

Solving hard cut problems via flow augmentation

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Warwick, 27 October 2020



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Overview



1. Flow augmentation
2. Application: Min CSP characterization
3. Application: Coupled Min-Cut



- Tractable...
 - MIN (S,T)-CUT
- ...and intractable
 - MULTIWAY CUT – separate at least 3 terminals from each other
 - MULTICUT – fulfil arbitrary set of cut requests (s_i, t_i)
 - BISECTION – find a min-cut with $n/2$ vertices per side
 - ...



- A **parameterized problem** is a problem where every input is given with a parameter k (e.g., solution size)
- A problem is **Fixed-Parameter Tractable (FPT)** parameterized by k if instances (for some function $f(k)$) can be solved in time
$$f(k) \cdot n^{O(1)}$$
- The contrast (intractable) is running times like n^k or 2^n

FPT algorithms for cut problems



- MULTIWAY CUT: $2^{O(k)} \cdot O(m)$ time
 - Marx [06]; Iwata, Oka, Yoshida [14]; Iwata, Yamaguchi [18] – various methods
- MULTICUT: $2^{\text{poly}(k)} \cdot n^{O(1)}$ time
 - Bousquet, Daligault, Thomassé [11/18]; Marx, Razgon [11/14] – various methods
- **Frameworks** for cut problems
 - **Treewidth reduction** [Marx, O'Sullivan, Razgon 13]
 - **Complete graph decomposition** [CLPPS 14/19; CKLPPSW 20+]



1. BI-OBJECTIVE (s,t) -CUT

- Given $G = (V, E)$, vertices s, t , integers k, W
- Find (s, t) -cut $Z \subseteq E$ such that
 1. $|Z| \leq k$
 2. $w(Z) \leq W$ (according to edge weights $w(e)$)

2. COUPLED MIN-CUT (defined later)

Bi-Objective (s,t)-Cut



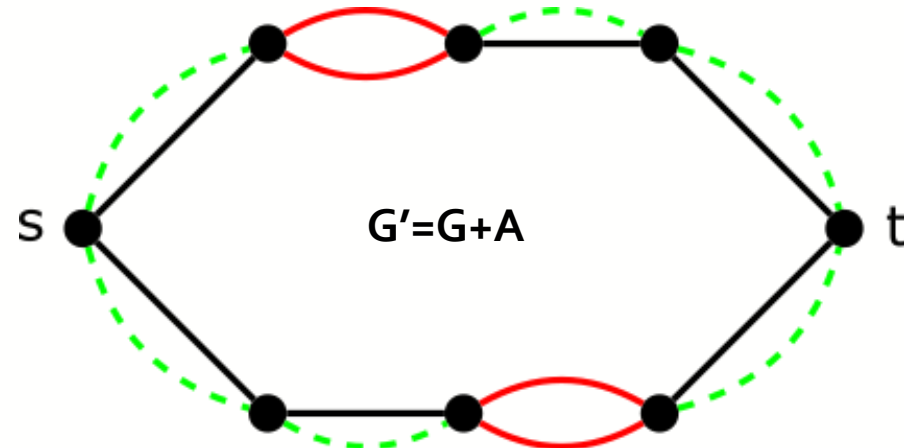
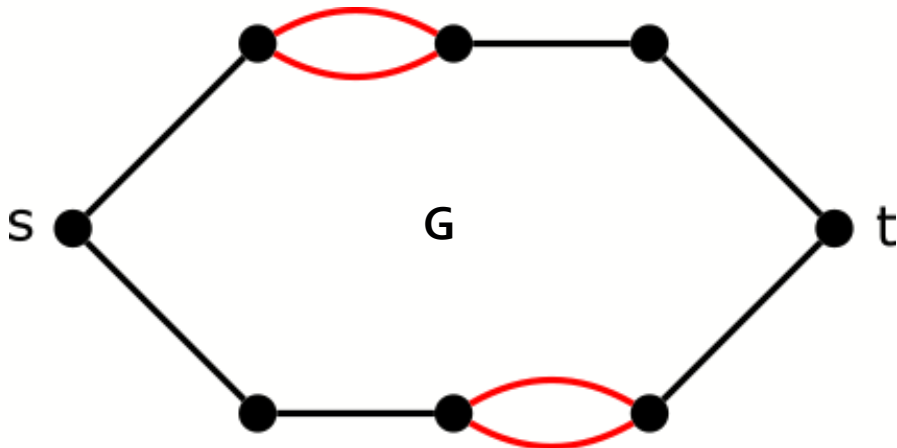
- NP-hard in general [Papadimitriou, Yannakakis 01]
- Tractable if solution is min-cut:
 - Set new edge weights $w'(e) = w(e) + W$
 - Compute (s,t)-min cut with weights w'
 - Cut budget $kW + W$
- Reduction fails if $\lambda_G(s, t) < k$

Flow Augmentation



Given $G = (V, E)$ with **unknown** minimal (s, t) -cut $Z \subseteq E$, and parameter $k=|Z|$, we can compute augmented graph $G' = G + A$ such that

1. Construction of G' takes $k^{O(1)} \cdot O(m)$ time
2. With probability at least one in $2^{O(k \log k)}$, Z is **(s,t)-min cut** in G'



Bi-objective (s,t)-cut



- FPT algorithm for Bi-objective (s,t)-cut: [Time $2^{O(k \log k)} \cdot O(m \log n)$]
 1. Repeat $2^{\Theta(k \log k)}$ times:
 - Compute augmentation $G'=G+A$ with target flow $\lambda_{G+A}(s, t) = k$
 - Solve problem as if Z is min-cut in G'
 2. Return cheapest solution found

This is a toy problem – but this is a competitive running time

Results overview



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Min SAT(Γ)
trichotomy

Flow Augmentation

Coupled Min-Cut
FPT algorithm





Min SAT(Γ) with constraint language Γ :

- **Input:** Formula $F = R_1(X_1) \wedge R_2(X_2) \wedge \dots$ of constraints $R_i \in \Gamma$; integer k
- **Question:** Is there an assignment to F with at most k false constraints $R_i(X_i)$?

Examples:

- **Undirected (s,t)-Cut:** Language $\Gamma = \{0, 1, (x = y)\}$:
 - $F = (s = 1) \wedge (s = v_1) \wedge (v_1 = v_2) \wedge \dots \wedge (t = 0)$
- **Edge Bipartization:** Language $\Gamma = \{(x \neq y)\}$
- **Almost 2-SAT:** Language $\Gamma = \{(x \vee y), (x \vee \bar{y}), (\bar{x} \vee \bar{y})\}$
- ℓ -Chain SAT: $\Gamma = \{0, 1, (x_0 \rightarrow x_1 \rightarrow \dots \rightarrow x_\ell)\}$

Min SAT characterization result



For every Boolean language Γ , one of the following holds:

1. Min SAT(Γ) is **FPT** (using flow augmentation)
2. Min SAT(Γ) is **W[1]-hard**
3. Min SAT(Γ) encompasses **directed cut problems** (such as ℓ -Chain SAT)

Define two relations:

- $R_4(a, b, c, d) \equiv (a = b) \wedge (c = d) \wedge (\bar{a} \vee \bar{c})$
- $R_{hard}(a, b, c, d) \equiv (a = b) \wedge (c = d)$

Then:

- **Min SAT(Γ)** with $\Gamma = \{0, 1, R_4\}$ defines **Coupled Min-Cut** and is **FPT**
- **Min SAT(Γ)** with $\Gamma = \{0, 1, R_{hard}\}$ is **W[1]-hard** (*double equality*)



Min SAT(Γ) falls into one of the following cases (up to duality):

1. Reduces to relation $(a = 0) \wedge (b = 1) \wedge (c \neq d)$ [EDGE BIPARTIZATION]
2. Reduces to $(a = 1) \wedge (c = d)$ and $(\bar{x}_1 \vee \dots \vee \bar{x}_d)$ [GRAPH PAIR CUT, etc.]
3. **Generalized Coupled Min-Cut (GCMC)**
4. W[1]-hard
5. Implements directed graph cut problems

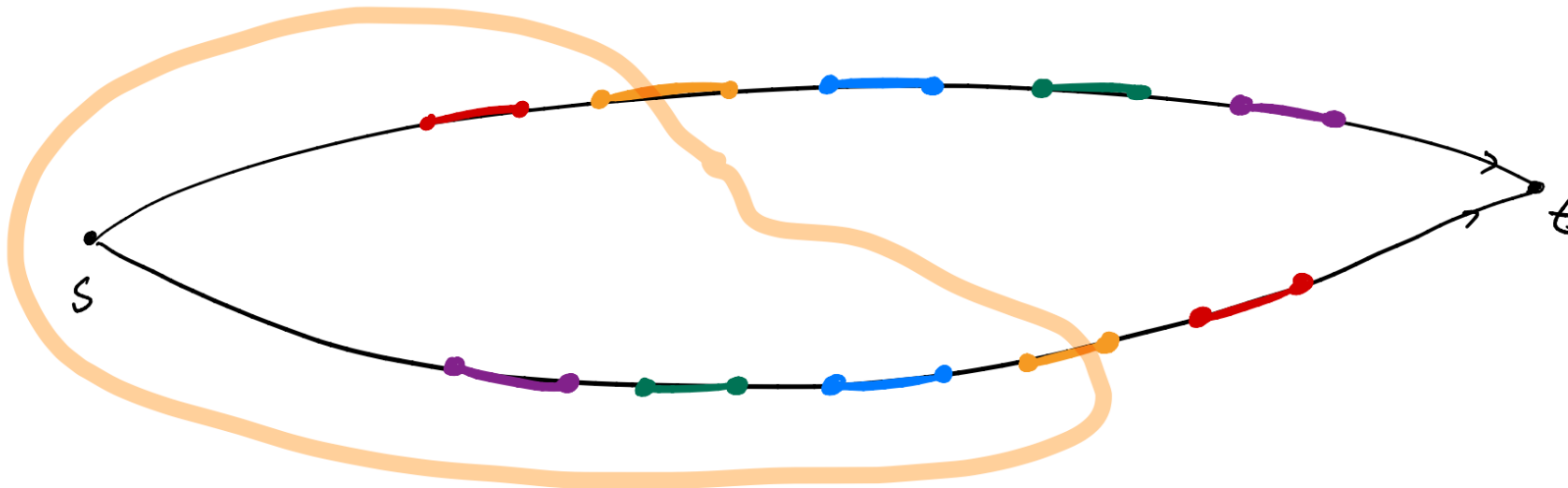
Cases 1-3 FPT, cases 1-2 use standard methods.

Coupled Min-Cut



Graph formulation:

- Graph $G = (V, E)$, vertices s, t , budget bound k
- Looking for (s, t) -cut Z of cost at most k
- Edges of E are grouped into **pairs** (e, f) where
 1. Edges e, f can be cut simultaneously at cost 1, but
 2. if edges e, f **not** cut then at most one edge is on the s -side of the cut



Coupled Min-Cut properties



- NP-hard in general
- Tractable (FPT) if Z is min-cut
- $\text{EDGE MULTICUT} \leq_{FPT} \text{COUPLED MIN-CUT}$
- Impervious (?) to previous methods
 - Important separators, shadow removal, decomposition methods don't apply
 - LP duality???

Flow Augmentation overview



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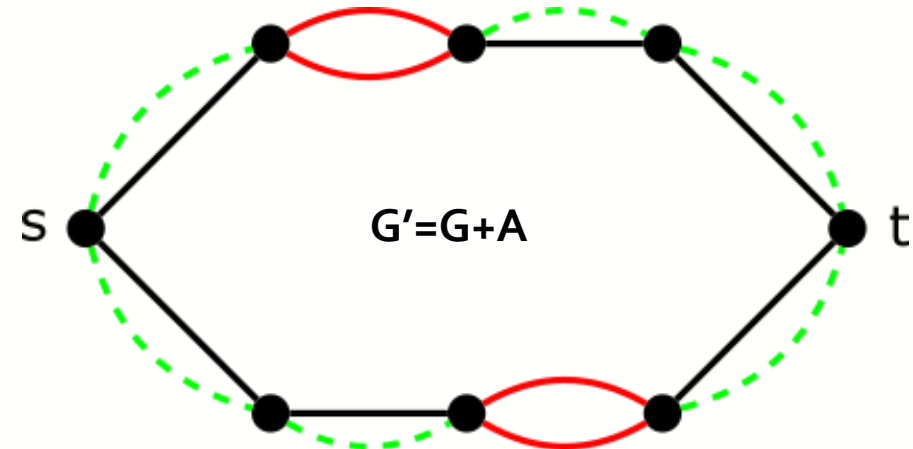
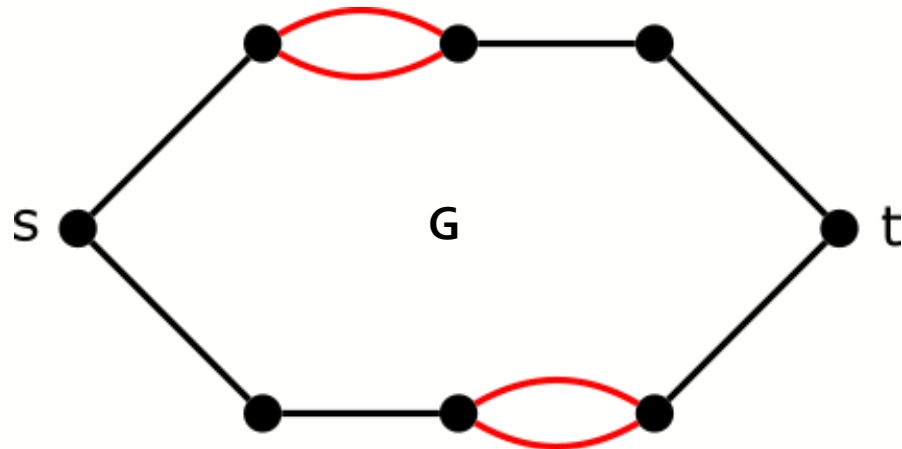
Flow Augmentation



- Graph $G = (V, E)$, vertices s, t , target flow k
- Unknown minimal (s, t) -cut Z , $|Z| = k$

Task:

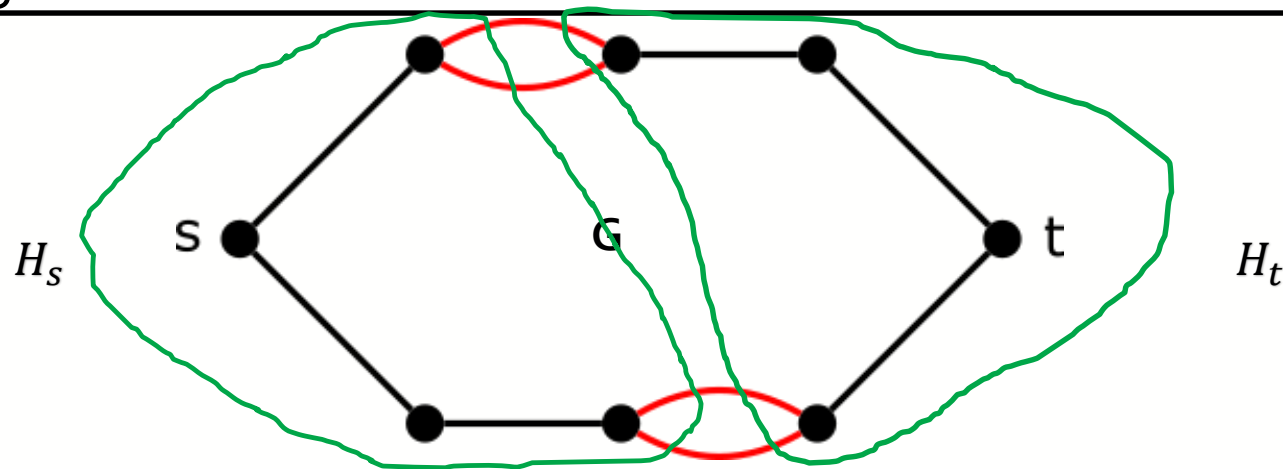
- Add edges A to G so that Z is (s, t) -min cut in $G + A$



Flow Augmentation observations



1. $G - Z$ partitions into two components H_s, H_t
2. Adding (u, v) to A **forbidden** if and only if $u \in H_s, v \in H_t$
3. Suffices to:
 - a. Add edges to A increasing (s, t) -max-flow, such that
 - b. No such edge is **forbidden**





- Trace “all” min-cuts from s to t
- Guess how Z interacts with each one
- Add augmenting edges consistent with the guess

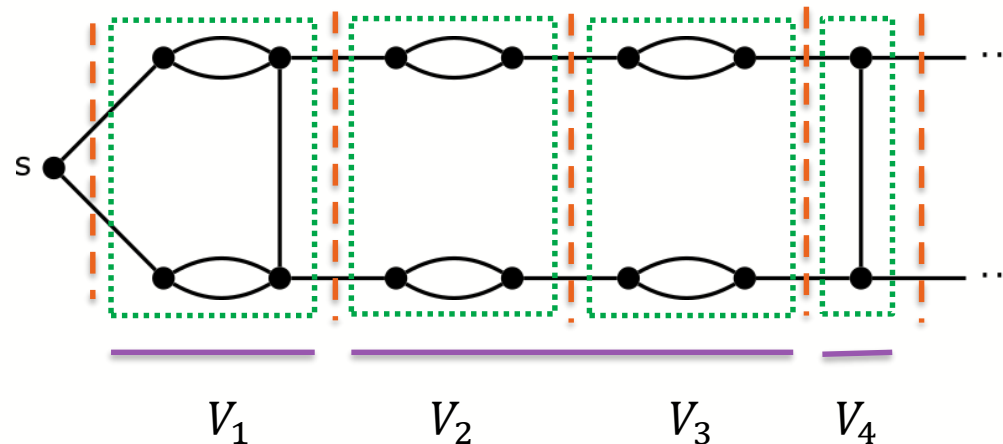
Reasons it could work:

- If a cut C is entirely in H_s (or H_t), there are no bad guesses
- If we can control “mixed” cuts C intersecting both components, there may be only $f(k)$ places where we need to guess correctly

Sequences of min-cuts



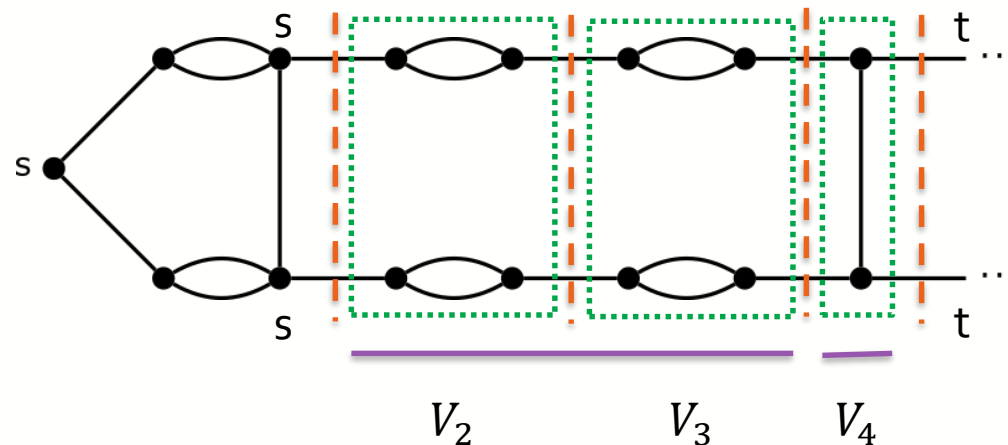
- Sequence of closest disjoint min-cuts C_1, C_2, \dots
- **Blocks** V_i : Sets of vertices between C_i 's
- **Bundles**: Either
 1. Single connected block, or
 2. Sequence of blocks, just short of inducing a connected subgraph



Flow augmentation: Recursion



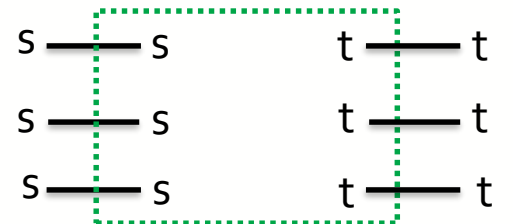
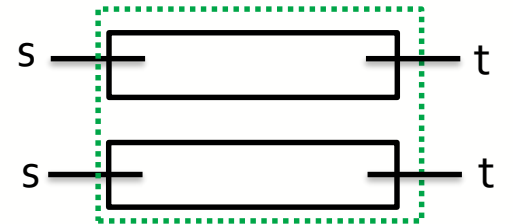
- **Claim:** At most $O(k)$ bundles are affected by Z
 - For any other bundle W , all vertices of $N[W]$ are in one component
- A maximal sequence of affected bundles defines a **new flow augmentation instance** (we “zoom in” on part of the graph)
- **Recursion state** (λ, k) : Progress = (increase λ) or (decrease k)



Single affected bundle



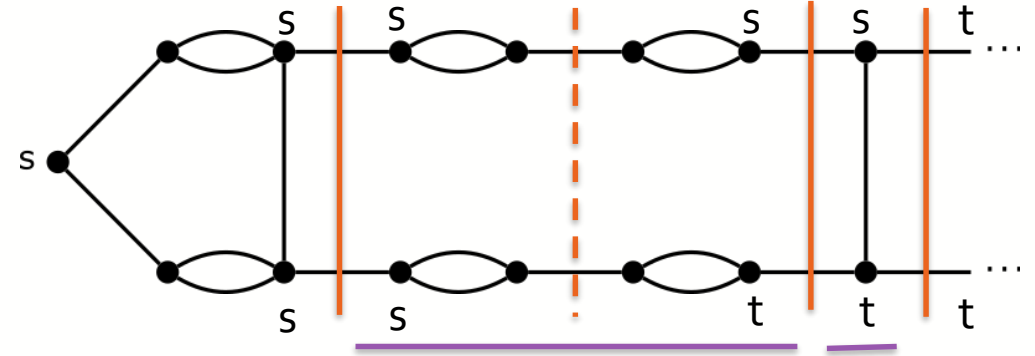
- **Disconnected** bundle:
 - Recurse independently into each component
 - Each call decreases k
- **Connected** bundle (single block):
 1. **Surrounding edge cut** – decreases k
 2. **No surrounding edge cut** – increases λ



Multiple affected bundles



- Select one cut C_i between bundles
- Guess assignment (s/t) for all vertices ($O(\lambda)$ many)
 - After initial cut
 - Surrounding the cut C_i
 - Before the final cut
- Recurse into both halves
- Progress:
 - Because all bundles were affected, cut budget k must be split two ways





- **Flow Augmentation** – solve hard cut problems by reducing them to **min-cut/max-flow** instances
- Applications shown:
 - Weighted cut problems
 - Coupled Min-Cut / **Min SAT(Γ)** trichotomy
- Is there **directed flow augmentation**?