a *set of multivariate units* and $d(u, v)$ a *dissimilarity* on it. They determine two types of networks:

The *k-nearest neighbors* network: $\mathcal{N}(k) = (\mathcal{V}, \mathcal{A}, d)$

$$(u, v) \in \mathcal{A} \Leftrightarrow v \text{ is among } k \text{ nearest neighbors of } u, \quad w(u, v) = d(u, v)$$

The *r-neighbors* network: $\mathcal{N}(r) = (\mathcal{V}, \mathcal{E}, d)$

$$(u : v) \in \mathcal{E} \Leftrightarrow d(u, v) \leq r, \quad w(u, v) = w(v, u) = d(u, v)$$

These networks provide a link between data analysis and network analysis.
Efficient algorithms ?!

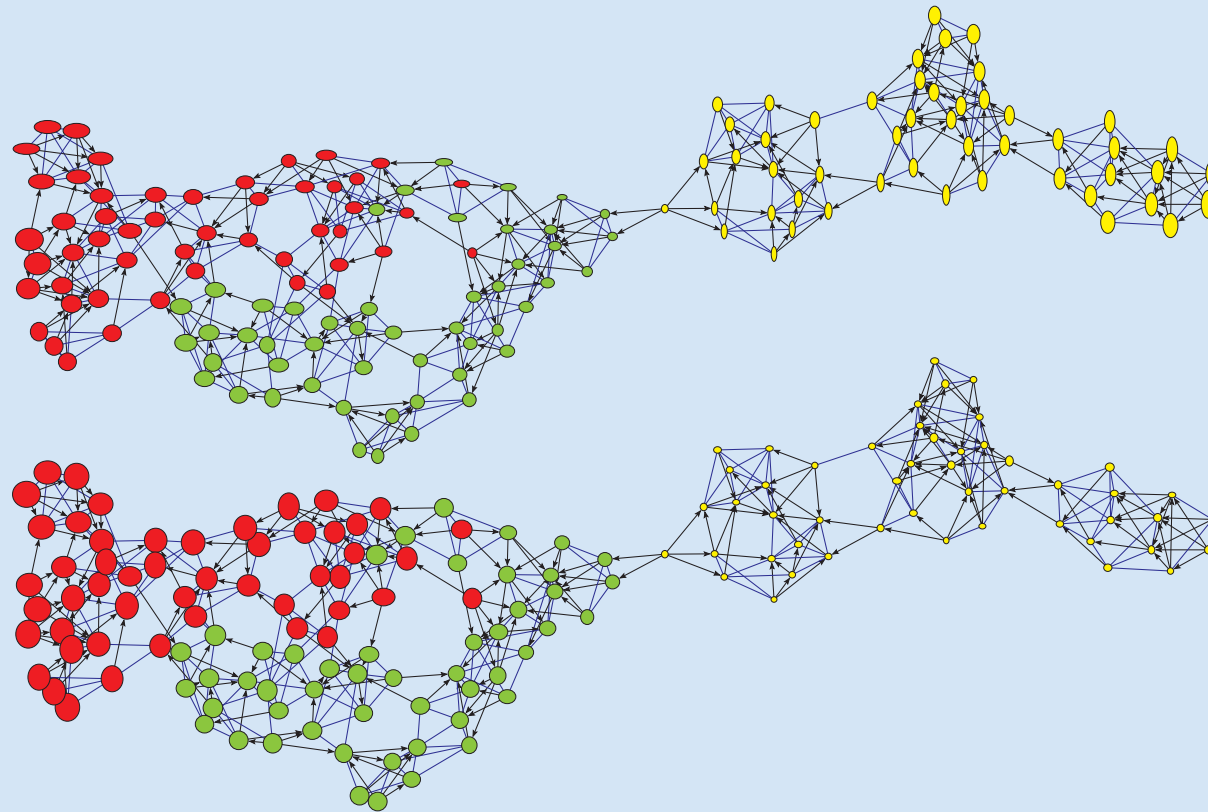
Fisher's *Iris data*.

Details on *Multivariate networks* and procedures in R.

Nearest k neighbors in R

```
k.neighbor2Net <-
# stores network of first k neighbors for
# dissimilarity matrix d to file fnet in Pajek format.
function(fnet,d,k){
  net <- file(fnet,"w")
  n <- nrow(d); rn <- rownames(d)
  cat("*vertices",n,"\n",file=net)
  for (i in 1:n) cat(i," \"",rn[i],"\"\\n",sep="",file=net)
  cat("*arcs\\n",file=net)
  for (i in 1:n) for (j in order(d[i,])[1:k+1]) {
    cat(i,j,d[i,j],"\\n",file=net)
  }
  close(net)
}
stand <-
# standardizes vector x .
function(x){
  s <- sd(x)
  if (s > 0) (x - mean(x))/s else x - x
}
data(iris)
ir <- cbind(stand(iris[,1]),stand(iris[,2]),stand(iris[,3]),
  stand(iris[,4]))
k.neighbor2Net("iris5.net",as.matrix(dist(ir)),5)
```


Fisher's Irises



Draw/Draw-Partition-2Vectors

The size of vertices is proportional to normalized (Sepal.Length, Sepal.Width) and (Petal.Length, Petal.Width). The color of vertices is determined by the original partition. *Iris data*.

Transformations

Words graph – words from a given set are vertices; two words are related iff one can be obtained from the other by change (add, delete, replace) of a single character. DIC28, *Paper*.

Text network – vertices are (selected) words from a given text; two words are related if they coappeared in the selected type of 'window' (same sentence, k consecutive words, ...) The weights count such coappearances. Example *CRA*.

Game graph – vertices are states in the game; two states are linked with an arc if the rules of the game allow the transition from first to the second state.

Networks from the Internet



KartOO network

Internet Mapping Project.

Links among WWW pages.

KartOO, TouchGraph.

Derived from archives of E-mail, blogs, . . . , server's logs.

Cybergeography, CAIDA.

Tools: *MedlineR, SocSciBot.*

Collecting Networks from WWW

Web wrappers are special programs for collecting information from web pages – often returned in XML format.

Examples in R: [Titles of patents from Nber](#), [Books from Amazon](#).

Several tools for automatic generation of wrappers: ([paper](#) / [list](#) / [LAPIS](#)).

Free programs: XWRAP ([description](#) / [page](#)) in TSIMMIS ([description](#) / [page](#)).

Among commercial programs it seems the best is [lixto](#).

Additional URLs [1](#), [2](#), [3](#).

Networks from Amazon in R

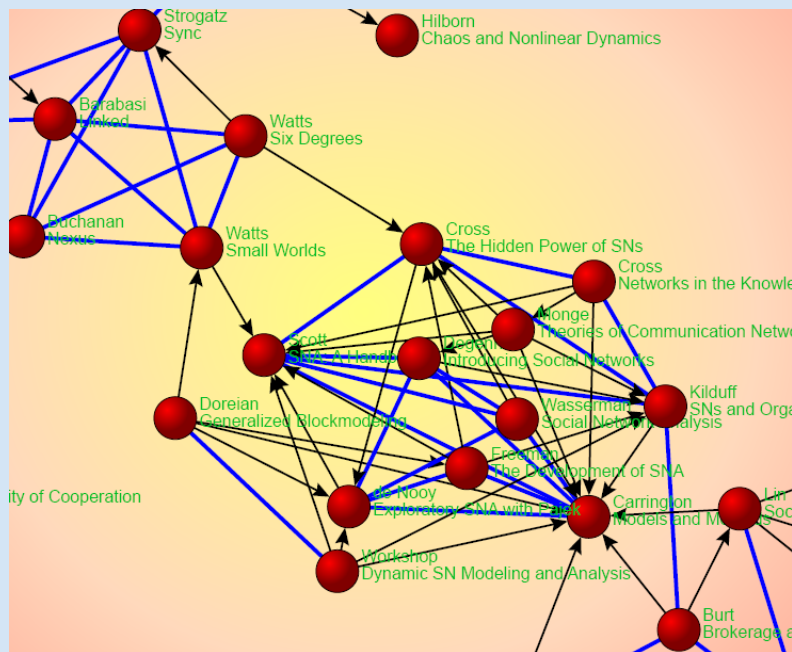
```
amazon <- function(fvtx,flnk,ftit,maxver){
# Creates a network of books from Amazon
# amazon('v.txt','a.txt','t.txt',10)
# Vladimir Batagelj, 20-21. nov. 2004 / 10. nov. 2006
  opis <- function(line){
    i <- regexpr('\>',line); l <- i[1]+attr(i,"match.length")[1]
    j <- regexpr('</a>',line); r <- j[1]-1; substr(line,l,r)
  }
  vid <- new.env(hash=TRUE,parent=emptyenv())
  vtx <- file(fvtx,"w"); cat('*vertices\n', file=vtx)
  tit <- file(ftit,"w"); cat('*vertices\n', file=tit)
  lnk <- file(flnk,"w"); cat('*arcs\n',file=lnk)
  url1 <- 'http://www.amazon.com/exec/obidos/tg/detail/-/'
  url2 <- '?v=glance';
  book <- '0521840856'
  auth <- "Patrick Doreian"
  titl <- "Generalized Blockmodeling"
  narc <- 0; nver <- 1
  page <- paste(url1,book,url2,sep='')
  cat(nver, ' "', book, '" URL "',page,'" \n', sep='', file=vtx)
  cat(nver, ' "', auth, ':\n',titl, '" \n', sep='', file=tit)
  assign(book,nver,env=vid)
  cat('new vertex ',nver,' - ',book,'\n')
  books <- c(book)
```

```

while (length(books)>0) {
  bk <- books[1]; books <- books[-1]
  vini <- get(bk,env=vid); cat(vini,'\n')
  page <- paste(url1,bk,url2,sep='')
  stran <- readLines(con<-url(page)); close(con)
  i <- grep("Customers who bought",stran,ignore.case=TRUE) [1]
  if (is.na(i)) break
  j <- grep("Explore Similar Items",stran,ignore.case=TRUE) [1]
  izrez <- stran[i:j]; izrez <- izrez[-which(izrez=="")]
  izrez <- izrez[-which(izrez==" ")]
  ik <- regexpr("/dp/",izrez); ii <- ik+attr(ik,"match.length")
  for (k in 1:length(ii)) {
    j <- ii[k];
    if (j > 0) {
      bk <- substr(izrez[k],j,j+9); cat('test',k,bk,'\n')
      if (exists(bk,env=vid,inherits=FALSE)){
        vter <- get(bk,env=vid,inherits=FALSE)
      } else {
        nver <- nver + 1; vter <- nver; line <- izrez[k]
        assign(bk,nver,env=vid)
        if (nver <= maxver) {books <- append(books,bk)}
        cat(nver,'"',bk,'" URL "',url1,bk,url2,'"'\n',sep='',file=vtx)
        cat('new vertex ',nver,' - ',bk,'\n');
        t <- opis(line); line <- izrez[k+1]
        if (substr(line,1,2)=='by') {a <- substr(line,4,100)}
        else { a <- 'UNKNOWN' }
        cat(nver,'"',a,':\n',t,'"'\n',sep='',file=tit)
      }
      narc <- narc + 1; cat(vini,vter,'\n', file=lnk)
    }
  }
  flush.console()
}
close(lnk); close(vtx); cat('Amazon - END\n')
}

```


Networks from Amazon – books on SNA



Books in SNA from Amazon, 10. november 2006; Starting point P. Doreian &: **Generalized Block-modeling**.

SVG picture. Files/ZIP.

The program in R is just a skeleton. Possible improvements: list of starting points; continuation after interrupts; ...

Random networks

Several types of networks can be produced randomly using special generators. The theoretical **background** of these generators is beyond the goals of this workshop.

Some of them are implemented in **Pajek** under

Net / Random network

but can be also described by the following **functions in R**.

Available is also a program **GeneoRnd** for generating random genealogies.

Random undirected graph of Erdős-Rényi type

```

dice <- function(n=6){return(1+trunc(n*runif(1,0,1)))}

ErdosRenyiNet <-
# generates a random undirected graph of Erdos-Renyi type
# with n vertices and m edges, and stores it on the file
# fnet in Pajek's format.
# Example:
#   ErdosRenyiNet('testER.net',100,175)
# -----
# by Vladimir Batagelj, R version: Ljubljana, 20. Dec 2004
# based on ALG.2 from: V. Batagelj, U. Brandes:
#   Efficient generation of large random networks
function(fnet,n,m){
  net <- file(fnet,"w"); cat("*vertices",n,"\n",file=net)
  cat('% random Erdos-Renyi undirected graph G(n,m) / m = ',
      m,'\n',file=net)
  # for (i in 1:n) cat(i," \"v\",i,\"\"\n",sep="",file=net)
  cat("*edges\n",file=net); L <- new.env(TRUE,NULL)
  for (i in 1:m){
    repeat { u <- dice(n); v <- dice(n)
      if (u!=v) {
        edge <- if (u<v) paste(u,v) else paste(v,u)
        if (!exists(edge,env=L,inherits=FALSE)) break }
    }
    assign(edge,0,env=L); cat(edge,'\n',file=net)
  }
  close(net)
}

```