

CELLULAR SHEAVES, COHOMOLOGY, & APPLICATIONS

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PART 3



OUTLINE



REASONING WITH SHEAVES

NONABELIAN GOALS

HARMONIC CONVERGENCE

SIMPLE EXAMPLES

CELLULAR SHEAVES

PROLOGUE : CALCULUS

BEYOND (VECTOR) SPACE

CO/SHEAVES TAKING VALUES IN...

THERE ARE SEVERAL APPLICATIONS FOR SHEAVES WITH MORE GENERAL DATA CATEGORIES

SETS

REEB GRAPHS, SMOOTHING, APPROXIMATION
CONTEXTUALITY, PARADOX

DE SILVA, MUNCH, & PATEL
ABRAMSKY + AL.

SEMIGROUPS

POSITIVE CO/HOMOLOGY FOR SHEAVES OVER TIME AXIS

G + KRISHNAN

\mathbb{Z} -MODULES

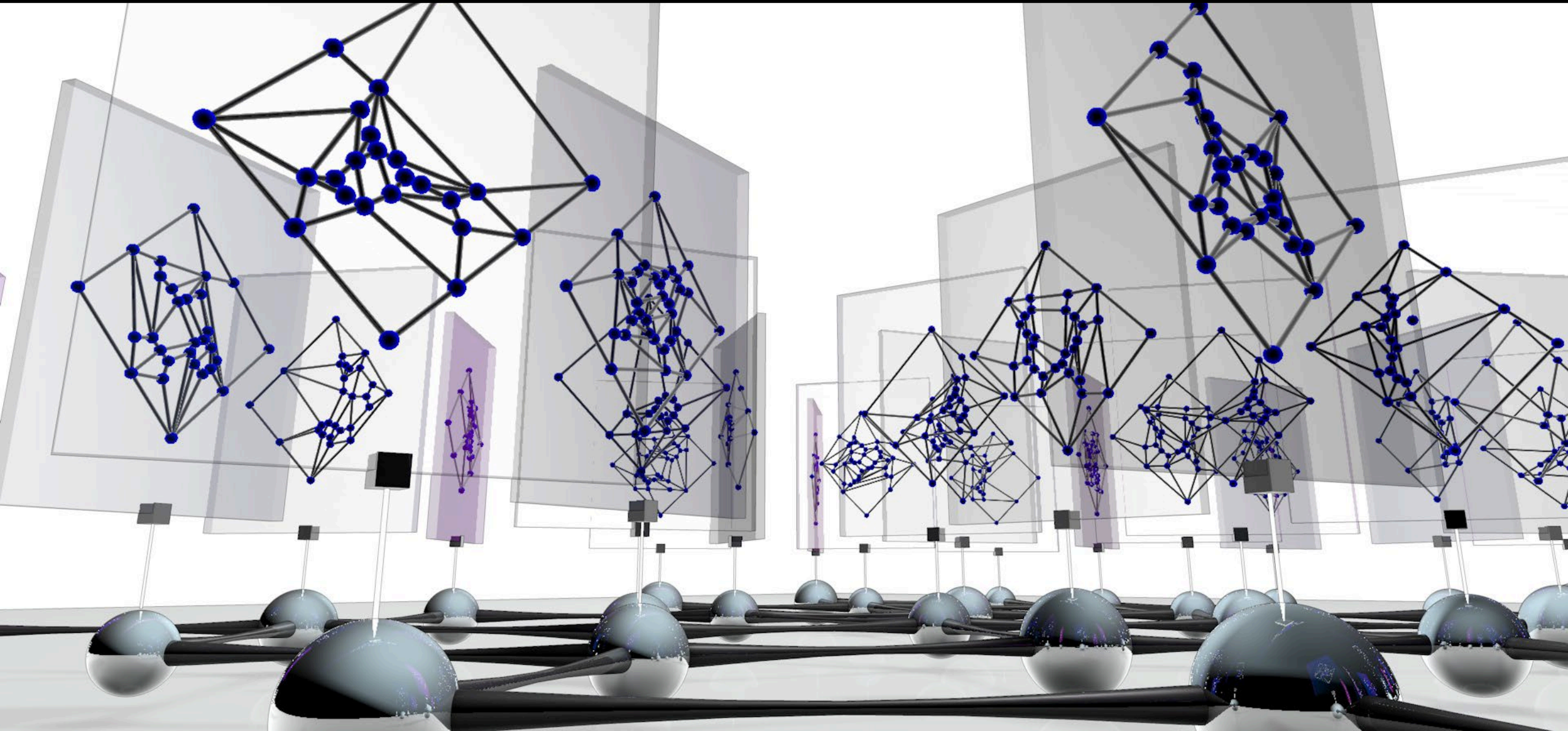
GENERALIZED PERSISTENCE ; BI-SHEAVES ; VERY GENERAL!

PATEL, MACPHERSON-PATEL

SOMETHING MORE GENERAL...

BASED ON WORK WITH HANS RIESS

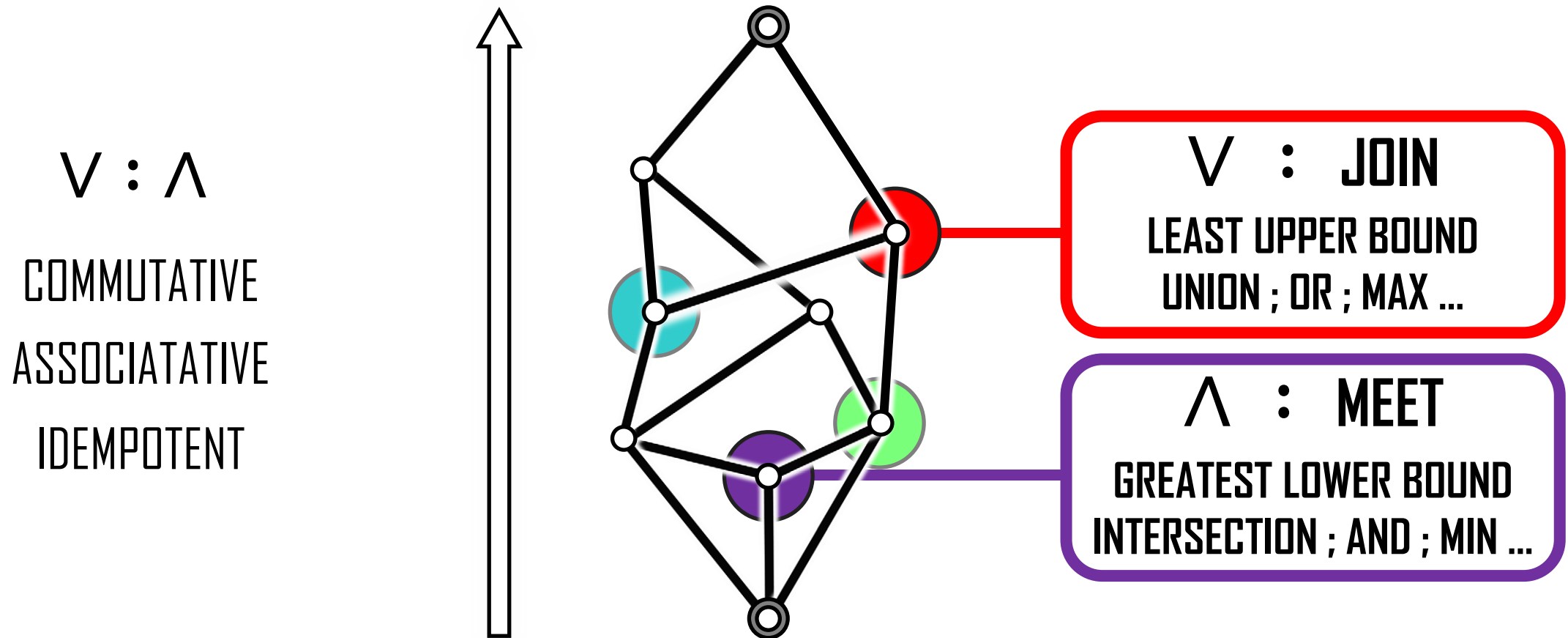
SHEAVES OF LATTICES



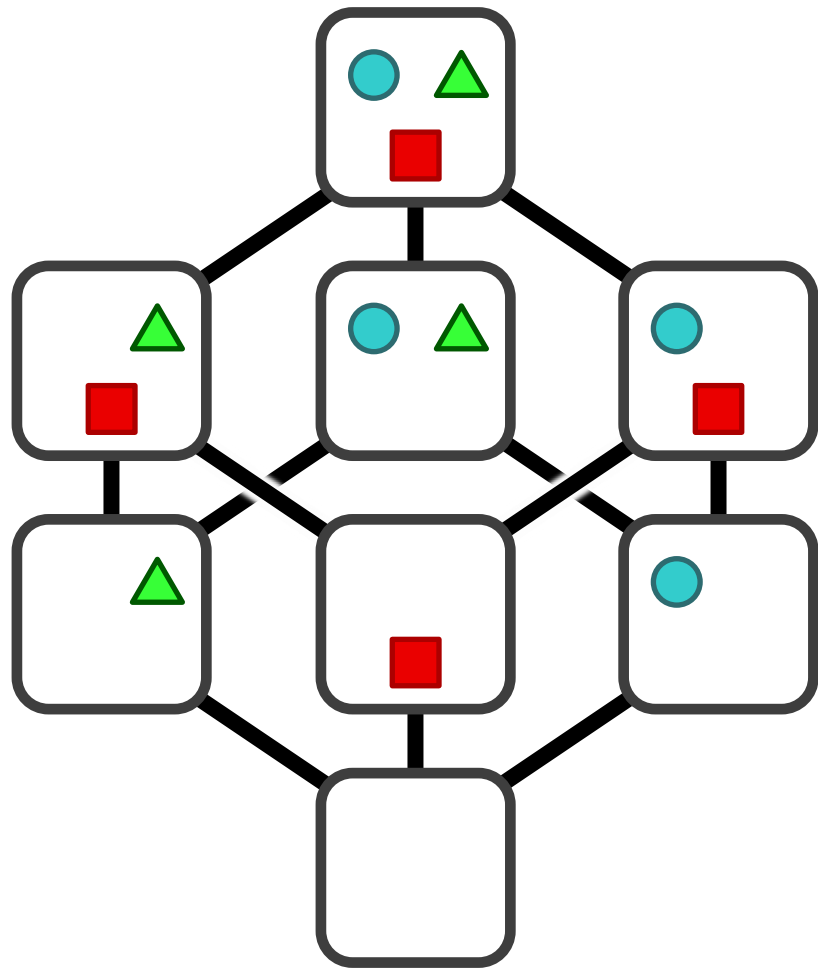
BACKGROUND : LATTICES

LATTICES

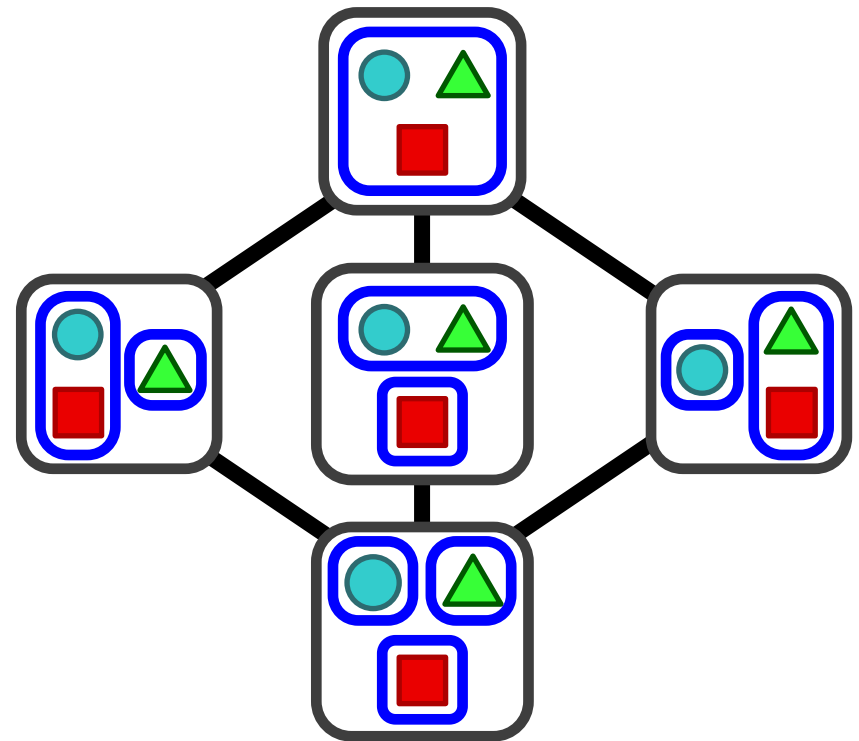
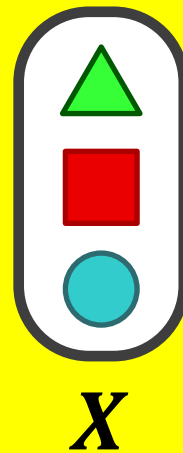
LATTICES are PARTIALLY ORDERED SETS with a PAIR of OPERATIONS



EXAMPLES OF LATTICES



$2^X = \text{POWER SET of } X$



$P(X) = \text{PARTITIONS of } X$

GALOIS / CONCEPT LATTICES

Given a binary relation on two sets:

Acts as a **KERNEL** for a transform...

Inverse blocks are called **CONCEPTS**

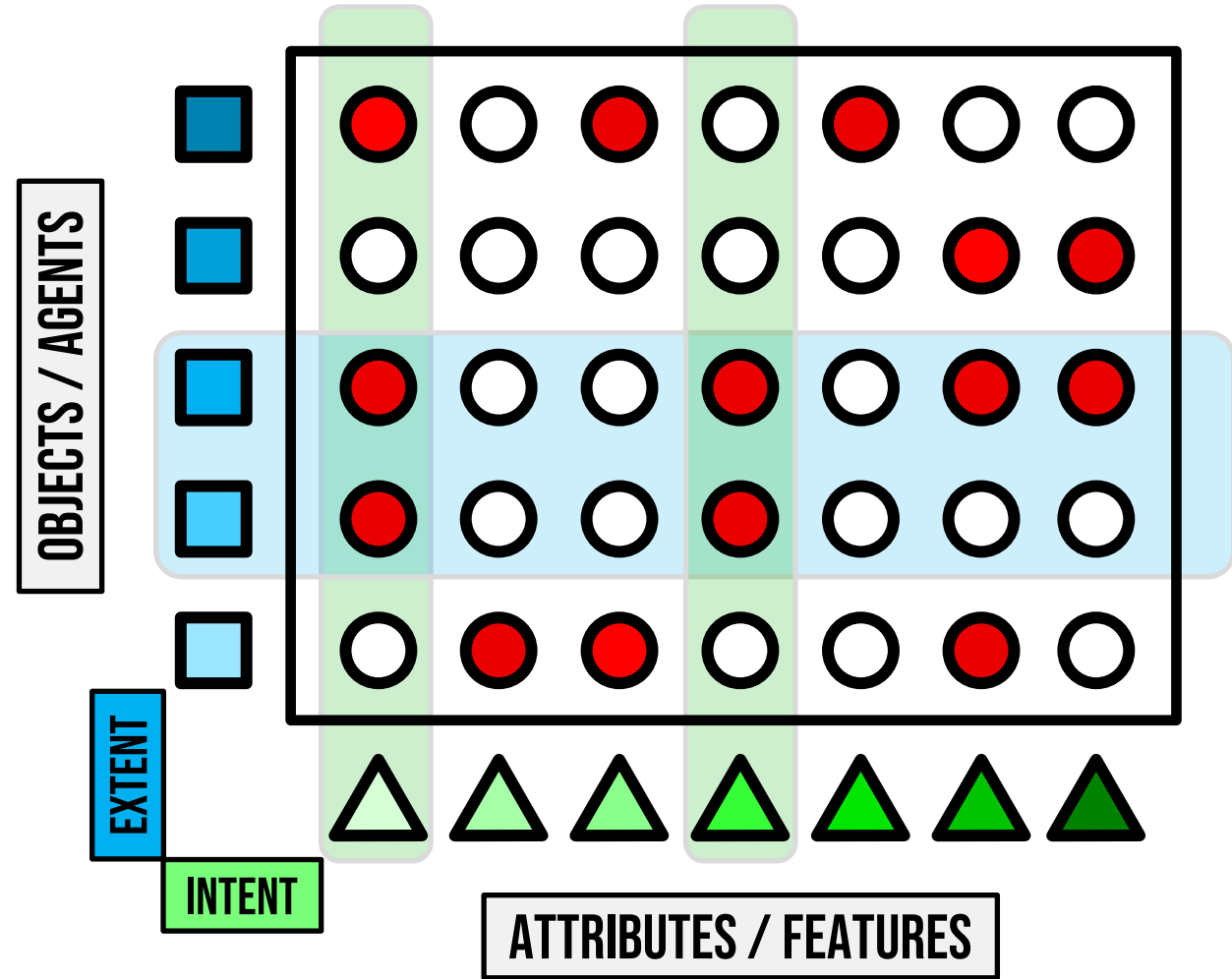
Concepts are pairs **INTENT-EXTENT**

Concepts generate an ordering/lattice

CONCEPT LATTICE

GALOIS LATTICE

[Wille, 1981 + more...]



ATTRIBUTES / FEATURES

GALOIS / CONCEPT LATTICES

EXAMPLE : Attributes of various social media platforms

[illegible]

GALOIS / CONCEPT LATTICES

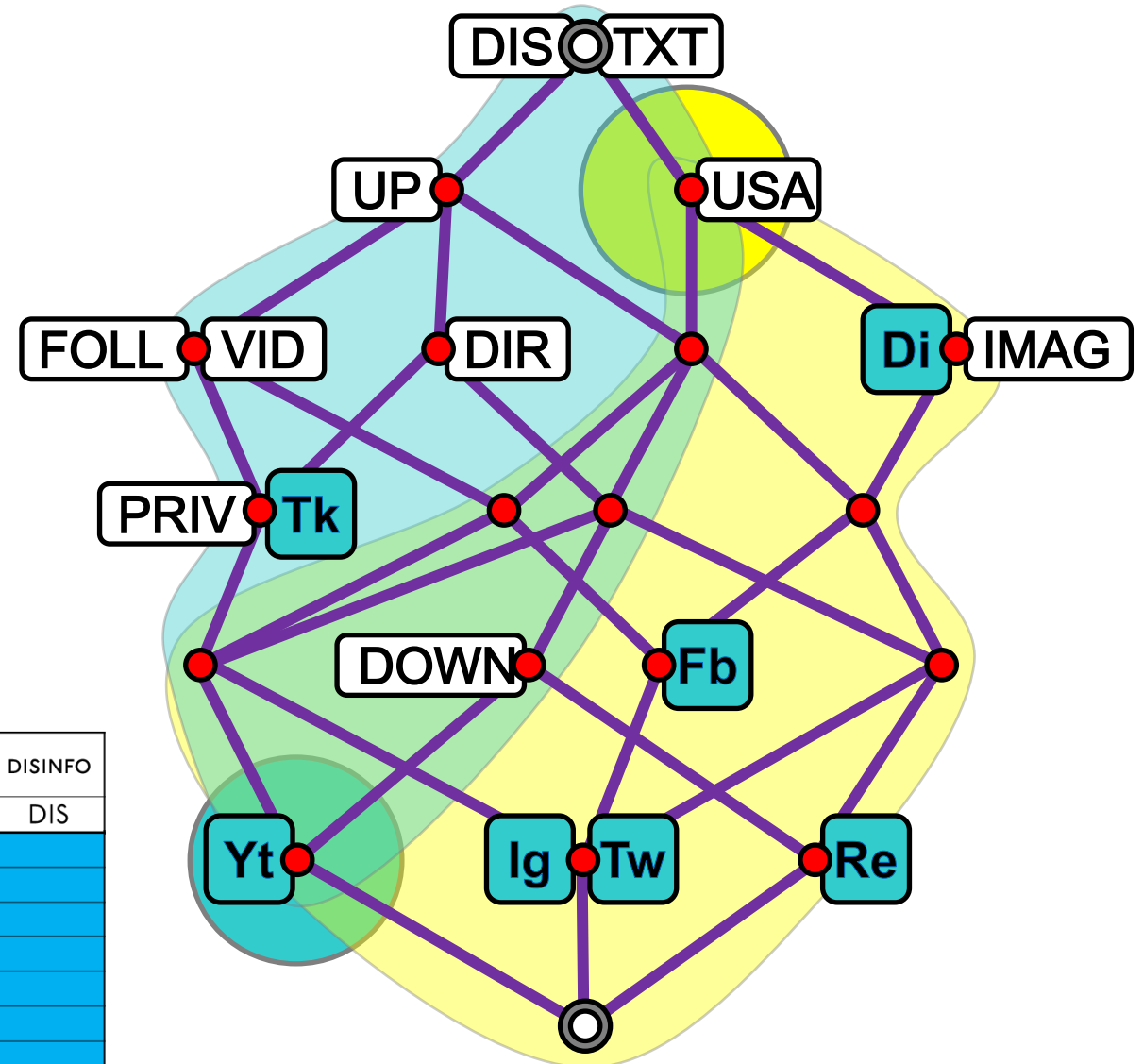
The GALOIS LATTICE arranges
the object-attribute pairs

It is illustrative to examine

UPSETS : 

DOWNSETS : 

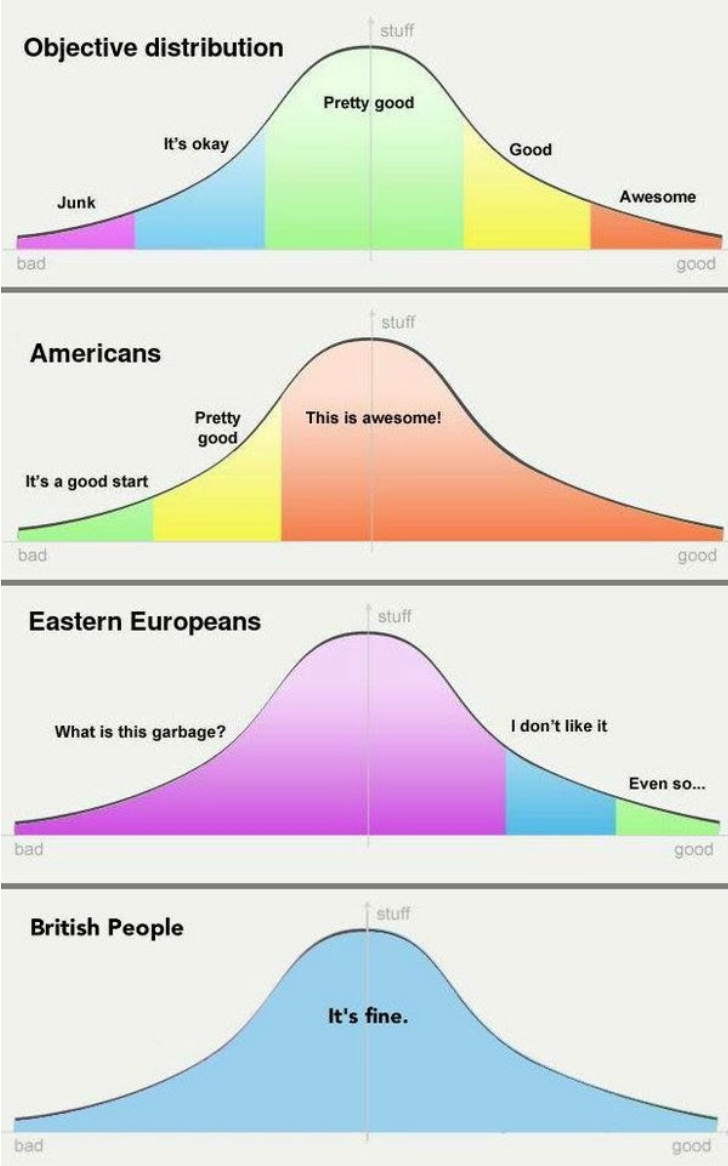
| | | TEXT | IMAGE | VIDEO | DIRECTED | FOLLOWS | UP/LIKE | DOWN | USA | PRIVATE | DISINFO |
|-----------|----|------|-------|-------|----------|---------|---------|------|-----|---------|---------|
| | | TXT | IMAG | VID | DIR | FOLL | UP | DOWN | USA | PRIV | DIS |
| TWITTER | Tw | | | | | | | | | | |
| FACEBOOK | Fb | | | | | | | | | | |
| INSTAGRAM | Ig | | | | | | | | | | |
| TIKTOK | Tk | | | | | | | | | | |
| DISCORD | Di | | | | | | | | | | |
| YOUTUBE | Yt | | | | | | | | | | |
| REDDIT | Re | | | | | | | | | | |



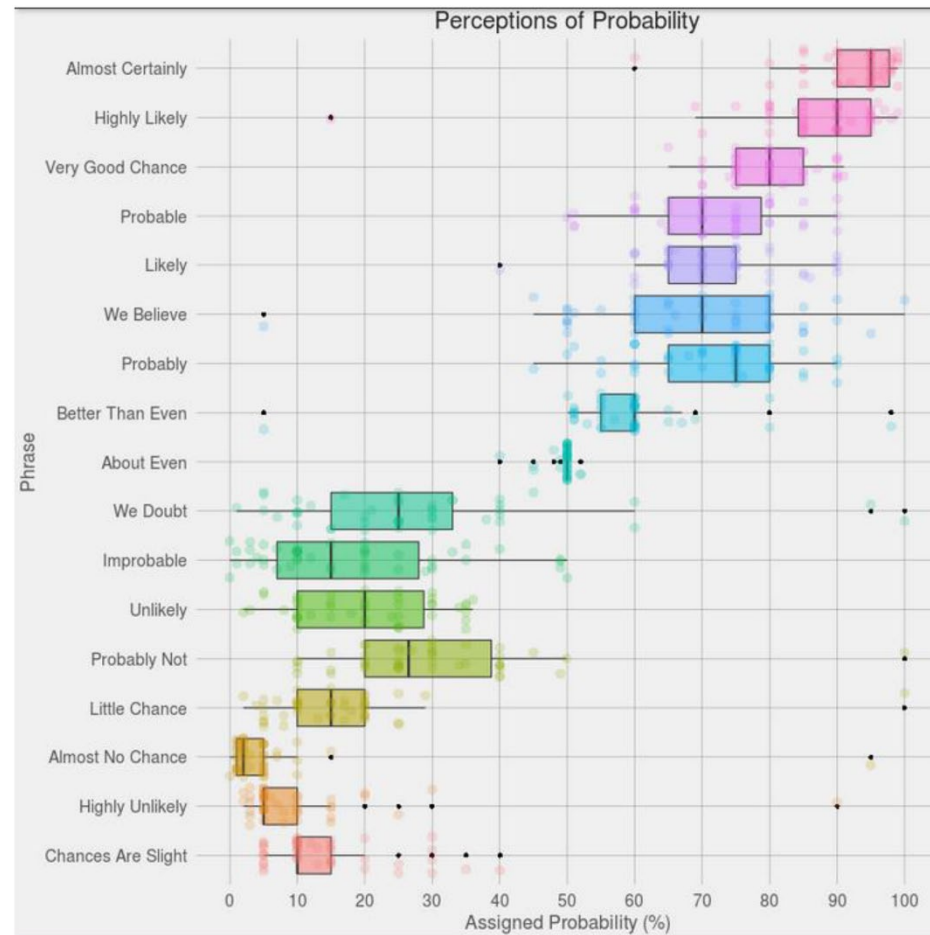
IDEA : NONLINEARITY VIA LATTICES

IF YOU WANT NONLINEAR SHEAVES, NON-LINEAR ORDERED SETS IS A GOOD START...

PERSONAL SCALES



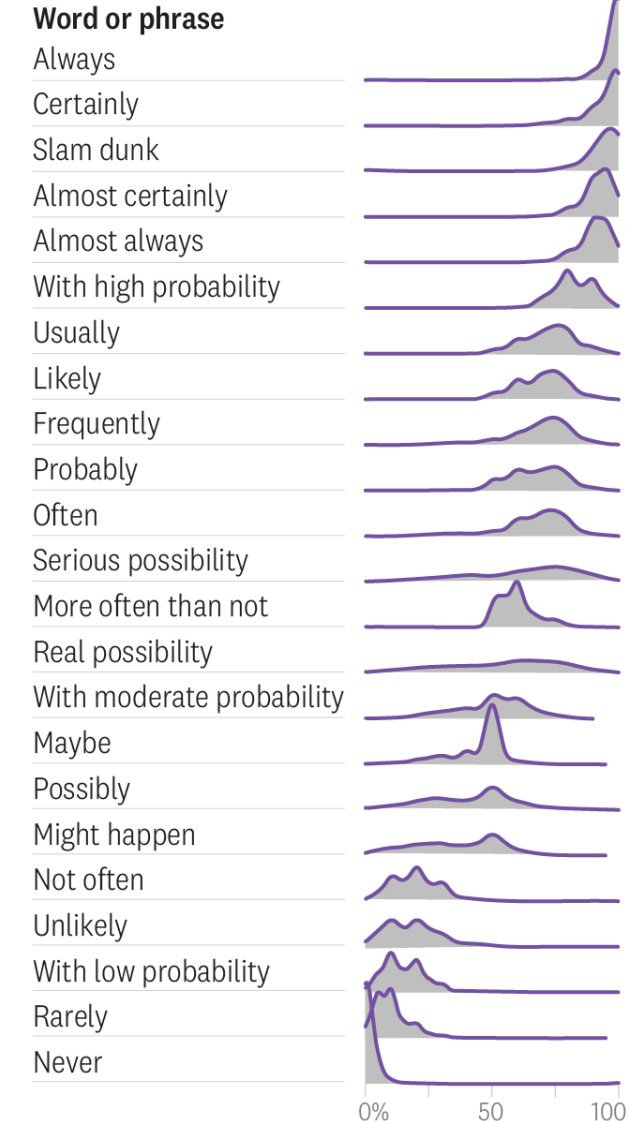
Individuals maintain personal structures for opinions, preferences, & perceptions



How People Interpret Probabilistic Words

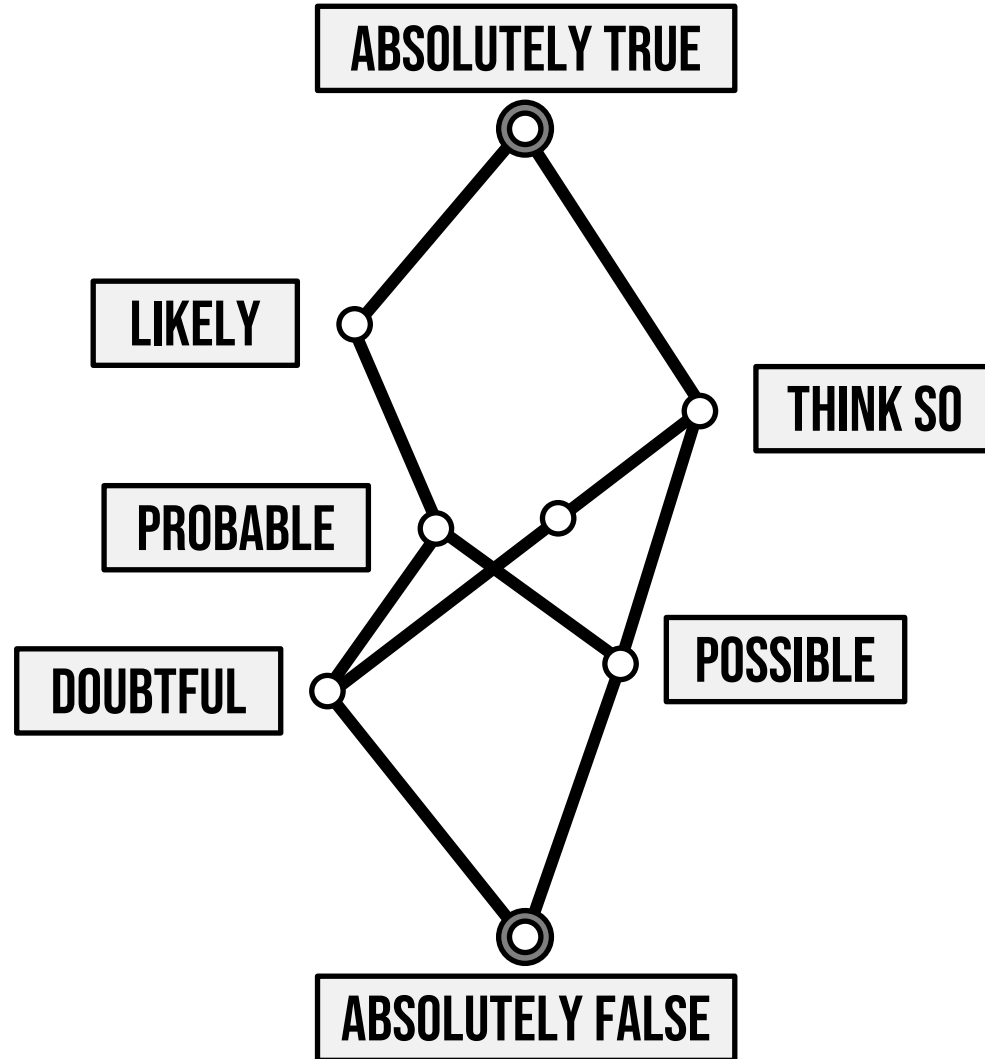
"Always" doesn't always mean always.

Distribution of responses according to respondents' estimate of likelihood

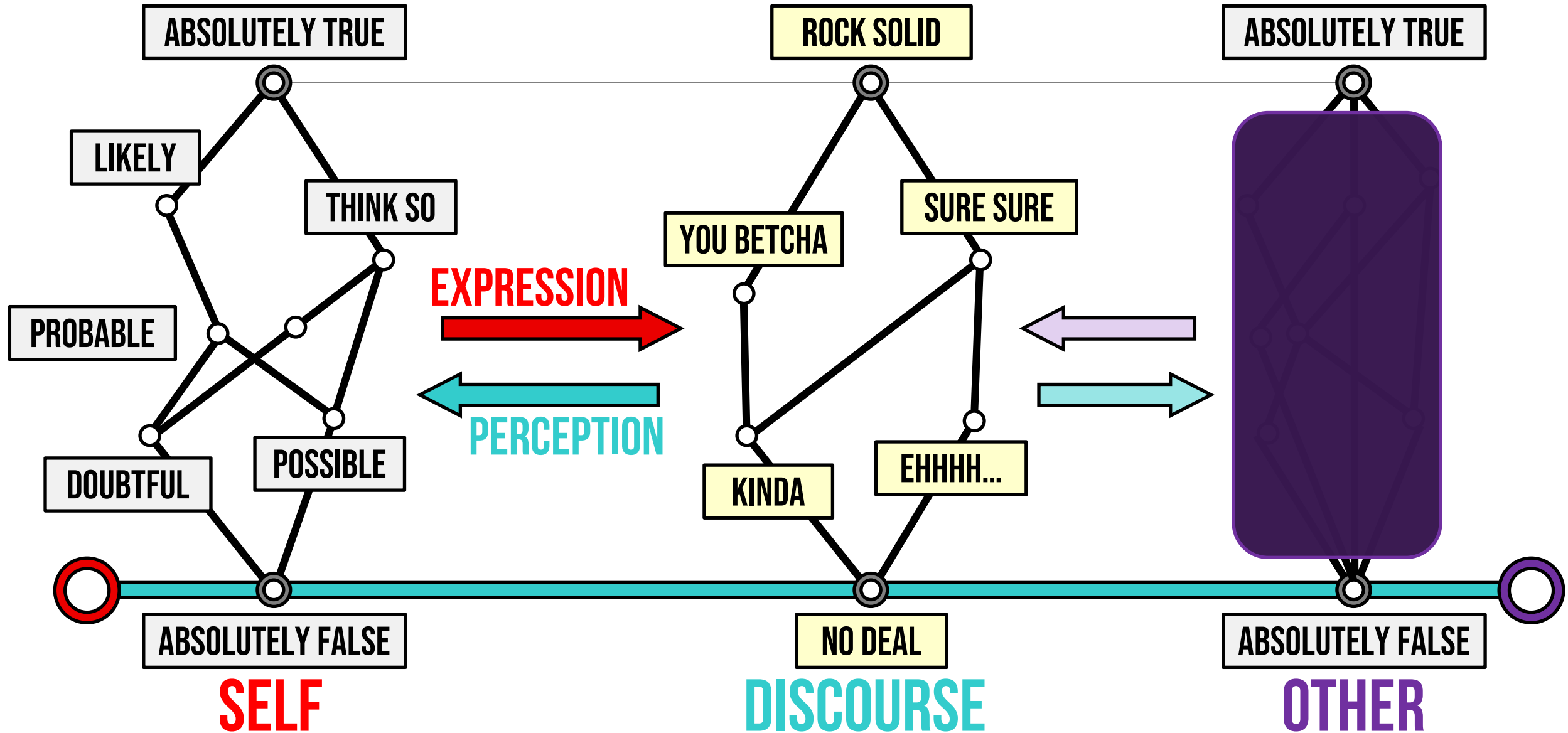


Source: Andrew Mauboussin and Michael J. Mauboussin

TRUTH & PERCEPTIONS & LATTICES



THE PROBLEM OF COMMUNICATION



EXAMPLE: COALITION-BUILDING

Consider a cover of a social network by neighborhoods N_v

Consider a sheaf \mathcal{F} whose stalks correspond to the NERVE POWERSETS

$$\mathcal{F}(v) = 2^{N_v} \quad \mathcal{F}(e) = 2^{N_u \cap N_v} \quad \partial e = \{u, v\}$$

Vertex stalks $\mathcal{F}(v)$ are private estimates of who supports the coalition

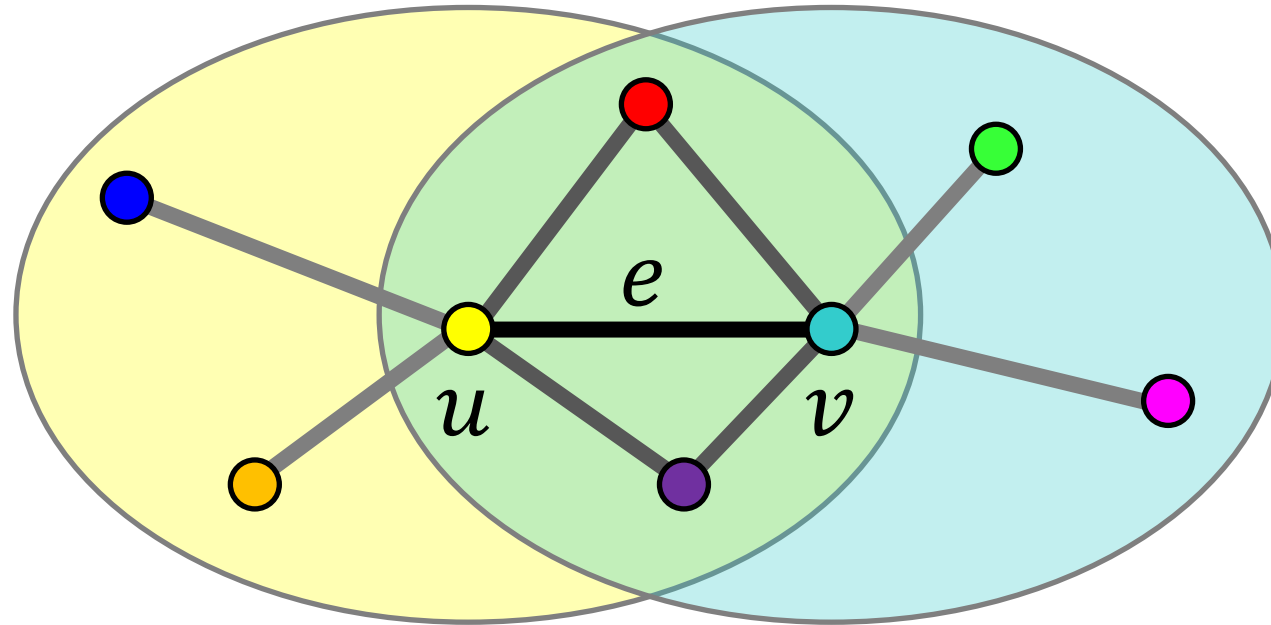
Maps to edge stalks $\mathcal{F}(e)$ are pairwise expressions of membership estimates

Think about how one could grow a coalition

EXAMPLE: COALITION-BUILDING

$$\mathcal{F}(v) = 2^{N_v}$$

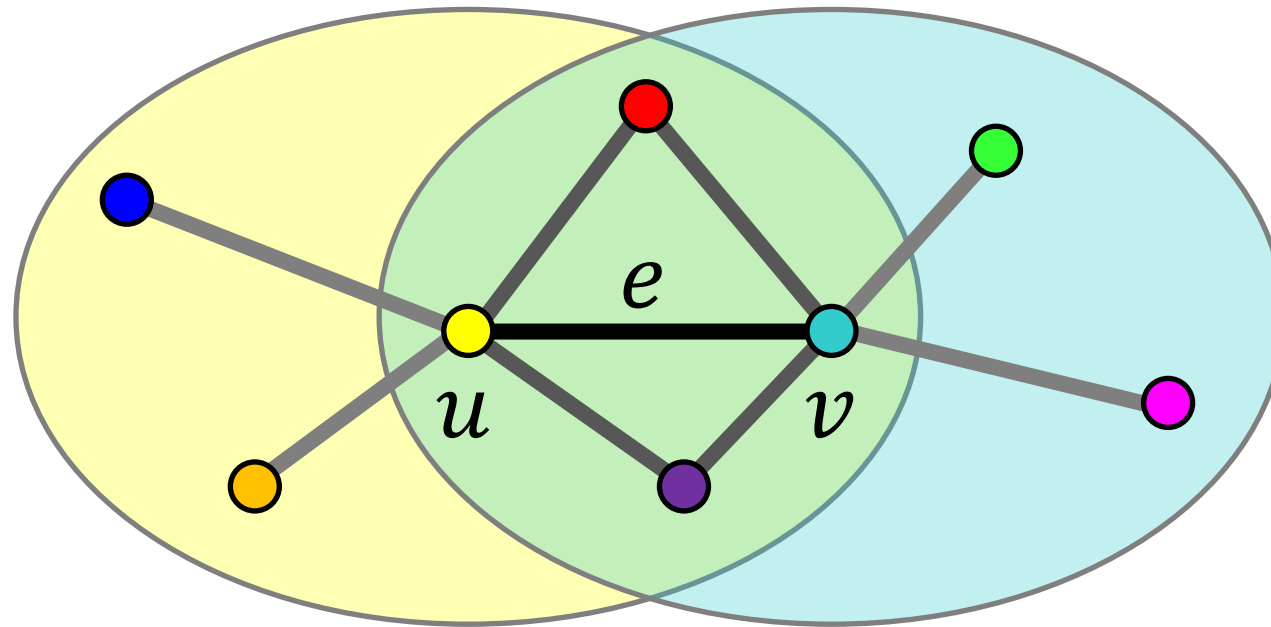
$$\mathcal{F}(e) = 2^{N_u \cap N_v}$$



Think about how one could grow a coalition

EXAMPLE: COALITION-BUILDING

Expression/Inference can be more subtle than *"Are you for or against?"*



Think about how one could grow a coalition

COMMUNICATION IN LATTICES

IS A BIT SUBTLE...

LATTICE MORPHISMS

GALOIS CONNECTIONS

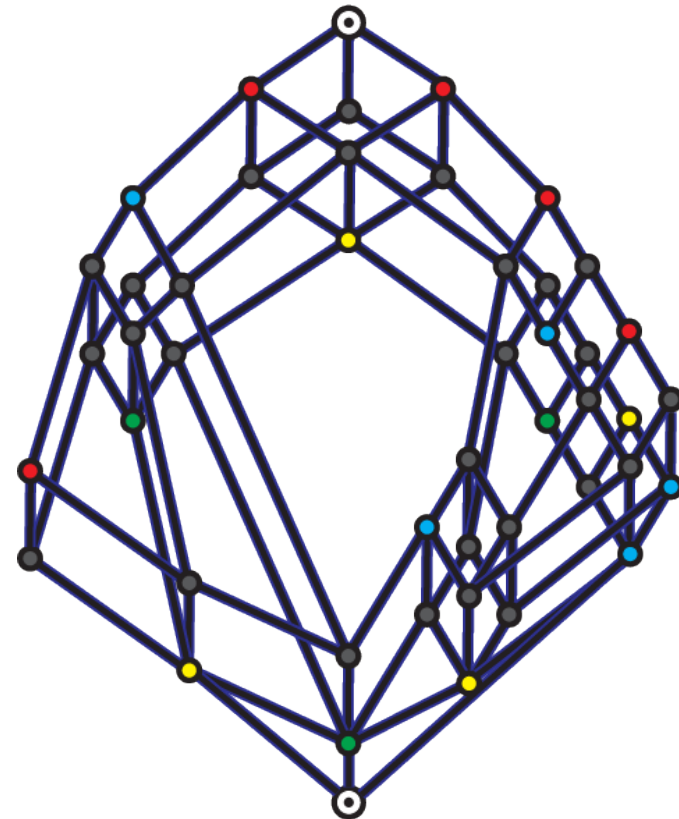
LATTICE CONNECTIONS come in ADJOINT PAIRS

$$X \begin{matrix} \xrightarrow{f_{\bullet}} \\ \xleftarrow{f^{\bullet}} \end{matrix} Y$$

$$f_{\bullet}(x) \leq y \iff x \leq f^{\bullet}(y)$$

$$f_{\bullet}(x) = \bigwedge f^{\bullet -1}(x^{\uparrow})$$

$$f^{\bullet}(y) = \bigvee f_{\bullet}^{-1}(y^{\downarrow})$$



WE WANT GLOBAL SECTIONS

BUT WE CAN'T JUST "RUN THE HEAT EQUATION" ON THESE SHEAVES
LIKE WE COULD DO FOR VECTOR-VALUED SYSTEMS

DEFINITION: TARSKI LAPLACIAN

DEFINITION : Let \mathcal{F} be a NETWORK SHEAF of LATTICES
Let $\mathbf{x} = (x_v) \in C^0$ be a 0-COCHAIN

TARSKI LAPLACIAN

$$L : C^0 \mathcal{F} \rightarrow C^0 \mathcal{F}$$

$$(L\mathbf{x})_v = \bigwedge_{e \in \delta v} (\mathcal{F}_{v \triangleleft e}) \cdot \left(\bigwedge_{w \in \partial e} (\mathcal{F}_{w \triangleleft e}) \cdot (x_w) \right)$$

WHY IS THIS A “LAPLACIAN”?

IT CERTAINLY DOESN'T LOOK LIKE IT...

DEFINITION: TARSKI LAPLACIAN

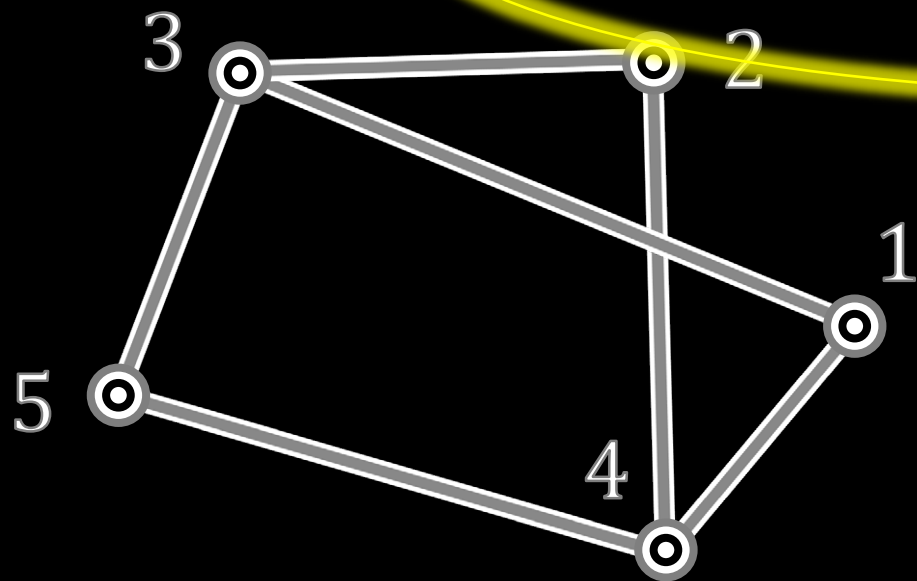
LEMMA : the TARSKI LAPLACIAN decomposes as

$$(L\mathbf{x})_v = \bigwedge_{e \in \delta v} (\mathcal{F}_{v \triangleleft e}) \cdot \left(\bigwedge_{w \in \partial e} (\mathcal{F}_{w \triangleleft e}) \cdot (x_w) \right)$$

$$= \underbrace{\left(\bigwedge_{e \in \delta v} (\mathcal{F}_{v \triangleleft e}) \cdot (\mathcal{F}_{v \triangleleft e}) \cdot (x_v) \right)}_{\text{EXPANSION}} \wedge \underbrace{\left(\bigwedge_{\substack{e \in \delta v \\ w \in \partial e - \{v\}}} (\mathcal{F}_{v \triangleleft e}) \cdot (\mathcal{F}_{w \triangleleft e}) \cdot (x_w) \right)}_{\text{MIXING}}$$

the Graph Laplacian

$$L = D - A = BB^T$$



$$L = \begin{bmatrix} 2 & 0 & -1 & -1 & 0 \\ 0 & 2 & -1 & -1 & 0 \\ -1 & -1 & 3 & 0 & -1 \\ -1 & -1 & 0 & 3 & -1 \\ 0 & 0 & -1 & -1 & 2 \end{bmatrix}$$

Very Useful Indeed!

DEFINITION: TARSKI LAPLACIAN

LEMMA : the TARSKI LAPLACIAN decomposes as

$$(L\mathbf{x})_v = \bigwedge_{e \in \delta v} (\mathcal{F}_{v \triangleleft e}) \cdot \left(\bigwedge_{w \in \partial e} (\mathcal{F}_{w \triangleleft e}) \cdot (x_w) \right) \quad \mathbf{BB}^T$$

$$= \underbrace{\left(\bigwedge_{e \in \delta v} (\mathcal{F}_{v \triangleleft e}) \cdot (\mathcal{F}_{v \triangleleft e}) \cdot (x_v) \right)}_{\text{EXPANSION}} \wedge \underbrace{\left(\bigwedge_{\substack{e \in \delta v \\ w \in \partial e - \{v\}}} (\mathcal{F}_{v \triangleleft e}) \cdot (\mathcal{F}_{w \triangleleft e}) \cdot (x_w) \right)}_{\text{MIXING}}$$

D - A

DIFFUSION

TARSKI FIXED POINT THEOREM

THEOREM : For a COMPLETE LATTICE, X ,
and an order-preserving endomorphism $f: X \rightarrow X$
the fixed point set $Fix(f)$ is a
NONEMPTY
COMPLETE
QUASI-SUBLATTICE
of X

HODGE-TARSKI FIXED POINT THEOREM

MAIN THEOREM : For a network sheaf \mathcal{F} of COMPLETE LATTICES,
the GLOBAL SECTIONS are computed via the TARSKI LAPLACIAN

$$H^0\mathcal{F} = \text{Fix}(Id \wedge L)$$

$$\text{COMPARE : } x_{n+1} = (I - L)x_n$$

$$\text{COMPARE : } dx/dt = -Lx$$

COROLLARY : $H^0\mathcal{F}$ is a NONEMPTY COMPLETE QUASI-SUBLATTICE of $C^0\mathcal{F}$

HIGHER DIMENSIONS

HODGE-TARSKI FIXED POINT THEOREM

THEOREM : For a network sheaf \mathcal{F} of COMPLETE LATTICES,
the GLOBAL SECTIONS are computed via the TARSKI LAPLACIAN

$$H^0 \mathcal{F} = \text{Fix}(\text{Id} \wedge L)$$

THINK : the global sections are really the 0-dimensional sheaf cohomology

DOES HIGHER DIMENSIONAL SHEAF COHOMOLOGY MAKE SENSE?

RECALL: TARSKI LAPLACIAN

TARSKI LAPLACIAN

$$L : C^0 \mathcal{F} \rightarrow C^0 \mathcal{F}$$

$$(L\mathbf{x})_v = \bigwedge_{e \in \delta v} (\mathcal{F}_{v \triangleleft e}) \cdot \left(\bigwedge_{w \in \partial e} (\mathcal{F}_{w \triangleleft e}) \cdot (x_w) \right)$$

EASY! Replace vertices with k -cells and edges with cofaces.

$$L_k : C^k \mathcal{F} \rightarrow C^k \mathcal{F}$$

HODGE-TARSKI COHOMOLOGY

DEFINITION : For a cellular sheaf \mathcal{F} of complete lattices on a cell complex, the TARSKI COHOMOLOGY is defined via the TARSKI LAPLACIAN

$$TH^k \mathcal{F} := \text{Fix}(Id \wedge L_k)$$

THEOREM : $TH^k \mathcal{F}$ IS A NONEMPTY COMPLETE QUASI-SUBLATTICE OF $C^k \mathcal{F}$

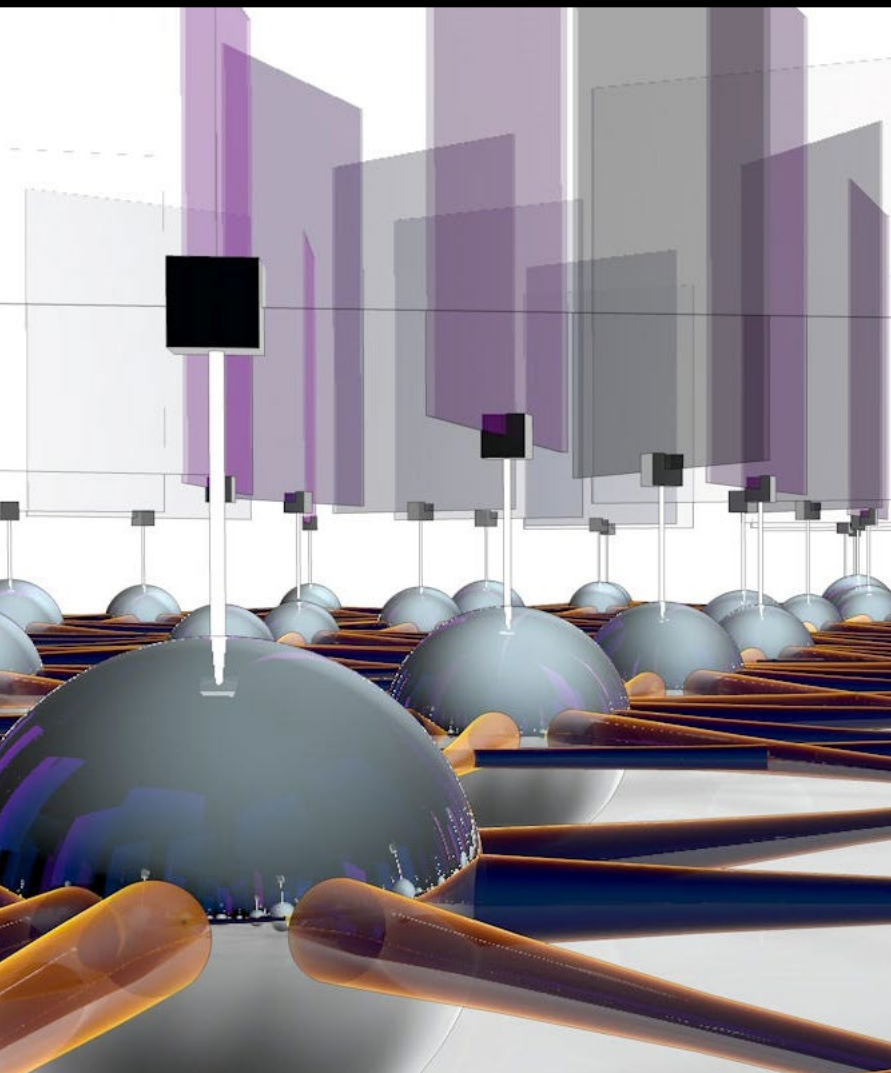
WHAT IS THIS?

THAT'S A GOOD QUESTION...

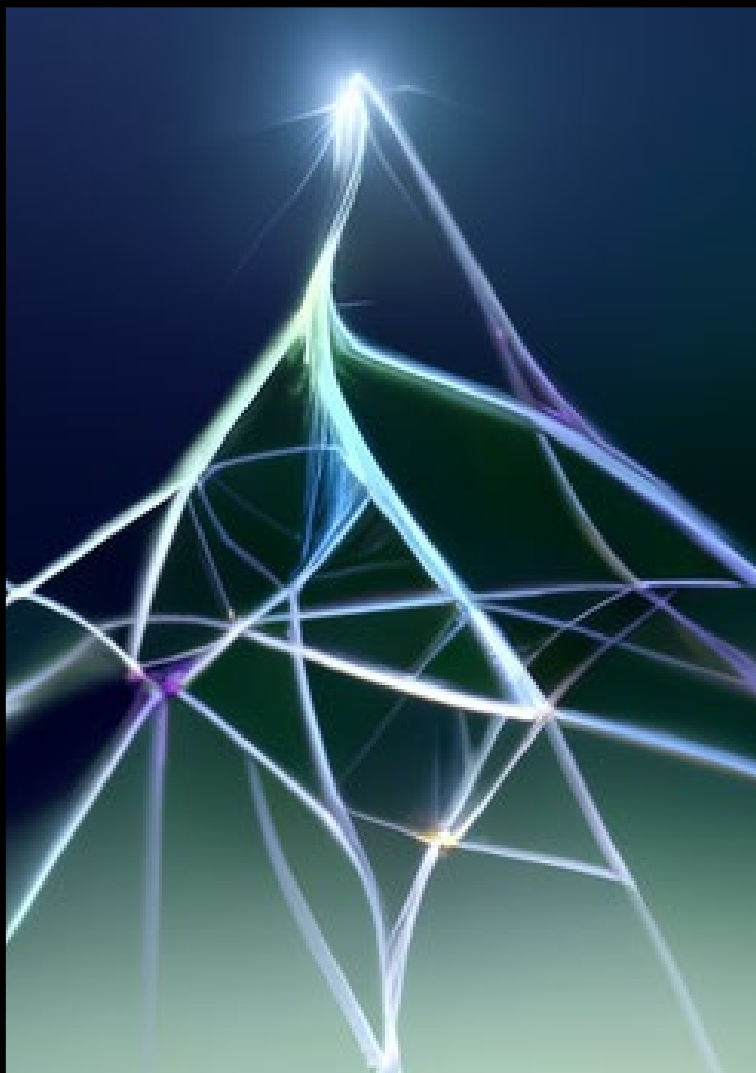
The background is a complex, abstract composition of numerous overlapping, semi-transparent geometric shapes, primarily cubes and rectangular prisms. These shapes are rendered in a variety of colors including shades of blue, cyan, magenta, and white, creating a sense of depth and movement. The overall effect is reminiscent of a digital or futuristic cityscape, with sharp angles and a high-tech aesthetic.

LET'S GO TO THE FUTURE...

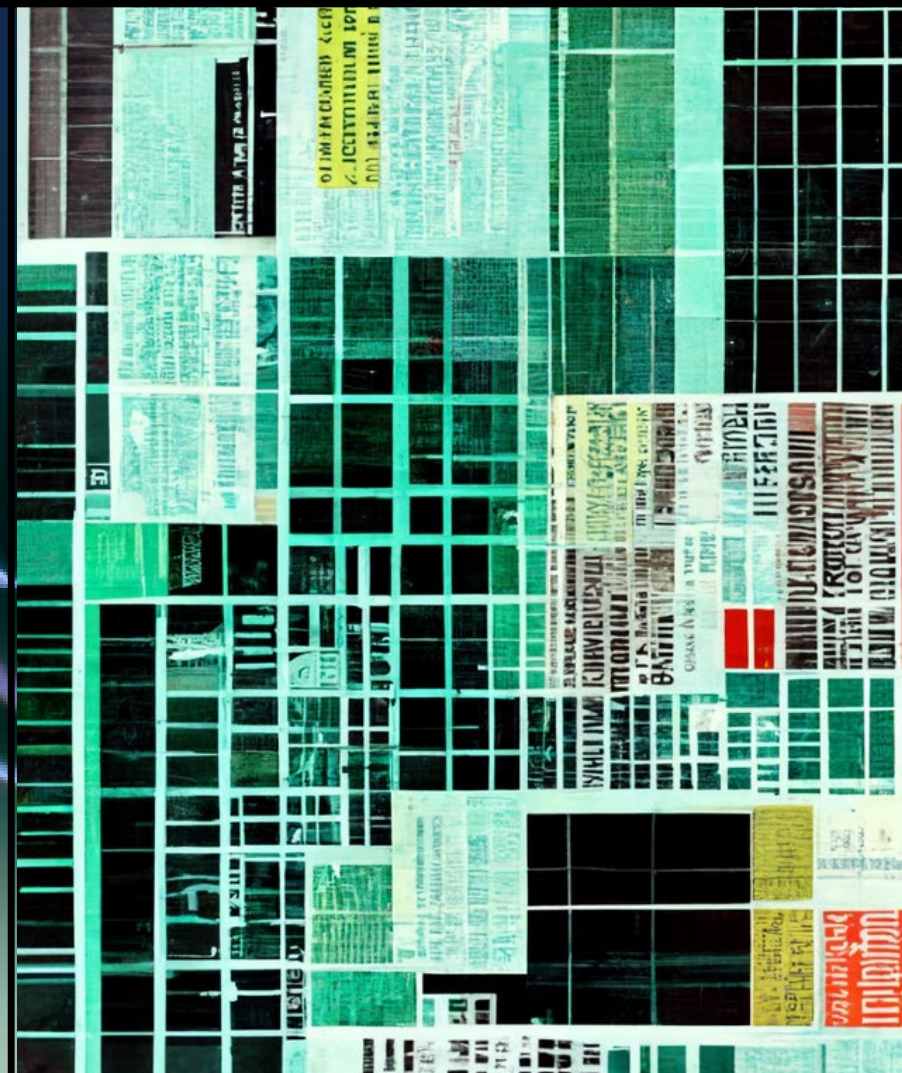
THREE PRINCIPAL TOOLS



SHEAVES



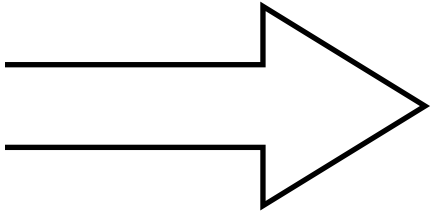
LATTICES



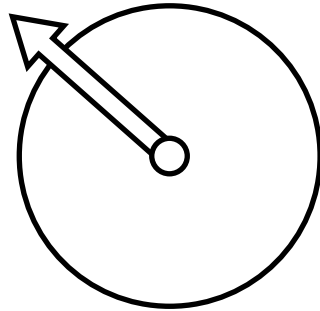
LAPLACIANS

TOWARDS REASONING

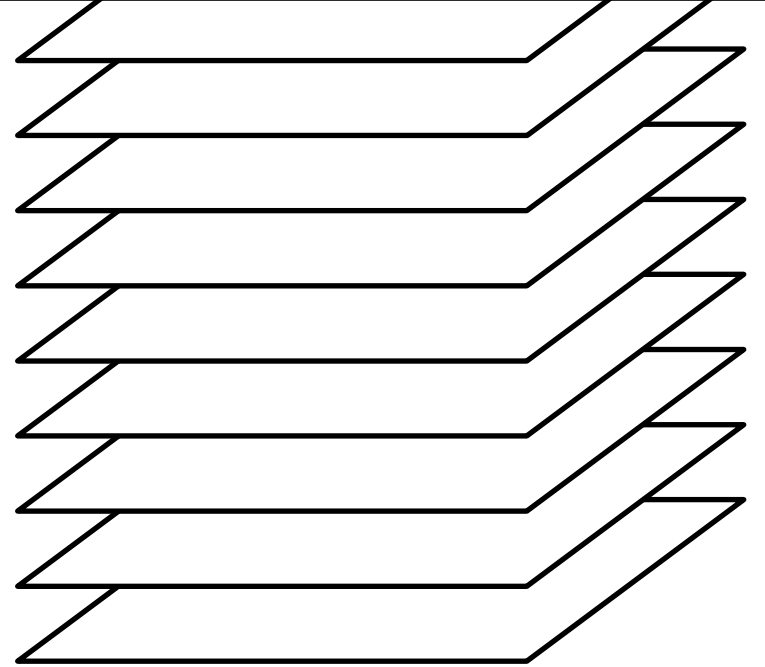
ADDITIONAL / HIGHER LATTICE STRUCTURES



SEMANTIC



RESIDUATED



CATEGORICAL

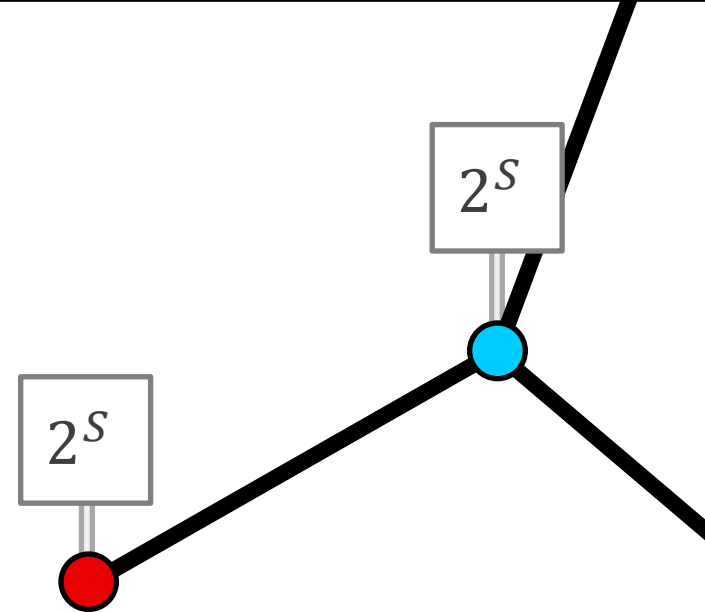
KRIPKE SEMANTICS & DISTRIBUTED REASONING

GIVEN : Kripke frame = $(S, \mathcal{K}_i \subset S \times S)$

DERIVE : Galois connections on powerset of states S

$$2^S \begin{array}{c} \xrightarrow{K_i^\exists} \\ \xleftarrow{K_i^\forall} \end{array} 2^S$$

SEMANTIC SHEAF : A graph of agents contemplating propositions S , communicating over the network to neighbors about knowledge/beliefs...



KRIPKE SEMANTICS & DISTRIBUTED REASONING

SEMANTIC SHEAF : A graph of agents contemplating propositions S , communicating over the network to neighbors about knowledge/beliefs...

DUAL TARSKI LAPLACIANS : For knowledge/possibility consensus

$$(L\sigma)_i = \bigwedge_{j \rightarrow i} \mathcal{K}_i^{\forall} \mathcal{K}_j^{\exists}(\sigma_j)$$
$$(L^* \sigma)_i = \bigvee_{j \rightarrow i} \mathcal{K}_j^{\exists} \mathcal{K}_i^{\forall}(\sigma_j)$$

HODGE-TARSKI THEOREM :
Iteration yields knowledge
(resp. possibility) consensus
over the network...

WORK IN PROGRESS

W/ PAIGE RANDALL NORTH : HANS RIESS : MIGUEL LOPEZ

THE PLAN...

LATTICE : of propositions and implications via up-sets

FUZZY LATTICE THEORY : strength of implication

cf. work of R. Bělohlávek (1999)

NEW INGREDIENT = ENRICHED CATEGORY THEORY

BOOLEAN LATTICE = category enriched in $\{0,1\}$ with all finite meets/joins

FUZZY LATTICE = category enriched in a commutative ordered monoid with all finite meets/joins (e.g., $[0,1]$ for the classic residuated theory)

WEIGHTED LIMITS/COLIMITS -->> Laplacian & Hodge-Tarski fixed point theorem

OPEN QUESTIONS/DIRECTIONS

BEYOND NETWORKS : higher-dimensional complexes & cohomologies

BEYOND DIFFUSION : waves, reaction-diffusion, patterns, drift, ...

BEYOND LATTICES : coefficients are the programming language of a sheaf

BEYOND SHEAVES : stacks! & more...

BEYOND MATHEMATICS : so much for us to do in AI/ML/GSP +

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MIGUEL LOPEZ, PHD CANDIDATE, AMCS, PENN

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