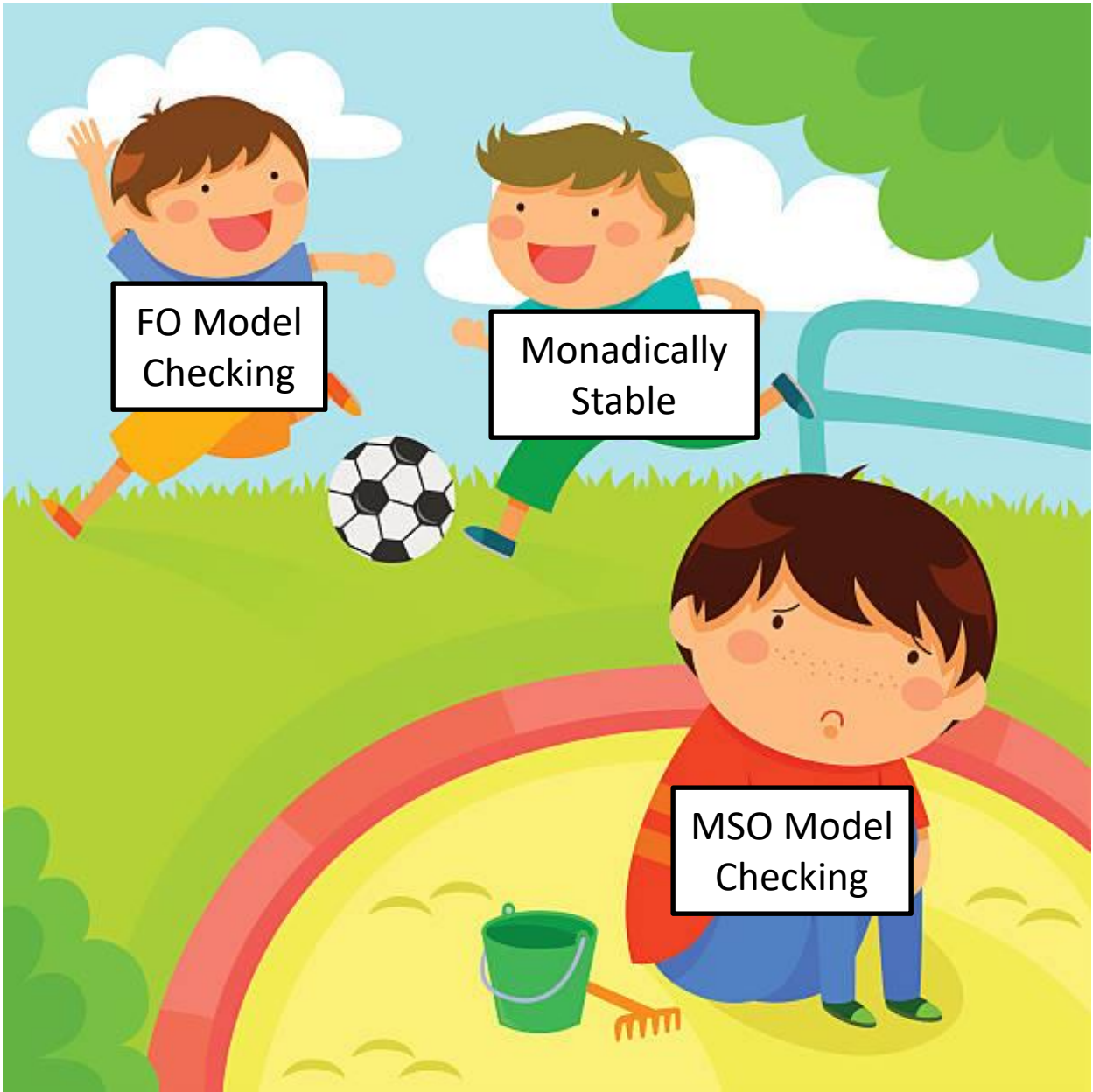


Approximate Evaluation of Quantitative Second Order Queries

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FO Model
Checking

Monadically
Stable

MSO Model
Checking

Motivation

- Many **NP**-hard problems are known to be fixed-parameter tractable parameterized by treewidth
 - **Minimum Vertex Cover, Maximum Independent Set, Hamiltonian Cycle, 3-Coloring...**
 - “Standard” dynamic programming
- But what if we want to show that some new problem is fixed-parameter tractable w.r.t. treewidth?
 - Solve it via dynamic programming, or...
 - use an algorithmic meta-theorem: **Courcelle’s Theorem**

Courcelle's Theorem

Given a LinCMSO_2 query (φ, \mathbf{t}) and a graph \mathbf{G} , we can compute an answer to (φ, \mathbf{t}) on \mathbf{G} in time at most $f(\varphi, \text{tw}(\mathbf{G})) \cdot |V(\mathbf{G})|$
where f is some computable function and $\text{tw}(\mathbf{G})$ is the treewidth of \mathbf{G}

- Why is this useful?
 - Instead of doing dynamic programming, we just need to encode problems as bounded-size “ LinCMSO_2 queries”
 - Can be much, much easier

Courcelle's Theorem

Given a LinCMSO_2 query (φ, \mathbf{t}) and a graph \mathbf{G} , we can compute an answer to (φ, \mathbf{t}) on \mathbf{G} in time at most $f(\varphi, \text{tw}(\mathbf{G})) \cdot |V(\mathbf{G})|$
where f is some computable function and $\text{tw}(\mathbf{G})$ is the treewidth of \mathbf{G}

- \mathbf{G} can be edge- and vertex-colored and weighted
- φ is a formula in CMSO_2 logic with some (optional) free vertex and/or edge set variables
 - Can quantify over single and set vertex/edge variables
 - Can count *modulo some fixed constant*
 - Can use usual logical connectives, query incidence etc.
- \mathbf{t} is some linear function over the free variables in φ
- Answer: assignment which maximizes \mathbf{t} or “**No**”

Courcelle's Theorem for Cliquewidth

Given a LinCMSO_1 query (φ, \mathbf{t}) and a graph \mathbf{G} , we can compute an answer to (φ, \mathbf{t}) on \mathbf{G} in time at most $f(\varphi, \text{cw}(\mathbf{G})) \cdot |V(\mathbf{G})|^2$
where f is some computable function and $\text{cw}(\mathbf{G})$ is the **cliquewidth** of \mathbf{G}

- \mathbf{G} can be ~~edge-and~~ vertex-colored and weighted
- φ is a formula in MSO_1 logic with some (optional) free vertex ~~and/or edge~~ set variables
 - Can quantify over single and set vertex/~~edge~~ variables
 - Can count *modulo some fixed constant*
 - Can use usual logical connectives, query incidence etc.
- \mathbf{t} is some linear function over the free variables in φ
- Answer: assignment which maximizes \mathbf{t} or “**No**”

Our Goal

- Courcelle's Theorem(s) = best meta-theorem(s) for exact fixed-parameter tractability w.r.t. tw and cw
 - Not “tight”: **Vertex Disjoint Paths** par. by treewidth

Our Goal

- Courcelle's Theorem(s) = best meta-theorem(s) for exact fixed-parameter tractability w.r.t. tw and cw

But what about approximate fixed-parameter tractability w.r.t. tw and cw?

- Could capture many problems which are not **FPT**, but are **FPT**-approximable w.r.t. tw and cw ([Lampis 2014](#))

Model Problems

- **Equitable k-Coloring / k-Connected Partition**
 - Partition graph into k equal-size parts
 - **Bounded-Degree**
 - Delete a smallest
 - **Capacitated Vert**
 - Cover/dominate vertices w.r.t. given capacity bounds
 - **Graph Motif**
 - Find a connected subgraph in a vertex-colored graph where each color occurs precisely a given number of times
- What are we approximating?
 - Not only optimization problems

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- What are we approximating?
 - Not only optimization problems
 - “Size constraints”





First Step: Enriching the Logic

- Goal: extend LinCMSO by allowing it to “count”
- But can't LinCMSO already count?
 - It can count *modulo some fixed constant*
 - Useful for, e.g., vertex minor detection. Not here.
- How can we let LinCMSO count?
 - Well, it already counts the optimization target \mathbf{t} :

$$t(X_1 \dots X_\ell) := a + \sum_{1 \leq i, j \leq \ell} a_{ij} w_i(X_j),$$

- a, a_{ij} are (large input-specified) numbers, w_i is sum of weights
- What if we add atoms that can compare these? **CMSO** [≡]

CMSO[\leq] on Model Problems

- **Equitable k-Coloring / k-Connected Partition** 
 - Partition graph into k **equal-size parts**
- **Bounded-Degree Vertex Deletion** 
 - Delete a smallest vertex set to achieve a **given max degree**
- **Capacitated Vertex Cover / Dominating Set** 
 - Cover/dominate vertices w.r.t. given **capacity bounds**
- **Graph Motif** 
 - Find a connected subgraph in a vertex-colored graph where each color occurs precisely **a given number of times**

... But It Is Too Powerful

- **W[1]**-hard to approximate for any constant ratio α :

$$\forall Y_1 \forall Y_2 (|Y_1| > |Y_2| \cdot \alpha^2 \vee |Y_2| > |Y_1| \cdot \alpha^2 \vee \psi(XY_1Y_2))$$

– on vertex-colored paths, parameterized by formula size

- **NP**-hard to find exact answers for:

$$\psi_1(X) \wedge \forall Y_1 \forall Y_2 \forall Y_3 \forall Y_4 ((|Y_1| \neq |Y_2|) \vee (|Y_3| = |Y_4|) \vee \psi_2(XY_3Y_4)),$$


– on vertex-colored bounded-depth trees

Taking a Step Back

$$\underbrace{\varphi(X_1 X_2 \dots) = \forall Y_1 \forall Y_2 \dots}_{\text{often used for comparisons}} \underbrace{\exists Z_1 \forall Z_2 \dots}_{\text{usually not used for comparisons}}$$

$$\underbrace{\varphi(X_1 X_2 \dots)}_{\text{comparisons allowed}} = \underbrace{\forall Y_1 \forall Y_2 \dots \exists Z_1 \forall Z_2 \dots}_{\text{comparisons forbidden}}$$

(first estimate) (first estimate)

- Restricting comparisons only to the first quantifier group would only capture some problems
- Restricting comparisons to the first two quantifier groups is too much
- Solution: only limited comparisons in the 2nd group 

(Blocked) \forall CMSO Formulas

- \forall CMSO_{1/2} logic contains all formulas that are built from *blocks* via \exists, \forall, \wedge
- Block: CMSO[\leq] formula with restricted comparisons:
 - Only non-negative coefficients and negation-normalized
 - At most one weight comparison has a weight term which involves variables in the 1st two quantifier groups
 - All other weight comparisons are exclusively for the first quantifier group

Meta-Theorem for Treewidth

What is this?

Theorem 11 (Approximation of \forall CMSO₂). *Given a \forall CMSO₂-query (φ, \hat{t}) , an accuracy $0 < \varepsilon \leq 0.5$ and a graph G with matching signature, we can compute a $(1 + \varepsilon)$ -approximate answer to (φ, \hat{t}) on G in time at most*

$$\left(\frac{\log(M)}{\varepsilon}\right)^{f_1(|\varphi|, \text{tw}(G))} \cdot |V(G)|^{1.001},$$

useful for exponential numbers encoded in binary

and in particular in time at most

$$\left(\frac{1}{\varepsilon}\right)^{f_2(|\varphi|, \text{tw}(G))} \cdot M^{0.001} \cdot |V(G)|^{1.001}$$

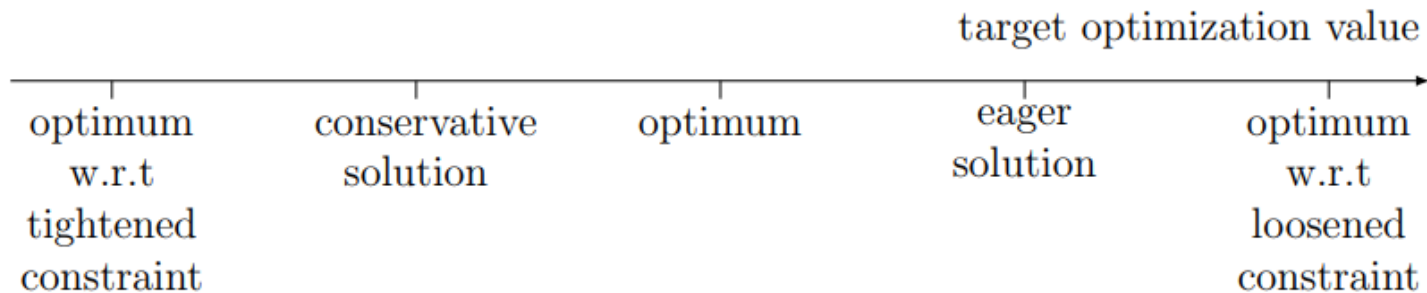
useful for unary-encoded numbers

where f_1, f_2 are computable functions and M is two plus the highest number that occurs in any weight term of φ or as any weight in G .

Approximate Solutions

- Original constraint: $|X| \leq |Y|$
 - Target value (=optimum): 0, $-\infty$, 50, ...
- Loosened constraint: $|X| \leq (1 + \varepsilon) \cdot |Y|$
 - Optimum could increase but not decrease
- Tightened constraint: $(1 + \varepsilon) \cdot |X| \leq |Y|$
 - Optimum could decrease but not increase
- A $(1+\varepsilon)$ -approximate solution contains two solutions:
 - An eager one that is at least as good as the original optimum while satisfying the loosened constraints
 - A conservative one that satisfies the original constraints while being at least as good as the optimum under the tightened constraints

Approximate Solutions



- A $(1+\varepsilon)$ -approximate solution contains two solutions:
 - An eager one that is at least as good as the original optimum while satisfying the loosened constraints
 - A conservative one that satisfies the original constraints while being at least as good as the optimum under the tightened constraints

Meta-Theorem for Cliquewidth

Theorem 10 (Approximation of $\forall\text{CMSO}_1$). *Given a $\forall\text{CMSO}_1$ -query (φ, \hat{t}) , an accuracy $0 < \varepsilon \leq 0.5$ and a graph G with matching signature, we can compute a $(1 + \varepsilon)$ -approximate answer to (φ, \hat{t}) on G in time at most*

$$\left(\frac{\log(M)}{\varepsilon}\right)^{f_1(|\varphi|, \text{cw}(G))} \cdot |V(G)|^2,$$

useful for exponential numbers encoded in binary

and in particular in time at most

$$\left(\frac{1}{\varepsilon}\right)^{f_2(|\varphi|, \text{cw}(G))} \cdot |V(G)|^2 \cdot M^{0.001}$$

useful for unary-encoded numbers

where f_1, f_2 are computable functions and M is two plus the highest number that occurs in any weight term of φ or as any weight in G .

- Notice that we have a lot of freedom in choosing ε
- What if we set ε to be so small that the loosened constraints are integer-equivalent to the original ones?

Exact Meta-Theorem

Theorem 12 (Evaluating Queries Exactly). *Given a \forall CMSO₁-query (or \forall CMSO₂-query) (φ, \hat{t}) and a graph G with matching signature and cliquewidth (or treewidth) at most k , we can compute an answer to (φ, \hat{t}) on G in time*

$$(M + |V(G)|)^{f(|\varphi|, k)}$$

where f is some computable function and M is the highest number that occurs in any weight term of φ or as any weight in G .

- Encoding a problem as a \forall CMSO_{1/2} query yields **XP**-tractability and **FPT**-time computation of “approximate” (eager/conservative) solutions

Some Examples

- **Equitable k-Coloring / k-Connected Partition**

- \forall CMSO₁ formula whose size depends on k
- When parameterizing by treewidth, we can also do:

$$\begin{aligned} \varphi_{\text{ECP-2}}(X) := & \left(\forall Y : (\text{"G[Y] is not a connected component of } G - X") \vee |Y| \geq \left\lfloor \frac{n}{k} \right\rfloor \right) \\ & \wedge \left(\forall Y : (\text{"G[Y] is not a connected component of } G - X") \vee |Y| \leq \left\lceil \frac{n}{k} \right\rceil \right) \\ & \wedge \left(\forall Y : (\text{"Y does not contain precisely one vertex from each"} \right. \\ & \quad \left. \text{connected component of } G - X") \vee |Y| \leq k \right) \\ & \wedge \left(\forall Y : (\text{"Y does not contain precisely one vertex from each"} \right. \\ & \quad \left. \text{connected component of } G - X") \vee |Y| \geq k \right), \end{aligned}$$

where X is an edge set variable, Y is a vertex set variable and $n = |V(G)|$.

Some Examples

- **Equitable k-Coloring / k-Connected Partition**
 - \forall CMSO₁ formula whose size depends on k
- **Bounded-Degree Vertex Deletion**

$$\varphi_{\text{BDVD}}(X) := \forall A : (\neg("X \cap A = \emptyset" \wedge "A \text{ contains an } A\text{-universal vertex}")) \vee |A| \leq p + 1.$$

- **Graph Motif**

$$\begin{aligned} \varphi_{\text{GM}}(X) := & \exists X_1, \dots, X_k : ("X \text{ is connected}") \\ & \wedge ("X_1, \dots, X_k \text{ is the partitioning of } X \text{ into colors } 1, \dots, k, \text{ respectively}") \\ & \wedge \left(\bigwedge_{i \in [k]} |X_i| \leq M(i) \wedge M(i) \leq |X_i| \right). \end{aligned}$$

Some Examples

- **Capacitated Vertex Cover / Capacitated Dom. Set**
 - Graph is preprocessed so that we can capture an “assignment” by the formula
- **Kidney Exchange**
 - Find a maximum-weight collection of vertex-disjoint cycles, each of length at most some given bound p
- **Non-graph problems**
 - Can capture known dynamic programming approximation algorithms for **Subset Sum** (incl. multidimensional variant), **Knapsack**, and other number problems via vertex weights

Last Examples

- **Max-Cut** and **Edge Dominating Set**
 - parameterized by cliquewidth
 - i.e., without edge set quantification
 - requires heavy preprocessing of input graph
 - probably harder to encode than to solve directly
 - only works for exact solutions (**XP**), not approximation
 - ...but still unexpected

Thank you for your attention

Here is a “mindmap” for the proof

