Exact Combinatorial Algorithms for Graph Bisection

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DIMACS Challenge

Minimum Bisection

- Input: undirected, unweighted graph G = (V, E)
- Output: partition V into sets A and B such that
 - $|A|, |B| \leq \lceil |V|/2 \rceil$
 - ② the number of edges between A and B is minimized
- Variants: ϵ -unbalanced, weights, more parts...



Example Instance

-----[3524 nodes, 5560, solution 30]

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Motivation

Applications:

- load balancing for parallel computing
- preprocessing step for some road network algorithms
- divide-and-conquer (e.g. VLSI design)

Solutions:

- NP-hard, O(log n) best approximation [Räcke '08].
- Heuristics:
 - numerous fast and good partitioners
 - often tailored to specific graph classes (e.g. road networks)
 - no approximation/optimality guarantees

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We want exact algorithms!

- Branch-and-bound: implicit enumeration.
- Subproblem = partial assignment (A, B), with $A, B \subseteq V$
 - ▶ represents bisections (A^+, B^+) with $A \subseteq A^+$ and $B \subseteq B^+$



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 - L: lower bound on all bisections consistent with (A, B)
 - U: best known bisection (updated on-line)
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Crucial ingredient: computing lower bounds.

Lower Bounds

Known bounds:

- linear programming [FMdSWW98, Sen01]
 - hundreds of nodes
- quadratic programming [HPZ11]
 - up to 3000 nodes
- semidefinite programming [AFHM08, Arm07]
 - up to 6000 nodes
- multicommodity flows [SST03]
 - hundreds of nodes
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We want to do better.

Summary

Our result:

- exact combinatorial algorithm for graph bisection
- works well for graphs with a small minimum bisection
 - ▶ road networks, VLSI instances, meshes...
- solves much larger instances than previous approaches

Main contributions:

- new lower bounds
- branching rules
- novel decomposition technique

• Goal: lower-bound all bisections consistent with (A, B).



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- Known bound: min-cut (max-flow) between A and B.
 - pros: simple, upper bound when balanced;
 - cons: weak if $|A| \ll |B|$, typically very unbalanced.



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We must reason about the worst possible extension A^+ .

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- "adversary" picks entire cells, from biggest to smallest
- \Rightarrow partition should have balanced cells (greedy + local search)

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Performance

- Proving 30 is a lower bound:
 - search tree: 1.3M nodes
 - time: 50 minutes



Branching Rule and Forced Assignments

- Branch on vertices likely to increase L the most:
 - If ar from A and B (to produce better cells)
 - e well connected to other vertices (to increase flow)
 - Ontained in large packing cells
- Forced assignments:
 - ▶ use logical implications to fix some vertices to A or B
 - works if upper and lower bounds are close
 - discards many potential branching nodes

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Branching Rule and Forced Assignments

- Proving 30 is a lower bound:
 - search tree: 1.3M nodes \rightarrow 90K nodes
 - ▶ time: 50 minutes → 3.5 minutes



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- Solution: decomposition!



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• Some *E_i* will cross the optimum bisection; at least one will not!



- Given an upper bound U:
 - split edges into U + 1 disjoint sets $E_1, E_2, \ldots, E_{U+1}$
 - **2** for every *i*, contract E_i and run branch-and-bound
 - return best solution found
- Some E_i will cross the optimum bisection; at least one will not!
- *E_i* should be a set of clumps (high degree, well spread).



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• Proving 30 is a lower bound:

- \blacktriangleright search tree: 90K nodes \rightarrow 1369 nodes
- time: 3.5 minutes \rightarrow 2 seconds



Experiments

- VLSI instance:
 - search tree: 20K nodes
 - time: 9 minutes



[34046 vertices, 54841 edges, answer 80]

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Walshaw Instances

Standard benchmark for graph partitioning ($\epsilon = 0$).

instance	n	т	opt	BB nodes	time [s]
add32	4 960	9 462	11	225	3
uk	4824	6837	19	1 624	4
3elt	4720	13722	90	12707	82
whitaker3	9 800	28 989	127	7 044	133
fe_4elt2	11143	32 818	130	10 391	224
4elt	15606	45 878	139	25 912	769
$data^*$	2851	15 093	189	495 569 759	5 750 388

[*: distributed execution using DryadOpt]

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Optimum bisections were known before, but without proofs.

Other Challenge Instances

series	instance	n	т	opt	BB nodes	time [s]
clustering	karate	34	78	10	4	0.00
	chesapeake	39	170	46	110 138	3.08
	dolphins	62	159	15	110	0.01
	lesmis	77	820	61	3 905 756	230.30
	polbooks	105	441	19	8	0.00
	football	115	613	61	7 301	1.08
	power	4941	6 594	12	94	0.21
delaunay	delaunay_n10	1024	3 0 5 6	63	14 361	18.25
	delaunay_n11	2 0 4 8	6 1 2 7	86	65 080	175.73
	delaunay_n12	4 0 9 6	12264	118	474 844	2711.73
	delaunay_n13	8 1 9 2	24 547	156	3 122 845	37 615.97
streets	luxembourg	114599	119 666	17	786	91.17

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We can solve very large instances with small bisections.

• NW (9th Challenge): 264 branch-and-bound nodes, 25 minutes



Instances from "exact" literature

- State-of-the-art approaches:
 - [Arm07]: semidefinite programming
 - [HPZ11]: quadratic programming

instance	п	т	opt	time [s]	[Arm07]	[HPZ11]
KKT_putt01_m2	115	433	28	0.81	1.67	1.51
mesh.274.469	274	469	37	0.03	8.52	24.62
gap2669.24859	2669	29037	55	0.15	348.95	—
taq170.424	170	4317	55	3.00	28.68	—
gap2669.6182	2669	12280	74	34.90	651.03	—
taq1021.2253	1021	4510	118	134.61	169.65	_

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Excellent performance for small bisections.

• Gargoyle: 2.6M branch-and-bound nodes, 13 hours



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[10002 vertices, 30000 edges, answer 175]

• Feline (mesh): 150K branch-and-bound nodes, 75 minutes



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Conclusion

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 - packing bound
 - decomposition
- Cut size matters
- Potential applications: evaluate/improve heuristics

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Thank you!