The combinatorics of monadic stability, monadic dependence, and related notions

Algomanet, Warsaw, September 9-13, 2024

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Understanding Monadic Stability and

# Monadic Dependence via

logic,
combinatorics, and
algorithms.

# Map of the Universe



 Monday morning: meta-theorems, logic, nowhere dense (+exercises)

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### MOTIVATION

#### INDEPENDENTSET

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#### INDEPENDENTSET is NP-complete

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INDEPENDENTSET is NP-complete on planar graphs.



Assign each instance a number, called the parameter. We hope that

- we can solve the instance if the parameter is small,
- interesting instances have a small parameter.

NP-hard problems may still be tractable for small parameter values!

PARAMETERIZED INDEPENDENTSET	
Input:	Graph $G$ and integer $k$
Parameter:	k
Question:	Does $G$ have an independent set of size $k$ ?

A parameterized problem is *fixed parameter tractable* (fpt) if instances with parameter k and size n can be solved in time  $f(k)n^c$ (for some fixed function f and constant c).

IS PARAMETERIZED INDEPENDENTSET fixed parameter tractable?

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check if v_1, \ldots, v_k is an IS of size k
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PARAMETERIZED INDEPENDENTSET is not fixed parameter tractable (unless FPT = W[1]).

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PARAMETERIZED INDEPENDENTSET is fixed parameter tractable on planar graphs.



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- $\bigcirc$  We guess w, place w in solution and remove N(w).
- $\bigcirc$  Then find a solution of size k-1 in remaining graph.


```
IS(G, k):

if G is empty return k == 0

find vertex v with degree \leq 5 in G

for all w \in N(v):

if IS(G \setminus N(w), k - 1) return True

return False
```

This solves parameterized independentSet on planar graphs in time  $O(6^k n)$ .

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But the problem becomes tractable if we both

- consider only planar graphs, and
- parameterize by the solution size.

Is parameterized Dominating Set FPT on planar graphs?
 Is parameterized Clique FPT on bounded genus graphs?
 ...

We would like a single mechanism that answers these and similar questions.

### Algorithmic Meta-Theorems:

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Our Goal:

- $\bigcirc$  L is first-order logic
- $\bigcirc$  P are monadically dependent graph classes

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  - the universe are the vertices
  - E denotes the binary adjacency relation between vertices
  - $\circ c_i$  denotes the unary relation "the vertex is colored with color i"

### Example

This graph is a structure G with

- $\bigcirc$  universe  $V = \{a, b, c\}$
- $\bigcirc$  symmetrical binary relation  $E := \{(a, b), (b, a), (b, c), (c, b), (a, c), (c, a)\}$
- $\bigcirc$  unary relations  $c_1 := \{a\}$ ,  $c_2 := \{c\}$



For a given signature  $\tau$ , first-order logic has ...

- $\bigcirc$  element-variables ( $x, y, z, \dots$ )
- $\bigcirc$  the equality relation = as well as the relations from  $\tau$ .
- $\bigcirc$  quantifiers  $\exists$  and  $\forall$ , as well as operators  $\land$ ,  $\lor$  and  $\neg$

We mostly work on colored undirected graphs with  $\tau = \{E, c_1, c_2, \dots\}.$ 

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### First-Order Model-Checking

Input: Graph G and first-order sentence  $\varphi$ Question:  $G \models \varphi$ ?

# Complexity

### Theorem (Vardi 1982)

The model-checking problem is PSPACE-complete.

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It is reasonable to assume that the length of the formula is small compared to the size of the graph. Parameterize by  $|\varphi|$ .

### Parameterized Complexity (Upper Bound)

#### Theorem

One can decide whether  $G \models \varphi$  in time  $O(|G|^{|\varphi|})$ .

Proof: Construct an evaluation tree of size  $O(|G|^{|\varphi|})$ .


Conjecture (based on SETH)

One cannot decide whether  $G \models \varphi$  in time  $O(|G|^{q-1-\varepsilon})$  for any  $\varepsilon > 0$  where q is the number of quantifiers of  $\varphi$ .

The previous algorithm is probably more or less optimal.

A faster model-checking algorithm would lead to an unexpected faster algorithm for many hard problems.

On certain graph classes, we can do much better though.

#### **Target Statement**

Let C be a "well-behaved" graph class. For an FO formula  $\varphi$ and graph  $G \in C$  one can decide whether  $G \models \varphi$  in time  $f(|\varphi|)n^{10}$  for some function f.

#### **Target Statement**

Let  $\mathcal{C}$  be a "well-behaved" graph class. For an FO formula  $\varphi$ and graph  $G \in \mathcal{C}$  one can decide whether  $G \models \varphi$  in time  $f(|\varphi|)n^{10}$  for some function f.

Examples of "well-behaved" classes:

- bounded degree
- planar graphs
- 0 ...

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...can be solved in time  $f(k) \cdot n^{10}$  on well-behaved graph classes.

Does the algorithic meta-theorem give an fpt algorithm for the following problem?

- $\bigcirc$  Input: SAT-instance with planar incidence graph,  $k \in \mathbb{N}$ .
- $\bigcirc$  Parameter: k.
- Question: is there a satisfying assignment with at most k variables set to true?

Does the algorithic meta-theorem give an fpt algorithm for the following problem?

- $\bigcirc$  Input: a planar graph G, and  $k \in \mathbb{N}$ .
- $\bigcirc$  Parameter: k.
- Is there a dominating set of size at most k that induces a connected subgraph?

Does the algorithic meta-theorem give an fpt algorithm for the following problem?

- $\bigcirc$  Input: a planar graph G, and  $k, d \in \mathbb{N}$ .
- $\bigcirc$  Parameter: k.
- Is it possible to remove k vertices such that every vertex has degree at most d?

### What other graph classes are "well-behaved"?

## Many Sparse Graph Classes



Figure by Felix Reidl

### Theorem (Grohe, Kreutzer, Siebertz 2017)

For a graph class C that is closed under subgraphs holds: C is nowhere dense iff the first-order model-checking problem on C is FPT (assuming FPT  $\neq$  AW[\*]).

Bounded Degree Model Checking: Seese, 1996 Planar Model Checking: Flum, Grohe 2001 Bounded Expansion Model Checking: Dvořák, Král, Thomas, 2010 Nowhere Dense Model Checking: Grohe, Kreutzer, Siebertz, 2017

A graph class C is *nowhere dense* if for every  $r \in \mathbb{N}$  there exists  $k \in \mathbb{N}$  such that no graph in C contains the r-subdivided clique of size k as a subgraph.

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Let C be nowhere dense. Prove that there exists t such that no graph in C contains the biclique  $K_{t,t}$  as a subgraph.

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Prove that the class of trees is nowhere dense

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Prove that the class of trees is nowhere dense

Prove that every class of bounded degree is nowhere dense.

## Many Sparse Graph Classes



Figure by Felix Reidl

# CAN WE GO BEYOND NOWHERE DENSE?

## Classes with FPT first-order model-checking?



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First-order model-checking is fpt on complements of nowhere dense classes by reduction.



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## Classes with FPT first-order model-checking?



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obtain  $\varphi'$  from  $\varphi$  by replacing  $x \sim y$  with dist(x, y) = 3

$$\Leftrightarrow$$



 $G\models\varphi$ 



obtain  $\varphi'$  from  $\varphi$  by replacing  $x \sim y$  with dist(x, y) = 3



also restrict quantifiers to black vertices



 $G\models\varphi$ 

 $\varphi$ -transduction: color vertices + apply  $\varphi$  + take induced subgraph



 $\varphi(x,y) := \operatorname{Red}(x) \wedge \operatorname{Red}(y) \wedge \operatorname{dist}(x,y) = 3$ 

A class  $\mathcal{D}$  is a *transduction* of a class  $\mathcal{C}$  if there exists  $\varphi$  such that every graph in  $\mathcal{D}$  is a  $\varphi$ -transduction of some graph in  $\mathcal{C}$ .










Gajarský, Kreutzer, Něsetřil, Ossona de Mendez, Pilipczuk, Siebertz, Toruńczyk, 2018. Něsetřil, Ossona de Mendez, 2016

A class is *structurally nowhere dense*, if it is a transduction of a nowhere dense graph class.

D, Mählmann, Siebertz, 2023

The first-order model-checking problem on  $\mathcal{C}$  is FPT on structurally nowhere dense graph classes.

## Classes with FPT first-order model-checking?



## Classes with FPT first-order model-checking?



## Monadic Stability/Dependence

Baldwin, Shelah, 1985

A class is *monadically stable*, if it does not transduce the class of all half-graphs.



A class is *monadically dependent*, if it does not transduce the class of all graphs.

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Every structurally nowhere dense class is monadically stable.

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D, Eleftheriadis, Mählmann, McCarty, Pilipczuk, Toruńczyk 2024

Let  $\mathcal C$  be monadically stable. The first-order model-checking problem is FPT on  $\mathcal C.$ 

## Classes with FPT first-order model-checking?



## Classes with FPT first-order model-checking?



#### You already know normal graphs.



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In trigraphs there are additional red error edges.

We can contract two (not neccessarily adjacent) vertices *a* and *b*. The edges of the new vertex *ab* follow this table.





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A *contraction sequence* is a sequence of contractions until only a single vertex is left.

e





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## 

#### Bonnet, Kim, Thomassé, Watrigant 2021



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Twinwidth: Smallest integer *d* such there is a contraction sequence where the red degree is *at all times* at most *d*.

'e t



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The following classes have bounded twinwidth

- planar graphs,
- $\bigcirc$  classes with bounded cliquewidth.

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The following classes do not have bounded twinwidth

 $\bigcirc$  graphs with degree three.

#### Trees have twinwidth at most two. Strategy:



Trees have twinwidth at most two. Strategy: When possible contract twin leafs.



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#### Bonnet, Kim, Thomassé, Watrigant 2021

Let C be a class of bounded twinwidth. Then first-order model-checking is fpt on C, if one is additionally provided a contraction sequence of bounded twinwidth.









## Map of the Universe



### EXERCISES

Show that the class of empty graphs is monadically stable/dependent.

### Argue: If C transduces D and D transduces E, then C transduces E.

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Argue that therefore monadically stable/dependent classes are closed under transductions.

Let  $\mathcal C$  be the class of graphs of degree at most three. Show that  $\mathcal C$  is monadically stable/dependent.

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Use the following theorem:

Corollary of Gaifman's Theorem

Let  $\varphi$  be a first-order formula. There is a number k with the following property. For every graph G there is a coloring  $c: V(G) \rightarrow [k]$  such that for all  $u, v \in V(G)$  with distance larger than k, the fact whether  $(u, v) \in E(G)$  depends only on (c(u), c(v)). Show that the class of "star matchings" and the class of "comparability grids" are not monadically dependent.

### APPENDIX

For every graph class  $\mathcal C$  that is closed under subgraphs:

 ${\mathcal C}$  is nowhere dense if and only if  ${\mathcal C}$  is monadically dependent.

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For a graph class C that is closed under subgraphs holds: C is monadically dependent iff the first-order modelchecking problem on C is FPT (assuming FPT  $\neq$  AW[\*]).

A graph class C is *unordered* if for some k it excludes the following graphs of order k as induced subgraphs.



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For every unordered graph class C:

 ${\mathcal C}$  is monadically stable if and only if  ${\mathcal C}$  is monadically dependent.

A graph class C is *unordered* if for some k it excludes the following graphs of order k as induced subgraphs.



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D, Eleftheriadis, Mählmann, McCarty, Pilipczuk, Toruńczyk 2024 For an unordered graph class C that is closed under induced subgraphs holds:  $\mathcal C$  is monadically dependent iff the first-order modelchecking problem on C is FPT (assuming FPT  $\neq$  AW[\*]).

An *ordered graph* is a graph together with a total order on its vertices (which can be queried by first-order logic).



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Bonnet, Giocanti, Ossona de Mendez, Simon, Thomassé, Toruńczyk 2021

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## Map of the Universe



Figure by Michał Pilipczuk