

ARTIFICIAL INTELLIGENCE



Solving search problems

Uninformed search strategies

Content

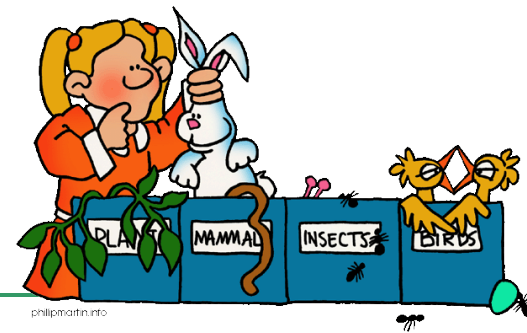
- Problems
- Problem solving
 - Steps of problem solving
- Solving problem by search
 - Steps of solving problem by search
 - Search strategies

Problems



- Two problem types:
 - Solving in a deterministic manner
 - Computing the sinus of an angle or the square root of a number
 - Solving in a stochastic manner
 - Real-world problems → design of ABS
 - Involve the search of a solution → AI's methods

Problems



□ Tipology

■ Search/optimisation problems

- Planning, satellite's design



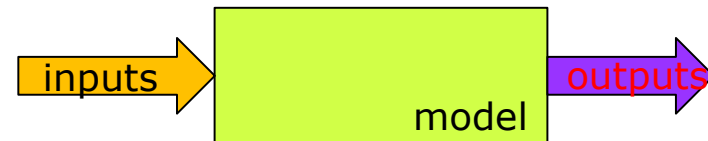
■ Modeling problems

- Predictions, classifications



■ Simulation problems

- Game theory



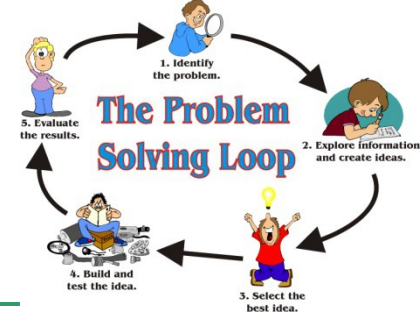
Problem solving



- Identification of a solution
 - In computer science (AI) → search process
 - In engineering and mathematics → optimisation process

- How?
 - Representation of (partial) solutions → points in the search space
 - Design of a search operators → map a potential solution into another one

Steps in problem solving



- ❑ Problem definition
- ❑ Problem analyses
- ❑ Selection of a solving technique
 - Search
 - Knowledge representation
 - Abstraction

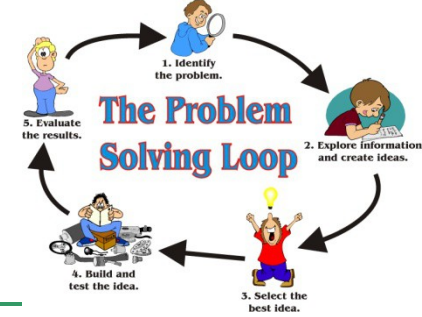


Solving problems by search

- Based on some objectives
- Composed by actions that accomplish the objectives
 - Each action changes a state of the problem
- More actions that map the initial state of problem into a final state

Steps in solving problems by search

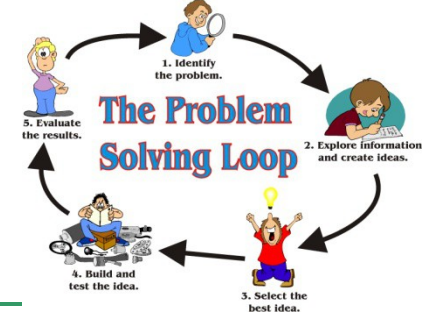
Problem definition



- Problem definition involves:
 - A search space
 - All possible states
 - Representation
 - Explicit – construction of all possible states
 - Default – by using some data structures and some functions (operators)
 - One or more initial state
 - One or more final states
 - One or more paths
 - More successive states
 - A set of rules (actions)
 - Successor functions (operators) – next state after a given one
 - Cost functions that evaluate
 - How a state is mapped into another state
 - An entire path
 - Objective functions that check if a state is final or not

Steps in solving problems by search

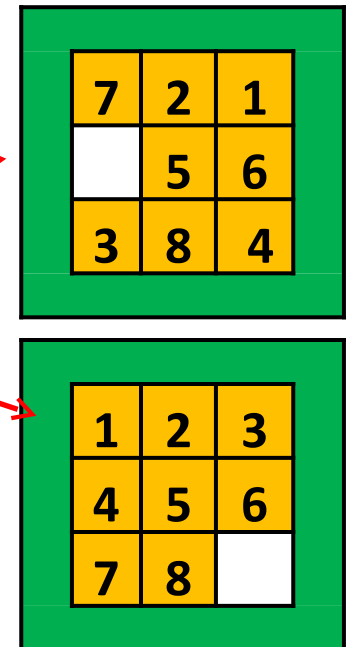
Problem definition



□ Examples

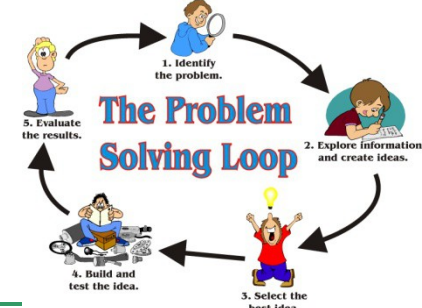
■ Puzzle game with 8 pieces

- State's space – different board configurations for a game with 8 pieces
- Initial state – a random configuration
- Final state – a configuration where all the pieces are sorted in a given manner
- Rules -> white moves
 - conditions: move inside the table
 - Transformations: the white space is moved up, down, to left or to right
- Solution - optimal sequence of white moves



Steps in solving problems by search

Problem definition



□ Examples

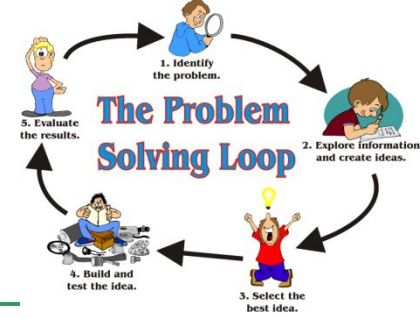
■ Queen's problem

- State's space – different board configurations for a game with n queens
- Initial state – a configuration without queens
- Final state – a configuration n queens so that none of them can hit any other in one move
- Rules -> put a queen on the table
 - conditions: the queen is not hit by any other queen
 - Transformations: put a new queen in a free cell of the table
- Solution - optimal placement of queens

	a	b	c	d	e	f	g	h	
1				♛					1
2							♛		2
3			♛						3
4								♛	4
5		♛							5
6					♛				6
7		♛							7
8						♛			8
	a	b	c	d	e	f	g	h	

Steps in solving problems by search

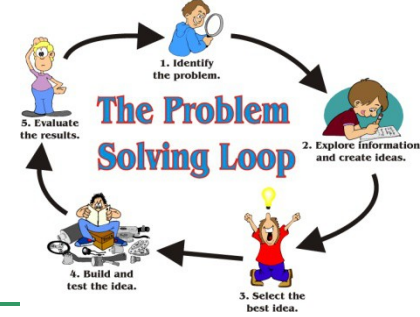
Problem analyse



- The problem can be decomposed?
 - The sub-problems are independent or not?
- The possible state's space is predictable?
- We want a solution or an optimal solution?
- The solution is represented by a single state or by more successive states?
- We require some knowledge for limiting the search or for identifying the solution?
- The problem is conversational or solitary?
 - Human interaction is required for problem solving?

Steps in solving problems by search

Selection of a solving technique

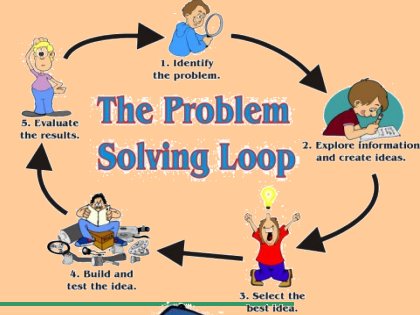


- Solving by moving rules (and control strategy) in the search space until we find a path from the initial state to the final state
- Solving by search
 - Examination of all possible states in order to identify
 - A path from the initial state to the final state
 - An optimal state
 - The search space = all possible states and the operators that maps the states



Steps in solving problems by search

Selection of a solving technique



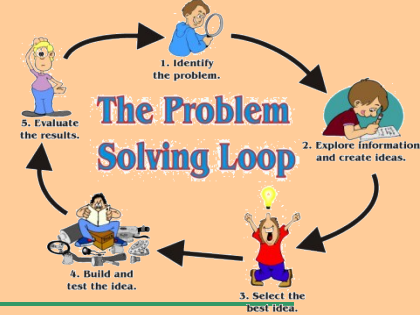
□ Solving by search

- More searching strategies → how we select one of them?
 - Computational complexity (temporal and spatial)
 - Completeness → the algorithms always ends and finds a solution (if it exists)
 - Optimal → the algorithms finds the optimal solution (the optimal cost of the path from the initial state to the final state)



Steps in solving problems by search

Selection of a solving technique



□ Solving by search

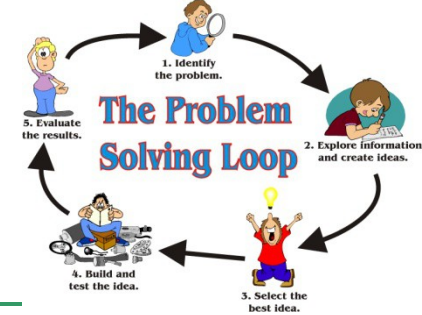
- More searching strategies → how we select one of them? → Computational complexity (temporal and spatial)
 - Strategy's performance depends on
 - Time for running
 - Memory for running
 - Size of input data
 - Computer's performance
 - Compiler's quality
 - Can be evaluated by complexity → computational efficiency
 - Spatial → required memory for solution identification
 - $S(n)$ – memory used by the best algorithms A that solves a decision problem f with n input data
 - Temporal → required time for solution identification
 - $T(n)$ – running time (number of steps) of the best algorithm A that solves a decision problem f with n input data

} Internal factors

} External factors

Steps in solving problems by search

Selection of a solving technique



■ Problem solving by search can be performed by:

■ Step by step construction of solution

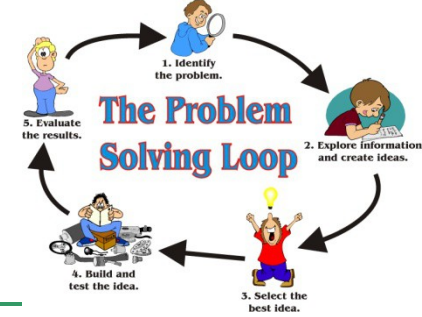


■ Optimal solution identification



Steps in solving problems by search

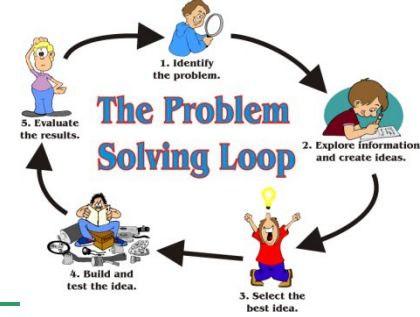
Selection of a solving technique



- Problem solving by search can be performed by:
 - Step by step construction of solution
 - Problem's components
 - Initial state
 - Operators (successor functions)
 - Final state
 - Solution = a path (of optimal cost) from the initial state to the final state
 - Search space
 - All the states that can be obtained from the initial state (by using the operators)
 - A state = a component of solution
 - Example
 - Traveling Salesman Problem (TSP)
 - Algorithms
 - Main idea: start with a solution's component and adding new components until a complete solution is obtained
 - Recurrent → until a condition is satisfied
 - The search's history (path from initial state to the final state) is retained in LIFO/FIFO containers
 - Advantages
 - Do not require knowledge (intelligent information)

Steps in solving problems by search

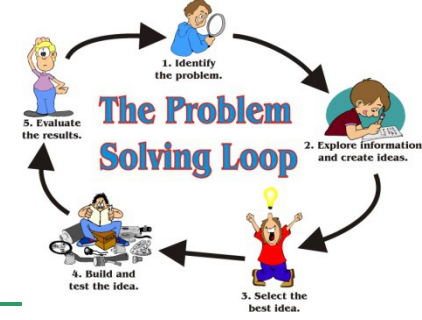
Selection of a solving technique



- Problem solving by search can be performed by:
 - Optimal solution identification
 - Problem's components
 - Conditions (constraints) that must be satisfied by the solution
 - Evaluation function for a potential solution → optimum identification
 - Search space
 - All possible and complete solutions
 - State = a complete solution
 - Example
 - Queen's problem
 - Algorithms
 - Main idea: start with a state that doesn't respect some conditions and change it for eliminating these violations
 - Iterative → a single state is retained and the algorithm tries to improve it
 - The searches history is not retained
 - Advantages
 - Simple
 - Requires a small memory
 - Can find good solutions in (continuous) search spaces very large (where other algorithms can not be utilised)

Steps in solving problems by search

Selection of a solving technique

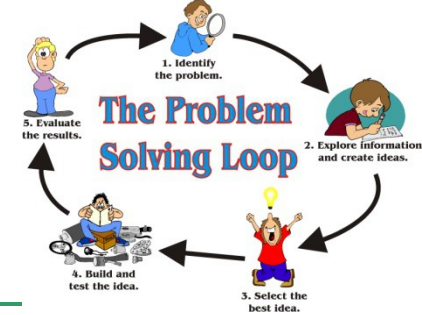


- Solving problem by search involves:
 - Very complex algorithms (NP-complete problems)
 - Search in an exponential space



Steps in solving problems by search

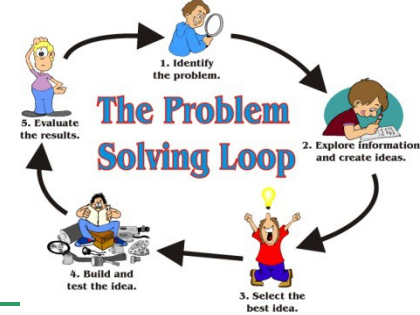
Selection of a solving technique



- Topology of search strategies:
 - Solution **generation**
 - **Constructive** search
 - Solution is identified step by step
 - Ex. TSP
 - **Perturbative** search
 - A possible solution is modified in order to obtain another possible solution
 - Ex. SAT - Propositional Satisfaction Problem
 - Search space **navigation**
 - **Systematic** search
 - The entire search space is visited
 - Solution identification (if it exists) → complete algorithms
 - **Local** search
 - Moving from a point of the search space into a neighbor point → incomplete algorithms
 - A state can be visited more times
 - **Certain** items of the search
 - **Deterministic** search
 - Algorithms that exactly identify the solution
 - **Stochastic** search
 - Algorithms that approximate the solution
 - Search space **exploration**
 - **Sequential** search
 - **Parallel** search

Steps in solving problems by search

Selection of a solving technique

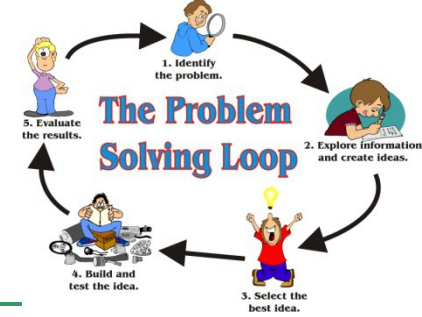


Topology of search strategies:

- **Number of objectives**
 - **Single-objective** search
 - The solution must respect a single condition/constraint
 - **Multi-objective** search
 - The solution must respect more conditions/constraints
- **Number of solutions**
 - **single-modal search**
 - There is a single optimal solution
 - **multi-modal search**
 - There are more optimal solutions
- **Algorithm**
 - Search over a **finite number of steps**
 - **Iterative search**
 - The algorithms converge through the optimal solutions
 - **Heuristic search**
 - The algorithms provide an approximation of the solution
- Search **mechanism**
 - **traditional search**
 - **modern search**
- where the search takes **place**
 - **local search**
 - **global search**

Steps in solving problems by search

Selection of a solving technique

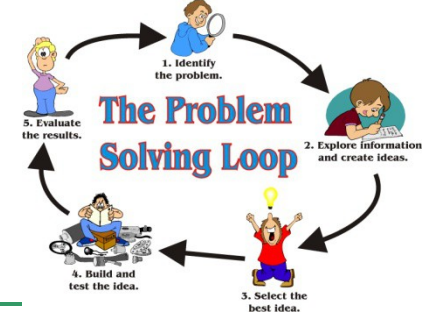


Topology of search strategies:

- Type **(linearity) of constraints**
 - **Linear search**
 - **non-linear search**
 - Classical (deterministic)
 - Direct – based on evaluation of the objective function
 - Indirect – based on derivative (I and/or II) of the objective function
 - Enumeration-based
 - How solution is identified
 - Uninformed – the solution is the final state
 - Informed – deals with an evaluation function for a possible solution
 - Search space type
 - Complete – the space is finite (if solution exists, then it can be found)
 - Incomplete – the space is infinite
 - Stochastic search
 - Based on random numbers
- **Agents** involves in search
 - Search by **a single agent** → without obstacle for achieving the objectives
 - **Adversarial search** → the opponent comes with some uncertainty

Steps in solving problems by search

Selection of a solving technique

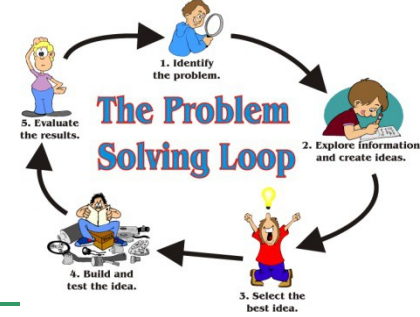


Example

- Topology of search strategies
 - Solution generation
 - **Constructive search**
 - Perturbative search
 - Search space navigation
 - **Systematic search**
 - Local search
 - Certain items of the search
 - **Deterministic search**
 - Stochastic search
 - Search space exploration
 - **Sequential search**
 - Parallel search

Steps in solving problems by search

Selection of a solving technique

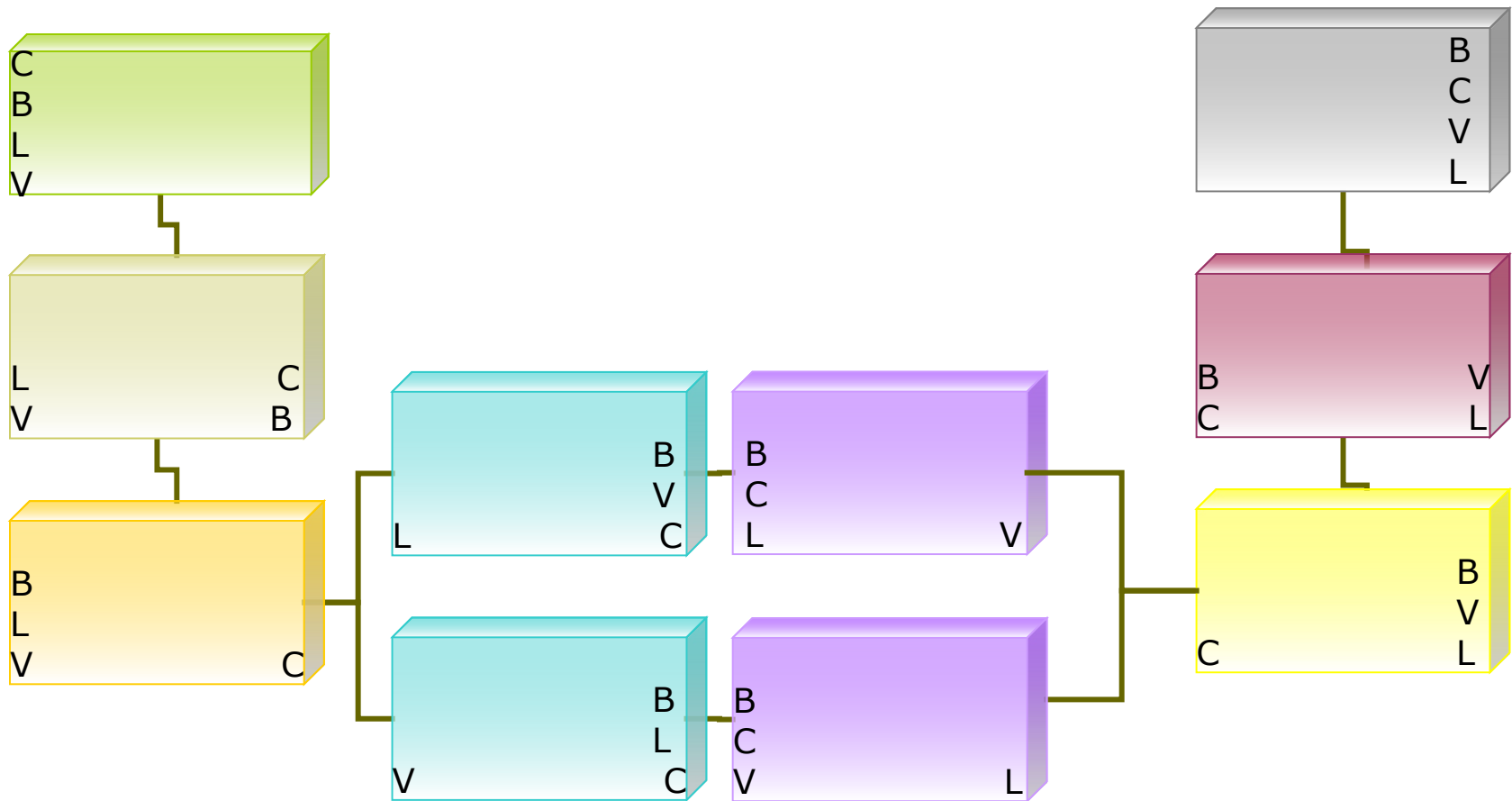
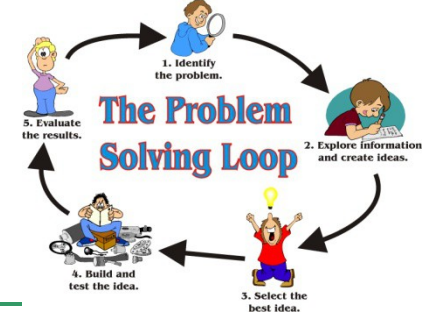


Example

- ❑ Constructive, global, determinist, sequential search
- ❑ Problem "capra, varza și lupul"
 - Input :
 - ❑ A goat, a cabbage and a wolf on a river-side
 - ❑ A boat with a boater
 - Output:
 - ❑ Move all the passengers on the other side of the river
 - ❑ Taking into account:
 - The boat has only 2 places
 - It is not possible to rest on the same side:
 - The goat and the cabbage
 - The wolf and the goat

Steps in solving problems by search

Selection of a solving technique



Search strategies – Basic elements



- Abstract data types (ADTs)
 - ADT list → linear structure
 - ADT tree → hierarchic structure
 - ADT graph → graph-based structure

- ADT
 - Domain and operations
 - Representation



Uninformed search strategies (USS)

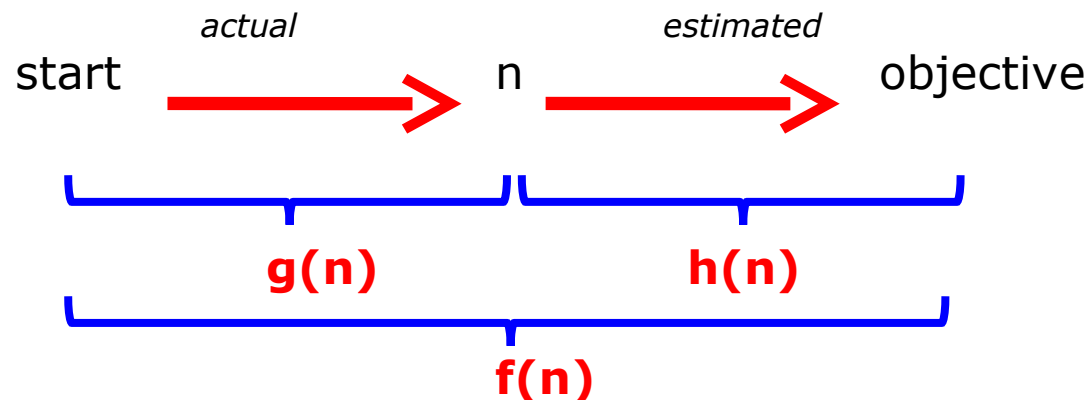
- Characteristics
 - Are NOT based on problem specific information
 - Are general
 - Blind strategies
 - Brute force methods
- Topology
 - Order of node exploration:
 - USS in linear structures
 - Linear search
 - Binary search
 - USS in non-linear structures
 - Breadth-first search
 - Uniform cost search (branch and bound)
 - Depth first search
 - Limited depth first search
 - iterative deepening depth-first search
 - Bidirectional search



SS in tree-based structures

□ Basic elements

- $f(n)$ – evaluation function for estimating the cost of a solution through node (state) n
- $h(n)$ – evaluation function for estimating the cost of a solution path from node (state) n to the final node (state)
- $g(n)$ – evaluation function for estimating the cost of a solution path from the initial node (state) to node (state) n
- $f(n) = g(n) + h(n)$



USS in tree-based structures

Breadth-first search – BFS

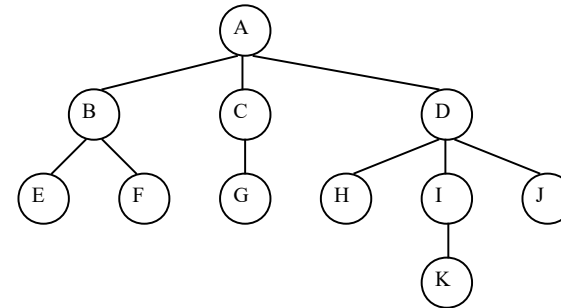


Basic elements

- All the nodes of depth d are visited before all the nodes of depth $d+1$
- All children of current node are added into a **FIFO** list (**queue**)

Exemplu

- Visiting order: A, B, C, D, E, F, G, H, I, J, K



Algorithm

```

bool BFS(elem, list){
    found = false;
    visited =  $\Phi$ ;
    toVisit = {start};          //FIFO list
    while((toVisit !=  $\Phi$ ) && (!found)){
        node = pop(toVisit);
        visited = visited  $\cup$  {node};
        if (node == elem)
            found = true;
        else{
            aux =  $\Phi$ ;
            for all (unvisited) children of node do
                aux = aux  $\cup$  {child};
            }
            toVisit = toVisit  $\cup$  aux;
        }
    }
    return found;
}
  
```

Vizitate deja	De vizitat
Φ	A
A	B, C, D
A, B	C, D, E, F
A, B, C	D, E, F, G
A, B, C, D	E, F, G, H, I, J
A, B, C, D, E	F, G, H, I, J
A, B, C, D, E, F	G, H, I, J
A, B, C, D, E, F, G	H, I, J
A, B, C, D, E, F, G, H	I, J
A, B, C, D, E, F, G, H, I	J, K
A, B, C, D, E, F, G, H, I, J	K
A, B, C, D, E, F, G, H, I, J, K	Φ

USS in tree-based structures

Breadth-first search – BFS



Search analyse:

- Time complexity:
 - b – ramification factor (number of children of a node)
 - d – length (depth) of solution
 - $T(n) = 1 + b + b^2 + \dots + b^d \Rightarrow O(b^d)$
- Space complexity
 - $S(n) = T(n)$
- Completeness
 - If solution exists, then BFS finds it
- Optimality
 - No

Advantages

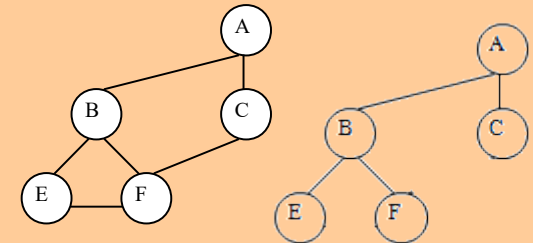
- Finds the shortest path to the objective node (the shallowest solution)

Disadvantages

- Generate and retain a tree whose size exponentially increases (with depth of objective node)
- Exponential time and space complexity
- [Russel&Norving experiment](#)
- Works only for small search spaces

Applications

- Identification of connex components in a graph
- Identification of the shortest path in a graph
- Optimisation in transport networks → [algorithm Ford-Fulkerson](#)
- Serialisation/deserialisation of a binary tree (vs. serialization in a sorted manner) allows efficiently reconstructing of the tree
- Collection copy (garbage collection) → [algorithm Cheney](#)



Vizitate deja	De vizitat
Φ	B
B	A, E, F
B, A	E, F, C
B, A, E	F, C
B, A, E, F	C
B, A, E, F, C	Φ

USS in tree-based structures

Uniform cost search – UCS



Basic elements

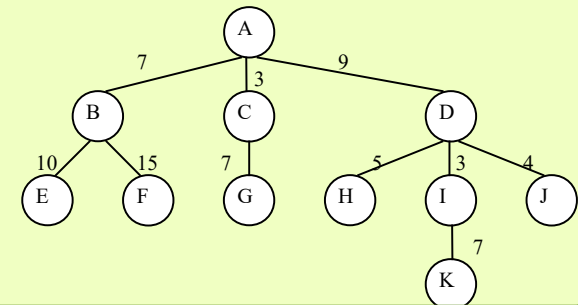
- BFS + special expand procedure (based on the cost of links between nodes)
- All the nodes of depth d are visited before all the nodes of depth $d+1$
- All children of current node are added into a **FIFO ordered** list
 - The nodes of minimum cost are firstly expanded
 - When a path to the final state is obtained, it became a candidat to the optimal solution
- *Branch and bound* algorithm

Example

- Visiting order: A, C, B, D, G, E, F, I, H, J, K

Algorithm

```
bool UCS(elem, list){
    found = false;
    visited =  $\Phi$ ;
    toVisit = {start};      //FIFO sorted list
    while((toVisit !=  $\Phi$ ) && (!found)){
        node = pop(toVisit);
        visited = visited U {node};
        if (node== elem)
            found = true;
        else
            aux =  $\Phi$ ;
        for all (unvisited) children of node do{
            aux = aux U {child};
        } // for
        toVisit = toVisit U aux;
        TotalCostSort(toVisit);
    } //while
    return found;
}
```



visited	toVisit
Φ	A
A	C(3), B(7), D(9)
A, C	B(7), D(9), G(3+7)
A, C, B	D(9), G(10), E(7+10), F(7+15)
A, C, B, D	G(10), I(9+3), J(9+4), H(9+5), E(17), F(22)
A, C, B, D, G	I(12), J(13), H(14), E(17), F(22)
A, C, B, D, G, I	J(13), H(14), E(17), F(22), K(9+3+7)
A, C, B, D, G, I, J	H(14), E(17), F(22), K(19)
A, C, B, D, G, I, J, H	E(17), F(22), K(19)
A, C, B, D, G, I, J, H, E	F(22), K(19)
A, C, B, D, G, I, J, H, E, F	K(19)
A, C, B, D, G, I, J, H, E, F, K	Φ

USS in tree-based structures

Uniform cost search – UCS



Complexity analyse

- Time complexity
 - b – ramification factor
 - d – length (depth) of solution
 - $T(n) = 1 + b + b^2 + \dots + b^d \Rightarrow O(b^d)$
- Space complexity
 - $S(n) = T(n)$
- Completeness
 - yes – if solutions exists, then UCS finds it
- Optimality
 - Yes

Advantages

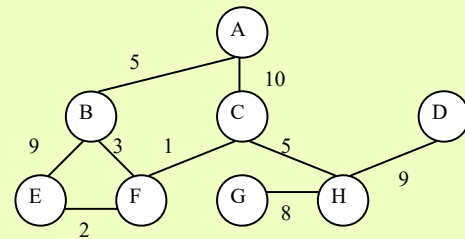
- Finding the minimum cost path to the objective node

Disadvantages

- Exponential time and space complexity

Applications

- Shortest path → [Dijkstra algorithm](#)



Vizitate deja	De vizitat
Φ	A(0)
A(0)	B(5), C(10)
A(0), B(5)	F(8), C(10), E(14)
A(0), B(5), F(8)	C(9), E(10)
A(0), B(5), F(8), C(9)	E(10), H(14)
A(0), B(5), F(8), C(9), E(10)	H(14)

USS in tree-based structures

depth-first search – DFS



Basic elements

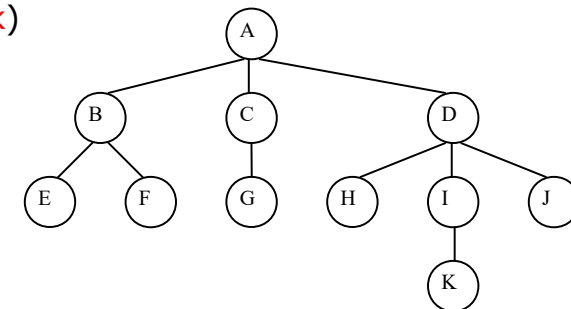
- Expand a child and depth search until
 - The final node is reached or
 - The node is a leaf
- Coming back in the most recent node that must be explored
- All the children of the current node are added in a **LIFO** list (**stack**)

Exemplu

- Visiting order: A, B, E, F, C, G, D, H, I, K, J

Algorithm

```
bool DFS(elem, list){
    found = false;
    visited =  $\Phi$ ;
    toVisit = {start}; //LIFO list
    while((toVisit !=  $\Phi$ ) && (!found)){
        node = pop(toVisit);
        visited = visited U {node};
        if (node== elem)
            found = true;
        else{
            aux =  $\Phi$ ;
            for all (unvisited) children of node do{
                aux = aux U {child};
            }
            toVisit = aux U toVisit;
        }
    } //while
    return found;
}
```



Vizitate deja	De vizitat
Φ	A
A	B, C, D
A, B	E, F, C, D
A, B, E	F, C, D
A, B, E, F	C, D
A, B, E, F, C	G, D
A, B, E, F, C, G	D
A, B, E, F, C, G, D	H, I, J
A, B, E, F, C, G, D, H	I, J
A, B, E, F, C, G, D, H, I	K, J
A, B, E, F, C, G, D, H, I, K	J
A, B, E, F, C, G, D, H, I, K, J	Φ

USS in tree-based structures

depth-first search – DFS



Complexity analyse

- Time complexity
 - b - ramification factor
 - d^{max} - maximal length (depth) of explored tree
 - $T(n) = 1 + b + b^2 + \dots + b^{d^{max}} \Rightarrow O(b^{d^{max}})$
- Space complexity
 - $S(n) = b * d_{max}$
- Completeness
 - No \rightarrow the algorithm does not end for infinite paths (there is no sufficient memory for all the nodes that are visited already)
- Optimality
 - No \rightarrow depth search can find a longer path than the optimal one

Advantages

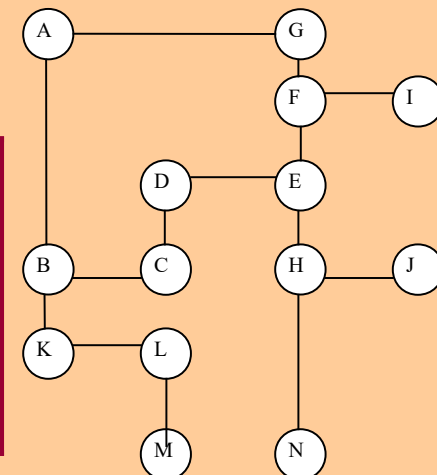
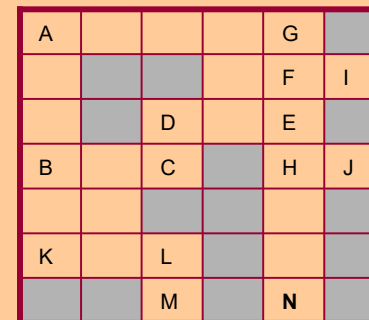
- Finding the shortest path with minimal resources (recursive version)

Disadvantages

- Dead paths
 - Infinite cycles
 - Longer solution than the optimal one

Applications

- Maze problem
- Identification of connex components
- Topological sorting
- Testing the graph planarity



USS in tree-based structures

depth-first search – DFS



```
bool DFS_edges(elem, list){
    discovered =  $\Phi$ ;
    back =  $\Phi$ ;
    toDiscover =  $\Phi$ ;    //LIFO
    for (all neighbours of start) do
        toDiscover = toDiscover U {(start, neighbour)}
    found = false;
    visited = {start};
    while((toDiscover !=  $\Phi$ ) && (!found)){
        edge = pop(toDiscover);
        if (edge.out != visited){
            discovered = discovered U {edge};
            visited = visited U {edge.out}
            if (edge.out == end)
                found = true;
            else{
                aux =  $\Phi$ ;
                for all neighbours of edge.out do{
                    aux = aux U {(edge.out, neighbour)};
                }
                toDiscover = aux U toDiscover;
            }
        }
        else
            back = back U {edge}
    } //while
    return found;
}
```

Muchia	Muchii vizitate deja	Muchii de vizitat	înapoi	Noduri vizitate
	Φ	AB, AF	Φ	A
AB	AB	BC, BK, AF	Φ	A, B
BC	AB, BC	CD, BK, AF	Φ	A, B, C
CD	AB, BC, CD	DE, BK, AF	Φ	A, B, C, D
DE	AB, BC, CD, DE	EF, EH, BK, AF	Φ	A, B, C, D, E
EF	AB, BC, CD, DE, EF	FI, FG, EH, BK, AF	Φ	A, B, C, D, E, F
FI	AB, BC, CD, DE, EF, FI	FG, EH, BK, AF	Φ	A, B, C, D, E, F, I
FG	AB, BC, CD, DE, EF, FI, FG	GA, EH, BK, AF	Φ	A, B, C, D, E, F, I, G
GA	AB, BC, CD, DE, EF, FI, FG	EH, BK, AF	GA	A, B, C, D, E, F, I, G
EH	AB, BC, CD, DE, EF, FI, FG	HJ, HN, BK, AF	GA	A, B, C, D, E, F, I, G, H
HJ	AB, BC, CD, DE, EF, FI, FG, HJ	HN, BK, AF	GA	A, B, C, D, E, F, I, G, H, J
HN	AB, BC, CD, DE, EF, FI, FG, HJ, HN	BK, AF	GA	A, B, C, D, E, F, I, G, H, N



USS in tree-based structures

depth-limited search – DLS

Basic elements

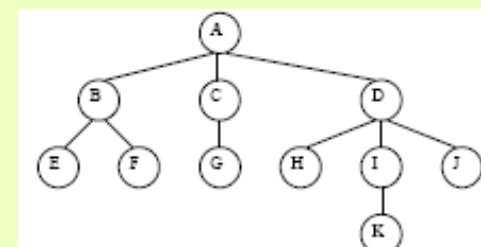
- DFS + maximal depth that limits the search (d_{lim})
- Solved the completeness problems of DFS

Example

- $d_{lim} = 2$
- Visiting order: A, B, E, F, C, G, D, H, I, J

Algorithm

```
bool DLS(elem, list, dlim){
    found = false;
    visited =  $\Phi$ ;
    toVisit = {start}; //LIFO list
    while((toVisit !=  $\Phi$ ) && (!found)){
        node = pop(toVisit);
        visited = visited U {node};
        if (node.depth <= dlim){
            if (node == elem)
                found = true;
            else{
                aux =  $\Phi$ ;
                for all (unvisited) children of node do{
                    aux = aux U {child};
                }
                toVisit = aux U toVisit;
            }
        }
    }
    return found;
}
```



Vizitate deja	De vizitat
Φ	A
A	B, C, D
A, B	E, F, C, D
A, B, E	F, C, D
A, B, E, F	C, D
A, B, E, F, C	G, D
A, B, E, F, C, G	D
A, B, E, F, C, G, D	H, I, J
A, B, E, F, C, G, D, H	I, J
A, B, E, F, C, G, D, H, I	J
A, B, E, F, C, G, D, H, I, K, J	Φ

USS in tree-based structures

depth-limited search – DLS



□ Complexity analyse

- Time complexity:
 - b – ramification factor
 - d_{lim} – limit of length (depth) allowed for the explored tree
 - $T(n) = 1 + b + b^2 + \dots + b^{d_{lim}} \Rightarrow O(b^{d_{lim}})$
- Space complexity
 - $S(n) = b * d_{lim}$
- Completeness
 - Yes, but $\Leftrightarrow d_{lim} > d$, where d = length (path) of optimal solution
- Optimality
 - No \rightarrow DLS can find a longer path than the optimal one

□ Advantages

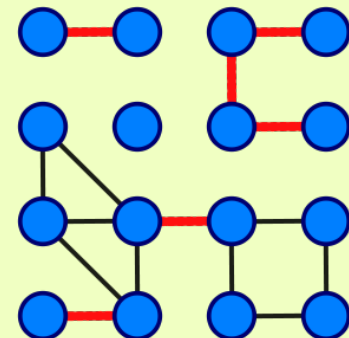
- Solves the completeness problems of DFS

□ Disadvantages

- How to choose a good limit d_{lim} ?

□ Applications

- Identification of bridges in a graph





USS in tree-based structures

iterative deepening depth search – IDDS

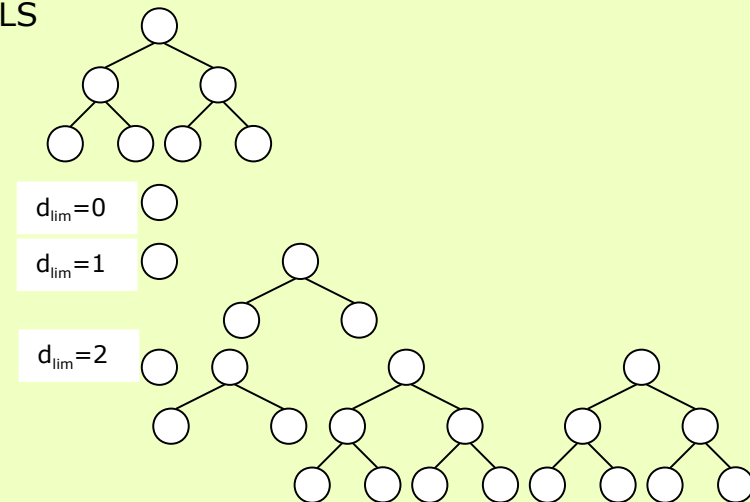
Basic elements

- $\cup \text{DLS}(d_{lim})$, where $d_{lim} = 1, 2, 3, \dots, d_{max}$
- Solves the identification of the optimal limit d_{lim} from DLS
- Usually, it works when:
 - The search space is large
 - The length (depth) of solution is known

Example

Algorithm

```
bool IDS(elem, list){  
    found = false;  
    dlim = 0;  
    while ((!found) && (dlim < dmax)){  
        found = DLS(elem, list, dlim);  
        dlim++;  
    }  
    return found;  
}
```





USS in tree-based structures

iterative deepening depth search – IDDS

□ Complexity analyse

■ Time complexity:

- $b^{d_{max}}$ nodes at depth d_{max} are expanded once $\Rightarrow 1 * b^{d_{max}}$
 - $b^{d_{max}-1}$ nodes at depth $d_{max}-1$ are expanded twice $\Rightarrow 2 * (b^{d_{max}-1})$
 - ...
 - b nodes at depth 1 are expanded d_{max} times $\Rightarrow d_{max} * b^1$
 - 1 node (the root) at depth 0 is expanded $d_{max}+1$ times $\Rightarrow (d_{max}+1)*b^0$
- $$T(n) = \sum_{i=0}^{d_{max}} (i+1)b^{d_{max}-i} \Rightarrow O(b^{d_{max}})$$

■ Space complexity

- $S(n) = b * d_{max}$

■ Completeness

- yes

■ Optimality

- yes

□ Advantages

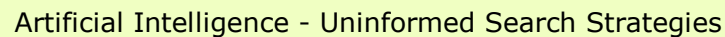
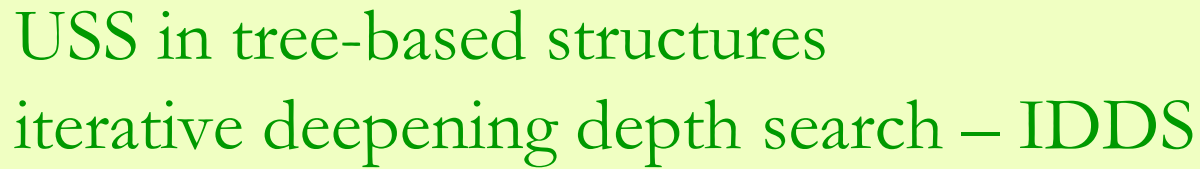
- Requires linear memory
- The goal state is obtained by a minimal path
- Faster than BFS and DFS

□ Disadvantages

- Requires to know the solution depth

□ Applications

- Tic tac toe game



USS in tree-based structures

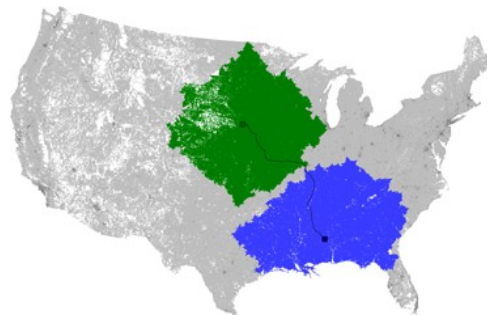
bi-directional search – BDS



□ Basic elements

- 2 parallel search strategies
 - *forward*: from root to leaves
 - *backward*: from leaves to root
- that end when they meet
- any SS can be used in a direction
- Requires establishing:
 - the parents and the children of each node
 - the meeting point

□ Example



□ Algorithm

- Depend on the SS used

USS in tree-based structures

bi-directional search – BDS



□ Complexity analyse

- Time complexity
 - b – ramification factor
 - d – solution length (depth)
 - $O(b^{d/2}) + O(b^{d/2}) \Rightarrow O(b^{d/2})$
- Space complexity
 - $S(n) = T(n)$
- Completeness
 - yes
- Optimality
 - yes

□ Advantages

- Good time and space complexity

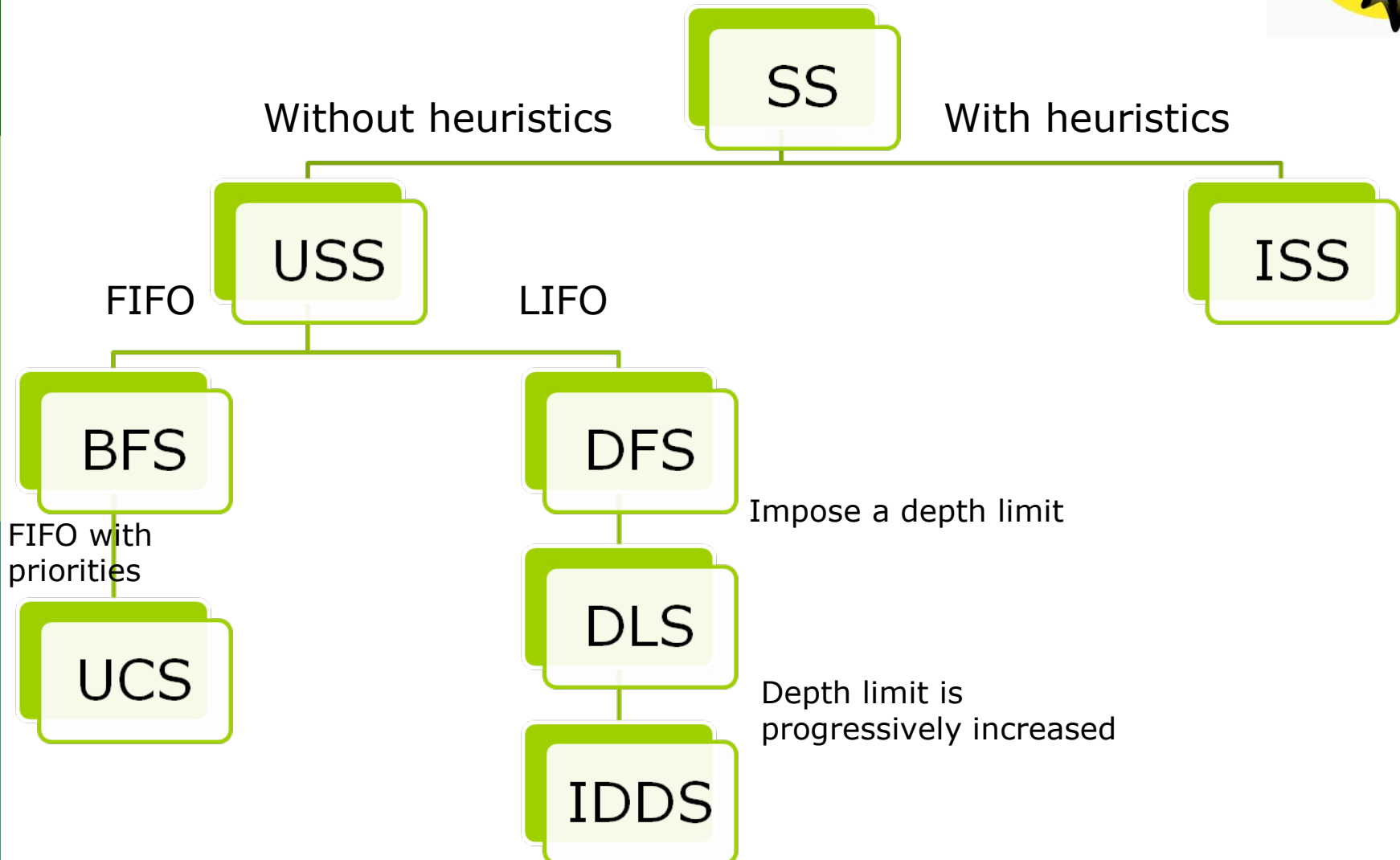
□ Disadvantages

- Each state must be reversed
 - From head to tail
 - From tail to head
- Difficult to implement
- Identification of parents and children for all the nodes
- The final state must be known

□ Applications

- Partitioning problem
- Shortest path

USS in tree-based structures





USS in tree-based structures

Comparison of performances

SS	Time complexity	Space complexity	Completeness	Optimality
BFS	$O(b^d)$	$O(b^d)$	Yes	Yes
UCS	$O(b^d)$	$O(b^d)$	Yes	Yes
DFS	$O(b^{d_{\max}})$	$O(b * d_{\max})$	No	No
DLS	$O(b^{d_{\lim}})$	$O(b * d_{\lim})$	Yes, if $d_{\lim} > d$	No
IDS	$O(b^d)$	$O(b * d)$	Da	Yes
BDS	$O(b^{d/2})$	$O(b^{d/2})$	Yes	Yes