



Algorithmic Metatheorems for Clique-Guarded First-Order Logic with Counting

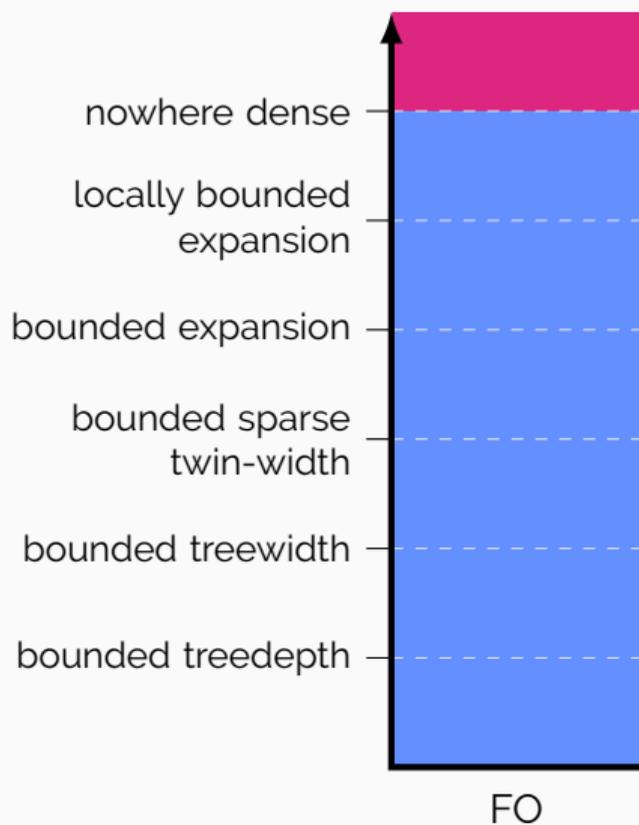
Steffen van Bergerem, Johannes Lange, and Nicole Schweikardt

ALMoTh 2026

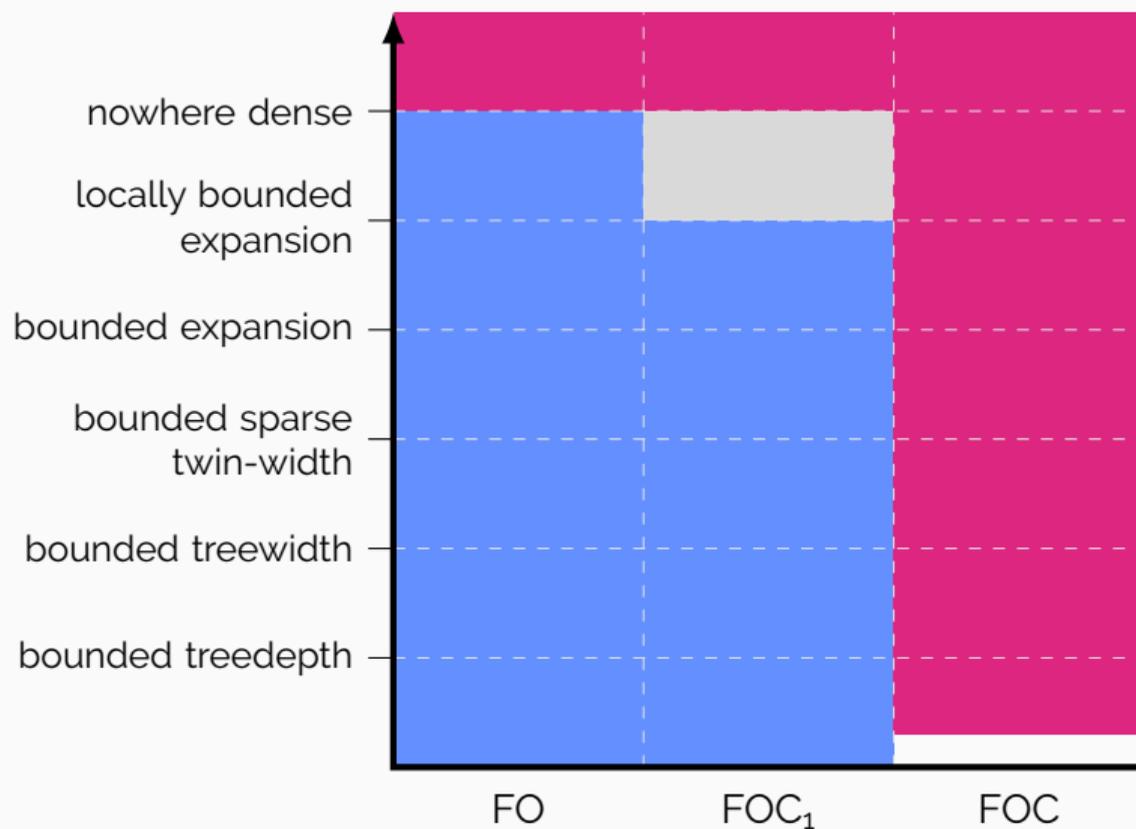
First-Order Model Checking

Given a relational structure \mathcal{A} and an FO sentence φ
Decide whether $\mathcal{A} \models \varphi$

Model Checking on Sparse Classes

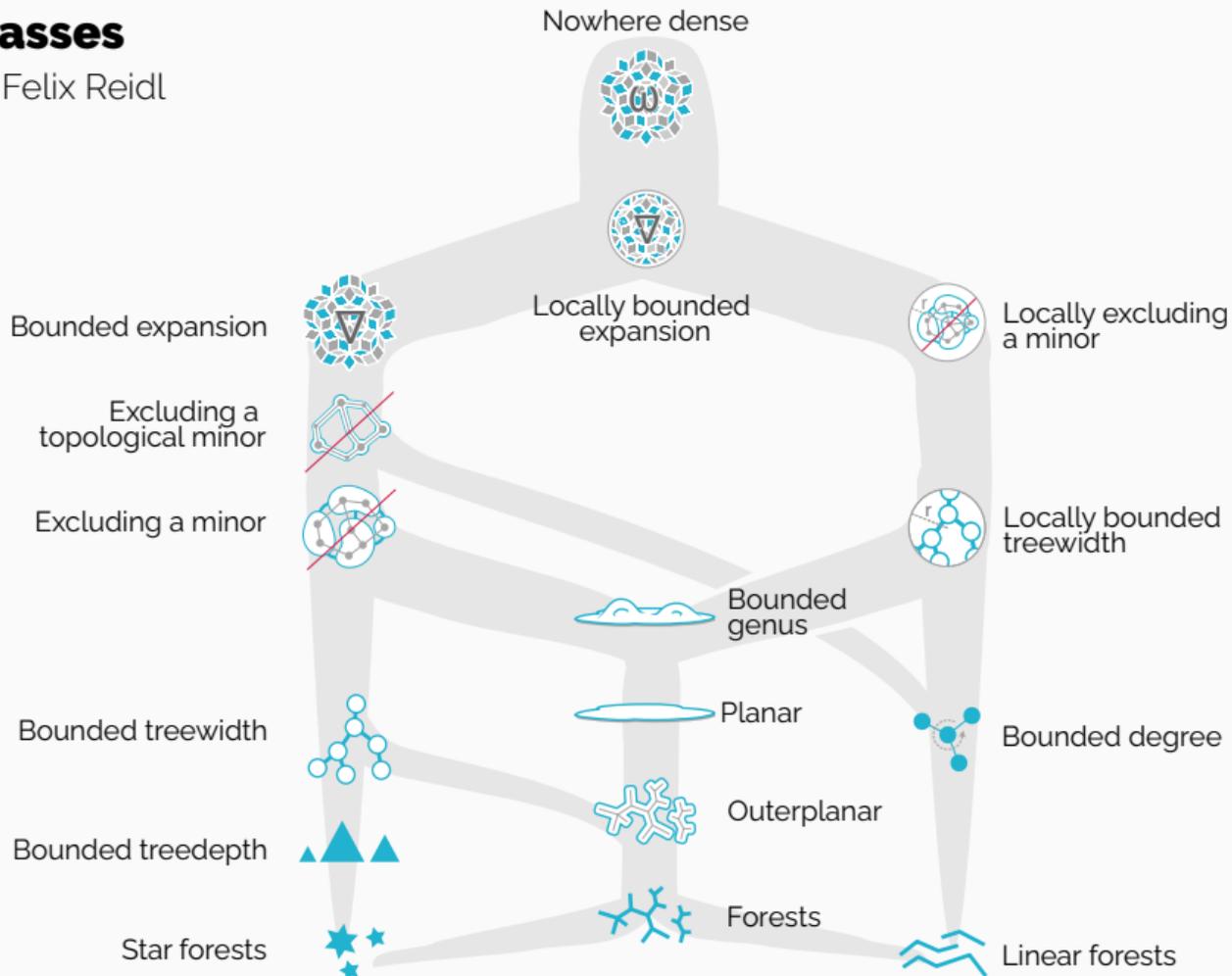


Model Checking on Sparse Classes



Sparse Classes

Illustration by Felix Reidl



First-Order Logic with Counting (FOC)

Counting terms

- i for every integer $i \in \mathbb{Z}$
- $\#(y_1, \dots, y_k). \varphi(\bar{x}, \bar{y})$ for every FOC formula $\varphi(\bar{x}, \bar{y})$
- $t_1 + t_2$ and $t_1 \cdot t_2$ for all FOC counting terms t_1, t_2

Formulae

First-Order Logic with Counting (FOC)

Counting terms

- i for every integer $i \in \mathbb{Z}$
- $\#(y_1, \dots, y_k). \varphi(\bar{x}, \bar{y})$ for every FOC formula $\varphi(\bar{x}, \bar{y})$
- $t_1 + t_2$ and $t_1 \cdot t_2$ for all FOC counting terms t_1, t_2

Formulae

- every FO formula
- $\neg \varphi_1, \varphi_1 \wedge \varphi_2, \varphi_1 \vee \varphi_2, \exists x \varphi_1$ for all FOC formulae φ_1, φ_2
- $P(t_1, \dots, t_m)$ for all FOC counting terms t_1, \dots, t_m and $P \in \mathbb{P}, P \subseteq \mathbb{Z}^m$

First-Order Logic with Counting (FOC)

Counting terms

- $t_1(x) = \#(y).E(x,y)$
- $t_2 = \#(x_1, \dots, x_k).(\bigwedge_{1 \leq i < j \leq k} E(x_i, x_j))$

Formulae

First-Order Logic with Counting (FOC)

Counting terms

- $t_1(x) = \#(y).E(x,y)$
- $t_2 = \#(x_1, \dots, x_k).(\bigwedge_{1 \leq i < j \leq k} E(x_i, x_j))$

Formulae

- $\varphi_1(x) = (\#(y).E(x,y) \leq \#(y).\neg E(x,y))$
- $\varphi_2(x) = \exists y (\#(z).(E(x,z) \wedge E(z,y)) \geq 2)$

First-Order Logic with Counting (FOC)

Counting terms

- $t_1(x) = \#(y).E(x,y)$
- $t_2 = \#(x_1, \dots, x_k).(\bigwedge_{1 \leq i < j \leq k} E(x_i, x_j))$
- $t_3 = t_2 + \#(x).\varphi_1(x)$

Formulae

- $\varphi_1(x) = (\#(y).E(x,y) \leq \#(y).\neg E(x,y))$
- $\varphi_2(x) = \exists y (\#(z).(E(x,z) \wedge E(z,y)) \geq 2)$

First-Order Logic with Counting (FOC)

Counting terms

- $t_1(x) = \#(y).E(x, y)$
- $t_2 = \#(x_1, \dots, x_k).(\bigwedge_{1 \leq i < j \leq k} E(x_i, x_j))$
- $t_3 = t_2 + \#(x).\varphi_1(x)$

Formulae

- $\varphi_1(x) = (\#(y).E(x, y) \leq \#(y).\neg E(x, y))$
- $\varphi_2(x) = \exists y (\#(z).(E(x, z) \wedge E(z, y)) \geq 2)$
- $\varphi_3(x) = \exists y E(x, y) \wedge (t_1(x) \leq t_1(y))$

First-Order Logic with Counting (FOC)

Counting terms

- $t_1(x) = \#(y).E(x,y)$
- $t_2 = \#(x_1, \dots, x_k).(\bigwedge_{1 \leq i < j \leq k} E(x_i, x_j))$
- $t_3 = t_2 + \#(x).\varphi_1(x)$

Formulae

- $\varphi_1(x) = (\#(y).E(x,y) \leq \#(y).\neg E(x,y))$
- $\varphi_2(x) = \exists y (\#(z).(E(x,z) \wedge E(z,y)) \geq 2)$
- $\varphi_3(x) = \exists y E(x,y) \wedge (t_1(x) \leq t_1(y))$

FOC₁

- introduced by Grohe and Schweikardt (PODS 2018)
- subformulae comparing counting terms have ≤ 1 free variable

First-Order Logic with Counting (FOC)

Counting terms

- $t_1(x) = \#(y).E(x,y)$
- $t_2 = \#(x_1, \dots, x_k).(\bigwedge_{1 \leq i < j \leq k} E(x_i, x_j))$
- $t_3 = t_2 + \#(x).\varphi_1(x)$

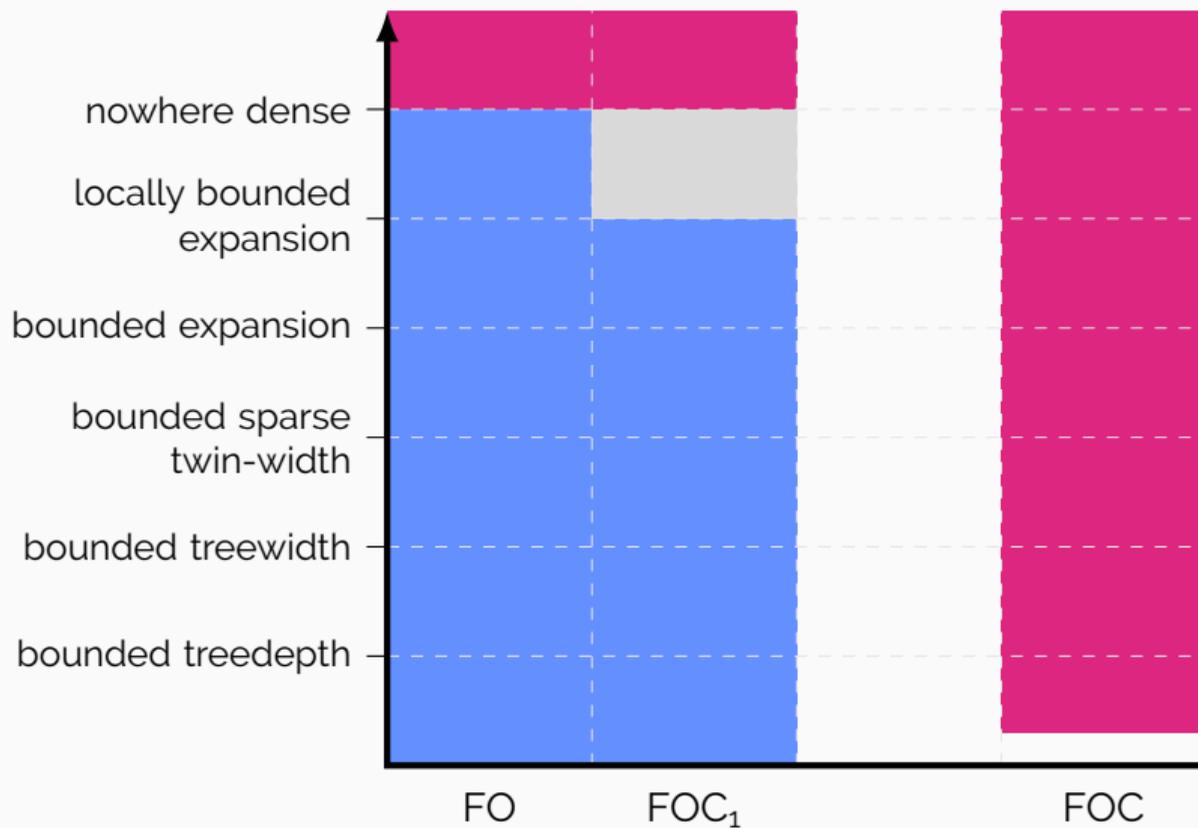
Formulae

- $\varphi_1(x) = (\#(y).E(x,y) \leq \#(y).\neg E(x,y))$
- $\varphi_2(x) = \exists y (\#(z).(E(x,z) \wedge E(z,y)) \geq 2)$
- $\varphi_3(x) = \exists y E(x,y) \wedge (t_1(x) \leq t_1(y))$

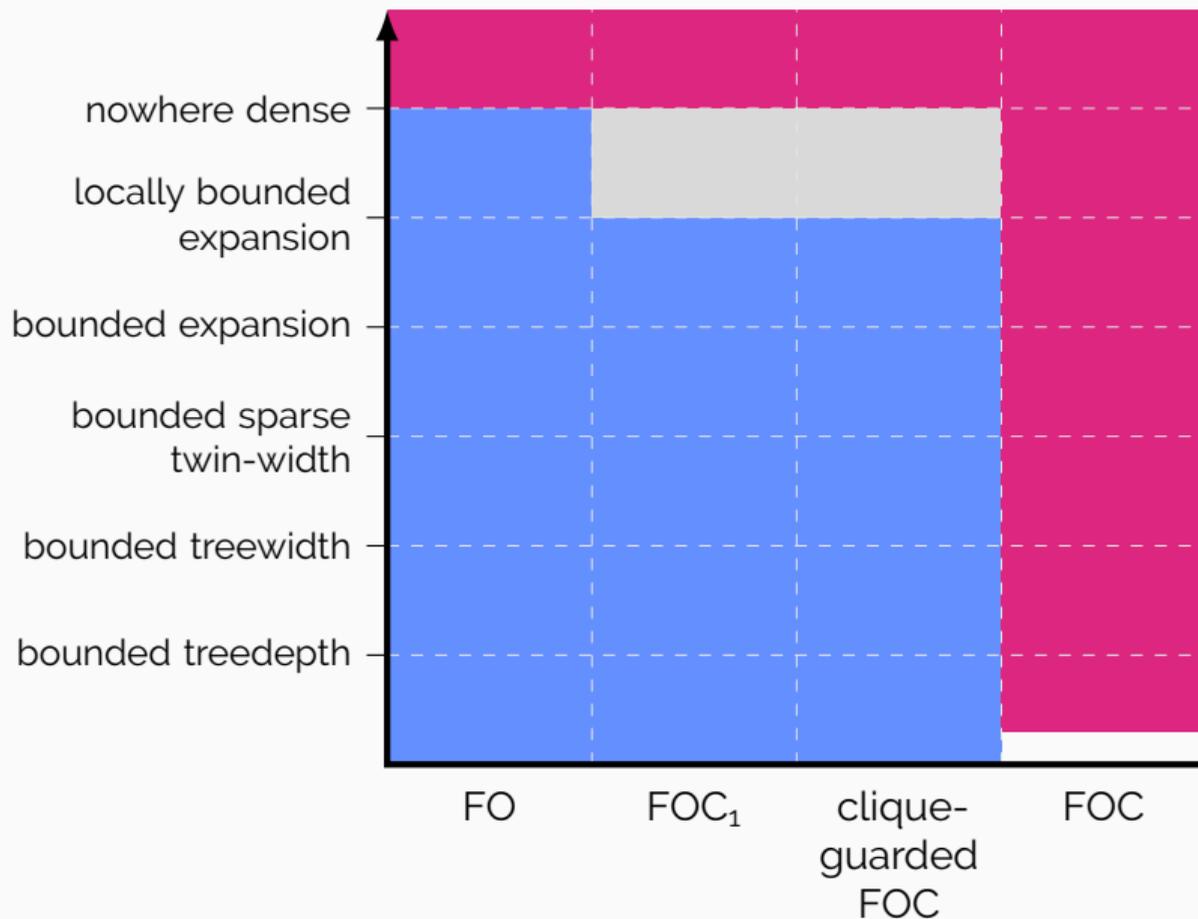
FOC₁

- introduced by Grohe and Schweikardt (PODS 2018)
- subformulae comparing counting terms have ≤ 1 free variable

Evaluation on Sparse Classes



Evaluation on Sparse Classes



Clique-Guarded First-Order Logic with Counting (cgFOC)

FOC₁

subformulae comparing counting terms have ≤ 1 free variable

Clique-Guarded First-Order Logic with Counting (cgFOC)

FOC₁

subformulae comparing counting terms have ≤ 1 free variable

clique-guarded FOC (“free variables have distance at most 1”)

every pair of free variables in subformulae comparing counting terms is guarded by an atom

Clique-Guarded First-Order Logic with Counting (cgFOC)

FOC₁

subformulae comparing counting terms have ≤ 1 free variable

clique-guarded FOC (“free variables have distance at most 1”)

every pair of free variables in subformulae comparing counting terms is guarded by an atom

- $\varphi_1(x_1, x_2, x_3) = E(x_1, x_2) \wedge E(x_1, x_3) \wedge E(x_2, x_3) \wedge (t_1(x_1, x_2) = t_2(x_3))$
for cgFOC counting terms t_1, t_2

Clique-Guarded First-Order Logic with Counting (cgFOC)

FOC₁

subformulae comparing counting terms have ≤ 1 free variable

clique-guarded FOC (“free variables have distance at most 1”)

every pair of free variables in subformulae comparing counting terms is guarded by an atom

- $\varphi_1(x_1, x_2, x_3) = E(x_1, x_2) \wedge E(x_1, x_3) \wedge E(x_2, x_3) \wedge (t_1(x_1, x_2) = t_2(x_3))$
for cgFOC counting terms t_1, t_2
- $\varphi_2(x_1, x_2, x_3, x_4) = R(x_1, x_3, x_4) \wedge R(x_2, x_3, x_4) \wedge E(x_1, x_2)$
 $\wedge (t_1(x_1, x_4) + t_2(x_2) = t_1(x_2, x_4))$

Clique-Guarded First-Order Logic with Counting (cgFOC)

FOC₁

subformulae comparing counting terms have ≤ 1 free variable

clique-guarded FOC (“free variables have distance at most 1”)

every pair of free variables in subformulae comparing counting terms is guarded by an atom

- $\varphi_1(x_1, x_2, x_3) = E(x_1, x_2) \wedge E(x_1, x_3) \wedge E(x_2, x_3) \wedge (t_1(x_1, x_2) = t_2(x_3))$
for cgFOC counting terms t_1, t_2
- $\varphi_2(x_1, x_2, x_3, x_4) = R(x_1, x_3, x_4) \wedge R(x_2, x_3, x_4) \wedge E(x_1, x_2)$
 $\wedge (t_1(x_1, x_4) + t_2(x_2) = t_1(x_2, x_4))$
- $t(\bar{x}) = \#\bar{y}.\varphi(\bar{x}, \bar{y})$ for a cgFOC formula φ

Clique-Guarded First-Order Logic with Counting (cgFOC)

FOC₁

subformulae comparing counting terms have ≤ 1 free variable

clique-guarded FOC (“free variables have distance at most 1”)

every pair of free variables in subformulae comparing counting terms is guarded by an atom

- $\varphi_1(x_1, x_2, x_3) = E(x_1, x_2) \wedge E(x_1, x_3) \wedge E(x_2, x_3) \wedge (t_1(x_1, x_2) = t_2(x_3))$
for cgFOC counting terms t_1, t_2
- $\varphi_2(x_1, x_2, x_3, x_4) = R(x_1, x_3, x_4) \wedge R(x_2, x_3, x_4) \wedge E(x_1, x_2)$
 $\wedge (t_1(x_1, x_4) + t_2(x_2) = t_1(x_2, x_4))$
- $t(\bar{x}) = \#\bar{y}.\varphi(\bar{x}, \bar{y})$ for a cgFOC formula φ

not clique-guarded FOC

- $\psi(x_1, x_2, x_3) = E(x_1, x_2) \wedge E(x_2, x_3) \wedge (t(x_1) \leq t(x_3))$

Main Results

Main Results

Tractability

- query answering
- enumeration
- agnostic PAC learning

Hardness

- model checking for variations of cgFOC

Query Answering

Preprocessing

Given a σ -structure \mathcal{A} ,
a cgFOC expression $\xi(\bar{x})$

Query answering

Given a tuple $\bar{v} \in A^{|\bar{x}|}$
Output $[[\xi(\bar{v})]]^{\mathcal{A}}$

Enumeration

Preprocessing

Given a σ -structure \mathcal{A} ,
a cgFOC formula $\xi(\bar{x})$

Enumeration

Output all tuples $\bar{v} \in A^{|\bar{x}|}$ with $\mathcal{A} \models \xi(\bar{v})$
in lexicographic order and
without duplicates

Query Answering

Preprocessing

Given a σ -structure \mathcal{A} ,
a cgFOC expression $\xi(\bar{x})$

Query answering

Given a tuple $\bar{v} \in A^{|\bar{x}|}$
Output $\llbracket \xi(\bar{v}) \rrbracket^{\mathcal{A}}$

Enumeration

Preprocessing

Given a σ -structure \mathcal{A} ,
a cgFOC formula $\xi(\bar{x})$

Enumeration

Output all tuples $\bar{v} \in A^{|\bar{x}|}$ with $\mathcal{A} \models \xi(\bar{v})$
in lexicographic order and
without duplicates

v. B., Lange, and Schweikardt, 2026+

For every class \mathcal{C} of **locally bounded expansion**, there are algorithms that solve the **query-answering** || **enumeration** problem for cgFOC on \mathcal{C} with

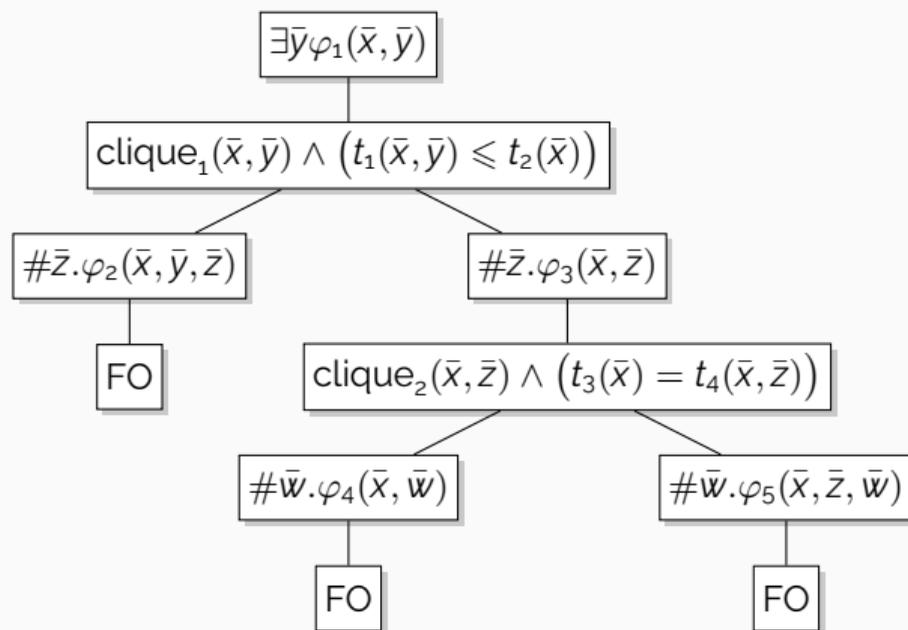
- preprocessing in time $f(\xi, \sigma, \varepsilon) \cdot |A|^{1+\varepsilon}$
- query answering in time $f(\xi, \sigma, \varepsilon)$ || $f(\xi, \sigma, \varepsilon)$ delay.

Query Answering and Enumeration: Main Proof Ideas

- Toruńczyk (PODS 2020): query answering on bounded expansion for queries of the form $\#\bar{y}.\varphi(\bar{x}, \bar{y})$ for an FO formula φ
- we generalise this to classes of locally bounded expansion

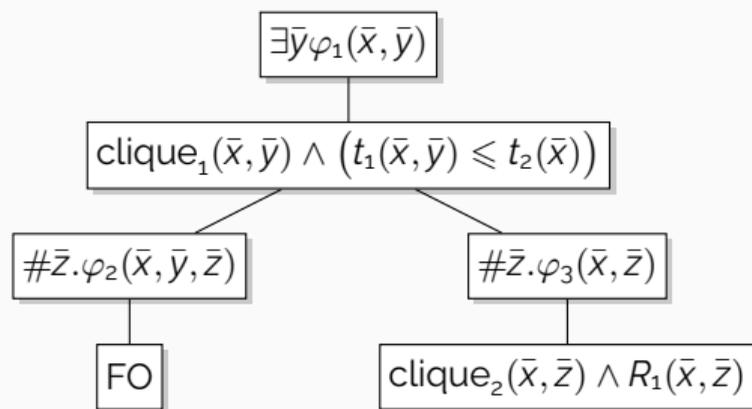
Query Answering and Enumeration: Main Proof Ideas

- Toruńczyk (PODS 2020): query answering on bounded expansion for queries of the form $\#\bar{y}.\varphi(\bar{x}, \bar{y})$ for an FO formula φ
- we generalise this to classes of locally bounded expansion



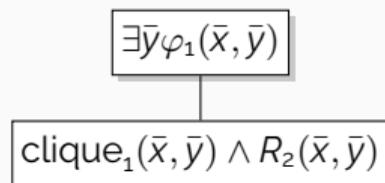
Query Answering and Enumeration: Main Proof Ideas

- Toruńczyk (PODS 2020): query answering on bounded expansion for queries of the form $\#\bar{y}.\varphi(\bar{x}, \bar{y})$ for an FO formula φ
- we generalise this to classes of locally bounded expansion



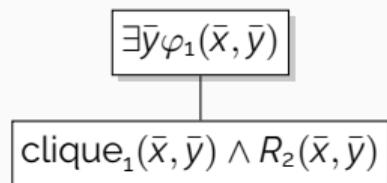
Query Answering and Enumeration: Main Proof Ideas

- Toruńczyk (PODS 2020): query answering on bounded expansion for queries of the form $\#\bar{y}.\varphi(\bar{x}, \bar{y})$ for an FO formula φ
- we generalise this to classes of locally bounded expansion



Query Answering and Enumeration: Main Proof Ideas

- Toruńczyk (PODS 2020): query answering on bounded expansion for queries of the form $\#\bar{y}.\varphi(\bar{x}, \bar{y})$ for an FO formula φ
- we generalise this to classes of locally bounded expansion



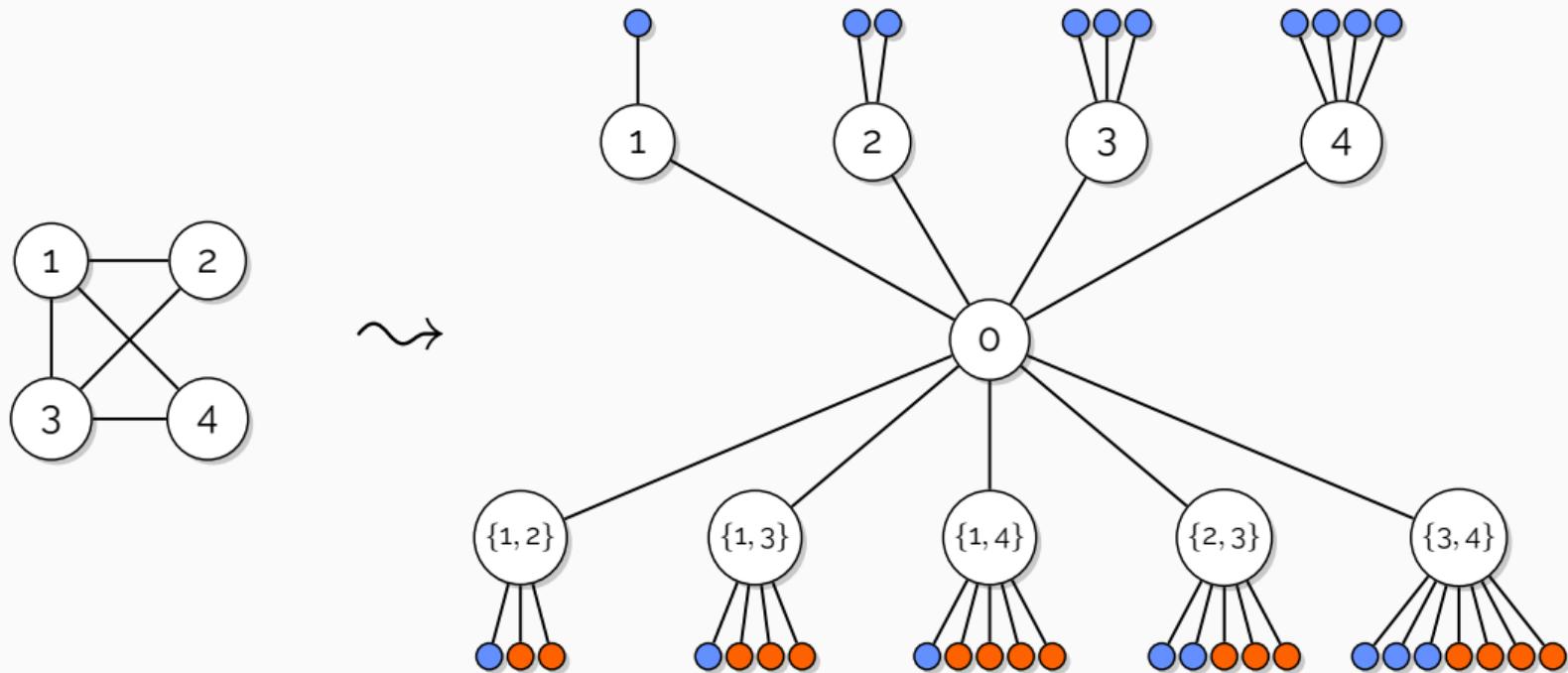
- iteratively replace subformulae comparing counting terms by new relation symbols
- because of “clique-guarded”, when computing the new relations,
 - we only need to consider tuples that form a clique, number of those is small in classes of locally bounded expansion
 - the new relations do not change the Gaifman graph
- in the end, we only need to handle a first-order formula

Hardness

v. B., Lange, and Schweikardt, 2026+

If we allow constructs of the form $E(x_1, x_2) \wedge E(x_2, x_3) \wedge (t(x_1) = t'(x_3))$, then the model-checking problem for the resulting logic ("**2-guarded**" FOC) is AW[*]-**hard** on the class of coloured **trees of height 2**.

Hardness of 2-Guarded FOC



Hardness

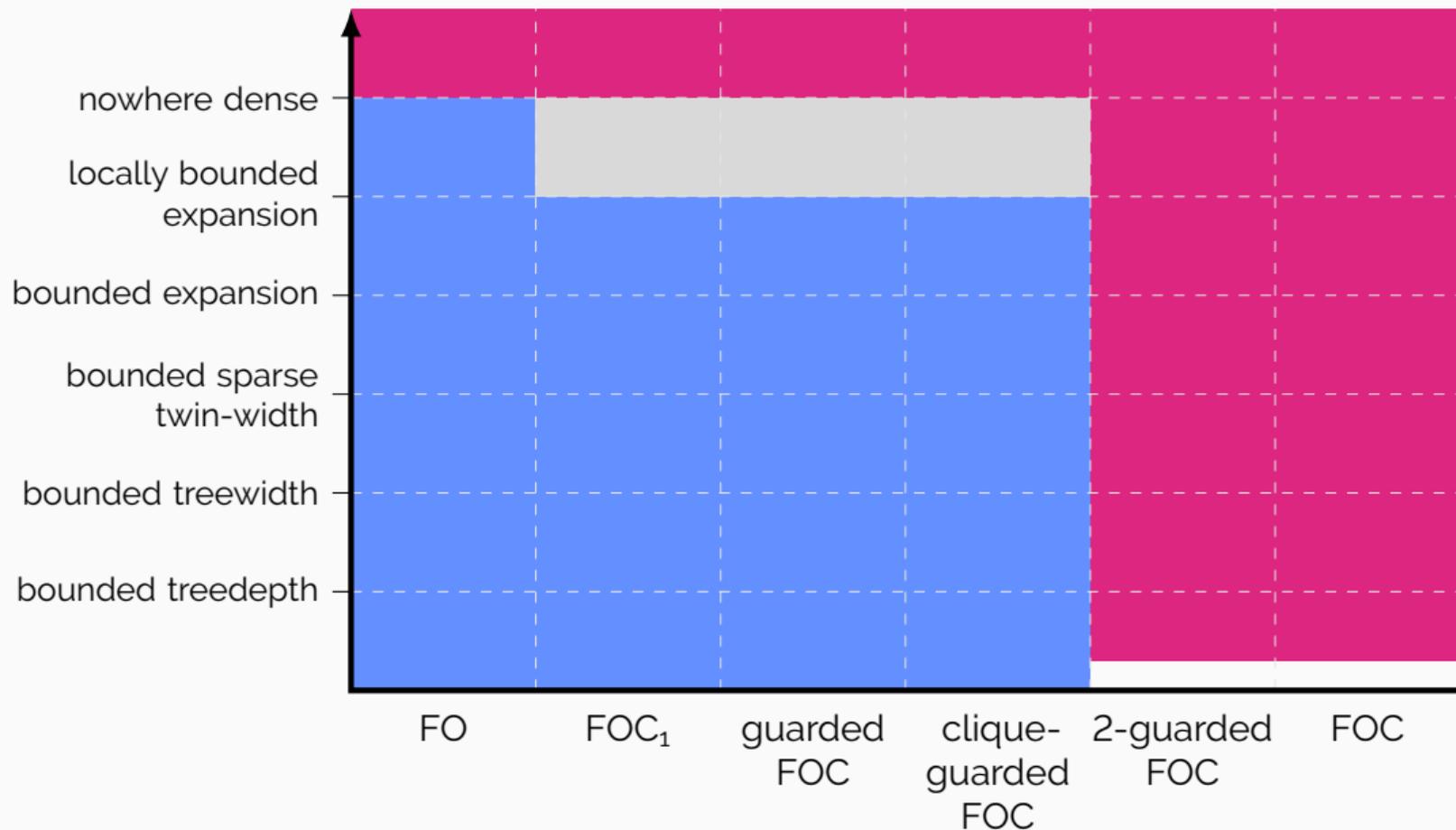
v. B., Lange, and Schweikardt, 2026+

If we allow constructs of the form $E(x_1, x_2) \wedge E(x_2, x_3) \wedge (t(x_1) = t'(x_3))$, then the model-checking problem for the resulting logic ("**2-guarded**" FOC) is AW[*]-**hard** on the class of coloured **trees of height 2**.

v. B., Lange, and Schweikardt, 2026+

Even if we only allow constructs of the form $E(x_1, x_2) \wedge (t(x_1) = t'(x_2))$, the model-checking problem for the resulting logic ("**guarded**" FOC) is AW[*]-**hard** on a class of coloured **graphs of shrub-depth 2**.

Evaluation on Sparse Classes



Agnostic PAC Learning

- agnostic Probably Approximately Correct learning
- fix $k \in \mathbb{N}_{\geq 1}$, $\ell, q \in \mathbb{N}$
- assume probability distribution \mathcal{D} on $A^k \times \{0, 1\}$
- we want to approximate \mathcal{D} by a cgFOC formula

Agnostic PAC Learning

- agnostic Probably Approximately Correct learning
- fix $k \in \mathbb{N}_{\geq 1}$, $\ell, q \in \mathbb{N}$
- assume probability distribution \mathcal{D} on $A^k \times \{0, 1\}$
- we want to approximate \mathcal{D} by a cgFOC formula
- algorithm is given a structure \mathcal{A} and draws examples from the distribution
- **Goal:** return a cgFOC formula $\varphi(x_1, \dots, x_k)$ with $\text{qr}(\varphi) \leq q$ that uses at most ℓ vertices from A as constants and has small expected error

$$\Pr_{(\bar{v}, \lambda) \sim \mathcal{D}} (\llbracket \varphi(\bar{v}) \rrbracket^{\mathcal{A}} \neq \lambda)$$

Agnostic PAC Learning

v. B., Lange, and Schweikardt, 2026+

On every class of locally bounded expansion, there is an algorithm that solves the agnostic-PAC-learning problem for cgFOC in time $\mathcal{O}(|A|^{1+\varepsilon})$.

Agnostic PAC Learning

v. B., Lange, and Schweikardt, 2026+

On every class of locally bounded expansion, there is an algorithm that solves the agnostic-PAC-learning problem for cgFOC in time $\mathcal{O}(|A|^{1+\epsilon})$.

- we even solve a more powerful problem, which we call **agnostic PAC enumeration**, with $\mathcal{O}(|A|^{1+\epsilon})$ preprocessing and $\mathcal{O}(1)$ delay
- this enumerates all formulae within the given complexity bounds, sorted by their (approximate) error

Agnostic PAC Learning

v. B., Lange, and Schweikardt, 2026+

On every class of locally bounded expansion, there is an algorithm that solves the agnostic-PAC-learning problem for cgFOC in time $\mathcal{O}(|A|^{1+\epsilon})$.

- we even solve a more powerful problem, which we call **agnostic PAC enumeration**, with $\mathcal{O}(|A|^{1+\epsilon})$ preprocessing and $\mathcal{O}(1)$ delay
- this enumerates all formulae within the given complexity bounds, sorted by their (approximate) error

Proof idea

- cgFOC formulae have bounded VC dimension on classes of locally bounded expansion (next talk)
- implies that constant number of examples suffices for approximation
- encode these examples in the structure, use enumeration result

Conclusion

Tractability for cgFOC on classes of locally bounded expansion

- $f(\xi, \sigma, \varepsilon)$ -time **query answering** after $f(\xi, \sigma, \varepsilon) \cdot |A|^{1+\varepsilon}$ -time preprocessing
- $f(\xi, \sigma, \varepsilon)$ -delay **enumeration** after $f(\xi, \sigma, \varepsilon) \cdot |A|^{1+\varepsilon}$ -time preprocessing
- $\mathcal{O}(1)$ -delay **agnostic PAC enumeration** after $\mathcal{O}(|A|^{1+\varepsilon})$ -time preprocessing

Hardness

- "2-guarded" FOC model checking is AW[*]-hard on trees of height 2
- "guarded" FOC model checking is AW[*]-hard on a class of shrub-depth 2

Conclusion

Tractability for cgFOC on classes of locally bounded expansion

- $f(\xi, \sigma, \varepsilon)$ -time **query answering** after $f(\xi, \sigma, \varepsilon) \cdot |A|^{1+\varepsilon}$ -time preprocessing
- $f(\xi, \sigma, \varepsilon)$ -delay **enumeration** after $f(\xi, \sigma, \varepsilon) \cdot |A|^{1+\varepsilon}$ -time preprocessing
- $\mathcal{O}(1)$ -delay **agnostic PAC enumeration** after $\mathcal{O}(|A|^{1+\varepsilon})$ -time preprocessing

Hardness

- "2-guarded" FOC model checking is AW[*]-hard on trees of height 2
- "guarded" FOC model checking is AW[*]-hard on a class of shrub-depth 2

Open Questions

- Tractability on nowhere dense classes?
- FOC₁ on dense classes?