

Polyhedral Geometry of Max-Linear Conditional Independence

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The work

Polyhedral Aspects of Maxoids (FN+25)



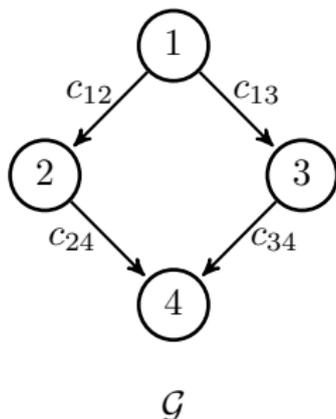
<https://arxiv.org/abs/2504.21068>

joint work with Tobias Boege, Kamillo Ferry, and Ben Hollering

Critical Paths and Polyhedral Geometry

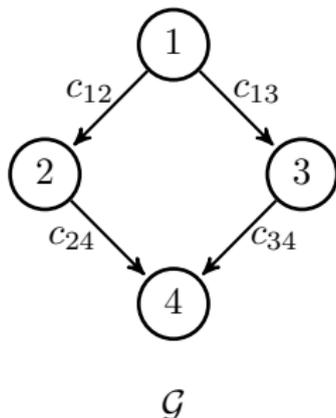
Weighted DAGs

Let $\mathcal{G} = (V, E)$ be a Directed acyclic graph (DAG), equipped with edge weights $c_{ij} \in \mathbb{R}$ for $i \rightarrow j \in E$.



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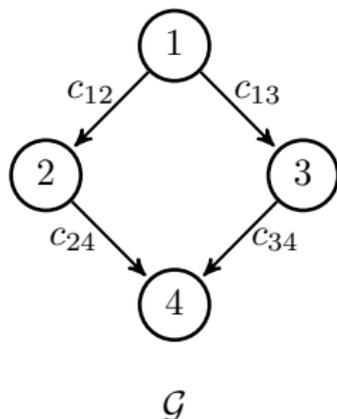


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The *coefficient matrix* of \mathcal{G}

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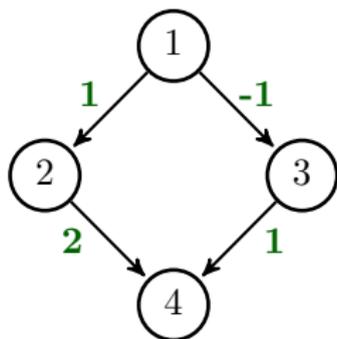
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The *weight* of a directed path from i to j in (\mathcal{G}, C) is

$$\omega_C(\pi) = \sum_{u \rightarrow v \in \pi} c_{uv}$$

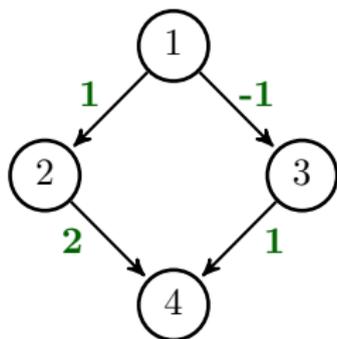
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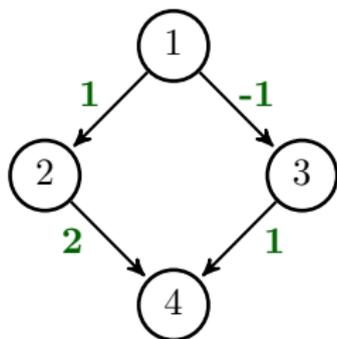
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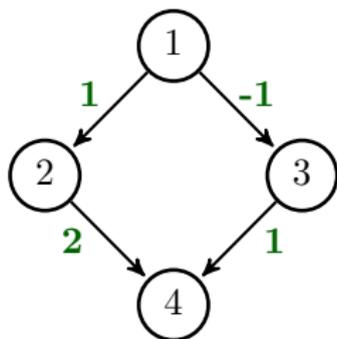
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$$\pi_{\text{crit}}^{14} = 1 \rightarrow 2 \rightarrow 4.$$

GOAL: For fixed C , characterize the set of all $C' \in \mathbb{R}^E$ which give rise to the same pairwise critical paths.

Weighted DAGs: example

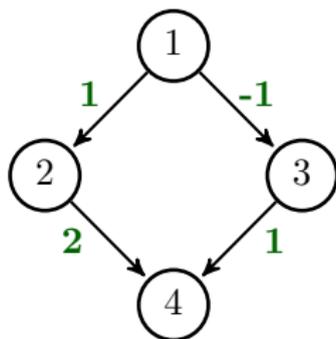


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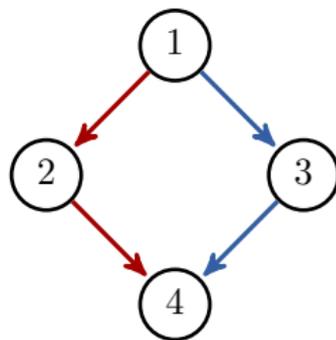
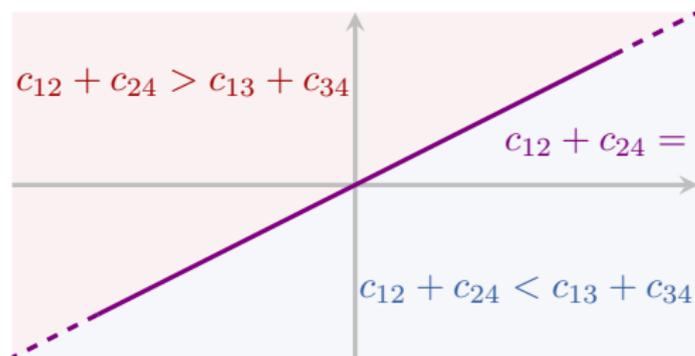
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Weighted DAGs: part II

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The critical path fan of \mathcal{G}

We call $C \in \mathbb{R}^E$ *generic* if no two paths $i \rightsquigarrow j$ have the same weight.

Theorem (FN+25, Theorem 2.9)

Let (\mathcal{G}, C) be a weighted DAG with generic $C \in \mathbb{R}^E$. The set

$$\text{cone}_{\mathcal{G}}(C) := \{C' \in \mathbb{R}^E : \pi_{crit}^{ij}(\mathcal{G}, C) = \pi_{crit}^{ij}(\mathcal{G}, C') \forall i, j \in V\} \quad (1)$$

is a *full-dimensional open polyhedral cone* defined by linear inequalities of the form

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Corollary (FN+25, Corollary 2.13)

The cones form a complete polyhedral *fan* in \mathbb{R}^E , which we denote by $\mathcal{F}_{\mathcal{G}}$.

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Path polynomials

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The weight of the critical $i - j$ path in (\mathcal{G}, C) can be computed by evaluating the max-plus *tropicalization* $F_{ij} := \text{Trop}(f_{ij})$ of f_{ij} at C :

$$F_{ij}(C) = \max_{\pi: i \rightsquigarrow j} \left\{ \sum_{u \rightarrow v \in \pi} c_{uv} \right\} = \max_{\pi: i \rightsquigarrow j} \omega_C(\pi),$$

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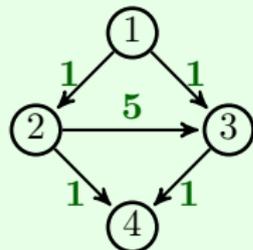
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Example



$$f_{24} = x_{24} + x_{23}x_{34}$$

$$F_{24} = \max\{x_{24}, x_{23} + x_{34}\}$$

$$F_{24}(C) = \max\{c_{24}, c_{23} + c_{34}\} = \max\{\mathbf{1}, \mathbf{5+1}\}$$

Tropical hypersurfaces

Every tropical polynomial $F \in \mathbb{R}[x_1, \dots, x_n]$ defines a *tropical hypersurface* $\mathcal{T}(F)$, which in turn induces a *polyhedral subdivision* of \mathbb{R}^n .

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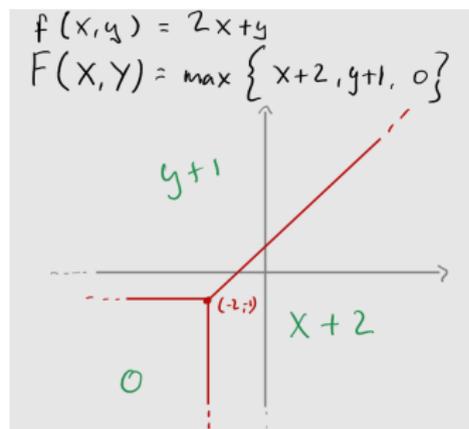
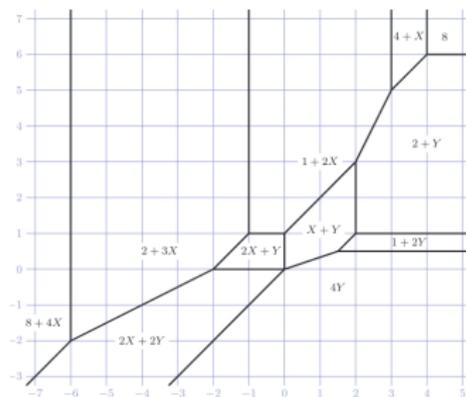


Figure: A tropical line



$$F(X, Y) = \min(8, 4 + X, 2 + Y, 1 + 2X, X + Y, 1 + 2Y, 2 + 3X, 2X + Y, X + 2Y, 4 + 3Y, 8 + 4X, 4 + 3X + Y, 2X + 2Y, 2 + X + 3Y, 4Y)$$

Figure: A tropical planar quartic. (Joswig (2021), Figure 1.4)

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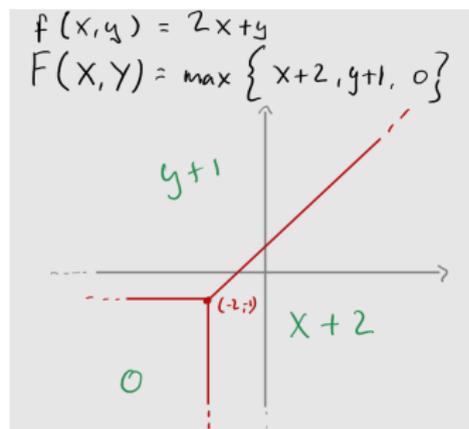


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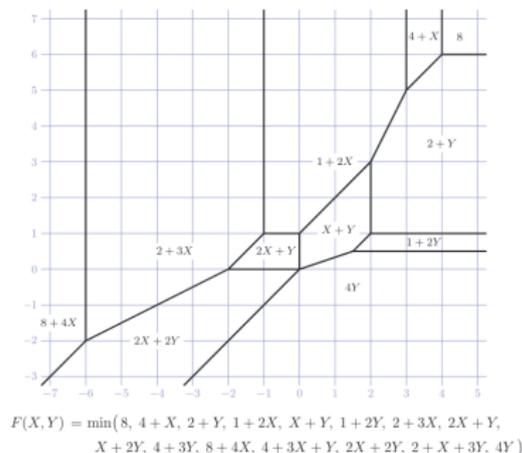


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$\mathcal{T}(F_{ij})$ induces a polyhedral subdivision of \mathbb{R}^E . For generic C it holds that

$$\pi_{\text{crit}}^{ij}(\mathcal{G}, C) = \pi \iff C \text{ lies in the interior of the region of } \mathcal{T}(F_{ij}) \text{ corresponding to } \pi.$$

Describing $\mathcal{T}(F)$

If $F = \text{Trop}(f)$ constant coefficients, (Joswig, 2021, Corollary 1.21) implies that

$\mathcal{T}(F)$ is the *Normal Fan* of $\text{Newt}(f)$,

where for $f = \sum c_\alpha x^\alpha \in \mathbb{R}[x_1, \dots, x_n]$, $\text{Newt}(f) := \text{conv}(\{\alpha \in \mathbb{Z}_{\geq 0}^n, c_\alpha \neq 0\})$

Example (Joswig (2021), Example 1.22)

Let $F(X, Y) = \max\{X, 2X, 2Y, X + 2Y, 2X + Y\}$.

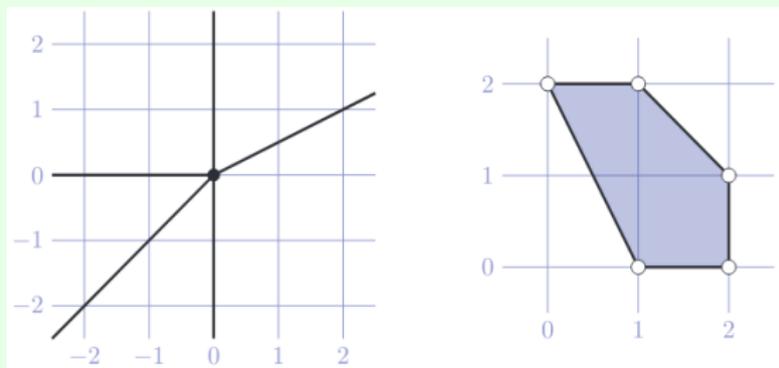


Figure: The rays of $\mathcal{T}(F)$, considered as a fan, are the inward pointing facet normals of the Newton polygon. (Source: Joswig (2021), Figure 1.7)

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Theorem (FN+25, Theorem 3.1)

The critical path fan $\mathcal{F}_{\mathcal{G}}$ coincides with the normal fan of the polytope $P_{\mathcal{G}}$, where

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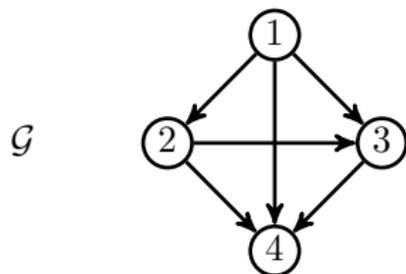
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Furthermore: the unique critical path structures of \mathcal{G} correspond bijectively to the *vertices* of $\mathcal{P}_{\mathcal{G}}$.

Example(sketch)



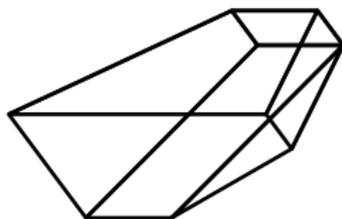
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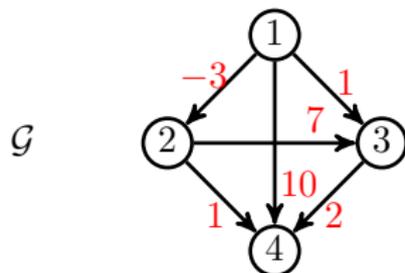
$$F_{14}(C) = \max\{c_{14}, c_{12} + c_{24}, c_{13} + c_{34}, c_{12} + c_{23} + c_{34}\}.$$

$\mathcal{F}_{\mathcal{G}} \subset \mathbb{R}^6 =$ common refinement of $\mathcal{T}(F_{13})$, $\mathcal{T}(F_{24})$, $\mathcal{T}(F_{14}) = \mathcal{N}(\mathcal{P}_{\mathcal{G}})$.



$$\mathcal{P}_{\mathcal{G}} = \text{Newt}(F_{13}) + \text{Newt}(F_{24}) + \text{Newt}(F_{14}) \subset (\mathbb{R}^6)^*$$

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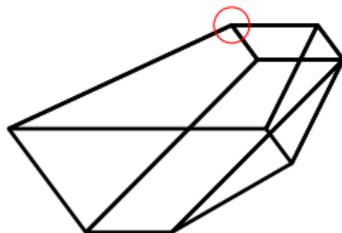
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C lies in a full-dimensional cone of $\mathcal{F}_{\mathcal{G}}$ corresponding to a *vertex* of $P_{\mathcal{G}}$.



$$P_{\mathcal{G}} = \text{Newt}(F_{13}) + \text{Newt}(F_{24}) + \text{Newt}(F_{14}) \subset (\mathbb{R}^6)^*$$

Max-Linear Conditional Independence

Max-Linear Bayesian Networks (MLBNs)

Let \mathcal{G} be DAG on n vertices with edge weights $c_{ij} \geq 0$ for $i \rightarrow j \in E(\mathcal{G})$. A random vector $X = (X_1, \dots, X_n)$ is distributed according to a max-linear model on \mathcal{G} if

$$X_j = \bigvee_{i \in \text{pa}(j)} c_{ij} X_i \vee Z_j, \quad c_{ij}, Z_j \geq 0 \quad (3)$$

where $\vee = \max$, $\text{pa}(j)$ is the set of parents of j in \mathcal{G} , and the Z_j are independent, atom-free, continuous random variables.

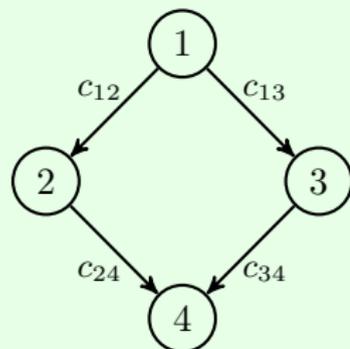
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Example



$$X_1 = Z_1$$

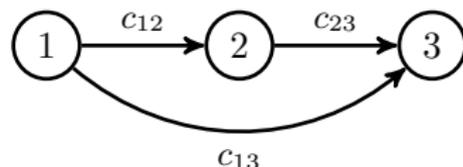
$$X_2 = c_{12} X_1 \vee Z_2$$

$$X_3 = c_{13} X_1 \vee Z_3$$

$$X_4 = c_{24} X_2 \vee c_{34} X_3 \vee Z_4$$

Challenges of the Max-Linear setting

The conditional independence structure of a MLBN depends on the choice of edge weights:

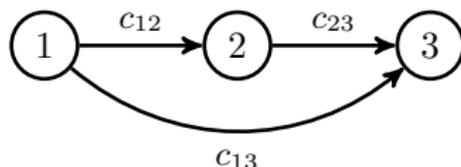


The CI statements of this model are $\begin{cases} \{1 \perp\!\!\!\perp 3|2\} & \text{if } c_{13} \leq c_{12}c_{23} \\ \emptyset & \text{if } c_{13} > c_{12}c_{23}. \end{cases}$

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This motivated the *C*-separation* criterion of Améndola et al. (2022).

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$$\{C' \in \mathbb{R}^E : \mathcal{M}_*(\mathcal{G}, C) = \mathcal{M}_*(\mathcal{G}, C')\}.$$

Lemma (FN+25, Lemma 2.2)

Let (\mathcal{G}, C) and (\mathcal{G}', C') be two weighted DAGs on the same node set and generic weights C and C' . Then

$$\mathcal{M}_*(\mathcal{G}, C) = \mathcal{M}_*(\mathcal{G}', C') \iff \pi_{crit}^{ij}(\mathcal{G}, C) = \pi_{crit}^{ij}(\mathcal{G}', C') \text{ for all } i \neq j.$$

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Lemma (FN+25, Lemma 2.2)

Let (\mathcal{G}, C) and (\mathcal{G}', C') be two weighted DAGs on the same node set and generic weights C and C' . Then

$$\mathcal{M}_*(\mathcal{G}, C) = \mathcal{M}_*(\mathcal{G}', C') \iff \pi_{crit}^{ij}(\mathcal{G}, C) = \pi_{crit}^{ij}(\mathcal{G}', C') \text{ for all } i \neq j.$$

⚠ We implicitly identify $(\mathbb{R}_{>0}^E, \vee, \cdot) \rightsquigarrow (\mathbb{R}^E, \vee, +)$ by taking logarithms.

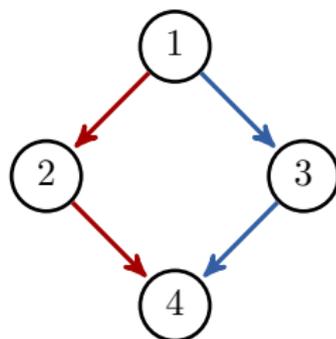
The diamond revisited

The three maxoids associated to the diamond graph are

$$\mathcal{M}_1 = \{[2 \perp\!\!\!\perp 3 \mid 1], [1 \perp\!\!\!\perp 4 \mid 2, 3], [1 \perp\!\!\!\perp 4 \mid 2]\} \quad c_{12} + c_{24} > c_{13} + c_{34}$$

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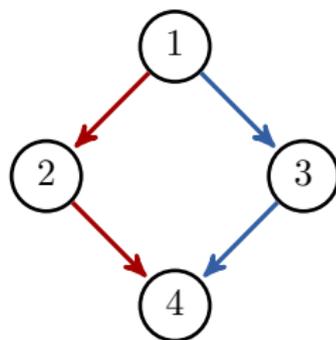
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$\mathcal{M}_3 = \mathcal{M}_1 \cup \mathcal{M}_2$. This is not a coincidence!

The Maxoid Fan

Theorem (FN+25, Corollary 2.13)

The maximal cones of $\mathcal{F}_{\mathcal{G}}$ are in bijection with the generic maxoids of \mathcal{G} , whereas lower-dimensional cones correspond to non-generic maxoids.

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$$F_1 \text{ is a face of } F_2 \implies \Phi(F_1) \supseteq \Phi(F_2) \quad \text{for all } F_1, F_2 \in \mathcal{F}_{\mathcal{G}}.$$

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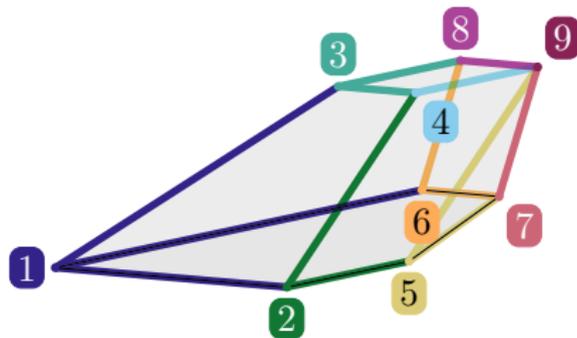
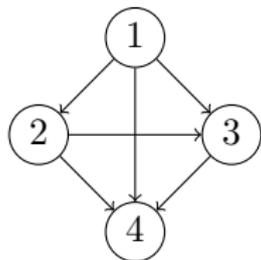
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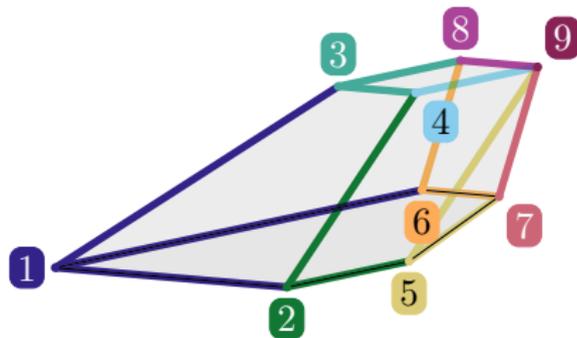
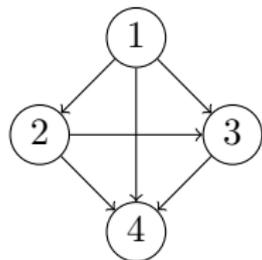
Bonus fact: the generic maxoids are in bijection with the vertices of $P_{\mathcal{G}}$. From now on, we refer to $P_{\mathcal{G}}$ and $\mathcal{F}_{\mathcal{G}}$ as the **maxoid polytope** and **maxoid fan** respectively.

Example: Continued



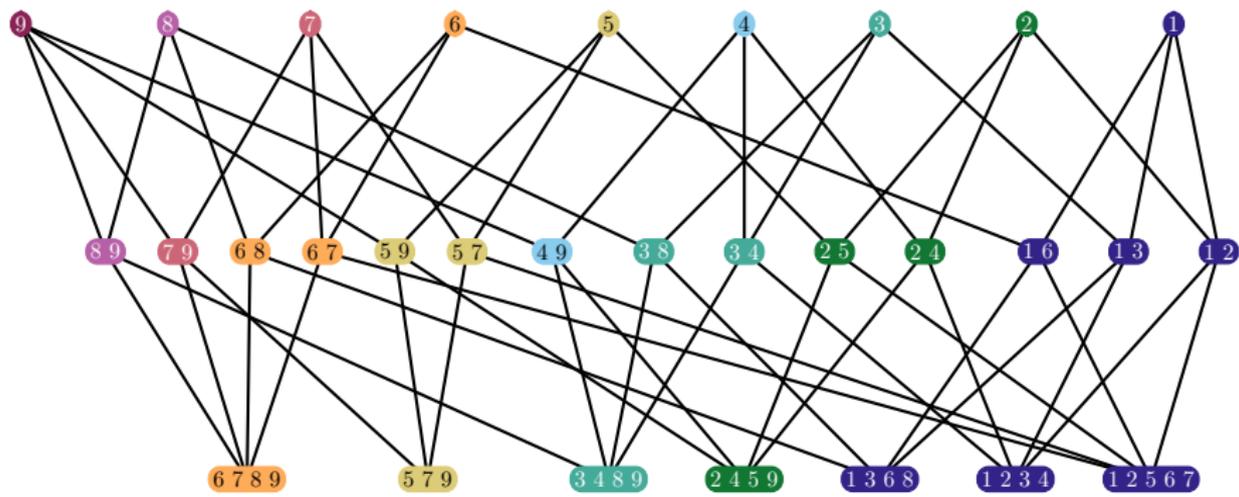
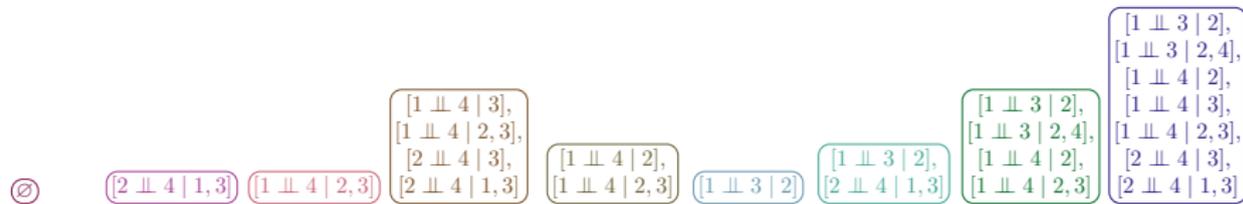
- Each colored vertex of P_G corresponds to a unique generic Max-Linear CI structure (Maxoid) associated to \mathcal{G} .

Example: Continued



- Each colored vertex of P_G corresponds to a unique generic Max-Linear CI structure (Maxoid) associated to \mathcal{G} .
- Weight matrices corresponding to higher-dimensional faces of P_G give rise to maxoids which are unions of generic maxoids.

Example: Continued



Applications

Application I: Enumeration of CI structures

The enumeration of all possible CI structures on n nodes requires computing them explicitly:

n	#TDAGs	#distinct maxoids	#distinct generic maxoids
3	3	4	4
4	18	41	40
5	181	987	892
6	32768	—	—

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Conversely, the number of maxoids which can arise from a specific DAG \mathcal{G} can be computed by constructing $P_{\mathcal{G}}$ and counting its vertices:

n	# E	$\dim(P_{\mathcal{G}})$	#vertices $P_{\mathcal{G}}$
3	3	1	2
4	6	3	9
5	10	6	103
6	15	10	3324

Table: Data for the perfect DAG on n nodes (i.e. $E = \{(i, j) : 1 \leq i < j \leq n\}$), for $3 \leq n \leq 6$

Application II: CI implication

The *local CI implication problem* states:

If a set of CI statements S holds in \mathcal{G} , does $[I \perp\!\!\!\perp J \mid K] \notin S$ also hold in \mathcal{G} ?

This problem can be solved using polyhedral methods.

```
julia> using Maxoids, Oscar;
julia> G = complete_DAG(4);
julia> maxoid_implication(G, [ CI"24|13" ] => [ CI"14|3" ])
(false, [-Inf 1/4 0 1/2; -Inf -Inf 0 0; -Inf -Inf -Inf 0; -Inf -Inf -Inf -Inf])
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Figure: Code snippet from Boege et al. (2025), which queries a CI implication using our package Maxoids.jl (<https://doi.org/10.5281/zenodo.17737553>).

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Generally, Maxoids are *compositional graphoids* which additionally fulfill:

$$\mathbf{Amalgamation:} \quad [i \perp\!\!\!\perp j \mid KM] \wedge [i \perp\!\!\!\perp j \mid LM] \implies [i \perp\!\!\!\perp j \mid KLM]$$

$$\mathbf{Strong Spohn:} \quad [i \perp\!\!\!\perp j \mid klm] \wedge [k \perp\!\!\!\perp l \mid iM] \wedge [k \perp\!\!\!\perp l \mid jM] \implies [k \perp\!\!\!\perp l \mid M]$$

$$[i \perp\!\!\!\perp j \mid klm] \wedge [k \perp\!\!\!\perp l \mid iM] \wedge [k \perp\!\!\!\perp l \mid M] \implies [k \perp\!\!\!\perp l \mid jM]$$

Thank you! Questions?



<https://arxiv.org/abs/2504.21068>

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- Boege, T., Ferry, K., Hollering, B., and Nowell, F. (2025). Polyhedral Aspects of Maxoids. arXiv:2504.21068.
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