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Co-evolution in Epistemic Networks: Reconstructing Social Complex Systems - A Summary Presentation

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Co-evolution in Epistemic Networks

Reconstructing Social Complex Systems

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CREA – CNRS / Ecole Polytechnique



Presentation of the thesis — Nov 19, 2005

The Reconstruction Problem

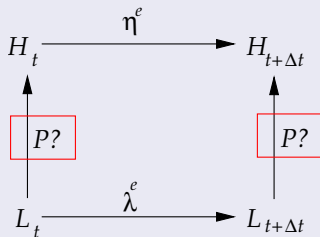
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Issues

- (i) Find P in order to deduce high-level observations H from strictly low-level phenomena L .
- (ii) Find a low-level dynamics λ that rebuilds high-level evolution η^e .

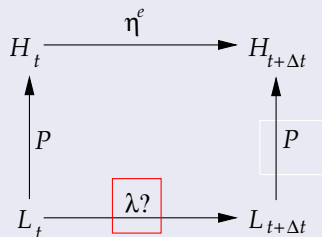


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Objectives

A socio-semantic complex system

- 1 Reproduce a hierarchic **epistemic hypergraph** of a knowledge community that fits a high-level expert-based description
- 2 Provide a low-level dynamics and a morphogenesis model that **rebuilds** the empirically observed high-level structure

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The structure of a knowledge community, and in particular its **epistemic hypergraph**, is primarily produced by the **co-evolution of agents and concepts**.

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The structure of a knowledge community, and in particular its **epistemic hypergraph**, is primarily produced by the **co-evolution of agents and concepts**.

Outline

- 1 **Epistemic communities**
 - Rationale & definitions
 - Epistemic community taxonomy and Galois lattices
 - Partial taxonomies: rebuilding history
- 2 **Micro-foundations of epistemic networks**
 - Networks
 - Towards a rebuilding model
 - Reconstruction of epistemic communities

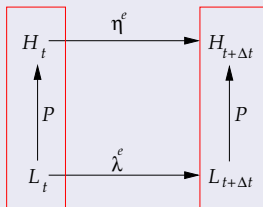
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Building taxonomies

Rationale

Describe the taxonomy of a knowledge community, in particular scientific communities, that matches high-level descriptions.



Building taxonomies

Epistemic communities

- **Epistemic Community**: group of agents sharing a common set of subjects, concepts, issues; sharing a common goal of knowledge creation — Haas (1992), Cowan et al. (2000)
- Definition here: “an epistemic community is the largest set of agents sharing a given set of concepts” – as such strongly linked with structural equivalence

Building taxonomies

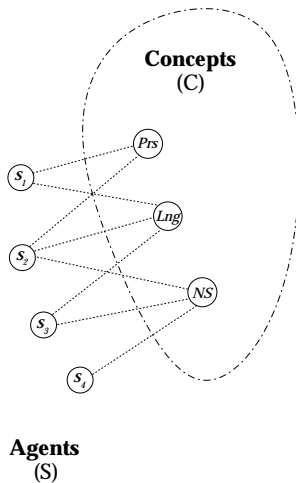
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Building taxonomies

Formal framework

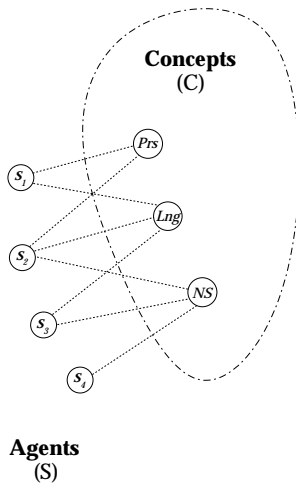
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- *Intent* S^\wedge of an agent set S : all concepts used by every agent in S
- *Extent* C^\star of a concept set C
- *Epistemic community*: the extent of a concept set C
- “ \wedge^\star ” is a *closure operation*:
 - 1 (idempotent) $(S^{\wedge^\star})^{\wedge^\star} = S^{\wedge^\star}$
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Building taxonomies

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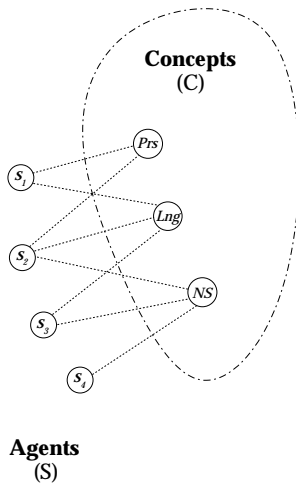
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Building taxonomies

Representing epistemic communities

- 1 structured into fields, with common concerns,
- 2 hierarchically: generalization / specialization,
- 3 overlapping.

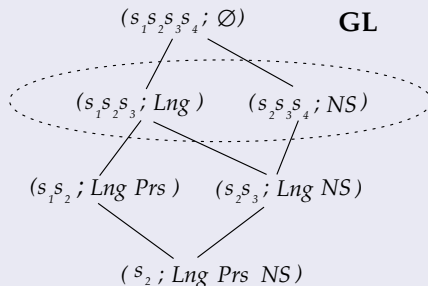
From trees to lattices

Building taxonomies

Galois lattices

$GL = \{(S^{\wedge*}, S) \mid S \subseteq \mathbf{S}\}$ is the partially-ordered set of all epistemic communities, with the partial order:

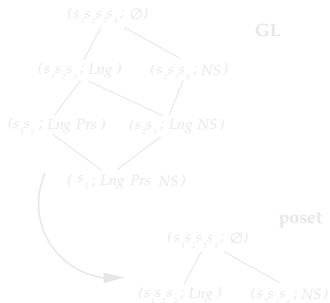
$$(X, X^{\wedge}) < (X', X'^{\wedge}) \Leftrightarrow X \subset X'$$



Managing taxonomies

Taxonomy selection & extraction

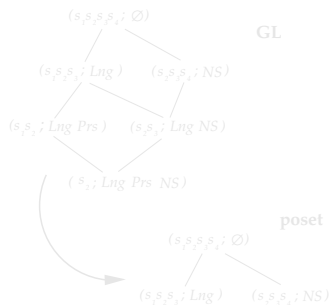
- Which ECs should we extract from the lattice?
- *Given the assumptions*, a first criterion is agent set size — Small isolated ECs could be interesting too.
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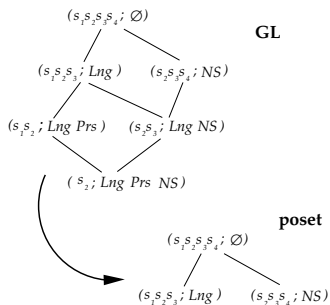
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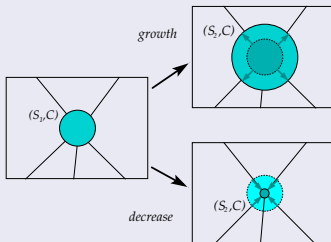
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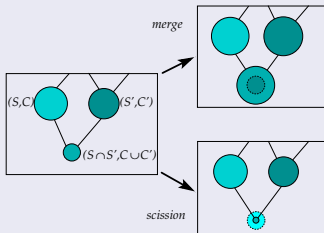
Managing taxonomies

Taxonomy evolution

- 1 Progress or decline of a field

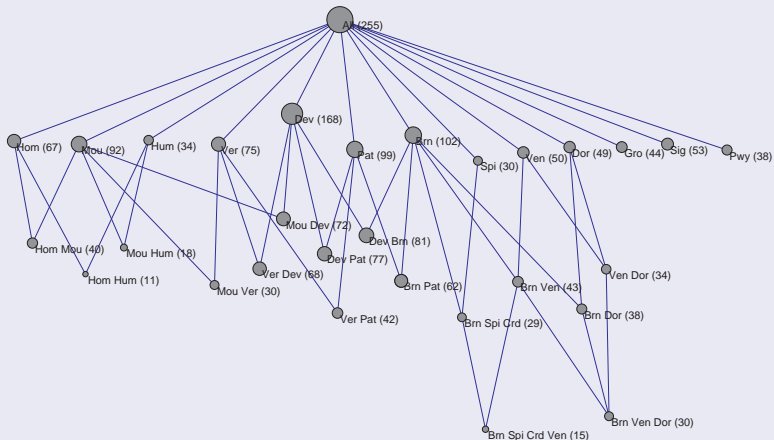


- 2 Merging or scission of a field



Empirical results

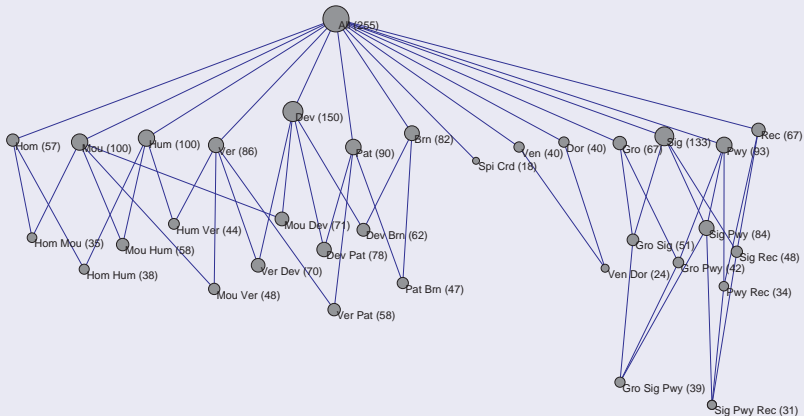
Hierarchical epistemic hypergraph 1990-1995



Partial taxonomies: rebuilding history

Empirical results

Hierarchical epistemic hypergraph 1998-2003



Empirical results

Historical description

- 1 Research on brain and spinal cord depreciated,
- 2 The community started to enquire relationships between signal, pathway, and receptors,
- 3 Mouse-related research is stable, yet significant stress on human-related topics & new relationship to homologous genes and vertebrates: growing focus on differential studies.

Matches expert-based descriptions

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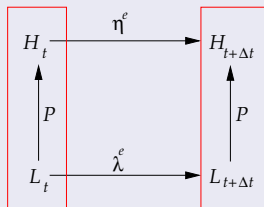
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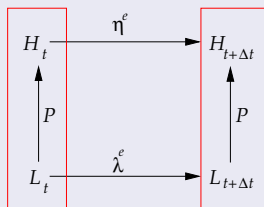
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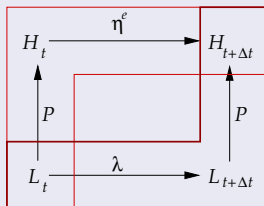


Micro-foundation

Reconstructing **high-level structure** from **low-level dynamics**:

— **reverse problem**: find λ such that $P \circ \lambda = \eta^e \circ P$.

Overview



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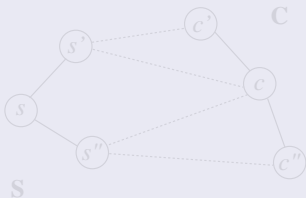
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Epistemic networks

Definitions

What is an epistemic network?

- A network of agents: $\mathcal{S}=(\mathbf{S}, E_{\mathbf{S}})$, evolving with time: $\mathcal{S}(t)$
- Semantic network: network of concepts, $\mathcal{C}=(\mathbf{C}, E_{\mathbf{C}})$
- Agents are linked to concepts they use, through \mathcal{R} .
- Three kinds of relations: $\mathcal{R}^{\mathbf{S}}$, $\mathcal{R}^{\mathbf{C}}$ and \mathcal{R} :

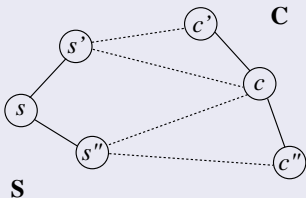


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Network morphogenesis

A brief survey

- 1 *Early times*: Erdos-Renyi, until unsatisfying power-law degree distribution and other statistical parameters
- 2 Pioneering models rebuild clustering, and **degree distribution** (preferential attachment (PA), network growth)
- 3 Since then and until now: models introducing various kinds of PA to rebuild diverse statistical parameters
- 4 But even with credible hypotheses, rare empirical validations, yet needed for realistic morphogenesis models

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High-level features

Degree distributions

- Four degree distributions: social, semantic, socio-semantic (from agents, from concepts)
- Power-law tail, log-normal fit

Clustering structure

High clustering both for monopartite coefficients and bipartite coefficients

Epistemic community structure

Many large ECs, particular distribution of EC sizes.

Suggest empirically credible low-level dynamics

Measuring interaction behavior

- Measure the interaction behavior of agents
- Have an essential preference f for nodes of kind m : $P(L|m)$
 → we may estimate f through $\hat{f}(m) = \frac{\nu(m)}{P(m)}$
- Check correlations between parameters: $\hat{c}_{m'}(m) = \frac{P(m|m')}{P(m)}$

Event-based modeling

Distinguish activity from attractivity: rich-get-richer or rich-work-harder?

Activity and interactivity: $f(m) = a(m)\iota(m)$

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Event-based low-level dynamics

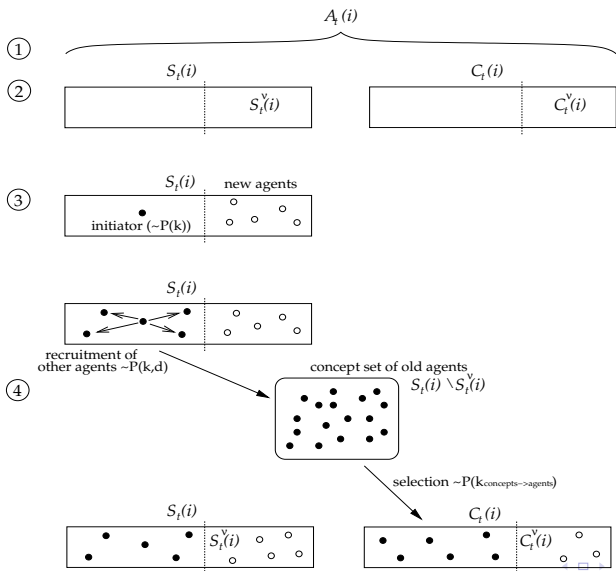
Choice of agents

Geometric distribution of agents, tri-modal distribution for newbies

Choice of concepts

Geometric distribution of concepts, uni-modal distribution of novel concepts

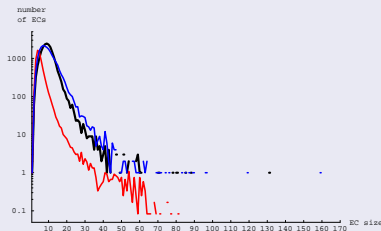
Model design



Reconstruction

Epistemic communities are produced by the co-evolution of agents and concepts

Degree distributions, clustering structure, epistemic structure are **reconstructed**.

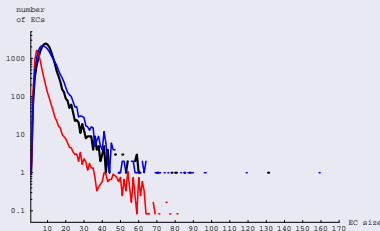


- Reconstruct high-level statistical parameters meaningful for epistemic networks
- Respecting low-level dynamics: **descriptive** rather than normative

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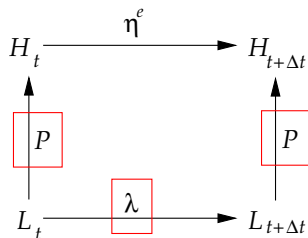
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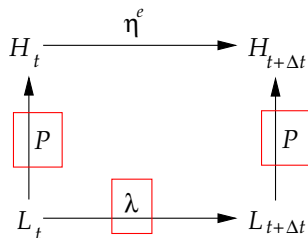
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Conclusion



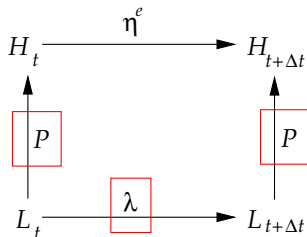
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Appraising levels

Relationships between different levels

- Dualism, reductionism ?
- **Emergentism**: low-level phenomena cause high-level phenomena, yet in turn not necessarily reduceable to low-level phenomena.
- Is it ok that a lower level creates a higher level, then the higher level in turn influences the lower level?

○ **Emergentism**: different levels of analysis to a same process

○ **Reductionism**: high level is reducible to low level

○ **Dualism**: high level and low level are independent

○ **Interactionism**: high level and low level are interdependent

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- Is it ok that a lower level creates a higher level, then the higher level in turn influences the lower level?
 - Rather, different modes of access to a same process: dual-mode of operational access.
"There may be emergence without emergent properties. Not asymmetric emergence of high-level properties out of basic properties, but symmetrical co-emergence of microscopic low-level features and high level behavior" (Bitbol, 2005)

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Levels as observations

Each level is an observation instrument (a *phenomenon*), and may provide information about some other observation gained through other instruments.

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“Observationism”

- 1 no substantial reality of levels
- 2 no reciprocal causation, but informational links
- 3 some phenomena cannot be rebuild from some given lower-level decriptions

Modeling links between levels

Reconstruction

- “Observationism” induces simply informational dependence between both levels: $\lambda(L|H)$, $\eta(L|H)$
- Thus, **reconstruction failure** may also come from ill-defined levels: **not yielding enough information about the given phenomenon** (e.g. learning & glial cells; concepts in addition to simple social interactions between agents)
- Reductionism only works when H is *fully deduceable*, not *reduceable*, from L .

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Stigmergence

Co-evolutionary framework

- Additionally, this viewpoint is not contradictory with some sort of causal retroaction: action of a group of neurons onto another group of neurons, agents creating & modifying their environment which in turn “acts upon them”: **stigmergence**.
- No downward causation either, simply influence of already existing environmental artifacts
- In our case, there is a co-evolution between semantic and social networks.

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