

Introduction to Data Science and Artificial Intelligence

Solving Problems by Search: Uninformed Search and Informed Search

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- Uninformed search strategies
- Informed search strategies
  - Greedy search
  - A \* search



#### **Review: Well-Defined Formulation**

Definition of a problem	The information used by an agent to decide what to do
Specification	<ul> <li>Initial state</li> <li>Action set, i.e. available actions (successor functions)</li> <li>State space, i.e. states reachable from the initial state <ul> <li>Solution path: sequence of actions from one state to another</li> </ul> </li> <li>Goal test predicate <ul> <li>Single state, enumerated list of states, abstract properties</li> </ul> </li> <li>Cost function <ul> <li>Path cost g(n), sum of all (action) step costs along the path</li> </ul> </li> </ul>
Solution	A path (a sequence of operators leading) from the Initial-State to a state that satisfies the Goal-Test

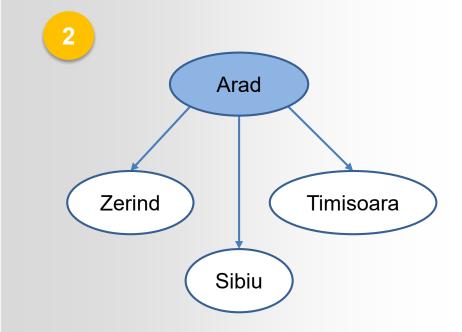


- Exploration of state space by generating successors of already-explored states
  - Frontier: candidate nodes for expansion
  - Explored set



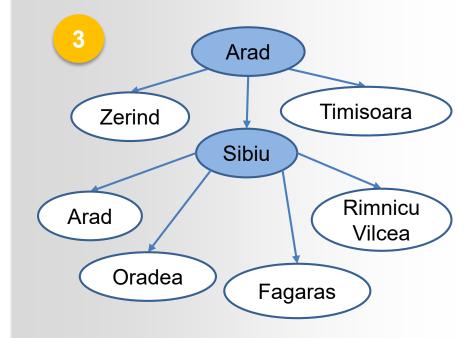


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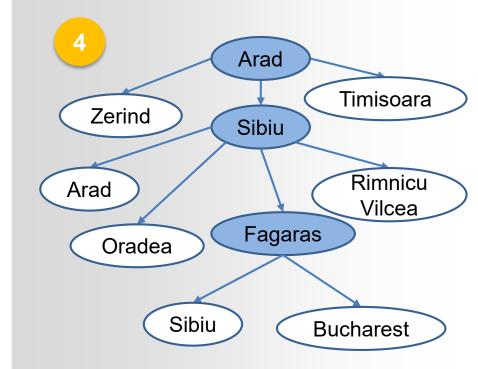


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# **Search Strategies**

- 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	How long does it take to find a solution: the number of nodes generated
Space Complexity	Maximum number of nodes in memory
Optimality	Does it always find the best (least-cost) solution?



### **Search Strategies**

- Branching factor
  - Maximum number of successors of any node
  - Or average branching factor

#### Uninformed vs Informed



#### **Uninformed** search strategies

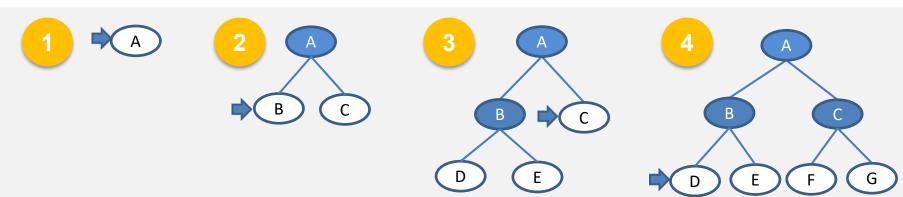
- Use only the information available in the problem definition
  - 1. Breadth-first search
  - 2. Uniform-cost search
  - 3. Depth-first search
  - 4. Depth-limited search
  - 5. Iterative deepening search

#### **Informed** search strategies

- Use problem-specific knowledge to guide the search
- Usually more efficient

#### **Breadth-First Search**

Expand shallowest unexpanded node which can be implemented by a First-In-First-Out (FIFO) queue



#### Denote

- b: maximum branching factor of the search tree
- d: depth of the least-cost solution
- Complete: Yes
- Optimal: Yes when all step costs equally





# **Complexity of BFS**

- Hypothetical state-space, where every node can be expanded into b new nodes, solution of path-length d
- Time:  $1 + b + b^2 + b^3 + \dots + b^d = O(b^d)$
- Space: (keeps every node in memory)  $O(b^d)$  are equal

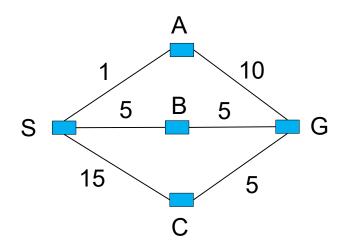
Nodes		Time		Memory
1	1	millisecond	100	bytes
111	0.1	seconds	11	kilobytes
11111	11	seconds	1	kilobytes
$10^{6}$	18	minutes	111	megabyt e
10 <sup>8</sup>	31	hours	11	gigabytes
$10^{10}$	128	days	1	terabyte
$10^{12}$	35	years	111	terabytes
10 <sup>14</sup>	3500	years	11111	terabytes
	$   \begin{array}{c}     1 \\     111 \\     11111 \\     10^6 \\     \hline     10^8 \\     10^{10} \\     10^{12}   \end{array} $	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	$egin{array}{cccccccccccccccccccccccccccccccccccc$	$\begin{array}{cccccccccccccccccccccccccccccccccccc$



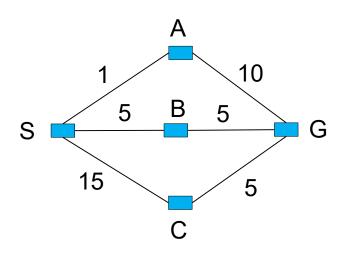
#### **Uniform-Cost Search**

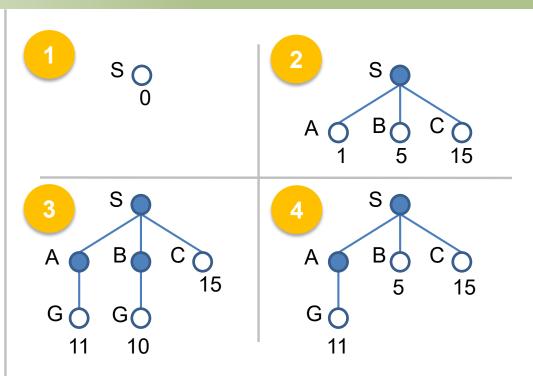
To consider edge costs, expand unexpanded node with the least path cost *g* 

- Modification of breath-first search
- Instead of First-In-First-Out (FIFO)
   queue, using a priority queue with
   path cost g(n) to order the elements
- BFS = UCS with g(n) = Depth(n)



#### **Uniform-Cost Search**





Here we do not expand notes that have been expanded.

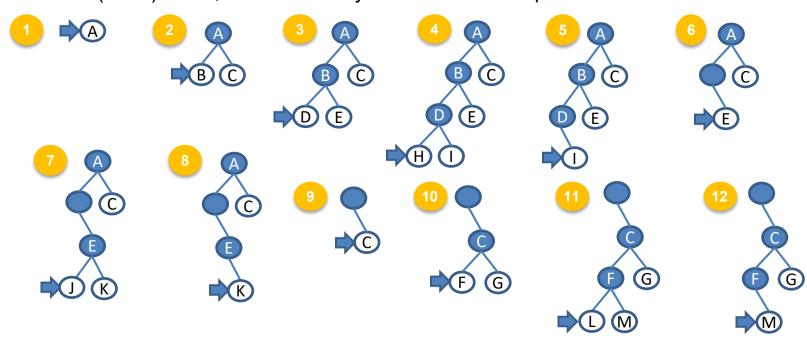


### **Uniform-Cost Search**

Complete	Yes
Time	# of nodes with path cost g <= cost of optimal solution (eqv. # of nodes pop out from the priority queue)
Space	# of nodes with path cost g <= cost of optimal solution
Optimal	Yes

### **Depth-First Search**

Expand deepest unexpanded node which can be implemented by a Last-In-First-Out (LIFO) stack, Backtrack only when no more expansion





# **Depth-First Search**

#### Denote

m: maximum depth of the state space

Complete	<ul> <li>infinite-depth spaces: No</li> <li>finite-depth spaces with loops: No</li> <li>with repeated-state checking: Yes</li> <li>finite-depth spaces without loops: Yes</li> </ul>
Time	$O(b^m)$ If solutions are dense, may be much faster than breadth-first
Space	O(bm)
Optimal	No



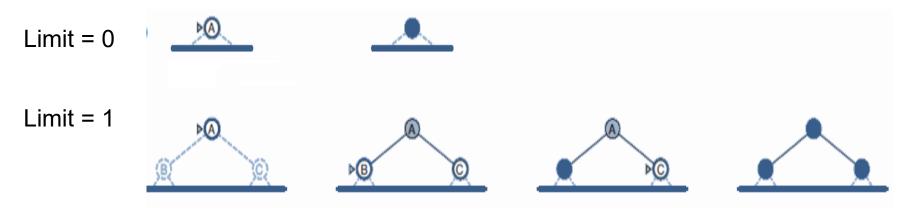
# **Depth-Limited Search**

To avoid infinite searching, Depth-first search with a cutoff on the max depth / of a path

Complete	Yes, if $I \ge d$
Time	$O(b^I)$
Space	O(bI)
Optimal	No

### **Iterative Deepening Search**

Iteratively estimate the max depth / of DLS one-by-one

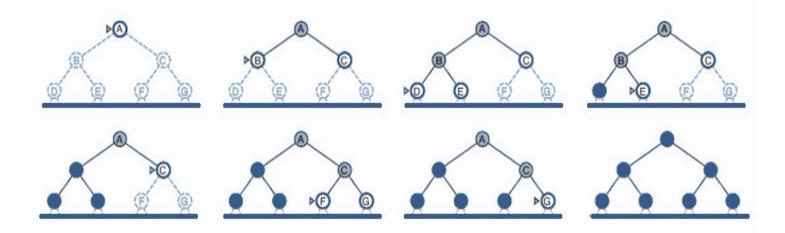




# **Iterative Deepening Search**

Iteratively estimate the max depth / of DLS one-by-one

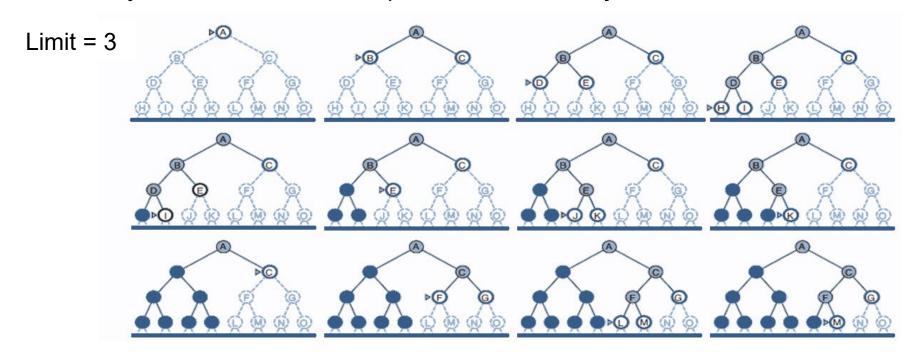
Limit = 2





# **Iterative Deepening Search**

Iteratively estimate the max depth / of DLS one-by-one





# **Iterative Deepening Search...**

```
Function ITERATIVE-DEEPENING-SEARCH(problem) returns a solution sequence
   inputs: problem, a problem
   for depth 0 to \infty do
    if DEPTH-LIMITED-SEARCH(problem, depth) succeeds then return its result
   end
   return failure
```

Complete	Yes
Time	$O(b^d)$
Space	O(bd)
Optimal	Yes



# Summary (we make assumptions for optimality)

Criterion	Breadth- first	Uniform- Cost	Depth-First	Depth- Limited	Iterative Deepening	Bidirectional (if applicable)
Time	$b^d$	$b^d$	$b^m$	$b^l$	$b^d$	$b^{d/2}$
Space	$b^d$	$b^d$	bm	bl	bd	$b^{d/2}$
Optimal	Yes	Yes	No	No	Yes	Yes
Complete	Yes	Yes	No	Yes, if $l \ge d$	Yes	Yes

#### Question to think:

• If a search strategy is optimal, is it also complete?





#### Uninformed search strategies

- Systematic generation of new states (→Goal Test)
- Inefficient (exponential space and time complexity)

#### Informed search strategies

- Use problem-specific knowledge
  - To decide the order of node expansion
- Best First Search: expand the most desirable unexpanded node
  - Use an evaluation function to estimate the "desirability" of each node





#### **Evaluation function**

- Path-cost function g(n)
  - Cost from initial state to current state (search-node) n
  - No information on the cost toward the goal
- Need to estimate cost to the closest goal
- "Heuristic" function h(n)
  - Estimated cost of the cheapest path from n to a goal state h(n)
    - Exact cost cannot be determined
  - depends only on the state at that node
  - h(n) is not larger than the real cost (admissible)



### **Greedy Search**

Expands the node that appears to be closest to goal

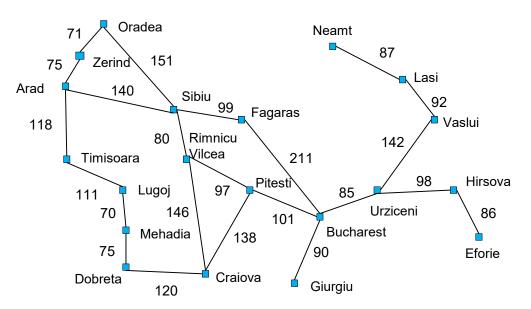
- Evaluation function h(n):estimate of cost from n to goal
- Function Greedy-Search(problem) returns solution
  - Return Best-First-Search(problem, h) // h(goal) = 0

Question: How to estimation the cost from n to goal?

Answer: Recall that we want to use problem-specific knowledge

#### **Example: Route-finding from Arad to Bucharest**

#### h(n) = straight-line distance from n to Bucharest



- Useful but potentially fallible (heuristic)
- Heuristic functions are problem-specific

366 0 160 242
161
176
77
151
226
244
241
234
380
98
193
253
329
80
199
374

# **Example**



The initial state



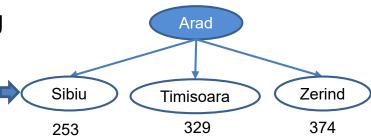
366

Arad	366
Bucharest	0
Craiova	160
Dobreta	242
Efoire	161
Fagaras	176
Giurgiu	77
Hirsova	151
Lasi	226
Lugoj	244
Mehadia	241
Neamt	234
Oradea	380
Pitesti	98
Rimnicu Vilcea	193
Sibiu	253
Timisoara	329
Urziceni	80
Vaslui	199
Zerind	374

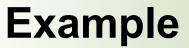




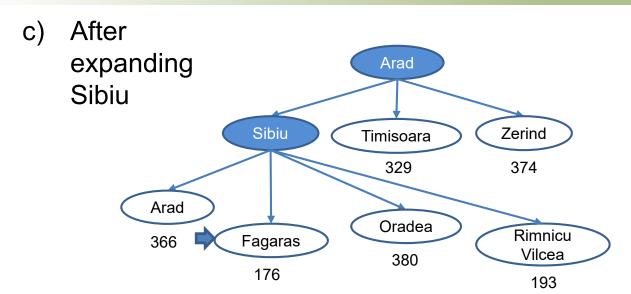
After expanding Arad



Arad	366
Bucharest	0
Craiova	160
Dobreta	242
Efoire	161
Fagaras	176
Giurgiu	77
Hirsova	151
Lasi	226
Lugoj	244
Mehadia	241
Neamt	234
Oradea	380
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Rimnicu Vilcea	193
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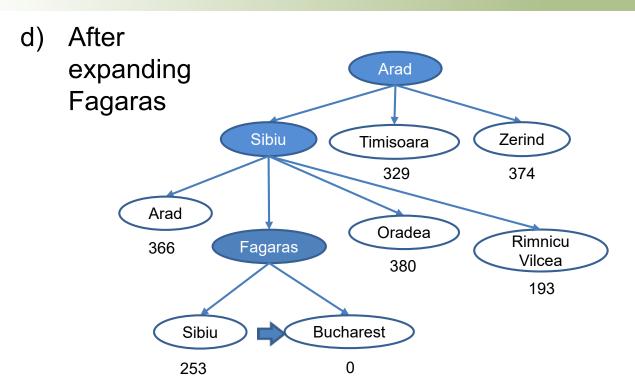




Arad	366
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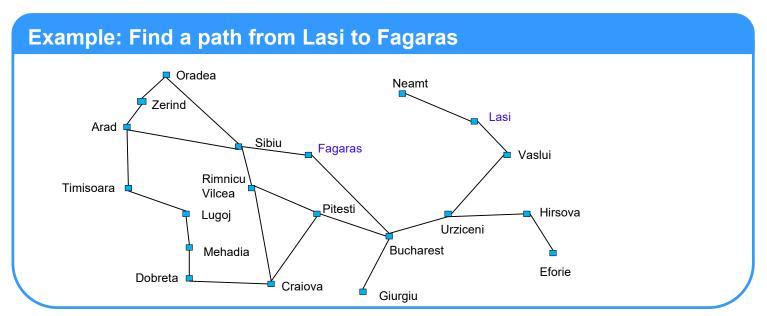


Arad	366
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Zerind	374



### Complete?

**Question:** Is this approach complete?



**Answer:** No





### **Greedy Search...**

• m: maximum depth of the search space

Complete	No
Time	$O(b^m)$
Space	$O(b^m)$ (keeps all nodes in memory)
Optimal	No

#### Question to think:

 Is it possible to combine functions g(n) and h(n) in one search strategy?





- Uniform-cost search
  - *g*(*n*): cost to reach n (Past Experience)
  - optimal and complete, but can be very inefficient
- Greedy search
  - h(n): cost from n to goal (Future Prediction)
  - neither optimal nor complete, but cuts search space considerably





Idea: Combine Greedy search with Uniform-Cost search

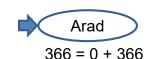
Evaluation function: f(n) = g(n) + h(n)

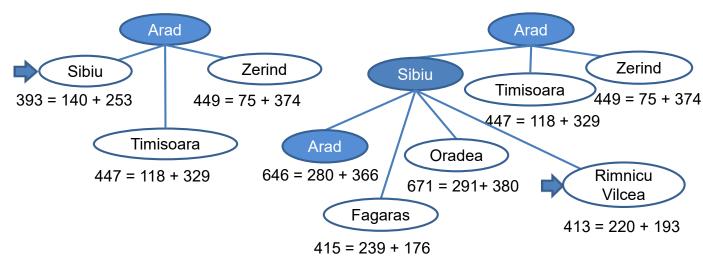
- f(n): estimated total cost of path through n to goal (Whole Life)
- If g = 0 → greedy search;
   If h = 0 → uniform-cost search
- Function A\* Search(problem) returns solution
  - Return Best-First-Search(problem, g + h)

Best-first-search with evaluation function g + h

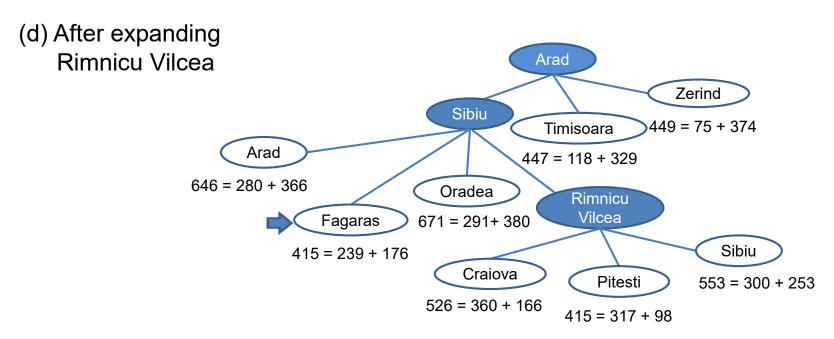
(a) The initial state (b) After expanding Arad

(c) After expanding Sibiu

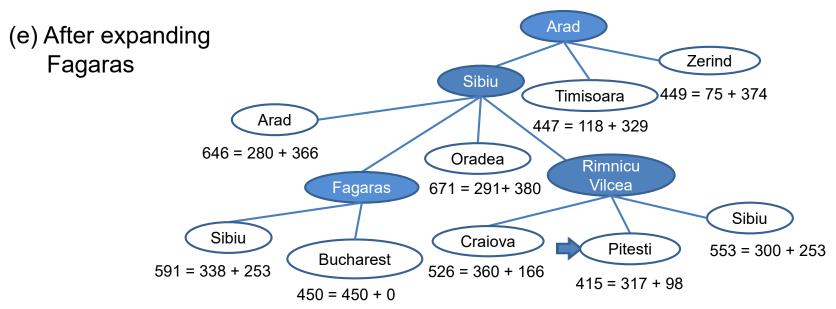


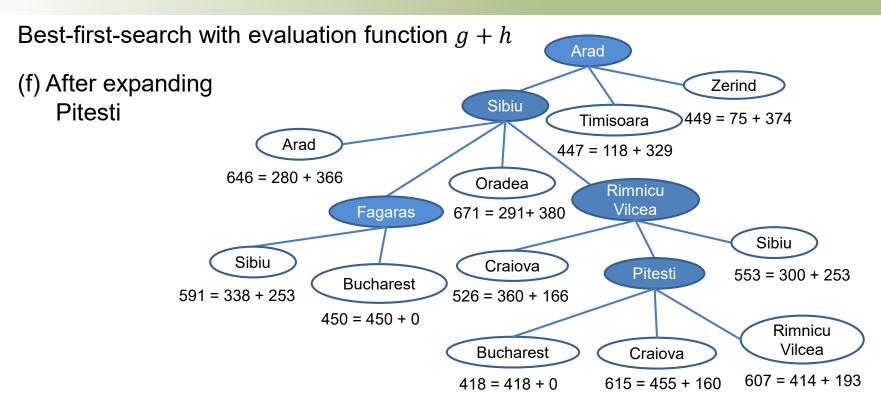


Best-first-search with evaluation function g + h



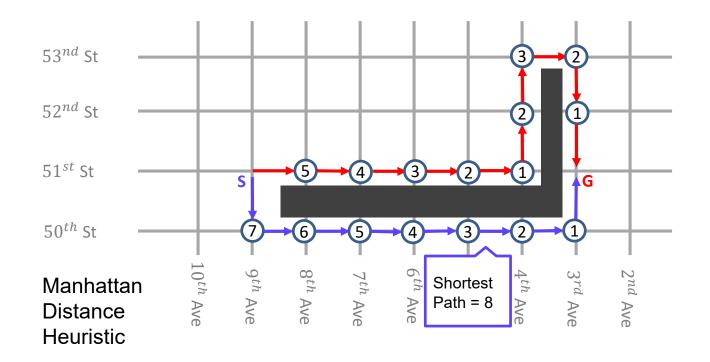
Best-first-search with evaluation function g + h



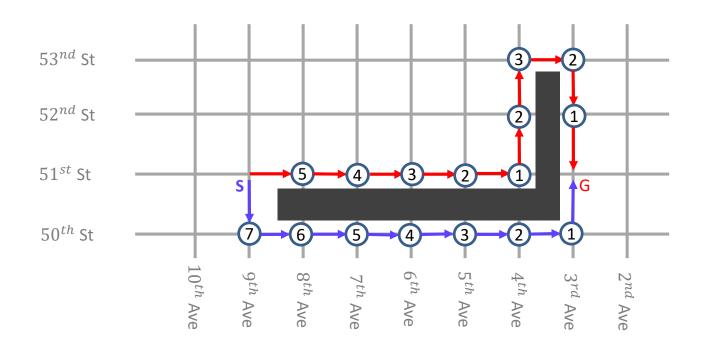




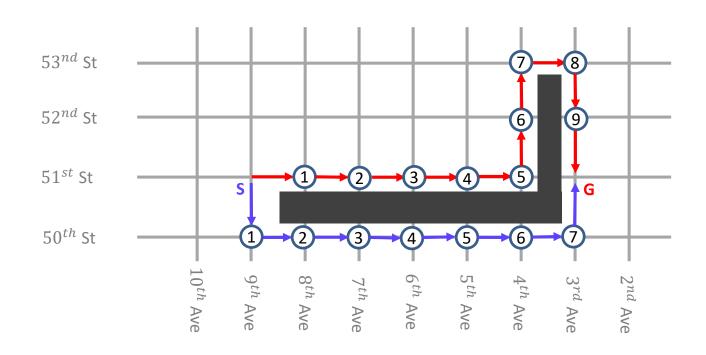
# **Example: Route-finding in Manhattan**



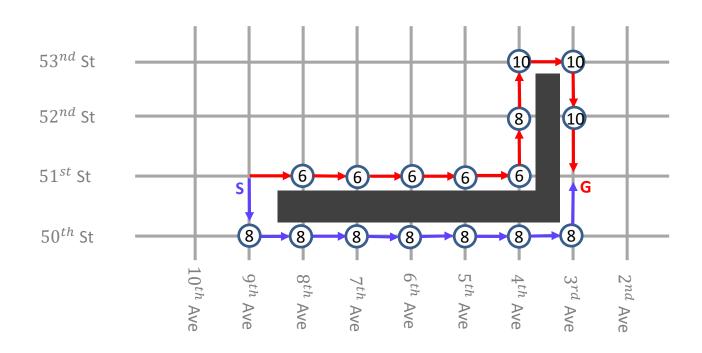
# **Example: Route-finding in Manhattan (Greedy)**



## **Example: Route-finding in Manhattan (UCS)**

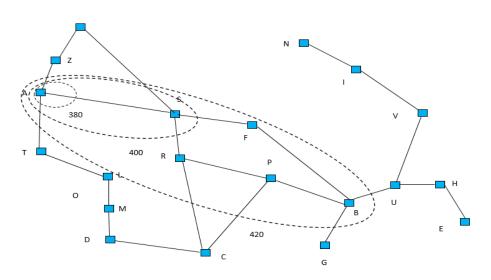


# **Example: Route-finding in Manhattan (A\*)**



# Complexity of A\*





Time	Exponential in length of solution
Space	(all generated nodes are kept in memory)  Exponential in length of solution

With a good heuristic, significant savings are still possible compared to uninformed search methods