

# Parameterized Complexity of MinCSP

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based on joint work with Eun Jung Kim, Stefan Kratsch, and Magnus Wahlström and many others



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For every finite D and  $\Gamma$ , the corresponding CSP is either NP-complete and polynomial time solvable.



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- Interesting if  $CSP(D, \Gamma)$  is P-time.
- Trivial  $n^{\mathcal{O}(k)}$  algorithm.
- Is it FPT parameterized by *k*?





MINCSP $(D, \Gamma)$ : Can you delete *k* constraints to make the instance satisfiable?

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- $D = \{0, 1\},\$   $\Gamma = \{x \to 0, 1 \to x, (x \to y) \land (y \to z) \land (z \to v)\}.$ 
  - 3-Chain SAT! FPT status was open for some time.







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  - · Minimum weight.
- This version is NP-hard.





Bi-objective (s, t)-Cut

**Input**: digraph  $G, s, t \in V(G), \omega : E(G) \to \mathbb{Z}_+, k, W \in \mathbb{Z}$ . **Question**: is there an (s, t)-cut Z with  $|Z| \le k$  and  $\omega(Z) \le W$ .





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• Figure: k = 3,  $W = 8 \longrightarrow YES$ .





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- Parameterization by *k* only?
  - Undirected: can hammer down with randomized contractions / treewidth reductions.





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- Set

 $M := 1 + \sum_{e \in E(G)} \omega(e),$  $cap(e) := M + \omega(e),$ and ask for cut of capacity  $\leq kM + W.$ 

### Directed flow-augmentation





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- Repeat  $2^{\mathcal{O}(k^4 \log k)}$  times:
  - Invoke flow-augmentation, obtaining *A*.
  - Find min-weight solution of cardinality  $\lambda_{G+A}(s, t)$  in G + A.
- BI-OBJECTIVE (*s*, *t*)-CUT is randomized FPT when parameterized by *k*.

# Solving bi-objective (s, t)-cut



#### Theorem (Kim, Kratsch, P., Wahlström)

There exists a randomized polynomial-time algorithm that, given a digraph  $G, s, t \in V(G)$ , and  $k \in \mathbb{Z}$ , outputs  $A \subseteq V(G) \times V(G)$  so that for every minimal (s, t)-cut Z of  $|Z| \leq k$ , with probability  $2^{-\mathcal{O}(k^4 \log k)}$  the set Z is a minimum (s, t)-cut in G + A.

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- Works in slightly more general setting of *star* (*s*, *t*)-*cut*.
  - (*s*, *t*)-cut whose every arc goes from reachable-from-*s* to non-reachable-from-*s*.



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  - (*s*, *t*)-cut whose every arc goes from reachable-from-*s* to non-reachable-from-*s*.
- Undirected graphs:  $2^{-\mathcal{O}(k \log k)}$  success probability.

# PoV: space of all mincuts



$P_1$		
P2		
$P_3$		
$P_4$		
$P_5$		
$P_6$		
$P_7$		
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- Write such clause for every choice of (i, j, u, v) for  $\exists Q$ .
- Space of all CSP solutions  $\equiv$  space of all *st*-mincuts.
- Flow-augmentation: cover the space of *minimal st*-cuts of size ≤ *k* with small number of such CSP instances.





#### Bundled cut

**Input**: digraph  $G, s, t \in V(G), k \in \mathbb{Z}$ , family  $\mathcal{B}$  of pairwise disjoint subsets of E(G).

**Question**: does there exist a minimal (s, t)-cut  $Z \subseteq \bigcup \mathcal{B}$  with  $|\{B \in \mathcal{B} \mid B \cap Z \neq \emptyset\}| \le k$ .

#### Figure: solution of cost 2.





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All  $B \in \mathcal{B}$  singletons  $\longrightarrow$  minimum (s, t)-cut.





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Every  $B \in \mathcal{B}$  of size at most 2  $\longrightarrow W[1]$ -hard with param. k.





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Every  $B \in \mathcal{B}$  is a path of length  $\leq \ell \longrightarrow \ell$ -CHAIN SAT.





#### *ℓ*-Chain SAT

**Input**: digraph *G*, *s*, *t*  $\in$  *V*(*G*), *k*  $\in$   $\mathbb{Z}$ , family  $\mathcal{B}$  of pairwise disjoint subsets of *E*(*G*), each being a path of length  $\leq \ell$ . **Question**: does there exist a minimal (*s*, *t*)-cut *Z*  $\subseteq \bigcup \mathcal{B}$  with  $|\{B \in \mathcal{B} \mid B \cap Z \neq \emptyset\}| \leq k$ .

Figure:  $\ell = 3$ .





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Case  $\ell = 1$ : MINIMUM (s, t)-CUT





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Case  $\ell = 2$ : still MINIMUM (*s*, *t*)-CUT.





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Case  $\ell = 3$ : starts to be interesting!





#### 3-Chain SAT

**Input**: digraph *G*, *s*, *t*  $\in$  *V*(*G*), *k*  $\in$   $\mathbb{Z}$ , family  $\mathcal{B}$  of pairwise disjoint subsets of *E*(*G*), each being a path of length  $\leq$  3. **Question**: does there exist a minimal (*s*, *t*)-cut *Z*  $\subseteq \bigcup \mathcal{B}$  with  $|\{B \in \mathcal{B} \mid B \cap Z \neq \emptyset\}| \leq k$ .

Focus on case  $\ell = 3$ .





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Flow-augmentation: focus on the case *Z* is a mincut ( $|Z| \le 2k$ ).



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$P_1$		
P2		· · · · · · · · · · · · · · · · · · ·
P3		(
$P_4$		
$P_5$		
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P7		
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- Collapse pair into one flow path.
- The same (P-time) description as the space of all mincuts.



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CHAIN  $\ell$ -SAT is FPT when param. by k and  $\ell$ .

Works also in the weighted setting. (Every  $B \in \mathcal{B}$  has its weight and we do not want to exceed total weight budget.)



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Works for slightly more general problem.

Theorem (Kim, Kratsch, P., Wahlström)

(WEIGHTED) BUNDLED CUT WITH PAIRWISE LINKED DELETABLE EDGES *is FPT when param. by k and maximum bundle size.* 

For every  $B \in \mathcal{B}$  and  $e, f \in B$ , there exists a directed path between an endpoint of *e* and an endpoint of *f* (in one of the directions) that does not use edge of a different bundle.

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 MIN SAT(Γ): For fixed boolean language Γ, given an instance and an integer k, can one delete k constraints to make it satisfiable?



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  - constraints of the form  $(x_0 \rightarrow x_1 \rightarrow \ldots \rightarrow x_\ell)$ ;
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• FPT!



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- Main algorithmic part: Each of these contain a new tractability island.
- Hardness: Marx-Razgon reduction is **the** hardness reduction.

### Tractability Island 1





- Constraints defined as an AND of:
  - implications (can use constants 1 and 0);
  - ORs of negated variables.

such that the implication graph is  $2K_2$ -free.

Then, MIN SAT is FPT when parameterized by *k* and max constraint arity (but only unweighted here; weighted is W[1]-hard).

### Tractability Island 2





- Constraints defined as an AND of 2-clauses (can use constants 1 and 0).
- Assumption: for every constraint, the graph of 2-clauses is  $2K_2$ -free (after deleting 1 and 0).

Then, MIN UNSAT is FPT when parameterized by *k* and max constraint arity (also in the weighted setting).



# MINCSP parameterized complexity classification programme

**Goal**: Provide FPT vs W[1]-hard dichotomy theorems for MINCSP parameterized by the number of unsatisfied constraints for various classes of languages  $\Gamma$ .



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*Note*: requires  $CSP(\Gamma)$  to be P-time.



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- Weighted Skew Multicut

 $(s_1, \ldots, s_\ell, t_1, \ldots, t_\ell \in V(G)$ , delete at most *k* arcs of minimum total weight to break all  $s_i \to t_j$  paths for  $1 \le i \le j \le \ell$ )



- Weighted Directed Feedback Arc Set (delete at most *k* arcs of minimum total weight to get a DAG)
- Weighted Directed Feedback Arc Set reduces to Weighted Skew Multicut with  $\ell \sim k$  by the standard Iterative Compression trick.
- Weighted Skew Multicut

   (s<sub>1</sub>,...,s<sub>ℓ</sub>, t<sub>1</sub>,...,t<sub>ℓ</sub> ∈ V(G), delete at most k arcs of minimum total weight to break all s<sub>i</sub> → t<sub>j</sub> paths for 1 ≤ i ≤ j ≤ ℓ)
- Weighted Skew Multicut unravels to a BUNDLED CUT WITH PAIRWISE LINKED DELETABLE EDGES instance.



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(WEIGHTED) DIRECTED SUBSET FEEDBACK ARC SET *reduces to* (WEIGHTED) BCWPLDE.

DIRECTED SUBSET FEEDBACK ARC SET: the input digraph *G* is equipped with  $R \subseteq E(G)$  and the goal is to delete at most *k* arcs so that no cycle contains an arc of *R*.



• UNDIRECTED MULTICUT: Graph *G*, terminal pairs  $\mathcal{T} \subseteq \binom{V(G)}{2}$ ,  $k \in \mathbb{Z}$ ; delete at most *k* arcs so that every  $st \in \mathcal{T}$  is separated.



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WEIGHTED UNDIRECTED MULTICUT can be solved using the algorithm of Tractability Island 2 as a black-box.





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  - Domain: Q.
  - Constraints have access to <, ≤, =, ≠ (and are FO formulae).</li>







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  - $\Gamma = \{<, \neq, =\}$  is FPT.
  - $\Gamma = \{<, \leq, \neq, =\}$  is W[1]-hard.
    - Graph formulation: SYMMETRIC MULTICUT. Directed graph *G*, unordered pairs of terminals  $\mathcal{T}$ , integer *k*. Delete *k* edges so that for every  $st \in \mathcal{T}$ , *s* and *t* are not in the same strong component.







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- **OPEN**: Full dichotomy.





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  - Highlight 1: there is always an FPT 2-approximation by using DIRECTED SUBSET FEEDBACK ARC SET separately on endpoints.
  - Highlight 2: {"precedes", "starts", "equals"} has an FPT algorithm via reduction to BCWPLDE.



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- (Dabrowski, Jonsson, Ordyniak, Osipov, Wahlström, SODA 2023)
  - FPT if two variables per equation and **K** is an Euclidean domain;
  - W[1]-hard for three variables per equation;
  - W[1]-hard for some commutative rings (e.g.,  $\mathbb{Z}/6\mathbb{Z}$ ).

### 3-terminal Directed Multicut





MULTICUT: graph *G*, terminal pairs  $(s_i, t_i)_{i=1}^{\ell}$ , integer *k*. *Goal*: delete *k* edges so that no  $s_i \rightarrow t_i$  path remains.

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- [Chitnis, Hajiaghayi, Marx, SODA'12]:  $\ell = 2$  is FPT.
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Theorem (Hatzel, Jaffke, Lima, Masařík, P., Sharma, Sorge, SODA'23)

 $\ell = 3$  case is FPT! (Uses twin-width and flow augmentation.)





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## Thanks!