

CHAPTER II

Problem-Solving & Search Algorithms



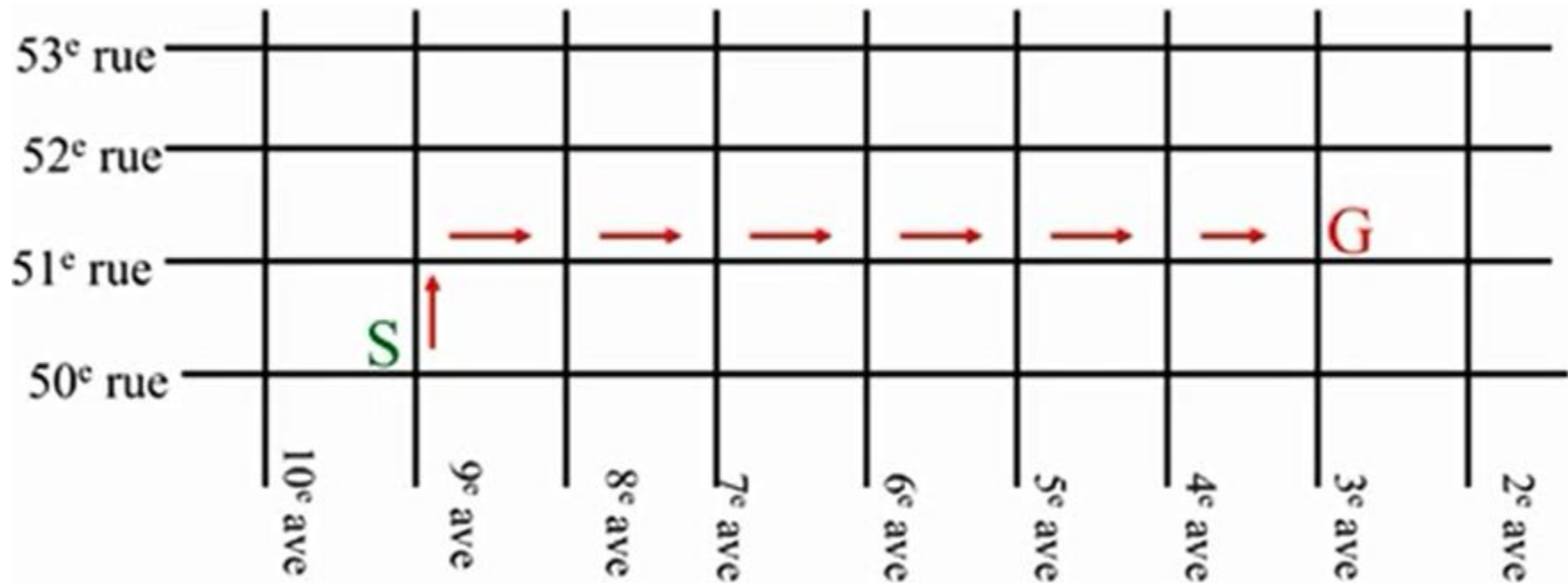
Solving a problem

- **Intuitive steps by a human**
 - Model the current situation
 - List possible solutions
 - Evaluate the value of each solution
 - Select the best option satisfying the goal
- **How to efficiently browse the list of solutions ?**
- **Several problems can be solved by searching in a graph :**
 - Each node represents a state of the environment
 - Each path through a graph represents a sequence of actions
 - The solution: simply look for the path that best satisfies our performance measurement

Problem-Solving

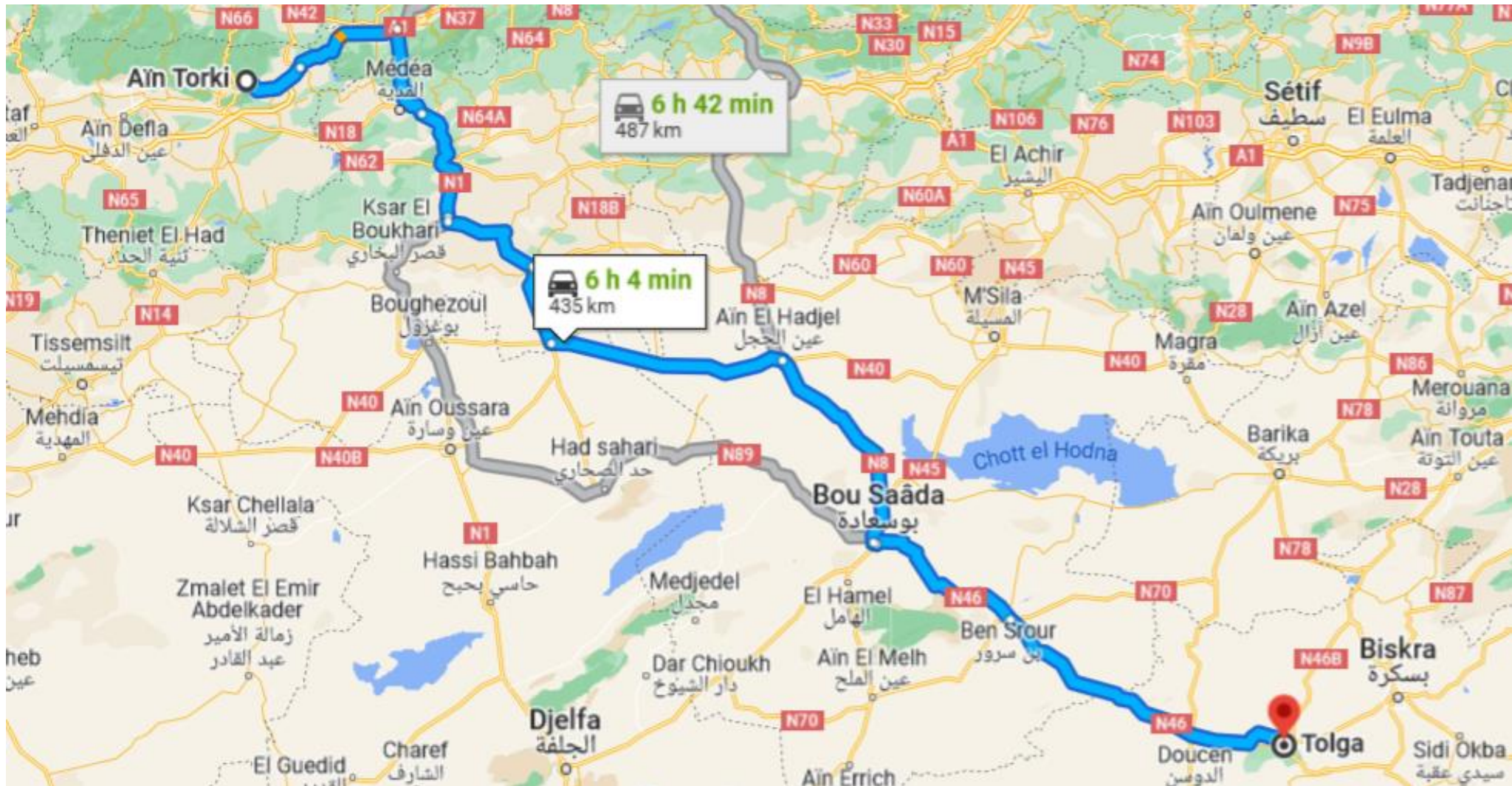
Example: Path-finding in a city

Find the best path between the 9th ave – 50th street to the 3rd ave -51st street



Problem-Solving

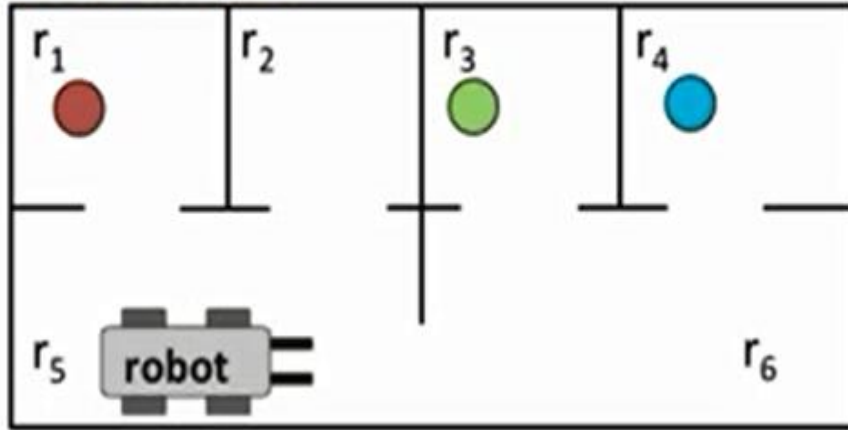
Example: Google Maps



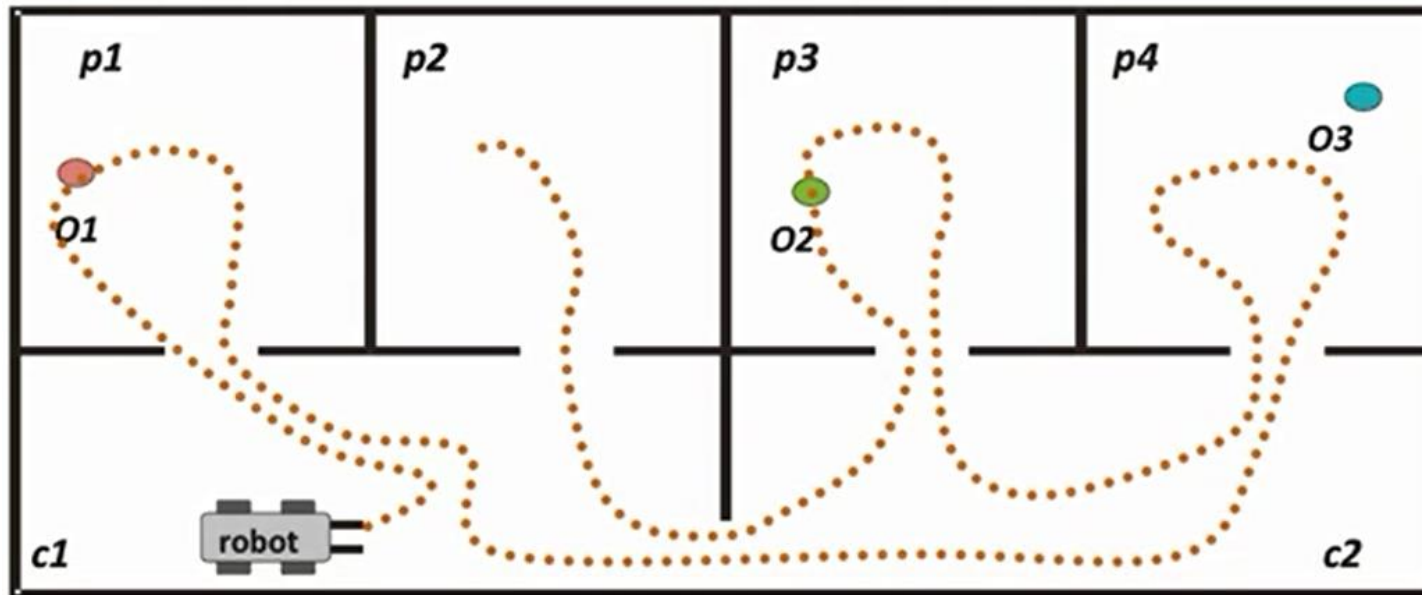
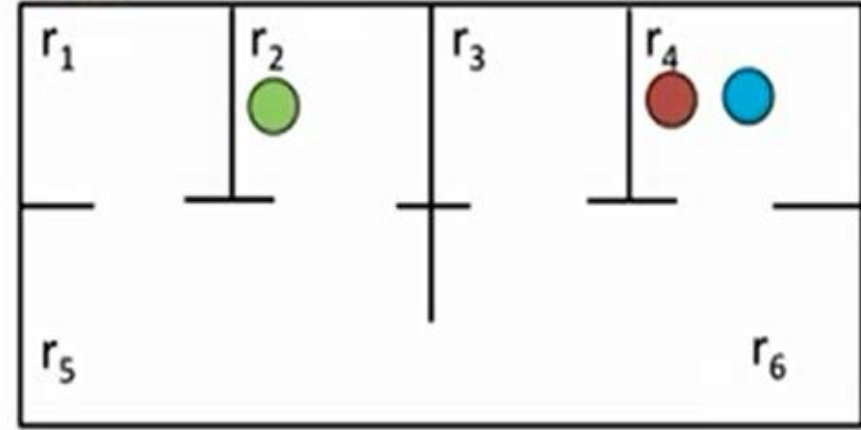
Problem-Solving

Example: Package delivery

Initial state



Goal



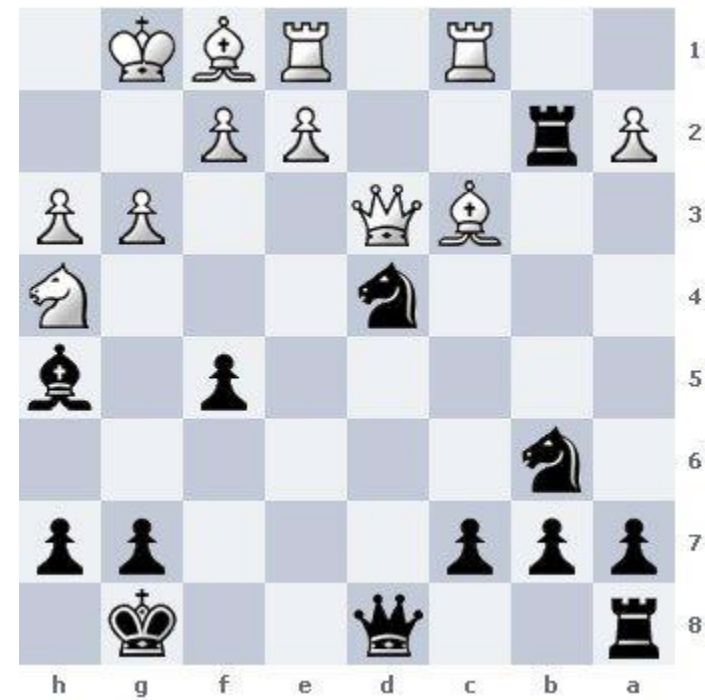
Problem-Solving

Example: Chess game

Initial state



Goal



Problem-Solving

Example: N-Puzzle

Initial state

2	8	3
1	6	4
7		5

?

Goal

1	2	3
8		4
7	6	5

Up

2	8	3
1	6	4
7		5



2	8	3
1		4
7	6	5



Up

2		3
1	8	4
7	6	5



Left

	2	3
1	8	4
7	6	5



Down

1	2	3
	8	4
7	6	5



Right

1	2	3
8		4
7	6	5

Problem-Solving

Graph search problem

■ Input:

- Initial node
- Goal function **Goal(n)** which returns **True** if the goal is achieved
- Transition function **Transition(n)** which returns the successor nodes of n
- Cost function $c(n,n')$ strictly positive, which returns the cost of going from n to n'

■ Output:

- A path in the graph (nodes and edges)
 - The path cost is the sum of all the edges cost in the graph
 - There can be several goal nodes

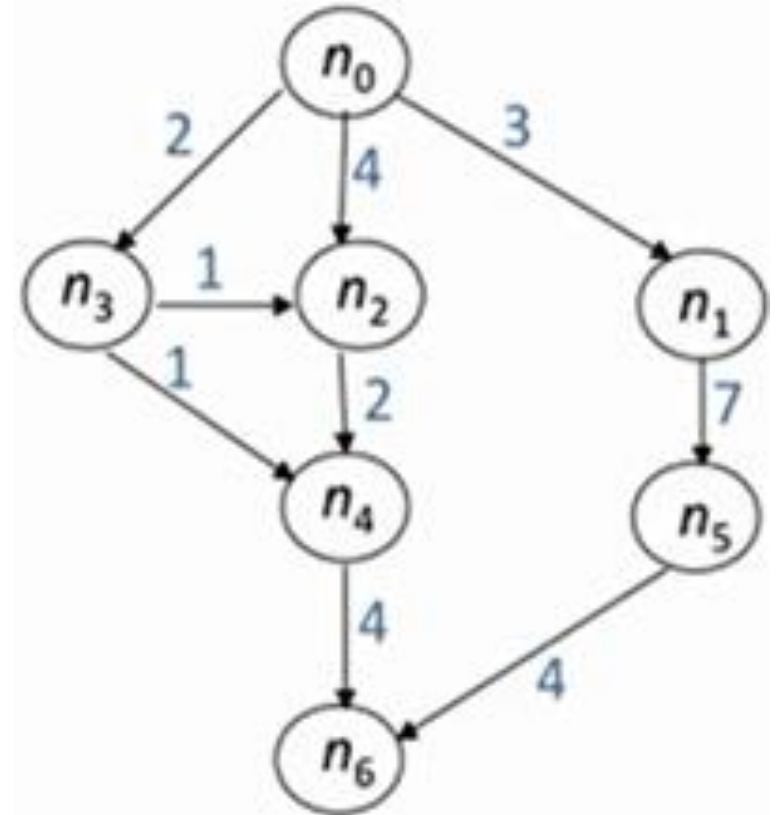
➤ Challenges:

- Find a solution path
- Find an optimal path
- Quickly find a path (in this case the optimality is not important)

Problem-Solving

A real world example: Find a path between two cities

- Cities: **Nodes**
- Paths between two cities: **Edges**
- Starting city: Initial node n_0
- Roads between cities: **Transition**(n_0) = (n_3, n_2, n_1)
- Distance between cities: $c(n_0, n_2) = 4$
- Destination city: **Goal**(n) = **True** if $n = n_6$ (n_6 is the destination city)



Search Algorithms

Any search problem is characterized by a **starting situation** and a **goal** to achieve and a **search space**:

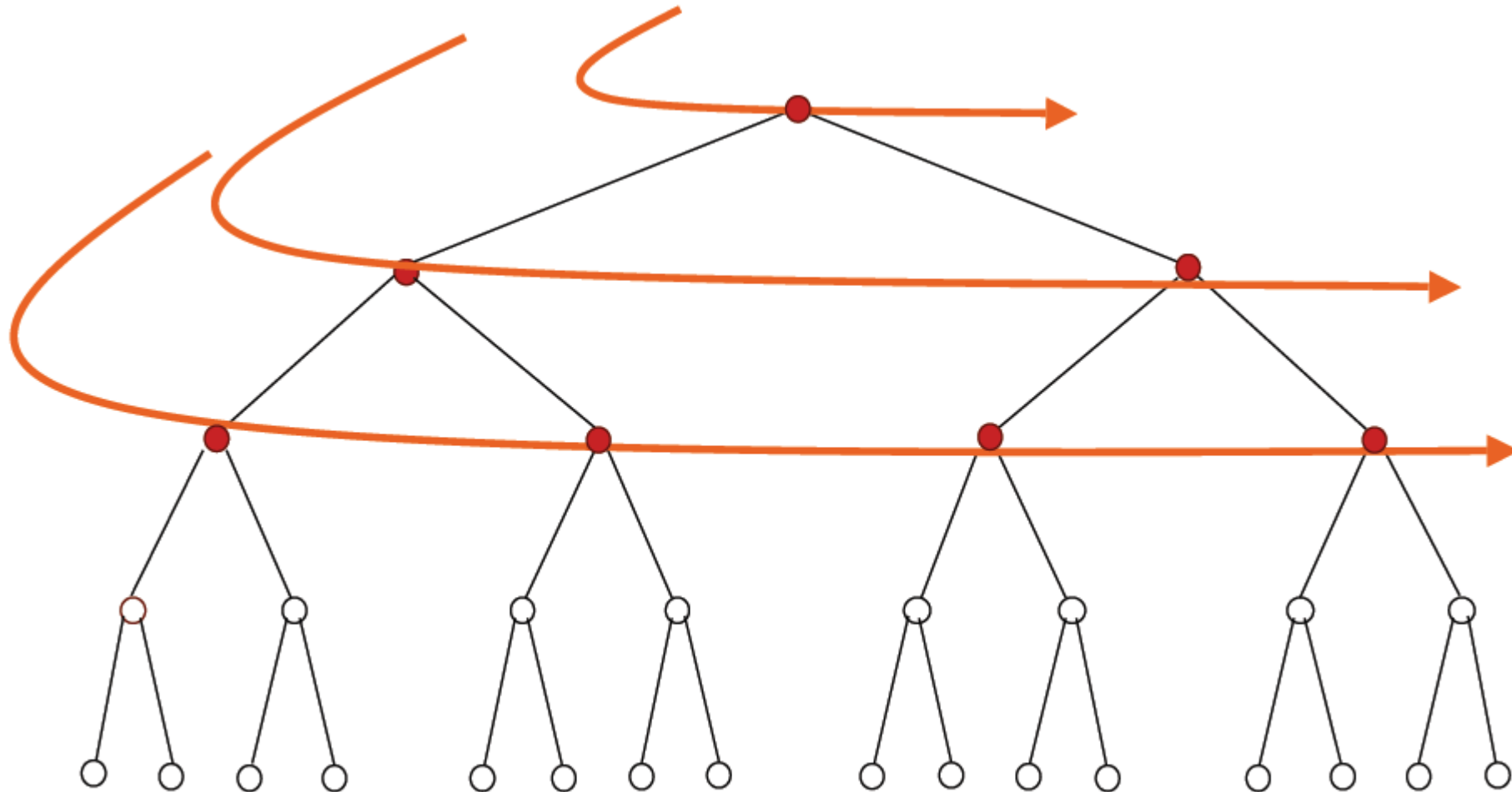
- The search space is composed of the set of all possible states.
- To determine possible operations to move from one state to another.
- To determine a search strategy.

There are different strategies:

- **Uninformed (Blind):**
 - Breadth-first search
 - Depth-first search (and its variations)
 - Uniform cost
- **Informed:**
 - Best-first search,
 - Greedy best-first search,
 - A* algorithm

Breadth-First Search (BFS)

- For a given node, explore the sibling nodes before exploring their children.



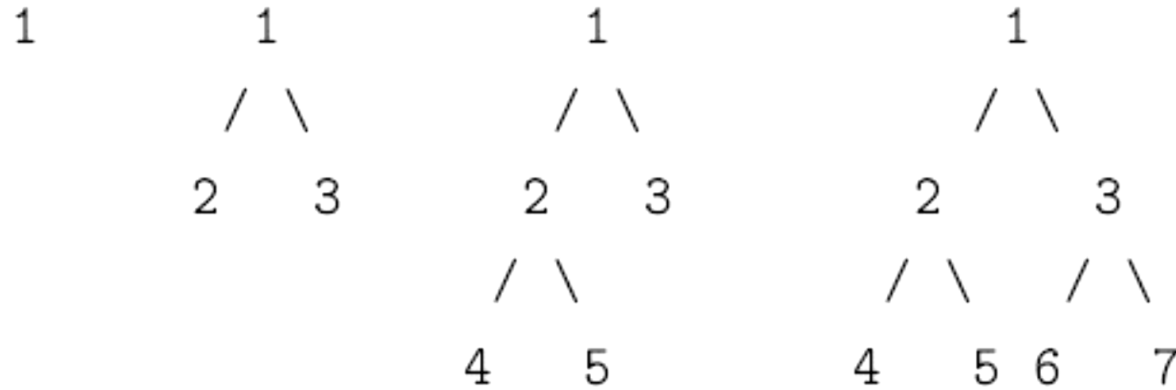
Breadth-First Search

Breadth-First Algorithm:

- 1- Put the **initial state** node in a **FIFO** queue: **OPEN**
- 2- If **n** corresponds to the **final state** then **Success**
- 3- If **OPEN** is **empty** then **Failure**
- 4- Remove **n** from **OPEN**
- 5- If **n** has **no successors** then go to **3**, otherwise:
 - Develop the successors of **n**
 - Insert them into **OPEN**
 - Establish chaining
 - Insert **n** into **Closed** (a queue containing the nodes already explored)
- 6- If among the **successors** there are **final states** then **Success**, otherwise go to **3**

Breadth-First Search

Tree traversal



1. Put 1 in the list. We get [1].

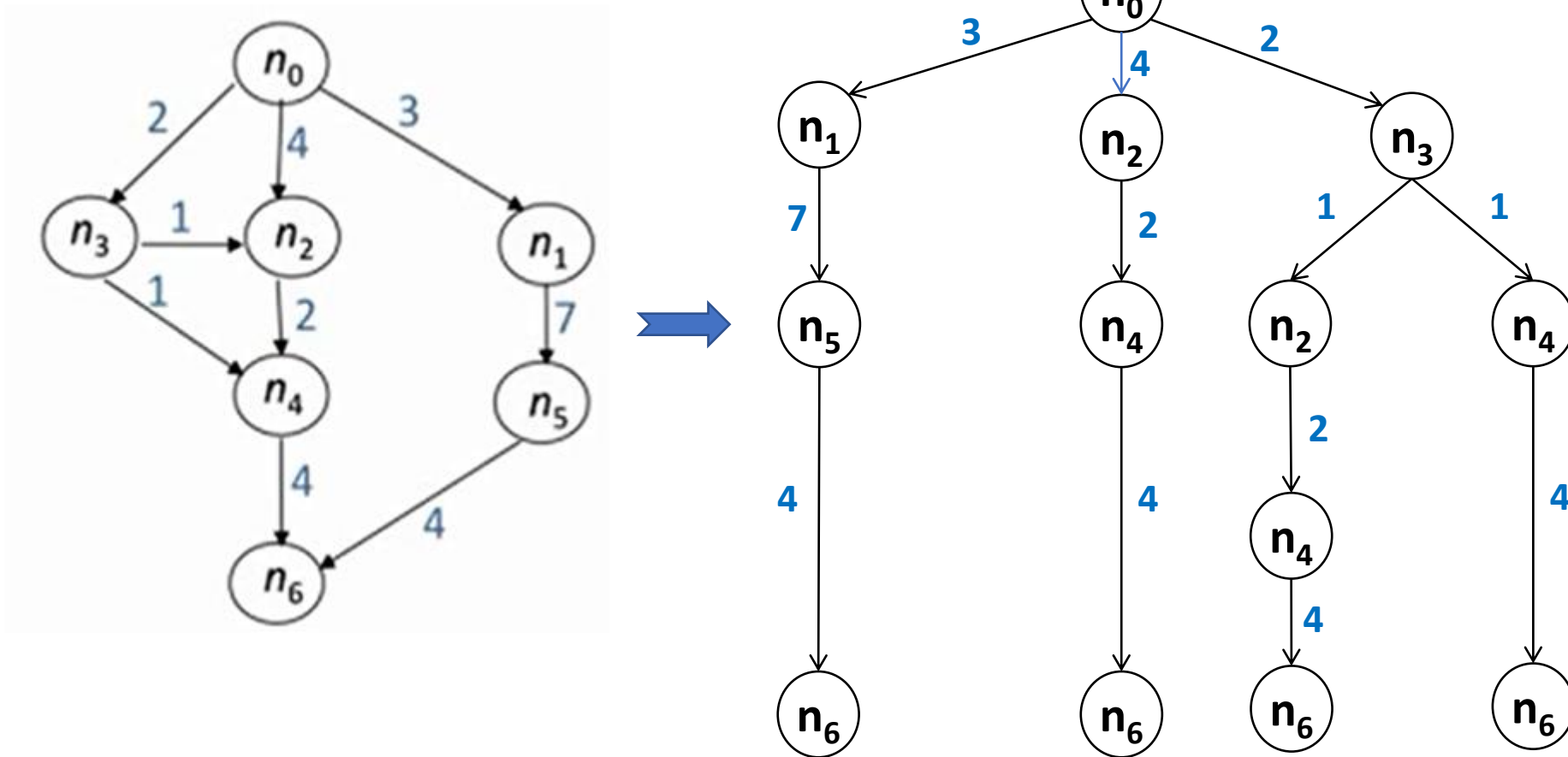
2. Remove the first element of the list (the 1) and add its successors 2, 3. We get [2,3].

3. Remove the first element of the list (the 2) and add its successors 4, 5. We get [3,4,5].

4. Remove the first element of the list (the 3) and add its successors 6, 7. We get [4,5,6,7].

Breadth-First Search

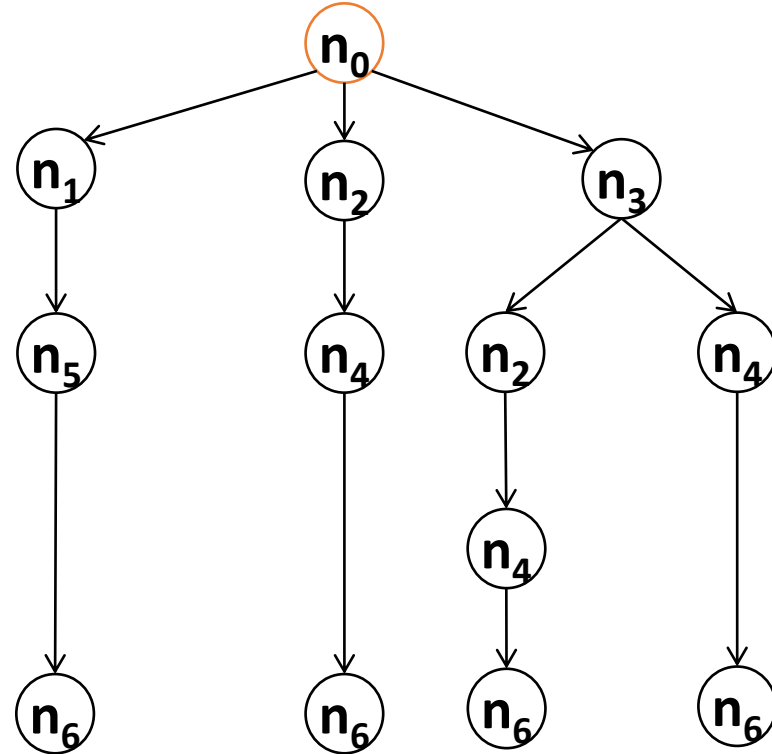
Illustrative example : Path between two cities n_0 and n_6



Breadth-First Search

- Illustration:

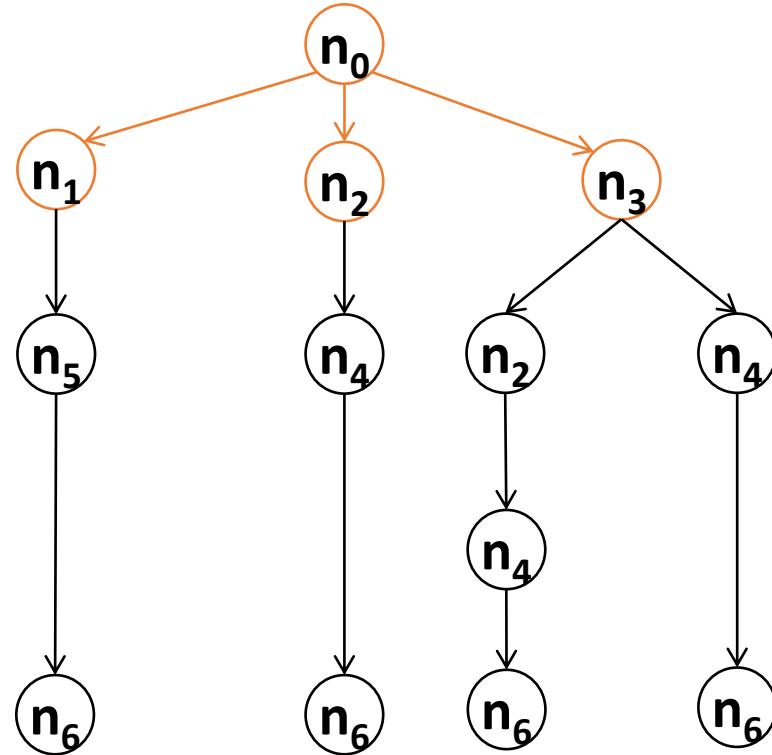
1. Put n_0 in the Open list. We get $[n_0]$.



Breadth-First Search

Illustration:

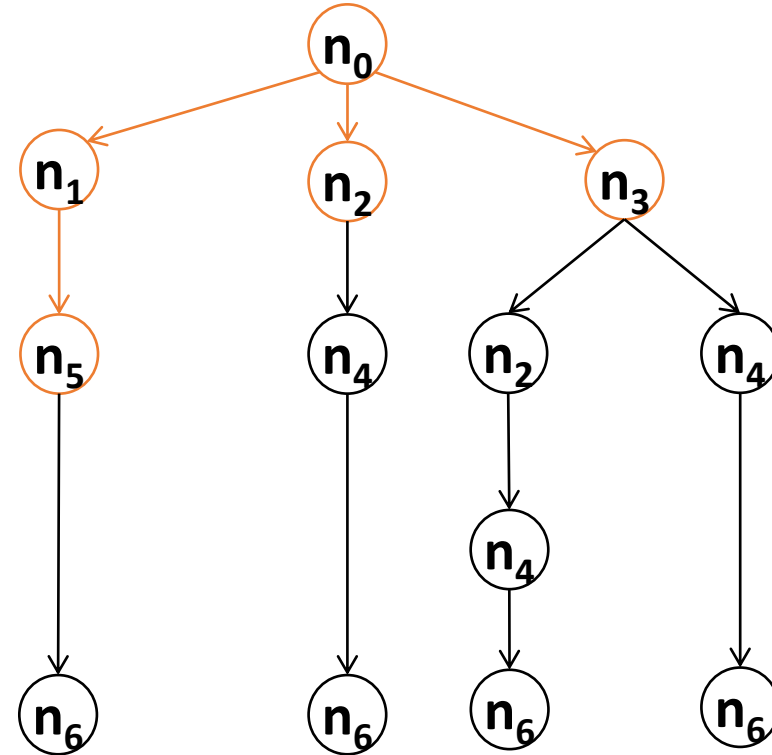
1. Put n_0 in the Open list. We get $[n_0]$.
2. Remove the first element of the list (the n_0) and add its successors n_1, n_2, n_3 . We get $[n_1, n_2, n_3]$.



Breadth-First Search

Illustration:

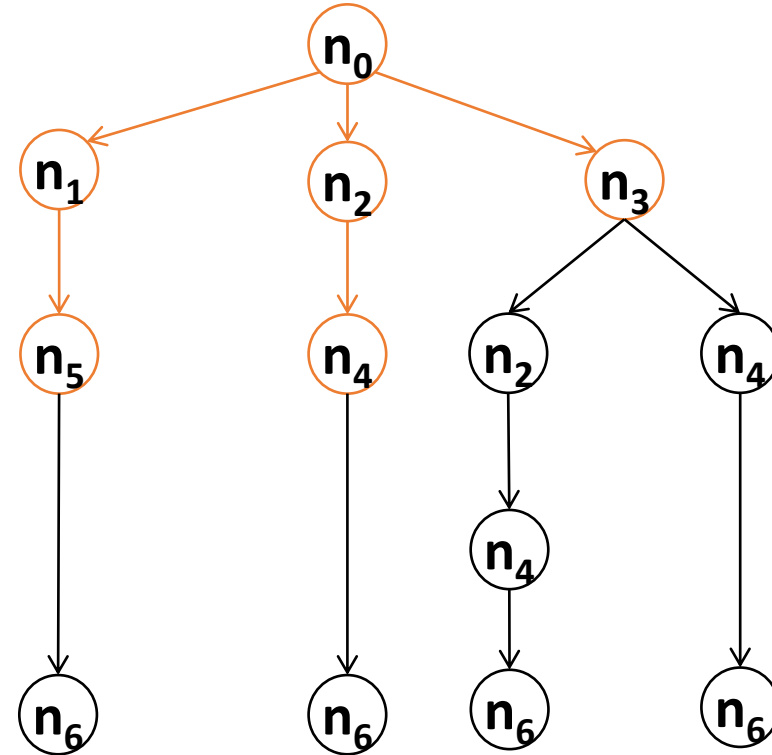
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2. Remove the first element of the list (the n_0) and add its successors n_1, n_2, n_3 . We get $[n_1, n_2, n_3]$.
3. Remove the first element of the list (the n_1) and add its successors n_5 . We get $[n_2, n_3, n_5]$.



Breadth-First Search

Illustration:

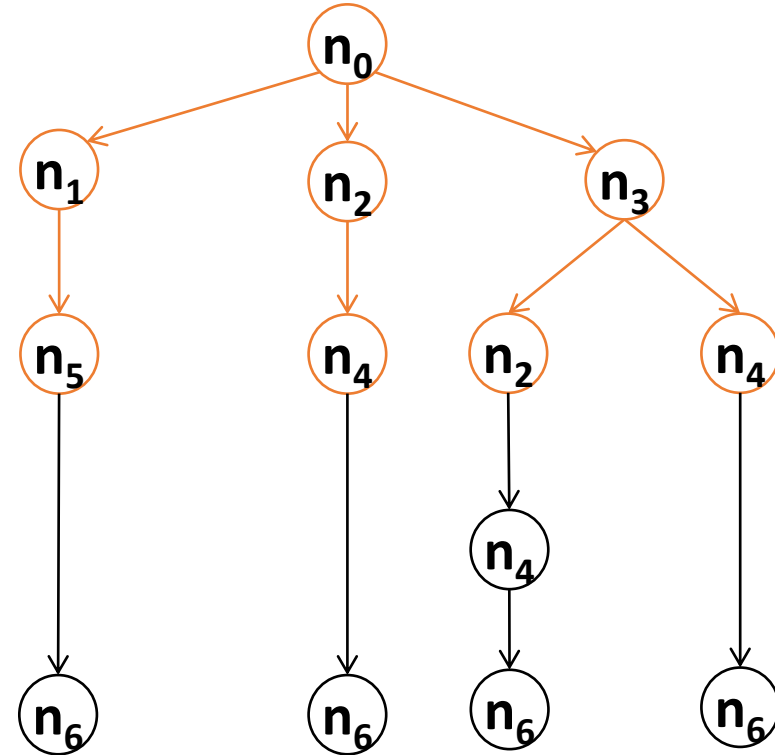
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3. Remove the first element of the list (the n_1) and add its successors n_5 . We get $[n_2, n_3, n_5]$.
4. Remove the first element of the list (the n_2) and add its successors n_4 . We get $[n_3, n_5, n_4]$.



Breadth-First Search

Illustration:

1. Put n_0 in the Open list. We get $[n_0]$.
2. Remove the first element of the list (the n_0) and add its successors n_1, n_2, n_3 . We get $[n_1, n_2, n_3]$.
3. Remove the first element of the list (the n_1) and add its successors n_5 . We get $[n_2, n_3, n_5]$.
4. Remove the first element of the list (the n_2) and add its successors n_4 . We get $[n_3, n_5, n_4]$.
5. Remove the first element of the list (the n_3) and add its successors n_2, n_4 . We get $[n_5, n_4, n_{2(n_3)}, n_{4(n_3)}]$.



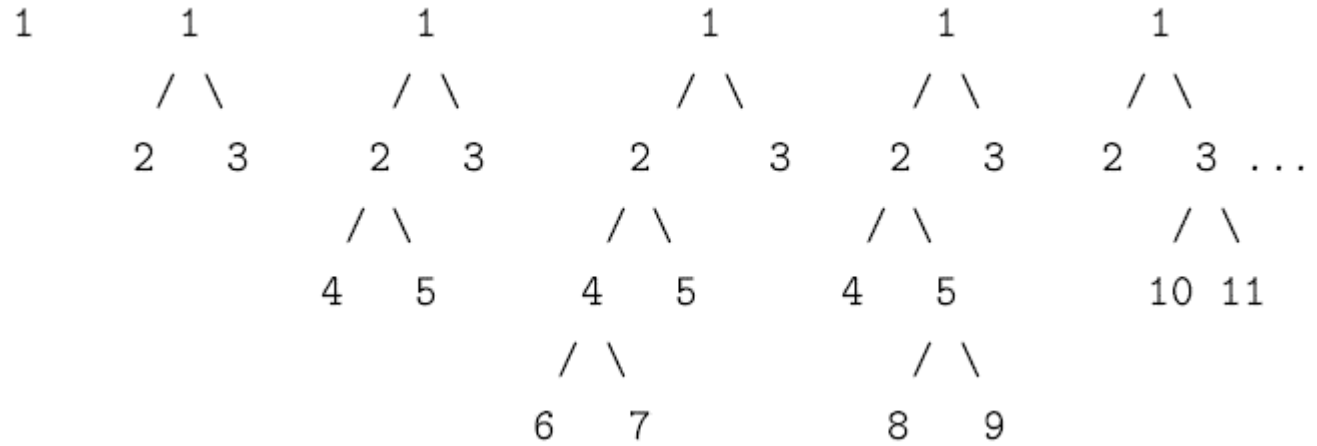
Depth-First Search

Depth-First Algorithm:

- 1- Put the **initial state** node in a **LIFO** stack: **OPEN**
- 2- If the stack is **empty** then **Failure**
- 3- Pop **n**
- 4- Develop the **successors** of **n**:
 - If **successors** exist then
 - Push the successors
 - Establish the successor chaining
 - Put **n** in **Closed**
- 6- If among the **successors** there are **final states** then **Success**, otherwise go to **2**

Depth-First Search

■ Tree traversal



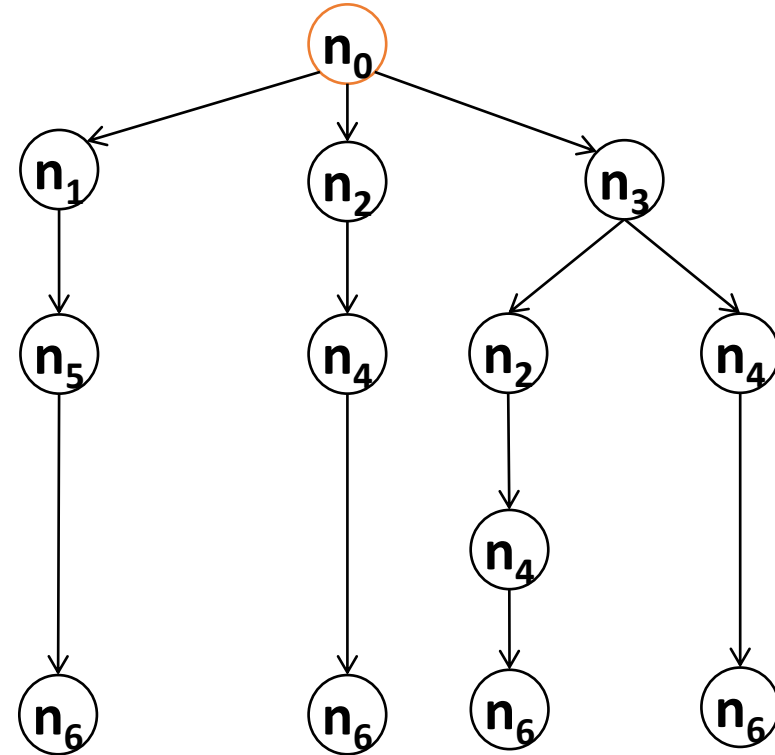
We are using a stack data structure:

1. Put 1 into the stack. We get [1].
2. Remove the first element from the stack (1) and add its successors 2, 3. We get [2,3].
3. Remove the first element from the stack (2) and add its successors 4, 5. We get [4,5,3].
4. Remove the first element from the stack (4) and add its successors 6, 7. We get [6,7,5,3].
5. Remove the first element from the stack (6) which has no successors. We get [7,5,3].
6. Remove the first element from the stack (7) which has no successors. We get [5,3].
7. Remove the first element from the stack (5) and add its successors 8, 9. We get [8,9,3].
8. Remove the first element from the stack (8) which has no successors. We get [9,3].
9. Remove the first element from the stack (9) which has no successors. We get [3].
10. Remove the first element from the stack (3) and add its successors 10, 11. We get [10,11].....

Depth-First Search

- **Illustration:**

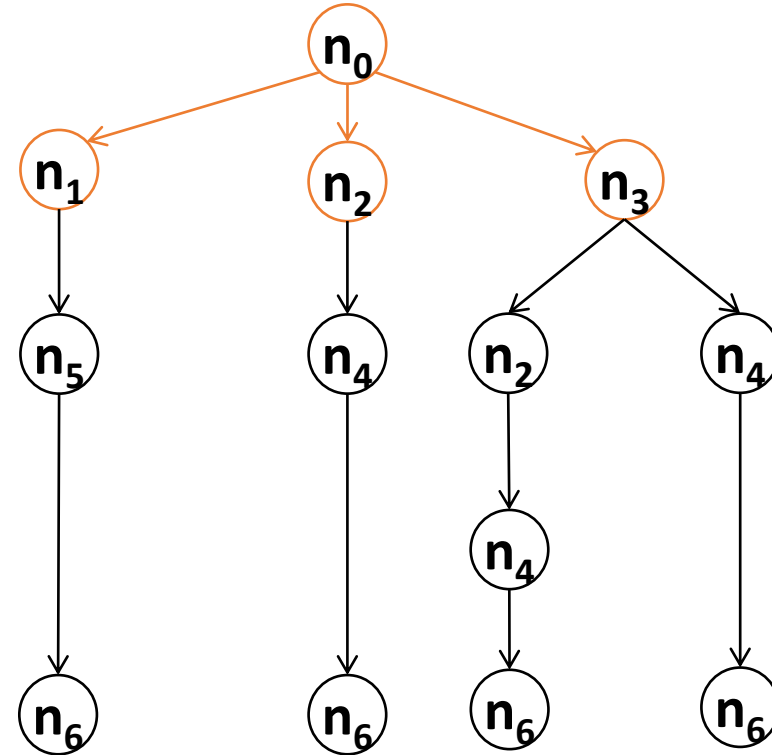
1. Put n_0 in **Open**. We get $[n_0]$.



Depth-First Search

- **Illustration:**

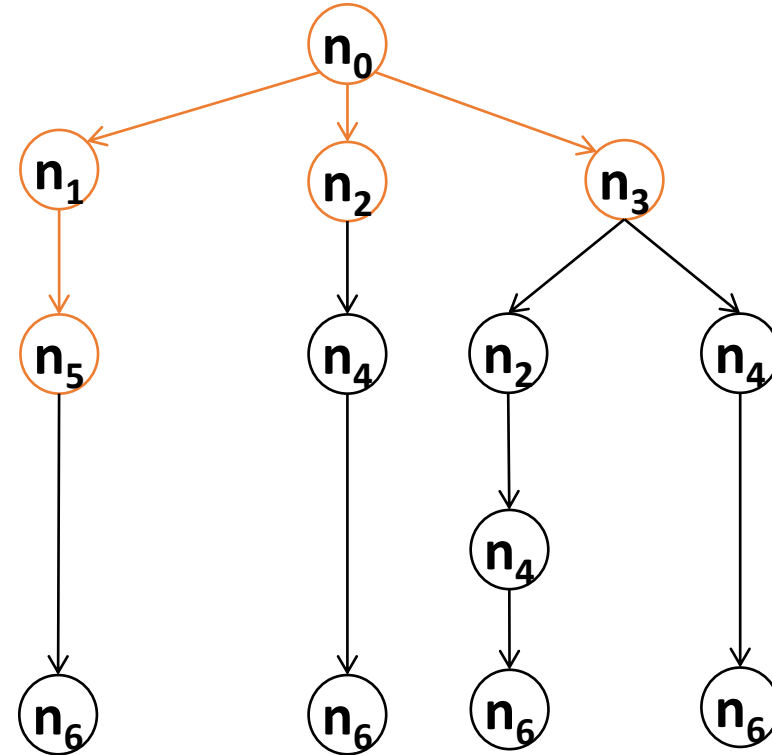
1. Put n_0 in **Open**. We get $[n_0]$.
2. Remove the first element of the stack (the n_0) and add its successors n_1, n_2, n_3 . We get $[n_1, n_2, n_3]$.



Depth-First Search

Illustration:

1. Put n_0 in **Open**. We get $[n_0]$.
2. Remove the first element of the stack (the n_0) and add its successors n_1, n_2, n_3 . We get $[n_1, n_2, n_3]$.
3. Remove the first element of the stack (the n_1) and add its successors n_5 . We get $[n_5, n_2, n_3]$.

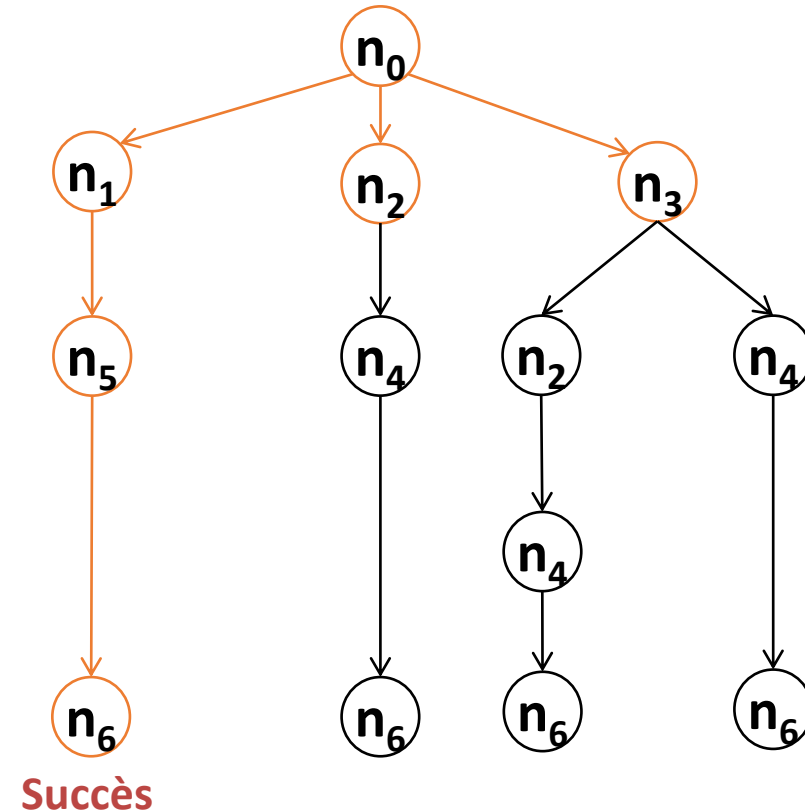


Depth-First Search

Illustration:

1. Put n_0 in **Open**. We get $[n_0]$.
2. Remove the first element of the stack (the n_0) and add its successors n_1, n_2, n_3 . We get $[n_1, n_2, n_3]$.
3. Remove the first element of the stack (the n_1) and add its successors n_5 . We get $[n_5, n_2, n_3]$.
4. Remove the first element of the stack (the n_5) and add its successors n_6 . We get $[n_6, n_2, n_3]$.

n_6 Appears in Open
then Stop

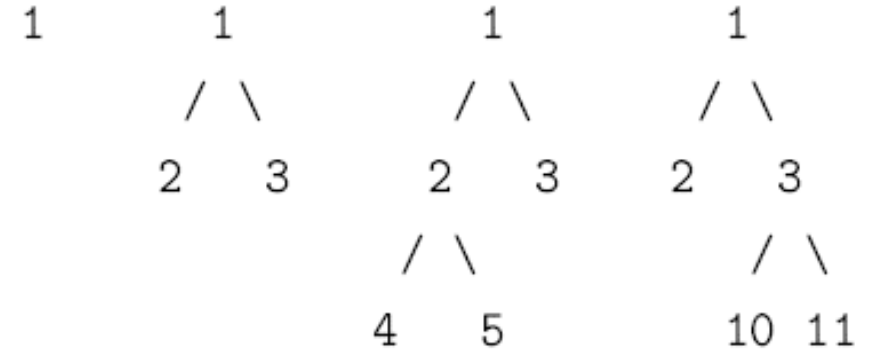


Limited Depth-First Search (LDFS)

It is a depth-first search where states are explored to a limited depth

Example with Limit = 2

We add the successors of a state only if we have not exceeded the limit of 2.



1. Put 1 in the stack with its depth. We get [(1,0)].
2. Remove the first element from the stack (1, 0) and add its successors 2, 3 if we have not exceeded the limit. We get [(2,1),(3,1)].
3. Remove the first element from the stack (2, 1) and add its successors 4, 5 if we have not exceeded the limit. We get [(4,2),(5,2),(3,1)].
4. Remove the first element from the stack (4, 2) and add nothing. We get [(5,2),(3,1)].
5. Remove the first element from the stack (5, 2) and add nothing. We get [(3,1)].
6. Remove the first element from the stack (3, 1) and add its successors 10, 11 if we have not exceeded the limit. We get [(10,2),(11,2)].
7. Remove the first element from the stack (10, 2) and add nothing. We get [(11,2)].
8. Remove the first element from the stack (11, 2) and add nothing. We get [].

Iterative Depth-First Search (IDFS)

It is an iteration of the limited depth-first search

For **p = 0** to **infinite** do

```
{  
Limited_DFS(p)  
}
```

BFS vs DFS

- ✓ The breadth-first search strategy is interesting because if there exists a path to the goal in the early levels, it finds the shortest path.
- ✓ The depth-first search strategy is also interesting because it returns the best path if the goal state is found in the early branches of the search tree.
- However, these two strategies are considered blind because they do not take into account the path leading to the goal state

Uniform-Cost Search Algorithm

Principle:

- Each arc of the graph is associated with a traversal cost.
- This algorithm provides an optimal cost solution.
 - $C(n_i, n_j)$ is the cost of an arc from n_i to n_j .
 - The cost function of a node ns , which is the successor node of n , is calculated as follows:

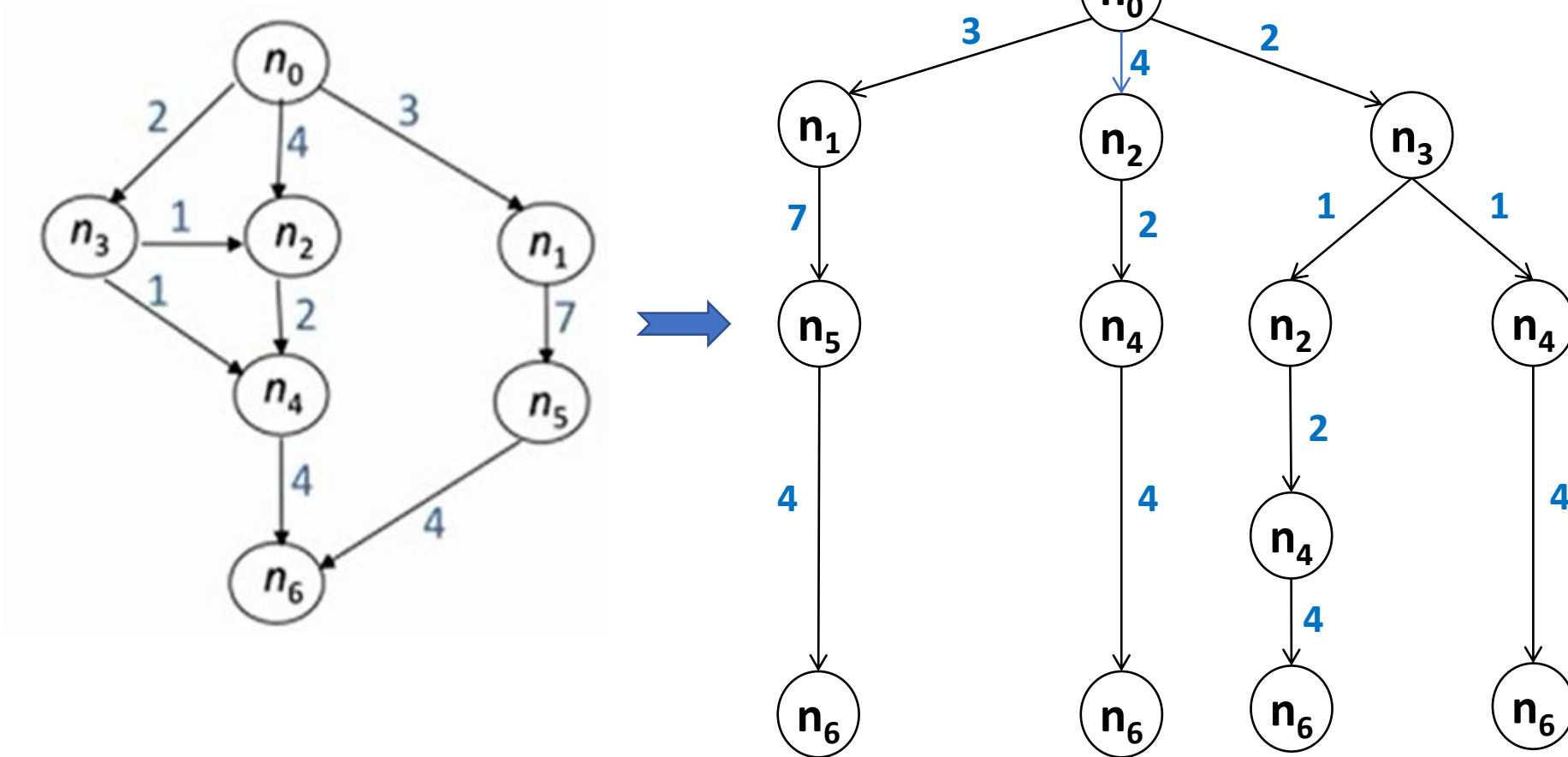
$$g(ns) = g(n) + c(n, ns)$$

Where $g(n)$ is the cost up to node n

The open nodes of the graph are ordered in ascending order.

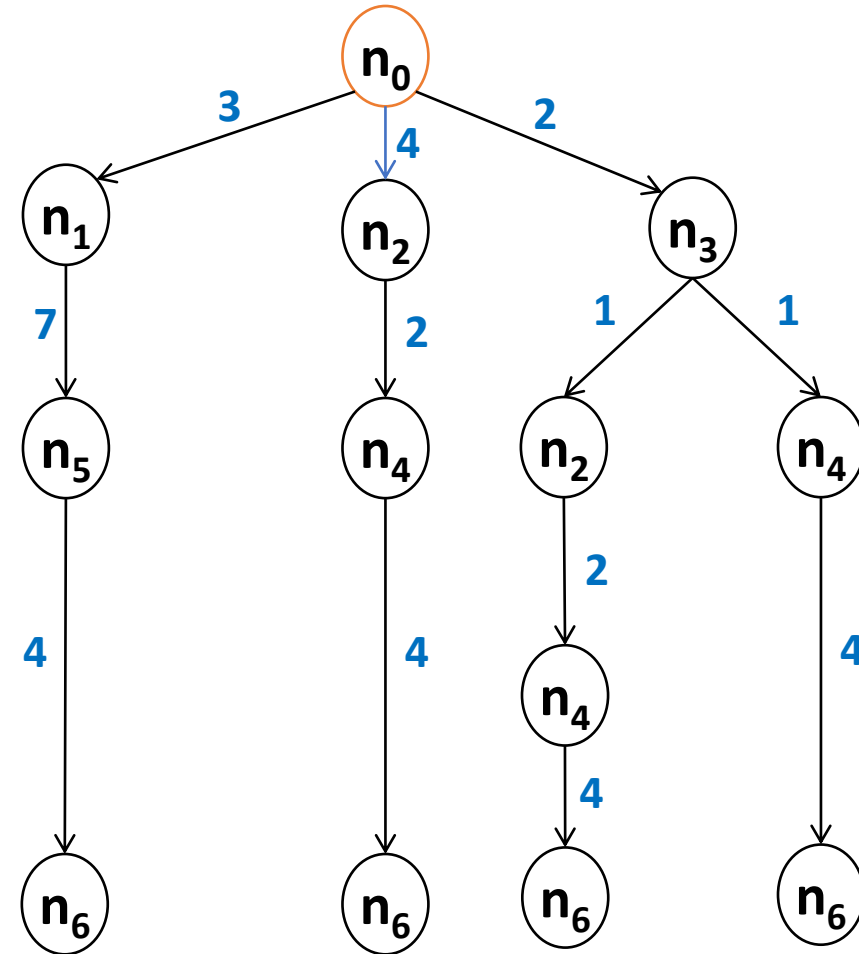
Uniform-Cost Search Algorithm

Illustrative example : Path between two cities n_0 and n_6



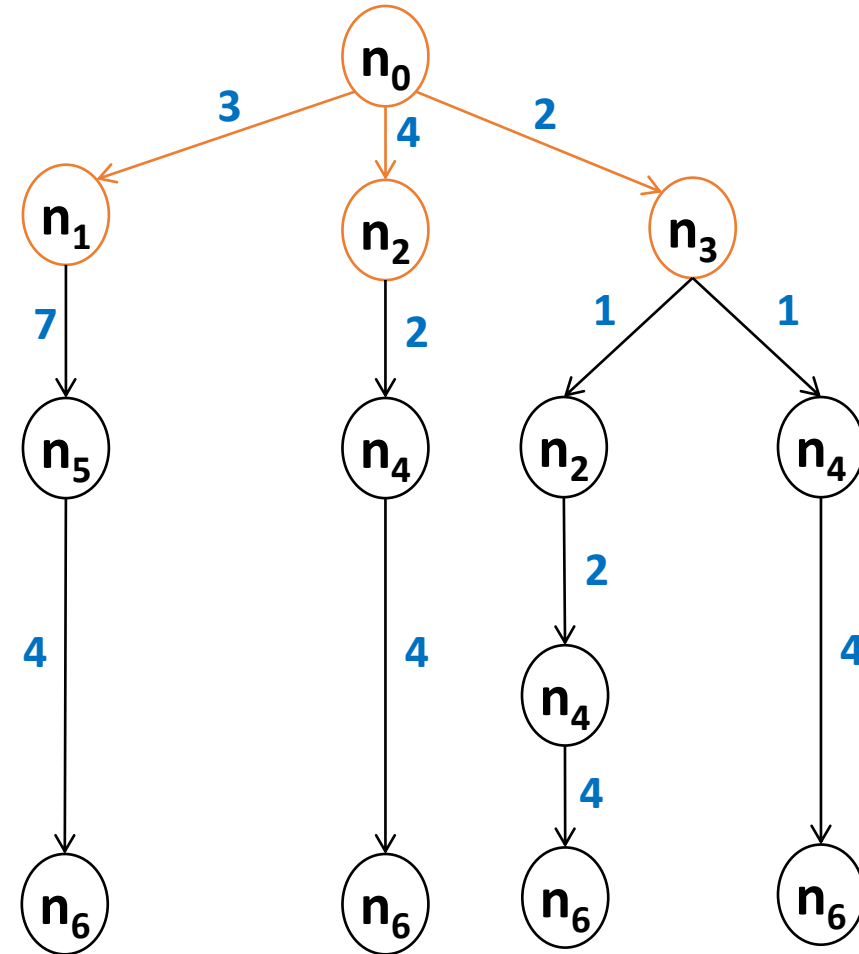
Uniform-Cost Search Algorithm

1. Put n_0 in Open with its initial cost. We get $[(n_0, 0)]$.



Uniform-Cost Search Algorithm

1. Put n_0 in Open with its initial cost. We get $[(n_0, 0)]$.
2. Remove the first element of the list $(n_0, 0)$ and add its successors n_1, n_2, n_3 to the list of states respecting the ascending order. To achieve this, we will calculate the total cost of each successor : $g(ns) = g(n) + c(n, ns)$
 $g(n_1) = g(n_0) + c(n_0, n_1) = 3$
We get: $[(n_3, 2, n_0), (n_1, 3, n_0), (n_2, 4, n_0)]$



Uniform-Cost Search Algorithm

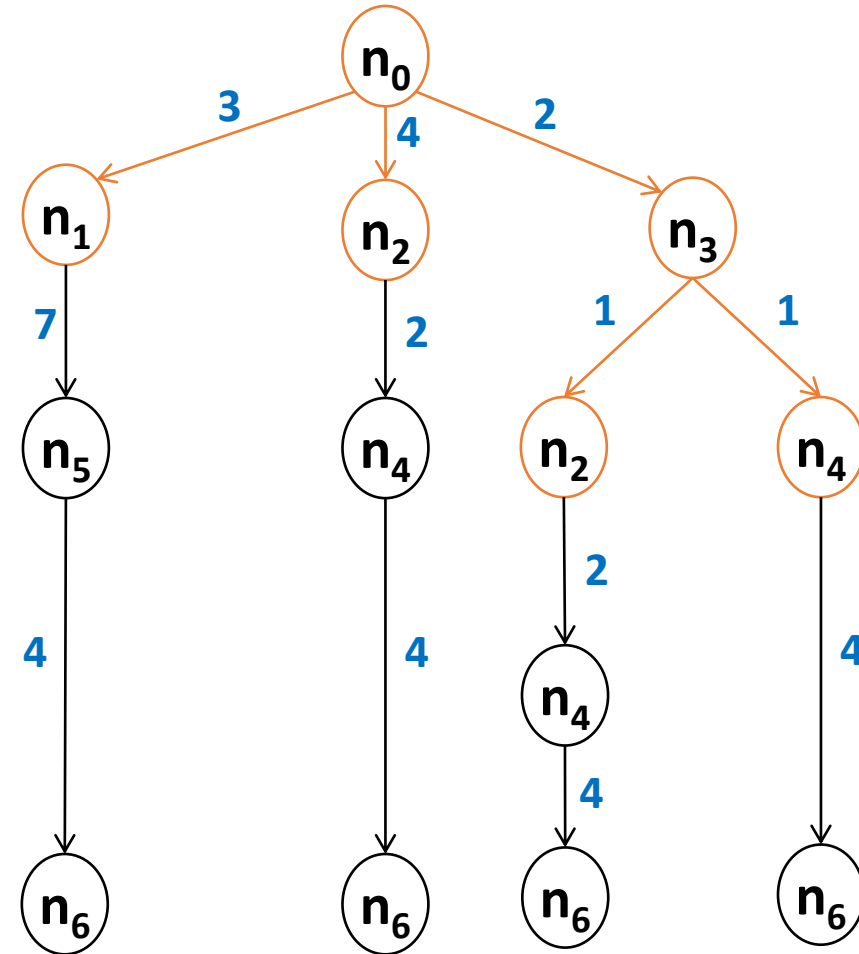
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$$g(n_1) = g(n_0) + c(n_0, n_1) = 3$$

We get: $[(n_3, 2, n_0), (n_1, 3, n_0), (n_2, 4, n_0)]$

3. Remove the first element of the list $(n_3, 2, n_0)$ and add its successors n_2, n_4 to the list of states respecting the ascending order. We get: $[(n_1, 3, n_0), (n_2, 3, n_3), (n_4, 3, n_3), (n_2, 4, n_0)]$



Uniform-Cost Search Algorithm

1. Put n_0 in Open with its initial cost. We get $[(n_0, 0)]$.

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$$g(n_1) = g(n_0) + c(n_0, n_1) = 3$$

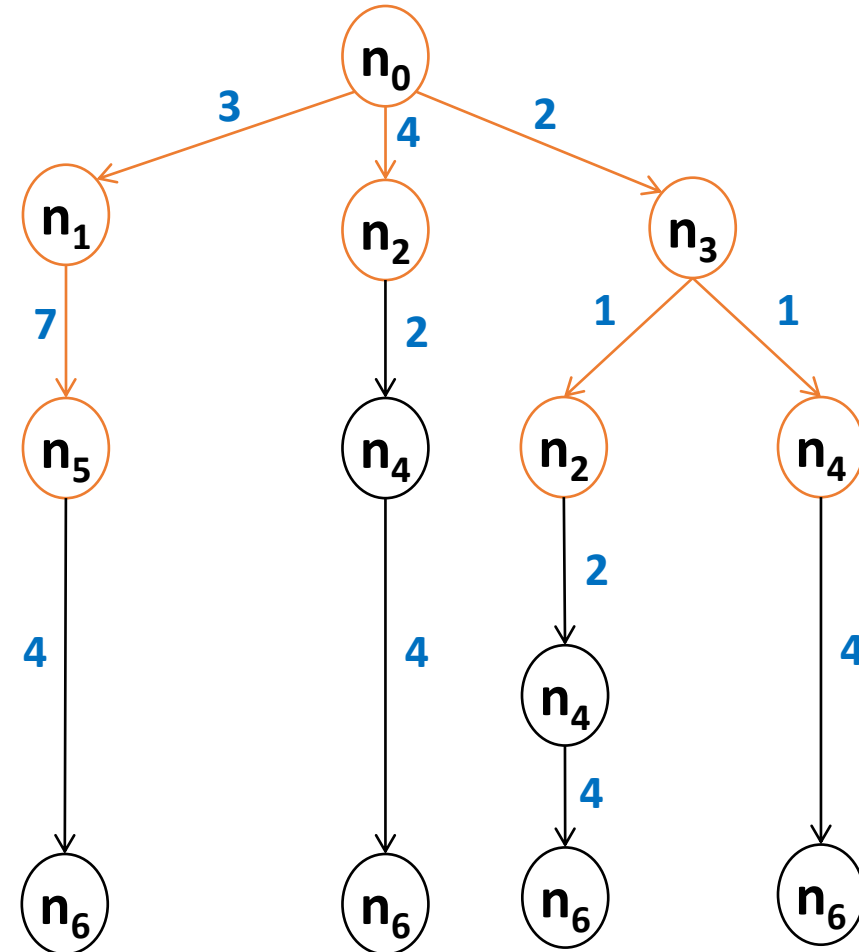
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3. Remove the first element of the list $(n_3, 2, n_0)$ and add its successors n_2, n_4 to the list of states respecting the ascending order.

We get: $[(n_1, 3, n_0), (n_2, 3, n_3), (n_4, 3, n_3), (n_2, 4, n_0)]$

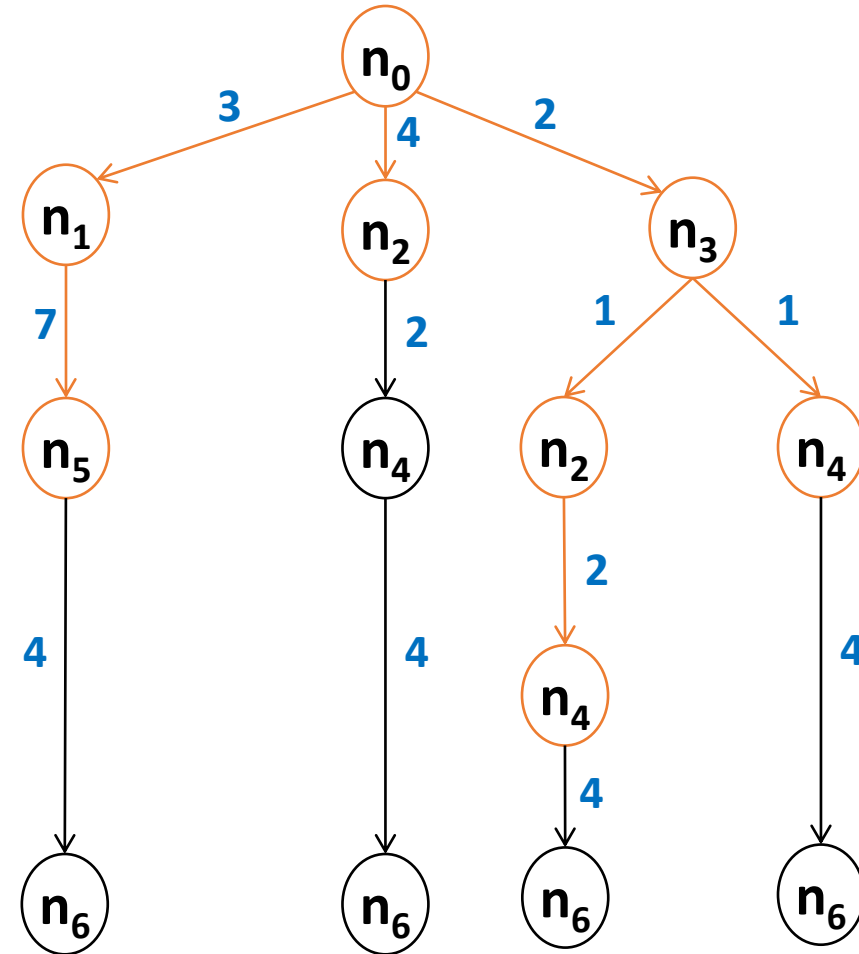
4. Remove the first element of the list $(n_1, 3, n_0)$ and add its successors n_5 to the list of states respecting the ascending order.

We get: $[(n_2, 3, n_3), (n_4, 3, n_3), (n_2, 4, n_0), (n_5, 10, n_1)]$



Uniform-Cost Search Algorithm

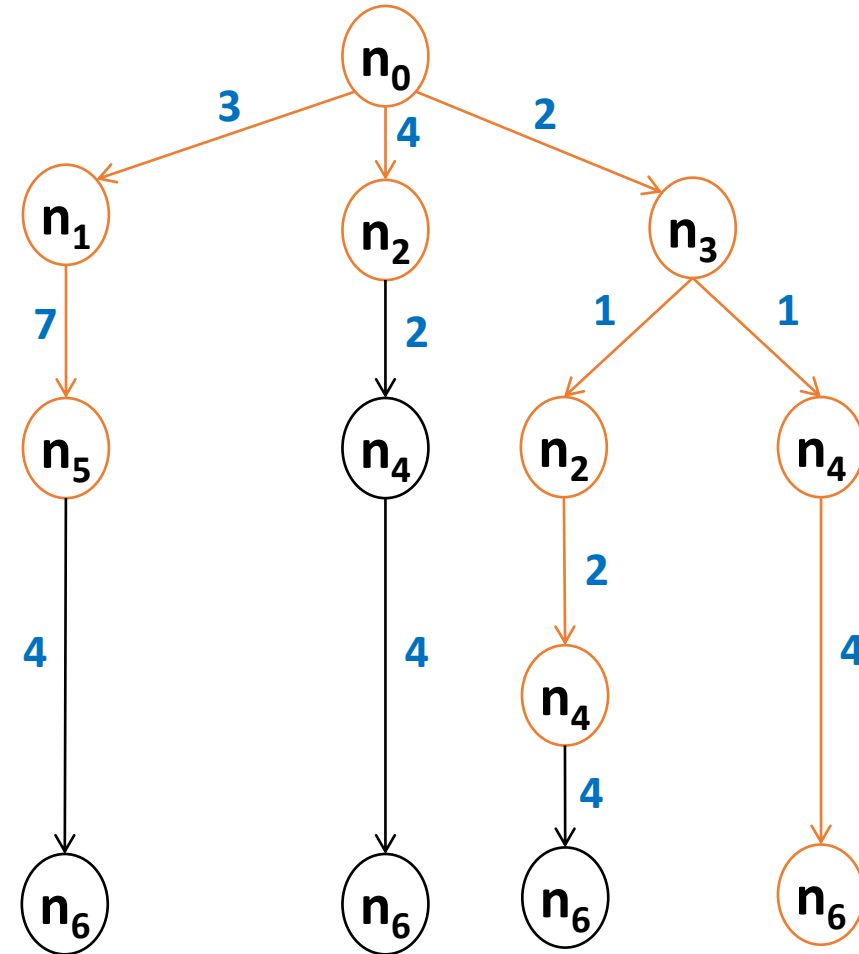
5. Remove the first element of the list $(n_2, 3, n_3)$ and add its successors n_4 to the list of states respecting the ascending order. We get: $[(n_4, 3, n_3), (n_2, 4, n_0), (n_4, 5, n_2), (n_5, 10, n_1)]$



Uniform-Cost Search Algorithm

5. Remove the first element of the list $(n_2, 3, n_3)$ and add its successors n_4 to the list of states respecting the ascending order. We get: $[(n_4, 3, n_3), (n_2, 4, n_0), (n_4, 5, n_2), (n_5, 10, n_1)]$

6. Remove the first element of the list $(n_4, 3, n_3)$ and add its successors n_6 to the list of states respecting the ascending order. We get: $[(n_2, 4, n_0), (n_4, 5, n_2), (n_6, 7, n_4), (n_5, 10, n_1)]$

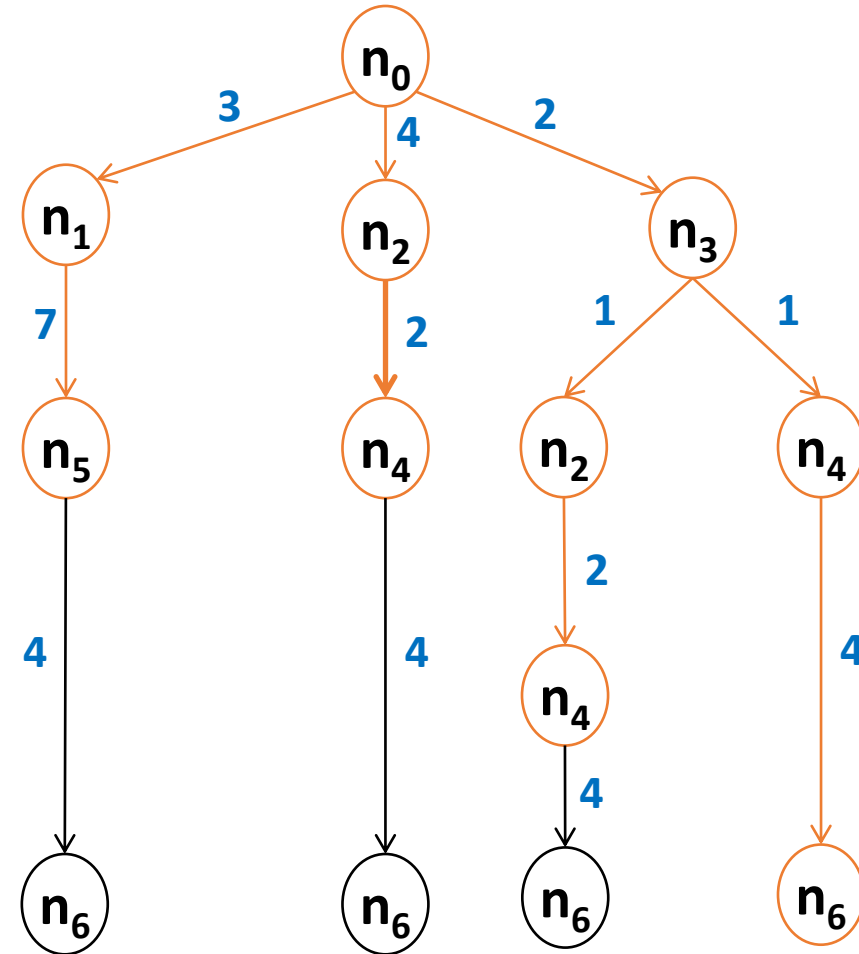


Uniform-Cost Search Algorithm

5. Remove the first element of the list $(n_2, 3, n_3)$ and add its successors n_4 to the list of states respecting the ascending order. We get: $[(n_4, 3, n_3), (n_2, 4, n_0), (n_4, 5, n_2), (n_5, 10, n_1)]$

6. Remove the first element of the list $(n_4, 3, n_3)$ and add its successors n_6 to the list of states respecting the ascending order. We get: $[(n_2, 4, n_0), (n_4, 5, n_2), (n_6, 7, n_4), (n_5, 10, n_1)]$

7. Remove the first element of the list $(n_2, 4, n_0)$ and add its successors n_4 to the list of states respecting the ascending order. We get: $[(n_4, 5, n_2 (n_3)), (n_4, 6, n_2 (n_0)), (n_6, 7, n_4), (n_5, 10, n_1)]$



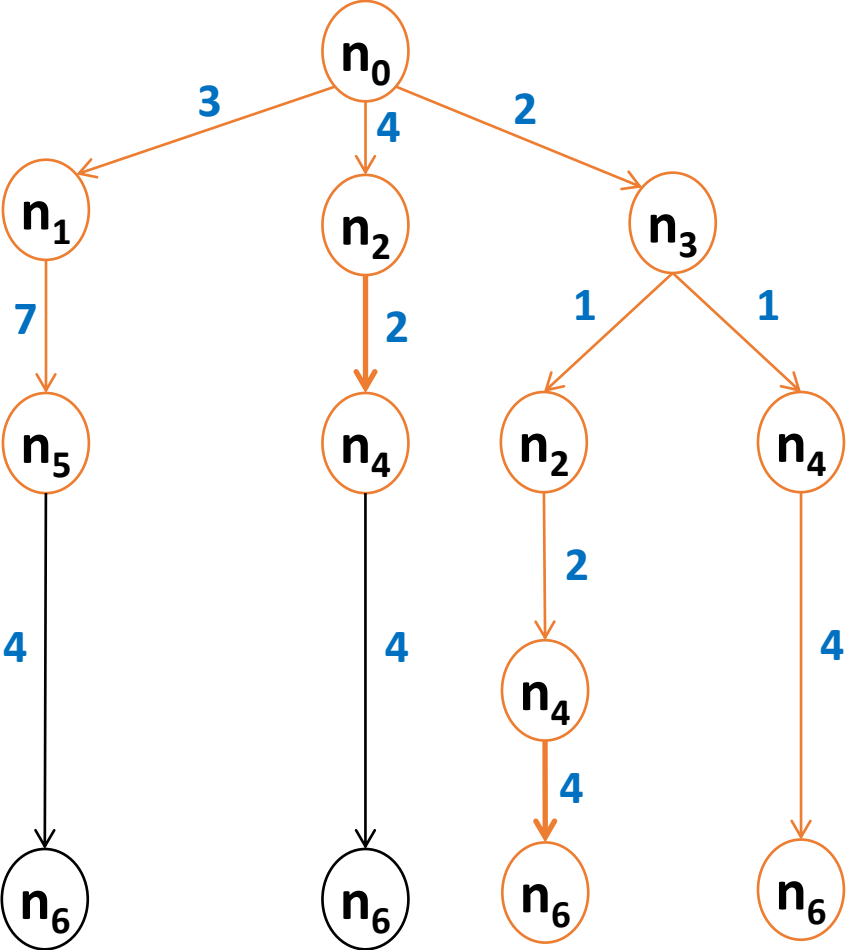
Uniform-Cost Search Algorithm

5. Remove the first element of the list $(n_2, 3, n_3)$ and add its successors n_4 to the list of states respecting the ascending order. We get: $[(n_4, 3, n_3), (n_2, 4, n_0), (n_4, 5, n_2), (n_5, 10, n_1)]$

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7. Remove the first element of the list $(n_2, 4, n_0)$ and add its successors n_4 to the list of states respecting the ascending order. We get: $[(n_4, 5, n_2 (n_3)), (n_4, 6, n_2 (n_0)), (n_6, 7, n_4), (n_5, 10, n_1)]$

8. Remove the first element of the list $(n_4, 5, n_2 (n_3))$ and add its successors n_6 to the list of states respecting the ascending order. We get: $[(n_4, 6, n_2 (n_0)), (n_6, 7, n_4 (n_3)), (n_6, 9, n_4 (n_2)), (n_5, 10, n_1)]$



Uniform-Cost Search Algorithm

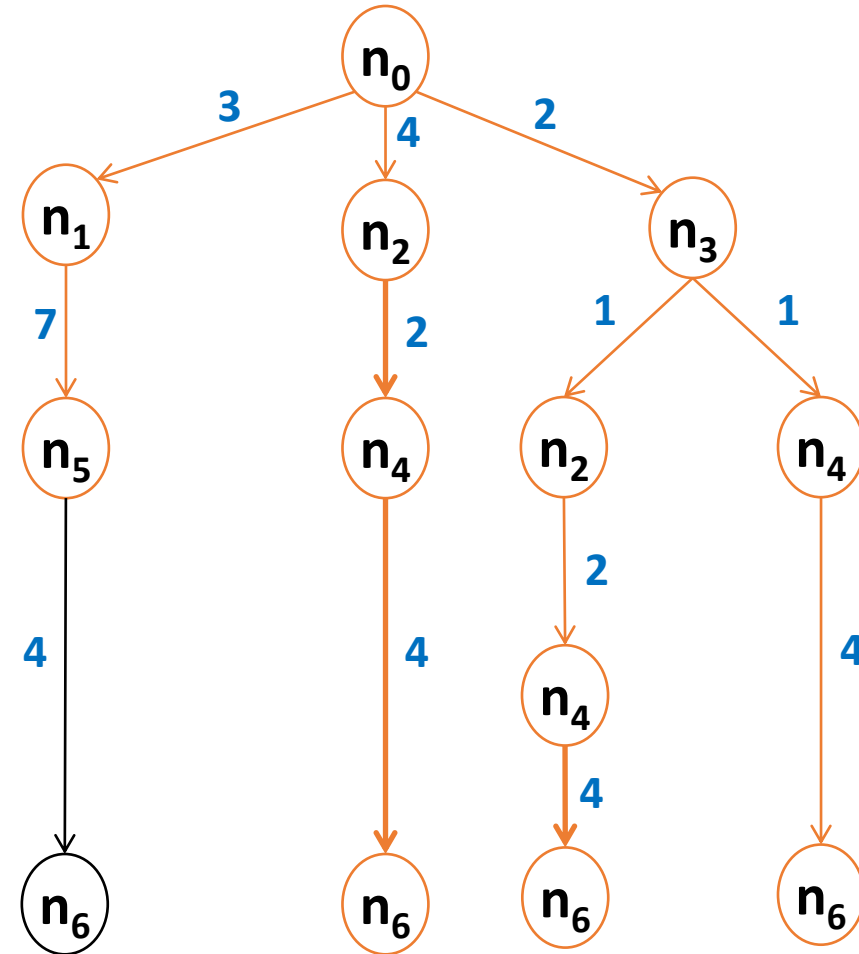
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6. Remove the first element of the list $(n_4, 3, n_3)$ and add its successors n_6 to the list of states respecting the ascending order. We get: $[(n_2, 4, n_0), (n_4, 5, n_2), (n_6, 7, n_4), (n_5, 10, n_1)]$

7. Remove the first element of the list $(n_2, 4, n_0)$ and add its successors n_4 to the list of states respecting the ascending order. We get: $[(n_4, 5, n_2(n_3)), (n_4, 6, n_2(n_0)), (n_6, 7, n_4), (n_5, 10, n_1)]$

8. Remove the first element of the list $(n_4, 5, n_2(n_3))$ and add its successors n_6 to the list of states respecting the ascending order. We get: $[(n_4, 6, n_2(n_0)), (n_6, 7, n_4(n_3)), (n_6, 9, n_4(n_2)), (n_5, 10, n_1)]$

9. Remove the first element of the list $(n_4, 6, n_2(n_0))$ and add its successors n_6 to the list of states respecting the ascending order. We get: $[(n_6, 7, n_4(n_3)), (n_6, 9, n_4(n_2, n_3)), (n_6, 10, n_4(n_2, n_0)), (n_5, 10, n_1)]$



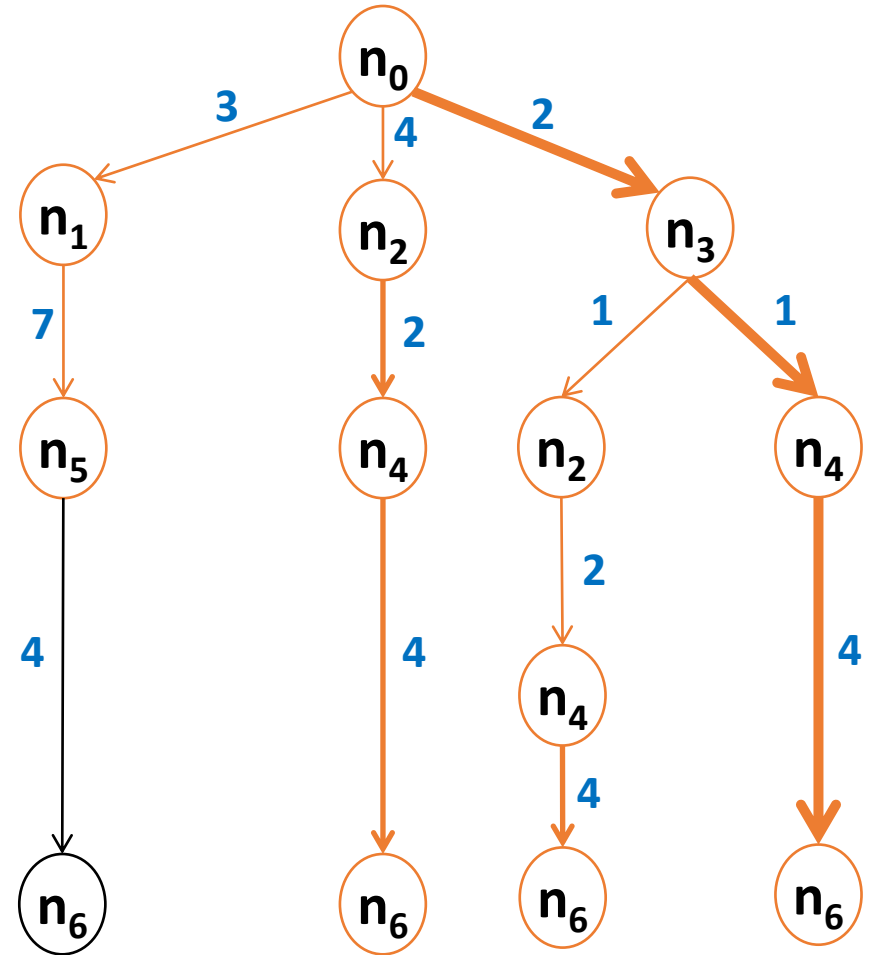
Uniform-Cost Search Algorithm

9. Remove the first element of the list $(n_4, 6, n_2(n_0))$ and add its successors n_6 to the list of states respecting the ascending order.

We get: $[(n_6, 7, n_4(n_3)), (n_9, 9, n_4(n_2, n_3)), (n_{10}, 10, n_4(n_2, n_0)), (n_5, 10, n_1)]$

n_6 Appears in the head of
Open then stop the search

The path with optimal cost: n_0, n_3, n_4, n_6



If the cost of **each arc = 1**, then **Uniform-Cost = BFS**

Heuristic-Based Search Algorithms

Best-First Search

1. Start the search by a **List** containing the starting state (**initial node**) of the problem
2. If **List** not empty:
 - Select a state **n** with **minimal** measure to expand
 - If **n** is a final state (**Goal node**) then return **Success**
 - Else, add all **n successor nodes** to the List with respect of ascending order according to the utility measure.
 - Restart at point 2.
3. Else return **Failure**.

Heuristic-Based Search Algorithms

Greedy Best-First Search

- The utility measure is given by an estimation function **h** .
- For each state **n** , **$h(n)$** represents the **estimated cost** from **n** to a **final state**.

For example, in the problem of the shortest path between two cities,

we can take **$h(n)$** = **direct distance** between **n** and the **destination city**.

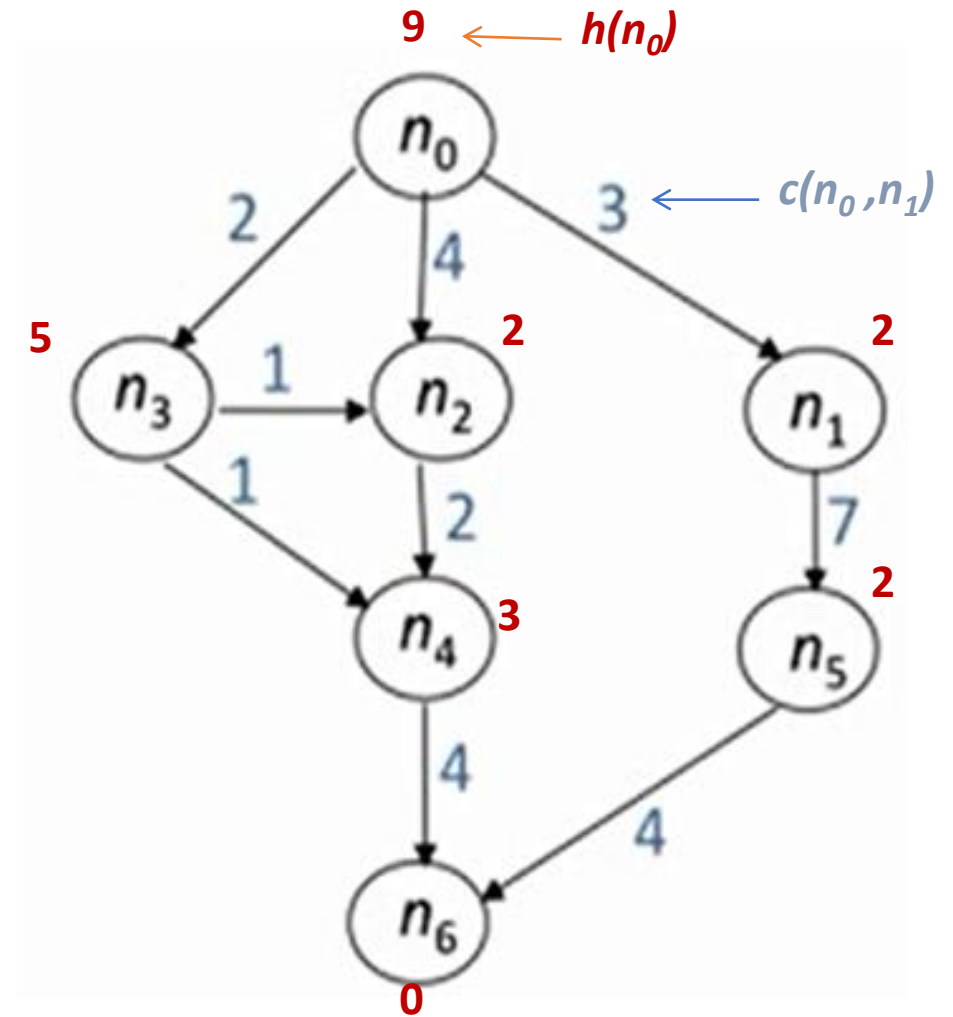
- Greedy search will choose the state that seems **closest** to a final state according to the **estimation function h** .

Greedy Best-First Search

Open List :

- $(n_0, 9, \text{void})$
- $(n_2, 2, n_0), (n_1, 2, n_0), (n_3, 5, n_0)$
- $(n_1, 2, n_0), (n_4, 3, n_2), (n_3, 5, n_0)$
- $(n_5, 2, n_1), (n_4, 3, n_2), (n_3, 5, n_0)$
- $(n_6, 0, n_5), (n_4, 3, n_2), (n_3, 5, n_0)$

Path : $n_0 \rightarrow n_1 \rightarrow n_5 \rightarrow n_6$



Heuristic-Based Search Algorithms

A* Search

- The utility measure is given by an evaluation function f
- For each node n : $f(n) = g(n) + h(n)$
 - $g(n)$ Is the cost till present to get n
 - $h(n)$ Is the estimated cost to go from n to the **goal node**.
 - $f(n)$ Is the total estimated cost to go from the **initial node** to the **goal node** going through n

h is said to be **admissible** if for all n : $h(n) \leq c(n)$

$c(n)$ being the real cost leading from n to the **final state**

A* Search Algorithm

1. Declare two nodes n, ns
2. Declare two lists **Open** and **Closed** (initially empty)
3. Add **initial node** to **Open**
4. **If Open** is empty **Then** Exit the loop with a **failure**
5. *Current node $n = \text{node at the head of Open}$*
6. Remove n from **Open** and add it to **Closed**.
7. **If $n = \text{goal}$** **Then** Exit the loop and **return the path**
Else : For each successor ns of n :
 - Initialize the value $g(ns) = g(n) + c(n, ns)$
 - Set parent of ns to n
 - **If *Open* or *Closed* contains a node $ns' = ns$ with $f(ns) \leq f(ns')$**
Then remove ns' from **Open** or **Closed** and insert ns into **Open** (with respect to the ascending order of f)
 - Else** : Insert ns into **Open** (with respect to the ascending order of f)
 - Go to 4.

A* Search Algorithm

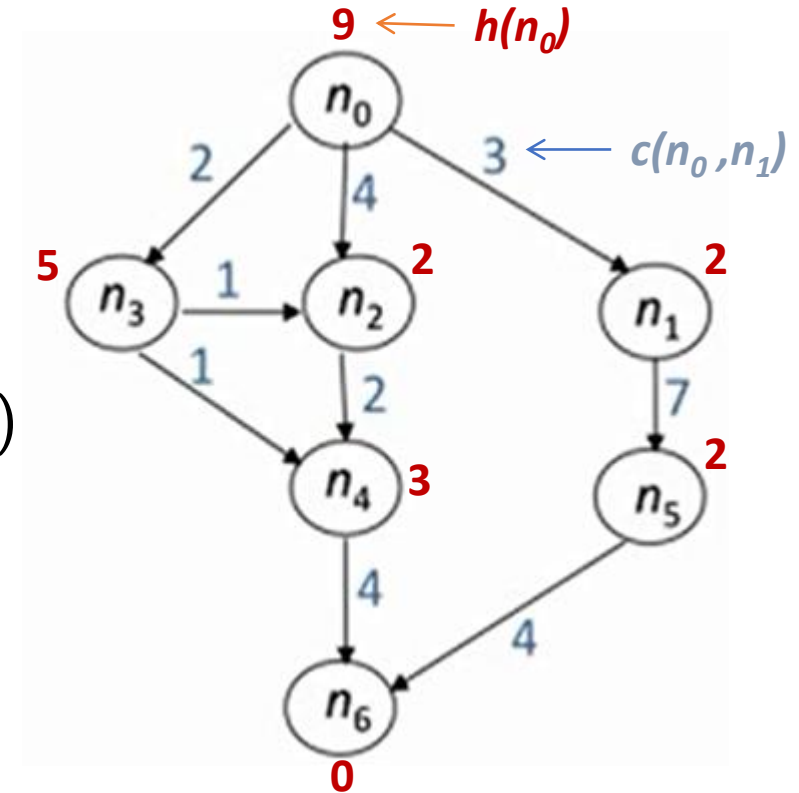
Illustrative example: Path-Finding between two cities

n_0 : Departure city (initial node)

n_6 : Destination city(goal node)

h : Direct distance between a city and the destination city (heuristic)

c : Real distance between two cities



A* Search Algorithm

Illustrative example: Path-Finding between two cities

State of Open in each iteration

(State, f, Parent)

1. $(n_0, 9, \text{void})$

2. $(n_1, 5, n_0), (n_2, 6, n_0), (n_3, 7, n_0)$

3. $(n_2, 6, n_0), (n_3, 7, n_0), (n_5, 12, n_1)$

4. $(n_3, 7, n_0), (n_4, 9, n_2), (n_5, 12, n_1)$

5. $(n_2, 5, n_3), (n_4, 6, n_3), (n_5, 12, n_1)$

6. $(n_4, 6, n_3), (n_5, 12, n_1)$

7. $(n_6, 7, n_4), (n_5, 12, n_1)$

8. Solution : n_0, n_3, n_4, n_6

State of Closed in each iteration

(State, f, Parent)

1. Vide

2. $(n_0, 9, \text{void})$

3. $(n_0, 9, \text{void}), (n_1, 5, n_0)$

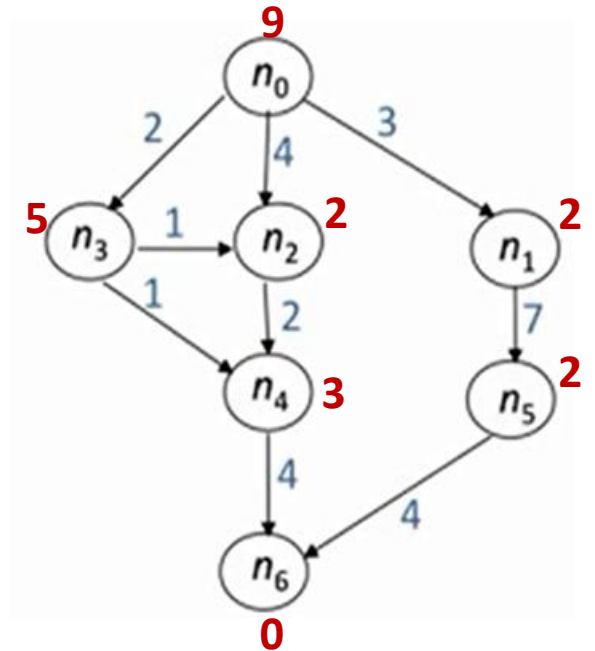
4. $(n_0, 9, \text{void}), (n_1, 5, n_0), (n_2, 6, n_0)$

5. $(n_0, 9, \text{void}), (n_1, 5, n_0), (n_3, 7, n_0)$

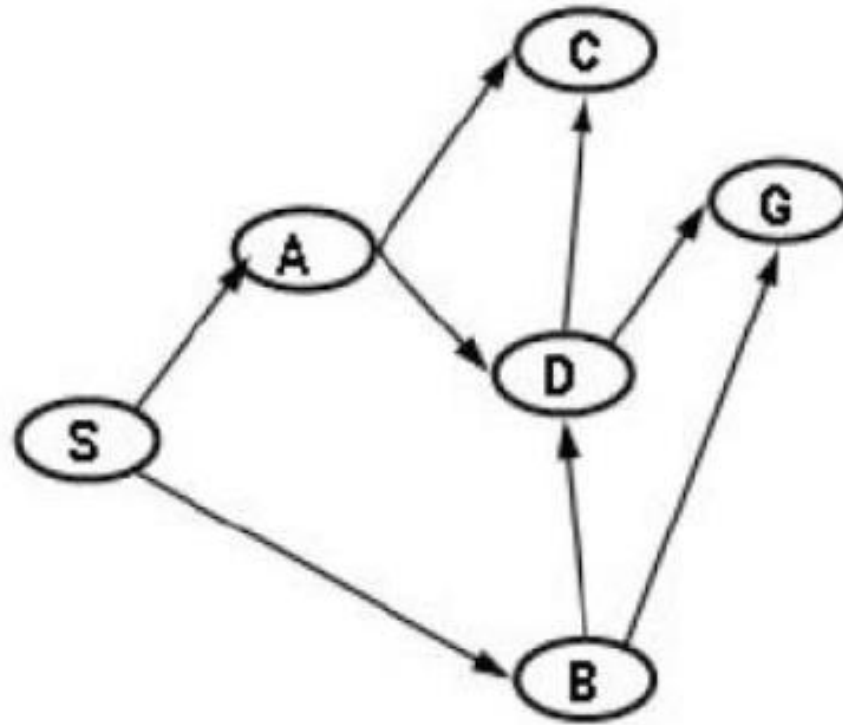
6. $(n_0, 9, \text{void}), (n_1, 5, n_0), (n_3, 7, n_0), (n_2, 5, n_3)$

7. $(n_0, 9, \text{void}), (n_1, 5, n_0), (n_3, 7, n_0), (n_2, 5, n_3), (n_4, 6, n_3)$

8. $(n_0, 9, \text{void}), (n_1, 5, n_0), (n_3, 7, n_0), (n_2, 5, n_3), (n_4, 6, n_3), (n_6, 7, n_4)$

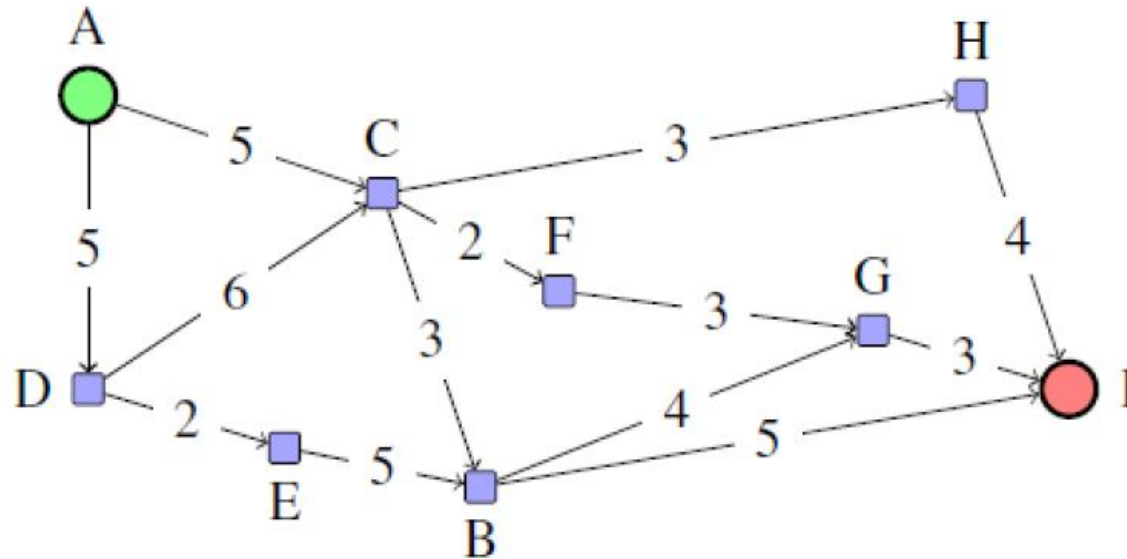


Exercise n°2: Transform the following graph into a search tree, then apply a breadth-first search followed by a depth-first search to find the state **G** from **S**. In case of conflicts between nodes, follow alphabetical order.



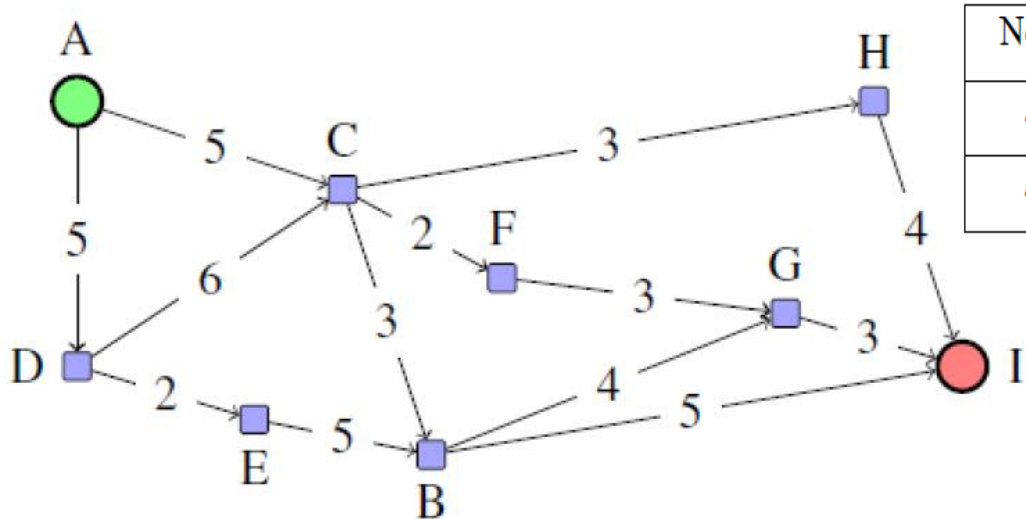
Exercise n° 3

We consider the following map. The objective is to find the optimal path between A and I. We also give two heuristics $h1$ and $h2$:



Nœud	A	B	C	D	E	F	G	H	I
$h1$	10	5	5	10	10	3	3	3	0
$h2$	10	2	8	11	6	2	1	5	0

1. Apply the strategies BFS, DFS and Uniform-Cost
2. Find the optimal (we minimize) path using the following algorithms:
 - a) Greedy Best-First Search using $h2$ as heuristic function
 - b) A* Search using $h1$



Nœud	A	B	C	D	E	F	G	H	I
<i>h1</i>	10	5	5	10	10	3	3	3	0
<i>h2</i>	10	2	8	11	6	2	1	5	0

1) Greedy Best-First Search using h2

[(A,10,void)]

[(C,8,A),(D,11,A)]

[(B,2,C),(F,2,C),(H,5,C),(D,11,A)]

[(I,0,B),(G,1,B),(F,2,C),(H,5,C)(D,11,A)]

The optimal path: A → C → B → I

2) A* Search using h1

[(A,10,void)]

[(C,10,A),(D,15,A)]

[(F,10,C), (H,11,C), (B,13,C),(D,15,A)]

[(H,11,C), (B,13,C), (G,13,F),(D,15,A)]

[(I,12,H)(B,13,C), (G,13,F),(D,15,A)]

Le chemin traversé est : A → C → H → I