

Chap 4 – Informed Search



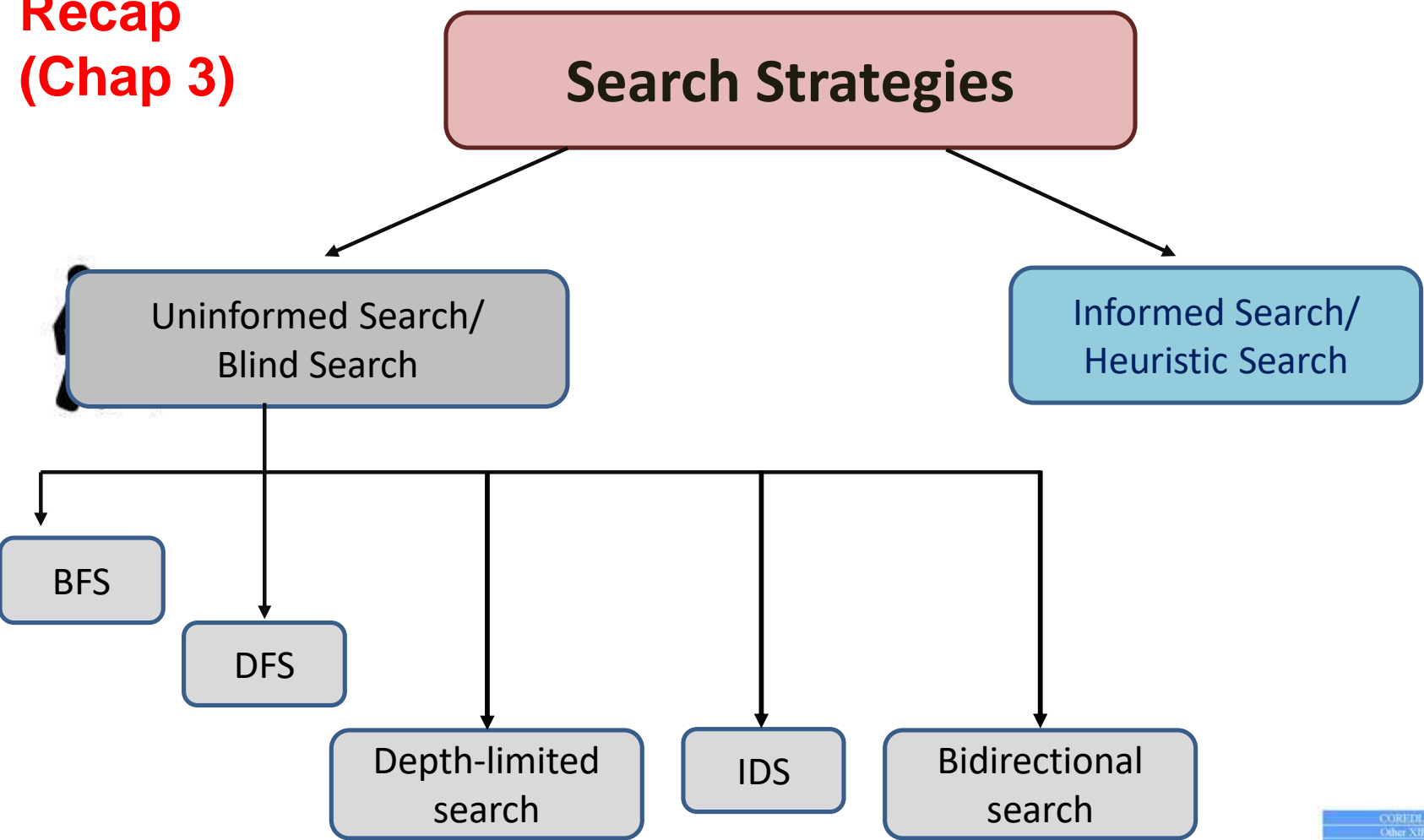
"I climb all this way, and you tell me *THAT'S* the meaning of life?!"



Today's Aims

- Define and evaluate informed (heuristic) search
- Heuristic Search Techniques
 - Hill climbing
 - Best-first Search
 - A* Search

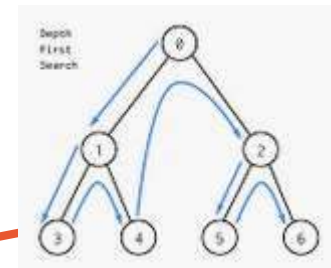
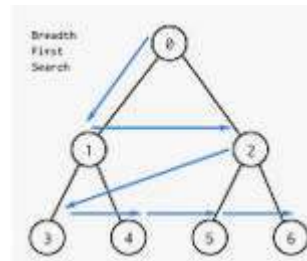
Recap
(Chap 3)





BFS vs. DFS

	BFS	DFS
Completeness	Complete	Complete (as long as not trapped in endless loop)
Optimality	Optimal (shortest path)	Not optimal
Time Complexity	Depends on the location of the goal node.	
Space Complexity	Consume more memory	Consume less memory





Search Strategies

Uninformed Search/
Blind Search



Informed Search/
Heuristic Search

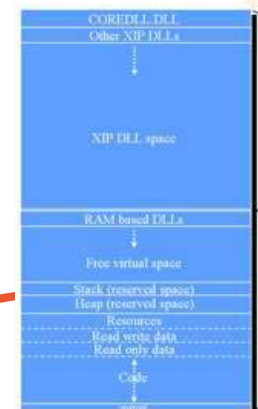
BFS

DFS

Depth-limited
search

IDS

Bidirectional
search

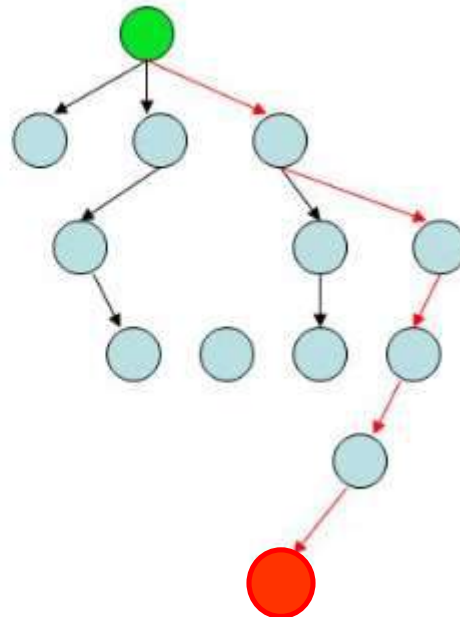




Heuristic

- A function that **ranks alternatives** at each branching step **based on available information** to decide which branch to follow [1].

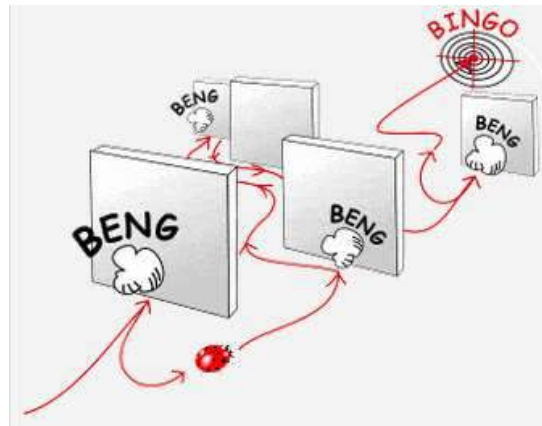
Idea: be **smart**
about what paths
to try.





Notes

- Design a heuristic often based on experience or intuition.



- Heuristic is only an informed guess of the next step to be taken – seldom able to predict the exact behavior of the state space.
- Can lead to a **suboptimal solution or fail**



With heuristic

- The agent is informed which branch(es) in a state space that is(are) **most likely** to lead to an acceptable problem solution.



Heuristic Function

- The heuristic function, $h(n)$ indicates how “close” a state n is to the goal.



Example

Let the heuristic function, $h(n)$, is given as follows:

$h(n)$ = estimated distance of state n from goal
= number of tiles at the **right place**.

* $h(n)$ determines the heuristic cost of a state

Initial:

$h = 4$

2	8	3
1	6	4
7		5



Final:

$h = 8$

1	2	3
8		4
7	6	5

Goal



Hill Climbing

- The simplest way in heuristic search
- 2 types:
 - Simple Hill Climbing
 - Steepest Ascent Hill Climbing





Limitations of Hill Climbing

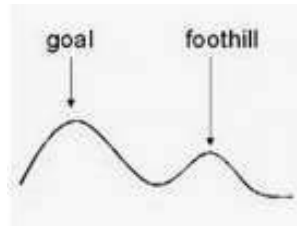
- It is less combinatorially explosive as it searches **locally** rather than globally;
- possibly very inefficient and ineffective .



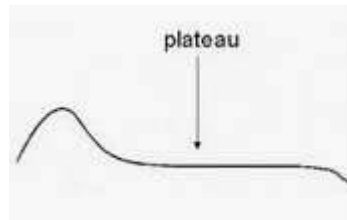
Limitations of Hill Climbing

- Hill climbing may encounter **local maximum**, which are

1. Foothills



2. Plateau



3. Ridges





Foothill

Foothill - a state that has no better move.

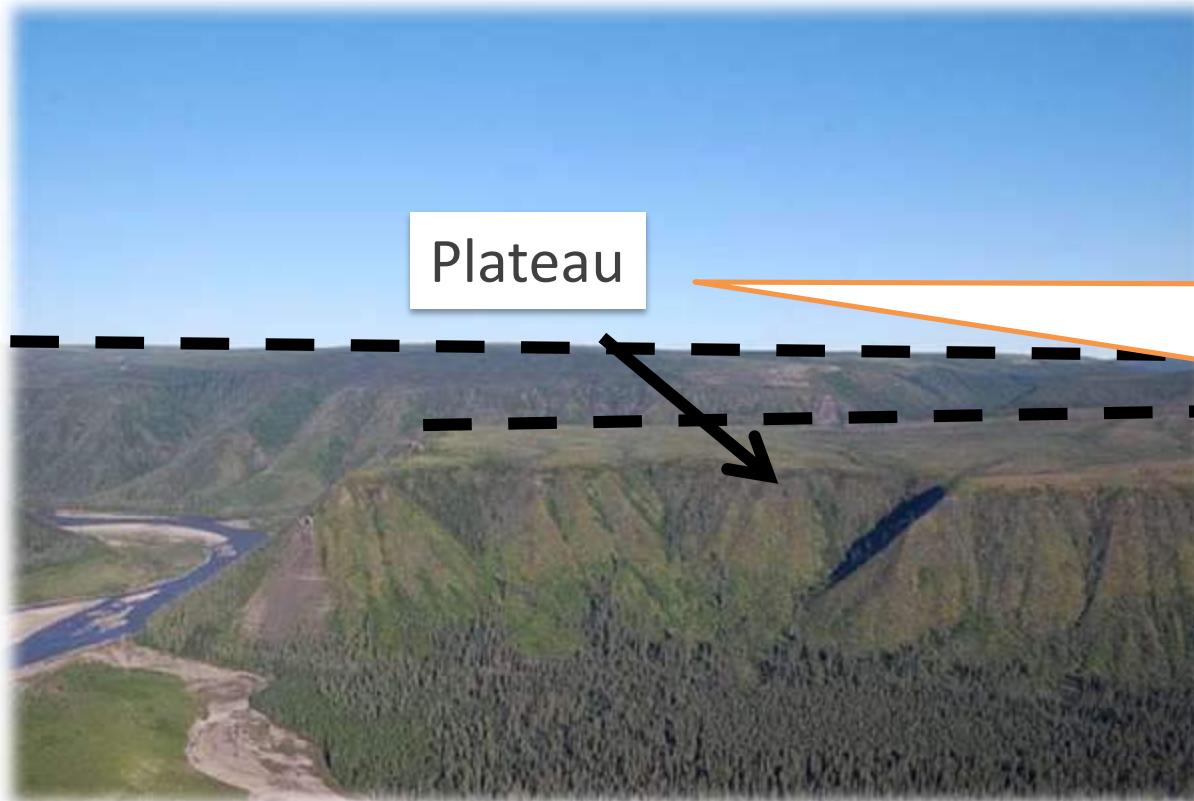


A simple hill climber will never find a solution though one exists.



Plateau

An area of state space where the heuristic values are **equal**.



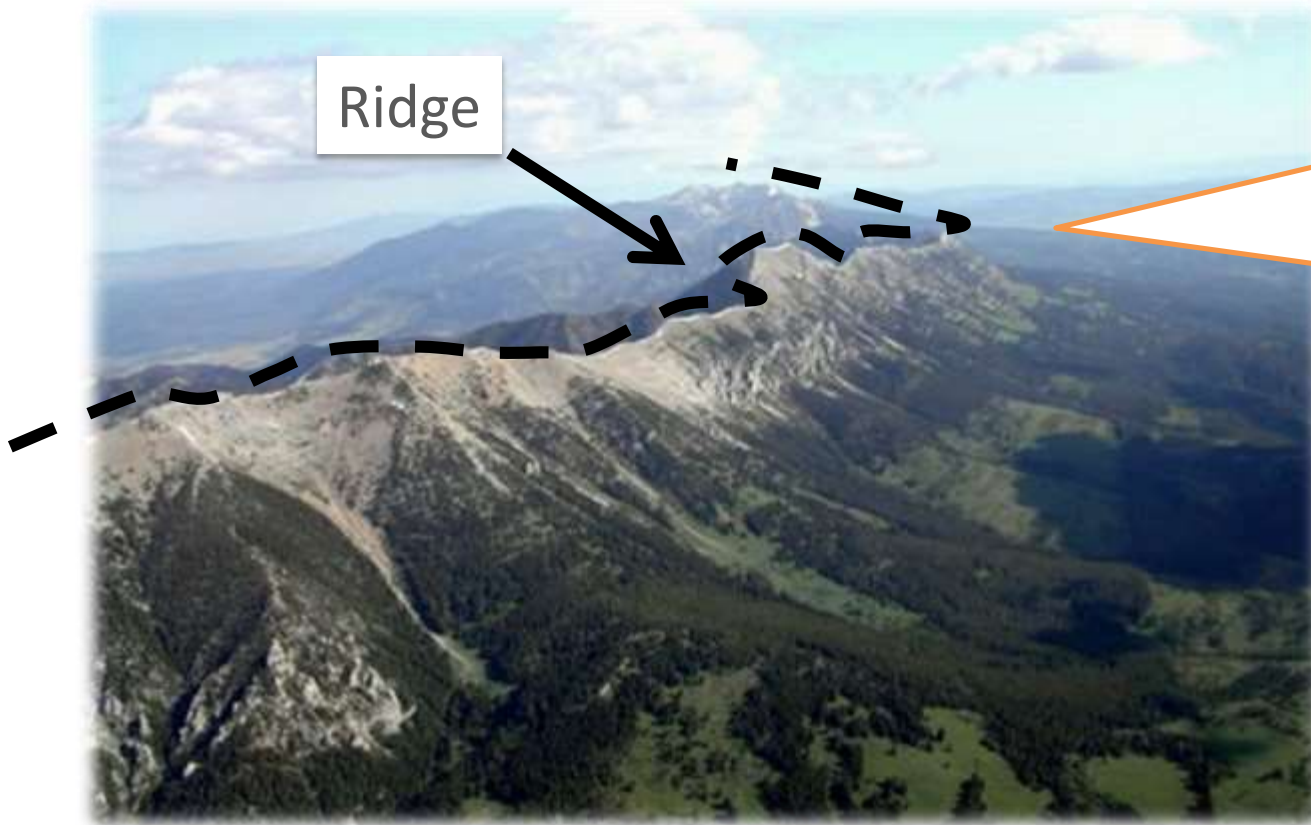
A hill climber will not know which direction to follow

To proceed, the search algorithm may conduct a random walk.



Ridge

An area of state space bordered by lower values in an descending search (or higher values in ascending search)



the ridge may slope very slowly toward the peak making it a **long route**.



Simple Hill Climbing

ALGORITHM

1. Set initial state to current
2. Loop on the following until goal is found or no more operators available
 - ▶ Select an operator and apply it to create a new state
 - ▶ Evaluate new state
 - ▶ If new state is better than current state, perform operator making new state the current state
3. Once loop is exited, either we have found the goal or return fail



Example

Let the heuristic function, $h(n)$, is given as follows:

$h(n)$ = estimated distance of state n from goal
= number of tiles at the **right place**.

* $h(n)$ determines the heuristic cost of a state

Initial:

$h = 4$

2	8	3
1	6	4
7		5



Final:

$h = 8$

1	2	3
8		4
7	6	5

Goal

Simple Hill Climbing

Example
1

$h = 4$

2	8	3
1	6	4
7		5

$h = 8$

1	2	3
8		4
7	6	5

Goal

$h = 5$

2	8	3
1		4
7	6	5

all states leading
from it have no
better values

$h = 5$

2		3
1	8	4
7	6	5

$h = 4$

2	8	3
1	4	
7	6	5

$h = 5$

2	8	3
	1	4
7	6	5



Ops! We're stuck!

- Simple hill climbing has faced a dead end.
- The child nodes are worse than the parent node.
- This problem is known as **foothill**.

Simple Hill Climbing

Example
2

$h = 6$

2		3
8	1	4
7	6	5

$h = 8$

1	2	3
8		4
7	6	5

Goal

$h = 7$

	2	3
8	1	4
7	6	5

Foothill

$h = 6$

8	2	3
	1	4
7	6	5



Steepest Ascend Hill Climbing

- Here, we attempt to improve on the previous Hill Climbing algorithm
 - Given initial state perform the following
 1. Set initial state to current
 2. Loop on the following until goal is found or a complete iteration occurs without change to current state
 - Generate all successor states to current state
 - Evaluate all successor states using heuristic
 - Select the successor state that yields the highest heuristic value and perform that operator

SAHC

Example
1

$h = 2$

2	6	3
1	8	4
7		5

$h = 8$

1	2	3
4	5	6
7	8	

Goal

$h = 3$

2	6	3
1		4
7	8	5

$h = 1$

2	6	3
1	8	4
	7	5

$h = 2$

2	6	3
1	8	4
7	5	

2		3
1	6	4
7	8	5

$h = 3$

2	6	3
1	4	
7	8	5

$h = 3$

2	6	3
	1	4
7	8	5

$h = 3$

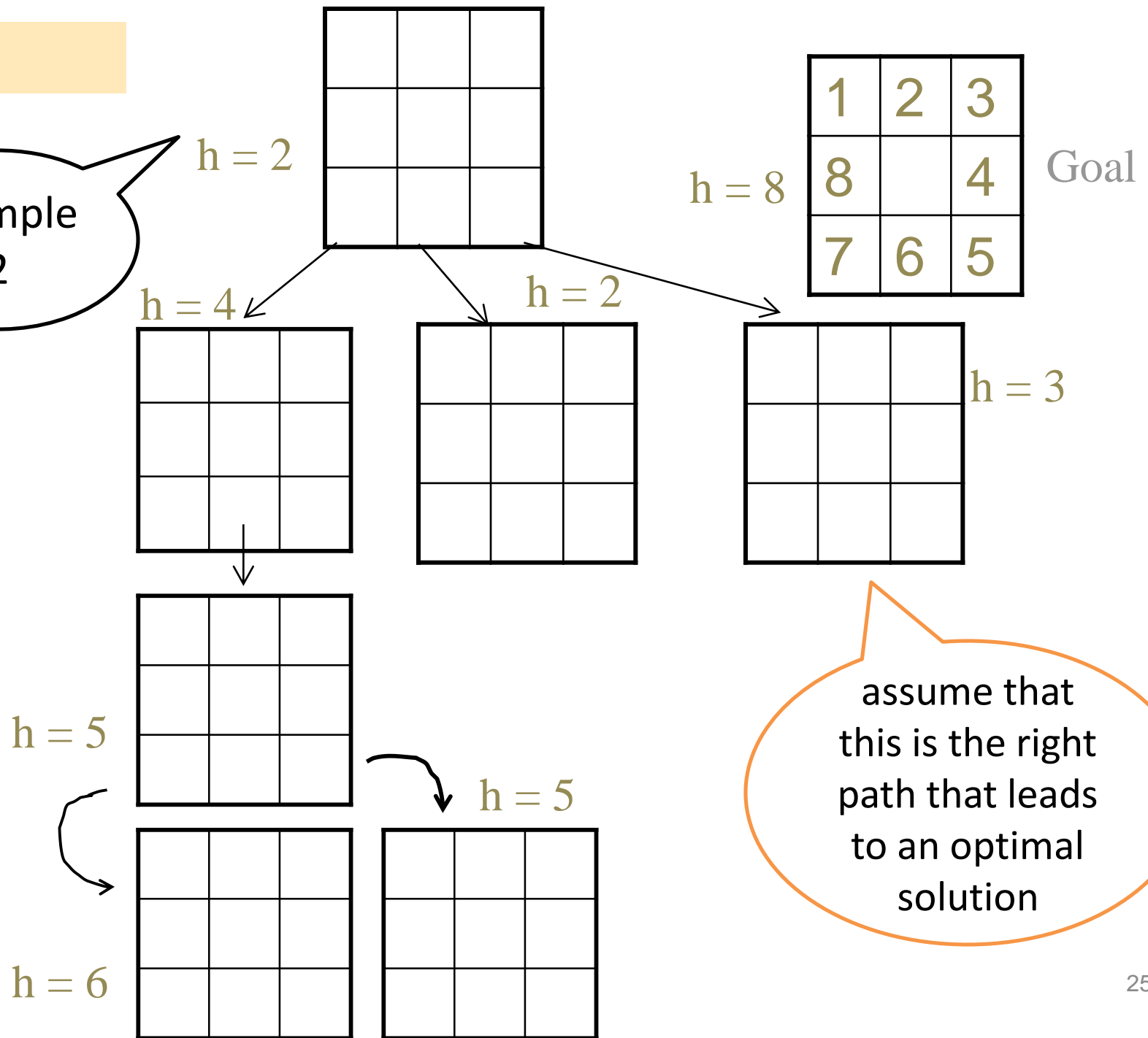


Ops! We're stuck again!

- SAHC has also faced a dead end.
- All heuristic costs are the same.
- This problem is known as **plateau**.

SAHC

Example
2

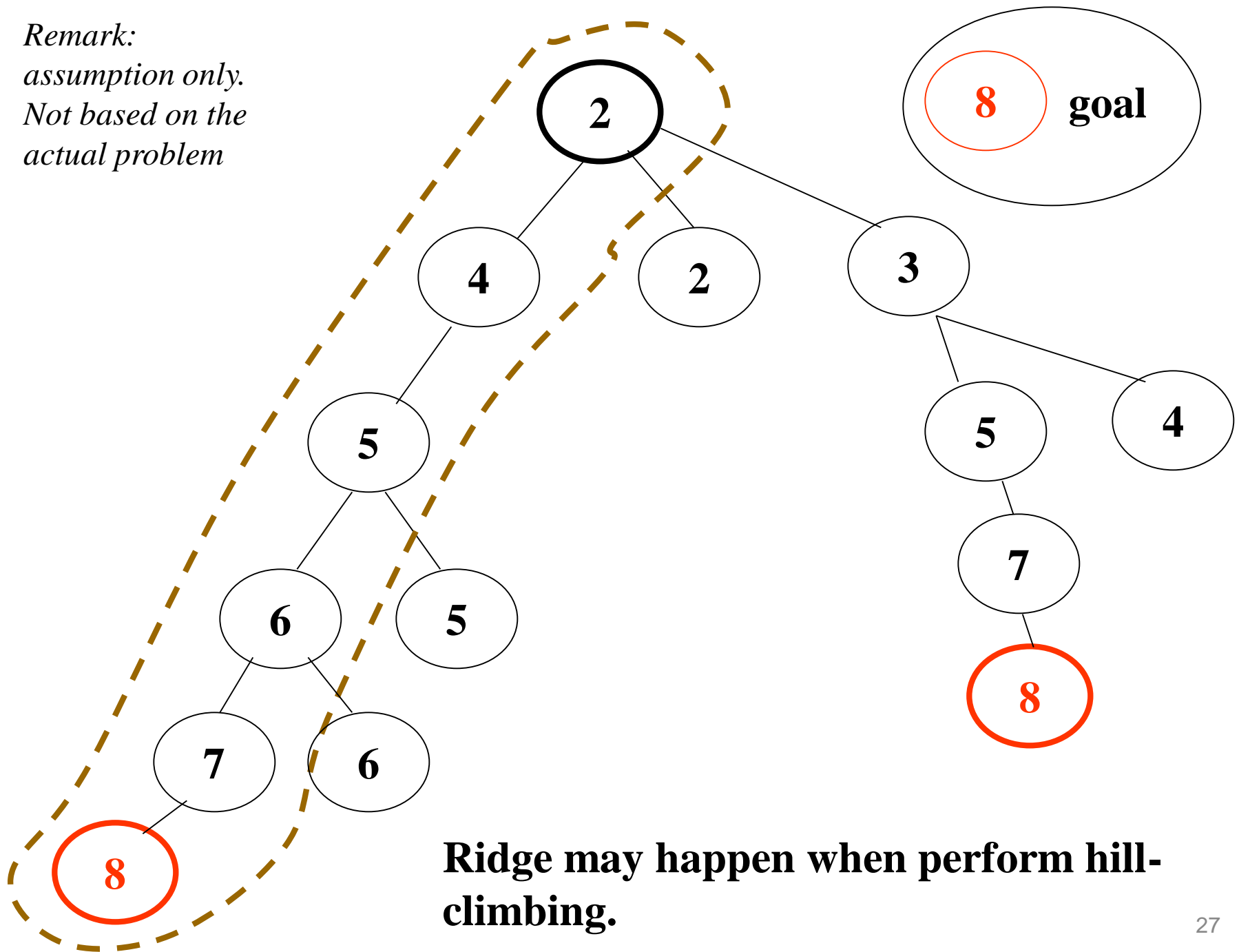




Ops! It took the wrong path!

- The actual path that the search took in Example 2 leads to a sub-optimal solution (longer path).
- The problem is known as **ridge**.

*Remark:
assumption only.
Not based on the
actual problem*





Designing a Heuristic Function

- **Local Heuristic**
 - less combinatorially explosive (less complex)
 - possibly very inefficient and ineffective
- **Global Heuristic**
 - Consider the global information
 - possibly more efficient than local heuristic

There is **NO GUARANTEE** that a heuristic can always produce a solution/optimal solution



Heuristic Search

Simple Hill Climbing

vs

Steepest Ascent Hill Climbing



- Similarity?

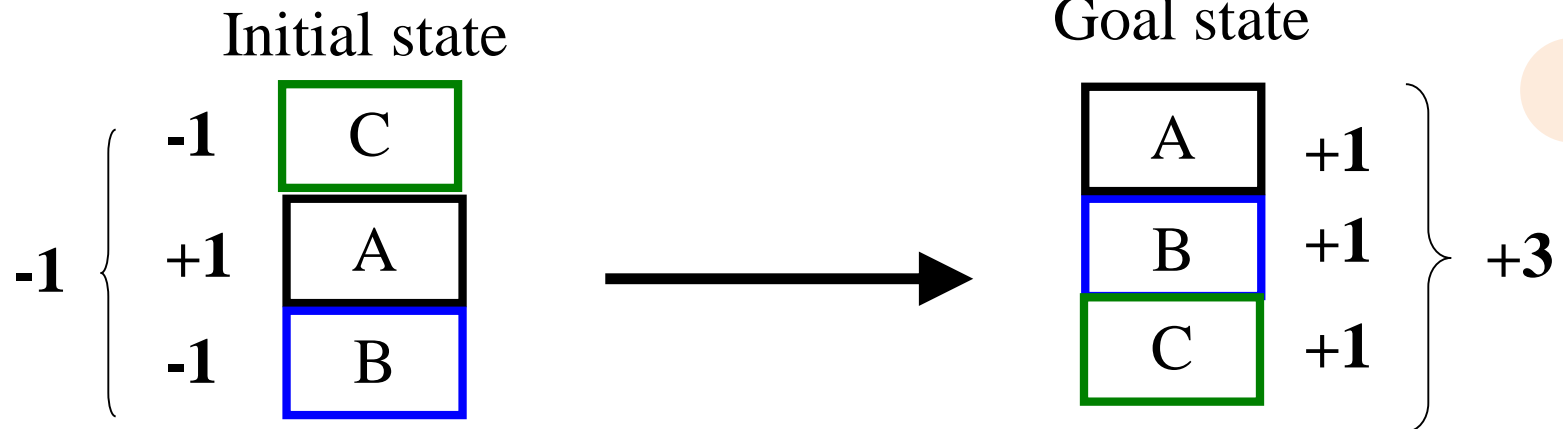
❖ Global information might be encoded in heuristic functions.



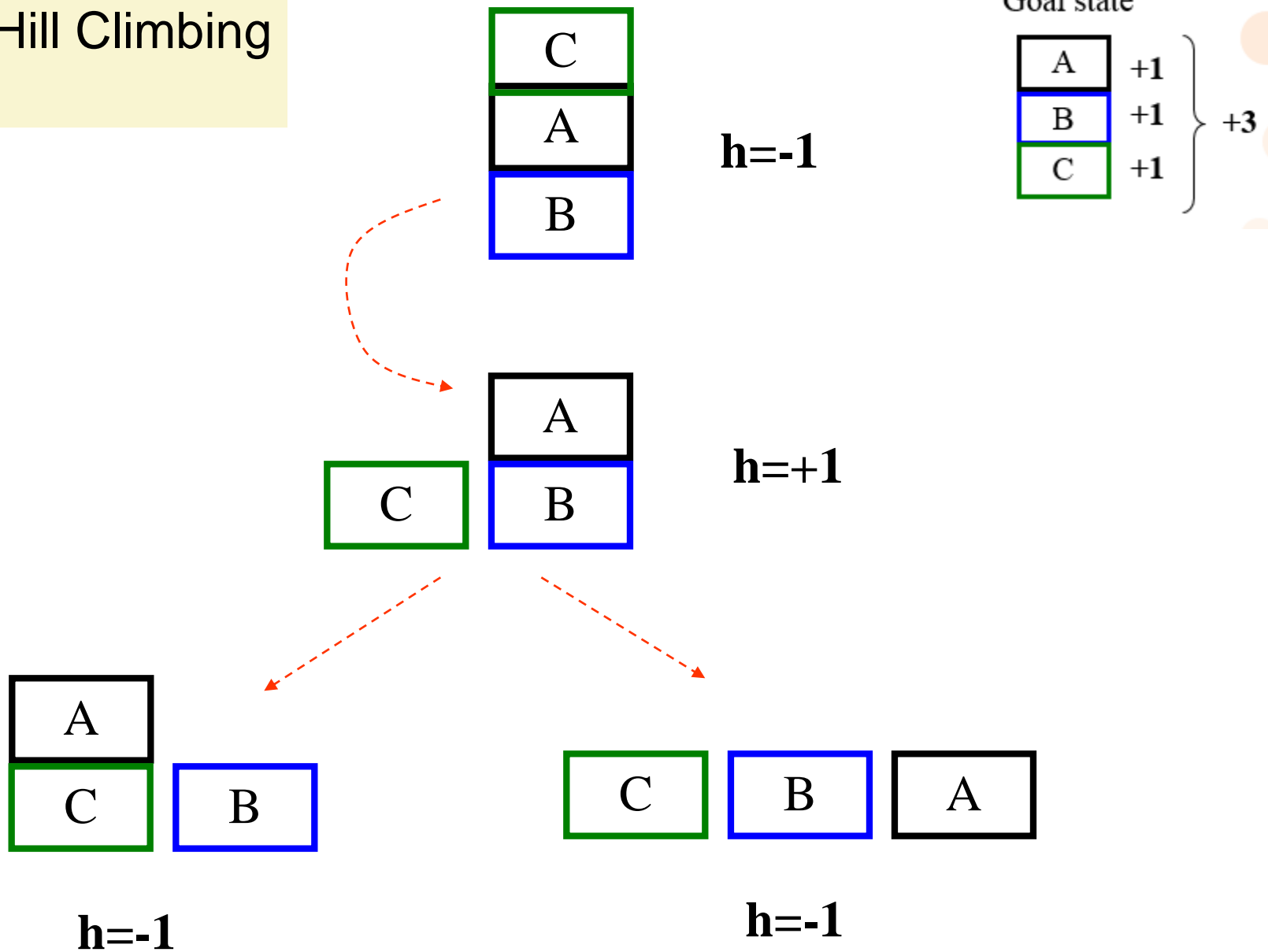
Local Heuristic

- Heuristic: IF the block is **rest on the right place**, Add 1. ELSE Minus 1.

👍: Easy to design 😊



Simple Hill Climbing SAHC





Ops! We are stuck!

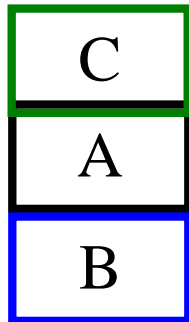
- 👎: It is proven local heuristic is ineffective and inefficient 😞



Global Heuristic

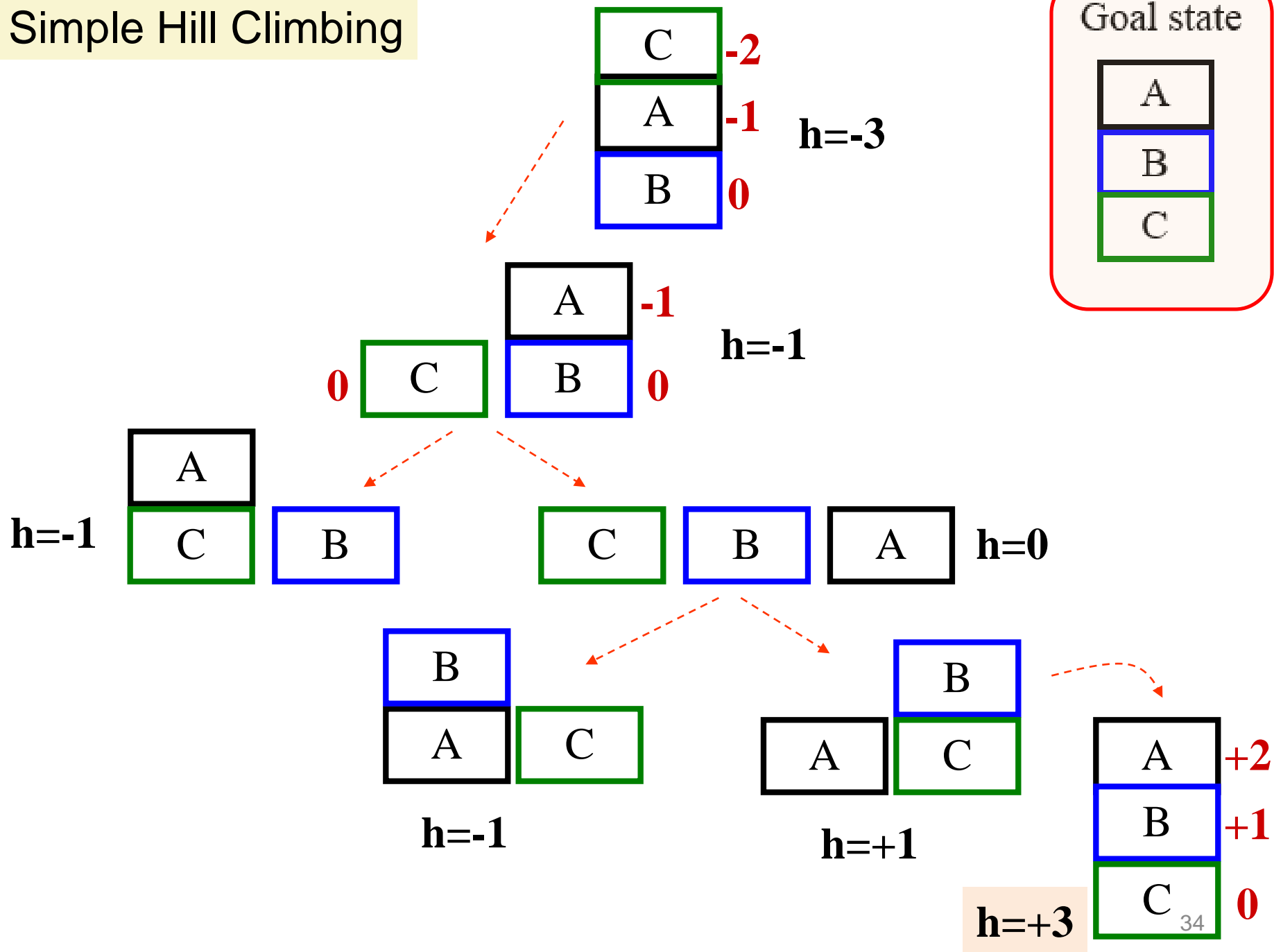
- Heuristic:

1. For each block that has **correct** support structure, **+ 1 for every block** in the support structure.
2. For each block that has an **incorrect** support structure, **- 1 for each block** in the each block in the existing structure.

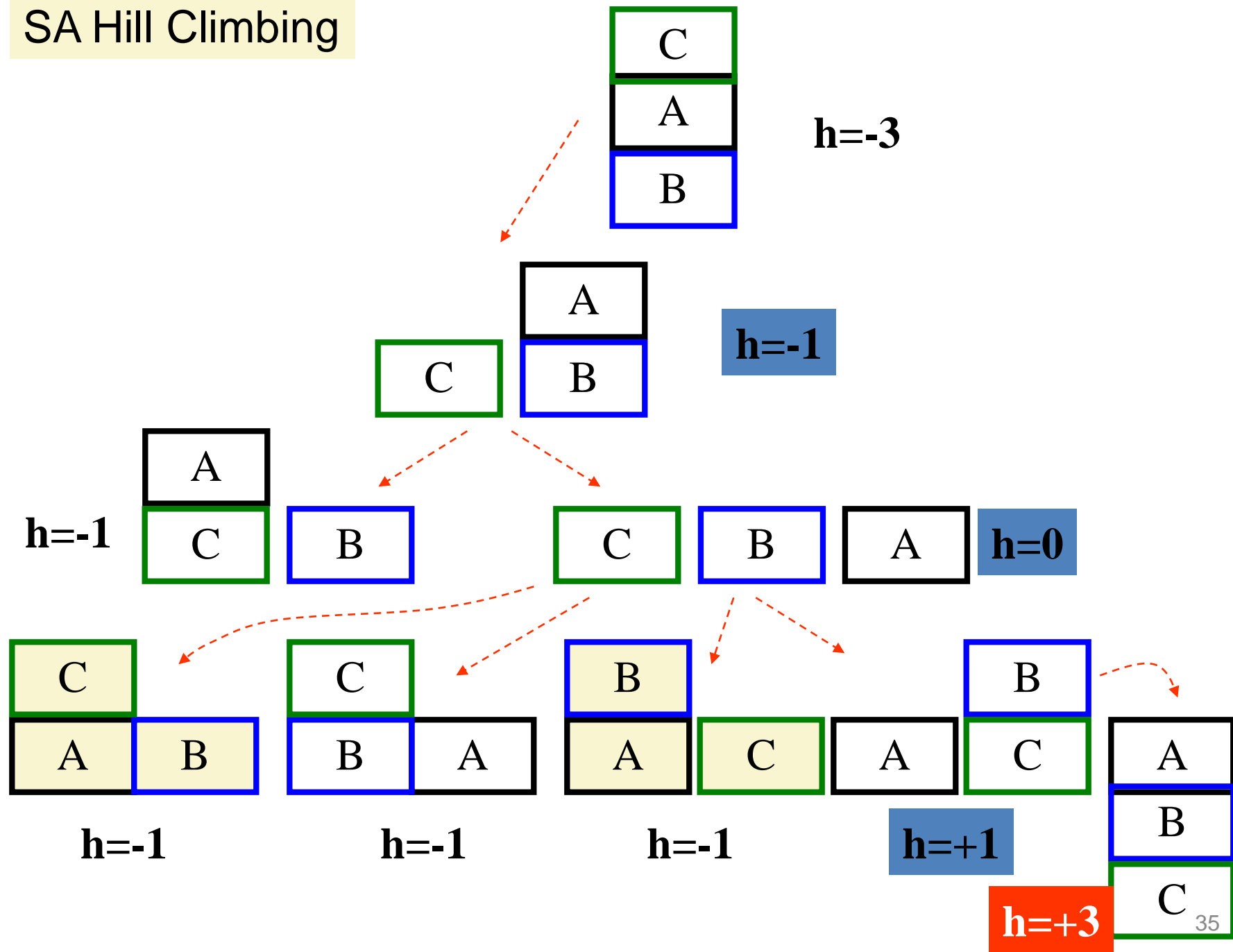


- The support for C is A & B
- The support for A is B
- The support for B is table, but the table is not a block

Simple Hill Climbing



SA Hill Climbing

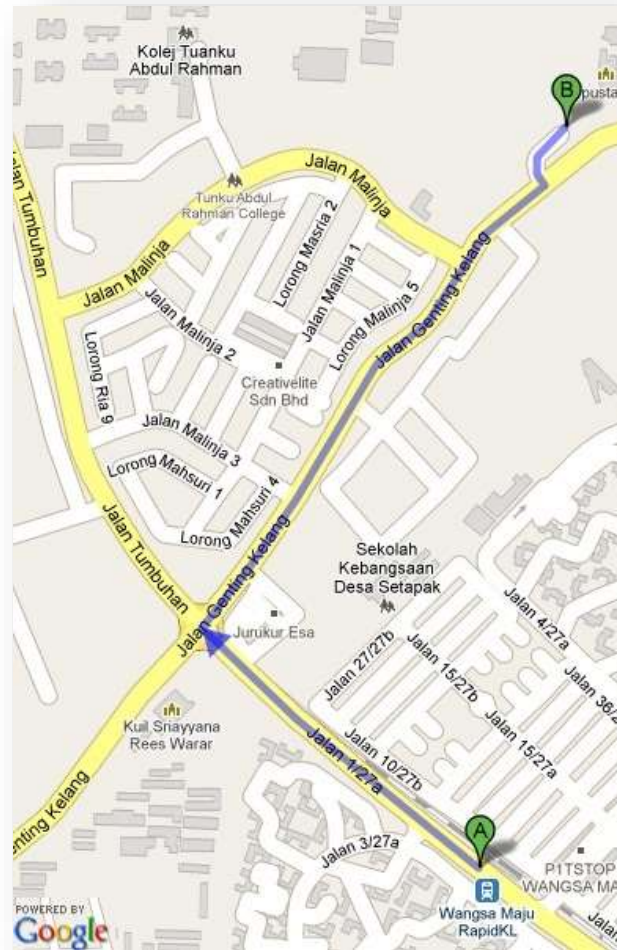




Conclusion

- Hill climbing is not always effective
- Heuristic function not always perfect
- Modify the heuristic function
- Inefficient in a large, rough problem space

Best First Search





Recall

- Depth-first search
 - 👍 without traverse all the branches.
- Breadth-first search
 - 👍 Not trapped into a dead-end path.

Best-first search =

👍 depth-first search + 👍 breadth-first search



Best First Search

- More flexible
- Use priority queue
- Heuristic is used (to **estimate** the **promising state/path**)



Best-first Search Algorithm

- ▶ Set $Open := [Start]$ and $closed := []$
 - ▶ Loop until $open := []$
 - ▶ Remove the LEFTMOST node X from $open$
 - ▶ If $X = goal$ then return the path from $Start$ to X
 - ▶ Else generate children of X
 - ▶ For each child of X
 - ▶ If the child is on *open* or *closed*
 - ☐ Assign the child its heuristic value
 - ☐ Add the child to *open*
 - ▶ Else if the child is on *open*
 - ☐ If the child was reached by a shorter path
 - ☐ Give the child a state on *open* the shorter path
- ...CONT NEXT PAGE



Best-first Search Algorithm -CONT

- ▶ Else if the child is already on closed
 - ▶ If the child was reached by a shorter path
 - ▶ Remove the state from closed
 - ▶ Add the child to open
- ▶ Put X on closed
- ▶ Sort states on open by heuristic merit (best leftmost)

Return fail

After loop till open = []

[Back to exercise](#)



heuristic

- Let:
- $h(n)$ = estimated distance of state n from goal
- = number of tiles **at the wrong place**.
- * $h(n)$ = heuristic cost
- Initial:

2	8	3
1	6	4
7		5



Final:

1	2	3
8		4
7	6	5

• Open=[a4]; Closed=[];

[Closed] = 4

2	8	3
1	6	4
7		5

a

$h = 0$

1	2	3
8		4
7	6	5

Goal

[Closed]
 $h = 3$

2	8	3
1		4
7	6	5

b

[Open]
 $h = 5$

2	8	3
1	6	4
	7	5

c

[Open]
 $h = 5$

2	8	3
1	6	4
7		5

d

Open=[b3,c5,d5]

Closed=[a4];

Open=[e3,g3,f4,c5,d5]

Closed=[a4,b3];

e

2	8	3
	1	4
7	6	5

[Closed]
 $h = 5$

2	8	3
1	4	
7	6	5

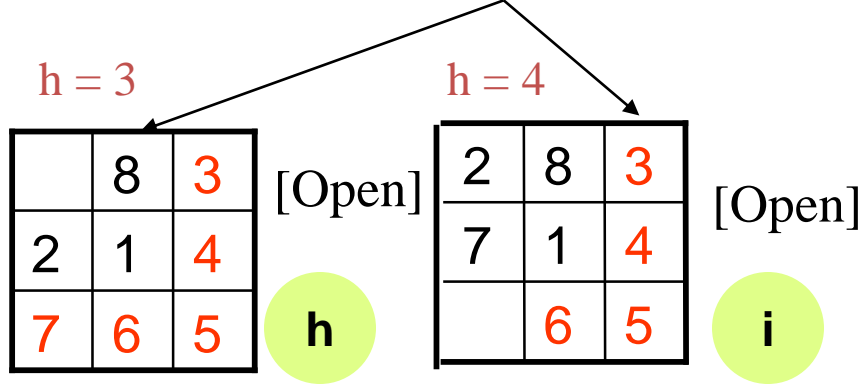
[Open]
 $h = 4$

f

2		3
1	8	4
7	6	5

[Open]
 $h = 3$

g



Open=[g3,h3,f4,i4,c5,d5]

Closed=[a4,b3,e3]

[Closed] = 4

2	8	3
1	6	4
7		5

a

$h = 0$

1	2	3
8		4
7	6	5

Goal

[Closed]

$h = 3$

2	8	3
1		4
7	6	5

b

[Open]

$h = 5$

2	8	3
1	6	4
	7	5

c

[Open]

$h = 5$

2	8	3
1	6	4
7	5	

d

e

2	8	3
	1	4
7	6	5

[Closed]

$h = 3$

2	8	3
1	4	
7	6	5

[Open]

$h = 4$

f

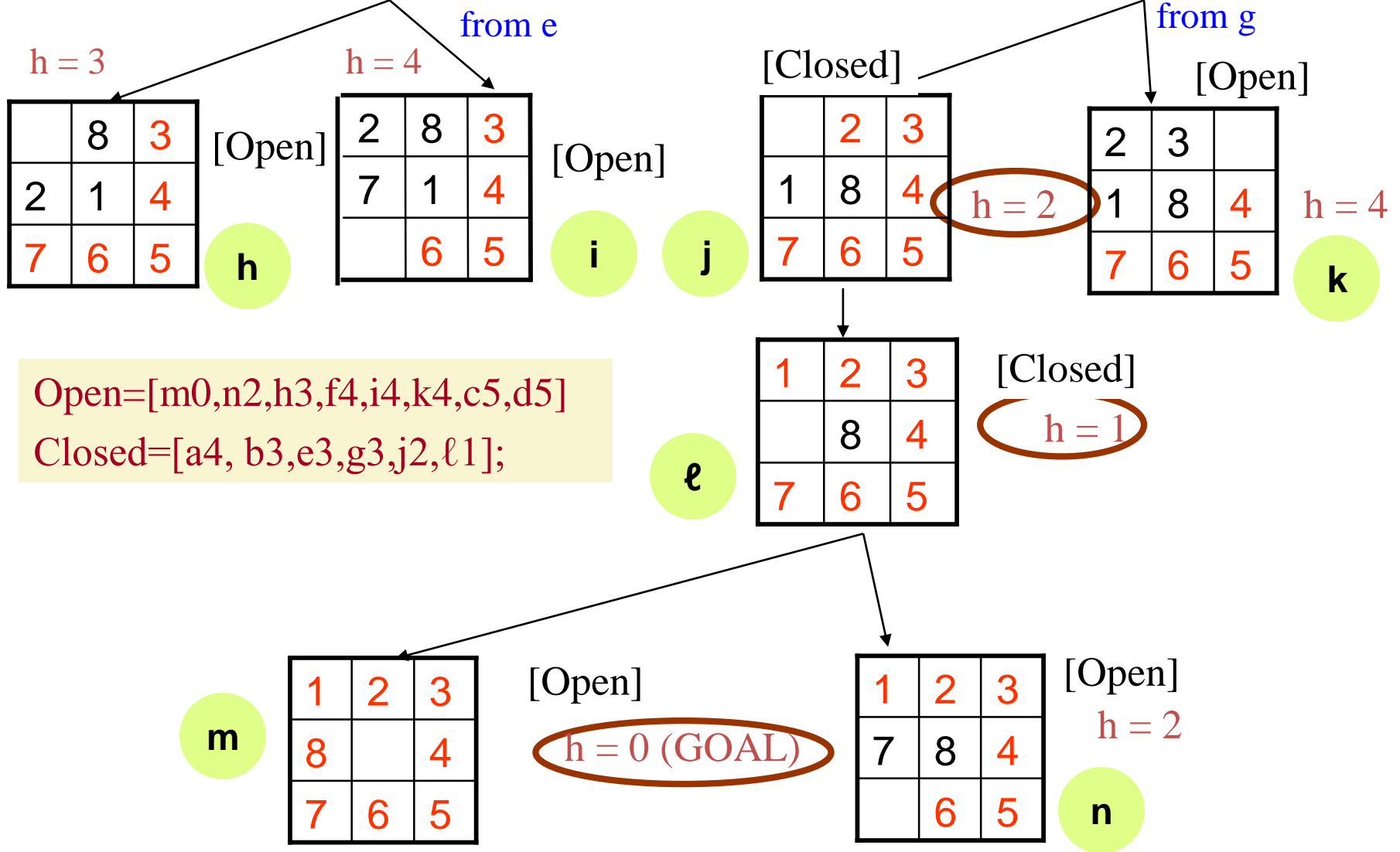
2		3
1	8	4
7	6	5

[Open]

$h = 3$

g

45



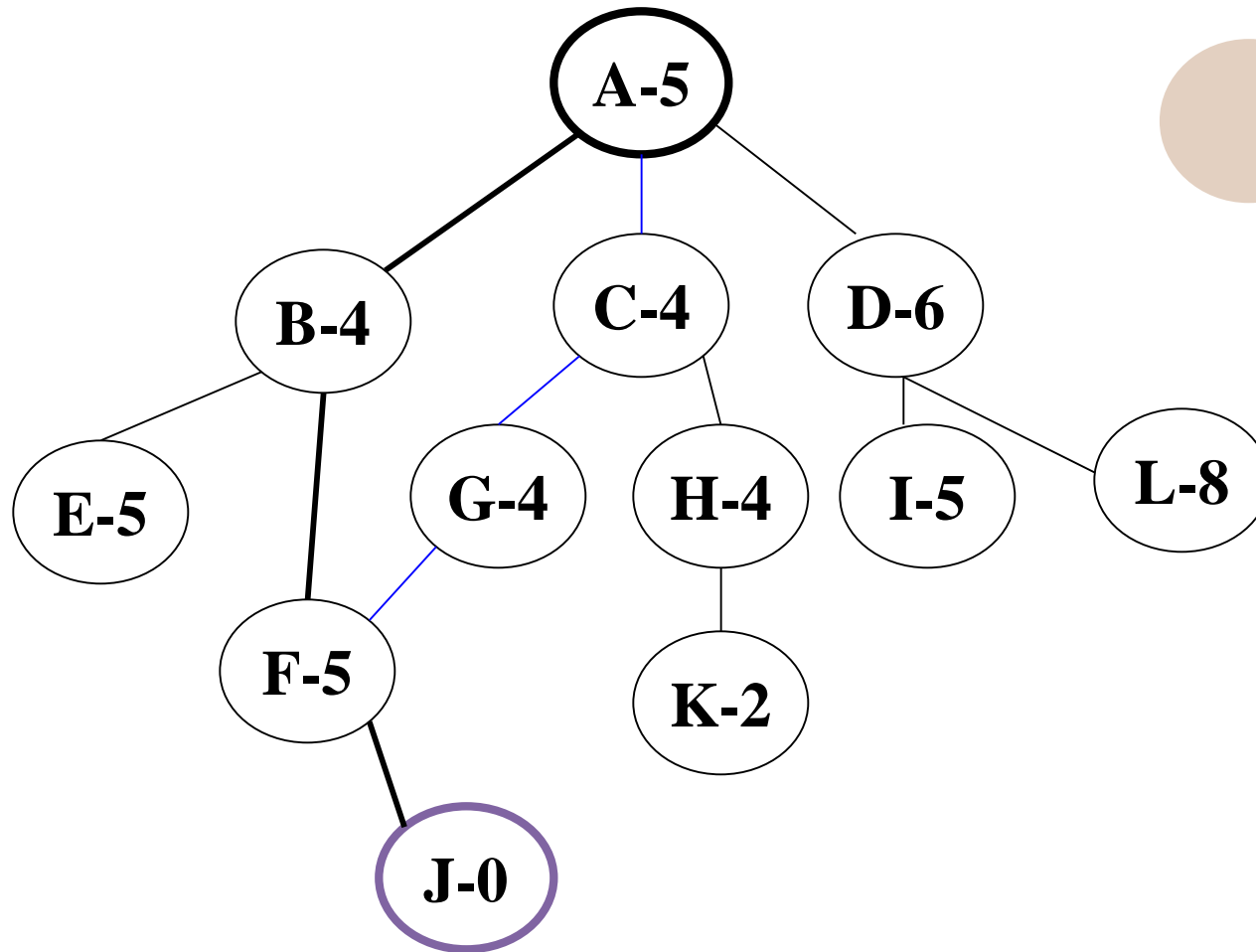


Open & Closed List

- Open=[a4]; Closed=[];
- Open=[b3,c5,d5]; Closed=[a4];
- Open=[e3,g3,f4,c5,d5]; Closed=[a4,b3];
- Open=[g3,h3,f4,i4,c5,d5]; Closed=[a4,b3,e3];
- Open=[j2,h3,f4,i4,k4,c5,d5]; Closed=[a4,b3,e3,g3];
- Open=[ℓ1,h3,f4,i4,k4,c5,d5];
Closed=[a4, b3,e3,g3,j2];
- Open=[m0,n2,h3,f4,i4,k4,c5,d5];
Closed=[a4, b3,e3,g3,j2,ℓ1];
- Evaluate m0; the solution is found; return path = a->b->g->j->ℓ->m



Exercise



[See algorithm](#)



Trace

- Open=[a5]; Closed=[];
- Eval A-5: Open=[b4,c4,d6]; Closed=[a5];
- Eval B-4: Open=[c4,e5,f5,d6]; Closed=[b4,a5];
- Eval C-4: Open=[g4,h4,e5,f5,d6]; Closed=[c4,b4,a5];
- Eval G-4: Open=[h4,e5,f5*,d6]; Closed=[g4,c4,b4,a5];
- Eval H-4: Open=[k2, e5,f5*,d6]; Closed=[h4, g4,c4,b4,a5];
- Eval K-2: Open=[e5,f5*,d6]; Closed=[k2,h4, g4,c4,b4,a5];
- Eval E-5: Open=[f5*,d6]; Closed=[e5,k2,h4, g4,c4,b4,a5];
- Eval F-5: Open=[j0,d6]; Closed=[f5*,e5,k2,h4, g4,c4,b4,a5];
- Eval j-0: Goal is found! Return path = a->b->f->j
- * f5 was reached by B-4



A* Search (Luger, pg 139)

Heuristic Function:

- $f(n) = g(n) + h(n)$

where:

$g(n)$ = distance n from the start state and
 $g(\text{start state}) = 0$

$h(n)$ = number of tiles **out of place**, hence

$h(\text{goal}) = 0$ $(h_1(n))$ **OR**

$h(n)$ = the sum of the Manhattan distances of the
tiles from their goal positions, hence

$h(\text{goal}) = 0$ $(h_2(n))$

look for minimum($f(n)$) and $h(n) = 0$

(h_1)

$g(n) = 0$

[Closed] $f(a) = 4$

2	8	3
1	6	4
7		5

$h(\text{goal}) = 0$

1	2	3
8		4
7	6	5

Goal

[Closed]

$f(b) = 4$

2	8	3
1		4
7	6	5

[Open]

$f(c) = 6$

2	8	3
1	6	4
	7	5

[Open]

$f(d) = 6$

2	8	3
1	6	4
7	5	

$g(n)=1$

$g(n)=2$

2	8	3
	1	4
7	6	5

[Closed]

$f(e)=5$

2	8	3
1	4	
7	6	5

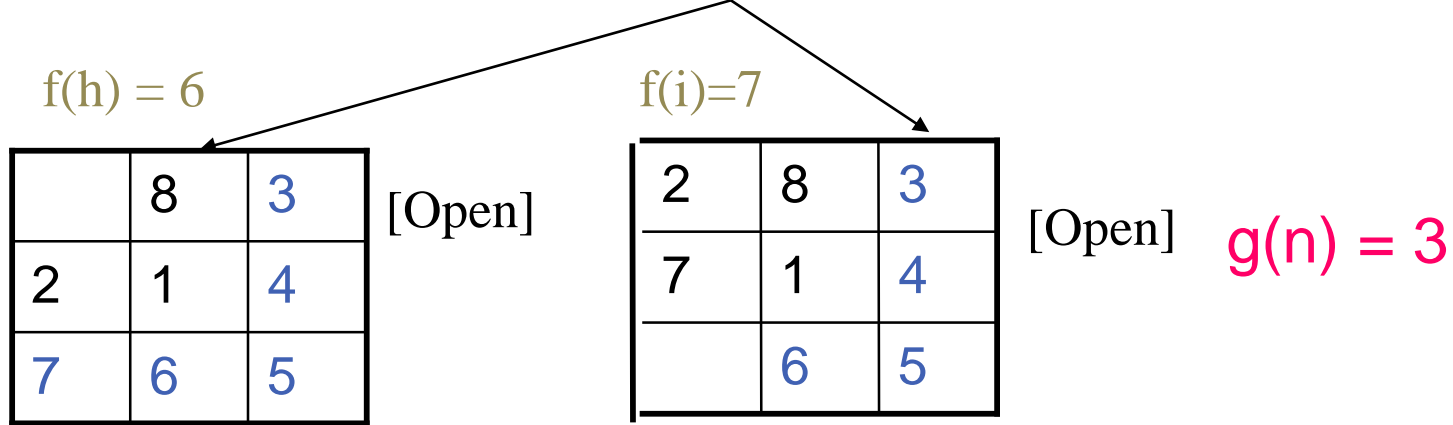
[Open]

$f(f)=6$

2		3
1	8	4
7	6	5

[Open]

$f(g)=5$



Open=[g5,c6,d6,f6,h6, i7]

Closed=[e5,b4,a4];

The same process will
be iterated until the
goal is found

(h_2)

$g(n) = 0$

$f(a) = 5$

2	8	3
1	6	4
7		5

$h(\text{goal}) = 0$

1	2	3
8		4
7	6	5

Goal

$f(b) = 5$

2	8	3
1		4
7	6	5

$f(c) = 7$

2	8	3
1	6	4
	7	5

$f(d) = 7$

2	8	3
1	6	4
7	5	

$g(n) = 1$

The same process will be iterated until the goal is found

$g(n) = 2$

2	8	3
	1	4
7	6	5

$f(e) = 7$

2	8	3
1	4	
7	6	5

$f(f) = 7$

2		3
1	8	4
7	6	5

$f(g) = 5$



Differences between SAHC and BFS

SA Hill-climbing

- ▶ one move is selected and all others are rejected, which will not be reconsidered.
- ▶ will quit if there is no better successor /children state than the current state.

Best-first search

- ▶ one move is selected, but others are still kept around so that they can be revisited later.
- ▶ will try on other nodes which initially was less promising

Summary

Search Strategies

Uninformed Search/
Blind Search



Informed Search/
Heuristic Search

Hill climbing

Simple Hill
Climbing

Steepest
Ascent Hill
Climbing

Best-first
Search

A* Search

$$f(n) = g(n) + h(n)$$

$(h_1(n))$ OR

$(h_2(n))$

Difference?

Similarity? Limitations?

Modify the heuristic function
(Local & Global Heuristic)

Next Week

Knowledge representation