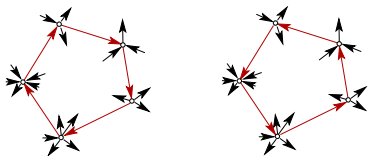


α -Orientations

Definition. Given $G = (V, E)$ and $\alpha : V \rightarrow \mathbf{N}$.

An α -orientation of G is an orientation with $\text{outdeg}(v) = \alpha(v)$ for all v .

- Reverting directed cycles preserves α -orientations.



Theorem. The set of α -orientations of a planar graph G has the structure of a distributive lattice.

Proof I: Essential Cycles

For the proof we assume that G is 2-connected.

Definition.

A cycle C of G is an **essential cycle** if

- C is chord-free and simple,
- the interior cut of C is **rigid**,
- there is an α -orientation X such that C is directed in X .

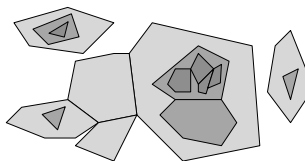
Lemma.

C is non-essential $\iff C$ has a directed chordal path in every α -orientation.

Proof II

Lemma.

Essential cycles are interiorly disjoint or contained.



Lemma.

If C is a directed cycle of X , then X^C can be obtained by a sequence of reversals of essential cycles.

Lemma.

If (C_1, \dots, C_k) is a flip sequence (ccw \rightarrow cw) on X then for every edge e the essential cycles $C^{l(e)}$ and $C^{r(e)}$ alternate in the sequence.

Proof III: Flip Sequences

Lemma.

The length of any flip sequence ($\text{ccw} \rightarrow \text{cw}$) is bounded and there is a unique α -orientation X_{\min} with the property that all cycles in X_{\min} are cw-cycles.

- $Y \prec X$ if a flip sequence $X \rightarrow Y$ exists.

Lemma.

Let $Y \prec X$ and C be an essential cycle. Every sequence $S = (C_1, \dots, C_k)$ of flips that transforms X into Y contains the same number of flips at C .

Proof IV: Potentials

Definition. An α -potential for G is a mapping $\wp : \text{Ess}_\alpha \rightarrow \mathbf{N}$ such that

- $|\wp(C) - \wp(C')| \leq 1$, if C and C' share an edge e .
- $\wp(C^{l(e)}) \leq \wp(C^{r(e)})$ for all e (orientation from X_{\min})

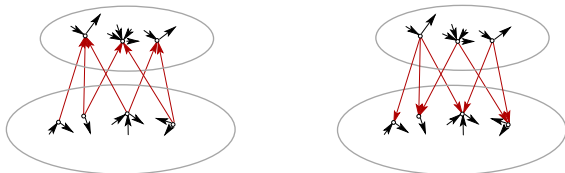
Lemma. There is a bijection between α -potentials and α -orientations.

Theorem. α -potentials are a distributive lattice with

- $(\wp_1 \vee \wp_2)(C) = \max\{\wp_1(C), \wp_2(C)\}$ and
- $(\wp_1 \wedge \wp_2)(C) = \min\{\wp_1(C), \wp_2(C)\}$ for all essential C .

A Dual Construction: c-Orientations

- Reorientations of directed cuts preserve **flow-difference** ($\#$ forward arcs $-$ $\#$ backward arcs) along cycles.



Theorem [Propp 1993]. The set of all orientations of a graph with prescribed flow-difference for all cycles has the structure of a distributive lattice.

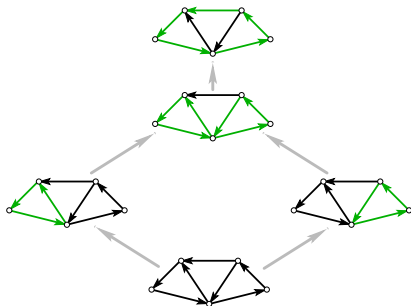
- Diagram edge \sim push a vertex ($\neq v_i$).

Circulations in Planar Graphs

Theorem [Khuller, Naor and Klein 1993].

The set of all integral flows respecting capacity constraints ($\ell(e) \leq f(e) \leq u(e)$) of a planar graph has the structure of a distributive lattice.

$$0 \leq f(e) \leq 1$$



- Diagram edge \sim add or subtract a unit of flow in ccw oriented facial cycle.

Δ -Bonds

$G = (V, E)$ a connected graph with a prescribed orientation.
With $x \in \mathbb{Z}^E$ and C cycle we define the circular flow difference

$$\Delta_x(C) := \sum_{e \in C^+} x(e) - \sum_{e \in C^-} x(e).$$

With $\Delta \in \mathbb{Z}^C$ and $l, u \in \mathbb{Z}^E$ define

$$\mathcal{B}_G(\Delta, l, u) = \{x \in \mathbb{Z}^E : \Delta_x = \Delta \text{ and } l \leq x \leq u\}.$$

Δ -Bonds as Generalization

$\mathcal{B}_G(\Delta, \ell, u)$ is the set of $x \in \mathbb{Z}^E$ such that

- $\Delta_x = \Delta$ (circular flow difference)
- $\ell \leq x \leq u$ (capacity constraints).

Special cases:

- c -orientations are $\mathcal{B}_G(\Delta, 0, 1)$
($\Delta(C) = \frac{1}{2}(|C^+| - |C^-| - c(C))$).
- Circular flows on planar G are $\mathcal{B}_{G^*}(0, \ell, u)$
(G^* the dual of G).
- α -orientations.

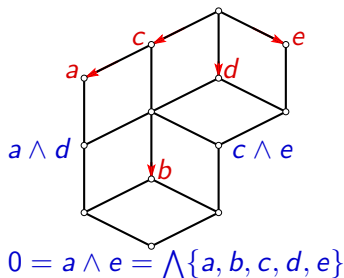
ULD Lattices

Definition. [Dilworth]

A lattice is an **upper locally distributive lattice (ULD)** if each element has a unique minimal representation as meet of meet-irreducibles.

i.e., there is a unique mapping $x \rightarrow M_x$ such that

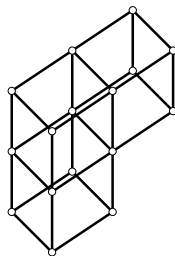
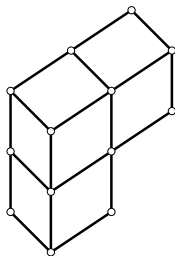
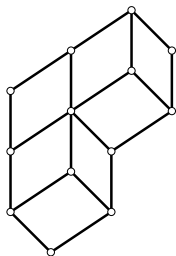
- $x = \bigwedge M_x$ (representation.)
- $x \neq \bigwedge A$ for all $A \subsetneq M_x$ (minimal).



ULD vs. Distributive

Proposition.

A lattice is ULD and LLD \iff it is distributive.



Diagrams of ULD lattices: A Characterization

A coloring of the edges of a digraph is a U -coloring iff

- arcs leaving a vertex have different colors.
- completion property:

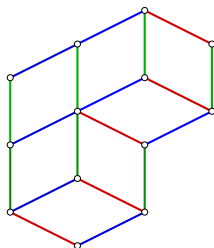


Theorem.

A digraph D is acyclic, has a unique source and admits a U -coloring $\iff D$ is the diagram of an ULD lattice.

\hookrightarrow Unique **1**.

Examples of U-colorings



- Δ -bond lattices, colors are the names of pushed vertices. (Connected, unique $\mathbf{0}$).
- Chip firing game with a fixed starting position (the source), colors are the names of fired vertices.

More Examples

Some LLD lattices with respect to inclusion order:

- Subtrees of a tree (Boulaye '67).
- Convex subsets of posets (Birkhoff and Bennett '85).
- Convex subgraphs of acyclic digraphs (Pfaltz '71).
(C is convex if with x, y all directed (x, y) -paths are in C).
- Convex sets of an abstract convex geometry (Edelman '80).
(This is an universal family of examples).

Outline

Orders and Lattices

Definitions

The Fundamental Theorem

Dimension and Planarity

Lattices and Graphs

α -orientations

Δ -Bonds and Further Examples

The ULD-Theorem

Distributive Lattices and Markov Chains

Coupling from the Past

Mixing time on α -orientations

A General problem: Sampling

- Ω a (large) finite set
- $\mu : \Omega \rightarrow [0, 1]$ a probability distribution, e.g. uniform distr.

Problem. Sample from Ω according to μ .

i.e., $\Pr(\text{output is } \omega) = \mu(\omega)$.

There are many hard instances of the sampling problem.

Relaxation: *Approximate sampling*

i.e., $\Pr(\text{output is } \omega) = \tilde{\mu}(\omega)$ for some $\tilde{\mu} \approx \mu$.

Applications of (approximate) sampling:

- Get hand on typical examples from Ω .
- Approximate counting.

Preliminaries on Markov Chains

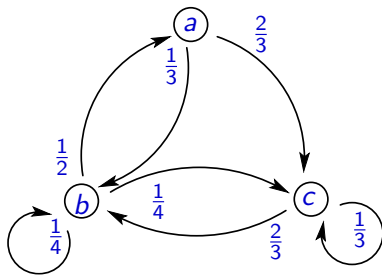
M transition matrix

- format $\Omega \times \Omega$
- entries $\in [0, 1]$
- row sums = 1 (stochastic)

Intuition:

$$\mathbf{M} = \begin{pmatrix} 0 & \frac{1}{3} & \frac{2}{3} \\ \frac{1}{2} & \frac{1}{4} & \frac{1}{4} \\ 0 & \frac{2}{3} & \frac{1}{3} \end{pmatrix}$$

M specifies a random walk



Ergodic Markov Chains

M is **ergodic** (i.e., irreducible and aperiodic)

\implies multiplicity of eigenvalue **1** is one

\implies unique π with $\pi = \pi \mathbf{M}$.

Fundamental Theorem.

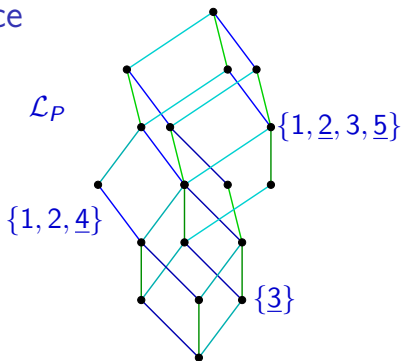
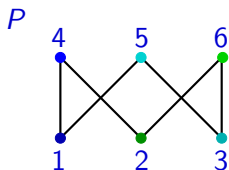
$$\mathbf{M} \text{ ergodic} \implies \lim_{t \rightarrow \infty} \mu_0 \mathbf{M}^t = \pi.$$

M symmetric and ergodic

$\implies \mathbf{M}^T \mathbb{1}^T = \mathbf{M} \mathbb{1}^T = \mathbb{1}^T$, hence $\mathbb{1} \mathbf{M} = \mathbb{1}$

$\implies \pi$ is the uniform distribution.

Example: Distributive Lattice



Lattice Walk

(A *natural* Markov chain on \mathcal{L}_P)

Identify state with downset D

- choose $x \in P$ & choose $s \in \{\uparrow, \downarrow\}$
- depending on s move to $D + x$ or $D - x$ (if possible)

Fact. The chain is ergodic and symmetric, i.e, π is uniform.

Mixing Time

$\mu_x^t = \delta_x \mathbf{M}^t$ the distrib. after t steps when start is in x

$$\Delta(t) := \max(\|\mu_x^t - \pi\|_{VD} : x \in \Omega)$$

$$\tau(\varepsilon) = \min(t : \Delta(t) \leq \varepsilon)$$

- $\tau(\varepsilon)$ is the **mixing time**.
- \mathbf{M} is **rapidly mixing** $\iff \tau(\varepsilon)$ is a polynomial function of $\log(\varepsilon^{-1})$ and the *problem size*.

Big Challenge. Find interesting rapidly mixing Markov chains

Example.

- Matchings (Jerrum & Sinclair '88)
- Linear Extensions (Karzanov & Khachiyan '91 / Bubley & Dyer '99)
- Planar Lattice Structures, e.g. Dimer Tilings (Luby et al. '93)