Foundations of Artificial Intelligence

17. State-Space Search: IDA*

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State-Space Search: Overview

Chapter overview: state-space search

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- ▶ 8.–12. Basic Algorithms
- ▶ 13.–19. Heuristic Algorithms
 - ► 13. Heuristics
 - ▶ 14. Analysis of Heuristics
 - ▶ 15. Best-first Graph Search
 - ▶ 16. Greedy Best-first Search, A*, Weighted A*
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17. State-Space Search: IDA*

17.1 IDA*: Idea

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IDA*

The main drawback of the presented best-first graph search algorithms is their space complexity.

Idea: use the concepts of iterative-deepening DFS

- depth-limited depth-first search with increasing limits
- instead of depth we limit f(in this chapter f(n) := g(n) + h(n.state) as in A^*)
- → IDA* (iterative-deepening A*)
- tree search, unlike the previous best-first search algorithms

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17. State-Space Search: IDA*

IDA*: Algorithm

Reminder: Iterative Deepening Depth-first Search

reminder: iterative deepening depth-first search

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17.2 IDA*: Algorithm

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IDA*: Algorithm

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First Attempt: IDA* Main Function

first attempt: iterative deepening A* (IDA*)

IDA* (First Attempt)
for f_limit ∈ {0,1,2,...}:
    solution := f_limited_search(init(), 0, f_limit)
    if solution ≠ none:
        return solution

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IDA*: Algorithm

First Attempt: *f*-Limited Search

```
function f_limited_search(s, g, f_limit):
if g + h(s) > f_{\text{-}}limit:
     return none
if is_goal(s):
     return ()
for each \langle a, s' \rangle \in \text{succ}(s):
     solution := f\_limited\_search(s', g + cost(a), f\_limit)
     if solution \neq none:
           solution.push_front(a)
           return solution
return none
```

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IDA*: Algorithm

Growing the *f* Limit

- ▶ In IDDFS, we grow the limit from the smallest limit that gives a non-empty search tree (0) by 1 at a time.
- ► This usually leads to exponential growth of the tree between rounds, so that re-exploration work can be amortized.
- ▶ In our first attempt at IDA*, there is no guarantee that increasing the f limit by 1 will lead to a larger search tree than in the previous round.
- ▶ This problem becomes worse if we also allow non-integer (fractional) costs, where increasing the limit by 1 would be very arbitrary.

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IDA* First Attempt: Discussion

- ▶ The pseudo-code can be rewritten to be even more similar to our IDDFS pseudo-code. However, this would make our next modification more complicated.
- ▶ The algorithm follows the same principles as IDDFS, but takes path costs and heuristic information into account.
- ▶ For unit-cost state spaces and the trivial heuristic $h: s \mapsto 0$ for all states s, it behaves identically to IDDFS.
- For general state spaces, there is a problem with this first attempt, however.

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IDA*: Algorithm

Setting the Next *f* Limit

idea: let the f-limited search compute the next sensible f limit

- \blacktriangleright Start with h(init()), the smallest f limit that results in a non-empty search tree.
- In every round, increase the f limit to the smallest value that ensures that in the next round at least one additional path will be considered by the search.
- → f_limited_search now returns two values:
 - ▶ the next f limit that would include at least one new node in the search tree (∞ if no such limit exists: none if a solution was found), and
 - the solution that was found (or none).

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Final Algorithm: f-Limited Search
    function f_limited_search(s, g, f_limit):
    if g + h(s) > f_{\text{-}}limit:
          return \langle g + h(s), none \rangle
    if is_goal(s):
          return (none, ())
     new\_limit := \infty
    for each \langle a, s' \rangle \in \text{succ}(s):
          \langle child\_limit, solution \rangle := f\_limited\_search(s', g + cost(a), f\_limit)
          if solution \neq none:
                solution.push_front(a)
                return (none, solution)
          new_limit := min(new_limit, child_limit)
    return (new_limit, none)
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      new limit := \infty
      for each \langle a, s' \rangle \in \text{succ}(s):
            \langle child\_limit, solution \rangle := f\_limited\_search(s', g + cost(a), f\_limit)
            if solution \neq none:
                  solution.push_front(a)
                  return (none, solution)
            new\_limit := min(new\_limit, child\_limit)
      return (new_limit, none)
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17.3 IDA*: Properties
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IDA*: Properties

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IDA*: Properties

Inherits important properties of A* and depth-first search:

- ightharpoonup semi-complete if h safe and cost(a) > 0 for all actions a
- optimal if h admissible
- **space complexity** $O(\ell b)$, where
 - \triangleright ℓ : length of longest generated path (for unit cost problems: bounded by optimal solution cost)
 - ▶ b: branching factor

→ proofs?

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IDA*: Properties

IDA*: Discussion

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compared to A* potentially considerable overhead because no duplicates are detected

- → exponentially slower in many state spaces
- → often combined with partial duplicate elimination (cycle detection, transposition tables)
- overhead due to iterative increases of f limit often negligible, but not always
 - especially problematic if action costs vary a lot: then it can easily happen that each new f limit only considers a small number of new paths

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17.4 Summary

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Summary

- ► IDA* is a tree search variant of A* based on iterative deepening depth-first search
- main advantage: low space complexity
- disadvantage: repeated work can be significant
- most useful when there are few duplicates

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