



Parameterized Complexity of CSPs

summary

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- Variables x_1, x_2, \dots, x_n , domain D , constraints.
- Each constraint has *arity* a , is applied to a tuple $(i_1, \dots, i_a) \in [n]^a$, and has relation $R \subseteq D^a$.
- An assignment $\alpha : [n] \rightarrow D$ satisfies a constraint if $(\alpha(i_j))_{j=1}^a \in R$.
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- **MAX SAT**
 - Reasonable parameters: all of the above, number of satisfied constraints, number of unsatisfied constraints, above/below guarantee parameterizations.

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- Positive results:
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- **Negative results:**
 - ETH \implies no $|D|^{o(k/\log k)}$ algorithm if the constraint graph is a cubic expander.
 - ETH \implies no $|D|^{o(\text{tw}/\log \text{tw})}$ algorithm.
 - Fundament of many fine-grained parameterized complexity results.

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- Large arity examples: CNF SAT, Not-All-Equal SAT.
 - Simple, rather uninteresting reduction to the previous slide.
 - Rather exotic regime of parameters.

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- (Crowston et al. 2012) If clauses have different length, parameterize above first-point lower bound: para-NP-hard.
- (Gutin et al. 2010) More similar results for constraints being linear equations, for permutation CSPs, etc.

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- Even very simple examples contain highly nontrivial flow/cut problems.
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- **ALMOST 2-SAT is FPT via a reduction to VERTEX COVER ABOVE LP.**
 - **and VERTEX COVER ABOVE LP is FPT thanks to very strong structural properties of the optimal LP solutions given by Nemhauser-Trotter.**

- Flow-augmentation completes dichotomy for boolean languages.
 - ℓ -CHAIN SAT is FPT $(x \rightarrow y \rightarrow p \rightarrow q)$.
 - COUPLED CUT is FPT $((x \rightarrow y) \wedge (p \rightarrow q) \wedge (\neg x \vee \neg p))$.
 - BUNDLED CUT with bundles of size 2 is W[1]-hard $((x \rightarrow y) \wedge (p \rightarrow q))$.
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- **Tractability isle in ID_2 and BUNDLED CUT WITH PAIRWISE LINKED DELETABLE EDGES useful for algorithms.**
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- **Potential good target for next dichotomy: Temporal CSPs.**

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- MINONES
 - Contains VERTEX COVER. The same branching for any bound on maximum arity.
 - (Kratsch, Wahlström 2010) Dichotomy for polynomial kernelization.
- MAXONES
 - Contains CLIQUE.
 - (Kratsch, Marx, Wahlström 2010) Dichotomy for polynomial kernelization.