

Starting a Fire with Twigs: Influence of Encapsulation Relations on Bottom-up Dynamics on Hypergraphs

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In collaboration with Prof.
Renaud Lambiotte



MATHEMATICAL
INSTITUTE

Multi-way/higher-order interactions

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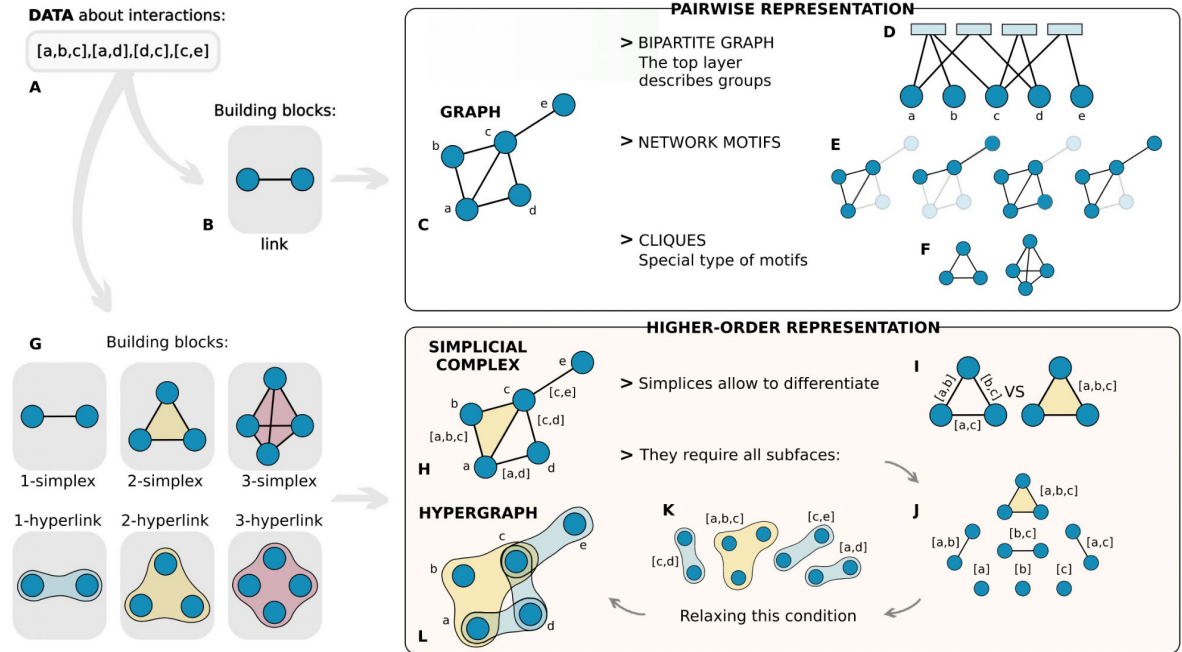


Fig. 1. Representations of higher-order interactions. A set of interactions of heterogeneous order (A) can be represented using only pairwise

Multi-way/higher-order interactions

Interactions occur between sets of nodes of arbitrary size.

F. Battiston, G. Cencetti, I. Iacopini et al. / Physics Reports 874 (2020) 1–92

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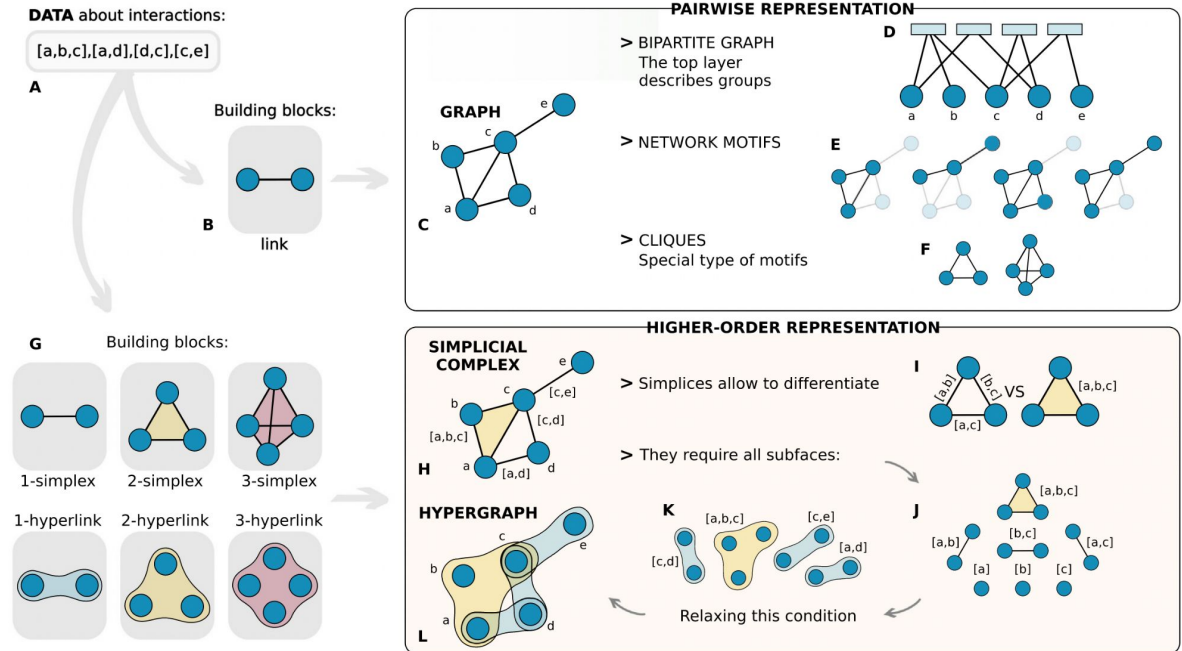


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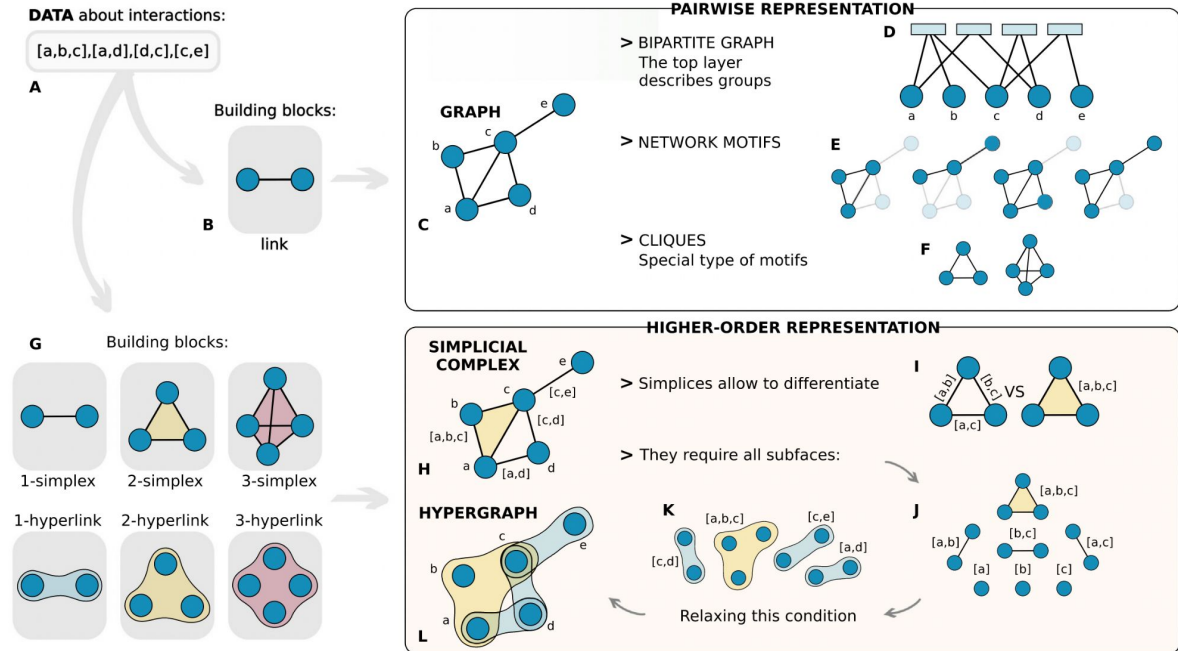


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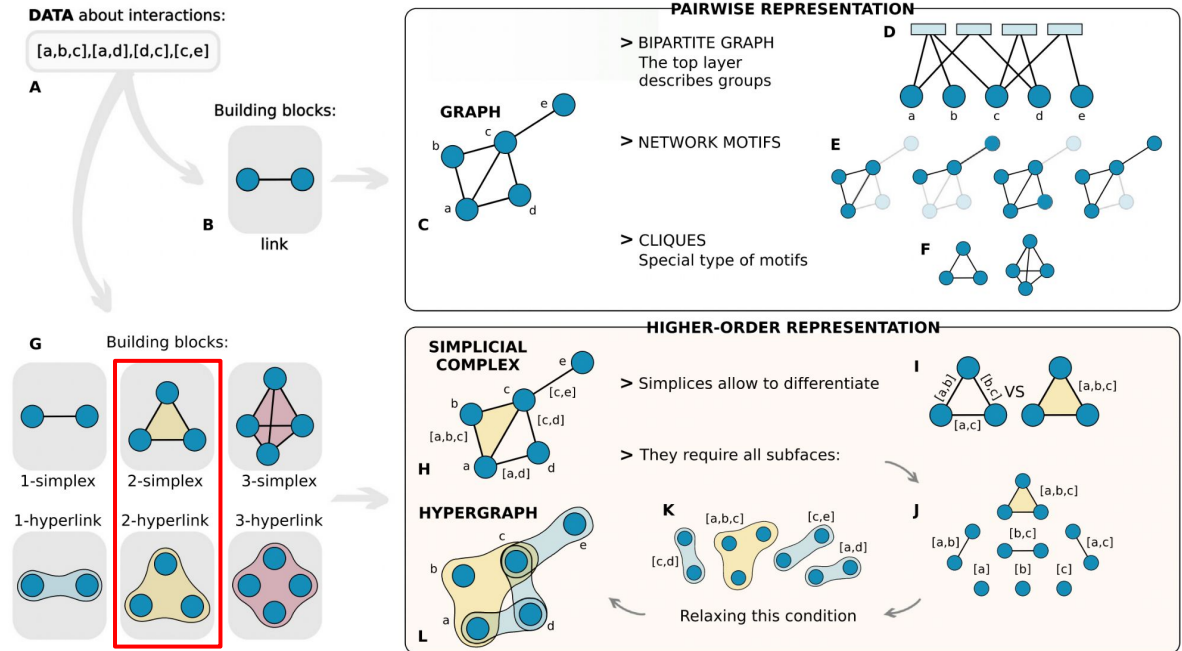


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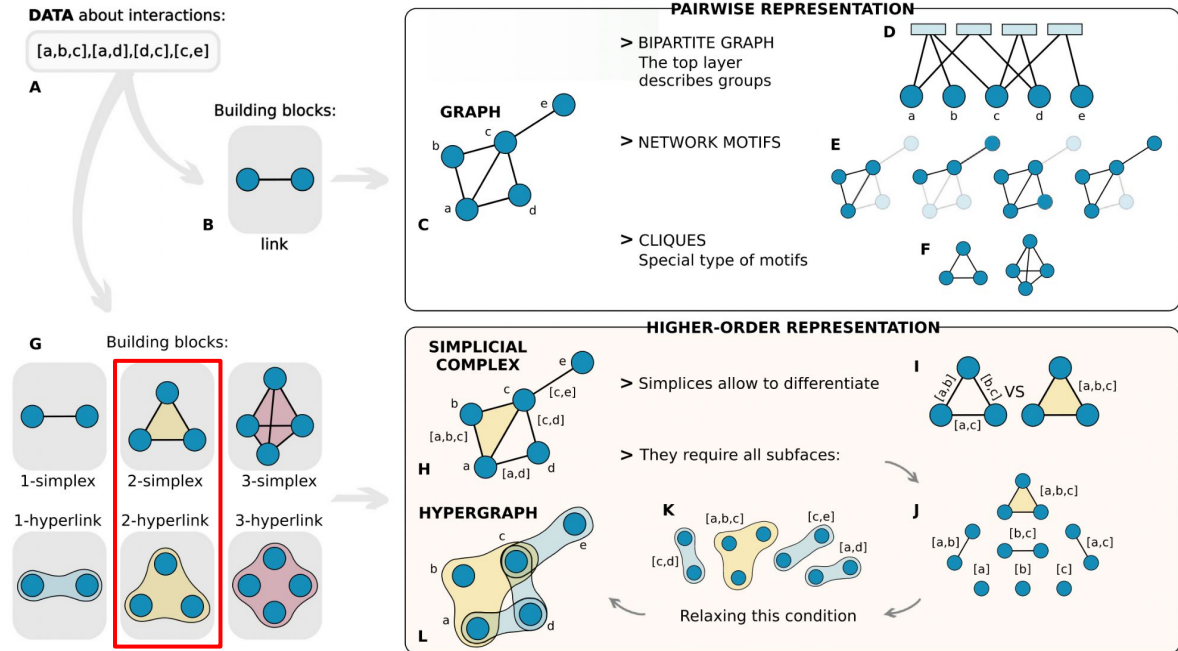


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But this is not often true!

Our goal: Characterize hypergraph structure based on **encapsulation**, then study a contagion process driven by encapsulation where activation occurs **on hyperedges**.

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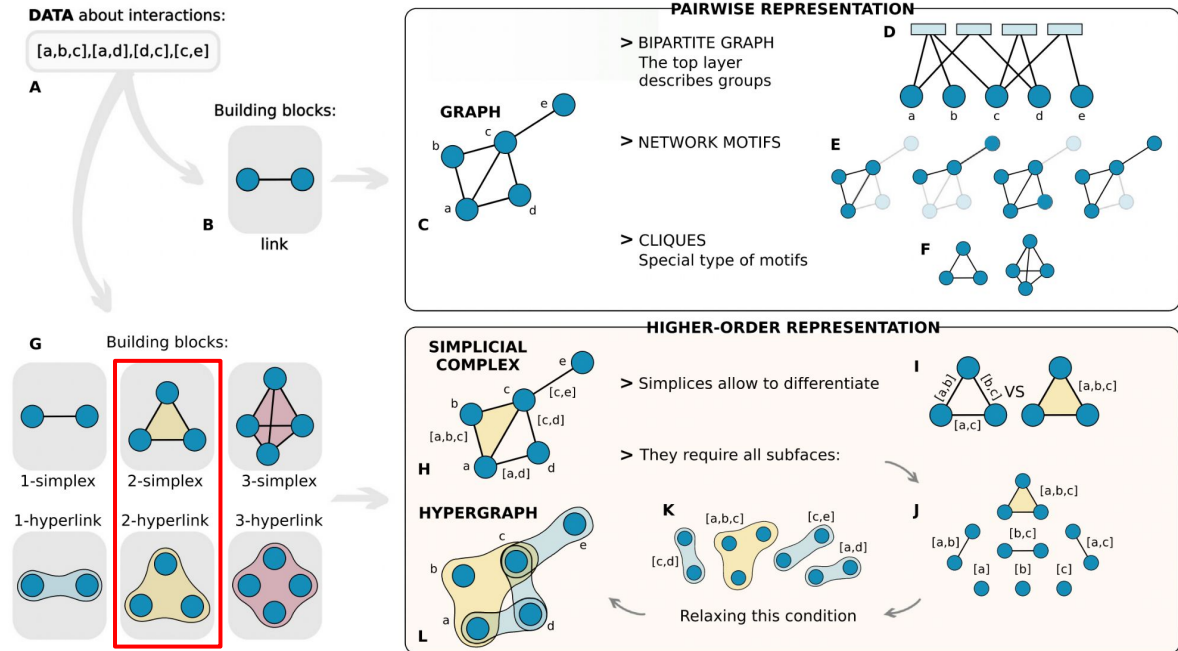
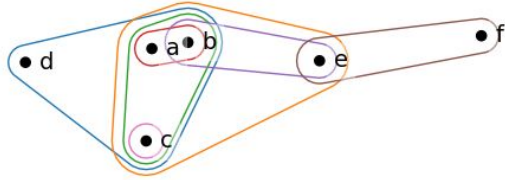


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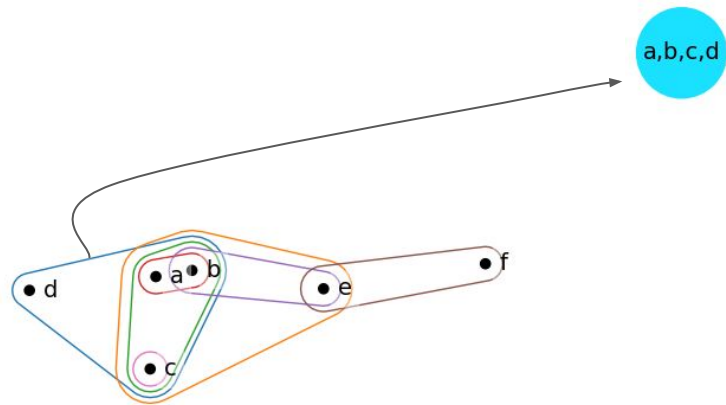
Encapsulation, Overlap, and Line Graphs of Hypergraphs

Hypergraph (Euler diagram)



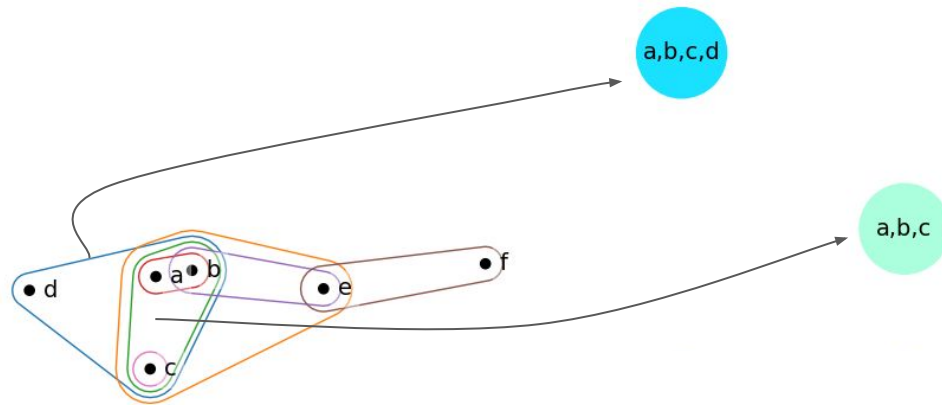
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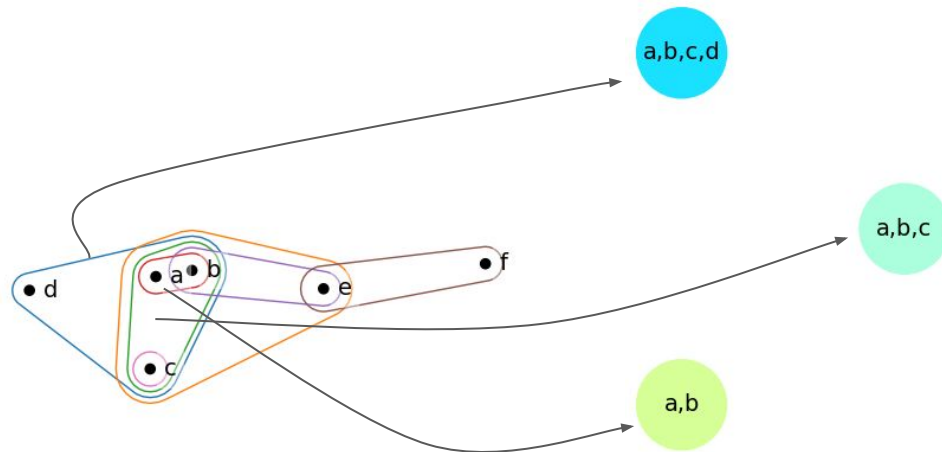
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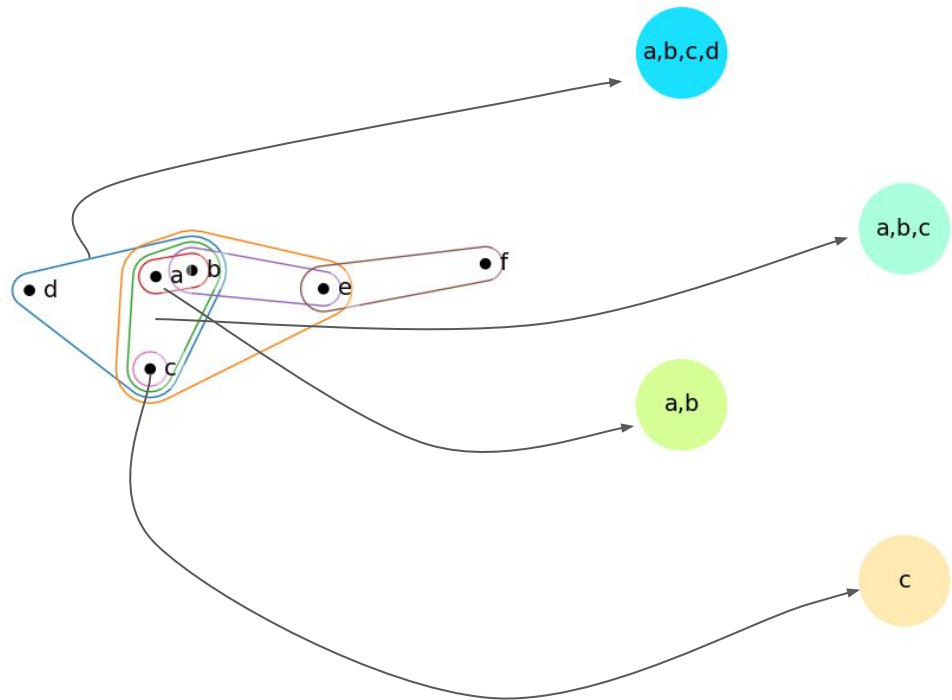
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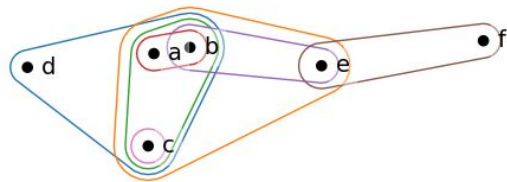
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k=4



k=3



k=2

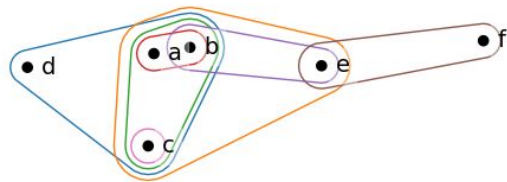


k=1

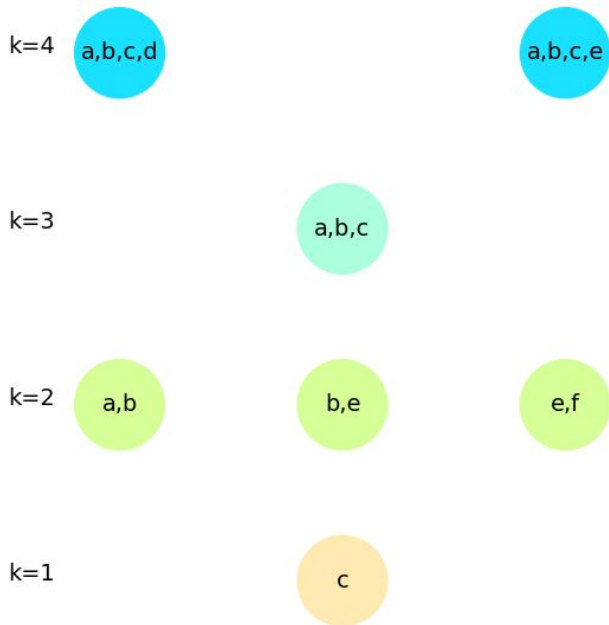


Encapsulation, Overlap, and Line Graphs of Hypergraphs

Hypergraph (Euler diagram)

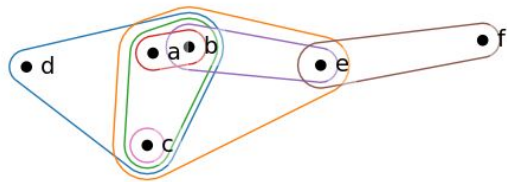


Construct the **line graph** of a hypergraph by treating hyperedges as nodes and drawing edges between them based on some relation.



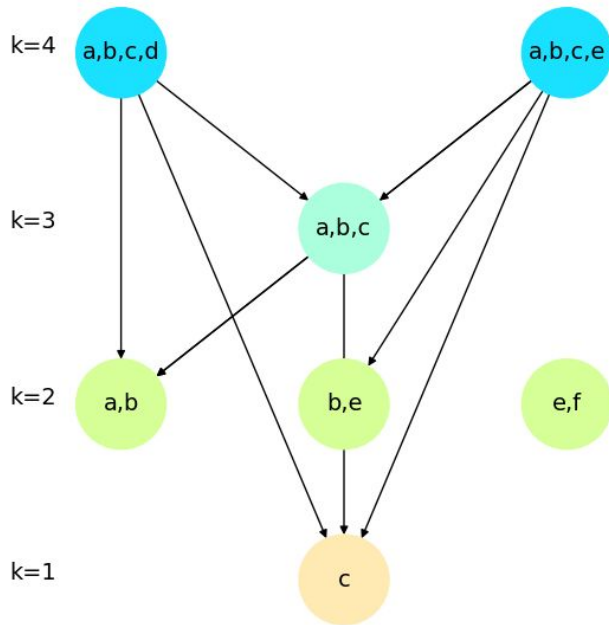
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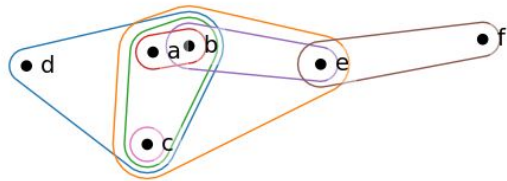
Subset (encapsulation)



Encapsulation DAG

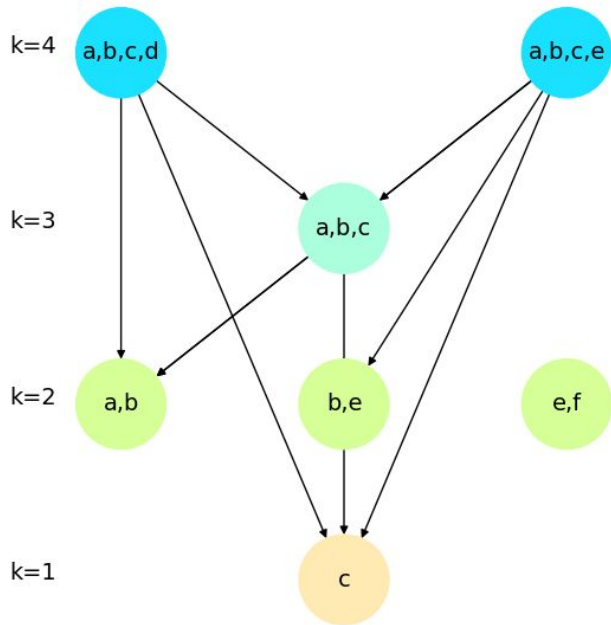
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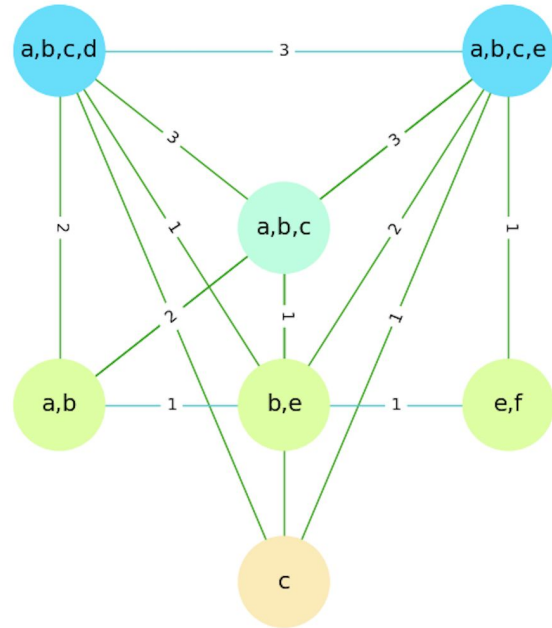
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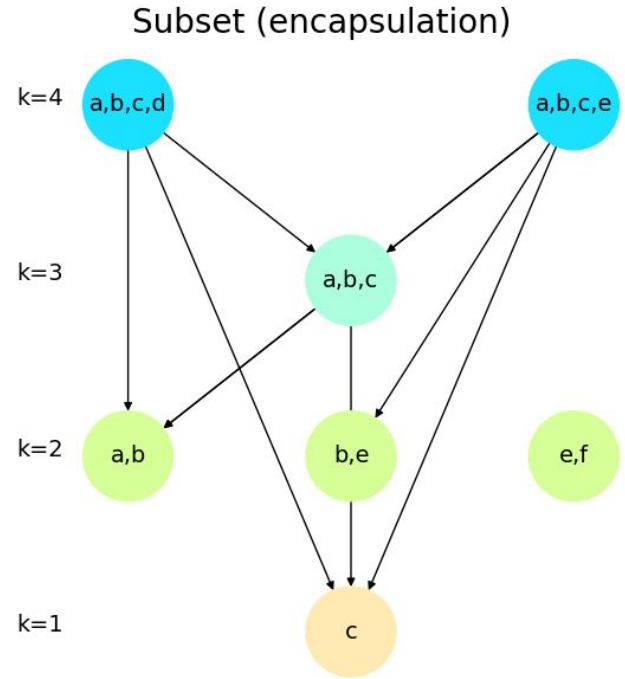
Intersection (overlap)



Overlap Line Graph

Why encapsulation DAGs?

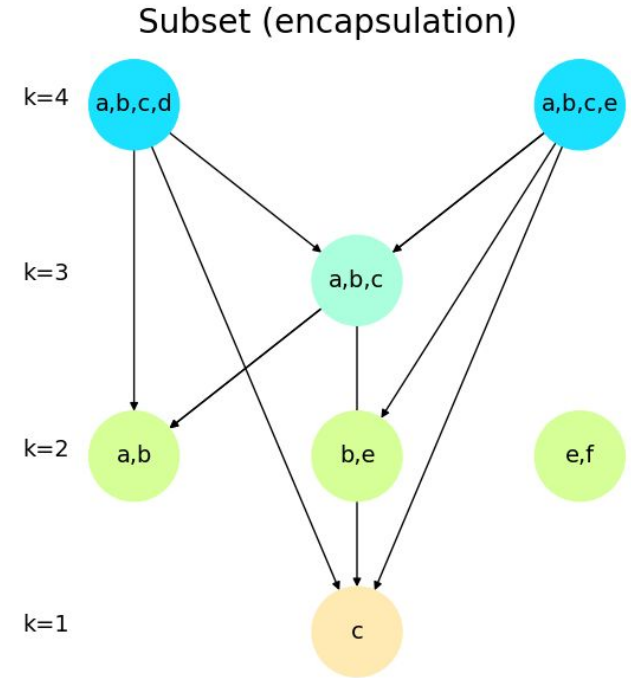
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Maximal edges in encapsulation DAG corresponds to “simplex assumption”

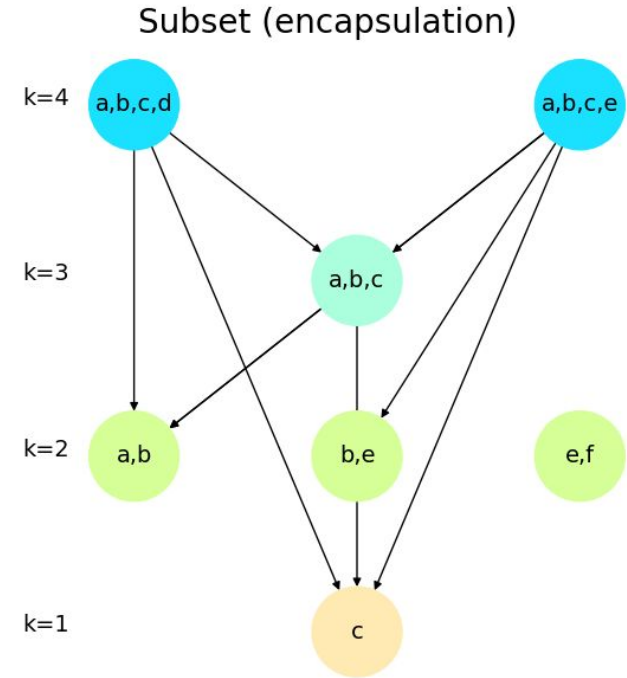


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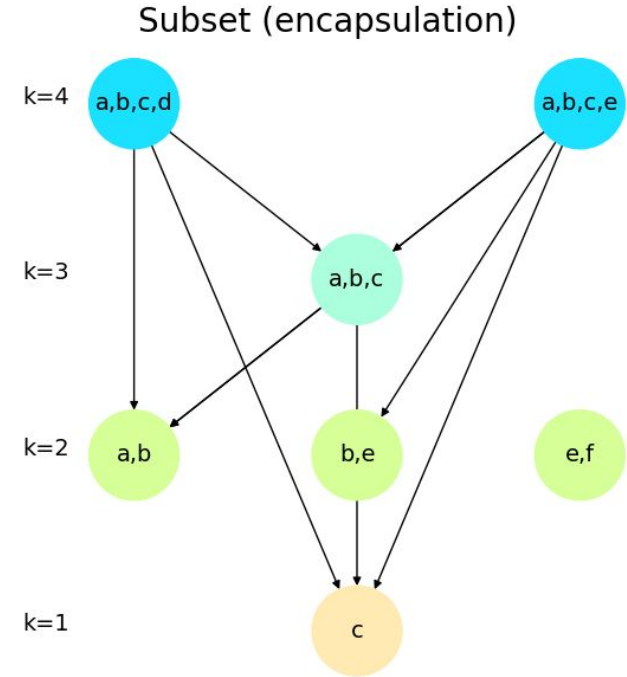
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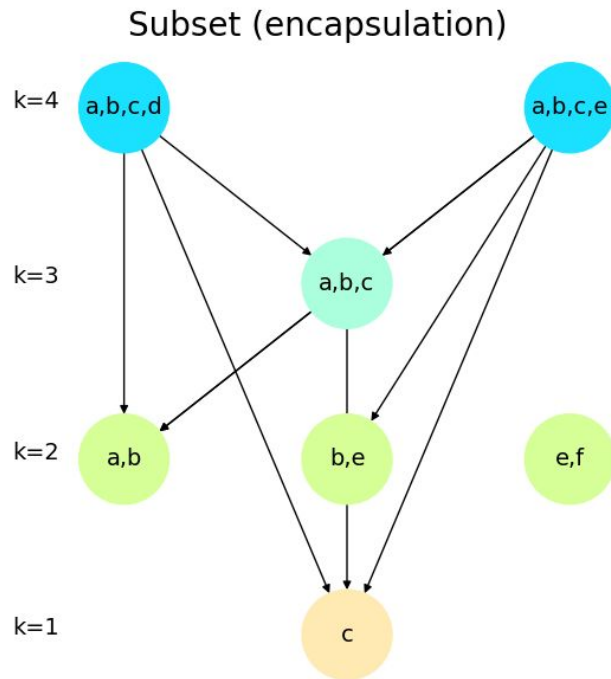
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Maximally assured

arXiv > cs > arXiv:2307.04613

Computer Science > Social and Information Networks

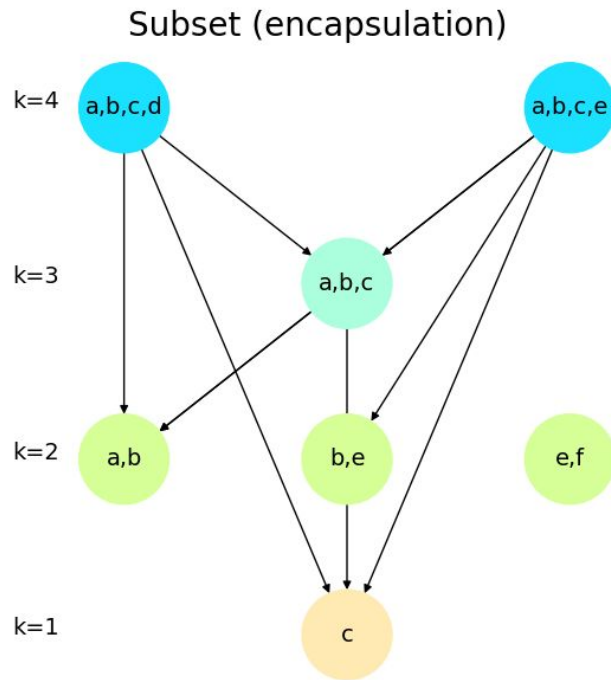
[Submitted on 10 Jul 2023]

In- an hyper **Encapsulation Structure and Dynamics in Hypergraphs**

Timothy LaRock, Renaud Lambiotte

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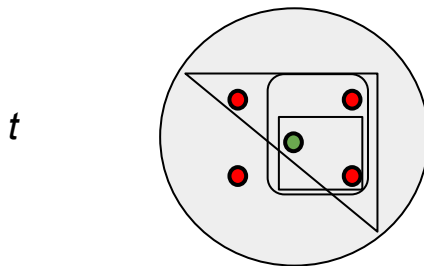
Dynamics on hypergraphs

Node-based Threshold Dynamics

A **node** becomes active if it participates in a hyperedge where more than a threshold of **nodes** become active.

Dynamics on hypergraphs

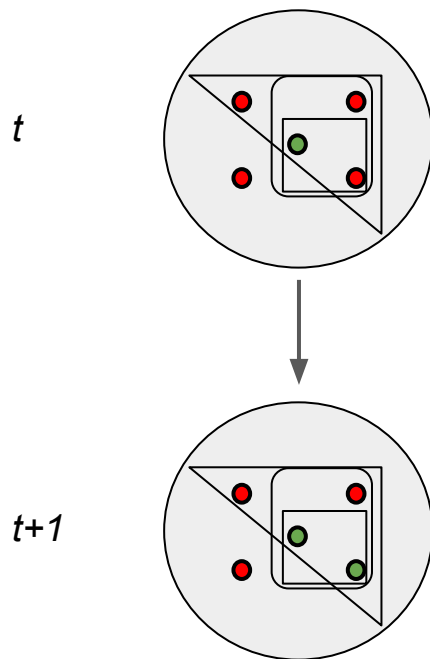
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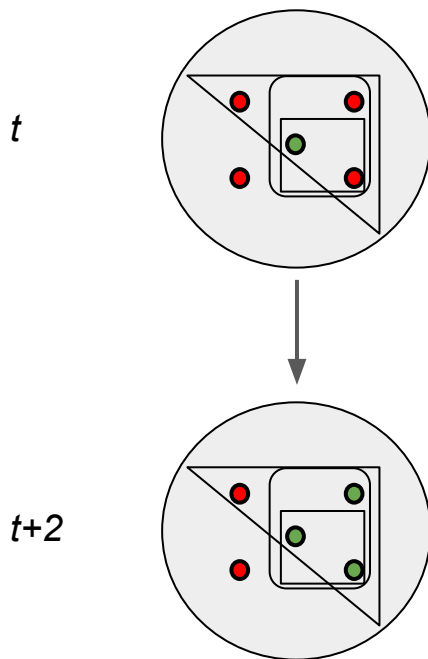
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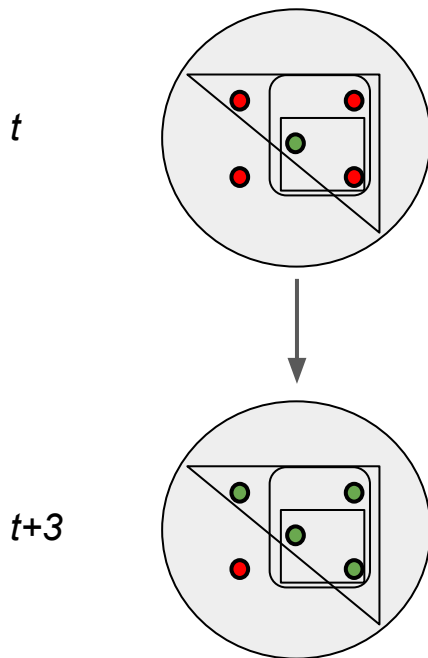
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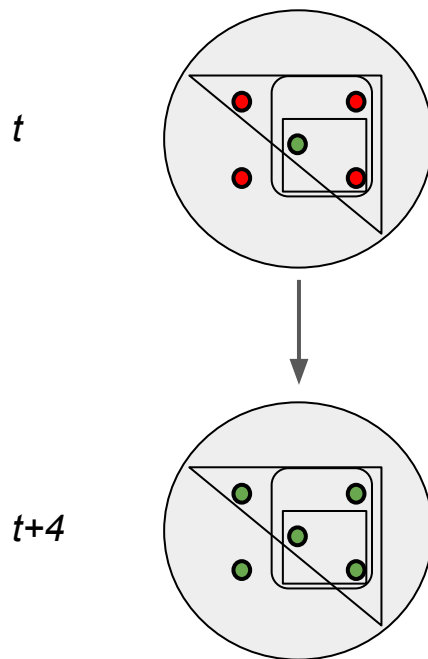
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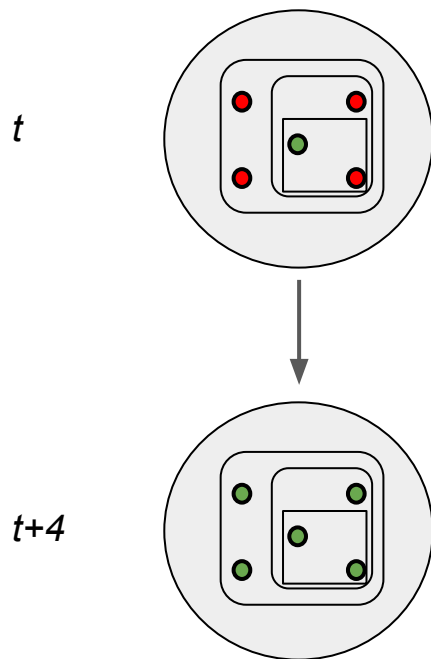
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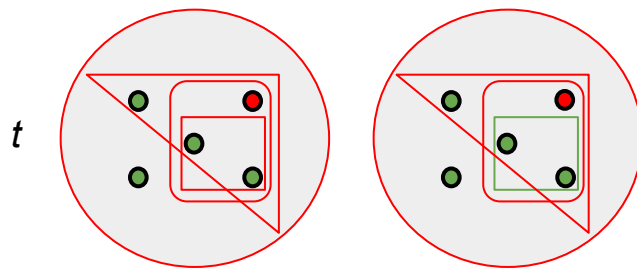
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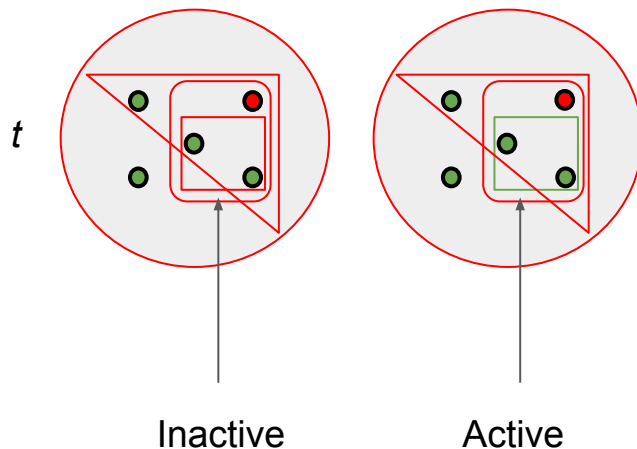


A **hyperedge** becomes active if more than a threshold of its **encapsulated hyperedges** becomes active.

Dynamics on hypergraphs

Nodes **and** edges in binary state,
active or inactive

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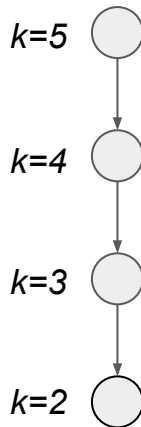


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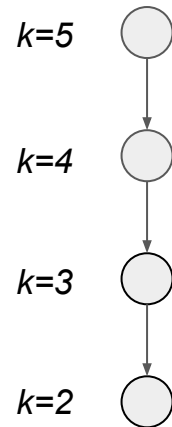
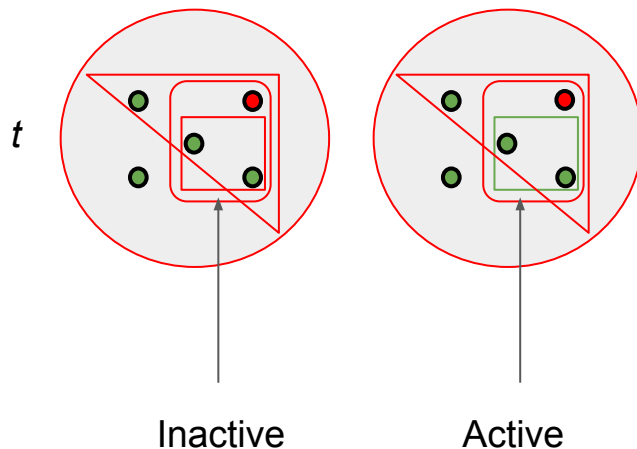
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Activation flows **upward** from **smallest to largest**



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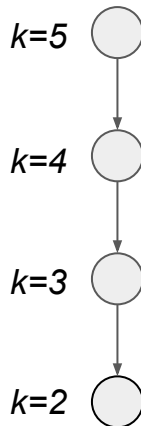
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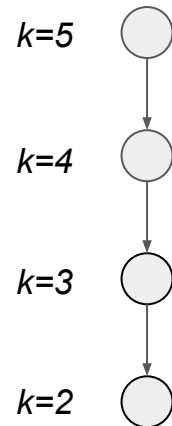
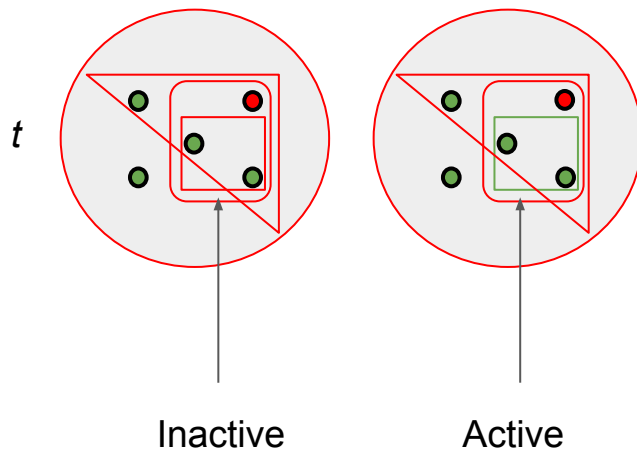
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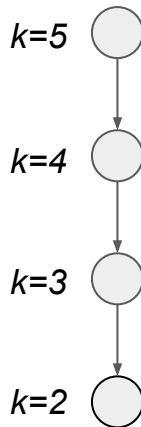
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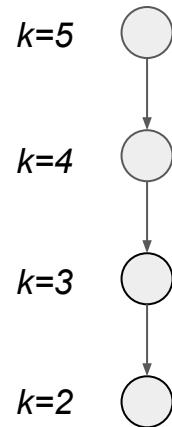
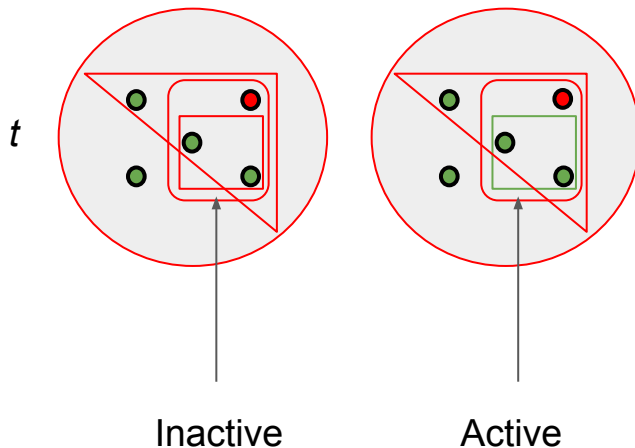
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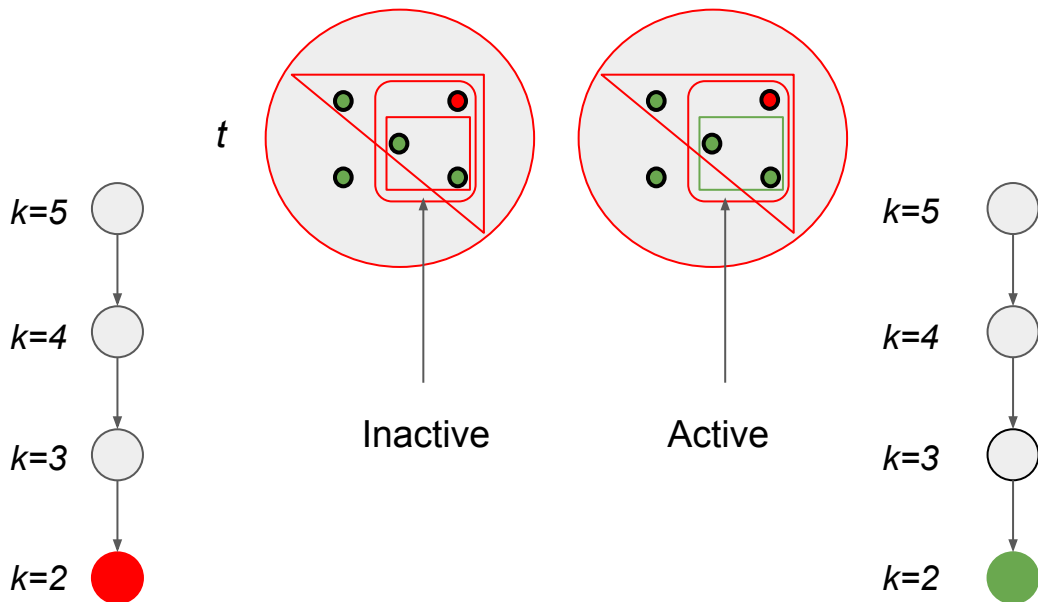
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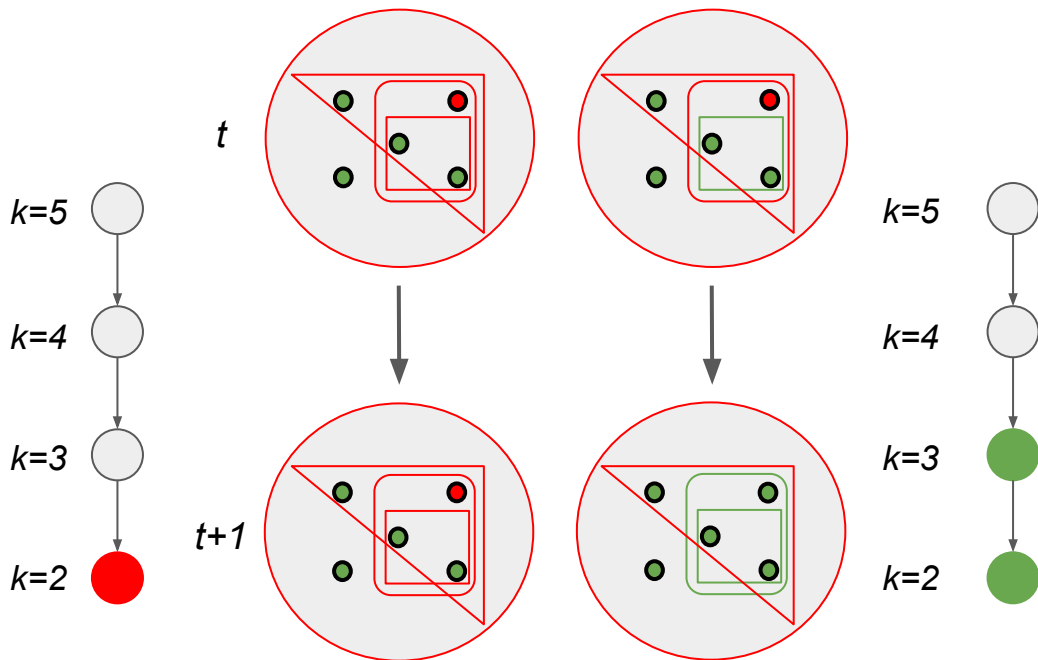
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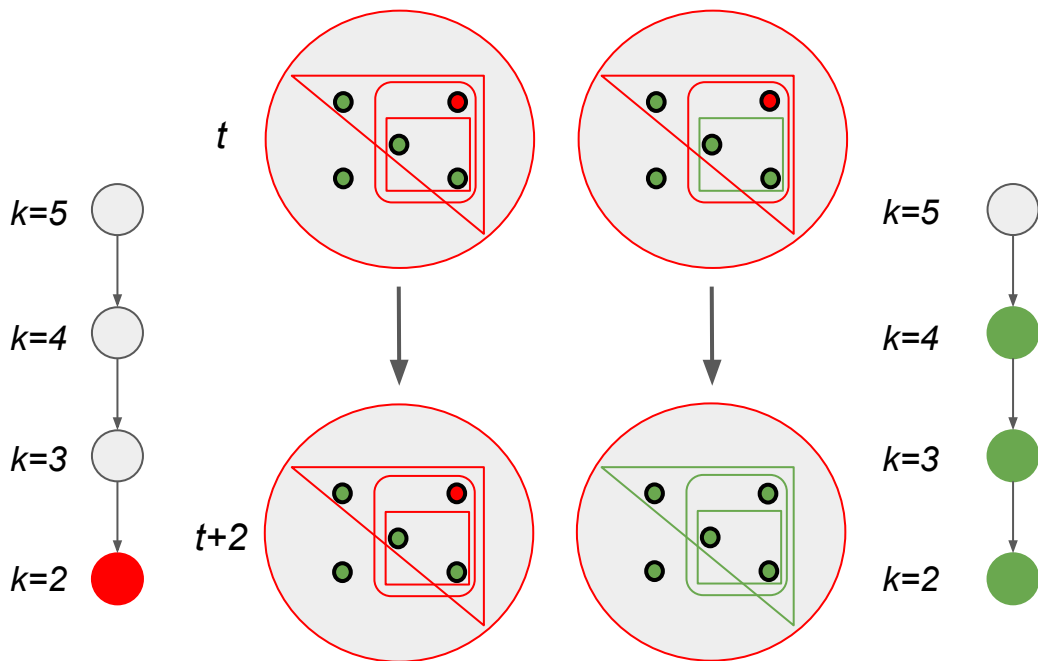
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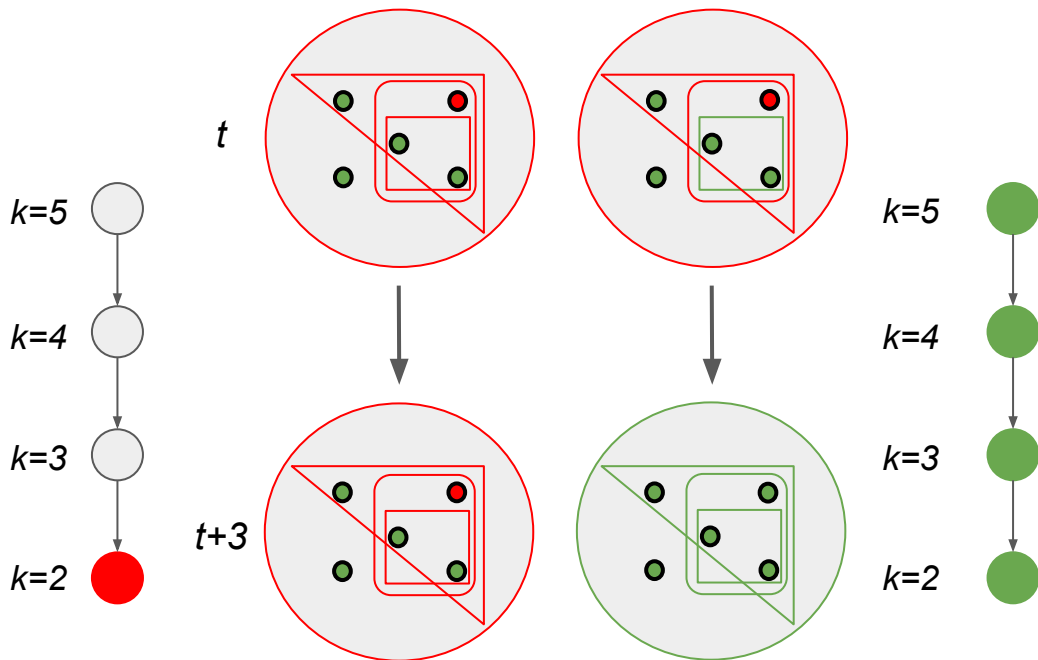
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(Campfire) Dynamics on hypergraphs

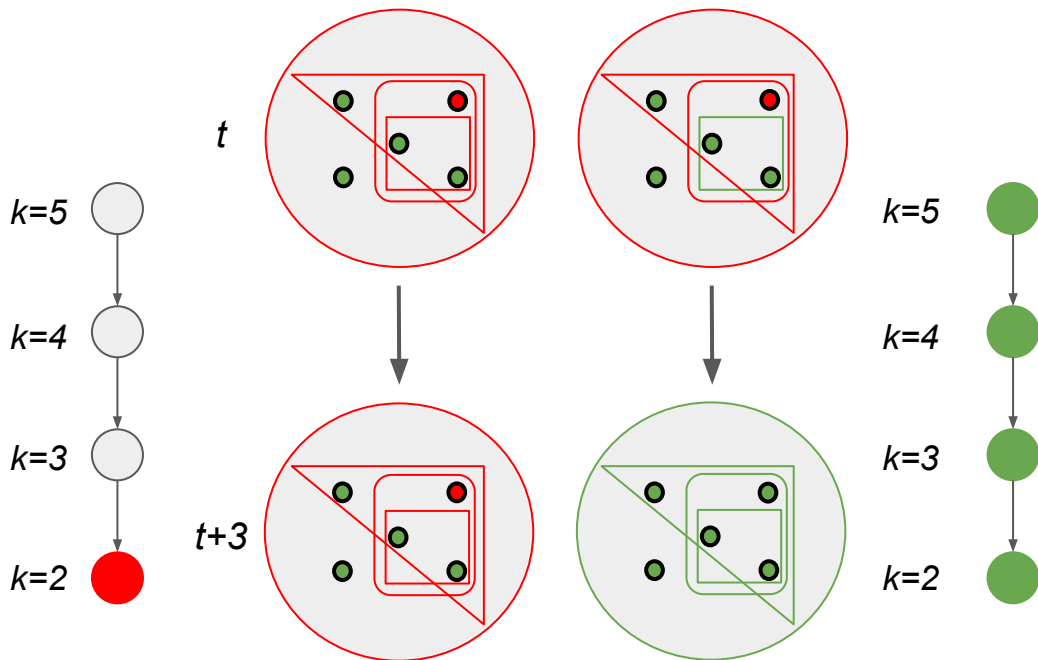
Analogy to lighting a campfire: the smallest fuel must be lit before the logs can catch on fire!



Image: <https://www.pelican.com/us/en/discover/pelican-flyer/post/how-to-start-a-campfire/>

Correspond to a type of **coordinated behavior** where nodes not only share goals/opinion/information, but coordinate to pass to other groups they are embedded within.

Encapsulation Dynamics



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Random Nested Hypergraph Model

Idea: Start from a fully encapsulated hypergraph (simplicial complex), then selectively rewire hyperedges to destroy encapsulation relationships

Parameters:

N : Number of nodes

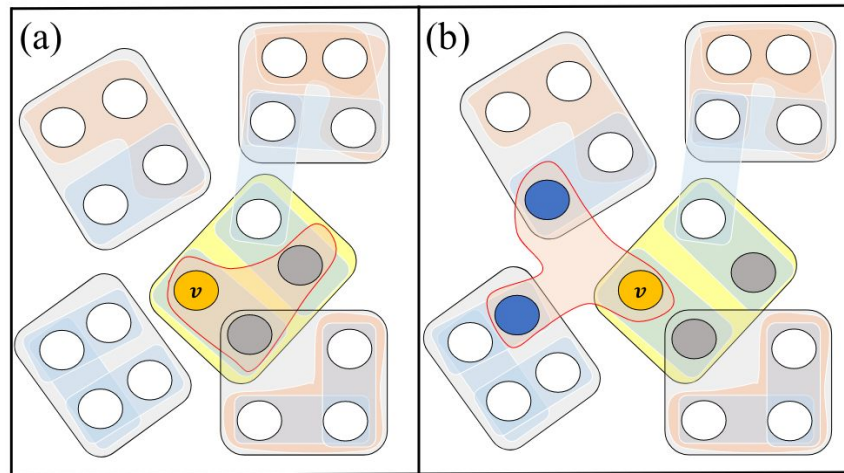
s_m : Maximum size hyperedge

H_s : Number of hyperedges of size s_m

ϵ_s : 1 minus probability of rewiring hyperedge of size s

Procedure:

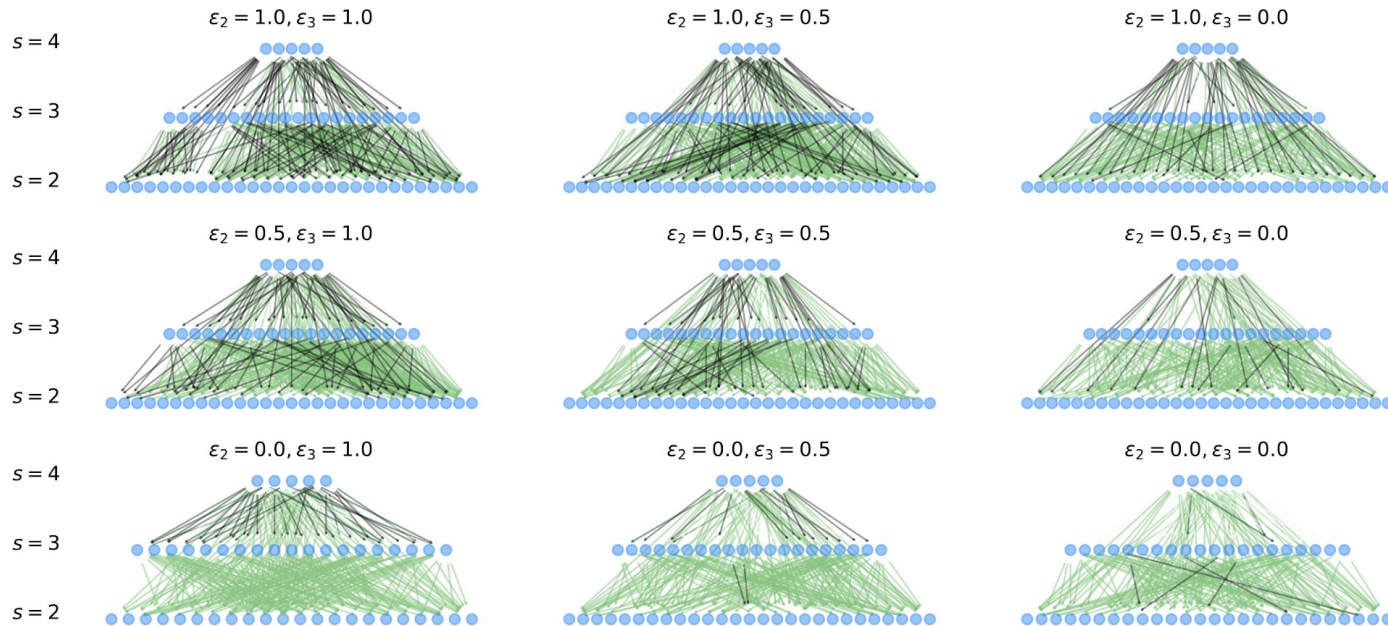
1. Generate random hyperedges of size s_m and all of their subhyperedges (power set)
2. For each hyperedge of size $s < s_m$, rewire with probability ϵ_s



Rewiring works by choosing a pivot node to keep, then randomizing other nodes by choosing nodes that are not in supersets of the original hyperedge.

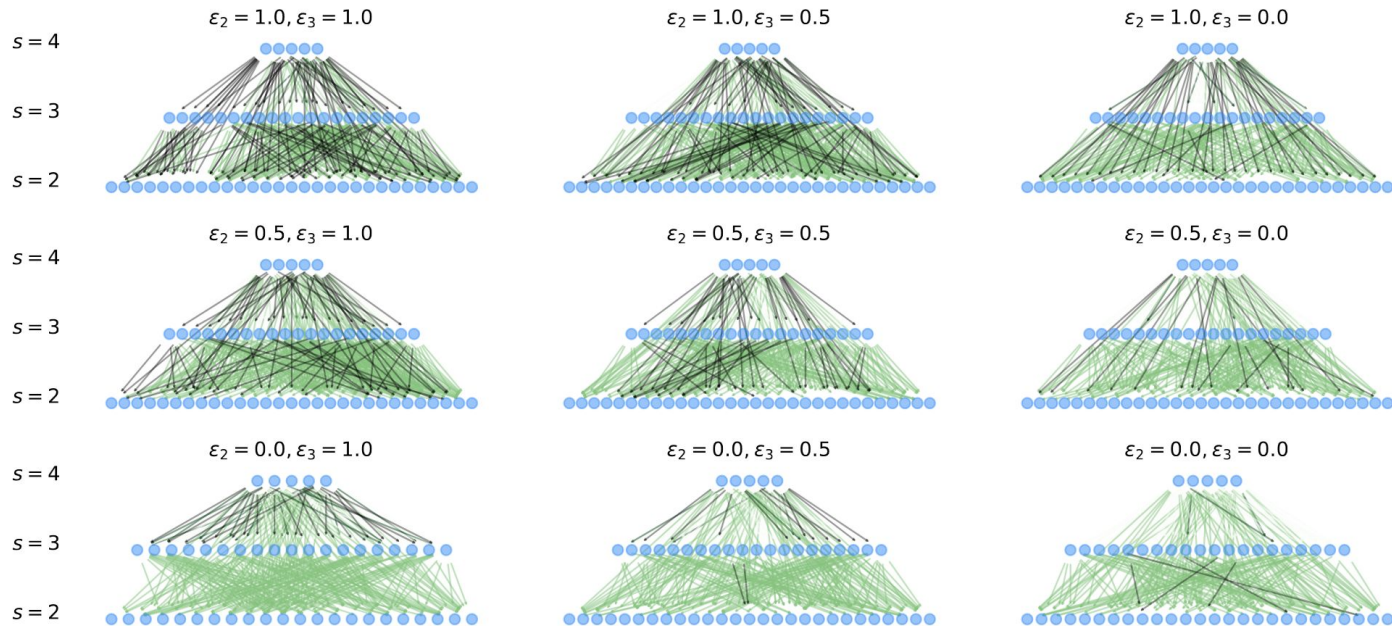
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Overlap Structures for Varying ε_s



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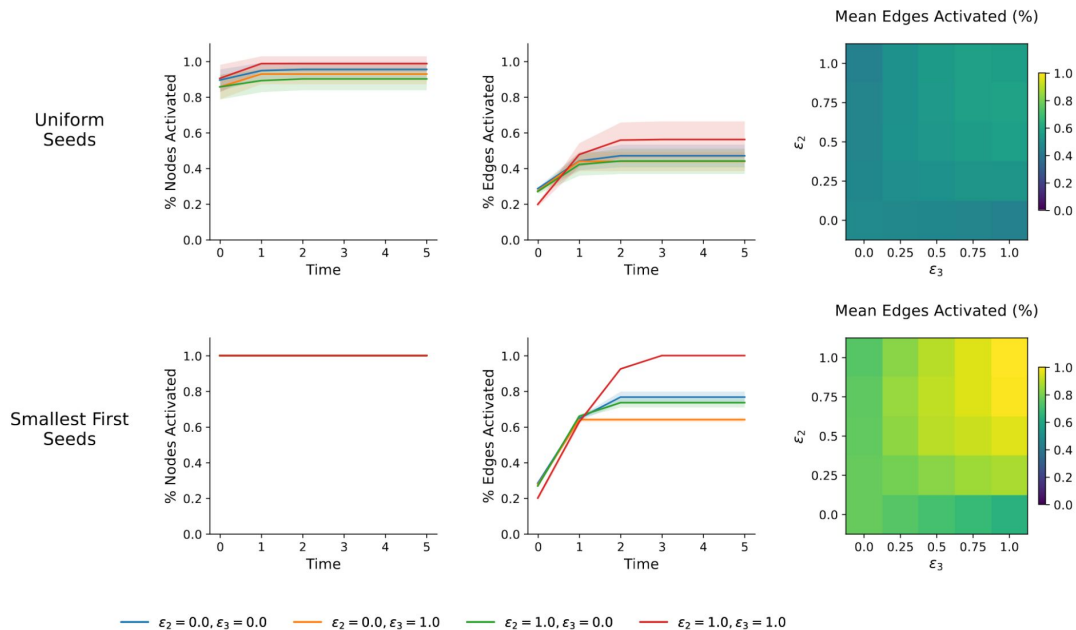
Overlap Structures for Varying ε_s



In strict encapsulation dynamics, **activation can only spread up the black edges!**

Simulation Results

Average results over 50 strict encapsulation dynamics simulations on 50 RNHM realizations



RNHM with $N = 20$, $s_m = 4$, $H_s = 5$, and varying ϵ_s , including individual nodes as hyperedges.

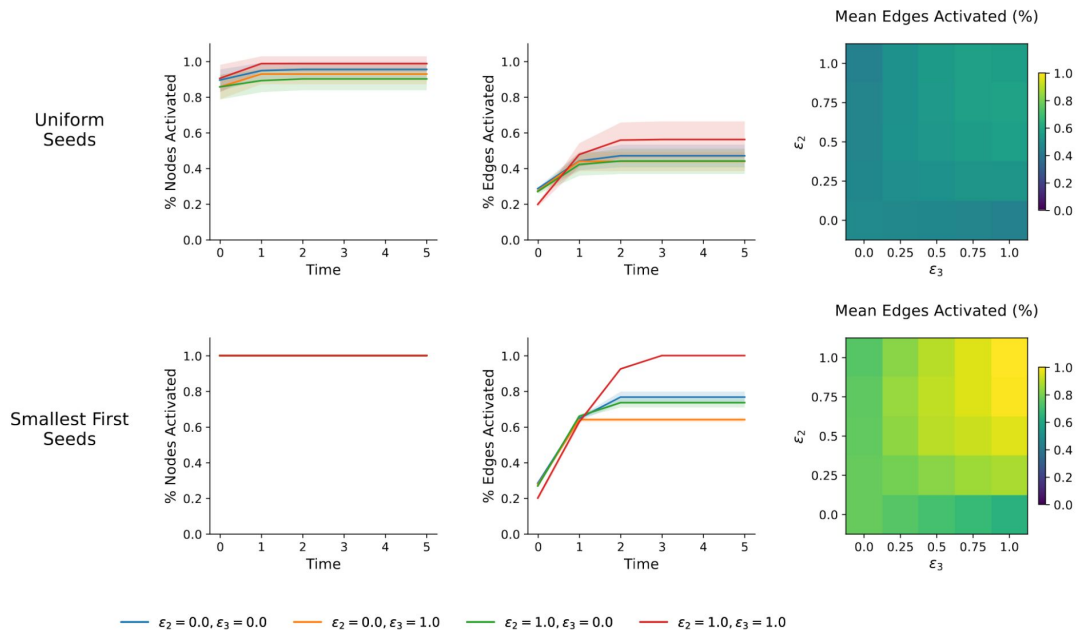
Number of seeds = N
(number of nodes)

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Uniform seeding:

- Even with high proportion of nodes activated, only half of edes in the best case

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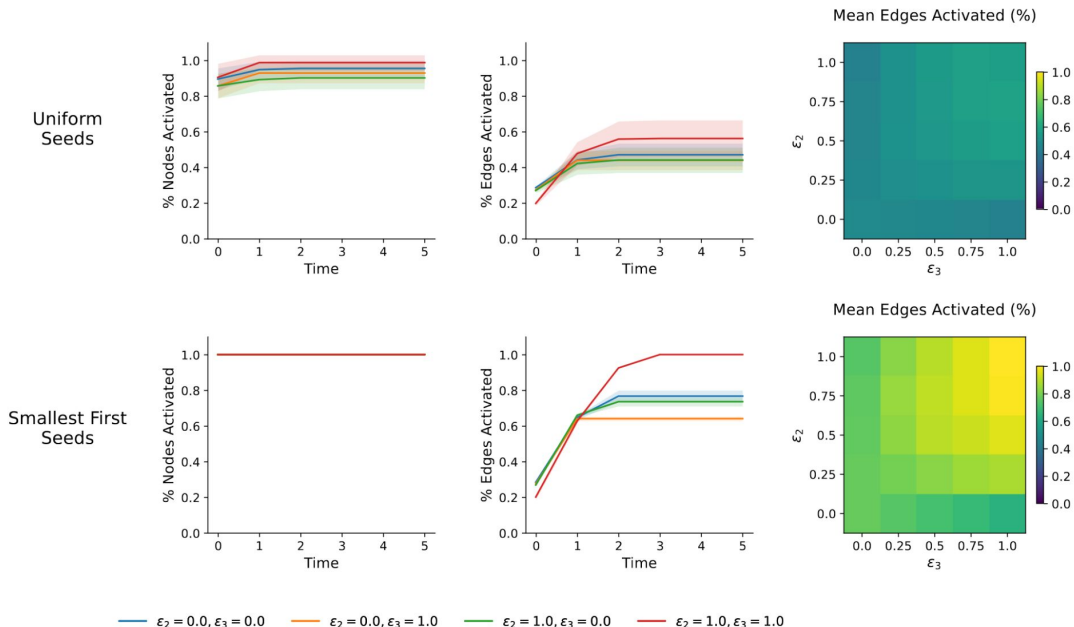
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Smallest first seeding:

- When no hyperedges are rewired, full hypergraph becomes activated (trivial but important)
- Even though nodes are activated by definition, all hyperedges do not become active. Key distinguishing feature from node-based threshold dynamics.

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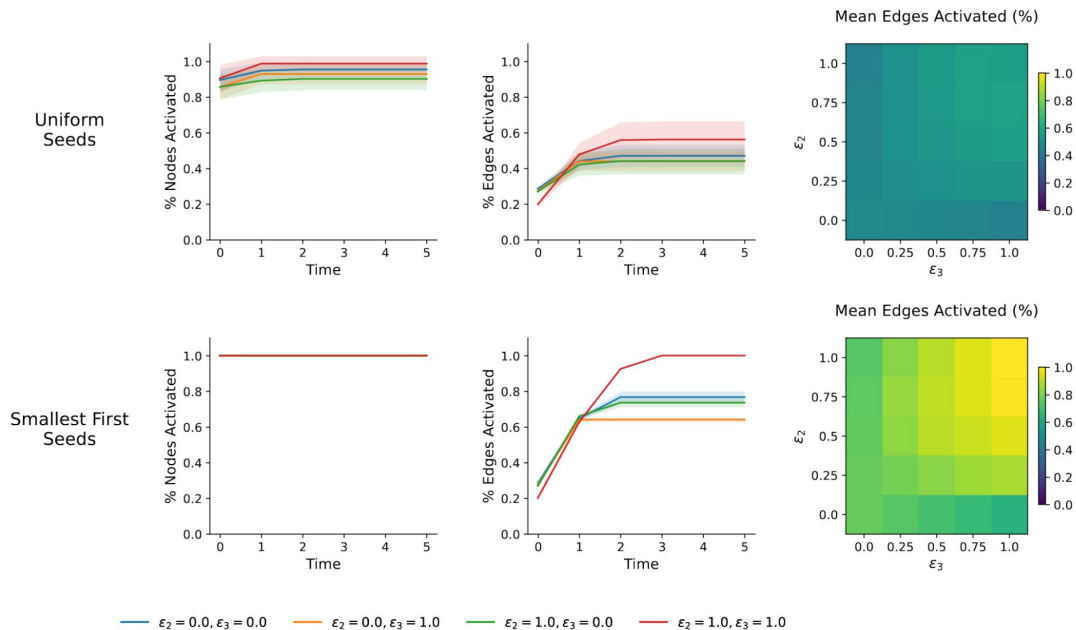
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These dynamics correspond not just to node influence, but to **coordinated behavior**!

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Thank you!

Contact

larock@maths.ox.ac.uk
<https://www.tlarock.github.io>

Code: @tlarock on GitHub

<https://www.github.com/tlarock/encapsulation-dynamics>



Also in the paper...

Measurements on empirical datasets

- DAG degrees, paths through encapsulation DAGs
- Comparison with randomization using a simple procedure that destroys encapsulation relationships

Comparison of strict vs. non-strict encapsulation dynamics on empirical hypergraphs, accounting for (non-)influence of individual nodes on pairwise interactions.

arXiv > cs > arXiv:2307.04613

Computer Science > Social and Information Networks

[Submitted on 10 Jul 2023]

Encapsulation Structure and Dynamics in Hypergraphs

Timothy LaRock, Renaud Lambiotte

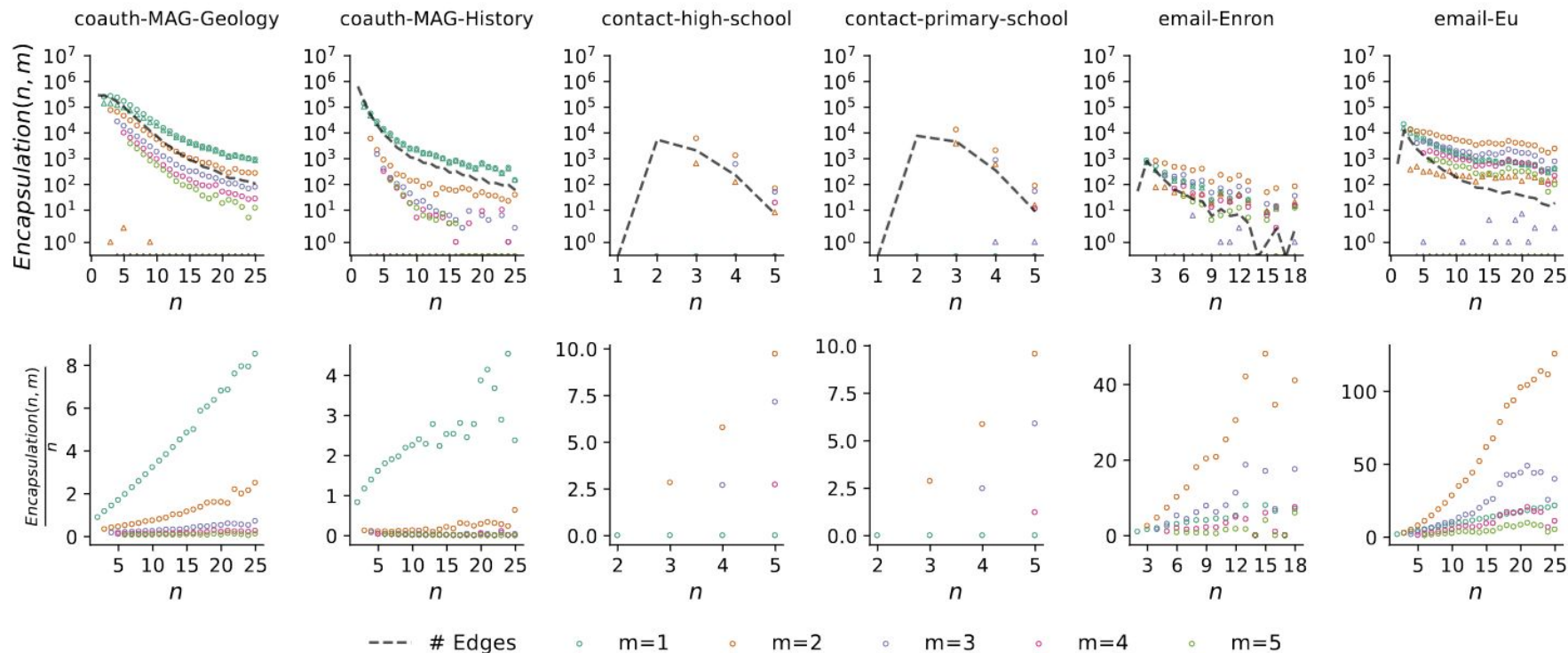
Future Work

- Structure
 - Temporal + dynamic interactions
 - Integration with work on hypergraph motifs, random walks, etc.
 - Further network analysis of encapsulation DAGs and other line graphs
- Dynamics
 - More complex models, including stochastic models
 - Combining node and hyperedge influence
 - Different models of influence between hyperedges (simulations w/ different line graphs)
 - Analytic results - encapsulation dynamics are node-based dynamics on line graphs
- Applications
 - Cooperation and competition
 - Social influence

Simulation Results



DAG Out-degree



Layer Randomization

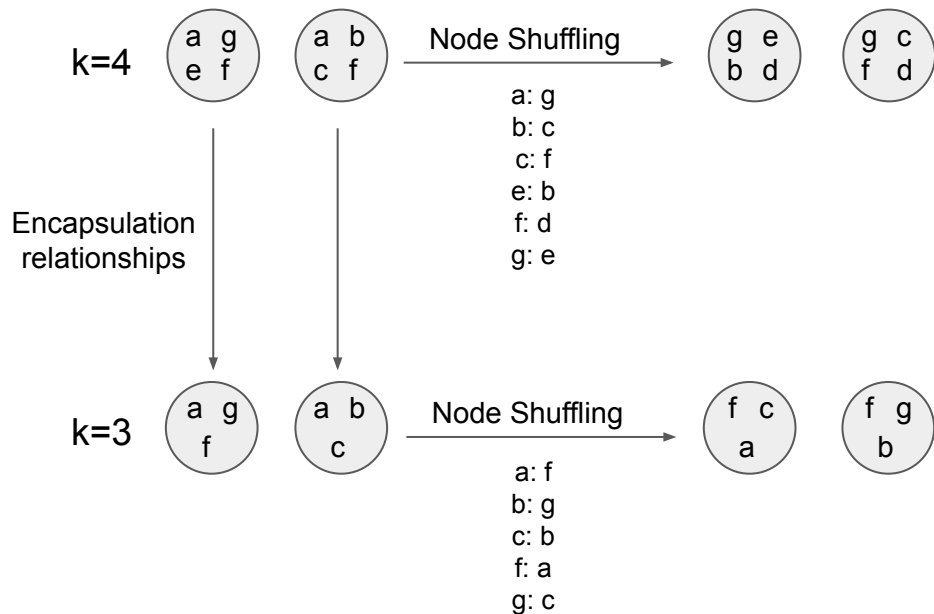
Idea: Shuffle node labels within each size layer

Randomizes:

- Encapsulation and overlap relationships
- Labeled node-degree distributions within and across size layers

Preserves:

- Hyperedge size distribution
- Unlabeled node-degree distribution within size layers



Layer Randomization

k=4



Node Shuffling



a: g
b: c
c: f
e: b
f: d
g: e

Encapsulation
relationships



Node Shuffling



a: f
b: g
c: b
f: a
g: c

Maximum Encapsulation

Layer Randomization

k = 4

k = 3

k = 2

coauth-MAG-Geology

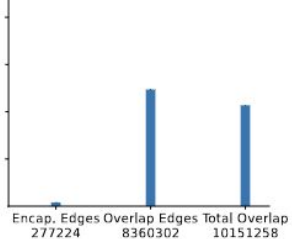
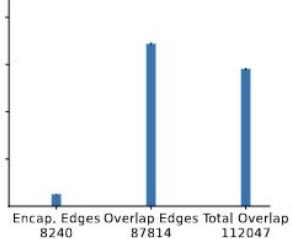
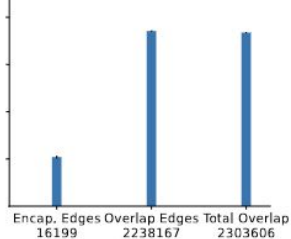
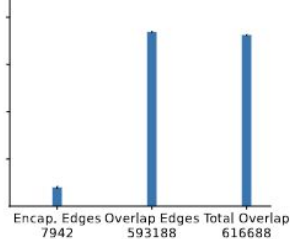
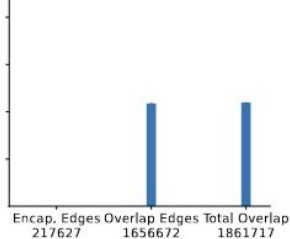
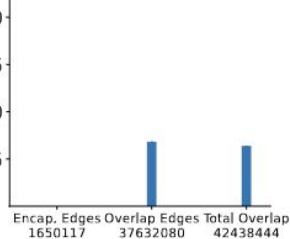
coauth-MAG-History

contact-high-school

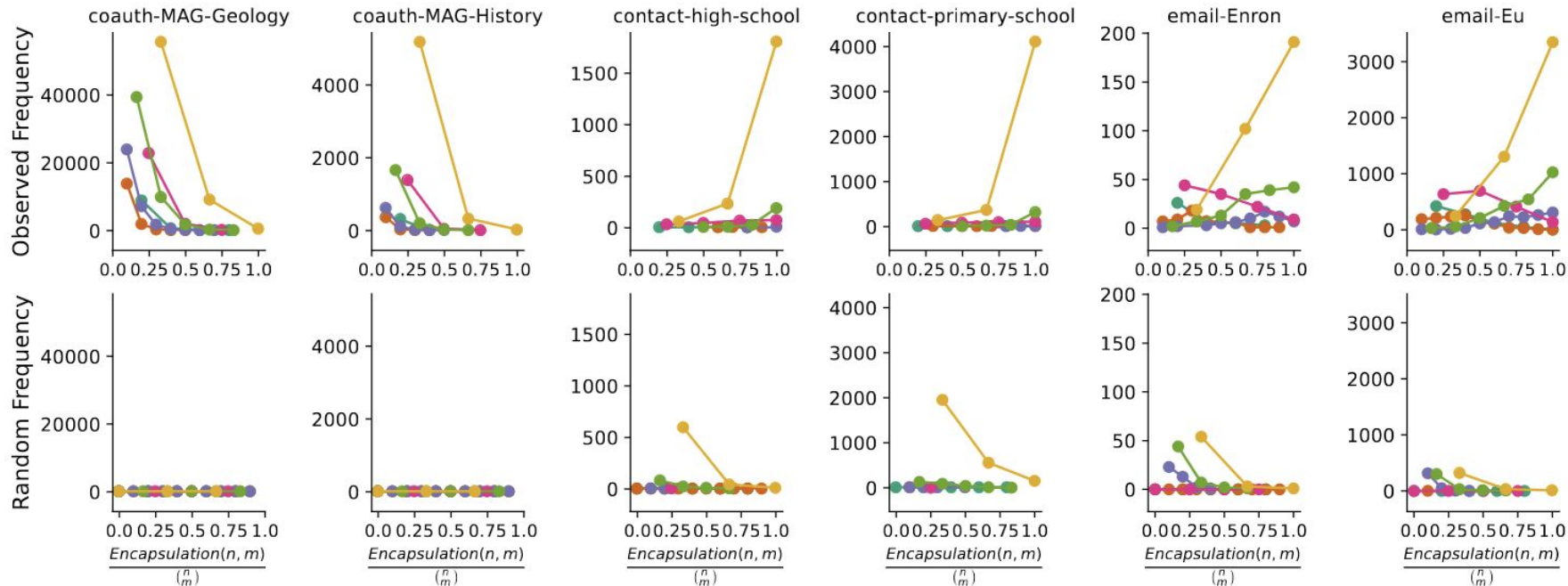
contact-primary-school

email-Enron

email-Eu

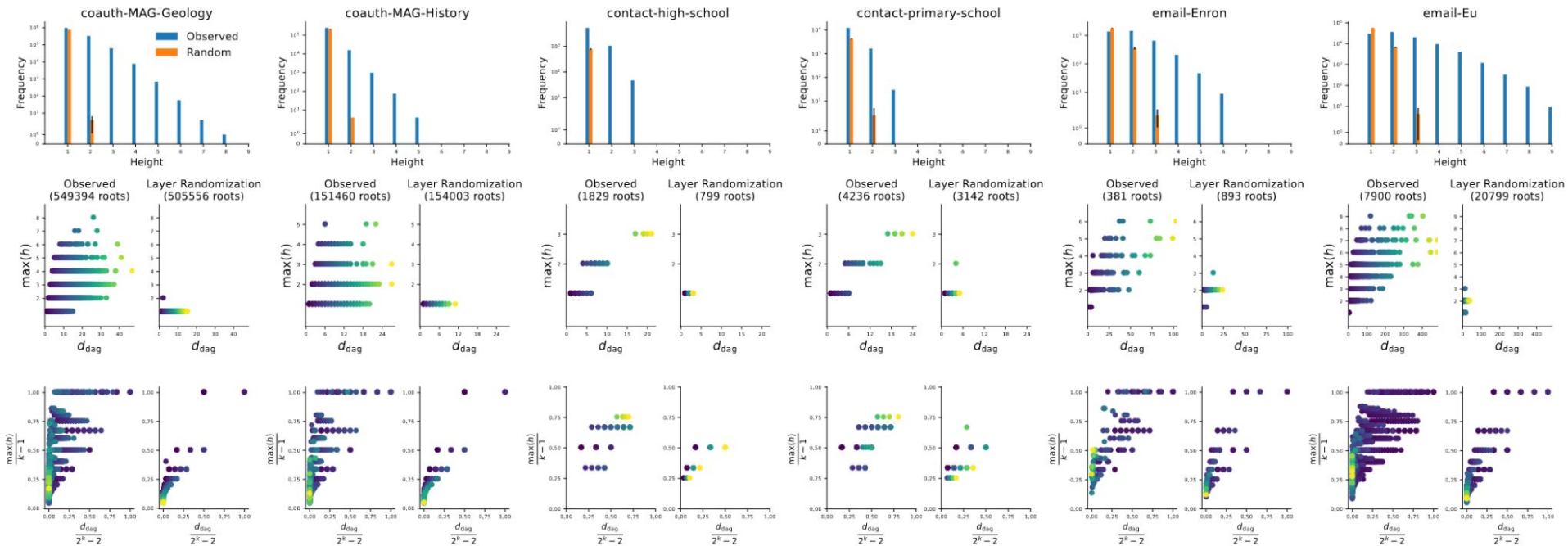


DAG Degree



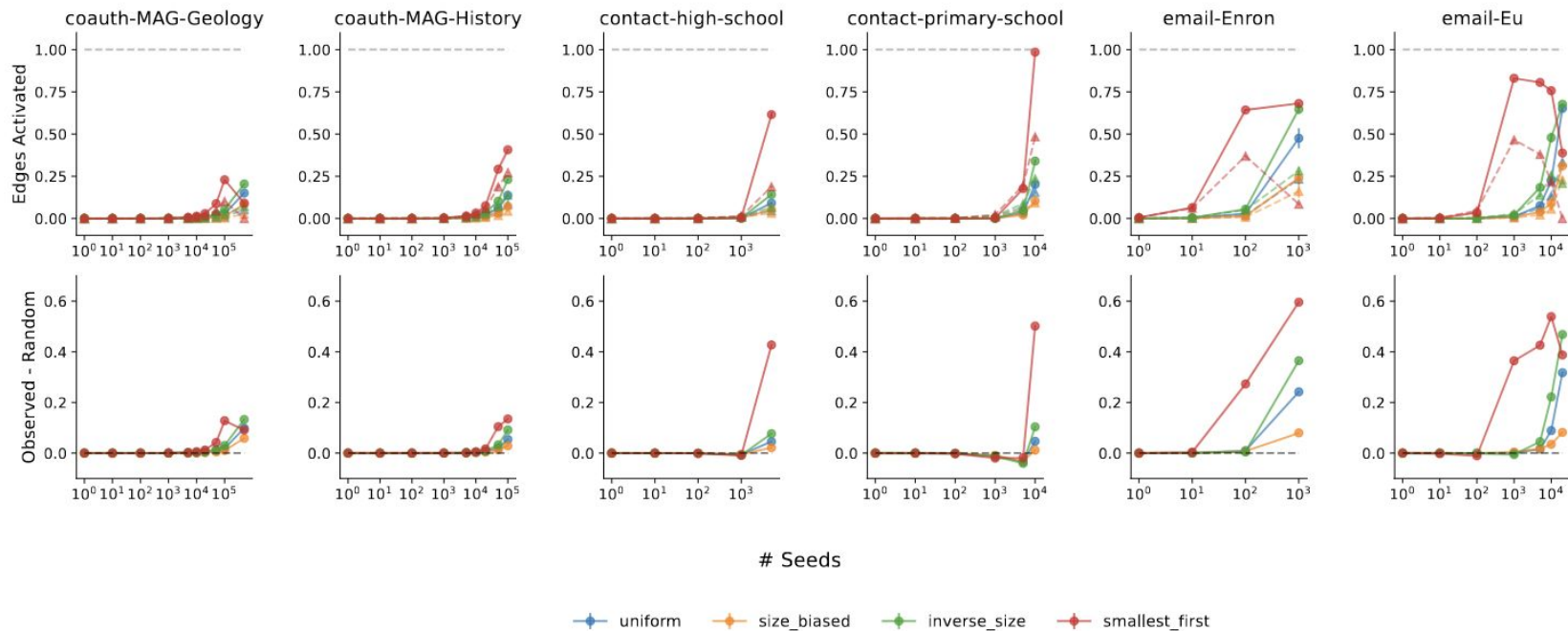
● $n=5, m=4$
● $n=5, m=3$
● $n=5, m=2$
● $n=4, m=3$
● $n=4, m=2$
● $n=3, m=2$

Paths through DAGs



Empirical Simulation Results

Strict Encapsulation Dynamics, 25 steps, $\tau = \text{all}$

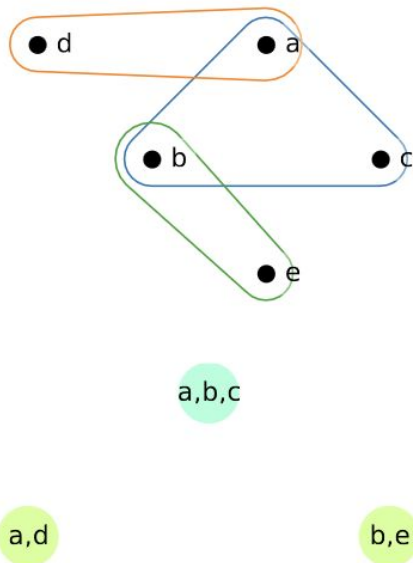


Encapsulation Dynamics

Hypergraph 1

In node-based threshold dynamics, all hyperedges will become active if nodes a and b become activated.

Full activation will be impossible in encapsulation dynamics.

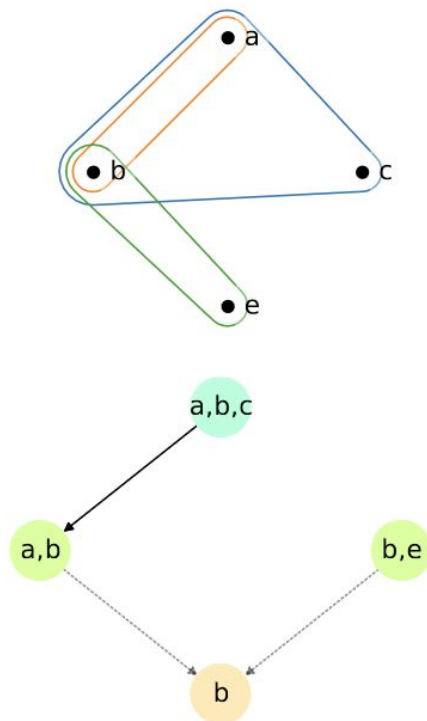


Hypergraph 2

In node-based threshold dynamics, activating any of a, b, e could activate all hyperedges.

In encapsulation dynamics, full activation can only happen if hyperedge $\{a, b\}$ is activated.

In non-strict dynamics, activation of b would activate a, b



There is tons of great work on higher-order structures and dynamics!

Networks beyond pairwise interactions: Structure and dynamics

Federico Battiston^{a,*}, Giulia Cencetti^b, Iacopo Iacopini^{c,d}, Vito Latora^{c,e,f,g},
Maxime Lucas^{h,i,j}, Alice Patania^k, Jean-Gabriel Young^l, Giovanni Petri^{m,n}

Simplicial closure and higher-order link prediction

Austin R. Benson^a, Rediet Abebe^a, Michael T. Schaub^{b,c}, Ali Jadbabaie^{b,d}, and Jon Kleinberg^{a,1}

Higher-order percolation processes on multiplex hypergraphs

Hanlin Sun¹ and Ginestra Bianconi^{1,2}

How Do Hyperedges Overlap in Real-World Hypergraphs? - Patterns, Measures, and Generators

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Contagion dynamics on hypergraphs with nested hyperedges

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Evolutionary dynamics of higher-order interactions in social networks

Unai Alvarez-Rodriguez^{1,2,*}, Federico Battiston^{3,4}, Guilherme Ferraz de Arruda⁵, Yamir Moreno^{5,6,7}, Matjaž Perc^{8,9,10}, and Vito Latora^{2,11,12}

Article

<https://doi.org/10.1038/s41467-023-37118-3>

Multistability, intermittency, and hybrid transitions in social contagion models on hypergraphs

Received: 13 April 2022

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Guilherme Ferraz de Arruda^{1,✉}, Giovanni Petri^{1,2},
Pablo Martín Rodríguez³ & Yamir Moreno^{1,4,5}

High-order Line Graphs of Non-uniform Hypergraphs: Algorithms, Applications, and Experimental Analysis

Xu T. Liu^{*†}, Jesun Firoz[‡], Sinan Aksoy[‡], Ilya Amburg[‡],

Andrew Lumsdaine^{†‡}, Cliff Joslyn[‡], Assefaw H. Gebremedhin[‡], Brenda Praggastis[‡]

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There is tons of graphs on higher-order structures?

micros!

Simplicial

Weighted Simplicial Complex: A Novel Approach for Predicting Small Group Evolution

Ankit Sharma^{1,3}, Terrence J. Moore², Ananthram Swami², and Jaideep Srivastava³

Networks pairwise interactions

Encetti



ARTICLE
<https://doi.org/10.1038/s41467-019-10431-6>
Simplicial models of social contagion

OPEN

Article
Multistability in

The effect of heterogeneity on hypergraph contagion models

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Hypergraph Motifs: Concepts, Algorithms, and Discoveries

Article
<https://doi.org/10.1038/s41467-023-37190-9>

Higher-order interactions shape collective dynamics differently in hypergraphs and simplicial complexes

High-order Line Graphs of Non-uniform Learning the effective order of a hypergraph dynamical system

Leonie Neuhäuser¹, Michael Scholkemper¹, Francesco Tudisco², and Michael T. Schaub¹

Andrew Lumsdaine^{1*}, Clift Joslyn¹, Assefaw H. Gebremedhin¹, Brenda Praggastis¹
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