

# Flow decompositions and directed graph minors

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29 August, 2024

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# Flow graphs

Given is an  $s$ - $t$  (multigraph) DAG  $G = (V; E)$  and a flow  $f : E \rightarrow \mathbb{N}$ .

Conservation of flow:  $\sum_{e \in \text{in}(v)} f(e) = \sum_{e \in \text{out}(v)} f(e) \quad \forall v \in V \setminus \{s, t\}$ .

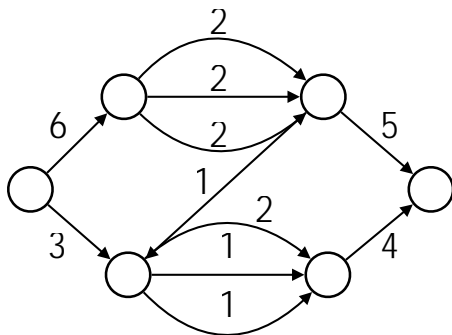


Figure: Simple flow graph.

# Minimum Flow Decomposition (MFD)

Minimum Flow Decomposition (MFD) of  $(G; f)$ : **minimum sized** set of  $s$ - $t$  **paths and weights**  $f(P_1; w_1); \dots; (P_k; w_k)g$  ( $w_i \in \mathbb{N}$ ) with

$$f = \sum_{i=1}^k w_i P_i$$

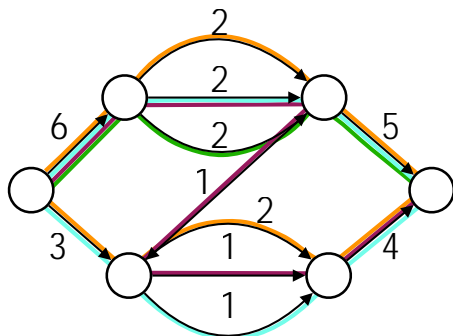


Figure: Flow decomposition in 6  $s$ - $t$  paths (weights omitted).

# Solving MFD, state of the art

- MFD is NP-hard, even if the flow values come from  $\{1; 2; 4\}$  [Hartman et al., 2012]
- MFD is APX-hard [Hartman et al., 2012]
- FPT in time  $O(2^{k^2} (n + \log k f k))$  [Kloster et al., 2018] and ILP [Dias et al., 2022] solvers exist
- Greedy approximation factor  $\Omega(m = \log m)$  [Cáceres et al., 2024]
- Greedy-weight commonly used in applications [Baaijens et al., 2020, Tomescu et al., 2013]

- Explore DAG structure with (minimal) flows
- $O(\log kf k)$ -approximation for graphs with forbidden minors
- Implication: Quasi-polynomial algorithm for some class of instances
- Bridges gap between application and theory

## Definition

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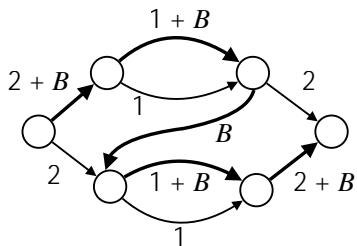
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We have  $\text{width}(G) = \text{mfd}(G; f)$  for all  $f > 0$ .

# Width hinders greedy

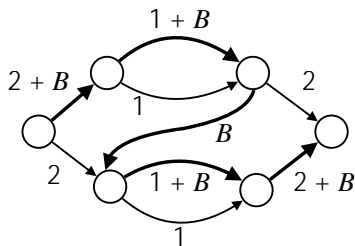


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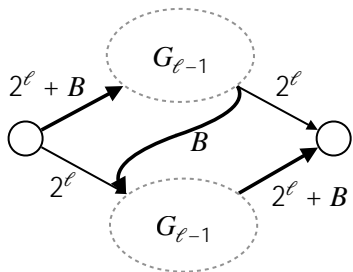


(a) The base case  $(G_1; X_{1;B})$ . Bold edges carry flow at least  $B$ .

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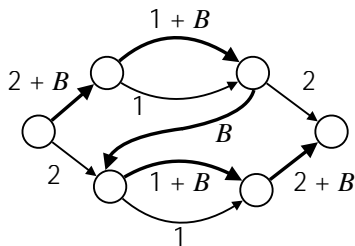


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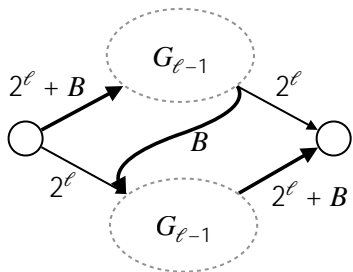


(b) Building  $(G_\ell; X_{\ell;B})$  from two copies of  $(G_{\ell-1}; X_{\ell-1;B})$  ( $\ell > 1$ ).

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$(G_\ell; X_{\ell;B})$  can be decomposed into  $\Theta(\ell)$  paths. Greedy-weight uses  $\Theta(2^\ell)$  paths.

! Approximation ratio for greedy-weight on MFD is  $\Omega(m = \log m)$ .

## Definition

$H$  is a *butterfly contraction* of  $H^0$  if  $H$  is obtained from contracting edges  $(u; v)$  in  $H^0$  where  $\deg^+(u) = 1$  or  $\deg^-(v) = 1$ .

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## Definition [Grigorjew et al., 2024]

We call  $H$  a *flow-minor* of an  $s$ - $t$  DAG  $G$ , if there exists  $H^0$  such that

- $E(H^0) = \text{supp}(f)$  for some flow  $f : E(G) \rightarrow \mathbb{N}$  (write  $H^0 = H_f$ ),
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From [Deligkas and Meir, 2017]: Delete edges  $(u; v)$ , where  $\deg^-(u) \geq 2$  and  $\deg^+(v) \geq 2$ , contract via butterfly contractions.

## Definition [Deligkas and Meir, 2017]

The *parallel width*  $pw$  of an  $s$ - $t$  DAG is the **largest minimal  $s$ - $t$  cut-set**.

- Equivalently, the maximum width( $G_f$ ) throughout all  $f \geq 0$ .
- Equivalently, the maximum possible value of a minimal flow.

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Let  $GP(c)$  be the  $s$ - $t$  DAG of  $c$  parallel edges  $(s; t)$ .

## Lemma [Deligkas and Meir, 2017]

$\text{pw}(G) < c$  iff  $G$  is  $GP(c)$ -f-minor free.



# $Ch_k$ DAG

We call the following DAGs  $Ch_k$ :

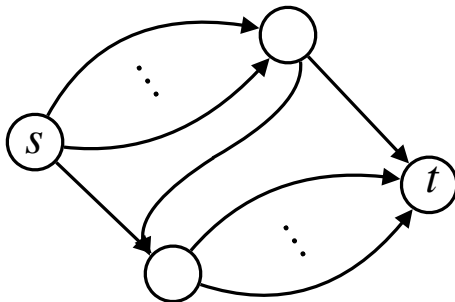


Figure: On each side,  $k$  parallel edges.

Lemma [Deligkas and Meir, 2017]

A DAG is series-parallel if and only if it is  $Ch_1$ -f-minor free.

## Definition ([Cáceres et al., 2024])

$G$  is called *width-stable* if  $\text{width}(G_f) \leq \text{width}(G_g)$  for all flows  $f \leq g$  on  $G$ .

The following are equivalent [Cáceres et al., 2024]:

- $G$  is width-stable,
- $G$  is  $Ch_2$ -f-minor free.

! Greedy is a  $O(\log \text{Val}(f))$ -approximation on width-stable graphs [Cáceres et al., 2024].

# Approximating MFD

## Power of two–approach

- 1 "Remove" the odd part: Given flow  $f$ , define flow  $g$  of the **same parity** as  $f$ ,
- 2 Choose a small  $g$  and decompose it to  $\text{Val}(g)$  paths,
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! We obtain a decomposition of size  $\text{Val}(g_1) + \text{Val}(g_2) + \dots + \text{Val}(g_{\log_2 k})$ .

# How to choose $g$ ?

How to choose a small, parity fixing flow  $g$ ?

$$\begin{aligned} & \min \text{Val}(g); \text{s.t.} \\ & g \text{ is a flow on } G; \\ & 0 \leq g \leq f; \\ & f - g \geq 0; \end{aligned} \tag{1}$$

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$$\begin{aligned} & \min \text{Val}(g); \text{s.t.} \\ & g \text{ is a flow on } G; \\ & 0 \leq g \leq f; \\ & f - g \geq 2 \cdot 0: \end{aligned} \tag{1}$$

Easier problem:

$$\begin{aligned} & \min \text{Val}(h); \text{s.t.} \\ & h \text{ is a flow on } G; \\ & 0 \leq h \leq f; \\ & 0 < h(e) \leq f(e) \quad \forall e \in E(G): f(e) \text{ is odd:} \end{aligned} \tag{2}$$

# How to choose $g$ ?

Lemma [Cáceres et al., 2024]

For every flow  $f : E \rightarrow \mathbb{N}$  we can find a flow  $\text{Unitary}(f) : E \rightarrow [0; 1g]$  such that  $\|f - \text{Unitary}(f)\|_2$ , in time  $O(m)$ .



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## Lemma [Cáceres et al., 2024]

For every flow  $f : E \rightarrow \mathbb{N}$  we can find a flow  $\text{Unitary}(f) : E \rightarrow \mathbb{N}$  such that  $\|f - \text{Unitary}(f)\|_2 \leq 1$ , in time  $O(m)$ .

## Lemma [Grigorjew et al., 2024]

$h$  is an optimal solution to (2)

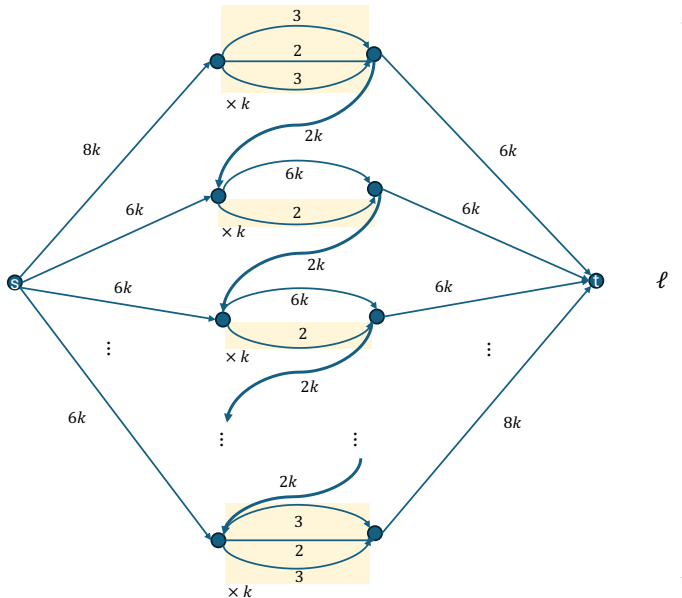
(1)  $g = h + \text{Unitary}(f + h)$  is an optimal solution to (1).

# How large can the $g$ get?

Lemma ([Grigorjew et al., 2024])

*For all  $c > 1$ , there are instances with  $\text{pw}(G) = \text{mfd}(G; f) > c$ .*

# How large can the $g$ get?



## Theorem [Grigorjew et al., 2024]

For all  $c > 1$ , MFD can be approximated with a factor of  $O(\log kf k)$  in runtime  $O(m \log jff jj (mfd(G; f) + n))$  on  $Ch_2$ -f-minor free DAGs and on  $GP(c)$ -f-minor free DAGs.

# $O(\log kf k)$ -approximation

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## Theorem

For all  $c > 1$ , MFD can be approximated with a factor of  $O(\log kf k)$  on instances with  $pw(G) = mfd(G; f) < c$ .

## Corollary [Grigorjew et al., 2024]

MFD can be solved in quasi-polynomial time on  $GP(c)$ -f-minor free DAGs, when the flow is coded in unary.

Proof:

- Approx. algorithm:  $\text{mfd}(G; f) \leq \text{pw}(G) \log kfk$ .
- MFD is in FPT [Kloster et al., 2018]:  
 $O(2^{k^2} (n + \log kfk)) = O(kfk^{\log kfk} (n + \log kfk))$ .

# Further problems

- Good approximation for MFD, when  $\text{pw}(G) = \text{mfd}(G; f) > c$ .
- Is MFD in APX? What about  $GP(c)$ -f-minor free graphs?
- PTAS on  $GP(c)$ -f-minor free graphs?
- Parameterized algorithms for (generalized) max flow or other flow problems?
- Structural graph theorems using flows?

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**Thank you!**





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