Confronting Hardness Using a Hybrid Approach

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SODA 2006

Joint work with Ryan Williams and Maverick Woo

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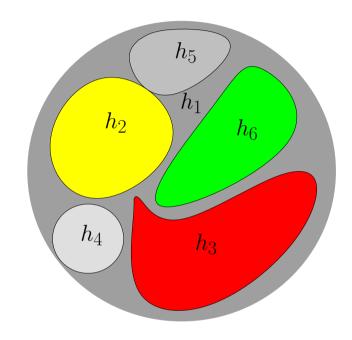
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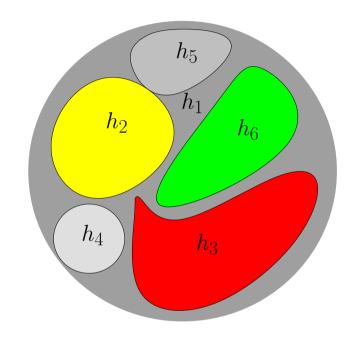
Space Approximation Ratio and Time...

Consider a set $H = \{h_1, \ldots, h_k\}$ of heuristics, good w.r.t. different complexity measures, partitioning the instance space.

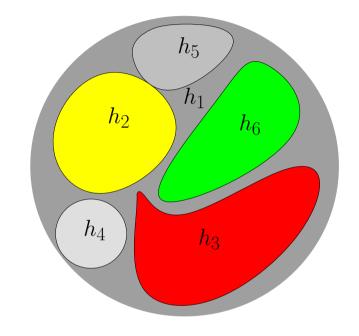


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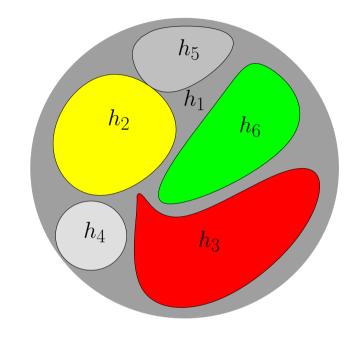
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 h_1 approximates the optimal solution within a factor of α and runs in polynomial time, on all dark gray instances.

 h_2 solves the problem exactly but runs in subexponential time ($2^{o(n)}$) on all yellow instances.

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A selector S which on each instance selects a heuristic in polynomial time.

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There exist hybrid algorithms for NP-Hard problems which for each h_i (on the instances on which S chooses to run h_i) do strictly better than the corresponding known hardness guarantees m_i .

MAX-CUT

Problem: Given a graph G, find a cut which maximizes the number of edges crossing it.

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No better than $(1/2 + \delta)$ -approximation is known which runs in less than quadratic time.

Hybrid Algorithm for Max-Cut

There's a simple *hybrid* algorithm which for any $\epsilon>0$, after a linear time test produces

- \bullet either a maximum cut in $\tilde{O}(2^{\epsilon m})$ time, or
- a $(\frac{1}{2} + \frac{\epsilon}{4})$ -approximation in linear time.

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If
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If
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for each edge in M, with probability 1/2 choose which of its endpoints to put in A. Put the other endpoint in B;

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for each vertex v not in M, with probability 1/2 choose whether to place it in A or B.

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the expected size of the cut is at least

$$(\varepsilon \frac{m}{2}) + \frac{1}{2}(m - \varepsilon \frac{m}{2}) = (\frac{1}{2} + \frac{\varepsilon}{4})m.$$

We get a *linear time* $(\frac{1}{2} + \frac{\varepsilon}{4})$ -approximation.

Karger, Motwani and Ramkumar, 1993: Longest Path is hard to approximate within $2^{O(\frac{\log n}{\log \log n})}$, unless $\text{NP} \subseteq \bigcap_{\delta>0} \text{DTIME}(2^{O(n^{\delta})})$.

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Note for $\ell=n/polylog(n)$ we get subexponential exact running time and a polylog approximation.

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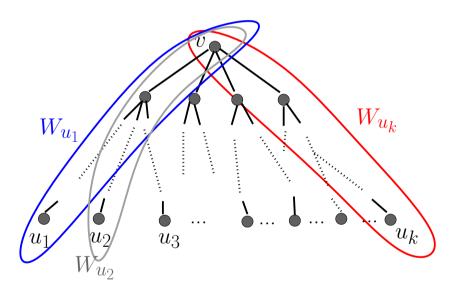
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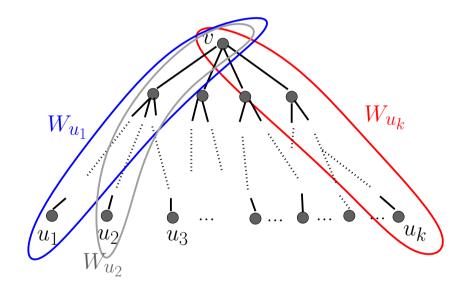
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- 3. Else, we have a DFS tree T of low depth. We can form a path decomposition $(P, \{W_{u_i}\})$ of width at most ℓ : For every leaf u let W_u contain u and its ancestors in T.



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$$P = \{(u_1, u_2), \dots, (u_{k-1}, u_k)\}$$

where u_1, u_2, \dots, u_k are the leaf
nodes in an inorder traversal of T .

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- 3. Otherwise the algorithm returns a path decomposition P of width at most ℓ .

Run an algorithm for Longest Path on graphs of bounded treewidth (based on dynamic programming) by Bodlaender, 1993 to get the longest path in $2^{O(\ell \log \ell)} n^{O(1)}$.

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Best Exact Algorithm: $\tilde{O}(10^n)$ by Feige and Killian, 2000.

For any unbounded constructible $\gamma(n)$, MINIMUM BANDWIDTH admits a hybrid algorithm which produces either

- ullet a linear arrangement achieving the minimum bandwidth in $4^{n+o(n)}$ time, or
- an $O(\gamma(n) \log^2(n) \log \log n)$ -approximation in polynomial time.

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One factor: a low diameter subgraph.

Simple Fact. If G contains a subgraph H of diameter d, then the bandwidth of G is at least (|H|-1)/d.

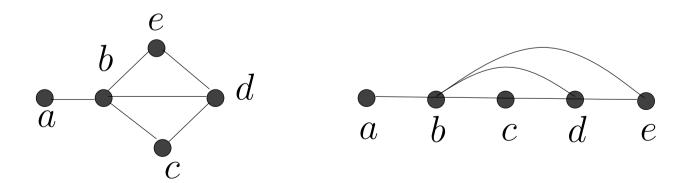
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Intuitively, the absence of a large subgraph with low diameter means that the graph does not expand by much, so it has a smallish node bisection.

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Overarching Idea:

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Two interesting problems arise in designing a hybrid algorithm for some Π

- ullet How to split the cases of Π ?
- How to select the right heuristic?

Thank You!