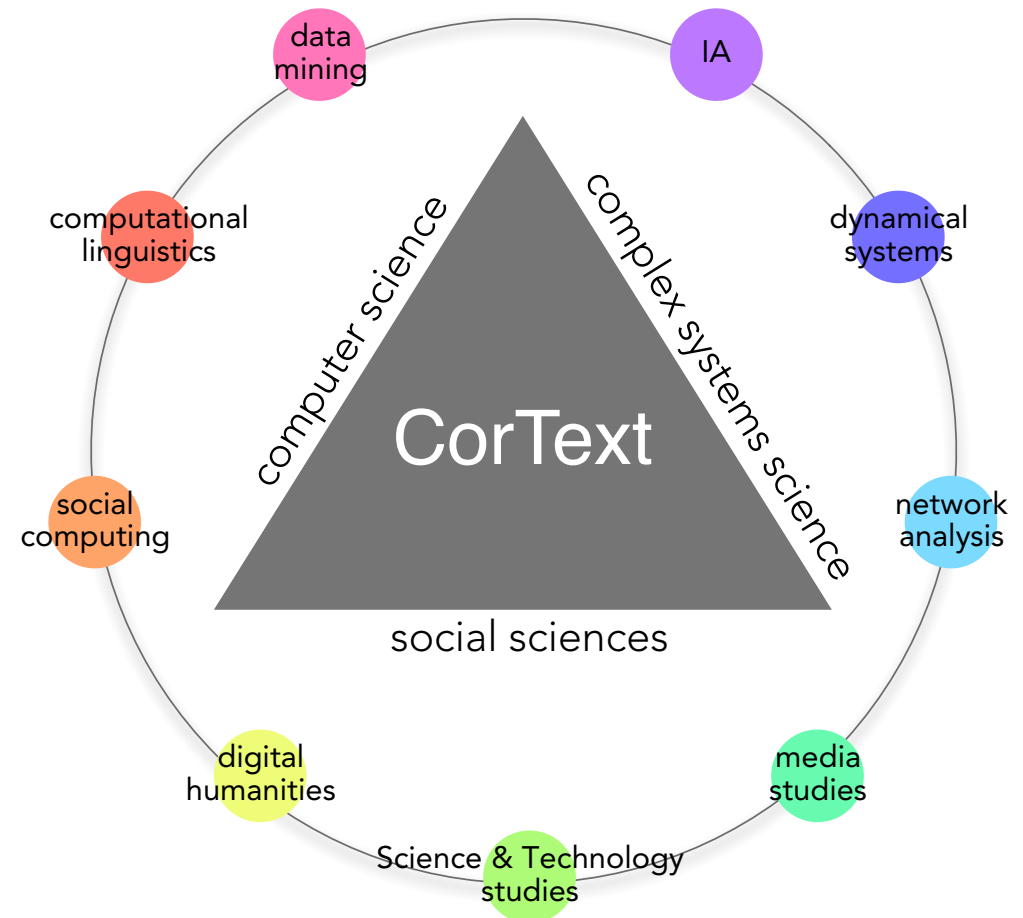


Modeling and mapping knowledge dynamics with CorText platform

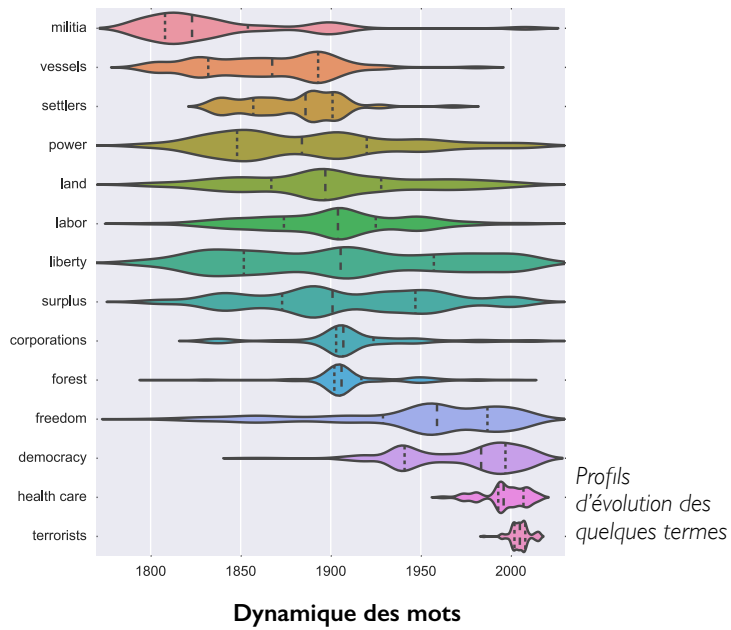
Outline

- A web based digital platform for social scientists conducting *empirical studies* in the fields of Media Studies, Science and Technology Studies, Digital Humanities, etc.
- CorText is an open online service for *heterogeneous data analysis, modeling and visualization*
- open registration:
<http://managerv2.cortext.net>

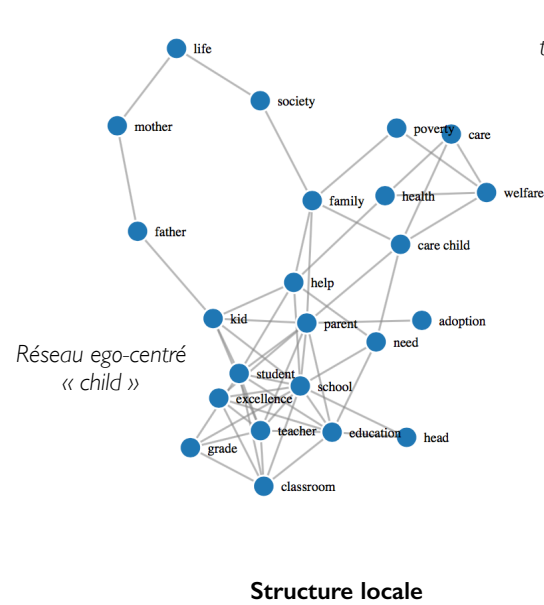


Multi-level perspective

Dynamics



Relationships



Extraction terminologique

Texts

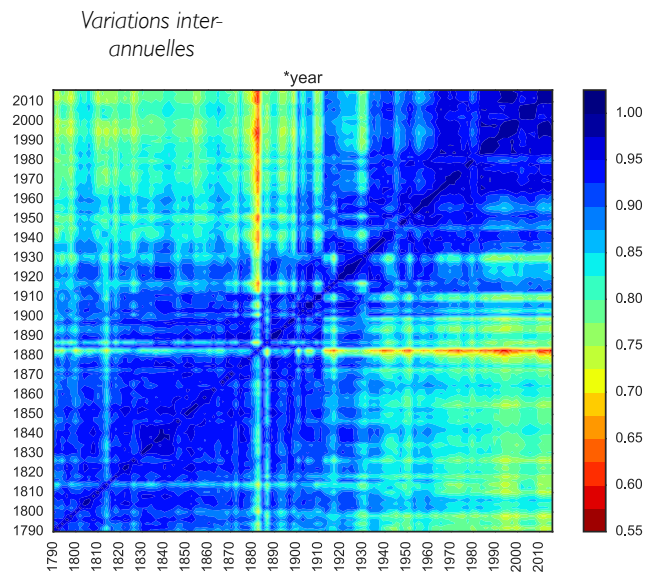
ECHELLE MICRO

label	forms
abundance	abundance
acres	acres lacre
aliens	aliens alien
armed forces	armed forces armed force force of arms arms and force
business men	business men business man men of business man of business businessman businessmen
children	children child
commerce and navigation	commerce and navigation navigation and commerce navigation or commerce
crews	crews crew
crime	crime crimes
crisis	crisis crises
crops	crops crop
cruisers	cruisers cruiser
democracy	democracy democracies
diplomatic relations	diplomatic relations
farm products	farm products products of the farm products of farm
great importance	great importance greater importance greatest importance
health care costs	health care costs cost of health care health care cost
income tax	income tax tax the income tax on the income
peace and freedom	peace and freedom freedom and peace peace with freedom
property rights	property rights property right rights of property right of property rights and property

Analyse textuelle

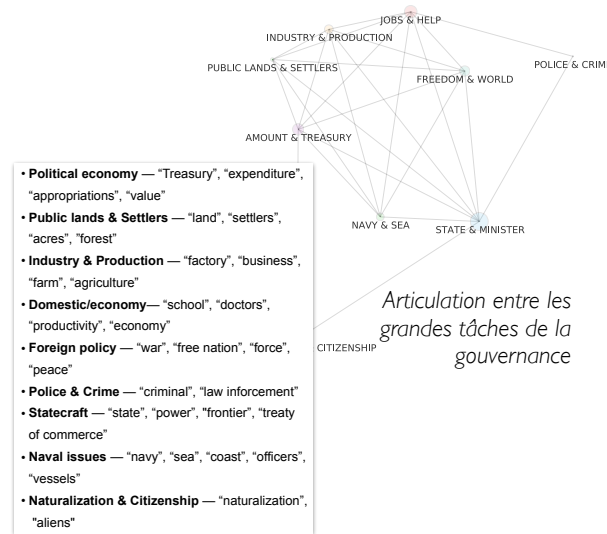
Multi-level perspective

Dynamics



Dynamique des discours

Relationships



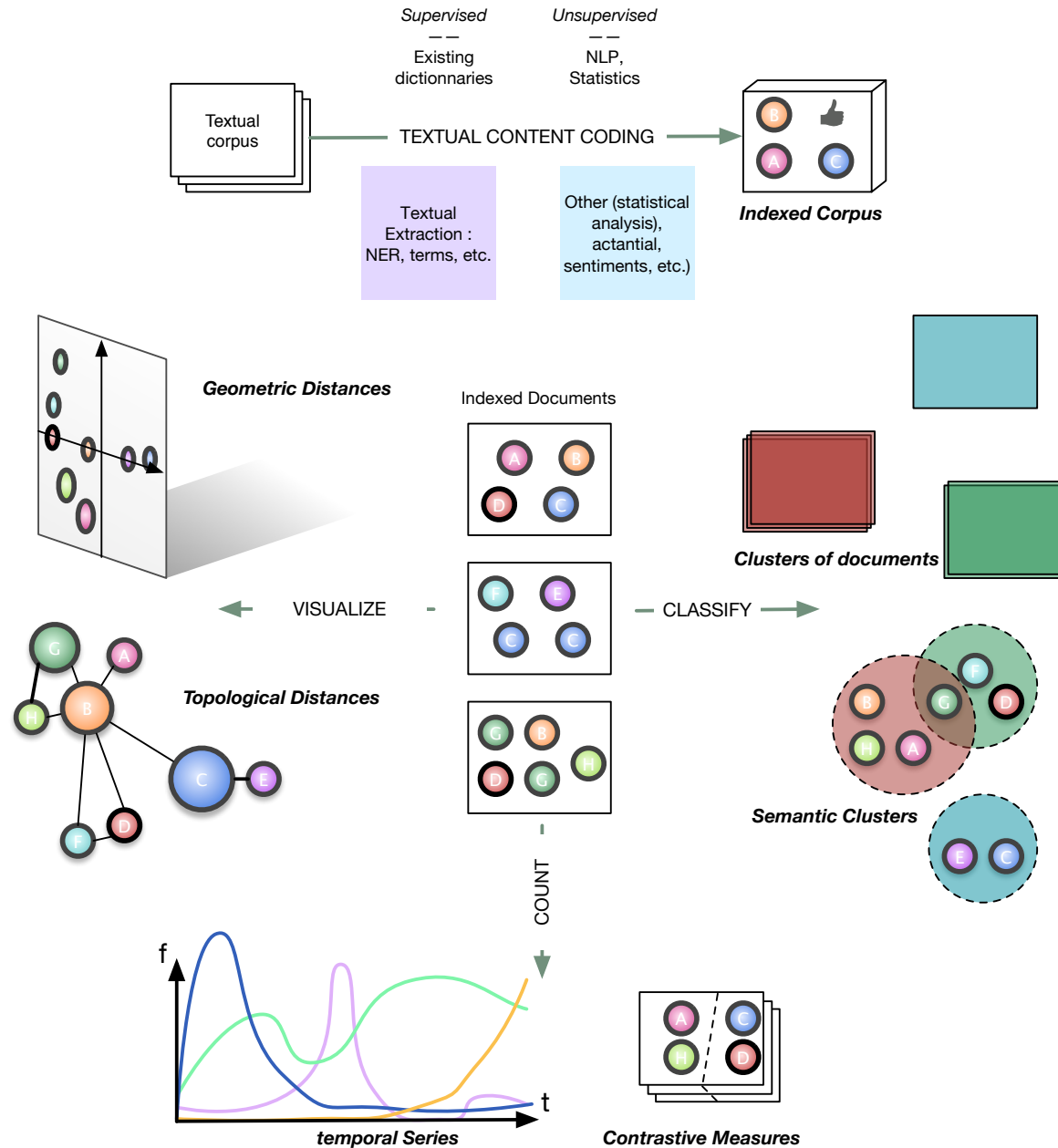
Structure macro des grands thèmes

Texts

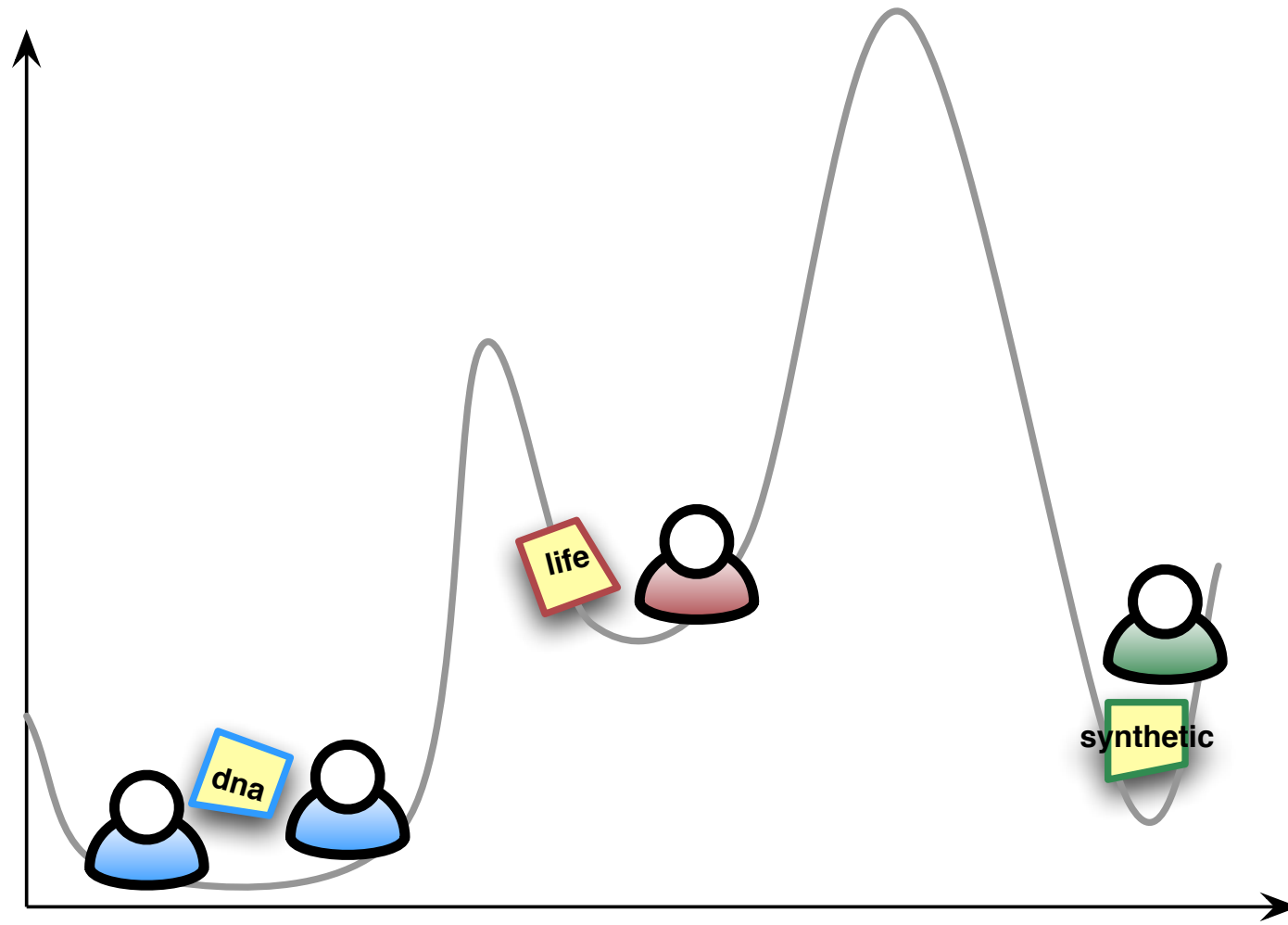


Corpus

Textual Corpus Processing

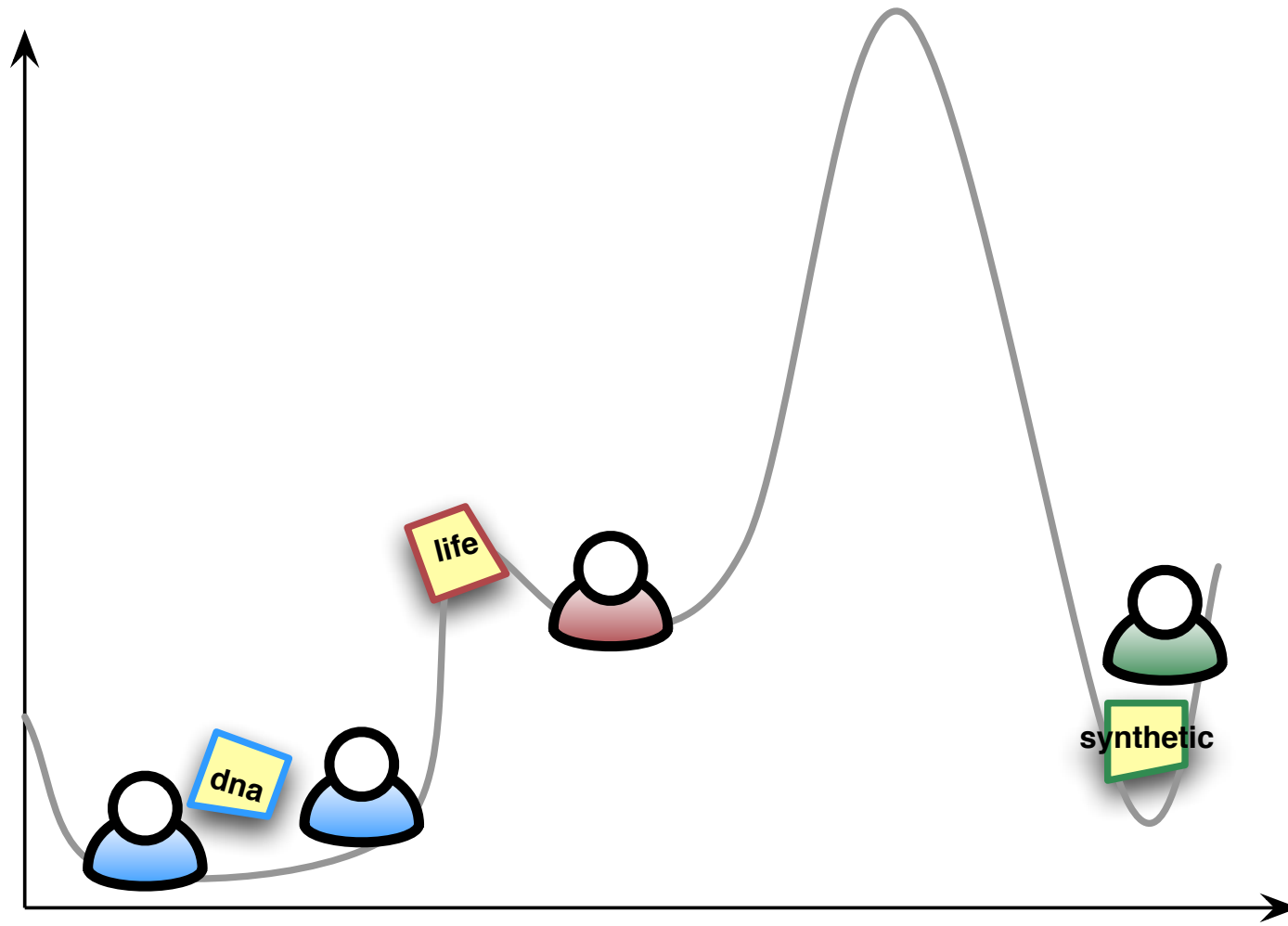


From traces to landscapes



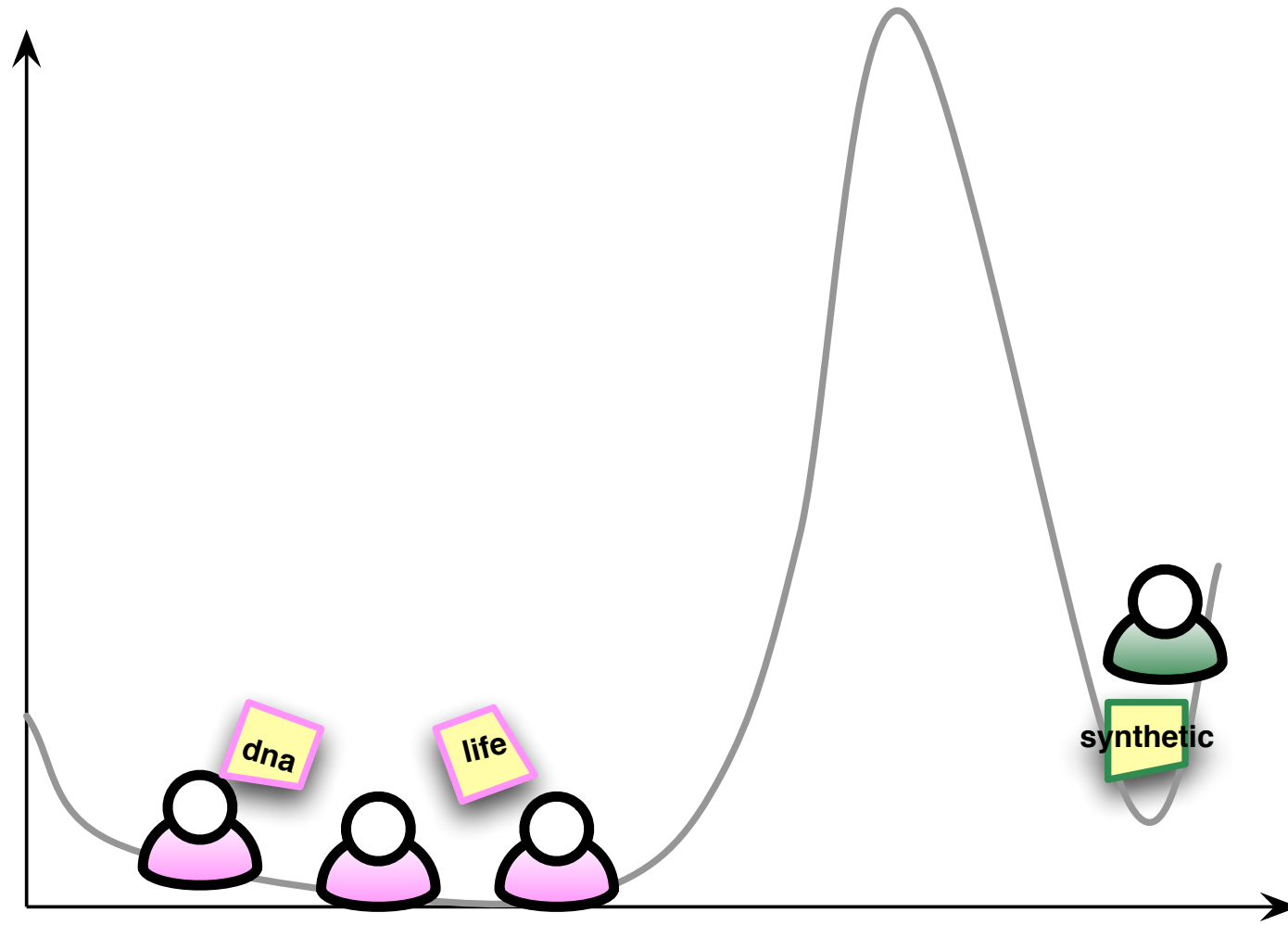
Landscape and its actors ($t=1$)

From traces to landscapes



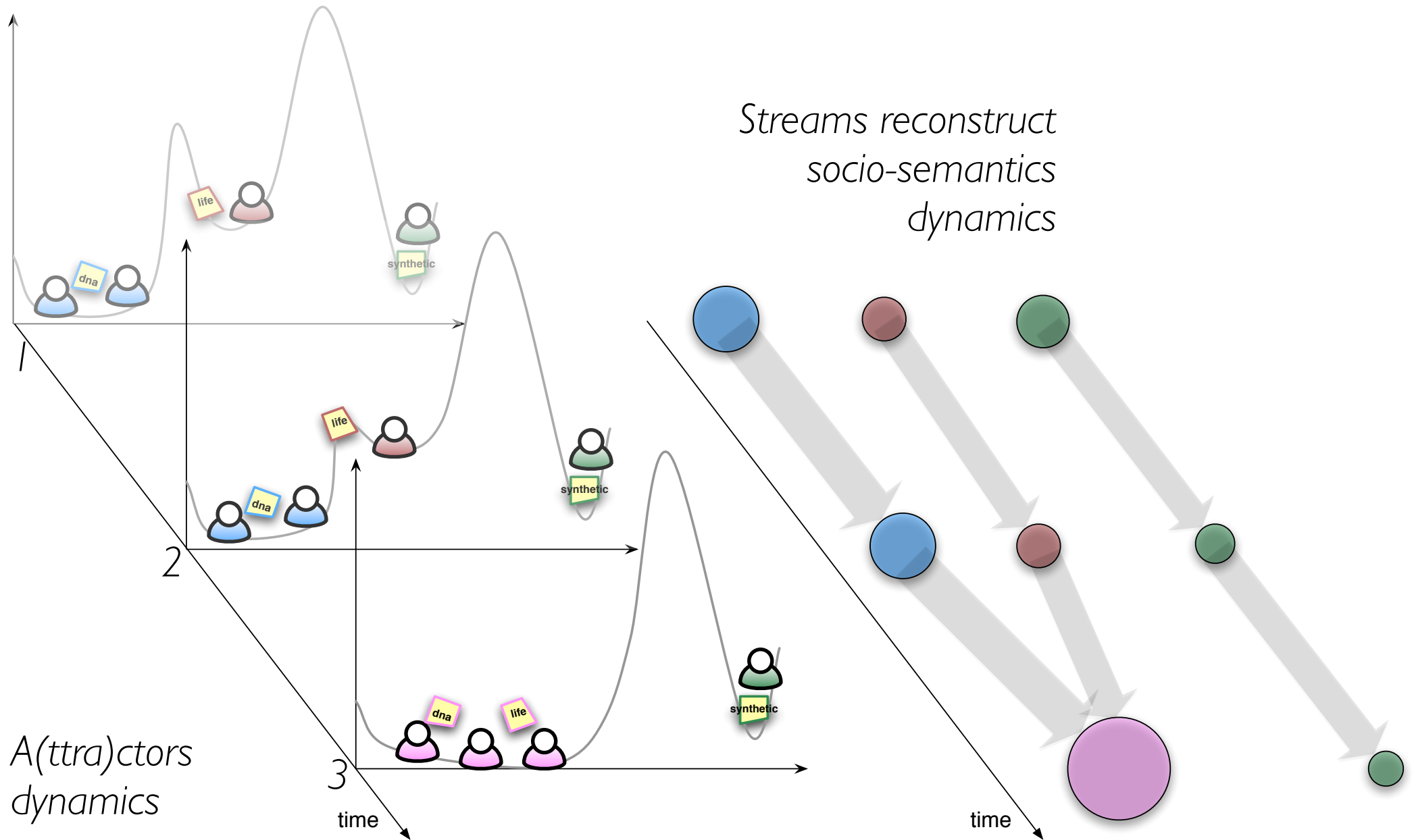
Landscape and its actors ($t=2$)

From traces to landscapes

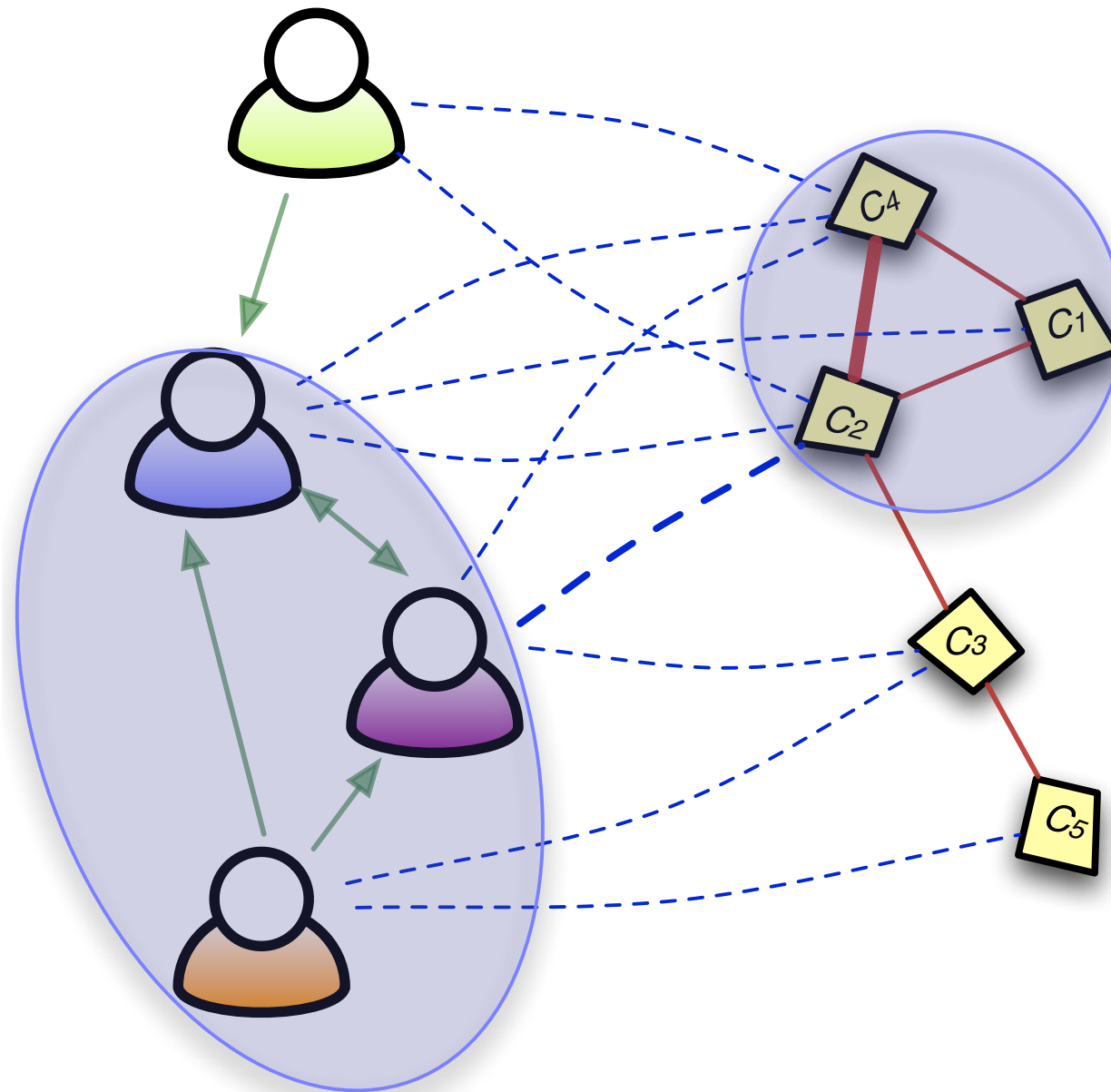


Landscape and its actors (t=3)

From traces to landscapes



Heterogeneous cooccurrences



- Generalized co-occurrences analysis framework mixing people, terms, countries, etc...

Understanding proximity metrics

It's all about context

“ Seule la proposition a un sens ; ce n'est que lié dans une proposition que le nom a une signification ”

Wittgenstein, "Tractatus logico-philosophicus" 1921, prop 3.3

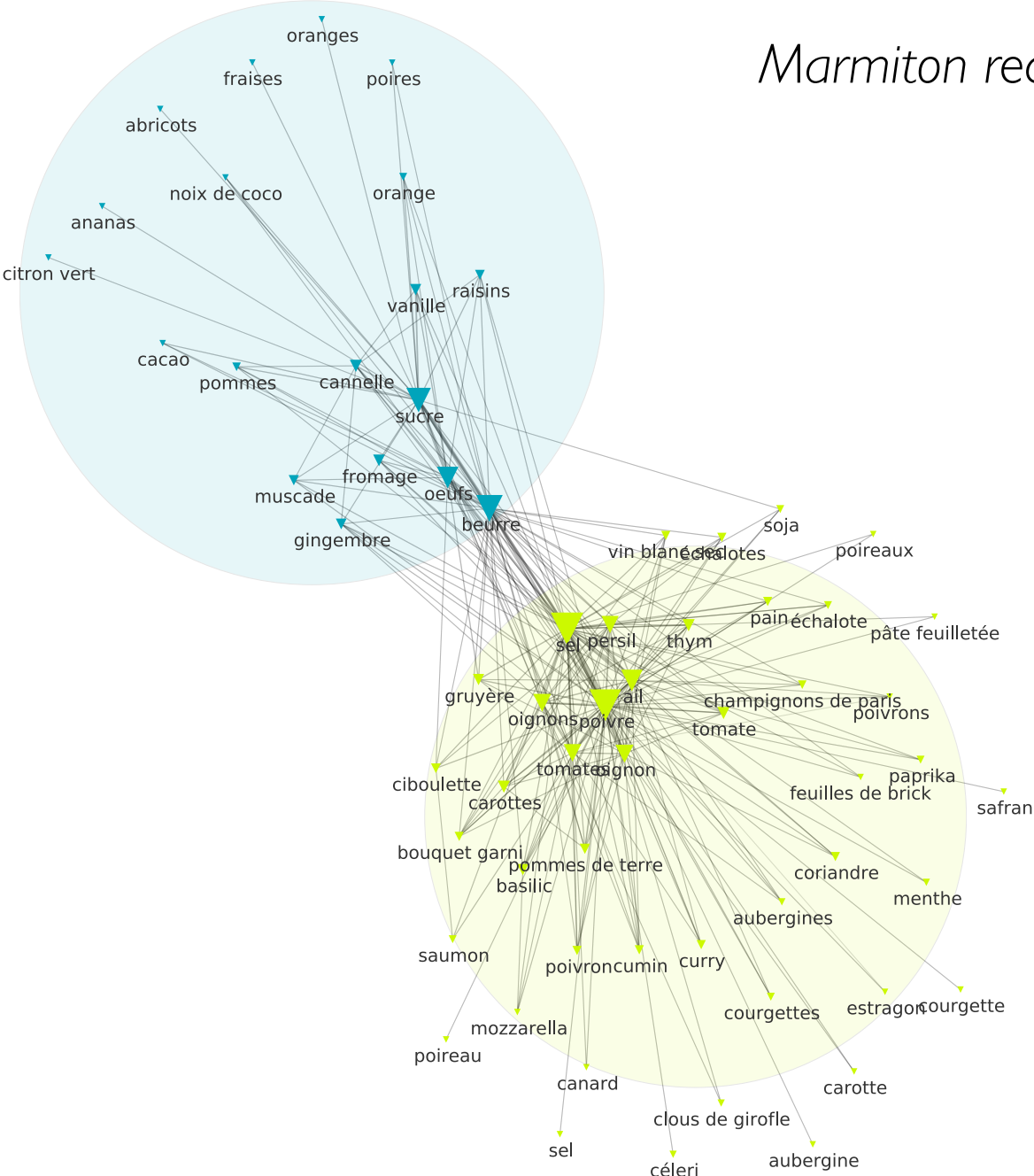
“ A word is characterized by the company it keeps ”

Firth, John R. "A synopsis of linguistic theory, 1930-1955."
Studies in linguistic analysis (1957) - 11.

Metrics

Marmiton recipes

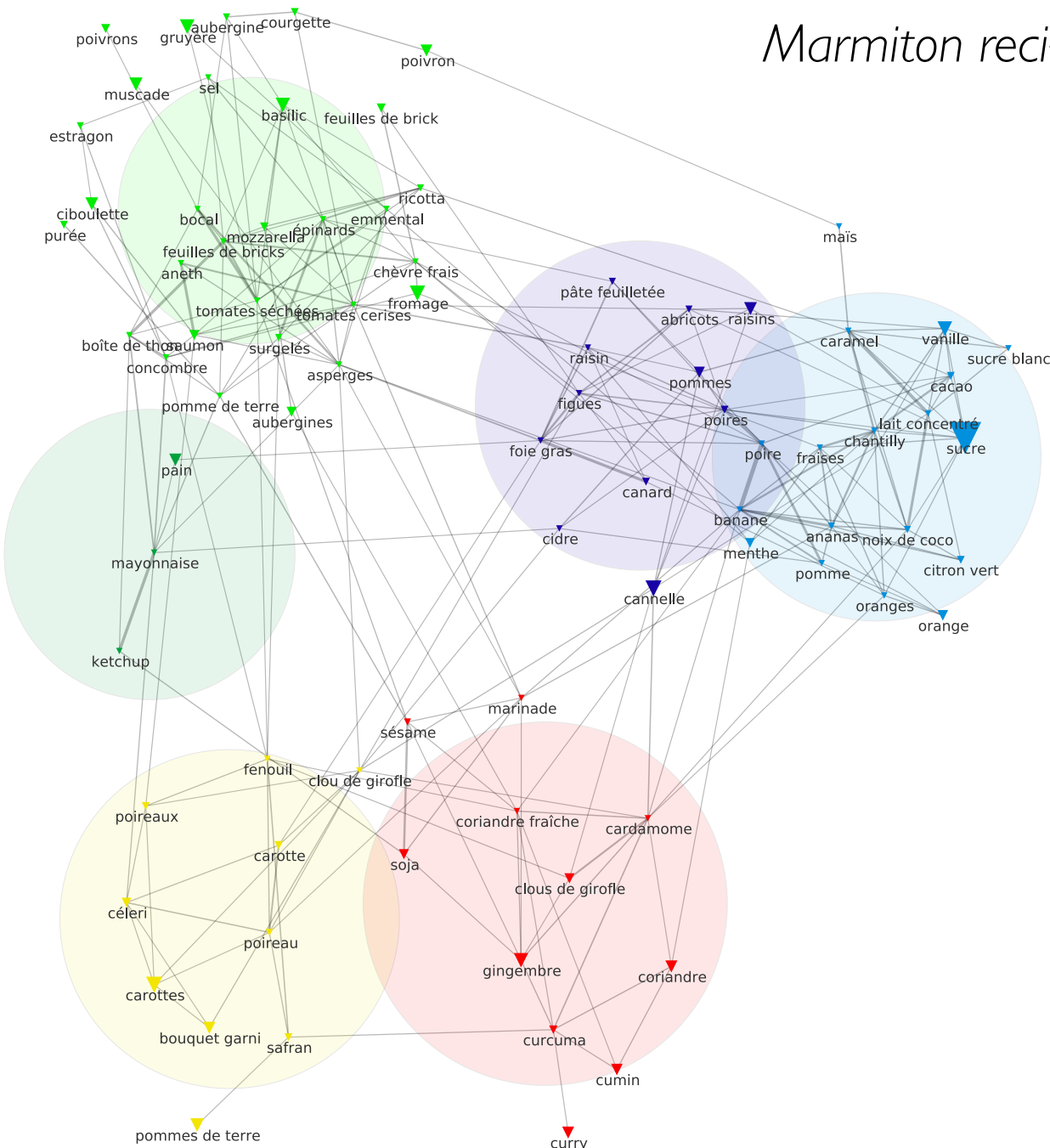
- Raw cooccurrence measure



Metrics

Marmiton recipes

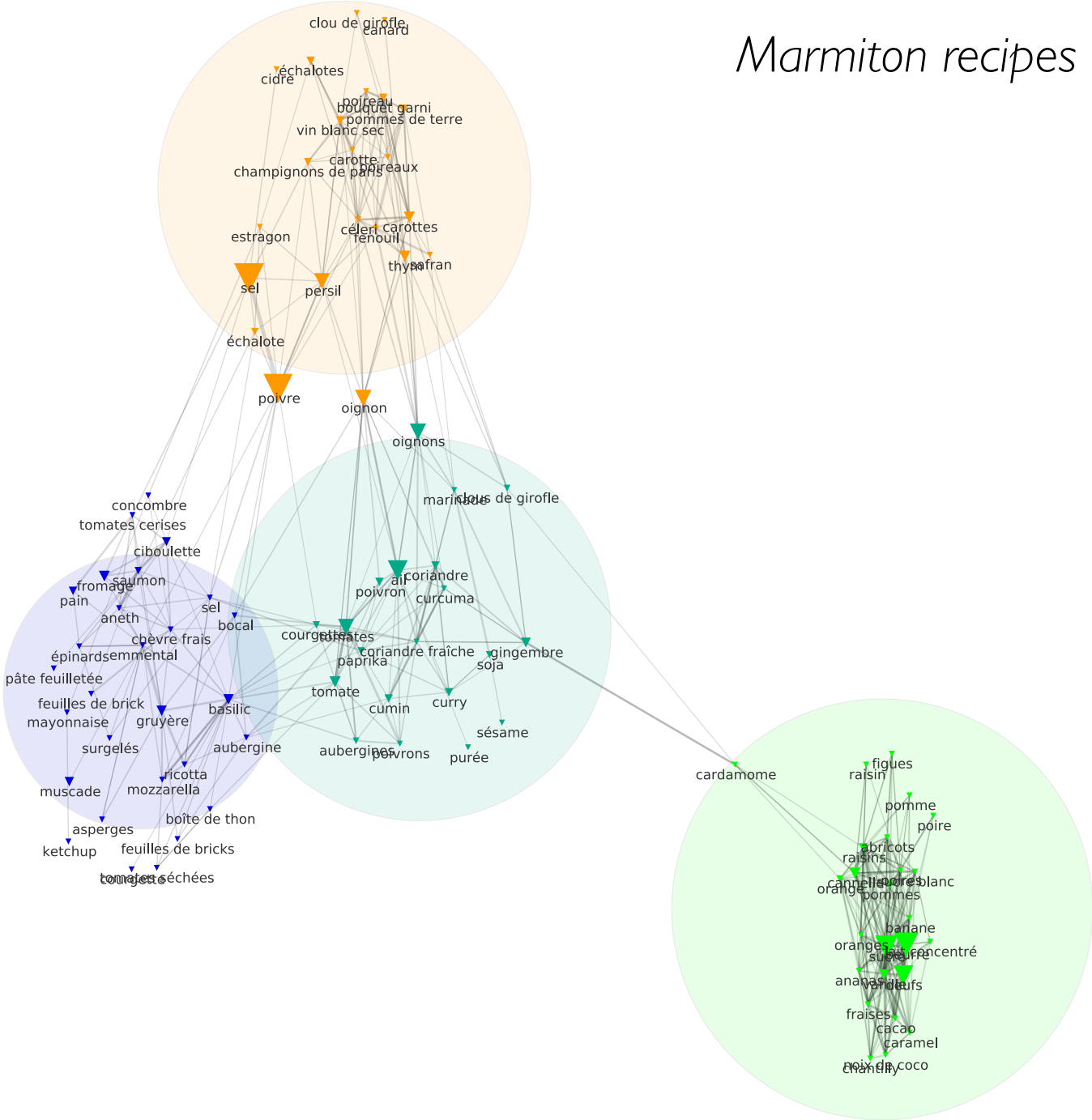
● Cramer



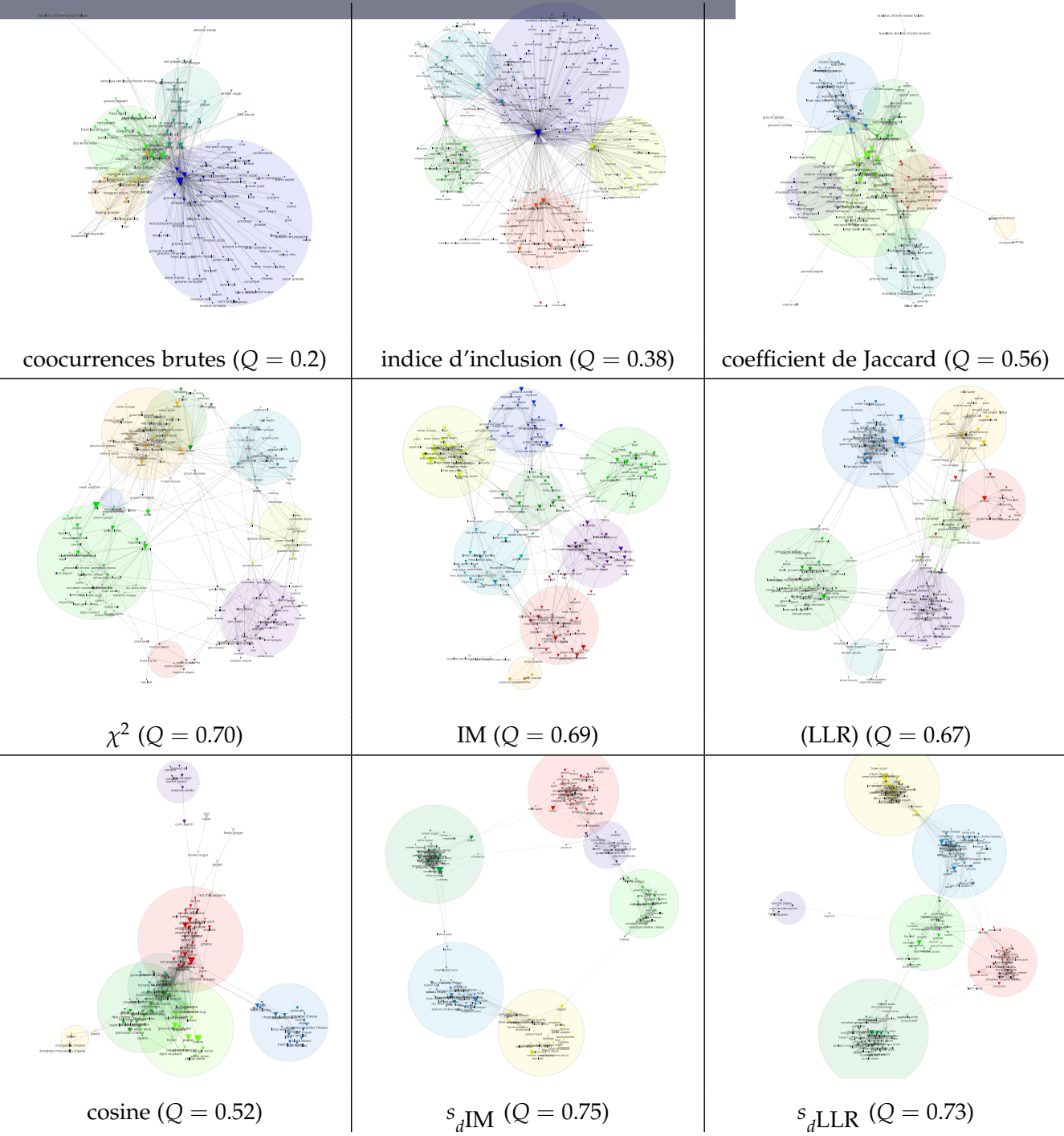
Metrics

Marmiton recipes

- Distributional Measure



Metrics

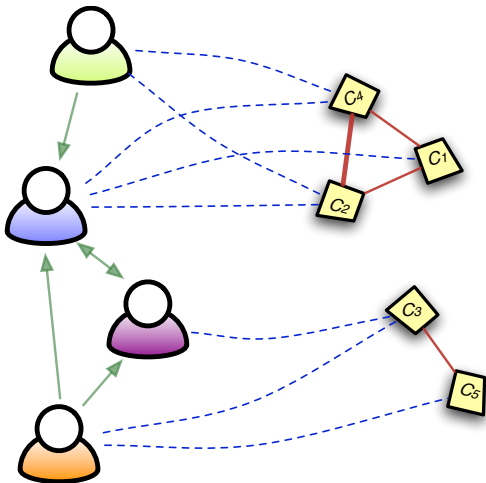


To summarize

Cooccurrences matrices

Cooccurrences matrices construction

- Occurrence matrix O : $O_{ij} = 1$ iff item i is used in document j , 0 otherwise
- The cooccurrence matrix enumerates every joint appearances of two items in the same document. $C = O^t O$

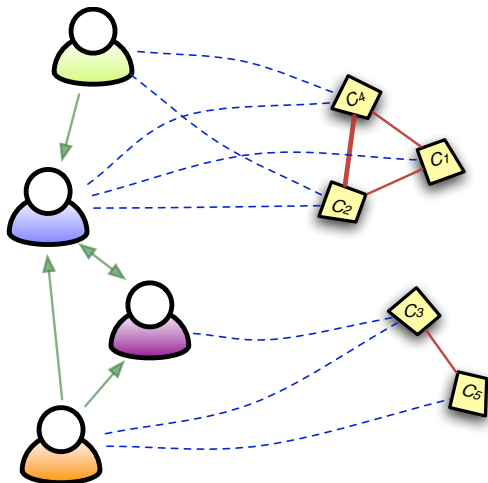


documents \ items	A	B	C	D
C1				
C2				
C3				
C4				
C5				

Cooccurrences matrices

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	C1	C2	C3	C4	C5
C1	x	1		1	
C2	1	x	1	2	
C3			x		1
C4	1	2		x	
C5			1		x

Proximity Measures

Cooccurrences variables

- *Co-Occurrence matrix C* :
 C_{ij} = number of joint occurrences of i and j in the same document
- total number of cooccurrences of i : $s_i = \sum_{j, j \neq i} C_{ij}$
- global number of co-occurrences : $N = \sum_i s_i$
- expected number of cooccurrences : $e_{ij} = \frac{s_i s_j}{N}$

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C1		1		1	
C2	1		1	2	
C3					1
C4	1	2			
C5			1		

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- expected number of cooccurrences : $e_{ij} = \frac{s_i s_j}{N}$

Direct Measures of Similarity :

- *Raw cooccurrences* : $S_R(i, j) = C_{ij}$
- *Association strength* : $S_A(i, j) = \frac{C_{ij}}{s_i s_j}$
- *Mutual Information* : $S_{MI}(i, j) = \log\left(\frac{C_{ij}}{e_{ij}}\right)$
- *Cosine* : $S_C(i, j) = \frac{C_{ij}}{\sqrt{s_i s_j}}$
- *Inclusion index* : $S_I(i, j) = \frac{C_{ij}}{\min(s_i, s_j)}$
- *Jaccard index* : $S_J(i, j) = \frac{C_{ij}}{s_i + s_j - C_{ij}}$
- χ^2 score : $S_{\chi^2}(i, j) = \frac{C_{ij} - e_{ij}}{\sqrt{e_{ij}}}$
- *Cramer index* : $S_{\text{cramer}}(i, j) = \frac{C_{ij} - e_{ij}}{e_{ij}}$

Proximity Measures

Cooccurrences variables

- *Co-Occurrence matrix C* :
 C_{ij} = number of joint occurrences of i and j in the same document
- total number of cooccurrences of i : $s_i = \sum_{j, j \neq i} C_{ij}$
- global number of co-occurrences : $N = \sum_i s_i$
- expected number of cooccurrences : $e_{ij} = \frac{s_i s_j}{N}$

Indirect Measures of Similarity :

- Cosine (distributional) :

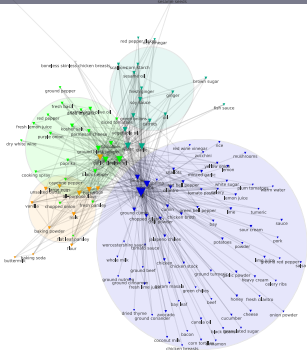
$$S_{Cd}(i, j) = \frac{\sum_{k \neq i, j} C_{ik} C_{jk}}{\sqrt{\sum_{k \neq i, j} C_{ik}^2 \sum_{k \neq i, j} C_{jk}^2}}$$

- Mutual Information (distributional) :

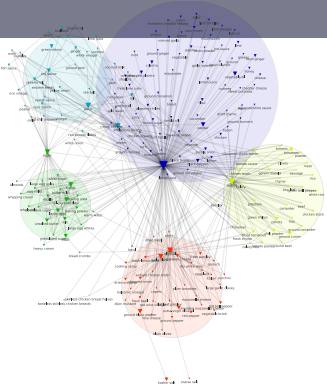
$$S_{MId} = \frac{\sum_{k \neq i, j; MI_{ik} > 0} \min(MI_{ik}, MI_{jk})}{\sum_{k \neq i, j; MI_{ik} > 0} MI_{ik}}$$

Metrics Summary

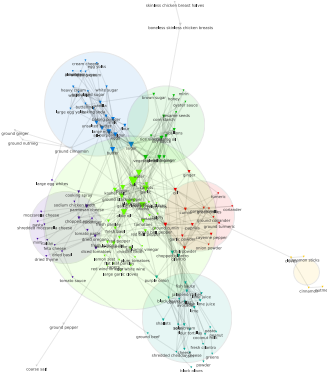
Co-word



cocurrences brutes ($Q = 0.2$)

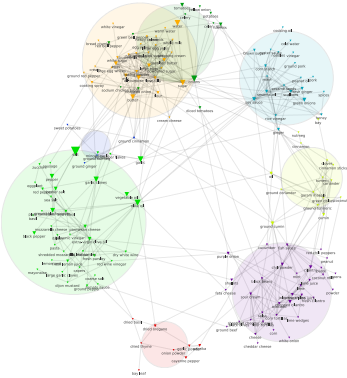


indice d'inclusion ($Q = 0.38$)

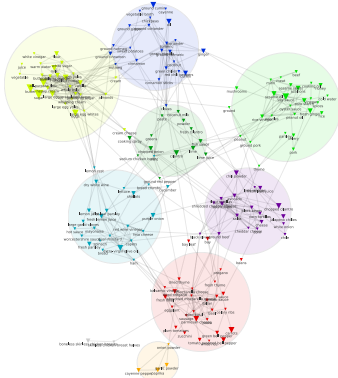


coefficient de Jaccard ($Q = 0.56$)

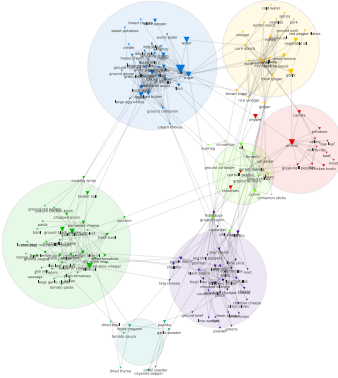
Direct (syntagmatic)



χ^2 ($Q = 0.70$)

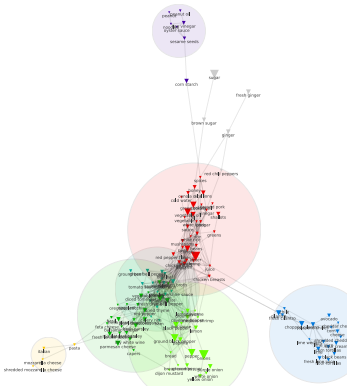


IM ($Q = 0.69$)

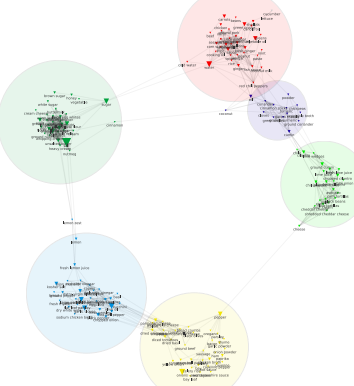


(LLR) ($Q = 0.67$)

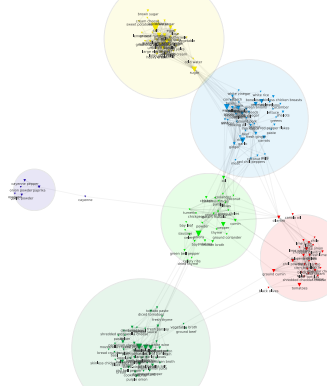
Indirect (paradigmatic)



cosine ($Q = 0.52$)



s_dIM ($Q = 0.75$)



s_dLLR ($Q = 0.73$)

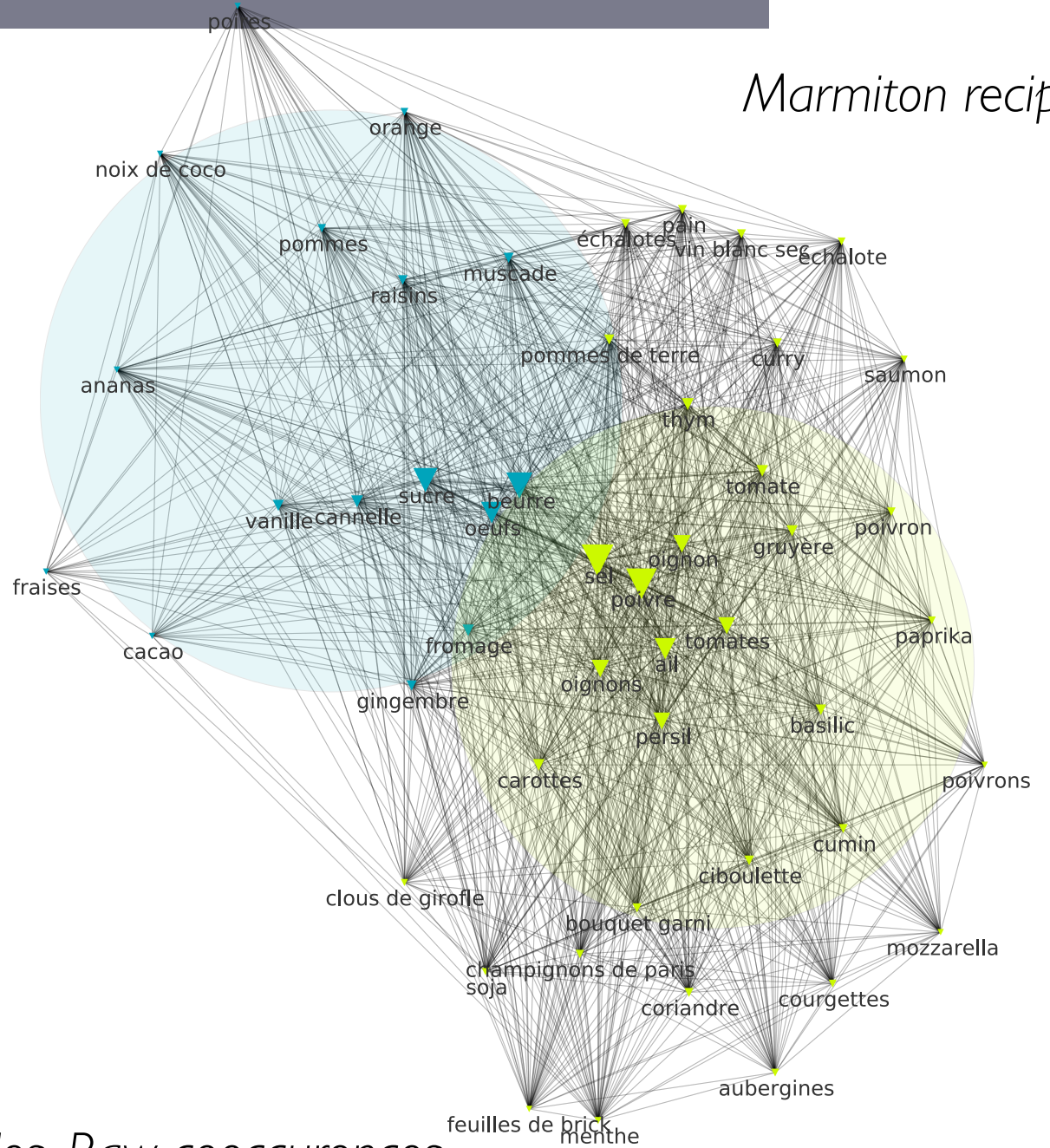
Metrics Summary

proximity measures	type of network	normalisation	special properties
raw	interaction network (<i>e.g.</i> social network)	no	-
χ^2	all	yes	normalization tend to create links toward higher degree nodes
MI	all	yes	Inspired from information theory
Cramer	all	yes	-
cosine	homogeneous network (eg. semantic)	yes	Classical measure (originating from scientometrics)
distributional	homogeneous network (eg. semantic)	yes	very robust measure (coming from computational linguistics)
cosine_het	affiliation network (eg. users sharing the same hashtags)	yes	two fields are required but the final network is homogeneous
dot_product_het	affiliation network (eg. users sharing the same hashtags)	no	two fields are required but the final network is homogeneous

Filtering

Marmiton recipes

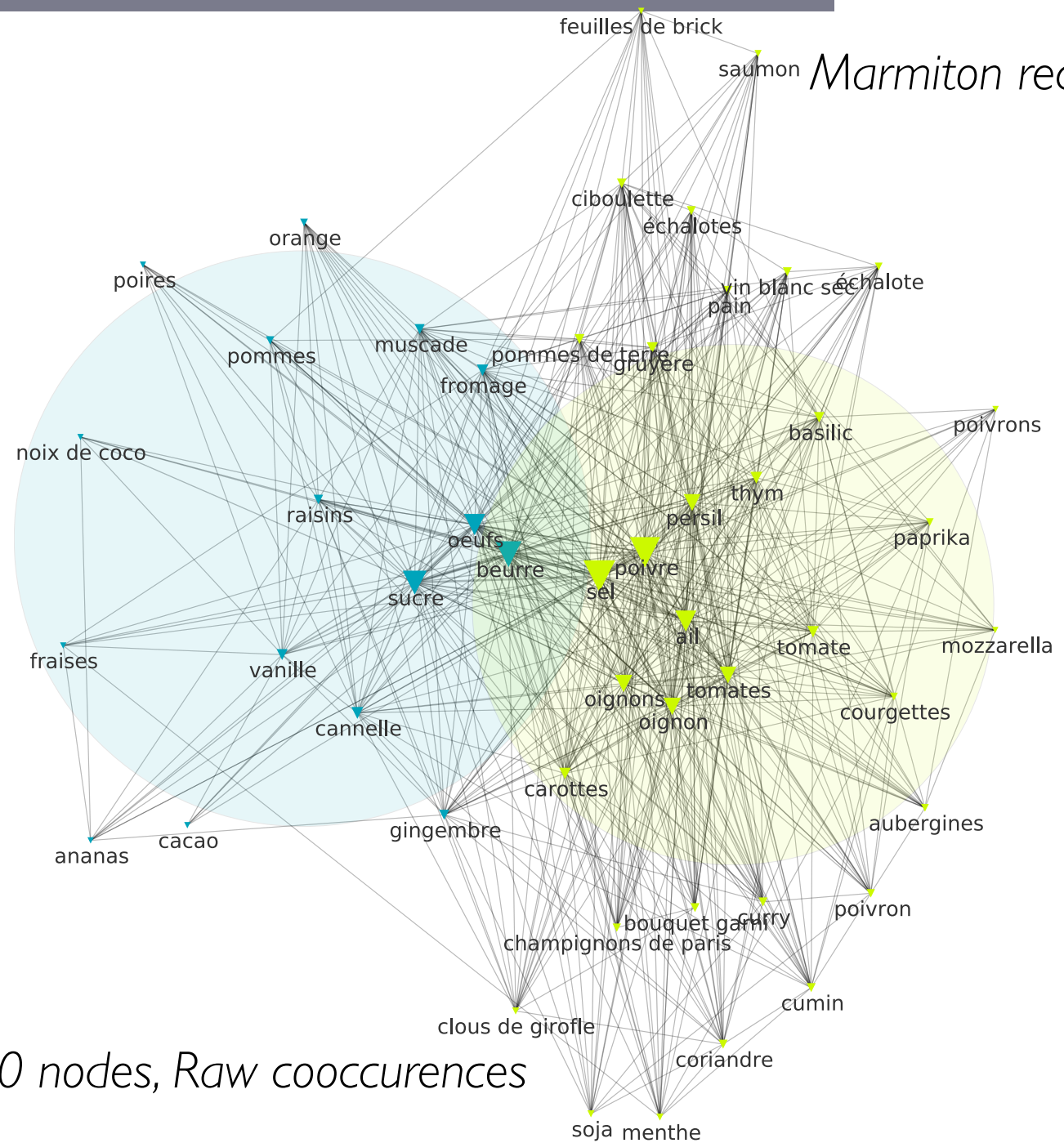
- No filtering



50 nodes, Raw cooccurrences

Filtering

- filtering on edge strength values: 10 cooc at least

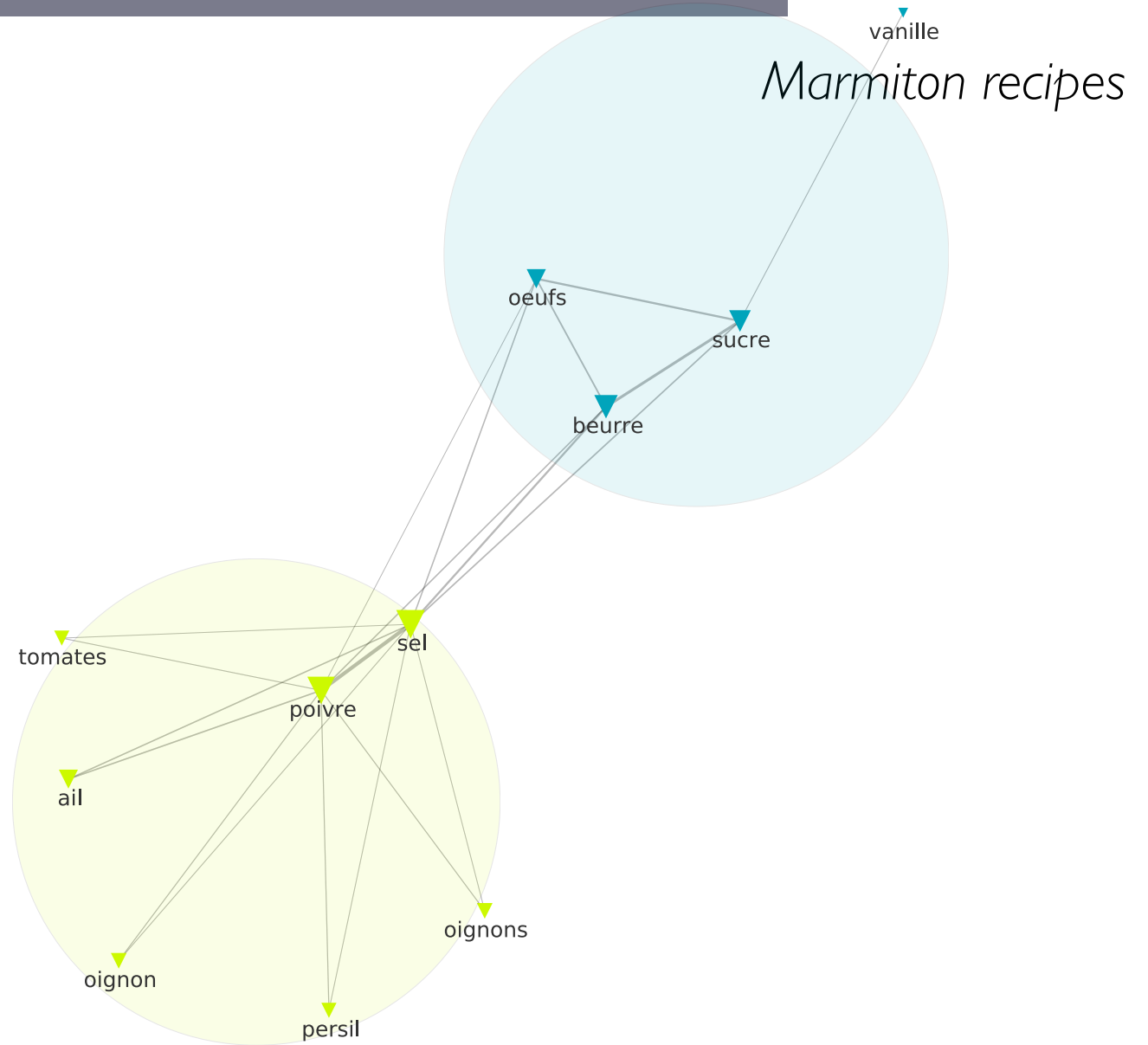


Marmiton recipes

50 nodes, Raw cooccurrences

Filtering

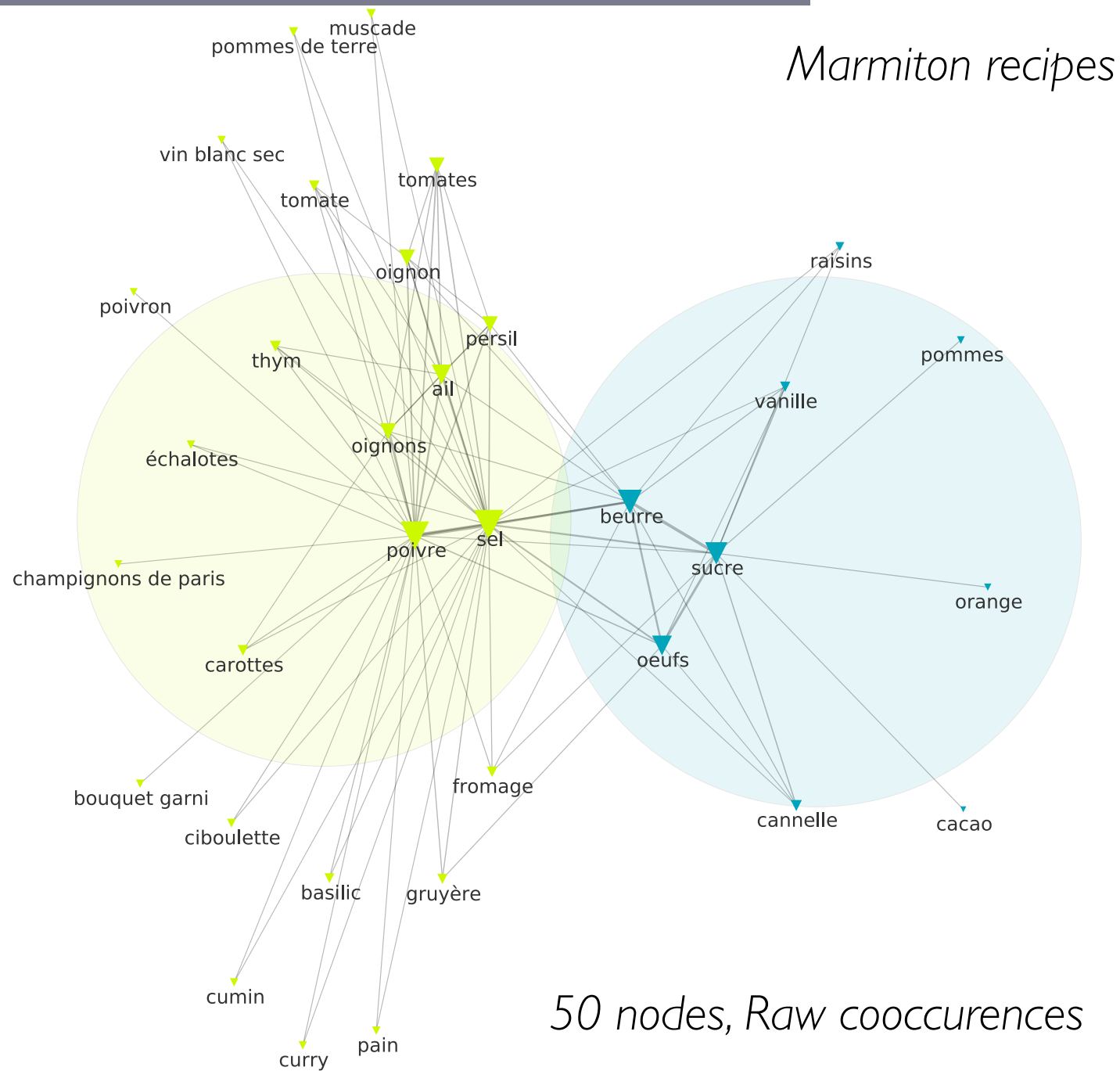
- filtering on the number of edges: 20 Top edges



50 nodes, Raw cooccurrences

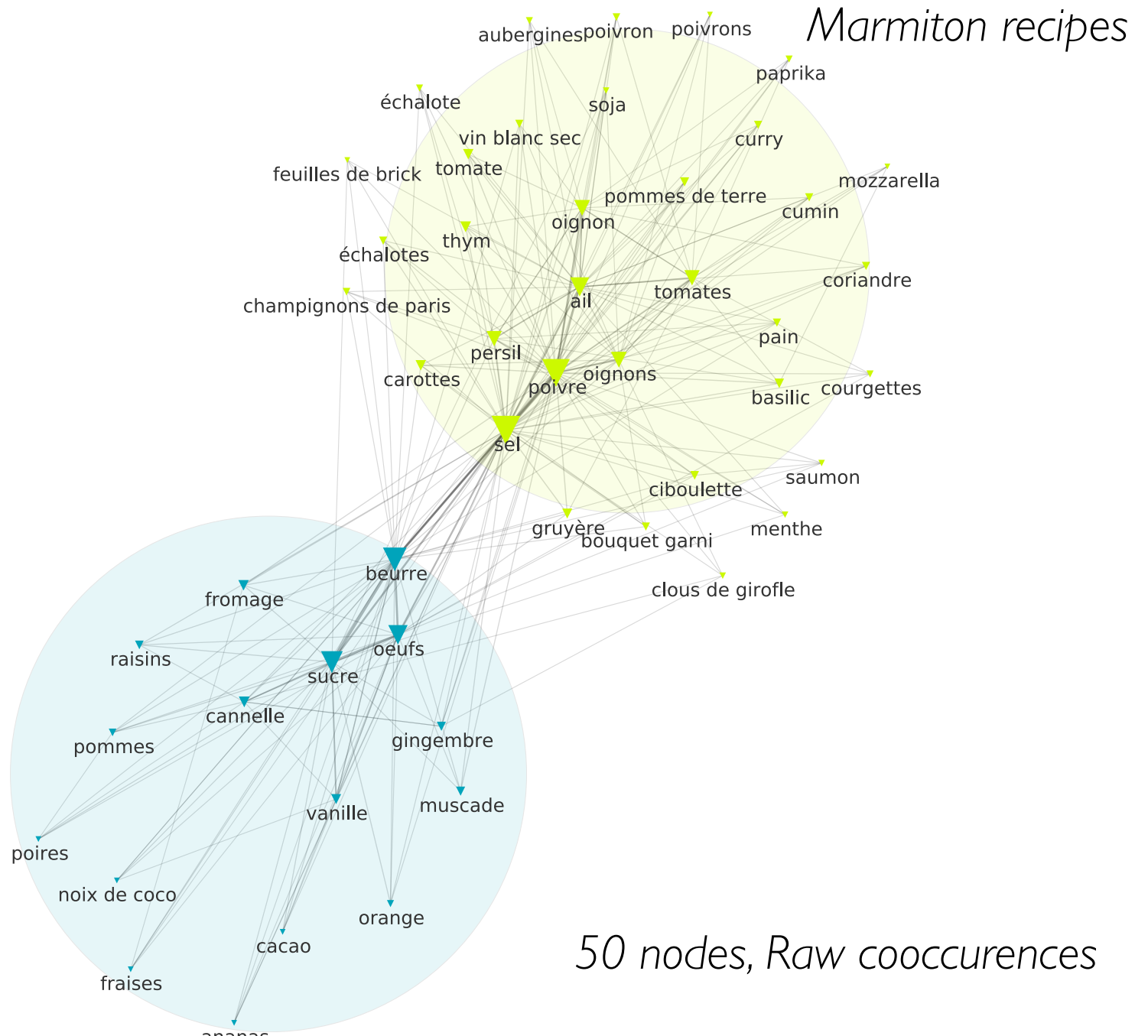
Filtering

- Auto-threshold (the edge strength threshold is automatically inferred from graph topology)

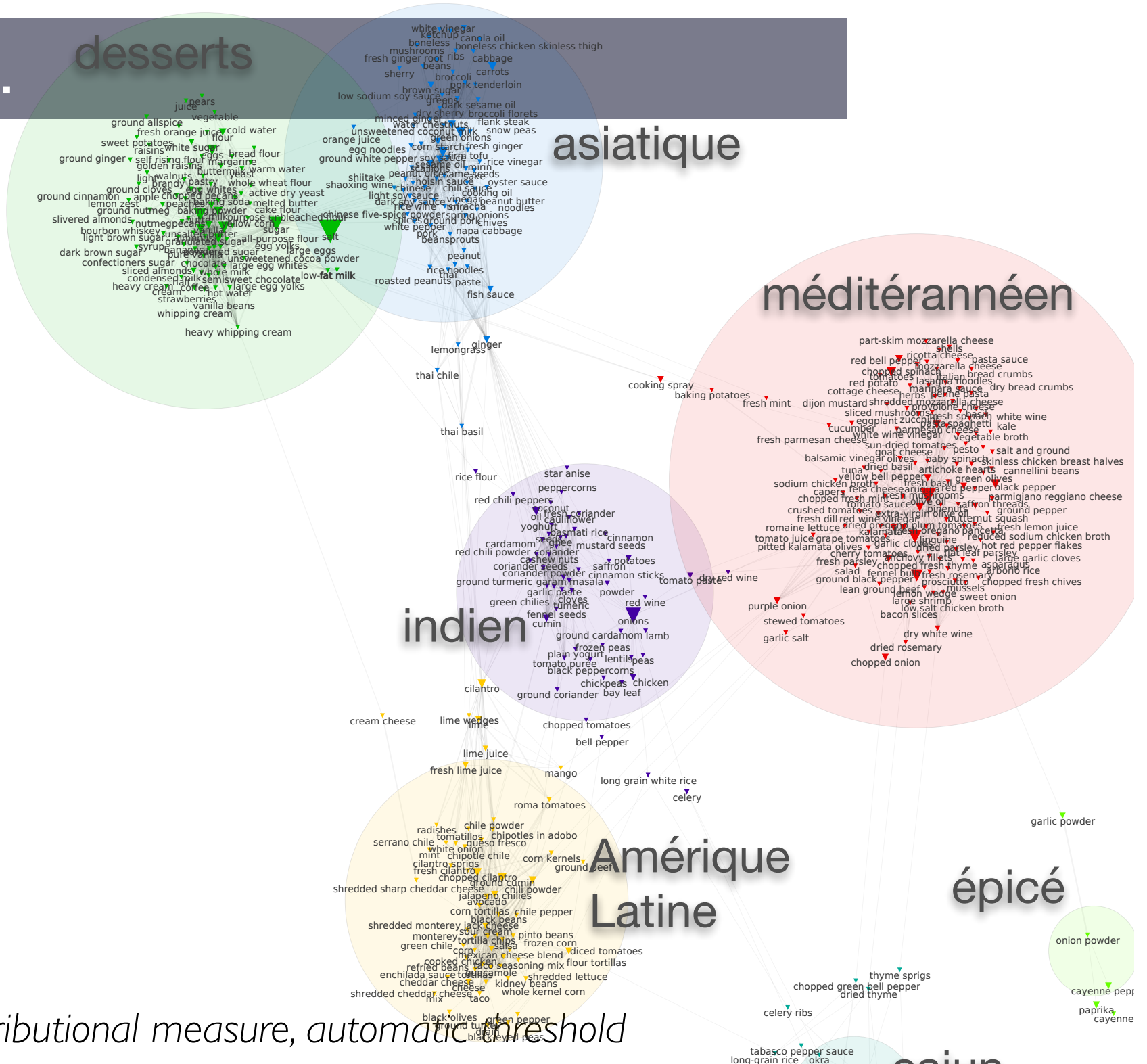


Filtering

- Local Filtering: 5 top neighbours



Recipes...

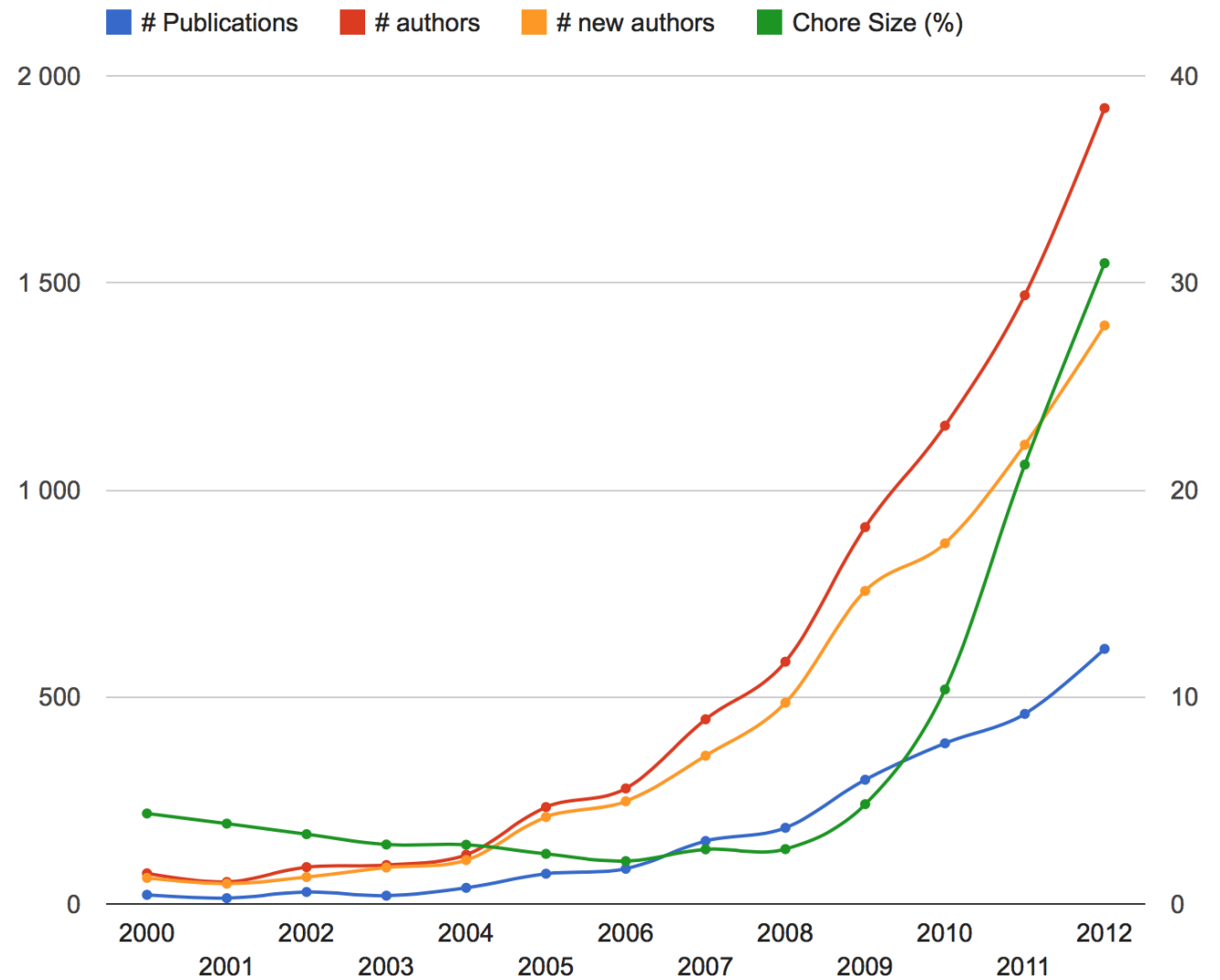
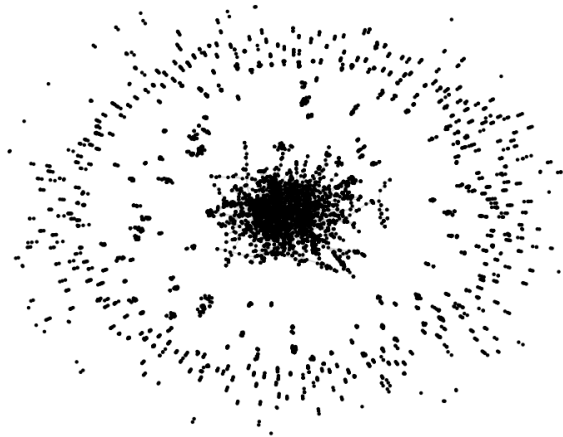


500 nodes, Distributional measure, automatic threshold

Synthetic Biology Emergence

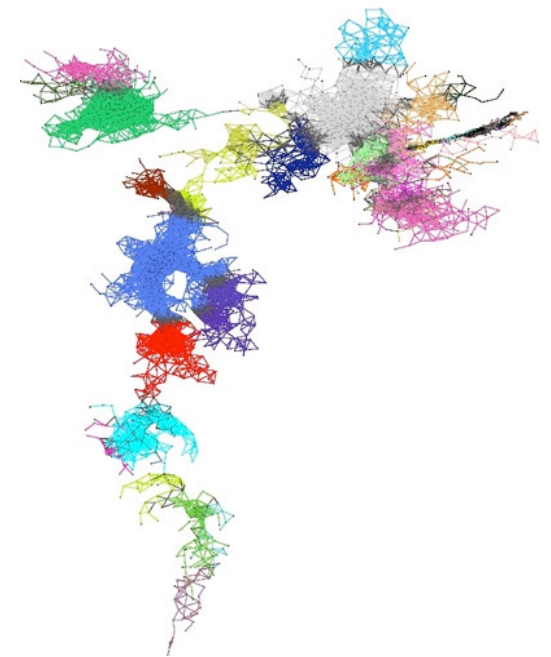
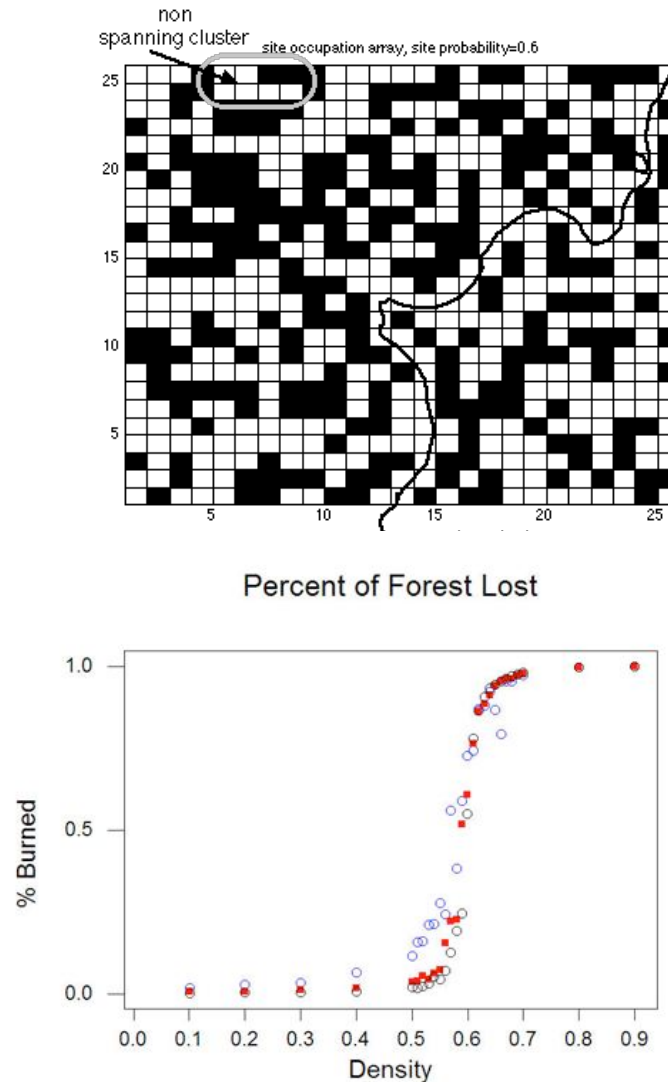
Collaboration Network Dynamics

- Unstabilized field at the crossroads between different disciplinary origins
- Exponential growth of publications with a very high rate of newcomers
- Although a central community is emerging



Collaboration Network Dynamics

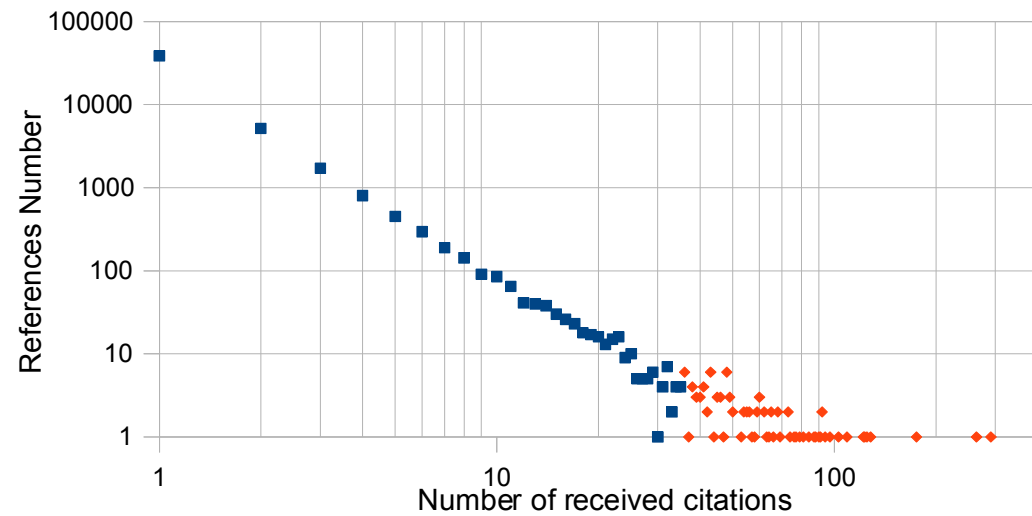
- Unstabilized field at the crossroads between different disciplinary origins
- Exponential growth of publications with a very high rate of newcomers
- Although a central community is emerging



«percolation threshold» above which the giant connected component collapse in disconnected subgraphs

Scientific references co-citation map

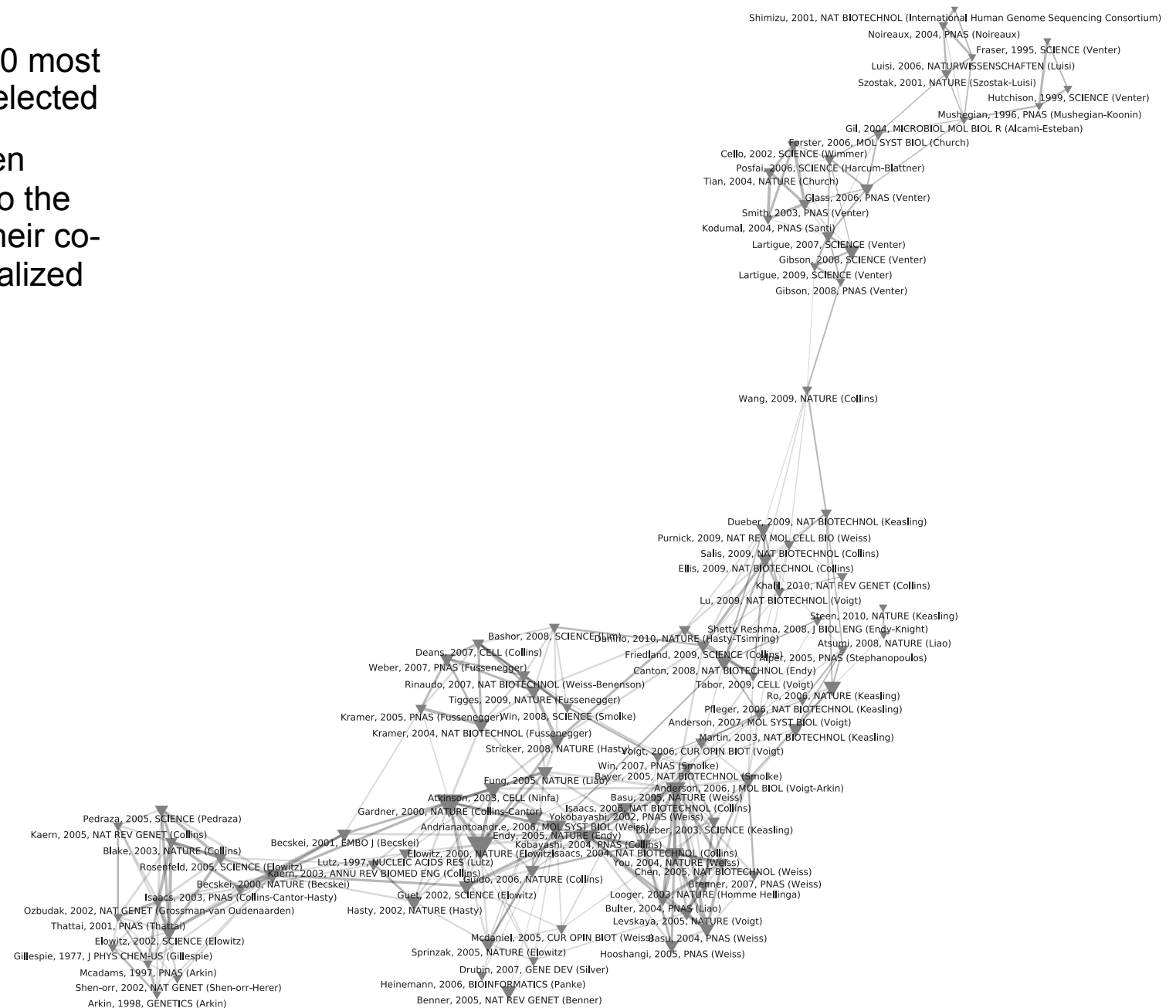
- **Selection:** only the 100 most cited references are selected



Received citations distributions

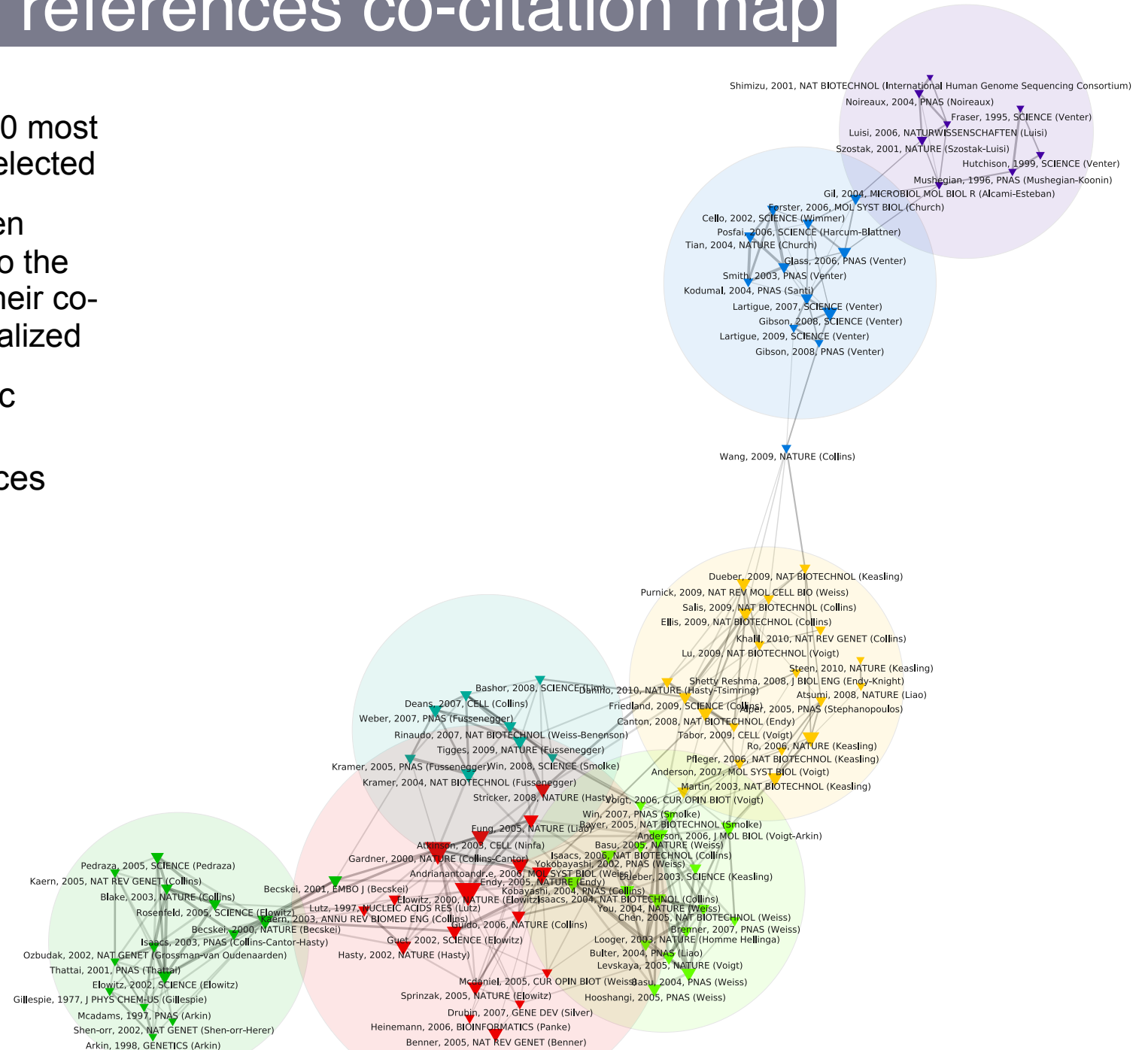
Scientific references co-citation map

- **Selection:** only the 100 most cited references are selected
- **Network:** They are then connected according to the relative frequency of their co-occurrences and spatialized



Scientific references co-citation map

- **Selection:** only the 100 most cited references are selected
- **Network:** They are then connected according to the relative frequency of their co-occurrences and spatialized
- **Clustering:** Automatic detection of cohesive subgroups of references



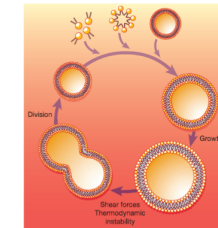
Scientific references co-citation map

- **Selection:** only the 100 most cited references are selected
- **Network:** They are then connected according to the relative frequency of their co-occurrences and spatialized
- **Clustering:** Automatic detection of cohesive subgroups of references
- **Interpretation:** intellectual schools, major actors

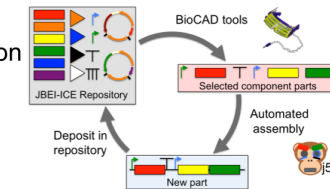
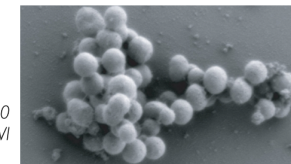
Keasling
Fussenegger
Endy
Venter
Weiss
Elowitz
Voigt
Collins

Minimal Cell/Sequence

Synthetic Genome



Réplication de vésicules
Synthesizing Life, Szostak 2001



Source: Site du Joint Bioenergy Institute

Stochasticity/Cell

Pedraza, 2005, SCIENCE
Kaern, 2005, NAT REV GENET (Collins)
Blake, 2003, NATURE
Rosenfeld, 2002, SCIENCE (Elowitz)
Isaacs, 2003, PNAS (Collins-Cantor-Hasty)
Ozbudak, 2002, NAT GENET (Grossman-van Oudenaarden)
Thattai, 2001, PNAS (Thattai)
Elowitz, 2002, SCIENCE (Elowitz)
Gillespie, 1977, J PHYS CHEM-US (Gillespie)
McAdams, 1997, PNAS (Arkin)
Shen-orr, 2002, NAT GENET (Shen-orr-Herer)
Arkin, 1998, GENETICS (Arkin)

Guy, 2002, SCIENCE (Elowitz)
Hasty, 2002, NATURE (Hasty)
Sprinzak, 2005, NATURE (Elowitz)
Drubin, 2007, GENE DEV (Silver)
Heinemann, 2006, BIOINFORMATICS (Panke)
Benner, 2005, NAT REV GENET (Benner)

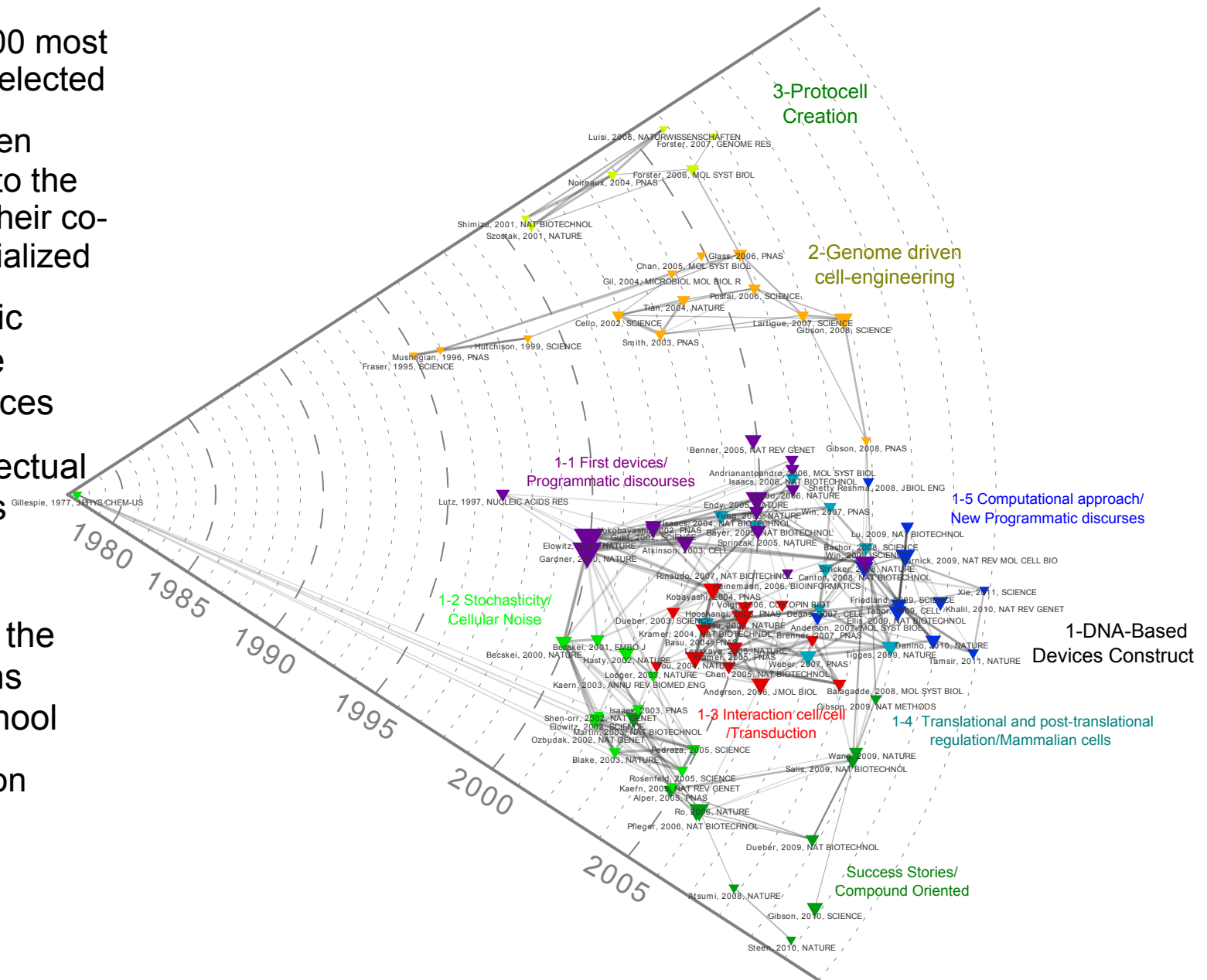
Shimizu, 2001, NAT BIOTECHNOL (International Human Genome Sequencing Consortium)
Noireaux, 2004, PNAS (Noireaux)
Fraser, 1995, SCIENCE (Venter)
Luisi, 2006, NATURWISSENSCHAFTEN (Luisi)
Szostak, 2001, NATURE (Szostak-Luisi)
Hutchison, 1999, SCIENCE (Venter)
Mushegian, 1996, PNAS (Mushegian-Koonin)
Gil, 2004, MICROBIOL MOL BIOL R (Alcami-Esteban)
Torster, 2006, MOL SYST BIOL (Church)
Cello, 2002, SCIENCE (Wimmer)
Posfai, 2006, SCIENCE (Harcum-Blattner)
Tian, 2004, NATURE (Church)
Glass, 2006, PNAS (Venter)
Smith, 2003, PNAS (Venter)
Kodumal, 2004, PNAS (Santini)
Lartigue, 2007, SCIENCE (Venter)
Gibson, 2008, SCIENCE (Venter)
Lartigue, 2009, SCIENCE (Venter)
Gibson, 2008, PNAS (Venter)

Control

Design/Programatic Discourse

Scientific references co-citation map

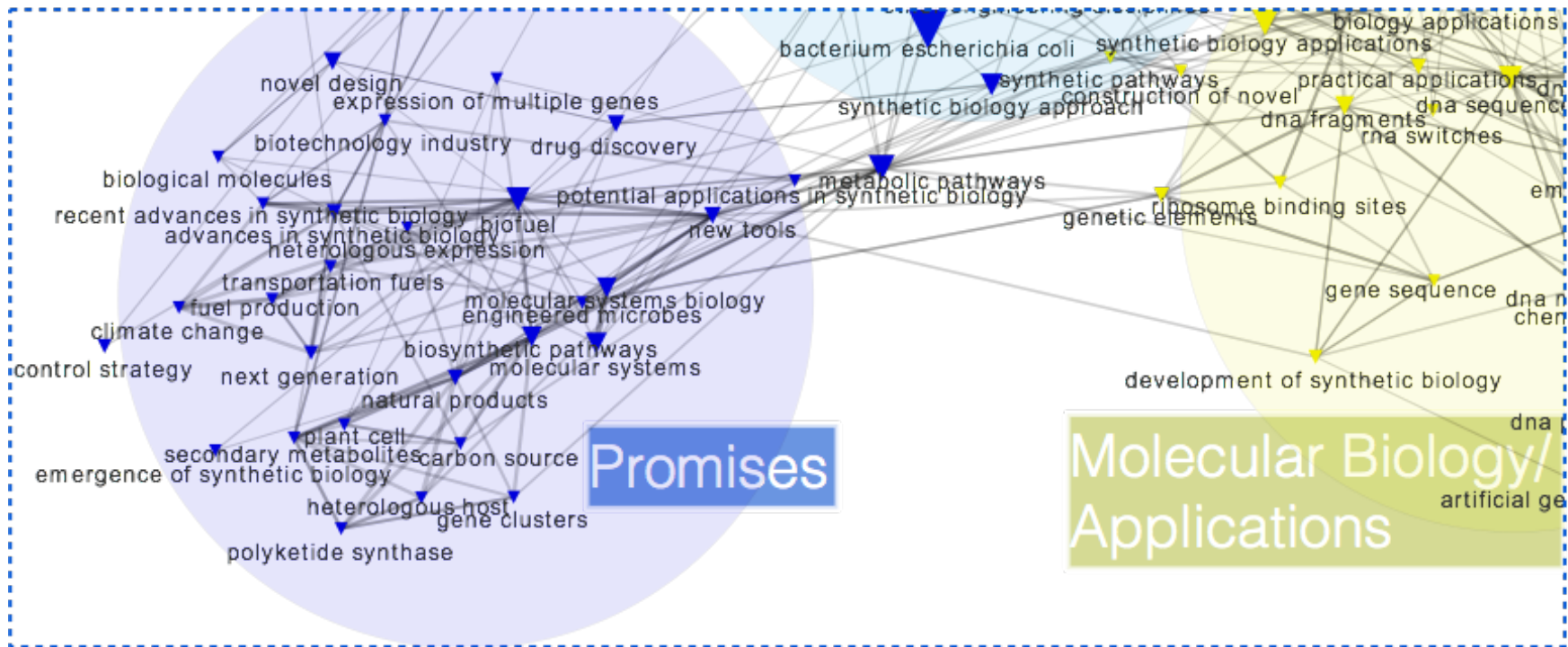
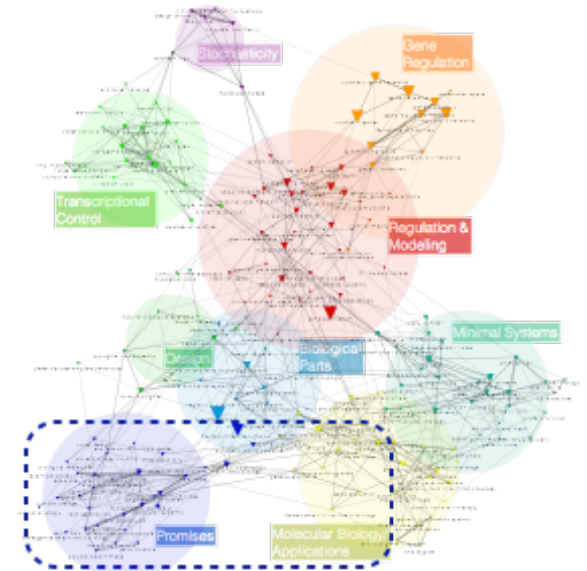
- **Selection:** only the 100 most cited references are selected
- **Network:** They are then connected according to the relative frequency of their co-occurrences and spatialized
- **Clustering:** Automatic detection of cohesive subgroups of references
- **Interpretation:** intellectual schools, major actors
- **Dynamics:**
 - > Time Evolution of the number of publications belonging to each school
 - > radar representation



Actors and discourses

- Determining role of a few **institutional entrepreneurs**
- Central role of **promises** discourses in the article

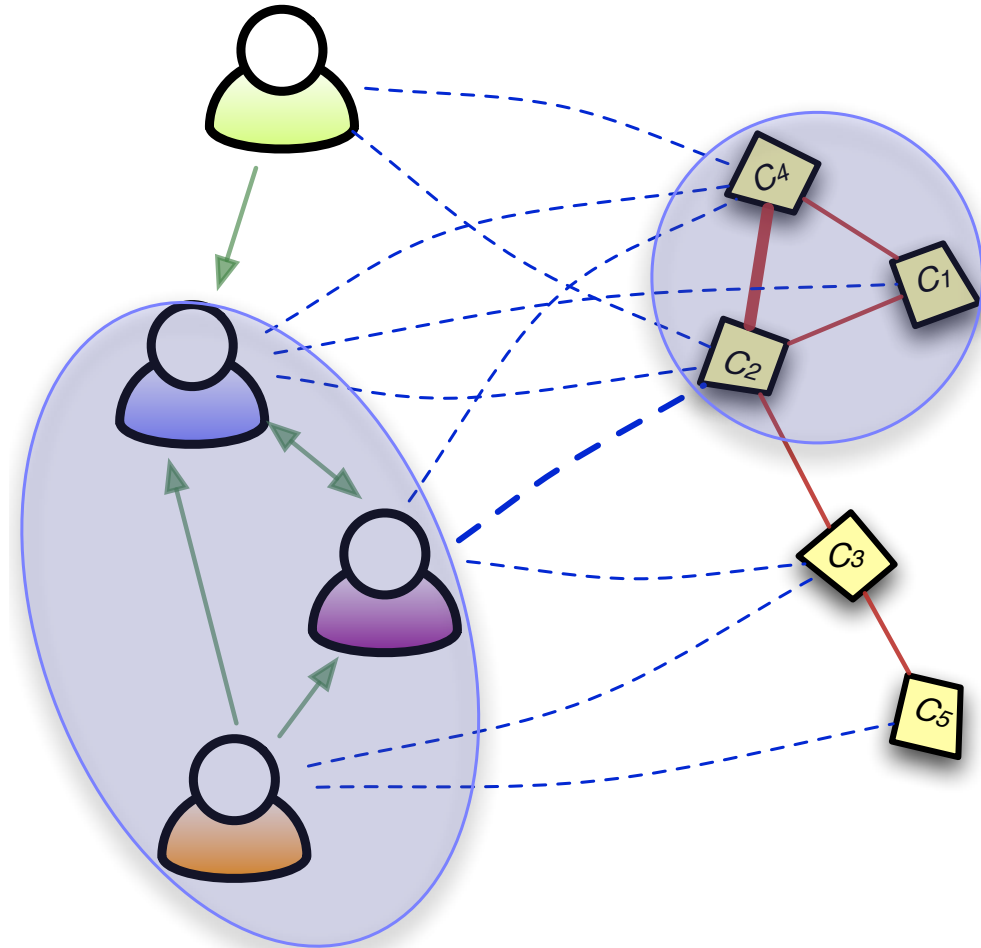
lexical map based on the analysis of titles and abstracts



Unveiling Networks Community Structure

Community Detection Algorithms

- Generalized co-occurrences analysis framework mixing people, terms, countries, etc...
- Clustering techniques are being used to circulate from micro to macro levels - clusters are made of possibly heterogeneous nodes associated in a singular manner.



Socio-semantic network and related clusters

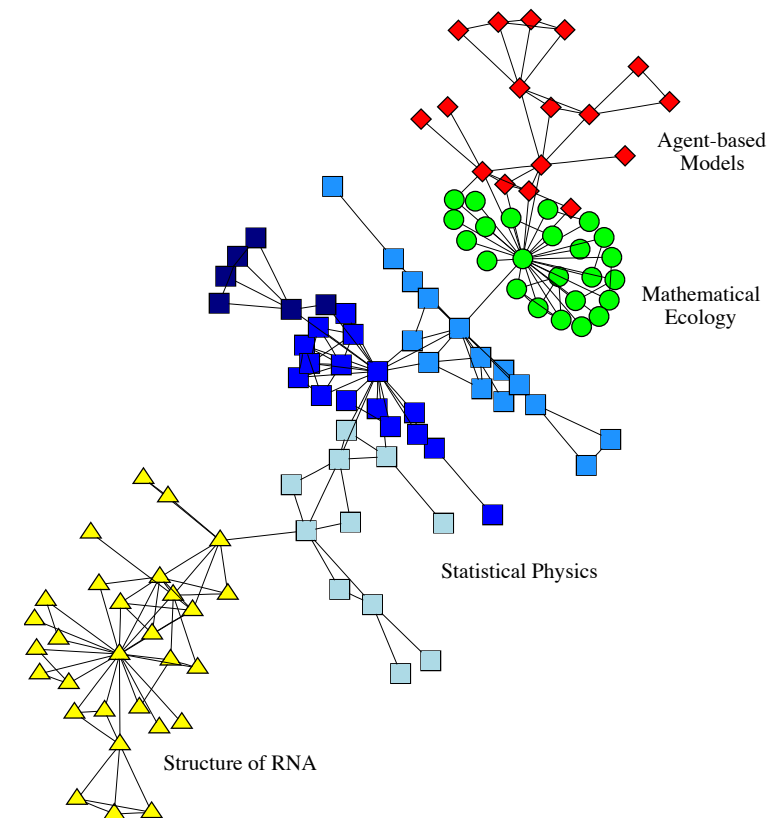
Community Detection Algorithms

- Modularity optimization (algorithmic definition)

$$Q \equiv \frac{1}{2|E|} \sum_c \sum_{i,j \in c} (A_{ij} - P_{ij})$$

$$P_{ij} \equiv \frac{d_i d_j}{2|E|}$$

Network Modularity Measure



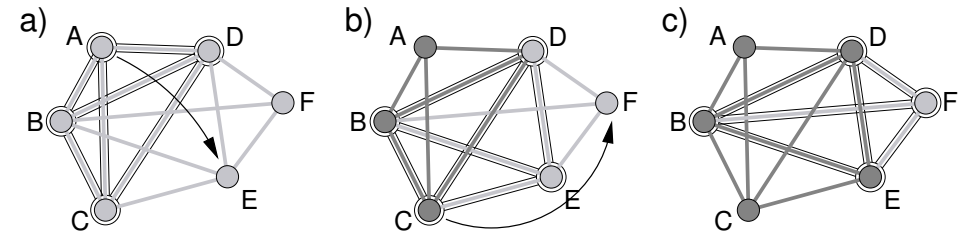
Girvan, M., & Newman, M. E. J. (2002). Community structure in social and biological networks. *Proceedings of the National Academy of Sciences of the United States of America*, 99, 7821–7826.

Santa Fe Collaboration Network

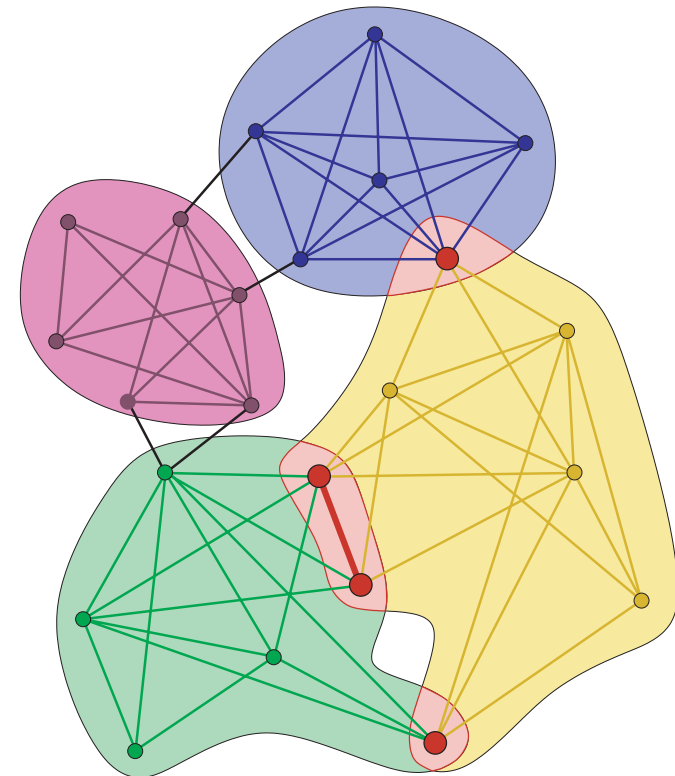
Community Detection Algorithms

- Modularity optimization (algorithmic definition)

- Clique Percolation (algebraic definition)



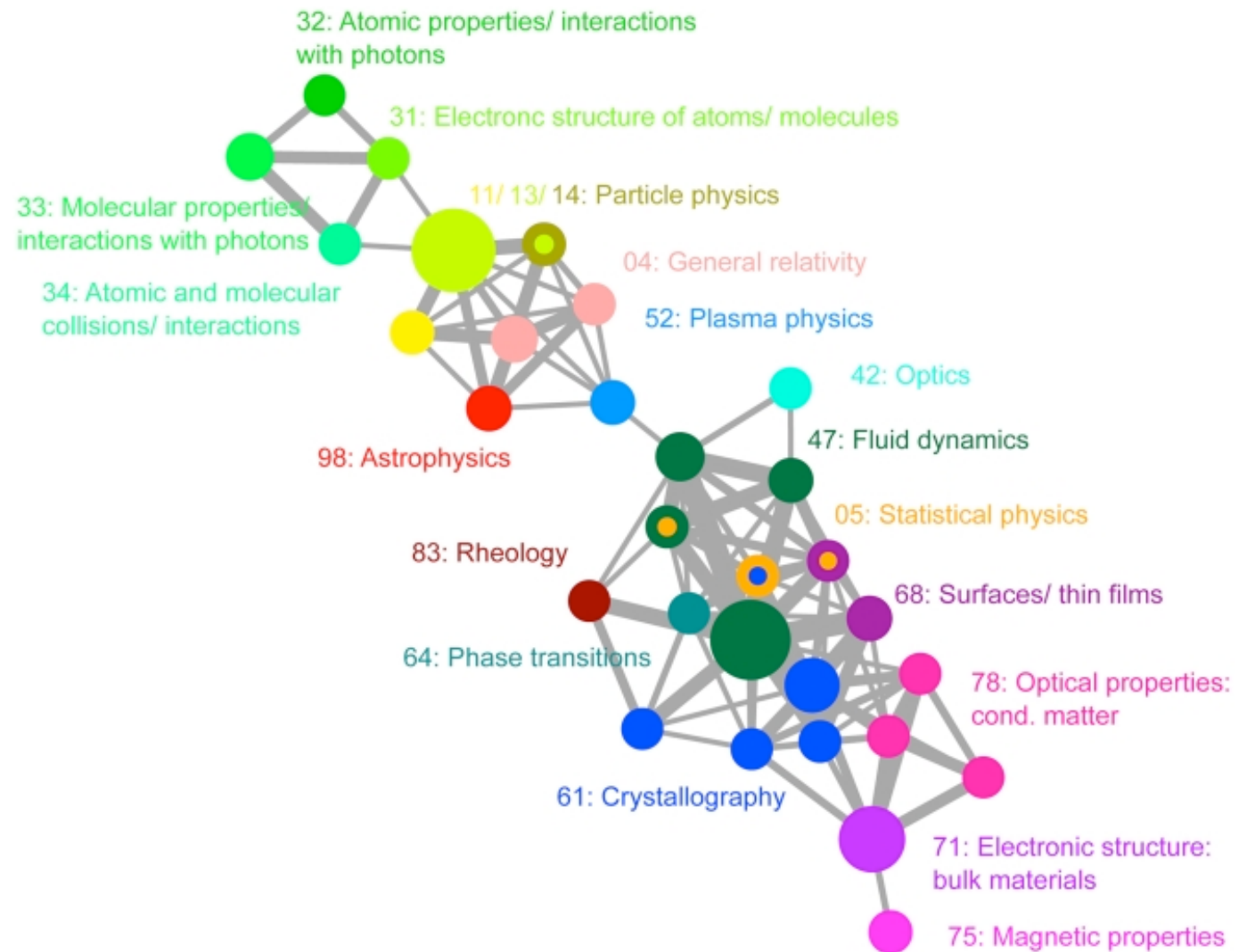
Rolling 4-clique



Clique percolation communities

Community Detection Algorithms

- Modularity optimization (algorithmic definition)
- Clique Percolation (algebraic definition)

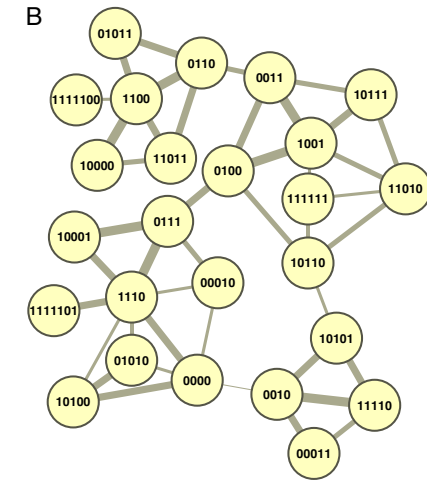
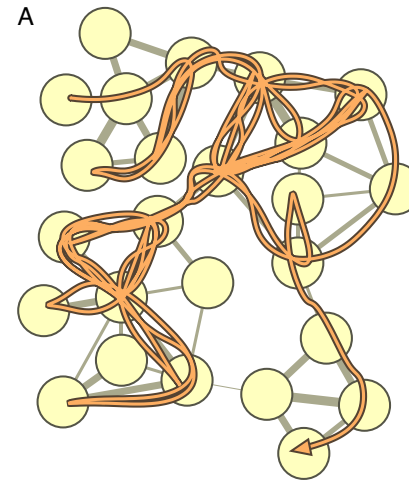


ex: Herrera, Mark, David C. Roberts, and Natali Gulbahce. Mapping the evolution of scientific fields. *PloS one* 5.5 (2010): e10355.

Map based on PACS code of APS publications

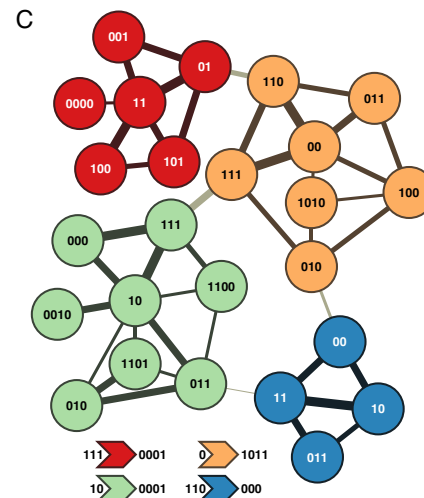
Community Detection Algorithms

- Modularity optimization (algorithmic definition)
- Clique Percolation (algebraic definition)
- Maps of Random walks (information theoretic approach)



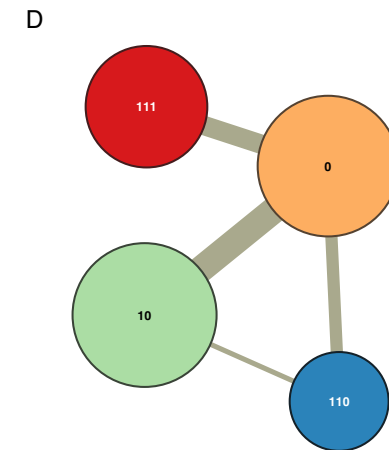
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1111100 1100 0110 11011 10000 11011 0110 0011 10111 1001 0011
1001 0100 0111 10001 1110 0111 10001 0111 1110 0000 1110 10001
0111 1110 0111 1110 1111101 1110 0000 10100 0000 1110 10001 0111
0100 10110 11010 10111 1001 0100 1001 10111 1001 0100 1001 0100
0011 0100 0011 0110 11011 0110 0011 0100 1001 10111 0011 0100
0111 10001 1110 10001 0111 0100 10110 1111111 10110 10101 11110
00011
    
```



```

111 0000 11 01 101 100 101 01 0001 0 110 011 00 110 00 111 1011 10
111 000 10 111 000 111 10 011 10 000 111 10 111 10 0010 10 011 010
011 10 000 111 0001 0 111 010 100 011 00 111 00 011 00 111 00 111
110 111 110 1011 111 01 101 01 0001 0 110 111 00 011 110 111 1011
10 111 000 10 000 111 0001 0 111 010 1010 010 1011 110 00 10 011
    
```



```

111 0000 11 01 101 100 101 01 0001 0 110 011 00 110 00 111 1011 10
111 000 10 111 000 111 10 011 10 000 111 10 111 10 0010 10 011 010
011 10 000 111 0001 0 111 010 100 011 00 111 00 011 00 111 00 111
110 111 110 1011 111 01 101 01 0001 0 110 111 00 011 110 111 1011
10 111 000 10 000 111 0001 0 111 010 1010 010 1011 110 00 10 011
    
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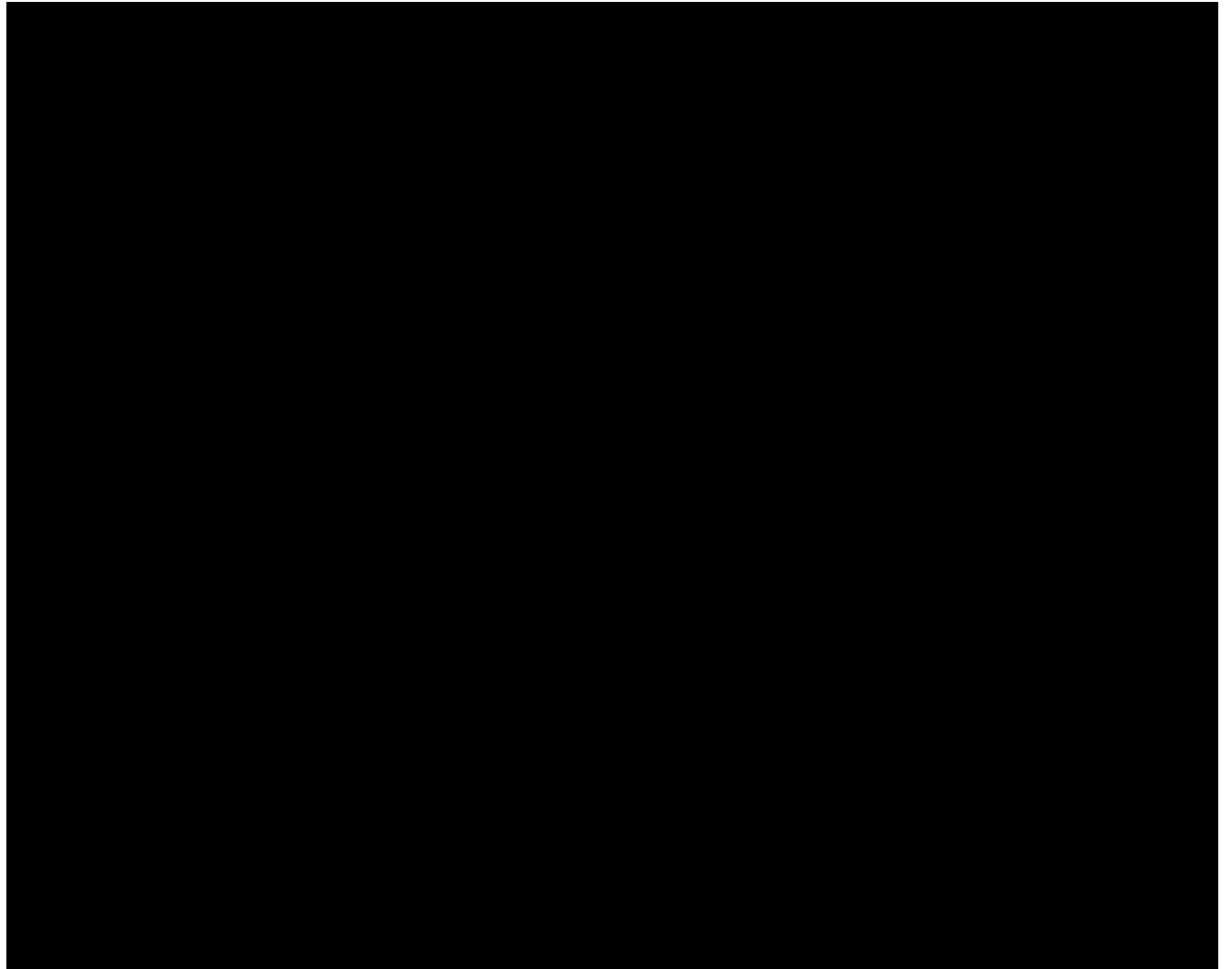
Rosvall, M., & Bergstrom, C.T. (2008). Maps of random walks on complex networks reveal community structure. *Proceedings of the National Academy of Sciences of the United States of America*, 105(4), 1118.

compressing probability flow description

Spatializing Networks

Mapping Networks

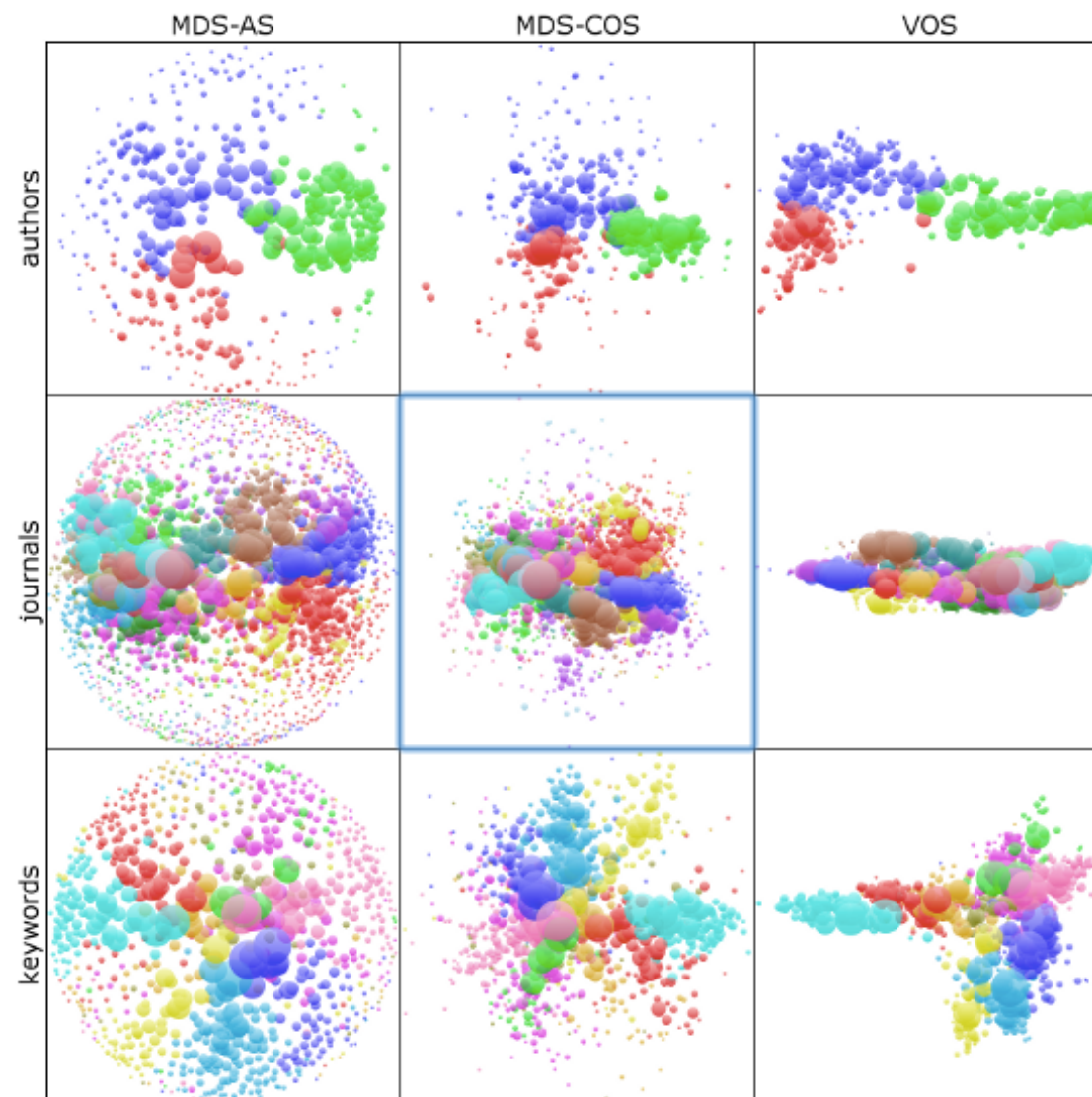
- Force-directed graph drawing



Gephi demo

Mapping Networks

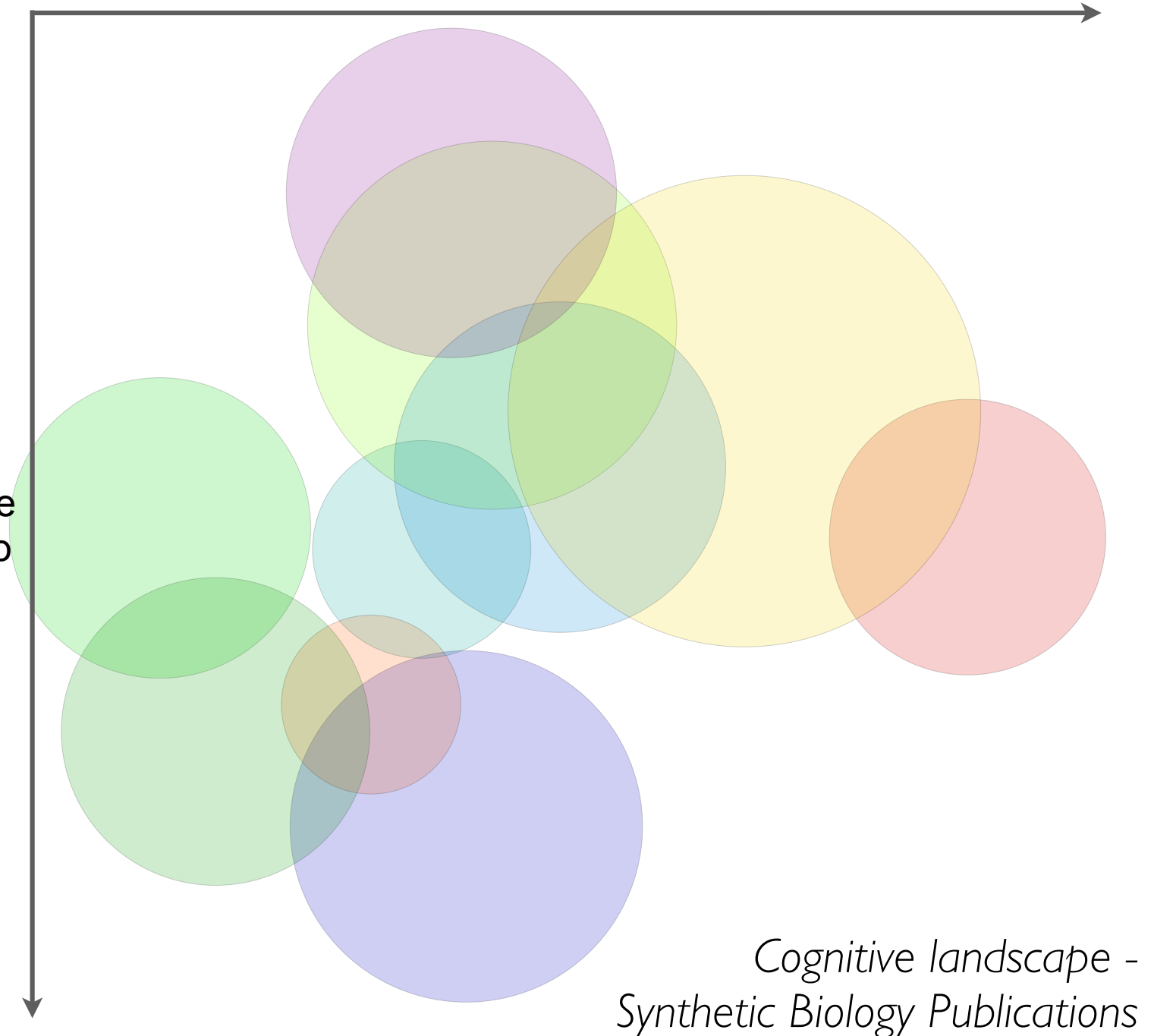
- Force-directed graph drawing
- Multidimensional Scaling and related techniques



Van Eck, N.J., & Waltman, L. (2010) "Software survey: VOSviewer, a computer program for bibliometric mapping", *Scientometrics*, Vol 84, No. 2, pp. 523–538.

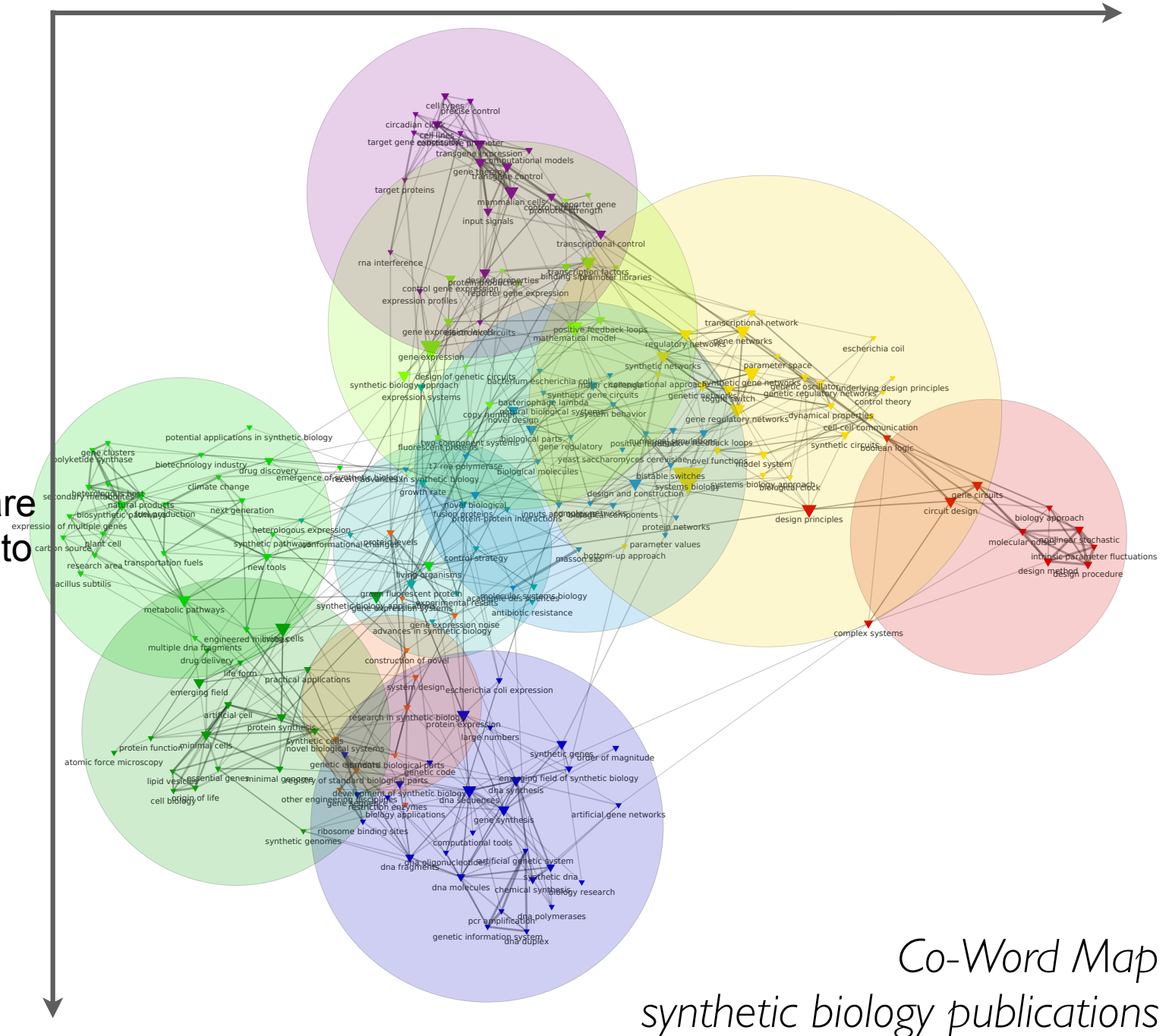
From Communities to Network Mapping

- Force-directed graph drawing
- Multidimensional Scaling and related techniques
- Actors, (here terms) are spatialized according to the position of their communities



From Communities to Network Mapping

- Force-directed graph drawing
- Multidimensional Scaling and related techniques
- Actors, (here terms) are spatialized according to the position of their communities



Textual Analysis

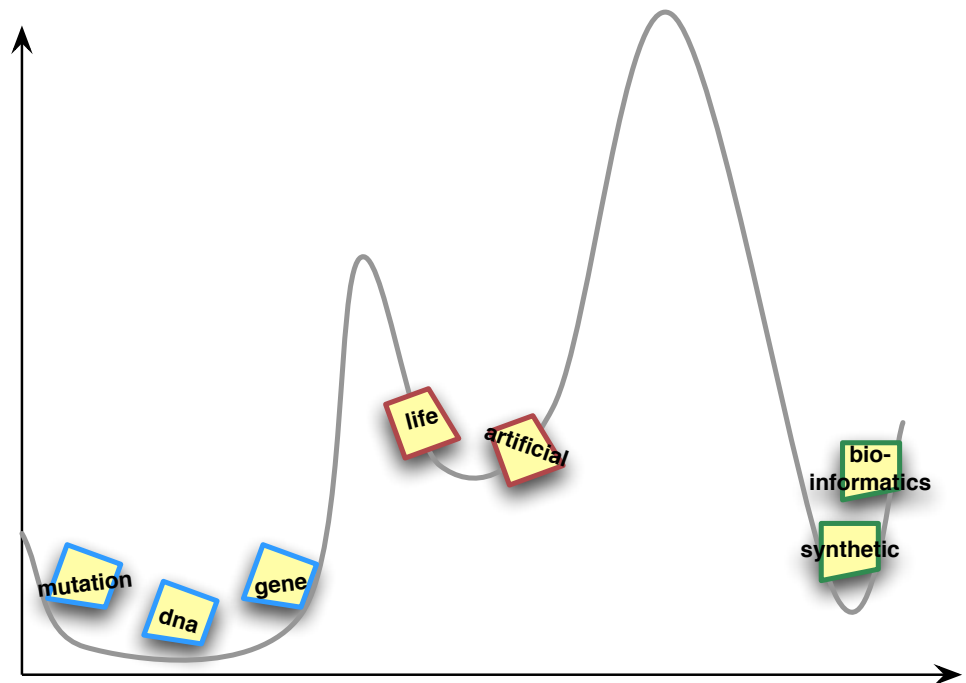
Textual Analysis

«Indexing is an intervention between the text and the co-word analysis, and the validity of the map will depend, to a certain extent, on the nature of the indexing. Yet since indexers try to capture what it is about a text that is interesting, they partially reproduce the readings that the texts are given within the field itself'. Thus, despite the fact that indexing is not entirely reliable, validity is never totally absent.»

Callon, M.; Law, J.; & Rip, A. (Eds.). (1986a). *Mapping the dynamics of science and technology: Sociology of Science in the real world*. London: The Macmillan Press Ltd.

what it is about a text that is interesting?

- **grammatical** criterion, candidate terms are usually limited *noun phrases*,
- **termhood**, terms should be domain specific to carry substantial information



Textual Analysis

We believed we could reduce our dependence on foreign oil and protect our planet. And today, America is number one in oil and gas.

Textual Analysis

Part-Of-Speech Tagging

PRP VBD PRP MD VB PRP NN IN JJ NN
We believed we could reduce our dependence on foreign oil
CC VB PRP NN. CC NN NNP VBZ NN CD IN
and protect our planet. And today, America is number one in
NN CC NN
oil and gas.

Textual Analysis

Chunking

PRP VBD PRP MD VB PRP NN IN JJ NN
We believed we could reduce our dependence on foreign oil
CC VB PRP NN. CC NN NNP VBZ NN CD IN
and protect our planet. And today, America is number one in
NN CC NN
oil and gas.

Extracted noun phrases:

- *dependence*
- *planet*
- *oil*
- *gas*

Textual Analysis

Chunking

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Extracted noun phrases:

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- *planet*
- *oil*
- *gas*
- *foreign oil*

Textual Analysis

Chunking

PRP VBD PRP MD VB PRP NN IN JJ NN
We believed we could reduce our dependence on foreign oil
CC VB PRP NN. CC NN NNP VBZ NN CD IN
and protect our planet. And today, America is number one in
NN CC NN
oil and gas.

Extracted noun phrases:

- *dependence*
- *planet*
- *oil*
- *gas*
- *foreign oil*
- *dependence on foreign oil*
- *oil and gas*

Textual Analysis

Stemming, Filtering and Standardizing

PRP VBD PRP MD VB PRP NN IN JJ NN
We believed we could reduce our dependence on foreign oil

CC VB PRP NN. CC NN NNP VBZ NN CD IN
and protect our planet. And today, America is number one in

NN CC NN
oil and gas.

Extracted classes:

- *dependence on foreign oil: {dependence on foreign oil ; foreign oil dependence}*
- *oil and gas: {oil and gas; gas and oil}*
- *planet: {planet, planets}*
- *etc.*

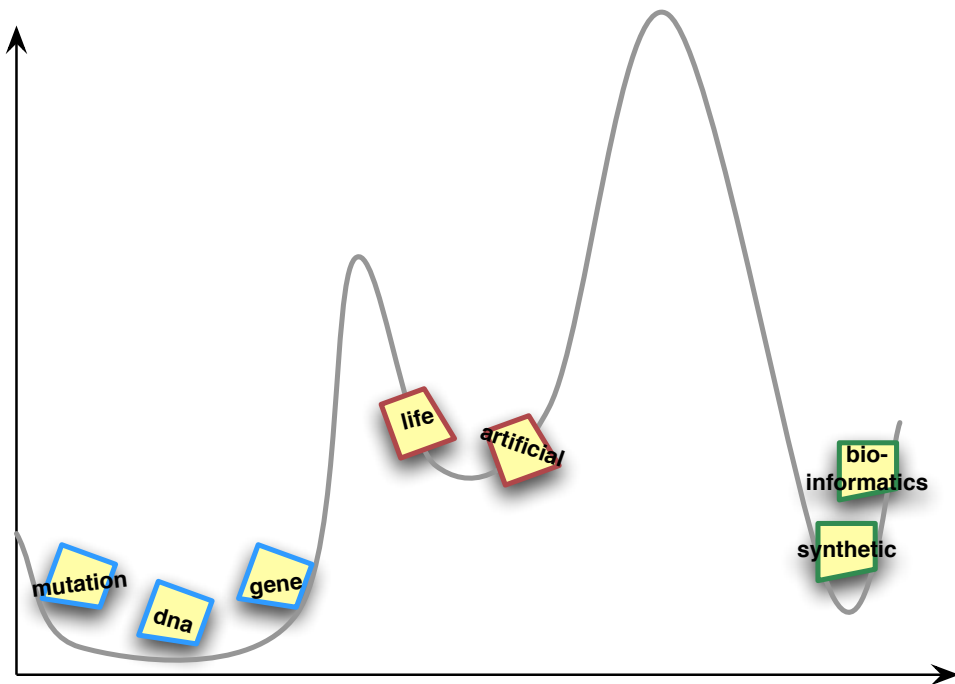
Textual Analysis

Termhood

- Candidate terms should be thematically specific ; terms not specific to a specific thematic subfield have neutral meaning given the whole domain and should be excluded
- Two possible strategies to assess how pertinent a term may be:
 - A term which local frequency in documents it occurs in is high is likely to be informative: *gf.idf*, pigeonhole score
 - A term which distribution is slanted toward certain type of documents or sub-vocabularies is likely to be pertinent: LLR, χ^2

Textual Analysis

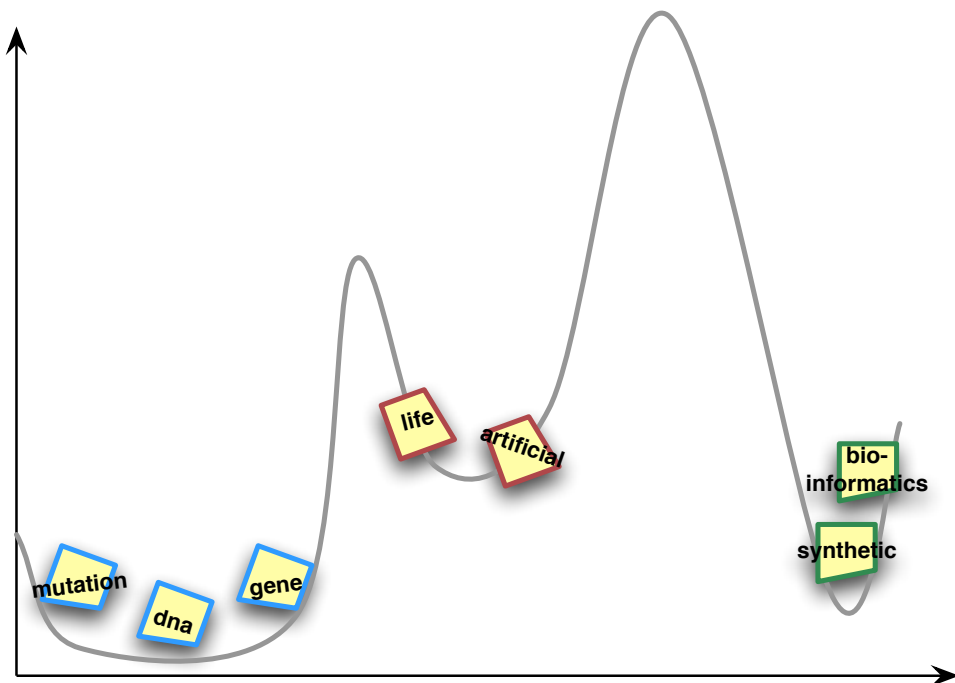
NLP analysis outcome



label	forms
heterologous expression	heterologous expression
complex networks	complex networks; complex network
design principles	design principle; design principles
transgene control	transgene control
gene expression noise	gene expression noise; noise in gene expression; gene-expression noise; noise in gene-expression
biosynthetic pathways	biosynthetic pathways; biosynthetic pathway
mathematical model	mathematical model; mathematical models; Mathematical models
metabolic pathways	metabolic pathway; metabolic pathways
control theory	control theory
reporter gene expression	reporter gene expression; reporter gene-expression
gene expression systems	gene expression systems; gene-expression systems
biological molecules	biological molecules; biological molecule
control circuit	control circuit; control circuits; circuit from a control; circuits control
toggle switch	Toggle switch; toggle switch; toggle switches
protein networks	protein networks; networks of proteins; proteins and networks
restriction enzymes	restriction enzymes; restriction enzyme
transcriptional network	transcriptional networks; transcriptional network; transcription networks
system design	system design; design these systems; systems design
Bacillus subtilis	Bacillus subtilis; bacillus subtilis
multiple DNA fragments	multiple DNA fragments; multiple dna fragments
drug delivery	drug delivery; delivery of drugs
biological clock	biological clock; biological clocks; Biological Clocks
mammalian cells	mammalian cells; Mammalian Cells
fuel production	production of fuels; Production of fuels; fuel production
artificial gene networks	artificial gene networks
control strategy	control strategy; control strategies; strategies that control
gene expression levels	gene expression levels; level of gene expression; gene-expression levels; level of gene-expression
computational approach	computational approach; Computational approaches; computational approaches; computational approach-; computational Approach; Computational approach-; Computational approach; Computational Approach
DNA synthesis	DNA synthesis; dna synthesis
gene networks	gene networks; gene network; Gene networks; networks of genes; Gene Network; genes and networks
biological parts	Biological Parts; biological parts
genetic information system	genetic information system; genetic -information system; genetic Information system
engineered microbes	engineered microbes; Engineering microbes; engineering microbes
Registry of Standard Biological Parts	registry of standard biological parts; Registry of Standard Biological Parts; Registry of Standard-Biological Parts; Registry of STANDARDS Biological Parts; Registry of standards Biological Parts; Registry of Standards Biological Parts

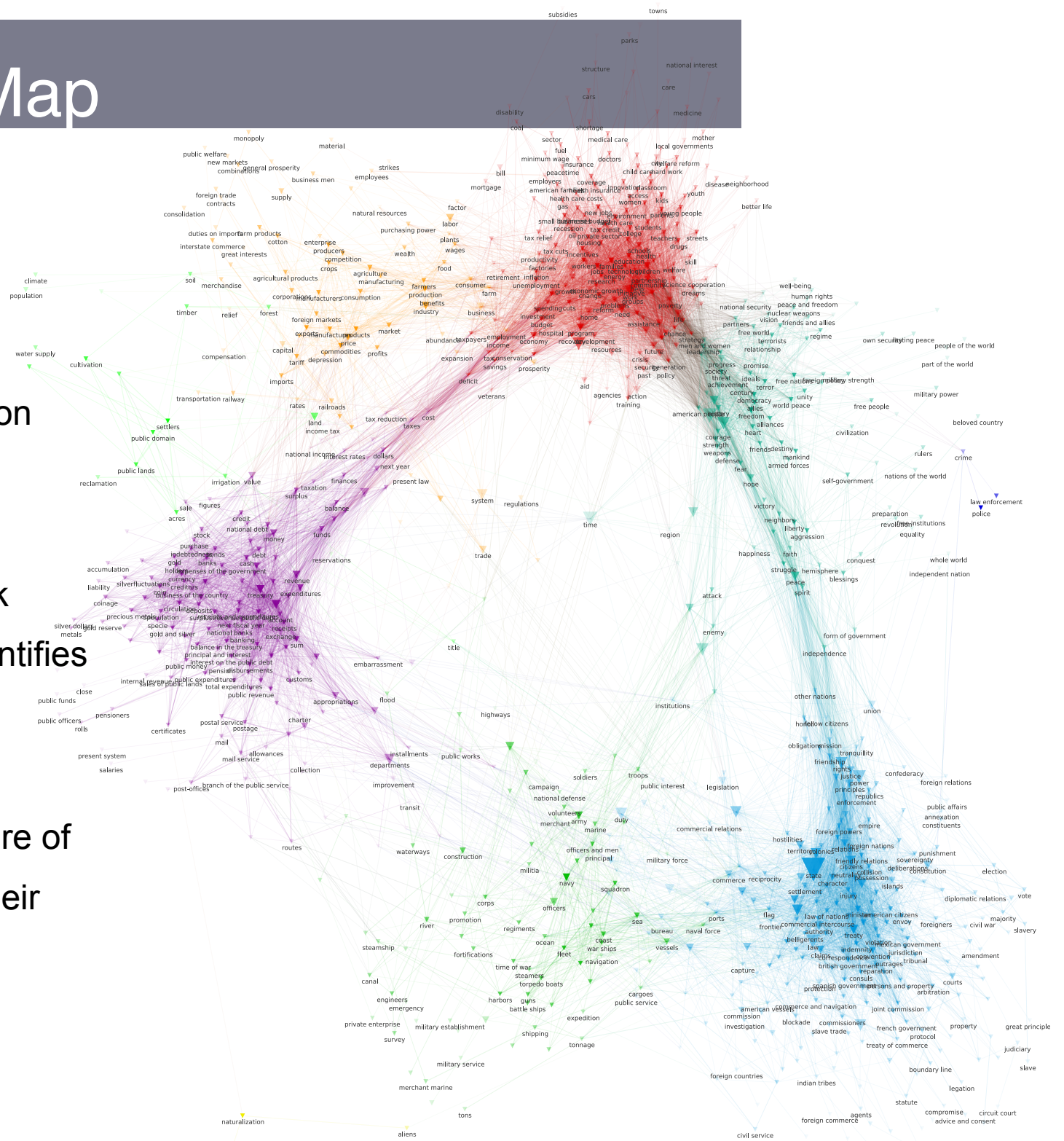
Textual Analysis

NLP analysis outcome



label	forms
abundance	abundance
acres	acres&lacre
aliens	aliens&lalien
alliances	alliances&lalliance
allies	allies&lally
American life	American life&lAmerican lives
armed forces	armed forces&larmed force&lforce of arms&larms and force
boundary line	boundary line&lline of boundary
business men	business men&lbusiness man&lmen of business&lman of business
children	children&lchild
commerce and navigation	commerce and navigation&lnavigation and commerce&lnavigation or commerce
construction work	construction work&lconstruction of works&lwork of construction&lworks construction&lconstruction of these works&lwork in construction
crews	crews&lcrew
crime	crime&lcrimes
crisis	crisis&lcrises
crops	crops&lcrop
cruisers	cruisers&lcruiser
democracy	democracyl&ldemocracies
diplomatic relations	diplomatic relations
expenses of the Government	expenses of the Government&lGovernment expenses&lGovernment expense&lexpenses of Government&lexpense of the Government
farm products	farm products&lproducts of the farm&lproducts of farm
great importance	great importancel&lgreater importancel&lgreatest importance
health care costs	health care costs&lcost of health care&lhealth care cost
income tax	income tax<he income<he tax on the income
peace and freedom	peace and freedom&lfreedom and peace&lpeace with freedom
property rights	property rights&lproperty right&lrights of property&lright of property&lrights and property

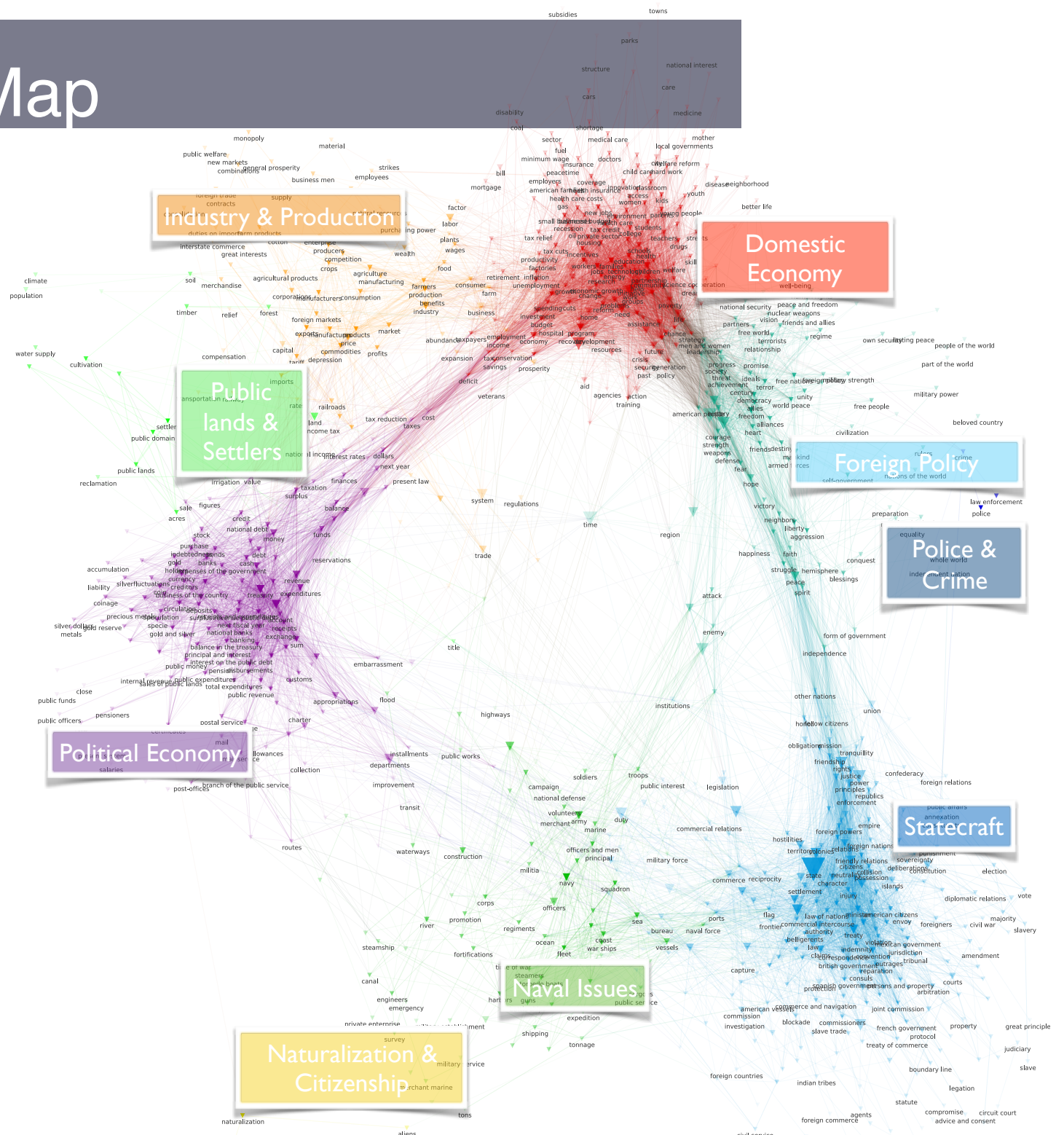
Semantic Map



Four steps process:

1. **Similarity Network** based on terms cooccurrences
2. **Filter Edges** to obtain the sparsest connected network
3. **Community detection** identifies cohesive thematic areas
4. **Mapping** allows to visually investigate the inner structure of those clusters, as well as their relationships

Semantic Map



9 Topics:

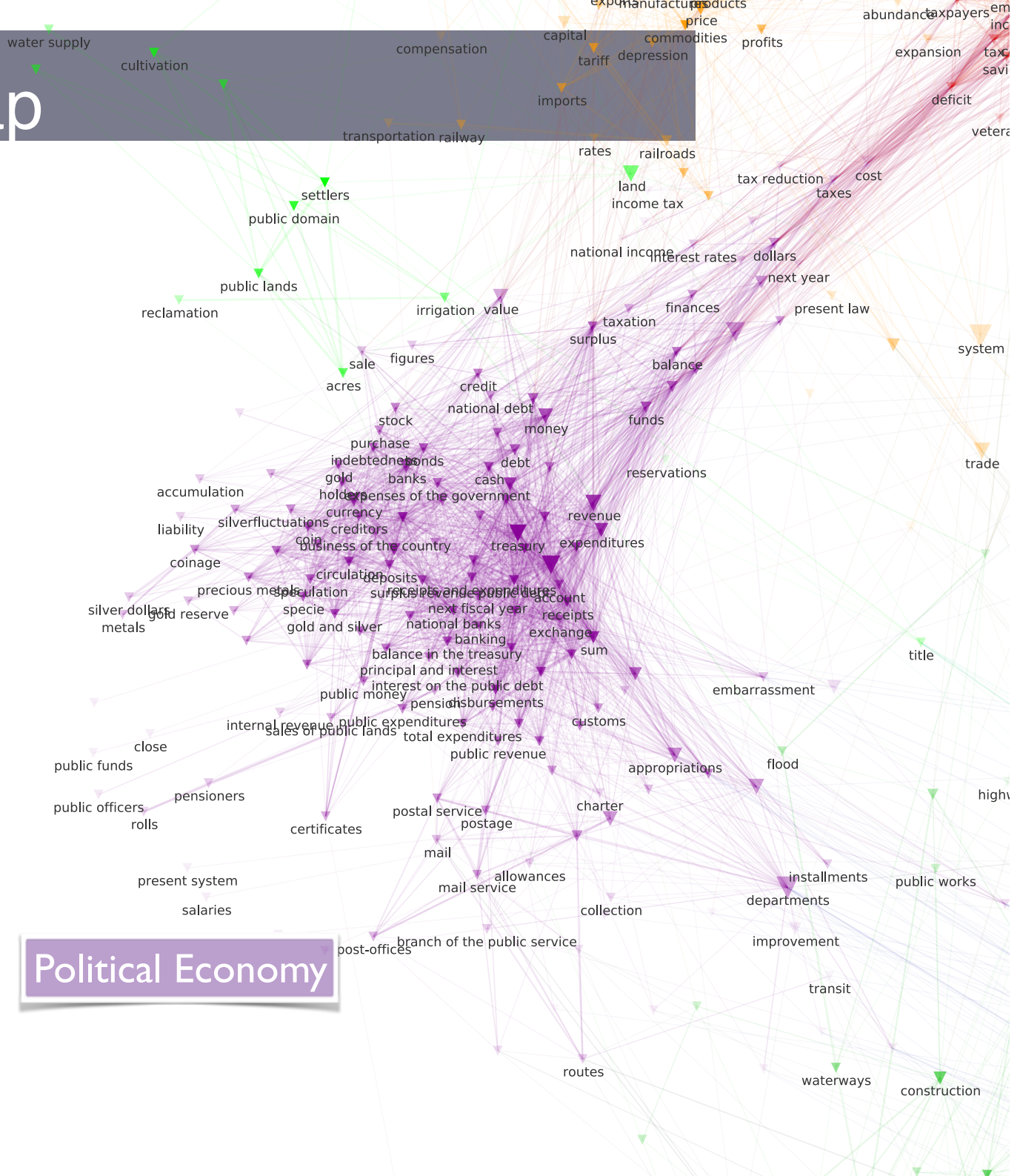
- **Statecraft** — “state”, “power”, “law”, “rights”, “act”, “purpose”
- **Naval issues** — “navy”, “sea”, “coast”, “officers”, “service”
- **Foreign policy** — “war”, “people”, “nations”, “force”, “peace”
- **Domestic/economy** — “Need”, “work”, “economy”, “years”, “policy”
- **Industry** — “system”, “trade”, “corporations,” “business”, “labor”
- **Political economy** — “Treasury”, “amount”, “appropriations”, “value”
- **Public lands & Settlers** — “land”, “settlers”, “acres”
- **Naturalization & Citizenship** — “naturalization”, “aliens”
- **Police & Crime** — “criminal”, “law enforcement”

Semantic Map

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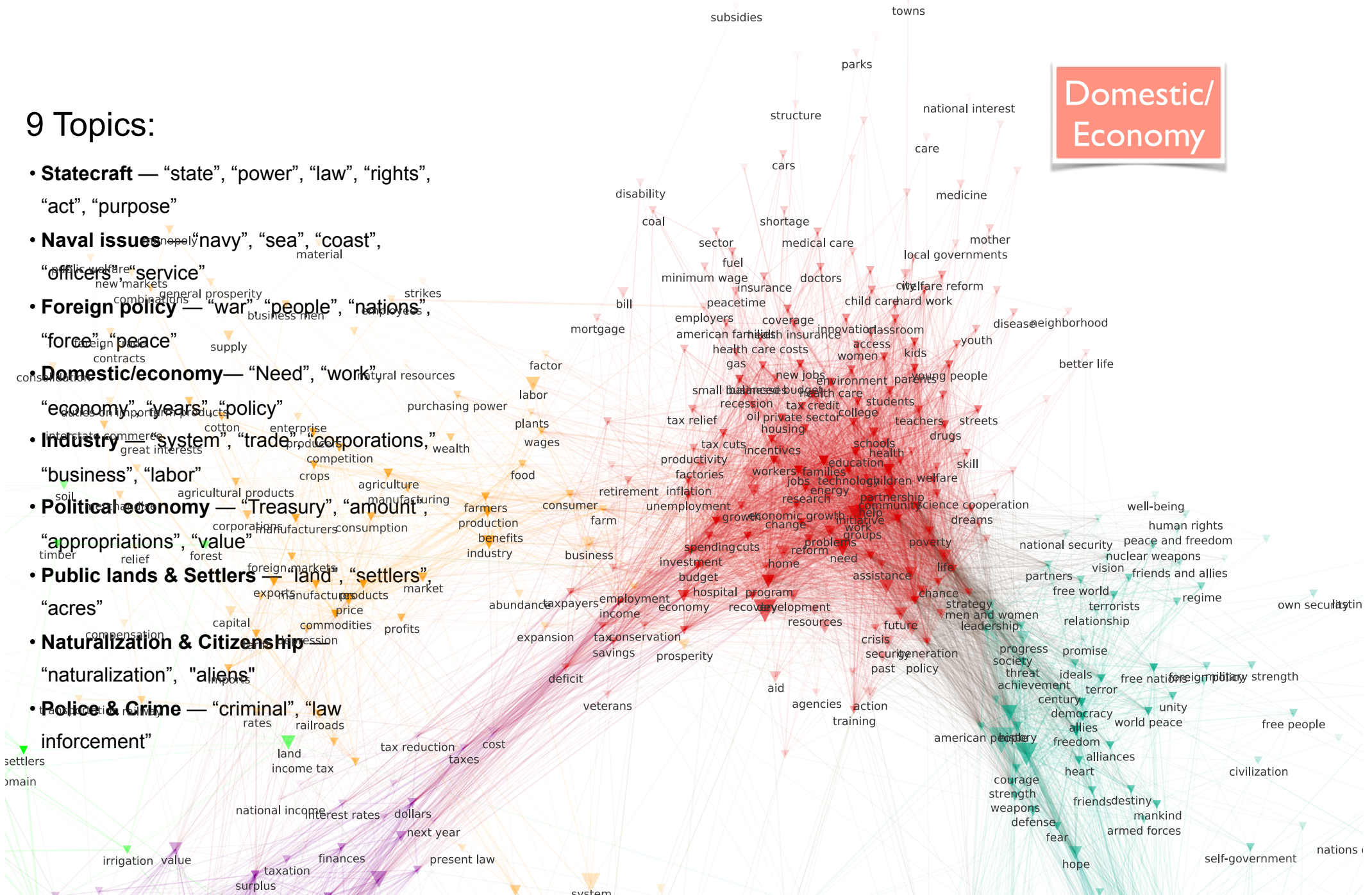
Political Economy



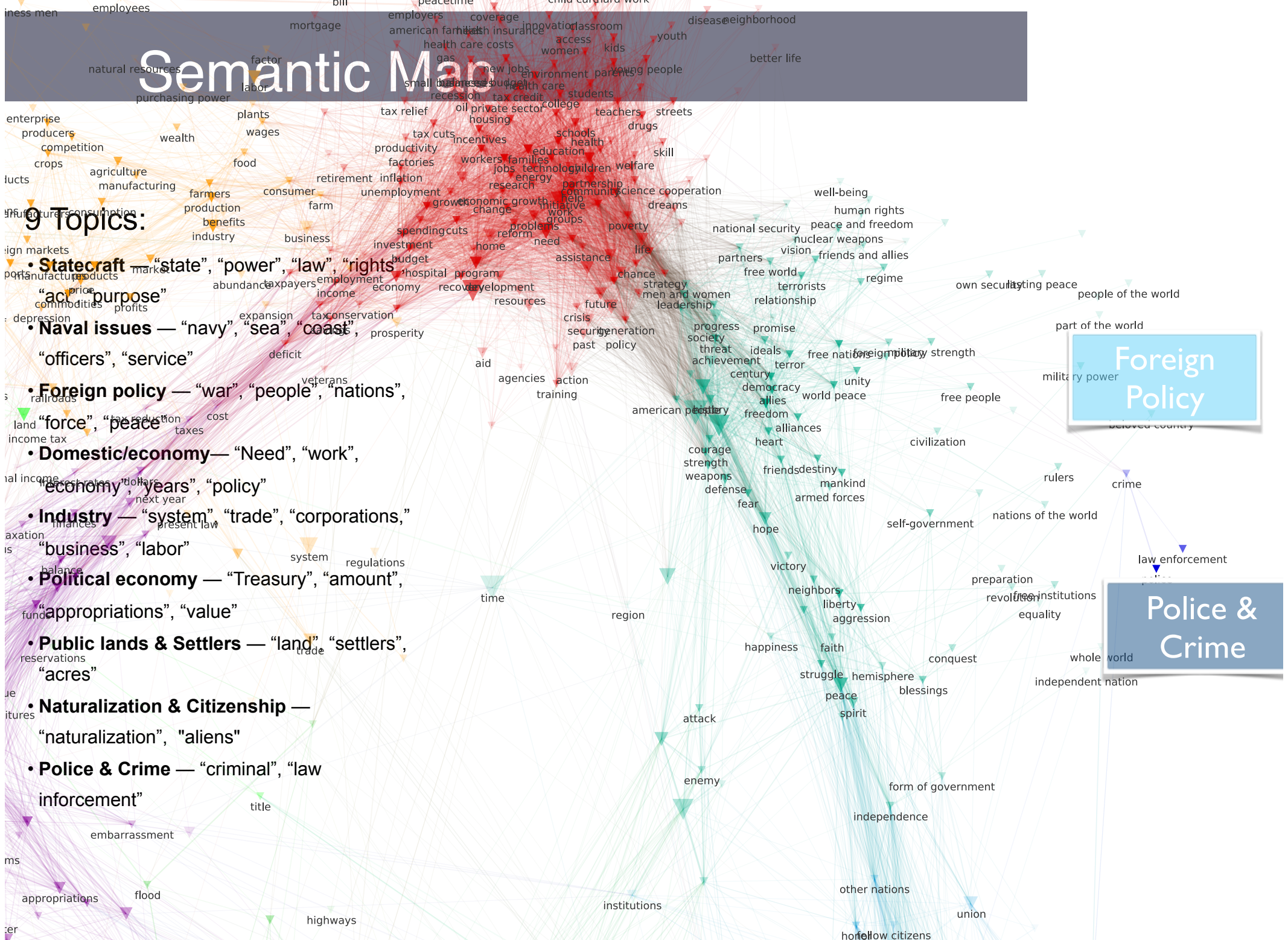
Semantic Map

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Semantic Map



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Foreign Policy

Police & Crime

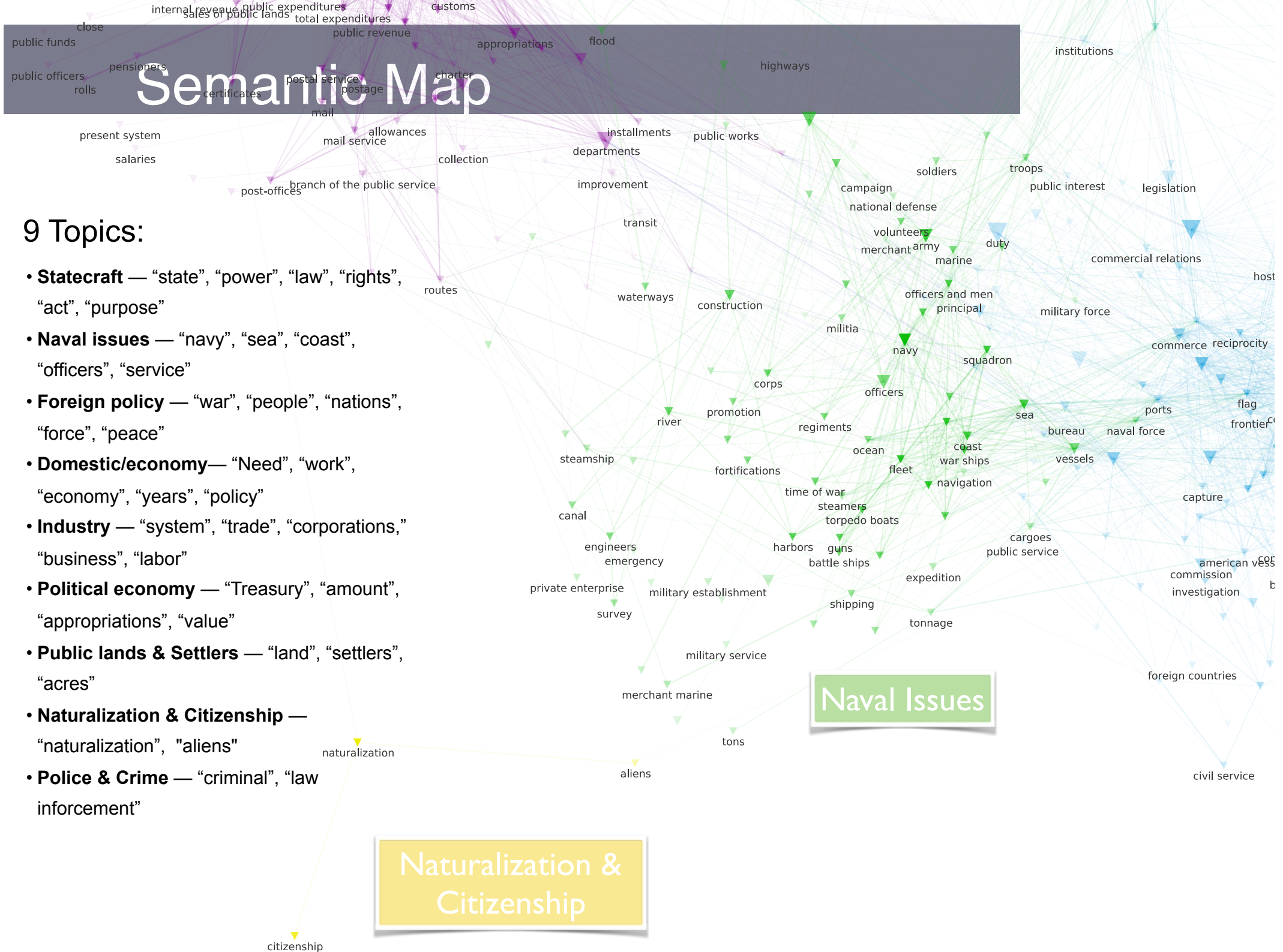
Semantic Map

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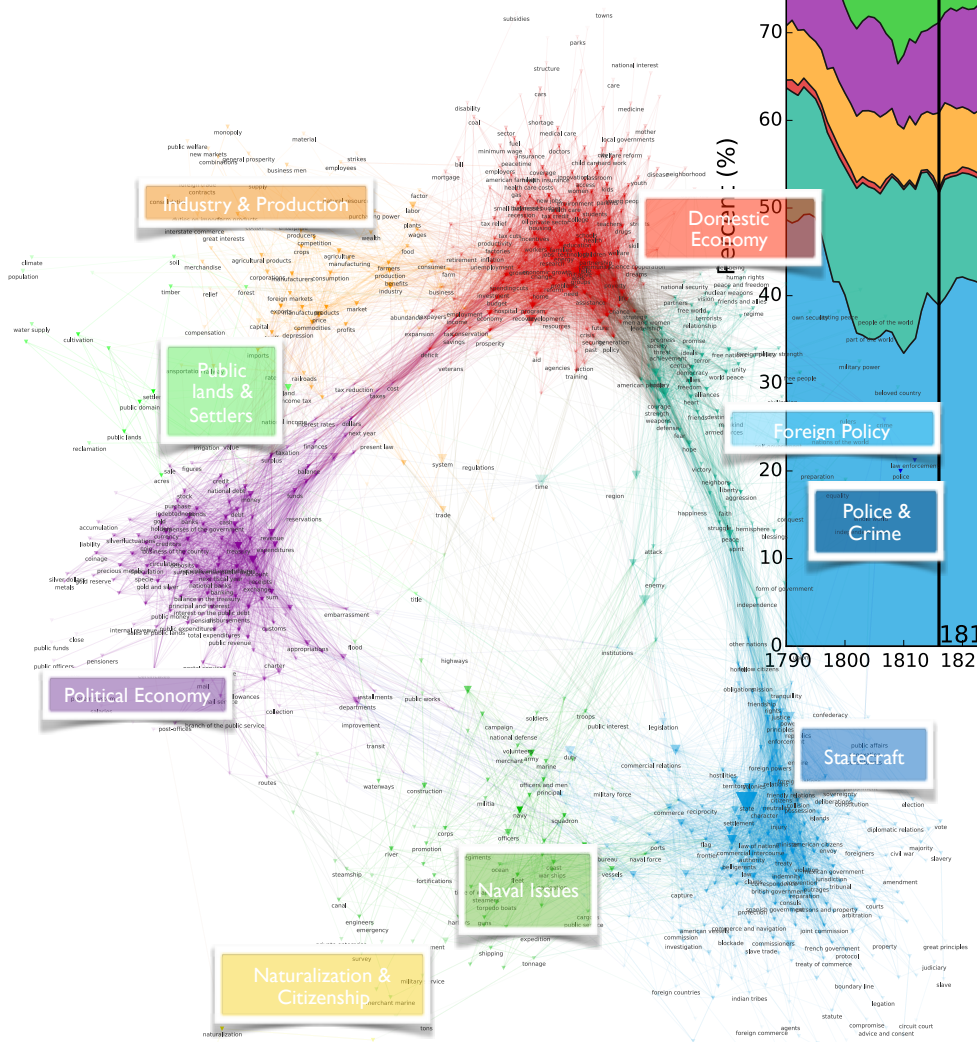
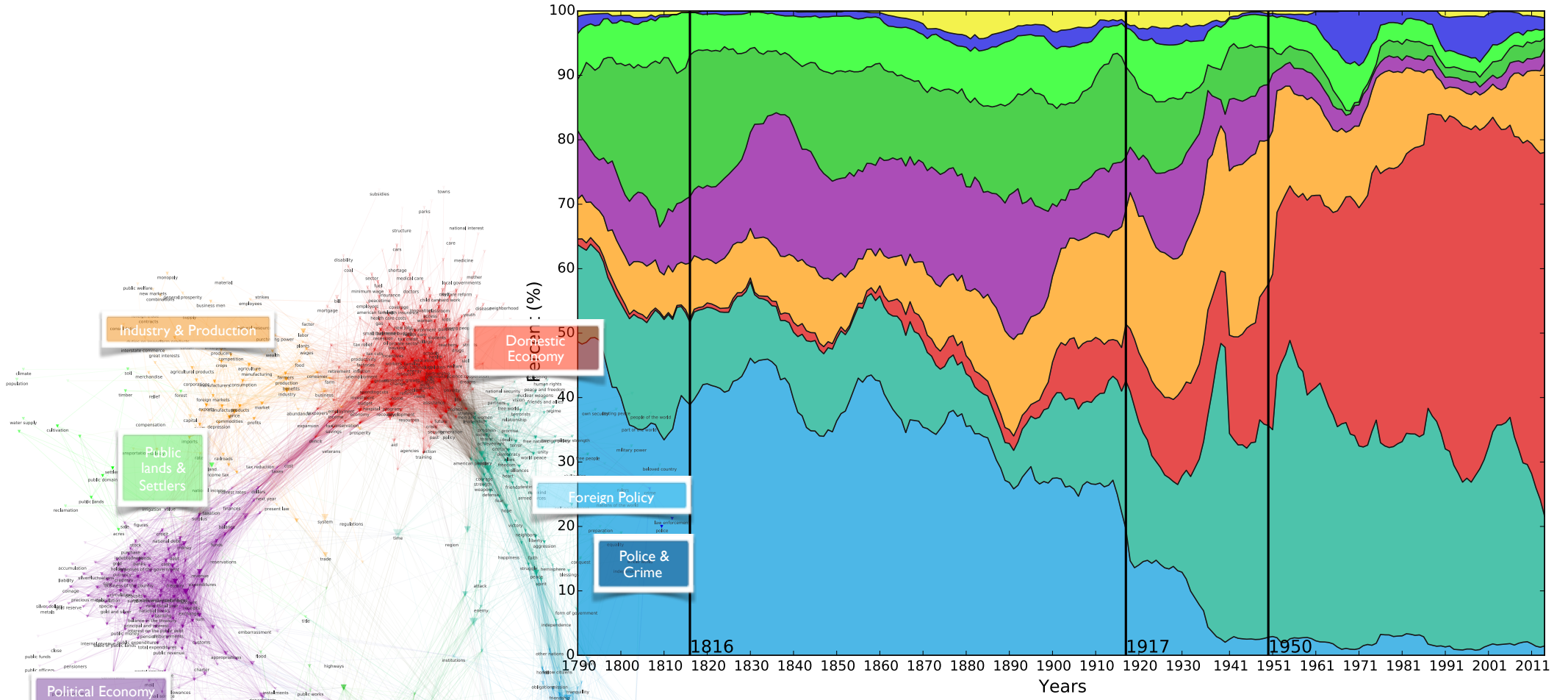
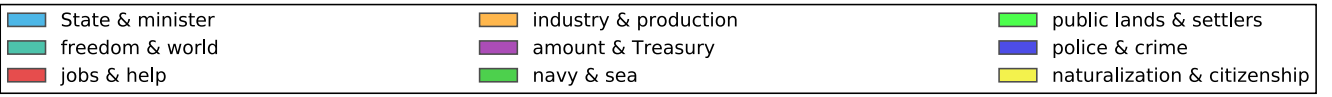
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Naturalization & Citizenship

Naval Issues



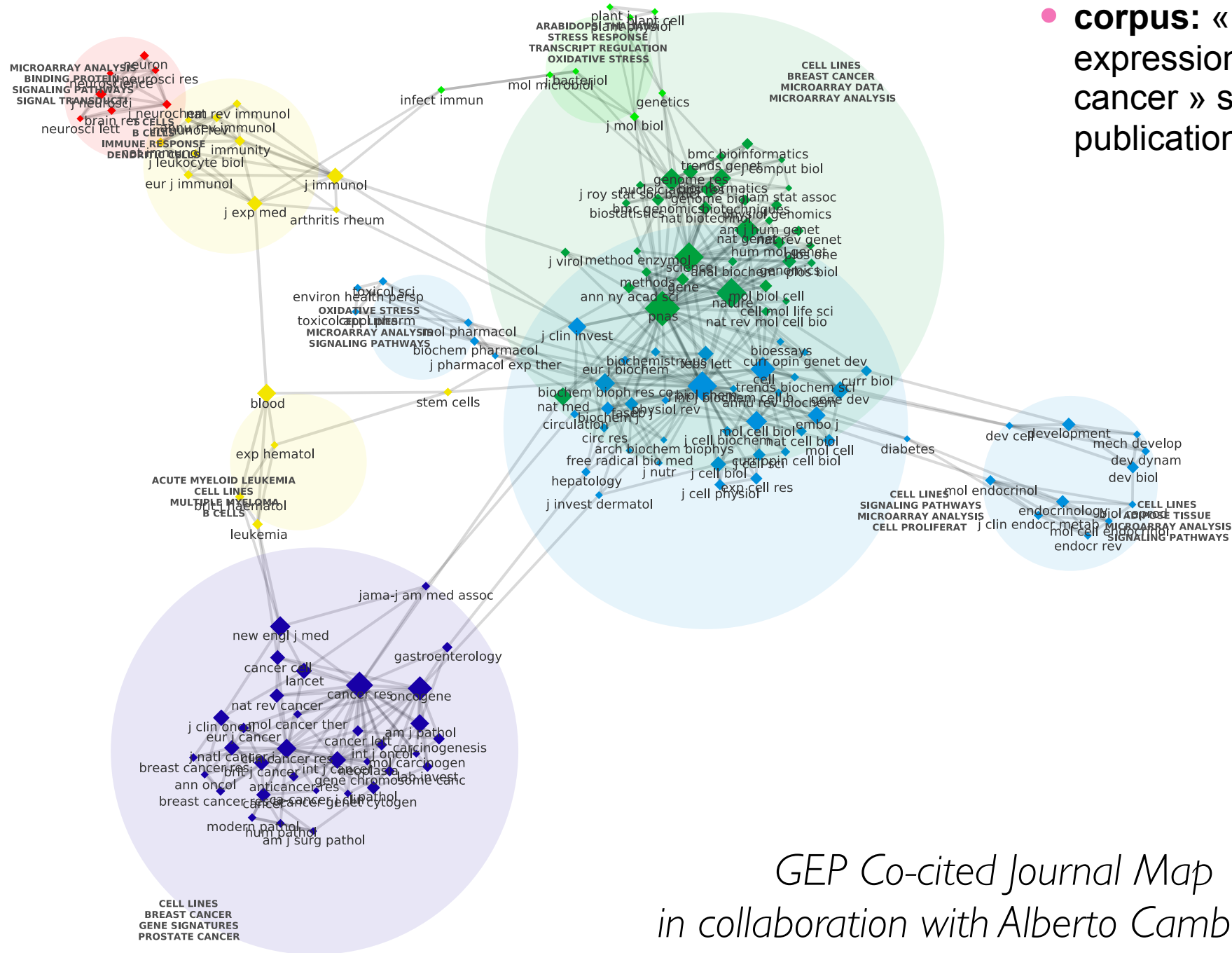
Categories in Time



From heterogeneous structure to dynamics

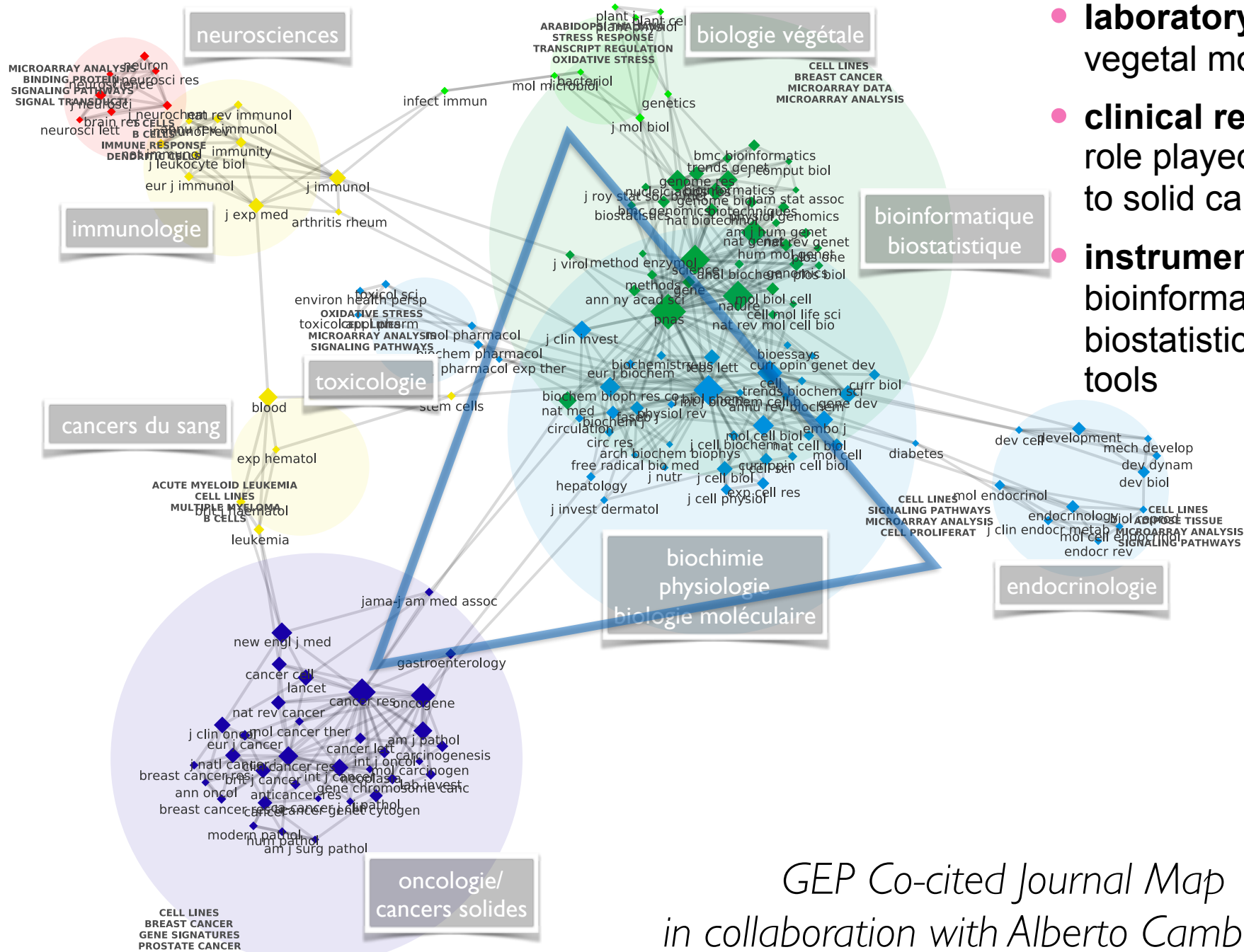
Translational Research Triangle

- **corpus:** « Gene expression profile & cancer » scientific publications



*GEP Co-cited Journal Map
in collaboration with Alberto Cambrosio*

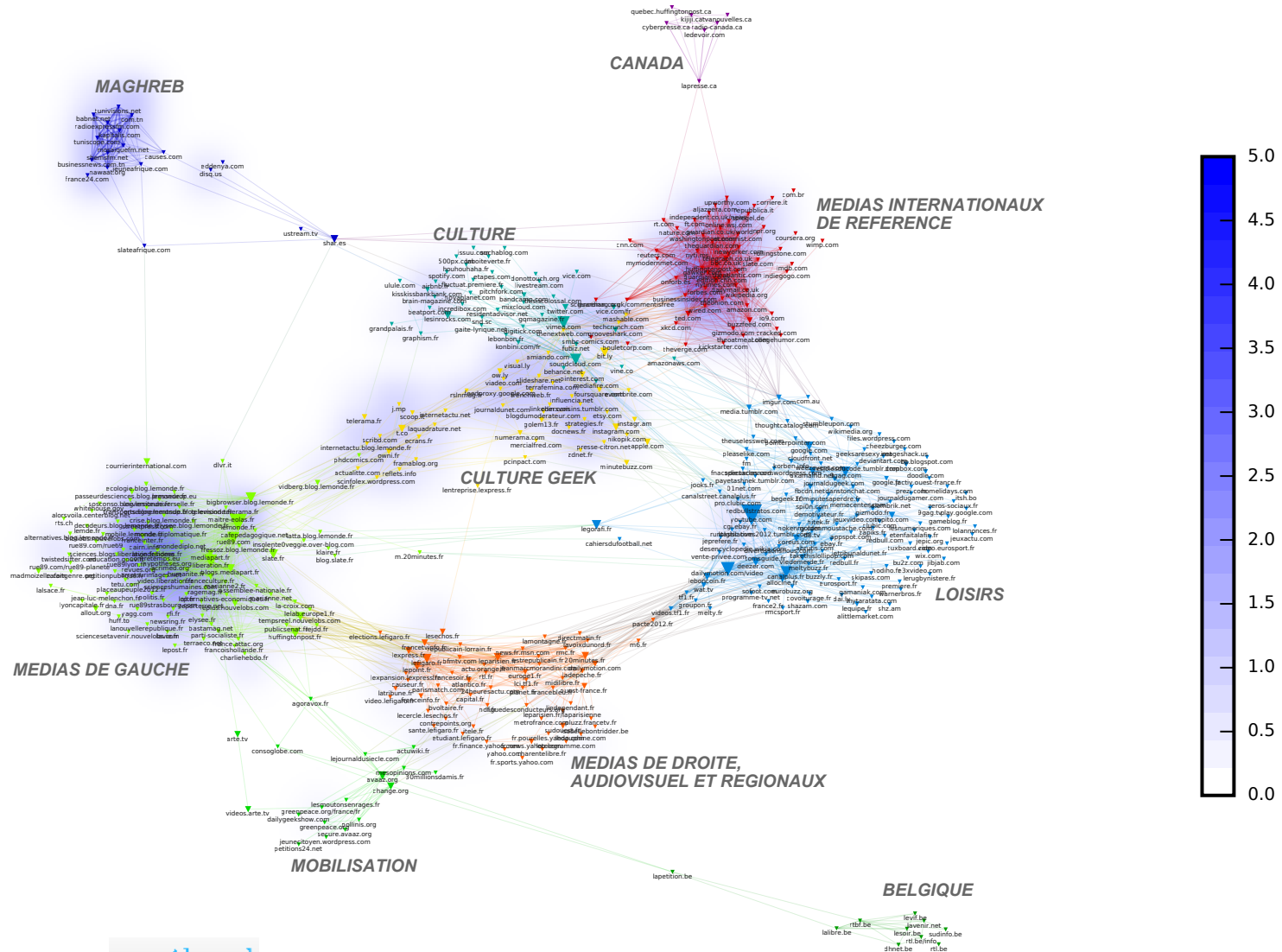
Translational Research Triangle



- **laboratory research** - vegetal model organism
- **clinical research** - central role played by application to solid cancer
- **instrumental research** - bioinformatics and biostatistics as necessary tools

*GEP Co-cited Journal Map
in collaboration with Alberto Cambrosio*

Mixing dimensions of analysis

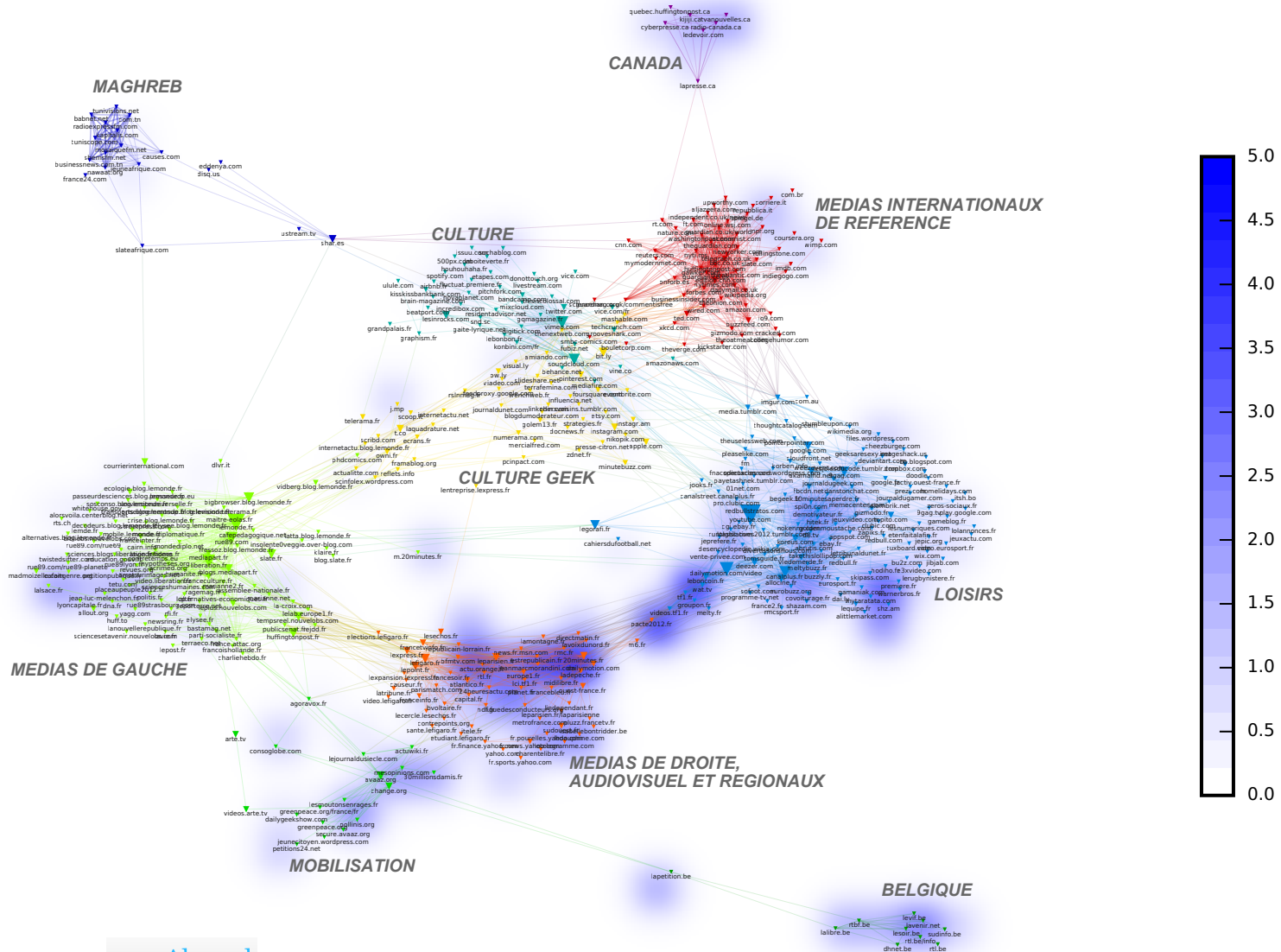


app. Algopol

CORTEXT

Facebook citation network

Mixing dimensions of analysis

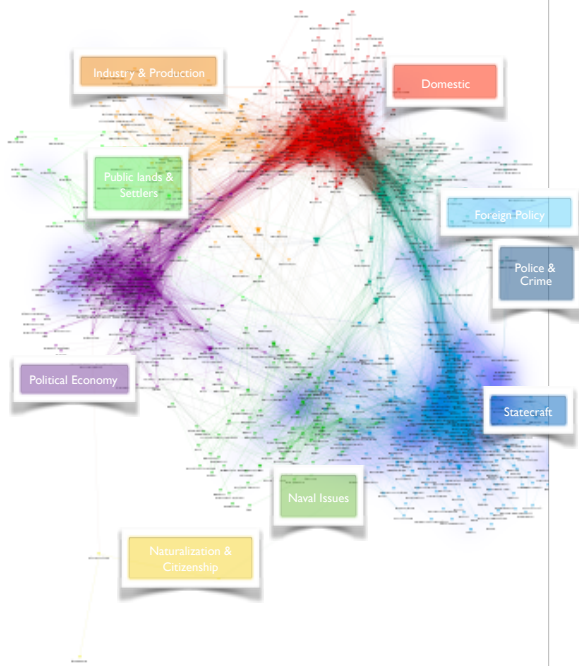


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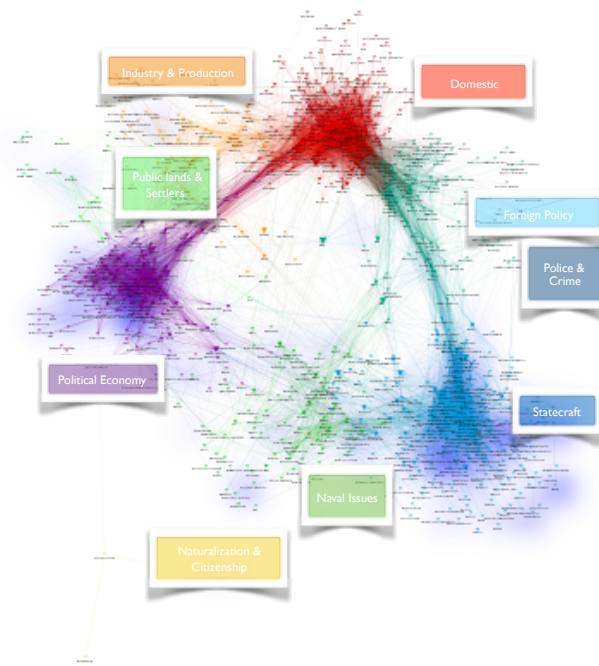
CORTEXT

Mapping individual presidents

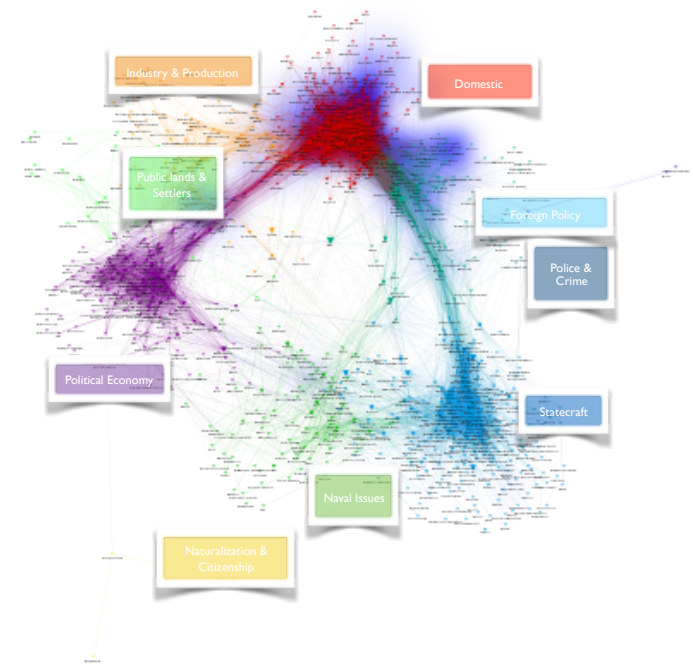
George Washington, 0-3



Abraham Lincoln, 0-3



Ronald Reagan, 0-3



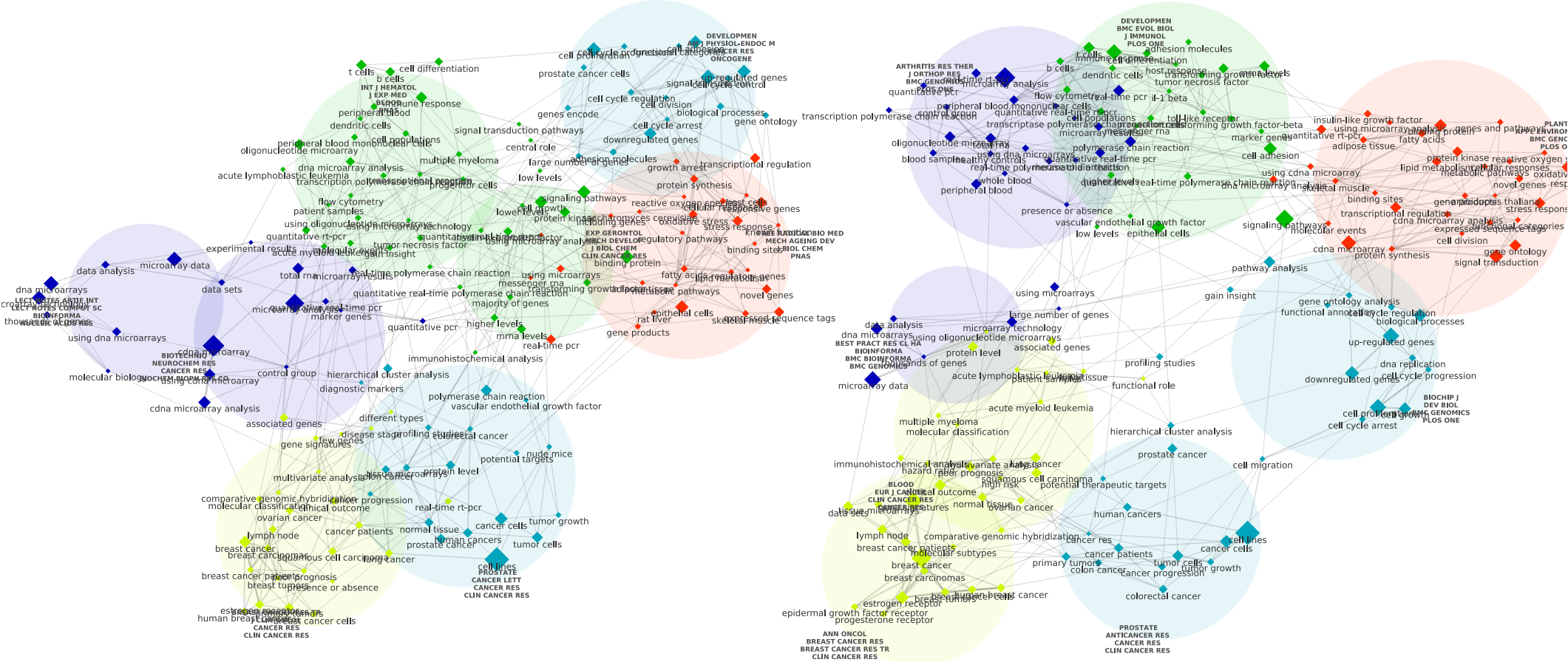
Comparing political discourses across time

Dynamical mapping

Inter-temporal matching between two maps

2001-2005

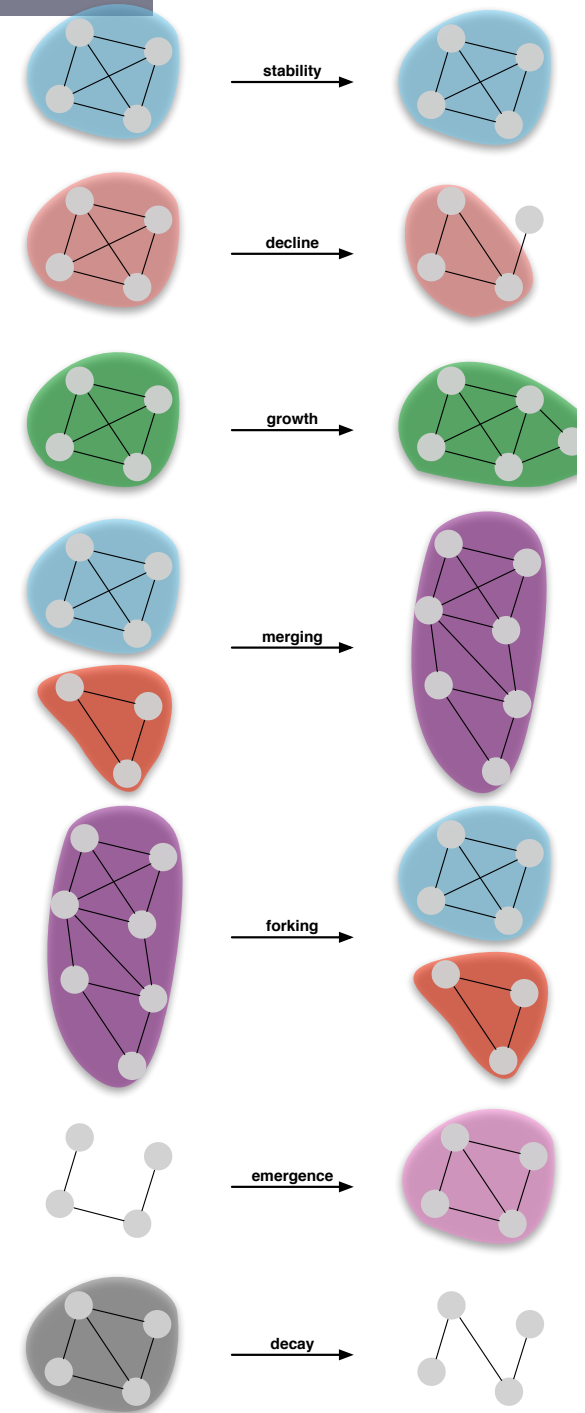
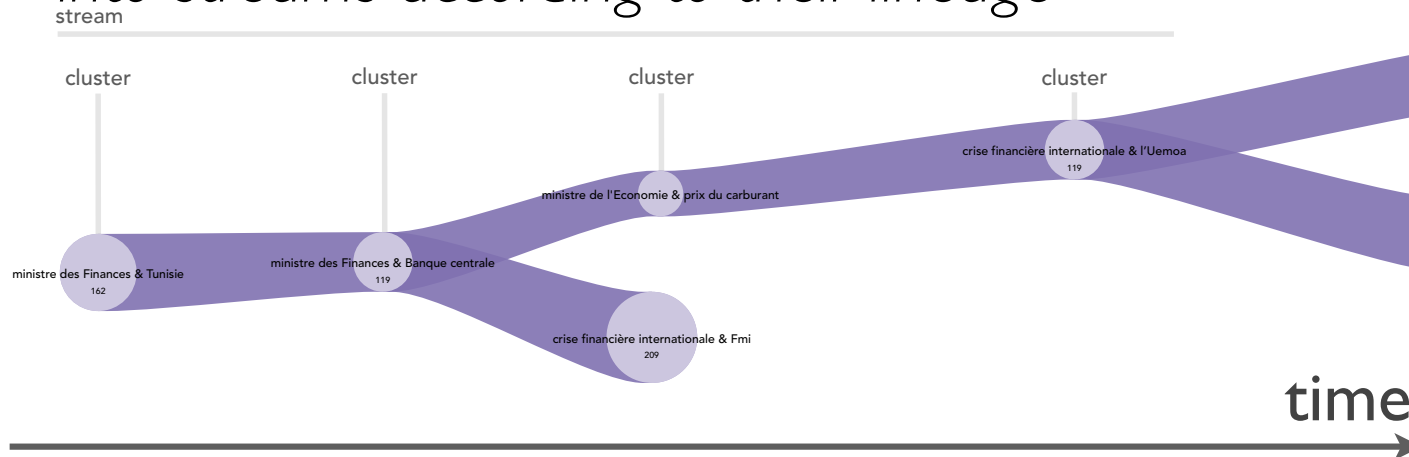
2006-2010



Stream Reconstruction

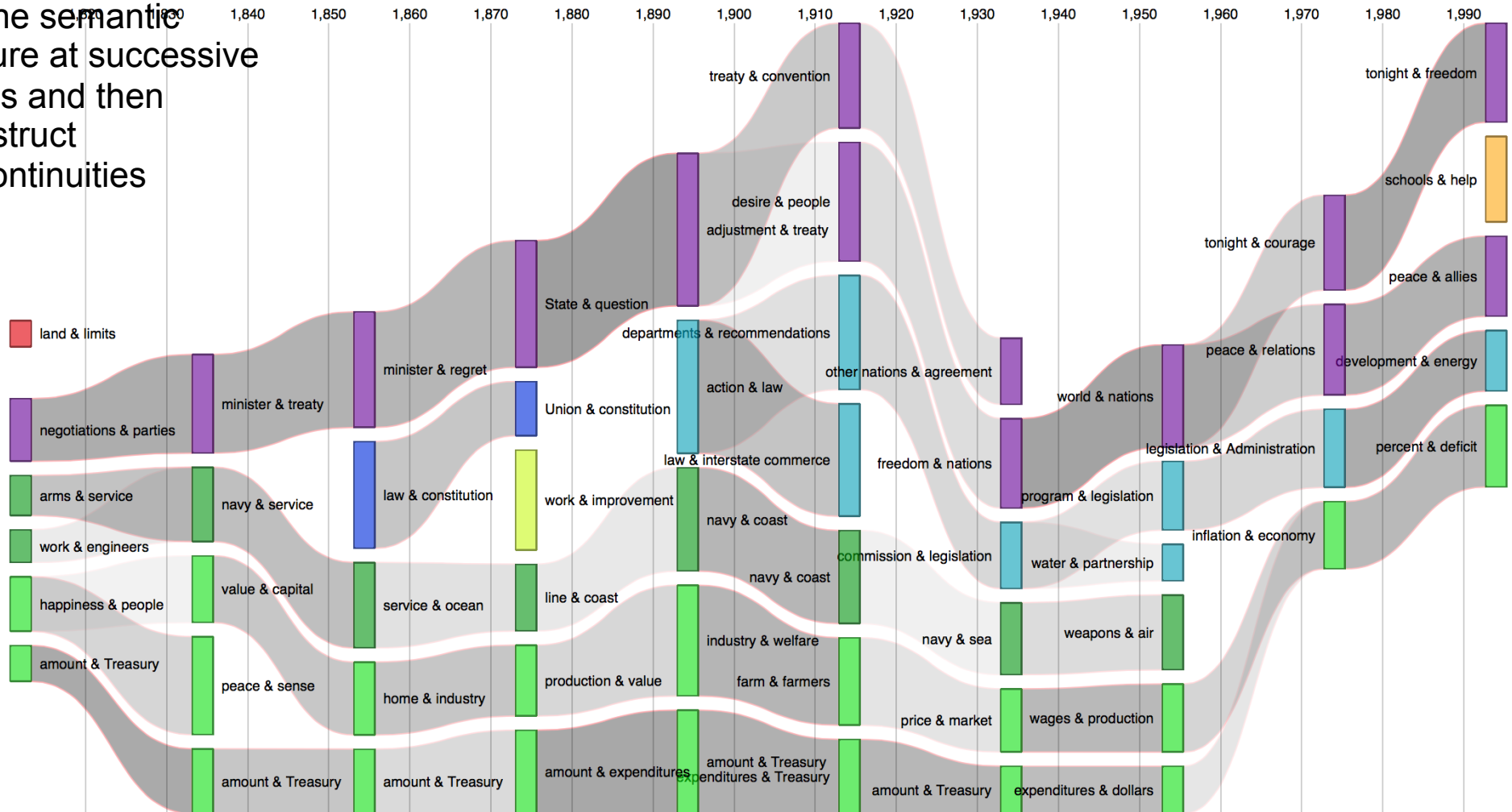
Possible events
at mesoscopic
level

Clusters are intertemporally grouped
into streams according to their lineage



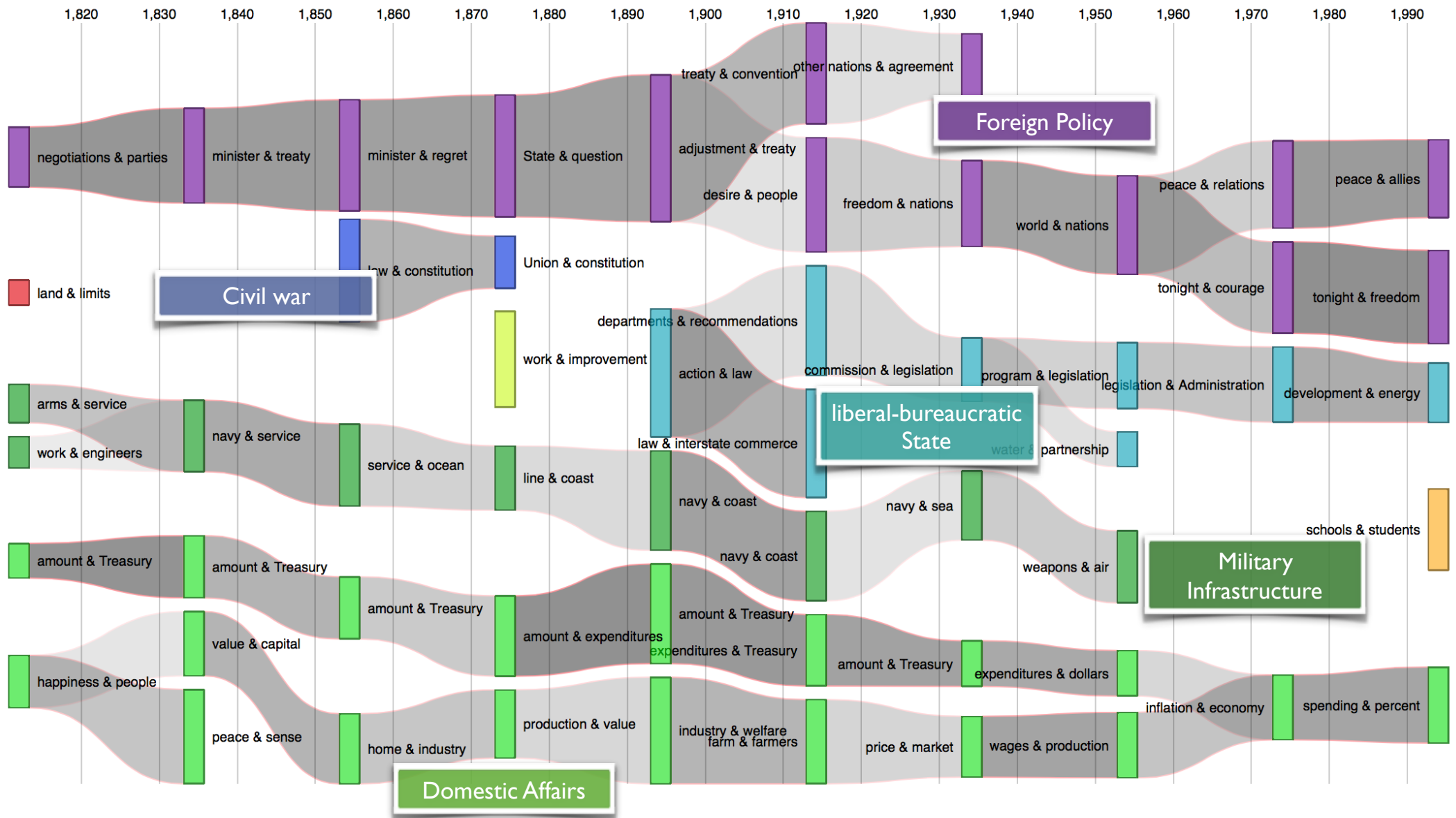
River Networks

- Capture topics with a delimited time span
- Conversely, river networks map the semantic structure at successive periods and then reconstruct (dis)continuities

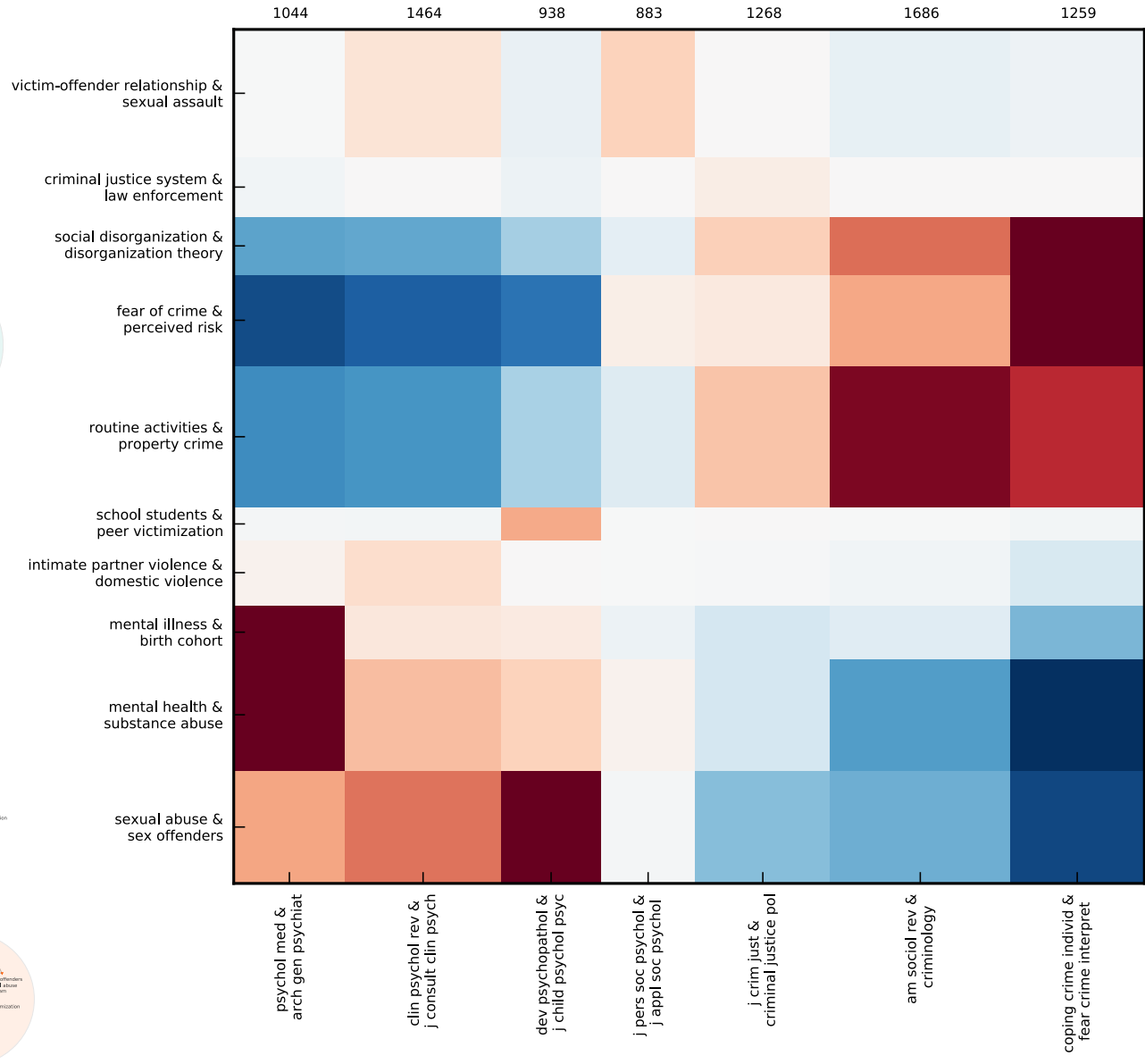
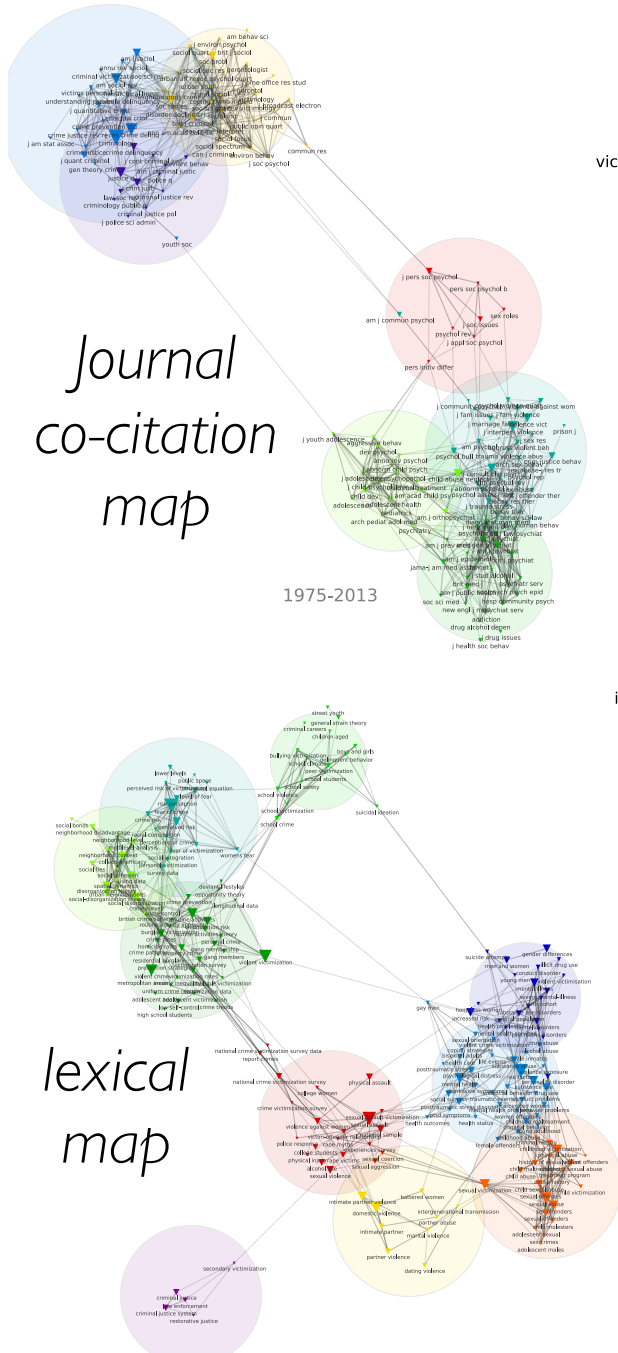


River Network

River Network



Mixing dimensions of analysis



crime science corpus