### Algorithms based on contraction sequences

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based on joint works with Colin Geniet, Eun Jung Kim, Amadeus Reinald, Stéphan Thomassé, and Rémi Watrigant

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Fixed-parameter tractable  $(f(k)n^{O(1)})$  algorithms for

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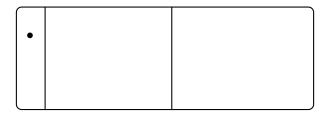
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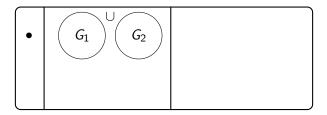
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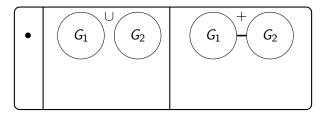
All these results are in fact part of the same framework



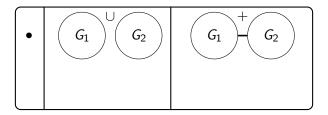
A single vertex is a cograph,



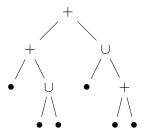
as well as the union of two cographs,

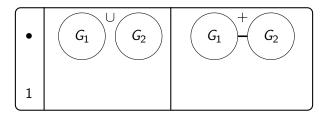


and the complete join of two cographs.

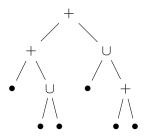


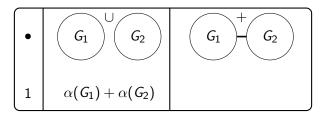
Many NP-hard problems are polytime solvable on cographs



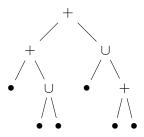


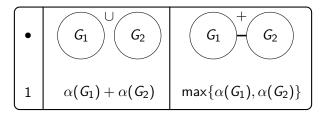
For instance the independence number  $\alpha(G)$  is polytime



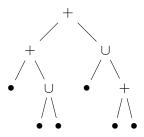


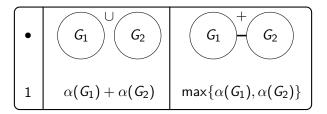
In case of a disjoint union: combine the solutions

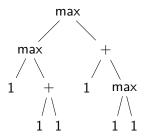




In case of a complete join: pick the larger one







Every induced subgraph has two twins

#### Every induced subgraph has two twins



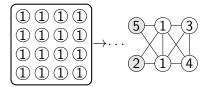
Is there another algorithmic scheme based on this definition?

#### Every induced subgraph has two twins



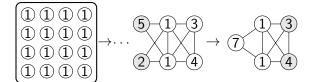
We store in each vertex its inner max independent set

#### Every induced subgraph has two twins



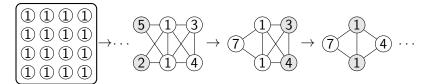
We can find a pair of false/true twins

#### Every induced subgraph has two twins



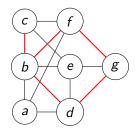
Sum them if they are false twins

#### Every induced subgraph has two twins



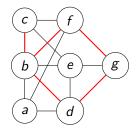
Max them if they are true twins

# Trigraphs



Three outcomes between a pair of vertices: edge, or non-edge, or red edge (error edge)

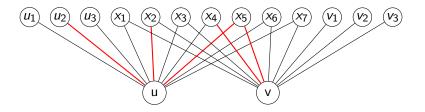
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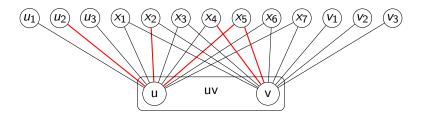
Red graph: trigraph minus its black edges

## Contractions in trigraphs



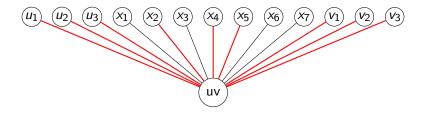
Identification of two non-necessarily adjacent vertices

## Contractions in trigraphs

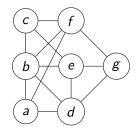


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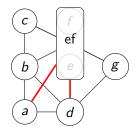
## Contractions in trigraphs



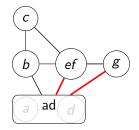
edges to  $N(u) \triangle N(v)$  turn red, for  $N(u) \cap N(v)$  red is absorbing



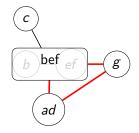
A contraction sequence of G: Sequence of trigraphs  $G = G_n, G_{n-1}, \ldots, G_2, G_1$  such that  $G_i$  is obtained by performing one contraction in  $G_{i+1}$ .



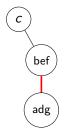
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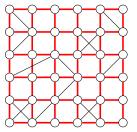


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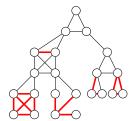


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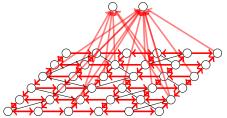
# Different conditions imposed in the sequence of red graphs



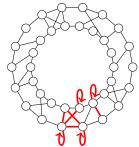
bd degree: defines bd twin-width



bd component: redefines bd cliquewidth

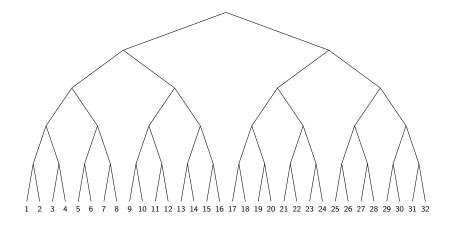


bd outdegree: defines bd oriented twin-width



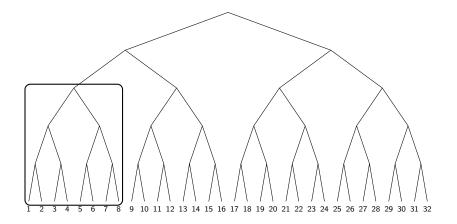
bd #edges: redefines bd linear cliquewidth

## Bd boolean-width $\Rightarrow$ bd component twin-width



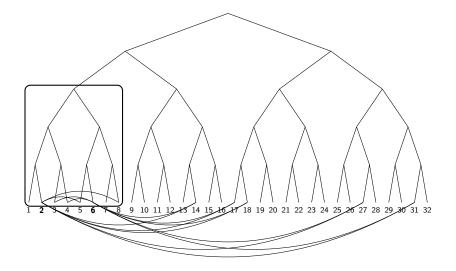
Bd boolean-width: binary tree layout s.t. every edge cut in the tree induces a bipartition with bd # distinct neighborhoods

### Bd boolean-width $\Rightarrow$ bd component twin-width



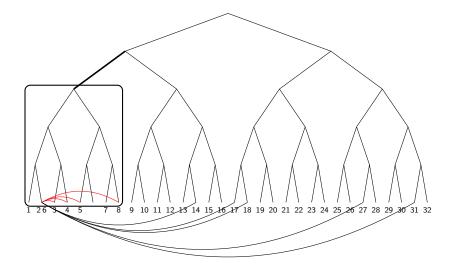
There is a subtree on  $\ell \in [d + 1, 2d]$  leaves, where d bounds the number of single-vertex neighborhoods in a bipartition

## Bd boolean-width $\Rightarrow$ bd component twin-width



Two vertices have the same neighborhood outside of this subtree

#### Bd boolean-width $\Rightarrow$ bd component twin-width



Contracting them preserves the upper bound at 2d on the size of red connected components

Component twin-width and boolean-width are tied

Theorem (B., Kim, Reinald, Thomassé '22)

A class has bounded component twin-width iff it has bounded boolean-width/cliquewidth/rank-width.

Proof.

We just saw one direction.

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When red components merge, their subtree gets a same parent.

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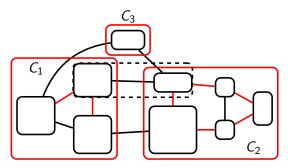
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Theorem (B., Kim, Reinald, Thomassé '22)

A class has bounded total twin-width iff it has bounded linear boolean-width/cliquewidth/rank-width.

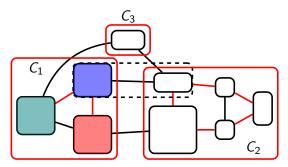
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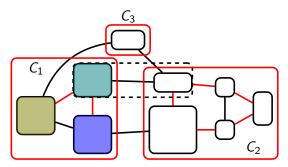
For every red component *C* keep every profile  $V(C) \rightarrow 2^{\{1,2,3\}} \setminus \{\emptyset\}$  realizable by a proper 3-coloring of G(C)

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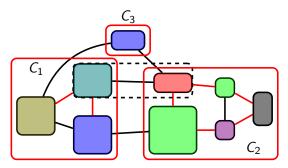
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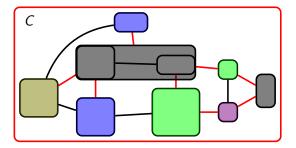
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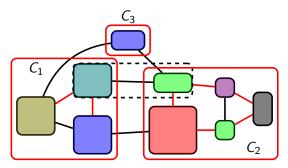
Some tuples of the at most d + 1 profiles corresponding to merging red components are compatible

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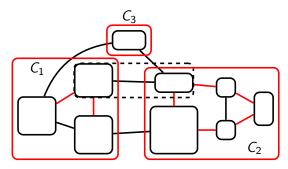
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Solve 3-COLORING on a graph G with a contraction sequence s.t. all red graphs have components of size at most d



Initialization: time 3nUpdate: time  $7^d d^2$ Total: time  $7^d d^2 n$ End: still a profile on the single vertex *containing* the whole graph?

GRAPH FO/MSO MODEL CHECKING **Parameter:**  $|\varphi|$  **Input:** A graph *G* and a first-order/monadic second-order sentence  $\varphi \in FO/MSO(\{E\})$ **Question:**  $G \models \varphi$ ?

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Example:

$$\varphi = \exists x_1 \exists x_2 \cdots \exists x_k \forall x \bigvee_{1 \leqslant i \leqslant k} x = x_i \lor \bigvee_{1 \leqslant i \leqslant k} E(x, x_i) \lor E(x_i, x)$$

 $G \models \varphi? \Leftrightarrow$ 

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 $G \models \varphi$ ?  $\Leftrightarrow$  k-Independent Set

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 $G \models \varphi$ ?  $\Leftrightarrow$  3-Coloring

# Courcelle's theorems

We will reprove with contraction sequences:

Theorem (Courcelle, Makowsky, Rotics '00)

MSO model checking can be solved in time  $f(|\varphi|, d) \cdot |V(G)|$  given a witness that the clique-width/component twin-width of the input G is at most d.

generalizes

#### Theorem (Courcelle '90)

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- as the incidence graph preserves bounded treewidth, possible edge-set quantification
- Inear FPT approximation for treewidth
- ▶ (polynomial) FPT approximation for clique-width

#### Rank-k m-types

Sets of non-equivalent formulas/sentences of quantifier rank at most k satisfied by a fixed structure:

$$\mathsf{tp}^\mathcal{L}_k(\mathscr{A}, ec{a} \in A^m) = \{ arphi(ec{x}) \in \mathcal{L}[k] : \mathscr{A} \models arphi(ec{a}) \},$$

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#### Theorem (folklore)

For  $\mathcal{L} \in \{FO, MSO\}$ , the number of rank-k m-types is bounded by a function of k and m only.

Proof.

" $\mathcal{L}[k+1]$  are Boolean combinations of  $\exists x \mathcal{L}[k]$ ."

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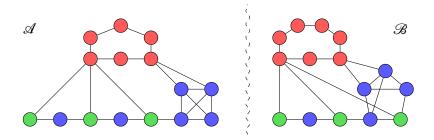
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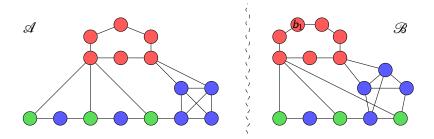
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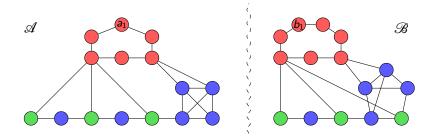
**Rank**-k types partition the graphs into g(k) classes. Efficient Model Checking = quickly finding the class of the input.



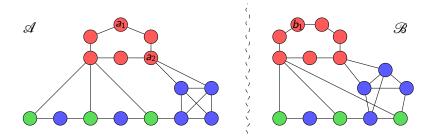
2-player game on two  $\sigma$ -structures  $\mathscr{A}, \mathscr{B}$  (for us, colored graphs)



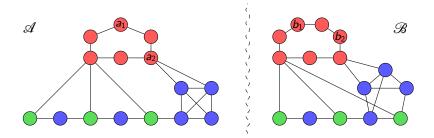
At each round, Spoiler picks a structure  $(\mathscr{B})$  and a vertex therein



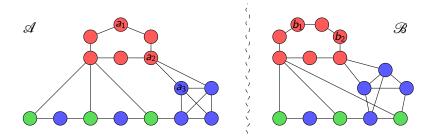
#### Duplicator answers with a vertex in the other structure



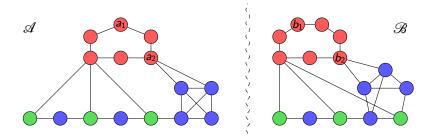
After q rounds, Duplicator wishes that  $a_i \mapsto b_i$  is an isomorphism between  $\mathscr{A}[a_1, \ldots, a_k]$  and  $\mathscr{B}[b_1, \ldots, b_k]$ 



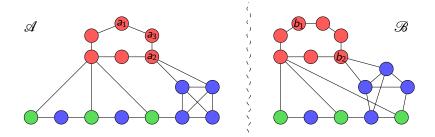
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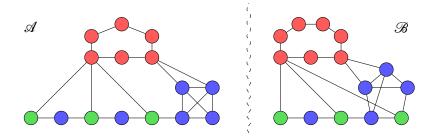
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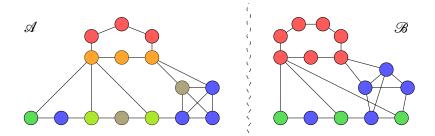
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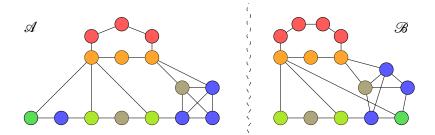
If Duplicator can survive k rounds, we write  $\mathscr{A} \equiv^{\mathsf{FO}}_k \mathscr{B}$ Here  $\mathscr{A} \equiv^{\mathsf{FO}}_2 \mathscr{B}$  and  $\mathscr{A} \not\equiv^{\mathsf{FO}}_3 \mathscr{B}$ 



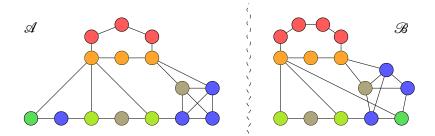
#### Same game but Spoiler can now play set moves



#### Same game but Spoiler can now play set moves



#### To which Duplicator answers a set in the other structure



Again we write  $\mathscr{A} \equiv_k^{\mathsf{MSO}} \mathscr{B}$  if Duplicator can survive k rounds

#### *k*-round EF games capture rank-*k* types

#### Theorem (Ehrenfeucht-Fraissé)

For every  $\sigma$ -structures  $\mathscr{A}, \mathscr{B}$  and logic  $\mathcal{L} \in \{FO, MSO\}$ ,

$$\mathscr{A} \equiv^{\mathcal{L}}_{k} \mathscr{B}$$
 if and only if  $tp^{\mathcal{L}}_{k}(\mathscr{A}) = tp^{\mathcal{L}}_{k}(\mathscr{B})$ .

k-round EF games capture rank-k types

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#### Proof.

Induction on k.

(⇒)  $\mathcal{L}[k+1]$  formulas are Boolean combinations of  $\exists x \varphi$  or  $\exists X \varphi$ where  $\varphi \in \mathcal{L}[k]$ . Use the answer of Duplicator to x = a or X = A. k-round EF games capture rank-k types

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( $\Leftarrow$ ) If  $\operatorname{tp}_{k+1}^{\mathcal{L}}(\mathcal{A}) = \operatorname{tp}_{k+1}^{\mathcal{L}}(\mathcal{B})$ , then the type  $\operatorname{tp}_{k}^{\mathcal{L}}(\mathcal{A}, a)$  is equal to some  $\operatorname{tp}_{k}^{\mathcal{L}}(\mathcal{B}, b)$ . Move *a* can be answered by playing *b*.

**Partitioned sentences:** sentences on  $(E, U_1, \ldots, U_d)$ -structures, interpreted as a graph vertex partitioned in *d* parts

Maintain for every red component C of every trigraph  $G_i$ 

 $\mathsf{tp}_k^{\mathsf{MSO}}(G,\mathcal{P}_i,C) = \{\varphi \in \mathsf{MSO}_{E,U_1,\dots,U_d}(k) : (G\langle C \rangle,\mathcal{P}_i \langle C \rangle) \models \varphi\}.$ 

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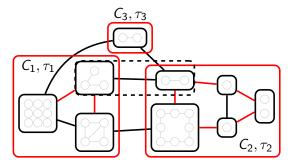
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For each  $v \in V(G)$ ,  $tp_k(G, \mathcal{P}_n, \{v\}) = type$  of  $K_1$  $tp_k(G, \mathcal{P}_1, \{V(G)\}) = type$  of G

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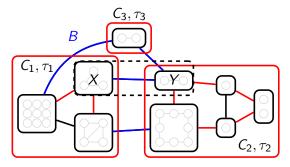


 $\tau = tp_k^{MSO}(G, \mathcal{P}_i, C)$  based on the  $\tau_j = tp_k^{MSO}(G, \mathcal{P}_{i+1}, C_j)$ ?

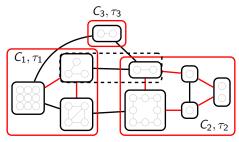
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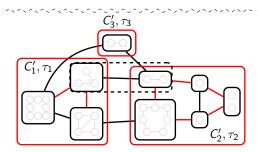
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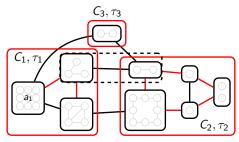


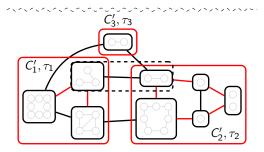
C arises from  $C_1, \ldots, C_{d'}$ :  $\tau = F(\tau_1, \ldots, \tau_{d'}, B, X, Y)$ 



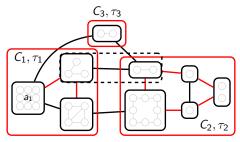


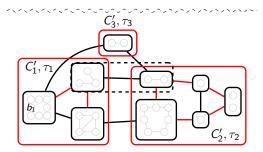
Duplicator combines her strategies in the red components



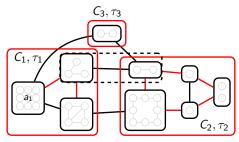


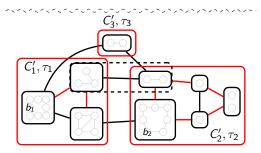
If Spoiler plays a vertex in the component of type  $\tau_1$ ,

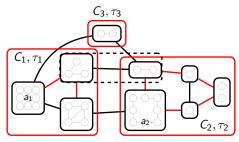


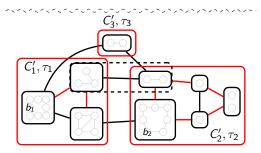


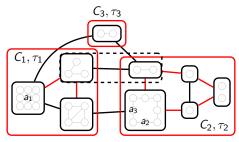
Duplicator answers the corresponding winning move

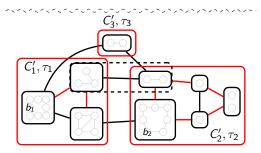


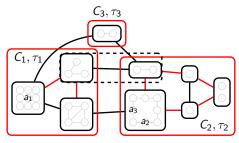


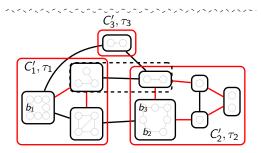


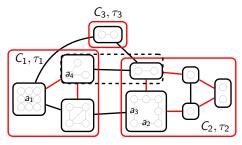


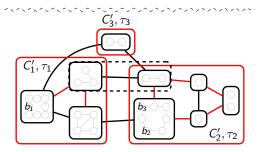


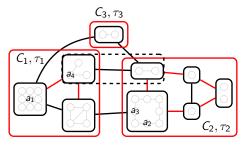


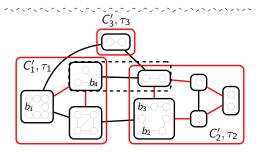


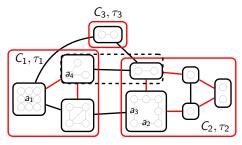


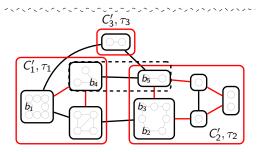


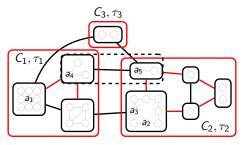


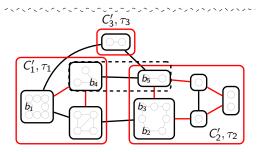


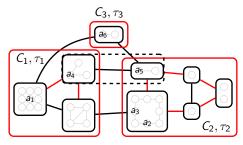


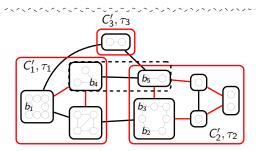


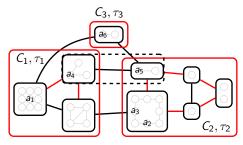


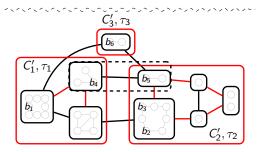


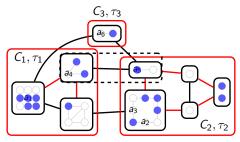


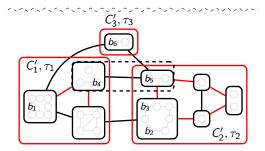




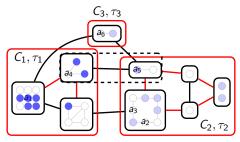


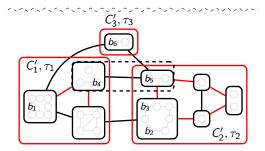




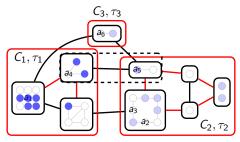


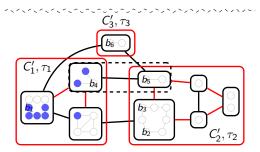
If Spoiler plays a set, Duplicator looks at the intersection with  $C_1$ ,



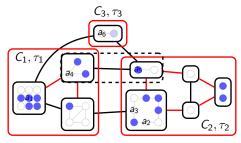


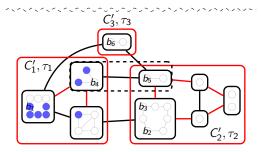
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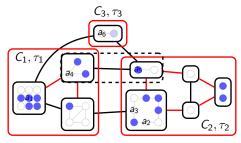


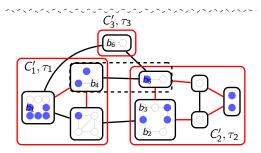
calls her winning strategy in  $C'_1$ 



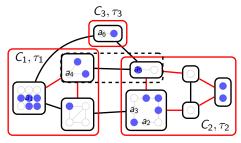


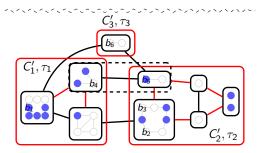
same for the other components



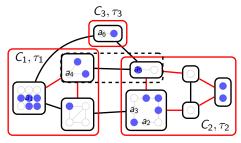


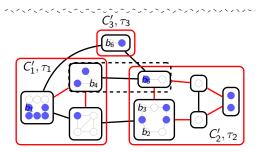
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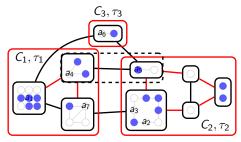


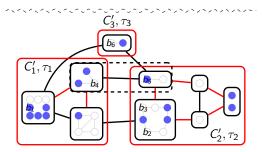
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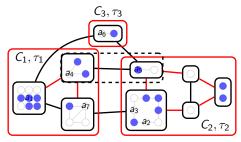


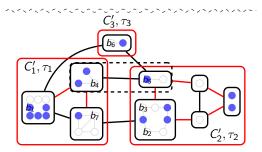


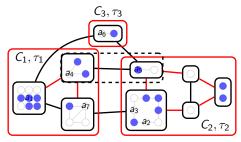
and plays the union

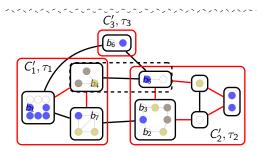


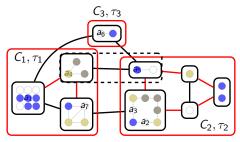


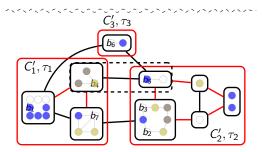












#### Turning it into a uniform algorithm

Reminder:

- #non-equivalent partitioned sentences of rank k: f(d, k)
- ▶ #rank-k partitioned types bounded by  $g(d, k) = 2^{f(d,k)}$

For each newly observed type  $\tau$ ,

- ▶ keep a representative  $(H, P)_{\tau}$  on at most  $(d+1)^{g(d,k)}$  vertices
- determine the 0, 1-vector of satisfied sentences on  $(H, \mathcal{P})_{\tau}$
- ▶ record the value of  $F(\tau_1, ..., \tau_{d'}, B, X, Y)$  for future uses

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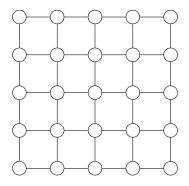
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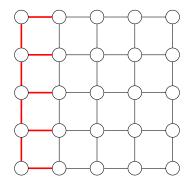
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To decide  $G \models \varphi$ , look at position  $\varphi$  in the 0, 1-vector of  $tp_k^{MSO}(G)$ 

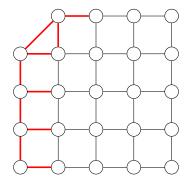
#### Twin-width is more general than the classic widths

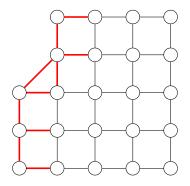


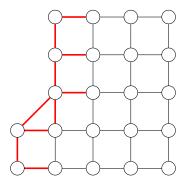
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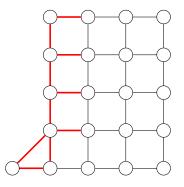


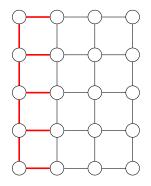
Twin-width is more general than the classic widths











4-sequence for planar grids

#### Theorem

The following classes have bounded twin-width, and O(1)-sequences can be computed in polynomial time.

- Bounded rank-width, and even, boolean-width graphs,
- every hereditary proper subclass of permutation graphs,
- posets of bounded antichain size (seen as digraphs),
- unit interval graphs,
- K<sub>t</sub>-minor free graphs,
- map graphs,
- subgraphs of d-dimensional grids,
- K<sub>t</sub>-free unit d-dimensional ball graphs,
- Ω(log n)-subdivisions of all the n-vertex graphs,
- cubic expanders defined by iterative random 2-lifts from K<sub>4</sub>,
- strong products of two bounded twin-width classes, one with bounded degree, etc.

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#### Can we solve problems faster, given an O(1)-sequence?

*k*-INDEPENDENT SET given a d = O(1)-sequence

*d*-sequence: 
$$G = G_n, G_{n-1}, ..., G_2, G_1 = K_1$$

Algorithm: For every connected subset D of size at most k of the red graph of every  $G_i$ , store in T[D, i] one largest independent set in  $G\langle D \rangle$  intersecting every vertex of D.

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Initialization:  $T[\{v\}, n] = \{v\}$ End:  $T[\{V(G)\}, 1] = IS$  of size at least k or largest IS in GRunning time:  $d^{2k}n^2$  red connected subgraphs, actually only  $d^{2k}n = 2^{O_d(k)}n$  updates *k*-INDEPENDENT SET given a d = O(1)-sequence

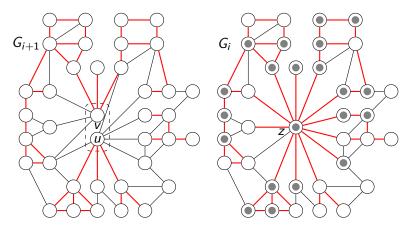
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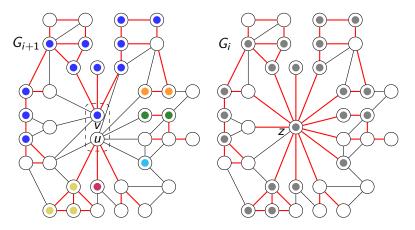
How to compute T[D, i] from all the T[D', i+1]?

# *k*-INDEPENDENT SET: Update of partial solutions



Best partial solution inhabiting •?

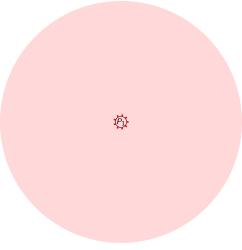
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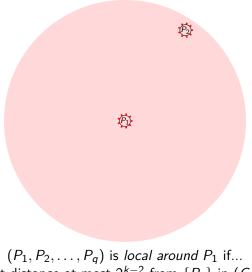
3 unions of  $\leqslant d + 2$  red connected subgraphs to consider in  $G_{i+1}$  with u, or v, or both

We will now generalize the previous algorithm to: Theorem (B., Kim, Thomassé, Watrigant '20) FO model checking can be solved in time  $f(|\varphi|, d) \cdot |V(G)|$  on graphs G given with a d-sequence. We will now generalize the previous algorithm to: Theorem (B., Kim, Thomassé, Watrigant '20) FO model checking can be solved in time  $f(|\varphi|, d) \cdot |V(G)|$  on graphs G given with a d-sequence.

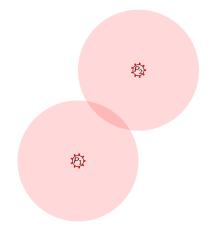
Add Gaifman's locality to our MSO model checking algorithm



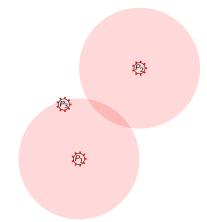
 $(P_1, P_2, \ldots, P_q)$  is local around  $P_1$  if...



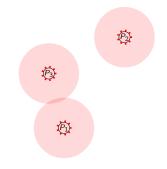
 $(P_1, P_2, \ldots, P_q)$  is local around  $P_1$  if...  $P_2$  is at distance at most  $2^{k-2}$  from  $\{P_1\}$  in  $(G, \mathcal{P}_i)$ 



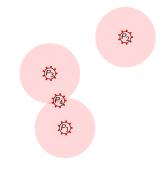
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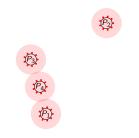
 $(P_1, P_2, \dots, P_q)$  is local around  $P_1$  if...  $P_3$  is at distance at most  $2^{k-3}$  from  $\{P_1, P_2\}$  in  $(G, \mathcal{P}_i)$ 



 $(P_1, P_2, \dots, P_q)$  is local around  $P_1$  if...  $P_3$  is at distance at most  $2^{k-3}$  from  $\{P_1, P_2\}$  in  $(G, \mathcal{P}_i)$ 



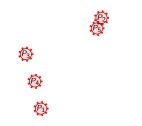
 $(P_1, P_2, \dots, P_q)$  is local around  $P_1$  if...  $P_4$  is at distance at most  $2^{k-4}$  from  $\{P_1, P_2, P_3\}$  in  $(G, \mathcal{P}_i)$ 



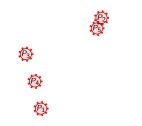
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 $(P_1, P_2, \dots, P_q)$  is local around  $P_1$  if...  $P_q$  is at distance at most  $2^{k-q}$  from  $\{P_1, \dots, P_{q-1}\}$  in  $(G, \mathcal{P}_i)$ 



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#### Partitioned local sentences and types

A prenex sentence is *partitioned local around* X in  $(G, \mathcal{P}_i)$  if of the form  $Qx_1 \in X \ Qx_2 \in P_2 \ \dots \ Qx_k \in P_k \ \psi(x_1, \dots, x_k)$  with

- $\blacktriangleright \psi$  is quantifier-free, and
- $(X, P_2, \ldots, P_k)$  local around X in  $(G, \mathcal{P}_i)$ .

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And the corresponding types:

$$\mathsf{ltp}_k^{\mathsf{FO}}(G,\mathcal{P}_i,X) = \{\varphi : \mathsf{qr}(\varphi) \leqslant k, \}$$

 $\varphi$  is partitioned local around X in  $(G, \mathcal{P}_i)$ ,  $(G, \mathcal{P}_i) \models \varphi$ . Partitioned local sentences/types in  $(G, \mathcal{P}_n)$  and  $(G, \mathcal{P}_1)$ 

#### Initialization of the dynamic programming

In 
$$(G, \mathcal{P}_n = \{\{v\} : v \in V(G)\})$$
: for every  $v \in V(G)$ ,  
 $Qx_1 \in \{v\} Qx_2 \in \{v\} \dots Qx_k \in \{v\} \psi \equiv \psi(v, v, \dots, v)$ 

Partitioned local types are easy to compute in  $(G, \mathcal{P}_n)$ 

Partitioned local sentences/types in  $(G, \mathcal{P}_n)$  and  $(G, \mathcal{P}_1)$ 

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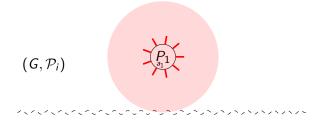
Partitioned local types are easy to compute in  $(G, \mathcal{P}_n)$ 

#### Output of the dynamic programming

In  $(G, \mathcal{P}_1 = \{V(G)\})$ :  $Qx_1 \in V(G) \ Qx_2 \in V(G) \ \dots \ Qx_k \in V(G) \ \psi \equiv \text{classic sentences}$ The partitioned local type in  $(G, \mathcal{P}_1)$  coincides with the type of G

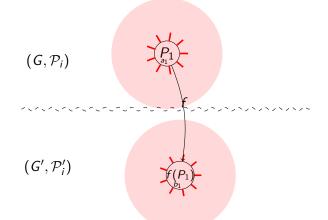


Local strategies win the global game

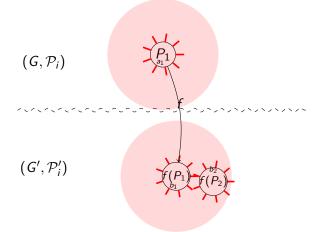


 $(G', \mathcal{P}'_i)$ 

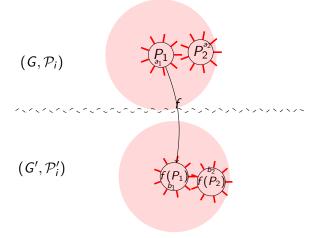
Say, Spoiler plays in  $P_1$ 



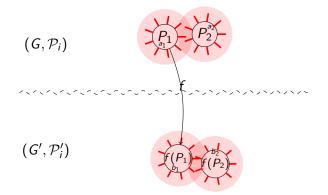
Duplicator answers in  $f(P_1)$  following the local game around  $P_1$ 



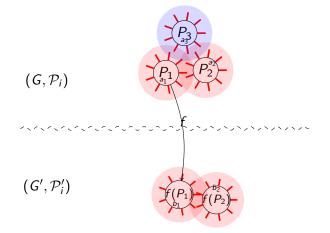
Now when Spoiler plays close to  $P_1$  or  $f(P_1)$ 



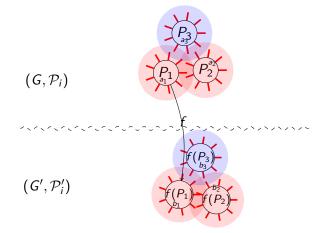
Duplicator follows the winning local strategy



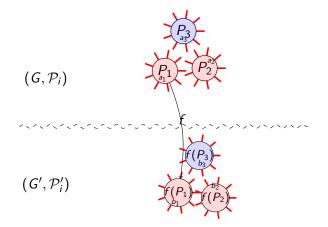
Duplicator follows the winning local strategy



If Spoiler plays too far



Duplicator starts a new local game around that new part



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 $(G, \mathcal{P}_{i+1}) \rightsquigarrow (G, \mathcal{P}_i) : X \text{ and } Y \text{ are merged in } Z$ 

Partitioned local types around P

• only needs an update if P is at distance at most  $2^{k-1}$  from Z

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Each contraction:  $O_{d,k}(1) = O(d^{2^k})$  updates in  $O_{d,k}(1) = f(d,k)$ Total time:  $O_{d,k}(n)$ 

# Conclusion

# Contraction sequences offer an interesting unifying and generalizing perspective

| Class of bounded  | MSO tr. of | FO tr. of      | seq. constraint | eff. MC |
|-------------------|------------|----------------|-----------------|---------|
| linear rank-width | paths      | linear order   | bd #edges       | MSO     |
| rank-width        | trees      | tree order     | bd component    | MSO     |
| twin-width        | not closed | perm. subclass | bd degree       | FO      |

Bounded degree, bounded expansion, nowhere denseness?

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#### Thank you for your attention!