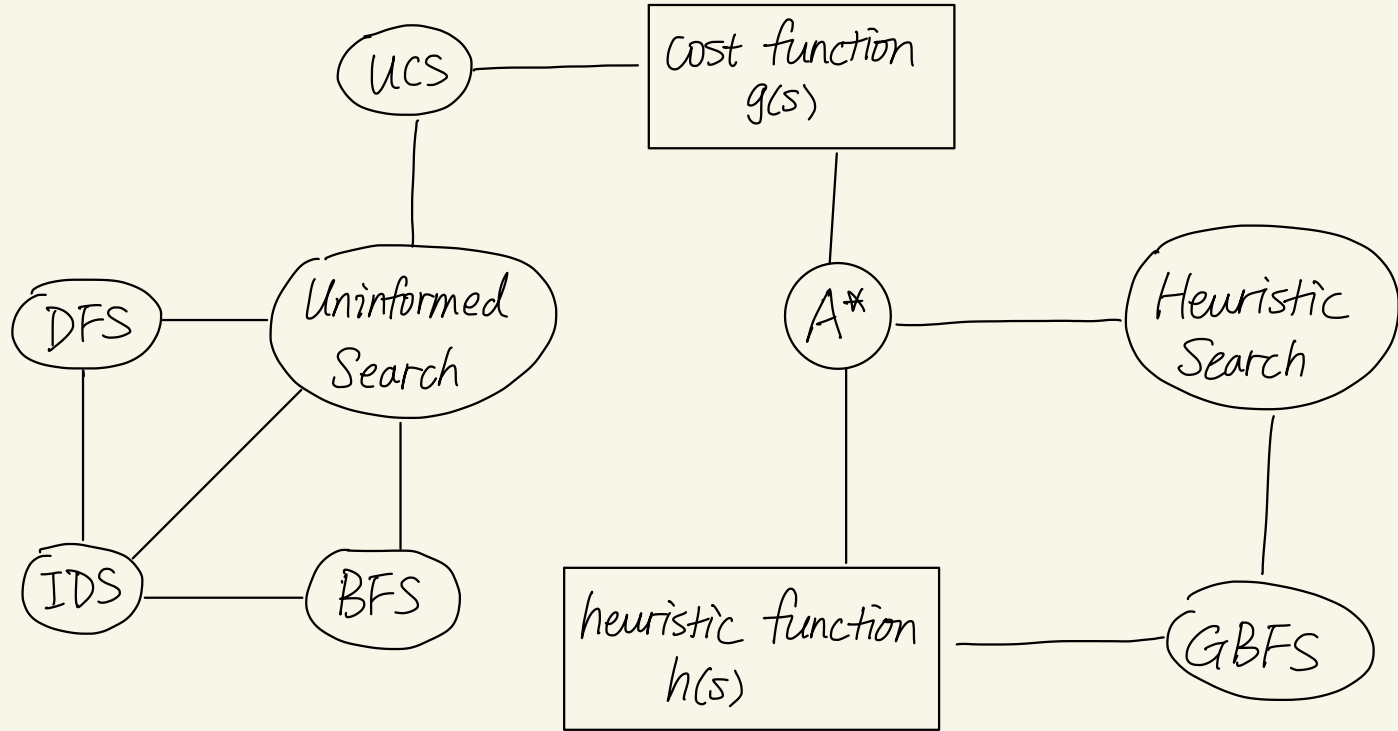


Search

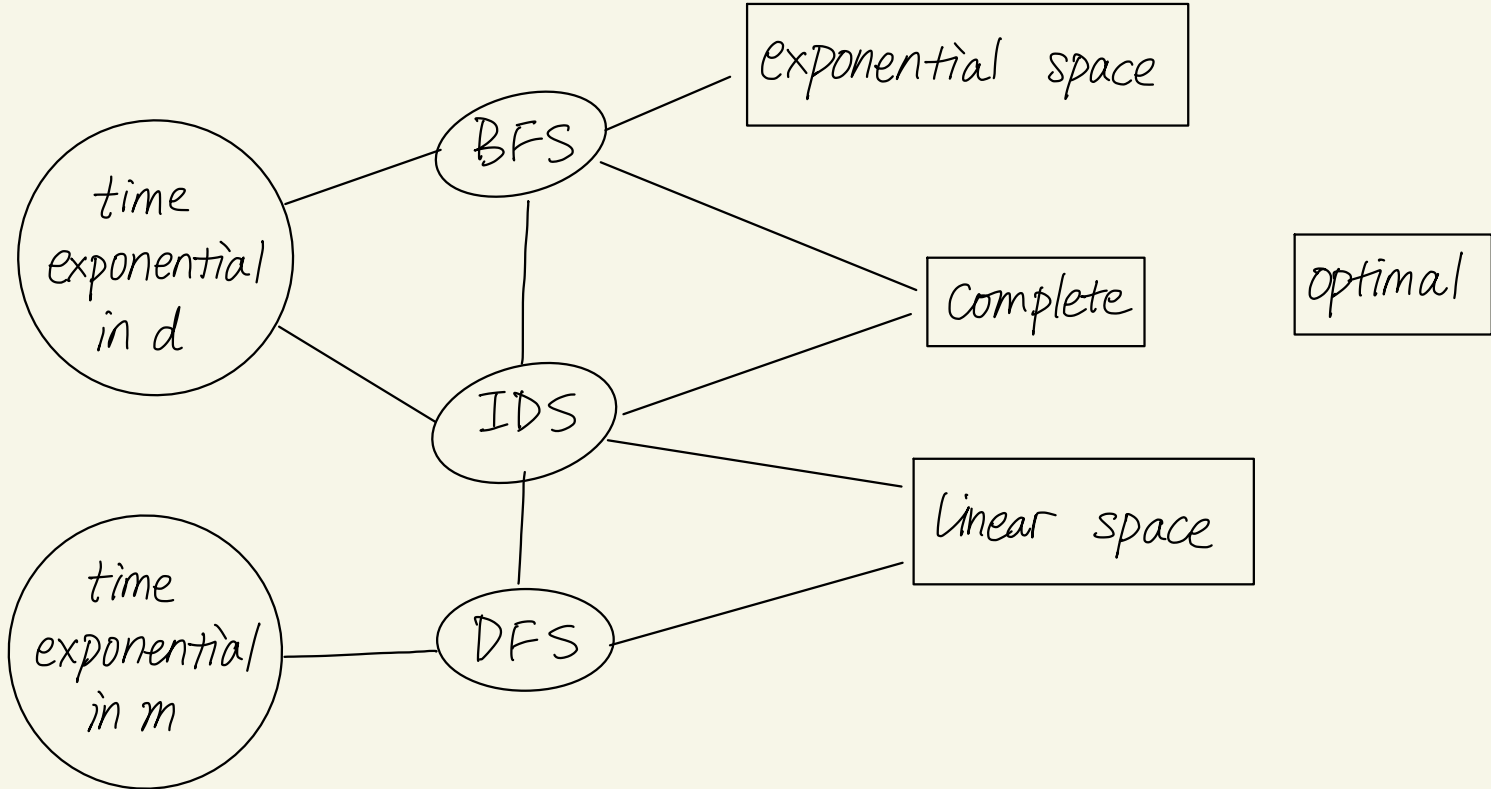
Summary and Questions

by Alice Gao ^_^

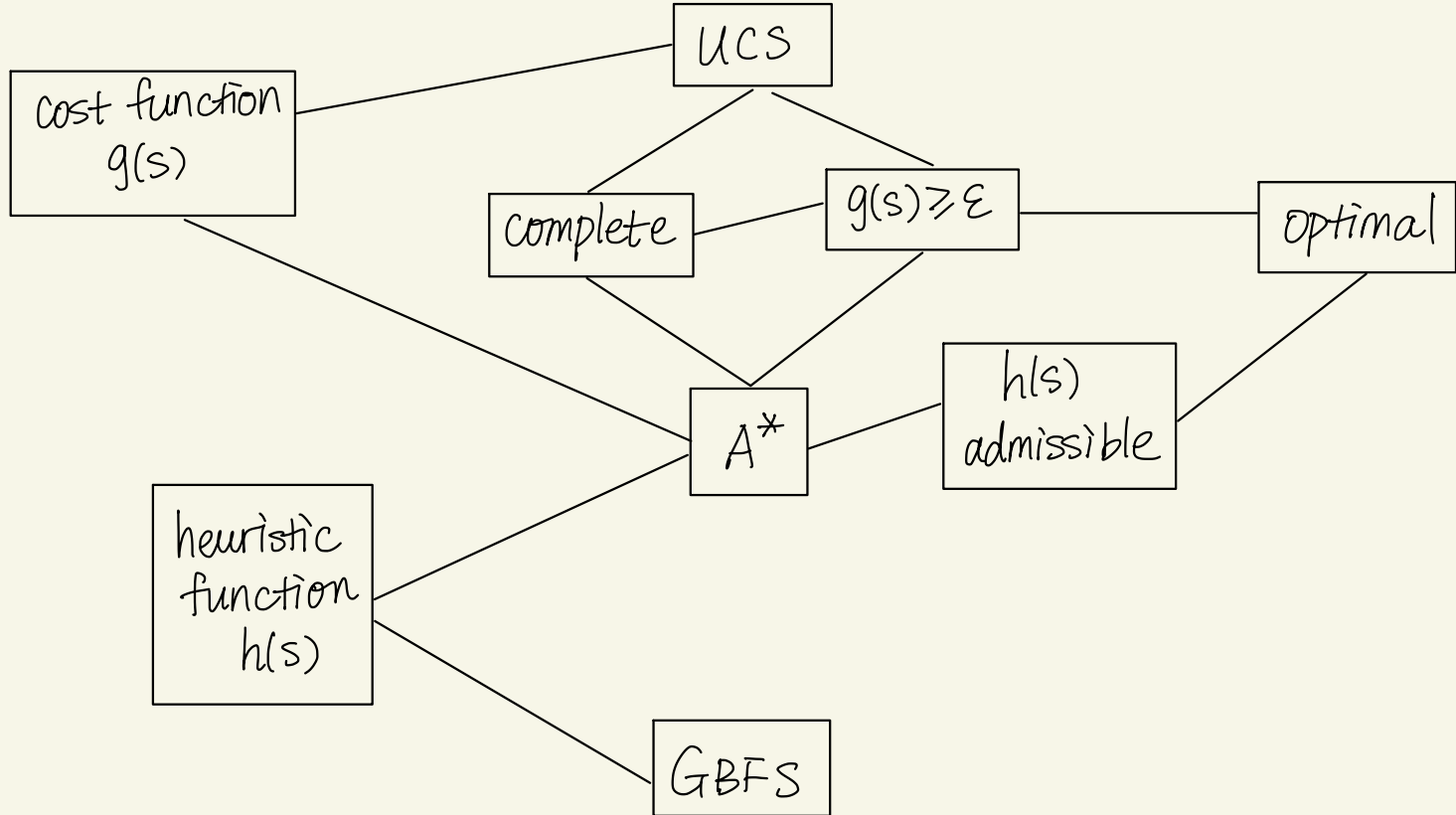
Uninformed v.s. Heuristic Search



Properties of BFS, DFS, and IDS



Properties of UCS, GBFS, A* Search.



Uninformed Search.
(Blind).

Depth - First Search.

- strategy : frontier is a stack (LIFO)
- complete? No.

stuck on infinite paths.

- optimal? No.

not optimizing costs.

- space : linear in m .

remembers one path.

- time : exponential in m .

visits the entire tree in the worst case.

Breadth-First Search.

- strategy: removes oldest state added.
- complete? Yes.
guaranteed to find shallowest goal node.
- optimal? No.
not optimizing costs.
- space: exponential in d .
of nodes on level d is exponential in d .
- time: exponential in d .
visits top d levels in the worst case.

Iterative - Deepening Search.

- strategy : for depth limit l from 0 to ∞ ,
perform DFS until depth limit l .
- complete? Yes. (\approx BFS)
terminates at level d .
- optimal? No. (\approx BFS & DFS)
not optimizing costs.
- space : linear in d . (\approx DFS)
remembers one path of length at most d .
- time : exponential in d . (\approx BFS)

Multiple-Path Pruning

- why? find a solution faster.
unnecessary to store multiple paths to same node.
- how?
use an explored set to remember visited states.
- effects on space complexity:
 - \uparrow space complexity to exponential
- effects on optimality:
 - UCS w/ multi-path pruning is still optimal.
 - A^* w/ multi-path pruning may not be optimal.

Uniform - Cost Search.

- strategy : remove path w/ smallest total cost.
- complete ? Yes.

edge cost is not arbitrarily small.

- optimal ? Yes.
- space : exponential in C^*/ϵ C^* : optimal path cost.
- time : ϵ : minimum edge cost.

Multi-path pruning.

why? only need to find one path to each state.

how? keep the first path found to each state.
discard all other paths found.

effects on algorithm properties.

- DFS : linear space \rightarrow exponential space.
- BFS : exponential space remains the same.
- UCS : still optimal.
- A* : may not be optimal depending on $h(n)$.

Heuristic Search

Greedy Best-First Search.

- strategy: expand state w/ smallest heuristic value.
- complete? No.
can get stuck in a cycle.
- optimal? No.
can return a suboptimal solution first.
- space and time: exponential.

1. Construct a search graph s.t. GBFS does not terminate.
2. Construct a search graph s.t. GBFS does not return the optimal solution.

A* Search

- strategy: expand state w/ smallest f value.

$$f(s) = \underset{\substack{\uparrow \\ \text{cost}}}{g(s)} + \underset{\substack{\uparrow \\ \text{heuristic}}}{h(s)}$$

- complete? Yes under mild conditions.
- optimal? Yes if $h(s)$ is admissible.
- space and time: exponential.

(A^* is optimal)

1. Construct a search graph w/ inadmissible heuristic and show that A^* search does not find the optimal solution.

or

(A^* is optimally efficient.)

2. Construct a search graph w/ admissible heuristic and show that

- ① UCS finds the optimal solution.
- ② A^* finds the optimal solution.
- ③ A^* visits fewer states than UCS.

Constructing Heuristics.

The cost of the optimal solution to the relaxed problem
= an admissible heuristic for the original problem.

Dominating Heuristics.

$h_1(n)$ dominates $h_2(n)$ iff

$$(1) \quad h_1(n) \geq h_2(n) \quad \forall n.$$

$$(2) \quad h_1(n) > h_2(n) \quad \exists n.$$

A^* with multi-path pruning

- if heuristic is **admissible** but **not consistent**, A^* w/ multi-path pruning is **not optimal**.

Definition of **Consistent Heuristic**.

- $h(n)$ is admissible, and
 - if there is an edge from n_1 to n_2 ,
 $h(n_1) - h(n_2) \leq g(n_1, n_2)$.
- if heuristic is **consistent**, A^* w/ multi-path pruning is **optimal**.

1. Given a search graph, verify that a heuristic function is admissible.
2. Given a search graph, verify that a heuristic function is consistent.
3. Construct a search graph and a heuristic that is admissible but not consistent, show that A^* w/ multi-path pruning is not optimal.