
Anytime AND/OR Depth-first Search for Combinatorial Optimization

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Outline

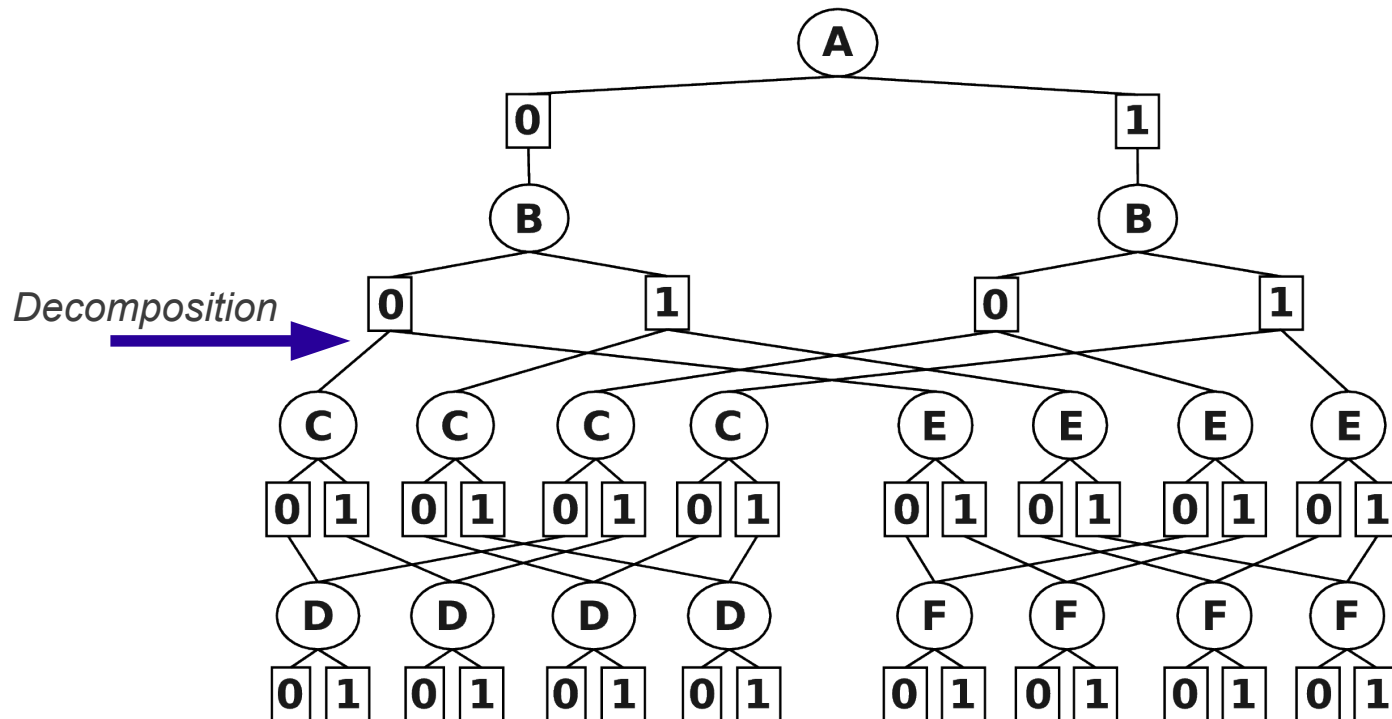
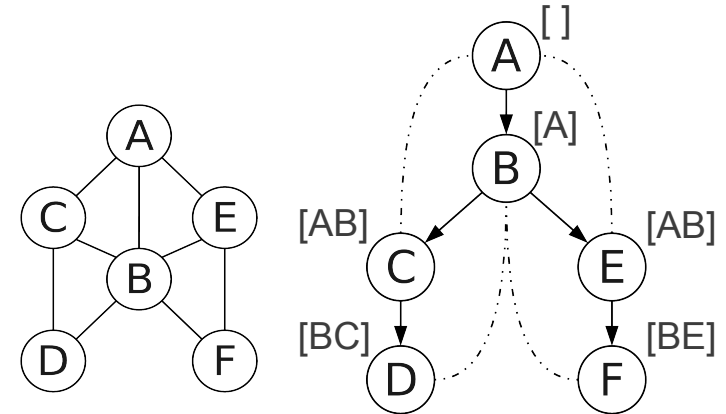


- AND/OR Search Spaces.
 - AND/OR Branch and Bound (AOBB).
- Conflict: Decomposition vs. Anytime.
 - Empirical evidence & analysis.
- Breadth-Rotating AOBB.
 - Example & analysis.
- Experimental Results.
 - Substantial improvements.



AND/OR Branch and Bound

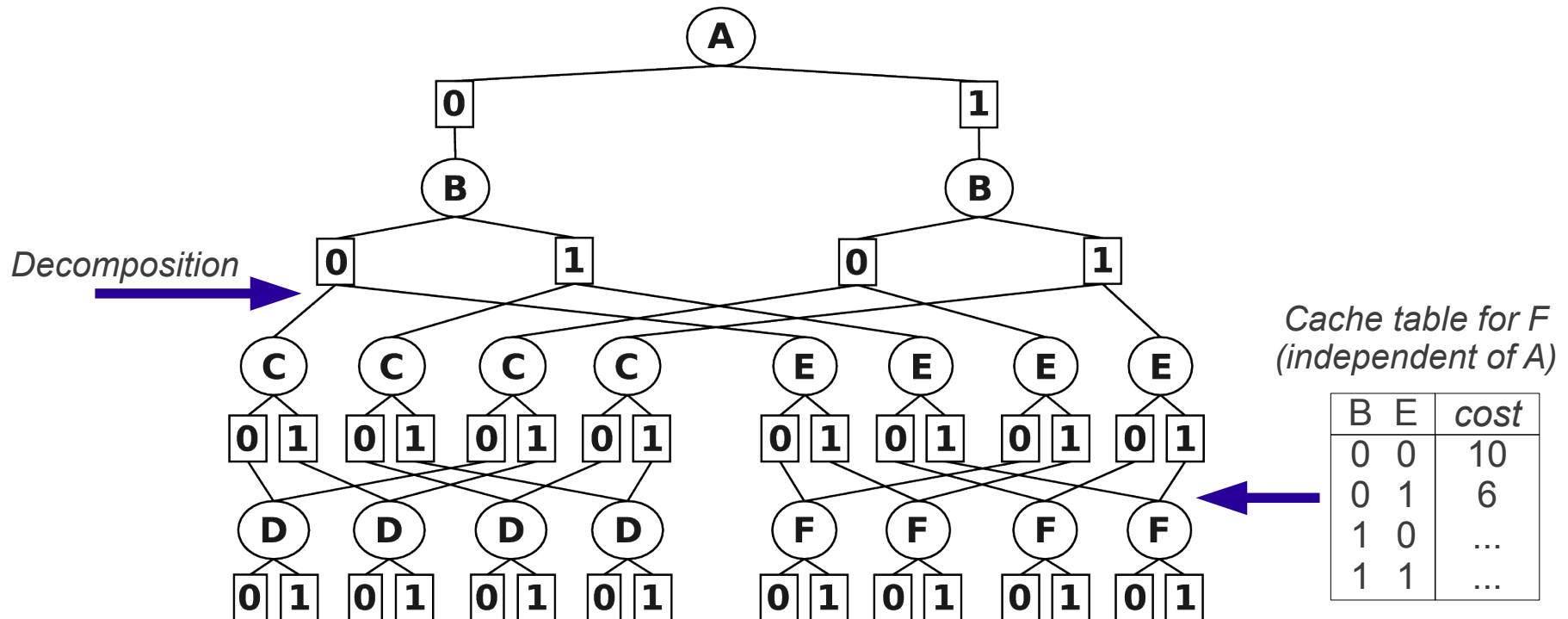
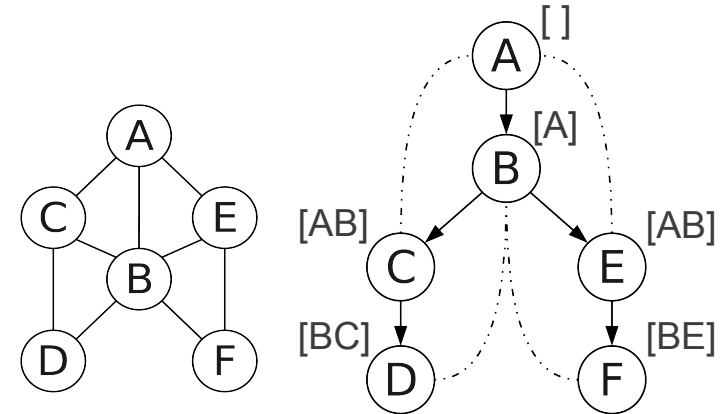
- Guided by pseudo tree:
 - Subproblem decomposition.





AND/OR Branch and Bound

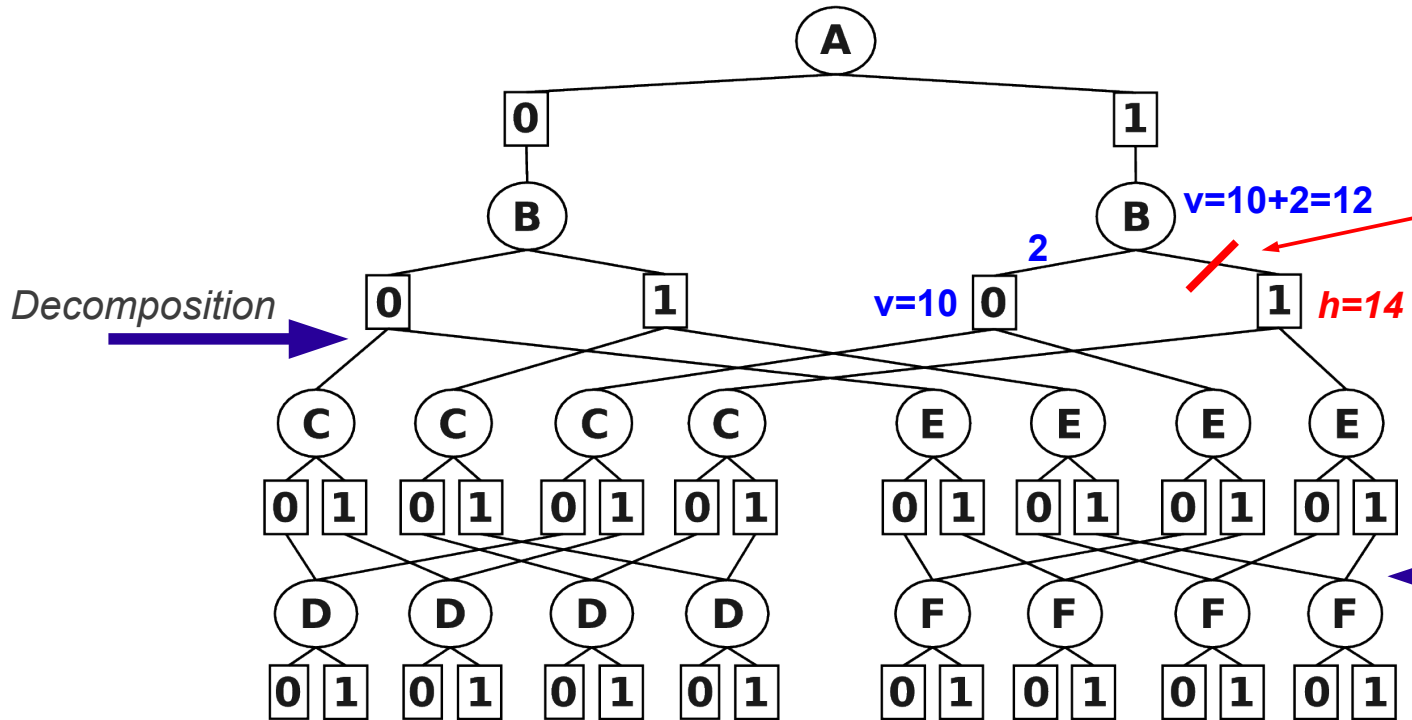
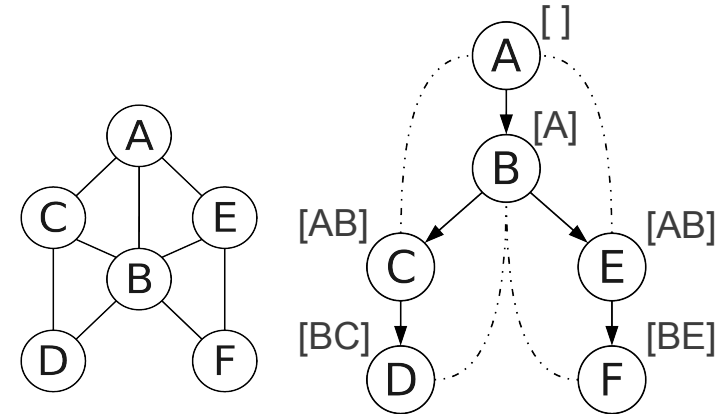
- Guided by pseudo tree:
 - Subproblem decomposition.
 - Merge unifiable subproblems.





AND/OR Branch and Bound

- Guided by pseudo tree:
 - Subproblem decomposition.
 - Merge unifiable subproblems.
- Mini-bucket heuristic.



Prune based on current best solution and heuristic estimate.

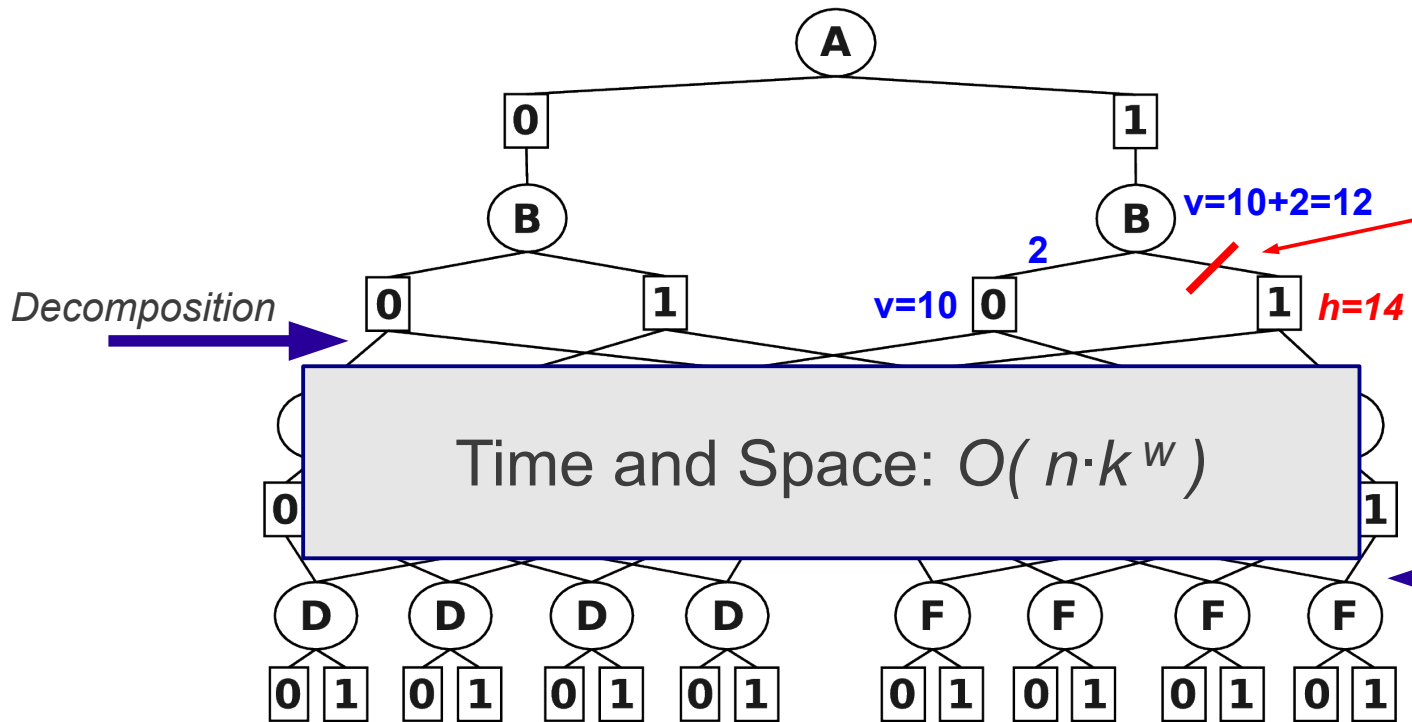
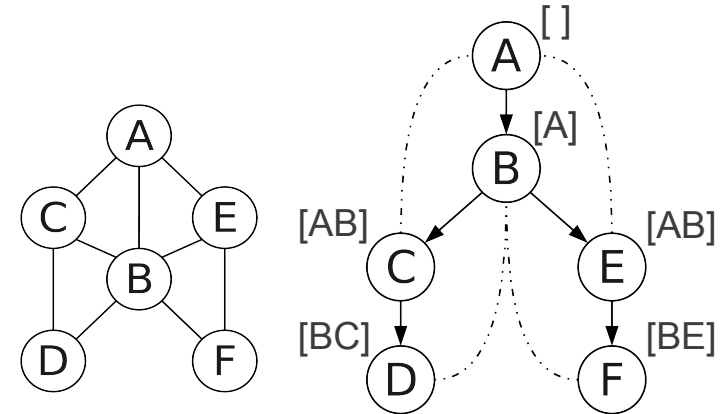
Cache table for F (independent of A)

B	E	cost
0	0	10
0	1	6
1	0	...
1	1	...



AND/OR Branch and Bound

- Guided by pseudo tree:
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Anytime Behavior of AOBB

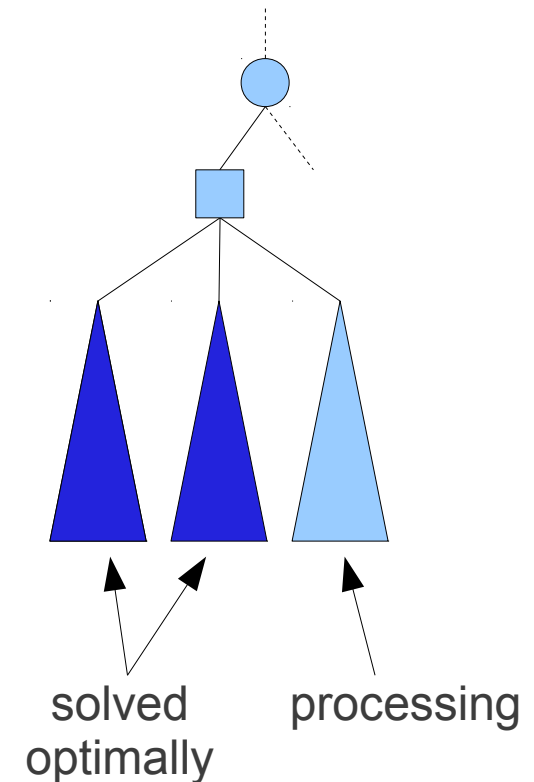


- Finding/proving optimal solution hard!
- Often easier:
 - Find any feasible solution quickly.
 - Improve with time, until optimum is found.
- Branch and Bound is anytime.
 - Depth-first to any solution and improve.
- Usage as approximation scheme:
 - AOBB competitive in UAI'10 evaluation.
 - But some cases: no solution within time bound.



Anytime vs. Decomposition

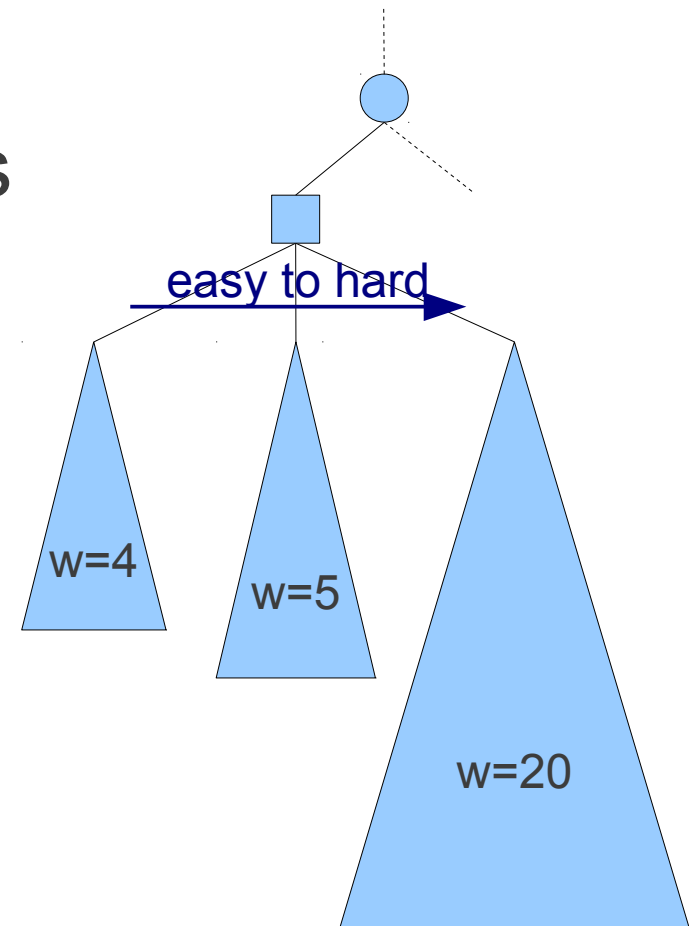
- Depth-first traversal of AND/OR space:
 - Subproblems successively solved to optimality.
- Breaks anytime behavior.
- First overall solution:
 - Is delayed until last subproblem starts processing.
 - Contains optimal solutions to all but last subproblem.





Remedy I: Subproblem Order

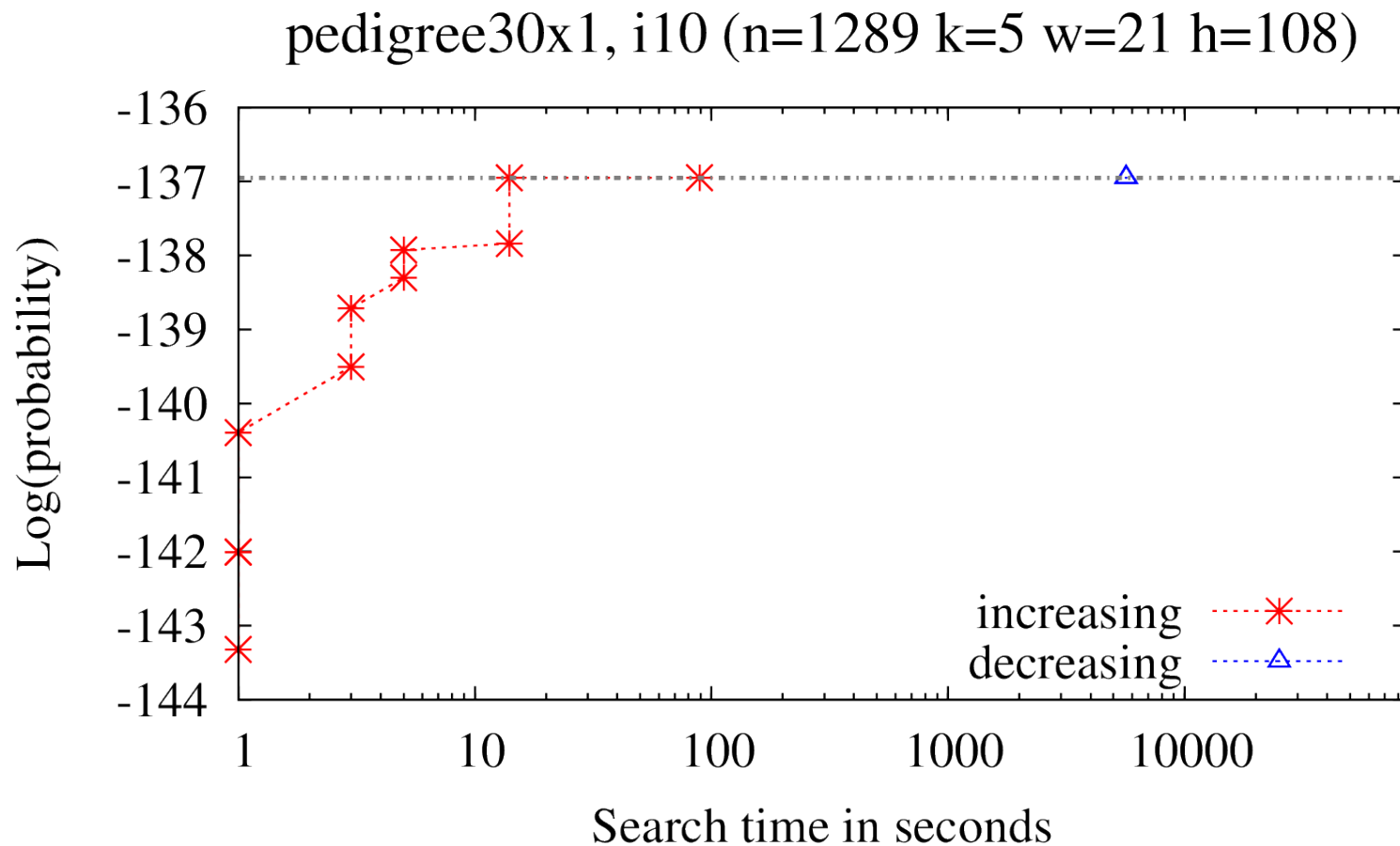
- One complex and several easy subproblems:
 - Solve easy ones optimally (fast).
 - Combine with anytime solutions from complex subproblem.
- Suggests ordering subproblems by increasing hardness:
 - Heuristic: subproblem induced width w .
 - AOBB has time $O(n \cdot k^w)$.
 - However, fails for multiple complex subproblems.





Remedy I: Subproblem Order

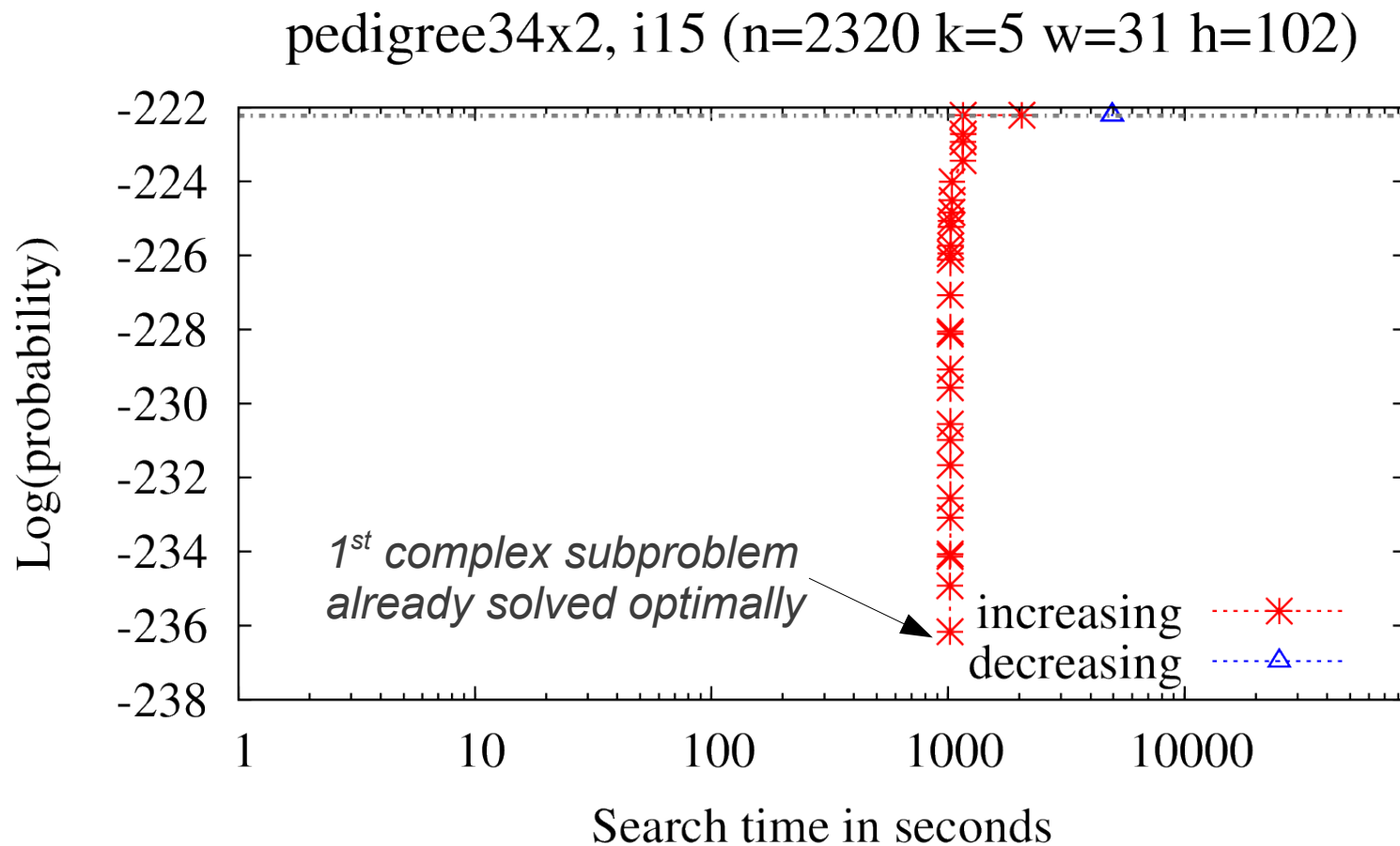
- 1 large and several smaller components.
 - Increasing order yields good performance.





Remedy I: Subproblem Order

- 2 large and several smaller components.
 - Even increasing order fails.



Remedy II: Subproblem Dive

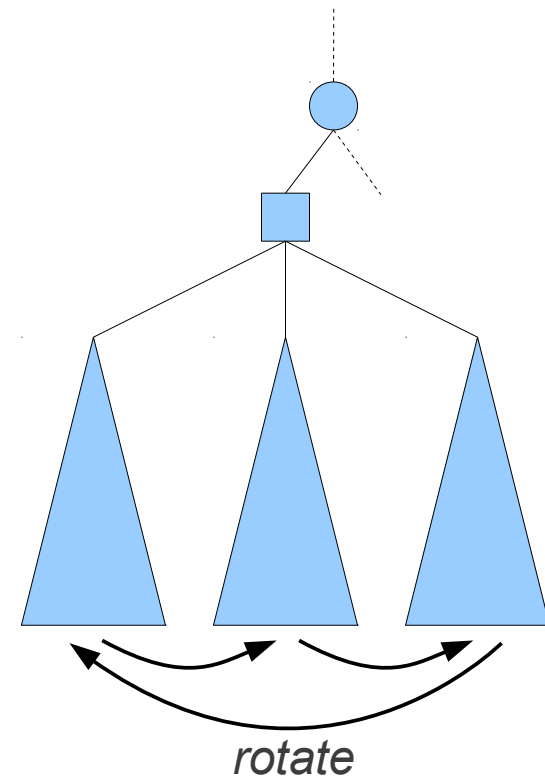


- Upon decomposition:
 - Greedily find solution for each subproblem.
 - Use mini-bucket heuristic for guidance.
 - Then solve each subproblem optimally, depth-first as before.
- Relies very much on heuristic:
 - Might fail due to dead end.
- Mixed performance in experiments.



Breadth-Rotating AOBB

- Construct all branches of solution tree “simultaneously”:
 - Take turns in processing subproblems.
 - Limit number of operations per visit.
 - Solve each depth-first as before.



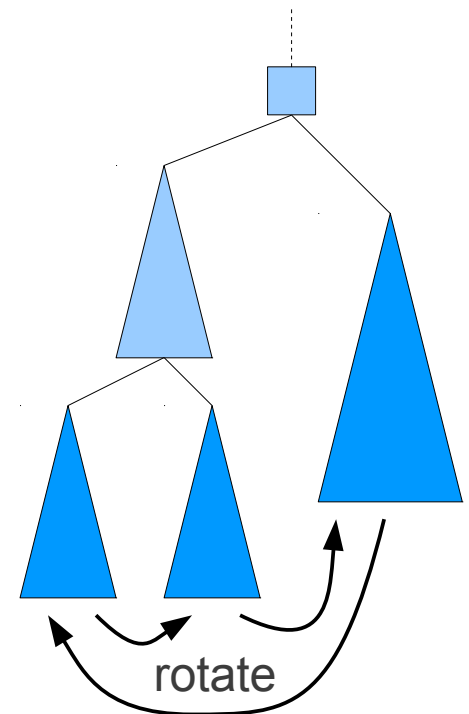


Breadth-Rotating AOBB

- High-level pseudo code:

1. Move breadth-first to next open subproblem P .
2. Process P depth-first, until either:
 - P is solved optimally,
 - P decomposes into child subproblems,
 - a predefined threshold of operations is reached.

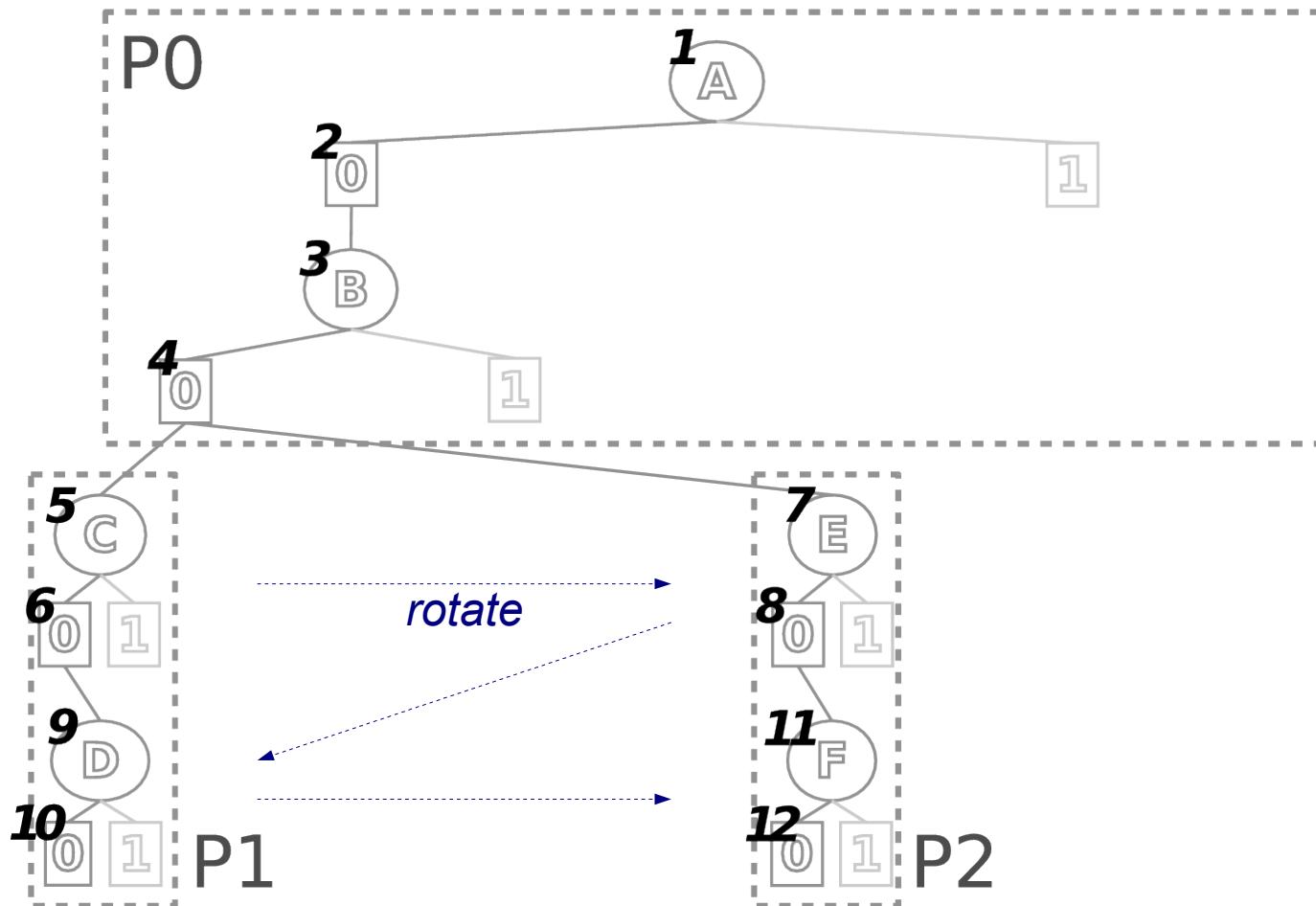
- Rotation skips subproblems with current child subproblems.





Breadth-Rotating AOBB

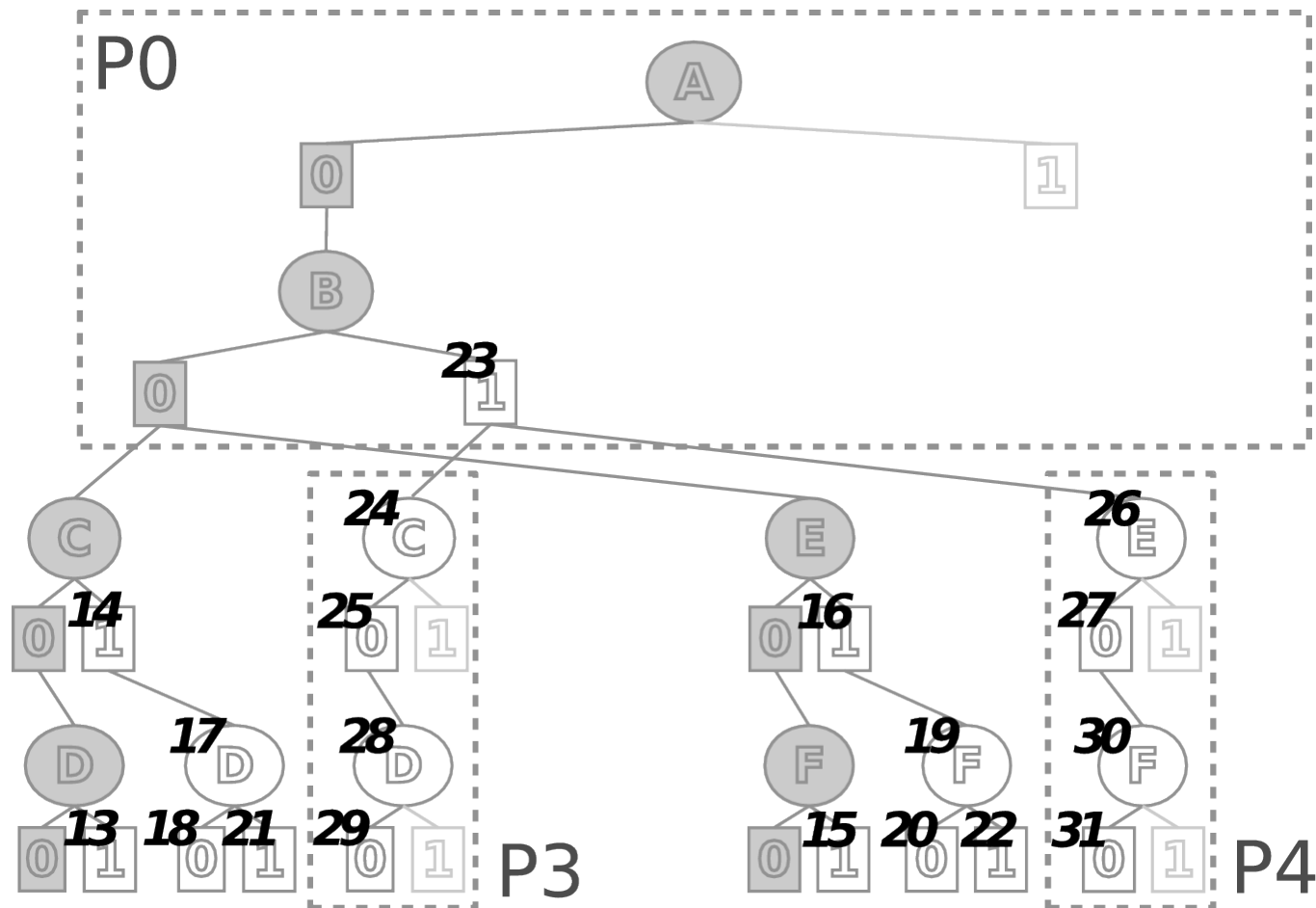
- Example problem with threshold $z = 2$.





Breadth-Rotating AOBB

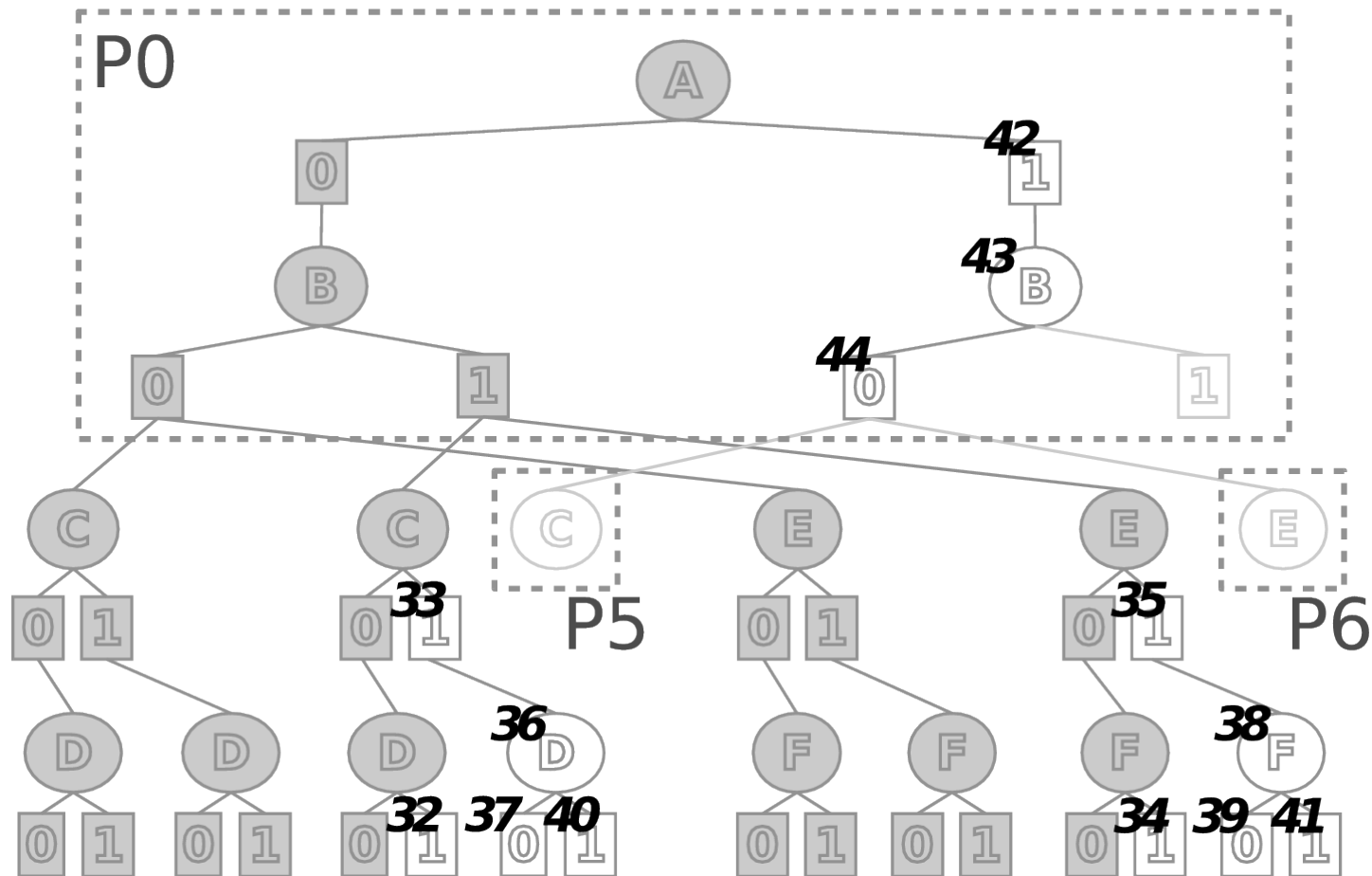
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Breadth-Rotating AOBB

- Example problem with threshold $z = 2$.



BRAOBB Analysis



- *Theorem:* BRAOBB maintains favorable asymptotic complexity of depth-first search.
 - Time: graph search $O(n \cdot k^w)$, tree search $O(n \cdot k^h)$.
 - Space: graph search $O(n \cdot k^w)$, tree search $O(n)$.
- Comparison with AOBB:
 - Anytime performance does not depend on subproblem order.
 - Overall performance (optimality proof) can increase or decrease.
 - Pruning impacted by node exploration.

Experimental Evaluation

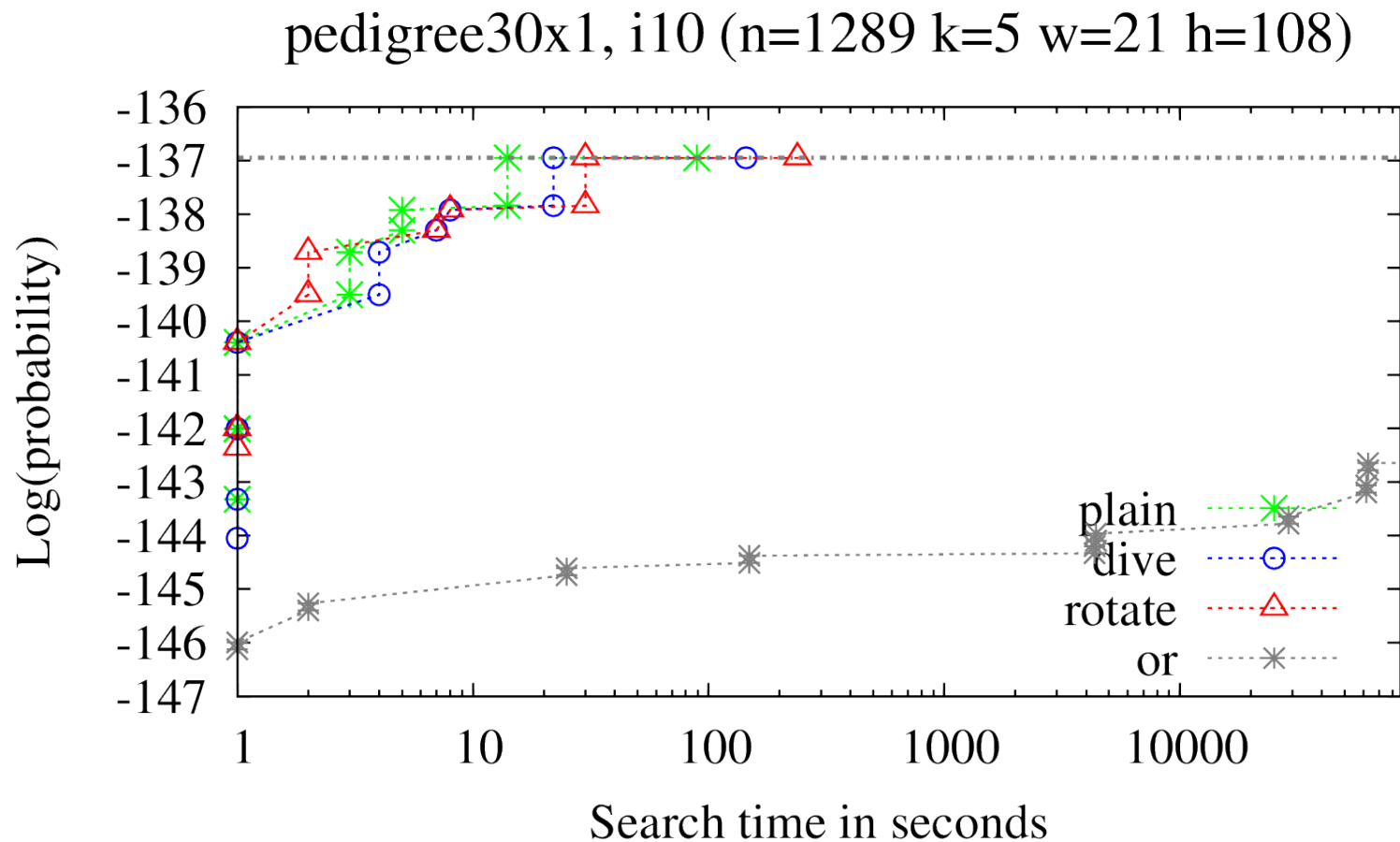


- Run and record anytime profiles (24h timeout):
 - Plain AOBB (increasing subproblem order).
 - AOBB with greedy dive.
 - Breadth-rotating AOBB.
 - OR graph search (no decomposition).
- Enforce decomposable problems:
 - Create network copies and connect at root.
 - 57 pedigree, 150 grid, 24 mastermind.
 - Combined over 60.000 CPU hours.
- UAI'10: Two very hard instances.



Experimental Evaluation

- “Sanity check”: 1 large subproblem
 - **AND/OR** schemes similarly good, **OR** search slow.

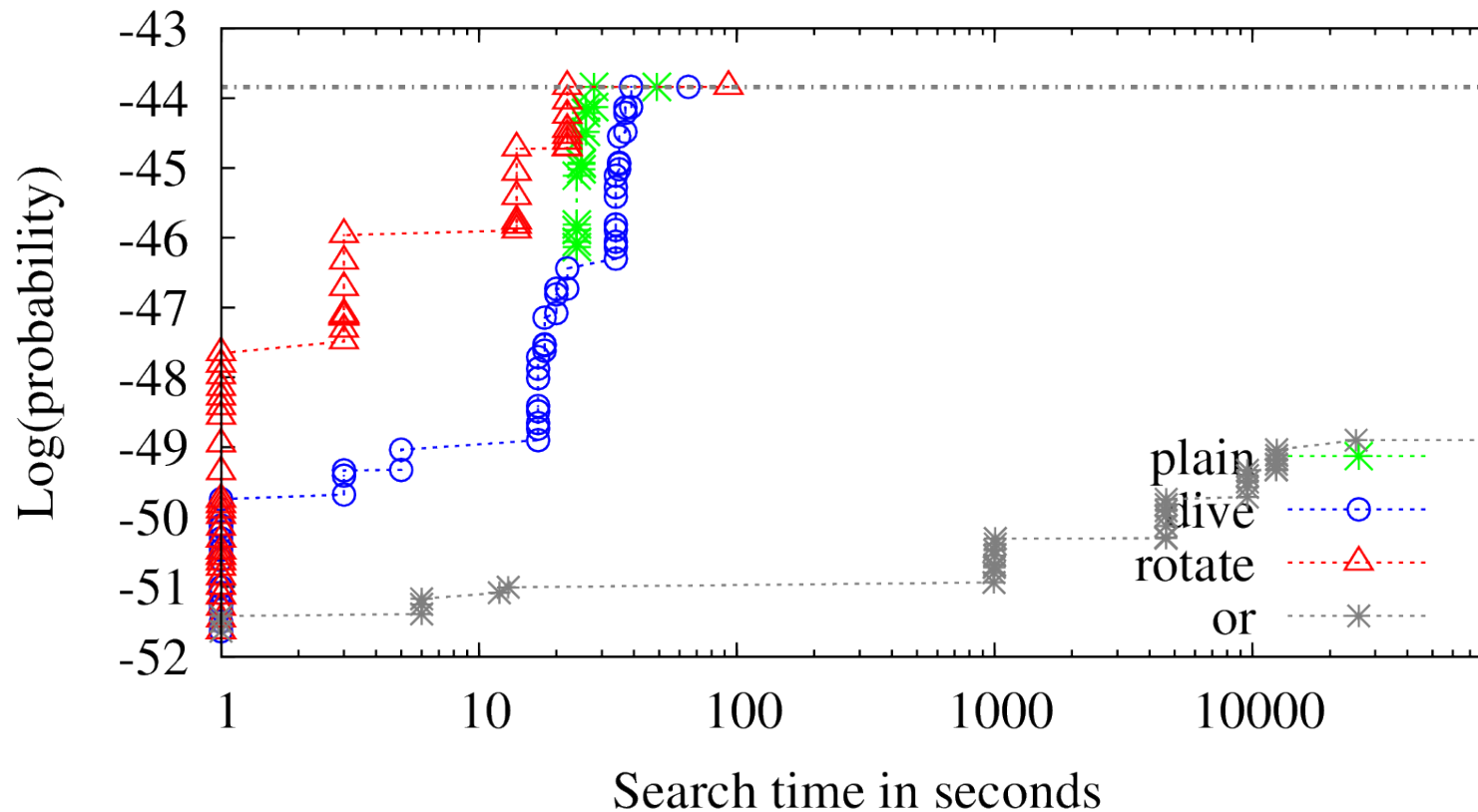




Experimental Evaluation

- Grid network with three large subproblems.
 - *rotate* outperforms *dive*, *plain* very late.

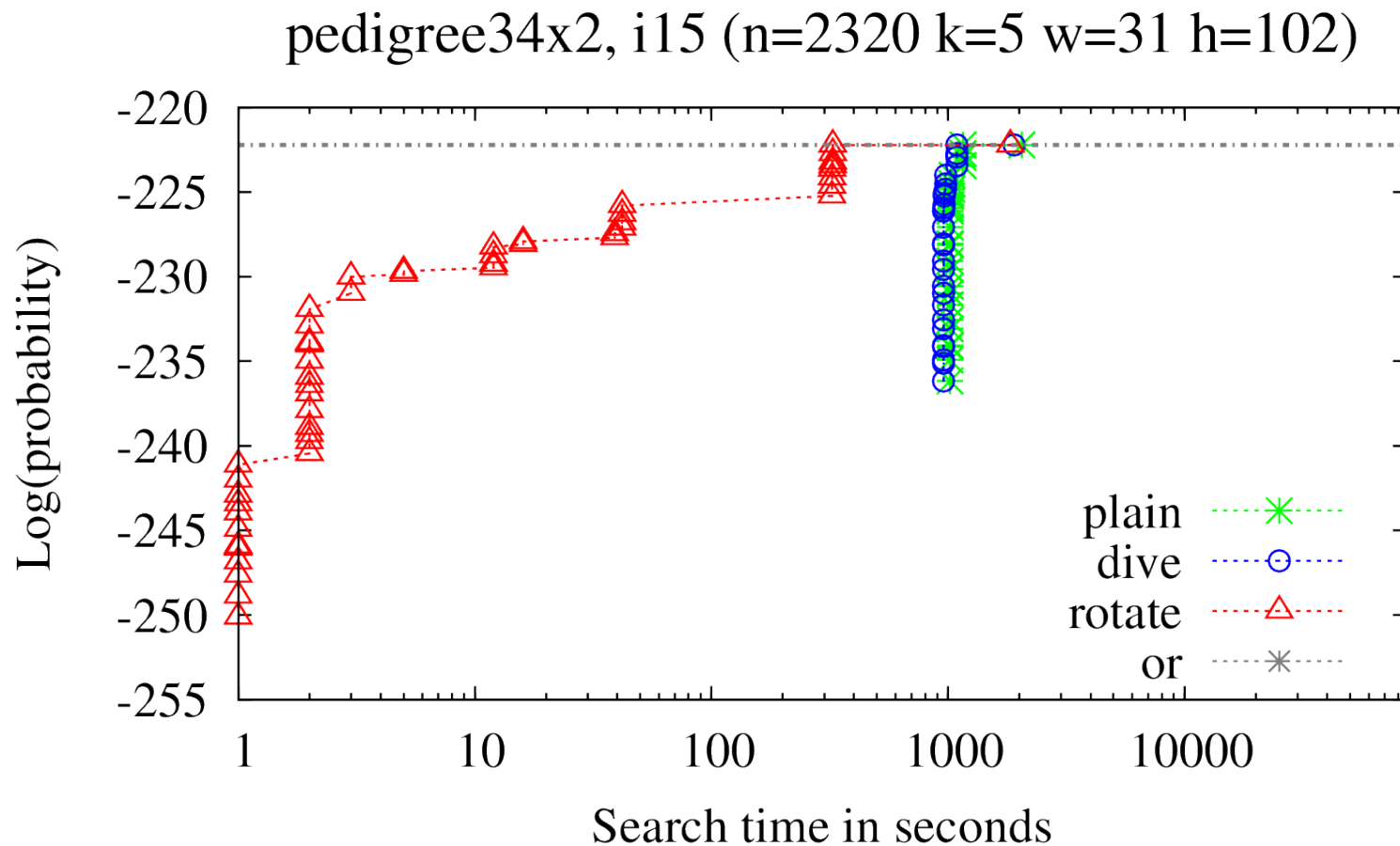
75-22-3x3, i20 (n=1452 k=2 w=32 h=94)





Experimental Evaluation

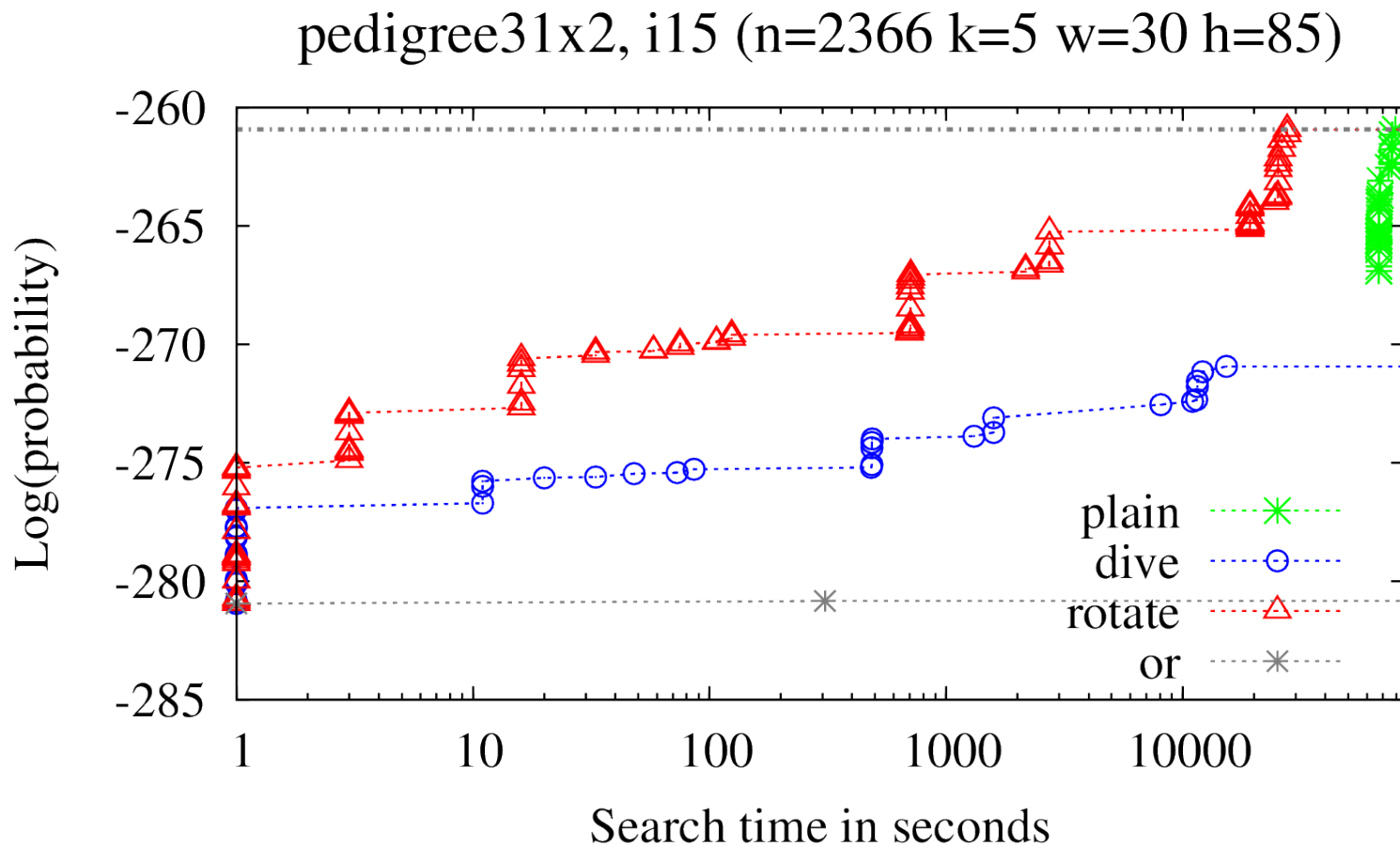
- Pedigree with two large subproblems.
 - *Dive* fails due to dead end, behaves like *plain*.





Experimental Evaluation

- Pedigree with two large subproblems.
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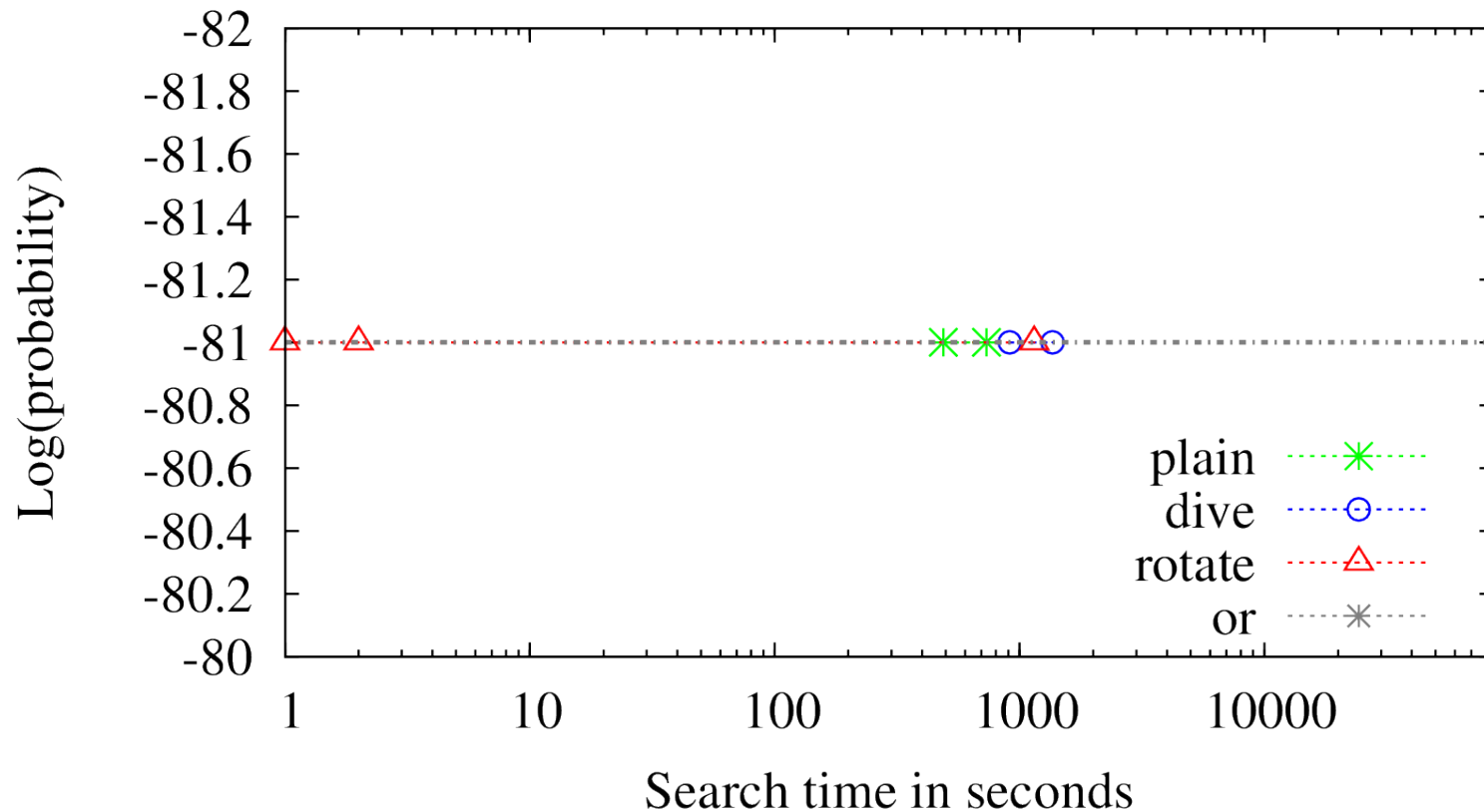




Experimental Evaluation

- Mastermind with three complex subproblems.
 - *rotate* finds solution much sooner.

mm-10-08-03-0012x3, i10 (n=7818 k=2 w=47 h=82)



Experimental Evaluation



- Summary table, entries give #instances:
 - Any solution / optimal solution / optimality proven

	Time bound						
	1 sec	5 sec	10 sec	1 min	5 min	1 hour	24 hours
Pedigree networks (171 total)							
plain	52 / 19 / 6	70 / 36 / 17	75 / 42 / 24	87 / 56 / 48	101 / 76 / 68	111 / 90 / 86	129 / 117 / 108
dive	76 / 16 / 5	86 / 29 / 13	94 / 38 / 20	105 / 53 / 48	116 / 69 / 64	127 / 89 / 86	135 / 114 / 105
rotate	153 / 26 / 2	157 / 40 / 15	160 / 47 / 24	162 / 59 / 48	164 / 74 / 60	165 / 98 / 84	167 / 127 / 102
or	73 / 6 / 1	76 / 7 / 5	77 / 10 / 5	79 / 10 / 9	82 / 12 / 11	87 / 16 / 15	90 / 22 / 21
Hybrid networks (150 total)							
plain	38 / 10 / 0	48 / 19 / 0	58 / 32 / 4	84 / 62 / 52	101 / 82 / 76	128 / 120 / 113	149 / 148 / 147
dive	47 / 6 / 0	52 / 12 / 0	55 / 24 / 1	82 / 54 / 37	97 / 78 / 71	121 / 111 / 104	147 / 147 / 146
rotate	122 / 16 / 0	128 / 27 / 0	129 / 35 / 1	136 / 69 / 38	143 / 86 / 73	146 / 126 / 110	149 / 149 / 147
or	45 / 0 / 0	45 / 1 / 0	46 / 1 / 0	53 / 2 / 0	57 / 4 / 2	64 / 10 / 9	74 / 21 / 21
Mastermind networks (24 total)							
plain	8 / 8 / 1	8 / 8 / 3	8 / 8 / 3	10 / 10 / 4	13 / 13 / 7	17 / 17 / 12	24 / 24 / 24
dive	8 / 8 / 1	8 / 8 / 3	8 / 8 / 3	11 / 11 / 5	12 / 12 / 6	21 / 21 / 19	24 / 24 / 24
rotate	18 / 18 / 1	18 / 18 / 3	18 / 18 / 3	18 / 18 / 3	21 / 21 / 4	24 / 24 / 19	24 / 24 / 24
or	0 / 0 / 0	0 / 0 / 0	0 / 0 / 0	0 / 0 / 0	0 / 0 / 0	0 / 0 / 0	0 / 0 / 0

First solutions quickly.



Experimental Evaluation

- Summary table, entries give #instances:
 - Any solution / optimal solution / optimality proven

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Grid networks (171 total)							
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dive	47 / 6 / 0	52 / 12 / 0	55 / 24 / 1	82 / 54 / 37	97 / 78 / 71	121 / 111 / 104	147 / 147 / 146
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or	0 / 0 / 0	0 / 0 / 0	0 / 0 / 0	0 / 0 / 0	0 / 0 / 0	0 / 0 / 0	0 / 0 / 0

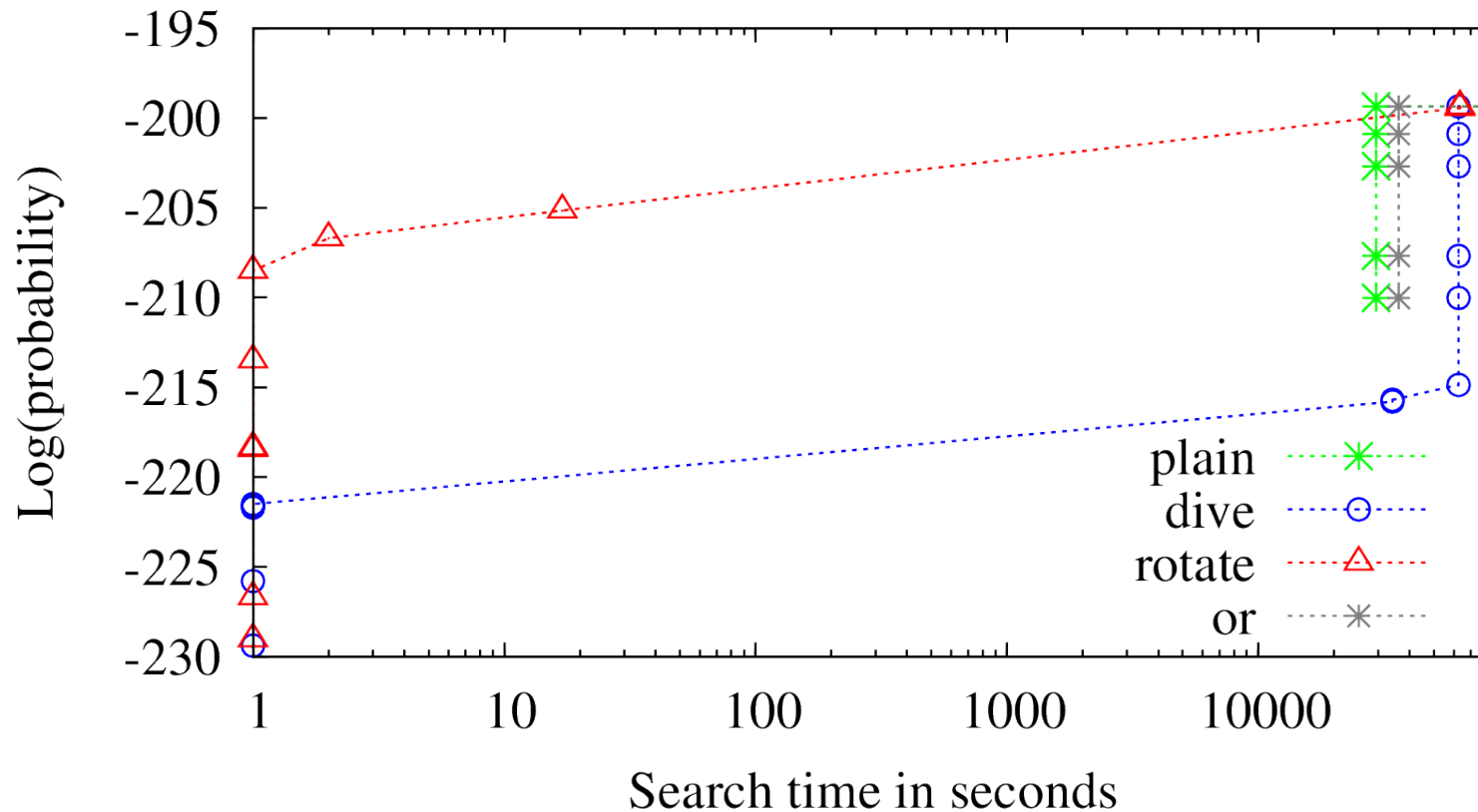
Faster to optimality,
a bit slower to prove.



Experimental Evaluation

- UAI'10: protein sidechain prediction.
 - Very hard due to large domains.

pdb1i24, i3 (n=337 k=81 w=33 h=57)

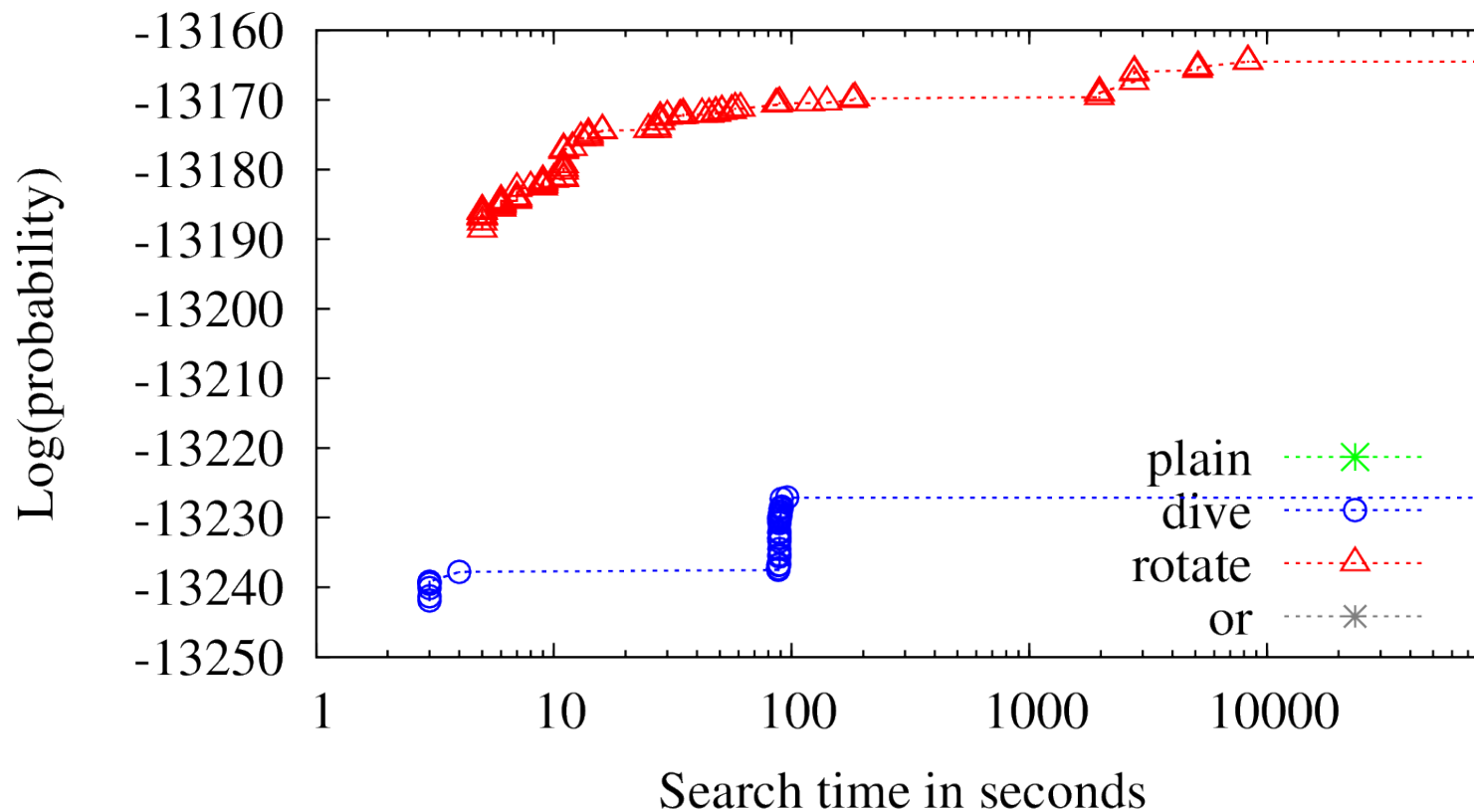




Experimental Evaluation

- UAI'10: protein folding.
 - Very complex, induced width >1000 .

protein1, i14 (n=14306 k=2 w=1122 h=1282)



Experiments Summary



- Plain AOBB:
 - Fails for more than one complex subproblem.
- AOBB with greedy dive:
 - Quick initial solution possible, if heuristic allows.
 - Slow to improve afterwards.
- Breadth-rotating AOBB:
 - Consistently good anytime performance.
 - Immediate initial solution on 293/345 instances.
 - Subsequent rapid improvements.
 - Useful as approximation scheme.

Conclusion



- Anytime behavior of depth-first BaB is compromised over AND/OR search spaces.
 - Suitable subproblem ordering and other remedies only partially viable.
- Introduced Breadth-rotating AOBB:
 - Rotate over subproblems, breadth-first.
 - Each explored depth-first.
 - Maintains asymptotic complexity.
- Greatly improved anytime performance.



Thank you!

Questions?



Impact of rotation threshold z

- BRAOBB: little change with different z .
 - Other rotation criteria trigger first.

