A Simple and Fast Bi-Objective Search Algorithm

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Bi-objective search has many applications



- Path-planning in robotics: distance and battery consumption
- HAZMAT transport in cities: travel time and risk of exposure for residents
- **Cycling** (path-finding): distance and driver safety
- Vehicle routing: monetary cost and travel time
 To our knowledge, bi-objective search not supported in PDDL

Bi-objective search

- **Two** objective cost functions g_1 , g_2
- Dominance relation:

 $(a,b)\prec (a',b')$ iff $a\leq a'$ and $b\leq b'$ but $(a,b)\neq (a',b')$

Pareto-optimal set: contains all *non-dominated* solutions



- State-of-the-art NAMOA*dr (Pulido et al., 2015).
- Dominance check: Does the newly found path to a state s is dominate (or is dominated by) a previously found path to s.



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Dominance checking has a big impact in performance.



The process takes linear time.

Highlights of Bi-Objective A* (BOA*)

- **1** Dominance checking in constant time (instead of linear time).
- 2 Simple. Resembling standard A*.
- **3** Empirical evaluation: find a Pareto set orders of magnitude faster.

- The heuristic values h_1 and h_2 are consistent.
- The *Open* list is sorted lexicographically by (f_1, f_2) .
- For each state s, BOA* maintains a $g_2^{\min}(s)$.



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In BOA*, dominance check takes constant time.

```
Algorithm 2: Bi-Objective A* (BOA*)
   Input : A search problem (S, E, c, s_{start}, s_{aoal}) and a
               consistent heuristic function h
   Output: A cost-unique Pareto-optimal solution set
 1 \text{ sols} \leftarrow \emptyset
 2 for each s \in S do
        q_2^{\min}(s) \leftarrow \infty
4 x \leftarrow new node with s(x) = s_{start}
5 \mathbf{g}(x) \leftarrow (0,0)
 6 parent(x) \leftarrow null
7 \mathbf{f}(x) \leftarrow (h_1(s_{start}), h_2(s_{start}))
 s Initialize Open and add x to it
9 while Open \neq \emptyset do
         Remove a node x from Open with the
10
          lexicographically smallest f-value of all nodes in
          Open
        if g_2(x) \ge g_2^{\min}(s(x)) \lor f_2(x) \ge g_2^{\min}(s_{goal}) then
11
              continue
12
         q_2^{\min}(s(x)) \leftarrow q_2(x)
13
         if s(x) = s_{goal} then
14
15
              Add x to sols
              continue
16
         for each t \in \text{Succ}(s(x)) do
              u \leftarrow new node with s(u) = t
18
              g(y) \leftarrow g(x) + c(s(x), t)
19
              parent(y) \leftarrow x
20
              \mathbf{f}(y) \leftarrow \mathbf{g}(y) + \mathbf{h}(t)
21
              if g_2(y) \ge g_2^{\min}(t) \lor f_2(y) \ge g_2^{\min}(s_{goal}) then
22
23
                   continue
24
              Add y to Open
25 return sols
```

Hernandez et al. (UNAB,PUC,WUSTL,USC): A Simple and Fast Bi-Objective Search Algorithm22 / 27

Theorem

BOA* computes a cost-unique Pareto-optimal solution set.

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We compare to:

- NAMOA*dr (Pulido et al., 2015)
- BOA* with standard linear-time dominance checking (sBOA*),
- Bi-Objective Dijkstra (BDijkstra), and Bidirectional Bi-Objective Dijkstra (BBDijkstra) (Sedeño et al., 2019).
- We use 5 road maps from the "9th DIMACS Implementation Challenge: Shortest Path".

Experimental Evaluation

Runtime (sec) on 50 instances. After 3,600 seconds, we use 3,600 seconds in the calculation of the average.

New York City (NY)						
264,346 states, 730,100 edges, $ sols = 199$ on average						
	Solved	Average	Max	Min		
NAMOA*	50/50	157.17	1,936.36	0.02		
sBOA*	50/50	9.75	148.65	0.10		
NAMOA*dr	50/50	0.65	4.99	0.11		
BOA*	50/50	0.32	1.95	0.11		
BBDijkstra	50/50	1.94	23.43	0.26		
BDijkstra	50/50	2.55	21.16	0.17		

Florida (FL)							
1,070,376 states, 2,712,798 edges, sols = 739 on average							
	Solved	Average	Max	Min			
NAMOA*	43/50	812.48	3,298.90	1.42			
sBOA*	46/50	349.64	1,238.25	0.43			
NAMOA*dr	50/50	19.66	329.79	0.43			
BOA*	50/50	4.59	60.54	0.43			
BBDijkstra	50/50	91.36	1,772.48	1.11			
BDijkstra	50/50	158.33	2,722.69	0.77			

BOA* versus the best algorithms of the state-of-the-art in the Great Lakes (LKS) map with 2,758,119 states and 6,885,658 edges.



- We present BOA* a simple and fast Bi-Objective A* search algorithm.
- BOA* resembling standard A*.
- BOA* is orders of magnitude faster than state-of-the-art.
- Research directions: Bounded-optimal Bi-Objective search and Multi-Objective search.