

# Solving Problems by Searching in Python

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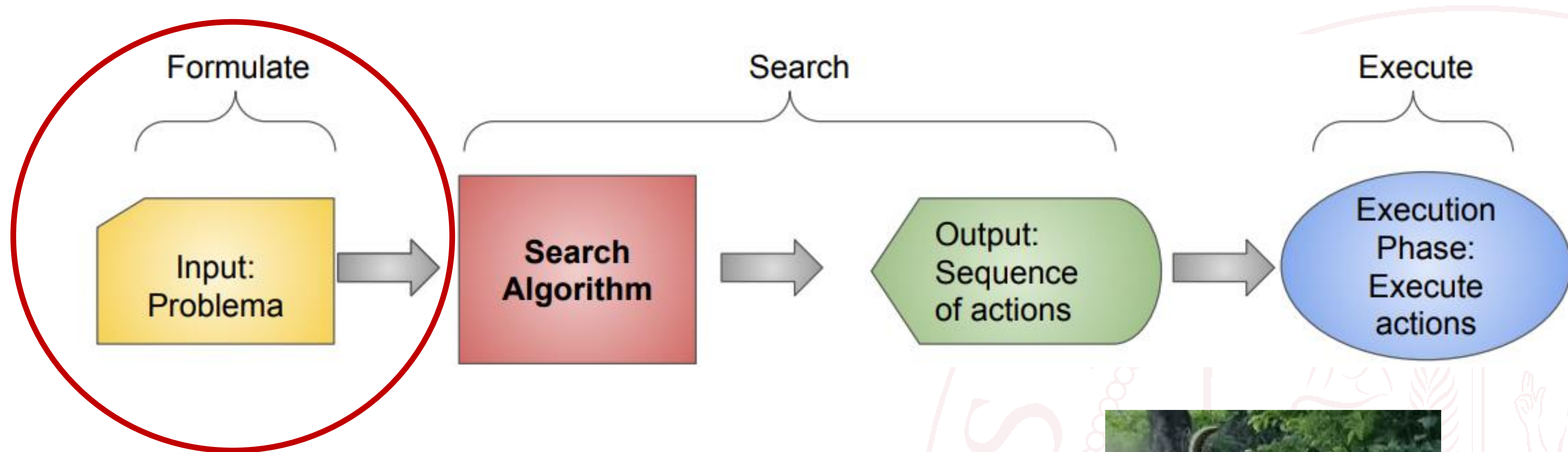
## Topics:

- Problem Formulation
- Search Tree
- Node
- Queues
- Best-First Search
- Greedy Best-First Search,  $A^*$ , Weighted  $A^*$ , Uniform-cost
- Breadth-First Search
- Iterative-Deepening Search



# Introduction to Search

Search can be defined as the steps to find a sequence of actions aiming the goal state



# Problem Formulation

- **Initial state:** the state of the agent starts in the beginning.
- **Actions:** set of actions applicable in the current state.

**ACTIONS(s):** return possible actions applicable in state  $s$ .

- **Transition model:** define the consequence of the each action application over the state.

**RESULT(s, a):** returns the state achieve after applying action  $a$  over state  $s$ .

- **State space:**
  - a) It is defined by the initial state, actions, and transition model.
  - b) It has all states which can be obtained by applying a sequence of actions.



# Problem Formulation

- **Goal test:** verify when a goal state is reached.

**IS\_GOAL(s):** return True when s is equal to the goal

- **Path cost function:** measures the solution quality by evaluating each path cost.



The path is the sequence of states within the state space provided by a sequence of actions

$c(s,a,s')$ : step cost to reach state  $s'$  from state  $s$  by applying action  $a$ .

**ACTION\_COST(s,a,s')**

The optimal solution will have the lowest path cost.

# class: Problem

```
class Problem(object):
    """The abstract class for a formal problem. A new domain subclasses this,
    overriding `actions` and `results`, and perhaps other methods.
    The default heuristic is 0 and the default action cost is 1 for all states.
    When you create an instance of a subclass, specify `initial`, and `goal` states
    (or give an `is_goal` method) and perhaps other keyword args for the subclass."""

    def __init__(self, initial=None, goal=None, **kwds):
        self.__dict__.update(initial=initial, goal=goal, **kwds)

    def actions(self, state):
        raise NotImplementedError
    def result(self, state, action):
        raise NotImplementedError
    def is_goal(self, state):
        return state == self.goal
    def action_cost(self, s, a, s1):
        return 1
    def h(self, node):
        return 0

    def __str__(self):
        return '{}({!r}, {!r})'.format(
            type(self).__name__, self.initial, self.goal)
```

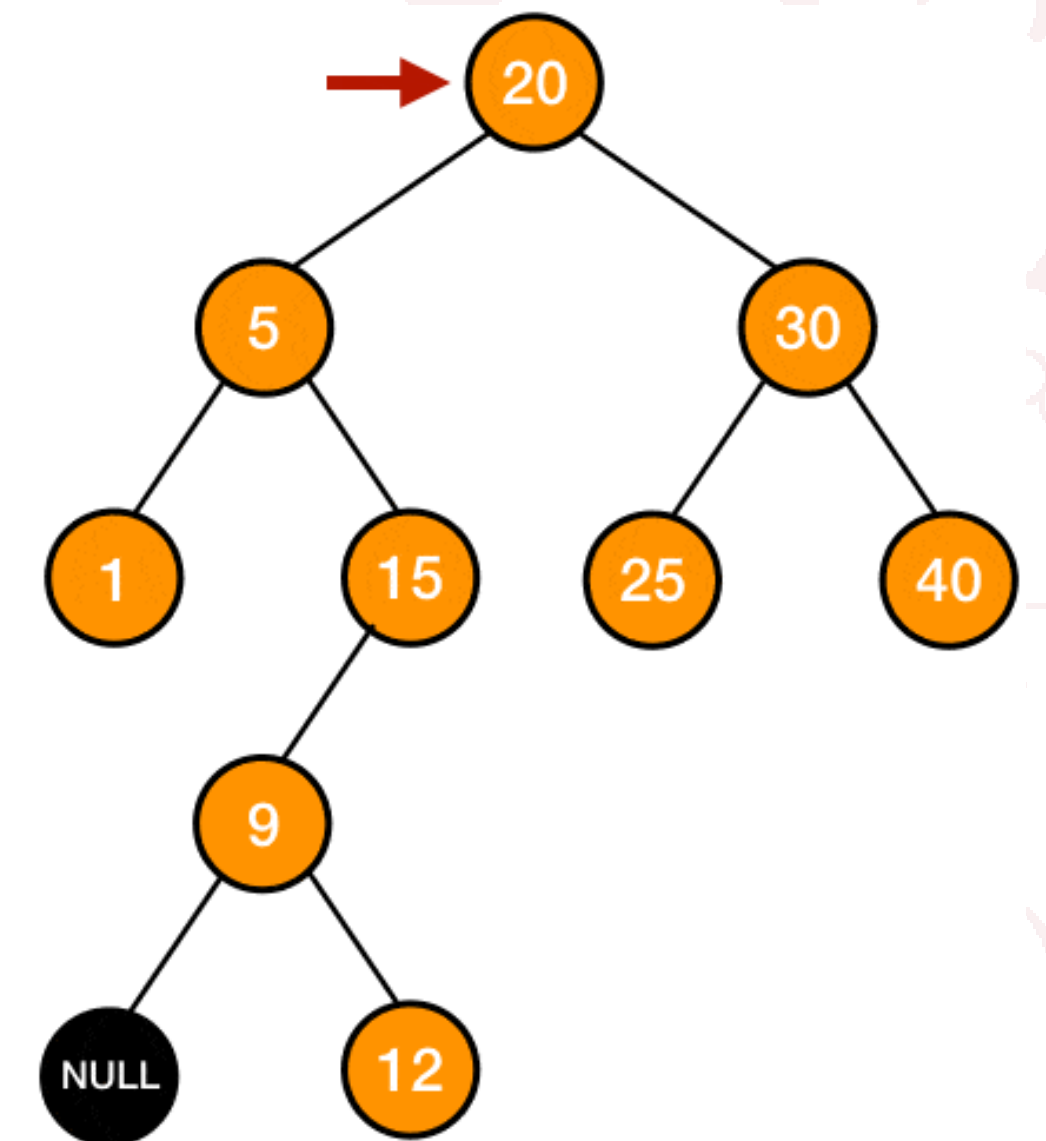
# Searching via Search Tree

we consider algorithms that superimpose a **search tree** onto the state space graph, forming various **paths starting from the initial state** and trying to **find one that reaches a target state**.

**Each node** in the search tree corresponds to a **state in the state space**,

and **branches** in the search tree correspond to **actions**.

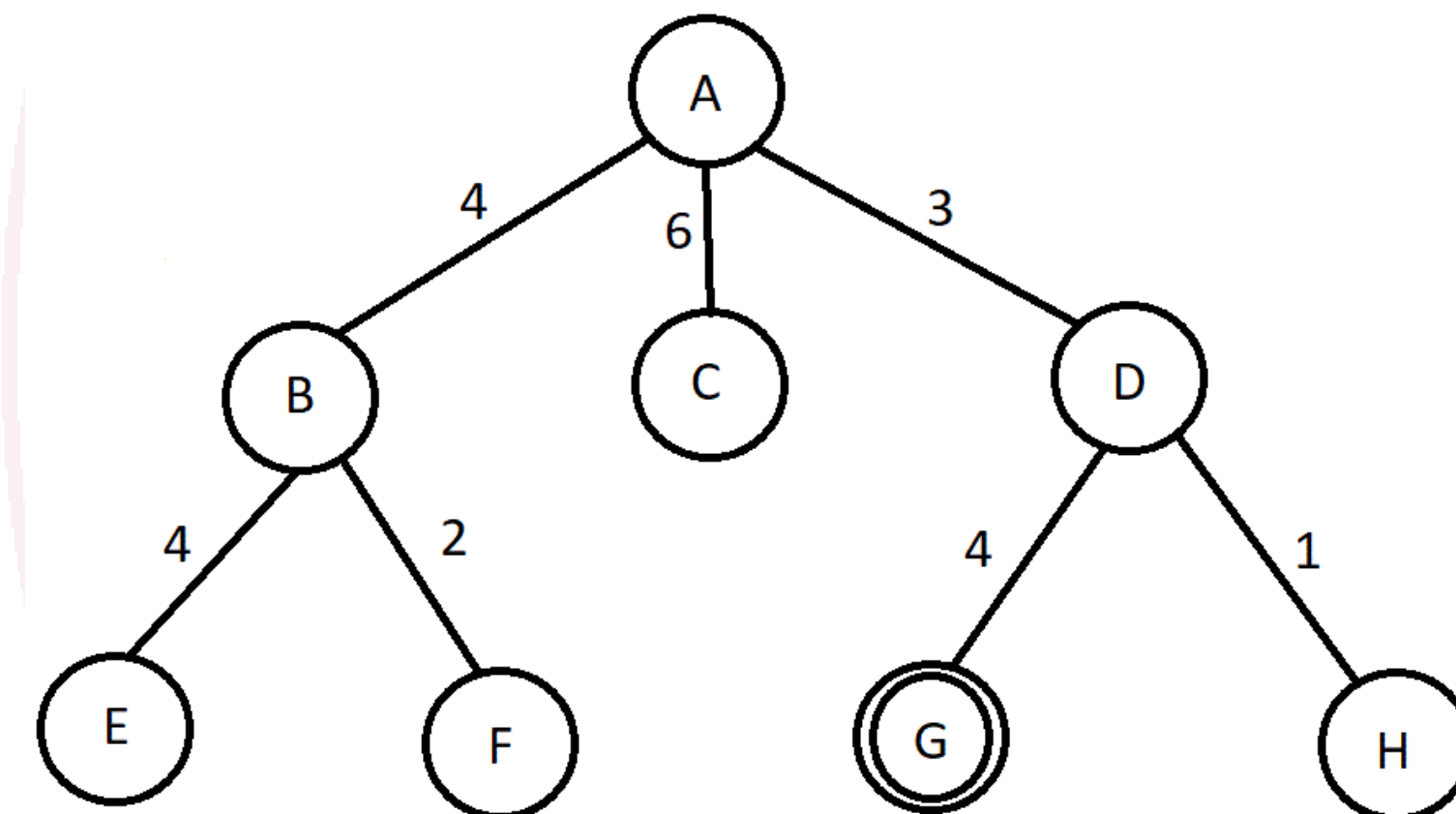
The **root of the tree** corresponds to the **initial state** of the problem.



# class: Node

```
class Node:
    "A Node in a search tree."
    def __init__(self, state, parent=None, action=None, path_cost=0):
        self.__dict__.update(state=state, parent=parent, action=action, path_cost=path_cost)

    def __repr__(self): return '<{}>'.format(self.state)
    def __len__(self): return 0 if self.parent is None else (1 + len(self.parent))
    def __lt__(self, other): return self.path_cost < other.path_cost
```



# class: Node

```
failure = Node('failure', path_cost=math.inf) # Indicates an algorithm couldn't find a solution.  
cutoff  = Node('cutoff',  path_cost=math.inf) # Indicates iterative deepening search was cut off.
```

```
def expand(problem, node):  
    "Expand a node, generating the children nodes."  
    s = node.state  
    for action in problem.actions(s):  
        s1 = problem.result(s, action)  
        cost = node.path_cost + problem.action_cost(s, action, s1)  
        yield Node(s1, node, action, cost)
```

```
def path_actions(node):  
    "The sequence of actions to get to this node."  
    if node.parent is None:  
        return []  
    return path_actions(node.parent) + [node.action]
```

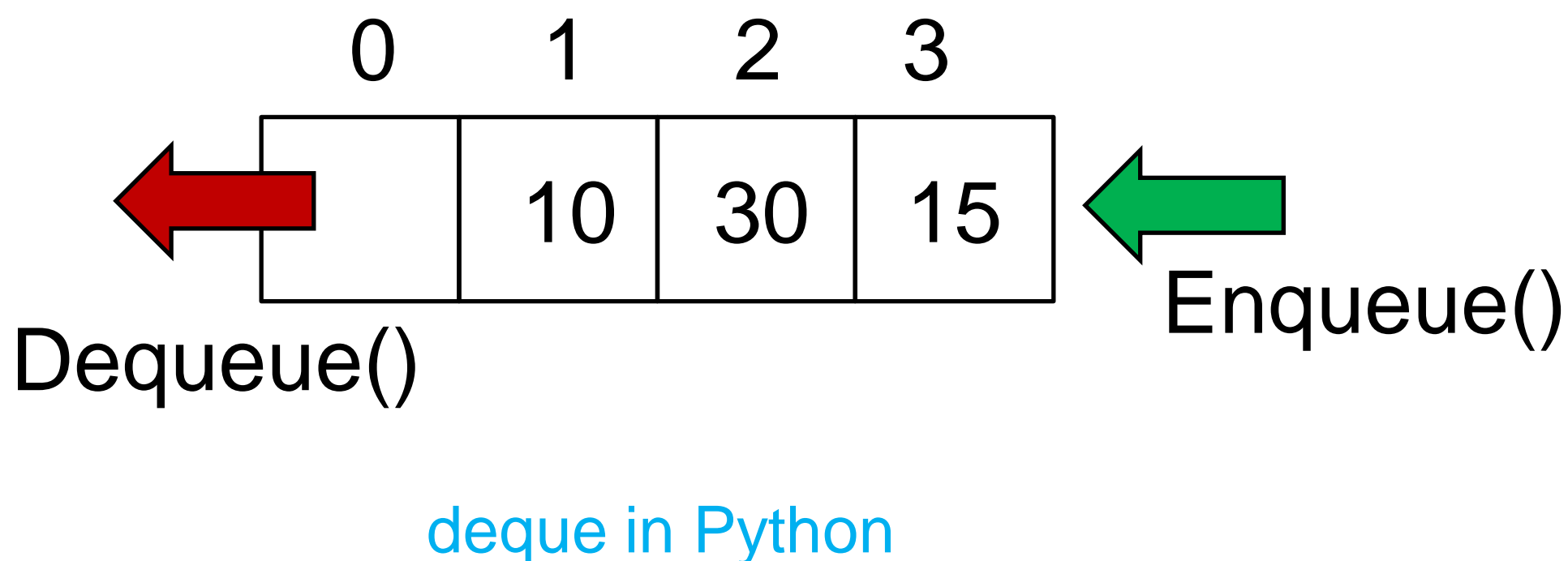
```
def path_states(node):  
    "The sequence of states to get to this node."  
    if node in (cutoff, failure, None):  
        return []  
    return path_states(node.parent) + [node.state]
```



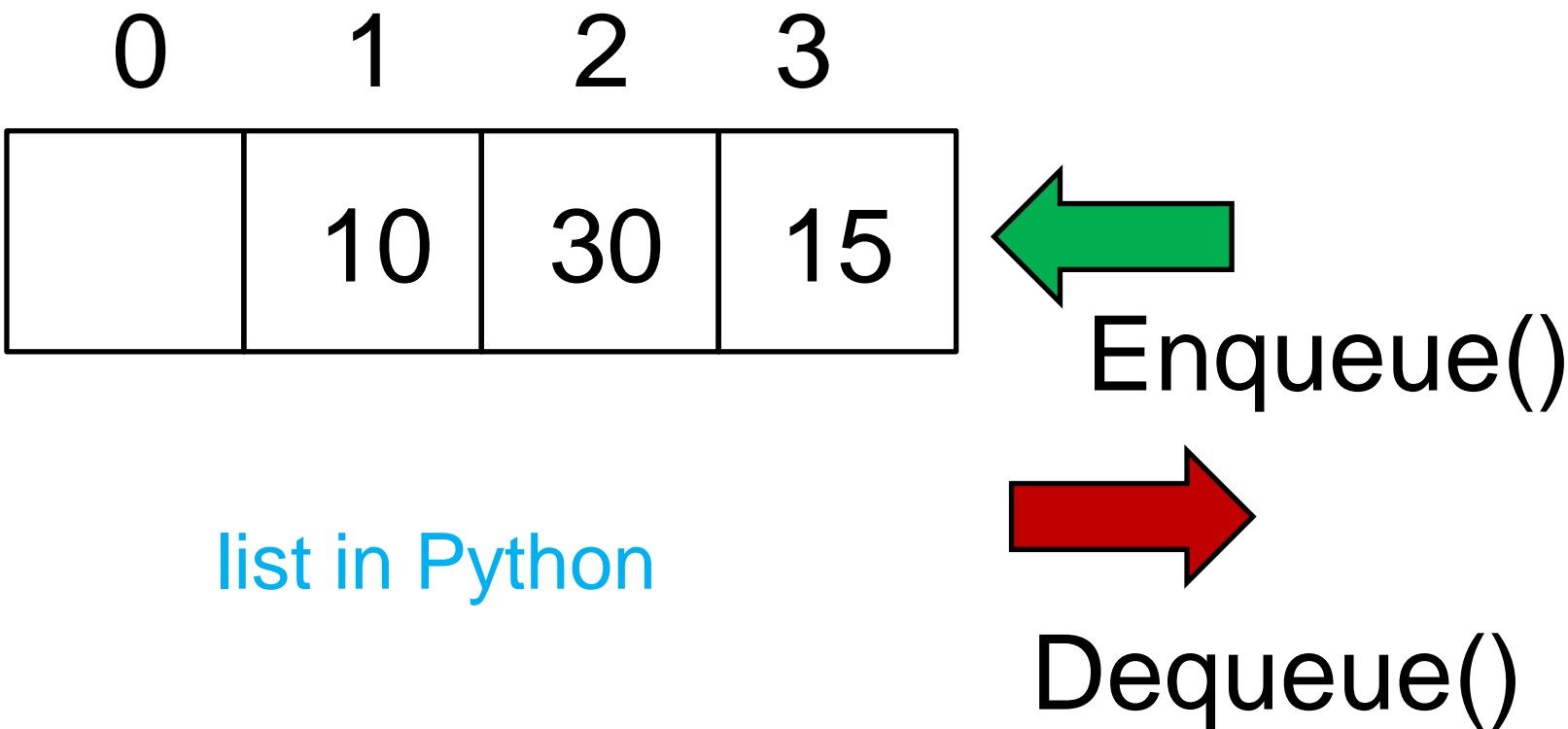
# Queues for Search Algorithms Implementation

In the search algorithms, we use three kinds of queues:

First-In, First-Out (FIFO)

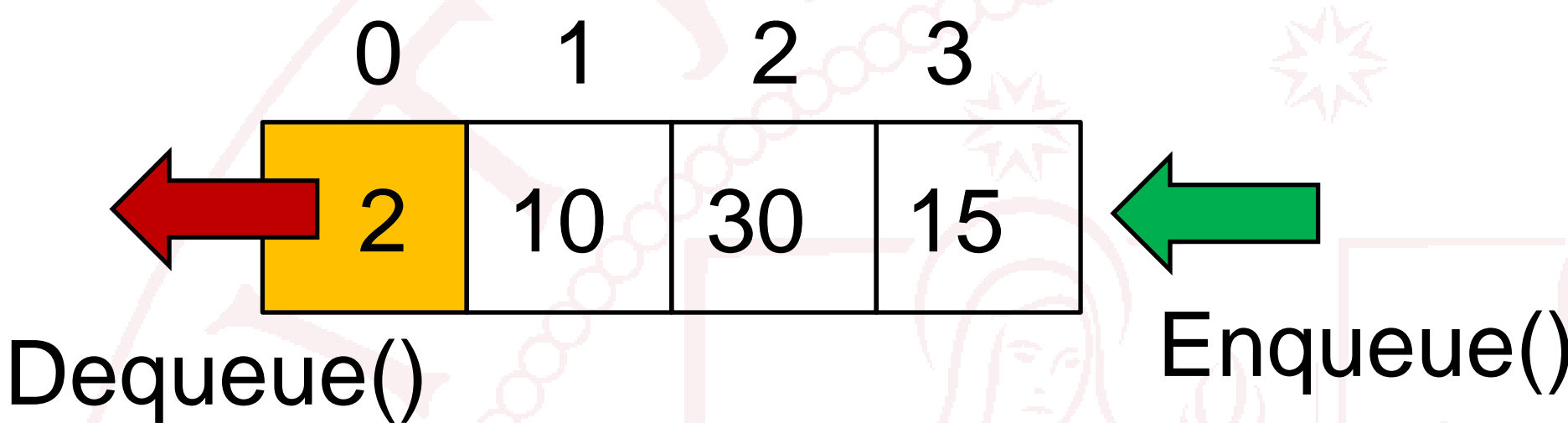


Last-In, First-Out (LIFO)



Priority Queue

Element with the highest priority



DECREASING PRIORITY ORDER

PriorityQueue class based on heap in Python

# Best-First Search (BFS): PseudoCode

```
function BEST-FIRST-SEARCH(problem, f) returns a solution node or failure
  node  $\leftarrow$  NODE(STATE=problem.INITIAL)
  frontier  $\leftarrow$  a priority queue ordered by f, with node as an element
  reached  $\leftarrow$  a lookup table, with one entry with key problem.INITIAL and value node
  while not IS-EMPTY(frontier) do
    node  $\leftarrow$  POP(frontier)
    if problem.IS-GOAL(node.STATE) then return node
    for each child in EXPAND(problem, node) do
      s  $\leftarrow$  child.STATE
      if s is not in reached or child.PATH-COST < reached[s].PATH-COST then
        reached[s]  $\leftarrow$  child
        add child to frontier
  return failure

function EXPAND(problem, node) yields nodes
  s  $\leftarrow$  node.STATE
  for each action in problem.ACTIONS(s) do
    s'  $\leftarrow$  problem.RESULT(s, action)
    cost  $\leftarrow$  node.PATH-COST + problem.ACTION-COST(s, action, s')
    yield NODE(STATE=s', PARENT=node, ACTION=action, PATH-COST=cost)
```



# Best-First Search in Python

```
def best_first_search(problem, f):  
    "Search nodes with minimum f(node) value first."  
    node = Node(problem.initial)  
    frontier = PriorityQueue([node], key=f)  
    reached = {problem.initial: node}  
    while frontier:  
        node = frontier.pop()  
        if problem.is_goal(node.state):  
            return node  
        for child in expand(problem, node):  
            s = child.state  
            if s not in reached or child.path_cost < reached[s].path_cost:  
                reached[s] = child  
                frontier.add(child)  
    return failure
```

each child node is added to the frontier if it has not been reached previously or is added again if it has now been reached with a lower cost path than the previous ones

# Best-First Search in Python

```
def best_first_search(problem, f):  
    "Search nodes with minimum f(node) value first."  
    node = Node(problem.initial)  
    frontier = PriorityQueue([node], key=f)  
    reached = {problem.initial: node}  
    while frontier:  
        node = frontier.pop()  
        if problem.is_goal(node.state):  
            return node  
        for child in expand(problem, node):  
            s = child.state  
            if s not in reached or child.path_cost < reached[s].path_cost:  
                reached[s] = child  
                frontier.add(child)  
    return failure
```



# From best\_first\_search to other search algorithms

```
def best_first_search(problem, f):  
    "Search nodes with minimum f(node) value first."  
    node = Node(problem.initial)  
    frontier = PriorityQueue([node], key=f)  
    reached = {problem.initial: node}  
    while frontier:  
        node = frontier.pop()  
        if problem.is_goal(node.state):  
            return node  
        for child in expand(problem, node):  
            s = child.state  
            if s not in reached or child.path_cost < reached[s].path_cost:  
                reached[s] = child  
                frontier.add(child)  
    return failure
```

We will modify the evaluation function  $f(n)$

# Breadth-First-BFS & Depth-First-BFS

```
def breadth_first_bfs(problem):  
    "Search shallowest nodes in the search tree first; using best-first."  
    return best_first_search(problem, f=len)
```

```
def depth_first_bfs(problem):  
    "Search deepest nodes in the search tree first; using best-first."  
    return best_first_search(problem, f=lambda n: -len(n))
```

Check how len is computed in the Node class

# Greedy best-first search in Python

It is possible to apply the [Best-First Search](#) algorithm

The evaluation function is  $f(n) = h(n)$

$h(n)$  = estimated cost to goal from  $n$  (i.e., the closest to reach to the goal)

```
def greedy_bfs(problem, h=None):  
    """Search nodes with minimum h(n)."""  
    h = h or problem.h  
    return best_first_search(problem, f=h)
```



# A\* Search in Python

It is possible to apply the Best-First Search algorithm

The evaluation function is  $f(n) = g(n) + h(n)$ :

$g(n)$  = cost so far to reach  $n$

$h(n)$  = estimated cost to goal from  $n$

```
def g(n): return n.path_cost
```

```
class Node:
    "A Node in a search tree."
    def __init__(self, state, parent=None, action=None, path_cost=0):
        self.__dict__.update(state=state, parent=parent, action=action, path_cost=path_cost)

    def __repr__(self): return '<{}>'.format(self.state)
    def __len__(self): return 0 if self.parent is None else (1 + len(self.parent))
    def __lt__(self, other): return self.path_cost < other.path_cost
```



# A\* Search in Python

It is possible to apply the [Best-First Search](#) algorithm

The evaluation function is  $f(n) = g(n) + h(n)$ :

$g(n)$  = cost so far to reach  $n$

$h(n)$  = estimated cost to goal from  $n$

```
def astar_search(problem, h=None):  
    """Search nodes with minimum  $f(n) = g(n) + h(n)$ ."""  
    h = h or problem.h  
    return best_first_search(problem, f=lambda n: g(n) + h(n))
```



# Weighted A\* Search in Python

It is possible to apply the [Best-First Search](#) algorithm

The evaluation function is  $f(n) = g(n) + \text{weight} * h(n)$ :

$g(n)$  = cost so far to reach  $n$

$h(n)$  = estimated cost to goal from  $n$

```
def weighted_astar_search(problem, h=None, weight=1.4):  
    """Search nodes with minimum  $f(n) = g(n) + \text{weight} * h(n)$ ."""  
    h = h or problem.h  
    return best_first_search(problem, f=lambda n: g(n) + weight * h(n))
```

# Uniform-cost Search in Python

It is possible to apply the [Best-First Search](#) algorithm

The evaluation function is  $f(n) = g(n)$

$g(n)$  = cost so far to reach  $n$

```
def uniform_cost_search(problem):  
    "Search nodes with minimum path cost first."  
    return best_first_search(problem, f=g)
```



# Breadth-First Search: PseudoCode

**function** BREADTH-FIRST-SEARCH(*problem*) **returns** a solution node or *failure*

*node*  $\leftarrow$  NODE(*problem*.INITIAL)

**if** *problem*.IS-GOAL(*node*.STATE) **then return** *node*

*frontier*  $\leftarrow$  a FIFO queue, with *node* as an element

*reached*  $\leftarrow$  {*problem*.INITIAL}

**while not** IS-EMPTY(*frontier*) **do**

*node*  $\leftarrow$  POP(*frontier*)

**for each** *child* **in** EXPAND(*problem*, *node*) **do**

*s*  $\leftarrow$  *child*.STATE

**if** *problem*.IS-GOAL(*s*) **then return** *child*

**if** *s* is not in *reached* **then**

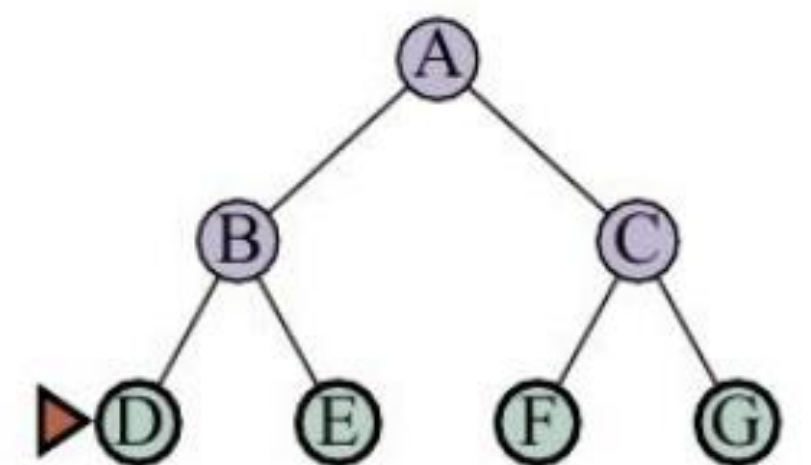
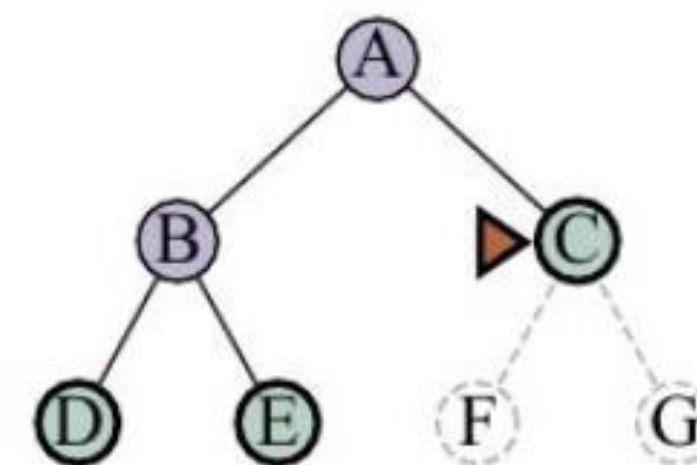
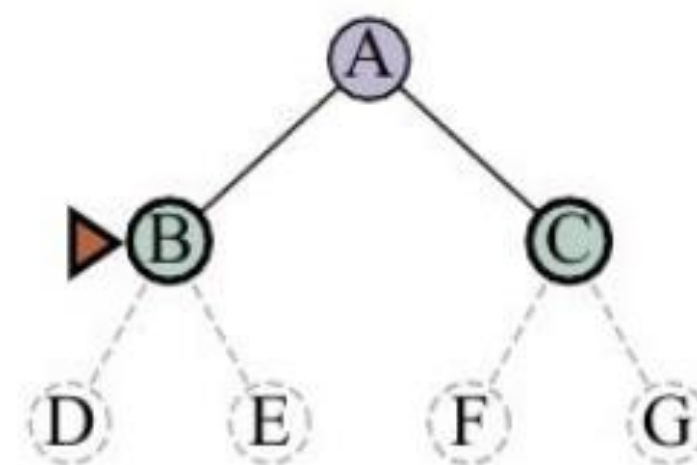
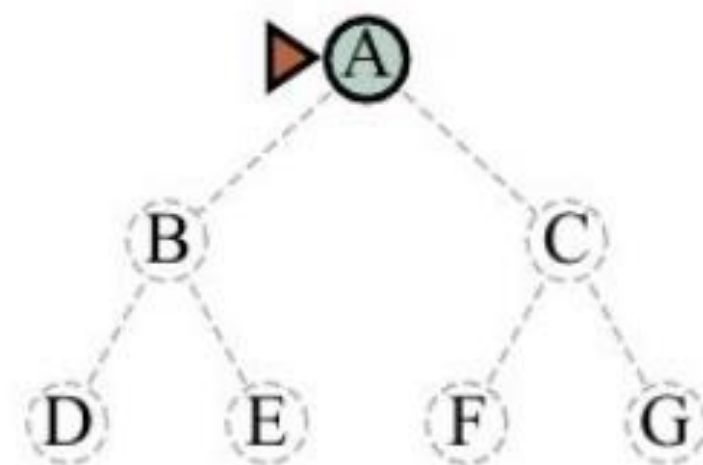
add *s* to *reached*

add *child* to *frontier*

**return** *failure*

NB: You add  
to the left

■ Frontier = {A}  $\rightarrow$  Frontier = {B, C}  $\rightarrow$  Frontier = {D, E, C}  $\rightarrow$  Frontier = {D, E, F, G}





# Breadth-First Search in Python

```
def breadth_first_search(problem):  
    "Search shallowest nodes in the search tree first."  
    node = Node(problem.initial)  
    if problem.is_goal(problem.initial):  
        return node  
    frontier = FIFOQueue([node])  
    reached = {problem.initial}  
    while frontier:  
        node = frontier.pop()  
        for child in expand(problem, node):  
            s = child.state  
            if problem.is_goal(s):  
                return child  
            if s not in reached:  
                reached.add(s)  
                frontier.appendleft(child)  
    return failure
```

deque: deque([1, 2, 3])

The deque after appending at right is :  
deque([1, 2, 3, 4])

The deque after appending at left is :  
deque([6, 1, 2, 3, 4])

# Iterative-Deepening Search: PseudoCode

**function** ITERATIVE-DEEPENING-SEARCH(*problem*) **returns** a solution node or *failure*

**for** *depth* = 0 to  $\infty$  **do**

*result*  $\leftarrow$  DEPTH-LIMITED-SEARCH(*problem*, *depth*)

**if** *result*  $\neq$  *cutoff* **then return** *result*

**function** DEPTH-LIMITED-SEARCH(*problem*,  $\ell$ ) **returns** a node or *failure* or *cutoff*

*frontier*  $\leftarrow$  a LIFO queue (stack) with NODE(*problem*.INITIAL) as an element

*result*  $\leftarrow$  *failure*

**while not** IS-EMPTY(*frontier*) **do**

*node*  $\leftarrow$  POP(*frontier*)

**if** *problem*.IS-GOAL(*node*.STATE) **then return** *node*

**if** DEPTH(*node*) >  $\ell$  **then**

*result*  $\leftarrow$  *cutoff*

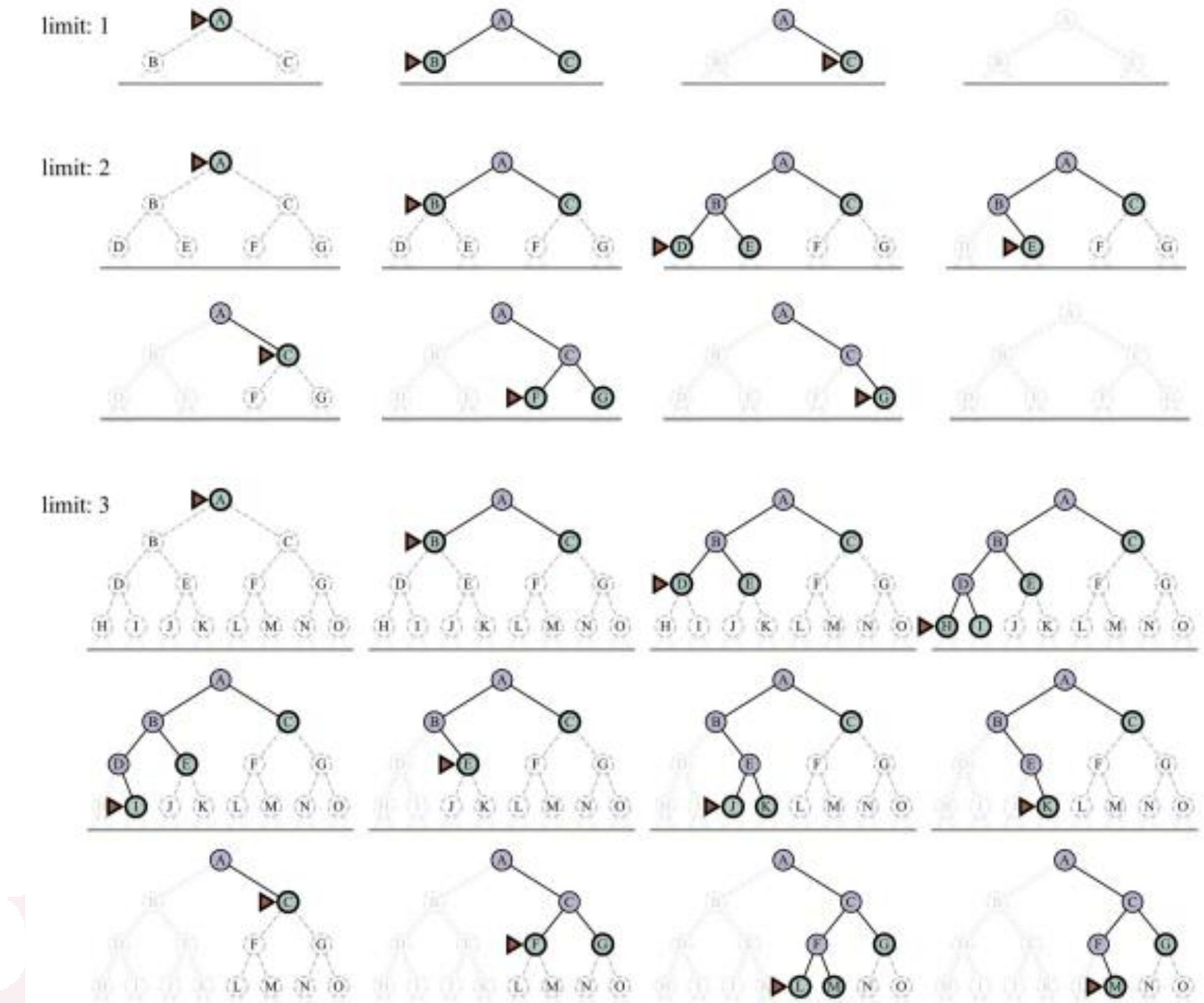
**else if not** IS-CYCLE(*node*) **do**

**for each** *child* **in** EXPAND(*problem*, *node*) **do**

                add *child* to *frontier*

**return** *result*

iteratively increasing the limit



# Iterative-Deepening Search in Python

```
def iterative_deepening_search(problem):
    "Do depth-limited search with increasing depth limits."
    for limit in range(1, sys.maxsize):
        result = depth_limited_search(problem, limit)
        if result != cutoff:
            return result

def depth_limited_search(problem, limit=10):
    "Search deepest nodes in the search tree first."
    frontier = LIFOQueue([Node(problem.initial)])
    result = failure
    while frontier:
        node = frontier.pop()
        if problem.is_goal(node.state):
            return node
        elif len(node) >= limit:
            result = cutoff
        elif not is_cycle(node):
            for child in expand(problem, node):
                frontier.append(child)
    return result
```

# Questions

