

Mixed minors, compact representations and χ -boundedness

Marek Sokołowski

25 May 2023

PART 1

MIXED MINORS

Twin-width and matrices

A 0/1-matrix can be:

Twin-width and matrices

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$$\begin{array}{cccc} 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{array}$$

horizontal

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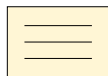
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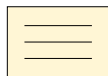


Twin-width and matrices

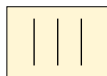
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mixed



Note: mixed \iff has a 2×2 contiguous mixed submatrix (**corner**).

Twin-width and matrices

Divisions

Division \mathcal{D} – partitioning of columns and rows into intervals (*blocks*).

0	1		0	0	0		0	0	1	1
1	1		0	0	1		1	1	0	0
<hr/>										
0	0		0	0	0		0	0	0	0
0	0		0	0	0		0	0	0	0
0	1		0	1	0		1	0	1	0
<hr/>										
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0	1		0	1	0		1	0	1	0
<hr/>										
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Zone – intersection of a row block and a column block

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\mathcal{D} is a **mixed minor** if each zone of \mathcal{D} is mixed.

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Mixed freeness

Matrix M is **d -mixed free** if it has no $d \times d$ mixed minor.

Grid theorem for twin-width

Mixed freeness

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Theorem (Twin-width I)

Let $d \in \mathbb{N}$ be an integer and G be a graph. Then:

- $\text{tww}(G) \leq d \implies G$ has a **$(2d + 2)$ -mixed free** adjacency matrix.
- G has a **d -mixed free** adjacency matrix $\implies \text{tww}(G) \leq 2^{2^{O(d)}}$.

Marcus–Tardos theorem and twin-width

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Theorem (Twin-width I, “Marcus–Tardos”)

If: M – a d -mixed free matrix,

\mathcal{D} – an $n \times n$ division of M

$\Rightarrow \mathcal{D}$ has at most $c_d \cdot n$ mixed zones ($c_d = \text{const}(d)$).

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Number of mixed zones: **linear** instead of **quadratic**!

PART 2

COMPACT REPRESENTATIONS

Pilipczuk, Sokołowski, Zych-Pawlewicz,
Compact Representation for Matrices of Bounded Twin-Width

Twin-width of matrices

Twin-width of matrices

t -twin-ordered matrices

0	1	1	0	0	0
1	0	1	1	1	0
0	0	1	0	0	0
0	0	1	0	0	1
1	1	1	1	1	1
0	0	0	1	1	1

Maximum \times 's in any row/column now: 0

Maximum so far: 0

Twin-width of matrices

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0	1	1	0	0	0
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Twin-width of matrices

t -twin-ordered matrices

0	1	1	0	0
1	0	1	1	0
0	0	1	0	×
1	1	1	1	1
0	0	0	1	1

Maximum ×'s in any row/column now: 1

Maximum so far: 1

Twin-width of matrices

t -twin-ordered matrices

0	1	1	0	0
1	0	1	1	0
0	0	1	0	×
1	1	1	1	1
0	0	0	1	1

Maximum ×'s in any row/column now: 1

Maximum so far: 1

Twin-width of matrices

t -twin-ordered matrices

0	1	1	0	0
1	0	1	1	0
0	0	1	0	×
1	1	1	1	1
0	0	0	1	1

Maximum ×'s in any row/column now: 1

Maximum so far: 1

Twin-width of matrices

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×	1	0	0
×	1	1	0
0	1	0	×
1	1	1	1
0	0	1	1

Maximum ×'s in any row/column now: 2

Maximum so far: 2

Twin-width of matrices

t -twin-ordered matrices

×	1	0	0
×	1	1	0
0	1	0	×
1	1	1	1
0	0	1	1

Maximum ×'s in any row/column now: 2

Maximum so far: 2

Twin-width of matrices

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×	1	×	0
0	1	0	×
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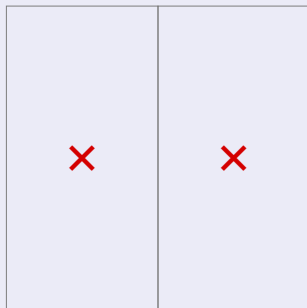
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×	1

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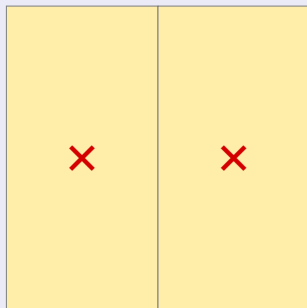


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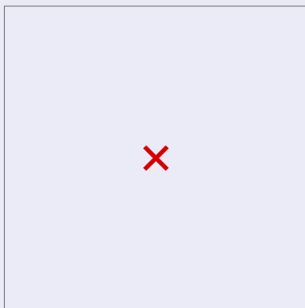
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Red number = maximum \times 's in any row/column during the contraction.

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Note: M is d -twin-ordered $\implies M$ is $(2d + 2)$ -mixed-free.

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Compact: bitsize $\mathcal{O}(S)$ bits if $S =$ information-theoretic min bitsize.

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≈ Twin-width II

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<i>adjacency labeling (Twin-width II)</i>	$\mathcal{O}_d(n \log n)$	$\mathcal{O}_d(\log n)$

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our result (PSZ-P, 2022)	$\mathcal{O}_d(n)$	$\mathcal{O}_d(\log \log n)$

Different zones in a division

M — a d -twin-ordered $n \times n$ matrix;

$s \mid n$;

\mathcal{D} — an $\frac{n}{s} \times \frac{n}{s}$ division of M where each zone is an $s \times s$ submatrix.

1	0	0	0	0	0	0	0	0	0	0	0	1	1	1	1	1	1	1
0	1	0	0	0	0	0	0	0	0	0	0	0	1	1	1	1	1	1
0	0	1	0	0	0	0	0	0	0	0	0	0	0	1	1	1	1	1
0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	1	1	1	1
0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	1	1	1
0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	1	1
0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	1
0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0
1	1	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0
1	1	1	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0
1	1	1	1	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0
1	1	1	1	1	0	0	0	0	0	0	0	0	0	0	0	1	0	0
1	1	1	1	1	1	0	0	0	0	0	0	0	0	0	0	0	1	0
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\mathcal{D} has at most $2^{\mathcal{O}_d(s)}$ different zones (*Twin-width II*).

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- “Marcus–Tardos”: at most $\mathcal{O}_d\left(\frac{n}{s}\right)$ **mixed** zones in total;
- Now (blackboard): at most $\mathcal{O}_d\left(\frac{n}{s}\right)$ different **non-mixed** zones.

Data structure

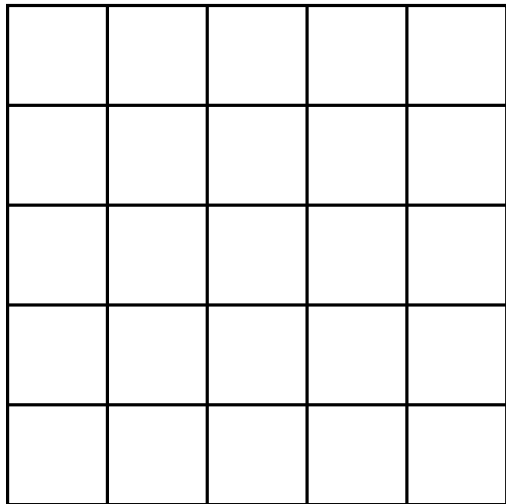
Reminder

- Fixed: $d \in \mathbb{N}$.
- Input: M – an $n \times n$ matrix that is d -twin-ordered.
- Target:
 - $\mathcal{O}_d(n)$ bits of memory,
 - $\mathcal{O}(\log \log n)$ per query.



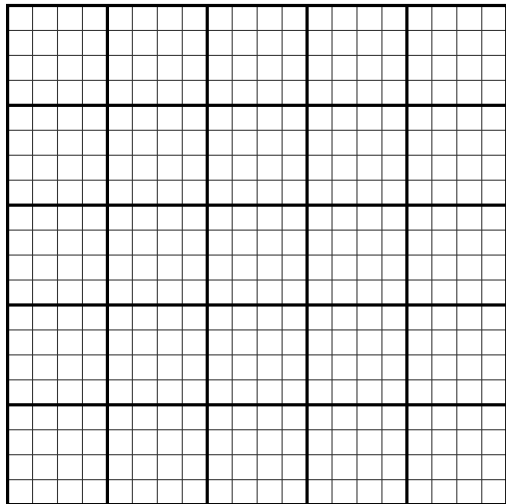
M

\mathcal{D}_1 – a division of M where each zone is an $n^{2/3} \times n^{2/3}$ submatrix.



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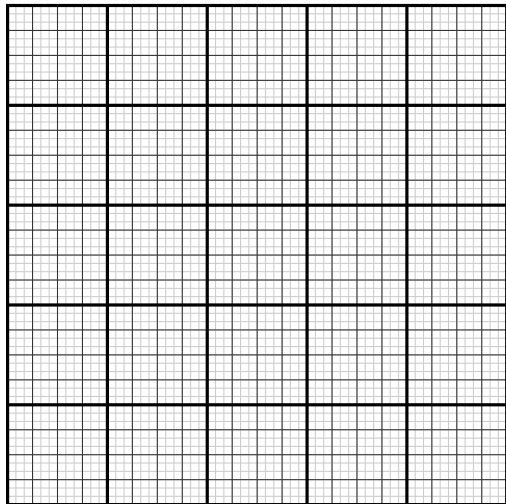
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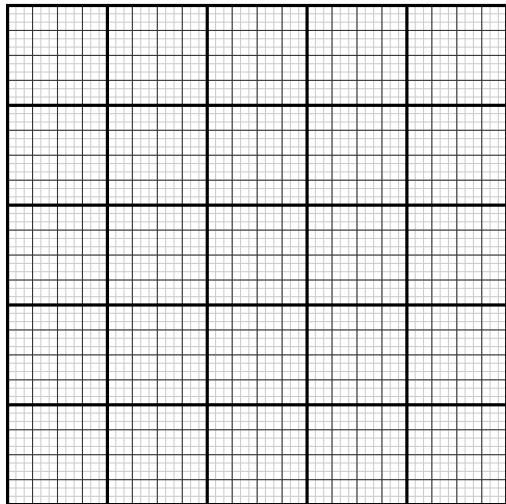


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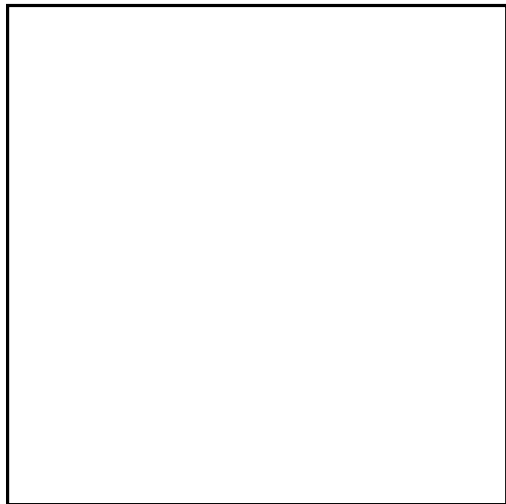
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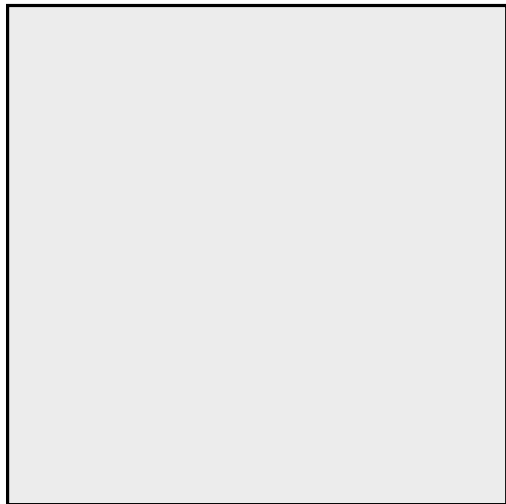
$\mathcal{D}_{\text{last}}$ – a division where each zone is an $\mathcal{O}_d(\log n) \times \mathcal{O}_d(\log n)$ submatrix.



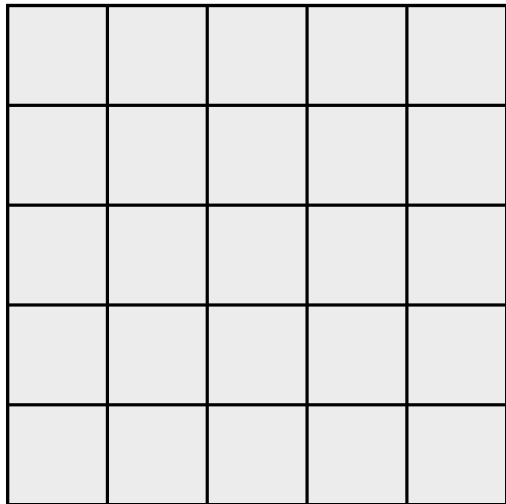
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Next, mark unique zones in each division.



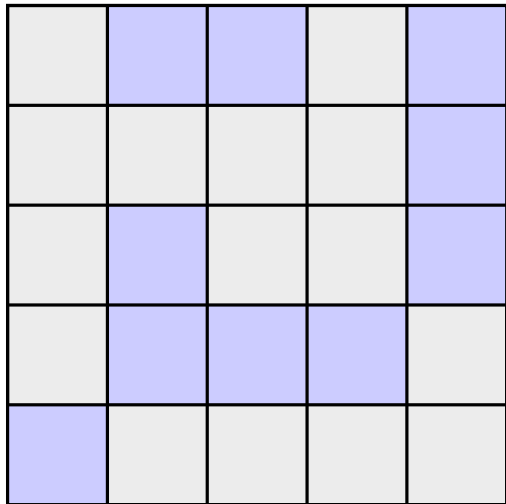
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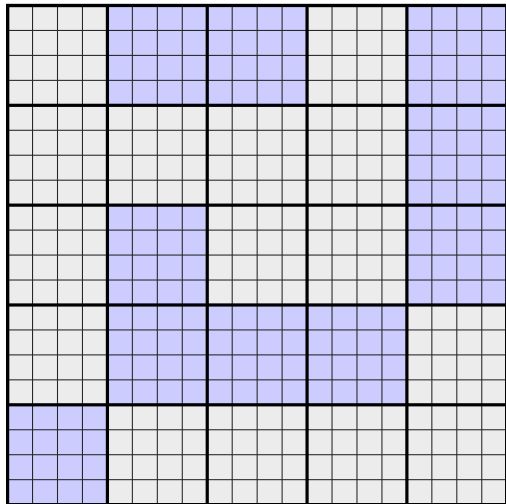
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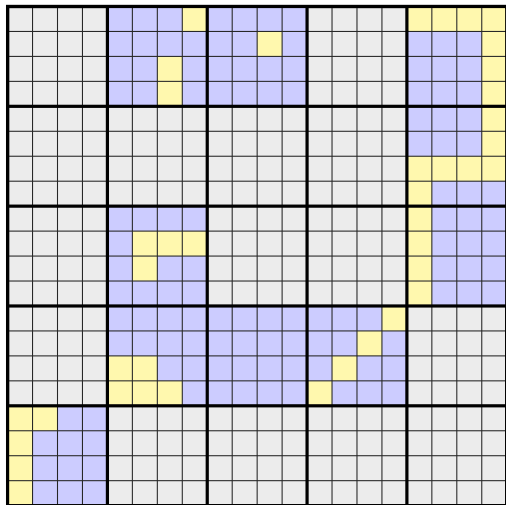
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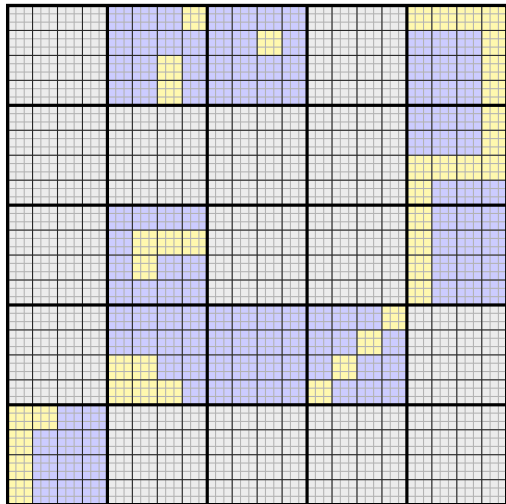
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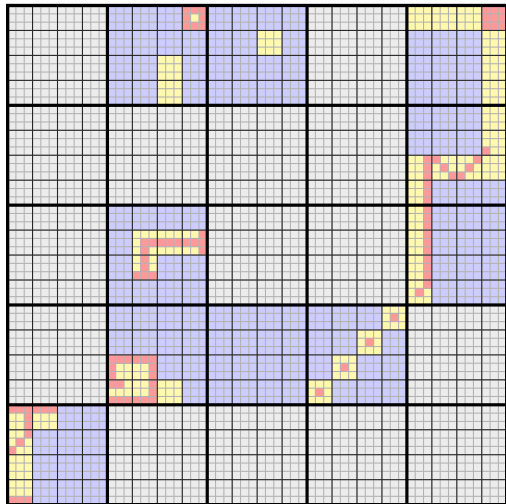
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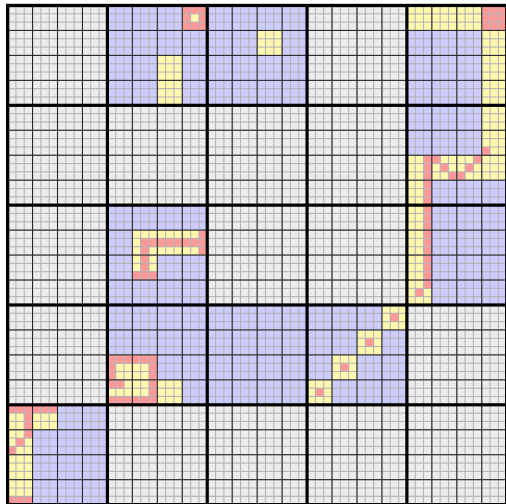


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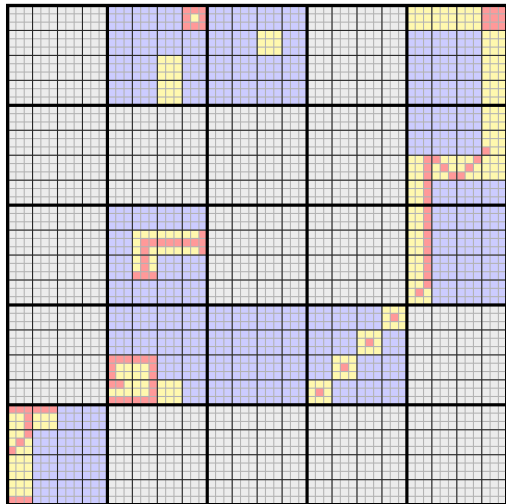
\mathcal{D}_k has at most $\mathcal{O}_d(n/n^{(2/3)^k})$ unique zones. $(n/n, n/n^{2/3}, n/n^{4/9}, \dots)$



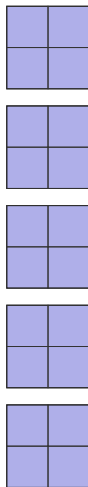
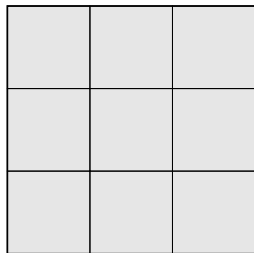
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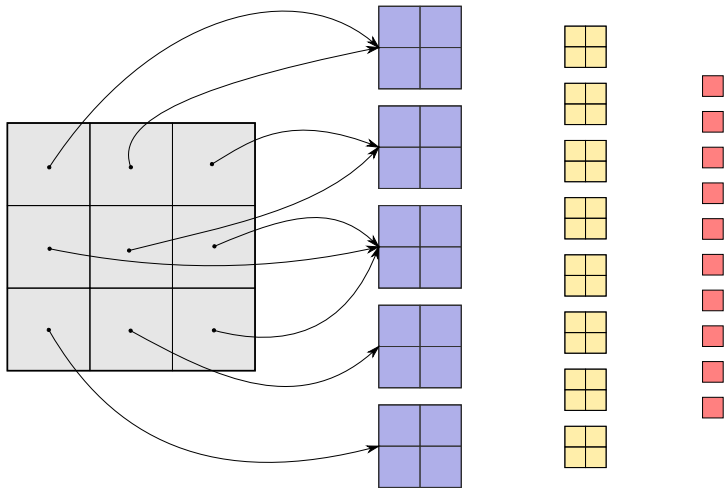
But: $\mathcal{D}_{\text{last}}$ has at most $\mathcal{O}(\sqrt{n})$ unique zones.



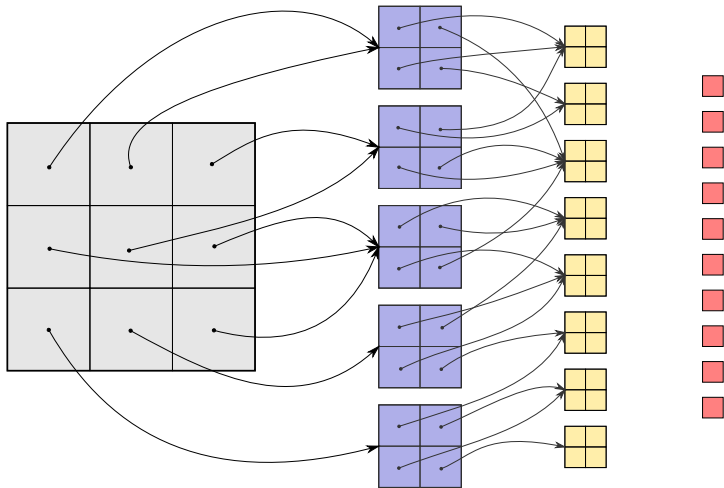
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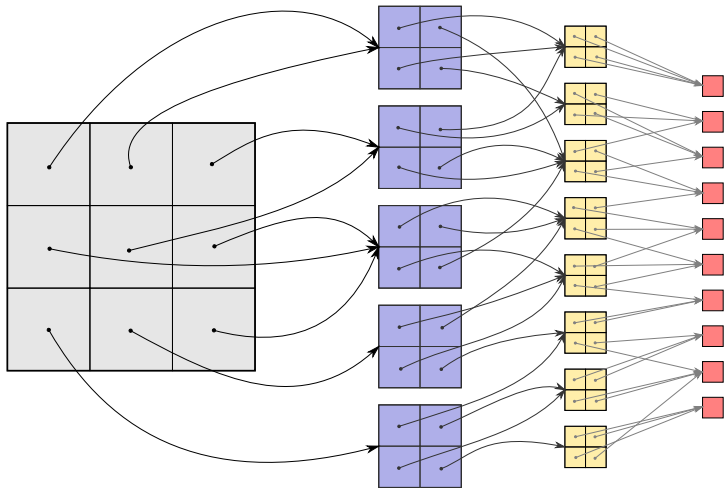
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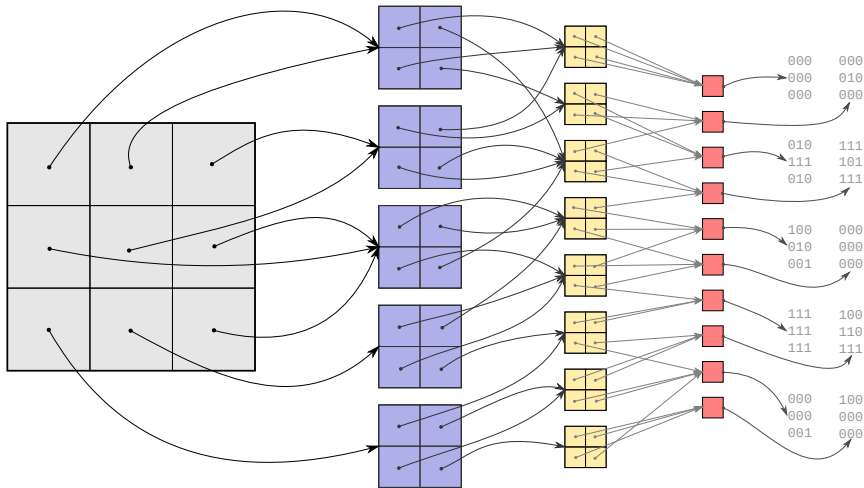
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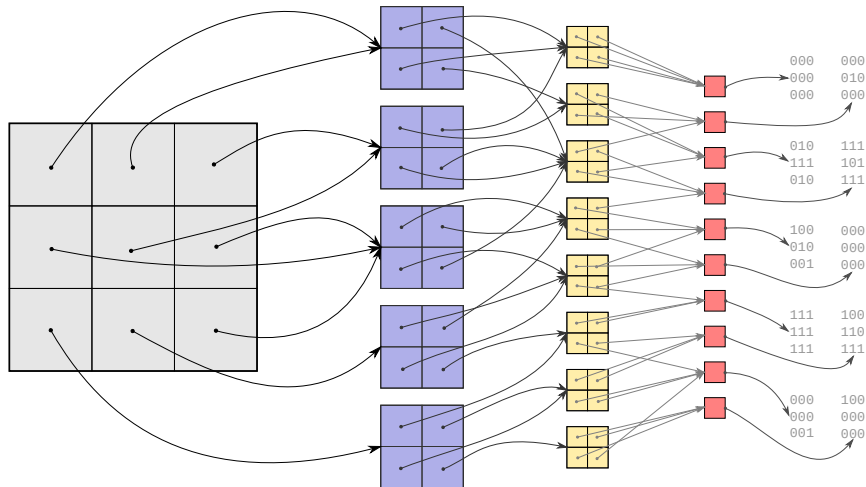


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 Now, create an object for each unique zone...
 Add pointers (each of size $\mathcal{O}(\log n)$ bits)...
 And store each unique zone of $\mathcal{D}_{\text{last}}$ explicitly.



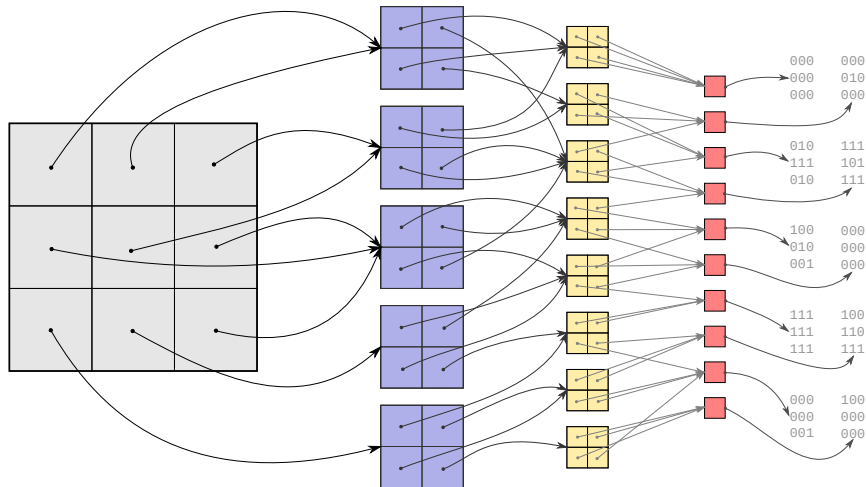
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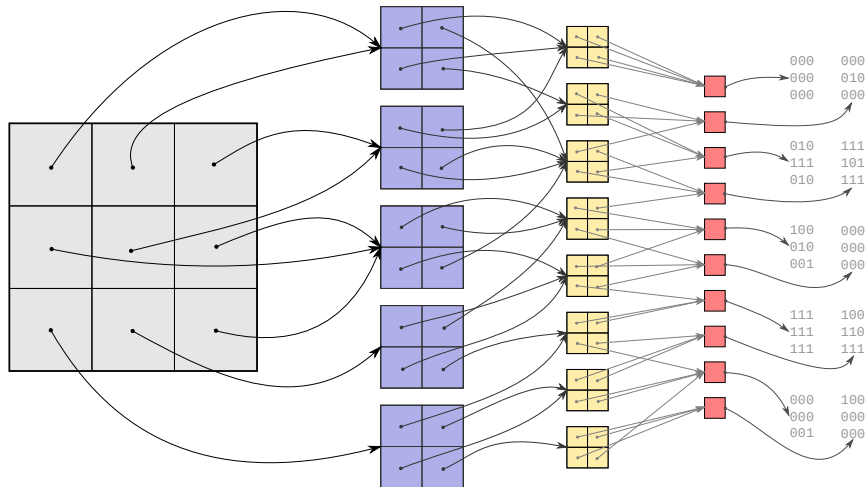
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Time complexity: $\mathcal{O}(\text{last}) = \mathcal{O}(\log \log n)$.



Memory complexity

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- $\mathcal{D}_{\text{last}}$ has only $\mathcal{O}(\sqrt{n})$ **small** zones and we can store them explicitly;
- The layers with **large** zones occupy $\mathcal{O}_d(n)$ bits in total;
- The data structure needs to be modified slightly for **medium** zones.

PART 3

χ -BOUNDEDNESS

Pilipczuk, Sokołowski,

Graphs of Bounded Twin-Width are Quasi-Polynomially χ -Bounded

χ -boundedness

Let \mathcal{C} – a hereditary class of graphs.

Definition

\mathcal{C} is χ -bounded by a function $f : \mathbb{N} \rightarrow \mathbb{N}$ if for every graph $G \in \mathcal{C}$,

$$\chi(G) \leq f(\omega(G)).$$

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Briański, Davies, Walczak (2022): **no**.

Twin-width and χ -boundedness

Theorem (Twin-width III)

Graphs of twin-width $\leq d$ are χ -bounded by $f_d(\omega) = (d + 2)^{\omega-1}$.

Twin-width and χ -boundedness

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Note: the proof shows that

$$f_d(\omega) \leq (d + 2) \cdot f_d(\omega - 1).$$

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Fix $d \in \mathbb{N}$. There exists a constant $\beta_d > 0$ such that graphs of twin-width $\leq d$ are χ -bounded by a **quasi-polynomial** function $f_d : \mathbb{N} \rightarrow \mathbb{N}$:

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Theorem (Bourneuf, Thomassé 2023)

Graphs of twin-width $\leq d$ are polynomially χ -bounded.

Ingredient: d -almost mixed minors

0	0	0	0	0	0	1	1	0	0	0	
1	1	0	0	1	1	1	0	0	1	1	0
<hr/>											
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	1	0	0	0	1	0	1	0	0	1	0
<hr/>											
1	0	0	0	1	1	0	0	1	1	1	1
0	1	0	1	1	1	1	1	1	0	0	1
<hr/>											
1	1	1	1	1	1	1	1	1	1	0	1
0	1	0	1	1	0	1	1	1	1	0	1

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0	0	0	0	0	0	1	1	0	0	0	
1	1	0	0	1	1	1	0	0	1	1	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	1	0	0	0	1	0	1	0	0	1	0
1	0	0	0	1	1	0	0	1	1	1	1
0	1	0	1	1	1	1	1	1	0	0	1
1	1	1	1	1	1	1	1	1	1	0	1
0	1	0	1	1	0	1	1	1	1	0	1

Fact

M has a $2d$ -almost mixed minor $\implies M$ has a d -mixed minor.

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0	0	0	0	0	0	1	1	0	0	0	
1	1	0	0	1	1	1	0	0	1	1	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	1	0	0	0	1	0	1	0	0	1	0
1	0	0	0	1	1	0	0	1	1	1	1
0	1	0	1	1	1	1	1	1	0	0	1
1	1	1	1	1	1	1	1	1	1	0	1
0	1	0	1	1	0	1	1	1	1	0	1

Fact

M has a $2d$ -almost mixed minor $\implies M$ has a d -mixed minor.

Corollary

$\text{tww}(G) \leq d \implies G$ has a $(2d + 2)$ -mixed free adjacency matrix
 $\implies G$ has a $(4d + 4)$ -almost mixed free adjacency matrix.

Idea

If we had

$$f_d(\omega) \leq \text{const}(d) \cdot f_d(0.9\omega),$$

then we would get $f_d(\omega) = \text{poly}(\omega)$!

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Induction on $\omega(G) \geq 1$.

G – a graph,

M – a d -almost mixed free adjacency matrix of G ,

$V(G) = \{1, \dots, n\}$ ordered according to M .

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M – a d -almost mixed free adjacency matrix of G ,

$V(G) = \{1, \dots, n\}$ ordered according to M .

Partition $V(G)$ into intervals $A_1 \cup A_2 \cup \dots \cup A_k$ (**blobs**) so that

$$\omega(G[A_i]) = 0.9\omega \quad \text{for } i = 1, 2, \dots, k.$$

$$V(G) = A_1 \cup \dots \cup A_k, \quad \omega(G[A_i]) = 0.9\omega \quad \text{for } i = 1, 2, \dots, k.$$

$\mathcal{D} :=$ a (symmetric) division of M from the partition A_1, \dots, A_k .

M	0	≡≡≡	≡	≡	M
0	0		0	0	0
	≡≡≡	M	0	0	
	0	0	M	M	
	0	0	M	0	
M	0	≡≡≡	≡	≡	M

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	≡≡≡	M	0	0	
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M	0	≡≡≡	≡	≡	M

First, paint each **blob**

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0	0		0	0	0
	≡≡≡	M	0	0	
	0	0	M	M	
	0	0	M	0	
M	0	≡≡≡	≡	≡	M

First, paint each **blob** using $f_d(0.9\omega)$ colors.

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0	0	≡≡≡	≡	≡	M
0	0		0	0	0
	≡≡≡	0	0	0	
	0	0	0	M	
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M	0	≡≡≡	≡	≡	0

First, paint each **blob** using $f_d(0.9\omega)$ colors.

For each color class C : each intersection $C \cap A_i$ is an **independent set!**

Blob-blob connections

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$\mathcal{D} :=$ a (symmetric) division of M from the partition A_1, \dots, A_k .

Given a set (color class) C s.t. each $A_i \cap C$ is an **independent set**.

0	0	≡≡≡	≡	≡≡	M
0	0		0	0	0
	≡≡	0	0	0	
	0	0	0	M	
	0	0	M	0	
M	0	≡≡≡	≡	≡≡	0

Blob-blob connections

$$V(G) = A_1 \cup \dots \cup A_k, \quad \omega(G[A_i]) = 0.9\omega \quad \text{for } i = 1, 2, \dots, k.$$

$\mathcal{D} :=$ a (symmetric) division of M from the partition A_1, \dots, A_k .

Given a set (color class) C s.t. each $A_i \cap C$ is an **independent set**.

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	≡≡	0	0	0	
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$$\chi(G[C]) \leq \underbrace{\text{const}(d)}_{\text{remove } \boxed{M} \text{ with Marcus-Tardos}} \cdot \underbrace{f_d(0.2\omega)}_{\text{afterwards, clique number } \leq 0.2\omega}$$

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Rich blobs

Fix a blob B_i ($i < k$), and let $B_i = \{\ell_i, \ell_i + 1, \dots, r_i\}$. We call B_i **rich** if:

- every vertex $v \in B_i$ is adjacent to any $s \in \{r_i + 1, r_i + 2, \dots, n\}$,
- no two consecutive vertices of B_i are twins with respect to $\{r_i + 1, r_i + 2, \dots, n\}$.

Solution: attempt 2 (rich blobs)

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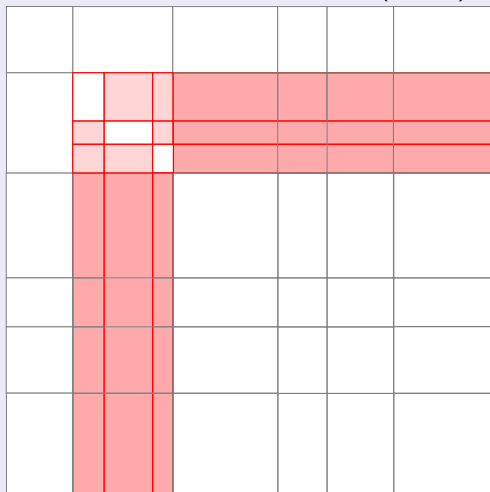
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By induction: quasi-polynomial!

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- Quasi-polynomial again!

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$$\begin{aligned} f_d(\omega) &\leq \text{const}(d) \cdot \{f_d(0.9\omega) + \max \chi(B_i) \cdot f_d(0.2\omega)\} \\ &\leq \text{const}(d) \cdot \left\{ f_d(0.9\omega) + f_d(0.1\omega) \cdot f_d(0.2\omega) \cdot f_{d-1}(\omega^d)^2 \right\} \end{aligned}$$

Subexponential ($2^{O(\omega^\varepsilon)}$ for any $\varepsilon > 0$ if parameters chosen carefully).

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Inspired by a work of Chudnovsky, Penev, Scott, Trotignon (*Substitution and χ -boundedness*, JCTB, 2013).

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We eventually get:

$$f_d(\omega) \leq \text{const}(d) \cdot \left\{ f_d(0.9\omega) + f_{d-1}(\omega^d)^2 \cdot \sum_{u=0}^{\lfloor \log_2(0.1\omega) \rfloor} f_d(2^{u+1}) \cdot f_d\left(\frac{0.2\omega}{2^u} + 1\right) \right\}$$

This resolves to $f_d(\omega) = 2^{\beta_d \cdot \log^d \omega}$.

Thank you!