# Algorithms based on contraction sequences 

Édouard Bonnet<br>based on joint works with Colin Geniet, Eun Jung Kim, Amadeus Reinald, Stéphan Thomassé, and Rémi Watrigant

ENS Lyon, LIP

November 15th, 2021, Journées Graphes et Algorithmes

## Unification and generalization via contraction sequences

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

- Hamiltonian Cycle parameterized by treewidth $k$ : Courcelle's theorem or progdyn on tree-decompositions


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- $k$-Permutation Pattern on all permutations: ad hoc algorithm of Guillemot and Marx
- $k$-Subgraph Isomorphism on bounded-width posets: Bova et al., Gajarský et al.
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All these results are in fact part of the same framework

## Cographs



A single vertex is a cograph,

## Cographs


as well as the union of two cographs,

## Cographs


and the complete join of two cographs.

## Cographs



Many NP-hard problems are polytime solvable on cographs


## Cographs



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


## Cographs



In case of a disjoint union: combine the solutions


## Cographs



In case of a complete join: pick the larger one


## Cographs



## Another cograph definition

Every induced subgraph has two twins

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Is there another algorithmic scheme based on this definition?

## Another cograph definition

Every induced subgraph has two twins
(1) (1) (1) (1)
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We store in each vertex its inner max independent set

## Another cograph definition

Every induced subgraph has two twins


We can find a pair of false/true twins

## Another cograph definition

Every induced subgraph has two twins


Sum them if they are false twins

## Another cograph definition

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

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Identification of two non-necessarily adjacent vertices

## Contractions in trigraphs


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

## Contraction sequence



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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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 $\ell \in[d+1,2 d]$ 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 $2 d$ 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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Theorem (B., Kim, Reinald, Thomassé '22)
A class has bounded total twin-width iff it has bounded linear boolean-width/cliquewidth/rank-width.

## Is it easier to design algorithms via this characterization?

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

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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 $d$


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

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

## Formulas, sentences, and model checking

Graph FO/MSO Model Checking Parameter: $|\varphi|$
Input: A graph $G$ and a first-order/monadic second-order sentence $\varphi \in F O / M S O(\{E\})$
Question: $G \models \varphi$ ?

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Example:

$$
\varphi=\exists x_{1} \exists x_{2} \cdots \exists x_{k} \forall x \bigvee_{1 \leqslant i \leqslant k} x=x_{i} \vee \bigvee_{1 \leqslant i \leqslant k} E\left(x, x_{i}\right) \vee E\left(x_{i}, x\right)
$$

$G \models \varphi ? \Leftrightarrow$

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

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\varphi=\exists x_{1} \exists x_{2} \cdots \exists x_{k} \bigwedge_{1 \leqslant i<j \leqslant k} \neg\left(x_{i}=x_{j}\right) \wedge \neg E\left(x_{i}, x_{j}\right) \wedge \neg E\left(x_{j}, x_{i}\right)
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$G \models \varphi ? \Leftrightarrow k$-Independent $\operatorname{Set}$

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$\varphi=\exists X_{1} \exists X_{2} \exists X_{3}\left(\forall x \bigvee_{1 \leqslant i \leqslant 3} X_{i}(x)\right) \wedge \forall x \forall y \bigwedge_{1 \leqslant i \leqslant 3}\left(X_{i}(x) \wedge X_{i}(y) \rightarrow \neg E(x, y)\right)$
$G \models \varphi ? \Leftrightarrow$

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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
- linear 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:

$$
\begin{aligned}
\operatorname{tp}_{k}^{\mathcal{L}}\left(\mathscr{A}, \vec{a} \in A^{m}\right) & =\{\varphi(\vec{x}) \in \mathcal{L}[k]: \mathscr{A} \models \varphi(\vec{a})\}, \\
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Theorem (folklore)
For $\mathcal{L} \in\{F O, M S O\}$, 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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Rank- $k$ types partition the graphs into $g(k)$ classes. Efficient Model Checking = quickly finding the class of the input.

## FO Ehrenfeucht-Fraissé game



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

## FO Ehrenfeucht-Fraissé game



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

## FO Ehrenfeucht-Fraissé game



Duplicator answers with a vertex in the other structure

## FO Ehrenfeucht-Fraissé game



After $q$ rounds, Duplicator wishes that $a_{i} \mapsto b_{i}$ is an isomorphism between $\mathscr{A}\left[a_{1}, \ldots, a_{k}\right]$ and $\mathscr{B}\left[b_{1}, \ldots, b_{k}\right]$

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When no longer possible, Spoiler wins

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If Duplicator can survive $k$ rounds, we write $\mathscr{A} \equiv{ }_{k}^{\mathrm{FO}} \mathscr{B}$ Here $\mathscr{A} \equiv{ }_{2}^{\mathrm{FO}} \mathscr{B}$ and $\mathscr{A} \not \equiv{ }_{3}^{\mathrm{FO}} \mathscr{B}$

## MSO Ehrenfeucht-Fraissé game



Same game but Spoiler can now play set moves

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Same game but Spoiler can now play set moves

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To which Duplicator answers a set in the other structure

## MSO Ehrenfeucht-Fraissé game



Again we write $\mathscr{A} \equiv{ }_{k}^{\mathrm{MSO}} \mathscr{B}$ 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 $\mathcal{L} \in\{F O, M S O\}$,

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\mathscr{A} \equiv \equiv_{k}^{\mathcal{L}} \mathscr{B} \text { if and only if } t p_{k}^{\mathcal{L}}(\mathscr{A})=t p_{k}^{\mathcal{L}}(\mathscr{B}) .
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Proof.
Induction on $k$.
$(\Rightarrow) \mathcal{L}[k+1]$ 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 $X=A$.

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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 $b$.

MSO model checking for component twin-width $d$
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 $G_{i}$

$$
\operatorname{tp}_{k}^{\mathrm{MSO}}\left(G, \mathcal{P}_{i}, C\right)=\left\{\varphi \in \mathrm{MSO}_{E, U_{1}, \ldots, U_{d}}(k):\left(G\langle C\rangle, \mathcal{P}_{i}\langle C\rangle\right) \models \varphi\right\} .
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For each $v \in V(G), \operatorname{tp}_{k}\left(G, \mathcal{P}_{n},\{v\}\right)=$ type of $K_{1}$

$$
\operatorname{tp}_{k}\left(G, \mathcal{P}_{1},\{V(G)\}\right)=\text { type of } G
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$$
\tau=\operatorname{tp}_{k}^{\mathrm{MSO}}\left(G, \mathcal{P}_{i}, C\right) \text { based on the } \tau_{j}=\operatorname{tp}_{k}^{\mathrm{MSO}}\left(G, \mathcal{P}_{i+1}, C_{j}\right) ?
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$C$ arises from $C_{1}, \ldots, C_{d^{\prime}}: \tau=F\left(\tau_{1}, \ldots, \tau_{d^{\prime}}, B, X, Y\right)$

Showing $\tau=F\left(\tau_{1}, \ldots, \tau_{d^{\prime}}, B, X, Y\right)$ via MSO EF game


Duplicator combines her strategies in the red components

Showing $\tau=F\left(\tau_{1}, \ldots, \tau_{d^{\prime}}, B, X, Y\right)$ via MSO EF game


If Spoiler plays a vertex in the component of type $\tau_{1}$,

Showing $\tau=F\left(\tau_{1}, \ldots, \tau_{d^{\prime}}, B, X, Y\right)$ via MSO EF game


Duplicator answers the corresponding winning move

Showing $\tau=F\left(\tau_{1}, \ldots, \tau_{d^{\prime}}, B, X, Y\right)$ via MSO EF game


Same in the component of type $\tau_{2}$

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## 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 $(H, \mathcal{P})_{\tau}$ 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 $F\left(\tau_{1}, \ldots, \tau_{d^{\prime}}, B, X, Y\right)$ for future uses


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

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,
- $K_{t}$-minor free graphs,
- map graphs,
- subgraphs of d-dimensional grids,
- $K_{t}$-free unit d-dimensional ball graphs,
- $\Omega(\log n)$-subdivisions of all the $n$-vertex graphs,
- cubic expanders defined by iterative random 2-lifts from $K_{4}$,
- 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: $G=G_{n}, G_{n-1}, \ldots, G_{2}, G_{1}=K_{1}$

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 $G$
Running time: $d^{2 k} n^{2}$ red connected subgraphs, actually only $d^{2 k} n=2^{O_{d}(k)} n$ updates

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How to compute $T[D, i]$ from all the $T\left[D^{\prime}, i+1\right]$ ?
k-Independent Set: Update of partial solutions


Best partial solution inhabiting •?
k-Independent Set: Update of partial solutions


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

## FO model checking on graphs of bounded twin-width

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.

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Add Gaifman's locality to our MSO model checking algorithm

## Local tuple of parts

$\left(P_{1}, P_{2}, \ldots, P_{q}\right)$ is local around $P_{1}$ if...

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## Local tuple of parts

( $P_{1}, P_{2}, \ldots, P_{q}$ ) is local around $P_{1}$ if...
$P_{4}$ is at distance at most $2^{k-4}$ from $\left\{P_{1}, P_{2}, P_{3}\right\}$ in ( $G, \mathcal{P}_{i}$ )

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## Partitioned local sentences and types

A prenex sentence is partitioned local around $X$ in $\left(G, \mathcal{P}_{i}\right)$ if of the form $Q x_{1} \in X Q x_{2} \in P_{2} \ldots Q x_{k} \in P_{k} \psi\left(x_{1}, \ldots, x_{k}\right)$ with

- $\psi$ is quantifier-free, and
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And the corresponding types:

$$
\operatorname{ltp}_{k}^{\mathrm{FO}}\left(G, \mathcal{P}_{i}, X\right)=\{\varphi: \operatorname{qr}(\varphi) \leqslant k
$$

$\varphi$ is partitioned local around $X$ in $\left(G, \mathcal{P}_{i}\right)$,

$$
\left.\left(G, \mathcal{P}_{i}\right) \models \varphi\right\} .
$$

## Partitioned local sentences/types in $\left(G, \mathcal{P}_{n}\right)$ and $\left(G, \mathcal{P}_{1}\right)$

Initialization of the dynamic programming
In $\left(G, \mathcal{P}_{n}=\{\{v\}: v \in V(G)\}\right)$ : for every $v \in V(G)$,
$Q x_{1} \in\{v\} Q x_{2} \in\{v\} \ldots Q x_{k} \in\{v\} \psi \equiv \psi(v, v, \ldots, v)$
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Output of the dynamic programming
$\ln \left(G, \mathcal{P}_{1}=\{V(G)\}\right):$
$Q x_{1} \in V(G) Q x_{2} \in V(G) \ldots Q x_{k} \in V(G) \psi \equiv$ classic sentences
The partitioned local type in $\left(G, \mathcal{P}_{1}\right)$ coincides with the type of $G$

## Partitioned local types give the partitioned types

Isom. $f: \mathcal{P}_{i} \rightarrow \mathcal{P}_{i}^{\prime}$ with $\operatorname{Itp}_{k}^{\mathrm{FO}}\left(G, \mathcal{P}_{i}, X\right)=\operatorname{ltp}_{k}^{\mathrm{FO}}\left(G^{\prime}, \mathcal{P}_{i}^{\prime}, f(X)\right)$
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Local strategies win the global game

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Say, Spoiler plays in $P_{1}$

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Duplicator answers in $f\left(P_{1}\right)$ following the local game around $P_{1}$

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Now when Spoiler plays close to $P_{1}$ or $f\left(P_{1}\right)$

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If Spoiler plays too far

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## Concluding as in the MSO model checking algorithm

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

Partitioned local types around $P$

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Each contraction: $O_{d, k}(1)=O\left(d^{2^{k}}\right)$ 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 |
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Bounded degree, bounded expansion, nowhere denseness?

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Thank you for your attention!

