

# ARTIFICIAL INTELLIGENCE

**Solving search problems**

Uninformed search strategies

# Topics

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## A. Short introduction in Artificial Intelligence (AI)

## A. Solving search problems

### A. Definition of search problems

### B. Search strategies

#### A. Uninformed search strategies

#### B. Informed search strategies

#### C. Local search strategies (Hill Climbing, Simulated Annealing, Tabu Search, Evolutionary algorithms, PSO, ACO)

#### D. Adversarial search strategies

## C. Intelligent systems

### A. Rule-based systems in certain environments

### B. Rule-based systems in uncertain environments (Bayes, Fuzzy)

### C. Learning systems

#### A. Decision Trees

#### B. Artificial Neural Networks

#### C. Support Vector Machines

#### • Evolutionary algorithms

### D. Hybrid systems

# Content

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

# Useful information

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- ❑ Chapters I.1, I.2 și II.3 of *S. Russell, P. Norvig, Artificial Intelligence: A Modern Approach, Prentice Hall, 1995*
- ❑ Chapters 1 and 2 of *C. Groșan, A. Abraham, Intelligent Systems: A Modern Approach, Springer, 2011*
- ❑ Chapters 2.1 – 2.4 of

<http://www-g.eng.cam.ac.uk/mmg/teaching/artificialintelligence/>

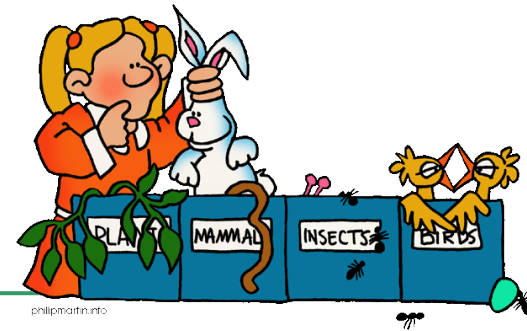
# Problems

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- 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/optimization problems

- Planning, satellite's design



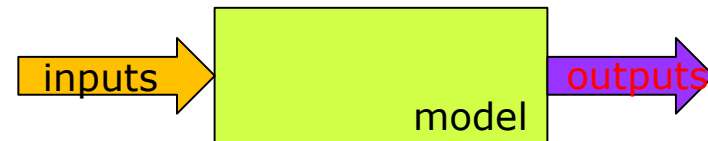
### ■ Modeling problems

- Predictions, classifications



### ■ Simulation problems

- Game theory



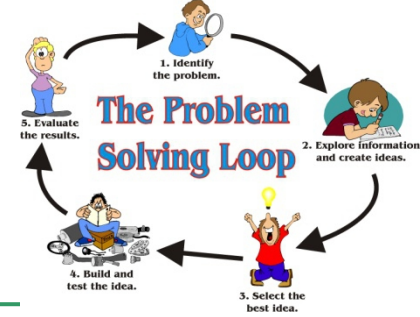
# Problem solving

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- 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
  - Abstract methods





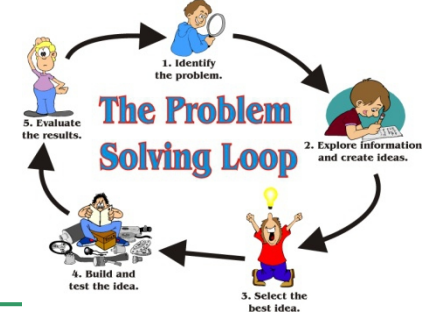
# Solving problems by search

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- ❑ 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

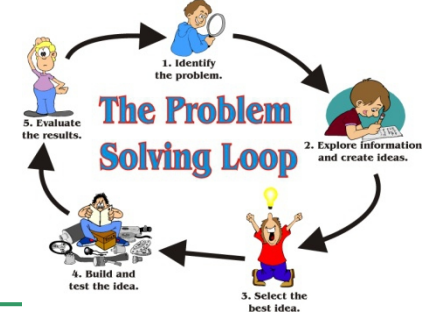
## Problem definition



- 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

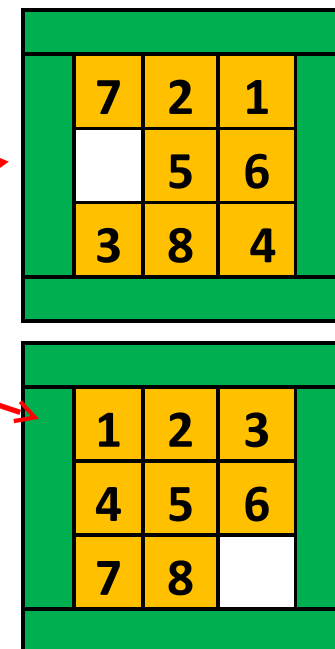
## Problem definition



### □ Examples

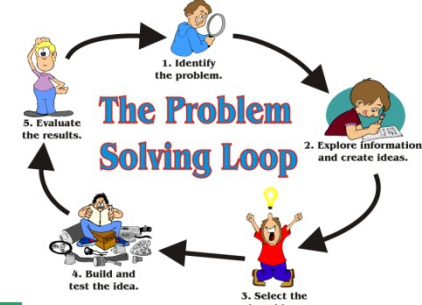
#### ■ Puzzle game with 8 pieces

- 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

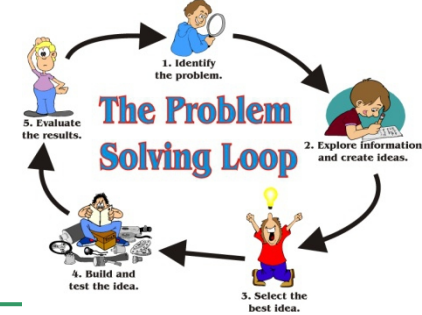
#### ■ Queen's problem

- 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

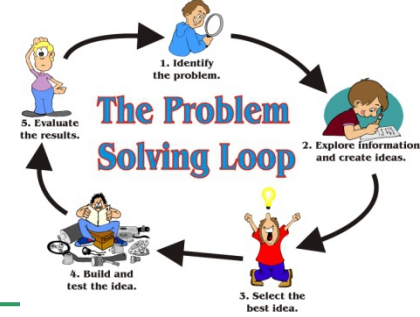
## Problem analyse



- 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

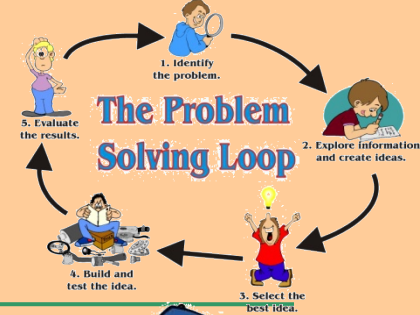


- Solving by moving rules (and control strategy) in the search space until we find a path from the initial state to the final state
- Solving by search
  - Examination of all possible states in order to identify
    - A path from the initial state to the final state
    - An optimal state
  - The search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

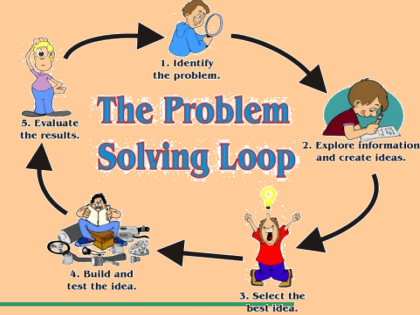
- More searching strategies → how we select one of them?

- Computational complexity (temporal and spatial)
- Completeness → the algorithms always ends and finds a solution (if it exists)
- Optimality → 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

} Internal factors

} External factors

- 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

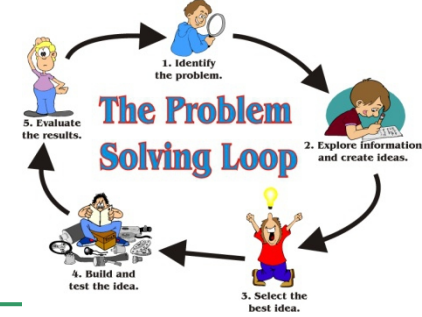
- $T(n)$  – running time (number of steps) of the best algorithm  $A$  that solves a decision problem  $f$  with  $n$  input data





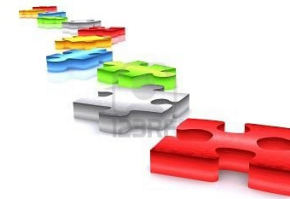
# Steps in solving problems by search

## Selection of a solving technique



■ Problem solving by search can be performed by:

■ Step by step construction of solution

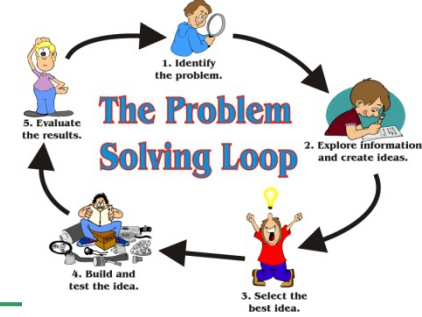


■ Optimal solution identification



# Steps in solving problems by search

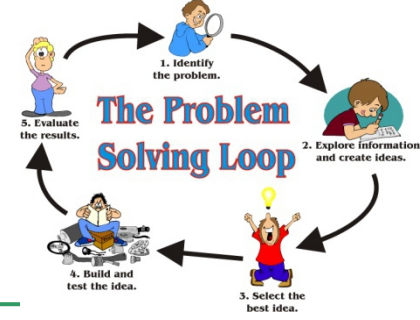
## Selection of a solving technique



- 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

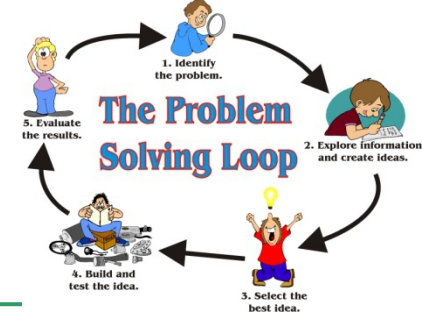
## Selection of a solving technique



- 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

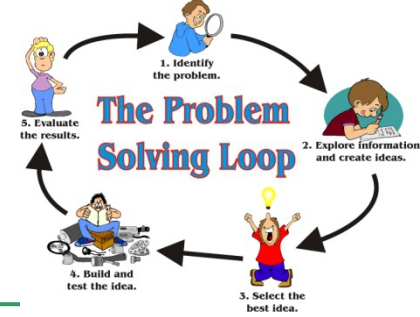


- Solving problem by search involves:
  - Very complex algorithms (NP-complete problems)
  - Search in an exponential space



# Steps in solving problems by search

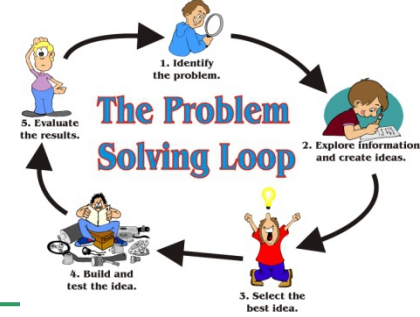
## Selection of a solving technique



- 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 Satisfiability 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

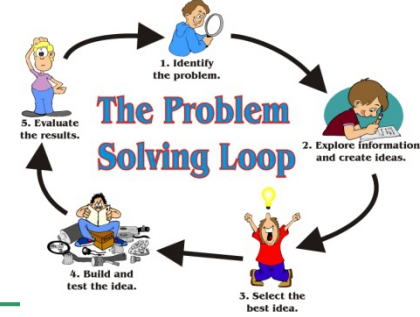


Topology of search strategies:

- **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

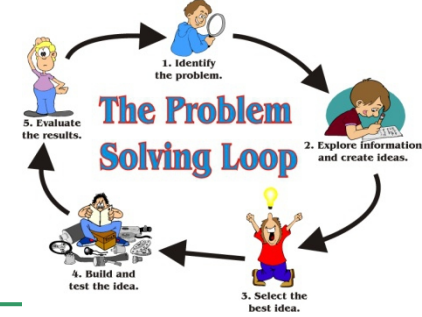


Topology of search strategies:

- Type **(linearity) of constraints**
  - **Linear search**
  - **non-linear search**
    - Classical (deterministic)
      - Direct – based on evaluation of the objective function
      - Indirect – based on derivatives (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



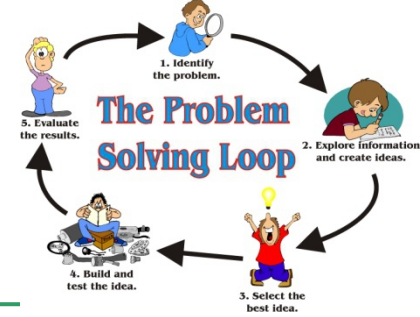
### Example

- 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

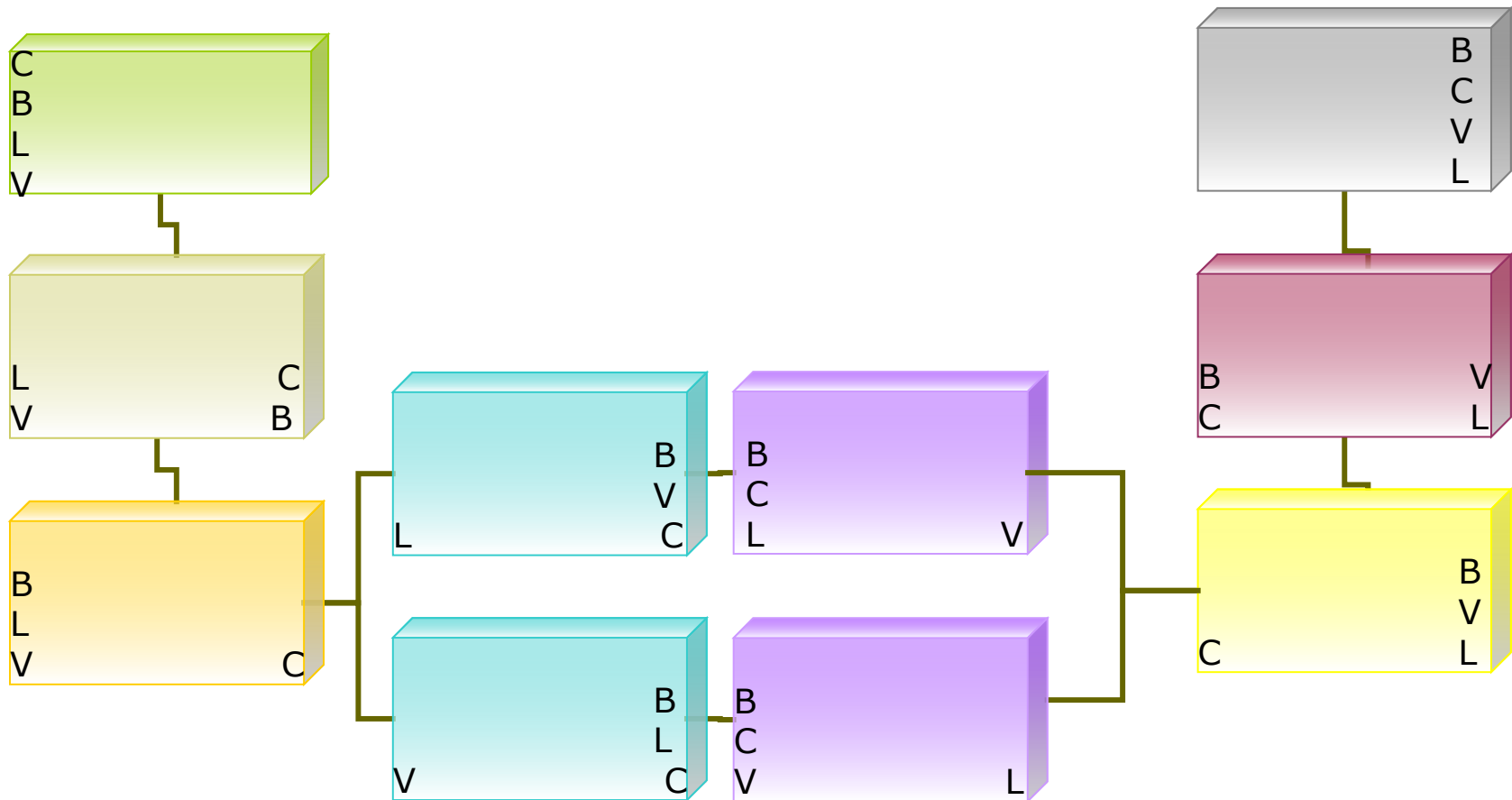
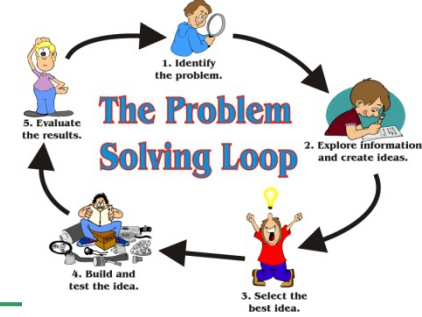


### Example

- Constructive, global, deterministic, 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

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

# Search strategies –

## Basic elements – ADT list



- Domain
  - $D = \{I \mid I = (el_1, el_2, \dots), \text{ where } el_i, i=1,2,3\dots, \text{ are of type } TE \text{ (type of element) and each element } el_i, i=1,2,3\dots, \text{ has a unique position in } I \text{ of type } TP \text{ (Type of position)}\}$
- Operations
  - Create(I)
  - First(I)
  - Last(I)
  - Next(I,p)
  - Prev(I,p)
  - Valid(I,p)
  - getElement(I,p)
  - getPoz (I,e)
  - Modify(I,p,e)
  - AddFirst(I,e)
  - AddToEnd(I,e)
  - AddAfter(I,p,e)
  - AddBefore(I,p,e)
  - Eliminate(I,p)
  - Search(I,e)
  - IsEmpty(I)
  - Dimension(I)
  - Destroy(I)
  - getIterator(I)
- Representation
  - Vector-based
  - Linked lists
- Special cases
  - **Stack – LIFO**
  - Queue – FIFO
  - Priority queue



# Search strategies –

## Basic elements – ADT list



- Domain
  - $D = \{l \mid l = (el_1, el_2, \dots), \text{ where } el_i, i=1,2,3\dots, \text{ are of type } TE \text{ (type of element) and each element } el_i, i=1,2,3\dots, \text{ has a unique position in } l \text{ of type } TP \text{ (Type of position)}\}$
- Operations
  - Create(l)
  - First(l)
  - Last(l)
  - Next(l,p)
  - Prev(l,p)
  - Valid(l,p)
  - getElement(l,p)
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  - Modify(l,p,e)
  - AddFirst(l,e)
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  - Eliminate(l,p)
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  - IsEmpty(l)
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  - Destroy(l)
  - getIterator(l)
- Representation
  - Vector-based
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- Special cases
  - Stack – LIFO
  - **Queue – FIFO**
  - Priority queue

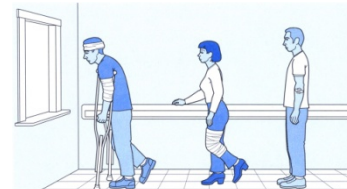


# Search strategies –

## Basic elements – ADT list



- Domain
  - $D = \{l \mid l = (el_1, el_2, \dots), \text{ where } el_i, i=1,2,3\dots, \text{ are of type } TE \text{ (type of element) and each element } el_i, i=1,2,3\dots, \text{ has a unique position in } l \text{ of type } TP \text{ (Type of position)}\}$
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  - Create(l)
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  - IsEmpty(l)
  - Dimension(l)
  - Destroy(l)
  - getIterator(l)
- Representation
  - Vector-based
  - Linked lists
- Special cases
  - Stack – LIFO
  - Queue – FIFO
  - **Priority queue**



# Search strategies –

## Basic elements – ADT Graph

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- Domain – container with nodes and links among nodes
  - $D = \{node_1, node_2, \dots, node_n, link_1, link_2, \dots, link_m\}$  where  $node_i, i=1,2,\dots,n$  are nodes and  $link_i, i=1,2,\dots,m$  are edges between nodes
- Operations
  - create
  - createNode
  - traverse
  - getIterator
  - destroy
- Representation
  - List of edges
  - List of adjacency
  - Matrix of adjacency
  - Incident matrix
- Special cases
  - Un-oriented and oriented graphs
  - Simple and multiple graphs
  - Connex and non-connex graphs
  - Complete or non-complete graph
  - With or without cycles (a-cycle  $\rightarrow$  forests, trees)

# Search strategies –

## Basic elements – ADT Tree

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- Domain – container with nodes and links between nodes
  - $D = \{node_1, node_2, \dots, node_n, link_1, link_2, \dots, link_m, \text{ where } node_i, i=1,2,\dots,n \text{ are nodes and } link_i, i=1,2,\dots,m \text{ are edges between nodes without cycles}\}$
- Operations
  - create
  - createLeaf
  - addSubtree
  - getInfoRoot
  - getSubtree
  - traverse
  - getIterator
  - destroy
- Representation
  - Vector-based
  - Linked lists of children
- Special cases
  - Binary trees (search trees)
  - N-ary trees



# Search strategies –

## Basic elements – paths in graphs

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- *path*
  - Unique nodes
- *trail*
  - Unique edges
- *walk*
  - Without restriction
- Close path
  - Initial node = final node
- circuit
  - A closed *trail*
- cycle
  - A closed *path*



# Uninformed search strategies (USS)

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- 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

# USS in linear structures

## Linear search

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### □ Theoretical aspects

- Checks each element of a list until the search one is found
- The list of elements can be sorted

### □ Example

- List = ( 2, 3, 1, ,7, 5)
- Elem = 7

### □ Algorithm

```
bool LS(elem, list){
    found = false;
    i = 1;
    while ((!found) && (i <= list.length)){
        if (elem == list[i])
            found = true;
        else
            i++;
    } //while
    return found;
}
```

# USS in linear structures

## Linear search



### □ Search analyse

- Time complexity
  - Best case:  $elem = list[1] \Rightarrow O(1)$
  - Worst case:  $elem \notin list \Rightarrow T(n) = n + 1 \Rightarrow O(n)$
  - Average case:  $T(n) = (1 + 2 + \dots + n + (n+1))/(n+1) \Rightarrow O(n)$
- Space complexity
  - $S(n) = n$
- Completeness
  - yes
- Optimality
  - yes

### □ Advantages

- Simplicity, good time complexity for small structures
- Containers can be un-sorted

### □ Disadvantages

- Bad time complexity for large structures

### □ Applications

- Search in real data bases

# USS in linear structures

## Binary search



### □ Theoretical aspects

- Identify an element in a sorted list
- *Divide et Conquer* strategy

### □ Example

- List = ( 2, 3, 5, 6, 8, 9, 13, 16, 18), Elem = 6
  - List = ( 2, 3, 5, 6, 8, 9, 13, 16, 18)
  - List = ( 2, 3, 5, 6)
  - List = ( 5, 6)
  - List = ( 6)

### □ Algorithm

```
bool BS(elem, list){
    found = false;
    left = 1;
    right = list.length;
    while((left < right) && (!found)){
        middle = left + (right - left)/2;
        if (element == list[middle])
            found = true;
        else
            if (element < list[middle])
                right = middle - 1;
            else
                left = middle + 1;
    } //while
    return found;
}
```

# USS in linear structures

## Binary search



### □ Search analyse

- Time complexity  $T(n) = 1$ , for  $n = 1$  and  $T(n) = T(n/2) + 1$ , otherwise  
Suppose that  $n = 2k \Rightarrow k = \log_2 n$   
Suppose that  $2k < n < 2k+1 \Rightarrow k < \log_2 n < k + 1$   
 $T(n) = T(n/2) + 1$   
 $T(n/2) = T(n/2^2) + 1$   
...  
 $T(n/2^{k-1}) = T(n/2^k) + 1$   
-----  
 $T(n) = k + 1 = \log_2 n + 1$
- Space complexity –  $S(n) = n$
- Completeness – yes
- Optimality – yes

### □ Advantages

- Low time complexity compare to linear search

### □ Disadvantages

- Work with sorted vectors

### □ Applications

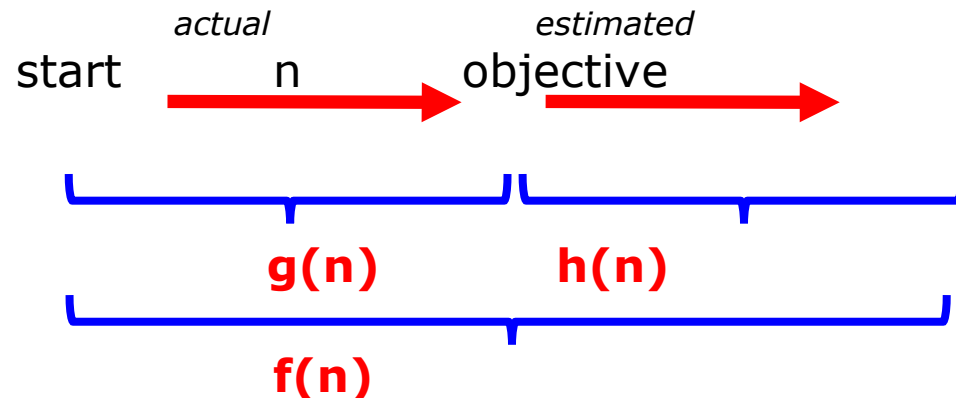
- Guess a number game
- Search in a phone book or in a dictionary



# 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

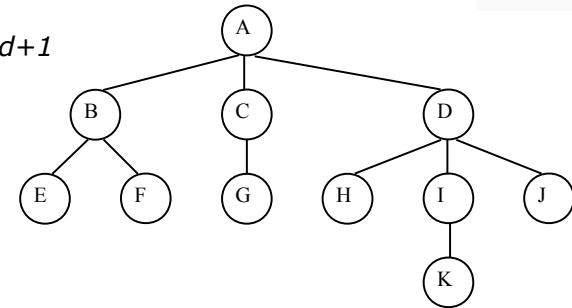


### Basic elements

- 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 = Φ;
    toVisit = {start};      //FIFO list
    while((toVisit != Φ) && (!found)){
        node = pop(toVisit);
        visited = visited U {node};
        if (node == elem)
            found = true;
        else{
            aux = Φ;
            for all (unvisited) children of node do{
                aux = aux U {child};
            }
            toVisit = toVisit U 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



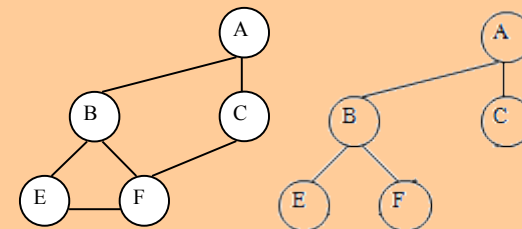
# USS in tree-based structures

## 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](#)
- Serialization/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
$\Phi$	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	$\Phi$

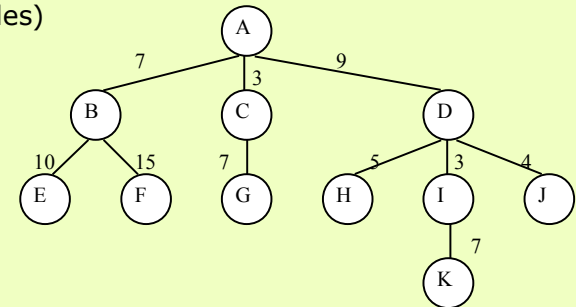
# USS in tree-based structures

## 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 becomes candidate 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
$\Phi$	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), <b>K(9+3+7)</b>
A, C, B, D, G, I, J	H(14), E(17), F(22), <b>K(19)</b>
A, C, B, D, G, I, J, H	E(17), F(22), <b>K(19)</b>
A, C, B, D, G, I, J, H, E	F(22), <b>K(19)</b>
A, C, B, D, G, I, J, H, E, F	K(19)
A, C, B, D, G, I, J, H, E, F, K	$\Phi$

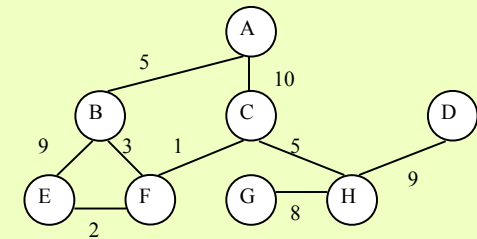
# USS in tree-based structures

## Uniform cost search – UCS



### Complexity analyses

- 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
$\Phi$	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)

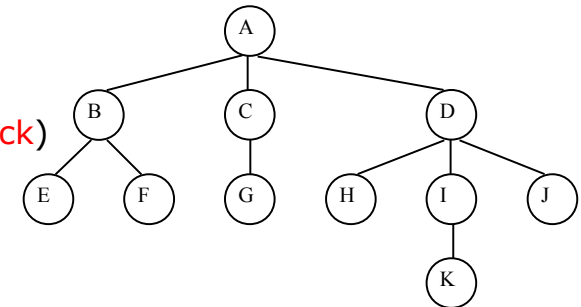
# USS in tree-based structures

## 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

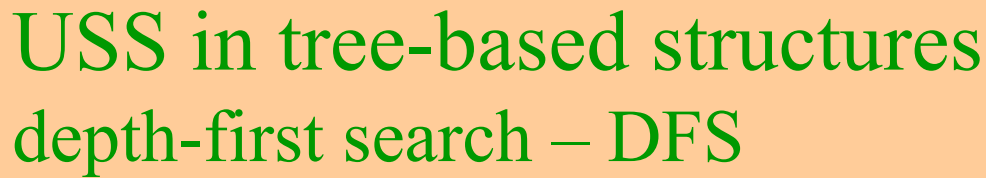
- Visiting order: A, B, E, F, C, G, D, H, I, K, J

### 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
$\Phi$	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	$\Phi$

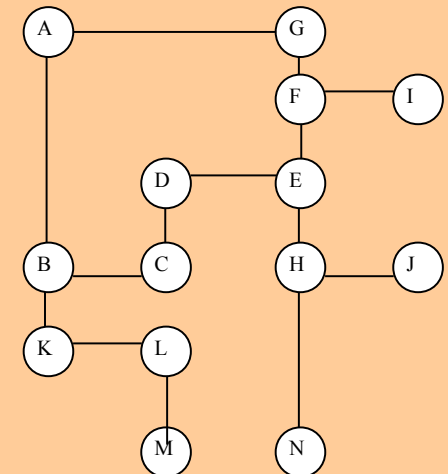


- No → depth search can find a longer path than the optimal one

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

- Longer solution than the optimal one

- 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
	$\Phi$	AB, AF	$\Phi$	A
AB	AB	BC, BK, AF	$\Phi$	A, B
BC	AB, BC	CD, BK, AF	$\Phi$	A, B, C
CD	AB, BC, CD	DE, BK, AF	$\Phi$	A, B, C, D
DE	AB, BC, CD, DE	EF, EH, BK, AF	$\Phi$	A, B, C, D, E
EF	AB, BC, CD, DE, EF	FI, FG, EH, BK, AF	$\Phi$	A, B, C, D, E, F
FI	AB, BC, CD, DE, EF, FI	FG, EH, BK, AF	$\Phi$	A, B, C, D, E, F, I
FG	AB, BC, CD, DE, EF, FI, FG	GA, EH, BK, AF	$\Phi$	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

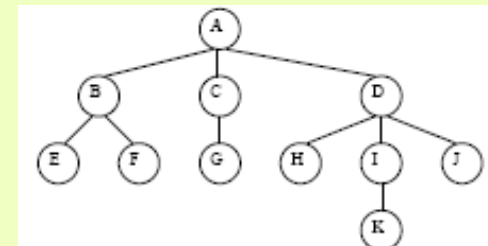
- DFS + maximal depth that limits the search ( $d_{lim}$ )
- Solved the completeness problems of DFS

### 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
$\Phi$	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	$\Phi$

# 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

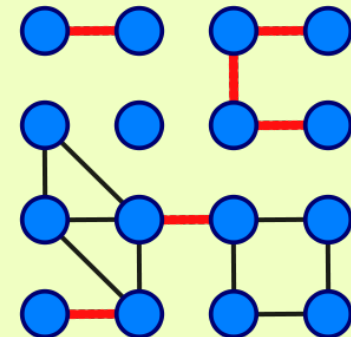
- Solves the completeness problems of DFS

### □ 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



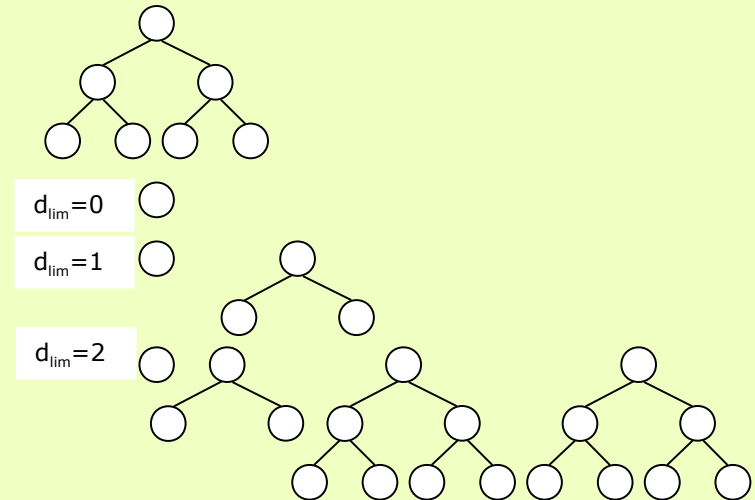
### Basic elements

- $U\ DLS(d_{lim})$ , 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 analysis

#### ■ 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

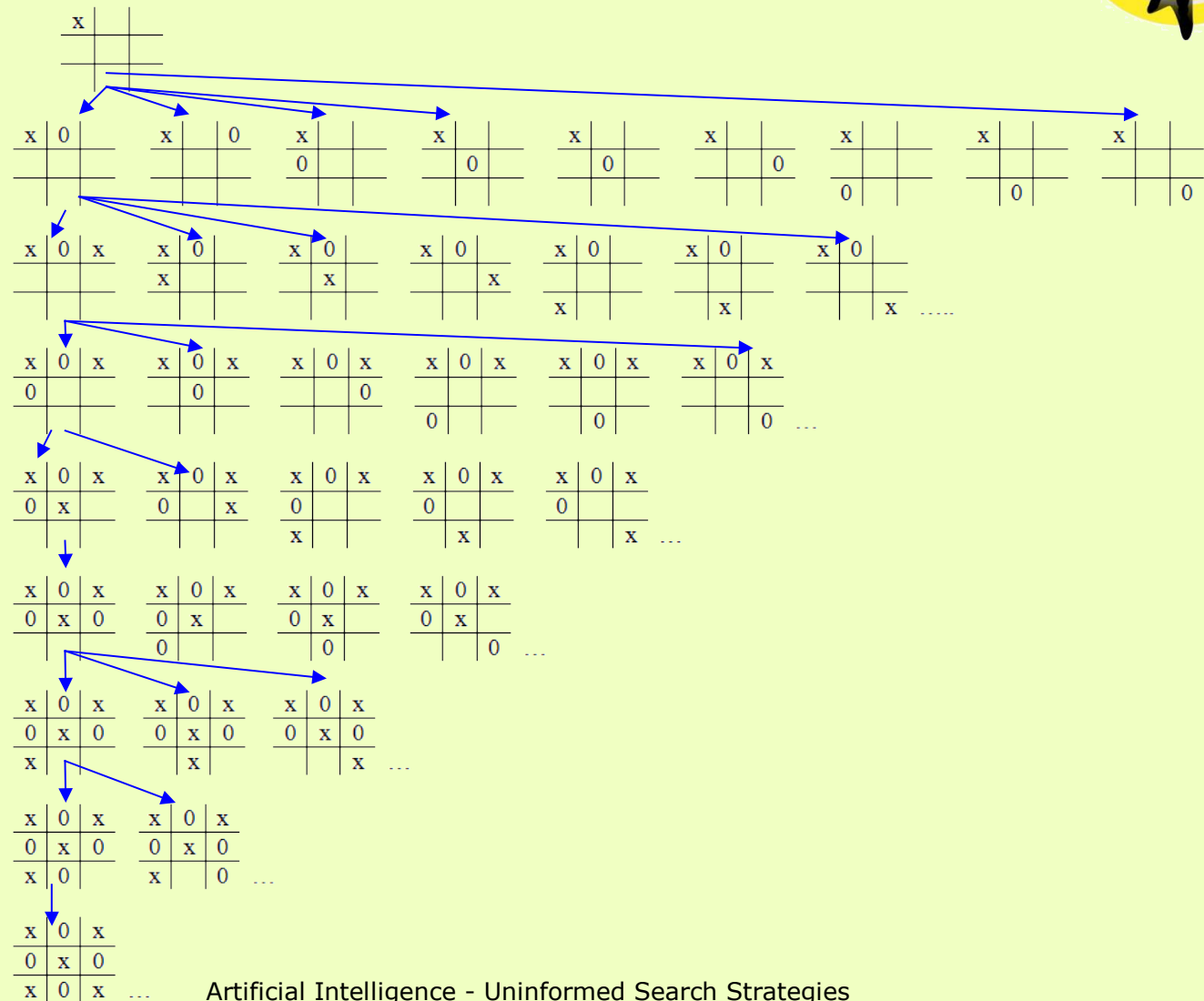
- Requires to know the solution depth

### □ Applications

- Tic tac toe game

# USS in tree-based structures

## iterative deepening depth search – IDDS



# USS in tree-based structures

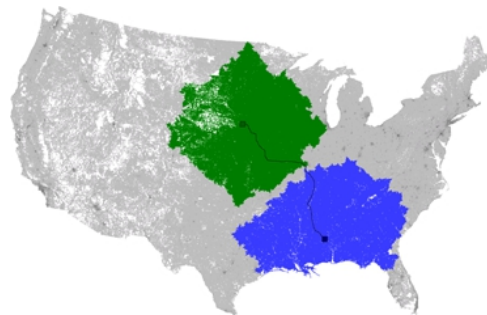
## bi-directional search – BDS



### □ 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

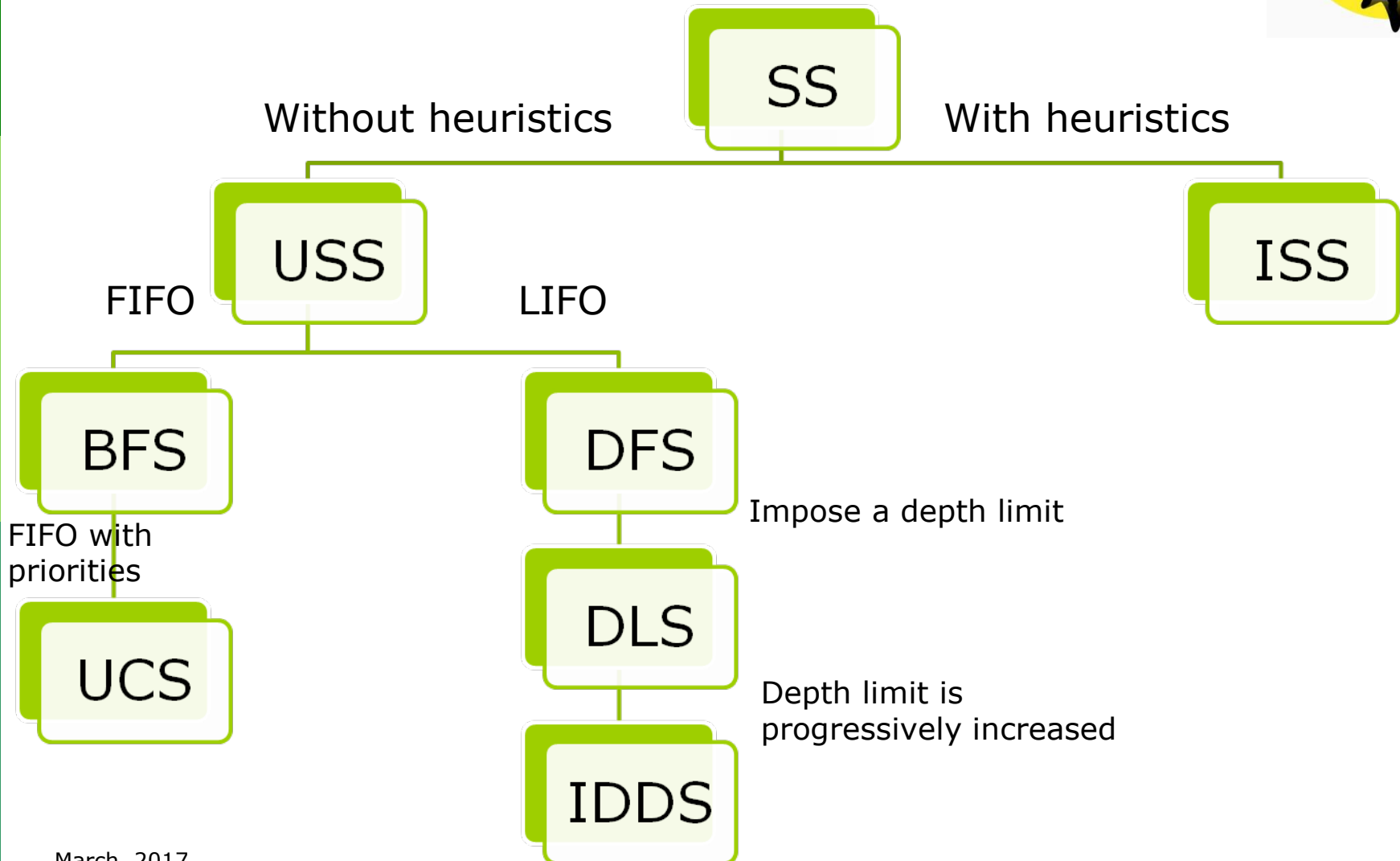
### □ Disadvantages

- 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

# Next lecture

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## A. Short introduction in Artificial Intelligence (AI)

### A. Solving search problems

#### A. Definition of search problems

#### B. Search strategies

A. Uninformed search strategies

B. **Informed search strategies**

C. Local search strategies (Hill Climbing, Simulated Annealing, Tabu Search, Evolutionary algorithms, PSO, ACO)

D. Adversarial search strategies

### C. Intelligent systems

A. Rule-based systems in certain environments

B. Rule-based systems in uncertain environments (Bayes, Fuzzy)

#### C. Learning systems

A. Decision Trees

B. Artificial Neural Networks

C. Support Vector Machines

• Evolutionary algorithms

#### D. Hybrid systems



## Next lecture – Useful information

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- ❑ Chapter II.4 of *S. Russell, P. Norvig, Artificial Intelligence: A Modern Approach, Prentice Hall, 1995*
- ❑ Chapters 3 and 4 of *C. Grosz, A. Abraham, Intelligent Systems: A Modern Approach, Springer, 2011*
- ❑ Chapter 2.5 of <http://www-g.eng.cam.ac.uk/mmg/teaching/artificialintelligence/>

# References

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- Presented information have been inspired from different bibliographic sources, but also from past AI lectures taught by:
  - PhD. Prof. Laura Diosan – [www.cs.ubbcluj.ro/~lauras](http://www.cs.ubbcluj.ro/~lauras)
  - PhD. Prof. Horia F. Pop - [www.cs.ubbcluj.ro/~hfpop](http://www.cs.ubbcluj.ro/~hfpop)