

CS 161 Fundamentals of Artificial Intelligence

Lecture 3

Uninformed Search

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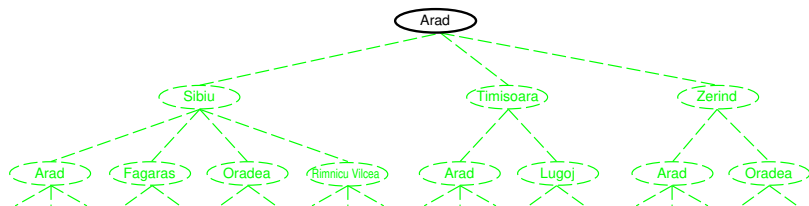
Tree search algorithms

Basic idea:

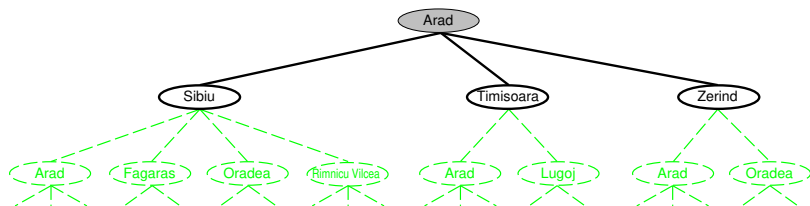
offline, simulated exploration of state space
by generating successors of already-explored states
(a.k.a. **expanding** states)

```
function TREE-SEARCH(problem, strategy) returns a solution, or failure
  initialize the search tree using the initial state of problem
  loop do
    if there are no candidates for expansion then return failure
    choose a leaf node for expansion according to strategy
    if the node contains a goal state then return the corresponding solution
    else expand the node and add the resulting nodes to the search tree
  end
```

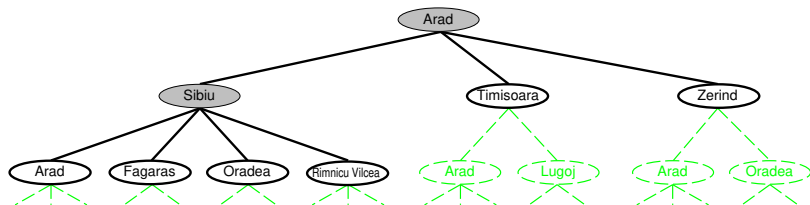
Tree search example



Tree search example



Tree search example



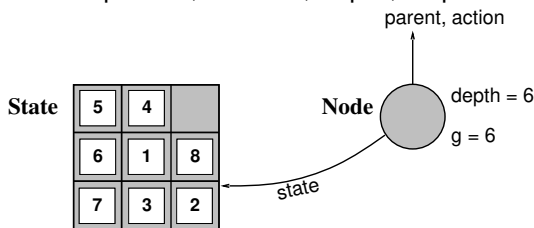
Implementation: states vs. nodes

A **state** is a (representation of) a physical configuration

A **node** is a data structure constituting part of a search tree

includes **parent**, **children**, **depth**, **path cost** $g(x)$

States do not have parents, children, depth, or path cost!



The **EXPAND** function creates new nodes, filling in the various fields and using the **SUCCESSORFN** of the problem to create the corresponding states.

Implementation: general tree search

```
function TREE-SEARCH(problem, fringe) returns a solution, or failure  
fringe  $\leftarrow$  INSERT(MAKE-NODE(INITIAL-STATE[problem]), fringe)  
loop do  
if fringe is empty then return failure  
node  $\leftarrow$  REMOVE-FRONT(fringe)  
if GOAL-TEST(problem, STATE(node)) then return node  
fringe  $\leftarrow$  INSERTALL(EXPAND(node, problem), fringe)
```

```
function EXPAND(node, problem) returns a set of nodes  
successors  $\leftarrow$  the empty set  
for each action, result in SUCCESSOR-FN(problem, STATE[node]) do  
s  $\leftarrow$  a new NODE  
PARENT-NODE[s]  $\leftarrow$  node; ACTION[s]  $\leftarrow$  action; STATE[s]  $\leftarrow$  result  
PATH-COST[s]  $\leftarrow$  PATH-COST[node] + STEP-COST(STATE[node], action, result)  
DEPTH[s]  $\leftarrow$  DEPTH[node] + 1  
add s to successors  
return successors
```

Search strategies

A strategy is defined by picking the **order of node expansion**

Strategies are evaluated along the following dimensions:

completeness—does it always find a solution if one exists?

time complexity—number of nodes generated/expanded

space complexity—maximum number of nodes in memory

optimality—does it always find a least-cost solution?

Time and space complexity are measured in terms of

b —maximum branching factor of the search tree

d —depth of the least-cost solution

m —maximum depth of the state space (may be ∞)

Uninformed search strategies

Uninformed strategies use only the information available in the problem definition

Breadth-first search (BFS)

Depth-first search (DFS)

Depth-limited search

Iterative deepening search

Uniform-cost search

Breadth-first search

```
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
```

Breadth-first search

← queue →

Expand shallowest unexpanded node

Implementation:

1. ϕ *fringe* is a FIFO queue, i.e., new successors go at end

2. A

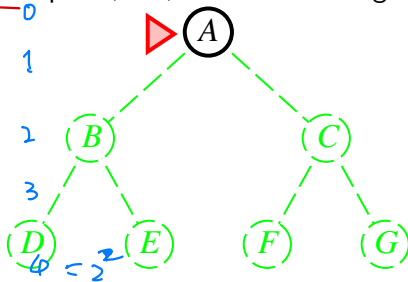
3. BC

4. CDE

5. DEFG

6. EFG

7. FG



8. G

9. ϕ

1

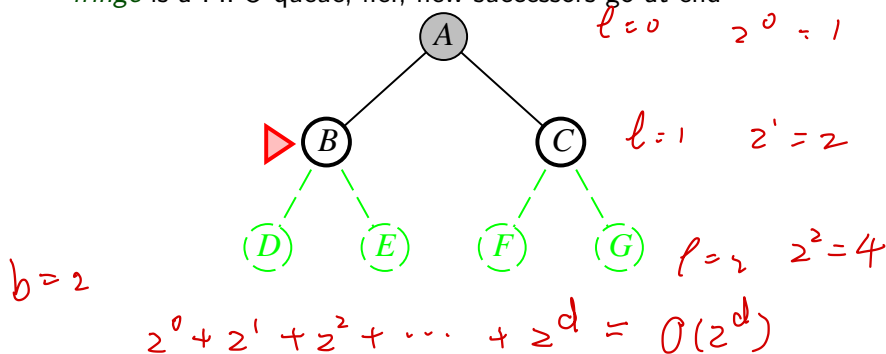
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Breadth-first search

Expand shallowest unexpanded node

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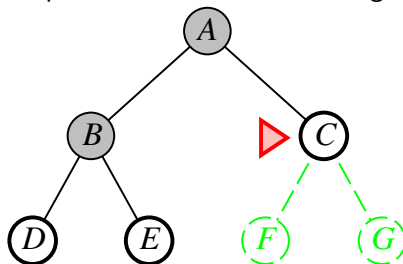


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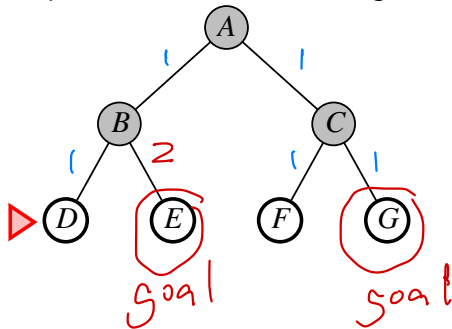


Breadth-first search

Expand shallowest unexpanded node

Implementation:

fringe is a FIFO queue, i.e., new successors go at end



Properties of breadth-first search

Complete?? Yes (if b is finite)

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Time?? $b + b^2 + b^3 + \dots + b^d = O(b^d)$, i.e., exponential in d

$b^0 +$

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Space?? $O(b^d)$ (keeps every node in memory)

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Optimal?? Yes (if cost = 1 per step); not optimal in general

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Space is the big problem; can easily generate nodes at 100MB/sec
so 24hrs = 8640GB.

Depth-first search

stack

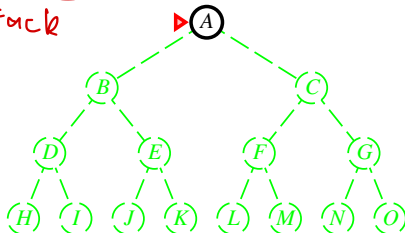


Expand deepest unexpanded node

Implementation:

fringe = LIFO queue, i.e., put successors at front

stack

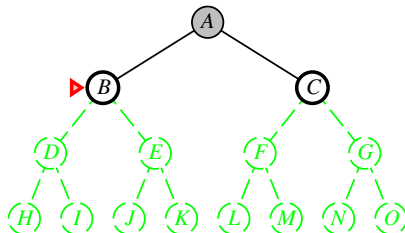


Depth-first search

Expand deepest unexpanded node

Implementation:

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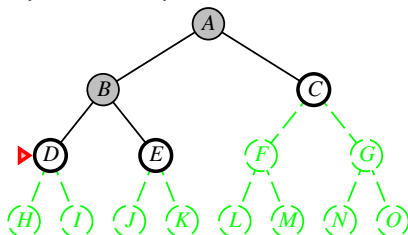


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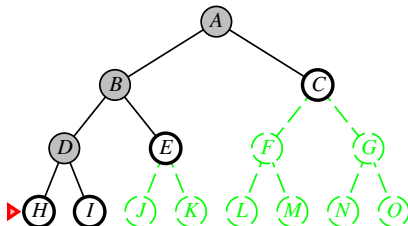


Depth-first search

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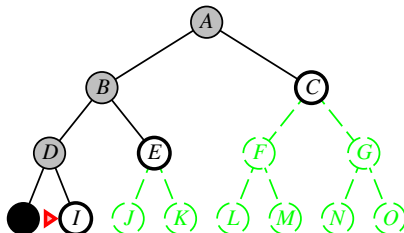


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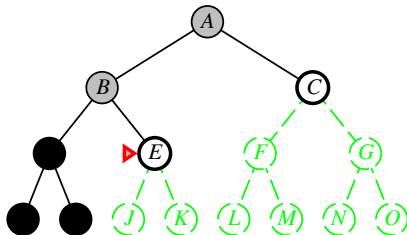


Depth-first search

Expand deepest unexpanded node

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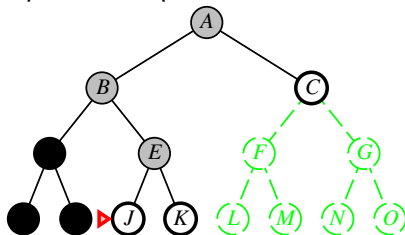


Depth-first search

Expand deepest unexpanded node

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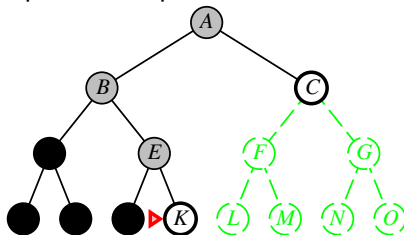


Depth-first search

Expand deepest unexpanded node

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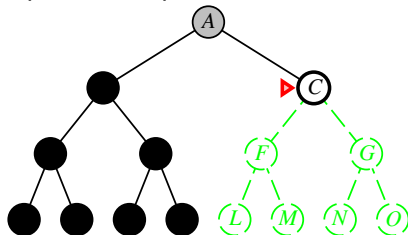


Depth-first search

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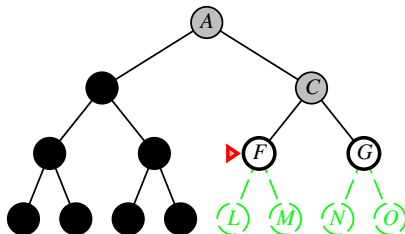


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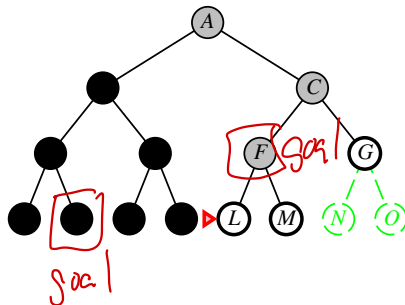


Depth-first search

Expand deepest unexpanded node

Implementation:

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Properties of depth-first search

Complete?? No: fails in infinite-depth spaces, spaces with loops
Modify to avoid repeated states along path
⇒ complete in finite spaces

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V.S. b^d for BFS

Time?? $O(b^m)$: terrible if m is much larger than d , where m is the maximum depth of any node

but if solutions are dense, may be much faster than breadth-first

Properties of depth-first search

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Time?? $O(b^m)$: terrible if m is much larger than d , where m is the maximum depth of any node

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Space?? $O(bm)$, i.e., linear space!

v.s. $O(b^d)$ in BFS

Properties of depth-first search

Complete?? No: fails in infinite-depth spaces, spaces with loops

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but if solutions are dense, may be much faster than breadth-first

Space?? $O(bm)$, i.e., linear space!

Optimal?? No

Depth-limited search

= depth-first search with depth limit ℓ ,
i.e., nodes at depth ℓ have no successors

Recursive implementation:

function ~~ITERATIVE-DEEPENING-SEARCH~~(*problem*) **returns** a solution node or *failure*
 for ~~*depth*~~ = 0 **to** ∞ **do**
 ~~*result*~~ \leftarrow ~~DEPTH-LIMITED-SEARCH~~(*problem*, ~~*depth*~~)
 if *result* \neq *cutoff* **then return** *result*

function DEPTH-LIMITED-SEARCH(*problem*, ℓ) **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*) > ℓ **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*

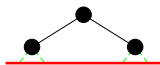
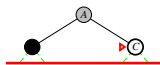
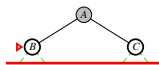
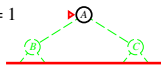
Iterative deepening search

Limit = 0



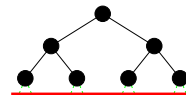
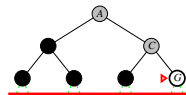
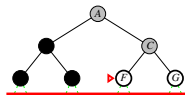
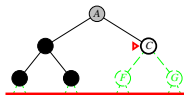
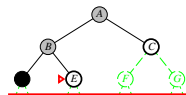
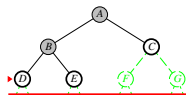
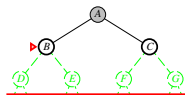
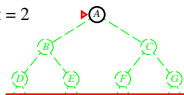
Iterative deepening search

Limit = 1



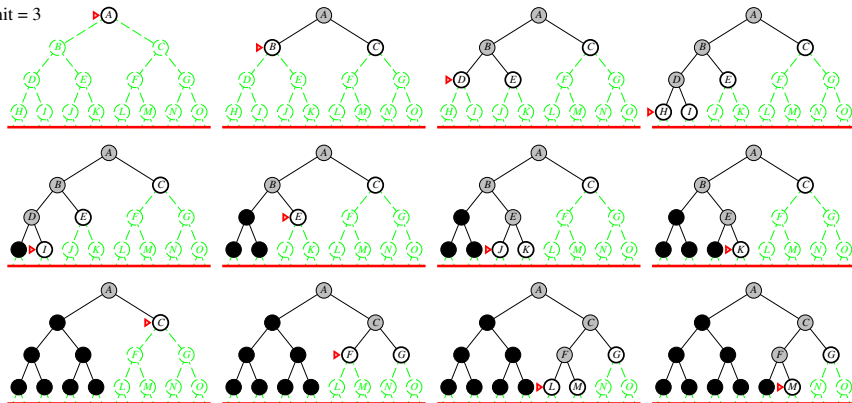
Iterative deepening search

Limit = 2



Iterative deepening search

Limit = 3



Properties of iterative deepening search

Complete?? Yes

Properties of iterative deepening search

Time complexity of RT.S
 $(b^0) + b^1 + b^2 + \dots + b^d = O(b^d)$

Complete?? Yes

Time?? $db^1 + (d-1)b^2 + \dots + b^d = O(b^d)$

$l=1$

b^1

$l=2$

$b^1 + b^2$

\vdots

$l=d$

$b^1 + b^2 + \dots + b^d$

$$d \cdot b^1 + (d-1)b^2 + \dots + b^d = O(b^d)$$

Properties of iterative deepening search

Complete?? Yes

Time?? $db^1 + (d-1)b^2 + \dots + b^d = O(b^d)$

Space?? $O(bd)$

Properties of iterative deepening search

Complete?? Yes

Time?? $db^1 + (d-1)b^2 + \dots + b^d = O(b^d)$

Space?? $O(bd)$

Optimal?? Yes, if step cost = 1

There is some extra cost for generating the upper levels multiple times, but it is not large. E.g., numerical comparison for $b = 10$ and $d = 5$, solution at far right leaf:

$$N(\text{IDS}) = 50 + 400 + 3,000 + 20,000 + 100,000 = 123,450$$

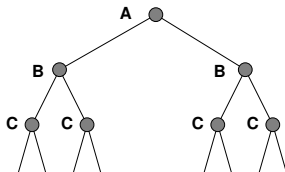
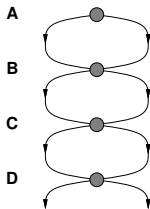
Handwritten notes: $d \cdot b^1$, $(d-1)b^2$, b^d

$$\begin{aligned} N(\text{BFS}) &= 10 + 100 + 1,000 + 10,000 + 100,000 \\ &= 111,110 \end{aligned}$$

IDS does better because other nodes at depth d are not expanded
BFS can be modified to apply goal test when a node is **generated**

Repeated states

Failure to detect repeated states can turn a linear problem into an exponential one!



Graph search

```
function GRAPH-SEARCH(problem) returns a solution, or failure
  initialize the frontier using the initial state of problem
  initialize the explored set to be empty
  loop do
    if the frontier is empty then return failure
    choose a leaf node and remove it from the frontier
    if the node contains a goal state then return the corresponding solution
    add the node to the explored set
    expand the chosen node, adding the resulting nodes to the frontier
    only if not in the frontier or explored set
  end
```

Uniform-cost search

- ▶ When all step costs are equal, breadth-first search is optimal

Uniform-cost search

- ▶ When all step costs are equal, breadth-first search is optimal
- ▶ What if all step costs are not equal?

Uniform-cost search

function UNIFORM-COST-SEARCH(*problem*) **returns** a solution, or failure

node \leftarrow a node with STATE = *problem*.INITIAL-STATE, PATH-COST = 0

frontier \leftarrow a priority queue ordered by PATH-COST, with *node* as the only element

explored \leftarrow an empty set

loop do

if EMPTY?(*frontier*) **then return** failure

node \leftarrow POP(*frontier*) /* chooses the lowest-cost node in *frontier* */

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

 add *node*.STATE to *explored*

for each *action* **in** *problem*.ACTIONS(*node*.STATE) **do**

child \leftarrow CHILD-NODE(*problem*, *node*, *action*)

if *child*.STATE is not in *explored* or *frontier* **then**

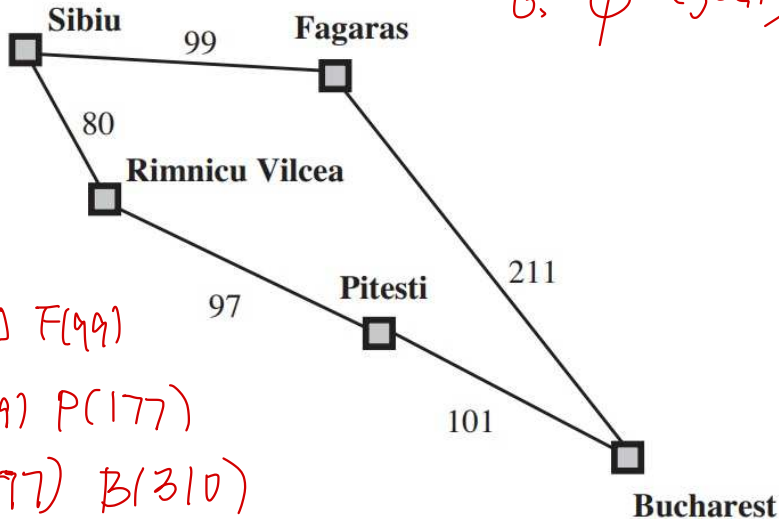
frontier \leftarrow INSERT(*child*, *frontier*)

else if *child*.STATE is in *frontier* with higher PATH-COST **then**

 replace that *frontier* node with *child*

Uniform-cost search

6. ϕ (goal)



1. $S(0)$

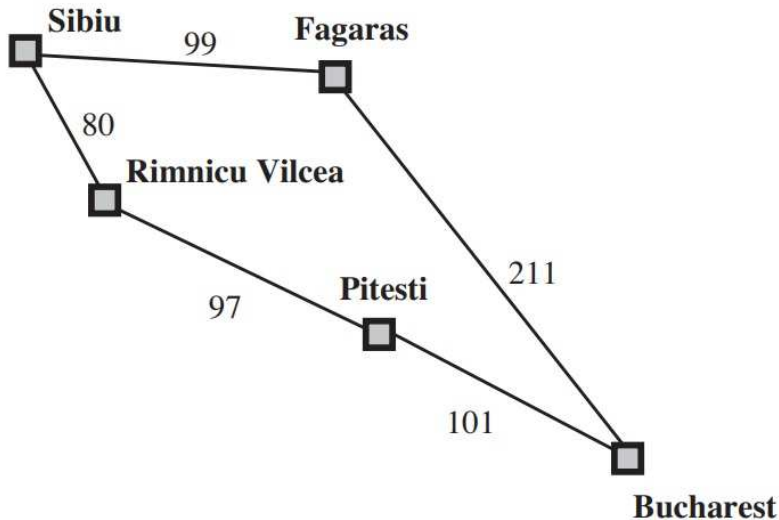
2. $R(80)$ $F(99)$

3. $F(99)$ $P(177)$

4. $P(177)$ $B(310)$

5. $B(278)$

Uniform-cost search



Sibiu → Rimnicu Vilcea → Fagaras → Pitesti → Bucharest

Uniform-cost search

Expand least-cost unexpanded node

Let $g(n)$ be the sum of the cost (path cost) from start to node n

Implementation:

fringe = queue ordered by path cost, lowest first

Equivalent to breadth-first if step costs all equal

Complete?? Yes, if step cost $\geq \epsilon$

Time?? # of nodes with $g \leq$ cost of optimal solution, $O(b^{\lceil C^*/\epsilon \rceil})$
where C^* is the cost of the optimal solution

Space?? # of nodes with $g \leq$ cost of optimal solution, $O(b^{\lceil C^*/\epsilon \rceil})$

Optimal?? Yes—nodes expanded in increasing order of $g(n)$

Summary of algorithms

Criterion	Breadth-First	Uniform-Cost	Depth-First	Depth-Limited	Iterative Deepening
Complete?	Yes*	Yes*	No	Yes, if $l \geq d$	Yes
Time	b^d	$b^{\lceil C^*/\epsilon \rceil}$	b^m	b^l	b^d
Space	b^d	$b^{\lceil C^*/\epsilon \rceil}$	bm	bl	bd
Optimal?	Yes*	Yes	No	No	Yes*

Summary

- Problem formulation usually requires abstracting away real-world details to define a state space that can feasibly be explored
- Variety of uninformed search strategies
- Iterative deepening search uses only linear space and not much more time than other uninformed algorithms
- Graph search can be exponentially more efficient than tree search

Acknowledgment

The slides are adapted from Stuart Russell et al.