# Constructive algorithm for path-width of matroids (and more)

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# Historical backgrounds

## Determining tree-width of graphs

- \* 1987: O(n<sup>k</sup>) alg. (Arnborg, Corneil, and Proskurowski)
- \* 1995: O(n<sup>2</sup>) alg. (Robertson and Seymour) (GM XIII)
  - \* Step 1: find an "approximate" tree-decomp. of width≤f(k)
  - \* Step 2: test forbidden minors for tree-width≤k
  - 1994 "Self-reduction" technique (Fellows and Langston) → one can construct such an algorithm without knowing the complete list of forbidden minors
- 1996 (Bodlaender and Kloks)
  - Dynamic Programming algorithm to do Step 2 without using forbidden minors

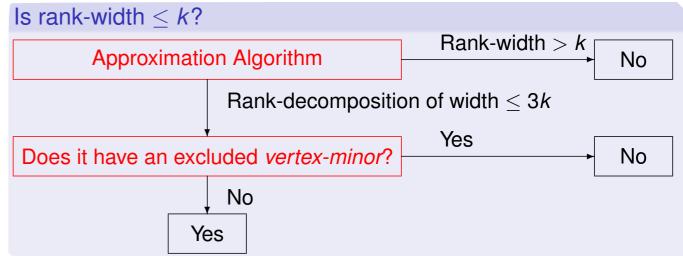
Similar method for path-width of graphs (dynamic programming)

This finds a path-decomposition or a tree-decomposition as well.

### Determining rank-width of

graphs

- 2006 (O., Seymour): O(n log n) algorithm to find an "approximate" rank-decomp. of width≤f(k)
   ...Step 1 (Improved to O(n) by O. 2008)
- 2005 (O.): #forbidden vertex-minors is finite for each k
- 2007 (Courcelle, O.): testing forbidden vertex-minors for graphs of bounded rank-width ... Step 2
- 2008 (Hlineny, O.): constructing a rankdecomp of width≤k by using algorithms based on forbidden minors (using matroids)



#### Q1: **Bodlaender-Kloks type algorithm** to find a rank-decomposition of width≤k?

Do we have a ...

Q2: **Constructive** algorithm to decide linear rank-width≤k for fixed k?



#### Solvable problems when rank-width is bounded (I)

#### Courcelle, Makowsky, and Rotics '00

Every graph problem expressible in monadic second-order logic formula (with no edge-set variables) is solvable in time  $O(n^3)$  for graphs having rank-width at most k for fixed k.

#### Solvable prob

CMR'00: Minimize w(X) satisfying  $\varphi(X)$  for graphs of bounded rank-width.

CMR'01: Counting the number of true assignments in polynomial time. (assuming unit time for arithmetic operations on  $\mathbb{R}$ .)

Many other proble solved in polynon

Can I find a partition of vertices into three subsets such that each set has no edges inside? (graph 3-coloring problem)

- Finding a chr
- Deciding whe
- Given a mon there is a par satisfied for a....

```
\exists X_1 \exists X_2 \exists X_3 \forall v \forall w (v, w \in X_1 \Rightarrow \neg \operatorname{adj}(v, w))
\land \forall v \forall w (v, w \in X_2 \Rightarrow \neg \operatorname{adj}(v, w))
\land \forall v \forall w (v, w \in X_3 \Rightarrow \neg \operatorname{adj}(v, w)) \cdots
```

All of these algorithms

- need the rank-decomposition of width  $\leq k$  as an input, and
- use the dynamic programming.

# Arranging vectors — Alternative view of path-width of matroids

### Arranging vectors: A problem in coding theory

- **Input**: n vectors in  $\mathbb{F}^m$ .
- Goal: Find the minimum k such that there is a linear layout
   v<sub>1</sub>, v<sub>2</sub>, ..., v<sub>n</sub> of the vectors such that

```
\begin{array}{l} \text{dim } (<\!v_1\!> \cap <\!v_2,\!v_3,...,\!v_n\!>) \leq k,\\\\ \text{dim } (<\!v_1,\!v_2\!> \cap <\!v_3,...,\!v_n\!>) \leq k,\\\\ \dots\\\\ \text{dim } (<\!v_1,\!v_2,\!v_3,...,\!v_{n-1}\!> \cap <\!v_n\!>) \leq k. \end{array}
```

 Minimum such k = "Trellis State-Complexity of a linear code" or "Trellis-Width" (coding theory)

or "Path-width" (matroid theory) — matroid representable over F

### Computing trellis-width (or path-width)

- Deciding trellis-width≤k is NP-complete (Kashyap07, 08)
- What if k is fixed? Remark in Kashyap's paper (07)
- 4 Concluding Remarks

The main contribution of this paper was to show that the decision problem TREL-LIS STATE-COMPLEXITY is NP-complete, thus settling a long-standing conjecture. Now, the situation is rather different if we consider a variation of the problem in which the integer w is not taken to be a part of the input to the problem. In other words, consider the following problem:

**Problem:** Weak Trellis State-Complexity

Let  $\mathbb{F}_q$  be a fixed finite field, and let w be a fixed positive integer.

**Instance:** An  $m \times n$  generator matrix for a linear code  $\mathcal{C}$  over  $\mathbb{F}_q$ .

**Question:** Is there a coordinate permutation of  $\mathcal{C}$  that yields a code  $\mathcal{C}'$  whose

YES

minimal trellis has state-complexity at most w?

There is good reason to believe that this problem is solvable in polynomial time.

#### For fixed k?

- ♦ WQO(Well-quasi-ordering) Conjecture: F-representable matroids are well-quasi-ordered
  by the minor relation. (no infinite antichain w.r.t. minors)
- \* If true, then for every class X of  $\mathbb{F}$ -representable matroids closed under taking minors, there are **finitely** many  $\mathbb{F}$ -representable matroids  $M_1, M_2, ..., M_n$  such that M is in X iff none of  $M_1, M_2, ..., M_n$  is a minor of M.
- ♦ Theorem (Geelen, Gerards, Whittle; 2012+): WQO Conj is true for finite F.
- Corollary: For each k and F, there exists a finite list of matroids such that an F-repre.
   matroid M has path-width≤k iff none in the list is a minor of M.
- Enough to check whether an input matroid has some minors in the finite list (which can be done in poly time for matroids of bounded branch-width, shown by Hlineny.)
- \* **Trouble**: 1. No algorithm known to construct the list of forbidden minors.
  - 2. Even if you know the list, this doesn't provide a linear ordering!



### Known algorithms for path-width/branch-width of matroids

• **STEP 1**: Find an <u>approximate</u> branch-decomposition (Hlineny 2006; O(n ) algorithm)

STEP 2: Use the dynamic programming to test all forbidden minors for branch-

width≤k or path-width≤k.

Branch-width: (the size of each forbidden minor) ≤ (6<sup>k</sup>-1)/5
 (Geelen, Gerards, Robertson, Whittle 2003)

- Path-width: (#forbidden minors) OPEN!
   No upper bound is known; Finite due to WQO.
- STEP 3: Use step 2 to construct a branch-decomp of width≤k (Hlineny, Oum)
- For path-width, an efficient algorithm exists but we did not know how to construct.



#### What's new? (1/2)

- Theorem [Jeong, Kim, O.]
   O( f(k) n³)-time algorithm to find
   a linear layout v<sub>1</sub>, v<sub>2</sub>, ..., v<sub>n</sub>
   of the input n vectors in F<sup>m</sup> such that
   dim (<v<sub>1</sub>,v<sub>2</sub>,...,v<sub>i</sub>> ∩ <v<sub>i+1</sub>,...,v<sub>n</sub>>)≤k for all i,
   if it exists (when F is a finite field)
- Outcome: Fixed Parameter Tractable to decide trellis-width≤k or path-width≤k of F-representable matroids



#### What's new (2/2) Extension to subspaces

For example, if

V<sub>i</sub>=<v<sub>i</sub>> for all ii,

if it exists (when F is a finite field)

### Corollary to Linear rank-width

- Cut-rank function cutrk<sub>G</sub>(X):=rank of X\*(V-X) submatrix of the adjacency matrix of a graph G
- Linear rank-width of a graph G:= min k such that  $\exists$  linear layout  $v_1, v_2, ..., v_n$  of the vertices with  $\mathsf{cutrk}_G(\{v_1, v_2, ..., v_i\}) \leq k$  for all i.
- NP-complete to decide linear rank-width≤k.
- THEOREM: For a fixed k,
   O(n³)-time algorithm to decide linear rank-width≤k. (NEW)

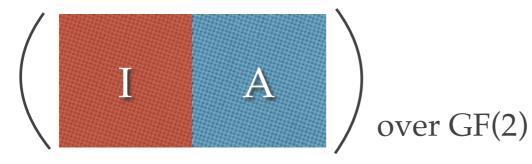
### Corollary to Linear rank-width

If A

is the adjacency matrix of G and

j

is the identity matrix, then



For a vertex v<sub>i</sub>,

let V<sub>i</sub>=span of the i-th and (i+n)-th column vectors

EASY FACT: 2 \* cutrk<sub>G</sub>(X)=dim (( $\Sigma$ {V<sub>i</sub>:v<sub>i</sub> $\in$ X})  $\cap$  ( $\Sigma$ {V<sub>i</sub>:v<sub>i</sub> $\notin$ X}))

Path-width of {V1,V2,...,Vn} = 2\*(linear rank-width of G)

### Corollary to Linear clique-width

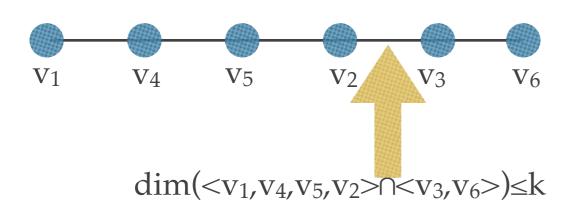
- Linear clique-width= "linearized version of clique-width"
- EASY FACT: If linear rank-width=k, then linear clique-width≤2<sup>k</sup>+1.
- NP-complete to decide linear clique-width≤k (when k is not fixed) (Fellows, Rosamond, Rotics, Szeider 2009)
- Corollary: The first approximation algorithm for linear cliquewidth.
  - For a fixed k,  $O(n^3)$ -time algorithm to find a linear clique-width expression of width  $\leq 2^k+1$  or confirms that linear clique-width>k.

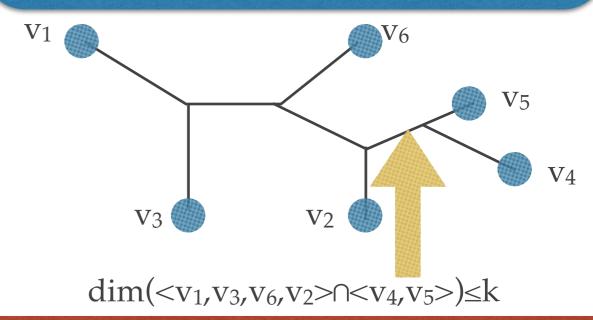


#### Bodlaender-Kloks type algorithm for path-width of "subspaces"

### Path-width vs Branch-width of (representable) matroids

Input:  $v_1, v_2, ..., v_n$ : vectors in  $\mathbb{F}^m$ . ( $\mathbb{F}$ : finite field) Output: Yes if  $\exists$  permutation  $\pi$  of  $\{1,2,\ldots,n\}$  s.t. ...



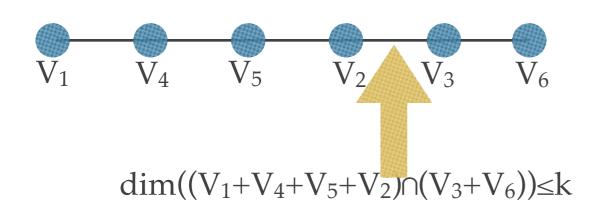


**Branch-decomposition** of width≤k

## Path-width vs Branch-width of a subspace arrangement

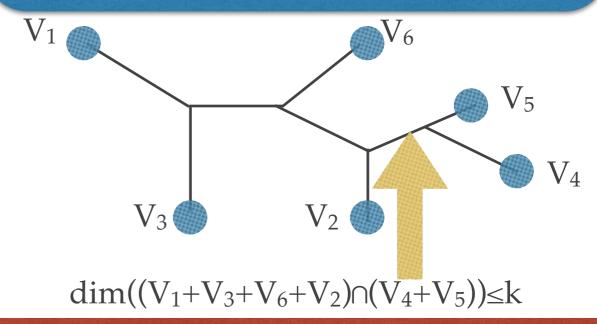
Input:  $V_1, V_2, ..., V_n$ : subspaces in  $\mathbb{F}^m$ . ( $\mathbb{F}$ : finite field)

Output: Yes if  $\exists$  permutation  $\pi$  of  $\{1,2,\ldots,n\}$  s.t. ...



Input:  $V_1, V_2, ..., V_n$ : subspaces in  $\mathbb{F}^m$ . (F: finite field)

Output: Yes if ∃ subcubic tree T with a bijection L:{leaves}→{subspaces} s.t. ...



**Branch-decomposition** of width≤k

#### Our algorithm

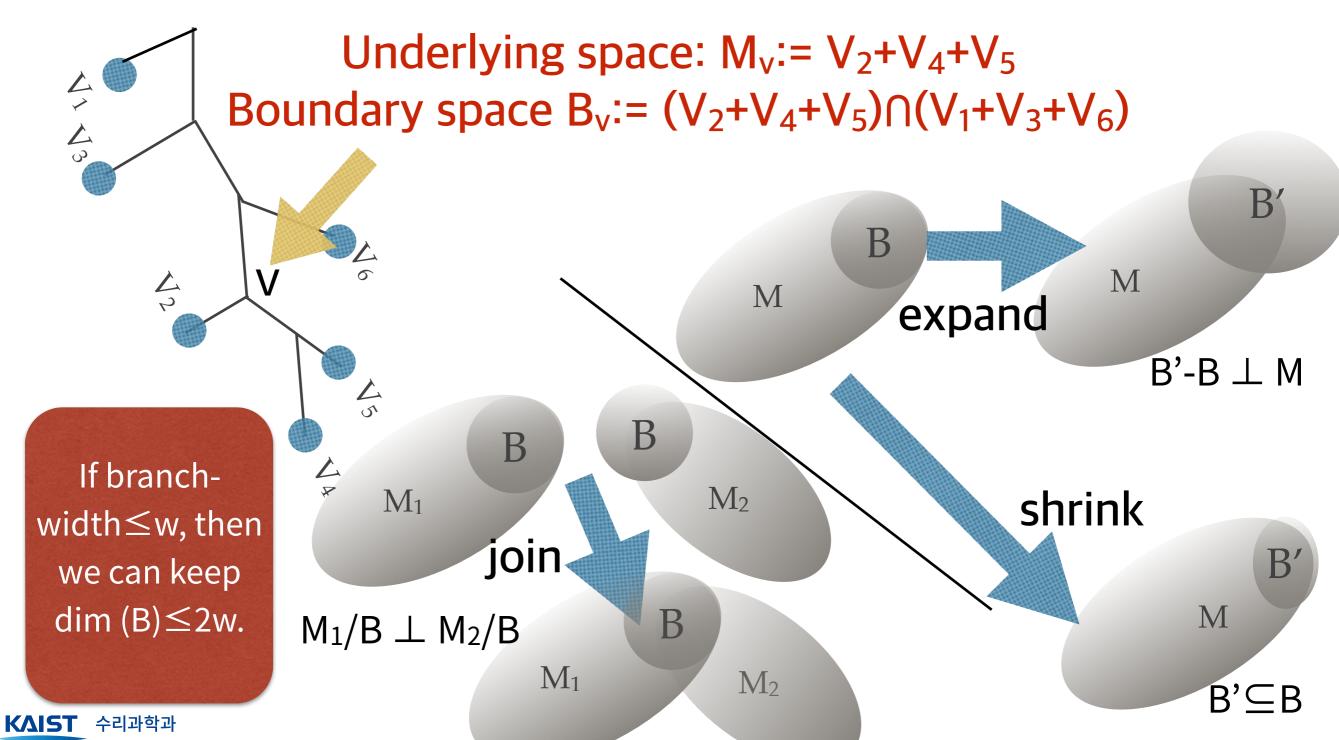
Input: n subspaces

We will discuss how to provide such a decomposition later

- Assume that we are given a branch-decomposition of width w.
- Task: Do the dynamic programming to enumerate all "partial solutions" of width at most k.

## How to run dynamic-programming on a "branch-decomposition of subspaces"

### Dynamic programming on a branch-decomposition



### Path-decomposition: an alternative definition of path-width

- Let  $\gamma = \{V_1, V_2, ..., V_n\}$  be a subspace arrangement (set of subspaces),  $\langle \gamma \rangle = V_1 + V_2 + ... + V_n$ .
- A path-decomposition of  $\gamma$ : := a sequence  $(S_1, S_2, ..., S_m)$  of subspaces of  $<\gamma$ >
  with an injective function  $\mu:\{1,2,\cdots,n\}\rightarrow\{1,2,...,m\}$  s.t.  $V_i\subseteq S_{\mu(i)}$ .
- Width of a path-decomposition := max  $(S_1+...+S_i) \cap (S_{i+1}+...+S_m)$ .
- S<sub>i</sub> is called a bag.
- THM:  $\exists$  linear layout of width $\leq$ k  $\Leftrightarrow$   $\exists$  path-dec. of width $\leq$ k.

#### What do we keep?

M

For a path-decomposition (S<sub>1</sub>,S<sub>2</sub>,...,S<sub>m</sub>)

For each "gap", there is a triple  $(L,R,\lambda)$ 

$$L_3 = (S_1 + S_2 + S_3) \cap B$$

S<sub>1</sub> S<sub>2</sub> S<sub>3</sub> S<sub>4</sub> S<sub>5</sub> S<sub>6</sub> S<sub>7</sub>

"Left subspace" "Right subspace" shown on B shown on B

$$L_3=(S_1+S_2+S_3)\cap B$$
  $R_3=(S_4+S_5+S_6+S_7)\cap B$ 

Extra connectivity not shown in B

 $\lambda_3 = \dim(S_1 + S_2 + S_3) \cap (S_4 + S_5 + S_6 + S_7) - \dim((S_1 + S_2 + S_3) \cap (S_4 + S_5 + S_6 + S_7) \cap B$ 

For (M,B), we only need to keep a sequence of (L,R, $\lambda$ ) in order to determine whether we have a path-decomposition

#### B-trajectory for a subspace B

- statistic:=triple (L,R,λ) of subspaces L, R of B and λ≥0.
- B-trajectory:=a finite sequence of statistics Γ=a<sub>1</sub>,a<sub>2</sub>,...,a<sub>n</sub> such that

$$L(a_1)\subseteq L(a_2)\subseteq ...\subseteq L(a_n),$$
  
 $R(a_1)\supseteq R(a_2)\supseteq ...\supseteq R(a_n).$ 

- Width of a B-trajectory:= max λ.
- The canonical B-trajectory of a path-decomposition (S<sub>1</sub>,S<sub>2</sub>,...,S<sub>n</sub>)

A path-decomposition  $(S_1, S_2, ..., S_{n-1})$ 

$$X_i := S_1 + S_2 + ... + S_i$$
  
 $Y_i := S_{i+1} + ... + S_{n-1}$ 



$$X_1 \cap B \subseteq X_2 \cap B \subseteq X_3 \cap B \subseteq ... \subseteq X_n \cap B$$

$$Y_1 \cap B \supseteq Y_2 \cap B \supseteq Y_3 \cap B \supseteq \dots \supseteq Y_n \cap B$$

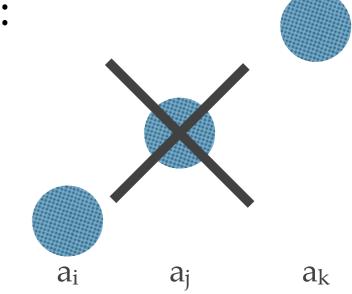
nonnegative integers

#### How to compress B-trajectories? Typical sequences of Bodlaender and Kloks

Reducing operation for a sequence of integers:
 In a sequence a<sub>1</sub>,a<sub>2</sub>,...,a<sub>n</sub> of integers,
 remove a<sub>j</sub> if a<sub>j</sub> is between a<sub>i</sub> and a<sub>k</sub> and i<j<k.</li>

• 
$$\tau(125392) = (192)$$





- A sequence is <u>typical</u> if no further reducing is possible.
- Lemma (Bodlaender and Kloks 1996):
   (#typical sequences consisting of {0,1,2,...,k})≤ (8/3)2<sup>2k</sup>.

### How to compress B-trajectories? "Compact" B-trajectories

- Reducing operation for a B-trajectory Γ=a<sub>1</sub>,a<sub>2</sub>,...,a<sub>n</sub> remove a<sub>j</sub> if
   L(a<sub>i</sub>)=L(a<sub>k</sub>), R(a<sub>i</sub>)=R(a<sub>k</sub>), and
   λ(a<sub>i</sub>) is between λ(a<sub>i</sub>) and λ(a<sub>k</sub>) and i<j<k.</li>
- A B-trajectory is <u>compact</u> if no further reducing is possible. The compactification  $\tau(\Gamma)$  is a compact B-trajectory obtained by reducing from  $\Gamma$ .

 $a_k$ 

• Lemma: if dim(B)=w and  $|\mathbb{F}|$ =q, then (#compact B-trajectories of width $\leq$ k)  $\leq$  ((8/3)  $2^{2k}$ )<sup>2w+1</sup>  $2^{2(2w+1)q^{2}w}$ 

#### What to store during dynamic programming? "Full Sets"

- For a subspace arrangement  $\gamma = \{V_1, V_2, ..., V_n\}$ , the **full set**  $FS_k(\gamma, B)$ :=set of all B-trajectories of width $\leq$ k that are better than some B-trajectories, realizable in  $\gamma$ .
- $FS_k(\gamma, \{0\}) \neq \emptyset$  if and only if path-width  $\leq k$ .

We aim to compute  $FS_k(\gamma, \{0\})$  by the dynamic programming

#### Computing the Full Set

Underlying space:  $M_v := V_2 + V_4 + V_5$ 

Boundary space  $B_v$ :=  $(V_2+V_4+V_5)\cap(V_1+V_3+V_6)$ 

At each leaf v, compute FS(M<sub>v</sub>, B<sub>v</sub>) "easy"

At each internal node v

with two children x and y,

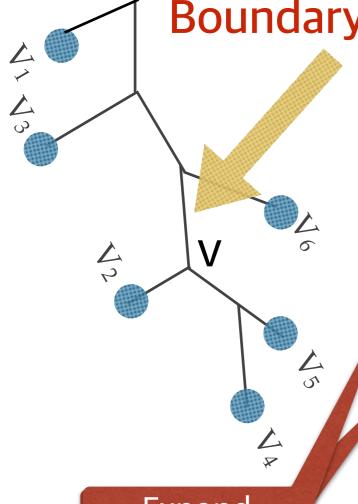
• compute  $FS(M_x, B_x+B_y)$  from  $FS(M_x, B_x)$ 

• compute  $FS(M_y, B_x+B_y)$  from  $FS(M_y, B_y)$ 

compute FS(M<sub>v</sub>, B<sub>x</sub>+B<sub>y</sub>) from

 $FS(M_x, B_x+B_y)$  and  $FS(M_y, B_x+B_y)$ 

compute  $FS(M_v,B_v)$  from  $FS(M_v,B_x+B_y)$ 



Expand

Join

Shrink

O(n)-time to compute  $FS_k(\gamma, \{0\})$ .

### How to provide an initial "approximate" branch-decomposition

 Method 1: Iterative compression; O(n) overhead Modify the output for V<sub>1</sub>,...,V<sub>i-1</sub> of width≤k to be the branch-decomposition of width≤k+1.
 TOTAL: O(n<sup>4</sup>) time (simpler but slower)

 Method 2: Use the algorithm by Hlineny and Oum (2008) that provides a branch-decomposition of width≤k in O(n³)-time (faster)

TOTAL: O(n<sup>3</sup>) time.



#### Concluding remarks

- By backtracking, we can construct a linear layout of width≤k.
- Similar idea works for rank-width of graphs and branch-width of matroids representable over a fixed finite field.
- Two bottlenecks for a faster algorithm:
  - Finding an approximate branch-decomposition.
  - Preprocessing the approximate branch-decomposition to make it more useful for dynamic programming.
     (e.g. Precompute a basis for each boundary space B)