On Finding a Sparse Subgraph in Subclasses of Perfect Graphs

Rémi Watrigant, Marin Bougeret and Rodolphe Giroudeau

LIRMM, Montpellier, France



1ères journées du GT CoA - Complexité et Algorithmes

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 - ► NP-hard in chordal graphs [Corneil and Perl, 1984]
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 - constant approximation algorithm in chordal graphs [Liazi et al., 2008]
- *k*-SS polynomial in:
 - split graphs (obvious)
 - ▶ bounded cliquewidth (⇒ trees, cographs, ...) [Boersma et al., 2012]

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Our results [W., Bougeret, Giroudeau, 2012]:

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- FPT algorithm in interval graphs (parameterized by the cost of the solution)
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Polynomial-Time Approximation Scheme (PTAS)

A *PTAS* for a minimization problem is an algorithm A_{ϵ} such that for any fixed $\epsilon > 0$, A_{ϵ} runs in polynomial time and outputs a solution of cost $< (1 + \epsilon)OPT$

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proper interval graph = no interval contains properly another one = unit interval graphs

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3 Open Problems and Future Work

Idea of the algorithm:

• sort intervals according to their right (or left) endpoints

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- greedy decomposition of the graph into a path of separators/cliques
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- dynamic programming processes the graph through the decomposition, enumerating all possible solutions.

 I_{m_1}









Restructuration of a solution : **compaction** $S \mapsto comp(S)$



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Remark

If for each block, the compaction produces a ρ -approximated solution, then it is a ρ -approximated solution for the **whole** graph.

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Re-structuration of optimal solutions

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Then define the compaction: for any $t \in \{1, ..., P\}$

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• we divide X_L into P consecutive subsets of same size $q_L \to X_1^L, ..., X_P^L$

• we divide X_R into P consecutive subsets of same size $q_R \to \tilde{X}_1^R, ..., \tilde{X}_P^R$ Then define the compaction: for any $t \in \{1, ..., P\}$

- Y_t^L are the q_L rightmost intervals of the t^{th} left block.
- Y_t^R are the q_R leftmost intervals of the t^{th} right block.



Re-structuration of optimal solutions

What do we need to construct such a solution ?



Re-structuration of optimal solutions

What do we need to construct such a solution ?

- the leftmost interval of the t^{th} left block for $t \in \{1, ..., P\}$
- the rightmost interval of the t^{th} right block for $t \in \{1, ..., P\}$
- x_R, x_L (plus remainders of divisions by P...)

 $\Rightarrow 2P + O(1)$ variables ranging in $\{0, ..., n\}$





Sketch of proof of the $(1 + \frac{4}{P})$ approximation ratio:



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•
$$SOL = \binom{x_L}{2} + \binom{x_R}{2} + \sum_{t=1}^{a} \sum_{u=1}^{u} E(Y_t^L, Y_u^U)$$

• $OPT = \binom{x_L}{2} + \binom{x_R}{2} + \sum_{t=1}^{a} \sum_{u=1}^{a} E(X_t^L, X_u^R)$

But:



Sketch of proof of the $(1 + \frac{4}{P})$ approximation ratio:

• $SOL = \begin{pmatrix} x_L \\ 2 \end{pmatrix} + \begin{pmatrix} x_R \\ 2 \end{pmatrix} + \sum_{t=1}^{r_a} \sum_{u=1}^{a} E(Y_t^L, Y_u^R)$ • $OPT = \begin{pmatrix} x_L \\ 2 \end{pmatrix} + \begin{pmatrix} x_R \\ 2 \end{pmatrix} + \sum_{t=1}^{a} \sum_{u=1}^{a} E(X_t^L, X_u^R)$

But:

• if some intervals of Y_t^L overlap some intervals of Y_u^R

Then:

• all intervals of X_{t+1}^{L} overlap all intervals of $\bigcup_{i=1}^{u-1} X_i^{R}$



Sketch of proof of the $(1 + \frac{4}{P})$ approximation ratio:

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But:

• if some intervals of Y_t^L overlap some intervals of Y_u^R

Then:

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Finally, we can prove that $\frac{SOL}{OPT} \leq 1 + \frac{4}{P}$

Conclusion:

Theorem

For any *P*, the previous algorithm outputs a $(1 + \frac{4}{P})$ -approximation for the *k*-Sparsest Subgraph in Proper Interval graphs in $O(n^{O(P)})$

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2 PTAS in Proper Interval Graphs

















Future work/open questions :

k-sparsest subgraph:

- extend FPT and/or approximation results to Chordal graphs
- ▶ NP-h/Poly on Interval, Proper interval ?

• k-densest subgraph:

- (NP-h/Poly on Interval, Proper interval)
- ▶ FPT/W[1]-hardness on Chordal graphs ?



Merci de votre attention !