Heuristic Search

Alice Gao Lecture 3 Readings: RN 3.5 (esp. 3.5.2), PM 3.6, 3.7.

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Outline

Learning Goals

Why Heuristic Search

LCFS, GBFS, and A* Lowest-Cost-First Search Greedy Best-First Search A* Search

Designing an Admissible Heuristic

Pruning the Search Space

Learning goals

By the end of the lecture, you should be able to

- Describe motivations for applying heuristic search algorithms.
- Trace the execution of and implement the Lowest-cost-first search, Greedy best-first search and A* search algorithm.
- Describe properties of the Lowest-cost-first, Greedy best-first and A* search algorithms.
- Design an admissible heuristic function for a search problem. Describe strategies for choosing among multiple heuristic functions.
- Describes strategies for pruning a search space.

Learning Goals

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Why Heuristic Search?

How would _____ choose which one of the two states to expand?

- an uninformed search algorithm
- humans

5	3	
8	7	6
2	4	1

1	2	3
4	5	
7	8	6

Why Heuristic Search

An uninformed search algorithm

- considers every state to be the same.
- does not know which state is closer to the goal.
- may not find the optimal solution.

An heuristic search algorithm

- uses heuristics to estimate how close the state is to a goal.
- try to find the optimal solution.

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Suppose that we are executing a search algorithm and we have added a path ending at \boldsymbol{n} to the frontier.

cost(n) is the actual cost of the path ending at n.

The Heuristic Function

Definition (search heuristic)

A search heuristic h(n) is an estimate of the cost of the cheapest path from node n to a goal node.

In general, h(n) can be arbitrary. However, a good heuristic function has the following properties.

- problem-specific.
- non-negative.
- h(n) = 0 if n is a goal node.
- ▶ *h*(*n*) must be easy to compute (without search).

LCFS, GBFS, and A*

- ▶ LCFS: remove the path with the lowest cost *cost*(*n*).
- GBFS: remove the path with the lowest heuristic value h(n).
- ► A*: remove the path with the lowest cost + heuristic value cost(n) + h(n).

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Lowest-cost-first search

- Frontier is a priority queue ordered by cost(n).
- Expand the path with the lowest cost(n).

Trace LCFS on a search graph

If there is a tie, remove nodes from the frontier in alphabetical order.



Space and Time Complexities

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Both complexities are exponential. LCFS examines a lot of paths to ensure that it returns the optimal solution first.

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Completeness and Optimality

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Completeness and Optimality

Yes and yes under mild conditions:(1) The branching factor is finite.(2) The cost of every edge is bounded below by a positive constant.

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Greedy Best-First Search

- Frontier is a priority queue ordered by h(n).
- Expand the node with the lowest h(n).

Trace GBFS on a search graph

If there is a tie, remove nodes from the frontier in alphabetical order.



Space and Time Complexities

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Completeness and Optimality

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Completeness and Optimality

No, GBFS is not complete. It could be stuck in a cycle. No, GBFS is not optimal. GBFS may return a sub-optimal path first.

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Completeness and Optimality

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A* Search

- Frontier is a priority queue ordered by f(n) = cost(n) + h(n).
- Expand the node with the lowest f(n).

Trace A* search on a search graph

If there is a tie, remove nodes from the frontier in alphabetical order.



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Completeness and Optimality

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Both complexities are exponential.

Completeness and Optimality

Yes and Yes, given mild conditions on the heuristic function.

Definition (admissible heuristic)

A heuristic h(n) is admissible if it never over-estimates the cost of the cheapest path from node n to a goal node.

Theorem (Optimality of A*)

If the heuristic h(n) is admissible, the solution found by A^* is optimal.

A* is Optimally Efficient

Among all optimal algorithms that start from the same start node and use the same heuristic, A^* expands the fewest nodes.

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Some Heuristic Functions for 8-Puzzle

- Manhattan Distance Heuristic: The sum of the Manhattan distances of the tiles from their goal positions
- Misplaced Tile Heuristic: The number of tiles that are NOT in their goal positions

Both heuristic functions are admissible.

 5
 3

 8
 7
 6

 2
 4
 1

Initial State

Goal State

1	2	3
4	5	6
7	8	

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Constructing an Admissible Heuristic

- 1. Define a relaxed problem by simplifying or removing constraints on the original problem.
- 2. Solve the relaxed problem without search.
- 3. The cost of the optimal solution to the relaxed problem is an admissible heuristic for the original problem.

Constructing an Admissible Heuristic for 8-Puzzle

8-puzzle: A tile can move from square A to square B

- ▶ if A and B are adjacent, and
- B is empty.

Which heuristics can we derive from relaxed versions of this problem?

CQ: Constructing an Admissible Heuristic

CQ: Which heuristics can we derive from the following relaxed 8-puzzle problem?

A tile can move from square A to square B if A and B are adjacent.

- (A) The Manhattan distance heuristic
- (B) The Misplaced tile heuristic
- (C) Another heuristic not described above

CQ: Constructing an Admissible Heuristic

- **CQ:** Which heuristics can we derive from the following relaxed 8-puzzle problem?
- A tile can move from square A to square B.
- (A) The Manhattan distance heuristic
- (B) The Misplaced tile heuristic
- (C) Another heuristic not described above

Which Heuristic is Better?

- We want a heuristic to be admissible.
- Want a heuristic to have higher values (close to h^*).
- Prefer a heuristic that is very different for different states.

Dominating Heuristic

Definition (dominating heuristic)

Given heuristics $h_1(n)$ and $h_2(n)$. $h_2(n)$ dominates $h_1(n)$ if

- $\blacktriangleright (\forall n \ (h_2(n) \ge h_1(n))).$
- $(\exists n \ (h_2(n) > h_1(n))).$

Theorem

If $h_2(n)$ dominates $h_1(n)$, A^* using h_2 will never expand more nodes than A^* using h_1 .

CQ: Which Heuristic of 8-puzzle is Better?

- CQ: Which of the two heuristics of the 8-puzzle is better?
- (A) The Manhattan distance heuristic dominates the Misplaced tile heuristic.
- (B) The Misplaced tile heuristic dominates the Manhattan distance heuristic.
- (C) Neither dominates the other one.

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Whenever we realize that we are following a cycle, we should stop expanding the path.

Why do we want to perform cycle pruning?

Cycles may cause an algorithm to not terminate, e.g. DFS. Exploring a cycle is a waste of time since it cannot be part of a solution.

How do we perform cycle pruning?

How do we perform cycle pruning?

Algorithm 2 Search w/ Cycle Pruning

```
1: ...

2: for every neighbour n of n_k do

3: if n \notin \langle n_0, \ldots, n_k \rangle then

4: add \langle n_0, \ldots, n_k, n \rangle to frontier;

5: ...
```

How do we perform cycle pruning?

Algorithm 3 Search w/ Cycle Pruning

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1: ...

2: for every neighbour n of n_k do

3: if n \notin \langle n_0, \ldots, n_k \rangle then

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What is the complexity of cycle pruning for DFS and BFS?

How do we perform cycle pruning?

Algorithm 4 Search w/ Cycle Pruning

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1: ...

2: for every neighbour n of n_k do

3: if n \notin \langle n_0, \ldots, n_k \rangle then

4: add \langle n_0, \ldots, n_k, n \rangle to frontier;

5: ...
```

What is the complexity of cycle pruning for DFS and BFS?

DFS: constant time. remember the nodes on the current path. BFS: linear in path length.

Why do we want to perform multi-path pruning?

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If we have already found a path to a node, we can discard other paths to the same node.

What is the relationship between cycle pruning and multi-path pruning?

Cycle pruning is a special case of multi-path pruning.

Search w/ Multi-Path Pruning

How do we perform multi-path pruning?

Search w/ Multi-Path Pruning

How do we perform multi-path pruning?

Algorithm 6 Search w/ Multi-Path Pruning

- 1: procedure SEARCH(Graph, Start node s, Goal test qoal(n)) frontier := { $\langle s \rangle$ }; 2: explored := $\{\}$; 3: while frontier is not empty do 4: select and remove path $\langle n_0, \ldots, n_k \rangle$ from frontier; 5: if $n_k \notin explored$ then 6: 7: add n_k to explored if $goal(n_k)$ then 8: return $\langle n_0, \ldots, n_k \rangle$; 9: for every neighbour n of n_k do 10: add $\langle n_0, \ldots, n_k, n \rangle$ to frontier; 11:
- 12: return no solution

A problem with multi-path pruning

- Multi-path pruning says that we keep the first path to a node and discard the rest.
- What if the first path to a node is not the least-cost path?
- Can multi-path pruning cause a search algorithm to fail to find the optimal solution?

Lowest-cost-first search w/ multi-path pruning

Can Lowest-Cost-First Search with multi-path pruning discard the optimal solution?

- (A) Yes, it is possible.
- (B) No, it is not possible.

A* search w/ multi-path pruning

Can A* search with an admissible heuristic and multi-path pruning discard the optimal solution?

- (A) Yes, it is possible.
- (B) No, it is not possible.

Finding optimal solution w/ multi-path pruning

What if a subsequent path to n is shorter than the first path found?

- Remove all paths from the frontier that use the longer path.
- Change the initial segment of the paths on the frontier to use the shorter path.
- Make sure that we find the least-cost path to a node first.

Consistent Heuristic

An admissible heuristic requires that:
 For any node m and any goal node g,

$$h(m) - h(g) \le cost(m, g).$$

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 To ensure that A* with multi-path pruning is optimal, we need a consistent heuristic function: For any two nodes m and n,

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 To ensure that A* with multi-path pruning is optimal, we need a consistent heuristic function: For any two nodes m and n,

$$h(m) - h(n) \le cost(m, n).$$

A consistent heuristic satisfies the monotone restriction:
 For any edge from m to n,

$$h(m) - h(n) \le cost(m, n).$$

Constructing Consistent Heuristics

- Most admissible heuristic functions are consistent.
- It's challenging to come up with a heuristic function that is admissible but not consistent.

Revisiting the learning goals

By the end of the lecture, you should be able to

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