

Possibilities determine the structure of the no-signalling polytope

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Introduction

- ▶ framework for non-locality and contextuality
- ▶ general measurement scenarios

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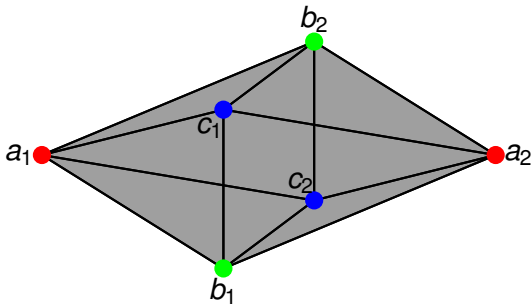
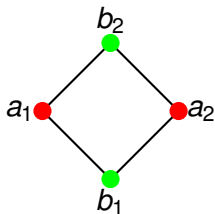
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- ▶ supports determine the **combinatorial structure** of the
no-signalling polytope.
- ▶ . . . but not quite possibilities alone!

Abramsky–Brandenburger framework

Measurement scenarios:

- ▶ a finite set of measurements X ;
- ▶ a finite set of outcomes O ;
- ▶ a cover \mathcal{M} of X , indicating **joint measurability**.



Examples: Bell-type scenarios, KS configurations, and more.

Abramsky-Brandenburger framework

No-signalling **empirical model**:

- ▶ a family $(e_C)_{C \in \mathcal{M}}$, where e_C is a probability distribution on the outcomes of measurements in context C .
- ▶ compatibility condition:

$$e_C|_{C \cap C'} = e_{C'}|_{C \cap C'}$$

(on multipartite scenarios: no-signalling)

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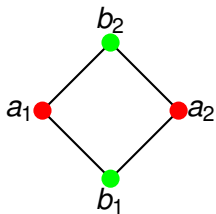
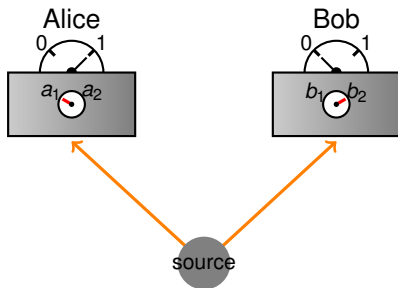
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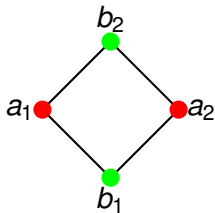
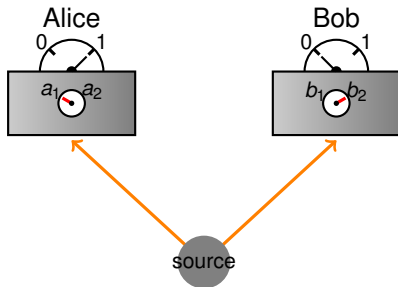
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An empirical model is **local/non-contextual** when there exists a **global distribution** p_X (i.e. for all measurements at the same time) that marginalises to all the e_C .

Bell scenario



Bell scenario



	00	01	10	11
$a_1 b_1$	$1/2$	0	0	$1/2$
$a_1 b_2$	$3/8$	$1/8$	$1/8$	$3/8$
$a_2 b_1$	$3/8$	$1/8$	$1/8$	$3/8$
$a_2 b_2$	$1/8$	$3/8$	$3/8$	$1/8$

Possibilistic collapse

- ▶ Given an empirical model e , define possibilistic model $\text{poss}(e)$ by taking the support of each distributions.
- ▶ Contains the possibilistic, or logical, information of that model.

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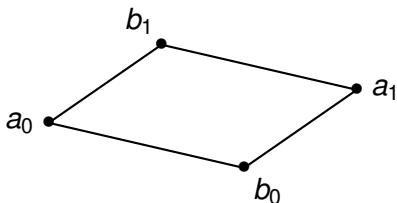
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$a_1 b_1$	$1/2$	0	0	$1/2$	\mapsto	$a_1 b_1$	1	0	0	1
$a_1 b_2$	$3/8$	$1/8$	$1/8$	$3/8$		$a_1 b_2$	1	1	1	1
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Logical contextuality: Hardy model

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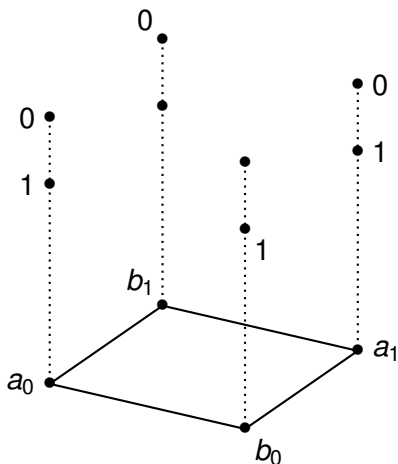
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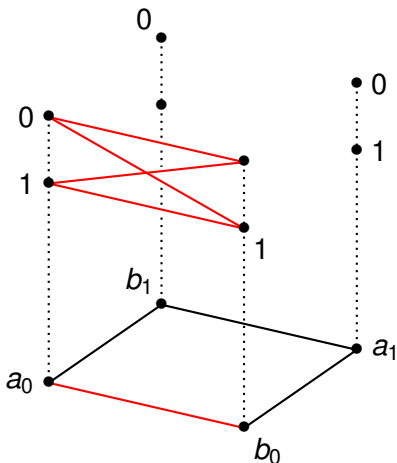
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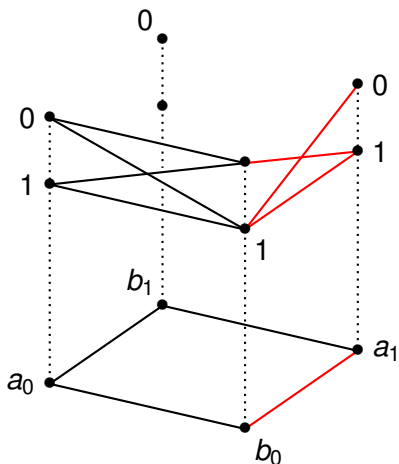
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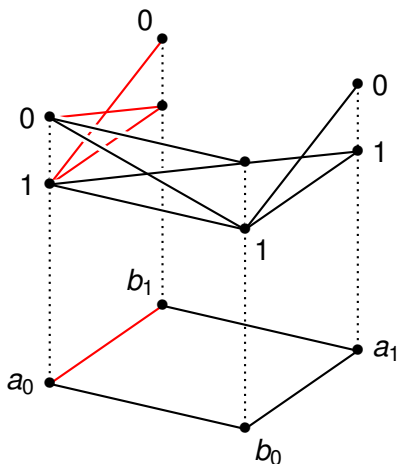
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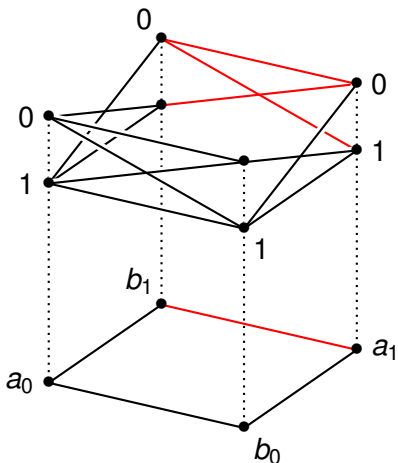
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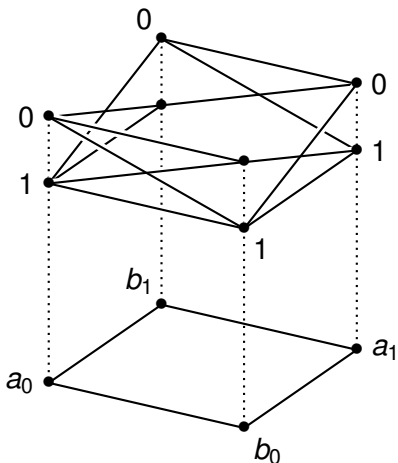
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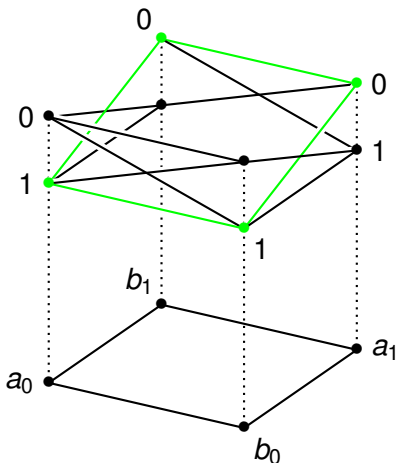
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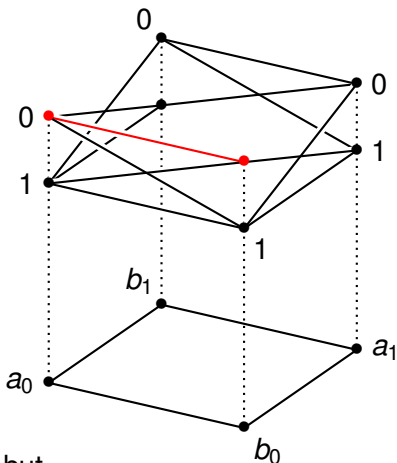
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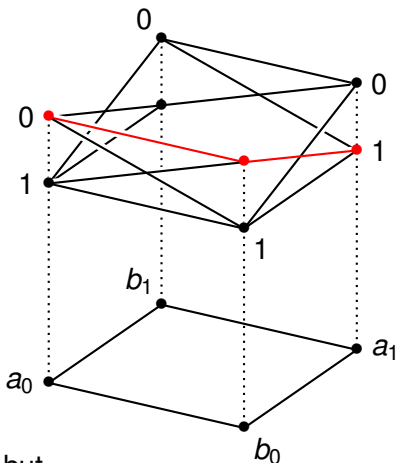
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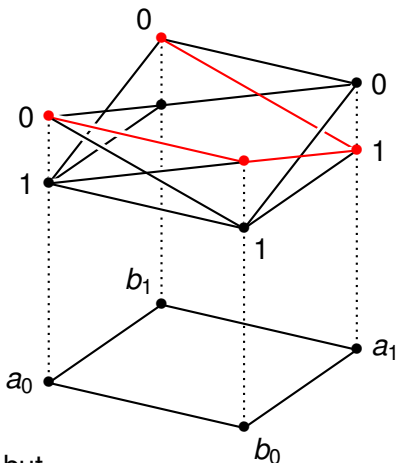
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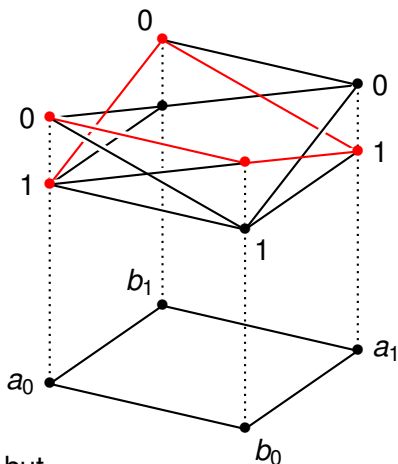
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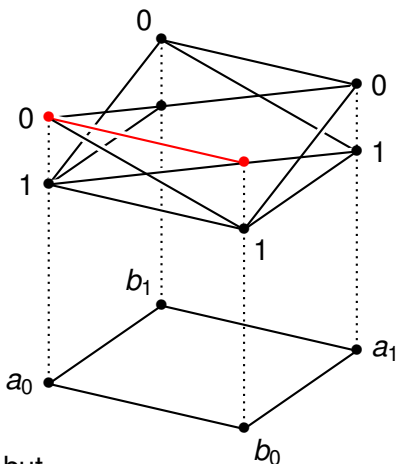
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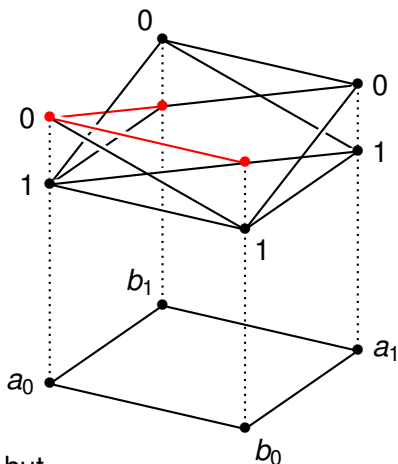
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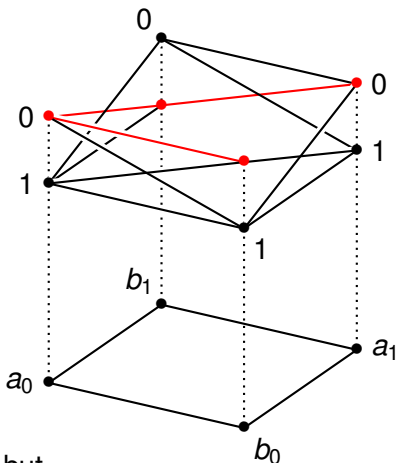
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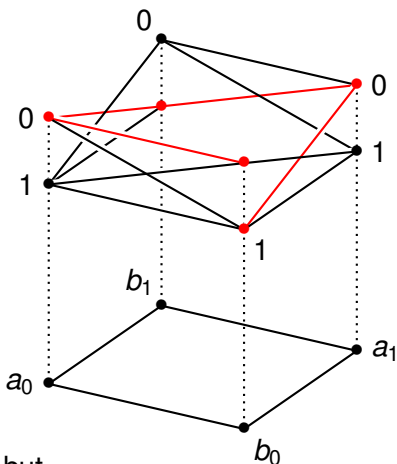
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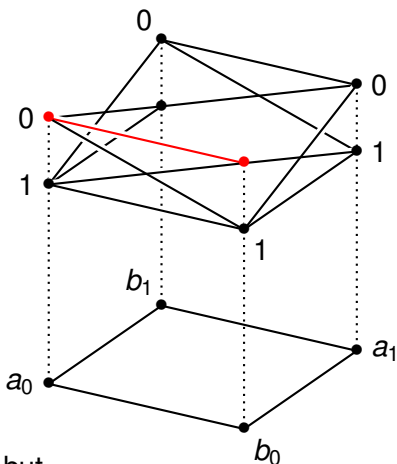
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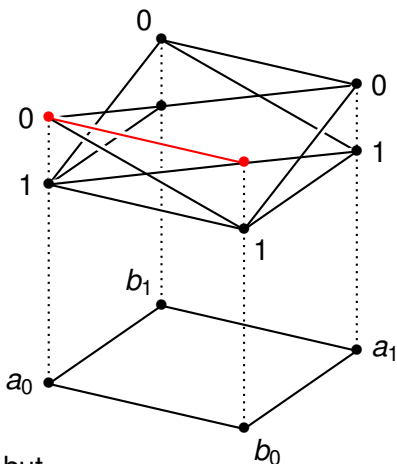


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Logical contextuality: Not all sections extend to global ones.

Also: strong contextuality, cohomological, All-versus-nothing.

The no-signalling polytope

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- ▶ explicitly represent models as points in \mathbb{R}^N , with $N = \sum_{C \in \mathcal{M}} |C|$.
- ▶ \mathcal{N} is a polytope: defined by a finite number of linear constraints.

The structure of the no-signalling polytope

- ▶ \mathcal{N} : set of probabilistic empirical models
- ▶ \mathcal{F} : the face lattice of this polytope (vertices, edges, ...)
- ▶ \mathcal{S} : possibilistic models of the form $\text{poss}(e)$ for some $e \in \mathcal{N}$
- ▶ ordered contextwise by support

Then

$$\mathcal{F} \cong \mathcal{S}_{\perp}$$

In fact, the result applies to a much wider class of polytopes.

\mathcal{N} is defined by constraints:

- ▶ Non-negativity;
- ▶ Linear equations: viz. normalisation and no-signalling.

In geometric terms: $\mathcal{N} = \mathcal{H}_{\geq \mathbf{0}} \cap \text{Aff}(\mathcal{N})$

where $\text{Aff}(\mathcal{N})$ is the affine subspace generated by \mathcal{N} ,
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For any P is **standard form**, there is an order-isomorphism between:

- ▶ $\mathcal{F}(P)$, the face lattice of P .
- ▶ $\mathcal{S}(P)$, set of “supports” of points in P , ordered by inclusion.

Polytopes

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Fundamental Theorem of Polytopes: the two notions coincide.

Face lattice

- ▶ $\mathbf{a} \cdot \mathbf{x} \geq b$ is **valid** for P if it is satisfied by every $\mathbf{x} \in P$.
- ▶ A valid inequality defines a **face** F of P :

$$F := \{\mathbf{x} \in P \mid \mathbf{a} \cdot \mathbf{x} = b\}.$$

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- ▶ It is atomistic, coatomistic, and graded.
- ▶ Meets in $\mathcal{F}(P)$ are given by intersection of faces, joins defined indirectly.

Called the **face lattice** of P , aka the **combinatorial type** of P .

Relative interior

Relative interior of a set S :

$$\text{relint}(S) = \{\mathbf{x} \in S \mid \exists \epsilon > 0. \text{Aff}(S) \cap B_\epsilon(\mathbf{x}) \subseteq S\}$$

For a convex set:

$$\text{relint}(S) = \{\mathbf{x} \in S \mid \forall \mathbf{y} \in S. \exists \epsilon > 0. (1 + \epsilon)\mathbf{x} - \epsilon\mathbf{y} \in S\}$$

Intuitively: a point \mathbf{x} is in the relative interior if the line segment $[\mathbf{y}, \mathbf{x}]$ from any point \mathbf{y} of S in to \mathbf{x} can be extended beyond \mathbf{x} while remaining in S .

Carrier face

Every polytope P can be written as the disjoint union of the relative interiors of its non-empty faces:

$$P = \bigsqcup_{F \in \mathcal{F}^+(P)} \operatorname{relint} F.$$

This means that for any polytope P we can define a map

$$\operatorname{carr} : P \longrightarrow \mathcal{F}^+(P)$$

which assigns to each point \mathbf{x} of P its **carrier face** — the unique face F such that $\mathbf{x} \in \operatorname{relint} F$.

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- ▶ $\mathcal{S}(P) := \{\text{supp } \mathbf{x} \mid \mathbf{x} \in P\}$, ordered componentwise.
- ▶ Join of \mathbf{u}, \mathbf{v} is componentwise boolean disjunction:
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- ▶ For $\mathbf{x}, \mathbf{y} \in P$ and $0 < \lambda < 1$,
 $\text{supp}(\lambda \mathbf{x} + (1 - \lambda)\mathbf{y}) = \text{supp } \mathbf{x} \vee \text{supp } \mathbf{y}$.

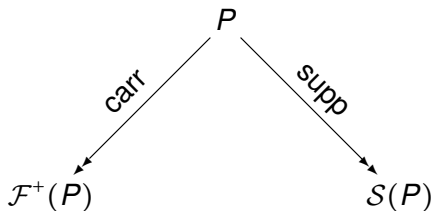
Supports

Polytope P in **standard form**: $P = \mathcal{H}_{\geq 0} \cap \text{Aff}(P)$.

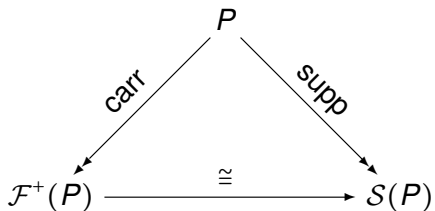
- ▶ Define a map $\text{supp} : \mathcal{H}_{\geq 0} \longrightarrow \{0, 1\}^n$:

$$(\text{supp } \mathbf{x})_i = \begin{cases} 0, & \mathbf{x}_i = 0 \\ 1, & \mathbf{x}_i > 0 \end{cases}$$

- ▶ $\mathcal{S}(P) := \{\text{supp } \mathbf{x} \mid \mathbf{x} \in P\}$, ordered componentwise.
- ▶ Join of \mathbf{u}, \mathbf{v} is componentwise boolean disjunction:
 $(\mathbf{u} \vee \mathbf{v})_i := \mathbf{u}_i \vee \mathbf{v}_i$.
- ▶ For $\mathbf{x}, \mathbf{y} \in P$ and $0 < \lambda < 1$,
 $\text{supp}(\lambda \mathbf{x} + (1 - \lambda)\mathbf{y}) = \text{supp } \mathbf{x} \vee \text{supp } \mathbf{y}$.
- ▶ So $\mathcal{S}(P)_\perp$ is a finite lattice.



WTS: $\text{carr } x \subseteq \text{carr } y \Leftrightarrow \text{supp } x \leq \text{supp } y$



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Clearly, $\mathbf{x}^\sigma \cdot \mathbf{z} \geq 0$ is valid for all $\mathbf{z} \in P$, and defines a face

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Show that $\mathbf{x} \in \text{relint } F_{\mathbf{x}}$:

- ▶ Let $\mathbf{z} \in F_{\mathbf{x}}$.
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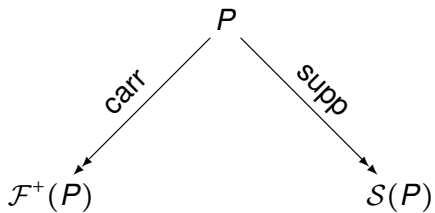
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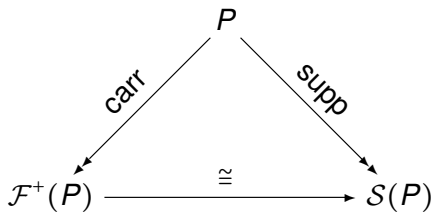
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- ▶ Let $\mathbf{z} \in F_{\mathbf{x}}$.
- ▶ Choose ϵ such that $\epsilon \mathbf{z} \leq \mathbf{x}$.
- ▶ $\mathbf{v} := (1 + \epsilon)\mathbf{x} - \epsilon \mathbf{z} \geq \mathbf{0}$.
- ▶ Hence, $\mathbf{v} \in F_{\mathbf{x}}$.



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Lattice iso: $\mathcal{F}(P) \cong \mathcal{S}(P)_\perp$

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- ▶ The vertices of the no-signalling polytope are exactly the probability models with minimal support. Moreover, there is only one probability model for each such minimal support.
- ▶ Therefore, the extremal empirical models are exactly those models which are completely and uniquely determined by their supports.
- ▶ These vertices of the polytope can be written as the disjoint union of the non-contextual, deterministic models – the vertices of the polytope of classical models – and the strongly contextual models with minimal support.

But ...

- ▶ Note the mention of support!
- ▶ We still start from probabilistic models and take their supports.

Can we characterise the combinatorial type of \mathcal{N} using **only** possibilistic notions?

- ▶ Recall that empirical models are families of **consistent distributions**.
- ▶ These can be defined over any commutative semiring R .
- ▶ $\mathbb{R}_{\geq 0}$ gives probabilistic models.
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The support lattice $\mathcal{S}(\mathcal{N}_{\mathbb{R}_{\geq 0}})$ is the image of this map.

Can we give an **intrinsic characterisation** of the image of the possibilistic collapse map, using only possibilistic notions?

$$\mathcal{S}(\mathcal{N}_{\mathbb{R}_{\geq 0}}) \neq \mathcal{N}_{\mathbb{B}}$$

i.e. there exist possibilistic empirical models that are not the support of any (probabilistic) empirical model (Abramsky, 2012).

A	B	00	01	10	11
a_1	b_1	1	0	0	1
a_1	b_2	1	1	0	1
a_2	b_1	1	0	0	1
a_2	b_2	1	0	0	1

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A	B	00	01	10	11
a_1	b_1	c	0	0	c'
a_1	b_2	d	g	0	d'
a_2	b_1	e	0	0	e'
a_2	b_2	f	0	0	f'

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- ▶ A sensible question would be: given a possibilistic empirical model, is there always a (probabilistic) empirical model whose support is at most the original one?
- ▶ That is, are minimal possibilistic models always realisable as supports?
- ▶ Also, NO!

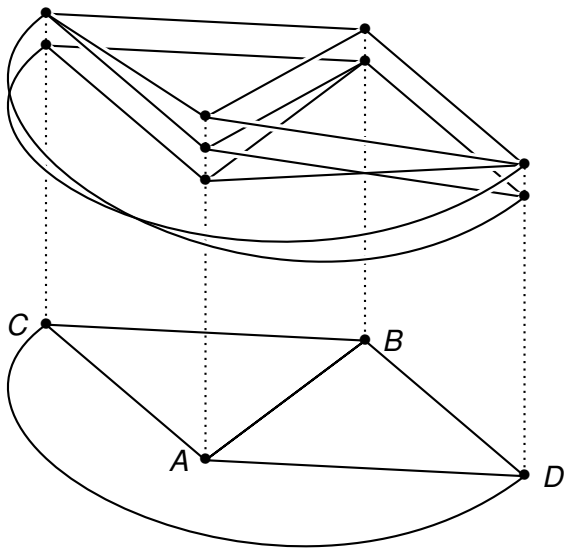
$$X = \{A, B, C, D\}$$

$$\mathcal{M} = \{\{A, B\}, \{A, C\}, \{A, D\}, \{B, C\}, \{B, D\}, \{C, D\}\}$$

$$O = \{0, 1, 2\}$$

Possible assignments:

AB	\mapsto	00,	10,	21
		a	b	c
AC	\mapsto	00,	11,	21
		d	e	f
AD	\mapsto	01,	10,	21
		k	l	m
BC	\mapsto	00,	11	
		g	h	
BD	\mapsto	00,	11	
		i	j	
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		n	o	

$a = k, \quad b = l, \quad g = i, \quad h = j, \quad c = n, \quad d = k, \quad e = l, \quad f = m$
 $c = h, \quad h = o, \quad g = n, \quad i = o, \quad j = n, \quad c = j, \quad l = o, \quad d = n.$

$$\begin{array}{l}
 AB \mapsto 00, 10, 21 \\
 \qquad \qquad a \quad b \quad c \\
 AC \mapsto 00, 11, 21 \\
 \qquad \qquad d \quad e \quad f \\
 AD \mapsto 01, 10, 21 \\
 \qquad \qquad k \quad l \quad m \\
 BC \mapsto 00, 11 \\
 \qquad \qquad g \quad h \\
 BD \mapsto 00, 11 \\
 \qquad \qquad i \quad j \\
 CD \mapsto 01, 10 \\
 \qquad \qquad n \quad o
 \end{array}$$

- ▶ All variables must be equated.
- ▶ Minimality: set any variable to zero, then all must be zero.
- ▶ Only remaining non-trivial equation is $a = a + a$.
- ▶ No non-zero, real solution!

A Bell-type example

$$X_{\text{Bell}} = \{A_1, B_1, C_1, D_1, A_2, B_2, C_2, D_2\}$$

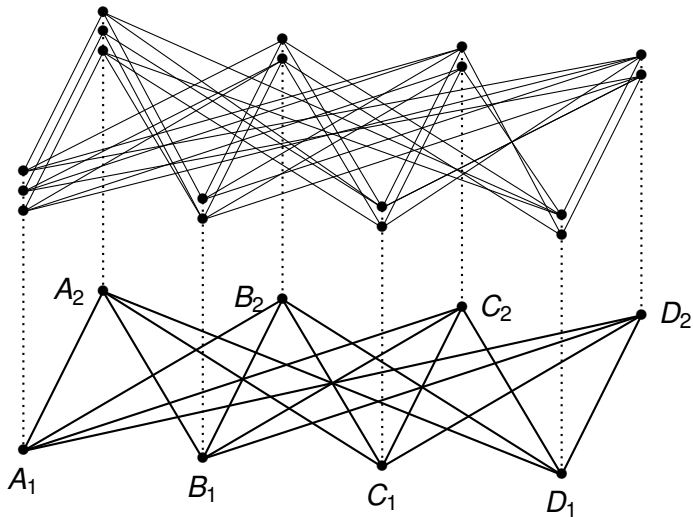
$$\mathcal{M}_{\text{Bell}} = \{A_1, B_1, C_1, D_1\} \times \{A_2, B_2, C_2, D_2\}$$

$$O = \{0, 1, 2\}$$

Possible sections:

$A_1 A_2$			\mapsto	00,	11,	22
$B_1 B_2,$	$C_1 C_2,$	$D_1 D_2$	\mapsto	00,	11	
$A_1 B_2,$	$A_2 B_1$		\mapsto	00,	10,	21
$A_1 C_2,$	$A_2 C_1$		\mapsto	00,	11,	21
$A_1 D_2,$	$A_2 D_1$		\mapsto	01,	10,	21
$B_1 C_2,$	$B_2 C_1$		\mapsto	00,	11	
$B_1 D_2,$	$B_2 D_1$		\mapsto	00,	11	
$C_1 D_2,$	$C_2 D_1$		\mapsto	01,	10	

A Bell-type example



Still an open question

Can we give an **intrinsic characterization** of the image of the possibilistic collapse map, using only possibilistic notions?