

Fast Algorithms for Pseudoarboricity

Meeting on Algorithm Engineering and Experiments – ALENEX 2016



JOHANNES GUTENBERG
UNIVERSITÄT MAINZ

Markus Blumenstock

Institute of Computer Science, University of Mainz

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A tree...



...an unrooted tree...

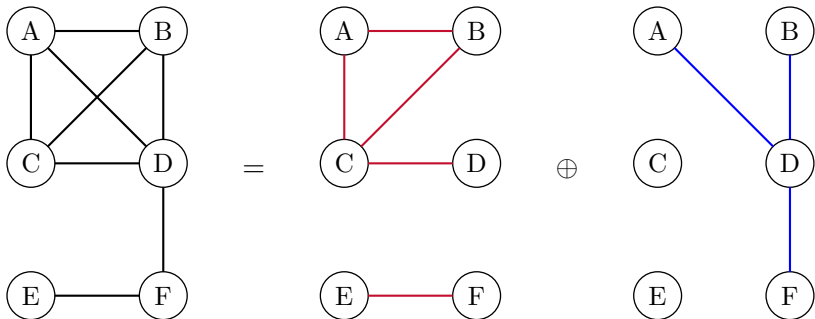


...a pseudotree!



Definition

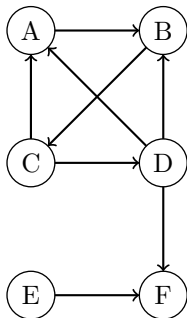
The pseudoarboricity $\rho(G)$ of an undirected graph G is the minimum number of pseudoforests into which the graph can be decomposed.



Pseudoarboricity and Orientations

Theorem (Frank–Gyárfás 1976 + Picard–Queyranne 1982)

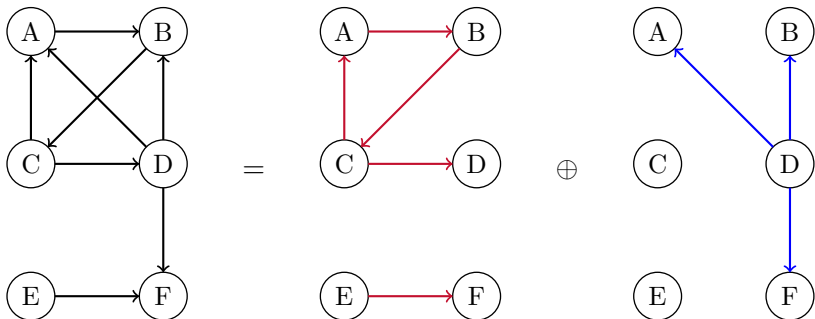
Let \vec{G} be an orientation of G such that the maximum indegree is minimal. Then this indegree equals the pseudoarboricity $p(G)$.



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- The test can be performed with a maximum flow algorithm, which 're-orient's' an arbitrary orientation to a d -orientation by reversing directed paths, if possible
- With Dinitz's algorithm, we need $\mathcal{O}(|E|^{3/2} \log |I|)$ time on a search interval I that contains p

- The algorithm can be turned into an approximation scheme which returns d satisfying $p \leq d \leq \lceil (1 + \epsilon)p \rceil$ in time

$$\mathcal{O}\left(|E| \frac{\log |V|}{\epsilon} \log |I|\right)$$

for $\epsilon > 0$ on a search interval I containing p

- The approximation is not achieved by stopping the binary search early, but by stopping Dinitz's algorithm after $2 + \log_{1+\epsilon} |V|$ phases (Kowalik 2006)

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- We will use the fact that always $p \in \mathcal{O}(\sqrt{|E|})$ holds

A Selection of New Results

Bound on p	Runtime bound to compute p
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Comparison: matroid partitioning algorithm (Westermann 1988):

$$\mathcal{O}(|E|^{3/2} \sqrt{\log p})$$

How to obtain $\mathcal{O}(|E|^{3/2} \log \log |V|)$ unconditionally

- Use a $(1 + \epsilon)$ -approximation to accelerate the exact algorithm!
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- A binary search in an exact algorithm needs $\mathcal{O}(\log(\epsilon p))$ tests on the narrowed search interval – the runtime is

$$\mathcal{O}(|E|^{3/2} \log \log^2 |V|).$$

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- Run the approximation scheme $i = 1, \dots, k$ times:

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- The initial interval size is $|I_0| \in \mathcal{O}(p)$ with the 2-approximation.
- In every phase $i = 1, \dots, k$, we reduce the interval to size

$$|I_i| \in \mathcal{O}(\log^{\times i} p).$$

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- The total runtime is $\mathcal{O}(|E|^{3/2} \log^* p)$

What have we shown?

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–	$\mathcal{O}\left(E ^{3/2} \sqrt{\log \log p}\right)$	(✓)
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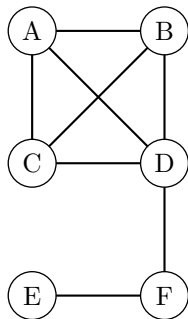
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The Densest Subgraph Problem

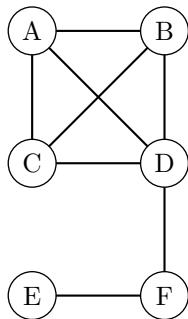


Definition

Let $G = (V, E)$ be an undirected graph. The maximum density is defined as

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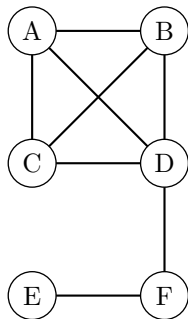
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Theorem (Picard–Queyranne 1982)

For any undirected graph G , we have $\lceil d^(G) \rceil = p$.*

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- For the resulting graph $G' = (V', E')$, we have $d^*(G') = d^*(G)$ and thus $p(G') = p(G)$ by the Picard–Queyranne theorem
- G' is possibly smaller (great!) and has a higher average density, i.e.

$$\frac{|E'|}{|V'|} \geq \frac{|E|}{|V|}.$$

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If $d^(G) \in \Omega(\sqrt{|E|})$, then we can obtain a subgraph $G' = (V', E')$ with $|E'| \in \Theta(|V'|^2)$ and $d^*(G') = d^*(G)$ in linear time.*

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Corollary

If $d^(G) \in \Omega(\sqrt{|E|})$, p can be determined in $\mathcal{O}(|E|^{3/2})$.*

The Re-orientation Algorithm in Practice (Dinitz's algorithm)

Graph	ρ	Without preprocessing		With preprocessing	
		$ E $	Runtime [s]	$ E' $	Runtime [s]
<i>Amazon</i>	5	900K	23	800K	1
<i>DBLP</i>	57	1M	4	14K	0
<i>YouTube</i>	46	3M	4	417K	0
<i>LiveJournal</i>	194	35M	251	540K	0
<i>Orkut</i>	228	117M	1012	13M	33

Note: Push-relabel algorithms are slower by an order of magnitude

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- A 2-approximation of p (and d^*) can be found in $\mathcal{O}(|E|)$ time with a greedy algorithm. Is a factor smaller than 2 possible in linear time?
(Any constant-factor approximation can be found in $\mathcal{O}(|E| \log |V| \log p)$ time)

Thank you for your attention!